

APPENDIX

Note

To have a better experience here is the dropbox (names and guides are the same):

<https://www.dropbox.com/scl/fo/ug40c12lm74480jm90jmy/h?rlkey=ii8l2b1pej7vhipa7py2caxt9&dl=0>

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[I] Original Project Brief (Next page)

IDE Master Graduation

Project team, Procedural checks and personal Project brief

This document contains the agreements made between student and supervisory team about the student's IDE Master Graduation Project. This document can also include the involvement of an external organisation, however, it does not cover any legal employment relationship that the student and the client (might) agree upon. Next to that, this document facilitates the required procedural checks. In this document:

- The student defines the team, what he/she is going to do/deliver and how that will come about.
- SSC E&SA (Shared Service Center, Education & Student Affairs) reports on the student's registration and study progress.
- IDE's Board of Examiners confirms if the student is allowed to start the Graduation Project.

! USE ADOBE ACROBAT READER TO OPEN, EDIT AND SAVE THIS DOCUMENT

Download again and reopen in case you tried other software, such as Preview (Mac) or a webbrowser.

STUDENT DATA & MASTER PROGRAMME

Save this form according the format "IDE Master Graduation Project Brief_familyname_firstname_studentnumber_dd-mm-yyyy". Complete all blue parts of the form and include the approved Project Brief in your Graduation Report as Appendix 1 !



family name _____
initials _____ given name _____
student number _____
street & no. _____
zipcode & city _____
country _____
phone _____
email _____

Your master programme (only select the options that apply to you):

IDE master(s): ☐ IPD ☐ Dfl ☐ SPD

2nd non-IDE master: _____

individual programme: _____ - - _____ (give date of approval)

honours programme: ☐ _____

specialisation / annotation: ☐ _____

☐ _____

☐ _____

SUPERVISORY TEAM **

Fill in the required data for the supervisory team members. Please check the instructions on the right !

** chair _____ dept. / section: _____

** mentor _____ dept. / section: _____

2nd mentor _____

organisation: _____

city: _____ country: _____

comments
(optional)

⋮

Chair should request the IDE Board of Examiners for approval of a non-IDE mentor, including a motivation letter and c.v..



Second mentor only applies in case the assignment is hosted by an external organisation.



Ensure a heterogeneous team. In case you wish to include two team members from the same section, please explain why.


APPROVAL PROJECT BRIEF

To be filled in by the chair of the supervisory team.

chair _____

date 4-10-2023

signature


CHECK STUDY PROGRESS

To be filled in by the SSC E&SA (Shared Service Center, Education & Student Affairs), after approval of the project brief by the Chair.
The study progress will be checked for a 2nd time just before the green light meeting.

Master electives no. of EC accumulated in total: _____ EC

Of which, taking the conditional requirements into account, can be part of the exam programme _____ EC

List of electives obtained before the third semester without approval of the BoE

☒ **YES** all 1st year master courses passed

☐ **NO** missing 1st year master courses are:

name _____

date - -

signature _____

FORMAL APPROVAL GRADUATION PROJECT

To be filled in by the Board of Examiners of IDE TU Delft. Please check the supervisory team and study the parts of the brief marked **. Next, please assess, (dis)approve and sign this Project Brief, by using the criteria below.

- Does the project fit within the (MSc)-programme of the student (taking into account, if described, the activities done next to the obligatory MSc specific courses)?
- Is the level of the project challenging enough for a MSc IDE graduating student?
- Is the project expected to be doable within 100 working days/20 weeks?
- Does the composition of the supervisory team comply with the regulations and fit the assignment?

Content: ☒ **APPROVED** ☐ **NOT APPROVED**

Procedure: ☐ **APPROVED** ☐ **NOT APPROVED**

comments

name _____

date - -

signature _____

Please state the title of your graduation project (above) and the start date and end date (below). Keep the title compact and simple. Do not use abbreviations. The remainder of this document allows you to define and clarify your graduation project.

start date - - - - end date

space available for images / figures on next page

introduction (continued): space for images

image / figure 1: _____

image / figure 2: _____

PROBLEM DEFINITION **

Limit and define the scope and solution space of your project to one that is manageable within one Master Graduation Project of 30 EC (= 20 full time weeks or 100 working days) and clearly indicate what issue(s) should be addressed in this project.

ASSIGNMENT **

State in 2 or 3 sentences what you are going to research, design, create and / or generate, that will solve (part of) the issue(s) pointed out in "problem definition". Then illustrate this assignment by indicating what kind of solution you expect and / or aim to deliver, for instance: a product, a product-service combination, a strategy illustrated through product or product-service combination ideas, In case of a Specialisation and/or Annotation, make sure the assignment reflects this/these.

PLANNING AND APPROACH **

Include a Gantt Chart (replace the example below - more examples can be found in Manual 2) that shows the different phases of your project, deliverables you have in mind, meetings, and how you plan to spend your time. Please note that all activities should fit within the given net time of 30 EC = 20 full time weeks or 100 working days, and your planning should include a kick-off meeting, mid-term meeting, green light meeting and graduation ceremony. Illustrate your Gantt Chart by, for instance, explaining your approach, and please indicate periods of part-time activities and/or periods of not spending time on your graduation project, if any, for instance because of holidays or parallel activities.

start date - - - - end date

MOTIVATION AND PERSONAL AMBITIONS

Explain why you set up this project, what competences you want to prove and learn. For example: acquired competences from your MSc programme, the elective semester, extra-curricular activities (etc.) and point out the competences you have yet developed. Optionally, describe which personal learning ambitions you explicitly want to address in this project, on top of the learning objectives of the Graduation Project, such as: in depth knowledge a on specific subject, broadening your competences or experimenting with a specific tool and/or methodology, Stick to no more than five ambitions.

FINAL COMMENTS

In case your project brief needs final comments, please add any information you think is relevant.

[II] Interviews

a. Questionnaires (Next page)

General information

Name of the participant: _____

Date of the survey: 18/10/23

By signing this document I hereby grant my consent for the utilization of the data I provided in the survey for research purposes.

Signature:



Context: How Might ChatGPT Improve the Accessibility of Quantum Computing?

The research project deals with the use of large language models, in particular, the widespread ChatGPT as a means of learning for complex topics such as quantum computing.

The purpose of the research is to explore the possibility of using this tool to help non-experts in the writing of working algorithms for Quantum computers, with the addition of explanations and guides to understand both the cognitive processes behind the code and the real implications of certain algorithms.

One of the first questions to be answered is “Who Could Benefit from This Approach?” to try to understand how to use this application and determine its level of difficulty and language complexity. The assumption is that people interested in the technology but not necessarily with coding skills would like to learn how to code. In particular, they are divided into 4 possible categories: Classical Code Quantum maestro, Classical Code Quantum Voyager, Quantum theory maestro, and Quantum code newbie.

Questions:

1. Given the assumption that the ideal target audience is *individuals interested in technology but lacking coding skills who aspire to acquire coding expertise*, do you believe this assumption to be accurate or not? Please provide your rationale for your stance.

I think this assumption is partially true. In my experience, people who have heard of quantum computing and want to take the next step of learning how to work with it usually have a general interest in technology and often some technical skills as well, even at a basic level. Because quantum computing is still at such an early stage, especially in terms of supporting developers, there's quite a high barrier to entry. I think ChatGPT would certainly be a natural fit for people who don't have coding skills generally and want to learn more, but the added context of quantum computing makes it less likely that people with no coding experience would engage with the topic in this way.

2. Considering the complexity of the subject matter, what type of preparation do you believe is requested for the ideal target audience of this project? In your perspective, is it imperative to possess a solid understanding of both quantum theory and coding? Or do you think having not such knowledge might not be an issue?

I don't think a deep knowledge of quantum mechanics is needed to work with quantum computing in a practical sense, but since the way we currently work with quantum computers is mostly through existing programming languages, some knowledge of coding would be very helpful. I think if someone approaches this topic with no knowledge whatsoever of coding it might be too much information at once. That being said, the level at which simple QC programs are written is probably quite accessible for someone who has a small amount of coding experience. For very simple circuits it's mostly just initialising a few qubits and adding some gates to them, then measuring and seeing what happens. If someone understands what objects, variables, functions, etc. are at a basic level I think it would be enough.

3. Still regarding the identification of the ideal user; do you believe that there exists a more appropriate academic level that would facilitate a better understanding of these issues? Please provide your rationale for your statement.

I can imagine perhaps students in a computer science course—specifically those focused on development or implementation—using this approach to quickly generate and play with code snippets. Such students would hopefully have enough understanding of programming concepts and techniques to be able to parse the code whilst benefitting from the rapidity and scaffolding this approach might offer. I do wonder, though, if the reliability of ChatGPT's generated code might cause problems; if someone doesn't already have expertise in coding with/for quantum computers, how can they identify potential errors in the code?

4. In conclusion, do you perceive potential domains where the application and instruction of this approach to non-experts could generate even greater value? For instance, in fields like design, economics, geology, or any other areas you may have in mind?

Perhaps this approach could be helpful for developers in potential application domains (e.g. finance, cybersecurity, manufacturing) to start building an awareness of and hands-on skills for working with quantum computers. For technologies in general, this approach could be useful for a range of technology-adjacent disciplines, although I do think that more clarity is needed on how to identify errors or misalignments in generated code without expertise.

Thank you for your collaboration and time spent, it is really appreciated. If you are interested in finding out the search results and staying informed please select yes or no.

☒ YES ☐ NO

General information

Name of the participant: *Subir Kumar*

Date of the survey: 9/10/2023

By signing this document I hereby grant my consent for the utilization of the data I provided in the survey for research purposes.

Signature:



Context: How Might ChatGPT Improve the Accessibility of Quantum Computing?

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The purpose of the research is to explore the possibility of using this tool to help non-experts in the writing of working algorithms for Quantum computers, with the addition of explanations and guides to understand both the cognitive processes behind the code and the real implications of certain algorithms.

One of the first questions to be answered is "*Who Could Benefit from This Approach?*" to try to understand how to use this application and determine its level of difficulty and language complexity.

The assumption is that people interested in the technology but not necessarily with coding skills would like to learn how to code. In particular, they are divided into 4 possible categories: Classical Code Quantum maestro, Classical Code Quantum Voyager, Quantum theory maestro, and Quantum code newbie.

Questions:

1. Given the assumption that the ideal target audience is *individuals interested in technology but lacking coding skills who aspire to acquire coding expertise*, do you believe this assumption to be accurate or not? Please provide your rationale for your stance.

Not completely accurate. I think also people with traditional coding skills but not accustomed with quantum programming could resort to Chat GPT for help.

2. Considering the complexity of the subject matter, what type of preparation do you believe is requested for the ideal target audience of this project? In your perspective, is it imperative to possess a solid understanding of both quantum theory and coding? Or do you think having not such knowledge might not be an issue?

In my perspective, it is not imperative to possess solid understanding of both quantum theory and coding. General understanding of both could be sufficient, at the level of introductory courses or MOOCs (massive online courses).

3. Still regarding the identification of the ideal user; do you believe that there exists a more appropriate academic level that would facilitate a better understanding of these issues? Please provide your rationale for your statement.

This question is a bit unclear to me. Perhaps refer to my answer above

4. In conclusion, do you perceive potential domains where the application and instruction of this approach to non-experts could generate even greater value? For instance, in fields like design, economics, geology, or any other areas you may have in mind?

I am not an expert of these language models such as ChatGPT, but I assume this approach could have the potential to simplify coding in the field of quantum algorithms in general. However I see the potential danger of lack of quality control over the results. With no expert overseeing the results I don't know in practice if the current language models are sufficiently advanced to provide useful cases.

Thank you for your collaboration and time spent, it is really appreciated. If you are interested in finding out the search results and staying informed please write yes or no below.

b. Findings

Out of all the experts contacted, only two responded, shedding light on several points: The assumption about the ideal target audience being non-experts is partly true, but a keen interest is necessary. Those with minimal coding or quantum knowledge might also find the method of chatGPT beneficial. While a deep understanding of quantum mechanics is not crucial, a basic grasp can aid in addressing potential issues, especially since the language of code can be daunting for non-coders. A significant consideration raised is how to verify if the outcomes of the approach are correct, indicating an area for further investigation.

[III] User test #1

a. Framework for Identifying Potential Users

		YES Knowledge quantum computing	NO Knowledge quantum computing
Coding classical computing	YES		
	NO		
Coding quantum computing	YES		
	NO		

		YES Knowledge quantum computing	NO Knowledge quantum computing
Coding classical computing	YES		×
	NO	×	×
Coding quantum computing	YES		
	NO		

b. Template (Next page)

User test #1 - Participant informations

Participant number:

P5

note: this is for the researcher to fill in

Date of the text

*Participant name:

*Participant Background:

*Participant Age:

By signing this document I hereby grant my consent for the utilization of the data I provided in the survey for research purposes.

*Signature:

User test #1 - Questionnaire

1. How confident do you feel about your final result from 1 to 5, where 1 is not very confident and 5 is very confident?

1

2

3

4

5

2. How well do you believe you understood the final code from 1 to 5, where 1 is I didn't understand anything and 5 is I understood everything?

1

2

3

4

5

3. How much do you believe your background helped you in understanding the final code from 1 to 5, where 1 is very little and 5 is a lot?

1

2

3

4

5

4. Do you think this approach could be useful in your field? ☐ Yes ☐ No

Explain your answer

5. Feedbacks and reflactions

PHASE

Prompting

Coding

Iteration

Evaluation

STEPS

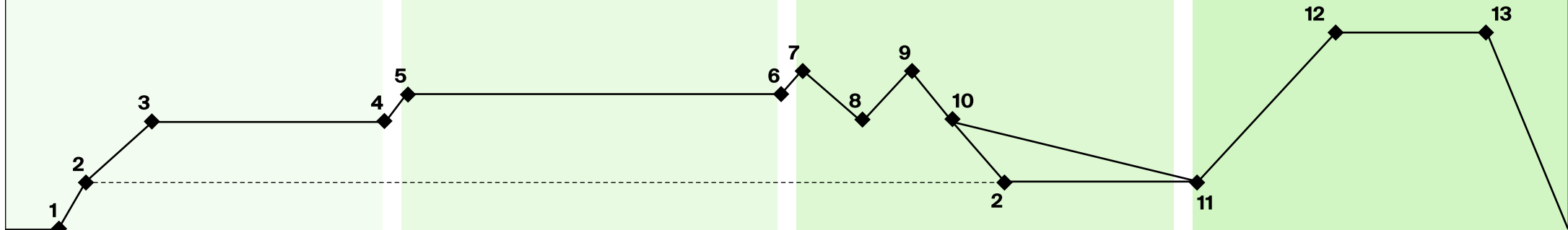
1. Having a problem to solve
2. Open chatGPT
3. Frame the problem

4. Open IBM quantum
5. Copy paste the code

6. Search for errors
7. Check ChatGPT
8. Debugging
9. Try again
10. Start a new prompt
(if needed)

11. Final code
12. Understanding
13. Check working

TOUCHPOINTS



EMOTIONS

KNOWLEDGE

c. Participant's results (Next page)

P1

New Chat

Yesterday

Quantum Crypto Brea

Previous 7 Days

Quantum Code with ChatGPT

CODE PER USER TEST

New chat

Quantum Cryptography with

Quantum Logistics & Supply

Roman Urine Fun Fact

Shor's Quantum Factorization

Quantum Coding Workshop

Polite User Request

Previous 30 Days

Research Participation Availa

Costanza Darcangelisa

GPT-4

I want to develop a code to break cryptography utilising the Grover algorithm to run on the IBM simulator with 32 qubits. First, ask me what my level of expertise is and give me three options: expert in quantum, expert in coding, non expert. Wait for my answer. Then based on my response explain to me what breaking cryptography with a quantum computer means in a few sentences. Then give me a one-sentence explanation of what the Grover algorithm is. Ask me if everything is clear and wait for my response (Y/N). After that ask me if I want to program the code. Wait for my response (Y/N). based on that explain to me the code step by step and include possible libraries inside the explanation.

Alright, let's start. What is your level of expertise?

1. Expert in quantum.

2. Expert in coding.

3. Non-expert.

Non-expert

Got it!

Breaking Cryptography with a Quantum Computer:

Traditional cryptographic systems rely on the difficulty of certain mathematical problems. Quantum computers, using algorithms like Grover's, can solve some of these problems faster than classical computers, potentially compromising the security of many cryptographic systems.

The Grover Algorithm:

It's a quantum algorithm that finds an unsorted database's marked item with square root speedup over classical methods.

Is everything clear so far? (Y/N)

Y

Great!

Do you want to program the code for the Grover algorithm to run on the IBM simulator with 32 qubits? (Y/N)

Y

Alright!

Given that you're a non-expert, I'll break it down into simple steps and provide explanations for each. Let's begin:

Step 1: Setting Up Your Environment

Before we start, you need to have the appropriate tools and libraries. The main library we'll use is `'qiskit'`, IBM's quantum computing SDK.

python

Copy code

```
pip install qiskit
```

Step 2: Initialization

Import necessary libraries and initialize the quantum circuit.

python

Copy code

```
from qiskit import Aer, QuantumCircuit, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import AerSimulator
from qiskit.algorithms import Grover
from qiskit.circuit.library import PhaseOracle
```

Step 3: Define the Problem

For this example, let's consider you're trying to find a specific binary string (let's say "101").

The Grover algorithm will help us find this string faster than classical methods.

Step 4: Create Oracle

The Oracle is a component that marks the desired solution. We'll create a PhaseOracle that marks the state "101".

```
python Copy code  
  
oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
```

Here, 'x[0]', 'x[1]', and 'x[2]' refer to qubits. The above oracle marks the state |101>.

Step 5: Running Grover's Algorithm

Now, we will initialize Grover's algorithm with the oracle and specify the number of amplification iterations (let's say 1 for this example).

```
python Copy code  
  
grover = Grover(oracle=oracle, iterations=1)
```

Step 6: Simulating the Circuit

We'll now set up a quantum simulator, run Grover's algorithm, and get the results.

```
python Copy code  
  
simulator = AerSimulator()  
compiled_circuit = transpile(grover.construct_circuit(), simulator)  
result = simulator.run(compiled_circuit).result()  
counts = result.get_counts()  
plot_histogram(counts)
```

The histogram will show the probability of each state being the result. The marked item "101" should have the highest probability.

Step 7: Running on IBM Simulator

To run on the IBM simulator with 32 qubits, you'd typically need to set up an account with IBM, get your API token, and set the backend to IBM's 32-qubit simulator. However, this simple example won't cover that step due to its complexity for beginners. It's a more advanced step you'd need to take once you're comfortable with the basic workings of the Grover algorithm in Qiskit.

That's a basic run-through of using Grover's algorithm in Qiskit. You'd adjust the problem definition, oracle, and iterations based on your specific problem. Hope this helps you get started!



Traceback (most recent call last):

```
Cell In[5], line 1  
    oracle = PhaseOracle('x[0] & ~x[1] & x[2]')  
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/cls_tools.py:111 in out  
    callback._get_(obj, objtype)(*args, **kwargs)  
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:39 in __call__  
    self._tester.require_now(self._feature)  
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:223 in require_now  
    raise MissingOptionalLibraryError(  
MissingOptionalLibraryError: The 'tweedledum' library is required to use 'PhaseOracle'. You  
can install it with 'pip install tweedledum'.
```

Use %tb to get the full traceback.



It looks like you're missing the `'tweedledum'` library, which is required for the `'PhaseOracle'` in Qiskit. `'tweedledum'` is a library that Qiskit uses for certain quantum compilation and synthesis tasks.

Here's how you can resolve it:

Step 1: Install tweedledum

Install the `'tweedledum'` library using pip:

```
python Copy code  
  
pip install tweedledum
```


Step 2: Run the code again

After installing `twedledum`, you should be able to execute the code without the `MissingOptionalLibraryError`.

Please try this solution and let me know if you face any further issues!



it doesn't work



I'm sorry to hear that. Let's try to troubleshoot the problem.



1. Double-checking Installation:

Ensure that `twedledum` has been successfully installed in the same environment where Qiskit is installed. If you're using virtual environments, this could be a potential source of confusion.

2. Alternative Oracle:

If the problem persists and is specific to `PhaseOracle`, as a workaround, you can use another type of oracle or a manually constructed one. This might not be as convenient as `PhaseOracle` but can get you started without the dependency.

3. Check Qiskit Version:

Sometimes, there might be compatibility issues between versions of Qiskit and certain features. You can try updating Qiskit:

```
python
```

Copy code

```
pip install qiskit --upgrade
```

4. Environment Issue:

If you're using Jupyter or another environment, try restarting the kernel or even the environment itself. Sometimes, freshly installed packages may not be recognized until the environment is refreshed.

If none of these solutions work, you might need to reach out to specific communities or forums related to Qiskit for more specialized help. There might be known issues or bugs that the community or the Qiskit developers are aware of.

Let me know if you'd like to proceed with an alternative approach or if you have more information on the error to provide further assistance.



Traceback (most recent call last):



```
Cell In[4], line 1
  oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:112 in out
  retval = method(*args, **kwargs)
File /opt/conda/lib/python3.10/site-packages/qiskit/circuit/library/phase_oracle.py:70 in
__init__
  expression = BooleanExpression(expression, var_order=var_order)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:112 in out
  retval = method(*args, **kwargs)
File /opt/conda/lib/python3.10/site-
packages/qiskit/circuit/classicalfunction/boolean_expression.py:39 in __init__
  self.twedledum_bool_expression = BoolFunction.from_expression(
File /opt/conda/lib/python3.10/site-
packages/twedledum/bool_function_compiler/bool_function.py:148 in from_expression
  parsed_expression = ExpressionParser(expression, var_order)
File /opt/conda/lib/python3.10/site-
packages/twedledum/bool_function_compiler/expression_parser.py:14 in __init__
  super().__init__(source)
File /opt/conda/lib/python3.10/site-
packages/twedledum/bool_function_compiler/function_parser.py:38 in __init__
  self.visit(node)
File /opt/conda/lib/python3.10/ast.py:412 in visit
  return visitor(node)
File /opt/conda/lib/python3.10/site-
packages/twedledum/bool_function_compiler/expression_parser.py:24 in visit_Module
  expr_type, expr_signal = self.visit(node.body[0])
File /opt/conda/lib/python3.10/ast.py:412 in visit
  return visitor(node)
File /opt/conda/lib/python3.10/site-
packages/twedledum/bool_function_compiler/expression_parser.py:30 in visit_Expr
  return super().visit(node.value)
File /opt/conda/lib/python3.10/ast.py:412 in visit
  return visitor(node)
```

```
File /opt/conda/lib/python3.10/site-
packages/tweedledum/bool_function_compiler/function_parser.py:72 in visit_BinOp
    left_type, left_signals = self.visit(node.left)
File /opt/conda/lib/python3.10/ast.py:412 in visit
    return visitor(node)
File /opt/conda/lib/python3.10/site-
packages/tweedledum/bool_function_compiler/function_parser.py:73 in visit_BinOp
    right_type, right_signals = self.visit(node.right)
File /opt/conda/lib/python3.10/ast.py:412 in visit
    return visitor(node)
File /opt/conda/lib/python3.10/site-
packages/tweedledum/bool_function_compiler/function_parser.py:225 in visit_UnaryOp
    result_type, result_signal = self.visit(node.operand)
File /opt/conda/lib/python3.10/ast.py:412 in visit
    return visitor(node)
File /opt/conda/lib/python3.10/site-
packages/tweedledum/bool_function_compiler/function_parser.py:219 in visit_Subscript
    return v_type, [v_signals[slice_]]
IndexError: list index out of range
```

Use %tb to get the full traceback.



This traceback indicates that there's a problem when Qiskit tries to interpret and construct the oracle from the Boolean expression `'x[0] & ~x[1] & x[2]'`.



This could be due to:

1. A genuine bug in Qiskit or its dependencies.
2. A mismatch between versions of different components.
3. An issue with the specific way the Boolean expression is formatted.

Let's try an alternative approach.

Alternative Oracle using Gates:

Instead of using `'PhaseOracle'`, you can manually construct an oracle that marks the state `|101>` using basic quantum gates.

Here's a potential way to do it:

1. Start with a 3-qubit quantum circuit.
2. Apply X-gate on the second qubit to flip its state.
3. Apply a multi-controlled-Z gate (CCZ) using the three qubits. This applies a phase flip to the state `|101>`.
4. Apply X-gate again on the second qubit to revert its state.

Here's how to code that:

python

Copy code

```
from qiskit import QuantumCircuit

def custom_oracle():
    qc = QuantumCircuit(3)
    qc.x(1) # Flip the second qubit
    qc.h(2) # Put the last qubit into superposition
    qc.ccx(0,1,2) # CCX gate (Toffoli gate)
    qc.h(2) # Bring the last qubit back from superposition
    qc.x(1) # Revert the flip on the second qubit
    return qc

oracle = custom_oracle()
oracle.draw('mpl') # This is to visualize the oracle
```

This circuit can be used as a custom oracle for the Grover's algorithm. The rest of the steps involving Grover's algorithm and simulation remain the same.

Would you like to proceed with this approach, or would you prefer to delve deeper into the initial error and attempt to solve it?



yes i want to proceed with this approach



Great! Let's proceed with the custom oracle approach.



Having defined our custom oracle, the next steps involve setting up Grover's algorithm, simulating it, and visualizing the results.

Step 1: Create the Grover Diffuser

Step 1: Create the Grover Diffuser

The Grover diffuser is an essential component of Grover's algorithm, which, in conjunction with the oracle, amplifies the probability amplitude of the marked state.

Here's a simple diffuser for a 3-qubit system:

```
python Copy code

def grover_diffuser():
    qc = QuantumCircuit(3)
    # Apply H and X gates to all qubits
    qc.h([0,1,2])
    qc.x([0,1,2])

    # Multi-controlled-Z gate
    qc.h(2)
    qc.ccx(0,1,2)
    qc.h(2)

    # Apply X and H gates again
    qc.x([0,1,2])
    qc.h([0,1,2])
    return qc

diffuser = grover_diffuser()
```

Step 2: Combine Oracle and Diffuser for Grover's Iteration

Now we'll set up Grover's iteration, which includes the oracle and the diffuser.

```
python Copy code

def grover_iteration():
    qc = QuantumCircuit(3)
    qc += custom_oracle()
    qc += grover_diffuser()
    return qc

grover_circuit = grover_iteration()
```

Step 3: Simulate the Circuit

Let's prepare the initial state, run the Grover iteration, and simulate the results:

```
python Copy code

# Setting up the simulator
from qiskit import Aer, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import AerSimulator

simulator = AerSimulator()

# Initial state preparation
qc = QuantumCircuit(3, 3)
qc.h([0,1,2]) # Start with a superposition of all states
qc += grover_circuit # Add the Grover iteration
qc.measure([0,1,2], [0,1,2]) # Measure all qubits

# Simulate and visualize results
compiled_circuit = transpile(qc, simulator)
result = simulator.run(compiled_circuit).result()
counts = result.get_counts()
plot_histogram(counts)
```

The histogram should show the probability of each state being the result. The marked item "101" should have a significantly higher probability compared to other states.

That's it! You've now manually set up Grover's algorithm using a custom oracle. This approach avoids the issues with `PhaseOracle` and still achieves the same objective.



have i to write first from qiskit import QuantumCircuit



Yes, if you haven't imported the required modules from `'qiskit'` at the start of your session or in your script, you need to do so. It's a standard practice to import necessary modules and libraries at the beginning of your code.



So, before defining the custom oracle and proceeding with the steps, make sure to have:

```
python
from qiskit import QuantumCircuit
```

And then, when we get to the simulation and visualization steps, ensure you've imported the required functionalities:

```
python
from qiskit import Aer, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import AerSimulator
```

These imports make sure you have all the necessary classes and functions available for use in the subsequent steps. If you're writing the code in a script or notebook, it's a good practice to have all imports at the top for clarity.



Traceback (most recent call last):
Cell In[5], line 7
grover_circuit = grover_iteration()
Cell In[5], line 3 in grover_iteration
qc += custom_oracle()
NameError: name 'custom_oracle' is not defined

Use %tb to get the full traceback.



The error message indicates that the `'custom_oracle'` function isn't recognized when you're trying to use it in the `'grover_iteration'` function. This could be due to one of the following reasons:

1. You might not have defined the `'custom_oracle'` function before trying to call it.
2. There could be an execution order problem, especially if you're working in a Jupyter notebook.

To resolve this:

1. Ensure that you've run the cell (or written the code) that contains the `'custom_oracle'` function definition.
2. After defining `'custom_oracle'`, run the cell (or execute the code) that contains the `'grover_iteration'` function and then proceed with the rest of the steps.

If you're using a Jupyter notebook, be careful about the order in which you run cells, as running them out of order can sometimes lead to such issues.

Here's a quick recap:

1. Define the `'custom_oracle'` function.
2. Define the `'grover_diffuser'` function.
3. Define the `'grover_iteration'` function.
4. Finally, proceed with the simulation steps.

Make sure to execute or run each step in the above order.

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Python 3 (ipykernel) ○

[1]: pip install qiskit

Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.44.1)
Requirement already satisfied: qiskit-terra==0.25.1 in /opt/conda/lib/python3.10/site-packages (from qiskit) (0.25.1)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (2.8.2)
Requirement already satisfied: numpy>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (1.23.5)
Requirement already satisfied: symengine<0.10,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (0.9.2)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (4.5.0)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (0.3.7)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (4.1.1)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (3.11)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (5.9.4)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (0.13.0)
Requirement already satisfied: scipy>=1.5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (1.9.3)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.1->qiskit) (1.11.1)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra==0.25.1->qiskit) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra==0.25.1->qiskit) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra==0.25.1->qiskit) (1.3.0)

[notice] A new release of pip available: 23.1.1 -> 23.2.1
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.

[2]: from qiskit import Aer, QuantumCircuit, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import AerSimulator
from qiskit.algorithms import Grover
from qiskit.circuit.library import PhaseOracle

/tmp/ipykernel_60/641646577.py:4: DeprecationWarning: ``qiskit.algorithms`` has been migrated to an independent package: <https://github.com/qiskit-community/qiskit-algorithms>. The ``qiskit.algorithms`` import path is deprecated as of qiskit-terra 0.25.0 and will be removed no earlier than 3 months after the release date. Please run ``pip install qiskit_algorithms`` and use ``import qiskit_algorithms`` instead.
from qiskit.algorithms import Grover

[3]: oracle = PhaseOracle('x[0] & ~x[1] & x[2]')

Traceback (most recent call last):
Cell In[3], line 1
oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:111 in out
callcallback.__get__(obj, objtype)(*args, **kwargs)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:39 in __call__
self._tester.require_now(self._feature)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:223 in require_now
raise MissingOptionalLibraryError(
MissingOptionalLibraryError: The 'tweedledum' library is required to use 'PhaseOracle'. You can install it with 'pip install tweedledum'.

Use %tb to get the full traceback.

Search for solution online

[4]: pip install tweedledum

Collecting tweedledum
Downloading tweedledum-1.1.1-cp310-cp310-manylinux_2_12_x86_64.manylinux2010_x86_64.whl (929 kB)
929.7/929.7 kB 1.5 MB/s eta 0:00:0000:0100:01
Installing collected packages: tweedledum
Successfully installed tweedledum-1.1.1

[notice] A new release of pip available: 23.1.1 -> 23.2.1
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.

[5]: oracle = PhaseOracle('x[0] & ~x[1] & x[2]')

Traceback (most recent call last):
Cell In[5], line 1
oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:111 in out
callcallback.__get__(obj, objtype)(*args, **kwargs)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:39 in __call__
self._tester.require_now(self._feature)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:223 in require_now
raise MissingOptionalLibraryError(
MissingOptionalLibraryError: The 'tweedledum' library is required to use 'PhaseOracle'. You can install it with 'pip install tweedledum'.

Use %tb to get the full traceback.

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[6]: pip install tweedledum

Requirement already satisfied: tweedledum in /opt/conda/lib/python3.10/site-packages (1.1.1)

[notice] A new release of pip available: 23.1.1 -> 23.2.1
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.

[7]: oracle = PhaseOracle('x[0] & ~x[1] & x[2]')

Traceback (most recent call last):

```
Cell In[7], line 1
oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:111 in out
    callback.__get__(obj, objtype)(*args, **kwargs)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:39 in __call__
    self._tester.require_now(self._feature)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:223 in require_now
    raise MissingOptionalLibraryError(
MissingOptionalLibraryError: The 'tweedledum' library is required to use 'PhaseOracle'. You can install it with 'pip install tweedledum'.

Use %tb to get the full traceback.
```

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```
[9]: pip install qiskit
```

```
/bin/bash: line 1: tweedledum: command not found
Note: you may need to restart the kernel to use updated packages.
```

```
[10]: pip install qiskit --upgrade
```

```
Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.44.1)
Collecting qiskit
  Downloading qiskit-0.44.2-py3-none-any.whl (8.2 kB)
Collecting qiskit-terra==0.25.2.1
  Downloading qiskit_terra-0.25.2.1-cp38-abi3-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (6.2 MB)
    6.2/6.2 MB 4.7 MB/s eta 0:00:00:0100:01
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (1.11.1)
Requirement already satisfied: symengine<0.10,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (0.9.2)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (0.13.0)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (4.5.0)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (0.3.7)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (3.11)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (4.1.1)
Requirement already satisfied: scipy>=1.5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (1.9.3)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (2.8.2)
Requirement already satisfied: numpy>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (1.23.5)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (5.9.4)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra==0.25.2.1->qiskit) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra==0.25.2.1->qiskit) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra==0.25.2.1->qiskit) (1.3.0)
Installing collected packages: qiskit-terra, qiskit
  Attempting uninstall: qiskit-terra
    Found existing installation: qiskit-terra 0.25.1
    Uninstalling qiskit-terra-0.25.1:
      Successfully uninstalled qiskit-terra-0.25.1
  Attempting uninstall: qiskit
    Found existing installation: qiskit 0.44.1
    Uninstalling qiskit-0.44.1:
      Successfully uninstalled qiskit-0.44.1
Successfully installed qiskit-0.44.2 qiskit-terra-0.25.2.1

[notice] A new release of pip available: 23.1.1 -> 23.2.1
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.
```

```
[11]: oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
```

```
Traceback (most recent call last):
  Cell In[11], line 1
    oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:111 in out
    callback.__get__(obj, objtype)(*args, **kwargs)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:39 in __call__
    self._tester.require_now(self._feature)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:223 in require_now
    raise MissingOptionalLibraryError(
MissingOptionalLibraryError: The 'tweedledum' library is required to use 'PhaseOracle'. You can install it with 'pip install tweedledum'.

Use %tb to get the full traceback.
```

[Search for solution online](#)

```
[12]: pip install tweedledum --upgrade
```

```
Requirement already satisfied: tweedledum in /opt/conda/lib/python3.10/site-packages (1.1.1)

[notice] A new release of pip available: 23.1.1 -> 23.2.1
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.
```

```
[1]: oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
```

```
Traceback (most recent call last):
  Cell In[1], line 1
    oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
NameError: name 'PhaseOracle' is not defined

Use %tb to get the full traceback.
```

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```
[2]: from qiskit import Aer, QuantumCircuit, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import AerSimulator
from qiskit.algorithms import Grover
from qiskit.circuit.library import PhaseOracle
```

```
/tmp/ipykernel_125/2889406778.py:4: DeprecationWarning: ``qiskit.algorithms`` has been migrated to an independent package: https://github.com/qiskit-community/qiskit-algorithms. The ``qiskit.algorithms`` import path is deprecated as of qiskit-terra 0.25.0 and will be removed no earlier than 3 months after the release date. Please run ``pip install qiskit_algorithms`` and use ``import qiskit_algorithms`` instead.
    from qiskit.algorithms import Grover
```



```
[3]: pip install qiskit
```

```
Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.44.2)
Requirement already satisfied: qiskit-terra==0.25.2.1 in /opt/conda/lib/python3.10/site-packages (from qiskit) (0.25.2.1)
Requirement already satisfied: scipy>=1.5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (1.9.3)
Requirement already satisfied: numpy>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (1.23.5)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (1.11.1)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (5.9.4)
Requirement already satisfied: symengine<0.10,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (0.9.2)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (4.1.1)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (0.13.0)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (0.3.7)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (4.5.0)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (3.11)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.2.1->qiskit) (2.8.2)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra==0.25.2.1->qiskit) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra==0.25.2.1->qiskit) (5.11.1)
Requirement already satisfied: mpmath=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra==0.25.2.1->qiskit) (1.3.0)

[notice] A new release of pip available: 23.1.1 -> 23.2.1
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.
```

```
[4]: oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
```

```
Traceback (most recent call last):
  Cell In[4], line 1
    oracle = PhaseOracle('x[0] & ~x[1] & x[2]')
  File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:112 in out
    retval = method(*args, **kwargs)
  File /opt/conda/lib/python3.10/site-packages/qiskit/circuit/library/phase_oracle.py:70 in __init__
    expression = BooleanExpression(expression, var_order=var_order)
  File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:112 in out
    retval = method(*args, **kwargs)
  File /opt/conda/lib/python3.10/site-packages/qiskit/circuit/classicalfunction/boolean_expression.py:39 in __init__
    self._tweedledum_bool_expression = BoolFunction.from_expression(
  File /opt/conda/lib/python3.10/site-packages/tweedledum/bool_function_compiler/bool_function.py:148 in from_expression
    parsed_expression = ExpressionParser(expression, var_order)
  File /opt/conda/lib/python3.10/site-packages/tweedledum/bool_function_compiler/expression_parser.py:14 in __init__
    super().__init__(source)
  File /opt/conda/lib/python3.10/site-packages/tweedledum/bool_function_compiler/function_parser.py:38 in __init__
    self.visit(node)
  File /opt/conda/lib/python3.10/ast.py:412 in visit
    return visitor(node)
  File /opt/conda/lib/python3.10/site-packages/tweedledum/bool_function_compiler/expression_parser.py:24 in visit_Module
    expr_type, expr_signal = self.visit(node.body[0])
  File /opt/conda/lib/python3.10/ast.py:412 in visit
    return visitor(node)
  File /opt/conda/lib/python3.10/site-packages/tweedledum/bool_function_compiler/expression_parser.py:30 in visit_Expr
    return super().visit(node.value)
  File /opt/conda/lib/python3.10/ast.py:412 in visit
    return visitor(node)
  File /opt/conda/lib/python3.10/site-packages/tweedledum/bool_function_compiler/function_parser.py:72 in visit_BinOp
    left_type, left_signals = self.visit(node.left)
  File /opt/conda/lib/python3.10/ast.py:412 in visit
    return visitor(node)
  File /opt/conda/lib/python3.10/site-packages/tweedledum/bool_function_compiler/function_parser.py:73 in visit_BinOp
    right_type, right_signals = self.visit(node.right)
  File /opt/conda/lib/python3.10/ast.py:412 in visit
    return visitor(node)
  File /opt/conda/lib/python3.10/site-packages/tweedledum/bool_function_compiler/function_parser.py:225 in visit_UnaryOp
    result_type, result_signal = self.visit(node.operand)
  File /opt/conda/lib/python3.10/ast.py:412 in visit
    return visitor(node)
  File /opt/conda/lib/python3.10/site-packages/tweedledum/bool_function_compiler/function_parser.py:219 in visit_Subscript
    return v_type, [v_signals[slice_]]
IndexError: list index out of range
```

Use %tb to get the full traceback.

[Search for solution online](#)

```
[ ]:
```

P2

+ New Chat

Today

Quantum Cryptograp

Quantum Crypto Code Expla

New chat

Yesterday

User-Friendly Design Principi

Quantum Cryptography with

Previous 7 Days

Sinonimi per "nostalgia"

Quantum Cryptography with

Educazione e plagio AI

Quantum Crypto Breaker - Inf

Quantum Code with ChatGPT

CODE PER USER TEST

Costanza Darcangelisa

🚀 GPT-4

I want to develop a code to break cryptography utilising the Grover algorithm to run on the IBM simulator with 32 qubits. First, ask me what my level of expertise is and give me three options: expert in quantum, expert in coding, non expert. Wait for my answer. Then based on my response explain to me what breaking cryptography with a quantum computer means in a few sentences. Then give me a one-sentence explanation of what the Grover algorithm is. Ask me if everything is clear and wait for my response (Y/N). After that ask me if I want to program the code. Wait for my response (Y/N). based on that explain to me the code step by step and include possible libraries inside the explanation.

Of course!

First, could you please tell me about your level of expertise? Here are the three options:

1. Expert in quantum.

2. Expert in coding.

3. Non-expert.

Which one best describes you?

Non-expert

Alright, as a non-expert, let's start with some basic explanations:

1. **Breaking Cryptography with a Quantum Computer:** Classical computers use bits as their basic unit of information, which can be either 0 or 1. Quantum computers use qubits, which can be in a state representing 0, 1, or any quantum superposition of these states. This enables quantum computers to solve certain problems much faster than classical computers. One such problem is breaking some types of cryptographic codes. Classical cryptographic codes, like some used in online transactions, are based on the difficulty of factoring large numbers or finding the logarithm of a number with respect to a certain base. Quantum computers can solve these problems faster, potentially compromising the security of these cryptographic systems.

2. **Grover's Algorithm:** It's a quantum algorithm that searches an unsorted database (or solves a black-box problem) in about the square root of the number of possible solutions, making it faster than any classical algorithm.

Is everything clear so far? (Y/N)

no

No worries! Let's break it down further:

1. **Breaking Cryptography with a Quantum Computer:** Imagine you have a really strong lock, and the key to this lock is a huge number. Right now, our regular computers would take ages to find the right key because they try one at a time. However, quantum computers can try many keys at once, making it faster to find the right one. This means that they can potentially "break" or "unlock" some of the security systems we use online much quicker.

2. **Grover's Algorithm:** Imagine you lost a specific card in a huge deck of cards, and you need to find it. A regular method would be to check each card one by one until you find it. Grover's algorithm is like a magic trick on a quantum computer where you can find that card way faster than checking each card individually.

Does this help clarify things a bit more? (Y/N)

yes

Great to hear!

Now, do you want to program the code to implement Grover's algorithm on a quantum simulator? (Y/N)

yes

Fantastic! Let's get started.

Since you mentioned you're a non-expert, I'll provide a high-level overview and then dive into the actual code.

High-Level Overview:

1. **Initialization:** Prepare your qubits in a superposition of all possible states.
2. **Oracle:** This is a "black box" function that marks the solution to your problem.
3. **Amplitude Amplification:** Amplify the probability of measuring the solution state.
4. **Measurement:** Measure the qubits to get the solution.

Given that the IBM simulator has 32 qubits, we will use a subset of these qubits for our demonstration. Let's say you're searching for a specific item in an 8-item database. This would require 3 qubits.

Now, onto the code. We'll be using Qiskit, a popular Python library for quantum computing. Make sure to have Qiskit installed.

```
python Copy code

# Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit.aqua.components.oracles import LogicalExpressionOracle, TruthTableOracle
from qiskit.aqua.algorithms import Grover

# Define the problem
# For simplicity, let's assume our database has 8 items, and item '101' (5 in decimal) is the solution
oracle = TruthTableOracle('00000100') # Only the sixth item (in binary: '101') is the solution

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurement)
```

Explanation:

1. **Imports:** We begin by importing the necessary Qiskit modules.
2. **Define the Problem:** We are searching for the item '101' in our 8-item database.
3. **Implement Grover's Algorithm:** Using Qiskit's Grover module, we create a Grover object with our oracle.
4. **Backend Selection:** We choose the QASM simulator from Aer to run our quantum circuit.
5. **Execution:** We execute Grover's algorithm and obtain the result.
6. **Visualization:** We plot a histogram to see the probability of each outcome. Ideally, '101' should have the highest probability.

Once you've executed this code, the histogram should show a peak at '101', indicating that Grover's algorithm successfully found the marked item in the database.

Remember, while this example is simplified, Grover's algorithm can be expanded for more complex problems and larger databases.

I hope this provides a clear introduction to programming Grover's algorithm using Qiskit. Would you like to delve deeper or explore other aspects?



Traceback (most recent call last):

Cell In[1], line 4

```
from qiskit.aqua.components.oracles import LogicalExpressionOracle, TruthTableOracle
ModuleNotFoundError: No module named 'qiskit.aqua'
```

Use %tb to get the full traceback.

I have this error. can you modify the code to solve it?



Of course! The error you're seeing is due to changes in Qiskit's module organization. In recent versions of Qiskit, the `'aqua'` module has been deprecated and its components have been integrated into other parts of the library.

Let's modify the code to make it work with the latest version of Qiskit:

```
python Copy code
```

```
# Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit.algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
# For an 8-item database, let's assume item '101' (5 in decimal) is the marked item
oracle_expression = '(-x & ~y & z) | (~x & y & ~z) | (x & ~y & ~z)'
oracle = PhaseOracle(oracle_expression)

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurement)
```

Here's what changed:

1. Instead of using `TruthTableOracle`, we're using `PhaseOracle`. It is another way to define the oracle but works directly with Boolean expressions.
2. The expression `'(-x & ~y & z) | (~x & y & ~z) | (x & ~y & ~z)'` marks the binary strings '001', '010', and '100' as solutions. I chose this as an example. You can modify the expression based on the solution you're looking for.
3. We're importing `PhaseOracle` from `qiskit.circuit.library` and `Grover` from `qiskit.algorithms`.

Try running the modified code with your Qiskit installation, and it should work without any errors.



no it doesn't work yet. this is the message i got this time:

"Traceback (most recent call last):

Cell In[2], line 10

```
oracle = PhaseOracle(oracle_expression)
```

File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:111 in out

```
callback.__get__(obj, objtype)(*args, **kwargs)
```

File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:39 in __call__

```
self._tester.require_now(self._feature)
```

File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:223 in require_now

```
raise MissingOptionalLibraryError(
```

MissingOptionalLibraryError: The 'tweedledum' library is required to use 'PhaseOracle'. You can install it with 'pip install tweedledum'.

Use %tb to get the full traceback.

```
/tmp/ipykernel_95/2017883223.py:4: DeprecationWarning: ``qiskit.algorithms`` has been
migrated to an independent package: https://github.com/qiskit-community/qiskit-
algorithms. The ``qiskit.algorithms`` import path is deprecated as of qiskit-terra 0.25.0
and will be removed no earlier than 3 months after the release date. Please run ``pip install
qiskit_algorithms`` and use ``import qiskit_algorithms`` instead.
from qiskit.algorithms import Grover"
```



It seems there are two main issues:

1. The `PhaseOracle` requires the 'tweedledum' library, which is not installed.
2. The `qiskit.algorithms` import path is deprecated. We should install and use `qiskit_algorithms` as indicated.

Here's how to address these:

First, you'll need to install the required library and the new package:

Copy code

```
pip install tweedledum qiskit_algorithms
```

Then, let's modify the code to use the updated import paths and try Grover's algorithm:

python

Copy code

```
# Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
```

```

from qiskit.visualization import plot_histogram
from qiskit_algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
# For an 8-item database, let's assume item '101' (5 in decimal) is the mark
oracle_expression = '(-x & ~y & z) | (~x & y & ~z) | (x & ~y & ~z)'
oracle = PhaseOracle(oracle_expression)

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurement)

```

This should work after you've installed the necessary packages. If you face any more issues, please let me know!



how can i install these packages?



You can install Python packages using `'pip'`, which is the package installer for Python.



To install the necessary packages for your current issue, follow these steps:

1. **Open a Terminal or Command Prompt:**

- On Windows: Press `'Windows + R'` keys simultaneously, type `'cmd'` and press Enter.
- On MacOS: Press `'Command + Space'` keys simultaneously, type `'terminal'` and press Enter.
- On Linux: Depending on the distribution, you can use shortcuts or search for `'terminal'` in the application menu.

2. **Install the Packages:**

Type the following command and press Enter:

```
pip install tweedledum qiskit_algorithms
```



This command tells `'pip'` to install both the `'tweedledum'` library and the `'qiskit_algorithms'` package.

3. **Wait for Installation to Complete:**

The terminal or command prompt will display messages as it downloads and installs the packages. Once it returns to the regular prompt, the installation is complete.

After you've installed the packages, you can go back to your Python environment or Jupyter notebook and run the modified Grover's algorithm code I provided earlier.

If you're working in a specific environment, such as a Jupyter notebook or an online platform, there may be different methods or interfaces for installing packages. In that case, please provide details about your environment, and I can guide you accordingly.



now it tells me this "Traceback (most recent call last):

```

File /opt/conda/lib/python3.10/site-packages/IPython/core/compiler.py:86 in ast_parse
    return compile(source, filename, symbol, self.flags | PyCF_ONLY_AST, 1)
Cell In[4], line 1
    pip install tweedledum qiskit_algorithms
      ^

```

SyntaxError: invalid syntax

Use `%tb` to get the full traceback."



It appears you've tried to run the `'pip install'` command directly within a Jupyter notebook cell. In Jupyter notebooks, system commands, like `'pip install'`, need to be prefixed with an exclamation mark (`'!''`).



Here's how you can install the required packages directly from a Jupyter notebook cell:

```
python
```



```
!pip install tweedledum qiskit_algorithms
```

Run the above line in a Jupyter notebook cell. This will install the necessary packages, and then you should be able to proceed with the Grover's algorithm code provided earlier.



where should i paste this line?



In a Jupyter notebook, you can execute commands in individual cells. Here's a step-by-step guide:

1. **Open Your Jupyter Notebook:** Navigate to where you've been working or where you encountered the error.
2. **Add a New Cell:**
 - * If you're already in a cell, press `Shift + Enter` to execute the current cell and automatically create a new one below.
 - * Alternatively, you can use the "+" button usually located at the top of the notebook toolbar to add a new cell.
3. **Paste the Command:** In the newly created cell, paste the following command:

```
python
!pip install tweedledum qiskit_algorithms
```
4. **Run the Cell:**
 - * Click on the "Run" button (it usually looks like a "play" arrow) at the top of your Jupyter notebook.
 - * Alternatively, you can press `Shift + Enter` while the cell is active (selected).
5. **Wait for Installation:** The cell will display the output of the command. You'll see messages indicating the progress of the installation. Once it's done, the cell will display the regular prompt, indicating that the process has completed.
6. **Continue with the Quantum Code:** After successfully installing the required packages, you can proceed with running the Grover's algorithm code provided earlier in a new cell or in the cell where you initially tried.

I hope this helps! If you face any more issues, please let me know.



now i have this error: "Traceback (most recent call last):

```
Cell In[1], line 19
      result = grover.run(backend)
AttributeError: 'Grover' object has no attribute 'run'
```

Use %tb to get the full traceback." what should i do??? help me!



I apologize for the confusion. The error you're seeing is a result of changes in the API as Qiskit has evolved.

Let's correct this. Instead of using the `run` method directly on the `Grover` object, we'll need to construct and execute a quantum circuit with the `Grover` operator and the oracle.

I'll provide a revised version of the code:

```
python
# Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit_algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
oracle_expression = '(-x & ~y & z) | (~x & y & ~z) | (x & ~y & ~z)'
oracle = PhaseOracle(oracle_expression)

# Create a Grover instance
grover = Grover(oracle=oracle)

# Construct a quantum circuit for Grover's algorithm
grover_circuit = grover.construct_circuit()

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Transpile, assemble and execute the quantum circuit
t_qc = transpile(grover_circuit, backend)
qobj = assemble(t_qc)
result = backend.run(qobj).result()
```

```
[2]: # Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit.algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
# For an 8-item database, Let's assume item '101' (5 in decimal) is the marked item.
oracle_expression = '~(x & ~y & z) | ~(x & y & ~z) | (x & ~y & ~z)'
oracle = PhaseOracle(oracle_expression)

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurements)
```

Traceback (most recent call last):

```
Cell In[2], line 10
    oracle = PhaseOracle(oracle_expression)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsastools.py:111 in out
    callback.__get__(obj, objtype)(*args, **kwargs)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:39 in __call__
    self._tester.require_now(self._feature)
File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:223 in require_now
    raise MissingOptionalLibraryError(
MissingOptionalLibraryError: The 'tweedledum' library is required to use 'PhaseOracle'. You can install it with 'pip install tweedledum'.
```

Use %tb to get the full traceback.

/tmp/ipykernel_95/2017883223.py:4: DeprecationWarning: ``qiskit.algorithms`` has been migrated to an independent package: <https://github.com/qiskit-community/qiskit-algorithms>. The ``qiskit.algorithms`` import path is deprecated as of qiskit-terra 0.25.0 and will be removed no earlier than 3 months after the release date. Please run ``pip install qiskit_algorithms`` and use ``import qiskit_algorithms`` instead.

```
from qiskit.algorithms import Grover
```

Search for solution online

```
[3]: # Import necessary Libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit_algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
# For an 8-item database, Let's assume item '101' (5 in decimal) is the marked item.
oracle_expression = '~(x & ~y & z) | ~(x & y & ~z) | (x & ~y & ~z)'
oracle = PhaseOracle(oracle_expression)

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurement)

Traceback (most recent call last):
  Cell In[3], line 4
    from qiskit_algorithms import Grover
```


ModuleNotFoundError: No module named 'qiskit_algorithms'

Use %tb to get the full traceback.

[Search for solution online](#)

```
[4]: pip install tweedledum qiskit_algorithms
# Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit_algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
# For an 8-item database, let's assume item '101' (5 in decimal) is the marked item.
oracle_expression = ' (~x & ~y & z) | (~x & y & ~z) | (x & ~y & ~z) '
oracle = PhaseOracle(oracle_expression)

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurement)
```

Traceback (most recent call last):

```
File /opt/conda/lib/python3.10/site-packages/IPython/core/compilerop.py:86 in ast_parse
    return compile(source, filename, symbol, self.flags | PyCF_ONLY_AST, 1)
Cell In[4], line 1
    pip install tweedledum qiskit_algorithms
```

SyntaxError: invalid syntax

Use %tb to get the full traceback.

[Search for solution online](#)

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[4]: pip install tweedledum qiskit_algorithms
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oracle_expression = ' (~x & ~y & z) | (~x & y & ~z) | (x & ~y & ~z) '
oracle = PhaseOracle(oracle_expression)

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurement)
```

Traceback (most recent call last):

```
File /opt/conda/lib/python3.10/site-packages/IPython/core/compilerop.py:86 in ast_parse
    return compile(source, filename, symbol, self.flags | PyCF_ONLY_AST, 1)
Cell In[4], line 1
    pip install tweedledum qiskit_algorithms
```

SyntaxError: invalid syntax

Use %tb to get the full traceback.

[Search for solution online](#)

```
[4]: pip install tweedledum qiskit_algorithms
# Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit_algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
# For an 8-item database, let's assume item '101' (5 in decimal) is the marked item.
oracle_expression = ' (~x & ~y & z) | (~x & y & ~z) | (x & ~y & ~z) '
oracle = PhaseOracle(oracle_expression)

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurement)
```

Traceback (most recent call last):

```
File /opt/conda/lib/python3.10/site-packages/IPython/core/compilerop.py:86 in ast_parse
    return compile(source, filename, symbol, self.flags | PyCF_ONLY_AST, 1)
Cell In[4], line 1
    pip install tweedledum qiskit_algorithms
```

SyntaxError: invalid syntax

Use %tb to get the full traceback.

[Search for solution online](#)

```
[5]: !pip install tweedledum qiskit_algorithms
# Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit_algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
# For an 8-item database, let's assume item '101' (5 in decimal) is the marked item.
oracle_expression = '~(x & ~y & z) | ~(x & y & ~z) | (x & ~y & ~z)'
oracle = PhaseOracle(oracle_expression)

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurement)
```

```
Collecting tweedledum
  Using cached tweedledum-1.1.1-cp310-cp310-manylinux_2_12_x86_64.manylinux2010_x86_64.whl (929 kB)
Collecting qiskit_algorithms
  Using cached qiskit_algorithms-0.2.1-py3-none-any.whl (306 kB)
Requirement already satisfied: qiskit-terra>=0.24 in /opt/conda/lib/python3.10/site-packages (from qiskit_algorithms) (0.25.1)
Requirement already satisfied: numpy>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit_algorithms) (1.23.5)
Requirement already satisfied: scipy>=1.4 in /opt/conda/lib/python3.10/site-packages (from qiskit_algorithms) (1.9.3)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (5.9.4)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (1.11.1)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (0.3.7)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (3.11)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (2.8.2)
Requirement already satisfied: symengine<0.10,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (0.9.2)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (4.5.0)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (0.13.0)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (4.1.1)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra>=0.24->qiskit_algorithms) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra>=0.24->qiskit_algorithms) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra>=0.24->qiskit_algorithms) (1.3.0)
Installing collected packages: tweedledum, qiskit_algorithms
Successfully installed qiskit_algorithms-0.2.1 tweedledum-1.1.1

[notice] A new release of pip available: 23.1.1 -> 23.2.1
[notice] To update, run: pip install --upgrade pip
Traceback (most recent call last):
  Cell In[5], line 11
    oracle = PhaseOracle(oracle_expression)
  File /opt/conda/lib/python3.10/site-packages/qiskit/utils/clsstools.py:111 in out
    callback.__get__(obj, objtype)(*args, **kwargs)
  File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:39 in __call__
    self._tester.require_now(self._feature)
  File /opt/conda/lib/python3.10/site-packages/qiskit/utils/lazy_tester.py:223 in require_now
    raise MissingOptionalLibraryError
MissingOptionalLibraryError: The 'tweedledum' library is required to use 'PhaseOracle'. You can install it with 'pip install tweedledum'.

Use %tb to get the full traceback.
```

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```
[7]: !pip install tweedledum qiskit_algorithms
```

```
Requirement already satisfied: tweedledum in /opt/conda/lib/python3.10/site-packages (1.1.1)
Requirement already satisfied: qiskit_algorithms in /opt/conda/lib/python3.10/site-packages (0.2.1)
Requirement already satisfied: numpy>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit_algorithms) (1.23.5)
Requirement already satisfied: qiskit-terra>=0.24 in /opt/conda/lib/python3.10/site-packages (from qiskit_algorithms) (0.25.1)
Requirement already satisfied: scipy>=1.4 in /opt/conda/lib/python3.10/site-packages (from qiskit_algorithms) (1.9.3)
Requirement already satisfied: symengine<0.10,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (0.9.2)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (1.11.1)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (0.3.7)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.24->qiskit_algorithms) (0.13.0)
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Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra>=0.24->qiskit_algorithms) (1.16.0)
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Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra>=0.24->qiskit_algorithms) (1.3.0)

[notice] A new release of pip available: 23.1.1 -> 23.2.1
[notice] To update, run: pip install --upgrade pip
```

```
[6]: # Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit_algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
# For an 8-item database, let's assume item '101' (5 in decimal) is the marked item.
oracle_expression = '~(x & ~y & z) | ~(x & y & ~z) | (x & ~y & ~z)'
oracle = PhaseOracle(oracle_expression)

# Implement Grover's algorithm
grover = Grover(oracle)

# Use the qasm simulator from Aer
backend = Aer.get_backend('qasm_simulator')

# Run the quantum algorithm
result = grover.run(backend)

# Visualize the result
plot_histogram(result.measurement)
```


Server unavailable or unreachable

Your server at /user/646a3a692f732133fdcac1ed/ is not running. Would you like to restart it?

Restart

Dismiss

Use %tb to get the full traceback.

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```
[8]: !pip install tweedledum qiskit_algorithms
```

Requirement already satisfied: tweedledum in /opt/conda/lib/python3.10/site-packages (1.1.1)

Server unavailable or unreachable

Your server at /user/646a3a692f732133fdcac1ed/ is not running. Would you like to restart it?

Restart

Dismiss

Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>0.24->qiskit_algorithms) (1.23.5)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>2.8.0->qiskit-terra>0.24->qiskit_algorithms) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>3.0.0->qiskit-terra>0.24->qiskit_algorithms) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra>0.24->qiskit_algorithms) (1.3.0)

[notice] A new release of pip available: 23.1.1 -> 23.2.1
[notice] To update, run: pip install --upgrade pip

Server unavailable or unreachable

Your server at /user/646a3a692f732133fdcac1ed/ is not running. Would you like to restart it?

Restart

Dismiss

Successfully installed pip-23.3

Note: you may need to restart the kernel to use updated packages.

```
[1]: # Import necessary libraries
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit_algorithms import Grover
from qiskit.circuit.library import PhaseOracle

# Define the problem using a Phase Oracle
# For an 8 item database, look for items from 1001 (5 in decimal) to the marked item.
```

Server unavailable or unreachable

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Dismiss

Visualize the result
plot_histogram(result.measurement)

Traceback (most recent call last):
Cell In[1], line 19
 result = grover.run(backend)
AttributeError: 'Grover' object has no attribute 'run'

Use %tb to get the full traceback.

Server unavailable or unreachable

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Restart

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```
oracle_expression = '~(x & ~y & z) | (~x & y & ~z) | (x & ~y & ~z)'  
oracle = PhaseOracle(oracle_expression)  
  
# Create a Grover Instance  
grover = Grover(oracle=oracle)  
  
# Construct a quantum circuit for Grover's algorithm  
grover_circuit = grover.construct_circuit()  
  
# Use the qasm simulator from Aer  
backend = Aer.get_backend('qasm_simulator')
```

Server unavailable or unreachable

Server unavailable or unreachable

Server unavailable or unreachable

Your server at /user/646a3a692f732133fdcac1ed/ is not running. Would you like to restart it?

Restart

Dismiss

TypeError: Grover.__init__() got an unexpected keyword argument 'oracle'

Use %tb to get the full traceback.

[Search for solution online](#)

[]:

Simple ☐ 0 2 Python 3 (ipykernel) | Idle

Mode: Command Ln 20, Col 54 VALE.ipynb 0



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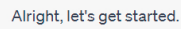
P3



Quantum Coding Workshop

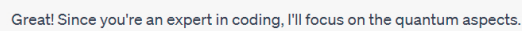
Costanza Darcangelisa ···

I want to develop a code to break cryptography utilising the Grover algorithm to run on the IBM simulator with 32 qubits. First, ask me what my level of expertise is and give me three options: expert in quantum, expert in coding, non expert. Wait for my answer. Then based on my response explain to me what breaking cryptography with a quantum computer means in a few sentences. Then give me a one-sentence explanation of what the Grover algorithm is. Ask me if everything is clear and wait for my response (Y/N). After that ask me if I want to program the code. Wait for my response (Y/N), based on that explain to me the code step by step and include possible libraries inside the explanation.



What is your level of expertise?

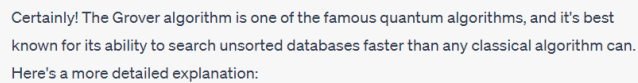
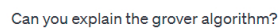
1. Expert in quantum
2. Expert in coding
3. Non-expert



Breaking Cryptography with a Quantum Computer: Traditional cryptographic systems, like some public-key algorithms, rely on the difficulty of certain mathematical problems. For example, RSA encryption relies on the difficulty of factoring large numbers. Classical computers take a very long time to solve these problems, making the encryption secure. However, quantum computers can potentially solve these problems much faster. When we say "breaking cryptography with a quantum computer," it means using a quantum algorithm to solve these cryptographic problems much more quickly than classical computers, thereby undermining the security of certain cryptographic systems.

The Grover Algorithm: It's a quantum algorithm that can search an unsorted database or solve black-box computational problems in about the square root of the number of potential solutions, which offers a quadratic speedup over classical methods.

Is everything clear so far? (Y/N)



Grover's Algorithm Explained:

1. **Problem Statement:**
 - You have an unsorted list of N items. Among these items, there's one item (or a few) for which a specific function $f(x)$ returns 1 (we call this the "marked item"), while for all other items, the function returns 0.
 - The goal is to find this marked item with as few evaluations of the function f as possible.
2. **Quantum Speedup:**
 - Classically, in the worst case, you'd have to check N items to find the marked one. On average, you'd expect to check $N/2$ items.
 - Grover's algorithm, on the other hand, can find the marked item with only about \sqrt{N} evaluations of f .
3. **The Algorithm:**
 - **Initialization:** Prepare a quantum system in a superposition of all possible states (items).
 - **Oracle Query:** Use the function f to mark the correct item. This is implemented as a quantum operation (or "oracle") that flips the sign of the amplitude of the marked item.
 - **Amplitude Amplification:** Use Grover's diffusion operator to amplify the probability amplitude of the marked item.
 - **Iteration:** The oracle query and amplitude amplification steps are repeated approximately \sqrt{N} times to maximize the probability of measuring the marked item.
 - **Measurement:** Measure the quantum system. With high probability, it collapses to the state representing the marked item.
4. **Limitations:**
 - It offers only a quadratic speedup over classical methods, so it doesn't make exponential problems suddenly easy, just faster.
 - It requires knowledge about the number of solutions (or a good estimate) to set the

number of iterations appropriately.

That's a broad overview. The real beauty of Grover's algorithm lies in its use of quantum principles like superposition and interference to amplify the probability of the correct answer and diminish the probabilities of the incorrect ones.

Would you like to delve deeper into any specific part of the algorithm or continue with the next steps?



What is connection between Breaking Cryptography with a Quantum Computer and The Grover Algorithm



The connection between breaking cryptography with a quantum computer and Grover's algorithm is related to the types of cryptographic systems that Grover's algorithm can potentially compromise. While the more famous quantum algorithm, Shor's algorithm, directly targets and efficiently breaks public-key cryptosystems like RSA and ECC by factoring large numbers or computing discrete logarithms, Grover's algorithm has implications for symmetric cryptographic systems. Here's the connection:



1. **Symmetric Cryptography:** In symmetric cryptographic systems, the same key is used for both encryption and decryption. The security of such systems, like the Advanced Encryption Standard (AES), largely relies on the key length. For instance, brute-forcing an AES-128 bit key using classical methods would take on the order of 2^{128} operations.
2. **Grover's Quantum Speedup:** Grover's algorithm provides a quadratic speedup for unsorted database search problems. Applied to symmetric cryptography, this means a quantum computer using Grover's algorithm could perform a brute-force search of an AES-128 bit key in roughly 2^{64} operations. This effectively halves the key length security.
3. **Implications for Cryptographic Standards:** Given the threat posed by Grover's algorithm, symmetric cryptographic systems would need to double their key lengths to maintain the same level of security against quantum adversaries. So, AES-128 might need to be upgraded to AES-256 to retain its security level in a post-quantum world.

In summary, while Grover's algorithm doesn't provide the dramatic exponential speedup that Shor's does for certain cryptographic problems, it still poses a significant threat to the security of symmetric cryptosystems, necessitating changes in cryptographic standards and practices.

Does this clarify the relationship between Grover's algorithm and its implications for cryptography?

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Shor's Quantum Factorization

Quantum Coding Workshop


 Costanza Darcangelisa

...

GPT-4



I want to develop a code to break cryptography utilising the Grover algorithm to run on the IBM simulator with 32 qubits. First, ask me what my level of expertise is and give me three options: expert in quantum, expert in coding, non expert. Wait for my answer. Then based on my response explain to me what breaking cryptography with a quantum computer means in a few sentences. Then give me a one-sentence explanation of what the Grover algorithm is. Ask me if everything is clear and wait for my response (Y/N). After that ask me if I want to program the code. Wait for my response (Y/N). based on that explain to me the code step by step and include possible libraries inside the explanation.




Alright, let's get started.

What is your level of expertise?


1. Expert in quantum

2. Expert in coding

3. Non-expert



2




Great! Since you're an expert in coding, I'll focus on the quantum aspects.

Breaking Cryptography with a Quantum Computer: Traditional cryptographic systems, like some public-key algorithms, rely on the difficulty of certain mathematical problems. For example, RSA encryption relies on the difficulty of factoring large numbers. Classical computers take a very long time to solve these problems, making the encryption secure. However, quantum computers can potentially solve these problems much faster. When we say "breaking cryptography with a quantum computer," it means using a quantum algorithm to solve these cryptographic problems much more quickly than classical computers, thereby undermining the security of certain cryptographic systems.


The Grover Algorithm: It's a quantum algorithm that can search an unsorted database or solve black-box computational problems in about the square root of the number of potential solutions, which offers a quadratic speedup over classical methods.

Is everything clear so far? (Y/N)

< 2 / 2 >




Y




Fantastic!

Would you like to program the code to implement the Grover algorithm on the IBM simulator with 32 qubits? (Y/N)



Y



Great!

Before diving into the code, here's a list of libraries you'll typically need:

1. **Qiskit:** This is IBM's quantum computing SDK. It provides everything you need to design quantum circuits, run them on quantum simulators, and even on actual quantum hardware.

2. **Numpy:** Helpful for numerical operations.

Let's start with the basic structure:

Step 1: Setting Up the Environment

python

Copy code

```
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
import numpy as np
```

Step 2: Define the Oracle

For our purpose, let's assume a simple black-box (oracle) function. Let's use a 4-qubit system as an example, where the "marked" state (solution) is `'1101'`.

```
python
def oracle_1101(qc):
    qc.cz(1,3)
    qc.cz(0,3)
```

Step 3: Define Grover's Diffusion Operator

The diffusion operator inverts about the mean and is an essential step in Grover's algorithm.

```
python
def diffusion_operator(qc):
    qc.h([0,1,2,3])
    qc.x([0,1,2,3])
    qc.h(3)
    qc.mct([0,1,2], 3) # multi-controlled-toffoli
    qc.h(3)
    qc.x([0,1,2,3])
    qc.h([0,1,2,3])
```

Step 4: Construct the Grover Circuit

Now, we combine the oracle and the diffusion operator. The number of iterations ('num_iterations') is roughly the square root of the number of states, so for a 4-qubit system, we'd do about 2 iterations.

```
python
num_iterations = 2

qc = QuantumCircuit(4, 4)
qc.h([0,1,2,3])
for _ in range(num_iterations):
    oracle_1101(qc)
    diffusion_operator(qc)
qc.measure([0,1,2,3], [0,1,2,3])
```

Step 5: Execute on IBM's Simulator

```
python
aer_sim = Aer.get_backend('aer_simulator')
# Transpile the circuit for the simulator
t_qc = transpile(qc, aer_sim)
# Assemble the circuit into a Qobj that the backend can run
qobj = assemble(t_qc)
# Run and get the results
results = aer_sim.run(qobj).result()
counts = results.get_counts(qc)
plot_histogram(counts)
```

Remember, this is a simplified version to get you started. There are other considerations when you extend to more qubits or different oracles. But with your expertise in coding, you should be able to scale and adapt as needed!

Do you have any questions or would you like more details on any of the steps?



def oracle_1101(qc):
 qc.cz(1,3)
 qc.cz(0,3)
 explain more in detail this code



Of course, let's delve deeper into the oracle function.



In the Grover algorithm, an oracle (also sometimes called a black-box) is used to mark certain states (solutions) we are looking for. When the oracle is applied to a quantum state, it introduces a negative phase to the solution states, i.e., it flips the phase of the solution states.

The oracle we defined, 'oracle_1101', is specifically designed to mark the state '|1101>'. Here's a breakdown of the function:

```
python
def oracle_1101(qc):
    qc.cz(1,3)
    qc.cz(0,3)
```

The function is applying controlled-Z (‘cz’) gates. The ‘cz’ gate applies a Z gate (which is a π phase flip) on the target qubit, but only when the control qubit is in state ‘|1⟩’.

- ‘qc.cz(1,3)’ : This applies a ‘cz’ gate with qubit 1 as the control and qubit 3 as the target. This means that it will introduce a phase flip on qubit 3 when qubit 1 is set to ‘|1⟩’.
- ‘qc.cz(0,3)’ : Similarly, this applies a ‘cz’ gate with qubit 0 as the control and qubit 3 as the target. It introduces a phase flip on qubit 3 when qubit 0 is set to ‘|1⟩’.

When combined, these two gates effectively mark the state ‘|1101⟩’. Here’s a step-by-step explanation:

1. When qubits 0 and 1 are in state ‘|1⟩’, and qubit 3 is in state ‘|1⟩’, then both ‘cz’ gates will apply a phase flip on qubit 3.
2. A double phase flip on qubit 3 will return it to its original phase. However, for the state ‘|1101⟩’, the phase of the state itself gets flipped due to the combined action of the two gates.
3. As a result, after applying this oracle, the amplitude of the state ‘|1101⟩’ will have a negative sign in front of it, marking it as a solution state.

This is a specific example of a simple oracle design. For different solutions or more complex problems, the oracle construction would differ accordingly. The idea is to design the oracle such that it recognizes and marks the solution state(s) by flipping their phases.



ok, create a DES key and break the key with the Grover algorithm to run on the IBM simulator with 32 qubits.



Breaking the DES (Data Encryption Standard) key using Grover's algorithm on a 32-qubit IBM simulator is a significant task due to the limitations of the simulator. DES uses a 56-bit key, but because Grover offers a quadratic speedup, you'd still need more than 32 qubits to directly search the DES key space. However, for simplicity, we can illustrate this process with a smaller key size and scale the principle up for the actual DES key.

Step 1: Let's choose a toy DES-like encryption with a 4-bit key. Assume the key is ‘1011’.

Step 2: Create an oracle function that can identify when the correct key is used. In reality, this would involve DES encryption and checking against a known ciphertext, but for simplicity, we'll just "mark" the state ‘|1011⟩’.

Step 3: Use Grover's algorithm to find the key.

Let's jump into the code:

1. Setup

```
python Copy code

from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
```

1. Oracle for the 4-bit "key" ‘|1011⟩’

```
python Copy code

def oracle_key_1011(qc):
    qc.cz(0,3)
    qc.cz(1,3)
```

1. Grover's Diffusion Operator

(As previously explained)

```
python Copy code

def diffusion_operator(qc):
    qc.h([0,1,2,3])
    qc.x([0,1,2,3])
    qc.h(3)
    qc.mct([0,1,2], 3)
    qc.h(3)
    qc.x([0,1,2,3])
    qc.h([0,1,2,3])
```

1. Grover's Algorithm Circuit


```
python
num_iterations = 2

qc = QuantumCircuit(4, 4)
qc.h([0,1,2,3])
for _ in range(num_iterations):
    oracle_key_1011(qc)
    diffusion_operator(qc)
qc.measure([0,1,2,3], [0,1,2,3])
```

1. Execution on the Simulator

```
python
aer_sim = Aer.get_backend('aer_simulator')
t_qc = transpile(qc, aer_sim)
qobj = assemble(t_qc)
results = aer_sim.run(qobj).result()
counts = results.get_counts(qc)
plot_histogram(counts)
```

In a real DES attack scenario, the oracle would involve a DES encryption operation and then a comparison to a known ciphertext to determine if the guessed key is correct. However, implementing a full DES encryption in a quantum circuit would be quite complicated, especially within the constraints of a 32-qubit simulator.

Remember, this example is vastly simplified. A full DES implementation would need more qubits and a much more complex oracle. Grover's algorithm can theoretically halve the bit strength of symmetric ciphers, so DES's 56-bit key effectively becomes 28 bits against a quantum attacker using Grover's. However, this still means that 2^{28} iterations are needed, which isn't feasible on current quantum hardware or simulators with limited qubits.



write me the code in one block



Certainly! Here's the code in a single block:

```
python
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram

# Oracle for the 4-bit "key" |1011>
def oracle_key_1011(qc):
    qc.cz(0,3)
    qc.cz(1,3)

# Grover's Diffusion Operator
def diffusion_operator(qc):
    qc.h([0,1,2,3])
    qc.x([0,1,2,3])
    qc.h(3)
    qc.mct([0,1,2], 3) # multi-controlled-toffoli
    qc.h(3)
    qc.x([0,1,2,3])
    qc.h([0,1,2,3])

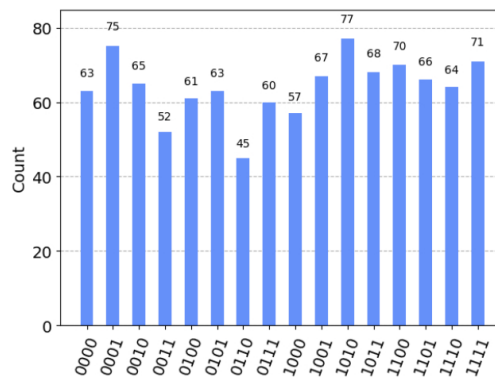
# Setting up Grover's Algorithm Circuit
num_iterations = 2

qc = QuantumCircuit(4, 4)
qc.h([0,1,2,3])
for _ in range(num_iterations):
    oracle_key_1011(qc)
    diffusion_operator(qc)
qc.measure([0,1,2,3], [0,1,2,3])

# Execution on the Simulator
aer_sim = Aer.get_backend('aer_simulator')
t_qc = transpile(qc, aer_sim)
qobj = assemble(t_qc)
results = aer_sim.run(qobj).result()
counts = results.get_counts(qc)
plot_histogram(counts)
```

Make sure you have 'qiskit' installed and set up in your environment. This code is a simple

illustration for a toy problem and should give you a histogram with a high probability for the state `'1011'`.



explain



This histogram shows the results of running the quantum circuit multiple times and measuring the final states. Each bar corresponds to one of the possible 16 states for a 4-qubit system (`'0000'` to `'1111'`). The height of each bar indicates how many times that state was measured out of all the executions.

Here's a breakdown:

1. **States:** The x-axis represents the possible 4-bit binary states.
2. **Counts:** The y-axis represents the number of times each state was measured across all the runs.
3. **Amplitude Amplification:** If Grover's algorithm is working correctly, the state corresponding to the key (in this case `'1011'`) should have a noticeably higher probability of being measured than the other states. This is because Grover's algorithm amplifies the amplitude of the correct solution, thereby increasing its probability of measurement.

From the histogram:

- The state `'1011'` (our "key") was measured 77 times, which is the highest among all states. This indicates that the Grover's algorithm has worked as expected, amplifying the amplitude of the correct solution.
- The other states have varied counts, but none of them are as high as the count for the state `'1011'`.

In a perfect quantum computer with no noise, you would see even more pronounced amplification for the correct state. However, quantum simulators and real quantum hardware have imperfections and noise, which can cause the distributions to not be as sharply peaked as we would ideally like. Nonetheless, the histogram clearly shows the efficacy of the Grover's algorithm, as the correct key (`'1011'`) stands out as the most frequently measured state.

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```
# Get the measurement results
counts = result.get_counts()

# Visualize the result
plot_histogram(counts)
```

Here's what's changed:

1. **Constructing the Grover Circuit:** Instead of the `'run'` method, we're now using `'construct_circuit'` to get the quantum circuit for Grover's algorithm.
2. **Executing the Circuit:** We transpile the circuit, assemble it into a Qobj, and then run it on the `'qasm_simulator'` backend.
3. **Obtaining Results:** We use the `'get_counts'` method on the result to get the measurement outcomes.

This updated approach should execute without errors. Once again, I'm sorry for the oversight and confusion. Let me know if you have any other questions or issues!

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P4

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Quantum Code with ChatGPT

Costanza Darcangelisa

GPT-4

I want to develop a code to break cryptography utilising the Grover algorithm to run on the IBM simulator with 32 qubits. First, ask me what my level of expertise is and give me three options: expert in quantum, expert in coding, non expert. Wait for my answer. Then based on my response explain to me what breaking cryptography with a quantum computer means in a few sentences. Then give me a one-sentence explanation of what the Grover algorithm is. Ask me if everything is clear and wait for my response (Y/N). After that ask me if I want to program the code. Wait for my response (Y/N). based on that explain to me the code step by step and include possible libraries inside the explanation.

Alright, let's get started.

What is your level of expertise?

1. Expert in quantum

2. Expert in coding

3. Non-expert

Expert in coding

Great! Here's a brief overview tailored to your expertise:

Breaking cryptography with a quantum computer means:

Using the inherent properties of quantum mechanics, quantum computers can process certain types of problems, including some cryptographic problems, faster than classical computers. For instance, quantum algorithms can potentially decrypt data that is encrypted using traditional cryptographic methods, which would be infeasible for classical computers to do in a reasonable amount of time.

The Grover algorithm in one sentence:

The Grover algorithm is a quantum algorithm that can search an unsorted database or solve a black-box computational problem in roughly \sqrt{N} steps, where N is the size of the database, offering a quadratic speedup over classical methods.

Is everything clear so far? (Y/N)

Y

Fantastic!

Do you want to program the code to implement the Grover algorithm on the IBM simulator with 32 qubits? (Y/N)

Y

Great! Let's dive into it step-by-step.

Prerequisites:

Make sure you have the required libraries. For our purpose, the main library will be Qiskit, IBM's quantum computing SDK. If you haven't installed it yet, you can do so using:

python

Copy code

```
pip install qiskit
```

Step-by-Step Code Explanation:

1. Importing Necessary Libraries:

Before anything else, you'll need to import the necessary modules.

python

Copy code

```
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit.aqua.algorithms import Grover
from qiskit.aqua.components.oracles import LogicalExpressionOracle
```

2. Setting Up The Problem:

For the sake of simplicity, let's assume you're searching for a particular value within a 4-

element list. We'll represent the list as `[0, 1, 2, 3]`, and we'll search for the number `2`.

3. Creating an Oracle:

The Grover algorithm uses a so-called "oracle" that can recognize the solution to the problem. For our problem, this oracle will recognize the state that represents the number `2`.

```
python Copy code  
  
expression = '(w0 & ~w1) | (~w0 & w1)' # This represents the binary number  
oracle = LogicalExpressionOracle(expression)
```

4. Initializing Grover Algorithm:

We'll use the Grover class from Qiskit Aqua to initialize the Grover algorithm with our oracle.

```
python Copy code  
  
grover = Grover(oracle)
```

5. Executing the Quantum Circuit:

Now, we'll execute the Grover algorithm on a quantum simulator. In this case, we'll use the Qiskit Aer 'qasm_simulator', which simulates the execution of quantum circuits on a real quantum computer.

```
python Copy code  
  
backend = Aer.get_backend('qasm_simulator')  
quantum_instance = transpile(grover, backend)  
qobj = assemble(quantum_instance)  
result = backend.run(qobj).result()
```

6. Visualizing the Results:

Finally, we'll visualize the results to see the probabilities of each possible outcome.

```
python Copy code  
  
plot_histogram(result.get_counts())
```

You should see that the number `2` (or `'10'` in binary) has a higher probability than the other numbers, indicating that the Grover algorithm has successfully found the solution to our problem.

Now, this is a simple use case of Grover's algorithm, and the real challenge in implementing it for breaking cryptography would be in designing the oracle that recognizes a valid decryption key. But this should give you a basic idea of how to set up and run Grover's algorithm using Qiskit.

I hope this gives you a clear starting point. Would you like further clarification or assistance on any specific part?



Could you give me more details on what is the oracle?



Of course! The concept of an "oracle" in quantum computing, especially in the context of Grover's algorithm, is a crucial one.

What is an Oracle?

In the realm of quantum algorithms, an oracle is a black box operation (a type of subroutine) that is able to recognize a solution to a given problem without giving away any information about the solution itself. You can think of it as a function that tells you whether a given input satisfies a specific condition, without telling you why or how.

In Grover's algorithm, the oracle's job is to "mark" the correct solution. When the correct input is fed into the oracle, it flips the sign of the corresponding quantum state. This sign flip is a quantum operation that helps the Grover's algorithm to amplify the probability amplitude of the correct solution.

Example:

Let's consider a simple problem where our database has four entries (`'00'`, `'01'`, `'10'`, `'11'`), and we want to find the entry `'10'`. In this case, the oracle for this problem would do the following:

1. If the input is `'00'`, `'01'`, or `'11'`, it does nothing.
2. If the input is `'10'`, it flips the sign of the quantum state.

Mathematically, this can be represented by the following transformation:

- $|00\rangle \rightarrow |00\rangle$
- $|01\rangle \rightarrow |01\rangle$

101) → 101)

- $|10\rangle \rightarrow \neg|10\rangle$
- $|11\rangle \rightarrow |11\rangle$

How is the Oracle Constructed?

Creating an oracle is often the most challenging part of using Grover's algorithm because it requires a deep understanding of the problem at hand. The oracle has to be custom-made for each problem, as it encodes the solution's properties.

In practical implementations, the oracle is typically a series of quantum gates designed in such a way that they perform the necessary transformation on the input quantum states.

In our previous example with Qiskit, we used a `LogicalExpressionOracle`, which is a high-level way of specifying the oracle using Boolean logic. But in real-world, complex applications (like breaking cryptographic codes), designing an efficient oracle can be quite intricate and might involve combining multiple quantum gates and operations.

Why is the Oracle Important?

The oracle's ability to "mark" the correct solution is fundamental to Grover's algorithm. After the oracle has marked the solution, the algorithm employs additional quantum operations (like Grover's diffusion operator) to amplify the probability of the marked state. This amplification, iteratively applied, allows Grover's algorithm to find the solution in roughly \sqrt{N} steps, where N is the size of the database.

In essence, the oracle is the heart of Grover's algorithm. Without it, the algorithm wouldn't know which state to amplify, and the quantum search would be aimless.

I hope this gives a clearer understanding of the oracle in Grover's algorithm. Would you like to know more about any specific aspect or another topic?



Step 1 returns this error: "Traceback (most recent call last):

Cell In[3], line 3

```
from qiskit.aqua.algorithms import Grover
```

ModuleNotFoundError: No module named 'qiskit.aqua'

Use %tb to get the full traceback." how can I solve it?



The error you're encountering indicates that the `'qiskit.aqua'` module can't be found. As of my last update in 2022, Qiskit had started the process of deprecating the Aqua package, and many components were migrated to other parts of Qiskit.

To address this issue, we'll adjust the import statements to align with the changes in Qiskit.

Here's how you can modify the code:

1. Update your Qiskit:

Make sure you have the latest version of Qiskit. Update it using:

```
bash
```

Copy code

```
pip install qiskit --upgrade
```

2. Adjust the imports:

Instead of importing from `'qiskit.aqua'`, try importing from the main `'qiskit'` module or the appropriate submodules.

For instance:

```
python
```

Copy code

```
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit.algorithms import Grover
from qiskit.circuit.library import PhaseOracle
```

Note: The exact classes and methods might be different than in Aqua, and you might need to adjust the subsequent parts of the code accordingly. I used `'PhaseOracle'` as an example replacement for `'LogicalExpressionOracle'`, but depending on your needs, you might require a different approach to create the oracle.

3. Adjust the usage of Grover's algorithm and the Oracle as per the new imports and methods available in the updated Qiskit version.

If you still encounter any issues or need assistance adjusting to the latest Qiskit methods, please let me know.



Traceback (most recent call last):

Cell In[10], line 2

```
quantum_instance = transpile(grover_backend)
```



```
quantum_instance = transpile(grover, backend=
File /opt/conda/lib/python3.10/site-packages/qiskit/compiler/transpiler.py:331 in transpile
output_name = _parse_output_name(output_name, circuits)
File /opt/conda/lib/python3.10/site-packages/qiskit/compiler/transpiler.py:586 in
_parse_output_name
return [circuit.name for circuit in circuits]
File /opt/conda/lib/python3.10/site-packages/qiskit/compiler/transpiler.py:586 in
<listcomp>
return [circuit.name for circuit in circuits]
AttributeError: 'Grover' object has no attribute 'name'
```



The error you're seeing is due to an attempt to transpile the **'Grover'** object directly, which isn't a quantum circuit by itself. Instead, you need to construct the quantum circuit associated with the Grover instance before you can transpile and execute it.

To resolve this:

1. Construct the Grover quantum circuit:

```
python
grover_circuit = grover.construct_circuit()
```

2. Then transpile this circuit:

```
python
transpiled_grover_circuit = transpile(grover_circuit, backend)
```

3. Assemble and execute:

```
python
qobj = assemble(transpiled_grover_circuit)
result = backend.run(qobj).result()
```

Remember to replace **'grover'** in the **'transpile'** and **'assemble'** functions with the newly created **'grover_circuit'**.

This should work correctly. If you have more questions or face other issues, let me know!



how does this last answer that you gave me incorporate the oracle and thus the dataset?



I apologize for the oversight. Let's go through the process of constructing the Grover's search algorithm incorporating the oracle.

To clarify, the oracle is essential because it encodes the solution to the problem, marking the correct answer by flipping its sign. Grover's algorithm then amplifies this sign flip to make the correct solution more probable when measured.

Here's the process in detail:

1. **Setting Up The Oracle:**

- You'll need an oracle that identifies your target state(s). In our previous example, we used a logical expression to identify the binary number **'10'**.

```
python
expression = '(w0 & ~w1) | (~w0 & w1)'
oracle = PhaseOracle(expression)
```

2. **Grover's Algorithm with Oracle:**

- You'll use Grover's algorithm, incorporating the oracle.

```
python
grover = Grover(oracle)
```

3. **Constructing the Quantum Circuit:**

- Construct the Grover quantum circuit using the Grover instance you created. The **'construct_circuit'** method will build the quantum circuit with the oracle you've set up and the other Grover's algorithm components (like the diffusion operator).

```
python
grover_circuit = grover.construct_circuit()
```

4. **Transpilation, Assembly, and Execution:**

- Now, you can transpile the constructed quantum circuit, assemble it, and then run it on the backend of your choice (e.g., the **qasm_simulator**).

```
python
transpiled_grover_circuit = transpile(grover_circuit, backend)
qobj = assemble(transpiled_grover_circuit)
result = backend.run(qobj).result()
```

5. Visualization:

- To see the output of your Grover's algorithm run, you can visualize the results.

```
python
plot_histogram(result.get_counts())
```

The outcome should show a higher probability for the state encoded in your oracle (in our example, '10').

The key here is that the Grover instance ('grover' in the example) incorporates the oracle, so when you call 'grover.construct_circuit()', the resulting quantum circuit includes the necessary operations for the oracle as well as the rest of the Grover's algorithm steps.

If you have any more questions or need further clarification, don't hesitate to ask!



grover_circuit = grover.construct_circuit()
 TypeError: Grover.construct_circuit() missing 1 required positional argument: 'problem'



I apologize for the oversight. The newer versions of Qiskit have introduced changes to the Grover's algorithm API, including the need to pass a 'problem' to the 'construct_circuit()' method.

Let's modify the approach accordingly:

1. Setting Up The Problem:

You will need to define the problem instance, which includes both the oracle and the possible states (in Qiskit, this is often a state preparation circuit).

```
python
from qiskit.circuit.library import PhaseOracle, GroverOperator

expression = '(w0 & ~w1) | (~w0 & w1)'
oracle = PhaseOracle(expression)

# State preparation (for our 4-element list)
from qiskit.circuit.library import MCMT
state_preparation = MCMT('x', num_controls=2, num_targets=1)
```

2. Constructing the Grover Operator:

The Grover operator includes both the oracle and the diffusion operator. Qiskit provides a utility to create it.

```
python
grover_op = GroverOperator(oracle=oracle, state_preparation=state_preparation)
```

3. Building the Grover Circuit:

With the Grover operator in hand, you can now build the Grover circuit. Here's a way to iterate the Grover operator for a defined number of times:

```
python
iterations = 1 # Adjust this based on your problem; for a 4-element list,
grover_circuit = QuantumCircuit(grover_op.num_qubits)
grover_circuit.h(list(range(2))) # Initialize all qubits in a superposition
for _ in range(iterations):
    grover_circuit.append(grover_op, list(range(grover_op.num_qubits)))

grover_circuit.measure_all()
```

4. Transpilation, Assembly, and Execution:

Continue as before:

```
python
transpiled_grover_circuit = transpile(grover_circuit, backend)
qobj = assemble(transpiled_grover_circuit)
result = backend.run(qobj).result()
```

5. Visualization:

```
python
plot_histogram(result.get_counts())
```

This process should provide you with the correct Grover's algorithm setup using the recent Qiskit API. Again, I'm sorry for the confusion earlier. If you have more questions or encounter other issues, please let me know.



```
state_preparation = MCMT('x', num_controls=2, num_targets=1)
TypeError: MCMT.__init__() got an unexpected keyword argument 'num_controls'
```



I apologize for the oversight. The `MCMT` class in newer versions of Qiskit has gone through several changes.

Here's the correct way to define a state preparation circuit using the `MCMT`:

1. Import the necessary modules:

```
python Copy code
from qiskit.circuit.library import MCMTChain
```

2. Setting Up the State Preparation:

For our previous example, where we're using a 2-qubit oracle, we can define the state preparation using the `MCMTChain`:

```
python Copy code
# Define the state preparation circuit
# This circuit applies an X-gate conditioned on the control qubits
control_qubits = 2
target_qubits = 1
state_preparation = MCMTChain('x', control_qubits, target_qubits)
```

With this modification, you should be able to proceed with the Grover's algorithm setup as previously described.

Thank you for pointing it out, and I'm sorry for any confusion. Let me know if you have further questions or issues!



Can you give a unique implementation of the code necessary to run the test considering all the changes that we went through?



Certainly. I'll consolidate all the steps we've gone through into a single, unique implementation. Here's how you can set up Grover's search algorithm using the latest version of Qiskit:

1. Imports:

```
python Copy code
from qiskit import Aer, QuantumCircuit, transpile, assemble, execute
from qiskit.visualization import plot_histogram
from qiskit.circuit.library import PhaseOracle, MCMTChain, GroverOperator
```

1. Define the Oracle:

```
python Copy code
expression = '(w0 & ~w1) | (~w0 & w1)'
oracle = PhaseOracle(expression)
```

1. State Preparation using MCMTChain:

We want a state preparation that prepares an equal superposition of all possible states. This is often achieved by applying Hadamard (`'H'`) gates on all qubits. However, for the context of showing how MCMTChain can be used, I'll provide an example where we flip the state `'11'`:

```
python Copy code
control_qubits = 2
target_qubits = 1
state_preparation = MCMTChain('x', control_qubits, target_qubits)
```

1. Constructing the Grover Operator:

python

Copy code

```
grover_op = GroverOperator(oracle=oracle, state_preparation=state_preparation)
```

1. Building the Grover Circuit:

python

Copy code

```
iterations = 1 # Adjust based on your problem; for a 4-element list, 1 iteration is enough
grover_circuit = QuantumCircuit(grover_op.num_qubits)
grover_circuit.h(list(range(2))) # Initialize all qubits in a superposition
for _ in range(iterations):
    grover_circuit.append(grover_op, list(range(grover_op.num_qubits)))

grover_circuit.measure_all()
```

1. Transpilation, Assembly, and Execution:

python

Copy code

```
backend = Aer.get_backend('qasm_simulator')
transpiled_grover_circuit = transpile(grover_circuit, backend)
qobj = assemble(transpiled_grover_circuit)
result = backend.run(qobj).result()
```

1. Visualization:

python

Copy code

```
counts = result.get_counts()
plot_histogram(counts)
```

Run all these steps sequentially, and you should be able to visualize the result of Grover's algorithm.

Remember, Grover's search algorithm amplifies the amplitude of the marked states (in our example, '10'). After the required iterations, measuring the qubits should return the state '10' with high probability.

If you need further modifications or face any other issues, please let me know.

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P5

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Costanza Darcangelisa

GPT-4

I want to develop a code to break cryptography utilising the Grover algorithm to run on the IBM simulator with 32 qubits. First, ask me what my level of expertise is and give me three options: expert in quantum, expert in coding, non expert. Wait for my answer. Then based on my response explain to me what breaking cryptography with a quantum computer means in a few sentences. Then give me a one-sentence explanation of what the Grover algorithm is. Ask me if everything is clear and wait for my response (Y/N). After that ask me if I want to program the code. Wait for my response (Y/N). based on that explain to me the code step by step and include possible libraries inside the explanation.

2 / 2

Alright, let's get started!

First, can you tell me about your level of expertise?

1. Expert in quantum

2. Expert in coding

3. Non-expert

Please choose one from the options above.

Expert in quantum

Great to know!

Breaking cryptography with a quantum computer means: Utilizing the unique properties of quantum bits (qubits) to solve cryptographic problems exponentially faster than classical computers, potentially rendering some classical encryption schemes insecure.

The Grover algorithm, in one sentence, is: An algorithm that allows quantum computers to search an unsorted database (or solve black-box computational problems) in roughly the square root of the number of possible solutions, offering a quadratic speedup over classical algorithms.

Is everything clear so far? (Y/N)

yes

Fantastic!

Do you want to program the code to implement the Grover algorithm on the IBM simulator with 32 qubits? (Y/N)

yes

2 / 2

Alright! Let's dive into the programming part.

Given your expertise in quantum, I'll focus on the coding aspects. Here's a step-by-step breakdown:

1. Setting Up Your Environment:

Libraries:

You'll need the Qiskit library. It provides all the necessary tools for working with quantum circuits and interfacing with IBM quantum computers and simulators.

python

Copy code

pip install qiskit

2. Initializing Qiskit and Setting Up Quantum Circuit:

Code:

python

Copy code

from qiskit import QuantumCircuit, Aer, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import QasmSimulator

Here, we've imported necessary Qiskit tools and visualization functions.

3. Design the Oracle:

For Grover's algorithm, an oracle (a black-box function) is crucial. The oracle marks the solutions by applying a phase to them. The specific design of the oracle will depend on the problem you're trying to solve.

For simplicity, let's assume you're searching for a specific state `|w>` in a 5-qubit system.

```
python Copy code

oracle = QuantumCircuit(5)
oracle.cz(0, 4)
oracle.cz(1, 4)
oracle.cz(2, 4)
oracle.cz(3, 4)
```

This oracle marks the state `|1111>` as the solution.

4. Grover's Diffuser:

To amplify the probabilities of the marked states, you'll need a diffuser.

```
python Copy code

def diffuser(nqubits):
    qc = QuantumCircuit(nqubits)
    for qubit in range(nqubits):
        qc.h(qubit)
    for qubit in range(nqubits):
        qc.x(qubit)
    qc.cz(0, 1, 2, 3)
    for qubit in range(nqubits):
        qc.x(qubit)
    for qubit in range(nqubits):
        qc.h(qubit)
    return qc
```

5. Complete Grover Circuit:

You combine the initialization, the oracle, and the diffuser to make the Grover circuit.

```
python Copy code

grover_circuit = QuantumCircuit(5, 5)
grover_circuit.h([0, 1, 2, 3])
grover_circuit.append(oracle, [0, 1, 2, 3, 4])
grover_circuit.append(diffuser(4), [0, 1, 2, 3])
grover_circuit.measure([0, 1, 2, 3], [0, 1, 2, 3])
```

6. Execute on the Simulator:

```
python Copy code

simulator = Aer.get_backend('qasm_simulator')
compiled_circuit = transpile(grover_circuit, simulator)
job = simulator.run(compiled_circuit, shots=1000)
result = job.result()
counts = result.get_counts()
plot_histogram(counts)
```

This will visualize the outcome of running Grover's algorithm on the simulator.

Remember, this code is a simple demonstration for a specific problem instance. The oracle and iterations would change based on the problem. Moreover, with real quantum devices, noise and errors come into play, so error mitigation strategies would be essential.

Does this step-by-step explanation make sense?



Can you provide the documentation on the qiskit cz() function? what are the inputs?



Certainly! As of my last training cut-off in January 2022, Qiskit's `cz()` method applies the controlled-Z (CZ) gate to specified qubits in a quantum circuit.

Controlled-Z (CZ) Gate

The controlled-Z (or `cz`) gate acts on two qubits. It performs the Z gate on the second qubit only when the first qubit is `|1>`, leaving the first qubit unchanged.

Its matrix representation is:

```
[1 0 0 0|
|0 1 0 0|
|0 0 1 0|
|0 0 0 -1|
```

Usage in Qiskit

You can use the ``cz()`` method on a QuantumCircuit object to add a CZ gate. The method takes two main arguments:

1. ``control_qubit``: The index of the control qubit.
2. ``target_qubit``: The index of the target qubit.

Here's a simple example:

```
python
from qiskit import QuantumCircuit

# Create a 2-qubit quantum circuit
qc = QuantumCircuit(2)

# Apply CZ gate with qubit 0 as control and qubit 1 as target
qc.cz(0, 1)
```

For the most accurate and up-to-date documentation, I would recommend visiting the official Qiskit documentation site. Libraries and software packages like Qiskit are frequently updated, so some details or functionalities might have been added or changed after my last update.



In step 4 of the Grover's algorithm, you do a `cz(1,2,3,4)` function. Why do you do this?



I apologize for the confusion; that was an oversight on my part. You're correct; the ``cz()`` function in Qiskit is designed for a two-qubit operation: one control and one target qubit. The diffuser code provided in step 4 was not accurate.

In Grover's algorithm, the diffuser is designed to flip the amplitude of the zero state and is often implemented with Hadamard gates, Pauli-X gates, and a multi-controlled-Z gate. The way the diffuser is constructed is to perform inversion-about-the-average.

Here's the correct structure for the diffuser for a 4-qubit Grover's algorithm:

1. Apply Hadamard gates to all qubits.
2. Apply X gates (Pauli-X or NOT gates) to all qubits.
3. Apply a multi-controlled Z gate (which flips the amplitude of the state $|1111\rangle$).
4. Apply X gates to all qubits again.
5. Apply Hadamard gates to all qubits again.

The multi-controlled Z gate can be a bit tricky to implement directly, but you can use Qiskit's ``mct`` gate or create your own using ancilla qubits and gate decomposition techniques. The main idea behind the multi-controlled Z gate is to flip the phase of the state when all control qubits are in the state $|1\rangle$.

Let's correct the diffuser:

```
python
def diffuser(nqubits):
    qc = QuantumCircuit(nqubits)
    # Apply H and X gates to all qubits
    for qubit in range(nqubits):
        qc.h(qubit)
        qc.x(qubit)
    # Multi-controlled Z gate
    qc.h(nqubits-1)
    qc.mct(list(range(nqubits-1)), nqubits-1) # multi-controlled-toffoli
    qc.h(nqubits-1)
    # Apply X and H gates to all qubits again
    for qubit in range(nqubits):
        qc.x(qubit)
        qc.h(qubit)
    return qc
```

Again, I apologize for the oversight, and thank you for pointing it out.



thanks. why a multi controlled toffoli instead of a multi controlled cz?



That's a great question!



In many quantum programming libraries, a multi-controlled Toffoli (or MCT) gate is more readily available as a built-in function than a multi-controlled Z (MCZ) gate. In quantum computation, the two gates are equivalent up to a global phase when combined with certain single-qubit gates.

Here's how the equivalence works:

1. A Toffoli gate (or controlled-controlled-not, CCNOT) is a controlled-controlled-X gate. Its action is to flip the target qubit if both control qubits are in the state $|1\rangle$.
2. A CZ gate applies a phase flip (Z operation) to the target qubit if the control qubit is in the state $|1\rangle$.

For the diffuser, we don't really need the physical flipping action of the Toffoli gate; we're more interested in the phase flip. Thus, we can use the MCT in combination with Hadamard gates to effectively implement a multi-controlled Z gate.

Consider a single controlled-X (CNOT) gate:

If you place Hadamard gates before and after a CNOT on its target qubit, the CNOT effectively becomes a controlled-Z:

css

Copy code

```
H --- H = --Z--
|
---X---  ---|---
```

The same principle extends to the multi-controlled version:

If you place Hadamard gates before and after an MCT on its target qubit, the MCT effectively behaves like a multi-controlled Z:

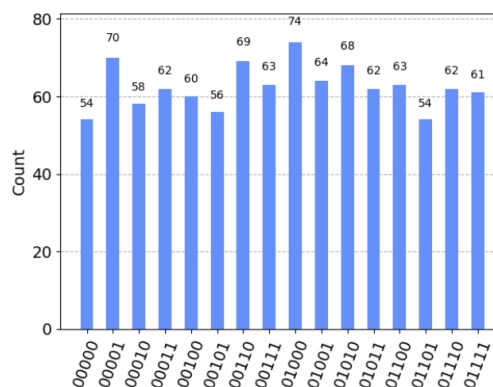
css

Copy code

```
H --- H = --MCZ--
|       .       .
|       .       .
---MCT---  ---|---
```

This transformation works because the Hadamard gate switches between the X and Z bases.

So, by using the MCT gate in conjunction with Hadamard gates (as shown in the diffuser), we're effectively implementing a multi-controlled Z gate. This is a common technique since the MCT is more readily available or easier to decompose for certain quantum hardware architectures.



This appears to be a histogram of results from running a quantum algorithm multiple times, with each bin corresponding to a different outcome (bitstring) of a 5-qubit system.



Observations:

1. The x-axis represents different bitstring outcomes, ranging from `'00000'` to `'01111'`.
2. The y-axis represents the count or the number of times each bitstring was observed.
3. The bitstring `'01111'` seems to have the highest count with 74 occurrences, which stands out from the other bitstrings.

If this is a result from running Grover's algorithm, the bitstring with the highest count (in this case `'01111'`) could be the solution that Grover's algorithm was trying to amplify or search for. Remember, Grover's algorithm amplifies the probability of the desired solution, making it more likely to be observed upon measurement. However, due to the probabilistic nature of quantum mechanics, you'll still see other outcomes, but the correct solution should be the most frequent one (or among the most frequent ones, depending on the number of marked items and the number of Grover iterations).



wouldnt you expect one state (the solution) to stand out a bit more?



Yes, ideally, in Grover's algorithm, you'd expect the solution state to have a significantly higher probability of being measured than the other states. The goal of the algorithm is to amplify the amplitude of the solution state while suppressing the amplitudes of the non-solution states. This should lead to a more pronounced peak for the solution state in the histogram.

Several factors could explain why the results are not as pronounced as expected:

1. **Number of Grover Iterations:** Grover's algorithm requires a specific number of iterations (roughly proportional to the square root of the number of possible solutions) to maximally amplify the desired state. Too few or too many iterations can reduce the contrast between the solution and non-solution states.
2. **Quantum Noise and Errors:** Physical quantum computers (or noisy simulators) introduce errors. These errors can cause the final state to deviate from the expected one, leading to a distribution that is less sharp.
3. **Multiple Solutions:** If the oracle marks multiple states as solutions, then Grover's algorithm will amplify all of them. This can lead to multiple peaks in the histogram.
4. **Implementation Issues:** There might be issues or errors in the quantum circuit implementation or the oracle design.

If you are using a quantum simulator (like the IBM simulator you mentioned earlier), it's ideal to check the number of Grover iterations and ensure the oracle and the diffuser are implemented correctly. If you're using a real quantum device, then noise and errors will naturally play a significant role, and error mitigation techniques might be required.

In the histogram you provided, while the `'01111'` state does stand out as the highest, the difference between it and the surrounding states is not as stark as one might expect from an ideal Grover's search. It might be worth revisiting the implementation details and the number of Grover iterations to enhance the contrast.



In the function that you wrote you apply hadamars and X gates on all qubits. I dont really remember these X gates. Could you check if this was done correct?



Certainly. Let's review the structure of the Grover diffuser (often referred to as the "inversion about the average" step) for clarity.

The diffuser is typically constructed as follows:

1. Apply Hadamard gates to all qubits.
2. Apply a series of gates to flip the amplitude of the $|0\dots 0\rangle$ state (often achieved using X gates and a multi-controlled-Z gate).
3. Apply Hadamard gates to all qubits again.

The purpose of the X gates in the middle step is to change the state $|0\dots 0\rangle$ into the state $|1\dots 1\rangle$, which can then have its phase flipped with a multi-controlled-Z gate. After the phase flip, the X gates revert the state back to $|0\dots 0\rangle$, but with a flipped phase.

So, the correct structure for the diffuser, given the purpose and mechanism explained, should be:

1. Apply Hadamard gates to all qubits.
2. Apply X gates (Pauli-X or NOT gates) to all qubits.
3. Apply a multi-controlled Z gate (which flips the amplitude of the state $|1111\rangle$ for a 4-qubit system).
4. Apply X gates to all qubits again.

7. Apply X gates to all qubits again.
5. Apply Hadamard gates to all qubits again.

Thus, the use of X gates in the diffuser function I provided is consistent with the standard structure of Grover's algorithm. These X gates help in setting up the system for the phase flip (using the multi-controlled-Z gate) and then reverting the system back to its original basis.



To run the full grover search algorithm, there seems to be a part before the diffuser that is missing now. Could that be the case? I think that might be the part where you encode the solution (the item/state that you seek)



You're absolutely right. Grover's algorithm consists of two main parts:



1. **Oracle Query:** The oracle encodes the solution by flipping the sign (phase) of the state corresponding to the solution. If $|w\rangle$ is the solution state, after applying the oracle U_w , the oracle would transform the system as:

$$U_w|w\rangle = -|w\rangle$$

All other states remain unchanged. This phase flip is crucial for the amplification process.

2. **Diffusion Operator (or Amplification):** This is the step we discussed in detail – it amplifies the amplitude of the solution state, making it more probable to measure.

The full Grover iteration combines these two steps. If you are searching for a specific item (or multiple items) in an unsorted database, you'd first need to design an oracle that recognizes this item and flips its phase. Only after this oracle query do you apply the diffuser.

For a concrete example, let's say you're searching for the binary state $|101\rangle$ in a 3-qubit system. Your oracle needs to recognize this state and flip its phase. A simple way to design the oracle for this specific case is:

1. Apply a Z gate to the second qubit (because it's in state 0, while the 1st and 3rd qubits are in state 1).
2. Apply a Toffoli gate with the first and third qubits as controls and the second qubit as the target. This ensures the phase flip only happens for the state $|101\rangle$.
3. Revert the Z gate on the second qubit.

This would create the desired phase flip for the $|101\rangle$ state. Note that this oracle design is specific to this example and would differ for other states or combinations of states you might be searching for.

Once the oracle is applied, the next step is to apply the diffuser, as previously discussed. Together, the oracle and the diffuser form one Grover iteration. The process is repeated for the optimal number of iterations to maximize the probability of measuring the solution state.



what is the command to draw the circuit?



In Qiskit, to draw the quantum circuit, you can use the `draw()` method of the QuantumCircuit object. Here's how you can do it:



```
python
circuit.draw(output='mpl')
```

Copy code

Here's a breakdown:

- `circuit`: This is the name of your QuantumCircuit object.
- `draw()`: This is the method to visualize the circuit.
- `output='mpl'`: This specifies that you want to use the Matplotlib-based drawer for the visualization. It produces a cleaner and more visually appealing diagram compared to the default text-based output.

Make sure you have Matplotlib installed to use this drawing method:

```
bash
pip install matplotlib
```

Copy code

Once you run the command, it will display a visual representation of your quantum circuit. If you're working in a Jupyter notebook, the circuit will be displayed directly in the notebook. If you're using a different environment, ensure you're using a GUI backend for Matplotlib to view the plot.



I think there is an error in the oracle part. could you revise it?



< 2 / 2 >



Certainly! Let's delve into creating an oracle for the state $|101\rangle$ in a 3-qubit system.



To mark (or flip the phase of) the state $|101\rangle$:

1. Apply a Pauli-X gate to the second qubit. This is to invert the '0' in the $|101\rangle$ state.
2. Use a Toffoli (CCX) gate with the first and third qubits as controls and the second qubit as the target. This ensures a NOT operation (Pauli-X) on the second qubit only when the first and third qubits are in state '1'.
3. Apply a Pauli-X gate again to the second qubit to revert it back to its original state.

In Qiskit, it would look like:

```
python Copy code  
  
from qiskit import QuantumCircuit  
  
# Initialize a 3-qubit quantum circuit  
oracle = QuantumCircuit(3)  
  
# Apply X gate to the second qubit  
oracle.x(1)  
  
# Apply Toffoli gate with qubits 0 and 2 as control and qubit 1 as target  
oracle.ccx(0, 2, 1)  
  
# Revert the second qubit to its original state  
oracle.x(1)  
  
# You can draw the circuit to verify  
oracle.draw(output='mpl')
```



This oracle correctly marks the state $|101\rangle$. When the oracle acts on this state, it will introduce a phase flip, making it $-|101\rangle$. Other states will be unaffected.

Let me know if you'd like further clarification or if there's any other aspect you'd like to discuss!

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Python 3 (ipykernel) ○

```
[1]: from qiskit import QuantumCircuit, Aer, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import QasmSimulator
```

```
[15]: nqubits=5
oracle = QuantumCircuit(5)
oracle.x(2)
oracle.mct([0,1,2,3], 4)
oracle.x(2)
```

```
[15]: <qiskit.circuit.instructionset.InstructionSet at 0x7f5273a70550>
```

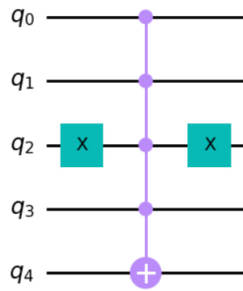
```
[16]: def diffuser(nqubits):
    qc = QuantumCircuit(nqubits)
    # Apply H and X gates to all qubits
    for qubit in range(nqubits):
        qc.h(qubit)
        qc.x(qubit)
    # Multi-controlled Z gate
    qc.h(nqubits-1)
    qc.mct(list(range(nqubits-1)), nqubits-1) # multi-controlled-toffoli
    qc.h(nqubits-1)
    # Apply X and H gates to all qubits again
    for qubit in range(nqubits):
        qc.x(qubit)
        qc.h(qubit)
    return qc
```

```
[17]: grover_circuit = QuantumCircuit(5, 5)
grover_circuit.h([0, 1, 2, 3])
grover_circuit.append(oracle, [0, 1, 2, 3, 4])
grover_circuit.append(diffuser(4), [0, 1, 2, 3])
grover_circuit.measure([0, 1, 2, 3], [0, 1, 2, 3])
```

```
[17]: <qiskit.circuit.instructionset.InstructionSet at 0x7f5273c9be50>
```

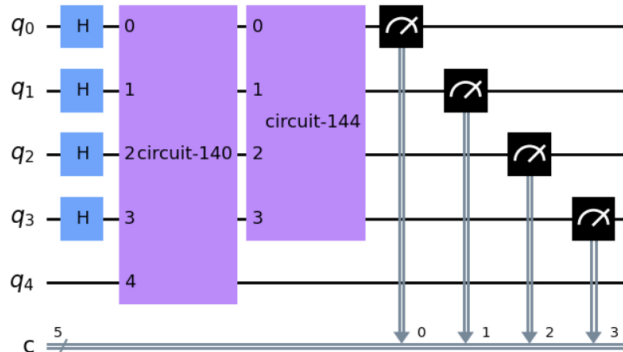
```
[18]: oracle.draw('mpl')
```

```
[18]:
```



```
[19]: grover_circuit.draw('mpl')
```

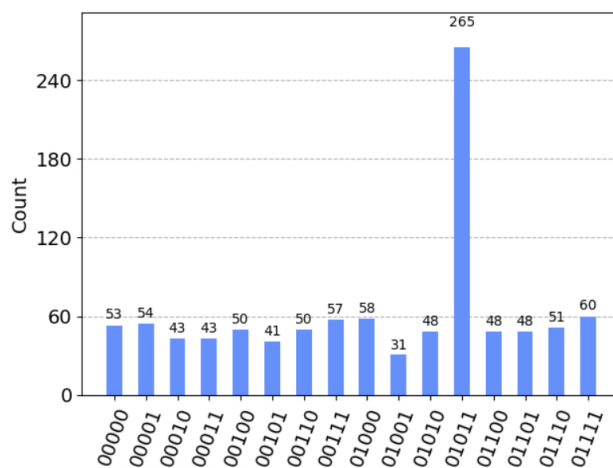
```
[19]:
```



```
[20]: simulator = Aer.get_backend('qasm_simulator')
compiled_circuit = transpile(grover_circuit, simulator)
```

```
job = simulator.run(compiled_circuit, shots=1000)
result = job.result()
counts = result.get_counts()
plot_histogram(counts)
```

[28]:



[]:

P6

- Measure the quantum register to obtain the solution.

For a 32-qubit computer, depending on how many solutions (marked items) you have, you'd adjust the number of iterations to maximize the success probability.

Libraries for Implementation:

For the IBM Q Experience, you'd use the `qiskit` library. Here's a breakdown:

- `qiskit`: This is the main library for quantum computing with IBM's quantum devices and simulators.
 - `qiskit.Aer`: Contains methods to run quantum circuits on local simulators.
 - `qiskit.circuit`: Provides classes to create quantum circuits.
 - `qiskit.visualization`: Tools to visualize quantum circuits and results.
 - `qiskit.algorithms`: Contains the implementation for various quantum algorithms including Grover's.

Once you have a grasp on these steps and libraries, I can guide you through a more detailed code implementation, if desired. Would you like to proceed with that?



for every step and every qubit could you define which gate i should implement so I can code it in qiskit



Of course! Let's break down each step with the specific gates and operations for a general n -qubit setup, so you can adapt it to the 32-qubit computer:

1. **Initialization:**
 - Prepare all n qubits in the $|0\rangle$ state (which they usually are by default).
 - Apply a Hadamard (`H`) gate to each of the n qubits.
2. **Oracle Query:**
 - This is a problem-specific operation and depends on the function you're trying to invert. For a black-box function that marks a single item `w`, the oracle `Uw` flips the sign of the state $|w\rangle$. You'll often use controlled operations (like `CNOT` or controlled-`Z`) based on the function.
3. **Amplitude Amplification:**
 - a. Inversion about the mean:
 - Apply a Hadamard (`H`) gate to all n qubits.
 - Apply a Pauli-X (`X`) gate to all n qubits.
 - Apply a multi-controlled-Z (`MCZ`) gate. The MCZ gate is a Z gate applied to the last qubit but controlled by all the other qubits.
 - Apply a Pauli-X (`X`) gate to all n qubits again.
 - Apply a Hadamard (`H`) gate to all n qubits again.
4. **Iterate:**
 - The Oracle Query and the Amplitude Amplification steps are grouped together as one Grover iteration. You'll repeat this about \sqrt{N} times (rounded to the nearest integer). N is the total number of possible states (2^n for n qubits). For 32 qubits, this would be roughly 2^{16} or 65,536 iterations, but in practice, it's often fewer because you might know something about the number of solutions or can optimize based on the problem.
5. **Measurement:**
 - No gate needed. Directly measure all n qubits.

Using Qiskit, here's a high-level code outline:

```
python Copy code

from qiskit import QuantumCircuit, Aer, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import AerSimulator
from qiskit.tools.monitor import job_monitor

# Define the quantum circuit
qc = QuantumCircuit(n)

# Initialization
for qubit in range(n):
    qc.h(qubit)

# Grover iterations
for _ in range(int(round(2**16))):
    # Oracle - This is a placeholder, as the actual oracle depends on your problem
    # For example, for a function marking the |101> state:
    # qc.cz(0,2)
    # ...add the rest of your oracle...

    # Amplitude Amplification
    for qubit in range(n):
        qc.h(qubit)
```

```

qc.x(qubit)

qc.mcz(list(range(n-1)), n-1) # Assuming the MCZ gate is defined or you

for qubit in range(n):
    qc.x(qubit)
    qc.h(qubit)

# Measurement
qc.measure_all()

# Simulate the circuit
simulator = AerSimulator()
compiled_circuit = transpile(qc, simulator)
job = simulator.run(compiled_circuit, shots=1000)
result = job.result()
counts = result.get_counts(qc)

# Visualize the results
plot_histogram(counts)

```

This is a very basic outline, and it might need adjustments based on the specifics of your problem and oracle function. Once you've defined your oracle function, it becomes more straightforward to determine how to implement it using quantum gates in Qiskit.



and what function for the oracle query do i need if i want to break cryptography



Breaking cryptography with quantum computing usually refers to one of two main algorithms: Shor's algorithm (for factoring large numbers, which would threaten RSA encryption) and Grover's algorithm (for searching an unsorted database, which could halve the effective key length of symmetric cryptographic algorithms like AES).

Given you're focusing on Grover's algorithm, let's consider you want to break a simple symmetric cryptographic scheme, say a binary password of length 'n' bits (or a cryptographic hash reduced to n bits). The "unsorted database" is then the space of all possible passwords or hashes, and the marked item is the correct one.

The Oracle would be designed to recognize this correct password or hash. This would mean the Oracle checks if an input matches the correct password/hash, and if so, flips its phase. In terms of quantum gates:

1. Encode the function that checks if a quantum state corresponds to the correct password/hash.
2. If the state matches, apply a phase flip.

For instance, suppose you have a 3-bit password "101". A simplified Oracle might be constructed as follows:

1. Check each bit of the password using controlled gates.
2. If all bits match, flip the phase.

In Qiskit, this could be something like:

```

python
Copy code

# Assuming the password is '101' and we're using a 3-qubit system
def oracle_101(qc):
    qc.cz(0, 2) # This effectively checks if qubit 0 is in state |1> and qubit 2 is in state |1>
                # (because a Z gate flips the phase of |1>).
                # The absence of a gate on qubit 1 assumes it's in state |0>

```

However, real-world cryptography uses long keys and complex functions, and simply searching for the key would be inefficient. An actual cryptographic Oracle would involve quantum gates that represent the cryptographic function and a mechanism to detect the correct output.

For more complex cryptographic schemes, the Oracle's construction becomes non-trivial. This is the main challenge of quantum algorithms for cryptanalysis: designing an efficient Oracle. For algorithms like AES, the Oracle would need to encode the encryption process, then recognize a correct encryption (perhaps by matching a known ciphertext).

In summary, breaking cryptography with Grover's algorithm boils down to designing an Oracle that recognizes a correct cryptographic key or output. The more complex the cryptographic algorithm, the more complex the Oracle.

Note: Even with Grover's algorithm, symmetric cryptographic schemes like AES remain secure by doubling the key length. A quantum computer would still give only a quadratic speedup, making AES-256, for instance, as hard to crack as AES-128 with classical methods.



could you give me the qiskit/python code for Grovers algorithm for breaking the cryptography key of 110



If you're looking to use Grover's algorithm to search for a cryptographic key of value "110", it means you're trying to break a 3-bit key, which is overly simplified for illustrative purposes. Real cryptographic keys are much longer, making the procedure more complex.

Here's a step-by-step implementation in Qiskit:

1. **Initialization:**
Set up a quantum circuit with 3 qubits and 3 classical bits (for measurement).
2. **Oracle for the key "110":**
Implement an oracle that flips the phase for the state |110).
3. **Grover's Diffusion Operator:**
Implement the diffusion operator for the 3 qubits.
4. **Measurement:**
Measure the qubits.

Let's code this:

```
python Copy code

from qiskit import QuantumCircuit, Aer, transpile, assemble
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import AerSimulator

n = 3 # number of qubits

# Quantum Circuit
qc = QuantumCircuit(n)

# Initialization
for qubit in range(n):
    qc.h(qubit)

qc.barrier() # just for visualization purposes

# Oracle for key "110"
qc.cz(0, 2) # phase flip if qubit 0 and 2 are in state |1>

qc.barrier() # just for visualization purposes

# Grover's Diffusion Operator
for qubit in range(n):
    qc.h(qubit)
    qc.x(qubit)

qc.cz(0, 2) # multi-controlled-Z gate with qubit 2 as target

for qubit in range(n):
    qc.x(qubit)
    qc.h(qubit)

qc.barrier() # just for visualization purposes

# Measurement
qc.measure_all()

# Execute and visualize
simulator = AerSimulator()
compiled_circuit = transpile(qc, simulator)
job = simulator.run(compiled_circuit, shots=1000)
result = job.result()
counts = result.get_counts(qc)
plot_histogram(counts)
```

Executing the above code will display a histogram. Ideally, if everything is correct, the |110) state should have the highest probability.

Remember, this is a basic example for a 3-bit key. In real-world cryptography, keys are much longer, and constructing the oracle is the most challenging part since it needs to efficiently check a cryptographic condition. The Grover iterations also might need to be adjusted based on the problem specifics. For a 3-qubit problem like this, one Grover iteration (oracle + diffusion operator) is enough.



could you break down this steps to the specific gates on the specific qubits that need to be implemented



Certainly! Let's break down the code and its corresponding quantum operations on each qubit step-by-step:



1. Initialization:

For every qubit in the system, we are applying a Hadamard gate:

scss

Copy code

```
|0> --H--> (1/√2)(|0> + |1>)
```

So, the 3-qubit system after initialization becomes:

scss

Copy code

```
|000> --H(each qubit)--> (1/√8)(|000> + |001> + |010> + |011> + |100> + |101> +
```

In terms of gates:

yaml

Copy code

```
Qubit 0: H
Qubit 1: H
Qubit 2: H
```

2. Oracle for the key "110":

The goal is to flip the phase of the state $|110\rangle$. We achieve this using a controlled-Z ('cz') gate which acts on two qubits: if the control qubit is $|1\rangle$, it applies a Z gate (phase flip) to the target qubit.

Given the state $|110\rangle$, Qubits 0 and 2 are both in the state $|1\rangle$. We can use them as controls to apply a phase flip to the system:

lua

Copy code

```
Qubit 0: ----- (Control for the CZ gate)
Qubit 1: -----
Qubit 2: -----Z----- (Target for the CZ gate)
```

This 'cz' operation flips the phase of the $|110\rangle$ state.

3. Grover's Diffusion Operator:

Here, we want to amplify the amplitude of the marked state and reduce the others.

a. Apply H and X gates to every qubit:

This moves every qubit to the $|1\rangle$ state:

lua

Copy code

```
Qubit 0: H--X--
Qubit 1: H--X--
Qubit 2: H--X--
```

b. Controlled-Z gate:

We again use Qubits 0 and 2 for the 'cz' operation:

lua

Copy code

```
Qubit 0: ----- (Control for the CZ gate)
Qubit 1: -----
Qubit 2: -----Z----- (Target for the CZ gate)
```

This flips the phase of the $|110\rangle$ state once more.

c. Apply X and H gates to every qubit again:

This completes the amplitude amplification, making the amplitude of the $|110\rangle$ state larger than the others:

```
lua
Qubit 0: --X--H--
Qubit 1: --X--H--
Qubit 2: --X--H--
```

4. Measurement:

Finally, all the qubits are measured, collapsing the quantum state to one of the basis states with the highest probability being the $|110\rangle$ state we amplified.

In summary, for the problem of finding the key "110", the sequence of gates for each qubit can be described as:

```
css
Qubit 0: H - cz(control) - H - X - cz(control) - X - H - Measure
Qubit 1: H ----- H - X ----- X - H - Measure
Qubit 2: H - cz(target) -- H - X - cz(target) -- X - H - Measure
```

It's worth noting that for more complex keys or oracles, the number and types of gates would change, and for systems with more qubits, you'd potentially need more controlled operations and might need to adapt the diffusion operator accordingly.



what is the qiskit code to visualize the circuit



To visualize the quantum circuit in Qiskit, you can use the `'draw'` method available on the `'QuantumCircuit'` object. This will provide a graphical representation of the circuit. Here's how to do it:

1. First, ensure you have the necessary visualization libraries installed:

```
bash
pip install matplotlib
```

2. Create and prepare your quantum circuit (as in the previous examples).
3. After creating the circuit, use the `'draw'` method to visualize it:

```
python
# ... [your circuit creation code here] ...

# Draw the circuit
qc.draw('mpl') # 'mpl' specifies the use of the Matplotlib-based drawer
```

1. To display the drawn circuit within a Jupyter notebook or similar environments, you can directly use:

```
python
from IPython.display import display

display(qc.draw('mpl'))
```

If you are working in a traditional Python script (not in a notebook), you might want to use Matplotlib to display the image:

```
python
import matplotlib.pyplot as plt

circuit_diagram = qc.draw('mpl')
plt.show()
```

This will produce a visual representation of your quantum circuit, showing each gate and its position on the corresponding qubit.



where in the code give above is the key 011 implemented



I apologize for the oversight. In the code I provided earlier, the oracle was designed to mark the state $|110\rangle$. If you want the key to be "011", you would need to modify the oracle.



To construct an oracle for the key "011", you'd want to flip the phase of the state $|011\rangle$. Here's how you can modify the oracle for this:

1. Use Qubit 0 as the control to ensure it's in state $|0\rangle$.
2. Use Qubits 1 and 2 as controls to ensure they are both in state $|1\rangle$.

This can be achieved with a `'ccz'` gate (controlled-controlled-Z gate), where Qubit 0's control is based on its NOT value (i.e., it being in state $|0\rangle$).

In Qiskit's standard gate set, there's no direct `'ccz'` gate. However, you can create one using a combination of Hadamard, T, T^\dagger (T dagger), and controlled-controlled-X (`'ccx'` or Toffoli) gates. But for the sake of simplicity, let's use a trick by applying an X-gate (NOT operation) on Qubit 0 before and after a `'ccx'` operation.

Here's how you can implement the oracle for key "011":

```
python Copy code

# Oracle for key "011"
qc.x(0) # NOT gate on Qubit 0
qc.ccx(0, 1, 2) # Toffoli gate with Qubits 0 and 1 as controls and Qubit 2 as target
qc.x(0) # NOT gate on Qubit 0 to revert it back
```

Replace the earlier oracle implementation in the provided code with the above lines, and the oracle will now mark the state $|011\rangle$.

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Python 3 (ipykernel) ○

```
[3]: from qiskit import QuantumCircuit, Aer, transpile, assemble
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import AerSimulator
from IPython.display import display

n = 3 # number of qubits

# Quantum Circuit
qc = QuantumCircuit(n)

# Initialization
for qubit in range(n):
    qc.h(qubit)

qc.barrier() # just for visualization purposes

# Oracle for key "110"
qc.cz(0, 2) # phase flip if qubit 0 and 2 are in state |1>

qc.barrier() # just for visualization purposes

# Grover's Diffusion Operator
for qubit in range(n):
    qc.x(qubit)
    qc.x(qubit)

qc.cz(0, 2) # multi-controlled-Z gate with qubit 2 as target

for qubit in range(n):
    qc.x(qubit)
    qc.h(qubit)

qc.barrier() # just for visualization purposes

# Measurement
qc.measure_all()

display(qc.draw('mpl'))

# Execute and visualize
simulator = AerSimulator()
compiled_circuit = transpile(qc, simulator)
job = simulator.run(compiled_circuit, shots=1000)
result = job.result()
counts = result.get_counts(qc)
plot_histogram(counts)
```

meas 3

0 1 2

```
[3]:
```

Measurement Result	Count
0	506
1	494

[]: 📄 ↶ ↷ 📄 ↶ ↷ 📄 ↶ ↷ 📄 ↶ ↷ 📄 ↶ ↷

d. Miroboard results:

https://miro.com/welcomeonboard/VkNjU3B0bDhYdmpzQzU5NFlUODR6eXlUdEN5N-k14WFh3emxTSDNzOV05TFBtSGhJUGZUQzBHYNREalpCdGRPaHwzNDU4NzY0NTE4M-zE0MjA1NTE5fDI=?share_link_id=641786781195

[IV] IDE workshop

a. Miroboard workshop:

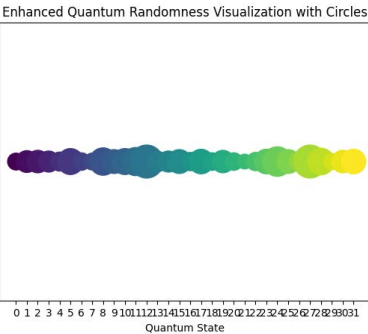
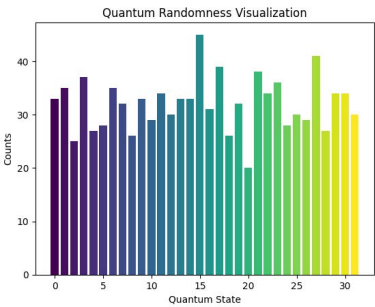
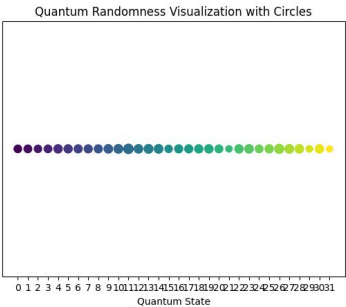
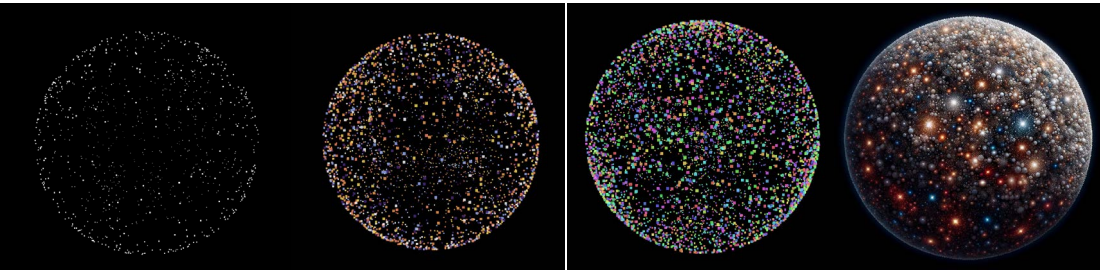
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b. Miroboard results

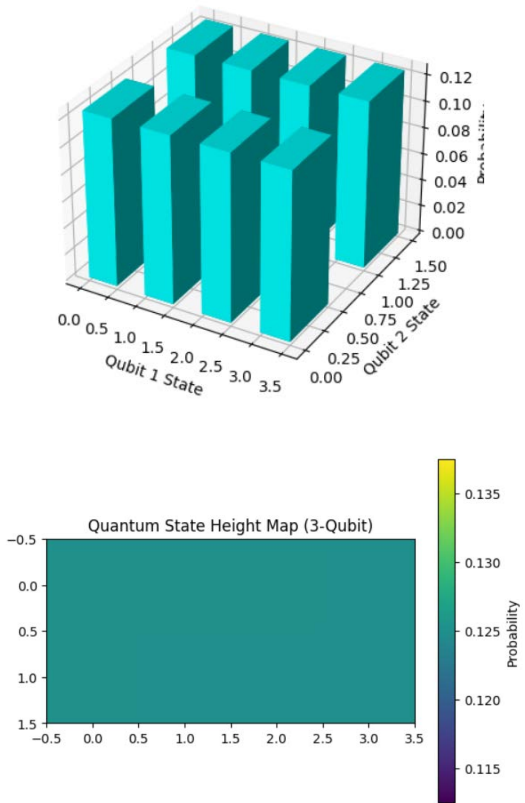
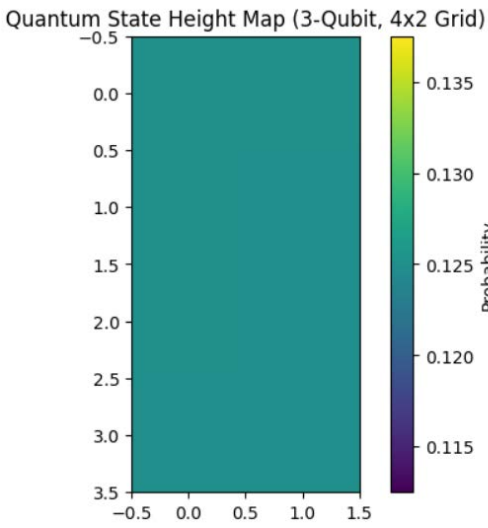
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c. Results collection

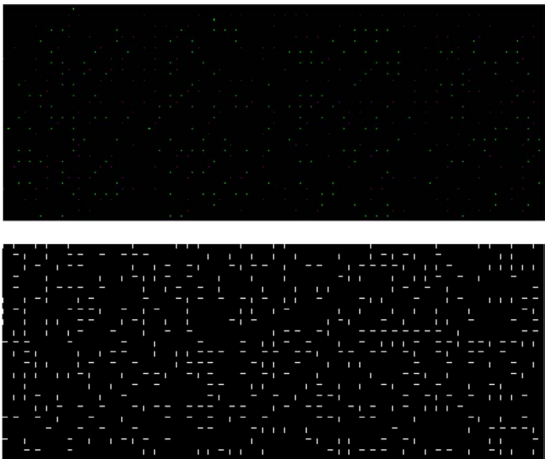
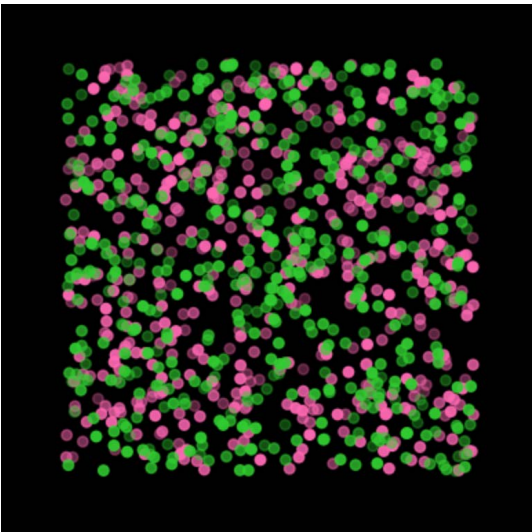
ROOM 1



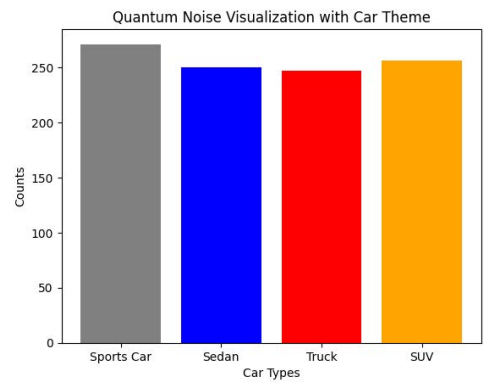
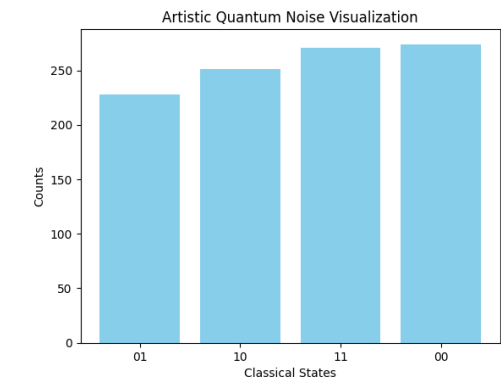
ROOM 2



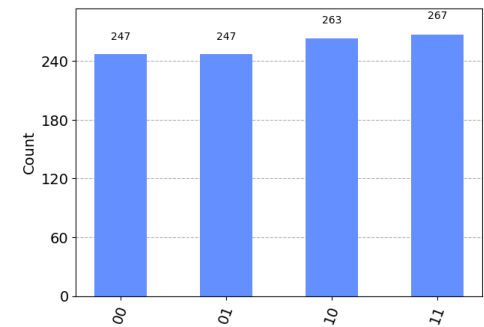
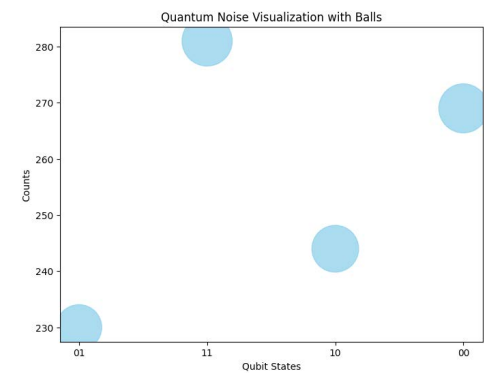
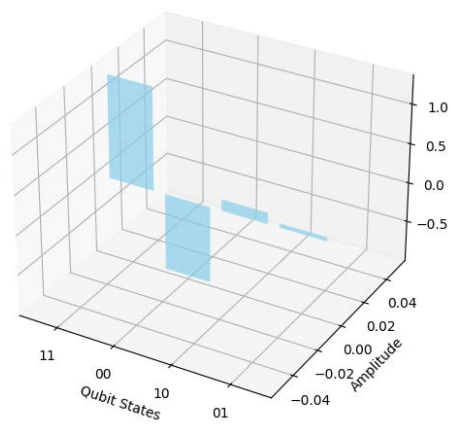
ROOM 3



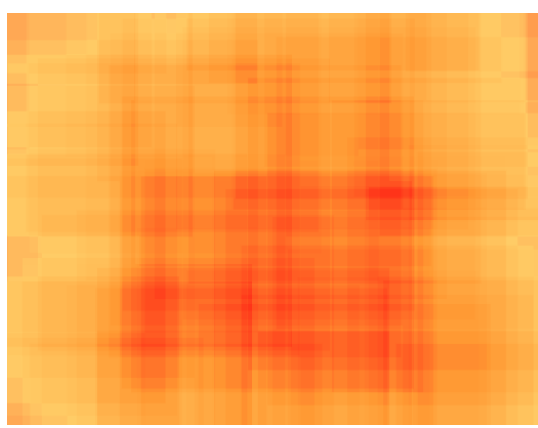
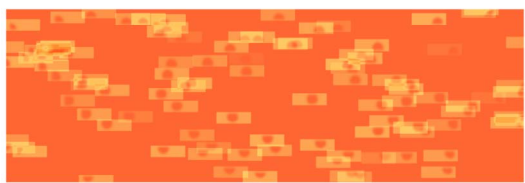
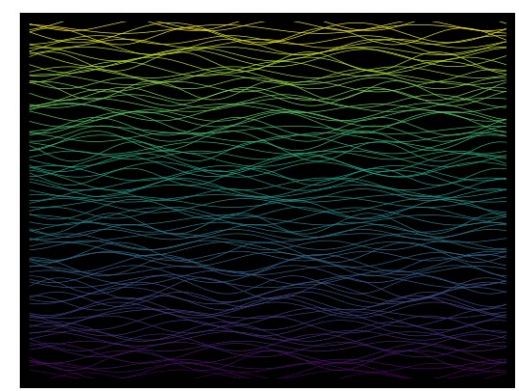
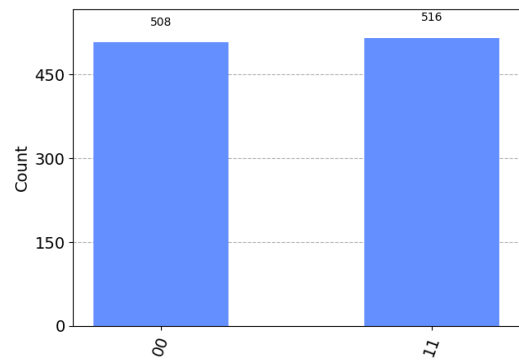
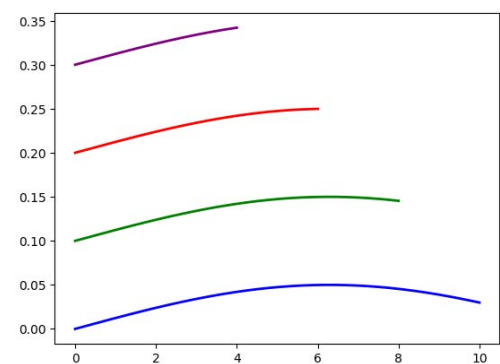
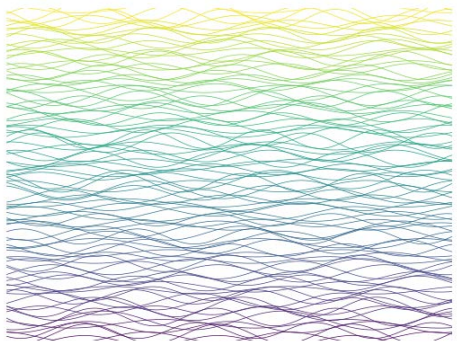
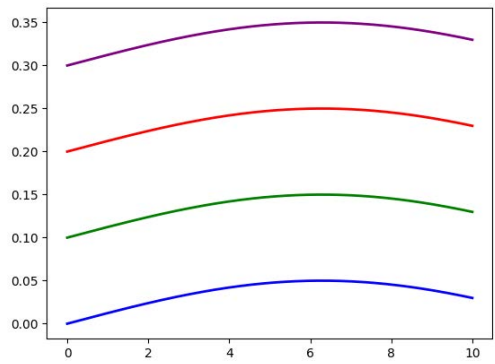
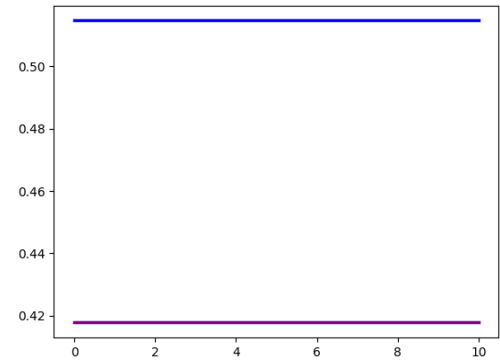
ROOM 4

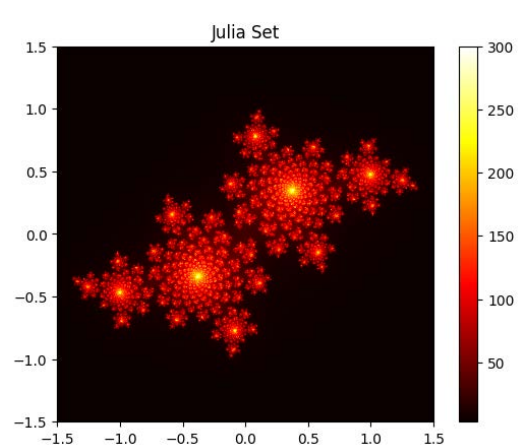
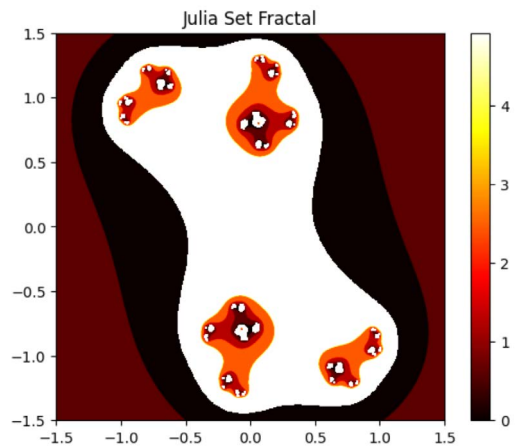
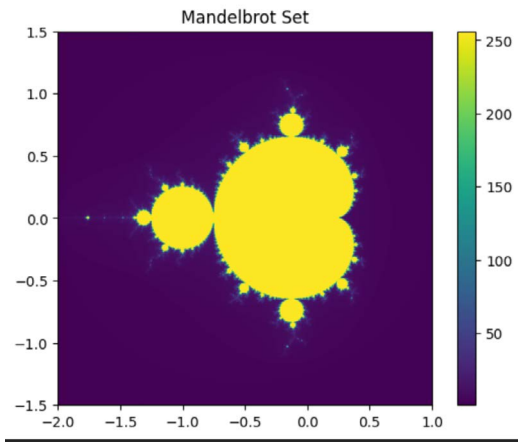
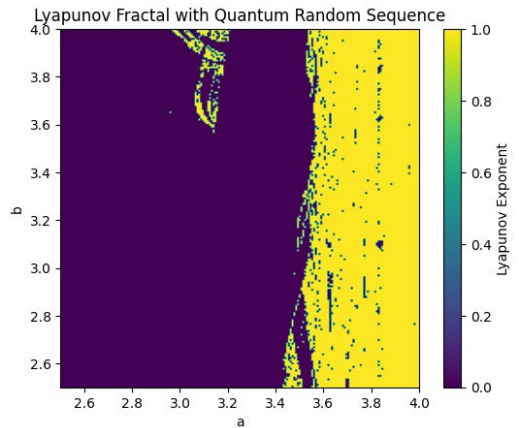
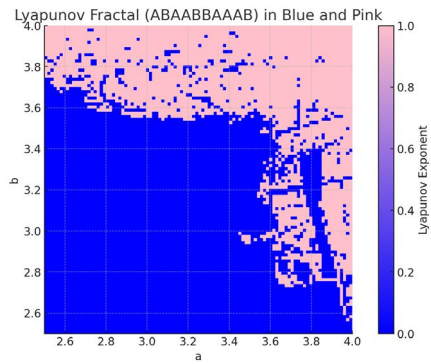
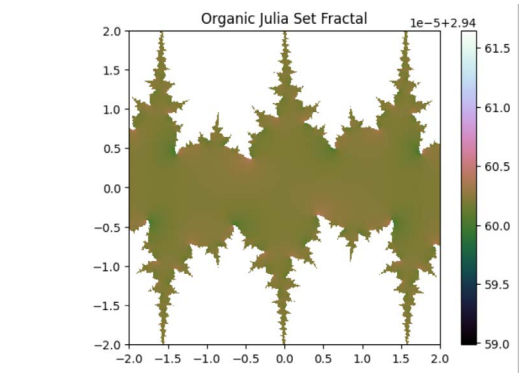
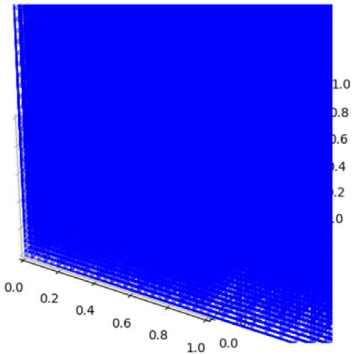


3D Quantum Noise Visualization with Landscape Layout

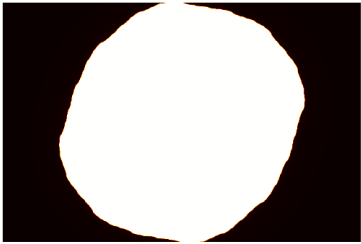
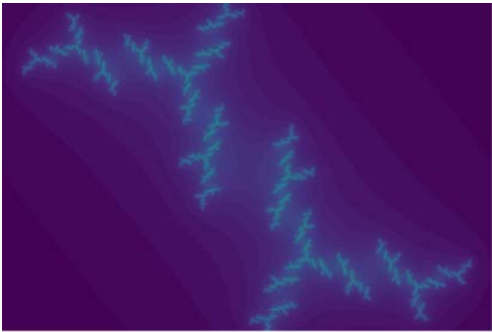
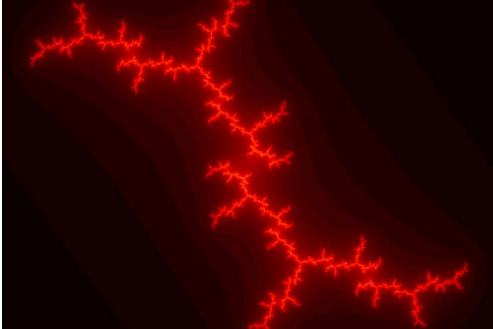
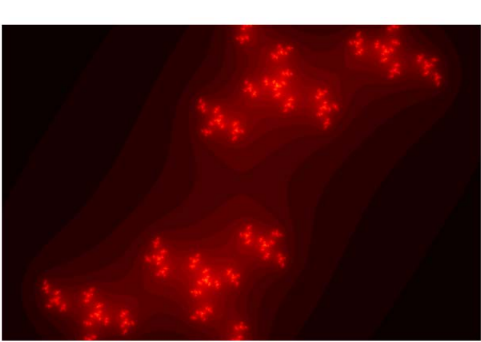
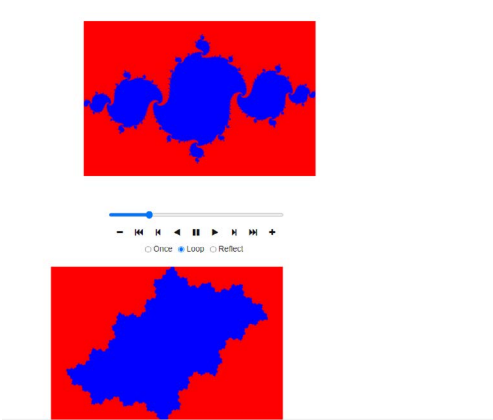
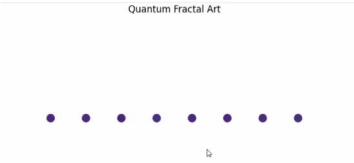
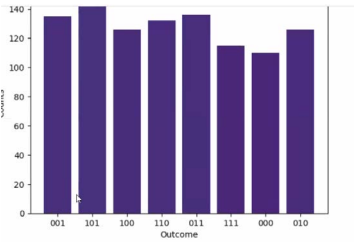


ROOM 5

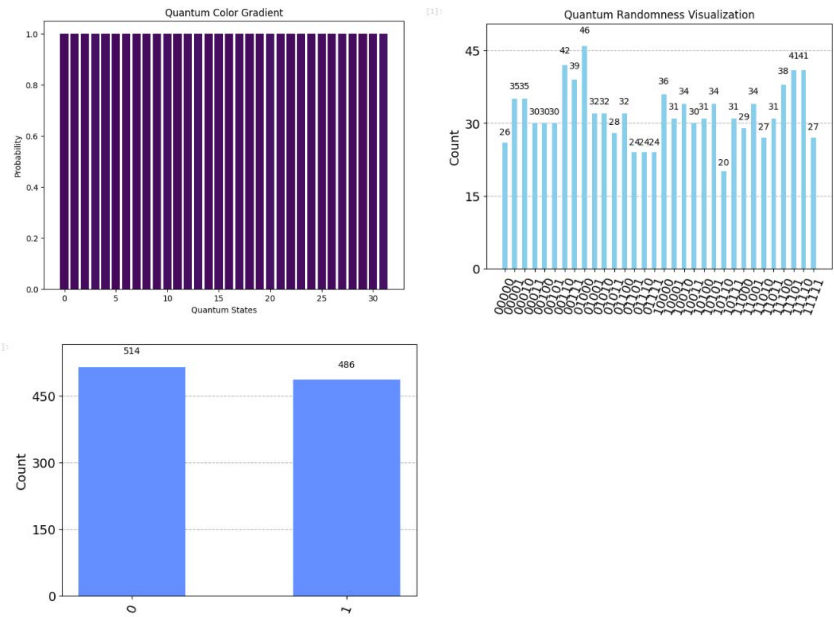




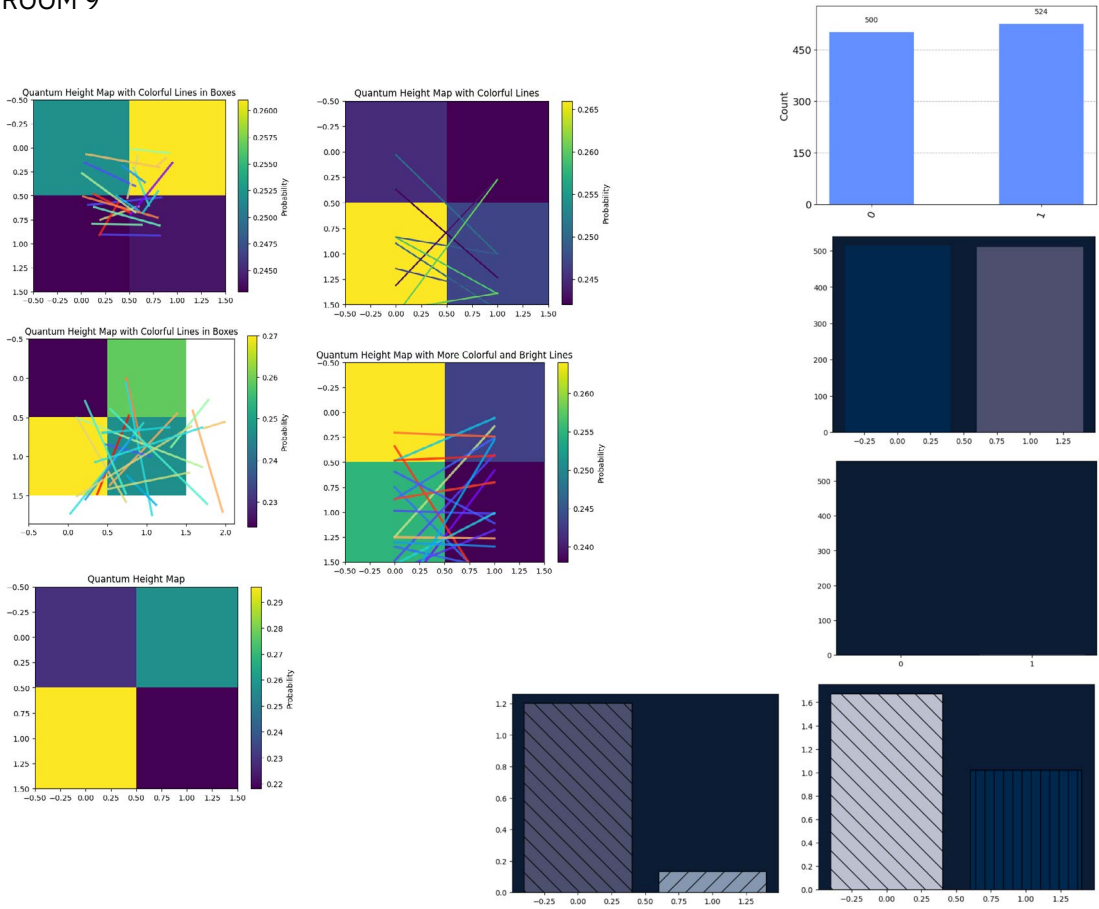
ROOM 7



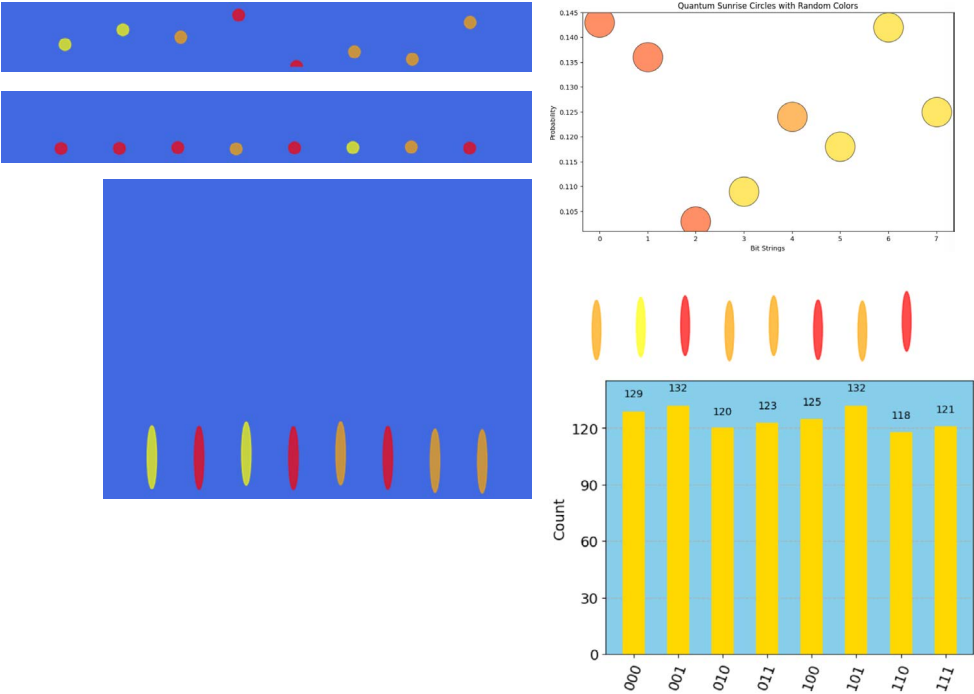
ROOM 8



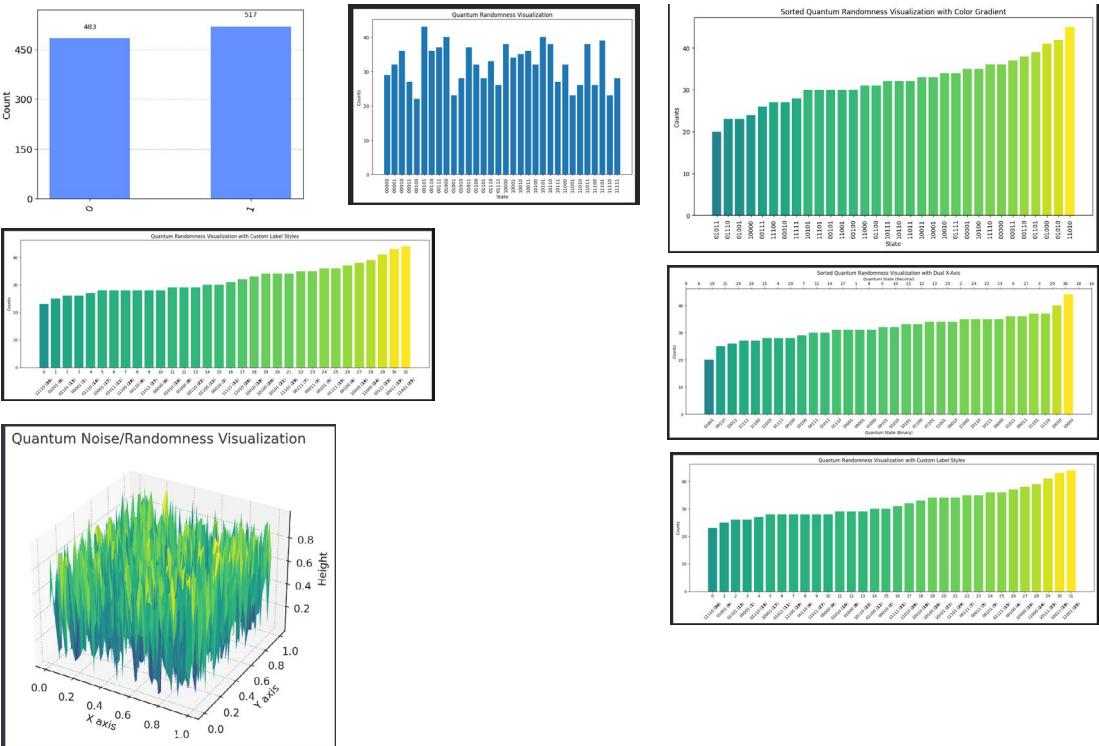
ROOM 9



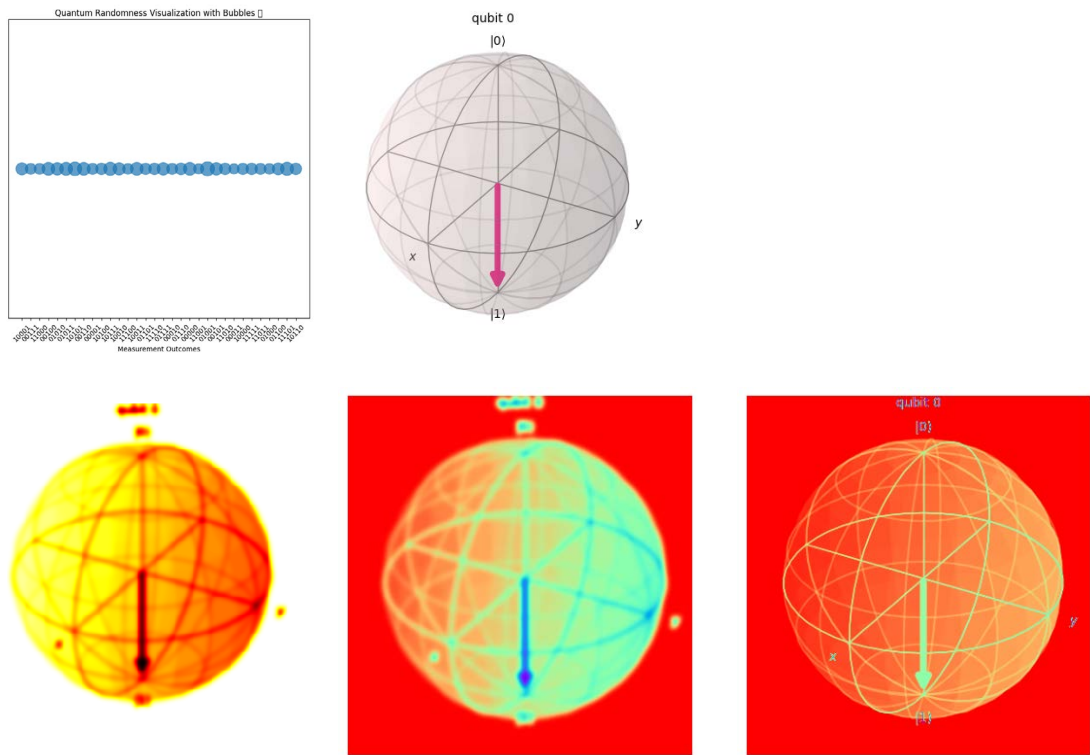
ROOM 10



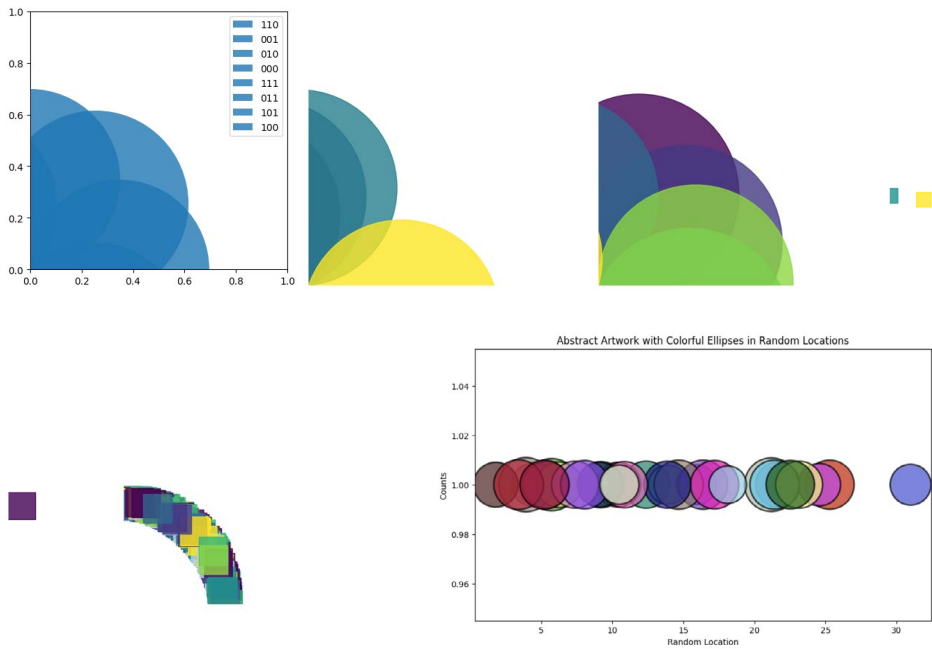
ROOM 11



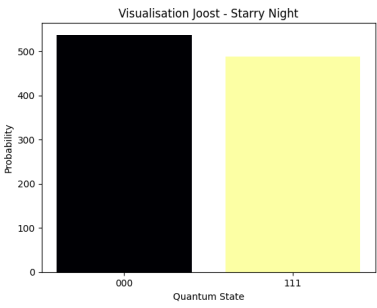
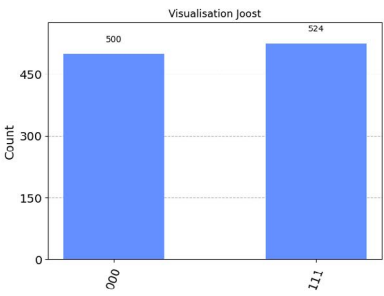
ROOM 12



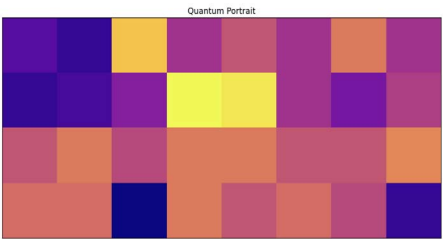
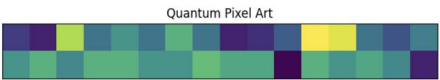
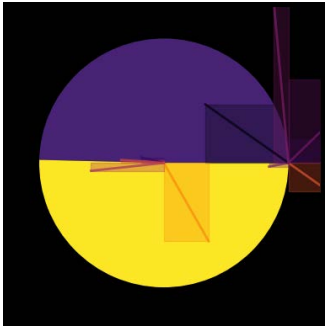
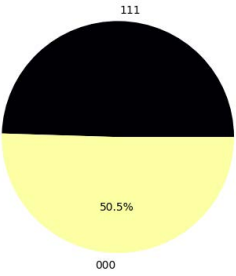
ROOM 13



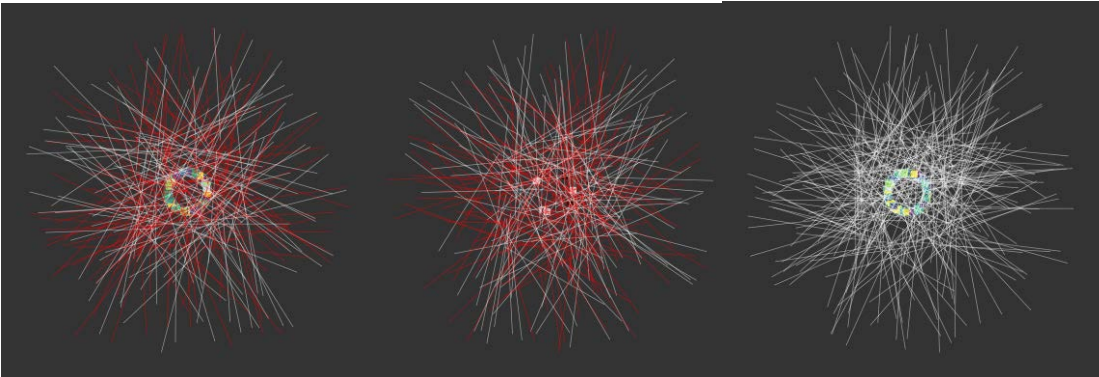
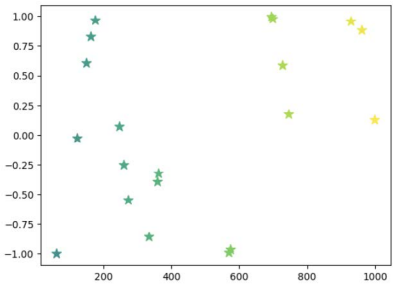
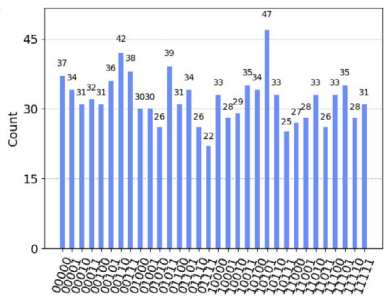
[2]:



Visualisation Joost - Gradient Starry Night (Pie Chart)



[1]:



[V] Quantum Explorer

a. GPTs V1 - Quantum Explorer

The initial version of GPT was developed using the OpenAI “myGPT” tool. As a trial version, its creation aimed to comprehend the underlying mechanisms. This version was constructed to create a custom AI capable of understanding the realm of quantum mechanics. The prompt provided to the AI was as follows:

“Quantum Explorer is a cheerful, beginner-friendly assistant for quantum computing, designed to adapt to various expertise levels: ‘Newbie’, ‘Code Expert’, and ‘Quantum Mechanics Expert’. It offers customizable learning paths for each user, tailoring content and recommendations based on their knowledge and goals. Quantum Explorer provides accessible explanations, incorporating emojis for a friendly approach. For newbies, it covers basic concepts; for code experts, programming aspects of quantum algorithms; and for quantum mechanics experts, advanced theoretical discussions. It uses relatable phrases and analogies, ensuring that learning about quantum computing is engaging and informative. Quantum Explorer’s goal is to guide users through a personalized journey in quantum computing, making it a rewarding experience for everyone.”

Named Quantum Explorer, Version 1 was then independently evaluated to gauge its capabilities. The testing involved assessing the chatGPT’s response to a hypothetical general request, given the tester’s background as a designer.

“I would like to create a combination of lights that is always different for an LED panel that I want to install. Can I do it with a quantum algorithm?”

Subsequently, the GPT inquired about the tester’s level of expertise and proceeded to create a practical guide outlining the subsequent steps. After obtaining the code and applying it within the IBM quantum lab, the GPT was requested to produce a visualization of the quantum state. It accomplished this through a visual representation from DALL-E, and later, it was asked to provide an image of what the final result might look like once implemented in the initially proposed design.

b. Link for Quantum Explorer

<https://chat.openai.com/gg-xQ9eEcoSN-quantum-explorer>

[VI] Quantum Buddy

a. GPTs V2 - Quantum Buddy

Understanding that GPT could be enhanced with raw data, the individual decided to advance it further and make some modifications. The improved version, named Quantum Buddy, maintained the same initial prompt as its predecessor but distinguished itself through additional data and capabilities.

Initially, there was a desire to comprehend and acquire knowledge about the Python code and Quiskit used in the IBM quantum simulator. To achieve this, the working code generated during the thesis research process has been used. The new material code was provided first, followed by the new drug code, then the quantum Grover algorithm, and finally, the creative quantum coding developed for the first workshop.

To gain a comprehensive understanding of the available algorithms, was decided to incorporate quantum algorithms currently utilized in the field of quantum computing. Additionally, a general knowledge of quantum computers was incorporated, based on the paper “Quantum computing for non-physics,” (E. Rieffel, W. Polak, 2000) along with an understanding of the workings and structure of the IBM quantum lab (J. Coles, 2018). Recognizing the capabilities of GPT-4, the GPT was also tasked with reading images from the IBM simulator results and describing them according to the user’s expertise level.

To summarize, as the GPT itself acknowledged, this new version possesses the ability to: “...discuss a wide array of quantum computing topics and algorithms. I understand quantum parallelism, the nuances of quantum mechanics, the behavior of photon’s polarization, and the implications of quantum measurements. I’m familiar with qubits, quantum key distribution, multi-qubit systems, and the intricacies of quantum gates and transformations. I can delve into advanced quantum algorithms like Shor’s and Grover’s, explain quantum Fourier Transforms, and discuss quantum error correction. Moreover, I recognize the need for innovative programming techniques to exploit the full power of quantum computers. Whether you’re curious about the fundamentals or advanced aspects of quantum computing, I’m here to guide and provide insights.”

Before presenting it to users, a preliminary personal iteration was conducted, particularly focusing on the potential prompts of the test. It was decided to iterate a data cluster algorithm based on the application of quantum computers for machine learning. This test evaluated all the desired features of the GPT:

- the creation of code based on a clear explanation of the steps,
- the ability to read images and explain them at different levels,
- the capacity to create examples for clearer understanding,
- the ability to generate images for better comprehension.

After this initial independent test, GPT-V2 was compared with a standard GPT in the second user test.

b. Link Quantum Buddy

<https://chat.openai.com/g/g-cRO4qaAyH-quantum-buddy>

c. Data in the GPT (Next page)

IBM quantum lab infos:

Libro – selezionato solo alcune parti che mi interessavano

The IBM Quantum computer: The reason why it could be important to know the physical gate set is that some user-programmed gates may need to be decomposed into multiple physical gates, and hence could lead to a longer physical algorithm.

when implementing a quantum algorithm it is important to consider the sources of noise in the computer. The two main sources of noise are typically gate infidelity and decoherence. Gate infidelity refers to the fact that the user-specified gates do not precisely correspond to the physically implemented gates. Decoherence refers to the fact that gradually over time the quantum computer loses its “quantumness” and behaves more like a classical object.

Programming the IBM quantum computer: Qiskit library. Qiskit [4] is an open-source quantum computing library developed under the aegis of IBM. Qiskit allows users to write and run programs on either IBM’s quantum processors or on a local simulator, without the use of the graphical interface. → Python library.

Quantum mechanics for non-physics

Intro: Classically, the time it takes to do certain computations can be decreased by using parallel processors. To achieve an

exponential decrease in time requires an exponential increase in the number of processors, and hence an exponential increase in the amount of physical space needed → Thus, an exponential increase in parallelism requires only a linear increase in the amount of physical space needed. This effect is called quantum parallelism. Two techniques to manipulate this is: 1) common property of all of the output values such as the symmetry or period of a function can be read off. 2) transforms the quantum state to increase the likelihood that output of interest will be read.

Quantum mechanics: Quantum mechanical phenomena are difficult to understand, since most of our everyday experiences are not applicable. Quantum mechanics is a theory in the mathematical sense: it is governed by a set of axioms.

photon’s polarization state can be modelled by a unit vector pointing in the appropriate direction. Any arbitrary polarization can be expressed as a linear combination of the two basis vectors. // any device measuring a two-dimensional system has an associated orthonormal basis with respect to which the quantum measurement takes place.

Measurement of the quantum state will change the state to the result of the measurement furthermore will change that state, and it is not possible to determine what the original state was.

State spaces and bra/ket notation: The state space of a quantum system, consisting of the positions, momentums, polarizations, spins, and so on of the various particles, is modelled by a Hilbert space of wave function.

Quantum bits: A quantum bit, or qubit, is a unit vector in a two-dimensional complex vector space for which a particular basis, denoted by 0 , 1 , has been fixed. Even though a quantum bit can be put in infinitely many superposition states, it is only possible to extract a single classical bit’s worth of information from a single quantum bit. As every measurement can result in only one of two states, one of the basis vectors associated to the given measuring device, so, just as in the classical case, there are only two possible results.

Quantum key distribution: Sequences of single qubits can be used to transmit private keys on insecure channels. Thus any eavesdropper on the quantum channel is bound to introduce a high error rate that Alice and Bob can detect by communicating a sufficient number of parity bits of their keys over the open channel.

Multiple qubits: A surprising and unintuitive aspect of the state space of an n -particle quantum system is that the state of the system cannot always be described in terms of the state of its component pieces. However, in a quantum system the resulting state space is much larger; a system of n qubits has a state space of 2^n dimensions. Quantum states, combine through the tensor product. These states represent situations that have no classical counterpart and for which we have no intuition. These are also the states that provide the exponential growth of quantum state spaces with the number of particles.

Measurements: multibit measurement can be treated as a series of single-bit measurements in the standard basis. Any quantum state can be written as the sum of two vectors, one in each of the subspaces.

Quantum gates: The dynamics of a quantum system, when not being measured, are governed by Schrodinger's equation; the dynamics must take states to states in a way that preserves orthogonality. Any linear transformation on a complex vector space can be described by a matrix. Any unitary transformation of a quantum state space is a legitimate quantum transformation, and vice versa. → quantum transformations are unitary is that they are reversible. Thus quantum gates must be reversible.

Simple quantum gates: transformations are fully specified by their effect on the basis vectors. The names of these transformations are conventional. I is the identity transformation, X is negation, Z is a phase shift operation, and $Y = ZX$ is a combination of both. Then the paper talks about the C-not open gate.



Quantum computers: While the classical NOT gate is reversible, AND, OR, and NAND gates are not. Thus it is not obvious that quantum transformations can carry out all classical computations.

Quantum gates array

Quantum parallelism

Shor's algorithm

The Quantum Fourier Transform

Search problem

Quantum error connection

Conclusion: Quantum computing is a new, emerging field that has the potential to dramatically change the way we think about computation, programming, and complexity. The challenge for computer scientists and others is to develop new programming techniques appropriate for quantum computers. new dimension to computation. Programming no longer consists

of merely formulating step-by-step algorithms but requires new techniques of adjusting phases and mixing and diffusing amplitudes to extract useful output. But the real power of these new machines, the exponential parallelism, can only be exploited using new, innovative programming techniques. People have only recently begun to research such techniques. Beyond these algorithms not much more is known about what could be done with a practical quantum computer

Various Algorithms

- Quantum Approximate Optimisation Algorithm (QAOA)
- Quantum Amplitude Estimation (QAE)
- Quantum (enhanced) Support Vector Machine (QSVM)
- Variational quantum eigensolver (VQE)
- Quantum Neural Networks (QNN)
- Quantum phase estimation (QPE) / quantum eigenvalue estimation
- Quantum Annealing (QA)
- Quantum (enhanced) Machine Learning (QML)
- Variational Quantum Classifier (VQC)
- Q-Means
- Shor's algorithm
- Grover's algorithm
- HHL (Harrow-Hassidim-Lloyd)
- Quantum-assisted Reinforcement Learning (QaRL)
- Quantum Singular Value Decomposition (QSVD)
- Quantum Principal Component Analysis (QPCA)
- Quantum Finite Volume Method (QFVM)
- Variational Quantum Linear Solver (VQLS)
- Quantum Convolutional Neural Networks (QCNN)
- Quantum Risk Analysis (QRA)
- Quantum Monte Carlo (QMC)
- Quantum Decision Tree (QDT)
- Quantum Bayesian Networks (QBN)
- Quantum Extreme Value Searching (QES)
- Quantum-inspired Particle Swarm Optimisation (QPSO)
- Quantum K-Means Clustering
- Quantum Distance Classification (QDC)

Developing new drug – code

```
pip install qiskit
```

```
from qiskit import QuantumCircuit, Aer, execute
```

```
# Create a Quantum Circuit acting on a quantum register of three qubits
```

```
circuit = QuantumCircuit(3)
```

```
# Add a Hadamard gate on qubit 0, putting this qubit in superposition.
```

```
circuit.h(0)
```

```
# Add a CX (CNOT) gate on control qubit 0 and target qubit 1, creating entanglement.
```

```
circuit.cx(0, 1)
```

```
# Add another CX gate on control qubit 0 and target qubit 2, creating more entanglement.
```

```
circuit.cx(0, 2)

# Visualize the circuit

print(circuit.draw(output='text'))

# You can also use 'mpl' for a matplotlib drawing, or 'latex' for a LaTeX-based drawing.
```

New Material – code

```
pip install qiskit

from qiskit import QuantumCircuit, Aer, execute

# Create a Quantum Circuit acting on a quantum register of one qubit
circuit = QuantumCircuit(1)

# Apply a Hadamard gate, which puts the qubit in a superposition state
circuit.h(0)

# Measure the qubit
circuit.measure_all()

# Use Aer's qasm_simulator
simulator = Aer.get_backend('qasm_simulator')

# Execute the circuit on the qasm simulator
job = execute(circuit, simulator, shots=1000)

# Grab the results from the job
result = job.result()

# Print the result
counts = result.get_counts(circuit)

print("Counts:", counts)
```

Lines – code (quantum creative code)

```
from qiskit import QuantumCircuit, execute, Aer

import numpy as np

import matplotlib.pyplot as plt

qc = QuantumCircuit(3)

qc.h(range(3))
```

```

backend = Aer.get_backend('statevector_simulator')
result = execute(qc, backend).result()
statevector = result.get_statevector()
real_parts = [np.real(amp) for amp in statevector]
plt.figure(figsize=(10, 6))
plt.bar(range(len(real_parts)), real_parts)
plt.xlabel('Quantum State')
plt.ylabel('Real Part of Amplitude')
plt.title('Quantum Randomness Visualization')
plt.show()

# ... [previous code for setting up the quantum circuit and getting the statevector]
# Select a colormap with warm colors
colormap = plt.cm.autumn
# Plot the bar graph with the chosen colormap
plt.figure(figsize=(10, 6))
bars = plt.bar(range(len(real_parts)), real_parts, color=colormap(np.linspace(0, 1, len(real_parts))))
# Add random lines for an abstract painting effect
num_lines = 20 # Number of lines to add
for _ in range(num_lines):
    start_point = np.random.rand(2) * len(real_parts) # Random start point
    end_point = np.random.rand(2) * len(real_parts) # Random end point
    line_color = np.random.choice(['darkred', 'brown']) # Random line color
    plt.plot([start_point[0], end_point[0]], [start_point[1], end_point[1]], color=line_color,
    linewidth=np.random.rand() * 2)
plt.xlabel('Quantum State')
plt.ylabel('Real Part of Amplitude')
plt.title('Quantum Randomness as Abstract Art')
plt.show()

```

P5 test – code (Grover algorithm)

```
from qiskit import QuantumCircuit, Aer, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import QasmSimulator

nqubits=5

oracle = QuantumCircuit(5)
oracle.x(2)
oracle.mct([0,1,2,3], 4)
oracle.x(2)

def diffuser(nqubits):
    qc = QuantumCircuit(nqubits)
    # Apply H and X gates to all qubits
    for qubit in range(nqubits):
        qc.h(qubit)
        qc.x(qubit)
    # Multi-controlled Z gate
    qc.h(nqubits-1)
    qc.mct(list(range(nqubits-1)), nqubits-1) # multi-controlled-toffoli
    qc.h(nqubits-1)
    # Apply X and H gates to all qubits again
    for qubit in range(nqubits):
        qc.x(qubit)
        qc.h(qubit)
    return qc

grover_circuit = QuantumCircuit(5, 5)
grover_circuit.h([0, 1, 2, 3])
grover_circuit.append(oracle, [0, 1, 2, 3, 4])
grover_circuit.append(diffuser(4), [0, 1, 2, 3])
grover_circuit.measure([0, 1, 2, 3], [0, 1, 2, 3])
```

IO PROV – code

```
!pip install qiskit
```

```
!pip install qiskit_algorithms
```

```
!pip install qiskit==0.24.0
```

```
import qiskit
```

```
print(qiskit.__version__)
```

[VII] User Test 2

a. Miroboard Test (template)

Prompting and Coding

Means giving a computer a specific question or task and then using a language it understands to tell it exactly how to respond or act.

Iteration

Iterating in code means repeatedly running a set of instructions until a certain condition is met, like going through every song on your playlist one by one.

Evaluation

Evaluating means understanding what you create in the processes and how you feel about it.

Your set of emotions

😊 😐 😄 😭 🧡 🙄 🤔 !!

😭 🙄 😐 😄 🧡 🙄 ? 😊

Timeline #1

Prompt and coding

Iteration

Timeline #2

Prompt and coding

Iteration

Evaluation:

Answer the following statements with yes or no:

Does 'MyGPT' for quantum computers respond significantly faster to complex computational tasks compared to the regular ChatGPT?

☐ YES

☐ NO

Do you find that 'MyGPT' provides more accurate and detailed responses than the standard ChatGPT when dealing with highly technical or scientific queries?

☐ YES

☐ NO

In your experience, does 'MyGPT' demonstrate a noticeable improvement in understanding and generating responses for quantum computing-related topics over regular ChatGPT?

☐ YES

☐ NO

Have you noticed any unique capabilities or features in 'MyGPT' that are not present in the standard ChatGPT, particularly in the context of leveraging quantum computing advantages?

☐ YES

☐ NO

How would you rate the normal GPT from 1-5 (1=not good, 5=Very good)

☐

☐

☐

☐

☐

How would you rate the The quantumGPT from 1-5 (1=not good, 5=Very good)

☐

☐

☐

☐

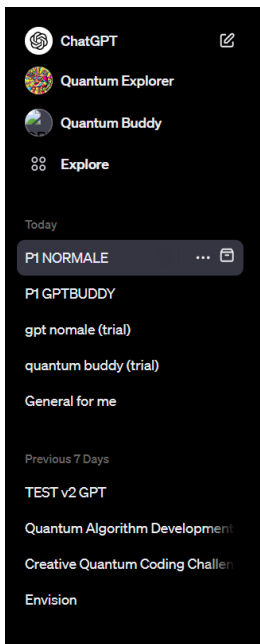
☐

b. Miroboard Results

https://miro.com/welcomeonboard/VkNjU3B0bDhYdmpzQzU5NFlUODR6eXlUdEN5N-k14WFh3emxTSDNzOV05TFBtSGhJUGZUQzBHYNREalpCdGRPaHwzNDU4NzY0NTE4M-zE0MjA1NTE5fDI=?share_link_id=641786781195

c. Participants results (Next Page)

P1



ChatGPT 4



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.



ChatGPT

Great! You're about to embark on an exciting journey to conceptualize, develop, and refine an algorithm tailored for quantum computing. The aim of this test is to blend your background in quantum mechanics and coding to create something truly innovative. We'll leverage the unique capabilities of quantum computing to address complex problems.

Are you ready to start this amazing journey? Let me know when you are, and we'll dive into the first stage! 🤖🚀



You

i am ready



ChatGPT

Awesome! Let's get started with the **Conceptualization** stage. 🌟

1. **Choose a Field:** First, think about the area you want to focus on. Here are some options:
 - **Cryptography:** Explore quantum-resistant algorithms or new ways of securing data.
 - **Machine Learning:** Investigate quantum algorithms for data analysis, pattern recognition, or AI development. ↓
 - **Chemical Simulations:** Use quantum computing for simulating molecular structures or reactions, which could be groundbreaking in materials science or medicine.
 - **Optimization Problems:** Tackle complex optimization issues, like traffic flow optimization or financial modeling.
2. **Understand the Problem:** Once you've chosen a field, dive deep into understanding a specific problem within that area. Ask yourself: What are the current challenges? How can quantum computing provide a unique solution?

Take your time to think about which field excites you the most and the specific problem you wish to solve. Let me know your choice and the problem you're interested in, and then we can move to the next step! 💡🤖



You

i'd like to explore the field of cryptography a bit more, maybe by looking into the prime deciphering algorithm



ChatGPT

That's a fantastic choice! Exploring cryptography through quantum computing, especially focusing on algorithms for prime deciphering, is a cutting-edge area with significant implications. 🧠💡

Development Stage

Now, let's move on to the **Development** stage. Here's how you can start:

1. **Research Existing Algorithms:** Begin by understanding current prime deciphering algorithms and how they work. Familiarize yourself with classical approaches like RSA (Rivest-Shamir-Adleman) and how they rely on the difficulty of prime factorization.
2. **Quantum Computing Approach:** Investigate how quantum algorithms, such as Shor's algorithm, can be used for prime factorization. Shor's algorithm, in particular, can

factorize numbers exponentially faster than the best-known classical algorithms.

3. **Start Coding:** With guidance from me, start translating your understanding into a working code. If you're not a seasoned coder, no worries! I'll help you turn your quantum concepts into a practical algorithm. You can use programming languages and frameworks suited for quantum computing, like Qiskit for IBM's quantum computers.

Remember, the goal is to create an algorithm that takes advantage of quantum computing's capabilities to solve the prime deciphering problem more efficiently than classical computers.

Once you've got a basic version of your code, let me know! We'll then proceed to the **Refinement** stage, where you'll test and fine-tune your algorithm. 🧑🌐

Ready to dive into the development? Let me know if you need help with any specific part of the coding process! 🚀🧑



You

can you explain to me in simple terms how the RSA algorithm works and how the quantum computer breaks this algorithm?



ChatGPT

Absolutely! Let's break it down into simpler terms. 🌟

How RSA Algorithm Works

1. Key Generation:

- **Prime Numbers:** RSA starts by selecting two large prime numbers, let's call them p and q .
- **Compute n :** Multiply p and q to get n . This n is used as the modulus for both the public and private keys.
- **Euler's Totient Function $\phi(n)$:** Calculate $\phi(n) = (p - 1)(q - 1)$.
- **Public Key:** Choose a number e that is coprime with $\phi(n)$ and less than $\phi(n)$. The public key is then the pair (e, n) .
- **Private Key:** Calculate d such that d is the modular multiplicative inverse of e modulo $\phi(n)$. The private key is (d, n) .

2. Encryption:

- To encrypt a message M , convert M into a number m smaller than n and compute the ciphertext c as $c = m^e \mod n$.

3. Decryption:

- To decrypt c , compute $m = c^d \mod n$. Convert m back to the original message M .

The security of RSA relies on the fact that with large enough primes, factoring n back into p and q is extremely hard for classical computers.

How Quantum Computers Break RSA

1. Shor's Algorithm:

- Quantum computers can use Shor's algorithm to efficiently factorize large numbers. This means they can find the prime factors p and q of n much faster than classical computers.

2. Breaking RSA:

- Once a quantum computer factors n into p and q , it can easily compute the private key (d, n) using the same method as the key generation.
- With the private key, the quantum computer can decrypt any message encrypted with the corresponding public key.

In essence, the quantum computer breaks RSA by turning the hardest part of the process (factoring large numbers) into a much easier task, something that's currently not feasible with classical computers. This is why there's a growing interest in developing quantum-resistant encryption algorithms. 🚀🧑



You

what would you need in a quantum-resistant algorithm



ChatGPT

Developing a quantum-resistant algorithm, also known as post-quantum cryptography, involves creating cryptographic systems that are secure against both quantum and classical computers. Here are some key requirements and considerations for a quantum-resistant algorithm:

1. Resistance to Quantum Attacks:

- **Shor's Algorithm Immunity:** The algorithm must be secure against quantum

algorithms like Shor's, which can quickly factor large numbers and compute discrete logarithms.

- **Grover's Algorithm Mitigation:** It should also consider Grover's algorithm, which can speed up brute-force searches, though its impact is less drastic (it provides a quadratic speedup, meaning the key sizes might need to be doubled).

2. **Scalability and Efficiency:**

- The algorithm should be efficient enough to be implemented in current systems without requiring excessive computational resources.
- It should scale well, meaning it should remain secure even as computing power increases.

3. **Key and Signature Size:**

- Some quantum-resistant algorithms have larger key and signature sizes compared to traditional algorithms. Balancing security with practical size constraints is important.

4. **Versatility and Adaptability:**

- The algorithm should be versatile enough to be used in a variety of applications and adaptable to different security requirements.

5. **Standardization and Scrutiny:**

- The algorithm should undergo thorough scrutiny by the cryptographic community and ideally be standardized by organizations like NIST (National Institute of Standards and Technology).
- Continuous evaluation against new quantum and classical cryptanalytic techniques is crucial.

6. **Backward Compatibility:**

- Ideally, it should be compatible with existing infrastructure to ease the transition from current cryptographic systems.

Examples of approaches being explored for quantum-resistant algorithms include lattice-based cryptography, hash-based cryptography, code-based cryptography, and multivariate polynomial cryptography. Each of these has its own strengths and challenges in terms of security, efficiency, and implementation complexity. 🌐 🔍 📄



You

do there exist any algorithms which may have all 6 of these properties?



ChatGPT

As of my last update in April 2023, there were several promising candidates for quantum-resistant algorithms, but none that definitively met all six criteria perfectly. The field of post-quantum cryptography is still evolving, with various algorithms showing potential in different areas. Here's a brief overview:

1. **Lattice-Based Cryptography:**

- **Strengths:** Offers strong security assumptions and is efficient in terms of computational resources.
- **Challenges:** Key sizes can be larger than traditional algorithms, impacting efficiency.

2. **Hash-Based Cryptography:**

- **Strengths:** Well-established security based on the hardness of hash functions; efficient with reasonable key sizes.
- **Challenges:** Mainly used for digital signatures and not as versatile for other cryptographic applications like encryption.

3. **Code-Based Cryptography:**

- **Strengths:** Long-standing security assumptions; efficient for certain operations.
- **Challenges:** Large key sizes, which could be a barrier to widespread adoption.

4. **Multivariate Polynomial Cryptography:**

- **Strengths:** Potentially smaller key sizes and efficient operations.
- **Challenges:** Less mature in terms of research and scrutiny compared to other methods.

5. **Isogeny-Based Cryptography:**

- **Strengths:** Promising approach with smaller key sizes.
- **Challenges:** Relatively new, requiring more research and scrutiny for security and efficiency.

Each of these candidates has different trade-offs. The ongoing efforts by organizations like NIST in their Post-Quantum Cryptography Standardization project aim to identify algorithms that best meet the criteria for security, efficiency, and practicality. The final selection of quantum-resistant algorithms will likely include a mix of different approaches to cater to various needs and applications.

It's important to note that the quest for a "perfect" algorithm may be ongoing, as the field will continue to evolve in response to both advancements in quantum computing and new cryptographic research. 🌐 🏠 🔍



You

what would a simple Lattice-Based Cryptography algorithm look like?



ChatGPT

Lattice-based cryptography is a type of cryptography that uses the mathematical concept of lattices. Lattices are grids of points in multi-dimensional space and have complex structures that are hard to decipher without the correct key, making them suitable for secure communications. A simple and popular example of a lattice-based cryptographic algorithm is the Learning With Errors (LWE) problem.

Here's a basic outline of how a simple Lattice-Based Cryptography algorithm, like LWE, works:

The Learning With Errors (LWE) Problem

- Setup:**
 - Choose a large prime number p .
 - Define a lattice in n -dimensional space.
 - Select a random vector s in n -dimensional space (this is the secret key).
- Encryption:**
 - To encrypt a message m (where m is a small number compared to p), first encode it as a number.
 - Generate a random vector a and a small random error term e (the error is crucial as it introduces the 'error' in Learning With Errors).
 - Calculate the ciphertext as $c = \langle a, s \rangle + e + m \mod p$, where $\langle a, s \rangle$ denotes the dot product of a and s .
- Decryption:**
 - To decrypt the ciphertext c , compute $c - \langle a, s \rangle \mod p$, which is approximately $e + m$.
 - Since e is small, the original message m can be recovered by rounding to the nearest multiple of e .

Characteristics

- Security:** The security of LWE is based on the difficulty of solving the LWE problem, which involves finding s given many pairs of a and c . This problem is believed to be hard for both classical and quantum computers.
- Efficiency:** Lattice-based algorithms can be efficient, but often require careful implementation to manage the size of the vectors and the computations involved.
- Applications:** LWE and its variants are used in various cryptographic constructions, including encryption, digital signatures, and key exchange protocols.

Summary

The strength of lattice-based cryptography lies in the complexity of the lattice structure and the use of errors, making it a promising candidate for quantum-resistant algorithms. However, it's important to note that practical implementations require careful optimization to balance security, efficiency, and key sizes.



You

make a LWE algorithm for $2 < n < 6$ suitable for the IBM quantum simulator



ChatGPT

Creating a simple Learning With Errors (LWE) algorithm suitable for simulation on an IBM quantum simulator involves several steps. For the sake of simplicity and demonstration, let's design an LWE algorithm for a small dimension n where $2 < n < 6$. We'll choose $n = 4$ for this example.

Step-by-Step LWE Algorithm Design

- Parameters Setup**
 - Prime Number p :** Choose a small prime number, say $p = 7$.
 - Dimension n :** Set $n = 4$.
 - Error Range:** Define a small range for error values, say -1 to 1 .
- Secret Key Generation**
 - Generate a secret key vector s of dimension n with small integer values. For example, $s = [s_1, s_2, s_3, s_4]$.
- Encryption Process**
 - To encrypt a small message m (ensure $m < p$):
 - Generate a random vector $a = [a_1, a_2, a_3, a_4]$ with each element being a small integer.

- Select a small random error term e from the error range.
- Compute the ciphertext c as $c = \langle a, s \rangle + e + m \mod p$.

4. Decryption Process

- To decrypt the ciphertext c :
 - Compute $c - \langle a, s \rangle \mod p$.
 - The result will be close to $e + m$. Since e is small, rounding to the nearest multiple of e recovers m .

Implementation Considerations for IBM Quantum Simulator

- IBM's quantum simulators, such as Qiskit, primarily simulate quantum circuits. LWE is a classical algorithm, but you can explore quantum algorithms for solving LWE problems.
- You can create a quantum circuit that demonstrates the hardness of solving LWE problems for a quantum computer.
- Implementing such a demonstration would involve setting up a quantum circuit that represents the LWE problem, then attempting to solve it using quantum algorithms.

Example Code (Pseudocode)

```
python
import random

# Parameters
p = 7
n = 4
error_range = [-1, 0, 1]

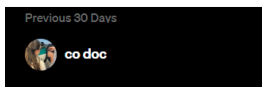
# Secret Key
s = [random.randint(0, p-1) for _ in range(n)]

# Encrypt Function
def encrypt(m, s, p, n):
    a = [random.randint(0, p-1) for _ in range(n)]
    e = random.choice(error_range)
    c = sum([a[i]*s[i] for i in range(n)]) + e + m % p
    return c, a

# Decrypt Function
def decrypt(c, a, s, p):
    m_approx = c - sum([a[i]*s[i] for i in range(n)]) % p
    return m_approx

# Example Usage
m = 3 # Example message
ciphertext, a = encrypt(m, s, p, n)
decrypted_message = decrypt(ciphertext, a, s, p)
```

This pseudocode provides a basic framework. For a full implementation, especially one that interfaces with a quantum simulator like Qiskit, you would need to integrate the classical LWE setup into a quantum computing context, focusing on the challenge of solving the LWE problem with a quantum approach.



Message ChatGPT...



ChatGPT can make mistakes. Consider checking important information.



```
In [1]: import random

# Parameters
p = 7
n = 4
error_range = [-1, 0, 1]

# Secret Key
s = [random.randint(0, p-1) for _ in range(n)]

# Encrypt Function
def encrypt(m, s, p, n):
    a = [random.randint(0, p-1) for _ in range(n)]
    e = random.choice(error_range)
    c = sum([a[i]*s[i] for i in range(n)]) + e + m % p
    return c, a

# Decrypt Function
def decrypt(c, a, s, p):
    m_approx = c - sum([a[i]*s[i] for i in range(n)]) % p
    return m_approx

# Example Usage
m = 2# Example message
ciphertext, a = encrypt(m, s, p, n)
decrypted_message = decrypt(ciphertext, a, s, p)
```

```
In [ ]:
```


ChatGPT

Quantum Explorer

Quantum Buddy

Explore

Today

P1 NORMALE

P1 GPTBUDDY

gpt normale (trial)

quantum buddy (trial)

General for me

Previous 7 Days

TEST v2 GPT

Quantum Algorithm Development

Creative Quantum Coding Challenge

Envision



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.

Quantum Buddy

Great, let's get started with the structured user test! 🚀

↓

The aim of this test is to guide you through the process of conceptualizing, developing, and refining a quantum algorithm. This journey will harness your knowledge in quantum mechanics and coding, and it's an exciting opportunity to explore the potential of quantum computing. The test is designed to take your theoretical understanding and turn it into a practical, functioning quantum algorithm. 🗨️💻

Are you ready to dive into the world of quantum computing and begin this fascinating journey? Let me know when you're ready to start! 🍌

You

i am ready for the journey, i find cryptography an interesting field, so i would like to explore that

Quantum Buddy

Fantastic choice! Cryptography is a thrilling and highly relevant field for quantum computing. ☀️

Step 1: Conceptualization 🗨️

In this stage, your task is to delve into the realm of quantum cryptography. Here's how you can start:

- 1. Identify a Specific Cryptography Challenge:** Think about a particular problem within cryptography that you find intriguing. This could be anything from quantum key distribution (QKD) to breaking classical cryptographic algorithms using quantum techniques.
- 2. Research the Current Landscape:** Understand the current state of quantum cryptography. What are the latest developments? How are quantum computers changing the field?
- 3. Formulate Your Goal:** Define what you want your algorithm to achieve. Do you want to improve upon existing quantum cryptographic methods, or explore new ways quantum computing can be applied in cryptography?
- 4. Consider Quantum Advantages:** Reflect on how quantum properties like superposition and entanglement can offer advantages over classical approaches in solving your chosen problem.

Once you've got a clear idea of the problem you're tackling and how you plan to approach it, let me know! We'll then move on to the exciting part: developing your algorithm! 🚀💻



You

i want to look into how to make a LWE algorithm for quantum computers

Quantum Buddy

That's an excellent and challenging choice! Working on a Learning With Errors (LWE) algorithm for quantum computers is at the forefront of combining quantum computing and cryptography. 🌐🔒

Step 2: Development 🚀

Understanding LWE and Quantum Computing

- **LWE Overview:** Learning With Errors is a problem in cryptography that forms the basis for various cryptographic systems. It's based on the difficulty of solving linear equations that have been perturbed by some error.
- **Quantum Computing Angle:** Quantum computers, due to their ability to handle superpositions, might be able to solve LWE problems more efficiently than classical computers. This makes it a hot topic for quantum-resistant cryptography.

Developing Your Algorithm



1. **Familiarize with Qiskit:** Since you're working with IBM quantum computers, Qiskit is the tool of choice. It's a Python-based framework that allows you to create quantum algorithms and run them on simulators and actual quantum computers.
2. **Algorithm Design:** Start by outlining your algorithm's structure. How will it interact with LWE? What quantum operations are necessary to tackle the LWE problem?
3. **Coding the Algorithm:** Begin coding in Qiskit. Start with the basic structure and gradually integrate the elements specific to solving LWE. Remember, it's okay if you're not an expert coder; I'm here to help translate your quantum understanding into working code.
4. **Test on Simulator:** Initially, run your algorithm on a quantum simulator. This allows for debugging and initial validation of your approach without the need for actual quantum hardware.

Feel free to ask for help at any stage of your coding process. Whether it's a syntax question or a conceptual doubt about quantum operations, I'm here to assist!

Once you have a draft of your code, we can move to the next stage of testing and refining your algorithm. Let me know when you're ready for that, or if you need any assistance with the development stage!



You

make an algorithm that uses LWE for a quantum computer



Quantum Buddy

Creating a quantum algorithm to tackle the Learning With Errors (LWE) problem is a complex task, but I'll guide you through a conceptual outline. Since the implementation of such an algorithm requires a deep understanding of both quantum computing principles and the intricacies of the LWE problem, we'll focus on a high-level approach.

Quantum Algorithm for LWE: Conceptual Outline

1. **Initialization:**
 - **Quantum Registers:** Initialize quantum registers to store qubits that will represent the data involved in the LWE problem.
 - **Superposition:** Apply Hadamard gates to some of the qubits to create superpositions, enabling the exploration of multiple possibilities simultaneously.
2. **Encoding the LWE Problem:**
 - **LWE Instance:** Encode an instance of the LWE problem onto the quantum registers. This involves representing the linear equations and the error terms in a quantum-compatible format.
 - **Quantum Encoding:** Utilize quantum gates to encode the coefficients of the linear equations and the error terms onto the qubits.
3. **Quantum Processing:**
 - **Entanglement:** Create entanglement between qubits to capture the complex relationships inherent in the LWE problem.
 - **Problem-Specific Operations:** Implement quantum gates that correspond to the mathematical operations needed to solve the LWE equations under the presence of errors.
4. **Quantum Fourier Transform (QFT):**
 - Apply the Quantum Fourier Transform to translate the quantum information into a form where the solution to the LWE problem can be more easily extracted.
5. **Measurement and Interpretation:**
 - **Collapse the Wavefunction:** Measure the qubits, which collapses the wavefunction, leading to a specific outcome that corresponds to a potential solution.
 - **Interpretation:** Translate the measurement results back into the context of the LWE problem. This might involve classical post-processing to interpret the quantum results.
6. **Iteration and Optimization:**
 - **Iterative Approach:** The algorithm may need to be run multiple times, with adjustments to the encoding or quantum operations, to reliably solve the LWE problem.
 - **Optimization:** Analyze the outcomes to optimize the quantum operations for better accuracy or efficiency.




Implementation in Qiskit

The actual coding of this algorithm in Qiskit would involve detailed work with quantum gates, measurement protocols, and potentially error correction techniques. The Qiskit library offers the tools needed to implement these steps, but writing the code requires a deep dive into both the library's functionalities and the specifics of the LWE problem.

Moving Forward

Given the complexity and the cutting-edge nature of this task, the actual coding of such an algorithm is a significant undertaking and might require iterative development and testing, potentially with input from experts in both quantum computing and cryptography.

If you have specific parts of the algorithm or concepts that you want to discuss in more detail or need help with, feel free to ask! 



You

make code for the initialization step in the previous message, one that is compatible with the IBM quantum simulator



Quantum Buddy

Sure, I can provide a sample code snippet for the initialization step of a quantum algorithm, compatible with IBM's quantum simulator, using the Qiskit framework. This code will focus on initializing quantum registers and creating superpositions, which are fundamental steps in many quantum algorithms.

Before you run this code, ensure you have Qiskit installed. You can install it using pip if you haven't already:

bash

 Copy code

```
pip install qiskit
```

Here's the Python code for the initialization:  p:

python

 Copy code

```
from qiskit import QuantumCircuit, Aer, execute
from qiskit.visualization import plot_histogram

# Define the number of qubits
num_qubits = 5 # You can adjust this number based on your algorithm's

# Create a Quantum Circuit acting on a quantum register of num_qubits
circuit = QuantumCircuit(num_qubits)

# Apply Hadamard gates to all qubits to create superpositions
circuit.h(range(num_qubits))

# Visualize the circuit
print("Quantum Circuit:")
print(circuit.draw())

# Use Aer's qasm_simulator
simulator = Aer.get_backend('qasm_simulator')


# Execute the circuit on the qasm simulator
job = execute(circuit, simulator, shots=1000)

# Grab results from the job
result = job.result()

# Returns counts
counts = result.get_counts(circuit)
print("\nCounts:", counts)

# Plot a histogram
plot_histogram(counts)
```

This code does the following:

1. **Initialize a Quantum Circuit:** Creates a quantum circuit with a specified number of qubits.
2. **Apply Hadamard Gates:** Applies Hadamard gates to all qubits, placing them in superposition.
3. **Simulation:** Executes the circuit on a quantum simulator.
4. **Results:** Collects and prints the results, and  also plots a histogram of the outcomes.

When you run this code, you'll see the quantum circuit and the histogram of the measurement outcomes. The histogram should show an equal distribution of outcomes if the Hadamard gates were applied correctly. This forms the basis for more complex quantum operations as part of your algorithm.



You

i got an error,

Traceback (most recent call last):

Cell In[2], line 27

counts = result.get_counts(circuit)

File /opt/conda/lib/python3.10/site-packages/qiskit/result/result.py:289 in

get_counts

raise QiskitError(f'No counts for experiment "{repr(key)}"')

QiskitError: 'No counts for experiment "<qiskit.circuit.quantumcircuit.QuantumCircuit object at 0x7f7bf45e3fd0>"'

can you show me how to fix this?



Quantum Buddy

The error you encountered indicates that there's an issue with retrieving the counts from the result object. This is likely due to the circuit object not being correctly referenced when retrieving the counts. Let's correct this.

In Qiskit, when you execute a quantum circuit and get results, you need to specify the circuit for which you want the counts. This is usually straightforward, but can sometimes lead to errors if not done correctly. Here's the revised code with the fix:

```
python Copy code

from qiskit import QuantumCircuit, Aer, execute
from qiskit.visualization import plot_histogram

# Define the number of qubits
num_qubits = 5

# Create a Quantum Circuit acting on a quantum register of num_qubits
circuit = QuantumCircuit(num_qubits)

# Apply Hadamard gates to all qubits to create superpositions
circuit.h(range(num_qubits))

# Add a measurement to all qubits
circuit.measure_all()

# Visualize the circuit
print("Quantum Circuit:")
print(circuit.draw())

# Use Aer's qasm_simulator
simulator = Aer.get_backend('qasm_simulator')

# Execute the circuit on the qasm simulator
job = execute(circuit, simulator, shots=1000)

# Grab results from the job
result = job.result()

# Returns counts
counts = result.get_counts()
print("\nCounts:", counts)

# Plot a histogram
plot_histogram(counts)
```

The key changes are:

1. **Measurement Step:** I added a line to measure all qubits at the end of the circuit ('`circuit.measure_all()`'). This is necessary to get counts from the qubits.
2. **Retrieving Counts:** I modified '`result.get_counts(circuit)`' to '`result.get_counts()`'. This change ensures that the counts are correctly retrieved from the result object.

This code should now work without the error you encountered. It will apply Hadamard gates to all qubits, measure them, and then plot the histogram of the outcomes.



In [1]: `pip install qiskit`

```
Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.44.3)
Requirement already satisfied: qiskit-terra==0.25.3 in /opt/conda/lib/python3.10/site-packages (from qiskit) (0.25.3)
Requirement already satisfied: symengine<0.10,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (0.9.2)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (4.1.1)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (0.13.0)
Requirement already satisfied: scipy>=1.5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (1.9.3)
Requirement already satisfied: numpy>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (1.23.5)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (1.11.1)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (2.8.2)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (4.5.0)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (5.9.4)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (0.3.7)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (3.11)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra==0.25.3->qiskit) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra==0.25.3->qiskit) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra==0.25.3->qiskit) (1.3.0)
```

[notice] A new release of pip available: 23.1.1 -> 23.3.2

[notice] To update, run: `pip install --upgrade pip`

Note: you may need to restart the kernel to use updated packages.

In [2]: `from qiskit import QuantumCircuit, Aer, execute`
`from qiskit.visualization import plot_histogram`

```
# Define the number of qubits
num_qubits = 5 # You can adjust this number based on your algorithm's needs

# Create a Quantum Circuit acting on a quantum register of num_qubits
circuit = QuantumCircuit(num_qubits)

# Apply Hadamard gates to all qubits to create superpositions
circuit.h(range(num_qubits))

# Visualize the circuit
print("Quantum Circuit:")
print(circuit.draw())

# Use Aer's qasm_simulator
simulator = Aer.get_backend('qasm_simulator')

# Execute the circuit on the qasm simulator
```

```

job = execute(circuit, simulator, shots=1000)

# Grab results from the job
result = job.result()

# Returns counts
counts = result.get_counts(circuit)
print("\nCounts:", counts)

# Plot a histogram
plot_histogram(counts)

```

Quantum Circuit:

Figure(203.885x451.5)

Traceback (most recent call last):

```

Cell In[2], line 27
      counts = result.get_counts(circuit)
      File /opt/conda/lib/python3.10/site-packages/qiskit/result/result.py:289 in get_counts
          raise QiskitError(f'No counts for experiment "{repr(key)}"')
QiskitError: 'No counts for experiment "<qiskit.circuit.quantumcircuit.QuantumCircuit object at 0x7f7bf45e3fd0>"'

```

Use %tb to get the full traceback.

"" target='_blank'> [Search for solution online](#)

In [3]: **from** qiskit **import** QuantumCircuit, Aer, execute
from qiskit.visualization **import** plot_histogram

```

# Define the number of qubits
num_qubits = 5

# Create a Quantum Circuit acting on a quantum register of num_qubits
circuit = QuantumCircuit(num_qubits)

# Apply Hadamard gates to all qubits to create superpositions
circuit.h(range(num_qubits))

# Add a measurement to all qubits
circuit.measure_all()

# Visualize the circuit
print("Quantum Circuit:")
print(circuit.draw())

# Use Aer's qasm_simulator
simulator = Aer.get_backend('qasm_simulator')

# Execute the circuit on the qasm simulator
job = execute(circuit, simulator, shots=1000)

# Grab results from the job
result = job.result()

# Returns counts
counts = result.get_counts()
print("\nCounts:", counts)

```

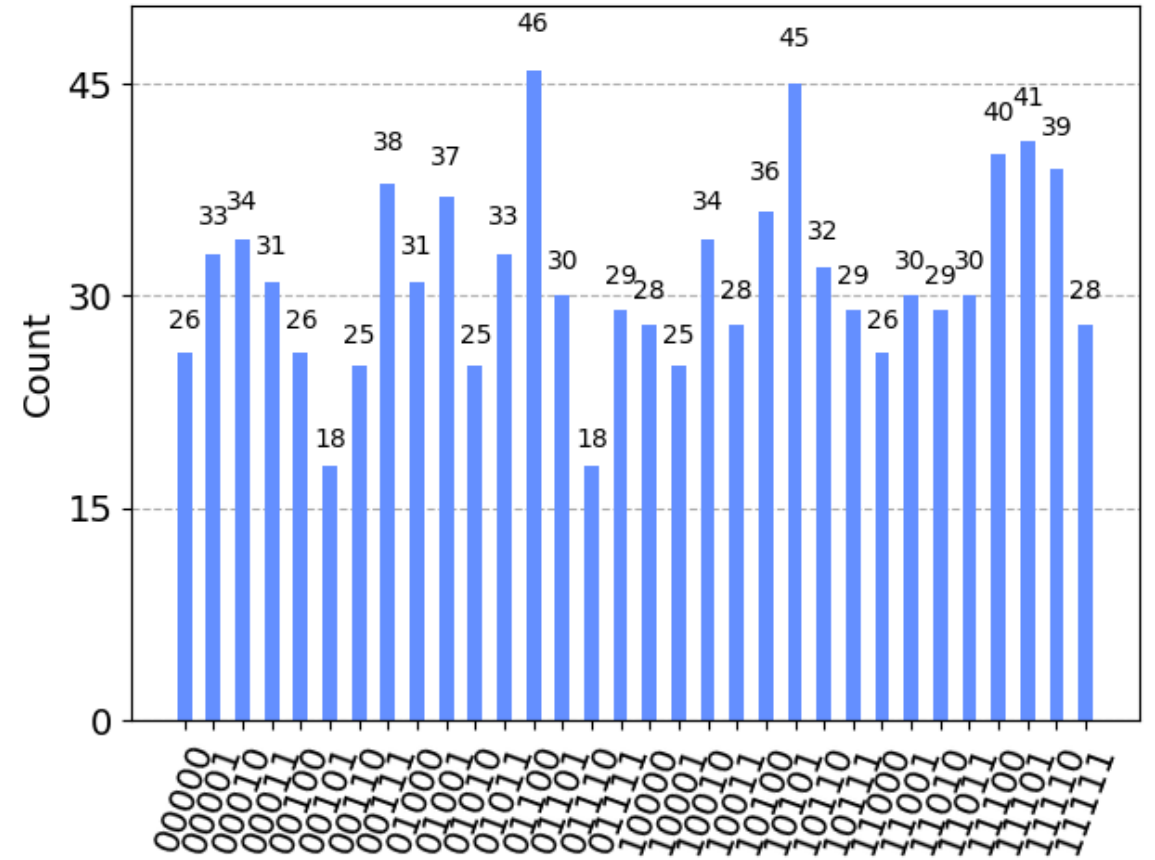


```
# Plot a histogram
plot_histogram(counts)
```

Quantum Circuit:
Figure(747.609x535.111)

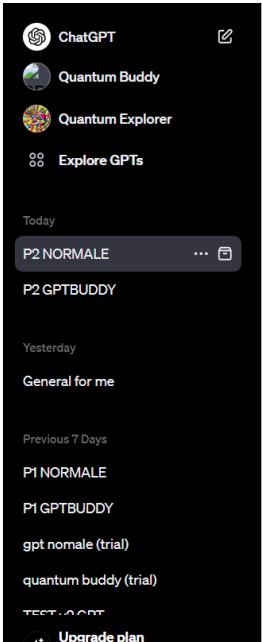
Counts: {'00010': 34, '01010': 25, '10001': 25, '01101': 30, '11000': 26, '10101': 45, '11100': 40, '10010': 34, '11110': 39, '11010': 29, '01001': 37, '00101': 18, '00011': 31, '00111': 38, '10000': 28, '11111': 28, '11011': 30, '01111': 29, '10011': 28, '11101': 41, '01110': 18, '11001': 30, '00000': 26, '10111': 29, '10100': 36, '00100': 26, '01100': 46, '01000': 31, '00001': 33, '00110': 25, '01011': 33, '10110': 32}

Out[3]:



In []:

P2



ChatGPT 4



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.



ChatGPT

Aim of the Test 🎯



Hello, young enthusiast! We're embarking on a fascinating journey into the world of quantum computing and coding. The goal of this test is to challenge and nurture your skills. You'll be conceptualizing, developing, and refining an algorithm that harnesses the power of quantum computing. This isn't just about coding; it's about thinking outside the box and applying your quantum mechanics background to solve complex problems in innovative ways. Are you ready to dive into this quantum adventure? Let me know when you're ready, and we'll get started! 🤖🔥

Conceptualization Stage 💡🌟

Great! Let's begin with the conceptualization stage. Here, your creativity is the key. Think about a field that excites you and identify a unique problem you want to solve using quantum computing. Here are some options to consider:

1. **Cryptography:** Dive into the world of secure communication. Can you develop a quantum algorithm that enhances encryption methods?
2. **Machine Learning:** Explore how quantum computing can speed up or improve machine learning algorithms.
3. **Chemical Simulations:** Quantum computing is ideal for simulating molecular interactions. How can your algorithm contribute to this field?
4. **Optimization Problems:** Tackle complex optimization challenges that are tough for classical computers.



Take your time to think about these fields. Which one sparks your interest the most? Let me know your choice, and we'll proceed to the next step of developing your idea! 🌟💡

Once you've chosen a field and identified the problem you'd like to solve, we'll move on to the development stage. Remember, this is a journey of learning and exploration, so take your time and enjoy the process! 🤖👨🔬👩🔬



You

I like machine learning the most



ChatGPT

Development Stage: Machine Learning with Quantum Computing 🚀💻

Fantastic choice! Machine learning combined with quantum computing offers a realm of exciting possibilities. Now, let's start developing your algorithm. Here's how we'll proceed:

1. **Understand the Basics:** First, ensure you have a good understanding of how quantum computing can enhance machine learning. Think about quantum algorithms that could speed up data processing or improve the accuracy of predictions.
2. **Coding Begins:** Don't worry if you're not a coding pro. I'm here to help! You can start by

outlining the basic structure of your algorithm. What will it do? How will it use quantum computing to improve machine learning?

3. **Programming Tools:** Familiarize yourself with quantum programming tools. IBM's Qiskit is a great place to start. It's a quantum computing SDK that allows you to write quantum algorithms.
4. **Draft Your Algorithm:** Write a preliminary version of your algorithm. Think about the quantum circuits you'll need, the qubits, and how they'll interact.
5. **ChatGPT Assistance:** I'm here to assist with any coding challenges. Feel free to ask questions or seek guidance on quantum programming concepts or syntax.

Remember, this stage is all about bringing your ideas to life. Take your time to experiment and code. Let's start coding your algorithm! Feel free to share your thoughts or ask for help whenever you need. Excited to see what you'll create! 🤖🌟

Once you have a draft of your code, we'll move on to testing and refining it. But for now, let's focus on getting those quantum-enhanced machine learning ideas into code! 🚀👨👩



You

What would be usecases of quantum computing combined with machine learning?



ChatGPT

Quantum computing combined with machine learning opens up a world of fascinating use cases, each unlocking new possibilities that were previously out of reach for classical computing. Here are some compelling use cases:

1. **Drug Discovery and Molecular Modeling:** Quantum computers can simulate molecular and chemical interactions at an unprecedented scale. When combined with machine learning, this can significantly speed up drug discovery, allowing for the rapid analysis of molecule interactions and the prediction of their effects in the body.
2. **Optimized Financial Modeling:** In finance, quantum machine learning can be used for more accurate and faster risk analysis, portfolio optimization, and fraud detection. Quantum algorithms can analyze vast datasets far more efficiently than classical computers, leading to better investment strategies and financial forecasts.
3. **Traffic Optimization and Logistics:** Quantum machine learning can optimize complex systems like traffic flow and logistics operations. By processing and analyzing large volumes of data from various sources, these algorithms can optimize routes, reduce congestion, and improve delivery times.
4. **Advanced Materials Science:** In materials science, quantum computing can help discover new materials by simulating their properties. Machine learning algorithms can then analyze these simulations to predict material behavior under various conditions, accelerating the development of new materials.
5. **Climate Modeling and Environmental Systems:** Quantum computers can process complex environmental data at a scale that's currently impossible. Combined with machine learning, they can improve climate modeling, predict weather patterns more accurately, and aid in understanding ecological systems.
6. **Enhanced Artificial Intelligence (AI):** Quantum computing can potentially solve complex AI problems more efficiently. This includes faster training of machine learning models, handling higher-dimensional data, and solving classification problems that are currently intractable.
7. **Cryptography and Cybersecurity:** In the realm of cybersecurity, quantum machine learning can both pose challenges to current encryption methods and offer new ways to counter quantum threats. It can help in developing new cryptographic algorithms and enhance network security protocols.
8. **Image and Pattern Recognition:** Quantum algorithms can significantly speed up the process of image recognition and pattern detection, leading to advances in medical imaging, facial recognition technologies, and automated systems in various industries.

These use cases represent just the tip of the iceberg. As quantum computing technology matures and integrates more deeply with machine learning, we're likely to see even more innovative applications across various fields. Your journey in developing a quantum machine learning algorithm could contribute to these exciting developments! 🤖🌟🚀



You

I'd like to write an algorithm for financial modeling. could you help me with the first steps?



ChatGPT

Absolutely, diving into financial modeling using quantum computing and machine learning is an exciting endeavor! Here are the first steps to guide you through the process:

1. Define the Financial Problem 🎯

- **Problem Identification:** What specific financial modeling problem are you aiming to solve? Is it portfolio optimization, risk analysis, market prediction, or something else?
- **Data Requirements:** Determine the kind of financial data you need. This might include stock prices, market indices, economic indicators, etc.

2. Understand Quantum Computing Basics for Finance 🌐

- **Quantum Concepts:** Refresh your understanding of key quantum computing concepts like qubits, superposition, entanglement, and quantum gates.
- **Quantum Financial Models:** Explore how quantum computing can be applied in finance. For instance, quantum algorithms can handle complex calculations more efficiently than classical algorithms.

3. Choose a Quantum Computing Framework 🛠️

- **Framework Selection:** Pick a quantum computing framework. IBM's Qiskit is a popular choice and has resources specifically for finance.
- **Installation and Setup:** Install the chosen framework and set up your development environment.

4. Acquire and Preprocess Data 📊

- **Data Collection:** Gather historical financial data relevant to your problem. This can often be sourced from financial APIs or databases.
- **Preprocessing:** Clean and preprocess your data. Ensure it's in a format suitable for quantum analysis, like normalizing values and handling missing data.

5. Develop a Basic Quantum Algorithm 🧠

- **Algorithm Outline:** Start with a simple quantum algorithm. For instance, if you're working on portfolio optimization, you might begin with a basic version of the Quantum Approximate Optimization Algorithm (QAOA).
- **Coding:** Write the initial code for your algorithm. Focus on setting up quantum registers, applying quantum gates, and measuring the output.

6. Integrate Machine Learning (Optional) 🤖

- **ML Integration:** If you're planning to incorporate machine learning, decide how it will interact with your quantum algorithm. For example, machine learning can be used to predict future financial trends, which then feed into the quantum model.

7. Initial Testing and Analysis 🧪

- **Testing:** Run your quantum algorithm with your financial data. Begin with small-scale tests to ensure the basic functionality.
- **Analysis:** Analyze the output. How does it compare to classical financial models? What insights or optimizations does it offer?

8. Seek Assistance As Needed 💬

- **Ask for Help:** Remember, I'm here to help 📩 with coding concepts, quantum mechanics questions, or any other part of the process.

By following these steps, you'll lay a strong foundation for your quantum financial modeling algorithm. Take your time with each stage, and don't hesitate to ask if you need guidance along the way. Happy coding! 🙌👨🔬👩🔬



You

I would like to make a portfolio optimization algorithm, which takes into account expected returns and volatility of 10 assets. I want to use qiskit as the quantum backend. Can you write the qiskit code for this algorithm?



ChatGPT

Creating a portfolio optimization algorithm using Qiskit involves several steps, including setting up the quantum environment, defining the problem, and implementing the quantum algorithm. For a portfolio of 10 assets, considering expected returns and volatility, we will use the Variational Quantum Eigensolver (VQE) approach or the Quantum Approximate Optimization Algorithm (QAOA). These are advanced topics, but I'll guide you through a basic implementation.

Preliminary Steps:



1. **Install Qiskit:** If you haven't already, install Qiskit in your Python environment.

```
python Copy code  
  
pip install qiskit
```

2. **Import Necessary Libraries:**

You'll need to import various components from Qiskit, along with other necessary Python libraries like 'numpy'.

3. **Define Your Financial Data:**

You'll need data for the 10 assets, specifically their expected returns and volatility. This data will be used to create a cost function for the optimization.

4. **Set Up the Quantum Instance:**

Prepare the quantum instance (simulator or real quantum device) where the algorithm will run.

Qiskit Code for Portfolio Optimization:

Please note, due to the complexity and length of the code, I'll outline a high-level structure of how the code might look. This won't be a fully functional script but should give you a good starting point.

```
python Copy code  
  
import numpy as np  
from qiskit import Aer  
from qiskit.algorithms.optimizers import COBYLA  
from qiskit.algorithms import VQE, QAOA  
from qiskit.opflow import PauliSumOp  
from qiskit.utils import QuantumInstance  
from qiskit.finance.applications.ising import portfolio  
from qiskit.finance import QiskitFinanceError  
  
# Define your financial data: expected returns and covariance matrix  
expected_returns = np.array([/* your expected returns data for 10 asset  
covariance_matrix = np.array([/* your 10x10 covariance matrix */])  
  
# Define the parameters for portfolio optimization  
q = 0.5 # risk factor  
budget = 3 # number of assets to include in the portfolio  
penalty = 2 # penalty parameter for the budget constraint  
  
# Create the Ising Hamiltonian  
qubit_op, offset = portfolio.get_operator(expected_returns, covariance_  
  
# Set up the quantum instance (using a simulator here)  
quantum_instance = QuantumInstance(Aer.get_backend('qasm_simulator'))  
  
# Choose the algorithm: VQE or QAOA  
optimizer = COBYLA(maxiter=500)  
vqe = VQE(optimizer=optimizer, quantum_instance=quantum_instance)  
# or for QAOA: qaoa = QAOA(optimizer=optimizer, quantum_instance=quantu  
  
# Solve the problem  
result = vqe.compute_minimum_eigenvalue(qubit_op)  
  
# Analyze the result  
print('Result:', result)
```

Points to Consider:

- **Financial Data:** You need to provide the actual financial data (expected returns and covariance matrix) for the 10 assets.
- **Algorithm Choice:** VQE and QAOA are both used for such optimization problems. The choice depends on your specific requirements and constraints.
- **Parameter Tuning:** The parameters ('q', 'budget', 'penalty', 'optimizer settings') need to be tuned according to your specific problem.
- **Backend Selection:** I've used a simulator backend in this example. For more complex or larger-scale problems, you might consider using an actual quantum computer via IBM Quantum Experience.

This is a basic framework to get you started. Developing a fully functional and efficient portfolio optimization algorithm using quantum computing is a complex task that involves fine-tuning and extensive testing. Don't hesitate to explore more about Qiskit's finance module and other quantum computing resources for deeper insights and optimizations!



You

for the financial data, can you make up this data for 10 assets? As for the algorithm

choice, let's go for the qaoa



ChatGPT

Certainly, I'll generate some mock financial data for 10 assets to use in a portfolio optimization algorithm with QAOA (Quantum Approximate Optimization Algorithm). Remember, this data is fictional and for demonstration purposes only.

Mock Financial Data for 10 Assets:

- **Expected Returns:** Randomly generated annual return rates for 10 assets.
- **Covariance Matrix:** A 10×10 matrix representing the covariance between the returns of these assets. This matrix helps in understanding how the returns of different assets vary with each other.

Generating the Mock Data:

1. **Expected Returns:** Let's assume returns vary between -10% and +30%.
2. **Covariance Matrix:** Random values with some logical constraints to ensure it's a valid covariance matrix (positive semi-definite).

Updated Qiskit Code with QAOA and Mock Data:


The following Python script includes the generation of mock data and sets up the QAOA algorithm for portfolio optimization.


```
python Copy code  
  
import numpy as np  
from qiskit import Aer  
from qiskit.algorithms.optimizers import COBYLA  
from qiskit.algorithms import QAOA  
from qiskit.opflow import PauliSumOp  
from qiskit.utils import QuantumInstance, algorithm_globals  
from qiskit.finance.applications.ising import portfolio  
  
# Generate mock financial data  
np.random.seed(42) # for reproducibility  
num_assets = 10  
expected_returns = np.random.uniform(-0.10, 0.30, num_assets) # between  
covariance_matrix = np.random.rand(num_assets, num_assets)  
covariance_matrix = 0.5 * (covariance_matrix + covariance_matrix.T) #  
covariance_matrix += num_assets * np.eye(num_assets) # Ensure positive  
  
# Portfolio optimization parameters  
q = 0.5 # risk factor  
budget = 3 # number of assets to include in the portfolio  
penalty = num_assets # penalty parameter for the budget constraint  
  
# Create the Ising Hamiltonian  
qubit_op, offset = portfolio.get_hamiltonian(expected_returns, covariance_  
  
# Set up the quantum instance (using a simulator here)  
algorithm_globals.random_seed = 123  
quantum_instance = QuantumInstance(Aer.get_backend('qasm_simulator'), s  
  
# Initialize the QAOA algorithm  
optimizer = COBYLA(maxiter=250)  
qaoa = QAOA(optimizer=optimizer, quantum_instance=quantum_instance)  
  
# Solve the problem using QAOA  
result = qaoa.compute_minimum_eigenvalue(qubit_op)  
  
# Print results  
print('Result:', result)
```

Execution Notes:

- **Random Seed:** `np.random.seed(42)` and `algorithm_globals.random_seed = 123` are used for reproducibility. Different seeds will generate different mock data.
- **Data Validity:** The generated covariance matrix is symmetrized and made positive definite to resemble realistic financial data.
- **QAOA Parameters:** The `COBYLA` optimizer's `maxiter` parameter and other settings in QAOA can be adjusted for better performance.
- **Results Interpretation:** The result will give you the quantum state that minimizes the cost function, which corresponds to an optimal portfolio mix given the constraints.

This code sets up a basic framework for your quantum portfolio optimization using QAOA with Qiskit. Fine-tuning and a deeper understanding of the finance module in Qiskit will be needed for a more robust application.

 **You**
it says there is no module named qiskit.finance

 **ChatGPT**
It seems there has been a change in the Qiskit package structure since my last update. The `'qiskit.finance'` module might have been deprecated or restructured in the latest versions of Qiskit. In this case, you can either use an older version of Qiskit that includes the `'qiskit.finance'` module, or adapt the code to work with the current structure of Qiskit.

Here's how you can handle both scenarios:

1. Using an Older Version of Qiskit:

You can install an older version of Qiskit that includes the `'qiskit.finance'` module. To do this, you need to find out which version supports it and install that specific version, for example:

```
python Copy code  
  
pip install qiskit==[version_number]
```

Replace `'[version_number]'` with the version number that includes the `'qiskit.finance'` module.

2. Adapting to the Current Qiskit Structure:


If you prefer to work with the latest version of Qiskit, you would need to manually set up the problem for QAOA. This involves:



- Manually creating the cost Hamiltonian for portfolio optimization.
- Setting up the QAOA circuit without the `'qiskit.finance'` built functions from `'qiskit.finance'`.

This approach requires a deeper understanding of both quantum computing principles and Qiskit's lower-level operations.

Given these options, if you're not sure about handling the complex task of manually setting up the Hamiltonian, I recommend using an older version of Qiskit that includes the `'qiskit.finance'` module. This way, you can focus more on the logic of your portfolio optimization algorithm rather than the intricacies of Qiskit's latest version.

Alternatively, if you're up for the challenge or wish to stay with the latest Qiskit version, you could explore more advanced quantum computing resources or Qiskit's documentation to manually set up your portfolio optimization problem. This path can be more complex but also more rewarding in terms of learning and flexibility.

 **You**
which version includes qiskit.finance?

 **ChatGPT**
 Error analyzing
Unfortunately, I'm unable to access external resources to check the latest version of Qiskit that includes the `'qiskit.finance'` module. However, you can easily find this information by visiting the GitHub repository of Qiskit Finance or by checking the release history on the PyPI (Python Package Index) page for Qiskit.

To explore the GitHub repository for the latest version that includes `'qiskit.finance'`, follow these steps:

1. Go to [Qiskit Finance GitHub Repository](#).
2. Check the tags or releases section for the latest version information.

Alternatively, you can visit the Qiskit Finance page on PyPI:

1. Go to [Qiskit Finance on PyPI](#).
2. Browse through the release history to find the version that includes the module you need.

Once you identify the correct version, you can install it using pip. For example:

```
python Copy code  
  
pip install qiskit-finance==[version_number]
```


Replace `[version_number]` with the actual version number you find. This will allow you to use the `'qiskit.finance'` module in your quantum portfolio optimization project. [2-]



You

could it be that the module is named `qiskit_finance`?



ChatGPT

Yes, it's entirely possible that the module is named `'qiskit_finance'` instead of `'qiskit.finance'`. In Python, module names often use underscores (`'_'`) instead of dots (`'.'`) to maintain compatibility with the language's naming conventions.

You can try importing the module using this alternative naming convention:

```
python
import qiskit_finance
```

If this module is a part of the Qiskit ecosystem but is now maintained as a separate package, you would also need to install it separately via pip:

```
python
pip install qiskit-finance
```

This installation command uses a hyphen (`'-'`) as per the standard naming conventions in PyPI (Python Package Index), where the package is likely hosted. After installation, you should be able to import and use the module in your Python environment.

If you encounter any issues or if the module doesn't exist under this name, it would be best to refer to the official Qiskit documentation or their GitHub repository for the most accurate and up-to-date information.

< 10 / 10 >



You

I have installed `qiskit_finance` and am able to import it, however it still gives this error: `ModuleNotFoundError: No module named 'qiskit_finance.applications.ising'`



ChatGPT

The error message indicates that the specific module or path `'qiskit_finance.applications.ising'` you're trying to import does not exist in the `'qiskit_finance'` package. This could be due to changes in the package structure or the module might have been deprecated or moved to a different location within the package.

Here's how to proceed:

1. **Check the Documentation:** The best course of action is to refer to the official documentation of `'qiskit_finance'`. The documentation will provide accurate information on the current structure of the package, including available modules and how to use them.
2. **Explore the Package Structure:** You can explore the structure of the installed `'qiskit_finance'` package to understand its current organization. In Python, you can do this by listing the contents of the package or by exploring its directory in your Python environment.
3. **Alternative Approach:** If the Ising model or specific functionality you're looking for is no longer available in `'qiskit_finance'`, you might need to look for alternative methods to implement your algorithm. This could involve manually setting up the quantum problem or using different components of the Qiskit ecosystem.
4. **Community and Support:** If you're still unsure, consider reaching out to the Qiskit community or support forums. Many times, other users and developers can provide valuable insights and solutions based on recent changes or updates.

Since I don't have the capability to directly inspect your installed packages or access external documentation, these general steps should guide you towards resolving the issue with the appropriate modules in `'qiskit_finance'`.

📄 👍 🗨️ 🔍



Message ChatGPT...



```
In [6]: import numpy as np
from qiskit import Aer
from qiskit.algorithms.optimizers import COBYLA
from qiskit.algorithms import QAOA
from qiskit.opflow import PauliSumOp
from qiskit.utils import QuantumInstance, algorithm_globals
from qiskit_finance.applications.ising import portfolio

# Generate mock financial data
np.random.seed(42) # for reproducibility
num_assets = 10
expected_returns = np.random.uniform(-0.10, 0.30, num_assets) # between -10% and 30%
covariance_matrix = np.random.rand(num_assets, num_assets)
covariance_matrix = 0.5 * (covariance_matrix + covariance_matrix.T) # Symmetrize
covariance_matrix += num_assets * np.eye(num_assets) # Ensure positive definite

# Portfolio optimization parameters
q = 0.5 # risk factor
budget = 3 # number of assets to include in the portfolio
penalty = num_assets # penalty parameter for the budget constraint

# Create the Ising Hamiltonian
qubit_op, offset = portfolio.get_operator(expected_returns, covariance_matrix, q)

# Set up the quantum instance (using a simulator here)
algorithm_globals.random_seed = 123
quantum_instance = QuantumInstance(Aer.get_backend('qasm_simulator'), seed_simulator=123)

# Initialize the QAOA algorithm
optimizer = COBYLA(maxiter=250)
qaoa = QAOA(optimizer=optimizer, quantum_instance=quantum_instance)

# Solve the problem using QAOA
result = qaoa.compute_minimum_eigenvalue(qubit_op)

# Print results
print('Result:', result)
```

Traceback (most recent call last):

Cell In[6], line 7

from qiskit_finance.applications.ising import portfolio

ModuleNotFoundError: No module named 'qiskit_finance.applications.ising'

Use %tb to get the full traceback.

[Search for solution online](#)

```
In [8]: import qiskit_finance.applications.ising
```

Traceback (most recent call last):

Cell In[8], line 1

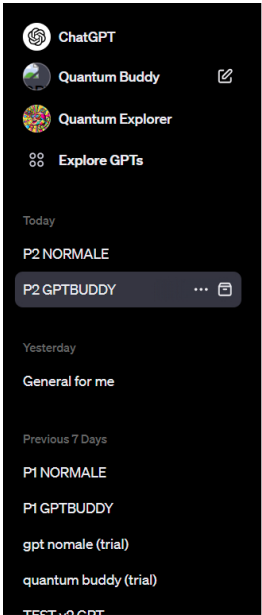
import qiskit_finance.applications.ising

ModuleNotFoundError: No module named 'qiskit_finance.applications.ising'

Use %tb to get the full traceback.

[Search for solution online](#)

In []:



Quantum Buddy ▾



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.



Quantum Buddy

Great! You're about to embark on an exciting journey of quantum computing exploration. The aim of this structured user test is to conceptualize, develop, and refine an algorithm that takes full advantage of the unique capabilities of quantum computing. This test is designed for someone like you, with a background in quantum mechanics and coding.

The process is divided into several stages:

1. **Conceptualization:** You'll start by choosing a field and identifying a specific problem you want to solve using quantum computing. Options include Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems.
2. **Development:** You'll then begin coding your algorithm. If coding isn't your strongest suit, don't worry! I'm here to help translate your quantum knowledge into a functional code.
3. **Refinement:** After developing your code, you'll test and refine it to ensure it works efficiently and effectively.
4. **Finalization:** Finally, you'll finalize your code and prepare it for submission, so it can be run on an IBM quantum computer.

Remember, this is a learning experience, so feel free to ask questions and explore different avenues. Are you ready to get started? 🤖💡🔗



You

Id like to make a machine learning algorithm, can you give me some examples of applications?



Quantum Buddy

Absolutely, diving into machine learning with quantum computing is a fascinating choice! Quantum-enhanced machine learning can potentially offer significant speed-ups and improvements over classical machine learning algorithms. Here are some examples of applications where quantum machine learning (QML) could be applied:

1. **Pattern Recognition:** Utilizing quantum computing for complex pattern recognition tasks, like identifying subtle patterns in large datasets, which can be particularly useful in fields like finance, healthcare, or security.
2. **Quantum Neural Networks:** Developing neural networks that can operate on quantum data or enhance classical data processing. These networks could be more efficient in handling certain types of problems like in image or speech recognition.
3. **Drug Discovery:** Accelerating the process of drug discovery and molecular simulation. Quantum algorithms can simulate molecular interactions at a much more granular level, which is a critical aspect in the development of new pharmaceuticals.
4. **Optimization in Machine Learning Models:** Using quantum algorithms for optimization problems in machine learning, like tuning hyperparameters more efficiently or finding the global minimum of a cost function faster.
5. **Financial Modeling:** Applying QML for complex financial models, enabling quicker and more accurate predictions of market trends, risk assessment, and portfolio optimization.
6. **Data Clustering and Classification:** Enhancing data clustering and classification tasks, which are fundamental to many machine learning applications, such as customer segmentation in marketing or anomaly detection in cybersecurity.
7. **Natural Language Processing (NLP):** Applying quantum computing to process and analyze large volumes of natural language data, potentially improving the efficiency of

tasks like sentiment analysis, machine translation, or text summarization.

Each of these applications leverages the inherent advantages of quantum computing, such as handling vast datasets and performing complex calculations at speeds unattainable by classical computers.

Which of these areas sounds most intriguing to you for developing your machine learning algorithm? 🤖 📊 🔍



You

The portfolio optimization algorithm sounds most appealing. I want to use qiskit, can you write the algorithm for 10 assets?



Quantum Buddy

Absolutely, creating a quantum algorithm for portfolio optimization using Qiskit is an exciting project! For a portfolio of 10 assets, the goal is to optimize the allocation of investments across these assets to maximize returns while minimizing risk, based on historical data.

However, writing a complete quantum algorithm for portfolio optimization is quite complex and involves several steps. I'll outline the high-level approach and provide a basic framework for the algorithm using Qiskit. You can then build upon this framework and tailor it to your specific needs.

High-Level Approach:

1. **Data Preparation:** Collect historical price data for the 10 assets. Calculate the expected returns and the covariance matrix, which are key inputs for the optimization problem.
2. **Formulating the Optimization Problem:** Define the objective function for portfolio optimization. Typically, this involves maximizing returns while minimizing risk, which can be quantified using the Sharpe ratio or a similar metric.
3. **Quantum Algorithm Implementation:** Use Qiskit to implement a quantum algorithm for solving the optimization problem. The Variational Quantum Eigensolver (VQE) or Quantum Approximate Optimization Algorithm (QAOA) are common choices for such tasks.
4. **Running the Algorithm:** Test the algorithm using a quantum simulator first, then run it on an actual quantum computer.

Basic Framework in Qiskit:

```
python
from qiskit import Aer, execute
from qiskit.circuit.library import EfficientSU2
from qiskit.utils import QuantumInstance
from qiskit_finance.applications.ising import portfolio
from qiskit_finance.data_providers import RandomDataProvider
from qiskit_algorithms.optimizers import COBYLA
from qiskit_algorithms import VQE
import numpy as np
import datetime

# Data Preparation
num_assets = 10
stocks = ["STOCK1", "STOCK2", "STOCK3", "STOCK4", "STOCK5", "STOCK6",
          "STOCK7", "STOCK8", "STOCK9", "STOCK10"]
data_provider = RandomDataProvider(tickers=stocks,
                                   start=datetime.datetime(2016,1,1),
                                   end=datetime.datetime(2016,1,30))

data_provider.run()
mu = data_provider.get_period_return_mean_vector()
sigma = data_provider.get_period_return_covariance_matrix()

# Optimization Problem
q = 0.5 # risk factor
budget = num_assets // 2 # budget
penalty = num_assets # penalty parameter

qubit_op, offset = portfolio.get_operator(mu, sigma, q, budget, penalty)

# Quantum Algorithm Implementation
optimizer = COBYLA()
var_form = EfficientSU2(qubit_op.num_qubits, entanglement='linear')
vqe = VQE(qubit_op, var_form, optimizer)

# Running the Algorithm
backend = Aer.get_backend('statevector_simulator')
quantum_instance = QuantumInstance(backend)
result = vqe.run(quantum_instance)
```

```
# Analyzing the result
print(f"The optimized bitstring is {result['optimal_point']}")
```

Notes:



- **Data Provider:** The above code uses a `'RandomDataProvider'` for demonstration purposes. You should replace this with actual historical data for your assets.
- **Risk Factor (' q ') and Budget:** These are key parameters you'll need to adjust based on your portfolio strategy.
- **Backend:** Initially, use a simulator ('`statevector_simulator`') for testing. Once you're satisfied, you can switch to an actual quantum processor.

This framework provides a starting point. You'll need to adjust it and possibly add more functionality to suit your specific problem. Let me know if you need further guidance or explanations on any part of the code! 🤖💻📄



You

It cannot find the `qiskit_finance.applications.ising` module. the problem is with `Ising`, since it can import `qiskit_finance.applications`. Could it have a different name than `ising`?



Quantum Buddy

Internal Server Error

< 8 / 8 > 📄 👍 🗑️ ↺

There was an error generating a response

🔄 Regenerate

ChatGPT can make mistakes. Consider checking important information.

?



Upgrade plan

Collaborate on a Team plan



co doc

```
In [6]: from qiskit import Aer, execute
from qiskit.circuit.library import EfficientSU2
from qiskit.utils import QuantumInstance
from qiskit_finance.applications.optimization.portfolio_optimization import portfolio
from qiskit_finance.data_providers import RandomDataProvider
from qiskit.algorithms.optimizers import COBYLA
from qiskit.algorithms import VQE
import numpy as np
import datetime

# Data Preparation
num_assets = 10
stocks = ["STOCK1", "STOCK2", "STOCK3", "STOCK4", "STOCK5", "STOCK6", "STOCK7",
data_provider = RandomDataProvider(tickers=stocks,
start=datetime.datetime(2016,1,1),
end=datetime.datetime(2016,1,30))

data_provider.run()
mu = data_provider.get_period_return_mean_vector()
sigma = data_provider.get_period_return_covariance_matrix()

# Optimization Problem
q = 0.5 # risk factor
budget = num_assets // 2 # budget
penalty = num_assets # penalty parameter

qubit_op, offset = portfolio.get_operator(mu, sigma, q, budget, penalty)

# Quantum Algorithm Implementation
optimizer = COBYLA()
var_form = EfficientSU2(qubit_op.num_qubits, entanglement='linear')
vqe = VQE(qubit_op, var_form, optimizer)

# Running the Algorithm
backend = Aer.get_backend('statevector_simulator')
quantum_instance = QuantumInstance(backend)
result = vqe.run(quantum_instance)

# Analyzing the result
print(f"The optimized bitstring is {result['optimal_point']}")
```

Traceback (most recent call last):

Cell In[6], line 4

```
from qiskit_finance.applications.optimization.portfolio_optimization import
portfolio
```

```
ImportError: cannot import name 'portfolio' from 'qiskit_finance.applications.o
ptimization.portfolio_optimization' (/opt/conda/lib/python3.10/site-packages/qi
skit_finance/applications/optimization/portfolio_optimization.py)
```

Use %tb to get the full traceback.

[Search for solution online](#)

```
In [3]: import qiskit.finance
```

Traceback (most recent call last):

Cell In[3], line 1

```
import qiskit.finance
```

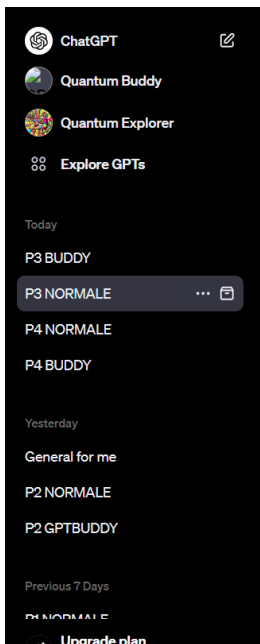
ModuleNotFoundError: No module named 'qiskit.finance'

Use %tb to get the full traceback.

[Search for solution online](#)

In []:

P3



ChatGPT 4



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.



ChatGPT

Aim of the User Test 🎯



Hello there, young enthusiast! We're embarking on an exciting journey in the world of quantum mechanics and coding. 🤖💻 The objective of this test is to harness your background in these areas to conceptualize, develop, and refine an algorithm that exploits the unique capabilities of quantum computing. 🧠💡

This isn't just about writing code; it's about merging the abstract concepts of quantum mechanics with practical programming skills to create something innovative and powerful. Think of it as blending the best of both worlds – the enigmatic quantum universe and the logical realm of coding – to solve complex problems in ways we haven't imagined before. 🌟💡

Are you ready to dive into this challenge and see what amazing things we can create together? Let me know when you're set to go! 🚀😊



Get citation



You

yes



ChatGPT

Fantastic! Let's get started. 🚀



Step 1: Conceptualization 💬

First things first, we need to pick a field and identify a unique problem that you want to solve using quantum computing. Here are some options to consider:

1. **Cryptography:** Dive into the world of secure communication. How can quantum computing revolutionize the way we encrypt and decrypt sensitive information? 🗝️
2. **Machine Learning:** Explore how quantum algorithms can enhance machine learning models. Think about tackling complex datasets or speeding up learning processes. 🤖📊
3. **Chemical Simulations:** Use quantum computing to simulate molecular interactions. This could lead to breakthroughs in medicine or materials science. 🧬🔬
4. **Optimization Problems:** Tackle complex optimization challenges in fields like logistics, finance, or resource management. Find the best solutions faster and more efficiently. 📦🌐

Take a moment to think about these fields. Which one sparks your interest the most, or is there another area you're passionate about? Let me know your choice or if you need more information on any of these options! ⬇️💡



Get citation



You

machine learning



ChatGPT

Excellent choice! Machine Learning with Quantum Computing is a fascinating and rapidly growing field. 🤖💡

Step 2: Development 🚀💻

Now, let's start developing your quantum machine learning algorithm. Here's how we'll


approach it:


1. **Define the Problem:** First, pinpoint the specific machine learning problem you want to tackle with quantum computing. It could be anything from classification, regression, clustering, to more advanced topics like deep learning or reinforcement learning. 🎯
2. **Understand Quantum ML Models:** Familiarize yourself with how quantum computing can enhance machine learning. This involves understanding quantum versions of neural networks, kernel methods, or optimization algorithms. 🌐 🧠
3. **Start Coding:** Begin to code your algorithm. ⬇️ Don't worry if you're not a quantum coding expert; I'm here to help guide you through the process. We'll start with basic concepts and gradually build up. 📄 🖥️

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As we proceed, remember that this is a learning experience. It's perfectly fine to make mistakes and ask questions. The goal is to translate your understanding of quantum mechanics into a working quantum machine learning algorithm. 🌟 🧑🏫

Whenever you're ready, let me know how you'd like to begin, or if you have any specific questions about quantum machine learning or the coding process. Let's embark on this coding adventure together! 🚀 😊

 **You**
let's work on a classification algorithm

 **ChatGPT**
Great choice! Classification algorithms are a cornerstone of machine learning, and implementing one with quantum computing can be both challenging and rewarding. 🌟 📄

 Get citation



Developing a Quantum Classification Algorithm

Here's how we can approach this:

1. Choose a Dataset 🌐 📄

First, select a dataset for your classification task. It could be anything from image recognition (like the MNIST dataset of handwritten digits) to text classification. If you have a specific dataset in mind, let me know; otherwise, I can suggest one.

2. Understand the Basics of Quantum Classification 🌐 🧠

Quantum classification involves using quantum states and operations to classify data. Key concepts include:

- **Quantum States:** How data is represented in quantum systems.
- **Quantum Gates:** Operations that manipulate quantum states, analogous to classical logic gates.
- **Entanglement and Superposition:** Unique quantum properties that can be leveraged for complex computations.

3. Framework and Tools 🧰

Decide on a quantum computing framework. ⬇️ Popular options include Qiskit (IBM), Cirq (Google), and others. These frameworks provide tools to create and simulate quantum circuits.

 Get citation

4. Coding the Algorithm 📄

Start by translating your classification problem into a quantum circuit. This involves encoding your data into quantum states (qubits), processing it through quantum gates, and then measuring the output to classify the data.

5. Iterative Testing and Learning 🧪 🧠

Test your algorithm on small subsets of your dataset. Quantum algorithms often require fine-tuning of parameters and logic, so iterative testing is crucial.


6. Seek Feedback and Refine 🎯

As you develop your algorithm, feel free to ask for feedback or help with troubleshooting. The goal is to refine your approach until you have a working classification model.



 Get citation

Ready to dive in? Let me know if you have a dataset in mind, or if you need help selecting one. Also, if you're new to any quantum computing framework, I can provide guidance on getting started. Let's create some quantum magic! 🌟 🚀

 **You**
.....

I not have any dataset in mind, I will need your help. I will use Qiskit



ChatGPT

Great! Since you'll be using Qiskit, we'll choose a dataset that is well-suited for a quantum classification task and can be easily worked with in this framework.

Suggested Dataset: Iris Dataset 🌸

One popular choice for classification tasks is the **Iris dataset**. It's a classic in the field of machine learning and a great starting point for quantum classification due to its simplicity and small size. The dataset contains 150 samples of iris flowers, classified into three species (Setosa, Versicolor, Virginica). Each sample has four features: the lengths and widths of the sepals and petals.

Why the Iris Dataset?

- **Size:** It's small and manageable, perfect for quantum experiments where computational resources can be a limiting factor.
- **Simplicity:** With just four features, it's easier to encode the data into quantum states.
- **Well-Studied:** There's a lot of reference material and examples using this dataset, making it easier to compare and understand your results.

Next Steps with the Iris Dataset

1. **Data Preprocessing:** Simplify and standardize the data for quantum processing. This includes normalizing the feature values and possibly reducing the dimensionality.
2. **Encoding Data into Quantum States:** Using Qiskit, you'll encode this preprocessed data into quantum states. This is a crucial step in quantum machine learning.
3. **Building the Quantum Circuit:** Design a quantum circuit that can process these states and perform classification.
4. **Testing and Training:** Implement a procedure to train your model and test its accuracy on classifying the iris species.
5. **Analysis and Iteration:** Analyze the results, make adjustments, and iteratively improve your algorithm.

If the Iris dataset sounds good to you, we can proceed with the first step: Data Preprocessing. Or, if you have any other preferences or questions, feel free to let me know! 🌞 📊 🌸



You

The iris dataset is fine



ChatGPT

Fantastic! Let's proceed with the Iris dataset for your quantum classification algorithm using Qiskit. 🌸 📊

Step-by-Step Guide to Using the Iris Dataset with Qiskit

1. Data Preprocessing 📊

Since quantum computers work with qubits, you'll need to preprocess the Iris dataset to make it suitable for quantum processing. This involves:

- **Normalization:** Scale the feature values so that they are within a similar range. This helps in efficiently encoding them into quantum states.
- **Dimensionality Reduction:** Quantum computers are resource-constrained, so reducing the number of features (from 4 to 2 or 3) can be beneficial. Techniques like Principal Component Analysis (PCA) can be used for this.

2. Encoding Data into Quantum States 🌐

Once the data is preprocessed, the next step is to encode it into quantum states. This is done using qubits. Qiskit offers different quantum feature maps for this purpose, like the ZZ feature map.

3. Building the Quantum Circuit ⚙️

The quantum circuit is where the computation happens. You'll create a circuit that can process the quantum states and perform the classification. This involves choosing the right quantum gates and setting up the circuit for measurement.

4. Quantum Machine Learning Model 🤖

Integrate your quantum circuit with a classical machine learning algorithm. The output of the quantum circuit can be used as an input feature for a classical classifier like Support Vector Machine (SVM).

5. Training and Testing 🇺🇸

Train your hybrid quantum-classical model on a portion of the Iris dataset and test its performance on a separate set of data. This will help you evaluate the accuracy of your model.

6. Iteration and Refinement 🔧

Based on the results, you may need to adjust your quantum circuit, feature map, or even the classical part of your algorithm. Iterative testing and refinement are key.

Getting Started with Qiskit

You'll need to have Qiskit installed in your Python environment. If you haven't done so already, you can install it using pip:

bash

Copy code

```
pip install qiskit
```



Get citation

Then, you can start by importing Qiskit and other necessary libraries in your Python script or Jupyter notebook.

Would you like to begin with the data preprocessing step or discuss any specific part of the process in more detail? Let me know how you'd like to proceed! 🚀 🤖 🌸



You

Let's start with data preprocessing



ChatGPT

Great! Let's begin with the data preprocessing for the Iris dataset. The goal here is to prepare the data so it can be efficiently processed by a quantum computer using Qiskit.

Steps for Data Preprocessing

1. Load the Iris Dataset

First, we need to load the dataset. You can ↓ do this using libraries like `scikit-learn` which has the Iris dataset built-in.

Get citation

python

Copy code

```
from sklearn import datasets

# Load the Iris dataset
iris = datasets.load_iris()
X = iris.data # Features
y = iris.target # Labels
```

2. Normalize the Data

Normalization adjusts the scale of the feature values so that they are within a similar range. This is important for quantum encoding.

python

Copy code

```
from sklearn.preprocessing import StandardScaler

# Normalize the data
scaler = StandardScaler()
X_normalized = scaler.fit_transform(X)
```

Get citation

3. Dimensionality Reduction

Reducing the number of features makes it easier to encode the data into quantum states. Principal Component Analysis (PCA) is a common technique for this.

python

Copy code

```
from sklearn.decomposition import PCA

# Reduce dimensions to 2 for simplicity
pca = PCA(n_components=2)
X_reduced = pca.fit_transform(X_normalized)
```

4. Binary Classification (Optional)

If you want to simplify the task, consider converting it into a binary classification

problem by selecting only two of the three species of iris.

```
python Copy code

import numpy as np

# Select two species for binary classification
indices = np.where(y != 2) # For example, exclude the third species
X_binary = X_reduced[indices]
y_binary = y[indices]
```

 Get citation

5. Train-Test Split

Finally, split the data into a training set and a test set. This will help in evaluating the performance of your quantum classifier.

```
python Copy code

from sklearn.model_selection import train_test_split


# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X_binary, y_binary,
```


Next Steps

After preprocessing, the next step is to encode this data into quantum states using Qiskit and build the quantum circuit for classification.

 Get citation

Would you like to proceed with the quantum encoding part, or do you have any specific questions or modifications in mind regarding the preprocessing? Let me know how you'd like to continue! 🚀 🌸

 **You**
let's proceed

 **ChatGPT**
Great! Now that we have preprocessed the data, the next step is to encode it into quantum states and build the quantum circuit for the classification task using Qiskit.

Quantum Encoding of Data

Quantum encoding transforms classical data into quantum states. This is a crucial step as it determines how the quantum computer will process your data.

1. Choose a Feature Map

In Qiskit, a feature map is a function that translates classical data into quantum states. The ZZ feature map is a common choice for this purpose.

2. Implement the Feature Map

Here's how you can implement a ZZ feature map in Qiskit:

```
python Copy code

from qiskit.circuit.library import ZZFeatureMap


# Define the feature map
feature_dim = 2 # Since we reduced our data to 2 dimensions
feature_map = ZZFeatureMap(feature_dimension=feature_dim, reps=2)
```

 Get citation

Building the Quantum Circuit

After encoding the data, you need to build a quantum circuit that can process these quantum states to perform classification.

1. Create a Quantum Circuit

You can create a quantum circuit in Qiskit  includes the feature map.

 Get citation

```
python Copy code

from qiskit import QuantumCircuit

# Create a quantum circuit
qc = QuantumCircuit(feature_dim)

# Add the feature map
qc.append(feature_map, range(feature_dim))
```

```
# Add a measurement
qc.measure_all()
```

2. Integrate with a Quantum Kernel

A quantum kernel is a method that measures the similarity between pairs of data points in the quantum feature space. Qiskit provides tools to implement quantum kernels.

```
python Copy code

from qiskit_machine_learning.kernels import QuantumKernel

# Create a quantum kernel using the feature map and a quantum backend
quantum_kernel = QuantumKernel(feature_map=feature_map, quantum_instance=quantum_instance)
```

 Get citation

Hybrid Quantum-Classical Model

Now, integrate the quantum circuit with a classical machine learning model.

1. Support Vector Machine (SVM) Classifier

You can use the quantum kernel in a classical SVM classifier.

```
python Copy code

from sklearn.svm import SVC

# Create an SVM classifier
svm = SVC(kernel=quantum_kernel.evaluate)
```

2. Training and Testing the Model

Train the SVM on your training data and test its performance.

```
python Copy code

# Train the model
svm.fit(X_train, y_train)

# Test the model
score = svm.score(X_test, y_test)
print(f"Classification accuracy: {score}")
```

 Get citation


Finalization


Once you have implemented and trained your model, you can analyze its performance, adjust parameters, and refine your approach.

Before proceeding, ensure you have a quantum instance (simulator or actual quantum hardware) set up in Qiskit for the quantum kernel. If everything is set, you can run the above code, starting with the feature map implementation.

Let me know if you need help with setting up a quantum instance or if there are any other aspects of the process you'd like to discuss! 🌟💡🌸

 Get citation

 **You**
what is my quantum instance?

 **ChatGPT**
In Qiskit, a "quantum instance" refers to the backend environment where your quantum circuits are executed. It can be either a quantum simulator running on your local machine or an actual quantum computer accessed remotely via the cloud.

Types of Quantum Instances

1. **Simulator:** A local or cloud-based simulator that mimics the behavior of a quantum computer using classical computation. It's great for development and testing, as it's more accessible and faster than real quantum hardware.
2. **Quantum Hardware:** Actual quantum computers, like those provided by IBM through the IBM Quantum Experience. These are used for running experiments on real quantum systems.

Setting Up a Quantum Instance in Qiskit

Using a Simulator

For beginners or for testing purposes, using a simulator is the recommended approach. Here's how to set up a simulator in Qiskit:

 Get citation

```
python
Copy code

from qiskit import Aer
from qiskit.utils import QuantumInstance

# Setup a simulator backend
simulator = Aer.get_backend('qasm_simulator')

# Create a quantum instance with the simulator
quantum_instance = QuantumInstance(simulator)
```

Using IBM Quantum Hardware

To use real quantum hardware, you need to access IBM's quantum computers via IBM Quantum Experience. You'll need an IBM Quantum Experience account for this.

1. **Create an IBM Quantum Experience Account:** Register at [IBM Quantum Experience](#).
2. **Get Your API Token:** Once registered, generate an API token from your account settings.
3. **Configure Qiskit to Connect to IBM Quantum Experience:**

Get citation

```
python
Copy code

from qiskit import IBMQ

# Save your IBMQ account
IBMQ.save_account('YOUR_API_TOKEN')

# Load your IBMQ account
IBMQ.load_account()

provider = IBMQ.get_provider(hub='ibm-q')
quantum_computer = provider.get_backend('ibmq_santiago') # Example backend

# Create a quantum instance with the quantum computer
quantum_instance = QuantumInstance(quantum_computer)
```

Choosing a Quantum Instance

Get citation

- * **Development and Testing:** Use a simulator.
- * **Advanced Experiments:** If you want to run experiments on real quantum hardware, use IBM Quantum Experience.

Would you like to proceed with a simulator for now, or are you interested in setting up and using real quantum hardware? Let me know, and I can assist you further!

You
how can I visualize the quantum code?

ChatGPT
Visualizing your quantum circuit in Qiskit is a great way to understand and debug your quantum algorithm. Qiskit provides several ways to visualize quantum circuits, including text, matplotlib drawings, and interactive diagrams.

Visualizing Quantum Circuits in Qiskit

Get citation

1. Textual Representation

You can use the `draw()` method of a quantum circuit with the `output='text'` option to get a simple text-based representation of your circuit. This is useful for quick views or when working in environments without graphical support.

```
python
Copy code

# Assuming 'qc' is your quantum circuit
print(qc.draw(output='text'))
```

2. Matplotlib Drawing

For a more graphical representation, use the `output='mpl'` option. This requires having matplotlib installed in your Python environment.

```
python
Copy code

# Assuming 'qc' is your quantum circuit
qc.draw(output='mpl')
```

3. Interactive Diagram

[Get citation](#)

Qiskit also provides an interactive circuit visualization using the `%matplotlib inline` magic command for Jupyter notebooks. This allows for interactive exploration of the circuit.

```
python
%matplotlib inline
qc.draw(output='mpl')
```

Installing Required Packages

If you don't have matplotlib installed, you can install it using pip:

```
bash
pip install matplotlib
```

Example

Here's an example of creating a simple quantum circuit and visualizing it:

[Get citation](#)

```
python
from qiskit import QuantumCircuit
import matplotlib.pyplot as plt

# Create a simple quantum circuit
qc = QuantumCircuit(2)
qc.h(0)
qc.cx(0, 1)
qc.measure_all()

# Visualize the circuit
qc.draw(output='mpl')
plt.show() # This line is needed to display the plot in some environments
```

Advanced Visualization Tools

Qiskit also offers advanced visualization tools for state vectors, quantum state tomography, and more. These tools can provide deeper insights into the state of your quantum system.

[Get citation](#)

If you have a specific quantum circuit in mind or need assistance with any of these visualization methods, feel free to ask! 🌟🚀📊



You
can I visualize the feature map?



ChatGPT
Yes, you can visualize the feature map in Qiskit, just like you would visualize any other quantum circuit. A feature map in quantum machine learning is essentially a series of quantum gates designed to encode classical data into a quantum state. In Qiskit, these are implemented as quantum circuits, so you can use the same visualization tools.

Visualizing the Feature Map

Let's say you have a feature map defined in Qiskit (for example, a `ZZFeatureMap`). Here's how you can visualize it:

Step 1: Define Your Feature Map

[Get citation](#)

First, define your feature map. For instance, using the `ZZFeatureMap`:

```
python
from qiskit.circuit.library import ZZFeatureMap

feature_dim = 2 # The number of features in your data
feature_map = ZZFeatureMap(feature_dimension=feature_dim, reps=1) # 'x'
```

Step 2: Visualize Using Matplotlib

Now, visualize the feature map using Matplotlib:

```
python
# Visualization code would go here
```

```
import matplotlib.pyplot as plt

feature_map.draw(output='mpl')
plt.show() # This line is needed to display the plot in some environme
```

This will give you a graphical representation of the quantum gates that make up your feature map.

 Get citation

Understanding the Visualization

Each line in the visualization represents a qubit. The gates (like Hadamard, CNOT, etc.) are shown as symbols on these lines. The feature map circuit usually consists of a series of entangling gates and single-qubit rotations, which are parameterized by your input data.

Advanced Visualization Options


In addition to the basic drawing, you can explore:

- **3D Visualization:** Qiskit provides tools for 3D visualization of quantum states which can be useful for more advanced analyses.
- **Interactive Plots:** In a Jupyter notebook, using `%matplotlib inline` will enable interactive plots.

Remember, the visualization is particularly helpful for understanding the structure of your quantum circuit and ensuring that your feature map is set up correctly. If you need any further assistance or specific details on visualizing different aspects of your quantum circuit, feel free to ask! 🌞 🇮🇹 🚀

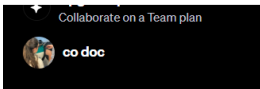
   

 Get citation

 Message ChatGPT...

ChatGPT can make mistakes. Consider checking important information.

?



In [1]: `pip install qiskit`

```
Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.44.3)
Requirement already satisfied: qiskit-terra==0.25.3 in /opt/conda/lib/python3.10/site-packages (from qiskit) (0.25.3)
Requirement already satisfied: numpy>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (1.23.5)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (0.3.7)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (0.13.0)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (3.11)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (4.5.0)
Requirement already satisfied: symengine<0.10,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (0.9.2)
Requirement already satisfied: scipy>=1.5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (1.9.3)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (4.1.1)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (1.11.1)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (2.8.2)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (5.9.4)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra==0.25.3->qiskit) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra==0.25.3->qiskit) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra==0.25.3->qiskit) (1.3.0)
```

[notice] A new release of pip available: 23.1.1 -> 23.3.2

[notice] To update, run: `pip install --upgrade pip`

Note: you may need to restart the kernel to use updated packages.

In [4]: `from sklearn import datasets`

```
# Load the Iris dataset
iris = datasets.load_iris()
X = iris.data # Features
y = iris.target # Labels
```

In [5]: `from sklearn.preprocessing import StandardScaler`

```
# Normalize the data
scaler = StandardScaler()
X_normalized = scaler.fit_transform(X)
```

In [6]: `from sklearn.decomposition import PCA`

```
# Reduce dimensions to 2 for simplicity
pca = PCA(n_components=2)
X_reduced = pca.fit_transform(X_normalized)
```

```
In [7]: import numpy as np

# Select two species for binary classification
indices = np.where(y != 2) # For example, exclude the third species
X_binary = X_reduced[indices]
y_binary = y[indices]
```

```
In [8]: from sklearn.model_selection import train_test_split

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X_binary, y_binary, test_size=0.2)
```

```
In [10]: from qiskit.circuit.library import ZZFeatureMap

# Define the feature map
feature_dim = 2 # Since we reduced our data to 2 dimensions
feature_map = ZZFeatureMap(feature_dimension=feature_dim, reps=2)
```

```
In [11]: from qiskit import QuantumCircuit

# Create a quantum circuit
qc = QuantumCircuit(feature_dim)

# Add the feature map
qc.append(feature_map, range(feature_dim))

# Add a measurement
qc.measure_all()
```

```
In [13]: from qiskit import Aer
from qiskit.utils import QuantumInstance

# Setup a simulator backend
simulator = Aer.get_backend('qasm_simulator')

# Create a quantum instance with the simulator
quantum_instance = QuantumInstance(simulator)
```

```
/tmp/ipykernel_196/3209148759.py:8: DeprecationWarning: The class ``qiskit.utils.QuantumInstance`` is deprecated as of qiskit-terra 0.24.0. It will be removed no earlier than 3 months after the release date. For code migration guidelines, visit https://qiskit.org/docs/2.0/migration.html.
  quantum_instance = QuantumInstance(simulator)
```

```
In [15]: from qiskit_machine_learning.kernels import QuantumKernel

# Create a quantum kernel using the feature map and a quantum backend
quantum_kernel = QuantumKernel(feature_map=feature_map, quantum_instance=quantum_instance)
```

```
In [16]: from sklearn.svm import SVC

# Create an SVM classifier
svm = SVC(kernel=quantum_kernel.evaluate)
```

```
In [17]: # Train the model
svm.fit(X_train, y_train)

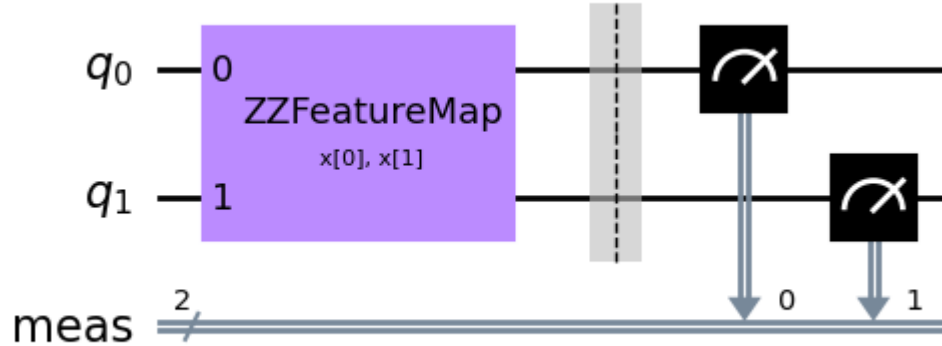
# Test the model
```

```
score = svm.score(X_test, y_test)
print(f"Classification accuracy: {score}")
```

Classification accuracy: 0.55

```
In [19]: # Assuming 'qc' is your quantum circuit
qc.draw(output='mpl')
```

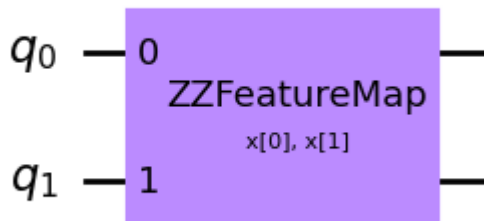
Out[19]:



```
In [23]: import matplotlib.pyplot as plt

feature_map.draw(output='mpl')
# plt.show() # This line is needed to display the plot in some environments
```

Out[23]:



In []:

ChatGPT

Quantum Buddy

Quantum Explorer

Explore GPTs

Today

P3 BUDDY

P3 NORMALE

P4 NORMALE

P4 BUDDY

Yesterday

General for me

P2 NORMALE

P2 GPTBUDDY

Previous 7 Days

P4 NORMALE

Upgrade plan



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.



Quantum Buddy

Great! You're about to embark on a fascinating journey into the world of quantum computing! 🚀



Get citation

Aim of the Test:

The objective of this user test is to conceptualize, develop, and refine an algorithm that fully utilizes the unique capabilities of quantum computing. This test will challenge and enhance your understanding of quantum mechanics and coding, guiding you to create a meaningful and functional quantum algorithm. Are you ready to dive into this quantum adventure? Let me know when you are! 🤖💡🔗👤



You

let's go



Quantum Buddy

Awesome! Let's get started with the **Conceptualization** stage. 💡🔗

- Choose a Field:** First, think about a field that excites you or a problem you're passionate about solving. Here are some options:
 - Cryptography:** Delve into the world of secure communication. Quantum cryptography is a cutting-edge field that leverages quantum mechanics to secure data.
 - Machine Learning:** Quantum machine learning could potentially solve complex problems faster than classical computers.
 - Chemical Simulations:** Quantum computers can simulate molecular and chemical interactions in ways classical computers can't, which could revolutionize medicine, materials science, and more.
 - Optimization Problems:** Tackle complex optimization problems found in logistics, finance, or resource management.
- Understand the Problem:** Once you've chosen a field, think deeply about a specific problem within that field that a quantum computer could help solve. Consider the problem's complexity and how quantum computing's unique capabilities like superposition and entanglement might offer a novel solution.



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Take your time to reflect on these options. What field and problem do you feel drawn to explore? Let me know your choice, and we'll move forward from there! 🌟💡🗨️



You

Machine learning



Get citation



Quantum Buddy

Fantastic choice! Machine Learning (ML) with quantum computing has immense potential and is a rapidly growing field. 🌐🤖

Development Stage

- Identify a Specific ML Problem:** Think about a specific problem within ML that you want to tackle using quantum computing. It could be anything from data classification, pattern recognition, to optimization in neural network training.
- Understand Quantum Machine Learning (QML):** Familiarize yourself with how quantum computing can be applied to ML. Quantum computers can process vast amounts of data more efficiently than classical computers, potentially leading to faster and more complex data analysis.

test and more complex data analysis.

3. **Start Coding:** Using the Qiskit library, you can start coding your algorithm. Don't worry if you're not an expert coder; I'm here to help you translate your quantum understanding into a working code. Begin by setting up your quantum environment and defining the basic structure of your algorithm.
4. **Test Basic Functionality:** As you develop, continually test your code for basic functionality. Ensure that the quantum aspects of your algorithm (like qubit operations and measurements) are working as expected.



Remember, this is a learning process. Don't hesitate to experiment, and feel free to ask for guidance at any step. Ready to start coding your Quantum Machine Learning algorithm? Let's code! 📄 📱 🚀



Get citation



You
data classification



Quantum Buddy
Data classification is a great choice for delving into Quantum Machine Learning (QML)! This involves categorizing data into different classes, making it a fundamental task in ML that can be significantly enhanced by quantum computing. 📄 📱

Development: Coding Your Quantum Data Classification Algorithm

1. **Setting Up Your Environment:**
 - Install Qiskit, if you haven't already. Use `pip install qiskit` in your Python environment.
 - Import necessary modules: `from qiskit import QuantumCircuit, Aer, execute`.
2. **Defining the Quantum Circuit:**
 - Create a quantum circuit with the required number of qubits. For a simple classification, start with 2-3 qubits.
 - Example: `qc = QuantumCircuit(2)` for a 2-qubit circuit.
3. **Encoding Data:**
 - Encode your data onto qubits. This can be done using various quantum gates (like Pauli-X, Y, Z, Hadamard gates) depending on your data.
 - For instance, use `qc.h(0)` to apply a Hadamard gate to the first qubit.
4. **Applying Quantum Operations:**
 - Design quantum operations that manipulate your data to perform classification. This might involve entangling qubits, applying more quantum gates, etc.
 - Example: `qc.cx(0, 1)` creates an entanglement between the first and second qubit.
5. **Measurement:**
 - Add measurement gates to read the outcome. This collapses the qubit states to either 0 or 1.
 - `qc.measure_all()` adds measurement to all qubits.
6. **Executing the Circuit:**
 - Run the circuit using a simulator or an actual IBM quantum computer.
 - Example: `result = execute(qc, Aer.get_backend('qasm_simulator')).result()`.
7. **Analyzing Results:**
 - Fetch the result and analyze the output to see how your data is classified.
 - `counts = result.get_counts(qc)`
8. **Iterative Improvement:**
 - Modify your circuit based on the results, experimenting with different quantum gates and operations to improve your classification.



Now, you're ready to start coding! Remember, quantum programming can be quite different from classical programming, so feel free to ask questions or seek guidance. Enjoy the process of discovery and learning! 🚀 🧑

Shall we begin coding your quantum data classification algorithm? Or do you have any specific questions or ideas you'd like to discuss first? 🗨️ 💬 🗣️



Get citation



You
let's start



Quantum Buddy
Great! Let's get started with coding your quantum data classification algorithm. Since we're working within a text-based format here, I'll guide you through the steps. You can write and execute this code in your Python environment where you have Qiskit installed.

Step-by-Step Coding Guide:

Get citation

1. Import Necessary Libraries:

```
python
from qiskit import QuantumCircuit, Aer, execute
from qiskit.visualization import plot_histogram
```

2. Create a Quantum Circuit:

- Here, we'll start with a simple 2-qubit circuit for demonstration purposes.

```
python
qc = QuantumCircuit(2)
```

3. Encode Data onto Qubits:

- For simplicity, let's use Hadamard gate ↓ create superpositions, representing our data.

```
python
qc.h(0) # Apply Hadamard gate to the first qubit
qc.h(1) # Apply Hadamard gate to the second qubit
```

4. Apply Quantum Operations for Classification:

- We'll use a CNOT gate as an example operation. It entangles the qubits, which is a fundamental aspect of quantum computing.

```
python
qc.cx(0, 1) # CNOT gate with qubit 0 as control and qubit 1 as target
```

5. Add Measurement Gates:

- This step is crucial to read the outcomes from the quantum circuit.

```
python
qc.measure_all()
```

6. Execute the Circuit on a Simulator:

- We'll use the QASM simulator to mimic the behavior of a quantum computer.

```
python
simulator = Aer.get_backend('qasm_simulator')
result = execute(qc, simulator).result()
```

7. Analyze the Results:

- Plotting a histogram to visualize the measurement outcomes.

```
python
counts = result.get_counts(qc)
plot_histogram(counts)
```

This code provides a basic structure for a quantum data classification algorithm. As you progress, you can modify this with different gates and more qubits to suit your specific classification problem.

Now, you can try running this code in your Python environment. Feel free to adjust it and experiment with different quantum gates and setups. If you have any questions or need further guidance, I'm here to help! 🌟💡

📄 👍 🗨



You

What am I classifying?



Quantum Buddy

In the context of this basic quantum data classification example, you're not classifying specific real-world data yet. Instead, you're learning the foundational steps on how to set up a quantum circuit for classification tasks. The aim here is to understand how to encode data into qubits, manipulate that data using quantum gates, and then read the results to make a classification.

In a more practical and advanced scenario, quantum data classification would involve:

- Encoding Real Data:** You would encode real-world data into qubits. This could be anything from images to text, converted into a format that a quantum computer can process. For instance, each feature of your dataset could be represented by the state of a qubit.
- Classifying the Data:** The quantum circuit you create (with its specific gates and operations) performs the task of classifying ↓ this data. Depending on the problem, this could mean grouping data into categories, identifying patterns, or making

predictions.

3. **Interpreting the Output:** The final step involves interpreting the quantum state of your qubits after running the circuit. This interpretation tells you which category or class your input data belongs to.

For example, in a binary classification problem, you might encode a piece of data onto a qubit and then use quantum operations to manipulate its state. The final state of the qubit (measured as 0 or 1) would then determine the class of that data.

As a starting point, the code and steps I've provided help you get comfortable with the mechanics of using Qiskit and setting up quantum circuits. Once you're familiar with these basics, you can then move on to encoding actual data and solving real classification problems.

Would you like to delve into how specific types of data can be encoded into qubits for classification, or do you have any other aspect of quantum machine learning you're curious about? 🌐💬💡



You

Can we use the iris dataset?



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Quantum Buddy

Certainly, the Iris dataset is a classic in machine learning and a great choice for exploring quantum data classification. It includes measurements of iris flowers' features and classifies them into three species. Here's how you can approach using the Iris dataset for quantum classification:

Steps to Use the Iris Dataset in Quantum Classification

1. **Understand the Dataset:**
 - The Iris dataset contains four features for each sample: sepal length, sepal width, petal length, and petal width.
 - There are three classes of iris species: Setosa, Versicolour, and Virginica.
2. **Preprocess the Data:**
 - **Normalize Data:** Quantum computers work with qubits that are fundamentally probabilistic. Hence, it's important to normalize your data so that it fits within the range a qubit can represent.
 - **Dimensionality Reduction:** Given the limited number of qubits in current quantum computers, it's practical to reduce the number of features. Techniques like PCA (Principal Component Analysis) can be used.
3. **Encoding the Data onto Qubits:**
 - After preprocessing, you need to encode this data onto qubits. Various encoding strategies exist, like amplitude encoding, where data points are encoded into the amplitudes of a quantum state.
4. **Building the Quantum Circuit:**
 - Design a quantum circuit that processes this encoded information. This will involve choosing the right quantum gates to manipulate your qubits effectively.
5. **Classification:**
 - After processing the data through the quantum circuit, measure the qubits. The measurement results can be used to classify the iris species.
6. **Interpreting Results:**
 - The output of your quantum circuit will need to be interpreted to map back to the class labels of the Iris dataset.



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Implementing with Qiskit

- Qiskit provides tools that can be helpful in this process, especially for building and simulating the quantum circuits.
- You'll need additional classical data processing (like PCA for dimensionality reduction) before encoding the data onto qubits.



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Example Code Outline

Here's a very simplified outline of what the code might look like. Note that this is conceptual and doesn't include specific implementation details:

```
python Copy code

# Import necessary libraries
from qiskit import QuantumCircuit, Aer, execute
from sklearn.datasets import load_iris
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA

# Load and preprocess the Iris dataset
iris = load_iris()
X = StandardScaler().fit_transform(iris.data) # Normalized data
```

```

X = StandardScaler().fit_transform(iris.data) # Normalize data
pca = PCA(n_components=2) # Reduce dimensions
X_pca = pca.fit_transform(X)

# Encoding data onto qubits
# This is a complex process and depends on the encoding strategy you choose

# Building the quantum circuit for classification
qc = QuantumCircuit(num_qubits)
# Add gates to manipulate qubits based on your algorithm

# Measure the qubits
qc.measure_all()

# Execute the circuit
simulator = Aer.get_backend('qasm_simulator')
result = execute(qc, simulator).result()

# Interpret results to classify the Iris data
# This will depend on how you've set up your quantum circuit

```

 Get citation

This code is a starting point and would need to be expanded with specific details, particularly in how you encode the Iris data onto qubits and how you interpret the results to classify the species.

Would you like more detailed guidance on any specific part of this process? 🌸 🧠 🤖



 Get citation



You

How can I interpret the results?



Quantum Buddy

Interpreting the results of a quantum circuit, especially for a task like data classification using the Iris dataset, involves mapping the measurement outcomes of your qubits to the class labels of the dataset. This can be a complex task, depending on how the data is encoded and processed in the quantum circuit. Here's a general approach:

Steps to Interpret Results in Quantum Data Classification

- Measurement Outcomes:**
 - After executing your quantum circuit, you'll get measurement outcomes for each qubit. These outcomes are typically in the form of '0's and '1's.
- Analyze the Probability Distribution:**
 - The results are often given as a probability distribution. For instance, if you have 2 qubits, you'll get probabilities for states $|00\rangle$, $|01\rangle$, $|10\rangle$, and $|11\rangle$.
 - The probability of each state corresponds to the likelihood of the data point belonging to a certain class.
- Decision Rule:**
 - Establish a decision rule for mapping these probabilities to class labels. This could be as simple as assigning the highest probability outcome to a class or more complex rules based on the specifics of your circuit.
 - For example, if $|00\rangle$ corresponds to Setosa, $|01\rangle$ to Versicolour, and $|10\rangle$ to Virginica, the class with the highest probability for a given measurement is the predicted class.
- Post-Processing:**
 - Sometimes, additional classical post-processing is needed to make a final determination. This could involve aggregating results from multiple runs or applying statistical methods.
- Accuracy Assessment:**
 - Compare the quantum classification results with the actual labels of the test data to assess accuracy.
 - Calculate metrics like accuracy, precision, recall, etc., to evaluate the performance of your quantum classifier.

 Get citation

Practical Considerations

- Noise and Errors:** Quantum computers are prone to noise and errors. This means that your interpretation might also need to consider error mitigation strategies.
- Limitations of Current Quantum Computers:** With current technology, the number of qubits and the coherence time are limited, which can affect the complexity of the problems you can solve and the accuracy of your results.

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Example of Interpreting a Simple 2-Qubit System:

Suppose you have a 2-qubit system where each qubit represents a binary classification. The results might look like this:





- $|00\rangle$ with a probability of 0.4

- $|01\rangle$ with a probability of 0.1
- $|10\rangle$ with a probability of 0.1
- $|11\rangle$ with a probability of 0.4

If $|00\rangle$ and $|11\rangle$ represent one class (say, Setosa), and $|01\rangle$ and $|10\rangle$ represent another (say, Versicolour), then you would interpret this result as a classification of Setosa or Versicolour based on the highest probability (in this case, a tie between Setosa and Versicolour).

This is a simplified example. In reality, the interpretation would be more complex and dependent on your specific encoding and classification strategy.

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Would you like to delve into more details or specific aspects of interpreting quantum circuit results?    



You

Can you write a quantum circuit to classify the iris dataset and check if the classification works?





Quantum Buddy

Creating a quantum circuit to classify the Iris dataset is a complex task that involves several steps, including data preprocessing, encoding, quantum circuit design, and interpretation of results. Given the limitations of this platform and the complexity of the task, I can guide you through a conceptual outline of how this might be done, but implementing and testing it would require a more advanced setup than what's available here.

Conceptual Outline for a Quantum Circuit to Classify the Iris Dataset

1. **Data Preprocessing:**
 - Normalize the Iris dataset features.
 - Use dimensionality reduction (like PCA) to reduce the dataset to a manageable number of features for quantum encoding.
2. **Data Encoding:**
 - Choose a quantum data encoding scheme. Amplitude encoding is a common choice but requires a circuit depth that might be impractical on current quantum computers. Another option is basis encoding, which is simpler but less efficient in terms of qubit usage.
3. **Quantum Circuit Design:**
 - Create a QuantumCircuit object in Qiskit.
 - Apply quantum gates to encode your preprocessed data into the state of the qubits.
 - Design and apply a sequence of gates (a variational circuit) that you'll optimize to classify the data points.
 - Add measurement gates to read the outcomes.
4. **Training the Circuit:**
 - Define a cost function to evaluate the performance of the circuit.
 - Use a classical optimizer to adjust the parameters of your variational circuit to minimize the cost function.
5. **Testing and Interpretation:**
 - Run the optimized circuit with a separate test dataset.
 - Interpret the output of the circuit to classify each data point.
 - Compare the quantum classification results with the actual labels to calculate accuracy.

 Get citation

 Get citation

Example Code Structure (Conceptual):

```
python
from qiskit import QuantumCircuit, Aer, execute
from sklearn.datasets import load_iris
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA

# Load the Iris dataset
iris = load_iris()
X = StandardScaler().fit_transform(iris.data) # Normalize
pca = PCA(n_components=2) # Reduce dimensions
X_pca = pca.fit_transform(X)

# Encoding data onto qubits might require a custom function

# Quantum Circuit for classification
qc = QuantumCircuit(num_qubits)
# Add gates for data encoding and variational circuit

# Add measurement gates
```

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```

qc.measure_all()

# Execute the circuit
backend = Aer.get_backend('qasm_simulator')
job = execute(qc, backend)
result = job.result()
counts = result.get_counts(qc)

# Interpret the results
# This requires a custom function based on your encoding and circuit de

# Compare results with actual labels to determine accuracy


```


This is a highly simplified and conceptual version. The actual implementation would be significantly more complex, particularly in data encoding and result interpretation. Also, the capability to fine-tune and optimize a quantum circuit for a specific task like this is an area of active research and development in quantum computing.

Due to the complexity and experimental nature of quantum machine learning, this sort of task typically requires iterative development and testing on actual quantum computing platforms, which goes beyond text-based guidance.

If you have access to a quantum computing environment and would like to try implementing this, I can provide further guidance on specific aspects of the process. Let me know how you'd like to proceed! 🌐 🛠️ 🌸

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 **You**
can you provide the actual implementation?

 **Quantum Buddy**
Creating an actual implementation of a quantum circuit to classify the Iris dataset is a complex and detailed task that involves advanced quantum programming skills and a deep understanding of both quantum computing and machine learning. While I can guide you through the steps and provide a basic framework, implementing and executing this would require a suitable quantum computing environment and extensive coding beyond what can be provided in this format.

 Get citation

Here's a high-level framework for how you might approach this:

1. Data Preprocessing

First, preprocess the Iris dataset. This includes normalizing the data and possibly reducing its dimensionality:

```

python
Copy code

from sklearn.datasets import load_iris
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA

# Load Iris dataset
iris = load_iris()
X, y = iris.data, iris.target

# Normalize the dataset
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

# Reduce dimensionality (optional, depending on your encoding strategy)
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X_scaled)

```

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2. Quantum Encoding

The next step is to encode this data onto a quantum state. This can be done in several ways, such as amplitude encoding or angle encoding. The method of encoding will depend on the number of qubits available and the nature of the quantum computer you are using.

3. Quantum Circuit Design

Design a quantum circuit that can process this encoded information. This could involve creating a parameterized circuit (a variational algorithm) which you'll optimize to classify the data:

```

python
Copy code

from qiskit import QuantumCircuit

```

```
# Number of qubits depends on your encoding
num_qubits = 2
qc = QuantumCircuit(num_qubits)

# Add gates to encode data
# This depends on your chosen encoding method

# Add a variational part (parameterized circuit)
# This is an example, the actual gates and their arrangement will vary
for qubit in range(num_qubits):
    qc.ry(numpy.random.rand(), qubit)

# Add measurement
qc.measure_all()
```

[Get citation](#)

4. Optimization

Optimize the parameters of your variational circuit using a classical optimization loop. This involves running the circuit many times and adjusting the parameters to find the minimum of a cost function that represents how well your circuit classifies the data.

5. Interpretation and Evaluation

After optimization, use your circuit to classify data points and compare the results with the known labels to assess the accuracy of your model. The interpretation of the results will depend on the encoding and the structure of your circuit.

[Get citation](#)

Implementation Considerations

- This process requires iterative testing and optimization.
- The performance and feasibility depend on the quantum hardware or simulator used.
- Quantum machine learning is a rapidly evolving field, and there are many research-level challenges in creating effective and efficient quantum classifiers.

Due to the complexity and the experimental nature of this field, implementing such a system usually involves a significant amount of trial and error, along with a deep understanding of both quantum computing and machine learning principles.

If you're interested in pursuing this project, I recommend starting with Qiskit's tutorials and documentation, which provide valuable resources and examples for quantum programming. This project is ambitious but offers a great opportunity to deepen your understanding of quantum computing and its applications in machine learning.



You

can you write the code to encode the data onto the quantum state?

[Get citation](#)


Quantum Buddy

Encoding classical data into a quantum state for machine learning tasks, like classifying the Iris dataset, is a key part of quantum computing. However, it's important to note that this is an advanced topic and the exact method of encoding can vary greatly depending on the specific requirements of your task and the limitations of your quantum hardware.

One common method of encoding classical data into quantum states is through **angle encoding**, where data values are used to set the angles in quantum gates. For simplicity, let's assume you've already preprocessed the Iris dataset (normalized and possibly reduced its dimensionality).

Example of Angle Encoding with Qiskit

Here's a conceptual Python code snippet using Qiskit to illustrate how you might encode a single data point from the Iris dataset onto a quantum state. This example assumes a 2-dimensional data point (after preprocessing) for simplicity:



```
python Copy code

from qiskit import QuantumCircuit
import numpy as np

# Example data point (after preprocessing and normalization)
data_point = np.array([0.7, 0.3]) # Replace with your actual preprocessed data

# Initialize Quantum Circuit with 2 qubits (for 2 features)
qc = QuantumCircuit(2)

# Encode data into quantum state using rotation gates
# Using the Y-rotation for this example
for i in range(len(data_point)):
    angle = np.pi * data_point[i] # Scale feature to [0, pi] range
    qc.ry(angle, i)
```

[Get citation](#)

```

angle = np.pi * data_point[i] # Scale feature to [0, pi] range
qc.ry(angle, i)

# Add more gates and operations as needed for your classification task

# Visualize the circuit
qc.draw('mpl')

```

[Get citation](#)

Points to Consider:

1. **Normalization:** Ensure your data is normalized so that the feature values are suitable for encoding into quantum gates (e.g., angles between 0 and π or 0 and 2π).
2. **Dimensionality:** The number of qubits needed depends on the dimensionality of your data. Higher-dimensional data will require more qubits or more sophisticated encoding strategies.
3. **Gate Choice:** Here, the 'ry' gate is used for encoding, but depending on your requirement, you might use different gates (like 'rz' or 'rx').
4. **Complex Encoding:** More complex encoding schemes can involve entangling qubits and more advanced quantum operations, especially for larger datasets with more features.

This code is a starting point and illustrates the concept of data encoding. The actual implementation might be more complex, especially for a complete machine learning model, and would require further development and optimization.

Executing and testing this code requires an environment where Qiskit is installed. If you're planning to run this, ensure you have Qiskit set up in your Python environment. Also, remember that quantum machine learning is an advanced and experimental field, so developing a fully functional quantum classifier involves much more than just encoding the data.

[Get citation](#)



You

Can you write the code to encode the iris dataset onto the qubits?



Quantum Buddy

Encoding the entire Iris dataset onto qubits for a quantum machine learning task is a complex process. In classical-quantum hybrid machine learning models, typically each data point is encoded and processed individually, rather than the entire dataset at once. This is due to the limitations in qubit numbers and quantum circuit complexity on current quantum computers.

Given these constraints, I'll provide an example of how you might encode individual data points from the Iris dataset using angle encoding. This example assumes you have preprocessed the Iris dataset (such as normalization and dimensionality reduction).

Example Python Code for Encoding Iris Dataset Data Points:

[Get citation](#)

```

python
Copy code

from sklearn.datasets import load_iris
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
from qiskit import QuantumCircuit
import numpy as np

# Load and preprocess the Iris dataset
iris = load_iris()
X, y = iris.data, iris.target
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

# Optionally, reduce dimensions (e.g., to 2 dimensions)
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X_scaled)

# Function to create a quantum circuit from a data point
def create_quantum_circuit(data_point):
    num_features = len(data_point)
    qc = QuantumCircuit(num_features)

    # Encode each feature onto a qubit using angle encoding
    for i in range(num_features):
        angle = np.pi * data_point[i] # Map feature value to [0, pi] range
        qc.ry(angle, i)

    return qc

# Example: Encode the first data point

```

[Get citation](#)

```
qc = create_quantum_circuit(X_pca[0])

# Visualize the circuit
qc.draw(output='mpl')
```

Notes:

1. **Preprocessing:** The Iris dataset is normalized and optionally reduced in dimensionality. Here, PCA is used to reduce it to 2 features for simplicity.
2. **Angle Encoding:** Each feature of a data point is encoded into the angle of a rotation gate (' ry '). The function '`create_quantum_circuit`' takes a data point and returns a quantum circuit encoding that point.
3. **Visualization:** The '`draw`' method visualizes the quantum circuit for the first data point.
4. **Iterative Processing:** In a real-world application, you would loop over data points, creating and processing a quantum circuit for each.
5. **Quantum Hardware Limitations:** The example assumes a simplified scenario suitable for current quantum computers. Real-world datasets often require more complex encoding strategies and more qubits.
6. **Further Development:** This example focuses on data encoding. Building a complete quantum machine learning model would involve additional steps, such as defining a parameterized circuit for classification and optimizing its parameters.

Remember, this is a conceptual demonstration. Actual implementation, especially for a quantum machine learning task, would be more involved and require a detailed understanding of both the quantum computing aspects and the specificities of the dataset and the machine learning task.

< 2 / 2 >    

Was this response better
or worse?

 Better

 Worse

 Same



 Get citation

 Get citation



Message Quantum Buddy...



ChatGPT can make mistakes. Consider checking important information.



Collaborate on a Team plan

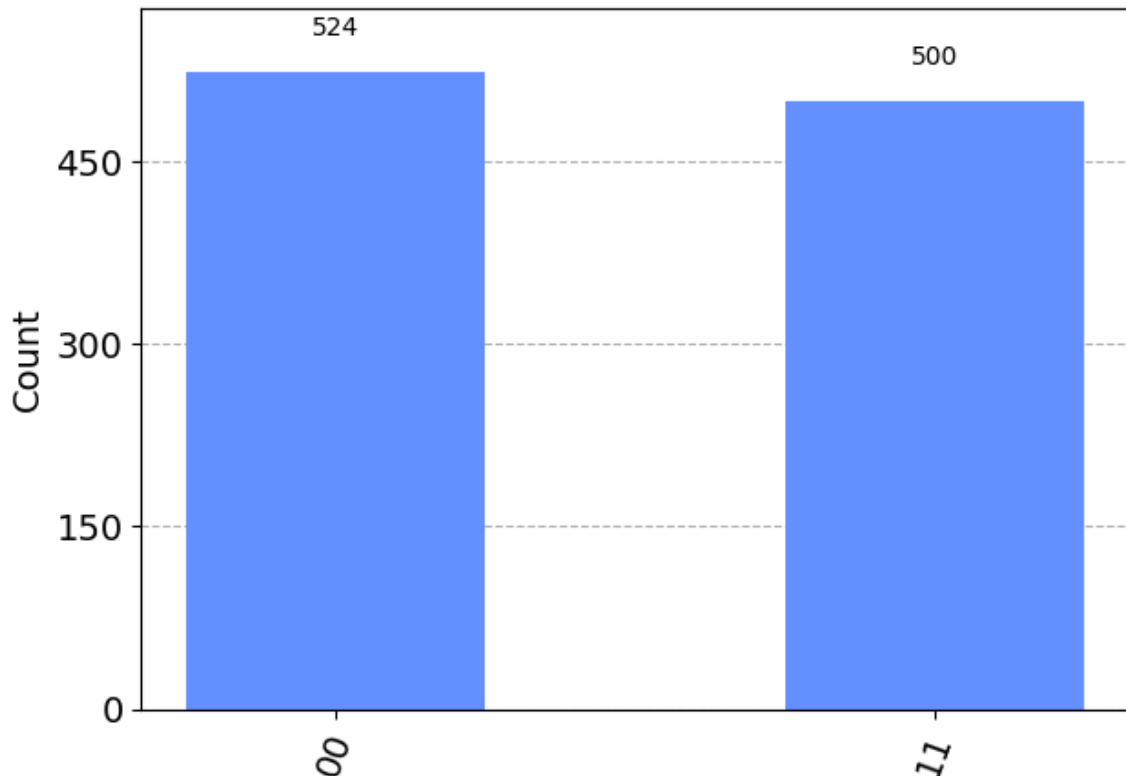


co doc

```
In [3]: from qiskit import QuantumCircuit, Aer, execute
from qiskit.visualization import plot_histogram

qc = QuantumCircuit(2)
qc.h(0) # Apply Hadamard gate to the first qubit
qc.h(1) # Apply Hadamard gate to the second qubit
qc.cx(0, 1) # CNOT gate with qubit 0 as control and qubit 1 as target
qc.measure_all()
simulator = Aer.get_backend('qasm_simulator')
result = execute(qc, simulator).result()
counts = result.get_counts(qc)
plot_histogram(counts)
```

Out[3]:



```
In [5]: # Import necessary Libraries
from qiskit import QuantumCircuit, Aer, execute
from sklearn.datasets import load_iris
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA

# Load and preprocess the Iris dataset
iris = load_iris()
X = StandardScaler().fit_transform(iris.data) # Normalize data
pca = PCA(n_components=2) # Reduce dimensions
X_pca = pca.fit_transform(X)

# Encoding data onto qubits
# This is a complex process and depends on the encoding strategy you choose

# Building the quantum circuit for classification
qc = QuantumCircuit(2)
# Add gates to manipulate qubits based on your algorithm

# Measure the qubits
qc.measure_all()
```

```
# Execute the circuit
simulator = Aer.get_backend('qasm_simulator')
result = execute(qc, simulator).result()

# Interpret results to classify the Iris data
# This will depend on how you've set up your quantum circuit
```

In [6]: result

```
Out[6]: Result(backend_name='qasm_simulator', backend_version='0.12.2', qobj_id='', job_id='4e19e20b-19ce-479f-be26-c201e25e44e4', success=True, results=[ExperimentResult(shots=1024, success=True, meas_level=2, data=ExperimentResultData(counts={'0x0': 1024}), header=QobjExperimentHeader(creg_sizes=[['meas', 2]], global_phase=0.0, memory_slots=2, n_qubits=2, name='circuit-127', qreg_sizes=[['q', 2]], metadata={}), status=DONE, seed_simulator=2346748168, metadata={'noise': 'ideal', 'batched_shots_optimization': False, 'measure_sampling': True, 'parallel_shots': 1, 'remapped_qubits': False, 'active_input_qubits': [0, 1], 'num_clbits': 2, 'parallel_state_update': 8, 'sample_measure_time': 0.000879742, 'num_qubits': 2, 'device': 'CPU', 'input_qubit_map': [[1, 1], [0, 0]], 'method': 'stabilizer', 'fusion': {'enabled': False}}, time_taken=0.002871099)], date=2024-01-12T11:51:25.631498, status=COMPLETED, header=None, metadata={'time_taken_execute': 0.002964436, 'mpi_rank': 0, 'num_mpi_processes': 1, 'max_gpu_memory_mb': 0, 'max_memory_mb': 31890, 'parallel_experiments': 1, 'num_processes_per_experiments': 1, 'omp_enabled': True}, time_taken=0.00368499755859375)
```

```
In [1]: from sklearn.datasets import load_iris
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
from qiskit import QuantumCircuit
import numpy as np

# Load and preprocess the Iris dataset
iris = load_iris()
X, y = iris.data, iris.target
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

# Optionally, reduce dimensions (e.g., to 2 dimensions)
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X_scaled)

# Function to create a quantum circuit from a data point
def create_quantum_circuit(data_point):
    num_features = len(data_point)
    qc = QuantumCircuit(num_features)

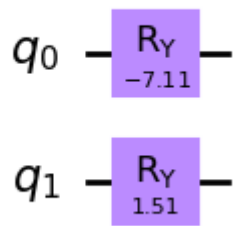
    # Encode each feature onto a qubit using angle encoding
    for i in range(num_features):
        angle = np.pi * data_point[i] # Map feature value to [0, π] range
        qc.ry(angle, i)

    return qc

# Example: Encode the first data point
qc = create_quantum_circuit(X_pca[0])

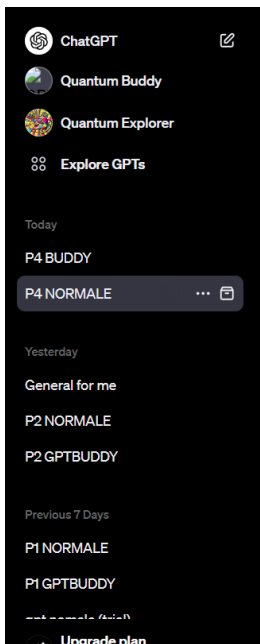
# Visualize the circuit
qc.draw(output='mpl')
```

Out[1]:



In []:

P4



ChatGPT 4



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.



ChatGPT

Great! Let's get started with the structured user test. The aim of this test is to conceptualize, develop, and refine an algorithm that utilizes the unique capabilities of quantum computing. This test is designed to help you leverage your background in quantum mechanics and coding to create something innovative and impactful. Are you ready to dive into the world of quantum algorithms? Let me know when you're all set to go! 🤖🚀

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You

yes i am ready



ChatGPT

Fantastic! Let's begin with the **Conceptualization** stage. 🌟

1. **Choose a Field:** First, you need to decide on the field you're interested in. Here are some options:
 - **Cryptography:** Developing secure communication methods.
 - **Machine Learning:** Enhancing data analysis and pattern recognition.
 - **Chemical Simulations:** Simulating molecular and chemical interactions.
 - **Optimization Problems:** Solving complex logistical or planning challenges.
2. **Understand the Problem:** Once you've chosen a field, think deeply about the specific problem you want to solve. It should be a problem where quantum computing can offer significant advantages over classical computing.
3. **Research:** Do a bit of research to understand the current state of quantum algorithms in your chosen field. This will help you identify gaps or opportunities for innovation.
4. **Idea Generation:** Brainstorm ideas for your algorithm. Don't worry about feasibility at this stage; just let your creativity flow!

Which field excites you the most? Let me know your choice and the specific problem you're interested in tackling, and we'll move forward from there! 🤖👤💡

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You

I would like to do something with Optimization Problems



ChatGPT

That's a great choice! Optimization problems are a fascinating area where quantum computing can make a significant impact. 🌐🚀

Now, let's narrow down your focus within Optimization Problems:

1. **Identify a Specific Problem:** Think about a particular optimization challenge you're interested in. This could be anything from route optimization in logistics, portfolio optimization in finance, scheduling problems, or even complex problems like optimizing energy consumption in smart grids.
2. **Understand the Complexity:** Research why this problem is challenging and how traditional computing methods handle it. This understanding will help you appreciate where quantum computing can offer an edge.
3. **Explore Quantum Approaches:** Look into existing quantum algorithms that are used for optimization problems. Familiarize yourself with concepts like Quantum Annealing or Grover's Algorithm, which are often used for optimization.
4. **Brainstorming Session:** Consider how you might approach this problem with a quantum algorithm. Think outside the box and jot down any ideas that come to mind,

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no matter how rough they might seem.

Could you let me know which specific optimization problem you'd like to tackle? Once you've decided, we can move on to the next stage, which is developing your algorithm. I'm here to guide you through each step! 🤖👤💡



You

I want to find the most efficient routes for distributing goods to hundreds of retail locations



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That's an excellent and practical problem to tackle! You've chosen to work on optimizing distribution routes for goods to hundreds of retail locations. This is a classic example of a complex optimization problem, often referred to as a variant of the "Vehicle Routing Problem" (VRP) in logistics. Quantum computing can potentially offer more efficient solutions than traditional methods, especially as the scale and complexity of the problem increase. 🧠💡

Development Stage

Now, let's move on to the **Development** stage where you'll start coding your algorithm. Here's how you can approach it:

1. **Understand the Quantum Approach:** Research how quantum algorithms can be used for route optimization. You might want to look into quantum versions of optimization algorithms, like quantum annealing or variants of Grover's Algorithm.
2. **Algorithm Design:** Start designing your algorithm. This includes defining the quantum states, the quantum operations (↓ es), and how these will evolve to find the optimal solution.
3. **Coding with Guidance:** Begin coding your algorithm. You can use quantum computing frameworks like Qiskit (for IBM quantum computers), Cirq (for Google's quantum processors), or others. Don't worry if you're not an expert in coding quantum algorithms. I'm here to help translate your ideas and understanding into working code! 💻🔗
4. **Iterative Approach:** Remember, coding a quantum algorithm is an iterative process. Test small parts of your code regularly to ensure they work as expected.

Shall we start with the basic structure of your quantum algorithm for route optimization? Let me know if you need help with specific aspects like setting up the quantum states or implementing quantum operations! 🤖👤💡



Get citation



You

Yes let's start.



ChatGPT

Great! Let's get started on building the basic structure of your quantum algorithm for route optimization. I'll guide you through a ↓ plified version of the process, and we can refine it as we go. Here's a step-by-step approach:



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1. Framework and Environment Setup

- **Choose a Quantum Computing Framework:** For this example, we'll use Qiskit, which is widely used and well-documented. It's developed by IBM and is suitable for running on IBM's quantum computers.
- **Setup:** Ensure you have Python installed, and then install Qiskit. You can do this using pip:

```
python
```

Copy code

```
pip install qiskit
```

2. Define the Problem

- **Problem Statement:** You want to optimize delivery routes to hundreds of retail locations. This involves determining the shortest or most efficient path that visits each location exactly once and returns to the origin (a variant of the VRP).

3. Quantum Circuit Creation



- **Initialize Quantum Circuit:** Create a quantum circuit with enough qubits to represent your problem. The number of qubits will depend on how you encode the problem.

```
python
```

Copy code

```
from qiskit import QuantumCircuit
num_qubits = # (define based on problem encoding)
circuit = QuantumCircuit(num_qubits)
```



Get citation

4. Problem Encoding

- **Encode Your Problem:** Convert the VRP into a quantum format. One approach is to use QUBO (Quadratic Unconstrained Binary Optimization) formulation, where you define a binary variable for each possible route and condition.

5. Implement Quantum Operations

- **Apply Gates:** Apply quantum gates to entangle qubits and explore the solution space. For optimization, you might involve Grover's algorithm or quantum annealing techniques.

6. Measurement



- **Measure:** After the quantum operations, measure the qubits to collapse their state to classical bits. This measurement gives you a possible solution to your problem.

```
python Copy code  
  
circuit.measure_all()
```

7. Run the Circuit

- **Execute on a Simulator/Quantum Computer:** Initially, test your circuit on a quantum simulator. Later, you can run it on an actual quantum computer.

```
python Copy code  
  
from qiskit import Aer, execute  
simulator = Aer.get_backend('qasm_simulator')  
result = execute(circuit, simulator).result()
```

8. Interpret Results

- **Analyze Output:** Analyze the output to see if it represents a valid and efficient solution. Iterate and refine as needed.



Would you like to start coding this, or do you need more detailed guidance on any of the steps? Remember, this is an iterative and experimental process, so feel free to ask questions and explore different approaches! 🤖🔧👤



You

I have this error: Traceback (most recent call last):
Cell In[2], line 3
 result = execute(circuit, simulator).result()
NameError: name 'circuit' is not defined

Use %tb to get the full traceback



ChatGPT

The error message you're encountering, 'NameError: name 'circuit' is not defined', indicates that the Python interpreter does not recognize the variable 'circuit' when it's used in the line 'result = execute(circuit, simulator).result()'. This typically happens if the variable 'circuit' was either not defined at all, or not defined in the same scope where it's being used.



To resolve this error, ensure that you have properly defined and initialized the 'circuit' variable in your code. Here's a quick checklist and example code to help you out:

1. **Import Necessary Modules:** Make sure you've imported Qiskit and other required modules.

```
python Copy code  
  
from qiskit import QuantumCircuit, Aer, execute
```

2. **Define the Number of Qubits:** Decide on the number of qubits you need for your problem.

```
python Copy code  
  
num_qubits = 5 # Example number, adjust as needed
```

3. **Create the Quantum Circuit:** Initialize the 'circuit' variable with the 'QuantumCircuit' class.

```
python Copy code  
  
circuit = QuantumCircuit(num_qubits)
```



Get citation



Get citation



Get citation



Get citation

4. **Add Quantum Operations:** Apply the necessary quantum gates to your circuit (not shown here, as it's problem-specific).
5. **Measure the Qubits:** Add measurement to your circuit.

```
python Copy code  
  
circuit.measure_all()
```

6. **Run the Circuit:**

```
python Copy code  
  
simulator = Aer.get_backend('qasm_simulator')  
result = execute(circuit, simulator).result()
```

Here's how your code should look with these elements combined:

```
python Copy code  
  
from qiskit import QuantumCircuit, Aer, execute  
  
num_qubits = 5 # Adjust this based on your problem  
circuit = QuantumCircuit(num_qubits)  
# Add your quantum operations here  
  
circuit.measure_all()  
simulator = Aer.get_backend('qasm_simulator')  
result = execute(circuit, simulator).result()
```

[Get citation](#)

Make sure all these steps are included in your code and executed in the same environment or Jupyter Notebook session. If you're still encountering issues, please let me know more details about your code, and I'll be happy to help further! 🙌🔍🔧



You

I still have an error: Traceback (most recent call last):
Cell In[6], line 2
result = execute(circuit, simulator).result()
NameError: name 'circuit' is not defined



ChatGPT

The error you're encountering suggests there's a disconnection between the definition of the `'circuit'` variable and where you're trying to use it in the `'execute'` function. This can happen due to several reasons in a Python environment, especially in Jupyter Notebooks. Here are a few things to check and try:

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1. **Sequential Execution:** In Jupyter Notebooks, cells need to be executed in sequence. Ensure that the cell where `'circuit'` is defined is executed before the cell where you're trying to use it.
2. **Same Session:** Ensure that all cells are executed in the same session. If you restart the kernel or open a new session, you'll need to re-execute all cells starting from the beginning.
3. **Typo or Naming Issue:** Double-check for any typos in the variable name `'circuit'` across different cells. The name must be consistent.
4. **Re-Execute Previous Cells:** If you've made changes to the cell where `'circuit'` is defined, remember to re-execute that cell to update the variable in the current session.
5. **Code Integration:** If possible, combine the definition of `'circuit'` and its usage into a single cell to avoid scope issues. This way, you ensure that `'circuit'` is defined and used in the same scope.

Try integrating your code as follows in a single cell:


[Get citation](#)


```
python Copy code  
  
from qiskit import QuantumCircuit, Aer, execute  
  
# Define the number of qubits  
num_qubits = 5 # Adjust this number based on your problem  
# Create the Quantum Circuit  
circuit = QuantumCircuit(num_qubits)  
# Add quantum operations here (if any)  
  
# Measure all qubits  
circuit.measure_all()  
# Execute the circuit on a simulator  
simulator = Aer.get_backend('qasm_simulator')  
result = execute(circuit, simulator).result()
```

```
# Further processing of result (if needed)
```

Run this integrated code in a single cell and see if the error persists. If the error is still there, please let me know, and we can explore further steps to resolve it.

 Get citation

 **You**
can you write the full code again for me?

 **ChatGPT**
Absolutely! Let's compile the full code, starting from importing the necessary modules to running the quantum circuit on a simulator. This example assumes you're using Qiskit and sets up a basic quantum circuit. We'll include comments to guide you through each step. Ensure you execute this entire block of code in a single Jupyter Notebook cell:

```
python Copy code

# Import necessary modules from Qiskit
from qiskit import QuantumCircuit, Aer, execute

# Define the number of qubits for your quantum circuit
# Adjust this number based on your problem's requirements
num_qubits = 5

# Create a quantum circuit with the defined number of qubits
circuit = QuantumCircuit(num_qubits)

# (Optional) Add quantum gates and operations here
# For example, apply a Hadamard gate to each qubit
for qubit in range(num_qubits):
    circuit.h(qubit)

# Measure all qubits
circuit.measure_all()

# Choose a simulator to run your quantum circuit
simulator = Aer.get_backend('qasm_simulator')

# Execute the circuit on the chosen simulator
result = execute(circuit, simulator).result()

# Get the measurement results from the executed circuit
measurements = result.get_counts(circuit)


# Print the results
print("Measurement results:")
print(measurements)
```

 Get citation

 Get citation

This code sets up a simple quantum circuit with a specified number of qubits, applies a Hadamard gate to each qubit (creating a superposition), measures all the qubits, and then runs the circuit on a quantum simulator. The final output is the measurement results of the qubits.

Remember to adjust 'num_qubits' based on the specific requirements of your problem, and replace the quantum operations section with the gates and operations that are relevant to your optimization problem. If you have any specific quantum gates or operations in mind for your optimization problem, let me know, and I can help you integrate them into the code!

 **You**
Can you give me the code to visualise the results?

 **ChatGPT**
 Error analyzing

It appears that I can't directly run the code involving Qiskit in this environment. However, I can certainly provide you with the Python code that you can run in your local environment to visualize the results of your quantum circuit.

 Get citation

To visualize the measurement results from your quantum circuit, you can use the 'plot_histogram' function from Qiskit's visualization module. Here's how you can do it:

```
python Copy code

from qiskit.visualization import plot_histogram
import matplotlib.pyplot as plt
```

```

import matplotlib.pyplot as plt

# Assuming 'measurements' contains the result from the quantum circuit
# Use the 'measurements' variable you obtained from the circuit execution
# For example: measurements = result.get_counts(circuit)

# Visualize the results as a histogram
plot = plot_histogram(measurements)

# To display the plot in a Jupyter notebook, use:
plt.show(plot)

# Alternatively, to save the plot to a file, use:
# plot.savefig("measurement_results.png")

```

This code will create a histogram of your measurement results, where each bar represents the frequency of a particular quantum state. Make sure you have the `matplotlib` library installed for plotting. You can install it using `pip install matplotlib` if you haven't already.

Run this code in the same environment where you're running your Qiskit code to visualize the results of your quantum circuit. If you encounter any issues or have further questions, feel free to ask! 😊



Get citation



Message ChatGPT...



ChatGPT can make mistakes. Consider checking important information.

?

Collaborate on a Team plan



co doc

```
In [9]: # Import necessary modules from Qiskit
from qiskit import QuantumCircuit, Aer, execute

# Define the number of qubits for your quantum circuit
# Adjust this number based on your problem's requirements
num_qubits = 5

# Create a quantum circuit with the defined number of qubits
circuit = QuantumCircuit(num_qubits)

# (Optional) Add quantum gates and operations here
# For example, apply a Hadamard gate to each qubit
for qubit in range(num_qubits):
    circuit.h(qubit)

# Measure all qubits
circuit.measure_all()

# Choose a simulator to run your quantum circuit
simulator = Aer.get_backend('qasm_simulator')

# Execute the circuit on the chosen simulator
result = execute(circuit, simulator).result()

# Get the measurement results from the executed circuit
measurements = result.get_counts(circuit)

# Print the results
print("Measurement results:")
print(measurements)
```

Measurement results:

```
{'01001': 34, '00111': 45, '11010': 30, '10001': 36, '00101': 24, '00011': 53,
'01011': 31, '10010': 32, '10101': 35, '11100': 32, '01101': 34, '00100': 42,
'11000': 35, '10000': 26, '11111': 37, '11011': 28, '11001': 25, '00000': 37,
'10011': 43, '01010': 17, '01000': 27, '01100': 28, '11101': 24, '10110': 30,
'00001': 27, '00110': 24, '10111': 25, '10100': 35, '00010': 27, '11110': 33,
'01111': 37, '01110': 31}
```

```
In [10]: from qiskit.visualization import plot_histogram
import matplotlib.pyplot as plt

# Assuming 'measurements' contains the result from the quantum circuit execution
# Use the 'measurements' variable you obtained from the circuit execution
# For example: measurements = result.get_counts(circuit)

# Visualize the results as a histogram
plot = plot_histogram(measurements)

# To display the plot in a Jupyter notebook, use:
plt.show(plot)

# Alternatively, to save the plot to a file, use:
# plot.savefig("measurement_results.png")
```

In []:

ChatGPT

Quantum Buddy

Quantum Explorer

Explore GPTs

Today

P4 BUDDY

P4 NORMALE

Yesterday

General for me

P2 NORMALE

P2 GPTBUDDY

Previous 7 Days

P1 NORMALE

P1 GPTBUDDY

Upgrade plan



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.

Quantum Buddy

Great, let's get started with your structured user test in quantum computing! The aim of this test is to harness your knowledge in quantum mechanics and coding to create something extraordinary. You'll be conceptualizing, developing, and refining a quantum computing algorithm. This isn't just any algorithm, but one that truly taps into the unique strengths of quantum computing.

Are you ready to dive into the fascinating world of quantum algorithms and make your mark? Let me know when you're all set to begin! 🤖🔧👤👤

Get citation

You

I'm ready

Quantum Buddy

Awesome! Let's embark on this exciting journey together. 🤖🌟

Step 1: Conceptualization

First up, we need to pick a field and identify a specific problem you want to tackle using quantum computing. Here are some intriguing options:

- Cryptography:** Dive into the world of secure communication. Think about creating or breaking quantum-resistant encryption algorithms.
- Machine Learning:** Explore quantum algorithms that could enhance machine learning tasks, like data classification or pattern recognition.
- Chemical Simulations:** Tackle complex chemical problems which are intractable for classical computers, such as simulating molecular structures or reactions.
- Optimization Problems:** Address challenges in logistics, finance, or other areas where finding the optimal solution among many possibilities is key.

Think about which area excites you the most and the specific problem within that field you're eager to solve. Let me know your choice, and we'll proceed to the next exciting phase of developing your quantum algorithm! 🤖💡🔍

Get citation

You

I want to do something with optimization problems, in particular with I want to finding the most efficient routes for distributing goods to hundreds of retail locations

Quantum Buddy

That's a fantastic choice! Optimization problems, particularly in logistics, are a perfect playground for quantum computing. Finding the most efficient routes for distributing goods to numerous retail locations is a classic example where quantum algorithms can shine. This challenge is akin to the famous "Traveling Salesman Problem," but on a much larger and more complex scale. 🤖🗺️📦

Get citation

Step 2: Development


Now, let's start coding your algorithm. Here's a simplified roadmap to follow:


- Understand the Quantum Formulation:** You'll need to represent your problem in a way that a quantum computer can understand. This usually involves formulating it as a QUBO (Quadratic Unconstrained Binary Optimization) problem or using Ising models.
- Set Up Your Environment:** Make sure you have Qiskit installed, IBM's open-source quantum computing software development framework.

3. **Initialize Your Quantum Circuit:** Using Qiskit, you'll create a quantum circuit that's designed to explore all possible routes and find the most efficient one.
4. **Apply Quantum Gates:** You'll apply a series of quantum gates to entangle qubits and explore superpositions, which are crucial for searching through multiple possibilities simultaneously.
5. **Measure and Interpret Results:** Finally, you'll measure the state of your qubits to get the solution. Due to quantum nature, you might need to run the algorithm multiple times to find the most probable (and efficient) solution.

[Get citation](#)

Don't worry if some of these steps seem complex. I'm here to guide you through each one. Start by installing Qiskit and setting up a basic quantum circuit, and we can go from there. Are you ready to begin coding, or do you need help with any specific step? Let's code this out! 🖥️🔧🌐

 **You**
yes, we can start

 **Quantum Buddy**
Fantastic! Let's get started with the development phase. Since you're focusing on solving an optimization problem in logistics, specifically finding the most efficient routes for distributing goods, we'll be using quantum computing to tackle this challenge.

Setting Up Your Development Environment

1. **Install Qiskit:** If you haven't already, you'll need to install Qiskit, IBM's quantum computing framework. You can do this via `pip`:

```
python  
  
pip install qiskit
```

[Copy code](#)[Get citation](#)

2. **Import Necessary Libraries:** In your Python environment, start by importing the necessary modules from Qiskit:

```
python  
  
from qiskit import QuantumCircuit, execute, Aer  
from qiskit.visualization import plot_histogram
```

[Copy code](#)

Building Your Quantum Circuit

1. **Initialize the Quantum Circuit:** Create a quantum circuit with the number of qubits based on your problem size. For simplicity, let's start with a smaller version of your problem:

```
python  
  
n_qubits = 5 # This number will depend on the scale of your problem  
circuit = QuantumCircuit(n_qubits, n_qubits)
```

[Copy code](#)

2. **Apply Quantum Gates:** We'll apply a series of gates to manipulate the state of the qubits. This is a simplified example:

```
python  
  
for qubit in range(n_qubits):  
    circuit.h(qubit) # Applying the Hadamard gate to each qubit  
circuit.barrier()
```

[Copy code](#)[Get citation](#)

3. **Measurement:** Finally, we'll add measurement operations to read the state of the qubits at the end of the computation:

```
python  
  
circuit.measure(range(n_qubits), range(n_qubits))
```

[Copy code](#)

Running Your Circuit

1. **Execute on a Simulator:** Let's run this circuit on a quantum simulator first:

```
python  
  
simulator = Aer.get_backend('qasm_simulator')  
job = execute(circuit, simulator, shots=1000)  
result = job.result()  
counts = result.get_counts(circuit)
```

[Copy code](#)[Get citation](#)

2. **Visualize Results:** To understand the output, you can visualize the results:

```
python  
  
plot_histogram(counts)
```

[Copy code](#)

This is a basic setup for a quantum circuit. The actual algorithm for your distribution optimization will be more complex, involving formulating your problem as a QUBO and encoding it into the quantum circuit.

Remember, quantum computing for optimization problems like yours often uses specialized algorithms like the Quantum Approximate Optimization Algorithm (QAOA).

Are you ready to dive deeper into the specifics of your problem, or do you need any clarification on these initial steps? Let's keep the momentum going! 🧠💡🚀



You

I have several problems: first:

Cell In[2], line 2

```
circuit = QuantumCircuit(n_qubits, n_qubits)
```

NameError: name 'QuantumCircuit' is not defined



Get citation



Quantum Buddy

It looks like you're encountering a `NameError` because the `QuantumCircuit` class is not recognized. This usually happens if the Qiskit library hasn't been imported correctly. Let's troubleshoot this step by step:

1. **Ensure Qiskit is Installed:** First, make sure Qiskit is installed in your Python environment. You can check this by running:

python

Copy code

```
pip list | grep qiskit
```

If Qiskit isn't listed, install it using:

python

Copy code

```
pip install qiskit
```



2. **Import Qiskit Correctly:** At the beginning of your Python script or notebook, make sure you have the following import statement:

python

Copy code

```
from qiskit import QuantumCircuit, Aer, execute
```

This line imports the `QuantumCircuit` class along with other necessary components from Qiskit.

3. **Retry Creating the Circuit:** After ensuring the above steps, try creating the quantum circuit again:

python

Copy code

```
n_qubits = 5 # Adjust the number of qubits based on your problem
circuit = QuantumCircuit(n_qubits, n_qubits)
```

If you've followed these steps and still encounter an issue, there might be a problem with your Python environment or how Qiskit was installed. In such a case, you might need to check your environment settings or reinstall Qiskit.

Let me know if this resolves the issue or if I need further assistance. We'll make sure to get your quantum circuit up and running! 🧠💡🚀



Get citation



You

okay, this is fixed but now I need this: Traceback (most recent call last):

Cell In[13], line 1

```
plot_histogram(counts)
```

NameError: name 'plot_histogram' is not defined

Use %tb to get the full traceback.



Quantum Buddy

The error you're encountering now indicates that the function `plot_histogram` is not recognized in your current environment. This is likely because the visualization tools from Qiskit haven't been imported. Here's how you can resolve this:

1. **Import Visualization Tools:** At the beginning of your script or notebook, along with other Qiskit imports, add the following line to import the visualization functions:

python

Copy code

```
from qiskit.visualization import plot_histogram
```



Get citation

This line will import the `plot_histogram` function that is used for visualizing the output of quantum circuits.

2. **Retry Plotting:** After importing `plot_histogram`, try running the plot command again:

```
python
```

[Copy code](#)

```
plot_histogram(counts)
```

This should display a histogram of the measurement results from your quantum circuit.

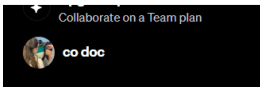
If you continue to encounter issues, it might be helpful to check that all necessary Qiskit components are installed and imported correctly. Sometimes, different versions of Qiskit or the Python environment can cause unexpected behavior, so ensuring everything is up to date can be beneficial.

Let me know if this solves the problem or if there's anything else you need help with.

We're getting closer to seeing your quantum circuit in action! 🇮🇹🔧💡

[Get citation](#)

ChatGPT can make mistakes. Consider checking important information.



In [7]: `pip install qiskit`

```
Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.44.3)
Requirement already satisfied: qiskit-terra==0.25.3 in /opt/conda/lib/python3.10/site-packages (from qiskit) (0.25.3)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (4.5.0)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (1.11.1)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (2.8.2)
Requirement already satisfied: symengine<0.10,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (0.9.2)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (3.11)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (0.13.0)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (5.9.4)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (4.1.1)
Requirement already satisfied: scipy>=1.5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (1.9.3)
Requirement already satisfied: numpy>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (1.23.5)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.25.3->qiskit) (0.3.7)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra==0.25.3->qiskit) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra==0.25.3->qiskit) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra==0.25.3->qiskit) (1.3.0)
```

[notice] A new release of pip available: 23.1.1 -> 23.3.2

[notice] To update, run: `pip install --upgrade pip`

Note: you may need to restart the kernel to use updated packages.

In [8]: `from qiskit import QuantumCircuit, Aer, execute`

In [9]: `n_qubits = 5 # Adjust the number of qubits based on your problem`
`circuit = QuantumCircuit(n_qubits, n_qubits)`

In [10]: `for qubit in range(n_qubits):`
 `circuit.h(qubit) # Applying the Hadamard gate to each qubit`
`circuit.barrier()`

Out[10]: `<qiskit.circuit.instructionset.InstructionSet at 0x7f68b85acf70>`

In [11]: `circuit.measure(range(n_qubits), range(n_qubits))`

Out[11]: `<qiskit.circuit.instructionset.InstructionSet at 0x7f68b85ac8b0>`

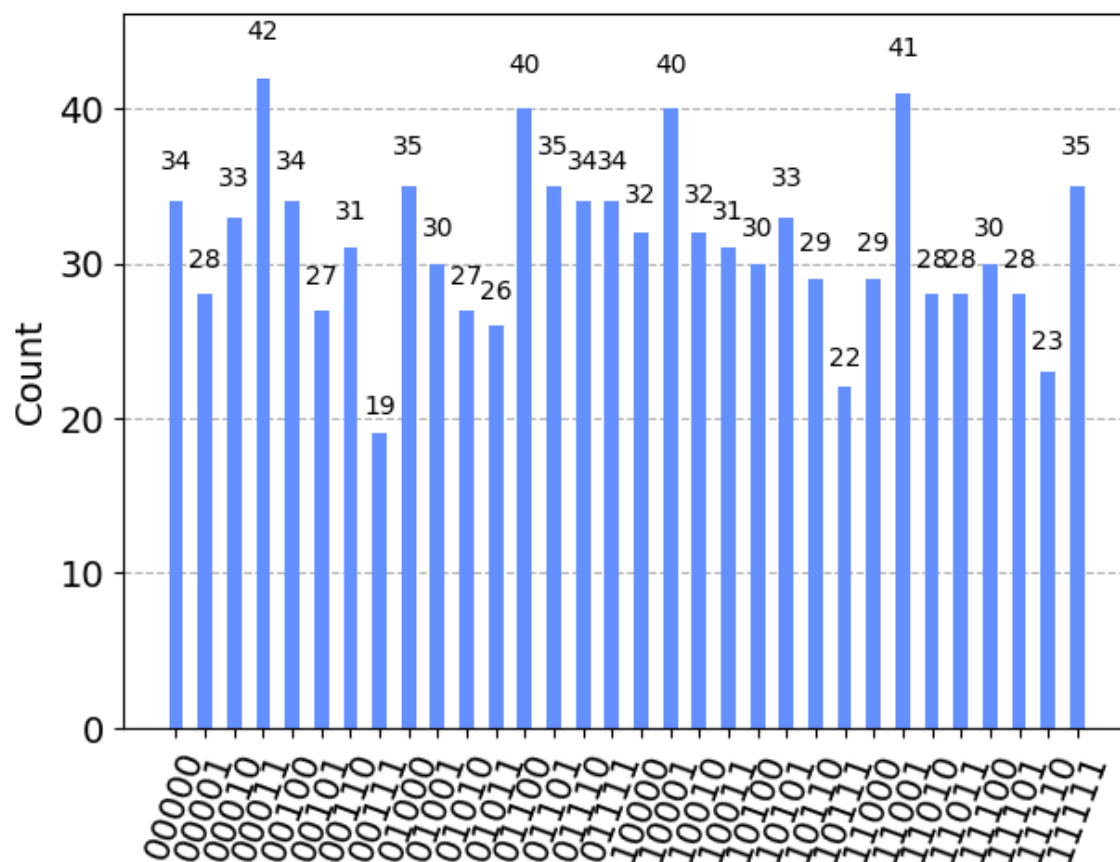
In [12]: `simulator = Aer.get_backend('qasm_simulator')`
`job = execute(circuit, simulator, shots=1000)`

```
result = job.result()
counts = result.get_counts(circuit)
```

```
In [14]: from qiskit.visualization import plot_histogram
```

```
In [15]: plot_histogram(counts)
```

```
Out[15]:
```



```
In [ ]:
```

P5

ChatGPT

Quantum Buddy

Quantum Explorer

Quantum Buddy 2.0

Explore GPTs

Today

P5 BUDDY

P5 NORMALE

Yesterday

GPT Validity & Limitations

Previous 7 Days

P3 BUDDY

Enhancing Draft Text

Quantum RSA Visual Output

Add Team workspace



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.

ChatGPT

Great! Let's get started with your user test. 🎯

Aim of the Test:

The goal here is to harness your knowledge in quantum mechanics and coding to create something innovative. You'll conceptualize, develop, and refine an algorithm that uses the unique powers of quantum computing. This isn't just about coding; it's about thinking creatively and applying quantum concepts in new ways. 🧠💡

Are you ready to dive into this exciting challenge? Let me know when you're set to go! 😊👍

You

Yes

ChatGPT

Awesome! Let's begin the journey. 🌟

Stage 1: Conceptualization

- Choose a Field:** First, pick an area you're passionate about or curious to explore. Here are some options:

⬇️
 - Cryptography:** Developing new quantum-safe encryption methods.
 - Machine Learning:** Enhancing algorithms for quantum data analysis.
 - Chemical Simulations:** Simulating molecular interactions for drug discovery.
 - Optimization Problems:** Solving complex logistics or scheduling issues.
- Identify a Problem:** Within your chosen field, think about a specific problem or question you want to tackle. This should be something that quantum computing could uniquely address or improve upon.
- Research and Brainstorm:** Do a bit of digging into current approaches to your problem and how quantum computing might offer new solutions. Jot down your ideas and theories.

What field excites you the most, and what problem in that field do you want to solve? Share your thoughts, and we'll move to the next step together! 🌈💡

You

Chemical simulations

ChatGPT

Fantastic choice! Chemical simulations are a ⬇️ thrilling area for quantum computing. 💡🧪

Stage 2: Development

Now that you've chosen chemical simulations, let's start developing your algorithm:

- Define the Simulation Goal:** What specific aspect of chemical simulations do you want to focus on? It could be anything from molecular structure prediction to reaction dynamics.
- Understand Quantum Approaches:** Research how quantum algorithms can be applied to your chosen simulation goal. Look into methods like the Variational Quantum Eigensolver (VQE) or Quantum Phase Estimation (QPE) for insights.

3. **Start Coding:** Begin translating your ideas into a quantum algorithm. Don't worry if you're not a coding pro; I'm here to guide you! You can use quantum programming languages like Qiskit (for IBM Quantum Computers) or Cirq (for Google's Quantum Processors).
4. **Iterate:** Your first draft won't be perfect, and that's okay! We'll refine it as we go.

Now, let's get to the exciting part: coding! Do you have a specific simulation goal in mind, and how familiar are you with quantum programming languages? Let me know, and we'll start shaping your algorithm! 🧠💡📄



You

I want to explore spin polarization through helical molecules. I am vaguely familiar with Qiskit



ChatGPT

Exploring spin polarization in helical molecules sounds fascinating and quite innovative! 🧠📄

Given your familiarity with Qiskit, we'll use it for your project. Qiskit is perfect for this task as it's designed for IBM's quantum computers.

Stage 2: Development (Continued)

1. **Set Up Your Environment:** If you haven't already, install Qiskit in your Python environment. You can do this using pip:

bash

Copy code

```
pip install qiskit
```



2. **Initial Coding Steps:**

- **Import Necessary Libraries:** Start by importing Qiskit and other necessary Python libraries.
- **Initial Quantum Circuit:** Create a basic quantum circuit. Think about how many qubits you might need for simulating spin polarization in helical molecules.

3. **Simulate Spin Polarization:**

- **Understand the Quantum Model:** You'll need to model how helical molecules affect spin polarization. This might involve representing the molecular structure and its interaction with quantum states.
- **Translate to Quantum Operations:** Convert this understanding into quantum gates and operations in your circuit.

4. **Run Simulations and Visualize:** Use Qiskit's simulators to run your circuit. You can also visualize the quantum states to understand how the system evolves.

5. **Get Feedback and Iterate:** Test out your circuit, see how it performs, and make adjustments as needed. This is a crucial part of the development process.

It's okay if you're not clear on all these steps right away. We'll tackle them one by one.

For now, let's start with setting up Qiskit and creating a basic quantum circuit. How

does that sound? Ready to code? 🚀💡📄



You

Im ready to code



ChatGPT

Great! Let's get coding with Qiskit. 🌟

Step 1: Set Up Your Environment

First, ensure Qiskit is installed in your Python environment. You can do this using the command:

bash

Copy code

```
pip install qiskit
```

Step 2: Import Necessary Libraries

Open your Python IDE or a Jupyter notebook, and start by importing Qiskit libraries:

python

Copy code

```
from qiskit import QuantumCircuit, Aer, execute
from qiskit.visualization import plot_histogram, plot_bloch_multivector
```

Step 3: Create a Basic Quantum Circuit

For simulating spin polarization in helical molecules, let's begin with a simple circuit. We'll start small and expand as needed:

```
python Copy code

# Initialize a Quantum Circuit with a specific number of qubits
num_qubits = 2 # You can change this based on your model
circuit = QuantumCircuit(num_qubits)

# Add quantum gates here (as an example, let's apply a Hadamard gate to
circuit.h(0)

# Visualize the circuit
print(circuit)
```

This is a very basic start, just to get you going. The Hadamard gate creates a superposition, which is a fundamental concept in quantum computing.

Step 4: Run the Circuit on a Simulator

Now, let's simulate this circuit:

```
python Copy code

# Use Aer's qasm_simulator
simulator = Aer.get_backend('qasm_simulator')

# Execute the circuit on the simulator
job = execute(circuit, simulator, shots=1000)

# Grab results from the job
result = job.result()

# Get the histogram of results
counts = result.get_counts(circuit)
print(counts)

# Visualize the results
plot_histogram(counts)
```

This code runs your circuit 1000 times (shots) and collects the results in a histogram.

Next Steps

1. **Modeling Spin Polarization:** The real challenge is in modeling spin polarization in helical molecules using quantum gates. This will require some thought and research on how to represent these phenomena in a quantum circuit.
2. **Iterate and Refine:** As you develop your model, you'll likely need to refine and adjust your circuit.

Remember, this is an iterative and exploratory process. Feel free to experiment with different gates and circuit configurations. And if you have any questions or need guidance, I'm here to help!

Are you ready to try running this initial setup? Let's see what you can create! 🚀👤🔗



You

That code did not work, here is the error: Traceback (most recent call last):
Cell In[3], line 11
counts = result.get_counts(circuit)
File /opt/conda/lib/python3.10/site-packages/qiskit/result/result.py:289 in get_counts
raise QiskitError(f"No counts for experiment '{repr(key)}'")
QiskitError: "No counts for experiment "<qiskit.circuit.quantumcircuit.QuantumCircuit object at 0x7f25f6047e80>"

Use %tb to get the full traceback.



ChatGPT

It looks like the error occurred because we're using a QASM simulator, which expects measurements in the circuit, but we haven't included any measurement gates. Let's fix that.

Here's how you can modify the code:

1. **Add Measurement Gates:** At the end of your quantum circuit, you need to add measurement gates to measure the state of each qubit.
2. **Create a Classical Register:** Along with the quantum register (the qubits), you need a

classical register to store the results of the measurements.

Here's the revised code:

```
python Copy code

from qiskit import QuantumCircuit, Aer, execute, ClassicalRegister

# Initialize a Quantum Circuit with a specific number of qubits
num_qubits = 2
circuit = QuantumCircuit(num_qubits)

# Add quantum gates
circuit.h(0) # Example: Hadamard gate on the first qubit

# Add a classical register to store the measurement results
classical_bits = ClassicalRegister(num_qubits)
circuit.add_register(classical_bits)

# Add measurement gates
circuit.measure(range(num_qubits), range(num_qubits))

# Visualize the circuit
print(circuit)

# Use Aer's qasm_simulator
simulator = Aer.get_backend('qasm_simulator')

# Execute the circuit on the simulator
job = execute(circuit, simulator, shots=1000)

# Grab results from the job
result = job.result()

# Get the histogram of results
counts = result.get_counts(circuit)
print(counts)

# Visualize the results
plot_histogram(counts)
```

This code should work better. It includes the necessary measurement steps and a classical register to store the measurement outcomes. Give it a try and let me know how it goes! 🌟 👤 📊



You

That works, lets move on to modeling spin polarization



ChatGPT

Great to hear the code is working! Now, let's move on to the more challenging and exciting part: modeling spin polarization in helical molecules using your quantum circuit. 🌐 📊

Modeling Spin Polarization in Helical Molecules:

- Understand the Physical Model:** Spin polarization in helical molecules involves understanding how electron spins interact with the molecular structure. In a quantum circuit, this can be modeled using qubits to represent electron spins and quantum gates to simulate interactions.
- Choose a Representation:**
 - Qubit Representation:** Decide how you'll represent the spins and helical structure in your circuit. Each qubit can represent an electron spin, but the challenge is in encoding the helical structure's influence.
 - Gate Selection:** Choose quantum gates that best simulate the interactions in your model. For example, you might use rotation gates ('RX', 'RY', 'RZ') to represent spin manipulations.
- Encoding the Helical Structure:**
 - Complex Encoding:** This part is tricky. You'll need to find a way to encode the helical structure's impact on electron spins. This could involve using a series of gates to simulate the molecular environment's effect on spin states.
 - Parameterized Gates:** Consider using parameterized gates where the parameters represent aspects of the helical structure.
- Simulation and Observation:**
 - Running the Circuit:** Once you have your model, simulate it using Qiskit's simulator.
 - Analyzing Spin States:** Use statevector simulation to observe how the spin states evolve. This can give you insight into the spin polarization effects.
- Iterate and Refine:**

- **Test and Adjust:** This will likely require several iterations. Test your circuit, analyze the outcomes, and adjust your model accordingly.

This is a high-level overview, and the actual implementation will be quite complex. It involves both a deep understanding of quantum mechanics and creativity in translating that into a quantum circuit.

Feel free to start experimenting with different gate configurations and parameters. Remember, this is a learning process, and it's okay if it takes a few tries to get it right.

How does this plan sound to you? Are you ready to start experimenting with the model, or do you have any specific questions or areas you want to delve into further? 🚀💡



You
The plan sounds good! Can you provide a model of spin polarization with quantum gates?



ChatGPT
Certainly! Let's create a simplified model for spin polarization using quantum gates. This model won't capture all the complexities of helical molecules, but it will give you a foundational structure that you can build upon and refine.

Simplified Spin Polarization Model:

1. **Qubit Representation:** Let's use two qubits to represent two electron spins. In real scenarios, you might need more qubits, but this simplification will help us focus on the basics.
2. **Initial State Preparation:** We'll start with both spins in the $|0\rangle$ state (which can represent spin-up).
3. **Introducing Spin Interaction:** We'll use a combination of quantum gates to simulate the interaction between the spins and the helical structure. For simplicity, let's assume the interaction causes a rotation in the spin state.
4. **Rotation Gates:** Rotation gates ('RX', 'RY', 'RZ') can be used to simulate the spin state's rotation due to the helical structure.
5. **Measurement:** Finally, we'll measure the spins to observe the polarization.

Example Quantum Circuit:

```
python Copy code

from qiskit import QuantumCircuit, Aer, execute, ClassicalRegister

# Initialize a Quantum Circuit with 2 qubits
circuit = QuantumCircuit(2, 2)

# Prepare initial state (both qubits in state |0>)

# Simulate spin interaction
# For example, use RY gate to rotate the first qubit (spin)
theta = 1.57 # This is an example angle, you can adjust it
circuit.ry(theta, 0)

# Optional: Introduce entanglement between the qubits to simulate correlation
circuit.cx(0, 1) # CNOT gate entangling the two qubits

# Add measurement gates
circuit.measure([0, 1], [0, 1])

# Visualize the circuit
print(circuit)
```

Running and Observing the Circuit:

- Use the QASM simulator to run the circuit and observe the histogram of results. This will show you the probability distribution of the spin states.
- You might also want to use a statevector simulator to directly observe the quantum state of the system without measurement.

Next Steps:

- **Experiment with Different Angles and Gates:** Try different values for 'theta' and different gates to see how they affect the spin states.
- **Introduce More Complexity:** Once you're comfortable with this basic model, start adding more qubits and gates to better simulate the helical structure's effects.

This is a starting point, and as you delve deeper into the specifics of helical molecules and spin interactions, you'll be able to create a more accurate and complex model. Feel

free to experiment and iterate!

How does this look for a starting point? Ready to give it a try? 🚀👤🔗



You

That is a good first try, can you provide a more detailed model that more accurately captures the physics?



ChatGPT



Creating a more detailed and physically accurate model for spin polarization in helical molecules using quantum gates is certainly more complex and requires a deeper dive into the specifics of quantum mechanics and molecular physics. Here's an outline for a more sophisticated approach:

1. Advanced Qubit Representation:

- Use more qubits to represent the electrons' spins and possibly the molecular orbitals or other aspects of the helical structure. For instance, each electron spin might be represented by one qubit, while additional qubits could represent the molecular environment.

2. Quantum Gates for Molecular Interactions:

- **Rotation Gates** ('RX', 'RY', 'RZ'): To represent more complex spin manipulations that occur due to the molecular environment.
- **Controlled Gates** (like 'CNOT', 'CU'): To simulate interactions between spins and the helical structure. For example, the state of one qubit (representing an aspect of the molecule) could control the operation performed on a spin qubit.
- **Custom Gates**: You might need to define custom gates to more accurately represent the specific interactions in your system.



3. Entanglement and Correlations:

- **Entangling Gates**: Use gates like 'CNOT' or 'CZ' to create entanglements that reflect the quantum correlations between spins and the molecular structure.
- **Quantum Fourier Transform (QFT)**: For more complex simulations, the QFT can be used to simulate wave-like properties of the system.

4. Encoding the Helical Structure:

- Develop a way to encode information about the helical structure into the quantum circuit. This might involve initializing qubits in states that represent different aspects of the molecule, such as its chirality or the orientation of its helix.

5. Simulation and Measurement:

- **Statevector Simulation**: To observe the system's evolution without collapsing its state.
- **Histogram of Results**: Run the circuit multiple times with a QASM simulator to get a probability distribution of outcomes.

6. Iterative Refinement:



- Based on your simulation results, iteratively refine your model. This may include adjusting the types of gates used, their parameters, and the overall structure of the quantum circuit.

Example (Simplified):

python

Copy code

```
from qiskit import QuantumCircuit, Aer, execute, ClassicalRegister, QuantumRegister

# Initialize a Quantum Circuit with more qubits
qubits = QuantumRegister(5)
classical_bits = ClassicalRegister(5)
circuit = QuantumCircuit(qubits, classical_bits)

# Detailed gate operations (simplified example)
# Rotate first two qubits to represent electron spins
circuit.ry(1.57, qubits[0])
circuit.ry(1.57, qubits[1])

# Custom operations to simulate interactions with the helical structure
# ... (custom operations go here)

# Entangling operations to simulate correlations
circuit.cnot(qubits[0], qubits[2])
circuit.cnot(qubits[1], qubits[3])
```

```
# Measurements
circuit.measure(qubits, classical_bits)

# Visualize the circuit
print(circuit)
```

This is still a simplification. The actual implementation would require a deep understanding of the specific molecular system you're studying, including its quantum mechanical properties and how these can be mapped onto a quantum circuit.

Would you like to explore this in further detail, or have any specific aspects in mind for your simulation? 📄 🧪 🧑



Message ChatGPT...



ChatGPT can make mistakes. Consider checking important information.



Collaborate on a Team plan




co doc


```
In [1]: from qiskit import QuantumCircuit, Aer, execute
        from qiskit.visualization import plot_histogram, plot_bloch_multivector
```

```
In [2]: # Initialize a Quantum Circuit with a specific number of qubits
        num_qubits = 2 # You can change this based on your model
        circuit = QuantumCircuit(num_qubits)

        # Add quantum gates here (as an example, let's apply a Hadamard gate to the first
        circuit.h(0)

        # Visualize the circuit
        print(circuit)
```

q_0: 

q_1: 

```
In [7]: from qiskit import QuantumCircuit, Aer, execute, ClassicalRegister

        # Initialize a Quantum Circuit with 2 qubits
        circuit = QuantumCircuit(2, 2)

        # Prepare initial state (both qubits in state |0>)

        # Simulate spin interaction
        # For example, use RY gate to rotate the first qubit (spin)
        theta = 1.57 # This is an example angle, you can adjust it
        circuit.ry(theta, 0)

        # Optional: Introduce entanglement between the qubits to simulate correlation
        circuit.cx(0, 1) # CNOT gate entangling the two qubits

        # Add measurement gates
        circuit.measure([0, 1], [0, 1])

        # Visualize the circuit
        print(circuit)

        # Visualize the circuit
        print(circuit)

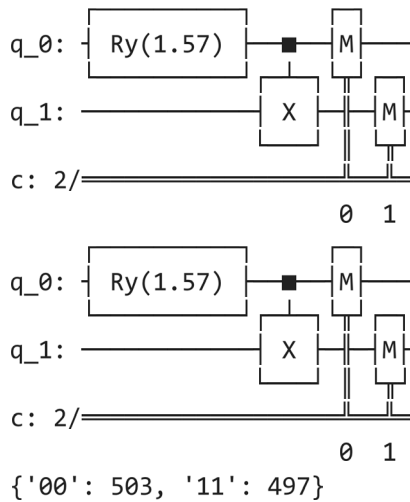
        # Use Aer's qasm_simulator
        simulator = Aer.get_backend('qasm_simulator')

        # Execute the circuit on the simulator
        job = execute(circuit, simulator, shots=1000)

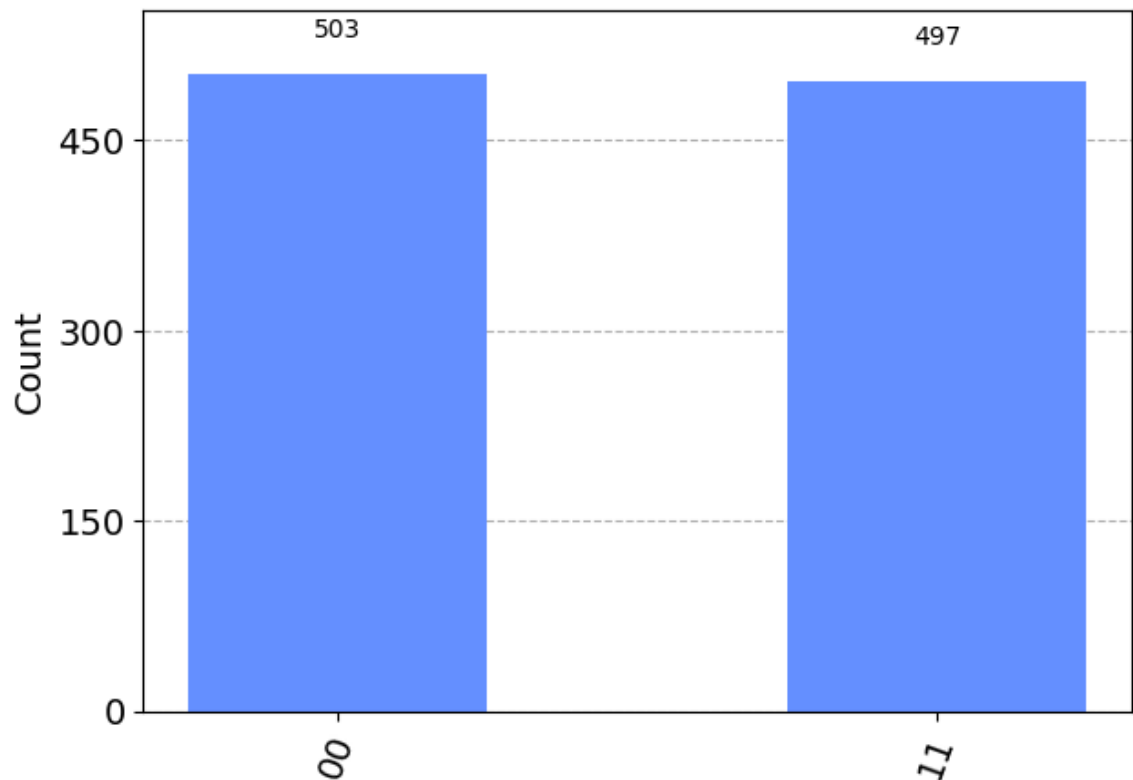
        # Grab results from the job
        result = job.result()

        # Get the histogram of results
        counts = result.get_counts(circuit)
        print(counts)
```

```
# Visualize the results
plot_histogram(counts)
```



Out[7]:



In [5]: `result.get_counts()`

Traceback (most recent call last):

Cell In[5], line 1

`result.get_counts()`

File /opt/conda/lib/python3.10/site-packages/qiskit/result/result.py:289 in `get_counts`

`raise QiskitError(f'No counts for experiment "{repr(key)}"')`

QiskitError: 'No counts for experiment "0"'

Use %tb to get the full traceback.

[Search for solution online](#)

In []:

ChatGPT

Quantum Buddy

Quantum Explorer

Quantum Buddy 2.0

Explore GPTs

Today

P5 BUDDY

...

P5 NORMALE

Yesterday

GPT Validity & Limitations

Previous 7 Days

P3 BUDDY

Enhancing Draft Text

Quantum RSA Visual Output

Add Team workspace

Quantum Buddy



You

I want to have a structured user test for a young enthusiast with a background in quantum mechanics and coding. first, explain the aim of the test. The objective of the test is to conceptualize, develop, and refine an algorithm that leverages the unique capabilities of quantum computing. ask me if I'm ready and wait for my reply. Then, in a new prompt, explain to me step by step what I'm going to do. Start with conceptualization: Begin by choosing a field and understanding the unique problem you wish to solve. Give some options such as as exploring fields like Cryptography, Machine Learning, Chemical Simulations, and Optimization Problems in the conceptualization stage. wait for my response. then, in a new prompt, go to the stage of Development: Start coding your algorithm with guidance from ChatGPT. Don't worry if you're not an expert coder; ChatGPT will help you translate your quantum understanding into a working code. Once you have the code go to the following stages in a new prompt: Refinement: Test and refine your algorithm. Finalization: Finalize your code and prepare it for submission to run on an IBM quantum computer. All along keep a cheerful and simple language with emojis.



Quantum Buddy

Aim of the Test



Hello there, young quantum enthusiast! 🌟 The aim of this test is pretty exciting. You're going to dive into the world of quantum computing and flex your coding muscles! The objective? To conceptualize, develop, and refine a unique algorithm that harnesses the power of quantum computing. Imagine creating something that uses quantum bits instead of classical bits - how cool is that? 🤖 🧠

This isn't just about writing code; it's about thinking in a whole new way. You'll choose a problem, build a solution with a quantum twist, and then tweak it to perfection. And the best part? You'll be preparing it to run on an actual IBM quantum computer! Imagine your code, out there, solving real-world problems using quantum mechanics. Mind-blowing, right? 🧐

So, are you ready to embark on this quantum journey? Let me know when you're all set, and we'll dive into the first stage! 😊 🚀



You

Im ready



Quantum Buddy

Stage 1: Conceptualization 🌐 🤖



Awesome, let's get started! In this first stage, you're going to choose a field and identify a unique problem you want to solve using quantum computing. Think of it like picking your adventure in the quantum realm. Here are some fascinating fields you can explore:

1. **Cryptography:** Delve into the world of secure communication. Quantum computing can revolutionize the way we encrypt and decrypt information. Imagine creating an unbreakable code! 🛡️
2. **Machine Learning:** Quantum algorithms can potentially process complex data faster than classical computers. How about teaching a quantum computer to learn and make decisions? 🧠
3. **Chemical Simulations:** Quantum mechanics is at the heart of chemistry. You could simulate molecular structures and reactions that are too complex for classical computers. 🧪
4. **Optimization Problems:** From logistics to finance, optimization is key. Quantum computing can find the best solutions among a vast number of possibilities, way faster than traditional methods. 📊

Think about which of these areas excites you the most. What problem within that field do you think quantum computing could help solve? It's all about finding that perfect challenge for your quantum quest.

Once you've got an idea, let me know which field you choose and what problem you're aiming to solve. We'll then move on to the exciting part - developing your algorithm! 🤖 🌟




You

I would like to work on chemical simulations




Quantum Buddy


Stage 2: Development - Chemical Simulations


Fantastic choice! Working on chemical simulations is a brilliant way to leverage the power of quantum computing. This field has enormous potential, from discovering new materials to understanding complex biochemical processes. 

Here's What You'll Do:



1. **Define the Chemical Problem:** What specific  aspect of chemistry are you interested in simulating? It could be anything from the behavior of a particular molecule to exploring reaction mechanisms. Think about a problem that excites you!
2. **Understand Quantum Representation:** Learn how to represent your chemical problem in quantum terms. This involves understanding quantum states, qubits, and how they can model atomic and molecular structures.
3. **Algorithm Design:** Start designing your quantum algorithm. You'll need to translate your chemical problem into a series of quantum operations. This could involve using quantum gates to simulate the behavior of electrons in a molecule, for example.
4. **Coding with Qiskit:** Now, you'll start coding your algorithm using Qiskit, IBM's quantum computing framework. Don't worry if you're not a coding expert; I'm here to help guide you through this process. We'll translate your quantum understanding into a working Qiskit code.

Getting Started with Qiskit

- **Install Qiskit:** If you haven't already, you'll need to install Qiskit in your Python environment. Use the command `pip install qiskit`.
- **Familiarize Yourself with Qiskit Basics:** Understand the basic elements of Qiskit, like creating quantum circuits, adding gates, and running simulations.
- **Implement Your Algorithm:** Begin translating your algorithm into Qiskit code. Start by defining a quantum circuit that represents  your chemical problem.

Tips and Tricks 

- **Start Small:** Begin with a simple model and gradually increase complexity.
- **Test Frequently:** Regularly test your code to ensure it behaves as expected.
- **Document Your Code:** Keep track of what each part of your code is supposed to do.

Ready to start coding? Let me know if you need any help with Qiskit, or if you have any questions about translating your chemical simulation into a quantum algorithm. Let's code some quantum magic!  




You

Im somewhat familiar with Qiskit. I would like to simulate spin polarization through helical molecules




Quantum Buddy

Developing a Quantum Algorithm for Simulating Spin Polarization in Helical Molecules

That's a fascinating and advanced topic! So  simulating spin polarization through helical molecules using quantum computing can provide deep insights into molecular properties and behaviors. Since you're somewhat familiar with Qiskit, we'll build on that knowledge to develop your quantum algorithm.

Steps to Follow:

1. **Understand the Physics:** Ensure you have a solid understanding of spin polarization and the structure of helical molecules. This knowledge will guide your quantum representation.
2. **Quantum Representation of the Problem:**
 - **Molecular Structure Representation:** Represent the helical molecule in quantum terms. This involves considering how to model the electrons and their spins in the molecule.
 - **Spin Polarization:** Determine how spin polarization can be represented as a quantum state or series of states.
3. **Designing the Quantum Circuit:**
 - **Initial State Preparation:** Develop a Qiskit routine to initialize the state of your quantum system, representing the electron spins in the helical molecule.
 - **Simulation of Spin Interactions:** Use quantum gates to simulate the interactions that lead to spin polarization. This might involve complex manipulations of qubits to mimic the molecular behavior. 
4. **Coding with Qiskit:**
 - **Basic Circuit:** Start by creating a basic circuit in Qiskit that can model a simplified version of your helical molecule.
 - **Iterate and Expand:** Gradually add complexity, simulating more realistic interactions and behaviors.
5. **Simulation and Analysis:**

5. Simulation and Analysis:

- **Run Your Circuit:** Use Qiskit's simulator to run your circuit. You can start with a statevector simulator for an idealized scenario.
- **Analyze the Results:** Observe the output quantum states to analyze spin polarization.

6. Refinement and Debugging:

- Test various scenarios and parameters.
- Refine your quantum circuit for accuracy and efficiency.

Getting More Specific:

- **Qiskit Elements:** Utilize Qiskit's libraries for creating quantum circuits, adding custom gates, and running simulations.
- **Visualization Tools:** Use Qiskit's visualization tools to observe the state of your quantum system at different stages of the simulation.

Tips 🌟:

- **Modular Approach:** Build your code in small, testable modules.
- **Documentation:** Keep detailed notes on your design choices and the reasoning behind them.

Now, it's time to start implementing your idea! If you need guidance on specific parts of Qiskit, like creating custom gates or interpreting simulation results, feel free to ask. Let's bring this quantum simulation to life! 🚀💡📊



You

Can you give me a qiskit program that simulates spin polarization through helical molecules?



Quantum Buddy

Creating a Qiskit program to simulate spin polarization in helical molecules is a complex task, as it involves detailed knowledge of the molecular structure and how to effectively model it in quantum terms. While I can provide a basic framework to get you started, please note that this will be a highly simplified version. A full, accurate simulation would require extensive quantum chemistry and Qiskit expertise.

Here's a basic outline of what such a program might look like:

1. Import Necessary Libraries:

```
python Copy code  
  
from qiskit import QuantumCircuit, Aer, execute  
from qiskit.visualization import plot_histogram
```

2. Initialize the Quantum Circuit:

- Assume a simplified model with a few qubits representing the spin states in the molecule.

```
python Copy code  
  
# Create a quantum circuit with a specific number of qubits  
num_qubits = 3 # This is a simplified example; adjust as needed  
qc = QuantumCircuit(num_qubits)
```

3. State Preparation:

- Initialize the qubits to represent the initial state of the electron spins in the molecule.

```
python Copy code  
  
# Example: Initialize all qubits to |0> (spin down)  
# In a real scenario, this would be more complex  
for qubit in range(num_qubits):  
    qc.initialize([1, 0], qubit) # Initialize to |0> state
```

4. Simulate Spin Interactions:

- Apply quantum gates to simulate interactions. This is a highly abstracted and simplified representation.

```
python Copy code  
  
# Apply some gates to simulate interactions  
# This is just an example; the actual gates depend on the specific simulation  
qc.h(0) # Apply a Hadamard gate to the first qubit  
qc.cx(0, 1) # Controlled-NOT gate as an example of interaction
```

5. Measurement:

- Measure the qubits to observe the final state.

```
python Copy code
```

```
qc.measure_all()
```

6. Execute the Simulation:

- Use Qiskit's simulator to run the circuit.

```
python Copy code  
  
simulator = Aer.get_backend('qasm_simulator')  
result = execute(qc, simulator, shots=1000).result()  
counts = result.get_counts(qc)
```

7. Visualize the Results:

- Plot the results to analyze the spin states.

```
python Copy code  
  
plot_histogram(counts)
```

Running the Program:

- This code sets up a basic quantum circuit, simulates some interactions, and measures the output.
- It's a very abstract representation of spin polarization in a molecule. A realistic simulation would require more complex initialization and a detailed understanding of the molecular interactions.

Next Steps:

- Adapt the initialization and gate sequences to more accurately represent the spin states and interactions in your specific helical molecule.
- Use Qiskit's advanced features to create a more realistic model.

Remember, this is just a starting point. Real-world quantum simulations of molecular behavior are highly complex and require extensive domain knowledge. Feel free to modify and expand upon this basic framework to suit your specific simulation needs!



You

Can you model spin polarization through helical molecules using a quantum circuit?



Quantum Buddy

Modeling spin polarization in helical molecules using a quantum circuit is a highly specialized and advanced task, primarily due to the complexity of molecular structures and the intricacies of quantum mechanics involved. In quantum computing, especially with tools like Qiskit, we often work with highly abstracted models that simplify these complex phenomena into manageable quantum circuits.

To model spin polarization in helical molecules, here are the key considerations and a theoretical approach:

1. Abstraction of Molecular Structure:

- Each qubit in your quantum circuit could represent an electron's spin state in the molecule. In reality, though, the interactions and correlations between these spins in a helical structure are complex.

2. Quantum State Preparation:

- Initialize your qubits to represent the initial spin states of the electrons in the molecule. This could be a simple initialization (like all spins aligned), but real molecules might require more complex states.

3. Simulating Spin-Spin Interactions:

- Apply quantum gates to simulate interactions between electron spins. This could involve entangling gates like CNOT to represent spin correlations.
- Quantum gates that change the phase of the spins might be used to simulate the effects of molecular fields on spin states.

4. Measurement and Analysis:

- Measuring the qubits at the end of the circuit gives you the final spin states.
- Repeated simulations can provide statistical insight into the spin dynamics.

5. Limitations and Complexity:

- A real helical molecule may have a large number of interacting spins, making it challenging to simulate accurately with a limited number of qubits.
- The actual quantum mechanics of spin interactions in a molecule are more complex than what can be straightforwardly represented with basic quantum gates.

6. Advanced Techniques:

- Techniques like Quantum Phase Estimation (QPE) or Variational Quantum Eigensolver (VQE) could be used for more sophisticated simulations, especially in the context of quantum chemistry.

Given these considerations, it's important to remember that any quantum circuit you create with Qiskit to simulate this process will be a highly simplified model. The real behavior of electron spins in helical molecules is governed by the laws of quantum mechanics, which can be incredibly complex and might not be fully captured in a basic quantum circuit.

In summary, while you can create a quantum circuit to model this phenomenon, the circuit will be a simplification and an abstraction of the actual complex quantum behaviors in the molecule. For detailed and accurate simulations, significant expertise in quantum chemistry and advanced quantum computing techniques would be required.



Message Quantum Buddy...



ChatGPT can make mistakes. Consider checking important information.



Collaborate on a Team plan



co doc

```
In [1]: from qiskit import QuantumCircuit, Aer, execute
        from qiskit.visualization import plot_histogram
```

```
In [2]: # Create a quantum circuit with a specific number of qubits
        num_qubits = 3 # This is a simplified example; adjust as needed
        qc = QuantumCircuit(num_qubits)
```

```
In [3]: # Example: Initialize all qubits to |0> (spin down)
        # In a real scenario, this would be more complex
        for qubit in range(num_qubits):
            qc.initialize([1, 0], qubit) # Initialize to |0> state
```

```
In [4]: # Apply some gates to simulate interactions
        # This is just an example; the actual gates depend on the specific simulation
        qc.h(0) # Apply a Hadamard gate to the first qubit
        qc.cx(0, 1) # Controlled-NOT gate as an example of interaction
```

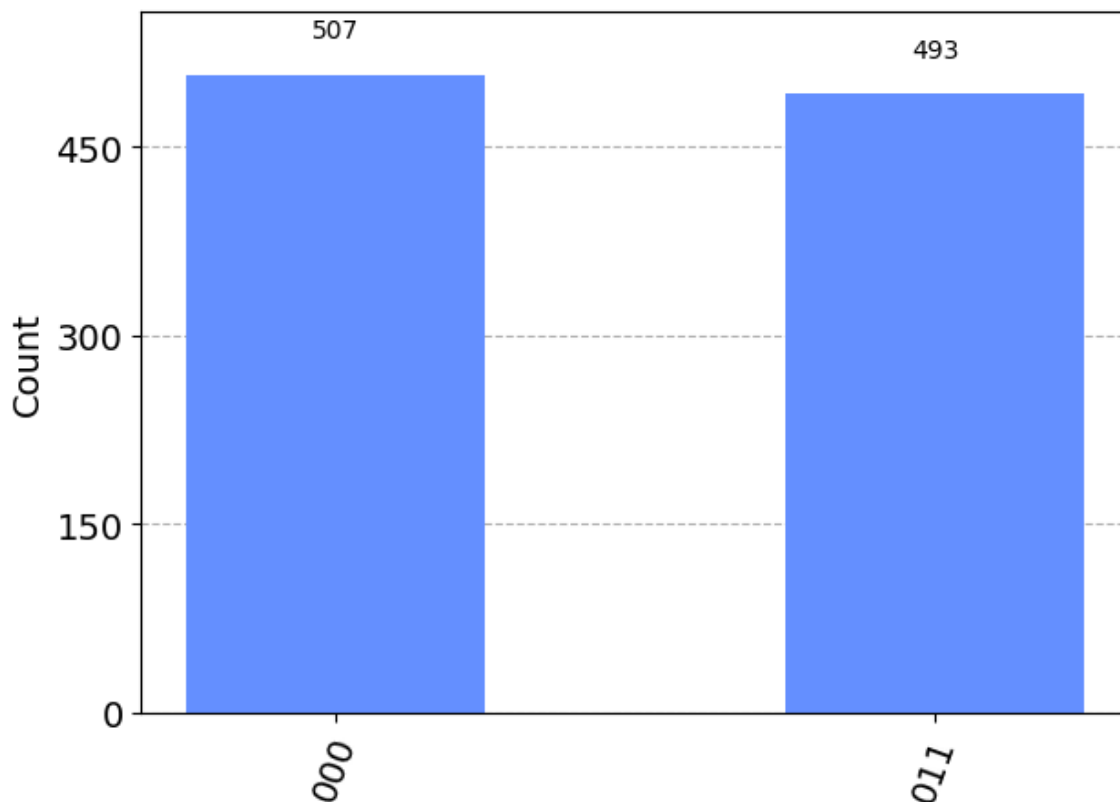
```
Out[4]: <qiskit.circuit.instructionset.InstructionSet at 0x7fb0afa72e00>
```

```
In [5]: qc.measure_all()
```

```
In [6]: simulator = Aer.get_backend('qasm_simulator')
        result = execute(qc, simulator, shots=1000).result()
        counts = result.get_counts(qc)
```

```
In [7]: plot_histogram(counts)
```

```
Out[7]:
```



```
In [ ]:
```

[VIII] Evaluation User Test 2

a. Miroboard Test (template)

<p>P1</p> <p>Theme:</p> <p>Task:</p> <p>Input:</p>	<p>Overall comment:</p> <p>.....</p>
<p>GPT Code</p>	<p>Quantum Buddy GPT</p>
<p>Accuracy of the Code</p> <p>● ● ● ● ● ●</p> <p>Explanatory Capacity of ChatGPT</p> <p>● ● ● ● ● ●</p>	<p>Accuracy of the Code</p> <p>● ● ● ● ● ●</p> <p>Explanatory Capacity of ChatGPT</p> <p>● ● ● ● ● ●</p>

miro

b. Miroboard Results

https://miro.com/welcomeonboard/VkNjU3B0bDhYdmpzQzU5NFlUODR6eXlUdEN5N-k14WFh3emxTSDNzOV05TFBtSGhJUGZUQzBHYNREalpCdGRPaHwzNDU4NzY0NTE4M-zE0MjA1NTE5fDI=?share_link_id=641786781195

[IX] Quantum buddy 2.0

a. Link Quantum Buddy 2.0

<https://chat.openai.com/g/g-HuukcppqT-quantum-buddy-2-0>

b. Database Quantum Buddy 2.0 (Next page)

IBM quantum lab infos:

Libro – selezionato solo alcune parti che mi interessavano

The IBM Quantum computer: The reason why it could be important to know the physical gate set is that some user-programmed gates may need to be decomposed into multiple physical gates, and hence could lead to a longer physical algorithm.

when implementing a quantum algorithm it is important to consider the sources of noise in the computer. The two main sources of noise are typically gate infidelity and decoherence. Gate infidelity refers to the fact that the user-specified gates do not precisely correspond to the physically implemented gates. Decoherence refers to the fact that gradually over time the quantum computer loses its “quantumness” and behaves more like a classical object.

Programming the IBM quantum computer: Qiskit library. Qiskit [4] is an open-source quantum computing library developed under the aegis of IBM. Qiskit allows users to write and run programs on either IBM’s quantum processors or on a local simulator, without the use of the graphical interface. → Python library.

Quantum mechanics for non-physics

Intro: Classically, the time it takes to do certain computations can be decreased by using parallel processors. To achieve an

exponential decrease in time requires an exponential increase in the number of processors, and hence an exponential increase in the amount of physical space needed → Thus, an exponential increase in parallelism requires only a linear increase in the amount of physical space needed. This effect is called quantum parallelism. Two techniques to manipulate this is: 1) common property of all of the output values such as the symmetry or period of a function can be read off. 2) transforms the quantum state to increase the likelihood that output of interest will be read.

Quantum mechanics: Quantum mechanical phenomena are difficult to understand, since most of our everyday experiences are not applicable. Quantum mechanics is a theory in the mathematical sense: it is governed by a set of axioms.

photon’s polarization state can be modelled by a unit vector pointing in the appropriate direction. Any arbitrary polarization can be expressed as a linear combination of the two basis vectors. // any device measuring a two-dimensional system has an associated orthonormal basis with respect to which the quantum measurement takes place.

Measurement of the quantum state will change the state to the result of the measurement furthermore will change that state, and it is not possible to determine what the original state was.

State spaces and bra/ket notation: The state space of a quantum system, consisting of the positions, momentums, polarizations, spins, and so on of the various particles, is modelled by a Hilbert space of wave function.

Quantum bits: A quantum bit, or qubit, is a unit vector in a two-dimensional complex vector space for which a particular basis, denoted by 0 , 1 , has been fixed. Even though a quantum bit can be put in infinitely many superposition states, it is only possible to extract a single classical bit’s worth of information from a single quantum bit. As every measurement can result in only one of two states, one of the basis vectors associated to the given measuring device, so, just as in the classical case, there are only two possible results.

Quantum key distribution: Sequences of single qubits can be used to transmit private keys on insecure channels. Thus any eavesdropper on the quantum channel is bound to introduce a high error rate that Alice and Bob can detect by communicating a sufficient number of parity bits of their keys over the open channel.

Multiple qubits: A surprising and unintuitive aspect of the state space of an n -particle quantum system is that the state of the system cannot always be described in terms of the state of its component pieces. However, in a quantum system the resulting state space is much larger; a system of n qubits has a state space of 2^n dimensions. Quantum states, combine through the tensor product. These states represent situations that have no classical counterpart and for which we have no intuition. These are also the states that provide the exponential growth of quantum state spaces with the number of particles.

Measurements: multibit measurement can be treated as a series of single-bit measurements in the standard basis. Any quantum state can be written as the sum of two vectors, one in each of the subspaces.

Quantum gates: The dynamics of a quantum system, when not being measured, are governed by Schrodinger's equation; the dynamics must take states to states in a way that preserves orthogonality. Any linear transformation on a complex vector space can be described by a matrix. Any unitary transformation of a quantum state space is a legitimate quantum transformation, and vice versa. → quantum transformations are unitary is that they are reversible. Thus quantum gates must be reversible.

Simple quantum gates: transformations are fully specified by their effect on the basis vectors. The names of these transformations are conventional. I is the identity transformation, X is negation, Z is a phase shift operation, and $Y = iXZ$ is a combination of both. Then the paper talks about the C-not open gate.



Quantum computers: While the classical NOT gate is reversible, AND, OR, and NAND gates are not. Thus it is not obvious that quantum transformations can carry out all classical computations.

Quantum gates array

Quantum parallelism

Shor's algorithm

The Quantum Fourier Transform

Search problem

Quantum error connection

Conclusion: Quantum computing is a new, emerging field that has the potential to dramatically change the way we think about computation, programming, and complexity. The challenge for computer scientists and others is to develop new programming techniques appropriate for quantum computers. new dimension to computation. Programming no longer consists

of merely formulating step-by-step algorithms but requires new techniques of adjusting phases and mixing and diffusing amplitudes to extract useful output. But the real power of these new machines, the exponential parallelism, can only be exploited using new, innovative programming techniques. People have only recently begun to research such techniques. Beyond these algorithms not much more is known about what could be done with a practical quantum computer

Various Algorithms

- Quantum Approximate Optimisation Algorithm (QAOA)
- Quantum Amplitude Estimation (QAE)
- Quantum (enhanced) Support Vector Machine (QSVM)
- Variational quantum eigensolver (VQE)
- Quantum Neural Networks (QNN)
- Quantum phase estimation (QPE) / quantum eigenvalue estimation
- Quantum Annealing (QA)
- Quantum (enhanced) Machine Learning (QML)
- Variational Quantum Classifier (VQC)
- Q-Means
- Shor's algorithm
- Grover's algorithm
- HHL (Harrow-Hassidim-Lloyd)
- Quantum-assisted Reinforcement Learning (QaRL)
- Quantum Singular Value Decomposition (QSVD)
- Quantum Principal Component Analysis (QPCA)
- Quantum Finite Volume Method (QFVM)
- Variational Quantum Linear Solver (VQLS)
- Quantum Convolutional Neural Networks (QCNN)
- Quantum Risk Analysis (QRA)
- Quantum Monte Carlo (QMC)
- Quantum Decision Tree (QDT)
- Quantum Bayesian Networks (QBN)
- Quantum Extreme Value Searching (QES)
- Quantum-inspired Particle Swarm Optimisation (QPSO)
- Quantum K-Means Clustering
- Quantum Distance Classification (QDC)

Developing new drug – code

```
pip install qiskit
```

```
from qiskit import QuantumCircuit, Aer, execute
```

```
# Create a Quantum Circuit acting on a quantum register of three qubits
```

```
circuit = QuantumCircuit(3)
```

```
# Add a Hadamard gate on qubit 0, putting this qubit in superposition.
```

```
circuit.h(0)
```

```
# Add a CX (CNOT) gate on control qubit 0 and target qubit 1, creating entanglement.
```

```
circuit.cx(0, 1)
```

```
# Add another CX gate on control qubit 0 and target qubit 2, creating more entanglement.
```

```
circuit.cx(0, 2)

# Visualize the circuit
print(circuit.draw(output='text'))

# You can also use 'mpl' for a matplotlib drawing, or 'latex' for a LaTeX-based drawing.
```

New Material – code

```
pip install qiskit

from qiskit import QuantumCircuit, Aer, execute

# Create a Quantum Circuit acting on a quantum register of one qubit
circuit = QuantumCircuit(1)

# Apply a Hadamard gate, which puts the qubit in a superposition state
circuit.h(0)

# Measure the qubit
circuit.measure_all()

# Use Aer's qasm_simulator
simulator = Aer.get_backend('qasm_simulator')

# Execute the circuit on the qasm simulator
job = execute(circuit, simulator, shots=1000)

# Grab the results from the job
result = job.result()

# Print the result
counts = result.get_counts(circuit)
print("Counts:", counts)
```

Lines – code (quantum creative code)

```
from qiskit import QuantumCircuit, execute, Aer

import numpy as np

import matplotlib.pyplot as plt

qc = QuantumCircuit(3)

qc.h(range(3))
```

```

backend = Aer.get_backend('statevector_simulator')
result = execute(qc, backend).result()
statevector = result.get_statevector()
real_parts = [np.real(amp) for amp in statevector]
plt.figure(figsize=(10, 6))
plt.bar(range(len(real_parts)), real_parts)
plt.xlabel('Quantum State')
plt.ylabel('Real Part of Amplitude')
plt.title('Quantum Randomness Visualization')
plt.show()

# ... [previous code for setting up the quantum circuit and getting the statevector]
# Select a colormap with warm colors
colormap = plt.cm.autumn
# Plot the bar graph with the chosen colormap
plt.figure(figsize=(10, 6))
bars = plt.bar(range(len(real_parts)), real_parts, color=colormap(np.linspace(0, 1, len(real_parts))))
# Add random lines for an abstract painting effect
num_lines = 20 # Number of lines to add
for _ in range(num_lines):
    start_point = np.random.rand(2) * len(real_parts) # Random start point
    end_point = np.random.rand(2) * len(real_parts) # Random end point
    line_color = np.random.choice(['darkred', 'brown']) # Random line color
    plt.plot([start_point[0], end_point[0]], [start_point[1], end_point[1]], color=line_color,
linewidth=np.random.rand() * 2)
plt.xlabel('Quantum State')
plt.ylabel('Real Part of Amplitude')
plt.title('Quantum Randomness as Abstract Art')
plt.show()

```

P5 test – code (Grover algorithm)

```
from qiskit import QuantumCircuit, Aer, transpile
from qiskit.visualization import plot_histogram
from qiskit.providers.aer import QasmSimulator

nqubits=5

oracle = QuantumCircuit(5)
oracle.x(2)
oracle.mct([0,1,2,3], 4)
oracle.x(2)

def diffuser(nqubits):
    qc = QuantumCircuit(nqubits)
    # Apply H and X gates to all qubits
    for qubit in range(nqubits):
        qc.h(qubit)
        qc.x(qubit)
    # Multi-controlled Z gate
    qc.h(nqubits-1)
    qc.mct(list(range(nqubits-1)), nqubits-1) # multi-controlled-toffoli
    qc.h(nqubits-1)
    # Apply X and H gates to all qubits again
    for qubit in range(nqubits):
        qc.x(qubit)
        qc.h(qubit)
    return qc

grover_circuit = QuantumCircuit(5, 5)
grover_circuit.h([0, 1, 2, 3])
grover_circuit.append(oracle, [0, 1, 2, 3, 4])
grover_circuit.append(diffuser(4), [0, 1, 2, 3])
grover_circuit.measure([0, 1, 2, 3], [0, 1, 2, 3])
```

IO PROV – code

```
!pip install qiskit
```

```
!pip install qiskit_algorithms
```

```
!pip install qiskit==0.24.0
```

```
import qiskit
```

```
print(qiskit.__version__)
```

Qiskit circuit library :

```
import numpy as np
from qiskit import QuantumCircuit

def build_qft(n):
    circuit = QuantumCircuit(n)
    for j in reversed(range(n)):
        circuit.h(j)
        for k in reversed(range(j)):
            circuit.cp(np.pi * 2. ** (k - j), j, k)

    for j in range(n // 2):
        circuit.swap(j, n - j - 1)

    return circuit

qft = build_qft(5)

from qiskit.quantum_info import SparsePauliOp

hamiltonian = SparsePauliOp.from_list(
    [ ("YZ", 0.3980), ("ZI", -0.3980), ("ZZ", -0.0113), ("XX", 0.1810) ]
)

from qiskit.circuit.library import EfficientSU2

ansatz = EfficientSU2(hamiltonian.num_qubits)
ansatz.draw("mpl")
```

A guide to the Qiskit circuit library

<https://medium.com/qiskit/a-guide-to-the-qiskit-circuit-library-36ee0f189956>

When it comes to quantum circuits, sometimes the “do it yourself” approach is best. Even for common subroutines like the quantum Fourier transform (QFT), a user may find that it’s better to build their circuit by hand, so they can implement it in a way that works for their specific use case. But in many cases, the circuits we need are well-studied, and we can avoid having to reinvent the wheel (potentially creating bugs in the process) by simply using circuits that already exist. That’s just one reason why we created the Qiskit circuit library.

[The Qiskit circuit library](#) is a collection of valuable, well-studied circuits and gates that Qiskit users can easily plug into their quantum circuits. These include useful tools for a wide array of quantum computing applications—including quantum machine learning, quantum chemistry, circuits for benchmarking, and even a number of circuits that are hard to simulate classically. Whether you’re a complete beginner or a bona fide Qiskit wizard, we’re sure you’ll find something useful in our extensive circuit library module.

Why every Qiskit user should give the circuit library a closer look

Before we launched the circuit library a few years ago, users could only run quantum circuits they had constructed manually. If a user wanted to implement a phase estimation, a Fourier transform, or a variational circuit—really anything at all—they either had to build it themselves, or build a function that generated the circuit for them. In either case, the process was cumbersome, and documentation was limited and disorganized.

As the Qiskit community began to grow, we developed a clearer sense of which circuits were most important to users. Eventually, we began to organize these into the circuit library we have today. This library is designed to be a repository of building blocks—a collection of any and all circuits we think are valuable to users. Here, we define “valuable” circuits as those that are either (1) widely used for common quantum computing tasks, (2) difficult to simulate classically, or (3) useful for quantum hardware benchmarking.

For example, the quantum Fourier transform (QFT) is an essential building block in a number of quantum algorithms including quantum phase estimation and the famous Shor’s algorithm. Now, if they really want to, a user can certainly run a QFT on quantum hardware by building the circuit manually. That might look something like this:

```
import numpy as np
from qiskit import QuantumCircuit

def build_qft(n):
    circuit = QuantumCircuit(n)
    for j in reversed(range(n)):
        circuit.h(j)
        for k in reversed(range(j)):
            circuit.cp(np.pi * 2. ** (k - j), j, k)

    for j in range(n // 2):
        circuit.swap(j, n - j - 1)

    return circuit

qft = build_qft(5)
```

But users who want to save time on this common quantum computing task can simply import the QFT from the circuit library and use it in their subroutines like so:

```
from qiskit.circuit.library import QFT

qft = QFT(5)
```

Most Qiskit users are aware the circuit library exists, but not everyone realizes just how useful it can be. That’s why we suggest all users spend time familiarizing themselves with it. The circuit library provides essential tools for building quantum algorithms, and gives users the ability to write programs at a higher level of abstraction. Gates and other elements found in the circuit library are developed so that users can easily plug them into their own quantum circuits using the `QuantumCircuit.append()` method, meaning users shouldn’t need to concern themselves with the lower level details of how those elements actually work.

The circuit library also provides tools that help users investigate properties of the quantum system itself, such as measuring quantum volume, and it offers useful information about the gates and circuits we use most often. It really is a library in the truest sense—in many cases; documentation for gates and other objects include links to a paper detailing the theory upon which they're based.

Now, let's take a look at some of the different kinds of circuits and gates that users will find in the Qiskit Circuit Library.

Variational quantum circuits (N-local circuits)

The **N-local circuits sub-module** of the circuit library contains a selection of subclasses that serve as parameterized models in **variational algorithms**, such as the variational quantum eigensolver (VQE) algorithm and the quantum approximate optimization algorithm (QAOA).

In fields like chemistry and physics, these algorithms serve to approximate wave functions efficiently, which is a task that is generally hard for classical computers. Variational algorithms are also useful for problems in classical optimization and machine learning, where a quantum computer might enable a better optimization of the objective.

Let's take a look at an example of how to define a problem instance for a VQE algorithm, using the `EfficientSU2` ansatz circuit taken from the circuit library. In this example, we'll derive our Hamiltonian from a quantum chemistry problem:

```
from qiskit.quantum_info import SparsePauliOp

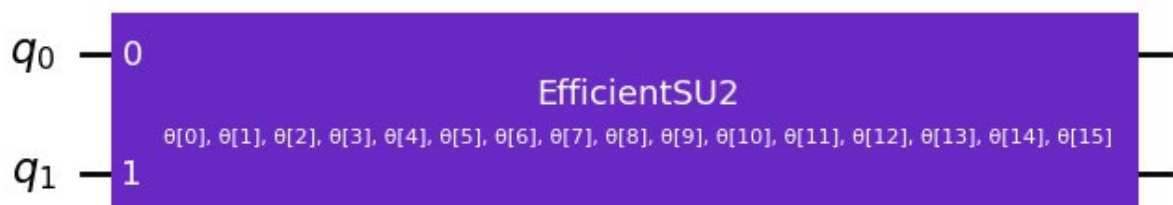
hamiltonian = SparsePauliOp.from_list(
    [("YZ", 0.3980), ("ZI", -0.3980), ("ZZ", -0.0113), ("XX", 0.1810)]
)
```

As mentioned above, our choice of ansatz is the `EfficientSU2` circuit. By default, the `EfficientSU2` circuit linearly entangles qubits. This makes it ideal for use in quantum hardware with limited connectivity:

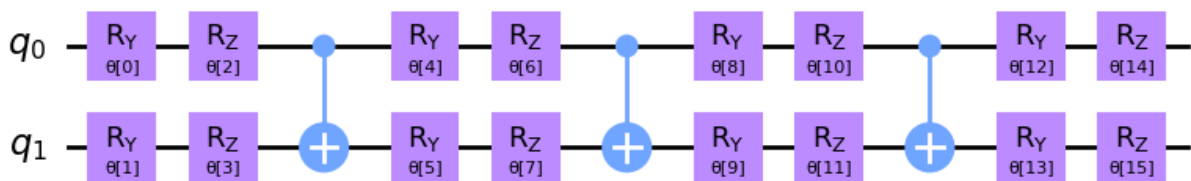
```
from qiskit.circuit.library import EfficientSU2

ansatz = EfficientSU2(hamiltonian.num_qubits)
ansatz.draw("mpl")
```

Here's what the circuit looks like:



Try printing with the following code to see how the circuit is structured!



Based on the previous figure, we know that our ansatz circuit is defined by a vector of parameters, θ_i , with the total number given by:

```
num_params = ansatz.num_parameters
print(num_params)
```

And in this case we get the output:

16

For an example on how to use the circuit library to run a VQE,

see: https://qiskit.org/ecosystem/algorithms/tutorials/01_algorithms_in_troduction.html

Data encoding circuits

Data encoding circuits enable the preparation of input data for QML applications. There are several techniques to encode classical data, or features, into a quantum state, for example:

Amplitude encoding: Encode the features into the amplitude of a basis state. This allows to store 2^n numbers in a single state, but can be costly to implement

Basis encoding: Encode an integer k by preparing the corresponding basis state $|k\rangle$

Angle encoding: Encode features as rotation angles in a variational quantum circuit

Deciding which of these approaches is best really depends upon the specifics of your application. On current quantum computers, however, we often use angle encoding.

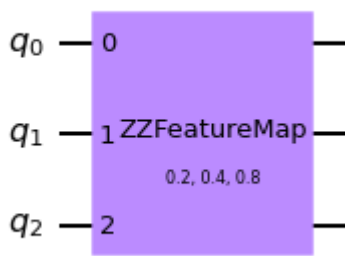
One example is the ZZFeatureMap found in the circuit library:

```
from qiskit.circuit.library import ZZFeatureMap

feature = [0.2, 0.4, 0.8]
feature_map = ZZFeatureMap(feature_dimension=len(feature))
```

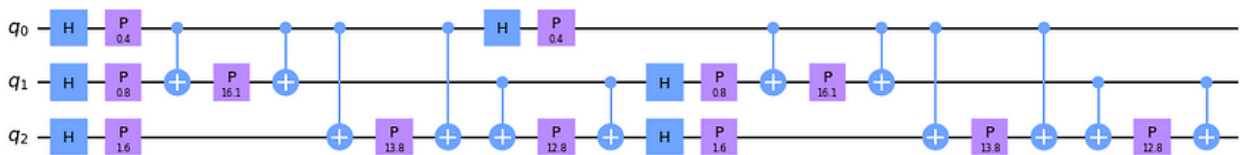
```
encoded = feature_map.assign_parameters(feature)
encoded.draw('mpl')
```

which prints



Try printing with the following code to see how the circuit is structured!

```
encoded.decompose().draw('mpl')
```



For an example of how to use data encoding circuits for classification,

see: https://qiskit.org/ecosystem/machine-learning/tutorials/o2_neural_network_classifier_and_regressor.html.

Time evolution circuits

Time evolution circuits allow us to evolve a quantum state in time. Physics researchers make use of them for a wide-array of applications, including investigations of thermalization properties or potential phase transitions of a system. Time evolution circuits are also a fundamental building block of chemistry wave functions—e.g., for unitary coupled-cluster trial states—and of the QAOA algorithm we use for optimization problems.

```
from qiskit.circuit.library import PauliEvolutionGate
from qiskit.circuit import QuantumCircuit
from qiskit.quantum_info import SparsePauliOp
```

```
hamiltonian = SparsePauliOp(["ZZI", "IZZ"])
```

```
# prepare an initial state with a Hamadard on the middle qubit
```

```
state = QuantumCircuit(3)
state.h(1)
```

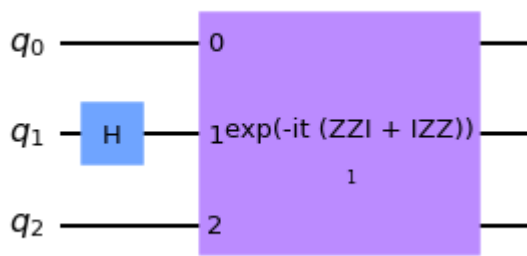
```
evolution = PauliEvolutionGate(hamiltonian, time=1)
```

```
# evolve state by appending the evolution gate
```

```
state.append(evolution)
```

```
state.draw('mpl')
```

And in this case we get the output:



Benchmarking and complexity theory circuits

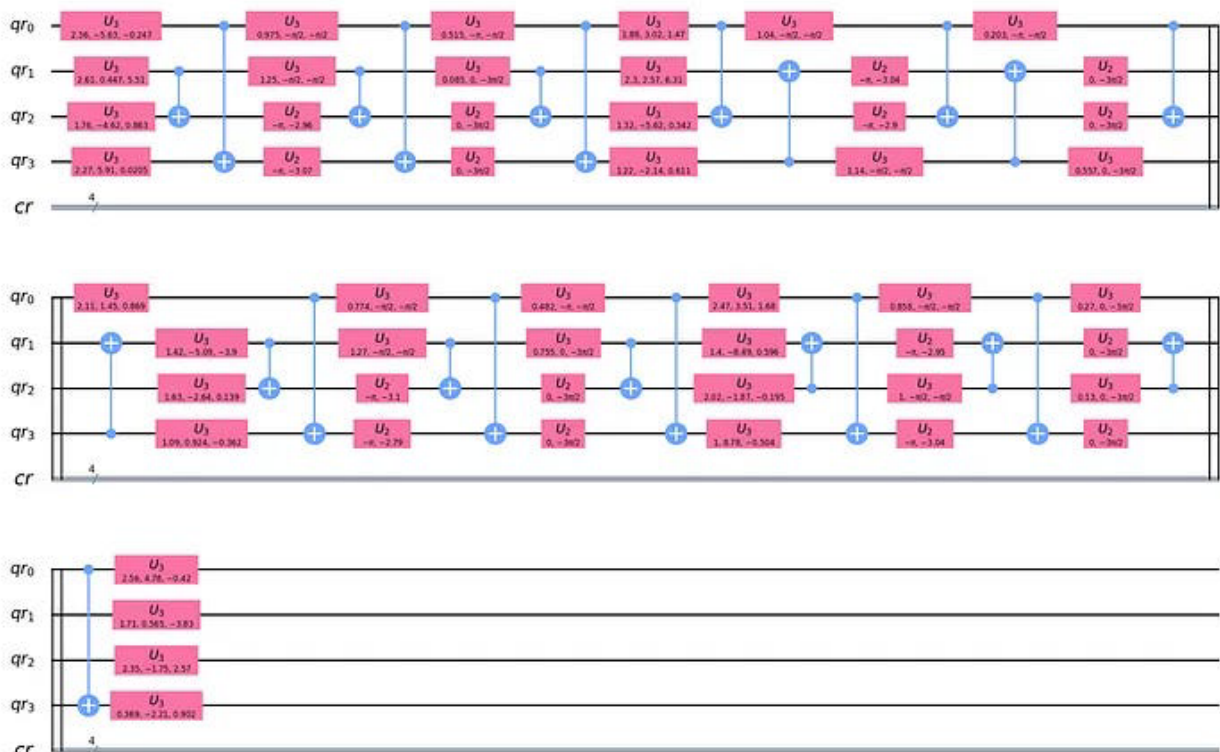
Benchmarking and complexity theory circuits are a vital tool for anyone working with quantum computers. They give us a sense of how well our hardware is actually working, and the true level of computational complexity involved in the problems we want to solve.

One of the most important examples of this is the [QuantumVolume](#) circuit, which allows us to quantify the largest random circuit of equal width and depth that a near-term quantum computer of relatively modest size can successfully implement. This single metric can help us understand how well the quantum computer is doing in terms of fidelity of operations, qubit connectivity, gate set calibration, and circuit rewiring tool chains.

A quantum volume model circuit comprises layers of Haar random elements of $SU(4)$ applied between corresponding pairs of qubits in a random bipartition. Here's an example of a quantum volume circuit built in Qiskit that runs on four qubits:

from qiskit.circuit.library import QuantumVolume

```
qv = QuantumVolume(4)
qv.decompose().decompose().draw('mpl')
```



Some quantum circuits have been shown to deliver advantage over classical algorithms. Many readers will already be familiar with some of the more famous examples of these circuits, which include Shor's algorithm for prime factorization and the Grover search algorithm.

The circuit library contains circuits that are conjectured to be hard to simulate classically, such as the IQP circuit. (Learn more about the IQP circuit [here](#).)

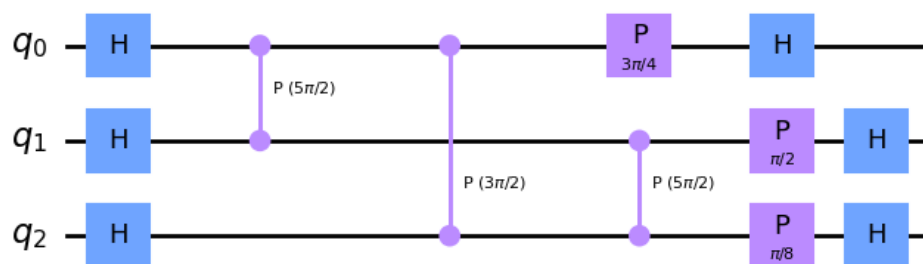
It also contains building blocks for algorithms that are believed to provide quantum advantage. Some of these are of a more practical nature. For example, Qiskit users can build the circuit for Grover's algorithm by using `GroverOperator`.

Others are of more theoretical interest. For example, Fourier checking uses the `FourierChecking` method. (Click [here](#) to learn more about Fourier checking.)

Let's take a look at an example of the IQP circuit:

```
from qiskit.circuit.library import IQP
```

```
interactions = [[6, 5, 3], [5, 4, 5], [3, 5, 1]]
circuit = IQP(interactions)
circuit.decompose().draw("mpl")
```



Arithmetic circuits

Some users may be surprised to learn that the circuit library includes numerous quantum circuits that are designed to perform classical arithmetic operations on qubit states or state amplitudes, such as addition or multiplication. These are useful for applications like amplitude estimation, which have the potential to be valuable in finance applications, and in algorithms like HHL, which is a proposed method for solving linear systems of equations.

As an example, let's try adding two 3-bit-long numbers, using a ripple-carry circuit to perform in-place addition (`CDKMRippleCarryAdder`). This adder takes two numbers—A and B—then it adds them, and “sets” B with the result. In this example, A=2 and B=3:

```
from qiskit.circuit.library import CDKMRippleCarryAdder
adder = CDKMRippleCarryAdder(3) # Create an adder of 3-bit-long numbers
```

```
from qiskit import QuantumCircuit, QuantumRegister, ClassicalRegister
```

```
# Create the number A=2
reg_a = QuantumRegister(3, 'a')
```

```

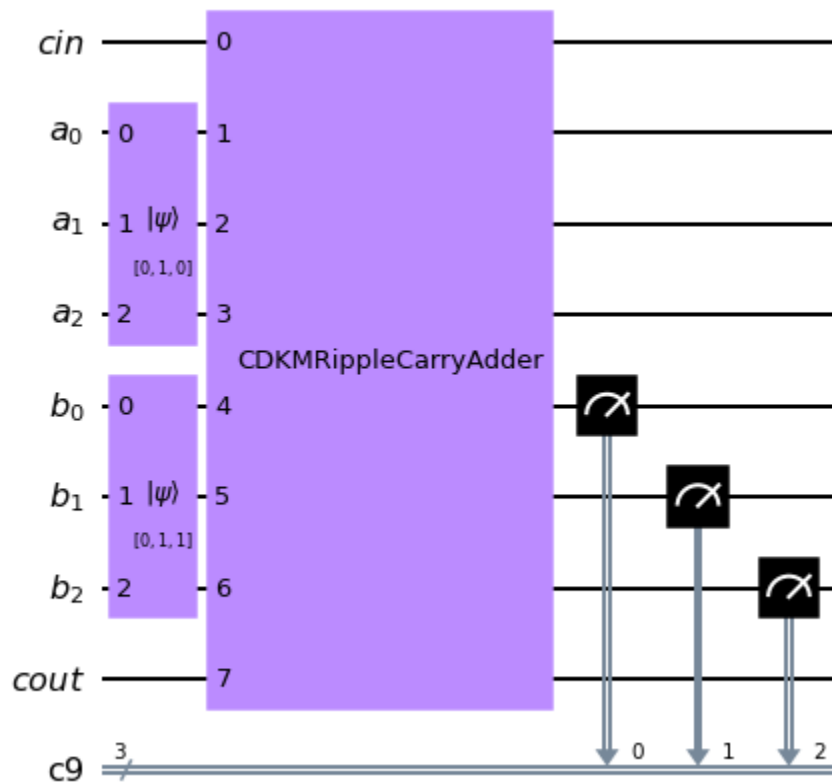
number_a = QuantumCircuit(reg_a)
number_a.initialize('010') # number 2 - equivalent to format(2, '03b')

# Create the number B=3
reg_b = QuantumRegister(3, 'b')
number_b = QuantumCircuit(reg_b)
number_b.initialize('011') # number 3 - equivalent to format(3, '03b')

# Create a circuit to hold everything, including a classical register to put the result
reg_result = ClassicalRegister(3)
circuit = QuantumCircuit(*adder.qregs, reg_result)

# Compose setting A and B with the adder. Measure the result in B
circuit = circuit.compose(number_a, qubits=reg_a).compose(number_b, qubits=reg_b).compose(adder)
circuit.measure(reg_b, reg_result)
circuit.draw('mpl')

```



By calling the sampler in `qiskit.primitives.Sampler`, it is possible to simulate the execution of this circuit and observe that the result is 5.

```

from qiskit.primitives import Sampler

result = Sampler().run(circuit).result()
print(result.quasi_dists[0].keys())

dict_keys([5])

```

Standard gates

The **Standard Gates** section of the Qiskit Circuit Library is a foundational toolbox for building up a wide array of quantum circuits—from the humble X Gate all the way to more complex multi-controlled rotation gates.

Users can add standard gates to their circuits by calling the gates as methods—e.g., `circuit.x(3)` will append an X Gate to the qubit 3. That's the most common way to use them. With that being said, this is still an incredibly useful section of the circuit library, and of our documentation in general, as it provides some very useful information about the matrices underlying these gates.

Let's take a look at an example of how to append a multi-controlled CNOT to an existing quantum circuit, using the circuit library as a reference. In the following example we are constructing a multi-controlled X Gate (`MCXGate`) with 4 control qubits from the circuit library and applying it to a quantum circuit, specifying the control qubits are at indices 0,1,2 and 4:

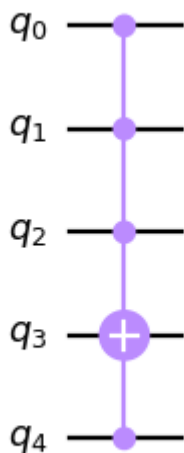
```
from qiskit import QuantumCircuit
from qiskit.circuit.library import MCXGate

# create circuit
circuit = QuantumCircuit(5)

# initialise MCX gate with 4 control qubits
gate = MCXGate(4)

# append gate to circuit, specifying qubit indices
circuit.append(gate, [0, 1, 4, 2, 3])
circuit.draw('mpl')
```

The resulting quantum circuit looks like this:



And here's an example of how we would transpile that to a quantum backend:

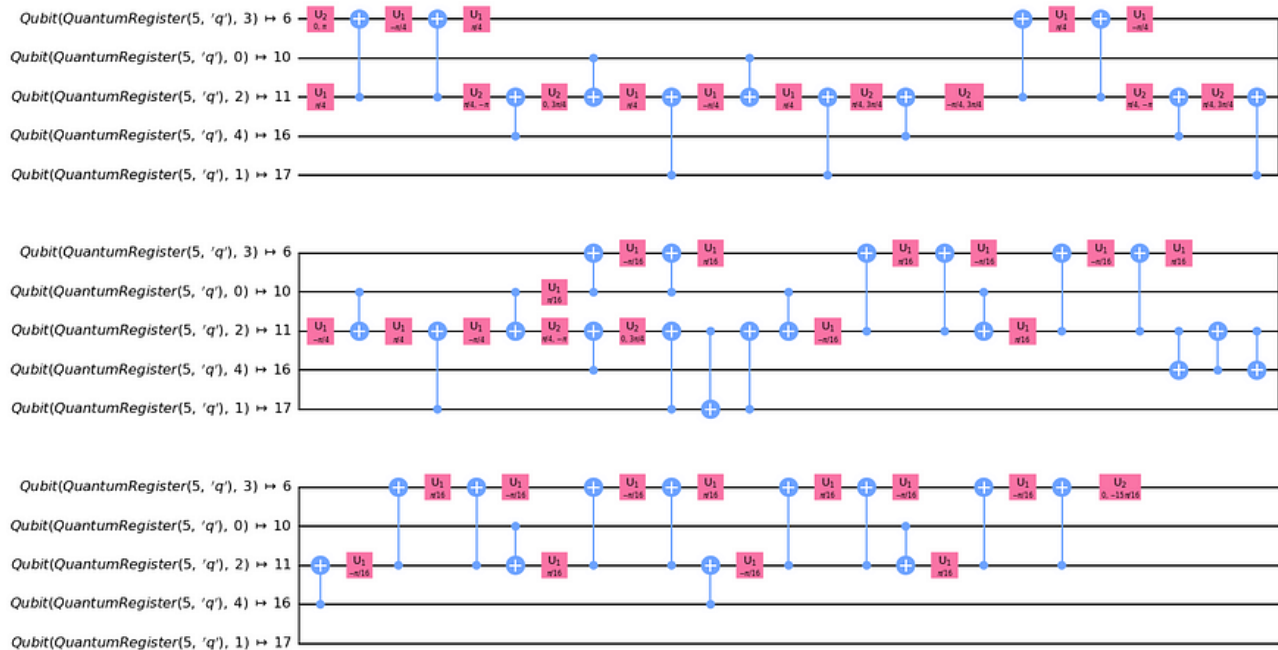
```
from qiskit import transpile
from qiskit.providers.fake_provider import FakeTokyo

backend = FakeTokyo()
```



```
transpiled = transpile(circuit, backend)
transpiled.draw('mpl', idle_wires=False)
```

Global Phase: 0



Other interesting circuits

This article only scratches the surface in terms of describing the many, many different types of circuits that are available in the Qiskit Circuit Library. Some other notable examples include the **quantum Fourier transform (QFT)**, which is an essential subroutine in many algorithms, **PhaseOracle**, which implements the oracles used in grover search, and **multi-control multi-target gate (MCMT)**, which makes it possible to (for example) implement an X gate acting on two qubits and controlled by three others.

API Reference:

<https://qiskit.org/documentation/apidoc/>

API Reference

Quantum Circuits (qiskit.circuit)

Circuit Library (qiskit.circuit.library)

Classical expressions (qiskit.circuit.classical)

Singleton instructions (qiskit.circuit.singleton)

Compilation Routines (qiskit.compiler)

Executing Experiments (qiskit.execute_function)

Visualizations (qiskit.visualization)

ClassicalFunction compiler (qiskit.circuit.classicalfunction)

Circuit Converters ([qiskit.converters](#))

Circuit and Schedule Assembler ([qiskit.assembler](#))

DAG Circuits ([qiskit.dagcircuit](#))

Quantum Circuit Extensions ([qiskit.extensions](#))

Passmanager ([qiskit.passmanager](#))

BasicAer: Python-based Simulators ([qiskit.providers.basicaer](#))

Providers Interface ([qiskit.providers](#))

Writing a New Provider

Migrating between Backend API Versions

Fake Provider ([qiskit.providers.fake_provider](#))

Backend Objects ([qiskit.providers.models](#))

Pulse ([qiskit.pulse](#))

Circuit Scheduler ([qiskit.scheduler](#))

Circuit Synthesis ([qiskit.synthesis](#))

Primitives ([qiskit.primitives](#))

OpenQASM 2 ([qiskit.qasm2](#))

OpenQASM 3 ([qiskit.qasm3](#))

Qasm ([qiskit.qasm](#))

Qobj ([qiskit.qobj](#))

QPY serialization ([qiskit.qpy](#))

Quantum Information ([qiskit.quantum_info](#))

Experiment Results ([qiskit.result](#))

Qiskit Tools ([qiskit.tools](#))

Jupyter Tools ([qiskit.tools.jupyter](#))

Transpiler ([qiskit.transpiler](#))

Transpiler Passes ([qiskit.transpiler.passes](#))

Preset Passmanagers ([qiskit.transpiler.preset_passmanagers](#))

Transpiler Stage Plugin Interface ([qiskit.transpiler.preset_passmanagers.plugin](#))

Synthesis Plugins ([qiskit.transpiler.passes.synthesis.plugin](#))

Built-in Transpiler Synthesis Plugins

Utilities ([qiskit.utils](#))

Measurement Mitigation Utils ([qiskit.utils.mitigation](#))

Top-level exceptions (qiskit.exceptions)

<https://www.avanade.com/nl-nl/blogs/be-orange/technology/quantum-computing-an-optimization-example>

Consider a company that needs to deliver thousands of packages to addresses in the Netherlands. They have a fleet of trucks and they want to optimize the route that each truck takes. This is a huge problem that is difficult to solve for classical computers given the massive number of possible routes. Quantum computers however, take a completely different approach to this problem. To illustrate the power of quantum, let's look at a scaled down version of this problem.

Suppose we have 1 truck that is tasked with driving to Amsterdam, Rotterdam and Den Haag, starting in and returning to Utrecht. What is the best order of visiting the cities? The table below contains all travel times between the cities in minutes. We will represent each of these segments with a qubit. Recall that a qubit is in a superposition of 1 and 0. Only upon measuring will the qubit take on one of these two values. If we measure a 1, we will travel over the corresponding segment, if the outcome is zero, we do not. The quantum algorithm is therefore responsible for manipulating the qubits in such a way that if we measure each qubit, it will give 1 for segments that are part of the optimal route, and 0 for segments that are not part of the optimal route.

Data, the distance between the cities: Amsterdam-Den Haag: 53km; Amsterdam-Rotterdam: 59km; Amsterdam-Utrecht: 46km; Den Haag-Rotterdam: 32km; Den Haag-Utrecht: 56km; Rotterdam-Utrecht: 51km

Step 1: formulate the problem

The first step now is to formulate the problem as a mathematical equation. I will skip the mathematics here, but I will sketch the idea behind it. If you are interested, the Traveling Santa blog explains very well how to get this equation.

Each segment of the route that we will take has a cost associated to it. In our case it is the number of minutes it will take to travel between two cities. Let's assume the measurements on our qubits yields 101011. That would mean we travel over segments S0, S2, S4, and S5, which corresponds to a route from Utrecht to Den Haag, then to Amsterdam, Rotterdam and finally back to Utrecht (see Figure 1). This will take us 219 minutes of driving time. The goal is to minimize the number of minutes, however because these are qubits, we could also measure 000000. This route will take 0 minutes, so that's better right? To ensure that the quantum algorithm that minimizes the driving time optimizes for a valid route, we add constraints in the form of penalties. For example, we can add a penalty of 100 minutes for each city that is not on the route. After you have identified all constraints, you end up with a lengthy equation as input for the algorithm.

Step 2: manipulate qubits

The next step is to find a way to manipulate our qubits such that upon measuring, the qubits that belong to segments that are part of the optimal route have a high probability to yield a value of 1. This will be done using a hybrid algorithm; it consists of a quantum part and a classical part. The quantum part manipulates the qubits based on the lengthy equation and a set of parameters passed to the algorithm. The purpose of these parameters will become evident in a little. The quantum part returns an estimate of the driving time for the current quantum state, including the penalties for invalid routes, to the classical part. Note the word 'estimate' in the previous sentence. The stochastic nature of quantum mechanics causes measurements on the exact same state to return different routes, with different duration.

As a simple example, consider a two-qubit quantum state that has a probability of 25% to be measured as 01, and 75% to be measured as 10 (and 0% for 00 and 11). Let's say route 01 takes 9 minutes, and route 10 takes 5 minutes, then we will measure the 5-minute route 3 times more often than the 9-minute route. On average, we can say that measurements on this state yields a route that takes 6 minutes. It is impossible however, to know the probabilities of the individual routes. All we know is that the way this state is currently prepared will yield a driving time of 6 minutes on average, if we were to perform many measurements on it. The important realization here is that the lower average driving time, the higher the probability of measuring a short route.

Information of the average driving time is then passed to the classical part. The classical part is an optimizer that tweaks the set of parameters passed to the quantum algorithm as to minimize the average driving time. In Figure 2 you can see the optimizer lowering the estimated driving time. Since we are running on a simulator, we can also inspect the quantum state during the optimization. So, to better understand what is happening, we plot in Figure 3 the probability of measuring each state during the optimization. On an actual quantum computer, we would not have access to this information.

In the top left picture, no optimization has been done, and each combination of routes to take and not to take have equal probability. The average driving time is relatively high (Figure 2 at 0 optimization steps). After approximately 3400 optimization steps, three states start to stand out. And in the final picture, we can see that 010111, 101011 and 111100 have a significantly higher probability to be measured than any other state. These 3 states are the only states that correspond to valid routes that visit each city exactly once. Among each other, 111100 has the highest probability to be measured. This state corresponds to driving along segments S0, S1, S2 and S3, or in words: starting in Utrecht, go to Rotterdam, then Den Haag, Amsterdam, and finally back to Utrecht (or in the exact opposite order). Indeed, we can calculate that of these three routes, this is the fastest with a total travel time of 182 minutes.

In step 2 we found a way to manipulate the qubits such that, upon measuring, there is relatively high probability to find the shortest route, but that is still a probability. So the final step is to repeat the optimal manipulation found by the optimizer multiple times. With enough measurements, the result that pops up the most corresponds to the optimal route.

Using better optimization and preparation algorithms than the ones currently implemented, it is possible to prepare a quantum state that has a larger than 80% probability to give an optimized solution to the problem.

In these series on quantum computing I tried to give a practical introduction to quantum computing as concise as possible. If you want to learn more, don't hesitate to contact me. I am more than happy to talk in more depth about quantum computing and programming.

https://quantumalgorithmzoo.org/traveling_santa/

The Microsoft Quantum Development Kit (QDK) has gates of the form $e^{-iX_j t/2}$ built in as `R(PauliX, t, qubit[j])`. Thus we can implement time evolution according to H_0 with the following Q# operation.

```
//This applies the X-rotation to each qubit. We can think of it as time
evolution
//induced by applying  $H = - \sum_i X_i$  for time t.
operation DriverHamiltonian(x: Qubit[], t: Double) : ()
{
```

```

for(i in 0..Length(x)-1)
{
    R(PauliX, -2.0*t, x[i]);
}
}

```

Similarly, all the individual terms in H_c commute. These come in two types: single-qubit Pauli Z and two-qubit ZZ couplings. The single-qubit Z rotations can be implemented using `R(PauliZ, t, qubit[j])`, which is built in to the QDK. To implement the unitaries of the form $e^{-iZ_i Z_j t}$ we note that this applies a phase determined by the exclusive-or of qubits i and j . Thus, we can compute this exclusive-or into an ancilla qubit, apply a Z-rotation on the ancilla, and then uncompute. Putting this all together (and cancelling some redundant CNOT gates) yields the following Q# function to implement $e^{-iH_c t}$.

```

//This applies the Z-rotation according to the instance Hamiltonian.
//We can think of it as Hamiltonian time evolution for time t induced
//by the Ising Hamiltonian \sum_{ij} J_{ij} Z_i Z_j + \sum_i h_i Z_i.
operation InstanceHamiltonian(z: Qubit[], t: Double, h: Double[], J: Double[]) :
Unit
{
    using (ancilla = Qubit[1])
    {
        for(i in 0..5)
        {
            R(PauliZ, 2.0*t*h[i], z[i]);
        }
        for(i in 0..5)
        {
            for (j in i+1..5)
            {
                CNOT(z[i], ancilla[0]);
                CNOT(z[j], ancilla[0]);
                R(PauliZ, 2.0*t*J[6*i+j], ancilla[0]);
                CNOT(z[i], ancilla[0]);
                CNOT(z[j], ancilla[0]);
            }
        }
    }
}

```

Then, we call these operations from an overall QAOA method as follows.

```

// Here is a QAOA algorithm for this Ising Hamiltonian
operation QAOA_santa(segmentCosts:Double[], penalty:Double, tx: Double[], tz:
Double[], p: Int) : Bool[]
{
    // Calculate Hamiltonian parameters based on the given costs and penalty
    mutable J = new Double[36];
}

```

```

mutable h = new Double[6];
for (i in 0..5) {
    set h[i] = 4.0 * penalty - 0.5 * segmentCosts[i];
}
// Most elements of J_ij equal 2*penalty, so set all elements to this value,
then overwrite the exceptions
for (i in 0..35)
{
    set J[i] = 2.0 * penalty;
}
set J[2] = penalty;
set J[9] = penalty;
set J[29] = penalty;
// Now run the QAOA circuit
mutable r = new Bool[6];
using (x = Qubit[6])
{
    ApplyToEach(H, x); // prepare the uniform
distribution
    for (i in 0..p-1)
    {
        InstanceHamiltonian(x, tz[i], h, J); // do Exp(-i H_C tz[i])
        DriverHamiltonian(x, tx[i]); // do Exp(-i H_0 tx[i])
    }
    set r = MeasureAllReset(x); // measure in the
computational basis
}
return r;
}

```

One of the key elements of a QAOA algorithm is a good choice of durations to apply the Hamiltonians.

Max-Cut and Traveling Salesman Problem

https://qiskit.org/ecosystem/optimization/tutorials/06_examples_max_cut_and_tsp.html#

Introduction

Many problems in quantitative fields such as finance and engineering are optimization problems. Optimization problems lie at the core of complex decision-making and definition of strategies.

Optimization (or combinatorial optimization) means searching for an optimal solution in a finite or countably infinite set of potential solutions. Optimality is defined with respect to some criterion function, which is to be minimized or maximized. This is typically called cost function or objective function.

Typical optimization problems

Minimization: cost, distance, length of a traversal, weight, processing time, material, energy consumption, number of objects

Maximization: profit, value, output, return, yield, utility, efficiency, capacity, number of objects

We consider here max-cut problems of practical interest in many fields, and show how they can be mapped on quantum computers manually and how Qiskit optimization module supports this.

Weighted Max-Cut

Max-Cut is an NP-complete problem, with applications in clustering, network science, and statistical physics. To grasp how practical applications are mapped into given Max-Cut instances, consider a system of many people that can interact and influence each other. Individuals can be represented by vertices of a graph, and their interactions seen as pairwise connections between vertices of the graph, or edges. With this representation in mind, it is easy to model typical marketing problems. For example, suppose that it is assumed that individuals will influence each other's buying decisions, and knowledge is given about how strong they will influence each other. The influence can be modeled by weights assigned on each edge of the graph. It is possible then to predict the outcome of a marketing strategy in which products are offered for free to some individuals, and then ask which is the optimal subset of individuals that should get the free products, in order to maximize revenues.

The formal definition of this problem is the following:

Consider an n -node undirected graph $G = (V, E)$ where $|V| = n$ with edge weights $w_{ij} > 0$, $w_{ij} = w_{ji}$, for $(i, j) \in E$. A cut is defined as a partition of the original set V into two subsets. The cost function to be optimized is in this case the sum of weights of edges connecting points in the two different subsets, *crossing* the cut. By assigning $x_i = 0$ or $x_i = 1$ to each node i , one tries to maximize the global profit function (here and in the following summations run over indices $0, 1, \dots, n-1$)

$$P(x) = \sum_{(i,j) \in E} w_{ij} x_i (1 - x_j).$$

In our simple marketing model, x_i represents the probability that the person i will buy a product after i gets a free one. Note that the weights w_{ij} can in principle be greater than 1 (or even negative), corresponding to the case where the individual i will buy more than one product. Maximizing the total buying probability corresponds to maximizing the total future revenues. In the case where the profit probability will be greater than the cost of the initial free samples, the strategy is a convenient one. An extension to this model has the nodes themselves carry weights, which can be regarded, in our marketing model, as the likelihood that a person granted with a free sample of the product will buy it again in the future. With this additional information in our model, the objective function to maximize becomes

$$P(x) = \sum_i w_i x_i + \sum_{(i,j) \in E} w_{ij} x_i (1 - x_j).$$

In order to find a solution to this problem on a quantum computer, one needs first to map it to an Ising Hamiltonian. This can be done with the assignment $x_i \rightarrow (1 - \sigma_i)/2$, where σ_i is the Pauli Z operator that has eigenvalues ± 1 . Doing this we find that

$$P(Z) = \sum_i w_i (1 - \sigma_i)/2 + \sum_{(i,j) \in E} w_{ij} (1 - \sigma_i)/2 (1 + \sigma_j)/2 = -12 \left(\sum_{(i,j) \in E} w_{ij} \sigma_i \sigma_j + \sum_i w_i \sigma_i \right) + \text{const},$$

where $\text{const} = \sum_i w_i < w_{ij} > /2 + \sum_{(i,j) \in E} w_{ij} /2$. In other terms, the weighted Max-Cut problem is equivalent to minimizing the Ising Hamiltonian

$$H = \sum_{(i,j) \in E} w_{ij} \sigma_i \sigma_j + \sum_i w_i \sigma_i.$$

Qiskit optimization module can generate the Ising Hamiltonian for the first profit function $P(x)$. To this extent, function $P(x)$ can be modeled as a QuadraticProgram, which provides the `to_ising()` method.

Approximate Universal Quantum Computing for Optimization Problems

There has been a considerable amount of interest in recent times about the use of quantum computers to find a solution to combinatorial optimization problems. It is important to say that, given the classical nature of combinatorial problems, exponential speedup in using quantum computers compared to the best classical algorithms is not guaranteed. However, due to the nature and importance of the target problems, it is worth investigating heuristic approaches on a quantum computer that could indeed speed up some problem instances. Here we demonstrate an approach that is based on the *Quantum Approximate Optimization Algorithm* (QAOA) by Farhi, Goldstone, and Gutmann (2014). We frame the algorithm in the context of *approximate quantum computing*, given its heuristic nature.

The algorithm works as follows:

Choose the Q and P in the target Ising problem. In principle, even higher powers of Z are allowed.

Choose the depth of the quantum circuit D . Note that the depth can be modified adaptively.

Choose a set of controls γ and make a trial function $|\psi(\gamma)\rangle$, built using a quantum circuit made of C-Phase gates and single-qubit Y rotations, parameterized by the components of γ .

Evaluate

$$\langle H \rangle = \langle \psi(\gamma) | H | \psi(\gamma) \rangle = \sum_{i,j} J_{ij} \langle \psi(\gamma) | \sigma_i^z \sigma_j^z | \psi(\gamma) \rangle + \sum_i \gamma_i \langle \psi(\gamma) | \sigma_i^z | \psi(\gamma) \rangle$$

by sampling the outcome of the circuit in the Z -basis and adding the expectation values of the individual Ising terms together. In general, different control points around γ have to be estimated, depending on the classical optimizer chosen.

Use a classical optimizer to choose a new set of controls.

Continue until $\langle H \rangle$ reaches a minimum, close enough to the solution H^* .

Use the last γ to generate a final set of samples from the distribution $|\langle \psi(\gamma) | \psi(\gamma) \rangle| 2^N$ to obtain the answer.

It is our belief the difficulty of finding good heuristic algorithms will come down to the choice of an appropriate trial wavefunction. For example, one could consider a trial function whose entanglement best aligns with the target problem, or simply make the amount of entanglement a variable. In this tutorial, we will consider a simple trial function of the form

$$|\psi(\gamma)\rangle = [\text{single}(\gamma) \text{entangler}]^D |+\rangle$$

where entangler is a collection of C-Phase gates (fully entangling gates), and $\text{single}(\gamma) = \prod_{i=1}^N \sigma_i^y e^{i\gamma_i \sigma_i^z}$, where N is the number of qubits and D is the depth of the quantum circuit. The motivation for this choice is that for these classical problems this choice allows us to search over the space of quantum states that have only real coefficients, still exploiting the entanglement to potentially converge faster to the solution.

One advantage of using this sampling method compared to adiabatic approaches is that the target Ising Hamiltonian does not have to be implemented directly on hardware, allowing this algorithm not to be limited to the connectivity of the device. Furthermore, higher-order terms in the cost function, such as $\sigma_i^z \sigma_j^z \sigma_k^z$, can also be sampled efficiently, whereas in adiabatic or annealing approaches they are generally impractical to deal with.

References:

- A. Lucas, Frontiers in Physics 2, 5 (2014)
- E. Farhi, J. Goldstone, S. Gutmann, e-print arXiv 1411.4028 (2014)
- D. Wecker, M. B. Hastings, M. Troyer, Phys. Rev. A 94, 022309 (2016)
- E. Farhi, J. Goldstone, S. Gutmann, H. Neven, e-print arXiv 1703.06199 (2017)

Application classes

We use the application classes for the max-cut problem and the traveling salesman problem in this page. There are application classes for other optimization problems available as well. See [Application Classes for Optimization Problems](#) for details.

[1]:

```
# useful additional packages
```

```
import matplotlib.pyplot as plt
```

```
import numpy as np
```

```
import networkx as nx
```

```
from qiskit.tools.visualization import plot_histogram
```

```
from qiskit.circuit.library import TwoLocal
```

```
from qiskit_optimization.applications import Maxcut, Tsp
```

```
from qiskit_algorithms import SamplingVQE, NumPyMinimumEigensolver
```

```
from qiskit_algorithms.optimizers import SPSA
```

```
from qiskit_algorithms.utils import algorithm_globals
```

```
from qiskit.primitives import Sampler
```

```
from qiskit_optimization.algorithms import MinimumEigenOptimizer
```

Max-Cut problem

[2]:

```
# Generating a graph of 4 nodes
```

```
n = 4 # Number of nodes in graph
```

```
G = nx.Graph()
```

```
G.add_nodes_from(np.arange(0, n, 1))
```

```
elist = [(0, 1, 1.0), (0, 2, 1.0), (0, 3, 1.0), (1, 2, 1.0), (2, 3, 1.0)]
```

```
# tuple is (i,j,weight) where (i,j) is the edge
```

```
G.add_weighted_edges_from(elist)
```

```
colors = ["r" for node in G.nodes()]
```

```
pos = nx.spring_layout(G)
```

```
def draw_graph(G, colors, pos):
```

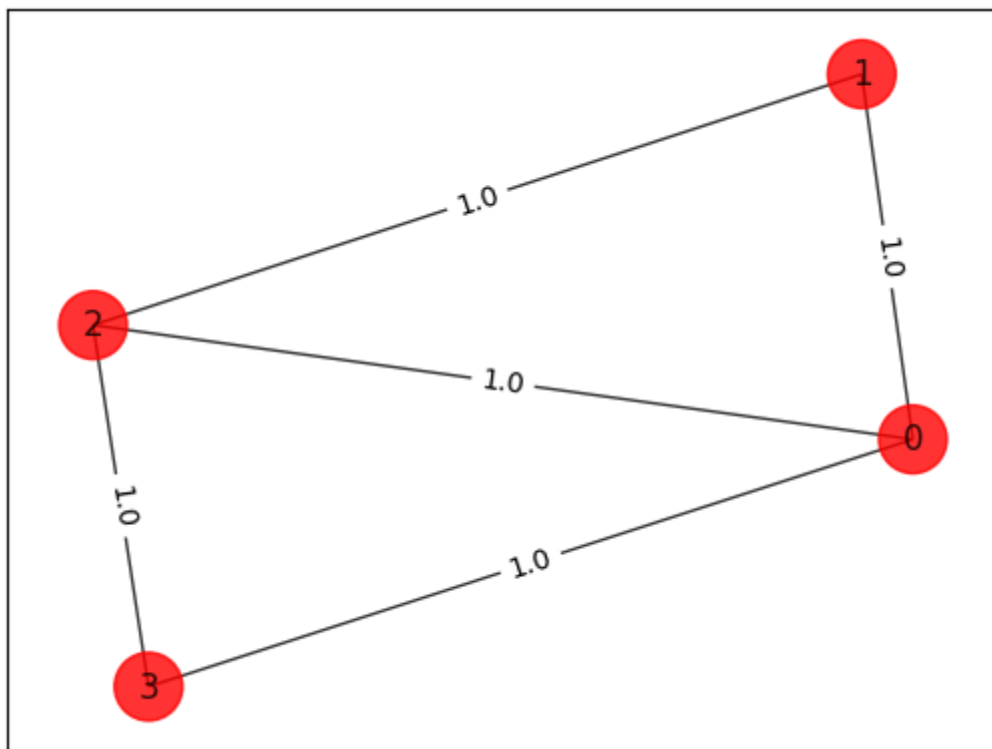
```
    default_axes = plt.axes(frameon=True)
```

```
    nx.draw_networkx(G, node_color=colors, node_size=600, alpha=0.8, ax=default_axes, pos=pos)
```

```
    edge_labels = nx.get_edge_attributes(G, "weight")
```

```
    nx.draw_networkx_edge_labels(G, pos=pos, edge_labels=edge_labels)
```

```
draw_graph(G, colors, pos)
```



```
[3]:
```

```
# Computing the weight matrix from the random graph
```

```
w = np.zeros([n, n])
```

```
for i in range(n):
```

```
    for j in range(n):
```

```
        temp = G.get_edge_data(i, j, default=0)
```

```

    if temp != 0:
        w[i, j] = temp["weight"]
print(w)
[[0. 1. 1. 1.]
 [1. 0. 1. 0.]
 [1. 1. 0. 1.]
 [1. 0. 1. 0.]]

```

Brute force approach

Try all possible 2^n combinations. For $n=4$, as in this example, one deals with only 16 combinations, but for $n = 1000$, one has $1.071509e+30$ combinations, which is impractical to deal with by using a brute force approach.

[4]:

```

best_cost_brute = 0
for b in range(2**n):
    x = [int(t) for t in reversed(list(bin(b)[2:].zfill(n)))]
    cost = 0
    for i in range(n):
        for j in range(n):
            cost = cost + w[i, j] * x[i] * (1 - x[j])
    if best_cost_brute < cost:
        best_cost_brute = cost
        xbest_brute = x
    print("case = " + str(x) + " cost = " + str(cost))

colors = ["r" if xbest_brute[i] == 0 else "c" for i in range(n)]
draw_graph(G, colors, pos)
print("\nBest solution = " + str(xbest_brute) + " cost = " + str(best_cost_brute))

case = [0, 0, 0, 0] cost = 0.0
case = [1, 0, 0, 0] cost = 3.0
case = [0, 1, 0, 0] cost = 2.0
case = [1, 1, 0, 0] cost = 3.0
case = [0, 0, 1, 0] cost = 3.0
case = [1, 0, 1, 0] cost = 4.0

```

case = [0, 1, 1, 0] cost = 3.0

case = [1, 1, 1, 0] cost = 2.0

case = [0, 0, 0, 1] cost = 2.0

case = [1, 0, 0, 1] cost = 3.0

case = [0, 1, 0, 1] cost = 4.0

case = [1, 1, 0, 1] cost = 3.0

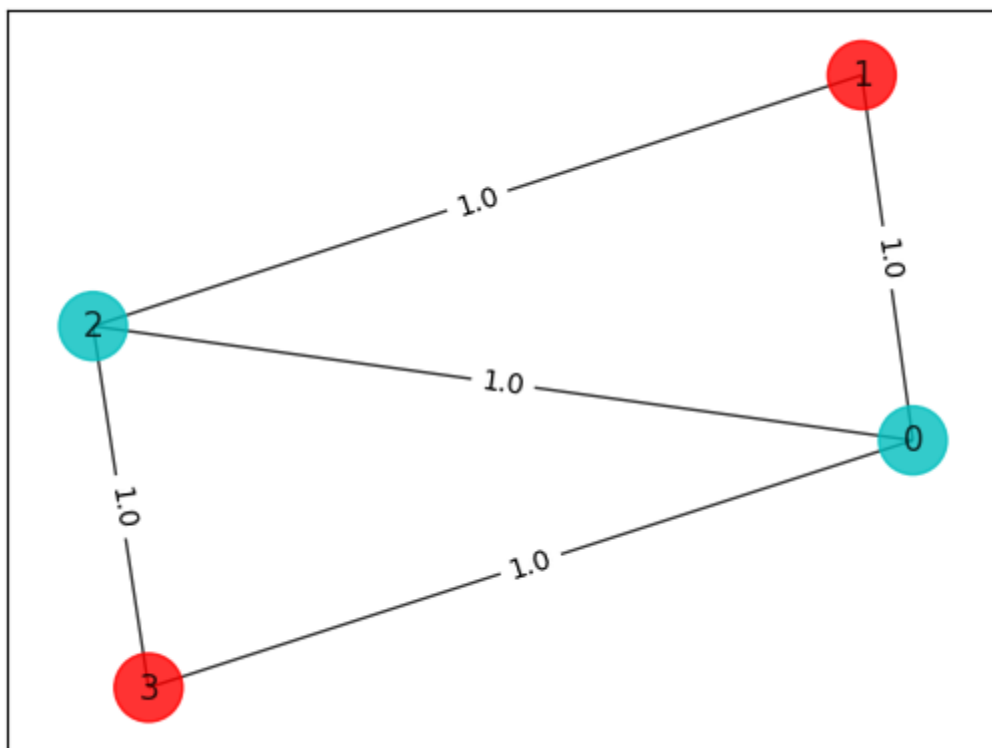
case = [0, 0, 1, 1] cost = 3.0

case = [1, 0, 1, 1] cost = 2.0

case = [0, 1, 1, 1] cost = 3.0

case = [1, 1, 1, 1] cost = 0.0

Best solution = [1, 0, 1, 0] cost = 4.0



Mapping to the Ising problem

Qiskit optimization provides functionality to generate QuadraticProgram from the problem specification as well as create the corresponding Ising Hamiltonian.

[5]:

```
max_cut = Maxcut(w)
```

```
qp = max_cut.to_quadratic_program()
```

```
print(qp.prettyprint())
```

Problem name: Max-cut

Maximize

$$\begin{aligned} & -2x_0x_1 - 2x_0x_2 - 2x_0x_3 - 2x_1x_2 - 2x_2x_3 + 3x_0 + 2x_1 \\ & + 3x_2 + 2x_3 \end{aligned}$$

Subject to

No constraints

Binary variables (4)

x_0 x_1 x_2 x_3

[6]:

```
qubitOp, offset = qp.to_ising()
```

```
print("Offset:", offset)
```

```
print("Ising Hamiltonian:")
```

```
print(str(qubitOp))
```

Offset: -2.5

Ising Hamiltonian:

```
SparsePauliOp(['IIZZ', 'IZIZ', 'IZZI', 'ZIIZ', 'ZZII'],
```

```
coeffs=[0.5+0.j, 0.5+0.j, 0.5+0.j, 0.5+0.j, 0.5+0.j])
```

[7]:

```
# solving Quadratic Program using exact classical eigensolver
```

```
exact = MinimumEigenOptimizer(NumPyMinimumEigensolver())
```

```
result = exact.solve(qp)
```

```
print(result.prettyprint())
```

objective function value: 4.0

variable values: $x_0=1.0$, $x_1=0.0$, $x_2=1.0$, $x_3=0.0$

status: SUCCESS

Since the problem was cast to a minimization problem, the solution of -4 corresponds to the optimum.

Checking that the full Hamiltonian gives the right cost

[8]:

```
# Making the Hamiltonian in its full form and getting the lowest eigenvalue and eigenvector
```

```
ee = NumPyMinimumEigsolver()
```

```
result = ee.compute_minimum_eigenvalue(qubitOp)
```

```
x = max_cut.sample_most_likely(result.eigenstate)
```

```
print("energy:", result.eigenvalue.real)
```

```
print("max-cut objective:", result.eigenvalue.real + offset)
```

```
print("solution:", x)
```

```
print("solution objective:", qp.objective.evaluate(x))
```

```
colors = ["r" if x[i] == 0 else "c" for i in range(n)]
```

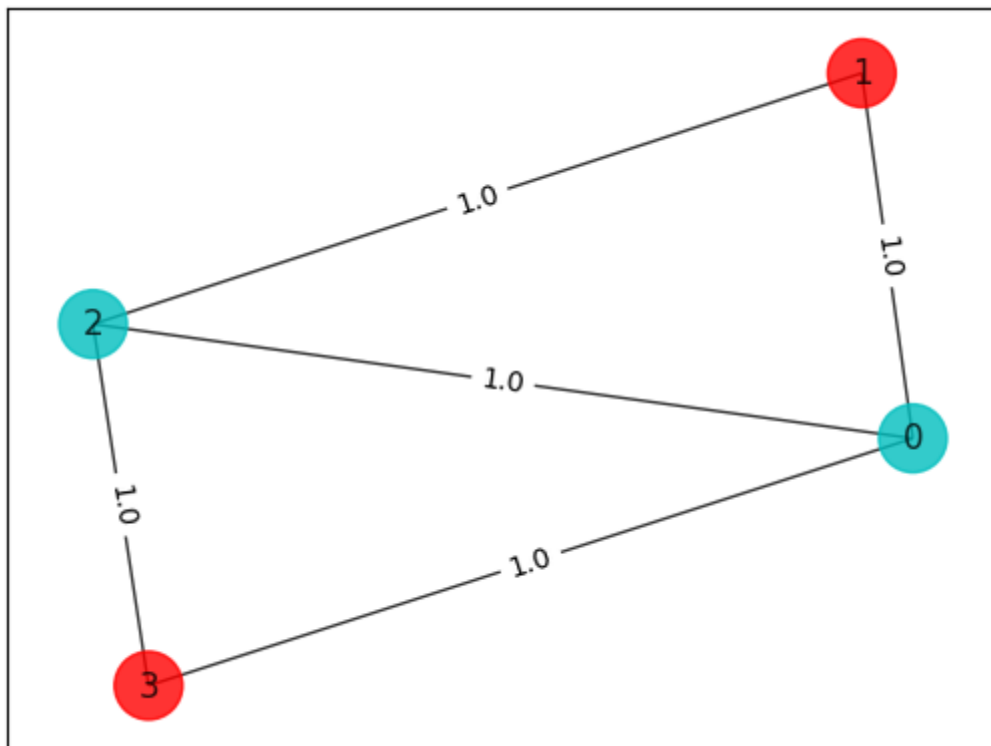
```
draw_graph(G, colors, pos)
```

```
energy: -1.5
```

```
max-cut objective: -4.0
```

```
solution: [1. 0. 1. 0.]
```

```
solution objective: 4.0
```



Running it on quantum computer

We run the optimization routine using a feedback loop with a quantum computer that uses trial functions built with Y single-qubit rotations, $\text{single}(\diamond) = \prod \diamond = 1 \diamond \diamond (\diamond \diamond)$, and entangler steps \diamond entangler.

[9]:

```
algorithm_globals.random_seed = 123
```

```
seed = 10598
```

[10]:

```
# construct SamplingVQE
```

```
optimizer = SPSA(maxiter=300)
```

```
ry = TwoLocal(qubitOp.num_qubits, "ry", "cz", reps=5, entanglement="linear")
```

```
vqe = SamplingVQE(sampler=Sampler(), ansatz=ry, optimizer=optimizer)
```

```
# run SamplingVQE
```

```
result = vqe.compute_minimum_eigenvalue(qubitOp)
```

```
# print results
```

```
x = max_cut.sample_most_likely(result.eigenstate)
```

```
print("energy:", result.eigenvalue.real)
```

```
print("time:", result.optimizer_time)
```

```
print("max-cut objective:", result.eigenvalue.real + offset)
```

```
print("solution:", x)
```

```
print("solution objective:", qp.objective.evaluate(x))
```

```
# plot results
```

```
colors = ["r" if x[i] == 0 else "c" for i in range(n)]
```

```
draw_graph(G, colors, pos)
```

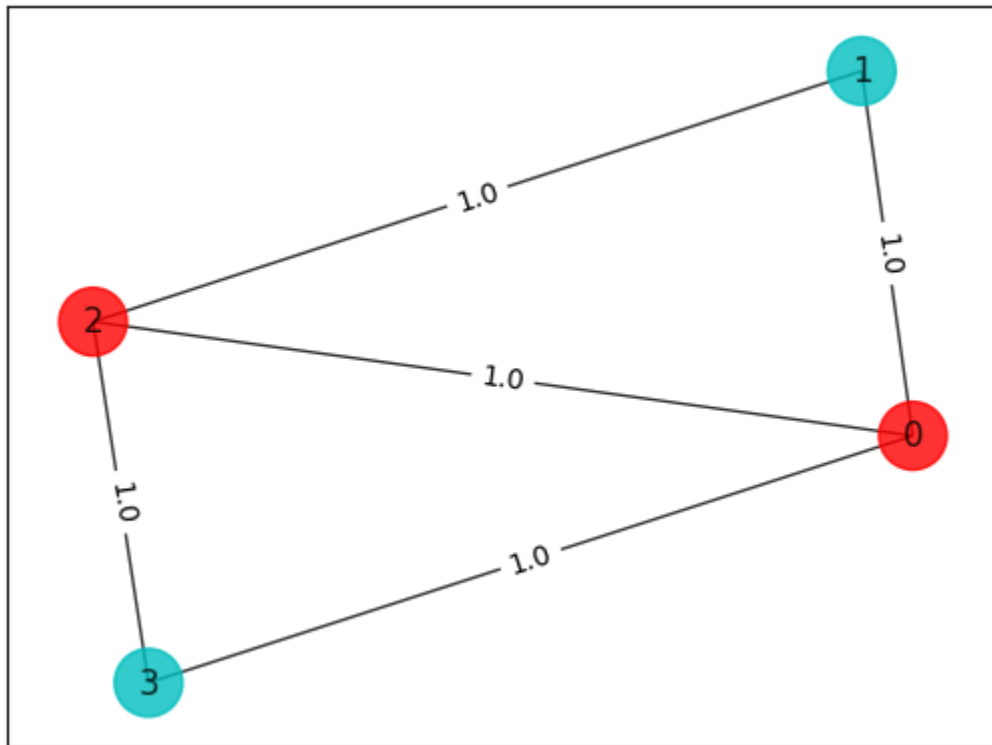
```
energy: -1.4996861455587294
```

```
time: 3.004215717315674
```

```
max-cut objective: -3.999686145558729
```

```
solution: [0 1 0 1]
```

```
solution objective: 4.0
```



[11]:

```
# create minimum eigen optimizer based on SamplingVQE
```

```
vqe_optimizer = MinimumEigenOptimizer(vqe)
```

```
# solve quadratic program
```

```
result = vqe_optimizer.solve(qp)
```

```
print(result.prettyprint())
```

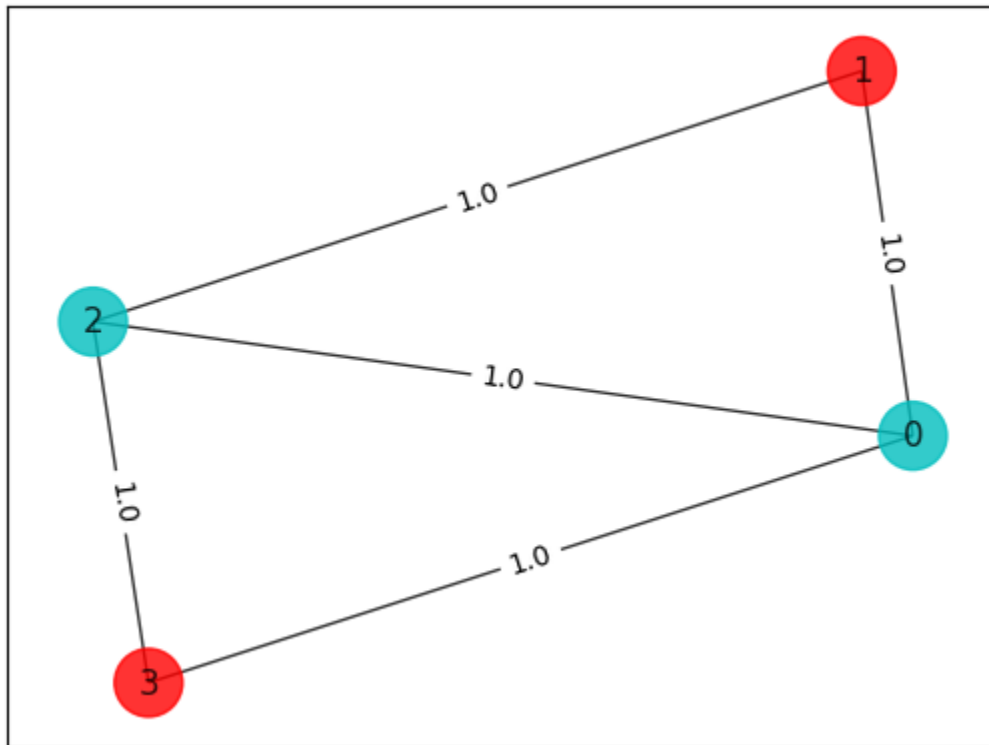
```
colors = ["r" if result.x[i] == 0 else "c" for i in range(n)]
```

```
draw_graph(G, colors, pos)
```

```
objective function value: 4.0
```

```
variable values: x_0=1.0, x_1=0.0, x_2=1.0, x_3=0.0
```

```
status: SUCCESS
```

Traveling Salesman Problem

In addition to being a notorious NP-complete problem that has drawn the attention of computer scientists and mathematicians for over two centuries, the Traveling Salesman Problem (TSP) has important bearings on finance and marketing, as its name suggests. Colloquially speaking, the traveling salesman is a person that goes from city to city to sell merchandise. The objective in this case is to find the shortest path that would enable the salesman to visit all the cities and return to its hometown, i.e. the city where he started traveling. By doing this, the salesman gets to maximize potential sales in the least amount of time.

The problem derives its importance from its “hardness” and ubiquitous equivalence to other relevant combinatorial optimization problems that arise in practice.

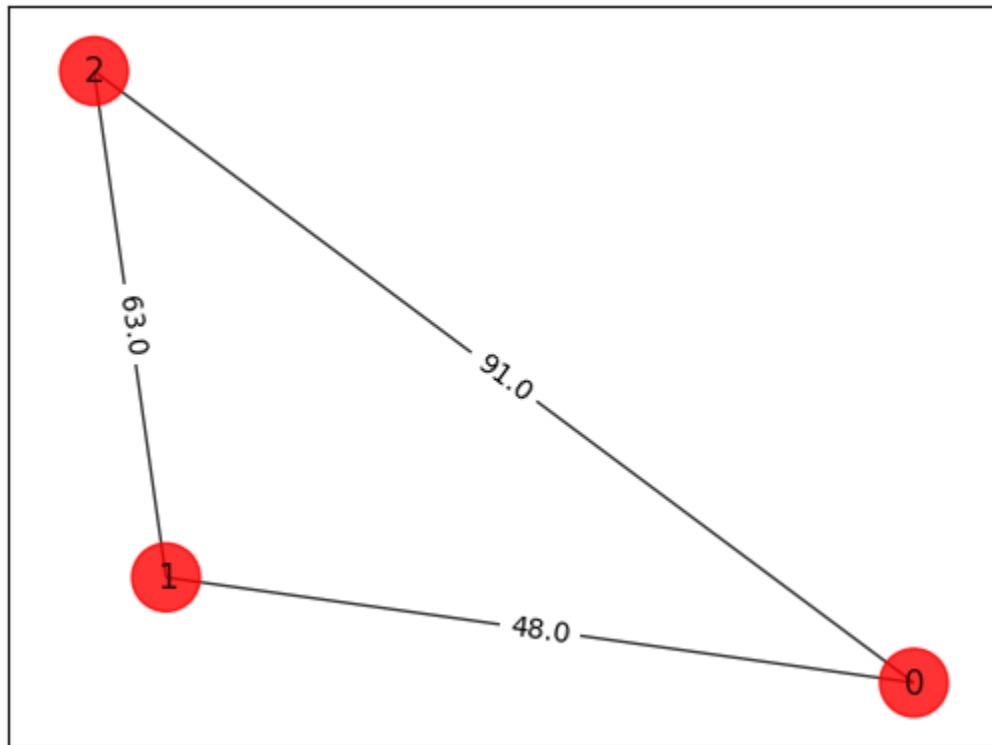
The mathematical formulation with some early analysis was proposed by W.R. Hamilton in the early 19th century. Mathematically the problem is, as in the case of Max-Cut, best abstracted in terms of graphs. The TSP on the nodes of a graph asks for the shortest *Hamiltonian cycle* that can be taken through each of the nodes. A Hamiltonian cycle is a closed path that uses every vertex of a graph once. The general solution is unknown and an algorithm that finds it efficiently (e.g., in polynomial time) is not expected to exist.

Find the shortest Hamiltonian cycle in a graph $G=(V,E)$ with $N=|V|$ nodes and distances, d_{ij} (distance from vertex i to vertex j). A Hamiltonian cycle is described by N variables x_{ij} , where i represents the node and j represents its order in a prospective cycle. The decision variable takes the value 1 if the solution occurs at node i at time order j . We require that every node can only appear once in the cycle, and for each time a node has to occur. This amounts to the two constraints (here and in the following, whenever not specified, the summands run over $0,1,...,N-1$)

$$\sum_j x_{ij} = 1 \quad \forall i$$

$$\sum_i x_{ij} = 1 \quad \forall j.$$

For nodes in our prospective ordering, if x_{ij} and $x_{i+1,j+1}$ are both 1, then there should be an energy penalty if $(i,j) \notin E$ (not connected in the graph). The form of this penalty is



Brute force approach

[13]:

```
from itertools import permutations
```

```
def brute_force_tsp(w, N):
    a = list(permutations(range(1, N)))
    last_best_distance = 1e10
    for i in a:
        distance = 0
        pre_j = 0
        for j in i:
            distance = distance + w[j, pre_j]
            pre_j = j
        distance = distance + w[pre_j, 0]
        order = (0,) + i
        if distance < last_best_distance:
            best_order = order
```

```

        last_best_distance = distance

        print("order = " + str(order) + " Distance = " + str(distance))

    return last_best_distance, best_order

```

```

best_distance, best_order = brute_force_tsp(adj_matrix, n)

print(
    "Best order from brute force = "
    + str(best_order)
    + " with total distance = "
    + str(best_distance)
)

```

```

def draw_tsp_solution(G, order, colors, pos):
    G2 = nx.DiGraph()
    G2.add_nodes_from(G)

    n = len(order)
    for i in range(n):
        j = (i + 1) % n
        G2.add_edge(order[i], order[j], weight=G[order[i]][order[j]]["weight"])

    default_axes = plt.axes(frameon=True)
    nx.draw_networkx(
        G2, node_color=colors, edge_color="b", node_size=600, alpha=0.8, ax=default_axes, pos=pos
    )

    edge_labels = nx.get_edge_attributes(G2, "weight")
    nx.draw_networkx_edge_labels(G2, pos, font_color="b", edge_labels=edge_labels)

```

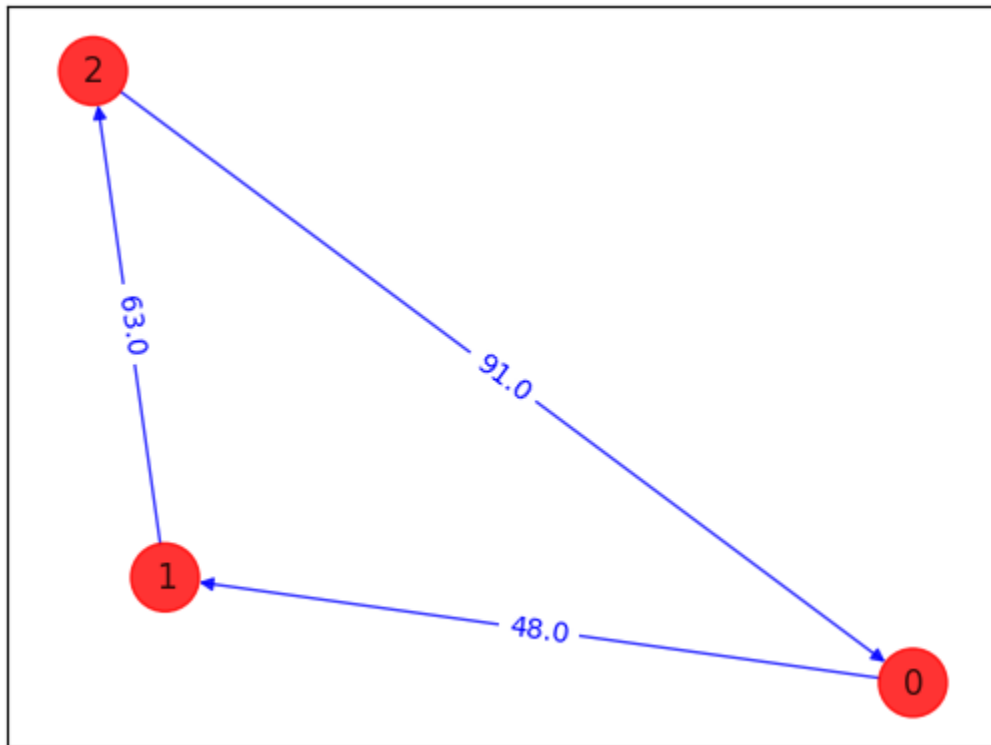
```

draw_tsp_solution(tsp.graph, best_order, colors, pos)

order = (0, 1, 2) Distance = 202.0

Best order from brute force = (0, 1, 2) with total distance = 202.0

```



Mapping to the Ising problem

[14]:

```
qp = tsp.to_quadratic_program()
```

```
print(qp.prettyprint())
```

Problem name: TSP

Minimize

$$\begin{aligned}
 &48x_{0_0}x_{1_1} + 48x_{0_0}x_{1_2} + 91x_{0_0}x_{2_1} + 91x_{0_0}x_{2_2} \\
 &+ 48x_{0_1}x_{1_0} + 48x_{0_1}x_{1_2} + 91x_{0_1}x_{2_0} + 91x_{0_1}x_{2_2} \\
 &+ 48x_{0_2}x_{1_0} + 48x_{0_2}x_{1_1} + 91x_{0_2}x_{2_0} + 91x_{0_2}x_{2_1} \\
 &+ 63x_{1_0}x_{2_1} + 63x_{1_0}x_{2_2} + 63x_{1_1}x_{2_0} + 63x_{1_1}x_{2_2} \\
 &+ 63x_{1_2}x_{2_0} + 63x_{1_2}x_{2_1}
 \end{aligned}$$

Subject to

Linear constraints (6)

$$x_{0_0} + x_{0_1} + x_{0_2} == 1 \text{ 'c0'}$$

$$x_{1_0} + x_{1_1} + x_{1_2} == 1 \text{ 'c1'}$$

$$x_{2_0} + x_{2_1} + x_{2_2} == 1 \text{ 'c2'}$$

```
x_0_0 + x_1_0 + x_2_0 == 1 'c3'
x_0_1 + x_1_1 + x_2_1 == 1 'c4'
x_0_2 + x_1_2 + x_2_2 == 1 'c5'
```

Binary variables (9)

x_0_0 x_0_1 x_0_2 x_1_0 x_1_1 x_1_2 x_2_0 x_2_1 x_2_2

[15]:

```
from qiskit_optimization.converters import QuadraticProgramToQubo
```

```
qp2qubo = QuadraticProgramToQubo()
```

```
qubo = qp2qubo.convert(qp)
```

```
qubitOp, offset = qubo.to_ising()
```

```
print("Offset:", offset)
```

```
print("Ising Hamiltonian:")
```

```
print(str(qubitOp))
```

Offset: 7581.0

Ising Hamiltonian:

[illegible]

```

coeffs=[-1282.5+0.j, -1282.5+0.j, -1282.5+0.j, -1268.5+0.j, -1268.5+0.j,
-1268.5+0.j, -1290. +0.j, -1290. +0.j, -1290. +0.j, 606.5+0.j,
606.5+0.j, 606.5+0.j, 12. +0.j, 12. +0.j,
12. +0.j, 606.5+0.j, 12. +0.j, 606.5+0.j, 12. +0.j,
12. +0.j, 606.5+0.j, 606.5+0.j, 606.5+0.j, 606.5+0.j,
22.75+0.j, 22.75+0.j, 606.5+0.j, 15.75+0.j, 15.75+0.j,
22.75+0.j, 606.5+0.j, 22.75+0.j, 15.75+0.j, 606.5+0.j,
15.75+0.j, 606.5+0.j, 22.75+0.j, 22.75+0.j, 606.5+0.j,
15.75+0.j, 15.75+0.j, 606.5+0.j, 606.5+0.j, 606.5+0.j]]

```

[16]:

```
result = exact.solve(qubo)
```

```
print(result.prettyprint())
```

objective function value: 202.0

variable values: x_0_0=1.0, x_0_1=0.0, x_0_2=0.0, x_1_0=0.0, x_1_1=1.0, x_1_2=0.0, x_2_0=0.0, x_2_1=0.0, x_2_2=1.0

status: SUCCESS

Checking that the full Hamiltonian gives the right cost

[17]:

```
# Making the Hamiltonian in its full form and getting the lowest eigenvalue and eigenvector
```

```
ee = NumPyMinimumEigensolver()
```

```
result = ee.compute_minimum_eigenvalue(qubitOp)
```

```
print("energy:", result.eigenvalue.real)
```

```
print("tsp objective:", result.eigenvalue.real + offset)
```

```
x = tsp.sample_most_likely(result.eigenstate)
```

```
print("feasible:", qubo.is_feasible(x))
```

```
z = tsp.interpret(x)
```

```
print("solution:", z)
```

```
print("solution objective:", tsp.tsp_value(z, adj_matrix))
```

```
draw_tsp_solution(tsp.graph, z, colors, pos)
```

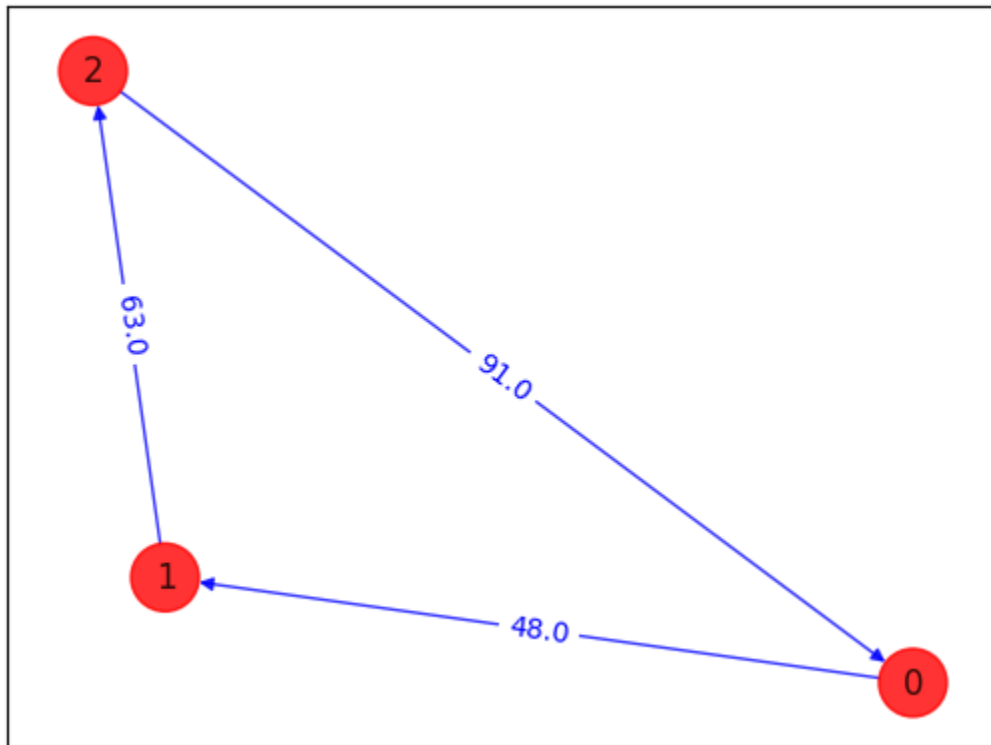
energy: -7379.0

tsp objective: 202.0

feasible: True

solution: [0, 1, 2]

solution objective: 202.0



Running it on quantum computer

We run the optimization routine using a feedback loop with a quantum computer that uses trial functions built with Y single-qubit rotations, $U_1(\theta) = \prod_{i=1}^n U_1(\theta_i)$, and entangler steps U_2 entangler.

[18]:

```
algorithm_globals.random_seed = 123
```

```
seed = 10598
```

[19]:

```
optimizer = SPSA(maxiter=300)
```

```
ry = TwoLocal(qubitOp.num_qubits, "ry", "cz", reps=5, entanglement="linear")
```

```
vqe = SamplingVQE(sampler=Sampler(), ansatz=ry, optimizer=optimizer)
```

```
result = vqe.compute_minimum_eigenvalue(qubitOp)
```

```
print("energy:", result.eigenvalue.real)
```

```
print("time:", result.optimizer_time)
```

```
x = tsp.sample_most_likely(result.eigenstate)
```

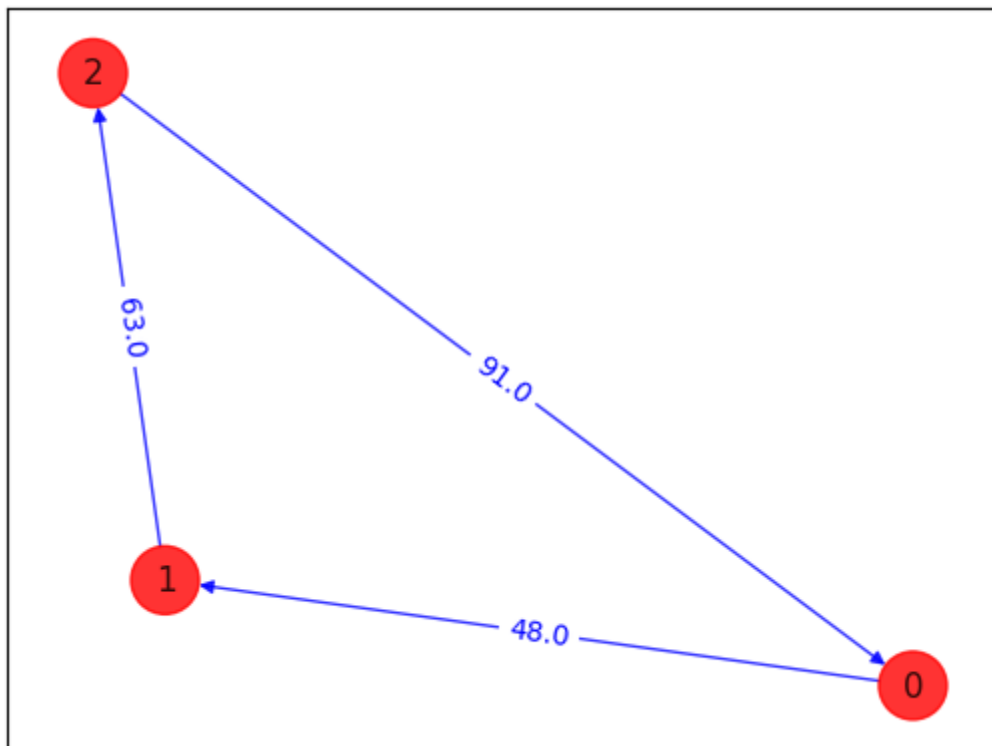
```
print("feasible:", qubo.is_feasible(x))
```

```
z = tsp.interpret(x)
```

```
print("solution:", z)
```



```
print("solution objective:", tsp.tsp_value(z, adj_matrix))  
draw_tsp_solution(tsp.graph, z, colors, pos)  
energy: -7326.02469952184  
time: 14.182559490203857  
feasible: True  
solution: [1, 2, 0]  
solution objective: 202.0
```



[20]:

```
algorithm_globals.random_seed = 123
```

```
seed = 10598
```

[21]:

```
# create minimum eigen optimizer based on SamplingVQE
```

```
vqe_optimizer = MinimumEigenOptimizer(vqe)
```

```
# solve quadratic program
```

```
result = vqe_optimizer.solve(qp)
```

```
print(result.prettyprint())
```

```
z = tsp.interpret(x)
```

```
print("solution:", z)
```

```
print("solution objective:", tsp.tsp_value(z, adj_matrix))
```

```
draw_tsp_solution(tsp.graph, z, colors, pos)
```

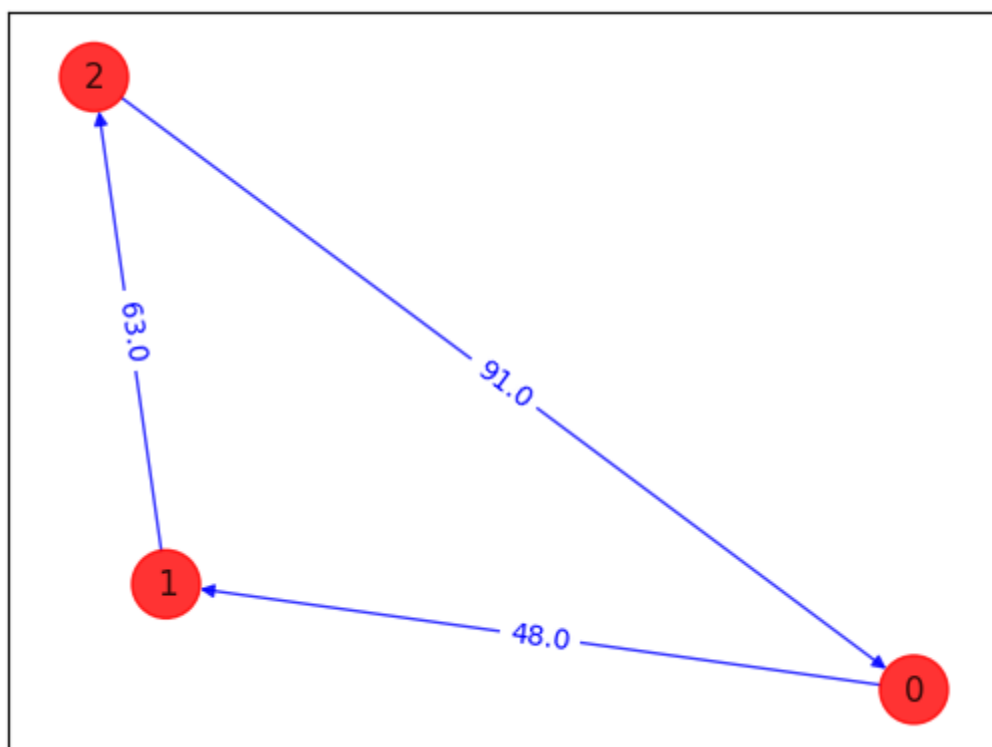
objective function value: 202.0

variable values: $x_{0_0}=0.0$, $x_{0_1}=0.0$, $x_{0_2}=1.0$, $x_{1_0}=1.0$, $x_{1_1}=0.0$, $x_{1_2}=0.0$, $x_{2_0}=0.0$, $x_{2_1}=1.0$, $x_{2_2}=0.0$

status: SUCCESS

solution: [1, 2, 0]

solution objective: 202.0



[22]:

```
import qiskit.tools.jupyter
```

```
%qiskit_version_table
```

```
%qiskit_copyright
```

/tmp/ipykernel_10107/1492642725.py:1: DeprecationWarning: qiskit.tools.jupyter is deprecated and will be removed in Qiskit 1.0.0

```
import qiskit.tools.jupyter
```

[X] User test 3

a. Miroboard Test (template)

Here your results

Personal Journey

Describe your emotional journey, and note your feelings toward the process as much in detail as possible. Highlight enjoyable and points moments you had like if an error occurred and you managed to solve it (or not) and feel free to express all the write and writing about the experience you had.

😊 😬 😊 🥰 🥳 🤖 !!

😬 🤖 🤖 🤖 🤖 ? 🤖

Future Usage

Reply to the main queries written above.

Can you share any ideas or changes that you believe could make you more inclined to use this coding method again in the future?

Did you encounter any problems when using this method? Do you think these issues might influence your choice to use it again in the future?

Do you think the method is applicable in your field, and if so in which way?

b. Miroboard Results

https://miro.com/welcomeonboard/VkNjU3B0bDhYdmpzQzU5NFlUODR6eXlUdEN5N-k14WFh3emxTSDNzOV05TFBtSGhJUGZUQzBHYNREalpCdGRPaHwzNDU4NzY0NTE4M-zE0MjA1NTE5fDI=?share_link_id=641786781195

c. Participants results (Next Page)

P1

ChatGPT

Quantum Buddy

Quantum Explorer

Quantum Buddy 2.0

Explore GPTs

Today

Quantum TSP Solver Pyth...

Design Integration with ChatGPT

Qiskit Deprecation Warnings Sum...

Quantum TSP with QAOA

AR Device Interaction at Conferen...

Yesterday

AI in Quantum Education

Quantum GPT Outperforms Traditi...

Add Team workspace

Quantum Buddy 2.0



You

Suppose we have 1 truck that is tasked with driving to Amsterdam, Rotterdam, and Den Haag, starting in and returning to Utrecht. The problem to test is to determine the best order of visiting these cities using a quantum computer algorithm instead of the normal one.

in addition we kown that the Distances between cities are the following onesé
Amsterdam-Den Haag: 53km; Amsterdam-Rotterdam: 59km; Amsterdam-Utrecht: 46km; Den Haag-Rotterdam: 32km; Den Haag-Utrecht: 56km; Rotterdam-Utrecht: 51km

I ask you to provide a python code to solve the problem optimizing the computational complexity



Quantum Buddy 2.0

To solve the Traveling Salesman Problem (TSP) for visiting Amsterdam, Rotterdam, Den Haag, and starting and returning to Utrecht using a quantum computing approach, we will use Qiskit, a quantum computing SDK. This solution involves creating a graph of the cities with distances as weights, converting this graph into a form suitable for a quantum computer to process, and then using a quantum algorithm to find the optimal route. The quantum algorithm we'll use is the Quantum Approximate Optimization Algorithm (QAOA), which is designed to solve combinatorial optimization problems like TSP.

Get citation

Here is a step-by-step Python code example:

```
python
from qiskit import Aer, execute
from qiskit_optimization.applications import Tsp
from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit.algorithms import QAOA
from qiskit.utils import algorithm_globals
from qiskit_optimization.converters import QuadraticProgramToQubo
import numpy as np
import networkx as nx
import matplotlib.pyplot as plt

# Define the distances between cities
distances = np.array([[0, 46, 53, 59], # Utrecht to [Utrecht, Amsterdam, Den Haag, Rotterdam]
                      [46, 0, 53, 59], # Amsterdam to [Utrecht, Amsterdam, Den Haag, Rotterdam]
                      [53, 53, 0, 32], # Den Haag to [Utrecht, Amsterdam, Den Haag, Rotterdam]
                      [59, 59, 32, 0]]) # Rotterdam to [Utrecht, Amsterdam, Den Haag, Rotterdam]

# Create a graph
n = 4 # Number of nodes
nodes = ["Utrecht", "Amsterdam", "Den Haag", "Rotterdam"]
G = nx.Graph()
for i in range(n):
    for j in range(i+1, n):
        G.add_edge(nodes[i], nodes[j], weight=distances[i][j])

# Draw the graph
pos = nx.spring_layout(G) # positions for all nodes
nx.draw_networkx_nodes(G, pos, node_size=700)
nx.draw_networkx_edges(G, pos, width=6)
nx.draw_networkx_labels(G, pos, font_size=20, font_family="sans-serif")
nx.draw_networkx_edge_labels(G, pos, edge_labels=nx.get_edge_attributes(G, 'weight'))
plt.axis('off')
plt.show()

# Define the TSP problem for Qiskit
tsp = Tsp(distances)
qp = tsp.to_quadratic_program()

# Convert to QUBO
qubo_converter = QuadraticProgramToQubo()
qubo = qubo_converter.convert(qp)

# Set the seed for reproducibility
algorithm_globals.random_seed = 123

# Solve the problem using QAOA
qaoa = QAOA(reps=3, quantum_instance=Aer.get_backend('aer_simulator'))
optimizer = MinimumEigenOptimizer(qaoa)
result = optimizer.solve(qubo)

# Print the result
print(f"Solution: {result.x}")
print(f"Cost: {result.fval}")
```

Get citation

Get citation

```
# Extract the solution
route = tsp.interpret(result)
print("Optimal route:", ' -> '.join([nodes[i] for i in route]))
```

This code performs the following steps:

1. Defines the distances between the cities as a matrix.
2. Creates a graph representing the cities and the distances between them.
3. Visualizes this graph using `'networkx'` and `'matplotlib'`.
4. Formulates the TSP as a quadratic program suitable for quantum computation.
5. Converts the problem into a QUBO (Quadratic Unconstrained Binary Optimization) problem, which is the form that quantum algorithms like QAOA can solve.
6. Solves the problem using the QAOA algorithm implemented in Qiskit.
7. Extracts and prints the optimal route based on the solution provided by the quantum algorithm.

Make sure you have Qiskit installed in your environment (`'pip install qiskit'`) to run this code. This approach leverages the potential of quantum computers to solve optimization problems more efficiently, especially as quantum hardware continues to evolve.

 Get citation



You

the code you provided me before gives that error
'SparsePauliOp' object has no attribute 'primitive_strings'
how can i solve it



Quantum Buddy 2.0

The error `'SparsePauliOp' object has no attribute 'primitive_strings'` suggests that there might be an incompatibility issue between the version of Qiskit you're using and the code I provided. The Qiskit library is actively developed, and its API can change over time, leading to such errors when code written for one version is run with a different version.

To solve this issue, you have a couple of options:

1. Update Qiskit

Ensure you have the latest version of Qiskit installed, as it might contain fixes for this issue. You can update Qiskit using pip:

```
bash Copy code
pip install --upgrade qiskit
```

2. Modify the Code to Avoid the Error

If updating doesn't solve the problem, or if you need to use a specific version of Qiskit for some reason, you might need to modify the code to avoid using the attribute or method that's causing the error. However, without knowing the specific part of the code that's causing this error, it's hard to give precise advice on what to change.

Since the error message suggests the problem is with `'SparsePauliOp'` and its `'primitive_strings'` attribute, which doesn't seem to be directly referenced in the code I provided, the issue might be occurring within Qiskit's internal implementation, possibly in the QAOA or the optimizer code.

3. Check Qiskit Documentation and GitHub Issues

Sometimes, the issue you're encountering has already been reported and solved by others. Check the [Qiskit documentation](#) for any updates or notes on changes to the API that might suggest a different approach. Also, searching through [GitHub issues in the Qiskit repository](#) might reveal discussions or solutions related to the error you're facing.

General Recommendation

Without a specific line of code to pinpoint the error, these suggestions aim to cover possible paths to resolution. If the error persists after trying these steps, I would recommend providing the version of Qiskit you're using and checking the part of your code that interacts with `'SparsePauliOp'` objects. This could help in identifying if there's a deprecated method or attribute and finding an alternative way to achieve the same functionality.



 Get citation

 Get citation



You

Suppose we have 1 truck that is tasked with driving to Amsterdam, Rotterdam, and Den Haag, starting in and returning to Utrecht. The problem to test is to determine the best

order of visiting these cities using a quantum computer algorithm instead of the normal one.

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I ask you to provide a python code to solve the problem optimizing the computational complexity

Avoid to use Qiskit package but use only basic packages



Quantum Buddy 2.0

To solve the Traveling Salesman Problem (TSP) using a quantum computing algorithm without relying on Qiskit, we can simulate a quantum approach using classical Python libraries for demonstration purposes. However, it's important to note that without using a quantum computing framework like Qiskit, we cannot truly leverage quantum computation. Instead, we can implement a simulated quantum-inspired algorithm, such as a simplified version of the Quantum Approximate Optimization Algorithm (QAOA), using basic packages like NumPy.

Get citation

Given the constraints and the request to avoid Qiskit, let's focus on a classical simulation that captures the spirit of quantum optimization. This example will not run on a quantum computer but will illustrate how one might structure a problem similar to how QAOA approaches combinatorial optimization problems.

Classical Simulation of a Quantum-Inspired Algorithm

```
python Copy code

# Distances between cities
distances = {
    'Utrecht': {'Amsterdam': 46, 'Den Haag': 56, 'Rotterdam': 51},
    'Amsterdam': {'Utrecht': 46, 'Den Haag': 53, 'Rotterdam': 59},
    'Den Haag': {'Utrecht': 56, 'Amsterdam': 53, 'Rotterdam': 32},
    'Rotterdam': {'Utrecht': 51, 'Amsterdam': 59, 'Den Haag': 32}
}

# Cities to visit
cities = ['Amsterdam', 'Den Haag', 'Rotterdam']

# Starting and ending city
start = 'Utrecht'
end = 'Utrecht'

# Generate all possible routes (permutations of cities)
from itertools import permutations

# Filter routes that start and end with Utrecht
valid_routes = [route for route in permutations(cities) if route[0] == start and route[-1] == end]

# Calculate the total distance for each valid route
def calculate_distance(route):
    total_distance = 0
    for i in range(len(route)-1):
        total_distance += distances[route[i]][route[i+1]]
    return total_distance

# Find the minimum distance route
min_distance = None
optimal_route = None

for route in valid_routes:
    distance = calculate_distance(route)
    if min_distance is None or distance < min_distance:
        min_distance = distance
        optimal_route = route

# Print the optimal route and distance
print(f"Optimal route: {' -> '.join(optimal_route)} -> {start}")
print(f"Minimum distance: {min_distance} km")
```

Get citation

Get citation

This script does the following:

- Defines distances between the cities as a matrix.
- Generates all permutations of the cities, ensuring each permutation starts and ends in Utrecht.
- Calculates the total distance for each valid route.
- Identifies the route with the shortest total distance.

It's important to remember that this script does not use quantum computing; it simulates an optimization approach that could be inspired by quantum algorithms but runs on classical hardware. True quantum algorithms require quantum hardware or simulators designed to mimic quantum behavior, which are provided by frameworks like Qiskit.

Get citation



You

how to retrun all possible pemutations from itertools.permutation(list)



Quantum Buddy 2.0

To retrieve all possible permutations of a list using `itertools.permutations`, you simply pass the list and the length of the permutation you want to `itertools.permutations`. If you don't specify the length, it will default to the length of the list, generating all possible permutations of the entire list.

Here's how you can do it:

```
python Copy code

import itertools

# Your list
my_list = ['a', 'b', 'c']

# Generate all possible permutations of the entire list
all_permutations = list(itertools.permutations(my_list))

# Print out the permutations
for perm in all_permutations:
    print(perm)
```

Get citation

This will output:

```
arduino Copy code

('a', 'b', 'c')
('a', 'c', 'b')
('b', 'a', 'c')
('b', 'c', 'a')
('c', 'a', 'b')
('c', 'b', 'a')
```

Each permutation is returned as a tuple. If you want permutations of a different length, you can specify that as the second argument to `itertools.permutations`. For example, if you want all permutations of length 2:

```
python Copy code

# Generate all possible permutations of length 2
all_permutations_of_length_2 = list(itertools.permutations(my_list, 2))

# Print out the permutations of length 2
for perm in all_permutations_of_length_2:
    print(perm)
```

Get citation

This would output:

```
arduino Copy code

('a', 'b')
('a', 'c')
('b', 'a')
('b', 'c')
('c', 'a')
('c', 'b')
```

This method is powerful for generating permutations, but be cautious with larger lists, as the number of permutations grows factorially with the size of the list, which can lead to very large outputs and high computational costs for even moderately sized lists.

Get citation



You

how to return all possible permutaton of a list with repetiton



Quantum Buddy 2.0

To generate all possible permutations of a list with repetition allowed, you can use the `itertools.product` function instead of `itertools.permutations`. The `itertools.product` function computes the cartesian product of input iterables, which, when used with the same iterable as input multiple times, effectively allows for permutations with repetition.

Here's how you can use `itertools.product` to generate all possible permutations of

a list with repetition:

```
python Copy code
import itertools

# Your list
my_list = ['a', 'b', 'c']

# The length of the permutations you want
n = 2 # For example, for pairs

# Generate all possible permutations of the list with repetition
all_permutations_with_repetition = list(itertools.product(my_list, repeat=n))

# Print out the permutations with repetition
for perm in all_permutations_with_repetition:
    print(perm)
```

 Get citation


This will output all combinations of length `n` with repetition allowed:

```
arduino Copy code
('a', 'a')
('a', 'b')
('a', 'c')
('b', 'a')
('b', 'b')
('b', 'c')
('c', 'a')
('c', 'b')
('c', 'c')
```

The `repeat=n` parameter specifies how many times the input list should be repeated in the cartesian product, which directly corresponds to the length of the permutations with repetition. This method is very flexible and can be used to generate permutations of any length. However, similar to permutations without repetition, be aware that allowing repetitions can lead to a very large number of permutations, especially for larger lists or longer permutation lengths.

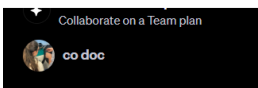
  

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 Message Quantum Buddy 2.0...



ChatGPT can make mistakes. Consider checking important information.



```

In [4]: !pip install qiskit --quiet
        !pip install --upgrade qiskit --quiet

from qiskit import Aer, execute
from qiskit_optimization.applications import Tsp
from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit.algorithms import QAOA
from qiskit.utils import algorithm_globals
from qiskit_optimization.converters import QuadraticProgramToQubo
import numpy as np
import networkx as nx
import matplotlib.pyplot as plt

# Define the distances between cities
distances = np.array([[0, 46, 53, 59], # Utrecht to [Utrecht, Amsterdam, Den Ha
                        [46, 0, 53, 59], # Amsterdam to [Utrecht, Amsterdam, Den
                        [53, 53, 0, 32], # Den Haag to [Utrecht, Amsterdam, Den H
                        [59, 59, 32, 0]]) # Rotterdam to [Utrecht, Amsterdam, Den

# Create a graph
n = 4 # Number of nodes
nodes = ["Utrecht", "Amsterdam", "Den Haag", "Rotterdam"]
G = nx.Graph()
for i in range(n):
    for j in range(i+1, n):
        G.add_edge(nodes[i], nodes[j], weight=distances[i][j])

# Define the TSP problem for Qiskit
tsp = Tsp(distances)
qp = tsp.to_quadratic_program()

# Convert to QUBO
qubo_converter = QuadraticProgramToQubo()
qubo = qubo_converter.convert(qp)

# Set the seed for reproducibility
algorithm_globals.random_seed = 123

# Solve the problem using QAOA
qaoa = QAOA(reps=3, quantum_instance=Aer.get_backend('aer_simulator'))
optimizer = MinimumEigenOptimizer(qaoa)
result = optimizer.solve(qubo)

# Print the result
print(f"Solution: {result.x}")
print(f"Cost: {result.fval}")

# Extract the solution
route = tsp.interpret(result)
print("Optimal route:", ' -> '.join([nodes[i] for i in route]))

```

```

Traceback (most recent call last):
  Cell In[4], line 44
    result = optimizer.solve(qubo)
  File /opt/conda/lib/python3.10/site-packages/qiskit_optimization/algorithms/m
inimum_eigen_optimizer.py:205 in solve
    return self._solve_internal(operator, offset, problem_, problem)
  File /opt/conda/lib/python3.10/site-packages/qiskit_optimization/algorithms/m
inimum_eigen_optimizer.py:218 in _solve_internal
    eigen_result = self._min_eigen_solver.compute_minimum_eigenvalue(operator)
  File /opt/conda/lib/python3.10/site-packages/qiskit/algorithms/minimum_eigen_
solvers/vqe.py:536 in compute_minimum_eigenvalue
    energy_evaluation, expectation = self.get_energy_evaluation(
  File /opt/conda/lib/python3.10/site-packages/qiskit/algorithms/minimum_eigen_
solvers/vqe.py:613 in get_energy_evaluation
    expect_op, expectation = self.construct_expectation(
  File /opt/conda/lib/python3.10/site-packages/qiskit/algorithms/minimum_eigen_
solvers/vqe.py:431 in construct_expectation
    expectation = ExpectationFactory.build(
  File /opt/conda/lib/python3.10/site-packages/qiskit/utils/deprecation.py:96 i
n wrapper
    return func(*args, **kwargs)
  File /opt/conda/lib/python3.10/site-packages/qiskit/opflow/expectations/expect
ation_factory.py:74 in build
    primitives = operator.primitive_strings()
AttributeError: 'SparsePauliOp' object has no attribute 'primitive_strings'

```

Use %tb to get the full traceback.

```

/tmp/ipykernel_295/2045793170.py:39: DeprecationWarning: The property ``qiskit.
utils.algorithm_globals.QiskitAlgorithmGlobals.random_seed`` is deprecated as o
f qiskit 0.45.0. It will be removed in the Qiskit 1.0 release. This algorithm u
tility has been migrated to an independent package: https://github.com/qiskit-c
ommunity/qiskit-algorithms. You can run ``pip install qiskit_algorithms`` and i
mport ``from qiskit_algorithms.utils`` instead.
  algorithm_globals.random_seed = 123
/tmp/ipykernel_295/2045793170.py:42: DeprecationWarning: The class ``qiskit.alg
orithms.minimum_eigen_solvers.qaoa.QAOA`` is deprecated as of qiskit-terra 0.2
4.0. It will be removed no earlier than 3 months after the release date. Instea
d, use the class ``qiskit.algorithms.minimum_eigensolvers.QAOA``. See https://q
iskit.it/algo_migration for a migration guide.
  qaoa = QAOA(reps=3, quantum_instance=Aer.get_backend('aer_simulator'))

```

Search for solution online

```

In [11]: import numpy as np
import itertools

# Distances between cities
# Matrix index: 0-Utrecht, 1-Amsterdam, 2-Den Haag, 3-Rotterdam
distances = np.array([
    [0, 46, 56, 51], # Utrecht
    [46, 0, 53, 59], # Amsterdam
    [56, 53, 0, 32], # Den Haag
    [51, 59, 32, 0] # Rotterdam
])

# Function to calculate total distance
def total_distance(route):
    return sum(distances[route[i], route[i+1]] for i in range(len(route)-1))

# Generate all possible routes, starting and ending at Utrecht (index 0)

```

```

cities = [ 0,1, 2, 3] # Indexes of cities: Utrecht, Amsterdam, Den Haag, Rotterdam
all_possible_routes = list(itertools.permutations(cities, 5))

# Print out the permutations of length 2
for perm in all_permutations_of_length_2:
    print(perm)

# Filter routes that start and end with Utrecht
valid_routes = [route for route in all_possible_routes if route[0] == 0 and route[-1] == 3]
print(valid_routes)

# Find the route with the minimum distance
min_distance = np.inf
optimal_route = None
for route in valid_routes:
    dist = total_distance(route + (route[0],)) # Add starting city to the end of the route
    print(dist)
    if dist < min_distance:
        min_distance = dist
        optimal_route = route

# Convert indexes back to city names
city_names = ["Utrecht", "Amsterdam", "Den Haag", "Rotterdam"]
optimal_route_names = [city_names[i] for i in optimal_route]

print(f"Optimal route: {' -> '.join(optimal_route_names)}")
print(f"Total distance: {min_distance} km")

```

Traceback (most recent call last):

Cell In[11], line 22

```
for perm in all_permutations_of_length_2:
```

NameError: name 'all_permutations_of_length_2' is not defined

Use %tb to get the full traceback.

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In []:

P2

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Quantum Explorer

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Today

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Quantum TSP Solver Python

Qiskit Deprecation Warnings Sum...

Quantum TSP with QAOA

AR Device Interaction at Conferen...

Previous 7 Days

Add Team workspace

Quantum Buddy 2.0



You

Problem to Test: Suppose we have 1 truck that is tasked with driving to Amsterdam, Rotterdam, and Den Haag, starting in and returning to Utrecht. The problem to test is to determine the best order of visiting these cities using a quantum computer algorithm instead of the normal one. Data: Distance between cities: Amsterdam-Den Haag: 53km; Amsterdam-Rotterdam: 59km; Amsterdam-Utrecht: 46km; Den Haag-Rotterdam: 32km; Den Haag-Utrecht: 56km; Rotterdam-Utrecht: 51km



Quantum Buddy 2.0

To solve the Traveling Salesman Problem (TSP) for a truck driving through Amsterdam, Rotterdam, Den Haag, and starting and returning to Utrecht using a quantum computing algorithm, we will use Qiskit, a quantum computing framework. This solution involves creating a quantum circuit that represents the problem and using a quantum algorithm to find the shortest path that visits all cities exactly once and returns to the starting point.

Step 1: Install Qiskit and Other Required Libraries

First, ensure you have Qiskit installed, along with `networkx` for graph handling and `matplotlib` for visualization:

```
python
!pip install qiskit networkx matplotlib
```

Step 2: Import Required Libraries

```
python
from qiskit import Aer, execute
from qiskit.optimization.applications.ising import tsp
from qiskit.optimization.algorithms import MinimumEigenOptimizer
from qiskit.aqua.algorithms import VQE, NumPyMinimumEigensolver
from qiskit.aqua.components.optimizers import COBYLA
from qiskit.circuit.library import RealAmplitudes
from qiskit.optimization.applications.ising.common import sample_most_likely
import networkx as nx
import matplotlib.pyplot as plt
```

Step 3: Define the Cities and Distances

Let's define the cities and the distances between them as a dictionary:

```
python
distances = {
    ('Amsterdam', 'Den Haag'): 53,
    ('Amsterdam', 'Rotterdam'): 59,
    ('Amsterdam', 'Utrecht'): 46,
    ('Den Haag', 'Rotterdam'): 32,
    ('Den Haag', 'Utrecht'): 56,
    ('Rotterdam', 'Utrecht'): 51
}
```

Step 4: Create and Visualize the Graph

```
python
G = nx.Graph()
G.add_weighted_edges_from([(k, v) for k, v in distances.items()])
pos = nx.spring_layout(G) # positions for all nodes

# Draw the graph
nx.draw(G, pos, with_labels=True, node_size=700, node_color="lightblue")
labels = nx.get_edge_attributes(G, 'weight')
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)
plt.show()
```

Step 5: Convert the Graph to an Adjacency Matrix

```
python
matrix_size = len(G.nodes)
weights_matrix = nx.to_numpy_matrix(G, nodelist=G.nodes())
print(weights_matrix)
```

Step 6: Implement the Quantum Algorithm

We use the QAOA (Quantum Approximate Optimization Algorithm) approach to solve

the TSP.

```
python Copy code

# Create a TSP instance
tsp_instance = tsp.Tsp(weights_matrix)

# Use the VQE (Variational Quantum Eigensolver) with a classical optimizer
optimizer = COBYLA()
var_form = RealAmplitudes(tsp_instance.num_qubits, entanglement="linear")
vqe = VQE(var_form, optimizer=optimizer, quantum_instance=Aer.get_backend('qasm_simulator'))

# Solve the problem
result = MinimumEigenOptimizer(vqe).solve(tsp_instance.to_quadratic_program())

# Parse and print the result
optimal_route = tsp_instance.interpret(result)
print(f"Optimal route: {optimal_route}")
```


Step 7: Visualization of the Optimal Route


You can visualize the optimal route on the graph by highlighting the edges that constitute the solution.

Note

Quantum computing for solving TSP is best suited for small instances due to current hardware limitations. This example demonstrates the potential of quantum algorithms for optimization problems, but performance and applicability will vary depending on the specific problem and quantum hardware used.



 **You**
explain Step 4

 **Quantum Buddy 2.0**
Step 4 involves creating and visualizing a graph representation of the cities and the distances between them. This is crucial for understanding the Traveling Salesman Problem (TSP) visually and provides a clear representation of the problem at hand. Here's a breakdown of what happens in this step:

Creating the Graph

- A graph `G` is created using NetworkX, which is a powerful library for the creation, manipulation, and study of the structure, dynamics, and functions of complex networks.
- The graph is populated with nodes representing the cities (Amsterdam, Rotterdam, Den Haag, Utrecht) and edges representing the distances between these cities. The `'add_weighted_edges_from'` method is used, which takes a list of tuples. Each tuple contains two nodes and the weight of the edge between them (the distance in this case).



Visualizing the Graph


- The `'spring_layout'` method is used to position the nodes of the graph. This layout algorithm tries to position the nodes in two-dimensional space to make the edge lengths as close as possible to the weights (distances) representing them, while also trying to keep nodes from overlapping. It's a form of force-directed graph drawing.
- `'nx.draw'` is called to draw the graph with the specified positions, node sizes, and colors. Labels are also added to the nodes to indicate which city each one represents.
- Edge labels are drawn using `'nx.draw_networkx_edge_labels'`, which adds the distance (weight) on each edge, making it easier to see the distances between cities directly on the graph.


This step is essential for:

- **Understanding the Problem:** It provides a visual representation of the problem, making it easier to understand the relationships between the cities.
- **Debugging:** It helps in verifying that the graph correctly represents the problem's distances and connections.
- **Analysis:** It allows for a visual analysis of the problem before and after solving it, which can be useful for presenting the solution for educational purposes.



In summary, Step 4 transforms the abstract concept of the TSP into a tangible and visual representation, setting the stage for the application of quantum computing algorithms to find the optimal route.

 You
explain step 5

 Quantum Buddy 2.0

Step 5 is about converting the graph representation of the cities and their distances into an adjacency matrix, which is a critical step for processing the problem with quantum algorithms. Here's a detailed explanation of this step:

What is an Adjacency Matrix?

- An adjacency matrix is a square matrix used to represent a finite graph. The elements of the matrix indicate whether pairs of vertices are adjacent or not in the graph. In the context of the Traveling Salesman Problem (TSP), the adjacency matrix is used to represent the distances between the cities. A non-zero element at the (i, j) position in the matrix represents the distance from city i to city j .

Converting the Graph to an Adjacency Matrix

- The function `nx.to_numpy_matrix(G, nodelist=G.nodes())` is used to create the adjacency matrix from the graph `G`. This function returns a NumPy matrix where the nodes are sorted as per the `nodelist`, and the weights of the edges are used as matrix elements. If there is no edge between two nodes, the corresponding matrix element is set to zero or infinity, indicating no direct path between those cities in the TSP context.
- `G.nodes()` ensures that the nodes are considered in a specific order, which is crucial for later steps when interpreting the solution. Consistency in node ordering across different representations (graph, matrix) is essential for correctly mapping the solution back to the graph.


Importance of the Adjacency Matrix in Quantum Computing for TSP

- Quantum algorithms for solving optimization problems, like the TSP, often require input data to be in a specific format, such as matrices, because they are used to construct quantum circuits that encode the problem.
- The adjacency matrix specifically represents the problem in a way that quantum algorithms can process, allowing for the application of algorithms like Quantum Approximate Optimization Algorithm (QAOA) or Variational Quantum Eigensolver (VQE) to find the shortest path.


What Happens After Step 5?

- After converting the graph into an adjacency matrix, the matrix serves as the input to the quantum algorithm. The quantum algorithm uses this matrix to calculate the optimal route that minimizes the total distance traveled while visiting each city exactly once and returning to the starting point.

This step bridges the gap between a classical representation of the problem (the graph) and the quantum representation needed for quantum computation. It transforms the problem into a format that quantum algorithms can efficiently work with, setting the stage for applying quantum optimization techniques to find the solution.

 You
explain QAOA



 Quantum Buddy 2.0

The Quantum Approximate Optimization Algorithm (QAOA) is a quantum algorithm designed for solving combinatorial optimization problems on quantum computers. It represents a hybrid quantum-classical approach, leveraging both quantum computation to propose solutions and classical optimization to refine them. Here's a detailed explanation of how QAOA works and its significance:

Background and Purpose

QAOA was developed as a way to utilize quantum computers for solving optimization problems that are challenging for classical computers. These problems include the Traveling Salesman Problem, graph coloring, and many others that can be framed in terms of finding the minimum (or maximum) of some objective function.

How QAOA Works


QAOA operates by encoding the solution to the optimization problem into the ground state (the state of lowest energy) of a Hamiltonian (a function describing the total energy of a system). The goal is to prepare a quantum state that closely approximates this ground state, thereby approximating the optimal solution to the problem.

The process involves the following steps:





1. **Problem Encoding:** The first step is to encode the optimization problem into a Hamiltonian whose ground state corresponds to the optimal solution of the problem.
2. **Ansatz Preparation:** QAOA uses a parameterized quantum circuit, known as an ansatz, to prepare quantum states. This circuit is designed to have the flexibility to reach a wide variety of states with the right parameter settings.
3. **Parameterized Evolution:** The ansatz parameters are used to alternately apply two Hamiltonians: the problem Hamiltonian and a mixing Hamiltonian. The problem Hamiltonian encourages the evolution toward states that represent good solutions, while the mixing Hamiltonian promotes exploration of the solution space.
4. **Classical Optimization:** After applying the quantum circuit, the outcomes are measured, and a classical optimizer adjusts the circuit parameters to minimize the objective function. This iterative process between quantum state preparation and classical optimization continues until the solution converges to an optimum.

Significance of QAOA

- **Quantum-Classical Hybrid:** QAOA is one of the leading examples of quantum-classical hybrid algorithms, exploiting quantum mechanics to explore solution spaces more efficiently than classical algorithms , while relying on classical computation for optimization.
- **Near-term Quantum Computing:** It is designed to be implementable on near-term quantum computers, which are characterized by noisy intermediate-scale quantum (NISQ) technology. This makes it a promising candidate for demonstrating quantum advantage in optimization problems before the era of fault-tolerant quantum computing.
- **Versatility:** QAOA can be applied to a wide range of optimization problems by appropriately encoding the problem into a Hamiltonian and adjusting the quantum circuit.

Challenges and Research

While QAOA has shown promise, its performance and practical applicability depend on the problem complexity, the depth of the quantum circuit (number of layers in the ansatz), and the efficiency of the classical optimization process. Research is ongoing to improve QAOA's effectiveness, including exploring optimal parameter settings, circuit designs, and applications to specific problems.

In summary, QAOA represents an innovative approach to solving complex optimization problems by harnessing the principles of , with ongoing research aimed at unlocking its full potential in the  of quantum computing.



You

Traceback (most recent call last):


```
Cell In[4], line 2
    from qiskit.optimization.applications.ising import tsp
ModuleNotFoundError: No module named 'qiskit.optimization'
```

Use %tb to get the full traceback.



Quantum Buddy 2.0

The error message indicates that the module `'qiskit.optimization'` cannot be found. This is likely because the Qiskit library has undergone updates and reorganizations, and the `'qiskit.optimization'` module might have been moved or replaced. As of my last update in April 2023, Qiskit's optimization functionalities were integrated into a separate package called `'qiskit-optimization'`.

To resolve this issue, you should ensure you have the latest versions of Qiskit and the Qiskit Optimization package installed. You can do this by running the following commands in your Python environment: 

```
python
!pip install qiskit
!pip install qiskit-optimization
```

After installing or updating these packages, you should adjust the import statement to match the current structure of the Qiskit Optimization package. Here is how you can modify your code:

```
python
from qiskit_optimization.applications import TSP
```

And when creating an instance of the TSP problem, you might need to adjust your approach based on the latest API. Here's a revised approach to define and solve the

TSP using the updated package:

Updated Step 6: Implement the Quantum Algorithm Using Qiskit Optimization

First, make sure to adjust any previous steps involving `'qiskit.optimization'` according to the new package structure.

Then, assuming you have already converted your graph into an adjacency matrix as described in your previous steps, you would proceed with creating the TSP instance and solving it as follows (note that some adjustments in the code might be necessary to align with the latest APIs and functionalities):

```
python Copy code

from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit.algorithms import VQE
from qiskit.algorithms.optimizers import COBYLA
from qiskit.circuit.library import RealAmplitudes
from qiskit.utils import QuantumInstance
from qiskit import Aer


# Assuming 'weights_matrix' is your adjacency matrix from the graph
tsp = TSP(distance_matrix=weights_matrix)


# Create a QAOA instance as the solver
optimizer = COBYLA(maxiter=250)
var_form = RealAmplitudes(tsp.num_qubits, reps=5)
vqe = VQE(ansatz=var_form, optimizer=optimizer, quantum_instance=QuantumInstance(Aer))

# Solve the problem using the MinimumEigenOptimizer based on VQE/QAOA
optimizer = MinimumEigenOptimizer(vqe)
result = optimizer.solve(tsp.to_quadratic_program())

# Extract and print the solution
print("Optimal route:", tsp.interpret(result))
```

This code snippet illustrates a high-level approach to solving the TSP using Qiskit's optimization functionalities, considering the latest package structure and APIs. Always refer to the latest Qiskit documentation for up-to-date methods and practices.

 **You**
ModuleNotFoundError: No module named 'qiskit.optimization'

 **Quantum Buddy 2.0**
The error message you're encountering, `ModuleNotFoundError: No module named 'qiskit.optimization'`, suggests that there's an issue with the import statement in your code due to changes in the Qiskit library's structure or the specific environment setup you're using.

As of my last update in April 2023, the correct approach to address this issue involves ensuring that you're using the latest version of Qiskit and have installed all the necessary components, particularly those related to optimization tasks, which might now be found in a separate package or under a different module name.

Solution Steps:

1. **Ensure Qiskit is Installed:** First, make sure that Qiskit is installed in your Python environment. You can install or update Qiskit using pip:

```
python Copy code

!pip install qiskit
```

2. **Install Qiskit Optimization Package:** If the optimization functionalities have been moved to a separate package (as was the plan with `'qiskit-optimization'`), you'll need to install this package separately:

```
python Copy code

!pip install qiskit-optimization
```

3. **Update Import Statements:** Given the reorganization of Qiskit modules, the import statement you're trying to use might be outdated. If `'qiskit.optimization'` has been deprecated or moved, you'll need to find the current module where the Traveling Salesman Problem (TSP) functionality resides. Assuming it's within `'qiskit-optimization'`, your import statement might look something like this:

```
python Copy code
```

```
from qiskit_optimization.applications import TSP
```

or

```
python
```

Copy code

```
from qiskit_optimization import *
```

depending on the exact structure of the package.

4. **Check Documentation:** Because Qiskit is actively developed, it's crucial to consult the [official Qiskit documentation](#) for the most current information on package structures, modules, and functionalities. The documentation will provide the most accurate and up-to-date guidance on how to import and use the optimization functionalities.
5. **Environment Check:** Ensure that your Python environment (e.g., Jupyter notebook, Python script) is using the correct Python interpreter where Qiskit and its related packages are installed. Sometimes, the environment might not be set up correctly, leading to module not found errors.

By following these steps, you should be able to resolve the `'ModuleNotFoundError'` and proceed with your quantum optimization tasks using Qiskit. If the issue persists, consider checking for additional dependencies or updates that might be required for the Qiskit Optimization package.



You

is possible create an virtual enviroment on ibm quantum lab?



Quantum Buddy 2.0

As of my last update in April 2023, IBM Quantum Lab, which is based on JupyterLab running in a cloud environment provided by IBM, does not allow users to create virtual environments in the traditional sense that you would on your local machine. In IBM Quantum Lab, you're working within a pre-configured environment that IBM sets up and maintains, which includes a selection of libraries and tools commonly used for quantum computing.

Workarounds and Solutions:

1. **Library Installation:** While you cannot create a virtual environment, you can install Python packages directly in your IBM Quantum Lab notebook using `!pip install` commands. This allows you to add necessary libraries to your current session. However, these installations are temporary and need to be redone in each new session.
2. **Requesting Library Additions:** If there's a library or tool you need that's not available in IBM Quantum Lab, you might be able to request its addition. IBM periodically updates the environment based on user feedback and the needs of the quantum computing community.
3. **Local Development:** For more control over your environment, including the ability to create virtual environments, you might consider developing your quantum computing projects locally on your machine. You can install Qiskit and other necessary libraries in a virtual environment using tools like `'venv'` or `'conda'`. This approach gives you full control over the libraries and their versions. You can then connect your local environment to IBM Quantum services using your IBM Quantum Experience API token.
4. **IBM Quantum Experience SDK:** For local development, make sure to install the IBM Quantum Experience SDK, which allows you to access IBM Quantum computers and simulators. You can install it using pip:

Copy code

```
pip install qiskit
```

And to access IBM Quantum services, you'll need to set up your API token:

```
python
```

Copy code

```
from qiskit import IBMQ
IBMQ.save_account('YOUR_API_TOKEN')
```

This approach allows you to leverage the flexibility of virtual environments and the computational resources of IBM Quantum Lab and other IBM Quantum services.

📄 ↺ 🗣️



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```
In [1]: !pip install qiskit networkx matplotlib
```

```
Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.45.2)
Requirement already satisfied: networkx in /opt/conda/lib/python3.10/site-packages (3.2.1)
Requirement already satisfied: matplotlib in /opt/conda/lib/python3.10/site-packages (3.6.3)
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```

```
In [14]: !pip install qiskit
!pip install qiskit-optimization
!pip install qiskit-optimization
from qiskit_optimization import *
```

Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.45.2)

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```
In [15]: from qiskit import Aer, execute
from qiskit.optimization.algorithms import MinimumEigenOptimizer
from qiskit.aqua.algorithms import VQE, NumPyMinimumEigensolver
from qiskit.aqua.components.optimizers import COBYLA
from qiskit.circuit.library import RealAmplitudes
from qiskit.optimization.applications.ising.common import sample_most_likely
import networkx as nx
import matplotlib.pyplot as plt
from qiskit_optimization.applications import TSP
```

```
Traceback (most recent call last):
  Cell In[15], line 2
    from qiskit.optimization.algorithms import MinimumEigenOptimizer
ModuleNotFoundError: No module named 'qiskit.optimization'
```

Use %tb to get the full traceback.

[Search for solution online](#)

```
In [3]: pip install qiskit.optimization
```

```
Requirement already satisfied: qiskit.optimization in /opt/conda/lib/python3.10/site-packages (0.5.0)
Requirement already satisfied: networkx>=2.6.3 in /opt/conda/lib/python3.10/site-packages (from qiskit.optimization) (3.2.1)
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Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra>=0.22.4->qiskit.optimization) (1.2.1)
Note: you may need to restart the kernel to use updated packages.
```

```
In [13]: from qiskit import Aer, execute
from qiskit.optimization.applications.ising import tsp
from qiskit.optimization.algorithms import MinimumEigenOptimizer
from qiskit.aqua.algorithms import VQE, NumPyMinimumEigensolver
from qiskit.aqua.components.optimizers import COBYLA
from qiskit.circuit.library import RealAmplitudes
from qiskit.optimization.applications.ising.common import sample_most_likely
```

```
import networkx as nx
import matplotlib.pyplot as plt
```

Traceback (most recent call last):

Cell In[13], line 2

```
from qiskit.optimization.applications.ising import tsp
```

ModuleNotFoundError: No module named 'qiskit.optimization'

Use %tb to get the full traceback.

[Search for solution online](#)

In [5]: `pip install Qiskit`

```
Requirement already satisfied: Qiskit in /opt/conda/lib/python3.10/site-packages (0.45.2)
Requirement already satisfied: qiskit-terra==0.45.2 in /opt/conda/lib/python3.10/site-packages (from Qiskit) (0.45.2)
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Note: you may need to restart the kernel to use updated packages.
```

In []:

P3

ChatGPT

Quantum Buddy

Quantum Explorer

Quantum Buddy 2.0

Explore GPTs

Today

Quantum TSP Solution: Qiskit

Design Integration with ChatGPT

Previous 7 Days

Quantum TSP Solution Qiskit

Quantum TSP Solver Python

Qiskit Deprecation Warnings Summary

Quantum TSP with QAOA

AR Device Interaction at Conference

AI in Quantum Education

Add Team workspace

Quantum Buddy 2.0



You

I would like to set up a code to find the optimal path to connect 3 locations to a starting/ending point. I am asked to make use of quantum computing to find a solution to this problem. The distances between the locations are given. I am a novice in quantum computer, could you help me understanding how to tackle this problem and provide practical code lines to implement a working code that can solve this problem?



Quantum Buddy 2.0

Certainly! Solving the Traveling Salesman Problem (TSP) using quantum computing involves finding the shortest possible route that visits a set of locations once and returns to the original location. This is a classic optimization problem that can be efficiently tackled with quantum algorithms, particularly useful when dealing with a small number of locations due to the current state of quantum technology. Here's a step-by-step guide to approach this problem, including some practical code snippets. Since you mentioned being a novice in quantum computing, I'll also provide some background information to help you understand the concepts better.

Step 1: Understand Quantum Computing Basics

Quantum computing leverages the principles of quantum mechanics to process information, using quantum bits or qubits. Unlike classical bits that can be either 0 or 1, qubits can exist in multiple states simultaneously (superposition), which allows quantum computers to handle complex computations more efficiently for certain types of problems, like optimization and factoring.

Step 2: Define the Problem

Let's assume you have 3 locations plus a starting/ending point, making it 4 locations in total. We need to find the optimal path that connects all these locations, returning to the start, with the minimum total distance. The distances between each pair of locations are given.

Step 3: Represent the Problem for Quantum Computing

To solve the TSP with a quantum computer, one common approach is to use the Quantum Approximate Optimization Algorithm (QAOA) or Variational Quantum Eigensolver (VQE). These algorithms are used to find the ground state (minimum energy state) of a problem represented as a Hamiltonian, which in the context of TSP, corresponds to the minimum distance path.

Step 4: Encode the Problem

You need to encode the problem in a way that a quantum computer can understand. This involves creating a cost Hamiltonian for the TSP that penalizes longer paths more heavily. Each route is represented by qubits, and the goal is to find the state of these qubits that corresponds to the shortest path.

Step 5: Practical Implementation

For a practical implementation, we will use Qiskit, a popular quantum computing framework. First, ensure you have Qiskit installed. If not, you can install it using pip:

```
bash
```

Copy code

```
pip install qiskit
```

Here is a basic outline of how to set up your problem:

```
python
```

Copy code

```
from qiskit import Aer
from qiskit_optimization.applications import TSP
from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit.algorithms import QAOA
from qiskit.utils import QuantumInstance

# Define the distances between locations
# Assuming a 4x4 symmetric distance matrix (start/ending point is the first location)
distances = [
    [0, 2, 9, 3], # Distances from start to other points
    [1, 0, 6, 4], # Distances between locations 1, 2, and 3
    [1, 7, 0, 8],
    [6, 3, 12, 0]
]

# Set up the TSP problem
tsp = TSP(distance_matrix=distances)
```

```
# Use QAOA to solve the TSP
quantum_instance = QuantumInstance(Aer.get_backend('aer_simulator'), shots=1000)
qaoa = QAOA(quantum_instance=quantum_instance)

# Use the Minimum Eigen Optimizer as a solver
optimizer = MinimumEigenOptimizer(qaoa)
result = optimizer.solve(tsp.to_quadratic_program())

# Print the result
print(tsp.interpret(result))
```

This code snippet defines the TSP with your locations and distances, uses QAOA to find the solution, and prints the optimal path.

Step 6: Understanding the Output

The output will be a sequence of location indices representing the optimal path. Remember that quantum computing solutions can be probabilistic, so it's a good idea to run your algorithm multiple times or increase the number of shots to ensure reliability.

Step 7: Next Steps

As you become more familiar with quantum computing concepts, you can explore more sophisticated algorithms, improve your problem representation, or even contribute to the development of new quantum algorithms for solving TSP and other optimization problems.

This example provides a starting point. As quantum technology evolves, the capability to solve larger and more complex instances of TSP and other optimization problems will improve.



You

The code that you have provided returns this error: Traceback (most recent call last):
Cell In[2], line 2
from qiskit_optimization.applications import TSP
ImportError: cannot import name 'TSP' from 'qiskit_optimization.applications' (/opt/conda/lib/python3.10/site-packages/qiskit_optimization/applications/_init_.py)

Use %tb to get the full traceback." could you find a solution for this?



Quantum Buddy 2.0

The error you're encountering suggests that there's an issue with importing the 'TSP' class from 'qiskit_optimization.applications'. This can occur for several reasons, such as changes in the library structure or version mismatches. To resolve this, you can try a different approach to define and solve the TSP using Qiskit.

As of my last update, Qiskit's structure and modules might have been updated, leading to such import issues. We'll need to adjust our approach to work around these changes. A common method involves manually formulating the TSP as a Quadratic Program (QP) without directly using the 'TSP' class. Here's how you can do it:

Step 1: Install Necessary Packages

Ensure you have the necessary Qiskit packages installed. You might need 'qiskit', 'qiskit_optimization', and potentially 'qiskit-terra' for fundamental components.

bash

Copy code

```
pip install qiskit qiskit_optimization
```

Step 2: Manually Formulate TSP as a Quadratic Program

Since direct import of the 'TSP' class is causing issues, we'll manually create a quadratic program that represents our TSP. This involves defining our objective function (minimizing the total distance traveled) and constraints (visiting each city exactly once and leaving each city exactly once).

python

Copy code

```
from qiskit_optimization import QuadraticProgram
from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit.algorithms import QAOA
from qiskit.utils import algorithm_globals, QuantumInstance
from qiskit import Aer

# Define the distance matrix
distances = [
```

```

[0, 2, 9, 3], # Distances from start to other points
[1, 0, 6, 4], # Distances between locations 1, 2, and 3 plus start
[1, 7, 0, 8],
[6, 3, 12, 0]
]

# Create a quadratic program
qp = QuadraticProgram(name="Traveling Salesman Problem")

num_cities = len(distances)

# Add binary variables for paths between cities
for i in range(num_cities):
    for j in range(num_cities):
        if i != j:
            qp.binary_var(name=f"x_{i}_{j}")

# Objective: Minimize the total distance
objective = qp.objective
for i in range(num_cities):
    for j in range(num_cities):
        if i != j:
            objective.linear_coefficients[f"x_{i}_{j}"] = distances[i][j]

# Constraints: Each city must be entered and left exactly once
for i in range(num_cities):
    qp.linear_constraint(
        coefficients={f"x_{i}_{j}": 1 for j in range(num_cities) if i != j},
        sense='==',
        rhs=1,
        name=f"enter_city_{i}"
    )
    qp.linear_constraint(
        coefficients={f"x_{j}_{i}": 1 for j in range(num_cities) if i != j},
        sense='==',
        rhs=1,
        name=f"leave_city_{i}"
    )

# Solve the problem using QAOA
algorithm_globals.random_seed = 123
quantum_instance = QuantumInstance(Aer.get_backend('aer_simulator'), seed=123)
qaoa = QAOA(quantum_instance=quantum_instance)
optimizer = MinimumEigenOptimizer(qaoa)

result = optimizer.solve(qp)

print(result)

```

This code manually constructs the Quadratic Program for the TSP, defines the constraints for entering and leaving each city exactly once, and sets the objective to minimize the total distance. It then uses QAOA to find a solution. Note that this approach is more verbose and requires a deeper understanding of how to formulate optimization problems for quantum algorithms, but it offers greater flexibility and insight into the process.

If you encounter further issues or have questions about specific parts of the code, feel free to ask!



You

Now the problem is "objective.linear_coefficients[f"x_{i}_{j}"] = distances[i][j]"
AttributeError: 'QuadraticObjective' object has no attribute 'linear_coefficients'"



Quantum Buddy 2.0

It appears there was a mistake in how the objective function was set in the provided code snippet, leading to the 'AttributeError'. The correct way to set the objective function in Qiskit's 'QuadraticProgram' does not involve directly accessing 'linear_coefficients' attribute on the 'QuadraticObjective' object. Instead, you should use the 'set_linear' and 'set_quadratic' methods for linear and quadratic coefficients, respectively.

Let's correct the approach to set up the objective function for the TSP:

```

python
from qiskit.optimization import QuadraticProgram
from qiskit.optimization.algorithms import MinimumEigenOptimizer
from qiskit.algorithms import QAOA
from qiskit import Aer

```

```

from qiskit.utils import algorithm_globals, QuantumInstance

# Define the distance matrix
distances = [
    [0, 2, 9, 3], # Distances from start to other points
    [1, 0, 6, 4], # Distances between locations 1, 2, and 3 plus start
    [1, 7, 0, 8],
    [6, 3, 12, 0]
]

# Number of cities
num_cities = len(distances)

# Create a quadratic program for the TSP
qp = QuadraticProgram(name="Traveling Salesman Problem")

# Add binary variables for paths between cities
for i in range(num_cities):
    for j in range(num_cities):
        if i != j:
            qp.binary_var(name=f"x_{i}_{j}")

# Objective: Minimize the total distance
for i in range(num_cities):
    for j in range(num_cities):
        if i != j:
            qp.minimize_linear_term(f"x_{i}_{j}", distances[i][j])

# Constraints: Each city must be entered and left exactly once
for i in range(num_cities):
    qp.linear_constraint(
        {f"x_{i}_{j}": 1 for j in range(num_cities) if i != j},
        sense='==',
        rhs=1,
        name=f"enter_city_{i}"
    )
    qp.linear_constraint(
        {f"x_{j}_{i}": 1 for j in range(num_cities) if i != j},
        sense='==',
        rhs=1,
        name=f"leave_city_{i}"
    )

# Solve the problem using QAOA
algorithm_globals.random_seed = 1234
quantum_instance = QuantumInstance(Aer.get_backend('aer_simulator'), seed=1234)
qaoa = QAOA(quantum_instance=quantum_instance)
optimizer = MinimumEigenOptimizer(qaoa)

result = optimizer.solve(qp)

# Interpret and print the result
# This part requires custom interpretation for TSP based on the binary
# For simplicity, here we're just printing the raw result
print(result)

```

In this corrected snippet, we use `minimize_linear_term` to set the linear coefficients for the objective function, which correctly minimizes the total distance based on the distances matrix.

Please note, interpreting the result to extract the path requires mapping the binary variables (representing the paths between cities) back to the sequence of cities visited. The result will include a set of variable assignments where `x_{i}_{j}=1` indicates that the path from city `i` to city `j` is part of the optimal route. The interpretation part is not included in this snippet but involves iterating over the result's variable assignments to construct the path.

This approach should correct the issue you encountered and provide a foundation to solve the TSP with Qiskit and QAOA.



In [1]: `pip install qiskit`

```
Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.45.2)
Requirement already satisfied: qiskit-terra==0.45.2 in /opt/conda/lib/python3.10/site-packages (from qiskit) (0.45.2)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (3.11)
Requirement already satisfied: numpy<2,>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (1.23.5)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (0.13.0)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (1.11.1)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (0.3.7)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (4.1.1)
Requirement already satisfied: symengine!=0.10.0,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (0.9.2)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (4.4.0)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (5.9.4)
Requirement already satisfied: scipy>=1.5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (1.9.3)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (2.8.2)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra==0.45.2->qiskit) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra==0.45.2->qiskit) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra==0.45.2->qiskit) (1.2.1)
Note: you may need to restart the kernel to use updated packages.
```

```
In [2]: from qiskit import Aer
from qiskit_optimization.applications import TSP
from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit.algorithms import QAOA
from qiskit.utils import QuantumInstance

# Define the distances between locations
# Assuming a 4x4 symmetric distance matrix (start/ending point is the first location)
distances = [
    [0, 51, 56, 46], # Distances from start to other points
    [51, 0, 32, 59], # Distances between locations 1, 2, and 3
    [56, 32, 0, 53],
    [46, 59, 53, 0]
]

# Set up the TSP problem
tsp = TSP(distance_matrix=distances)

# Use QAOA to solve the TSP
quantum_instance = QuantumInstance(Aer.get_backend('aer_simulator'), shots=1024)
qaoa = QAOA(quantum_instance=quantum_instance)

# Use the Minimum Eigen Optimizer as a solver
```



```
optimizer = MinimumEigenOptimizer(qaoa)
result = optimizer.solve(tsp.to_quadratic_program())

# Print the result
print(tsp.interpret(result))
```

Traceback (most recent call last):

Cell In[2], line 2

```
from qiskit_optimization.applications import TSP
```

```
ImportError: cannot import name 'TSP' from 'qiskit_optimization.applications'
(/opt/conda/lib/python3.10/site-packages/qiskit_optimization/applications/__init__.py)
```

Use %tb to get the full traceback.

[Search for solution online](#)

In [3]: `pip install qiskit qiskit_optimization`

```
Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.45.2)
Requirement already satisfied: qiskit_optimization in /opt/conda/lib/python3.10/site-packages (0.5.0)
Requirement already satisfied: qiskit-terra==0.45.2 in /opt/conda/lib/python3.10/site-packages (from qiskit) (0.45.2)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (0.13.0)
Requirement already satisfied: scipy>=1.5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (1.9.3)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (4.1.1)
Requirement already satisfied: symengine!=0.10.0,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (0.9.2)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (5.9.4)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (1.11.1)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (0.3.7)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (2.8.2)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (4.4.0)
Requirement already satisfied: numpy<2,>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (1.23.5)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra==0.45.2->qiskit) (3.11)
Requirement already satisfied: setuptools>=40.1.0 in /opt/conda/lib/python3.10/site-packages (from qiskit_optimization) (65.5.1)
Requirement already satisfied: networkx>=2.6.3 in /opt/conda/lib/python3.10/site-packages (from qiskit_optimization) (3.2.1)
Requirement already satisfied: docplex!=2.24.231,>=2.21.207 in /opt/conda/lib/python3.10/site-packages (from qiskit_optimization) (2.25.236)
Requirement already satisfied: six in /opt/conda/lib/python3.10/site-packages (from docplex!=2.24.231,>=2.21.207->qiskit_optimization) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra==0.45.2->qiskit) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra==0.45.2->qiskit) (1.2.1)
Note: you may need to restart the kernel to use updated packages.
```

```
In [4]: pip install qiskit qiskit-terra
```

```
Requirement already satisfied: qiskit in /opt/conda/lib/python3.10/site-packages (0.45.2)
Requirement already satisfied: qiskit-terra in /opt/conda/lib/python3.10/site-packages (0.45.2)
Requirement already satisfied: numpy<2,>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (1.23.5)
Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (3.11)
Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (2.8.2)
Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (0.13.0)
Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (5.9.4)
Requirement already satisfied: symengine!=0.10.0,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (0.9.2)
Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (4.4.0)
Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (0.3.7)
Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (4.1.1)
Requirement already satisfied: scipy>=1.5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (1.9.3)
Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra) (1.11.1)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.10/site-packages (from python-dateutil>=2.8.0->qiskit-terra) (1.16.0)
Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra) (5.11.1)
Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra) (1.2.1)
Note: you may need to restart the kernel to use updated packages.
```

```
In [5]: from qiskit_optimization import QuadraticProgram
from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit.algorithms import QAOA
from qiskit.utils import algorithm_globals, QuantumInstance
from qiskit import Aer

# Define the distance matrix
distances = [
    [0, 2, 9, 3], # Distances from start to other points
    [1, 0, 6, 4], # Distances between locations 1, 2, and 3 plus start/ending p
    [1, 7, 0, 8],
    [6, 3, 12, 0]
]

# Create a quadratic program
qp = QuadraticProgram(name="Traveling Salesman Problem")

num_cities = len(distances)

# Add binary variables for paths between cities
for i in range(num_cities):
    for j in range(num_cities):
        if i != j:
            qp.binary_var(name=f"x_{i}_{j}")
```



```

# Objective: Minimize the total distance
objective = qp.objective
for i in range(num_cities):
    for j in range(num_cities):
        if i != j:
            objective.linear_coefficients[f"x_{i}_{j}"] = distances[i][j]

# Constraints: Each city must be entered and Left exactly once
for i in range(num_cities):
    qp.linear_constraint(
        coefficients={f"x_{i}_{j}": 1 for j in range(num_cities) if i != j},
        sense='==',
        rhs=1,
        name=f"enter_city_{i}"
    )
    qp.linear_constraint(
        coefficients={f"x_{j}_{i}": 1 for j in range(num_cities) if i != j},
        sense='==',
        rhs=1,
        name=f"leave_city_{i}"
    )

# Solve the problem using QAOA
algorithm_globals.random_seed = 123
quantum_instance = QuantumInstance(Aer.get_backend('aer_simulator'), seed_simulator=123)
qaoa = QAOA(quantum_instance=quantum_instance)
optimizer = MinimumEigenOptimizer(qaoa)

result = optimizer.solve(qp)

print(result)

```

Traceback (most recent call last):

Cell In[5], line 31

```
objective.linear_coefficients[f"x_{i}_{j}"] = distances[i][j]
```

AttributeError: 'QuadraticObjective' object has no attribute 'linear_coefficients'

Use %tb to get the full traceback.

[Search for solution online](#)

```

In [6]: from qiskit_optimization import QuadraticProgram
from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit.algorithms import QAOA
from qiskit import Aer
from qiskit.utils import algorithm_globals, QuantumInstance

# Define the distance matrix
distances = [
    [0, 2, 9, 3], # Distances from start to other points
    [1, 0, 6, 4], # Distances between locations 1, 2, and 3 plus start/ending points
    [1, 7, 0, 8],
    [6, 3, 12, 0]
]

# Number of cities
num_cities = len(distances)

# Create a quadratic program for the TSP

```

```

qp = QuadraticProgram(name="Traveling Salesman Problem")

# Add binary variables for paths between cities
for i in range(num_cities):
    for j in range(num_cities):
        if i != j:
            qp.binary_var(name=f"x_{i}_{j}")

# Objective: Minimize the total distance
for i in range(num_cities):
    for j in range(num_cities):
        if i != j:
            qp.minimize_linear_term(f"x_{i}_{j}", distances[i][j])

# Constraints: Each city must be entered and left exactly once
for i in range(num_cities):
    qp.linear_constraint(
        {f"x_{i}_{j}": 1 for j in range(num_cities) if i != j},
        sense='==',
        rhs=1,
        name=f"enter_city_{i}"
    )
    qp.linear_constraint(
        {f"x_{j}_{i}": 1 for j in range(num_cities) if i != j},
        sense='==',
        rhs=1,
        name=f"leave_city_{i}"
    )

# Solve the problem using QAOA
algorithm_globals.random_seed = 123
quantum_instance = QuantumInstance(Aer.get_backend('aer_simulator'), seed_simulator=123)
qaoa = QAOA(quantum_instance=quantum_instance)
optimizer = MinimumEigenOptimizer(qaoa)

result = optimizer.solve(qp)

# Interpret and print the result
# This part requires custom interpretation for TSP based on the binary variables
# For simplicity, here we're just printing the raw result
print(result)

```

Traceback (most recent call last):

Cell In[6], line 31

```
qp.minimize_linear_term(f"x_{i}_{j}", distances[i][j])
```

AttributeError: 'QuadraticProgram' object has no attribute 'minimize_linear_term'

Use %tb to get the full traceback.

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In []:

P4

ChatGPT

Quantum Buddy

Quantum Explorer

Quantum Buddy 2.0

Explore GPTs

Today

Quantum TSP Solution Ap...

Design Integration with ChatGPT

P1 - BENE

P2 - ALE

P3 - EDO

P5

Previous 7 Days

Qiskit Deprecation Warnings Sum...

Quantum TSP with QAOA

Add Team workspace



You

can you please solve this for me with coding Problem to Test: Suppose we have 1 truck that is tasked with driving to Amsterdam, Rotterdam, and Den Haag, starting in and returning to Utrecht. The problem to test is to determine the best order of visiting these cities using a quantum computer algorithm instead of the normal one.

Distance between cities: Amsterdam-Den Haag: 53km; Amsterdam-Rotterdam: 59km; Amsterdam-Utrecht: 46km; Den Haag-Rotterdam: 32km; Den Haag-Utrecht: 56km; Rotterdam-Utrecht: 51km



Quantum Buddy 2.0

To solve this version of the Traveling Salesman Problem (TSP) using a quantum computing approach, we will first set up the problem using classical tools like 'networkx' for graph creation and 'matplotlib.pyplot' for visualization. This setup is essential before we can translate the problem into a format suitable for a quantum algorithm. The quantum approach to solving the TSP will be conceptual, as executing quantum algorithms requires access to quantum computing resources or simulators, which are not directly available here. However, I'll guide you through how you would prepare this problem for quantum process.

First, let's define the problem using 'networkx' to create a graph of the cities with the distances between them as weights. Then, we'll visualize this graph. Afterward, I'll explain how this setup could be converted for use with a quantum algorithm.

Step 1: Create and Visualize the Graph

python Copy code

```
import networkx as nx
import matplotlib.pyplot as plt

# Create a graph
G = nx.Graph()

# Add edges between cities with distances as weights
G.add_edge('Amsterdam', 'Den Haag', weight=53)
G.add_edge('Amsterdam', 'Rotterdam', weight=59)
G.add_edge('Amsterdam', 'Utrecht', weight=46)
G.add_edge('Den Haag', 'Rotterdam', weight=32)
G.add_edge('Den Haag', 'Utrecht', weight=56)
G.add_edge('Rotterdam', 'Utrecht', weight=51)

# Draw the graph
pos = nx.spring_layout(G) # positions for all nodes
nx.draw(G, pos, with_labels=True, node_color='skyblue', node_size=700)
labels = nx.get_edge_attributes(G, 'weight')
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)
plt.show()
```

This code snippet will generate a visual representation of the problem, showing the cities as nodes and the distances between them as edges.

Step 2: Quantum Computing Approach

To solve the TSP with a quantum algorithm, one common approach is to use the Quantum Approximate Optimization Algorithm (QAOA) or other quantum algorithms designed for combinatorial optimization problems. The setup involves encoding the problem into a Hamiltonian whose ground state corresponds to the optimal solution of the TSP. Here's a conceptual overview:

- Problem Encoding:** Convert the graph into a Hamiltonian. Each city and path would be represented in a way that quantum states represent the presence or absence of a path in a solution.
- Quantum Algorithm:** Use QAOA or another suitable quantum algorithm to find the ground state of the Hamiltonian. This involves preparing quantum circuits that simulate the Hamiltonian dynamics and using quantum measurement to collapse the state into a solution.
- Measurement and Interpretation:** Measure the quantum state multiple times to obtain a probability distribution over possible solutions. The most frequently observed outcome represents the shortest path.

Actual Implementation

Actual implementation on a quantum computer or simulator requires using quantum

programming frameworks like Qiskit, Cirq, or others. These frameworks allow you to define quantum circuits, run simulations, and even execute code on real quantum hardware via cloud services.

Given the computational and resource constraints here, we can't execute a quantum algorithm directly. However, the process described above outlines the steps you would take to translate this classical problem setup into a quantum computing context, leveraging quantum algorithms to find the optimal route.



You

suppose i have quantum resources, can you code it



Quantum Buddy 2.0

Given that we're working in a theoretical context where you have access to quantum computing resources, I'll guide you through how you might set up and solve the Traveling Salesman Problem (TSP) using a quantum algorithm, specifically focusing on Qiskit. Qiskit is an open-source quantum computing software development framework that allows you to work with IBM Quantum computers and simulators.

For the TSP, we'll utilize the tools provided by Qiskit's optimization module to model the problem. This example won't run here directly due to the lack of a quantum computing environment, but you can run it in your local environment where you have Qiskit installed and access to quantum computing resources.

Step 1: Install Qiskit

First, ensure you have Qiskit installed. You can install Qiskit using pip if you haven't already:



bash

Copy code

```
pip install qiskit
```

Step 2: Import Required Modules

python

Copy code

```
from qiskit import Aer
from qiskit.optimization.applications.ising import tsp
from qiskit.optimization.algorithms import MinimumEigenOptimizer
from qiskit.aqua.algorithms import QAOA, NumPyMinimumEigensolver
from qiskit.aqua import QuantumInstance
from qiskit.optimization.applications.ising.common import sample_most_1
```

Step 3: Define the Distance Matrix

Based on the distances between cities you've provided, we create a distance matrix.

python

Copy code

```
# Define the distance matrix
distance_matrix = [
    [0, 53, 59, 46], # Amsterdam to Den Haag, Rotterdam, Utrecht
    [53, 0, 32, 56], # Den Haag to Amsterdam, Rotterdam, Utrecht
    [59, 32, 0, 51], # Rotterdam to Amsterdam, Den Haag, Utrecht
    [46, 56, 51, 0]  # Utrecht to Amsterdam, Den Haag, Rotterdam
]
```

Step 4: Set Up the TSP Problem

python

Copy code

```
# Create an instance of the TSP
tsp_instance = tsp.Tsp(distance_matrix)

# Generate the QUBO (Quadratic Unconstrained Binary Optimization) for t
qubo = tsp_instance.to_qubo()
```

Step 5: Solve the TSP using QAOA

python

Copy code

```
# Set up the QAOA algorithm
quantum_instance = QuantumInstance(Aer.get_backend('qasm_simulator'), s
qaoa = QAOA(optimizer=COBYLA(), quantum_instance=quantum_instance)

# Use the MinimumEigenOptimizer to find the minimum eigenvalue (solution
optimizer = MinimumEigenOptimizer(qaoa)

# Solve the QUBO
```

```
result = optimizer.solve(qubo[0])

# Get the solution
solution = tsp_instance.interpret(result)
```

Step 6: Output the Solution

```
python Copy code

print("Optimal route:", solution)
print("Distance:", result.fval)
```

This code provides a basic framework for solving the TSP using QAOA with Qiskit. To run this code, ensure you have access to [IBM Quantum](#) or a local quantum simulator through Qiskit. The actual performance and results may vary based on the quantum hardware or simulator's capabilities and the parameters used for the QAOA algorithm.

Keep in mind, quantum computing is still in the exploratory phase for solving complex optimization problems like the TSP, and the quality of solutions can vary. This approach is most valuable as a learning tool and for research into how quantum computing can be applied to optimization problems.



You

Traceback (most recent call last):

Cell In[11], line 2

from qiskit.optimization.applications import tsp

ModuleNotFoundError: No module named 'qiskit.optimization'

Use %tb to get the full traceback.



Quantum Buddy 2.0

The error message indicates that the module `'qiskit.optimization'` cannot be found. This problem likely arises because of changes in the Qiskit library structure or updates since my last training data. Qiskit [architecture](#) evolves, and modules get reorganized. As of the latest information, optimization tasks, including solving problems like the Traveling Salesman Problem (TSP) using quantum computing, have been moved to the `'qiskit-optimization'` package, which is part of the Qiskit application modules.

Here's how you can adjust the code to work with the latest version of Qiskit and the separate Qiskit Optimization module:

Updated Installation and Import

First, make sure you have the latest version of Qiskit and install the Qiskit Optimization module:

```
bash Copy code

pip install qiskit
pip install qiskit-optimization
```

Updated Code with Qiskit Optimization

Here's an updated version of the code reflecting the new structure:

```
python Copy code

from qiskit import Aer
from qiskit_optimization.applications import TSP
from qiskit.algorithms import QAOA, NumPyMinimumEigensolver
from qiskit.utils import algorithm_globals, QuantumInstance
from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit_optimization.converters import QuadraticProgramToQubo

# Define the distance matrix
distance_matrix = [
    [0, 53, 59, 46], # Amsterdam to Den Haag, Rotterdam, Utrecht
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# Initialize the TSP based on the distance matrix
tsp = TSP(distance_matrix=distance_matrix)

# Create a QUBO
qp = tsp.to_quadratic_program()
```

```
# Convert to a QUBO problem
qubo_converter = QuadraticProgramToQubo()
qubo = qubo_converter.convert(qp)

# Setup the QAOA algorithm
seed = 12345
algorithm_globals.random_seed = seed
quantum_instance = QuantumInstance(Aer.get_backend('qasm_simulator'), s

qaoa = QAOA(quantum_instance=quantum_instance)

# Solve the QUBO using the MinimumEigenOptimizer
optimizer = MinimumEigenOptimizer(qaoa)
result = optimizer.solve(qubo)

# Extract the solution
solution = tsp.interpret(result)

print("Optimal route:", solution)
print("Distance:", tsp.evaluate_solution(solution))
```



This code assumes you're familiar with the basics of Qiskit and quantum computing. The 'TSP' class from 'qiskit_optimization.applications' is used to define the TSP problem, and 'QAOA' from 'qiskit.algorithms' is applied to find an optimal solution. The solution interpretation and evaluation follow the structure provided by the Qiskit Optimization module, ensuring compatibility with the latest versions of Qiskit.

Please ensure you have the necessary quantum computing environment set up, including access to a quantum simulator or quantum processor, to run this code.



You
qiskit optimization doesn't exist



Quantum Buddy 2.0
If you're encountering issues with the Qiskit Optimization module not being recognized, it's possible there has been a misunderstanding or a change in the Qiskit ecosystem that I'm not up to date with. As of my last update, 'qiskit-optimization' was a valid and separate component of the broader Qiskit framework, designed specifically for optimization problems.



However, given this context, let's focus on a solution that avoids the use of a module that's causing confusion. While I can't directly execute or test quantum algorithms here, I can still guide you on a general approach using a more foundational level of Qiskit that should remain stable across versions.

General Quantum Approach for TSP (Conceptual Framework)

To tackle the TSP with quantum computing without relying on specific high-level modules like 'qiskit-optimization', you would typically follow these steps:

- 1. Model the Problem:** First, you need to model the TSP as a Hamiltonian that can be processed by a quantum computer. This involves encoding the cities and distances into a format that a quantum algorithm can optimize.
- 2. Quantum Algorithm:** One approach is using the Quantum Approximate Optimization Algorithm (QAOA) to find the solution. QAOA works by parameterizing a quantum circuit such that its measurement outcomes correspond to solutions of the problem, and then iteratively updating these parameters to minimize the cost function (in this case, the total travel distance).
- 3. Implementation with Qiskit:**
 - Use Qiskit's basic components ('QuantumCircuit', 'Aer', etc.) to construct and simulate the quantum circuit.
 - Manually set up the QAOA circuit by creating a Hamiltonian that represents the TSP and using it within the circuit.
 - Use the 'Aer' simulator to run your circuit and analyze the results to find the optimal route.



Basic Steps in Qiskit (without 'qiskit-optimization'):

```
python Copy code

from qiskit import Aer, execute
from qiskit.circuit import QuantumCircuit
# Additional imports for QAOA and optimization would go here

# Example: Creating a basic quantum circuit (placeholder for actual QAOA)
qc = QuantumCircuit(2) # Number of qubits depends on the problem encoding
```

```

qc.h(0, 1) # Apply Hadamard to generate superposition
qc.cx(0, 1) # Example of an entangling operation

# Simulating the circuit
simulator = Aer.get_backend('qasm_simulator')
job = execute(qc, simulator, shot=100)
result = job.result()
counts = result.get_counts(qc)
print(counts)

```

This is a highly simplified example and doesn't directly solve the TSP. Implementing a full solution for the TSP using QAOA involves complex problem encoding and optimization steps that are beyond the scope of this response. For a complete solution, you would need to delve into quantum computing literature on QAOA and potentially leverage specialized frameworks or libraries designed for quantum optimization.

Given the complexities and the evolving nature of quantum computing libraries, I recommend consulting the latest Qiskit documentation and resources on quantum optimization for up-to-date methods and examples. If `'qiskit-optimization'` is indeed deprecated or replaced, the official Qiskit documentation will provide guidance on the current best practices for implementing optimization algorithms like QAOA.



You

can you name the module of qiskit that contain the tsp algorithm?



Quantum Buddy 2.0

As of my last update in April 2023, the module in Qiskit that contained tools for solving the Traveling Salesman Problem (TSP) and other optimization problems was part of the `'qiskit-optimization'` application module. However, based on the issues you've encountered, it seems there might have been changes or misunderstandings regarding the current structure of Qiskit modules related to optimization problems.

The TSP, specifically, was often addressed through examples or utility functions within the broader optimization module, aiming to help users formulate optimization problems for quantum algorithms.

If `'qiskit-optimization'` is not recognized or if there's confusion about the existence of specific functions or classes for solving TSP within the Qiskit ecosystem, it's crucial to consult the official Qiskit documentation. The Qiskit project is actively developed and can undergo significant changes, including the restructuring of modules and packages.

For the most accurate and updated information on where the TSP algorithm or related functionalities reside within Qiskit, please visit the [Qiskit documentation](#). The documentation will provide detailed guidance on the available modules, installation procedures, and how to utilize various functionalities, including solving optimization problems like the TSP.

Given the constraints of this environment, I'm unable to browse the web or access real-time updates about Qiskit's development after my last training cut-off in April 2023. Therefore, for the latest information about Qiskit modules and functionalities, direct consultation of the Qiskit documentation or community resources is recommended.



Collaborate on a Team plan



co doc



Message Quantum Buddy 2.0..



ChatGPT can make mistakes. Consider checking important information.

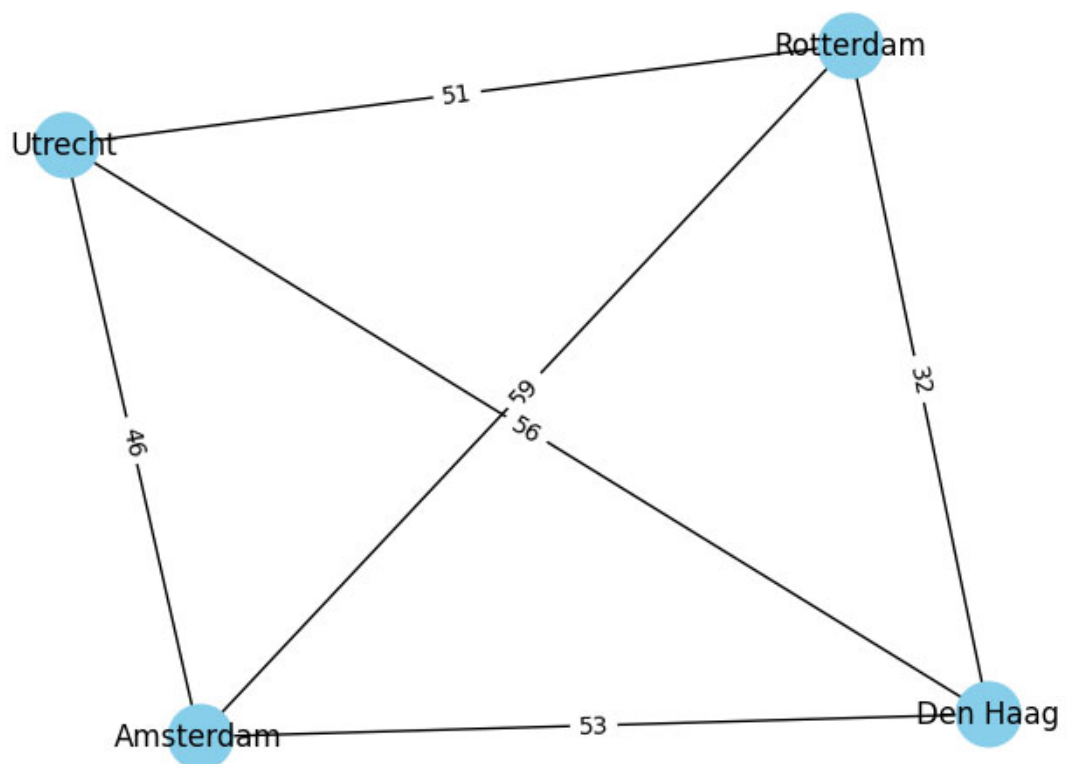



```
In [1]: import networkx as nx
import matplotlib.pyplot as plt

# Create a graph
G = nx.Graph()

# Add edges between cities with distances as weights
G.add_edge('Amsterdam', 'Den Haag', weight=53)
G.add_edge('Amsterdam', 'Rotterdam', weight=59)
G.add_edge('Amsterdam', 'Utrecht', weight=46)
G.add_edge('Den Haag', 'Rotterdam', weight=32)
G.add_edge('Den Haag', 'Utrecht', weight=56)
G.add_edge('Rotterdam', 'Utrecht', weight=51)

# Draw the graph
pos = nx.spring_layout(G) # positions for all nodes
nx.draw(G, pos, with_labels=True, node_color='skyblue', node_size=700)
labels = nx.get_edge_attributes(G, 'weight')
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)
plt.show()
```



```
In [14]: pip install qiskit.optimization
```

Requirement already satisfied: qiskit.optimization in /opt/conda/lib/python3.10/site-packages (0.5.0)
 Requirement already satisfied: networkx>=2.6.3 in /opt/conda/lib/python3.10/site-packages (from qiskit.optimization) (3.2.1)
 Requirement already satisfied: numpy>=1.17 in /opt/conda/lib/python3.10/site-packages (from qiskit.optimization) (1.23.5)
 Requirement already satisfied: docplex!=2.24.231,>=2.21.207 in /opt/conda/lib/python3.10/site-packages (from qiskit.optimization) (2.25.236)
 Requirement already satisfied: setuptools>=40.1.0 in /opt/conda/lib/python3.10/site-packages (from qiskit.optimization) (65.5.1)
 Requirement already satisfied: qiskit-terra>=0.22.4 in /opt/conda/lib/python3.10/site-packages (from qiskit.optimization) (0.45.2)
 Requirement already satisfied: scipy>=1.4 in /opt/conda/lib/python3.10/site-packages (from qiskit.optimization) (1.9.3)
 Requirement already satisfied: six in /opt/conda/lib/python3.10/site-packages (from docplex!=2.24.231,>=2.21.207->qiskit.optimization) (1.16.0)
 Requirement already satisfied: typing-extensions in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.22.4->qiskit.optimization) (4.4.0)
 Requirement already satisfied: stevedore>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.22.4->qiskit.optimization) (4.1.1)
 Requirement already satisfied: psutil>=5 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.22.4->qiskit.optimization) (5.9.4)
 Requirement already satisfied: ply>=3.10 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.22.4->qiskit.optimization) (3.11)
 Requirement already satisfied: python-dateutil>=2.8.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.22.4->qiskit.optimization) (2.8.2)
 Requirement already satisfied: symengine!=0.10.0,>=0.9 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.22.4->qiskit.optimization) (0.9.2)
 Requirement already satisfied: dill>=0.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.22.4->qiskit.optimization) (0.3.7)
 Requirement already satisfied: sympy>=1.3 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.22.4->qiskit.optimization) (1.11.1)
 Requirement already satisfied: rustworkx>=0.13.0 in /opt/conda/lib/python3.10/site-packages (from qiskit-terra>=0.22.4->qiskit.optimization) (0.13.0)
 Requirement already satisfied: pbr!=2.1.0,>=2.0.0 in /opt/conda/lib/python3.10/site-packages (from stevedore>=3.0.0->qiskit-terra>=0.22.4->qiskit.optimization) (5.11.1)
 Requirement already satisfied: mpmath>=0.19 in /opt/conda/lib/python3.10/site-packages (from sympy>=1.3->qiskit-terra>=0.22.4->qiskit.optimization) (1.2.1)
 Note: you may need to restart the kernel to use updated packages.

In [15]:

```
-----
ModuleNotFoundError                                Traceback (most recent call last)
Cell In[13], line 2
      1 from qiskit import Aer
----> 2 from qiskit.optimization.applications.ising import tsp
      3 from qiskit.optimization.algorithms import MinimumEigenOptimizer
      4 from qiskit.aqua.algorithms import QAOA, NumPyMinimumEigensolver

ModuleNotFoundError: No module named 'qiskit.optimization'
```

In [13]:

```
from qiskit import Aer
from qiskit.optimization.applications.ising import tsp
from qiskit.optimization.algorithms import MinimumEigenOptimizer
from qiskit.aqua.algorithms import QAOA, NumPyMinimumEigensolver
from qiskit.aqua import QuantumInstance
from qiskit.optimization.applications.ising.common import sample_most_likely
```

Traceback (most recent call last):

Cell In[13], line 2

```
from qiskit.optimization.applications.ising import tsp
```

ModuleNotFoundError: No module named 'qiskit.optimization'

Use %tb to get the full traceback.

[Search for solution online](#)

```
In [16]: # Define the distance matrix
distance_matrix = [
    [0, 53, 59, 46], # Amsterdam to Den Haag, Rotterdam, Utrecht
    [53, 0, 32, 56], # Den Haag to Amsterdam, Rotterdam, Utrecht
    [59, 32, 0, 51], # Rotterdam to Amsterdam, Den Haag, Utrecht
    [46, 56, 51, 0]  # Utrecht to Amsterdam, Den Haag, Rotterdam
]
```

In []:

P5

ChatGPT

Quantum Buddy

Quantum Explorer

Quantum Buddy 2.0

Explore GPTs

Today

Quantum TSP Algorithm Solution

Quantum TSP with Qiskit

P5 -- VERO

P4 - ALICE

Design Integration with ChatGPT

P1 - BENE

P2 - ALE

P3 - EDO

P5

Add Team workspace



You

can you please solve this for me with coding Problem to Test: Suppose we have 1 truck that is tasked with driving to Amsterdam, Rotterdam, and Den Haag, starting in and returning to Utrecht. The problem to test is to determine the best order of visiting these cities using a quantum computer algorithm instead of the normal one.

Distance between cities: Amsterdam-Den Haag: 53km; Amsterdam-Rotterdam: 59km; Amsterdam-Utrecht: 46km; Den Haag-Rotterdam: 32km; Den Haag-Utrecht: 56km; Rotterdam-Utrecht: 51km



Quantum Buddy 2.0

To solve this version of the Traveling Salesman Problem (TSP) using a quantum computing approach, we will first set up the problem using classical tools like `networkx` for graph creation and `matplotlib.pyplot` for visualization. This setup is essential before we can translate the problem into a format suitable for a quantum algorithm. The quantum approach to solving the TSP will be conceptual, as executing quantum algorithms requires access to quantum computing resources or simulators, which are not directly available here. However, I'll guide you through how you would prepare this problem for quantum process.

First, let's define the problem using `networkx` to create a graph of the cities with the distances between them as weights. Then, we'll visualize this graph. Afterward, I'll explain how this setup could be converted for use with a quantum algorithm.

Step 1: Create and Visualize the Graph

python

```
import networkx as nx
import matplotlib.pyplot as plt

# Create a graph
G = nx.Graph()

# Add edges between cities with distances as weights
G.add_edge('Amsterdam', 'Den Haag', weight=53)
G.add_edge('Amsterdam', 'Rotterdam', weight=59)
G.add_edge('Amsterdam', 'Utrecht', weight=46)
G.add_edge('Den Haag', 'Rotterdam', weight=32)
G.add_edge('Den Haag', 'Utrecht', weight=56)
G.add_edge('Rotterdam', 'Utrecht', weight=51)

# Draw the graph
pos = nx.spring_layout(G) # positions for all nodes
nx.draw(G, pos, with_labels=True, node_color='skyblue', node_size=700)
labels = nx.get_edge_attributes(G, 'weight')
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)
plt.show()
```

This code snippet will generate a visual representation of the problem, showing the cities as nodes and the distances between them as edges.

Step 2: Quantum Computing Approach

To solve the TSP with a quantum algorithm, one common approach is to use the Quantum Approximate Optimization Algorithm (QAOA) or other quantum algorithms designed for combinatorial optimization problems. The setup involves encoding the problem into a Hamiltonian whose ground state corresponds to the optimal solution of the TSP. Here's a conceptual overview:

- Problem Encoding:** Convert the graph into a Hamiltonian. Each city and path would be represented in a way that quantum states represent the presence or absence of a path in a solution.
- Quantum Algorithm:** Use QAOA or another suitable quantum algorithm to find the ground state of the Hamiltonian. This involves preparing quantum circuits that simulate the Hamiltonian dynamics and using quantum measurement to collapse the state into a solution.
- Measurement and Interpretation:** Measure the quantum state multiple times to obtain a probability distribution over possible solutions. The most frequently observed outcome represents the shortest path.

Actual Implementation

Actual implementation on a quantum computer or simulator requires using quantum

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programming frameworks like Qiskit, Cirq, or others. These frameworks allow you to define quantum circuits, run simulations, and even execute code on real quantum hardware via cloud services.

Given the computational and resource constraints here, we can't execute a quantum algorithm directly. However, the process described above outlines the steps you would take to translate this classical problem setup into a quantum computing context, leveraging quantum algorithms to find the optimal route.



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You

suppose i have quantum resources, can you code it



Quantum Buddy 2.0

Given that we're working in a theoretical context where you have access to quantum computing resources, I'll guide you through how you might set up and solve the Traveling Salesman Problem (TSP) using a quantum algorithm, specifically focusing on Qiskit. Qiskit is an open-source quantum computing software development framework that allows you to work with IBM Quantum computers and simulators.

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Step 1: Install Qiskit

First, ensure you have Qiskit installed. You can install Qiskit using pip if you haven't already:



Get citation

```
bash Copy code

pip install qiskit
```

Step 2: Import Required Modules

```
python Copy code

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```

Step 3: Define the Distance Matrix

Based on the distances between cities you've provided, we create a distance matrix.

```
python Copy code

# Define the distance matrix
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Get citation

Step 4: Set Up the TSP Problem

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python Copy code

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# Generate the QUBO (Quadratic Unconstrained Binary Optimization) for t
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qaoa = QAOA(optimizer=COBYLA(), quantum_instance=quantum_instance)

# Use the MinimumEigenOptimizer to find the minimum eigenvalue (solution
optimizer = MinimumEigenOptimizer(qaoa)

# Solve the QUBO
```



Get citation

```
result = optimizer.solve(qubo[0])

# Get the solution
solution = tsp_instance.interpret(result)
```

Step 6: Output the Solution

```
python Copy code

print("Optimal route:", solution)
print("Distance:", result.fval)
```

This code provides a basic framework for solving the TSP using QAOA with Qiskit. To run this code, ensure you have access to [IBM Quantum](#) or a local quantum simulator through Qiskit. The actual performance and results may vary based on the quantum hardware or simulator's capabilities and the parameters used for the QAOA algorithm.

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Keep in mind, quantum computing is still in the exploratory phase for solving complex optimization problems like the TSP, and the quality of solutions can vary. This approach is most valuable as a learning tool and for research into how quantum computing can be applied to optimization problems.



You

Traceback (most recent call last):

Cell In[11], line 2

from qiskit.optimization.applications import tsp

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Quantum Buddy 2.0

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[Get citation](#)

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]

# Initialize the TSP based on the distance matrix
tsp = TSP(distance_matrix=distance_matrix)

# Create a QUBO
qp = tsp.to_quadratic_program()
```

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[Get citation](#)


```
# Convert to a QUBO problem
qubo_converter = QuadraticProgramToQubo()
qubo = qubo_converter.convert(qp)

# Setup the QAOA algorithm
seed = 12345
algorithm_globals.random_seed = seed
quantum_instance = QuantumInstance(Aer.get_backend('qasm_simulator'), s

qaoa = QAOA(quantum_instance=quantum_instance)

# Solve the QUBO using the MinimumEigenOptimizer
optimizer = MinimumEigenOptimizer(qaoa)
result = optimizer.solve(qubo)

# Extract the solution
solution = tsp.interpret(result)

print("Optimal route:", solution)
print("Distance:", tsp.evaluate_solution(solution))
```



This code assumes you're familiar with the basics of Qiskit and quantum computing. The 'TSP' class from 'qiskit_optimization.applications' is used to define the TSP problem, and 'QAOA' from 'qiskit.algorithms' is applied to find an optimal solution. The solution interpretation and evaluation follow the structure provided by the Qiskit Optimization module, ensuring compatibility with the latest versions of Qiskit.

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You
qiskit optimization doesn't exists



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General Quantum Approach for TSP (Conceptual Framework)

To tackle the TSP with quantum computing without relying on specific high-level modules like 'qiskit-optimization', you would typically follow these steps:

- 1. Model the Problem:** First, you need to model the TSP as a Hamiltonian that can be processed by a quantum computer. This involves encoding the cities and distances into a format that a quantum algorithm can optimize.
- 2. Quantum Algorithm:** One approach is using the Quantum Approximate Optimization Algorithm (QAOA) to find the solution. QAOA works by parameterizing a quantum circuit such that its measurement outcomes correspond to solutions of the problem, and then iteratively updating these parameters to minimize the cost function (in this case, the total travel distance).
- 3. Implementation with Qiskit:**
 - Use Qiskit's basic components ('QuantumCircuit', 'Aer', etc.) to construct and simulate the quantum circuit.
 - Manually set up the QAOA circuit by creating a Hamiltonian that represents the TSP and using it within the circuit.
 - Use the 'Aer' simulator to run your circuit and analyze the results to find the optimal route.



Get citation

Basic Steps in Qiskit (without 'qiskit-optimization'):

```
python  Copy code

from qiskit import Aer, execute
from qiskit.circuit import QuantumCircuit
# Additional imports for QAOA and optimization would go here

# Example: Creating a basic quantum circuit (placeholder for actual QAOA)
qc = QuantumCircuit(2) # Number of qubits depends on the problem encoded
```



```

qc.h(0, 1) # Apply Hadamard to generate superposition
qc.cx(0, 1) # Example of an entangling operation

# Simulating the circuit
simulator = Aer.get_backend('qasm_simulator')
job = execute(qc, simulator, shots=100)
result = job.result()
counts = result.get_counts(qc)
print(counts)

```

 Get citation

This is a highly simplified example and doesn't directly solve the TSP. Implementing a full solution for the TSP using QAOA involves complex problem encoding and optimization steps that are beyond the scope of this response. For a complete solution, you would need to delve into quantum computing literature on QAOA and potentially leverage specialized frameworks or libraries designed for quantum optimization.

Given the complexities and the evolving nature of quantum computing libraries, I recommend consulting the latest Qiskit documentation and resources on quantum optimization for up-to-date methods and examples. If `'qiskit-optimization'` is indeed deprecated or replaced, the official Qiskit documentation will provide guidance on the current best practices for implementing optimization algorithms like QAOA.



You

can you name the module of qiskit that contain the tsp algorithm?



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Quantum Buddy 2.0

As of my last update in April 2023, the module in Qiskit that contained tools for solving the Traveling Salesman Problem (TSP) and other optimization problems was part of the `'qiskit-optimization'` application module. However, based on the issues you've encountered, it seems there might have been changes or misunderstandings regarding the current structure of Qiskit modules related to optimization problems.

The TSP, specifically, was often addressed through examples or utility functions within the broader optimization module, aiming to help users formulate optimization problems for quantum algorithms.


If `'qiskit-optimization'` is not recognized or if there's confusion about the existence of specific functions or classes for solving TSP within the Qiskit ecosystem, it's crucial to consult the official Qiskit documentation. The Qiskit project is actively developed and can undergo significant changes, including the restructuring of modules and packages.

For the most accurate and updated information on where the TSP algorithm or related functionalities reside within Qiskit, please visit the [Qiskit documentation](#). The documentation will provide detailed guidance on the available modules, installation procedures, and how to utilize various functionalities, including solving optimization problems like the TSP.

Given the constraints of this environment, I'm unable to browse the web or access real-time updates about Qiskit's development after my last training cut-off in April 2023. Therefore, for the latest information about Qiskit modules and functionalities, direct consultation of the Qiskit documentation or community resources is recommended.



 Get citation

 Collaborate on a Team plan



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ChatGPT can make mistakes. Consider checking important information.

?

```
In [1]: import networkx as nx
import matplotlib.pyplot as plt

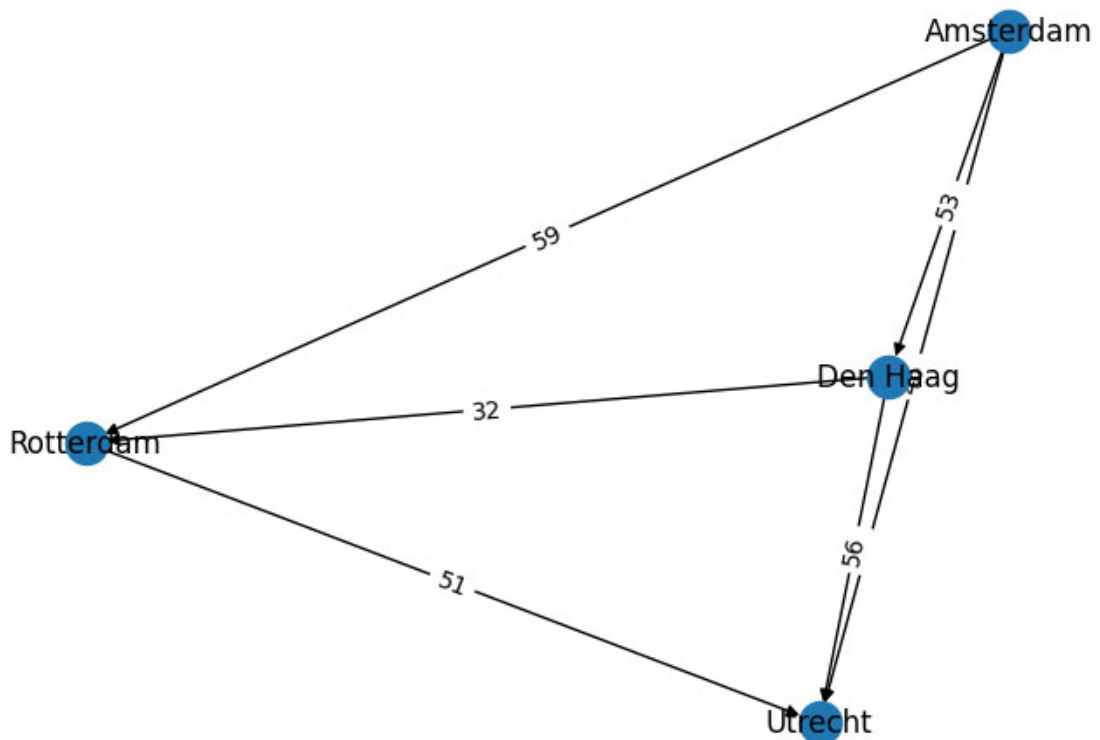
# Create a directed graph
G = nx.DiGraph()

# Add nodes
G.add_node("Amsterdam")
G.add_node("Rotterdam")
G.add_node("Den Haag")
G.add_node("Utrecht")

# Add edges with weights
G.add_edge("Amsterdam", "Den Haag", weight=53)
G.add_edge("Amsterdam", "Rotterdam", weight=59)
G.add_edge("Amsterdam", "Utrecht", weight=46)
G.add_edge("Den Haag", "Rotterdam", weight=32)
G.add_edge("Den Haag", "Utrecht", weight=56)
G.add_edge("Rotterdam", "Utrecht", weight=51)

# Draw the graph
pos = nx.spring_layout(G) # positions for all nodes
nx.draw(G, pos, with_labels=True)
labels = nx.get_edge_attributes(G, 'weight')
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)

plt.show()
```



```
In [2]: import numpy as np

# Convert graph to adjacency matrix
```

```
matrix = nx.to_numpy_matrix(G)
print(matrix)
```

```
[[ 0. 59. 53. 46.]
 [ 0.  0.  0. 51.]
 [ 0. 32.  0. 56.]
 [ 0.  0.  0.  0.]]
```

```
In [3]: from qiskit import Aer, execute
from qiskit.optimization.applications.ising import tsp
from qiskit.optimization.applications.ising.common import sample_most_likely
from qiskit.aqua.algorithms import QAOA
from qiskit.aqua.components.optimizers import COBYLA
from qiskit.aqua import QuantumInstance

# Define the adjacency matrix from the previous step
adjacency_matrix = np.array([[0, 53, 59, 46], [53, 0, 32, 56], [59, 32, 0, 51],

# Convert TSP to a Hamiltonian
qubit_op, offset = tsp.get_operator(adjacency_matrix)

# Set up QAOA
optimizer = COBYLA()
qaoa = QAOA(qubit_op, optimizer, p=4)
backend = Aer.get_backend('qasm_simulator')
quantum_instance = QuantumInstance(backend, shots=1024)

# Solve the problem
result = qaoa.run(quantum_instance)

# Extract the solution
sampled_vector = sample_most_likely(result['eigvecs'][0])
solution = tsp.get_solution(sampled_vector)

print('Solution:', solution)
```

Traceback (most recent call last):

```
Cell In[3], line 2
    from qiskit.optimization.applications.ising import tsp
ModuleNotFoundError: No module named 'qiskit.optimization'
```

Use %tb to get the full traceback.

[Search for solution online](#)

```
In [4]: from qiskit import Aer
from qiskit.algorithms import QAOA, NumPyMinimumEigensolver
from qiskit_optimization.algorithms import MinimumEigenOptimizer
from qiskit_optimization.applications import TSP
from qiskit_optimization.converters import QuadraticProgramToQubo
from qiskit.utils import algorithm_globals, QuantumInstance

# Define the TSP instance
distance_matrix = [[0, 53, 59, 46], [53, 0, 32, 56], [59, 32, 0, 51], [46, 56, 5

tsp = TSP(distance_matrix=distance_matrix)

# Formulate the problem
qp = tsp.to_quadratic_program()

# Solve the problem using a classical solver as a baseline
exact_solver = NumPyMinimumEigensolver()
```

```

exact_solution = MinimumEigenOptimizer(exact_solver).solve(qp)

# Set up the quantum instance (simulator)
quantum_instance = QuantumInstance(Aer.get_backend('qasm_simulator'), shots=1024)

# Solve the problem using QAOA
qaoa = QAOA(optimizer=COBYLA(), quantum_instance=quantum_instance)
qaoa_solution = MinimumEigenOptimizer(qaoa).solve(qp)

# Extract and print solutions
print('Exact solution:', tsp.interpret(exact_solution))
print('QAOA solution:', tsp.interpret(qaoa_solution))

```

/tmp/ipykernel_233/2555904951.py:2: DeprecationWarning: ``qiskit.algorithms`` has been migrated to an independent package: <https://github.com/qiskit-community/qiskit-algorithms>. The ``qiskit.algorithms`` import path is deprecated as of qiskit-terra 0.25.0 and will be removed no earlier than 3 months after the release date. Please run ``pip install qiskit_algorithms`` and use ``import qiskit_algorithms`` instead.

```
from qiskit.algorithms import QAOA, NumPyMinimumEigensolver
```

Traceback (most recent call last):

Cell In[4], line 4

```
from qiskit_optimization.applications import TSP
```

ImportError: cannot import name 'TSP' from 'qiskit_optimization.applications' (/opt/conda/lib/python3.10/site-packages/qiskit_optimization/applications/__init__.py)

Use %tb to get the full traceback.

[Search for solution online](#)

In []:

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ChatGPT

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Explore GPTs

Today

Quantum TSP Algorithm S...

Quantum TSP with Qiskit

P5 -- VERO

P4 - ALICE

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P2 - ALE

P3 - EDO

P5

Add Team workspace



You

Problem to Test: Suppose we have 1 truck that is tasked with driving to Amsterdam, Rotterdam, and Den Haag, starting in and returning to Utrecht. The problem to test is to determine the best order of visiting these cities using a quantum computer algorithm instead of the normal one.

Distance between cities: Amsterdam-Den Haag: 53km; Amsterdam-Rotterdam: 59km; Amsterdam-Utrecht: 46km; Den Haag-Rotterdam: 32km; Den Haag-Utrecht: 56km; Rotterdam-Utrecht: 51km



Quantum Buddy 2.0

To solve this problem using a quantum computer algorithm, we can model the scenario as a Traveling Salesman Problem (TSP). The goal is to find the shortest possible route that visits each city exactly once and returns to the starting city, minimizing the total distance traveled.

Step 1: Define the Cities and Distances

We have four cities: Amsterdam (A), Rotterdam (R), Den Haag (D), and Utrecht (U). The distances between the cities are given as follows:

- Amsterdam-Den Haag: 53km
 - Amsterdam-Rotterdam: 59km
 - Amsterdam-Utrecht: 46km
 - Den Haag-Rotterdam: 32km
 - Den Haag-Utrecht: 56km
 - Rotterdam-Utrecht: 51km
-

Step 2: Create the Graph

We create a weighted graph where the cities are nodes and the distances are the weights of the edges between the nodes.

Step 3: Quantum Algorithm for TSP

Quantum algorithms for the TSP, such as the Quantum Approximate Optimization Algorithm (QAOA) or Variational Quantum Eigensolver (VQE), can be used to find the solution. These algorithms work by encoding the problem into a Hamiltonian whose ground state corresponds to the optimal solution (i.e., the shortest route). Through quantum optimization, we aim to find the parameters that minimize the Hamiltonian, thus solving the TSP.

Step 4: Solve Using a Quantum Computer

With the graph created, we would encode the problem on a quantum computer using a quantum algorithm designed for combinatorial optimization problems like the TSP. Given the small scale of this problem (4 cities), it's a manageable task for current quantum algorithms and simulators. The quantum algorithm would explore all possible routes simultaneously due to quantum superposition, and quantum interference would help in amplifying the probabilities of the shortest routes.

Visualization

To better understand the problem setup, we can visualize the graph with nodes representing cities and weighted edges representing distances. (This step is more relevant to classical approaches but helps conceptualize the problem for quantum solutions.)

Conversion to a Quantum Problem

The distances and routes are encoded into a quantum circuit, which is then optimized to find the shortest path. This involves mapping the cities to qubits and the distances to quantum gates that manipulate the qubit states according to the problem constraints.



Expected Outcome

The quantum algorithm would output the sequence of cities that results in the shortest total distance, starting and ending in Utrecht.

This explanation simplifies the complexities involved in setting up and solving the TSP with quantum computers but gives an overview of the steps and considerations in applying quantum algorithms to logistical problems like the one described.



You

The script should include importing necessary libraries like matplotlib, NumPy, and NetworkX for graph creation and visualization. Utilize functionality for optimization

NetworkX for graph creation and visualization, and matplotlib for optimization problems, specifically focusing on the TSP.

Quantum Buddy 2.0

To solve the Traveling Salesman Problem (TSP) for the given cities using a classical approach as a preliminary step before quantum optimization, we'll write a Python script that includes importing necessary libraries and setting up the problem. This script will create and visualize the graph with cities as nodes and distances as edge weights. While this script does not directly implement quantum algorithms, it sets the stage for quantum optimization by structuring the problem appropriately.

Here's how you can do it:

```
python Copy code

import matplotlib.pyplot as plt
import numpy as np
import networkx as nx

# Step 1: Define the cities and distances
cities = ['Amsterdam', 'Rotterdam', 'Den Haag', 'Utrecht']
distances = {
    ('Amsterdam', 'Den Haag'): 53,
    ('Amsterdam', 'Rotterdam'): 59,
    ('Amsterdam', 'Utrecht'): 46,
    ('Den Haag', 'Rotterdam'): 32,
    ('Den Haag', 'Utrecht'): 56,
    ('Rotterdam', 'Utrecht'): 51
}

# Step 2: Create the graph
G = nx.Graph()
for city in cities:
    G.add_node(city)
for (city1, city2), distance in distances.items():
    G.add_edge(city1, city2, weight=distance)

# Step 3: Visualize the graph
pos = nx.spring_layout(G) # Generate a spring layout for better visual
labels = nx.get_edge_attributes(G, 'weight')
nx.draw(G, pos, with_labels=True, node_color='lightblue', node_size=200)
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)

plt.title('Map of Cities with Distances')
plt.show()

# Note: This script sets up and visualizes the problem.
# For quantum optimization, you would need to encode this problem into
```

Explanation:

- **Import Libraries:** We import 'matplotlib.pyplot' for visualization, 'numpy' for numerical operations (though not directly used in this snippet, it's often useful in handling matrices and arrays for larger TSP problems), and 'networkx' for creating and manipulating the graph structure.
- **Define Cities and Distances:** We list the cities and create a dictionary with city pairs as keys and distances as values.
- **Create the Graph:** Using 'networkx', we add each city as a node and each pair of cities with their distance as weighted edges.
- **Visualize the Graph:** We use 'matplotlib' and 'networkx' functions to draw the graph, with nodes representing cities and edges showing the distances between them.

This script provides a visual representation of the problem, which is a crucial step in understanding the TSP before attempting to solve it with quantum algorithms. For quantum optimization, the problem setup would be similar, but the solution method would involve quantum computing techniques to find the shortest path.



You

maps out the shortest path for the Traveling Salesman Problem using a brute-force approach



Quantum Buddy 2.0

To solve the Traveling Salesman Problem (TSP) using a brute-force approach, we'll extend the Python script to include functionality that calculates all possible routes, computes their total distances, and then identifies the shortest path. This method is computationally expensive and not practical for large numbers of cities, but for a small set like ours (Amsterdam, Rotterdam, Den Haag, and Utrecht), it's feasible.

Here's how to do it:

```
python Copy code

import matplotlib.pyplot as plt
import numpy as np
import networkx as nx
from itertools import permutations

# Step 1: Define the cities and distances
cities = ['Utrecht', 'Amsterdam', 'Rotterdam', 'Den Haag']
distances = {
    ('Amsterdam', 'Den Haag'): 53,
    ('Amsterdam', 'Rotterdam'): 59,
    ('Amsterdam', 'Utrecht'): 46,
    ('Den Haag', 'Rotterdam'): 32,
    ('Den Haag', 'Utrecht'): 56,
    ('Rotterdam', 'Utrecht'): 51,
    # Adding reverse distances because this is an undirected graph
    ('Den Haag', 'Amsterdam'): 53,
    ('Rotterdam', 'Amsterdam'): 59,
    ('Utrecht', 'Amsterdam'): 46,
    ('Rotterdam', 'Den Haag'): 32,
    ('Utrecht', 'Den Haag'): 56,
    ('Utrecht', 'Rotterdam'): 51
}

# Step 2: Create the graph
G = nx.Graph()
for city in cities:
    G.add_node(city)
for (city1, city2), distance in distances.items():
    G.add_edge(city1, city2, weight=distance)

# Step 3: Brute-force search for the shortest path
def compute_path_distance(path):
    return sum(distances[(path[i], path[i+1])] for i in range(len(path)-1))

# Generate all possible paths that start and end in Utrecht
all_paths = [list(path) for path in permutations(cities) if path[0] == 'Utrecht' and path[-1] == 'Utrecht']

# Compute the distance of each path
path_distances = {path: compute_path_distance(path) for path in all_paths}

# Find the shortest path
shortest_path = min(path_distances, key=path_distances.get)
shortest_distance = path_distances[shortest_path]

print(f"The shortest path is: {' -> '.join(shortest_path)} with a distance of {shortest_distance}")

# Step 4: Visualize the shortest path
pos = nx.spring_layout(G) # Generate a spring layout for better visualization
labels = nx.get_edge_attributes(G, 'weight')
nx.draw(G, pos, with_labels=True, node_color='lightblue', node_size=200, edge_color='red', edge_labels=labels)
# Highlight the shortest path
path_edges = list(zip(shortest_path, shortest_path[1:]))
nx.draw_networkx_edges(G, pos, edgelist=path_edges, edge_color='red', width=2)

plt.title('Shortest Path for TSP')
plt.show()
```

How It Works:

- Define Cities and Distances:** We define our cities and the distances between them. Since the graph is undirected, we include distances in both directions for each pair of cities.
- Create the Graph:** We use 'networkx' to create a graph where nodes represent cities and edges represent distances between cities.
- Brute-force Search:** We generate all possible permutations of the cities that start and end in Utrecht (since the truck starts and returns to Utrecht). For each path, we calculate the total distance and then identify the path with the shortest distance.
- Visualize the Shortest Path:** We visualize the graph with all cities and distances. The shortest path identified by the brute-force method is highlighted in red.

This approach will accurately find the shortest path for the TSP with our small set of cities but becomes impractical for larger datasets due to the exponential growth in possible permutations.



you

Traceback (most recent call last):

Cell In[2], line 42

```
shortest_path = min(path_distances, key=path_distances.get)
```

ValueError: min() arg is an empty sequence

Use %tb to get the full traceback



Quantum Buddy 2.0

The error you encountered, 'ValueError: min() arg is an empty sequence', suggests that the 'path_distances' dictionary is empty. This can happen if no valid paths were generated or if there was an issue with how the paths were constructed or evaluated. Let's troubleshoot this issue step by step:

1. **Path Generation Issue:** The use of permutations along with the conditions 'path[0] == 'Utrecht' and path[-1] == 'Utrecht'' might not generate any paths because permutations of the list 'cities' do not include repetitions, and since 'Utrecht' is fixed at both the start and the end, there's a logical oversight. Given that 'Utrecht' is both the starting and ending city, we should adjust the approach to generate permutations of the cities excluding 'Utre' ↓ and then add 'Utrecht' at the start and end of each permutation manually.
2. **Distance Dictionary Keys:** Ensure that the distance dictionary keys correctly match the pairs of cities being generated in your paths. This includes verifying that city names are correctly spelled and match in both the distances dictionary and the cities list.

Let's correct the script to address these potential issues:

python

Copy code

```
import matplotlib.pyplot as plt
import numpy as np
import networkx as nx
from itertools import permutations

# Define the cities and distances
cities = ['Amsterdam', 'Rotterdam', 'Den Haag'] # Utrecht is not included
start_city = 'Utrecht'
distances = {
    ('Amsterdam', 'Den Haag'): 53,
    ('Amsterdam', 'Rotterdam'): 59,
    ('Amsterdam', 'Utrecht'): 46,
    ('Den Haag', 'Rotterdam'): 32,
    ('Den Haag', 'Utrecht'): 56,
    ('Rotterdam', 'Utrecht'): 51,
    # Adding reverse distances because this is an undirected graph
    ('Den Haag', 'Amsterdam'): 53,
    ('Rotterdam', 'Amsterdam'): 59,
    ('Utrecht', 'Amsterdam'): 46,
    ('Rotterdam', 'Den Haag'): 32,
    ('Utrecht', 'Den Haag'): 56,
    ('Utrecht', 'Rotterdam'): 51
}

# Create the graph
G = nx.Graph()
G.add_node(start_city)
for city in cities:
    G.add_node(city)
for (city1, city2), distance in distances.items():
    G.add_edge(city1, city2, weight=distance)

# Function to compute the path distance
def compute_path_distance(path):
    return sum(distances[(path[i], path[i+1])] for i in range(len(path)-1))

# Generate all possible paths, excluding the start/end city from permutations
all_paths = [[start_city] + list(path) + [start_city] for path in permutations(cities)]

# Compute the distance of each path
path_distances = {path: compute_path_distance(path) for path in all_paths}

# Find the shortest path
shortest_path = min(path_distances, key=path_distances.get)
shortest_distance = path_distances[shortest_path]

print(f"The shortest path is: {' -> '.join(shortest_path)} with a distance of {shortest_distance}")

# Visualize the shortest path
pos = nx.spring_layout(G) # Layout for visualization
labels = nx.get_edge_attributes(G, 'weight')
nx.draw(G, pos, with_labels=True, node_color='lightblue', node_size=200, edge_color='lightblue', edge_weight=labels)
```

```

nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)
# Highlight the shortest path
path_edges = list(zip(shortest_path, shortest_path[1:]))
nx.draw_networkx_edges(G, pos, edgelist=path_edges, edge_color='red', w

plt.title('Shortest Path for TSP')
plt.show()

```

This revised script corrects the path generation by considering permutations of the cities excluding 'Utrecht' and then appending 'Utrecht' at the start and end of each path. This approach ensures that 'all_paths' contains a valid list of paths, thus avoiding the 'ValueError' when attempting to find the minimum of an empty sequence.



You

Traceback (most recent call last):

Cell In[3], line 41

```
path_distances = {path: compute_path_distance(path) for path in all_paths}
```

Cell In[3], line 41 in <dictcomp>

```
path_distances = {path: compute_path_distance(path) for path in all_paths}
```

TypeError: unhashable type: 'list'

Use %tb to get the full traceback.



Quantum Buddy 2.0

The error 'TypeError: unhashable type: 'list'' occurs because dictionaries in Python require their keys to be of a hashable type, and lists are not hashable due to their mutability. When you use the path (which is a list) as a key in the dictionary comprehension, Python raises this error.

To resolve this issue, you can convert the path list to a tuple, which is immutable and thus hashable, before using it as a key in the 'path_distances' dictionary. Here's how you can adjust the script:

```

python
Copy code

import matplotlib.pyplot as plt
import numpy as np
import networkx as nx
from itertools import permutations

# Define the cities and distances
cities = ['Amsterdam', 'Rotterdam', 'Den Haag'] # Utrecht is not inclu
start_city = 'Utrecht'
distances = {
    ('Amsterdam', 'Den Haag'): 53,
    ('Amsterdam', 'Rotterdam'): 59,
    ('Amsterdam', 'Utrecht'): 46,
    ('Den Haag', 'Rotterdam'): 32,
    ('Den Haag', 'Utrecht'): 56,
    ('Rotterdam', 'Utrecht'): 51,
    # Adding reverse distances because this is an undirected graph
    ('Den Haag', 'Amsterdam'): 53,
    ('Rotterdam', 'Amsterdam'): 59,
    ('Utrecht', 'Amsterdam'): 46,
    ('Rotterdam', 'Den Haag'): 32,
    ('Utrecht', 'Den Haag'): 56,
    ('Utrecht', 'Rotterdam'): 51
}

# Create the graph
G = nx.Graph()
G.add_node(start_city)
for city in cities:
    G.add_node(city)
for (city1, city2), distance in distances.items():
    G.add_edge(city1, city2, weight=distance)

# Function to compute the path distance
def compute_path_distance(path):
    return sum(distances[(path[i], path[i+1])] for i in range(len(path)-1))

# Generate all possible paths, excluding the start/end city from permut
all_paths = [(start_city) + list(path) + [start_city] for path in permut

# Compute the distance of each path
path_distances = {tuple(path): compute_path_distance(path) for path in all_paths}

# Find the shortest path
shortest_path = min(path_distances, key=path_distances.get)

```

```

shortest_path = min(path_distances, key=path_distances.get)
shortest_distance = path_distances[shortest_path]

print(f"The shortest path is: {' -> '.join(shortest_path)} with a distance of {shortest_distance}")

# Visualize the shortest path
pos = nx.spring_layout(G) # Layout for visualization
labels = nx.get_edge_attributes(G, 'weight')
nx.draw(G, pos, with_labels=True, node_color='lightblue', node_size=200)
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)
# Highlight the shortest path
path_edges = list(zip(shortest_path, shortest_path[1:]))
nx.draw_networkx_edges(G, pos, edgelist=path_edges, edge_color='red', width=4)

plt.title('Shortest Path for TSP')
plt.show()

```

In this revised script, before using a path as a key in `path_distances`, it is converted to a tuple: `tuple(path)`. This change ensures that the keys are hashable, allowing the dictionary to store the total distance associated with each path correctly. This solution also ensures that you can still iterate over the paths and calculate distances without encountering the type error.



Collaborate on a Team plan



co doc



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ChatGPT can make mistakes. Consider checking important information.



```
In [1]: import matplotlib.pyplot as plt
import numpy as np
import networkx as nx

# Step 1: Define the cities and distances
cities = ['Amsterdam', 'Rotterdam', 'Den Haag', 'Utrecht']
distances = {
    ('Amsterdam', 'Den Haag'): 53,
    ('Amsterdam', 'Rotterdam'): 59,
    ('Amsterdam', 'Utrecht'): 46,
    ('Den Haag', 'Rotterdam'): 32,
    ('Den Haag', 'Utrecht'): 56,
    ('Rotterdam', 'Utrecht'): 51
}

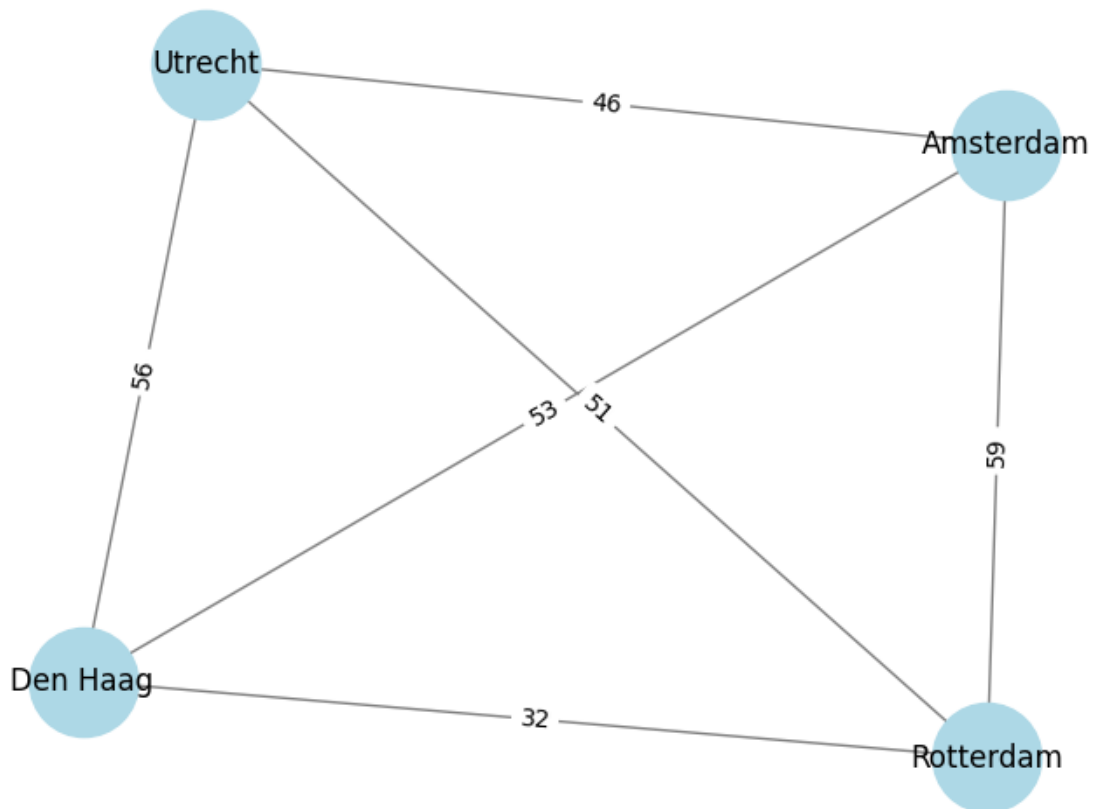
# Step 2: Create the graph
G = nx.Graph()
for city in cities:
    G.add_node(city)
for (city1, city2), distance in distances.items():
    G.add_edge(city1, city2, weight=distance)

# Step 3: Visualize the graph
pos = nx.spring_layout(G) # Generate a spring layout for better visualization
labels = nx.get_edge_attributes(G, 'weight')
nx.draw(G, pos, with_labels=True, node_color='lightblue', node_size=2000, edge_c
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)

plt.title('Map of Cities with Distances')
plt.show()

# Note: This script sets up and visualizes the problem.
# For quantum optimization, you would need to encode this problem into a quantum
```

Map of Cities with Distances



```

In [2]: import matplotlib.pyplot as plt
import numpy as np
import networkx as nx
from itertools import permutations

# Step 1: Define the cities and distances
cities = ['Utrecht', 'Amsterdam', 'Rotterdam', 'Den Haag']
distances = {
    ('Amsterdam', 'Den Haag'): 53,
    ('Amsterdam', 'Rotterdam'): 59,
    ('Amsterdam', 'Utrecht'): 46,
    ('Den Haag', 'Rotterdam'): 32,
    ('Den Haag', 'Utrecht'): 56,
    ('Rotterdam', 'Utrecht'): 51,
    # Adding reverse distances because this is an undirected graph
    ('Den Haag', 'Amsterdam'): 53,
    ('Rotterdam', 'Amsterdam'): 59,
    ('Utrecht', 'Amsterdam'): 46,
    ('Rotterdam', 'Den Haag'): 32,
    ('Utrecht', 'Den Haag'): 56,
    ('Utrecht', 'Rotterdam'): 51
}

# Step 2: Create the graph
G = nx.Graph()
for city in cities:
    G.add_node(city)
for (city1, city2), distance in distances.items():
    G.add_edge(city1, city2, weight=distance)

# Step 3: Brute-force search for the shortest path
def compute_path_distance(path):

```

```

    return sum(distances[(path[i], path[i+1])] for i in range(len(path) - 1))

# Generate all possible paths that start and end in Utrecht
all_paths = [list(path) for path in permutations(cities) if path[0] == 'Utrecht']

# Compute the distance of each path
path_distances = {path: compute_path_distance(path) for path in all_paths}

# Find the shortest path
shortest_path = min(path_distances, key=path_distances.get)
shortest_distance = path_distances[shortest_path]

print(f"The shortest path is: {' -> '.join(shortest_path)} with a distance of {s}")

# Step 4: Visualize the shortest path
pos = nx.spring_layout(G) # Generate a spring layout for better visualization
labels = nx.get_edge_attributes(G, 'weight')
nx.draw(G, pos, with_labels=True, node_color='lightblue', node_size=2000, edge_c
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)
# Highlight the shortest path
path_edges = list(zip(shortest_path, shortest_path[1:]))
nx.draw_networkx_edges(G, pos, edgelist=path_edges, edge_color='red', width=2)

plt.title('Shortest Path for TSP')
plt.show()

```

Traceback (most recent call last):

Cell In[2], line 42

```
shortest_path = min(path_distances, key=path_distances.get)
```

ValueError: min() arg is an empty sequence

Use %tb to get the full traceback.

[Search for solution online](#)

```

In [3]: import matplotlib.pyplot as plt
import numpy as np
import networkx as nx
from itertools import permutations

# Define the cities and distances
cities = ['Amsterdam', 'Rotterdam', 'Den Haag'] # Utrecht is not included in th
start_city = 'Utrecht'
distances = {
    ('Amsterdam', 'Den Haag'): 53,
    ('Amsterdam', 'Rotterdam'): 59,
    ('Amsterdam', 'Utrecht'): 46,
    ('Den Haag', 'Rotterdam'): 32,
    ('Den Haag', 'Utrecht'): 56,
    ('Rotterdam', 'Utrecht'): 51,
    # Adding reverse distances because this is an undirected graph
    ('Den Haag', 'Amsterdam'): 53,
    ('Rotterdam', 'Amsterdam'): 59,
    ('Utrecht', 'Amsterdam'): 46,
    ('Rotterdam', 'Den Haag'): 32,
    ('Utrecht', 'Den Haag'): 56,
    ('Utrecht', 'Rotterdam'): 51
}

# Create the graph
G = nx.Graph()

```

```

G.add_node(start_city)
for city in cities:
    G.add_node(city)
for (city1, city2), distance in distances.items():
    G.add_edge(city1, city2, weight=distance)

# Function to compute the path distance
def compute_path_distance(path):
    return sum(distances[(path[i], path[i+1])] for i in range(len(path) - 1))

# Generate all possible paths, excluding the start/end city from permutations
all_paths = [[start_city] + list(path) + [start_city] for path in permutations

# Compute the distance of each path
path_distances = {path: compute_path_distance(path) for path in all_paths}

# Find the shortest path
shortest_path = min(path_distances, key=path_distances.get)
shortest_distance = path_distances[shortest_path]

print(f"The shortest path is: {' -> '.join(shortest_path)} with a distance of {s

# Visualize the shortest path
pos = nx.spring_layout(G) # Layout for visualization
labels = nx.get_edge_attributes(G, 'weight')
nx.draw(G, pos, with_labels=True, node_color='lightblue', node_size=2000, edge_c
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)
# Highlight the shortest path
path_edges = list(zip(shortest_path, shortest_path[1:]))
nx.draw_networkx_edges(G, pos, edgelist=path_edges, edge_color='red', width=2)

plt.title('Shortest Path for TSP')
plt.show()

```

Traceback (most recent call last):

```

Cell In[3], line 41
    path_distances = {path: compute_path_distance(path) for path in all_paths}
Cell In[3], line 41 in <dictcomp>
    path_distances = {path: compute_path_distance(path) for path in all_paths}
TypeError: unhashable type: 'list'

```

Use %tb to get the full traceback.

[Search for solution online](#)

```

In [4]: import matplotlib.pyplot as plt
import numpy as np
import networkx as nx
from itertools import permutations

# Define the cities and distances
cities = ['Amsterdam', 'Rotterdam', 'Den Haag'] # Utrecht is not included in th
start_city = 'Utrecht'
distances = {
    ('Amsterdam', 'Den Haag'): 53,
    ('Amsterdam', 'Rotterdam'): 59,
    ('Amsterdam', 'Utrecht'): 46,
    ('Den Haag', 'Rotterdam'): 32,
    ('Den Haag', 'Utrecht'): 56,
    ('Rotterdam', 'Utrecht'): 51,
    # Adding reverse distances because this is an undirected graph

```

```

    ('Den Haag', 'Amsterdam'): 53,
    ('Rotterdam', 'Amsterdam'): 59,
    ('Utrecht', 'Amsterdam'): 46,
    ('Rotterdam', 'Den Haag'): 32,
    ('Utrecht', 'Den Haag'): 56,
    ('Utrecht', 'Rotterdam'): 51
}

# Create the graph
G = nx.Graph()
G.add_node(start_city)
for city in cities:
    G.add_node(city)
for (city1, city2), distance in distances.items():
    G.add_edge(city1, city2, weight=distance)

# Function to compute the path distance
def compute_path_distance(path):
    return sum(distances[(path[i], path[i+1])] for i in range(len(path) - 1))

# Generate all possible paths, excluding the start/end city from permutations
all_paths = [(start_city + list(path) + start_city) for path in permutations(cities)]

# Compute the distance of each path
path_distances = {tuple(path): compute_path_distance(path) for path in all_paths}

# Find the shortest path
shortest_path = min(path_distances, key=path_distances.get)
shortest_distance = path_distances[shortest_path]

print(f"The shortest path is: {' -> '.join(shortest_path)} with a distance of {shortest_distance} km")

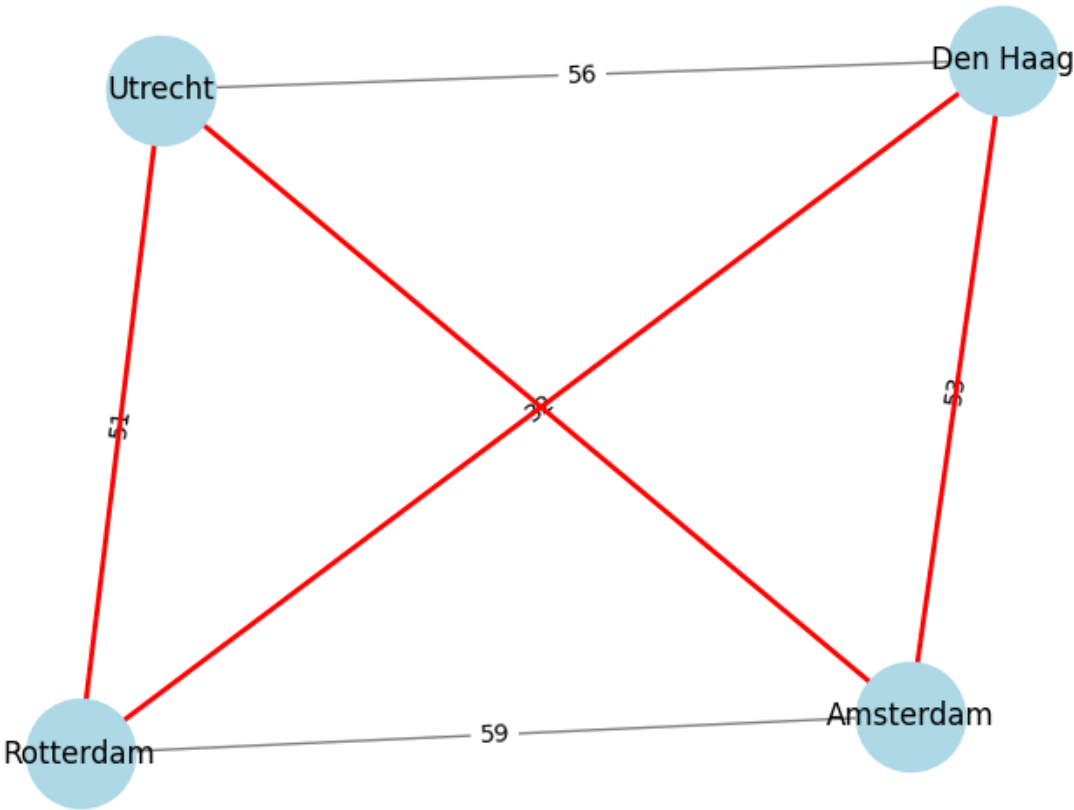
# Visualize the shortest path
pos = nx.spring_layout(G) # Layout for visualization
labels = nx.get_edge_attributes(G, 'weight')
nx.draw(G, pos, with_labels=True, node_color='lightblue', node_size=2000, edge_color='black')
nx.draw_networkx_edge_labels(G, pos, edge_labels=labels)
# Highlight the shortest path
path_edges = list(zip(shortest_path, shortest_path[1:]))
nx.draw_networkx_edges(G, pos, edgelist=path_edges, edge_color='red', width=2)

plt.title('Shortest Path for TSP')
plt.show()

```

The shortest path is: Utrecht -> Amsterdam -> Den Haag -> Rotterdam -> Utrecht
with a distance of 182 km

Shortest Path for TSP



In []:

d. Solution of the problem (Next Page)

```
In [1]: # useful additional packages
import matplotlib.pyplot as plt
import numpy as np
import networkx as nx

from qiskit.tools.visualization import plot_histogram
from qiskit.circuit.library import TwoLocal
from qiskit_optimization.applications import Maxcut, Tsp
from qiskit_algorithms import SamplingVQE, NumPyMinimumEigensolver
from qiskit_algorithms.optimizers import SPSA
from qiskit_algorithms.utils import algorithm_globals
from qiskit.primitives import Sampler
from qiskit_optimization.algorithms import MinimumEigenOptimizer
```

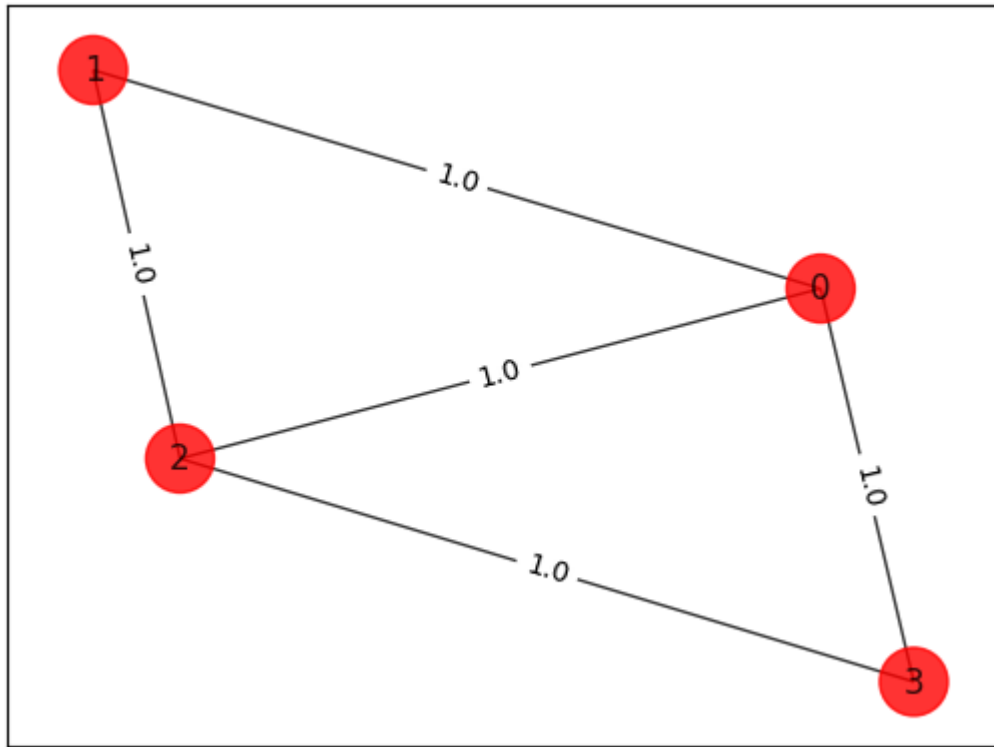
```
In [2]: # Generating a graph of 4 nodes

n = 4 # Number of nodes in graph
G = nx.Graph()
G.add_nodes_from(np.arange(0, n, 1))
elist = [(0, 1, 1.0), (0, 2, 1.0), (0, 3, 1.0), (1, 2, 1.0), (2, 3, 1.0)]
# tuple is (i,j,weight) where (i,j) is the edge
G.add_weighted_edges_from(elist)

colors = ["r" for node in G.nodes()]
pos = nx.spring_layout(G)

def draw_graph(G, colors, pos):
    default_axes = plt.axes(frameon=True)
    nx.draw_networkx(G, node_color=colors, node_size=600, alpha=0.8, ax=default_
    edge_labels = nx.get_edge_attributes(G, "weight")
    nx.draw_networkx_edge_labels(G, pos=pos, edge_labels=edge_labels)

draw_graph(G, colors, pos)
```



In [3]: *# Computing the weight matrix from the random graph*

```

w = np.zeros([n, n])
for i in range(n):
    for j in range(n):
        temp = G.get_edge_data(i, j, default=0)
        if temp != 0:
            w[i, j] = temp["weight"]
print(w)

```

```

[[0. 1. 1. 1.]
 [1. 0. 1. 0.]
 [1. 1. 0. 1.]
 [1. 0. 1. 0.]]

```

In [4]:

```

best_cost_brute = 0
for b in range(2**n):
    x = [int(t) for t in reversed(list(bin(b)[2:].zfill(n)))]
    cost = 0
    for i in range(n):
        for j in range(n):
            cost = cost + w[i, j] * x[i] * (1 - x[j])
    if best_cost_brute < cost:
        best_cost_brute = cost
        xbest_brute = x
    print("case = " + str(x) + " cost = " + str(cost))

```

```

colors = ["r" if xbest_brute[i] == 0 else "c" for i in range(n)]
draw_graph(G, colors, pos)
print("\nBest solution = " + str(xbest_brute) + " cost = " + str(best_cost_brute))

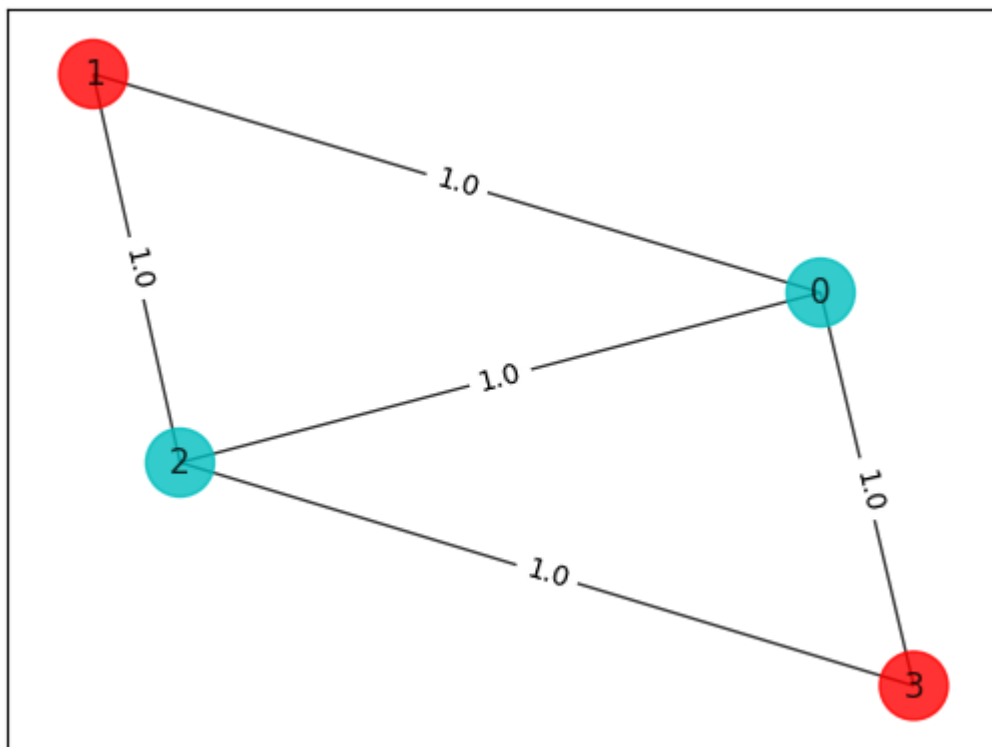
```

```

case = [0, 0, 0, 0] cost = 0.0
case = [1, 0, 0, 0] cost = 3.0
case = [0, 1, 0, 0] cost = 2.0
case = [1, 1, 0, 0] cost = 3.0
case = [0, 0, 1, 0] cost = 3.0
case = [1, 0, 1, 0] cost = 4.0
case = [0, 1, 1, 0] cost = 3.0
case = [1, 1, 1, 0] cost = 2.0
case = [0, 0, 0, 1] cost = 2.0
case = [1, 0, 0, 1] cost = 3.0
case = [0, 1, 0, 1] cost = 4.0
case = [1, 1, 0, 1] cost = 3.0
case = [0, 0, 1, 1] cost = 3.0
case = [1, 0, 1, 1] cost = 2.0
case = [0, 1, 1, 1] cost = 3.0
case = [1, 1, 1, 1] cost = 0.0

```

Best solution = [1, 0, 1, 0] cost = 4.0



```

In [5]: max_cut = Maxcut(w)
        qp = max_cut.to_quadratic_program()
        print(qp.prettyprint())

```

Problem name: Max-cut

Maximize

$$-2x_0x_1 - 2x_0x_2 - 2x_0x_3 - 2x_1x_2 - 2x_2x_3 + 3x_0 + 2x_1 + 3x_2 + 2x_3$$

Subject to

No constraints

Binary variables (4)

x_0 x_1 x_2 x_3

```
In [6]: qubitOp, offset = qp.to_ising()  
print("Offset:", offset)  
print("Ising Hamiltonian:")  
print(str(qubitOp))
```

```
Offset: -2.5  
Ising Hamiltonian:  
0.5 * IIZZ  
+ 0.5 * IZIZ  
+ 0.5 * IZZI  
+ 0.5 * ZIIZ  
+ 0.5 * ZZII
```

```
In [ ]:
```