

Vario-scale visualization of the AHN2 point cloud

P5 presentation

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Second mentor:	Dr.ir. B.M. Meijers
Examiner:	Dr. H.M.H. van der Heijden

What is a point cloud?

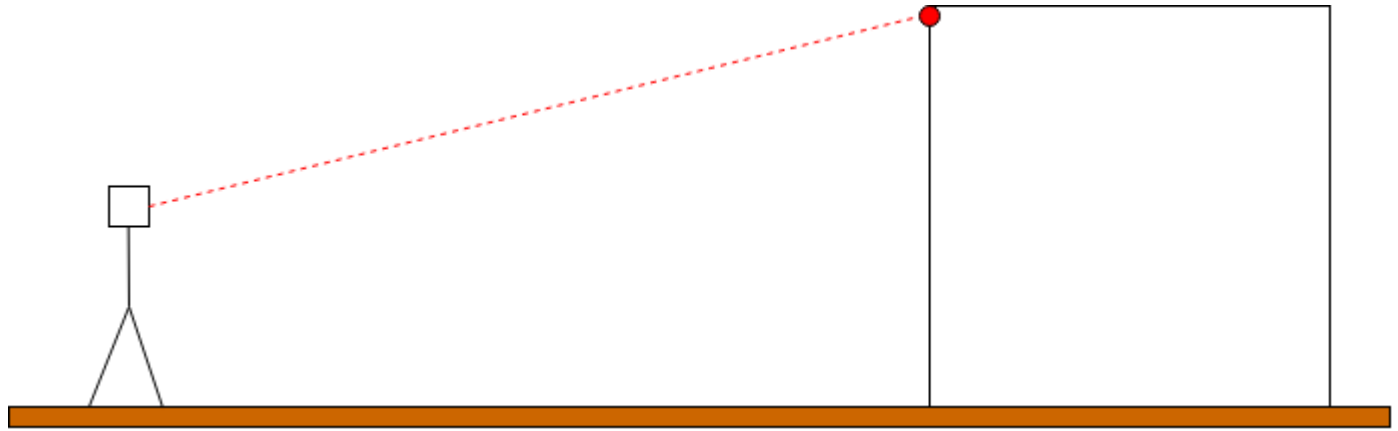
What is a point cloud?

- Collection of geo-referenced points

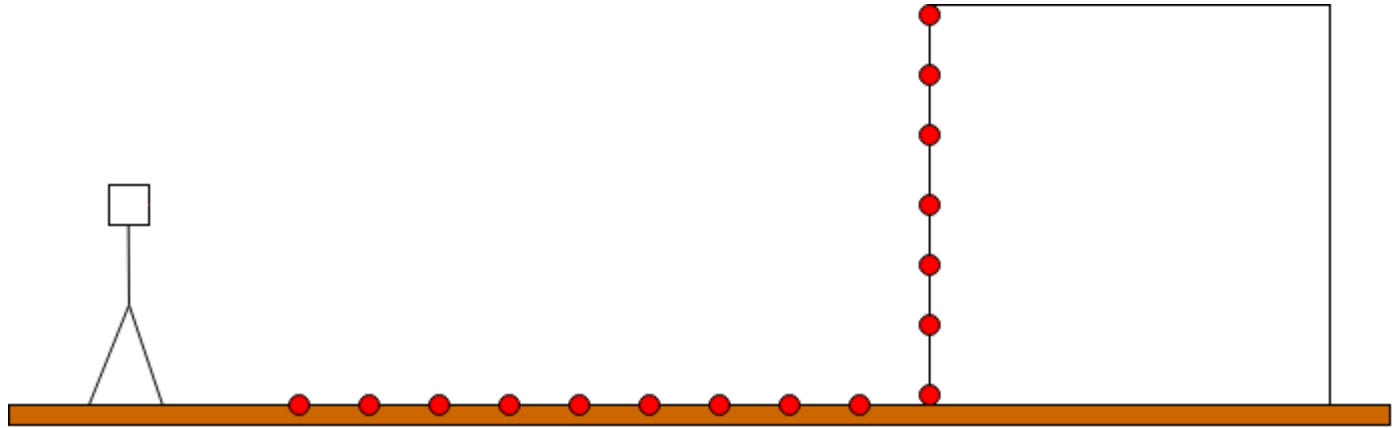
What is a point cloud?

- Collection of geo-referenced points
- Collected using LiDAR (Light Detection and Ranging)

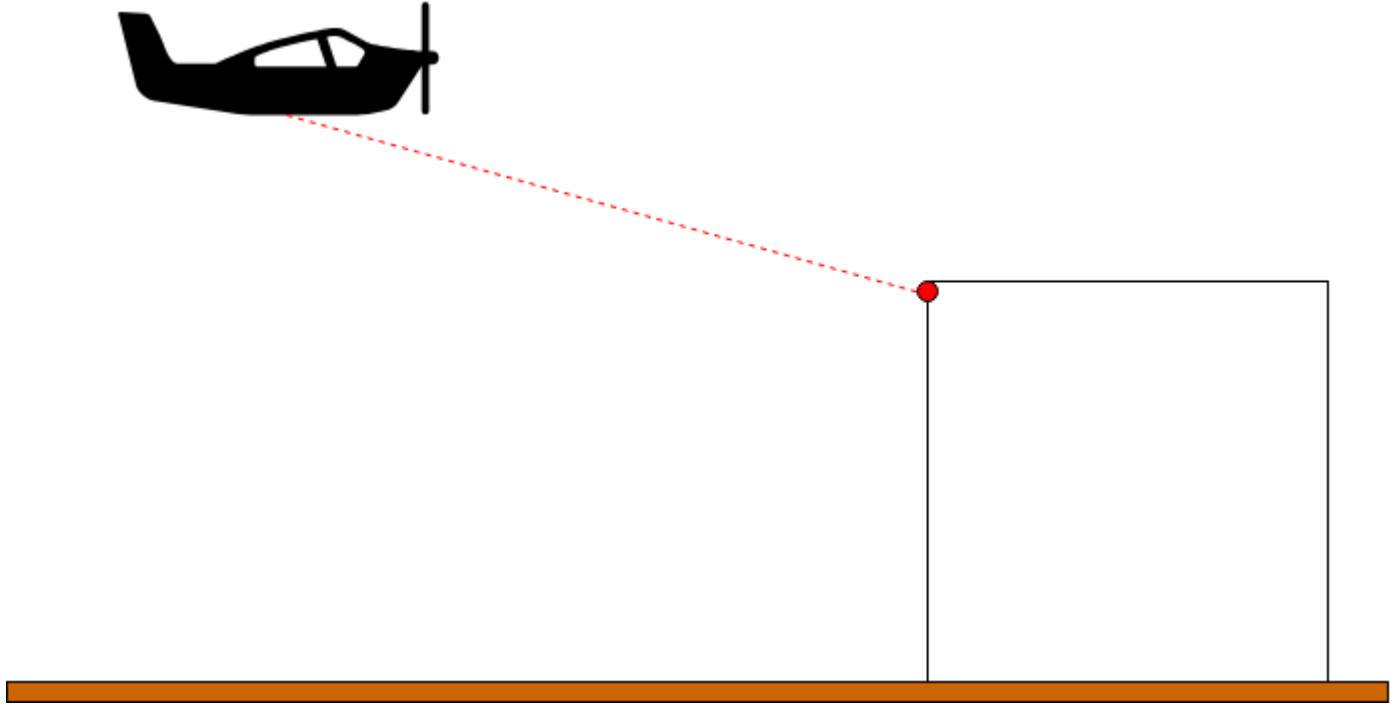
What is a point cloud?



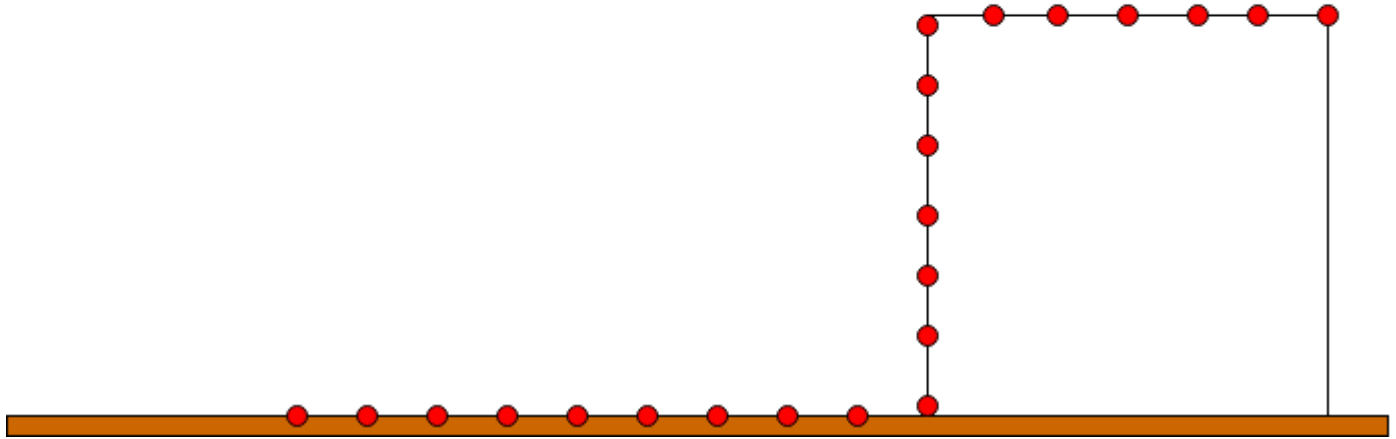
What is a point cloud?



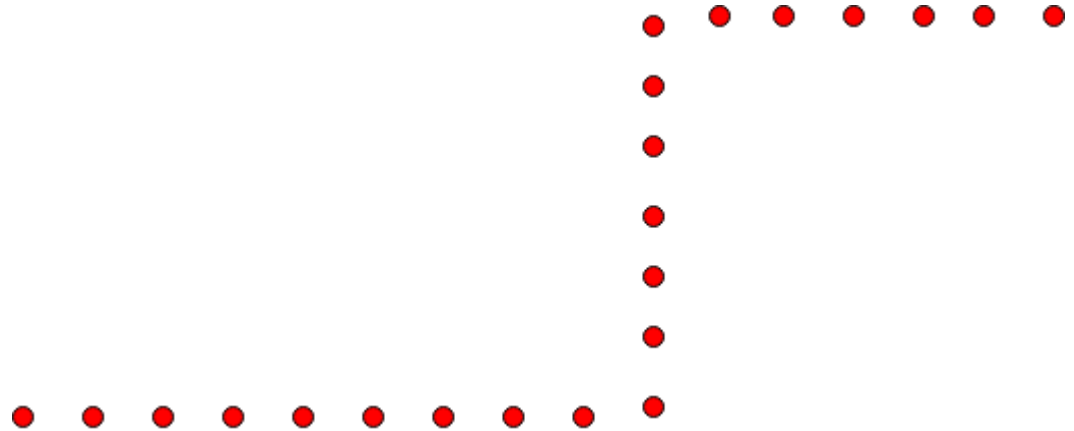
What is a point cloud?



What is a point cloud?



What is a point cloud?

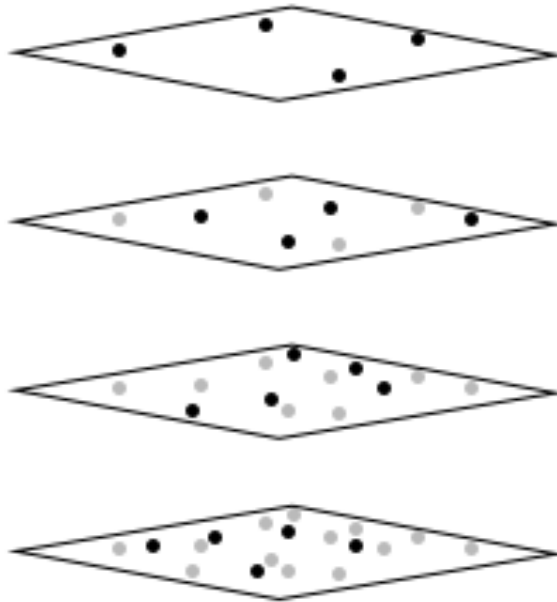


Thesis goal

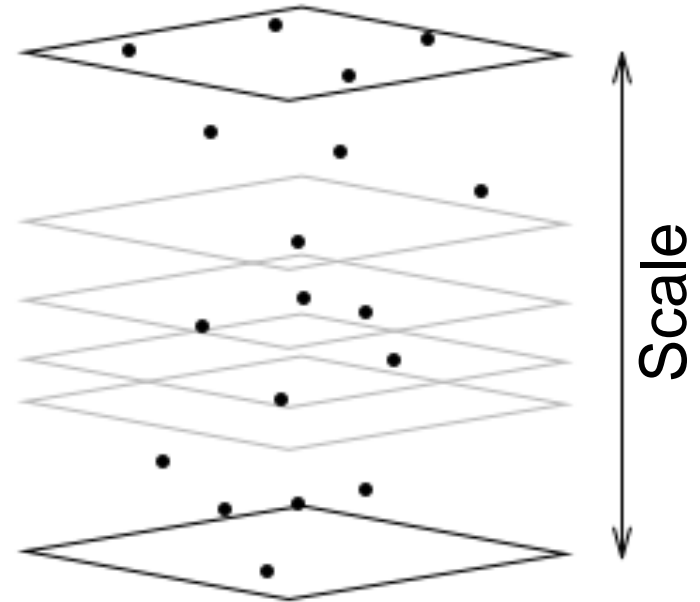
- From discrete visualisation of massive point clouds, to the continuous visualization of massive point clouds.

What is vario-scale visualization?

Vario-scale visualization

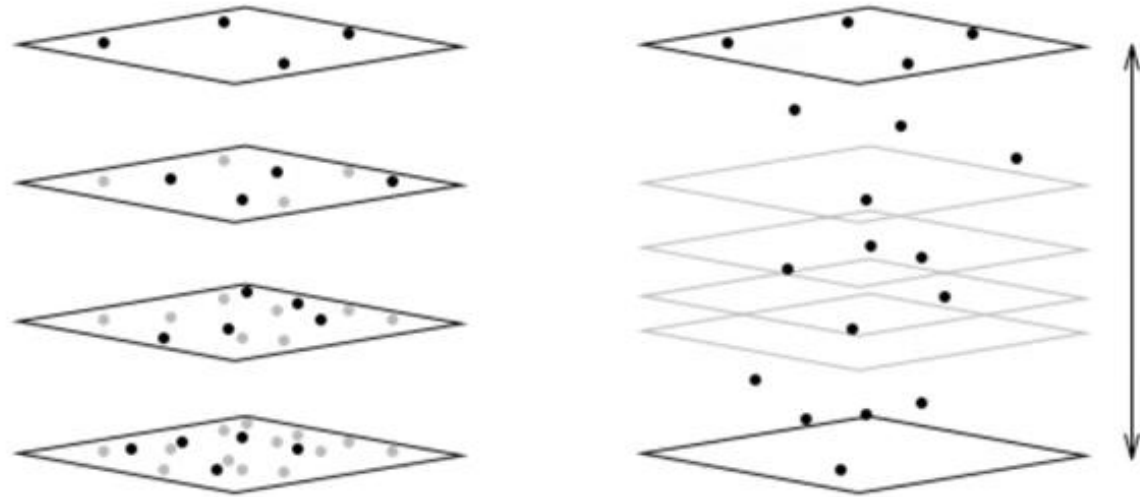


Discrete

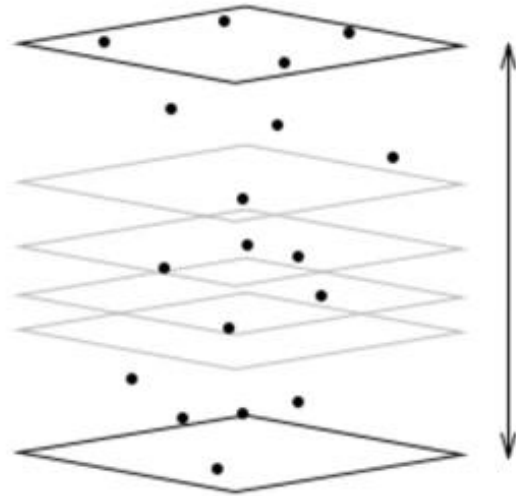


Continuous

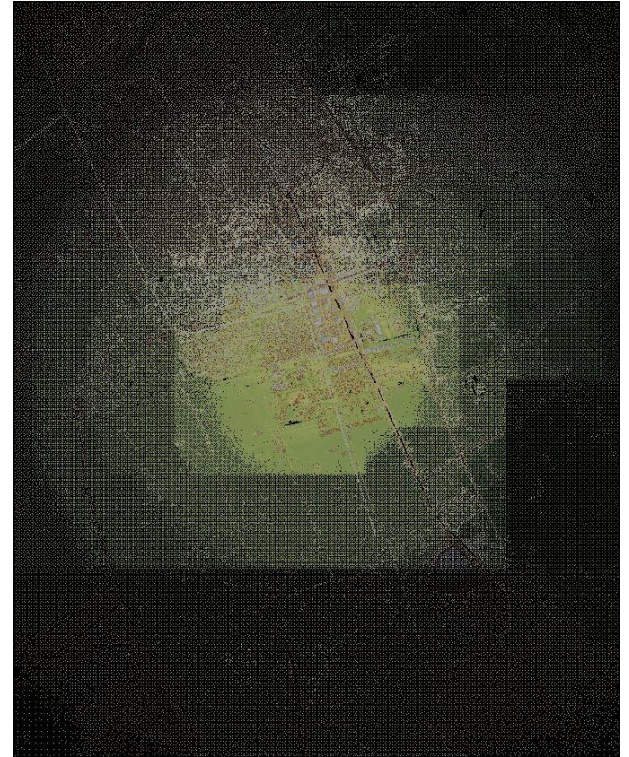
Discrete visualization



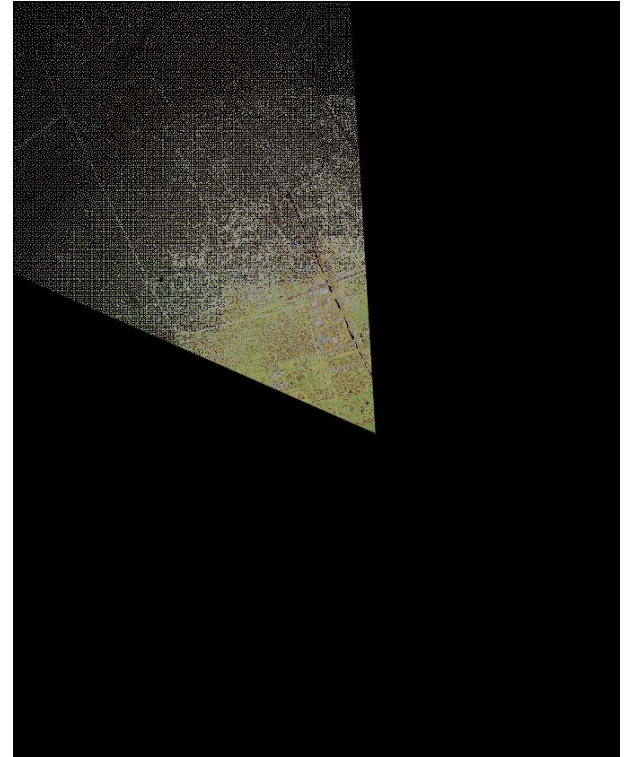
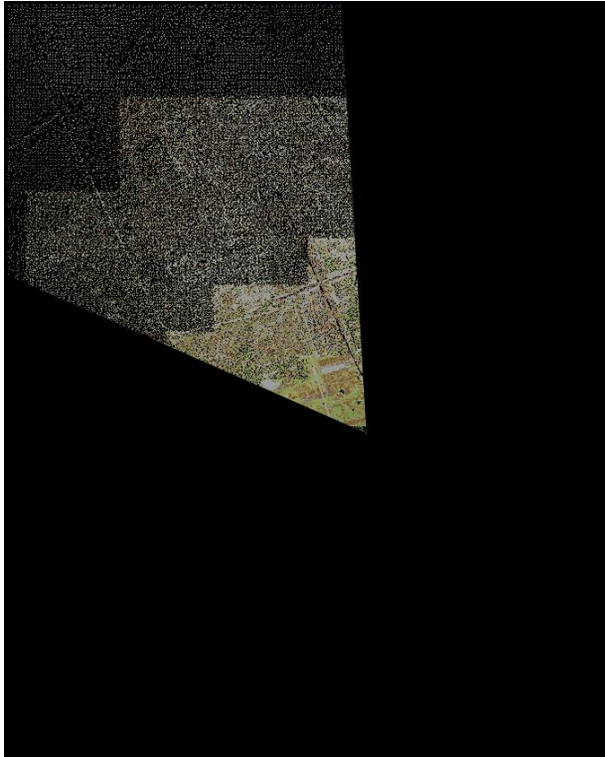
Continuous visualization



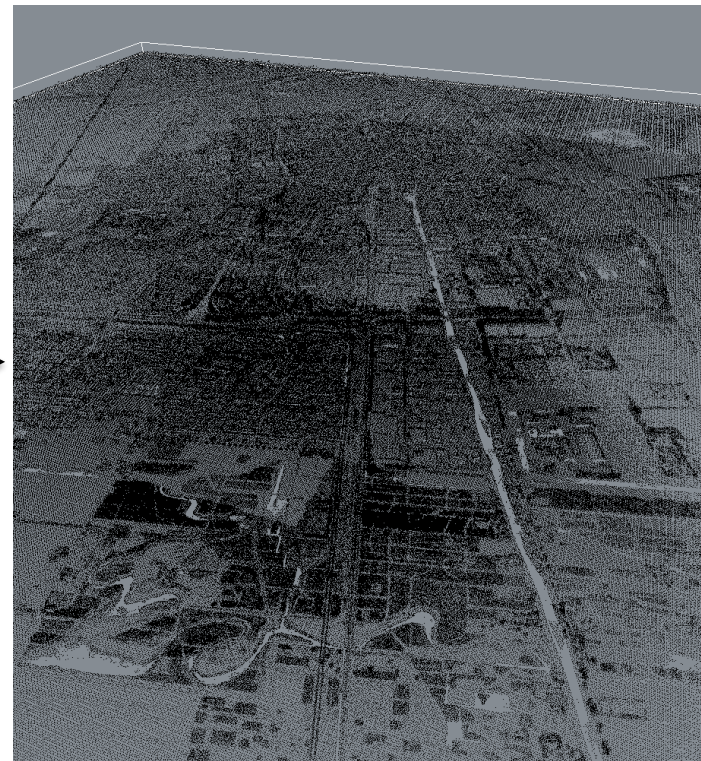
Top-down data set



Top-down camera angle



Perspective



- Introduction
- Research
- Theory
- Implementation

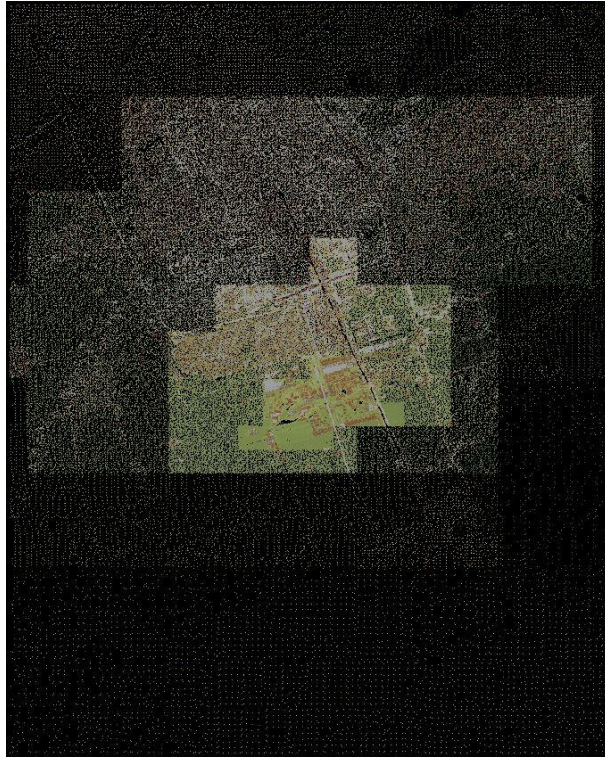
- Introduction
- Research
- Theory
- Implementation

What is the problem?

What is the problem?

Density jumps in web based point cloud visualization

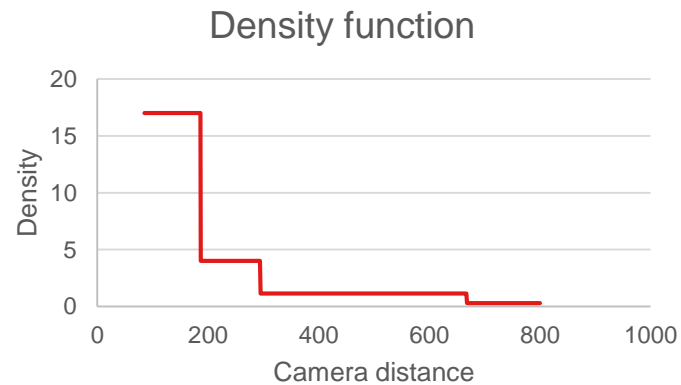
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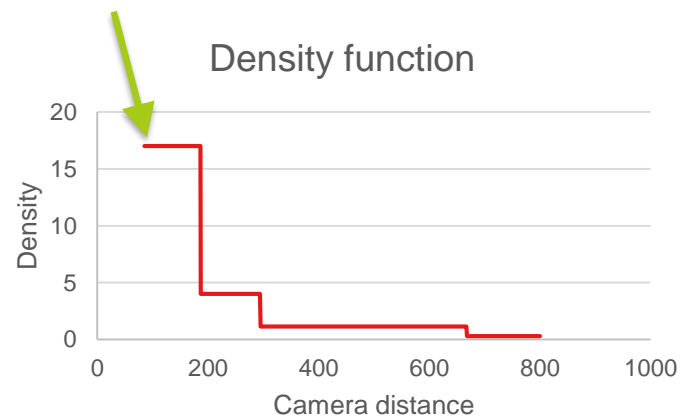
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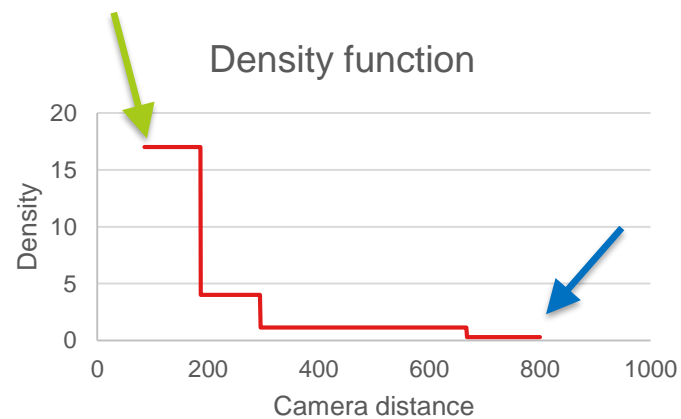
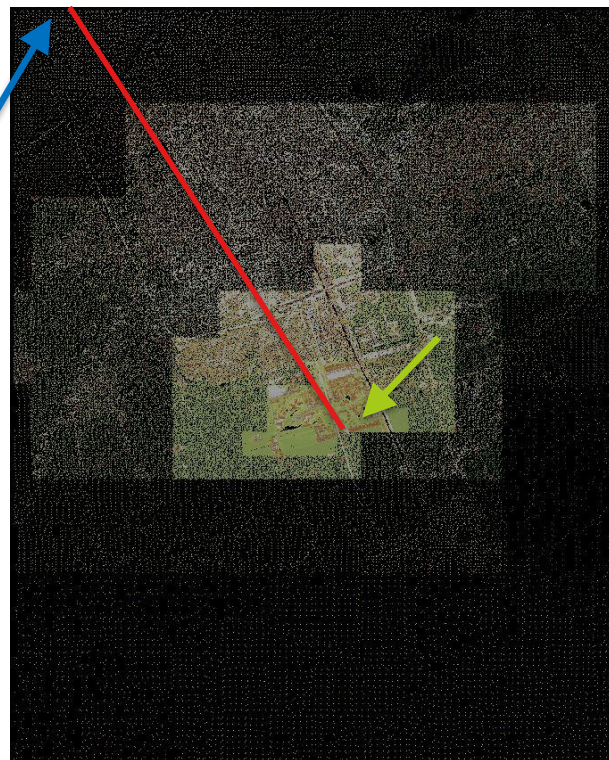
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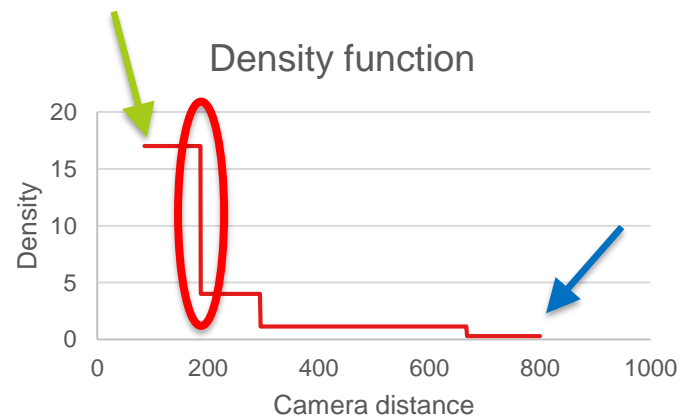
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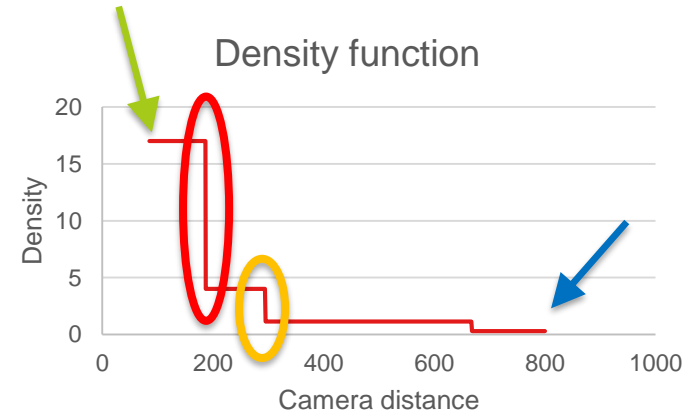
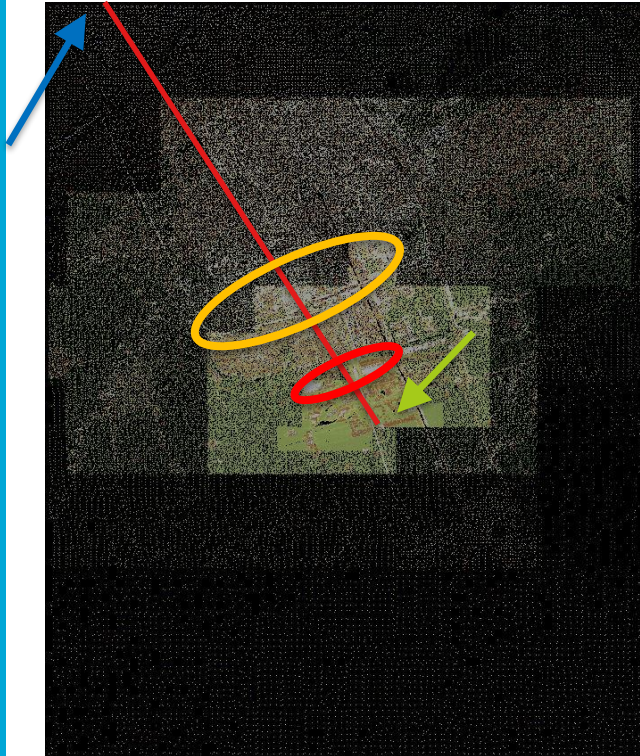
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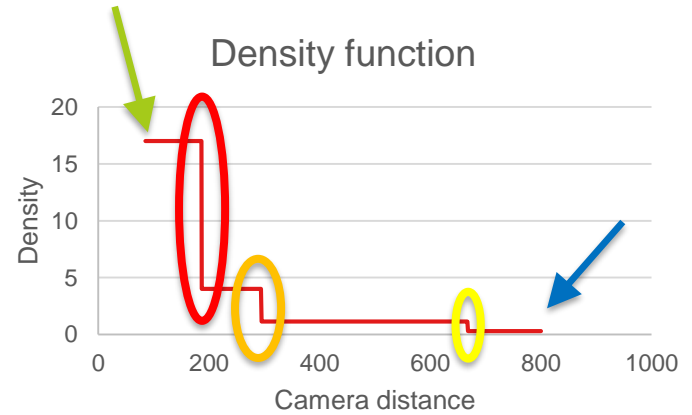
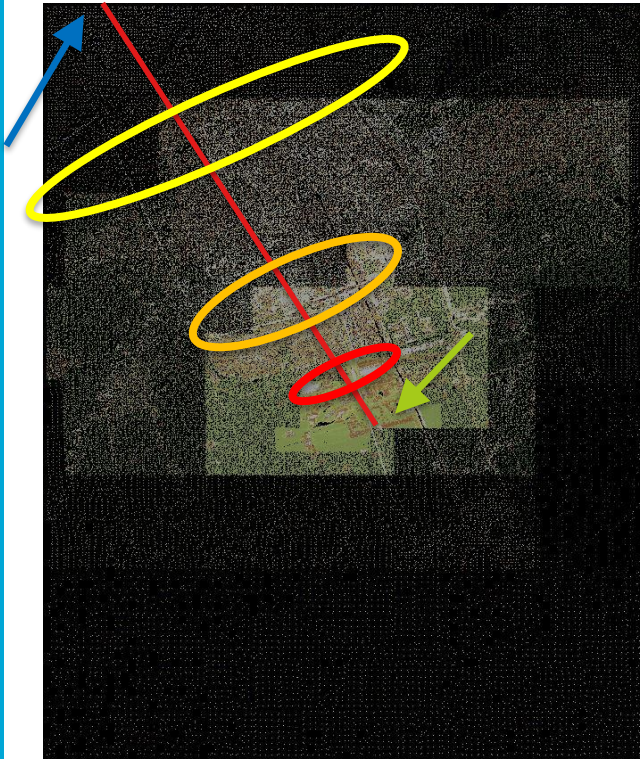
What is the problem?



What is the problem?



What is the problem?



What is the problem?

- In different media these density jumps manifest themselves differently

What is the problem?

- Single frame perspective

What is the problem?



What is the problem?



What is the problem?



What is the problem?

- Single frame perspective
- 30 fps web visualization

What is the problem?

The screenshot shows a web browser window displaying a 3D point cloud viewer. The browser's developer tools are open, showing the Network tab. The network log indicates a significant performance issue, with a total load time of 2.79 seconds and 335 requests. The log shows a large number of 'read' requests for depth and hierarchy data, which are taking a long time to load. The viewer interface on the left shows various settings like 'Point budget: 100,000' and 'Field of view: 100'. The 3D view shows a complex point cloud structure.

Name	Status	Type	Ir	Size	Time	Waterfall
read?depth?Begin=13&depth?End=1...	200	xhr	✓	268 KB	2.44 s	
read?depth?Begin=13&depth?End=1...	200	xhr	✓	283 KB	1.91 s	
read?depth?Begin=13&depth?End=1...	200	xhr	✓	261 KB	1.93 s	
read?depth?Begin=14&depth?End=1...	200	xhr	✓	65.3 KB	808 ...	
read?depth?Begin=14&depth?End=1...	200	xhr	✓	87.9 KB	2.27 s	
hierarchy?bounds= -23375;-11687.5...	200	xhr	✓	880 B	1.72 s	
read?depth?Begin=13&depth?End=1...	200	xhr	✓	310 KB	17.0...	
read?depth?Begin=13&depth?End=1...	200	xhr	✓	330 KB	17.4...	
read?depth?Begin=12&depth?End=1...	200	xhr	✓	391 KB	14.2...	
read?depth?Begin=12&depth?End=1...	200	xhr	✓	582 KB	15.9...	
ui-bg_glass_65_fff_1x400.png	200	png	✓	541 B	3.38 s	
read?depth?Begin=14&depth?End=1...	200	xhr	✓	39.2 KB	11.0...	
read?depth?Begin=14&depth?End=1...	200	xhr	✓	91.2 KB	13.9...	
hierarchy?bounds= 0.0;-11687.5;116...	200	xhr	✓	1.6 KB	2.86 s	
ui-bg_glass_100_fff_1x400.png	200	png	✓	662 B	297 ...	
read?depth?Begin=13&depth?End=1...	200	xhr	✓	371 KB	1.28 s	
read?depth?Begin=13&depth?End=1...	200	xhr	✓	551 B	45 ms	
hierarchy?bounds= -23375;-11687.5...	200	xhr	✓	355 B	67 ms	
hierarchy?bounds= -35062.5;-11687...	200	xhr	✓	338 B	44 ms	
read?depth?Begin=12&depth?End=1...	200	xhr	✓	503 KB	2.62 s	
read?depth?Begin=13&depth?End=1...	200	xhr	✓	4.1 KB	47 ms	
read?depth?Begin=13&depth?End=1...	200	xhr	✓	723 B	2.59 s	
read?depth?Begin=11&depth?End=1...	200	xhr	✓	(from disk cache)	4 ms	
read?depth?Begin=12&depth?End=1...	200	xhr	✓	6.4 KB	2.16 s	

335 requests | 26.3 MB transferred | Finish: 23.7 min | DOMContentLoaded: 1.84 s | Load: 2.79 s

Search across all network headers
Press Control + F in the Network panel to open the Network Search pane.

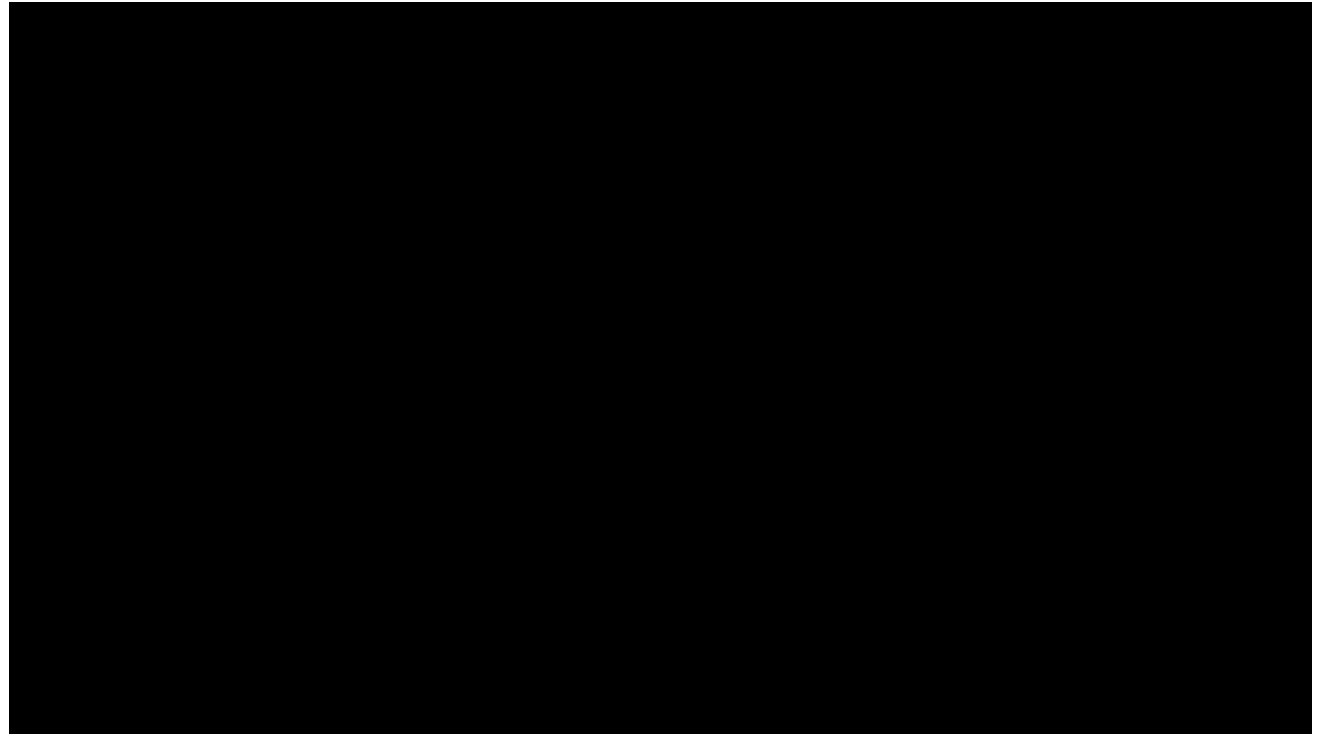
CSS variable value previews in the Styles pane
When a property value is a CSS variable, DevTools now shows a color preview next to the variable.

Stop infinite loops

What is the problem?

- Single frame perspective
- 30 fps popping
- VR inconsistent loading

What is the problem?

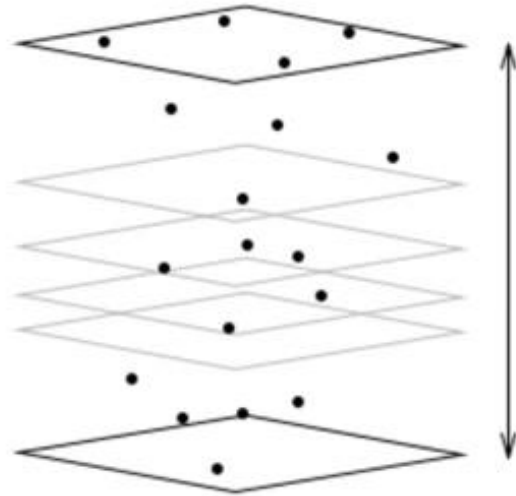


What is the solution?

What is the solution?

Vario-scale visualization

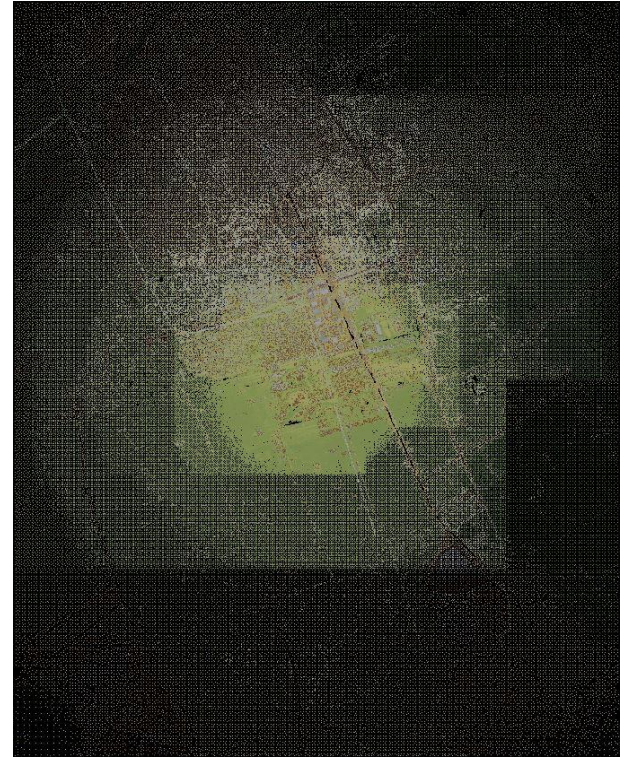
What is the solution?



What is the proposed solution?



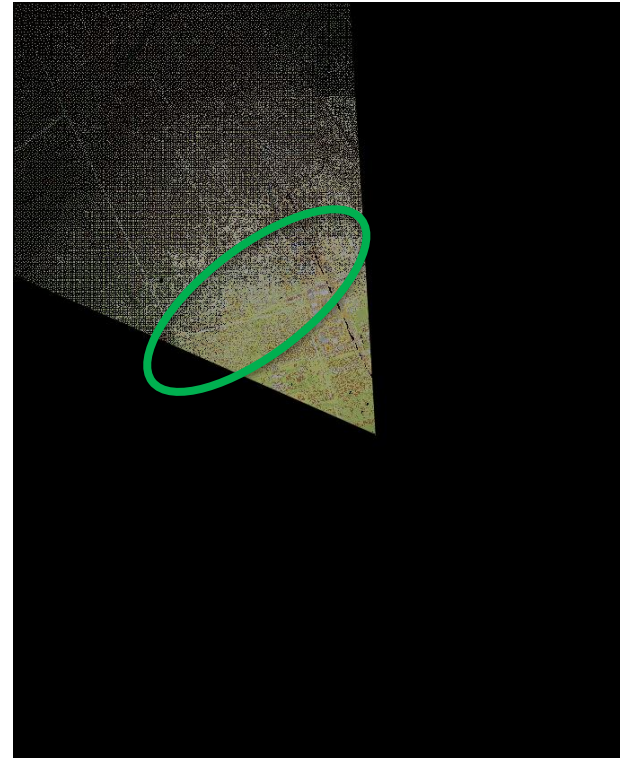
What is the proposed solution?



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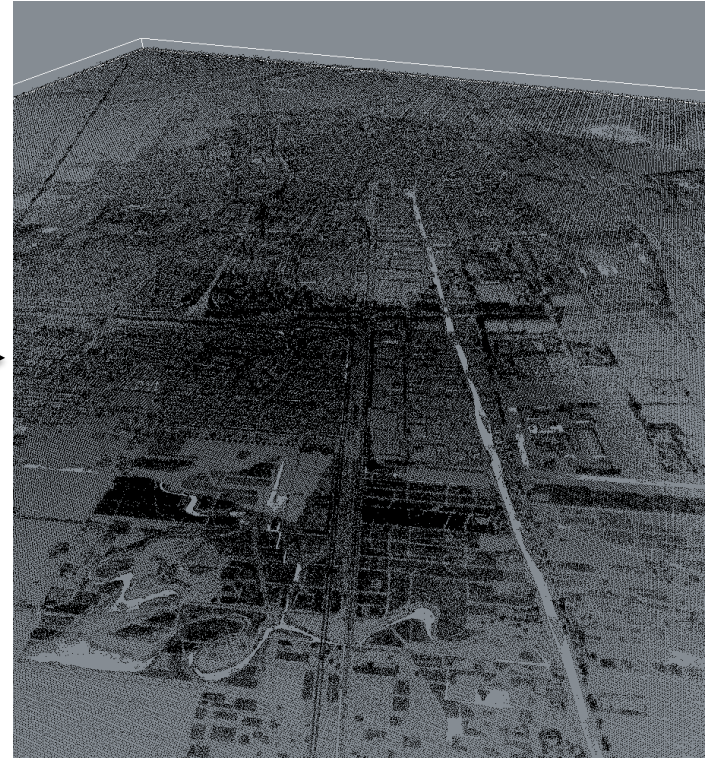
What is the proposed solution?



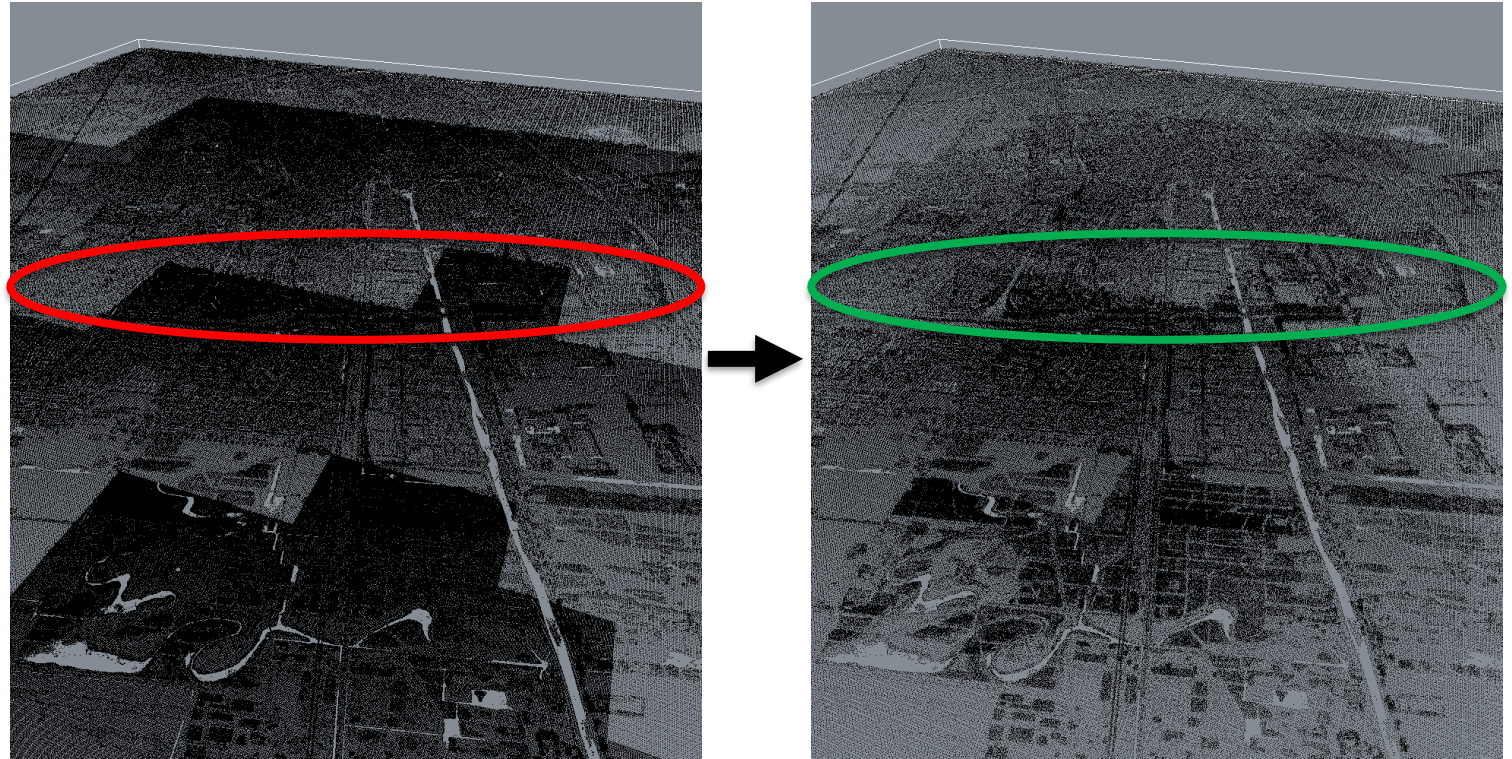
What is the proposed solution?



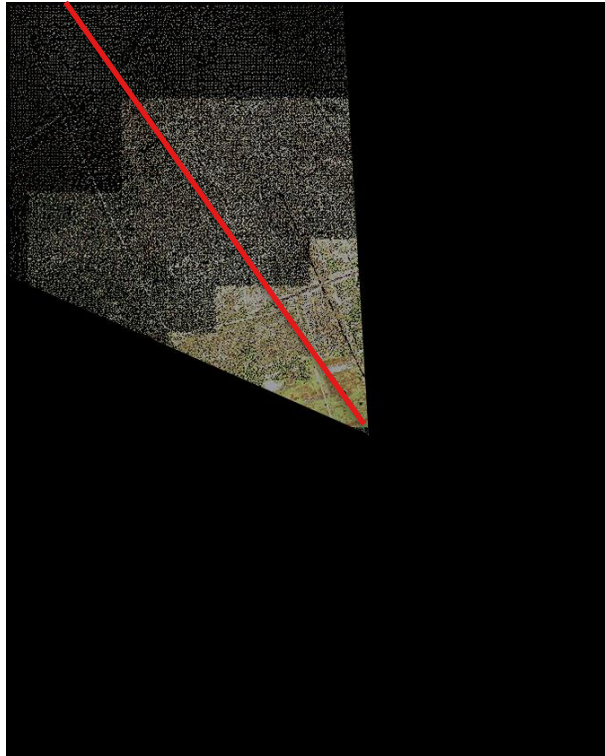
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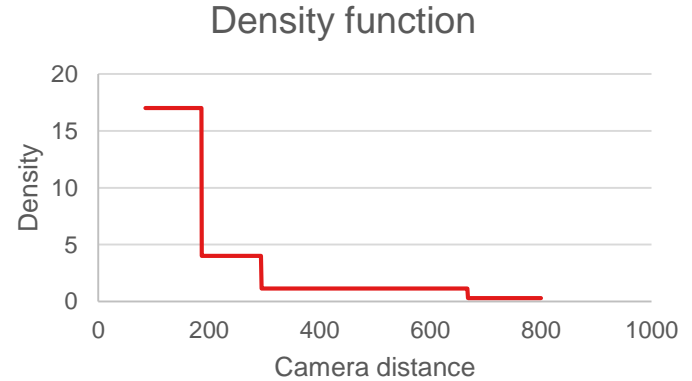
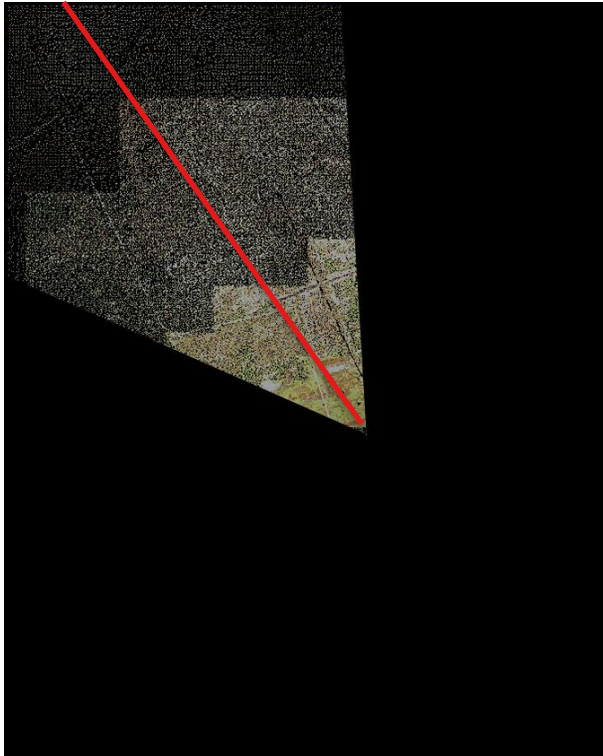
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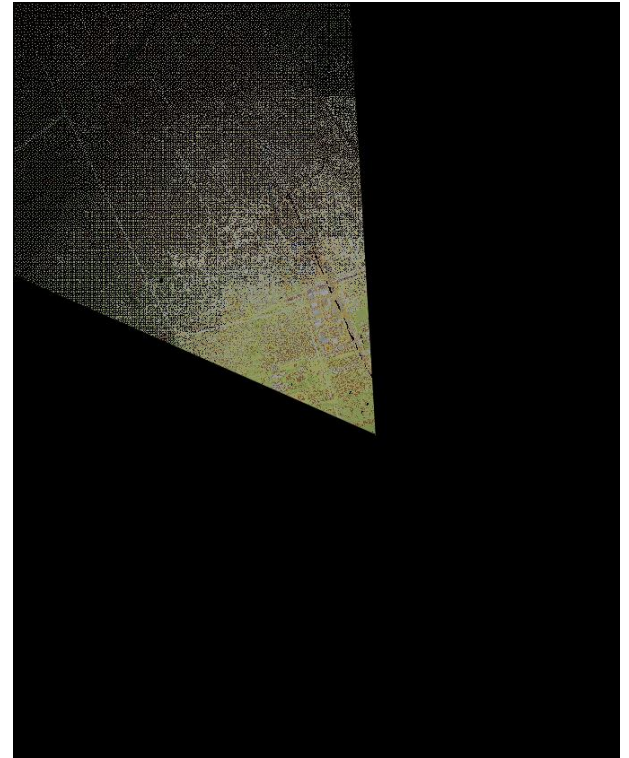
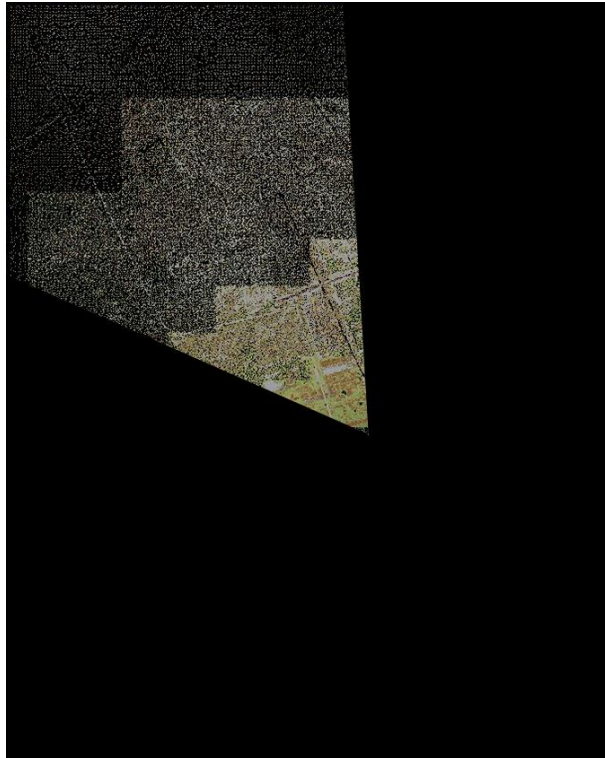
What is the proposed solution?



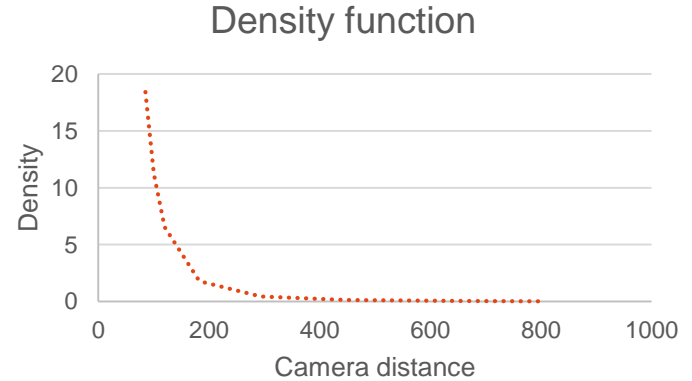
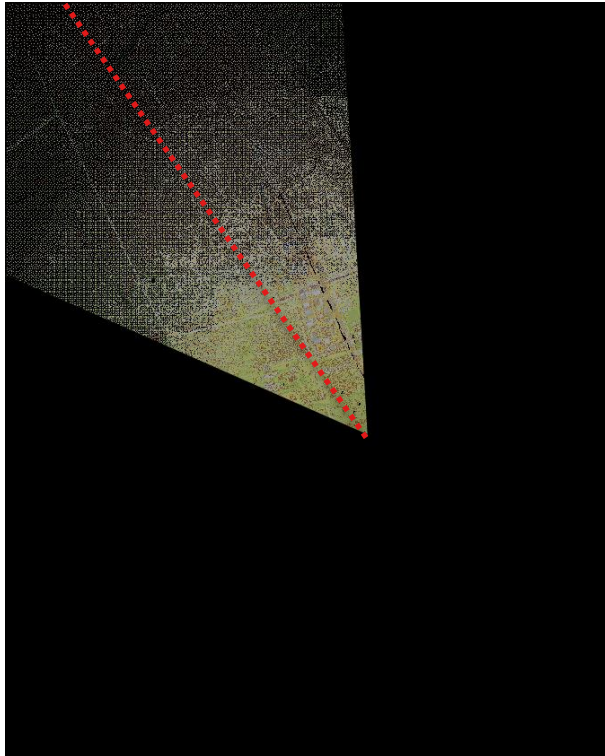
What is the proposed solution?



What is the proposed solution?



What is the proposed solution?



- Introduction
- **Research**
- Theory
- Implementation

Research objective

- To create a vario-scale visualization method for the AHN2 point cloud

Research question

To what extent can a **vario-scale visualization method** be created that eliminates **density jumps** from the **web-based visualization** of the **AHN2 point cloud**?

Supporting research questions

1. To what extent is the current body of research done on the vario-scale visualization of vector data sets relevant for the vario-scale visualization of point cloud data sets?

Supporting research questions

2. To what extent can a theoretical post-processing approach be created for vario-scale visualization of point cloud data sets?

Supporting research questions

3. Which point-cloud processing framework is best suited to create a proof-of-concept vario-scale visualization platform for the AHN2 point cloud?

Supporting research questions

4. To what extent can the theoretical approach be implemented in an existing point cloud web visualization framework?

Presentation

- Introduction
- Research
- Theory
- Implementation

Supporting research question

1. To what extent is the current body of research done on the vario-scale visualization of vector data sets relevant for the vario-scale visualization of point cloud data sets?

Theory – Vector data sets

2002 – Adaptive zooming (Cecconi and Galanda)

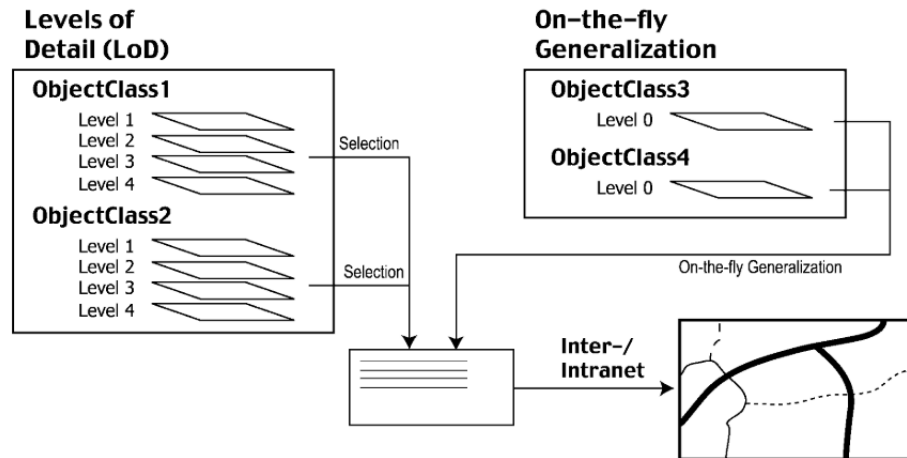
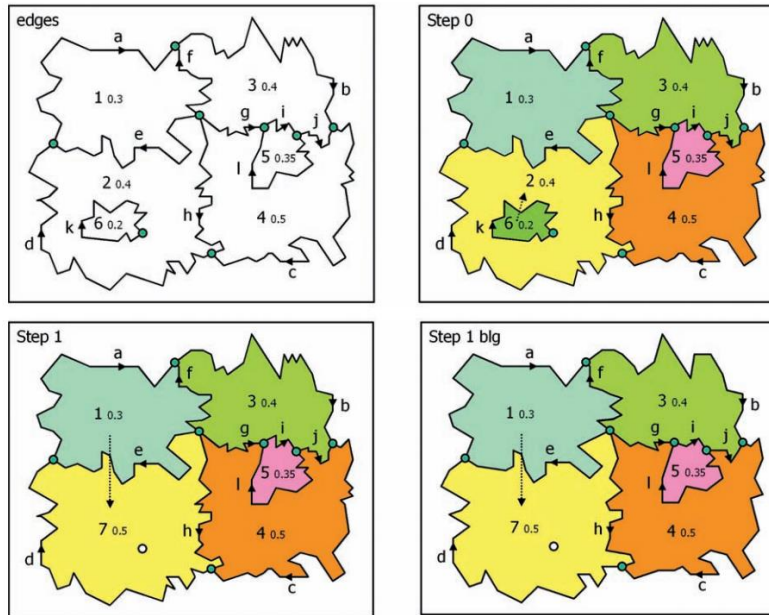


Figure 1: Principle of adaptive zooming and web map generation based on LoD and on-the-fly generalization (Source: Cecconi and Galanda [14]).

Theory – Vector data sets

2005 – Variable scale (van Oosterom)



Theory – Vector data sets

2005 – Variable scale (van Oosterom)

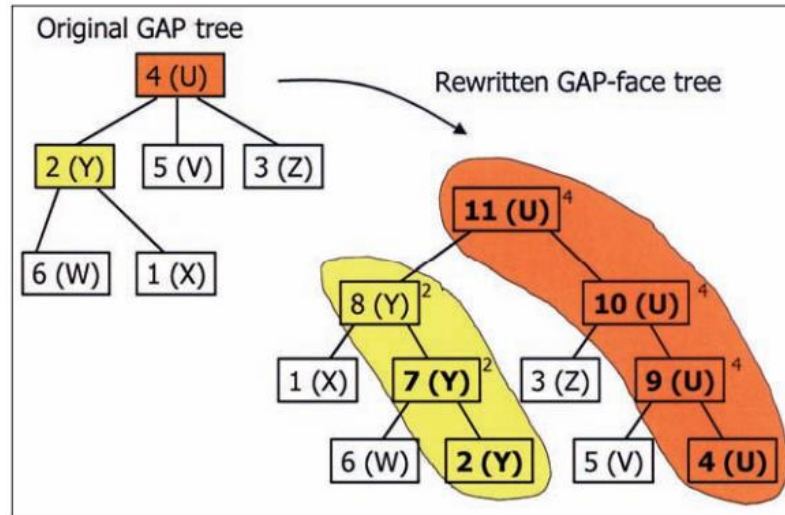


Figure 5. The classic GAP tree rewritten as the GAP-face tree (with a new object Id whenever a face changes and the old object Id appearing in a small font to the upper right of a node). The class is shown in brackets after the object Id.

Theory – Vector data sets

2005 – Variable scale (van Oosterom)

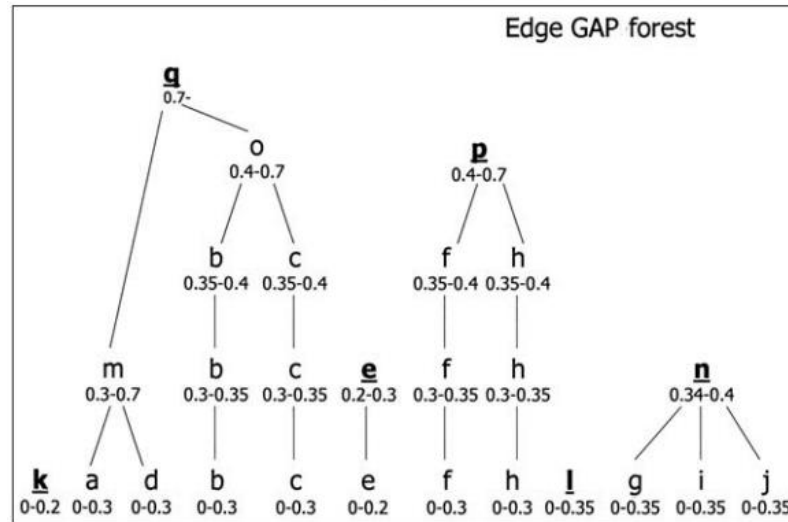
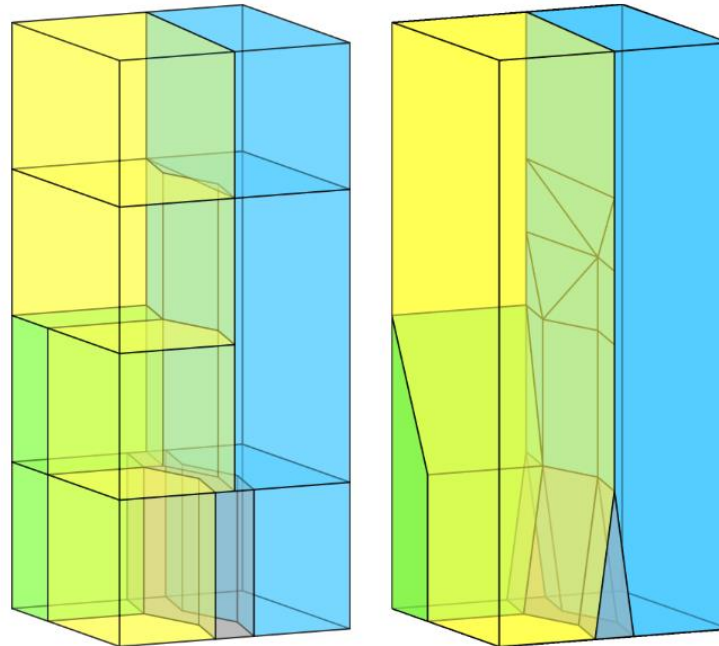


Figure 6. GAP-edge forest (with important ranges). Note that the edges shown in bold and the underlined letters **k**, **q**, **e**, **p**, **l**, and **n** are the roots of the different GAP-edge trees.

Theory – Vector data sets

2013 – Variable scale (Suba)



(a) SSC for classic tGAP

(b) SSC for smooth tGAP

Theory – Point cloud data sets

1. To what extent is the current body of research done on the vario-scale visualization of vector data sets relevant for the vario-scale visualization of point cloud data sets?

Theory – Point cloud data sets

- 2002: Simplification by semantics
- 2005: Simplification by attributes (faces & edges)
- 2013: Continuous simplification in 3 dimensions (X, Y, Scale)

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Theory – Point cloud data sets

Simplification in 3 dimensions (X, Y, camera distance) using the density formula

Theory – Simplification in 3 dimensions

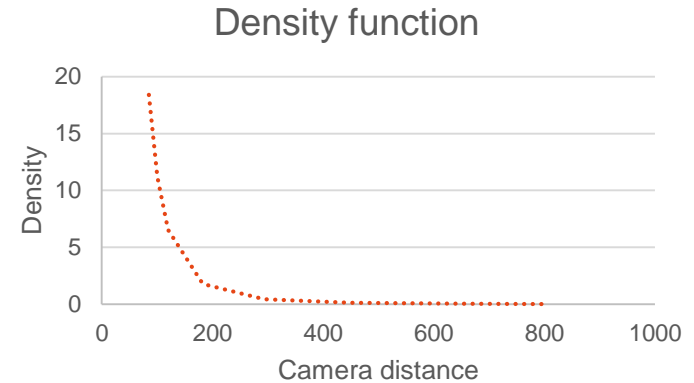
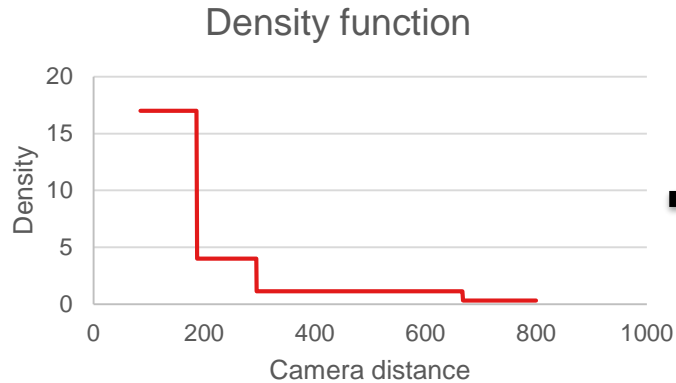
- Camera origin

Theory – Simplification in 3 dimensions

- Camera origin
- Point distance from camera

Theory – Simplification in 3 dimensions

- Camera origin
- Point distance from camera

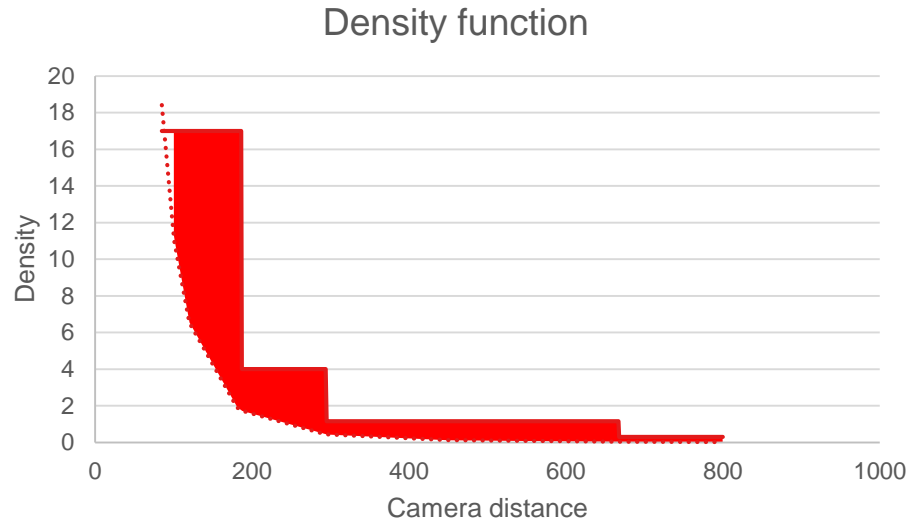


Theory – Simplification in 3 dimensions

- Perform a per-point evaluation to determine whether the point should be rendered or not

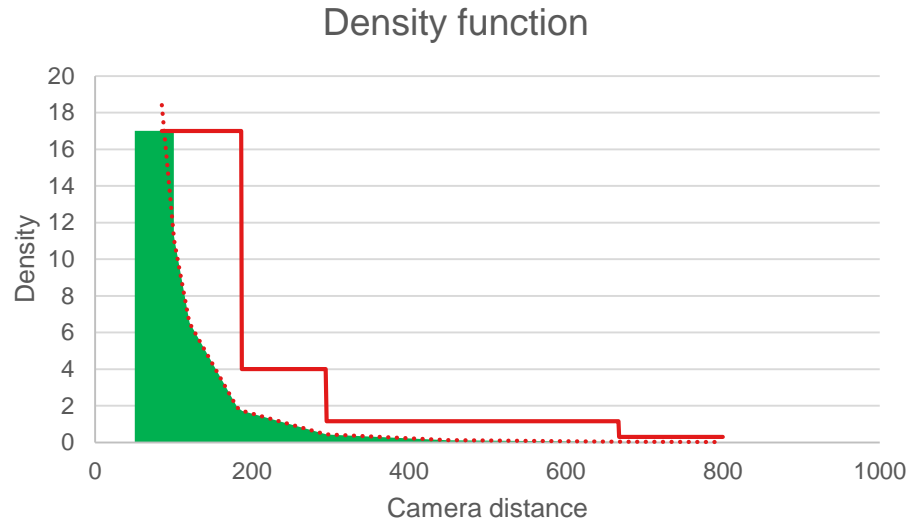
Theory – Simplification in 3 dimensions

- Perform a per-point evaluation



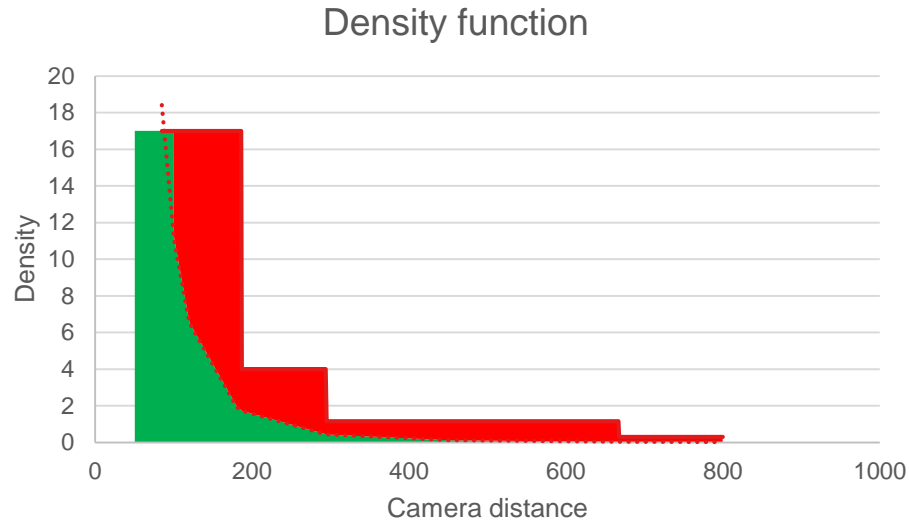
Theory – Simplification in 3 dimensions

- Perform a per-point evaluation



Theory – density function

- We know what we want to keep and what we want to remove



Theory – density function

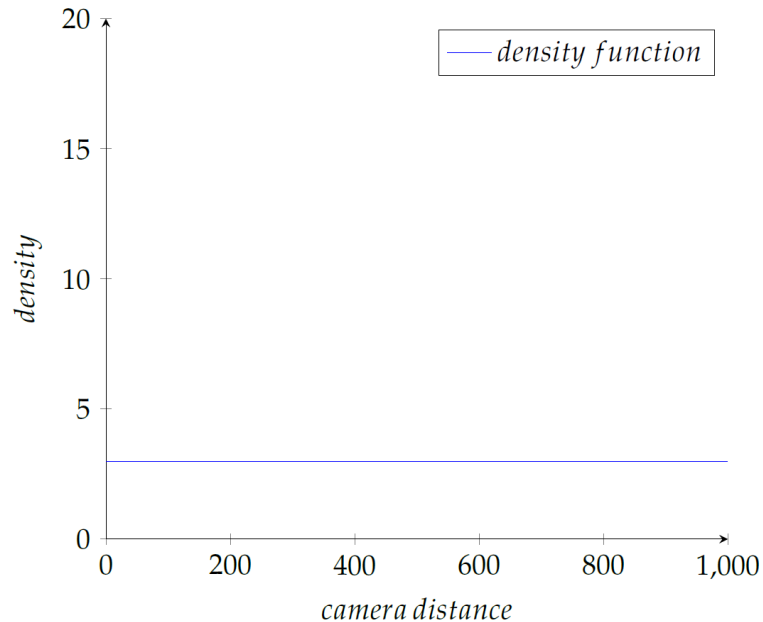
- We know what we want to keep and what we want to remove
- How do we know the density function?

Theory – basic density function

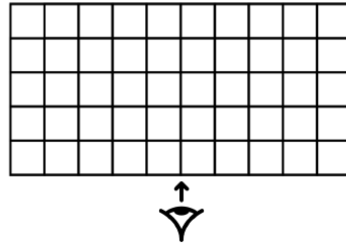
- Lets start with a basic example

Theory – basic density function

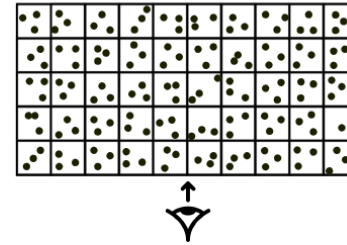
- Lets start with a basic example



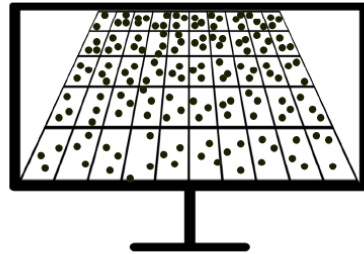
Theory – basic density function



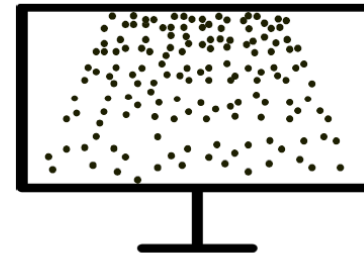
(a) 2D spatial extent of LiDAR data set, with density grid



(b) 2D spatial extents, with a density of 3 points per $n * n$ units



(c) Perspective translation of point cloud with a density of 3 points per $n * n$ units



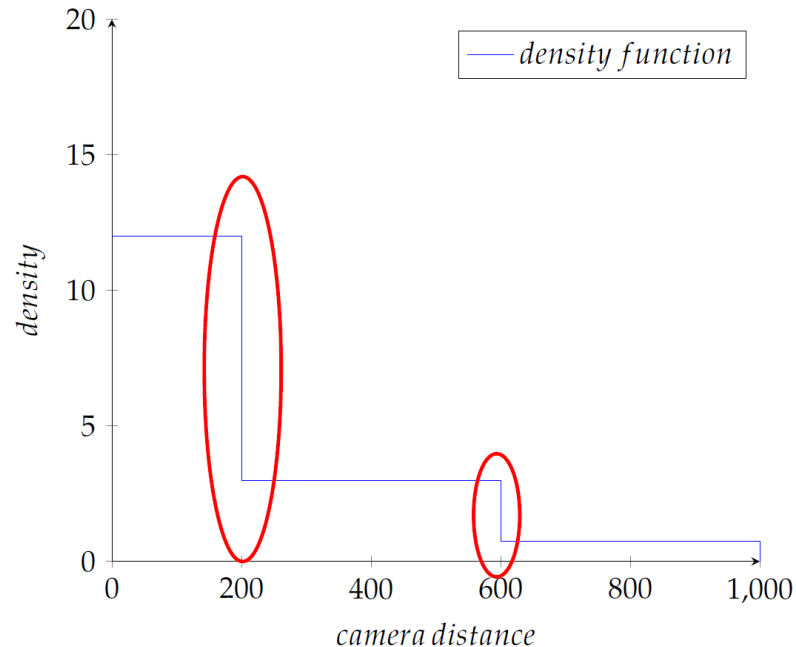
(d) User's visualization of the point cloud

Theory – octree density function

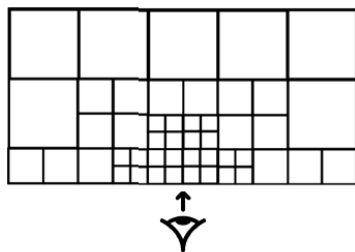
- The current situation

Theory – octree density function

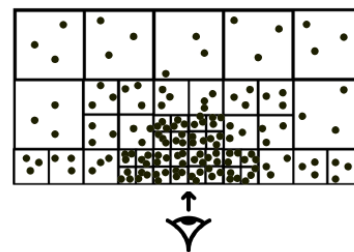
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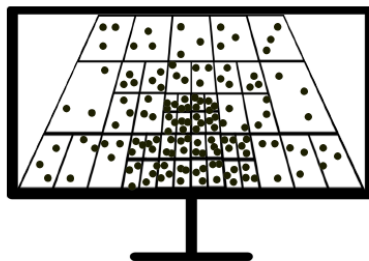
Theory – octree density function



(a) 2D spatial extents, overlaid with an Octree selection grid



(b) 2D spatial extents, with an increased density closer to the camera

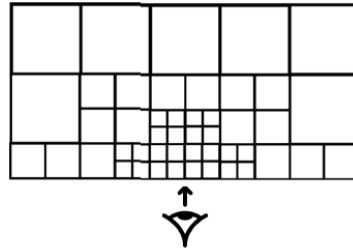


(c) Perspective translation of point cloud using octree selection grid

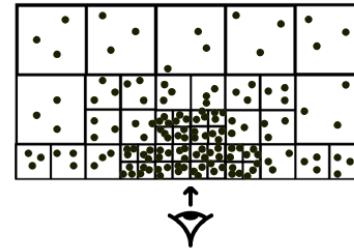


(d) User's visualization of the point cloud

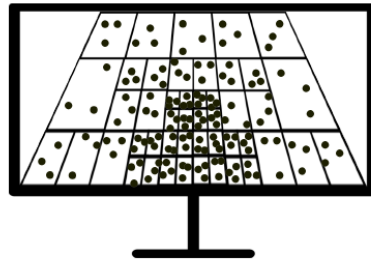
Theory – octree density function



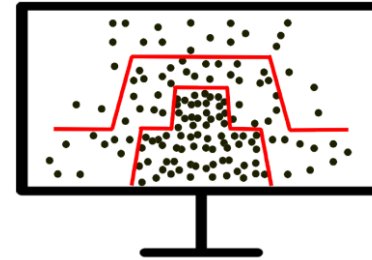
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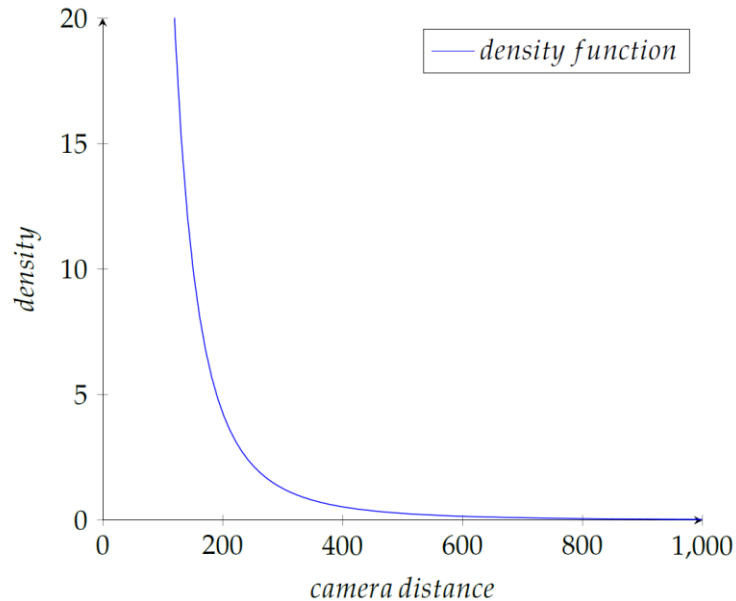
(d) User's visualization of the point cloud

Theory – vario-scale density function

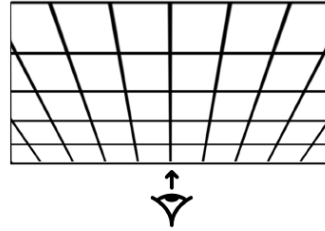
- The desired situation

Theory – vario-scale density function

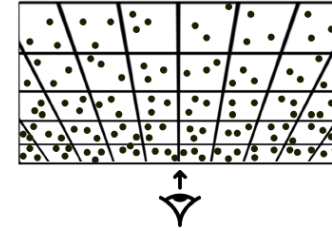
- The desired situation



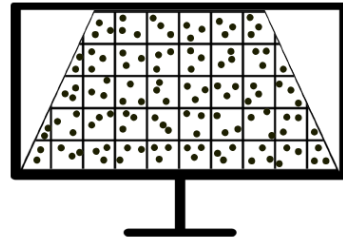
Theory – vario-scale density function



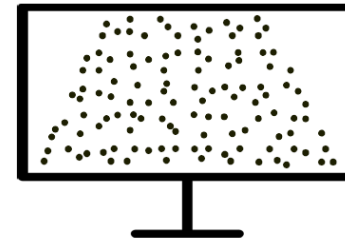
(a) 2D spatial extents, overlaid with the inverse density grid



(b) 2D spatial extents, with the inverse density grid filled with points



(c) Perspective translation of point cloud using the perspective matrix translation



(d) User's visualization of the point cloud

Theory – vario-scale density function

- The desired situation
- A global vario-scale function

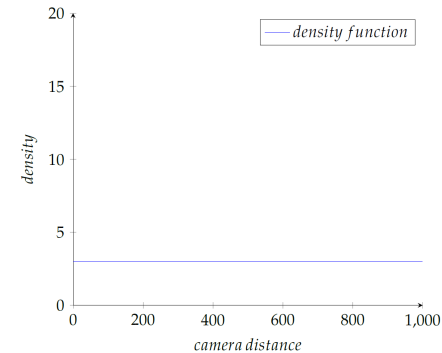
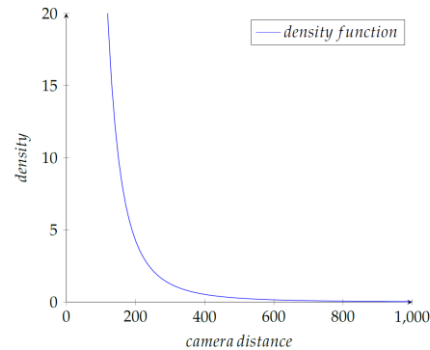
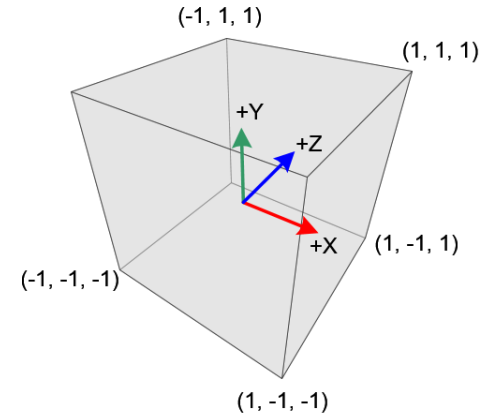
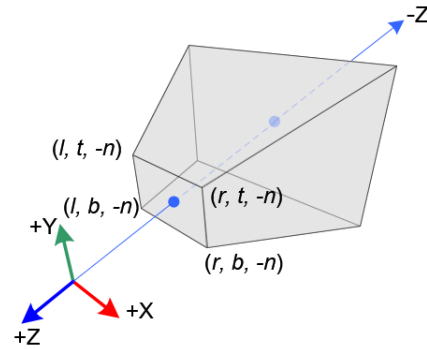
Theory – vario-scale density function

- The desired situation
- A global vario-scale function
- Dependant on a translation through the perspective matrix

Theory – vario-scale density function

- The desired situation
- A global vario-scale function
- Dependant on a translation through the perspective matrix
- Dependant on
 - Near-plane
 - Far-plane
 - Field of View (FoV)

Theory – vario-scale density function



Theory – vario-scale density function

- Finding how the density is transformed through the perspective translation matrix is out of the scope of this thesis

Theory – vario-scale density function

- Finding how the density is transformed through the perspective translation matrix is out of the scope of this thesis
- The results presented are created by using a function specific to each frame

Theory

- We know the density function
- We know what to keep and what to remove
- How do we do this?

Supporting research question

2. To what extent can a theoretical post-processing approach be created for vario-scale visualization of point cloud data sets?

Different approaches

- Random point removal
- Filtering bands point removal
- Circle packing point removal

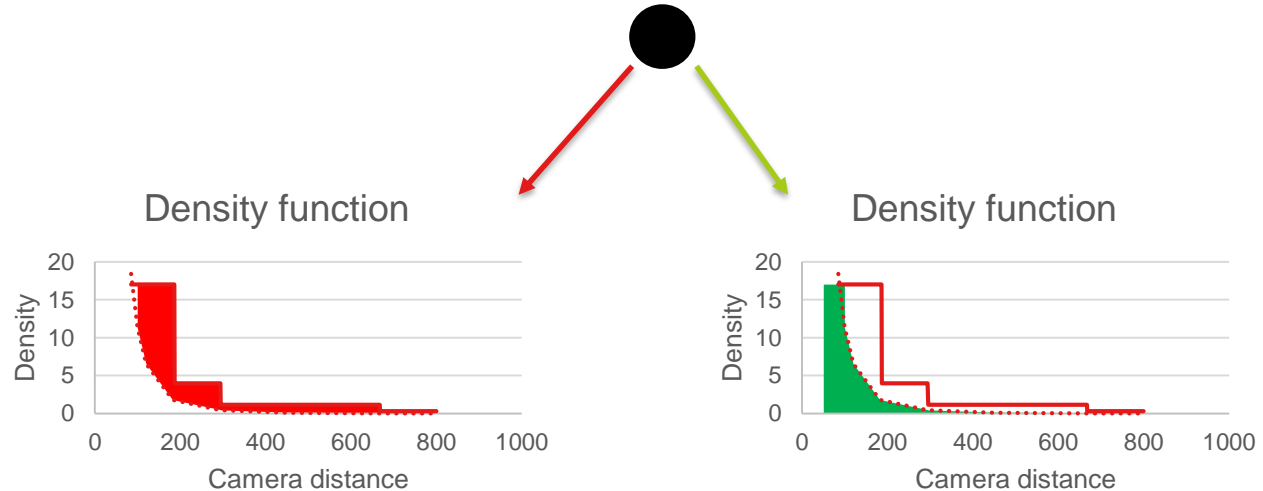
Random point removal

Random point removal

- Determine random value per point

Random point removal

- Determine random value per point
- Assed to threshold value



Random point removal

Algorithm 1 Random removal

```
import numpy as np
import random
import scipy

def random_removal(points, camera_parameters, density_function):
    """
    """
    camera_origin = camera_parameters.origin
    selected_points = set()
    kdtree = scipy.spatial.KDTree(points)

    for point in points:
        d = distance_from_camera(point, camera_origin)
        r = random(0,1)
        local_density = len(kdtree.query_ball_point(point, 0.5642))
        t = density_function(d) / local_density

        if r >= t:
            continue
        if r < t:
            selected_points.add(point)

    return selected_points
```

Random point removal

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    return selected_points
```

Random point removal - advantage

- Random selection

Random point removal - disadvantage

- Creation of global kd-tree to compute local density

Random point removal - disadvantage

- Creation of global kd-tree to compute local density
- Computationally heavy because of per point computations:
 - Distance to the camera
 - Random value
 - Local density
 - Threshold

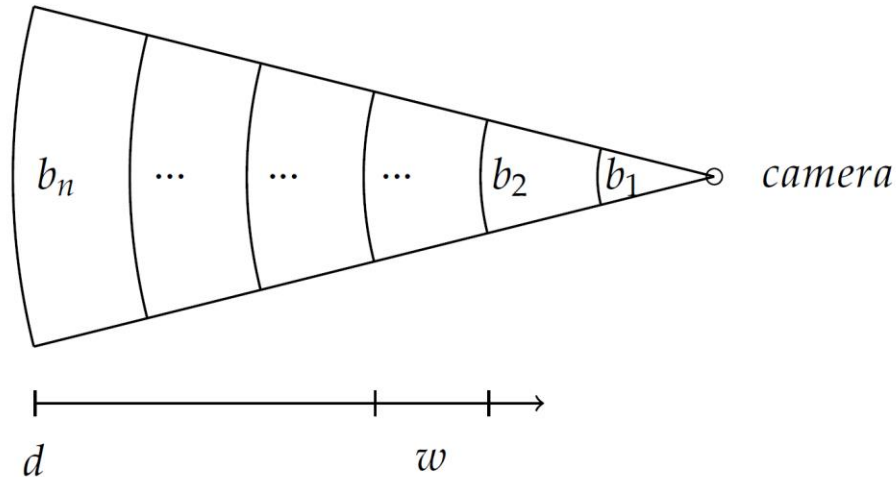
Filtering bands point removal

Filtering bands point removal

- Create bands extending outward from camera origin

Filtering bands point removal

- Create bands extending outward from camera origin



Filtering bands point removal

- Create bands extending outwards from camera origin
- Filter points on local density

Filtering bands point removal

Algorithm 2 Filtering bands

```
def filtering_bands(points ,
                    camera_parameters ,
                    density_function ,
                    width = 1.0):
    """
    """
    distance = camera_parameters.distance
    origin = camera_parameters.origin
    selected_points = set()

    for i in range(distance / width):
        band_radius_center = i * width + 0.5 * width
        local_points_array = points_in_band(origin ,
                                           band_radius_center ,
                                           points)
        allowed_points = density_function(band_radius_center)

        selected_points.add(local_points_array[allowed_points])

    return selected_points
```

Filtering bands - advantage

- No local density calculation per point

Filtering bands - advantage

- No local density calculation per point
- Per band density calculation

Filtering bands - advantage

- No local density calculation per point
- Per band density calculation
- No global computation needed such as kd-tree building

Filtering bands - disadvantage

- (Small) discrete steps

Filtering bands - disadvantage

- (Small) discrete steps
- Requires band creation

Filtering bands - disadvantage

- (Small) discrete steps
- Requires band creation
- Per band clipping operation

```
for i in range(distance / width):
    band_radius_center = i * width + 0.5 * width
    local_points_array = points_in_band(origin,
                                       band_radius_center,
                                       points)
    allowed_points = density_function(band_radius_center)

    selected_points.add(local_points_array[allowed_points])

return selected_points
```

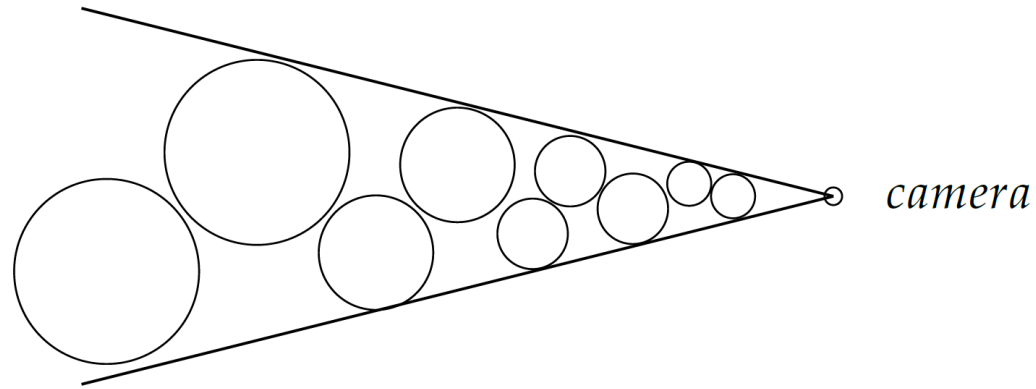
Circle packing point removal

Circle packing point removal

- Determine region in which no two points can exist

Circle packing point removal

- Determine region in which no two points can exist



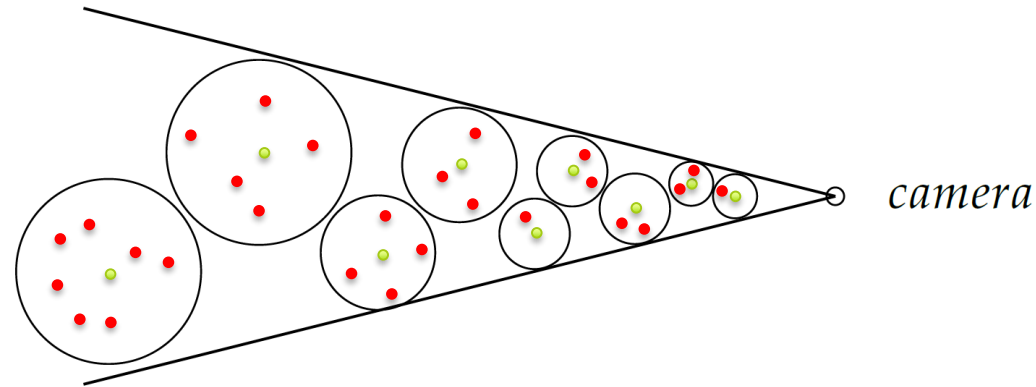
Circle packing point removal

- Determine region in which no two points can exist
- Discard all neighbours in that region as potential points

Circle packing point removal

- Determine region in which no two points can exist
- Discard all neighbours in that region as potential points to evaluate
- Based on the circle packing study of arrangement of circles in a given surface, so that no overlap occurs

Circle packing point removal



- Based on the circle packing study of arrangement of circles in a given surface, so that no overlap occurs

Circle packing point removal

Algorithm 3 Circle packing

```
import scipy

def circle_packing(points, camera_parameters, density_function):
    """
    """
    used_points = [False] * len(points)
    selected_points = set()

    kdtree = scipy.spatial.KDTree(points)

    for j, point in enumerate(points):
        if used[j] == True:
            continue

        d = distance_from_camera(point, camera_parameters)
        r = radius_function(density_function(d))
        nn = kdtree.query_ball_point(point, r)

        for i in nn:
            used[i] = True

        selected_points.add(point)

    return selected_points
```

Circle packing - advantage

- Only computations needed for accepted points

Circle packing - advantage

- Only computations needed for accepted points

```
for j, point in enumerate(points):
    if used(j) == True:
        continue

    d = distance_from_camera(point, camera_parameters)
    r = radius_function(density_function(d))
    nn = kdtree.query_ball_point(point, r)

    for i in nn:
        used[i] = True

    selected_points.add(point)

return selected_points
```


Circle packing - disadvantage

- Requires the creation of a kd-tree for fast nn-search

Theoretic summary

- Determine the density function
- Eliminate points according to this density function
- Using one of the three described methods

Presentation

- Introduction
- Research
- Theory
- **Implementation**

Supporting research question

3. Which point-cloud processing framework is best suited to create a proof-of-concept vario-scale visualization platform for the AHN2 point cloud?

Web based visualization framework

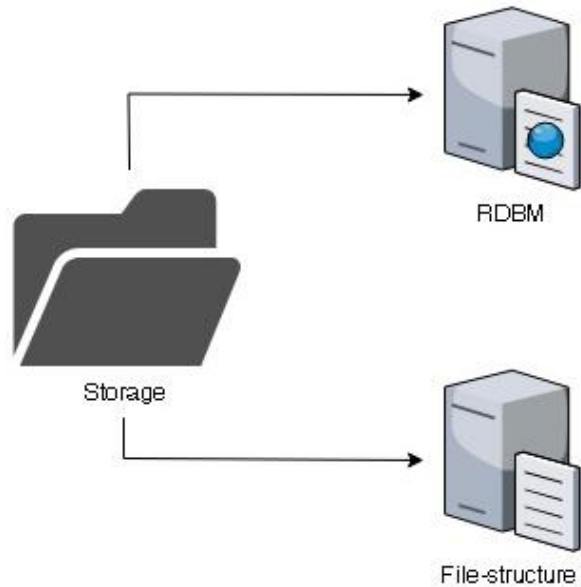
How does web based point cloud visualization currently work?

Web based visualization framework

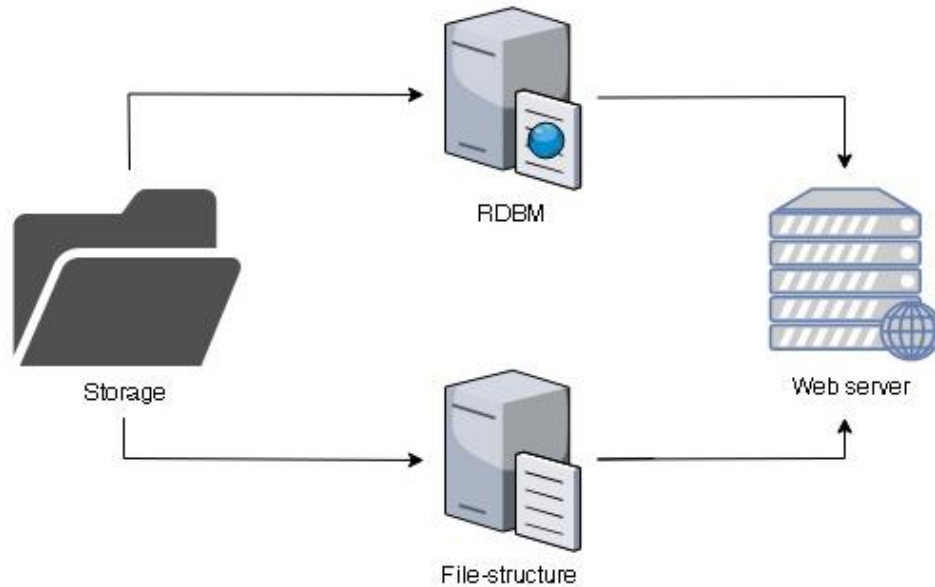


Storage

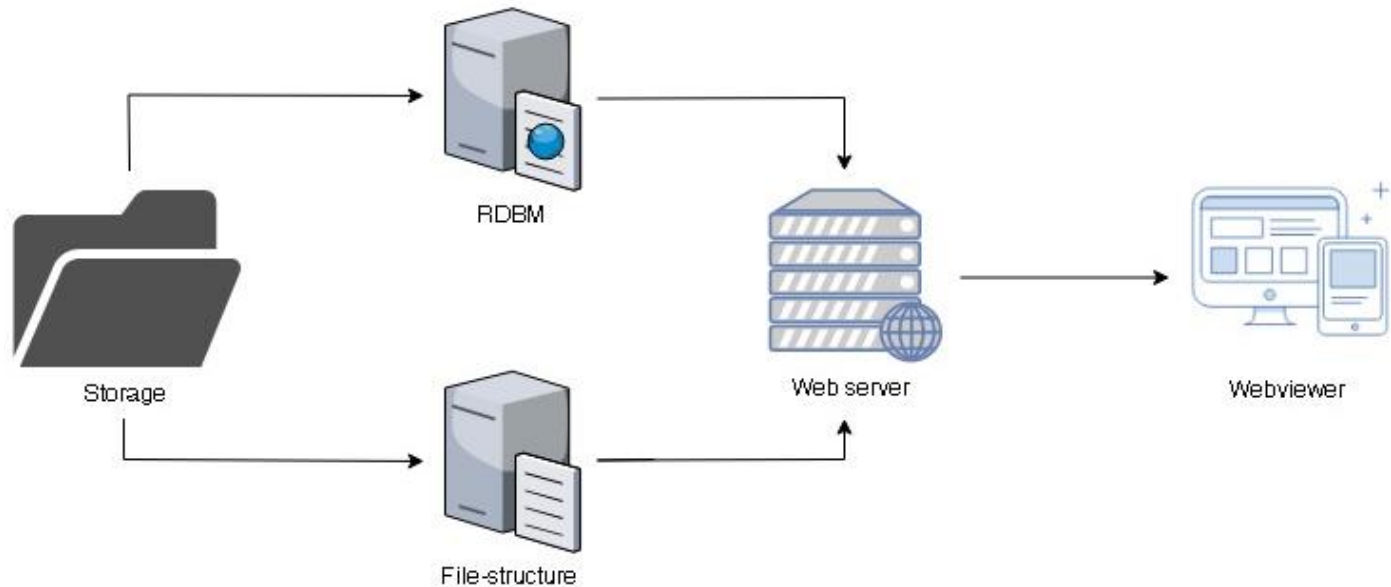
Web based visualization framework



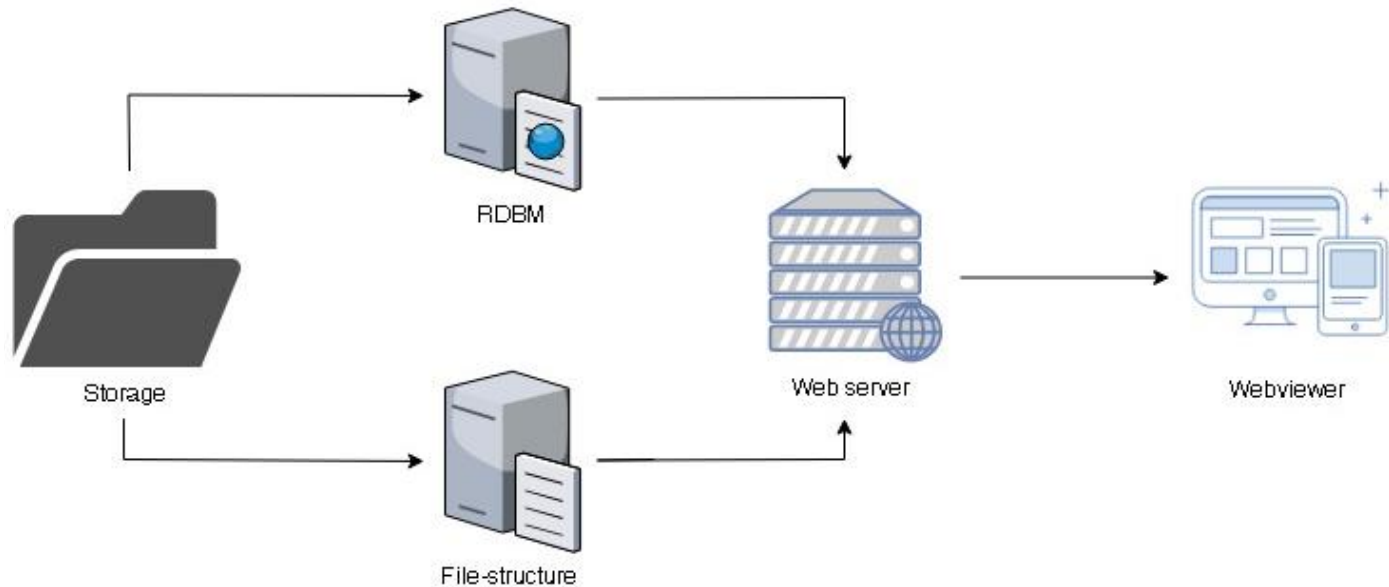
Web based visualization framework



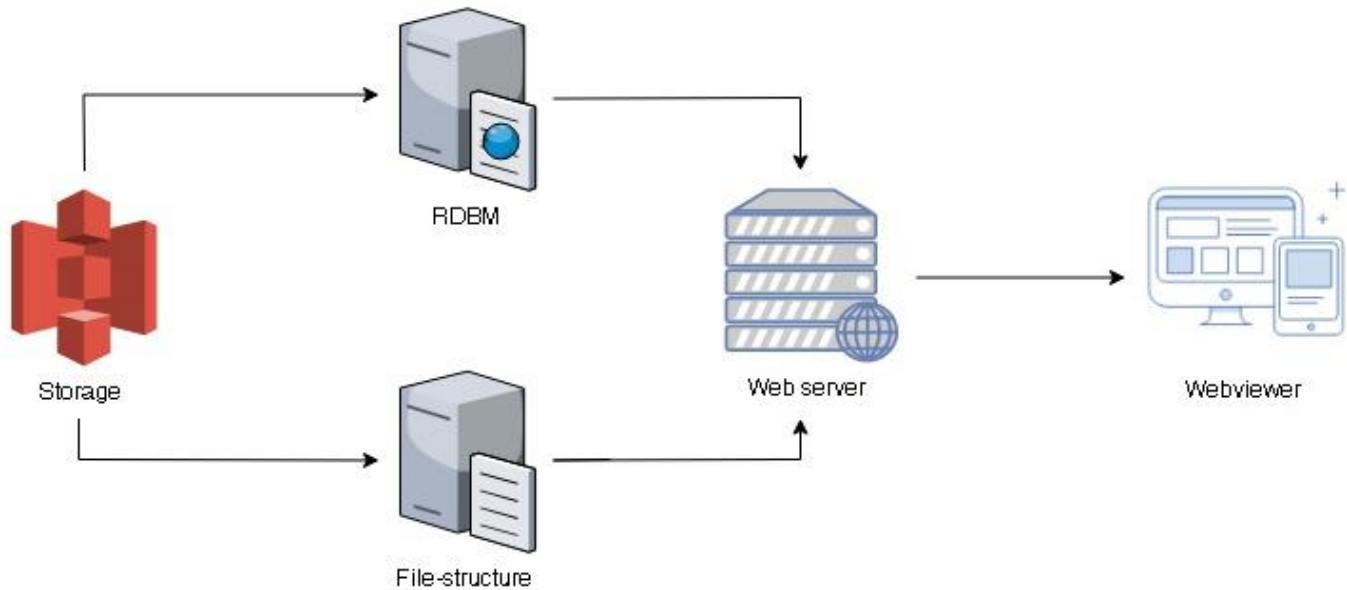
Web based visualization framework



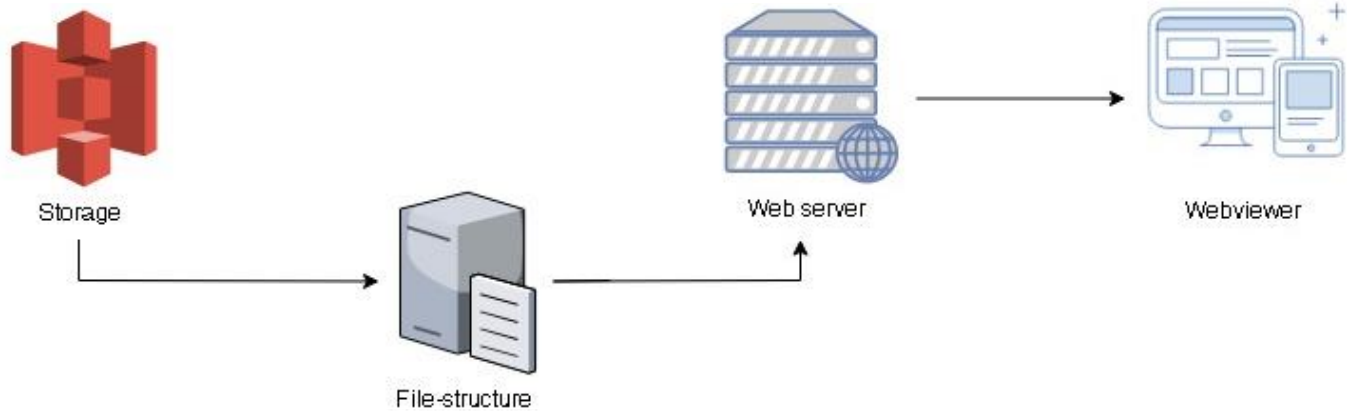
Chosen framework



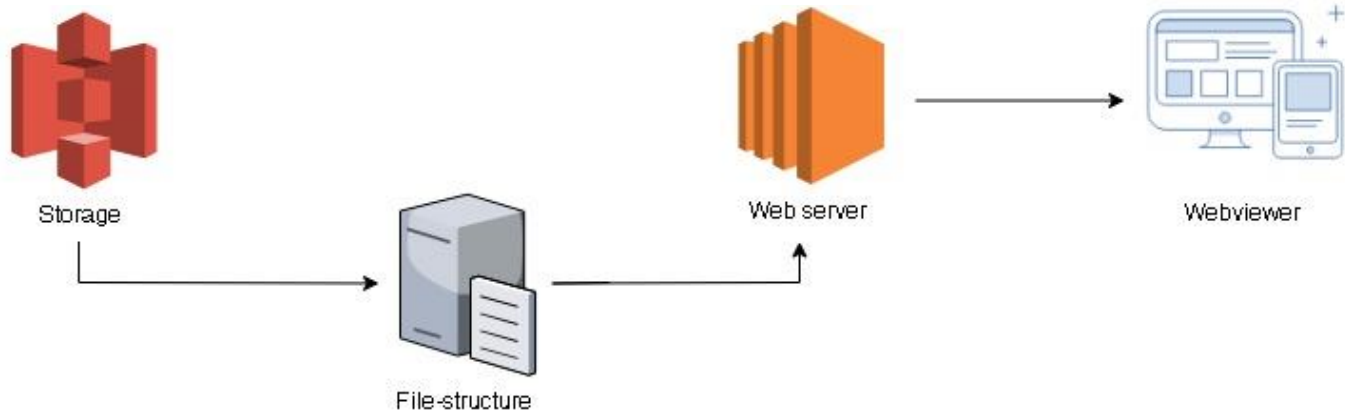
Chosen framework



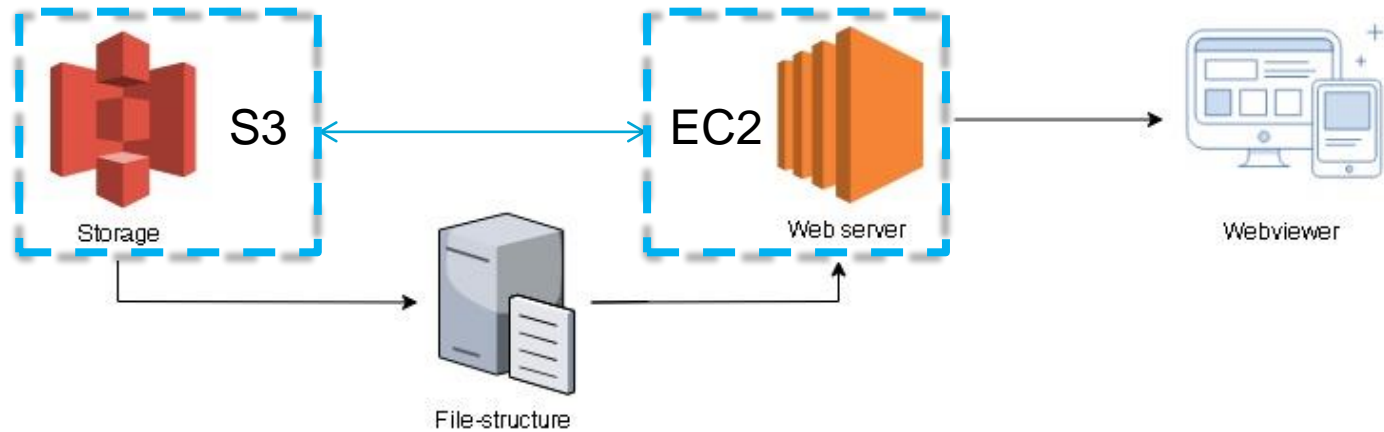
Chosen framework



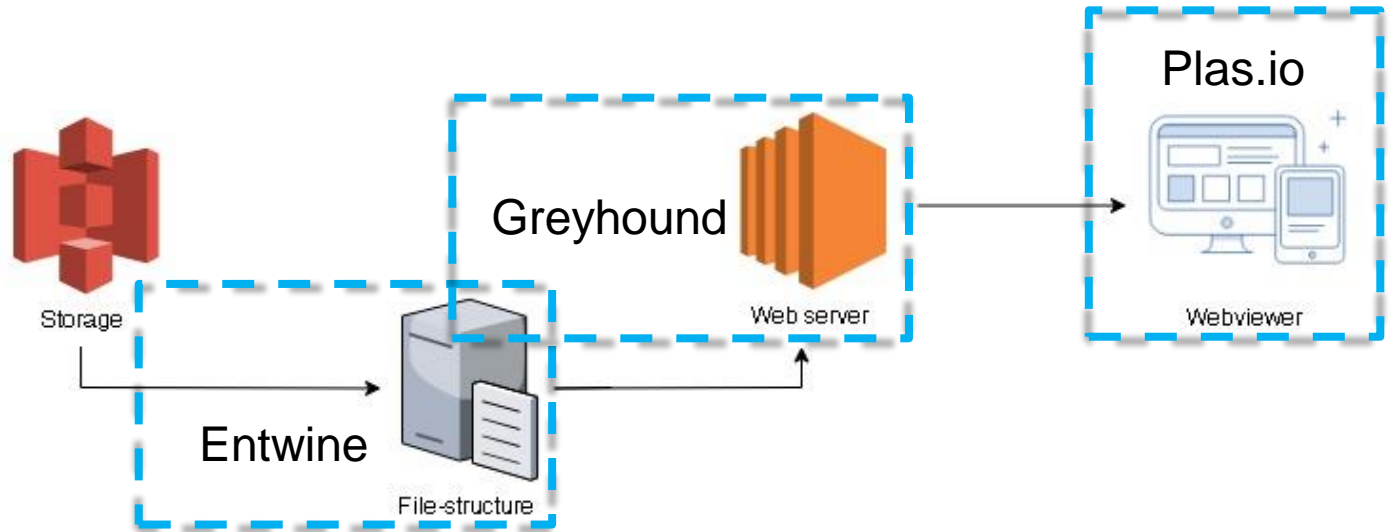
Chosen framework



Chosen framework



Chosen framework

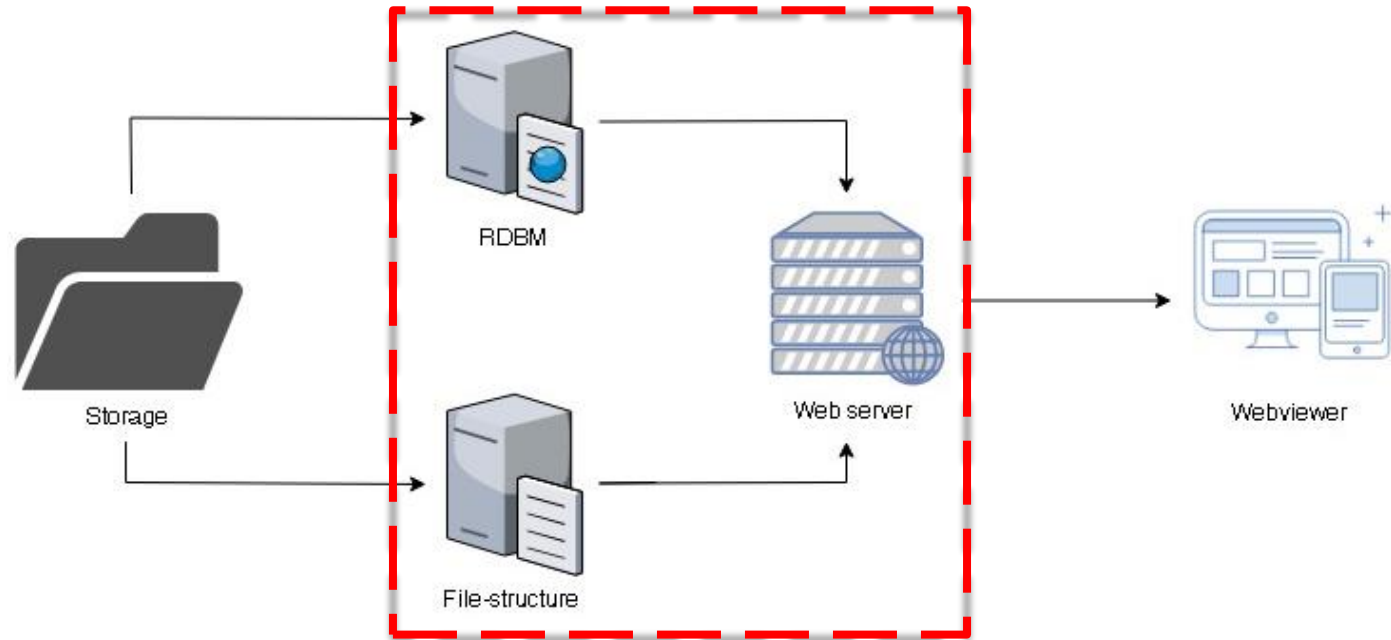


Supporting research questions

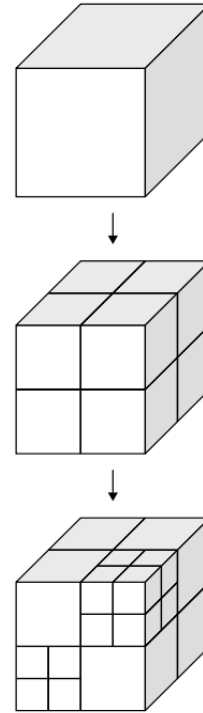
4. To what extent cant the theoretical approach be implemented in an existing point cloud web visualization framework?

Where is the issue?

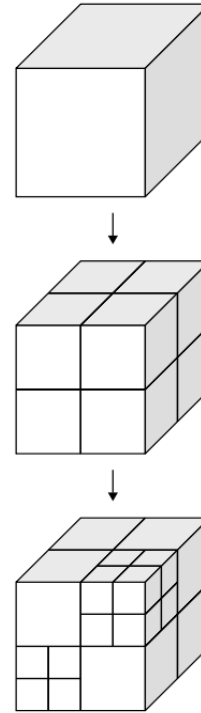
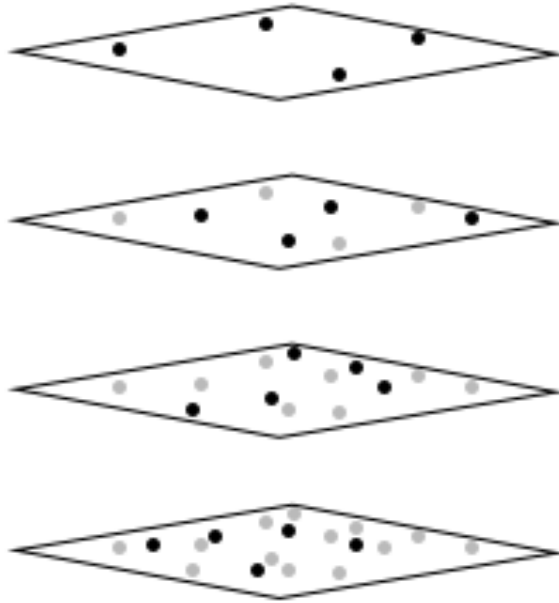
Where is the issue?



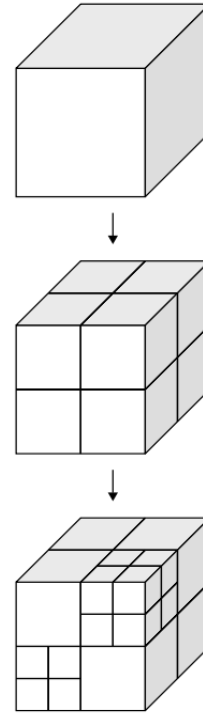
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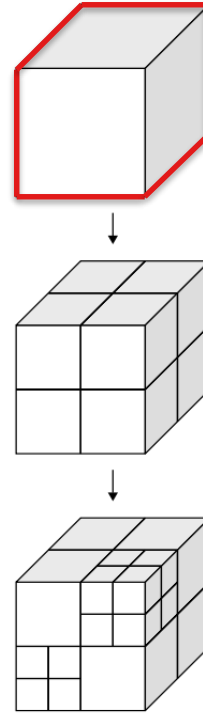
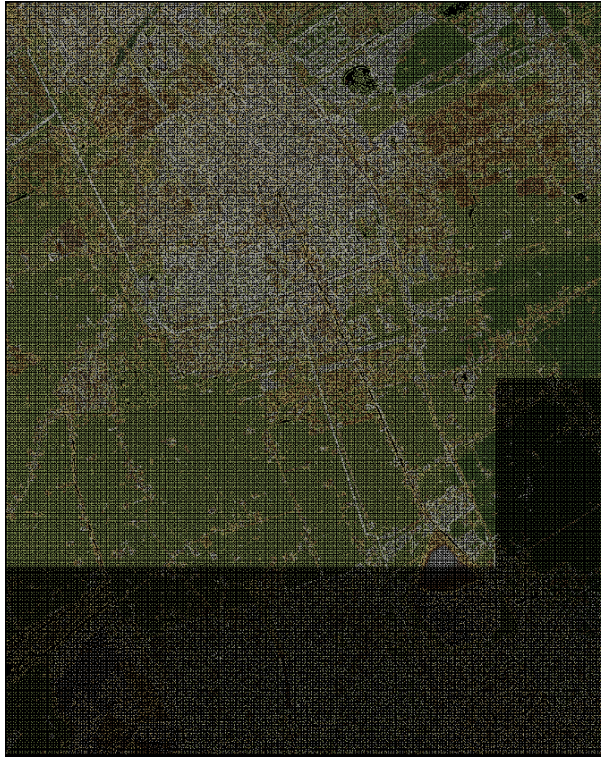
Where is the issue?



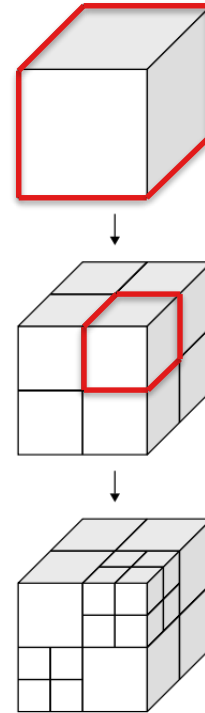
Where is the issue?



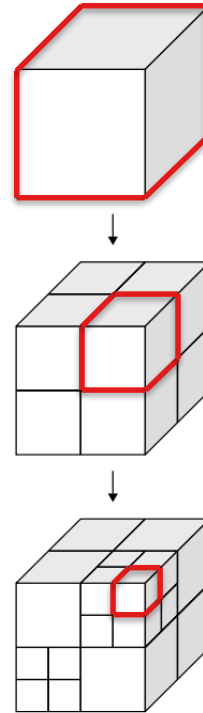
Where is the issue?



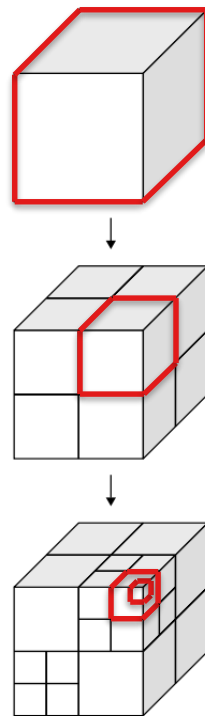
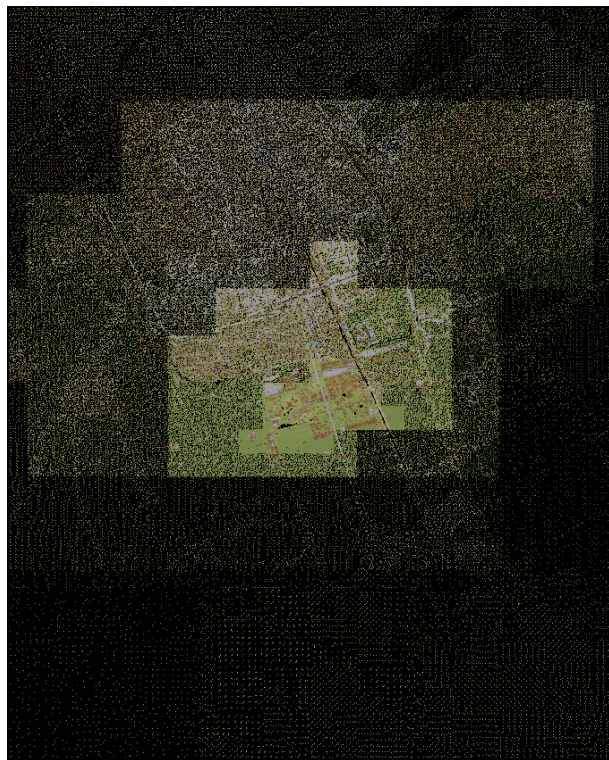
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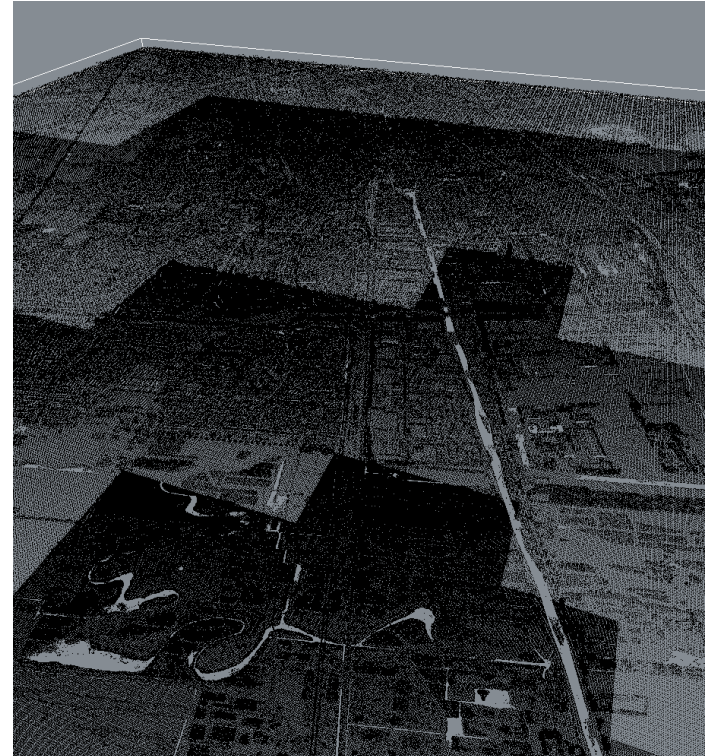
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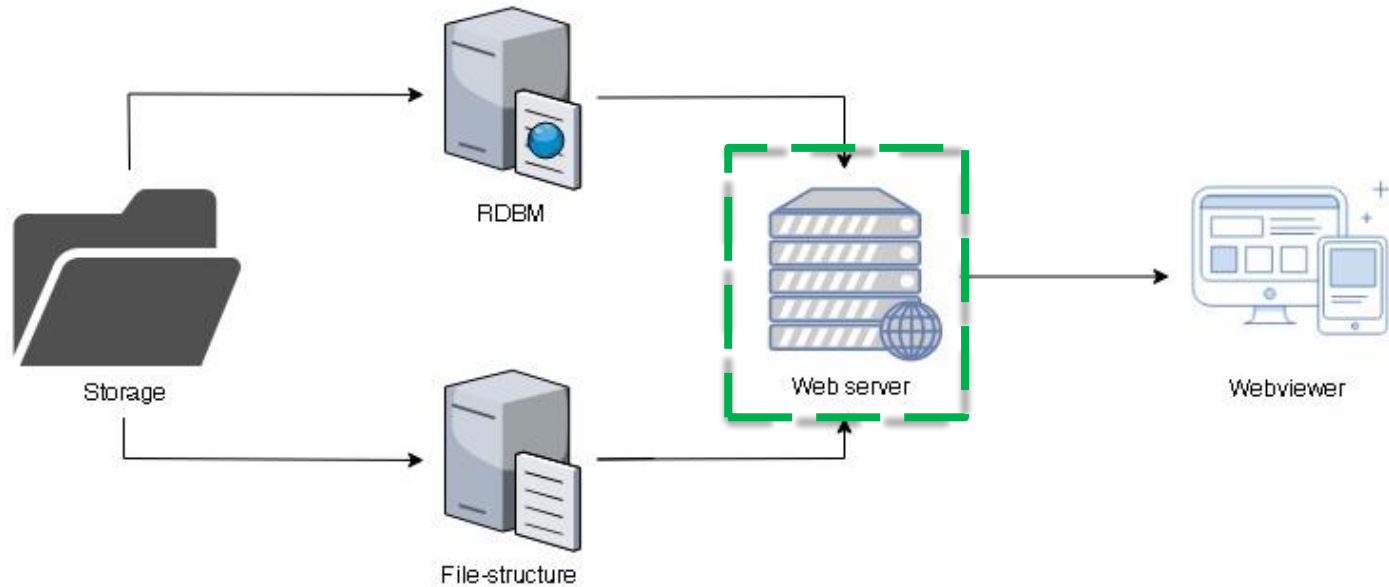


Where is the issue?

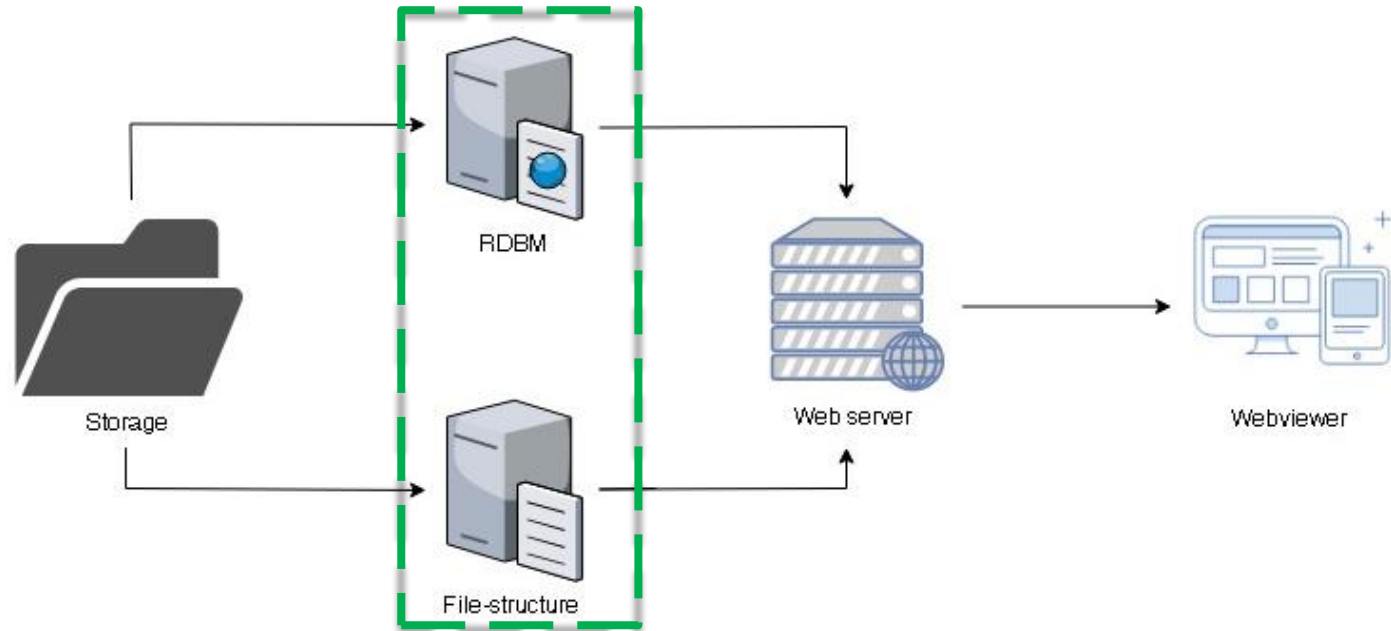


Where is the solution?

Where is the solution?



Why not here?



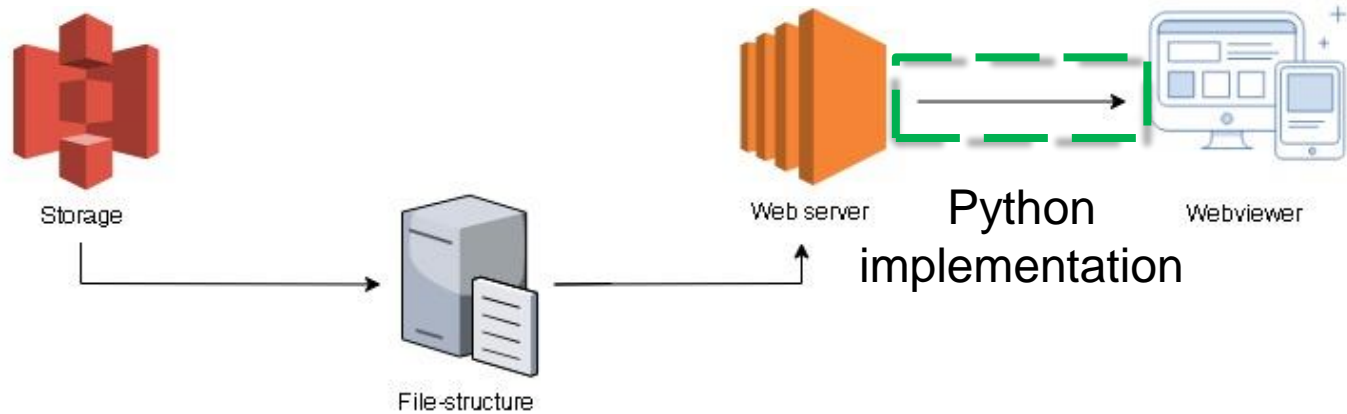
Why not here?

- Current octree index most efficient for web based querying

Why not here?

- Current octree index most efficient for web based querying
- Creating a new indexing method for vario-scale visualization of point clouds is outside of the scope of this thesis

Method



Python implementation

- Circle packing method

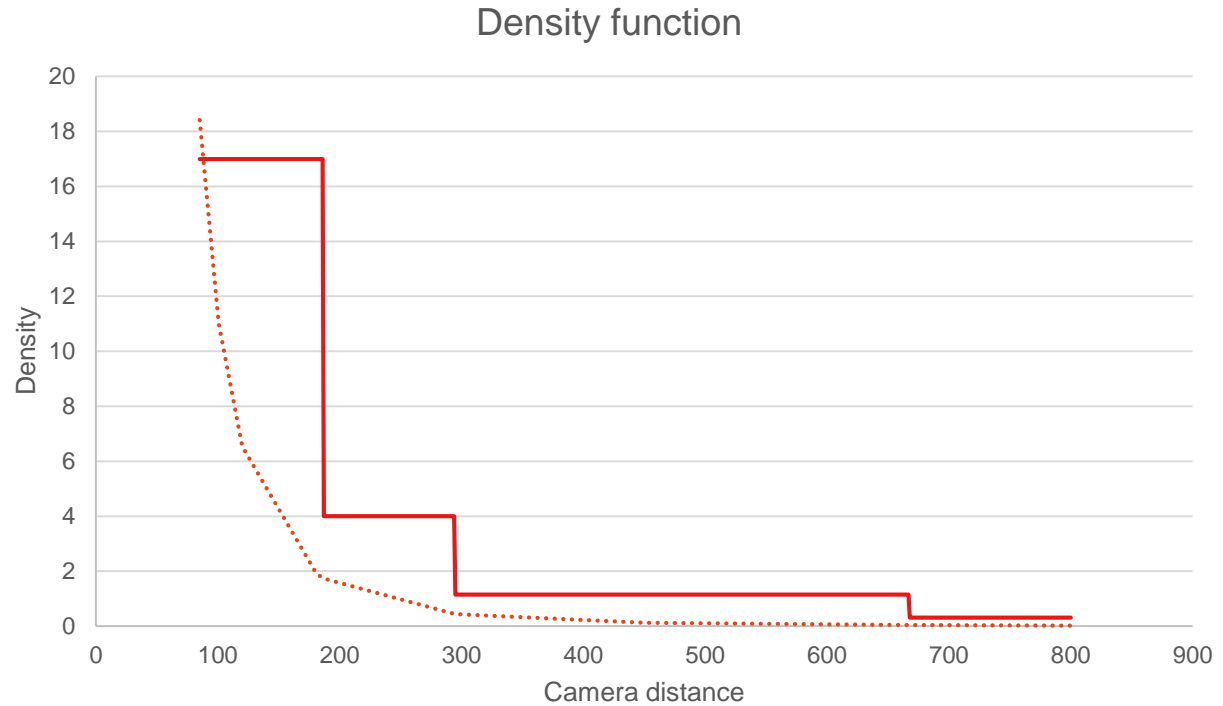
Python implementation

- Circle packing method
- Density function to match the frame's density jumps

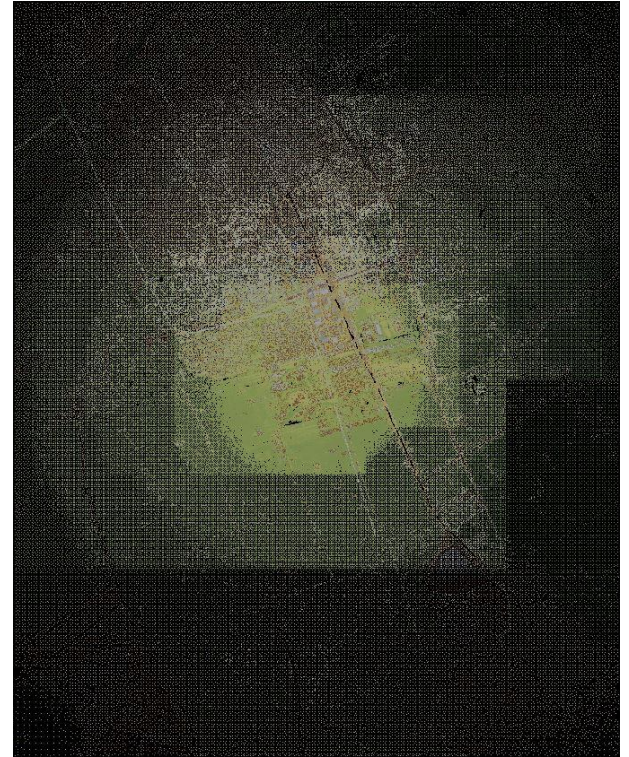
Python implementation

- Circle packing method
- Density function to match the frame's density jumps
- Density formula =
$$1/(0,7*(0,005*camera_distance)^3)$$

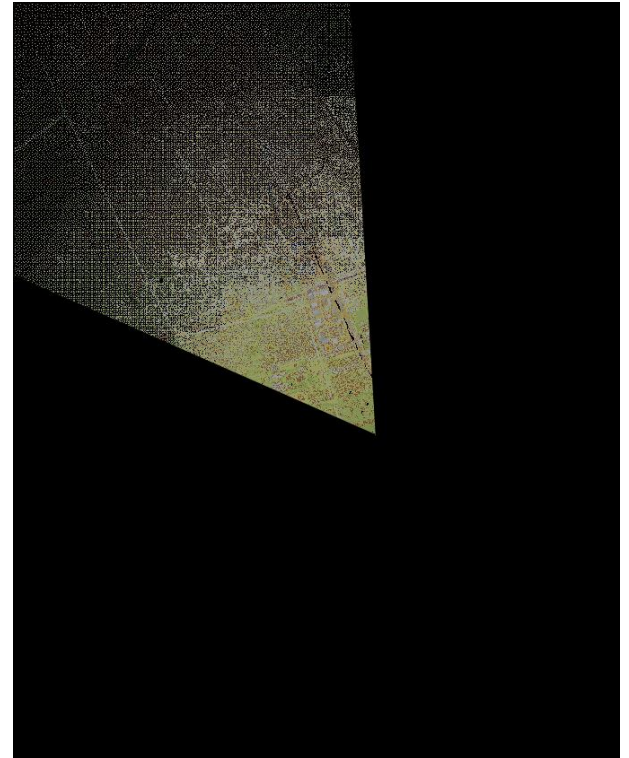
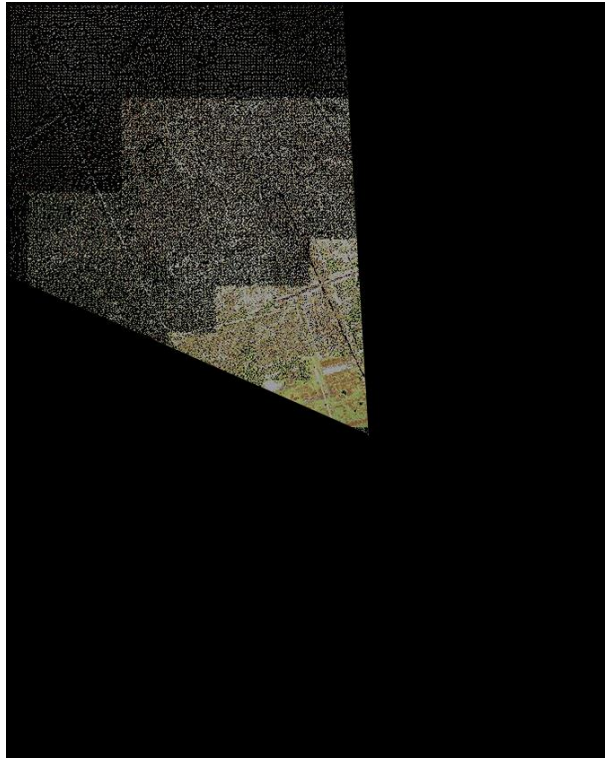
Results



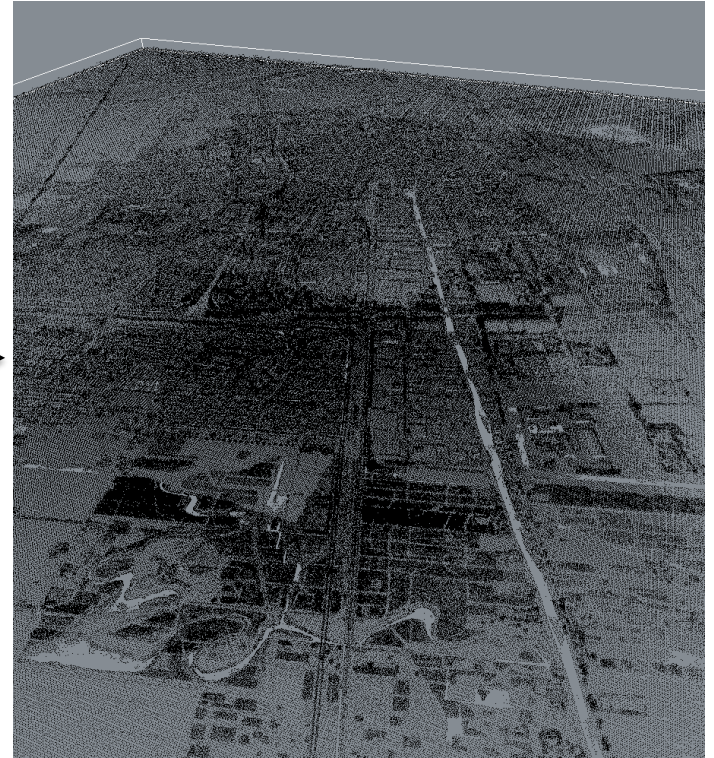
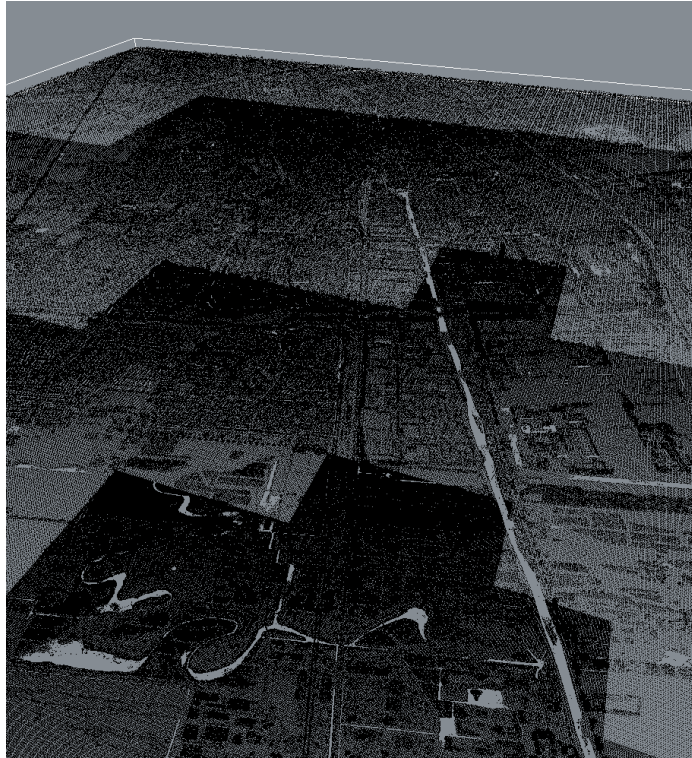
Results – top-down



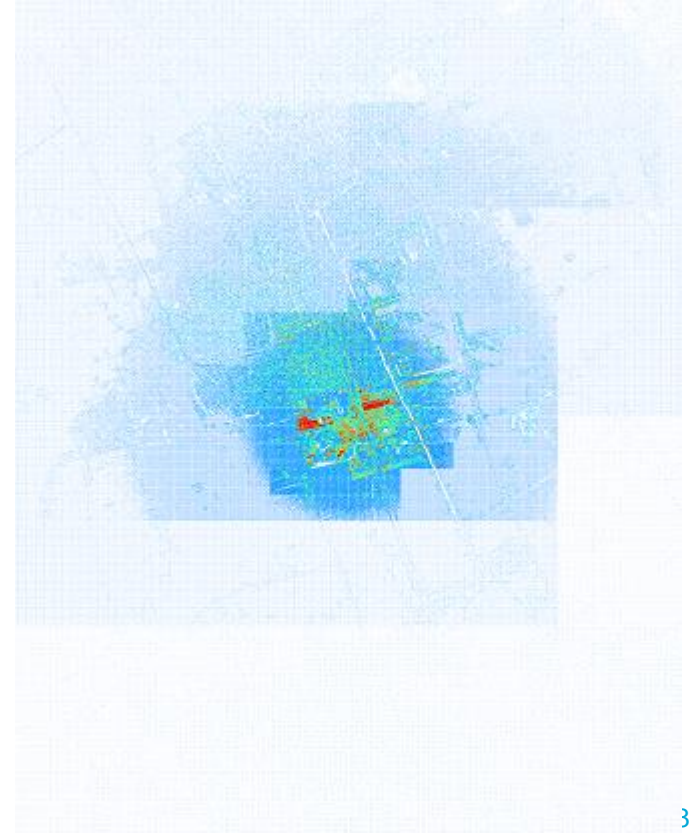
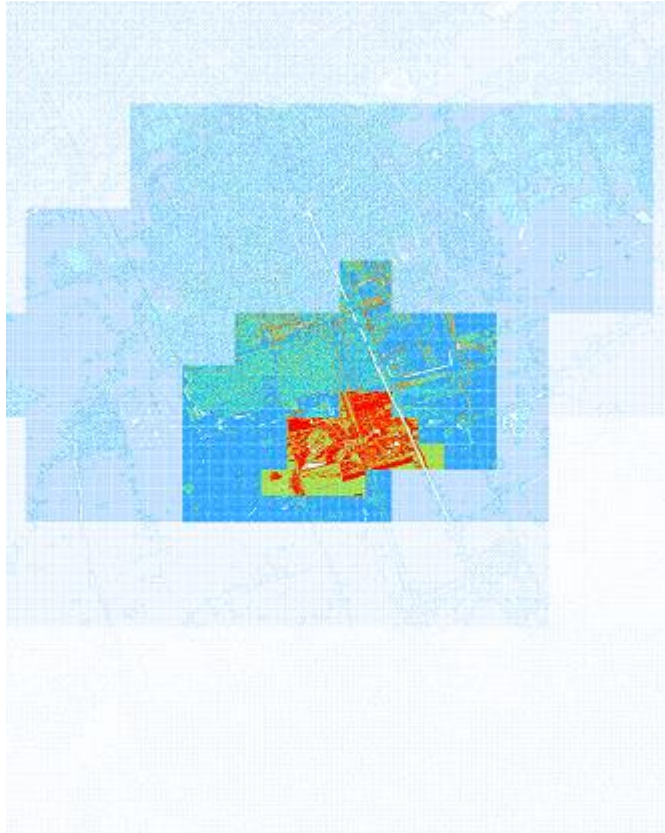
Results – top-down camera frustum



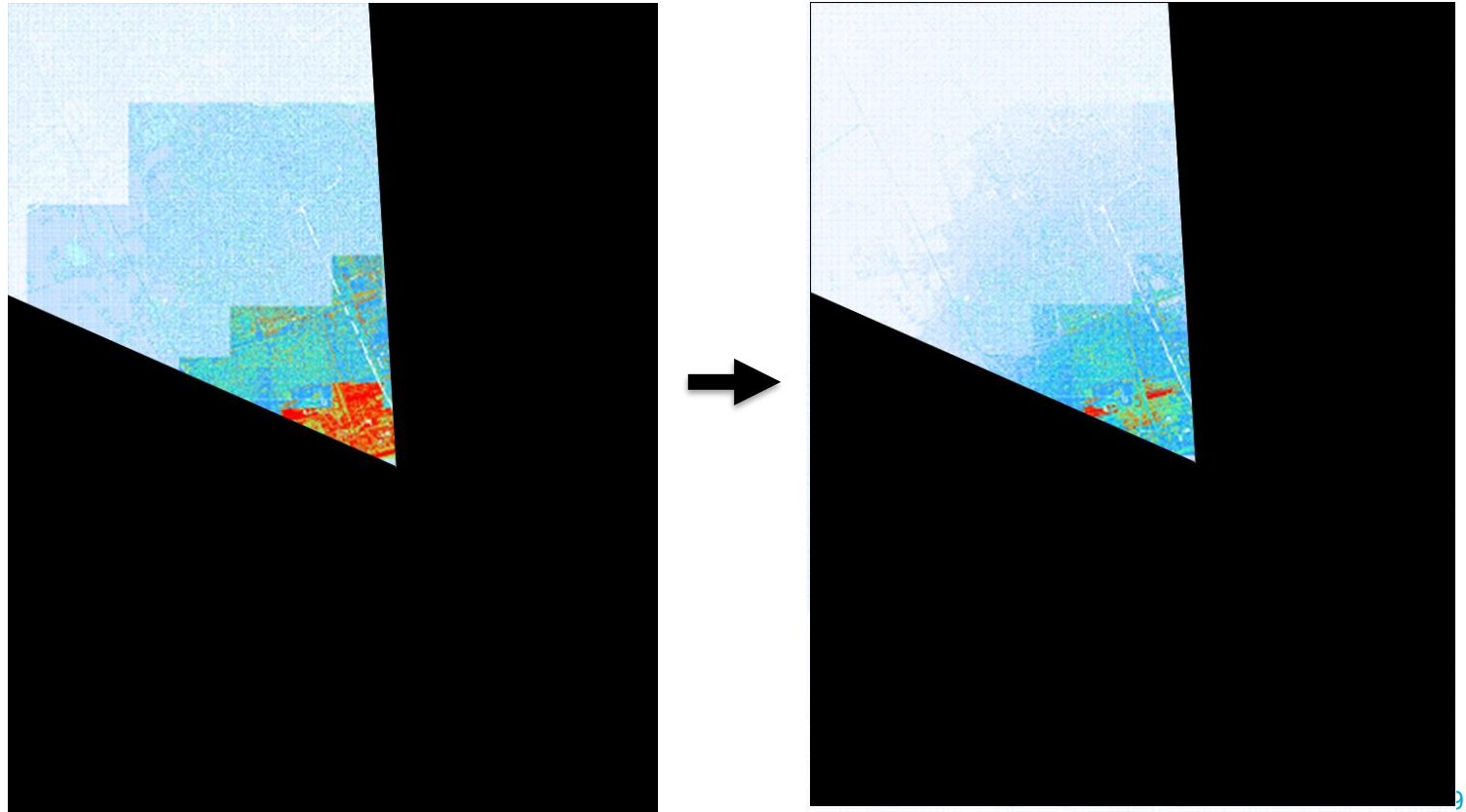
Results – top-down camera frustum



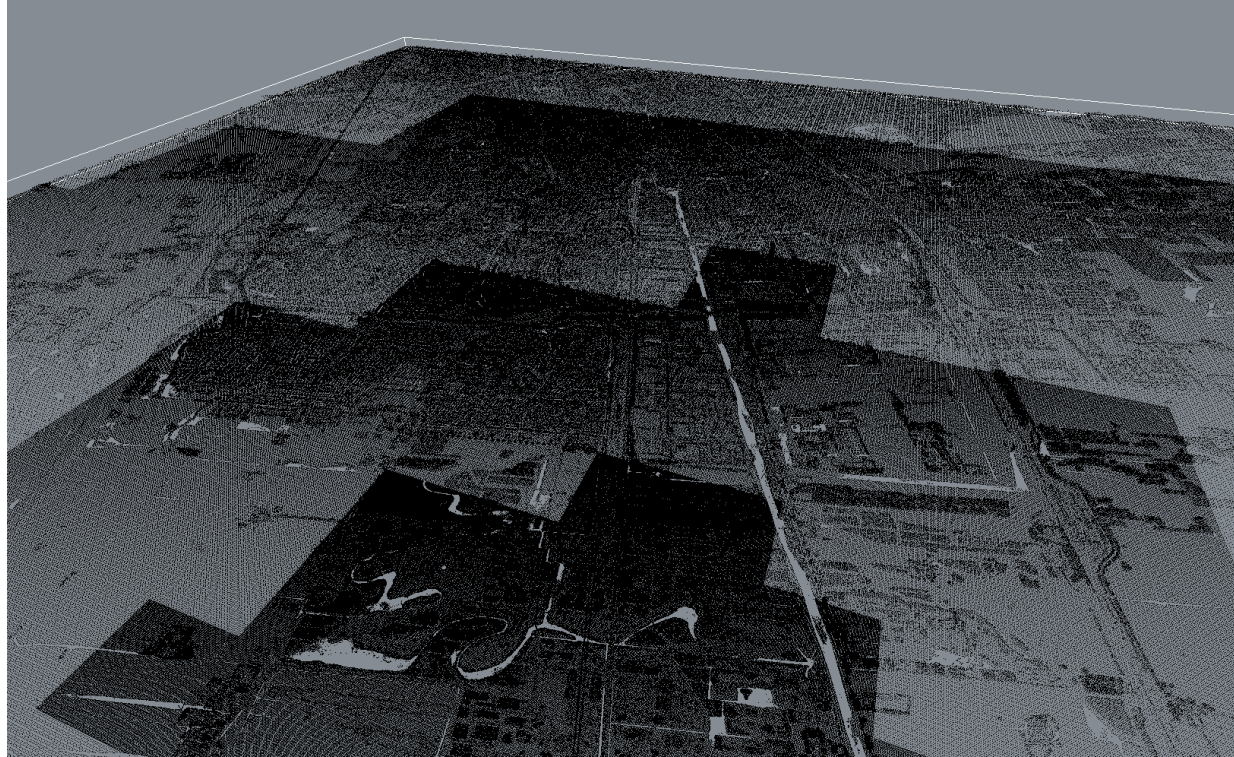
Results – density



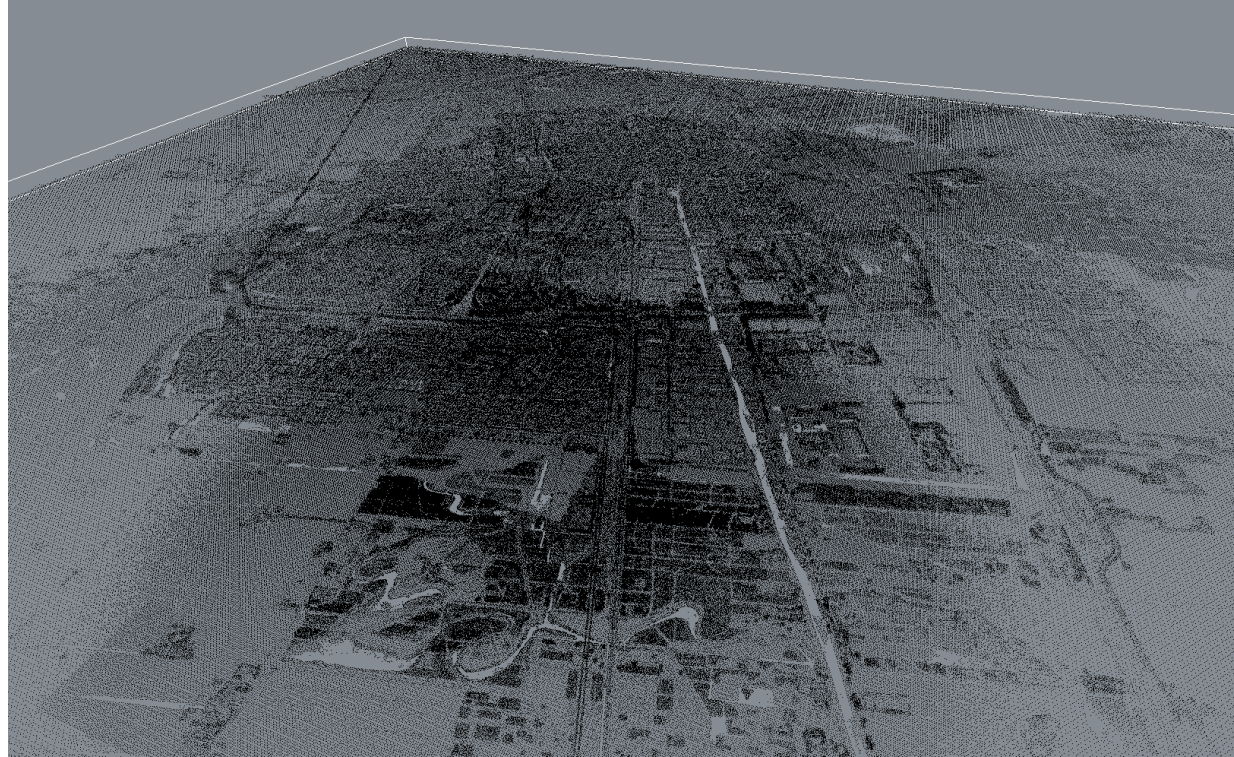
Results – density camera frustum



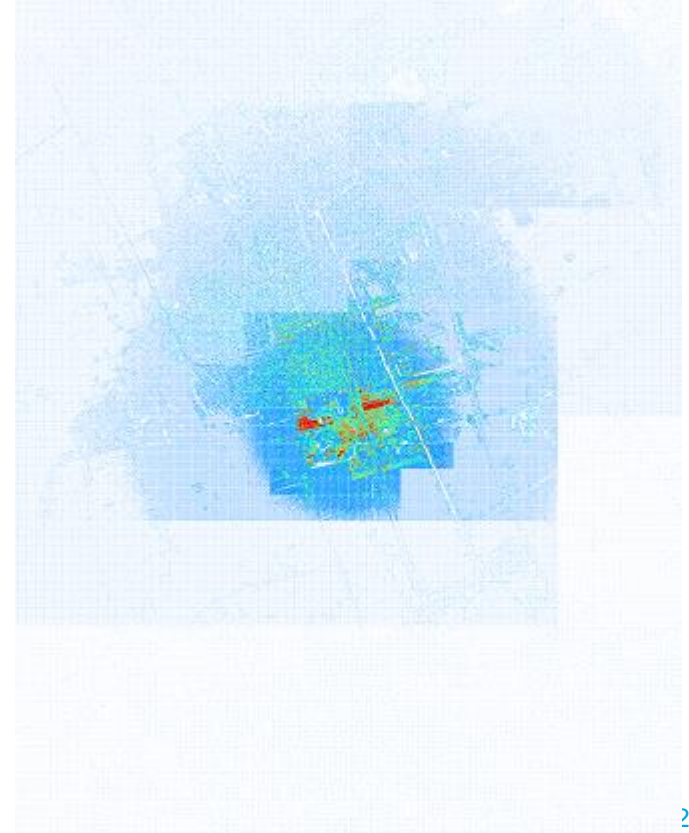
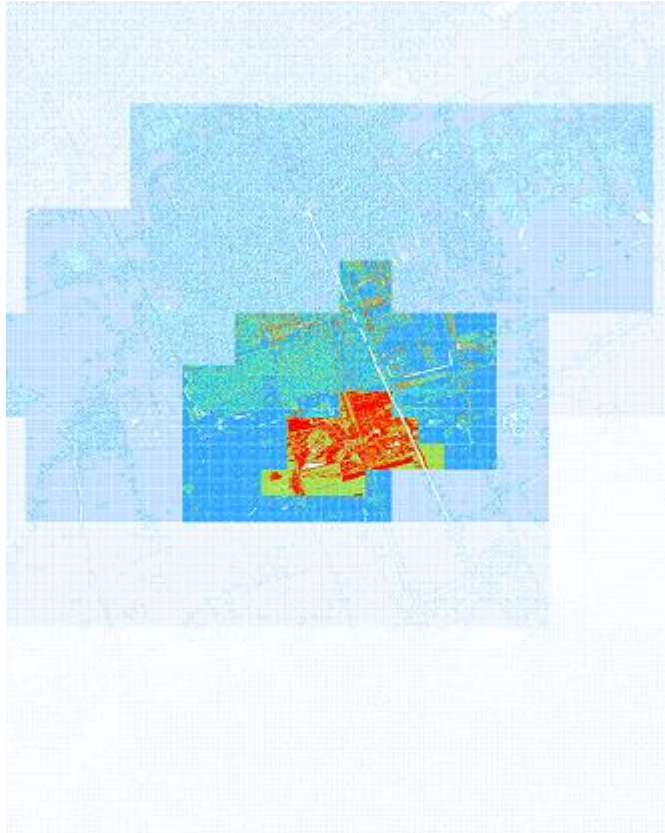
Results – perspective original



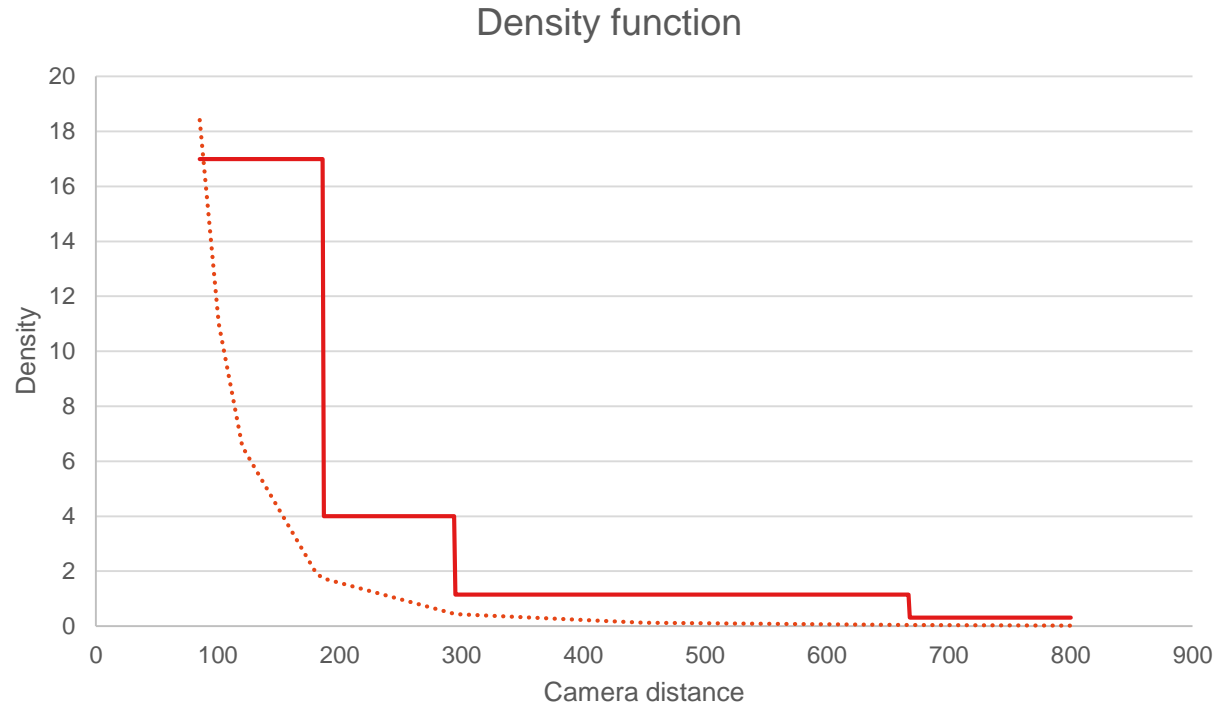
Results – perspective vario-scale



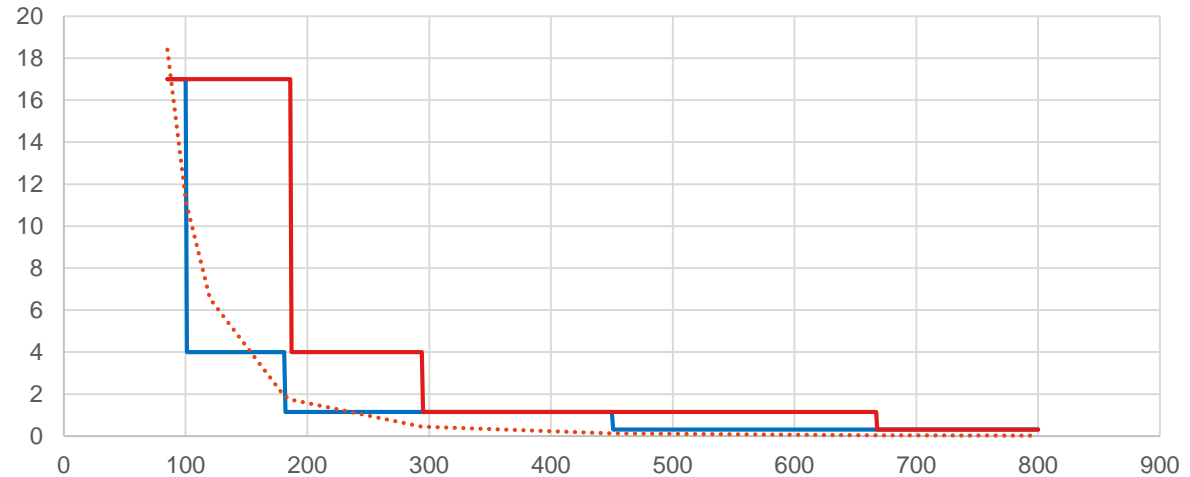
Results – analysis



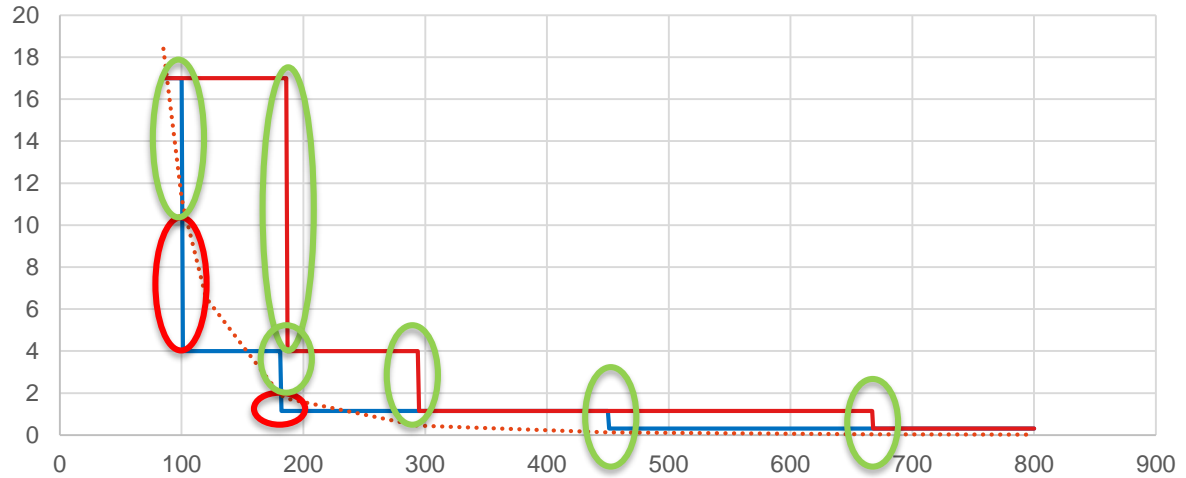
Results



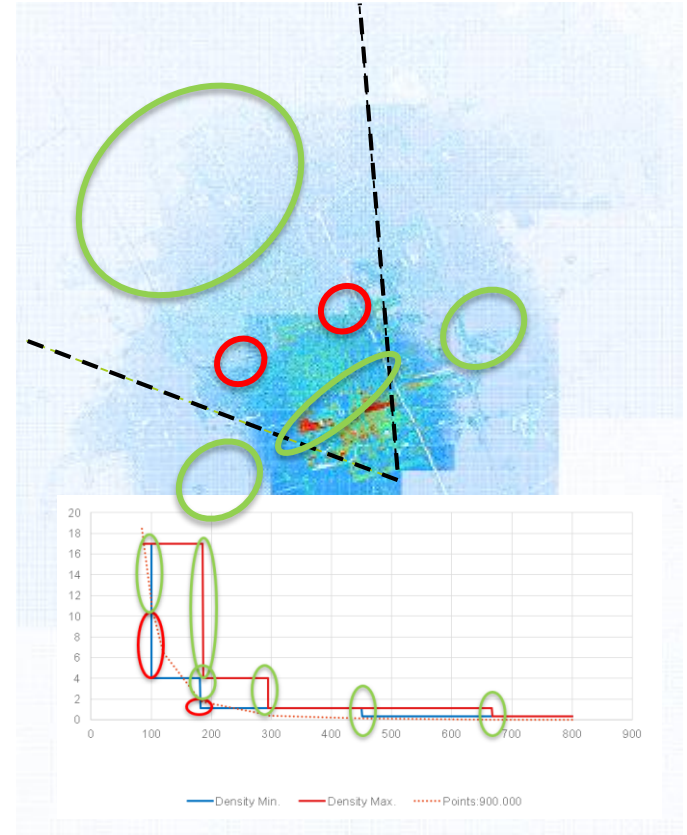
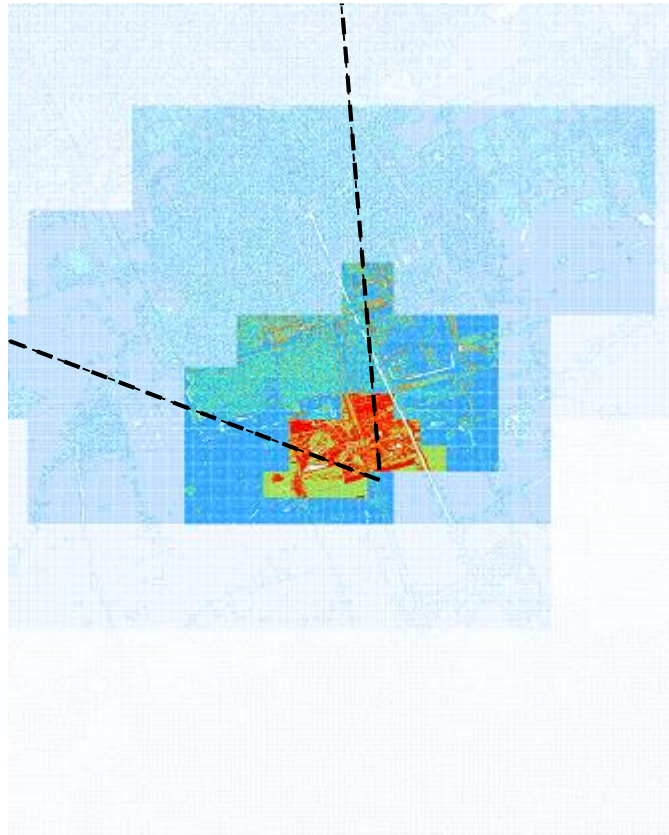
Results – analysis



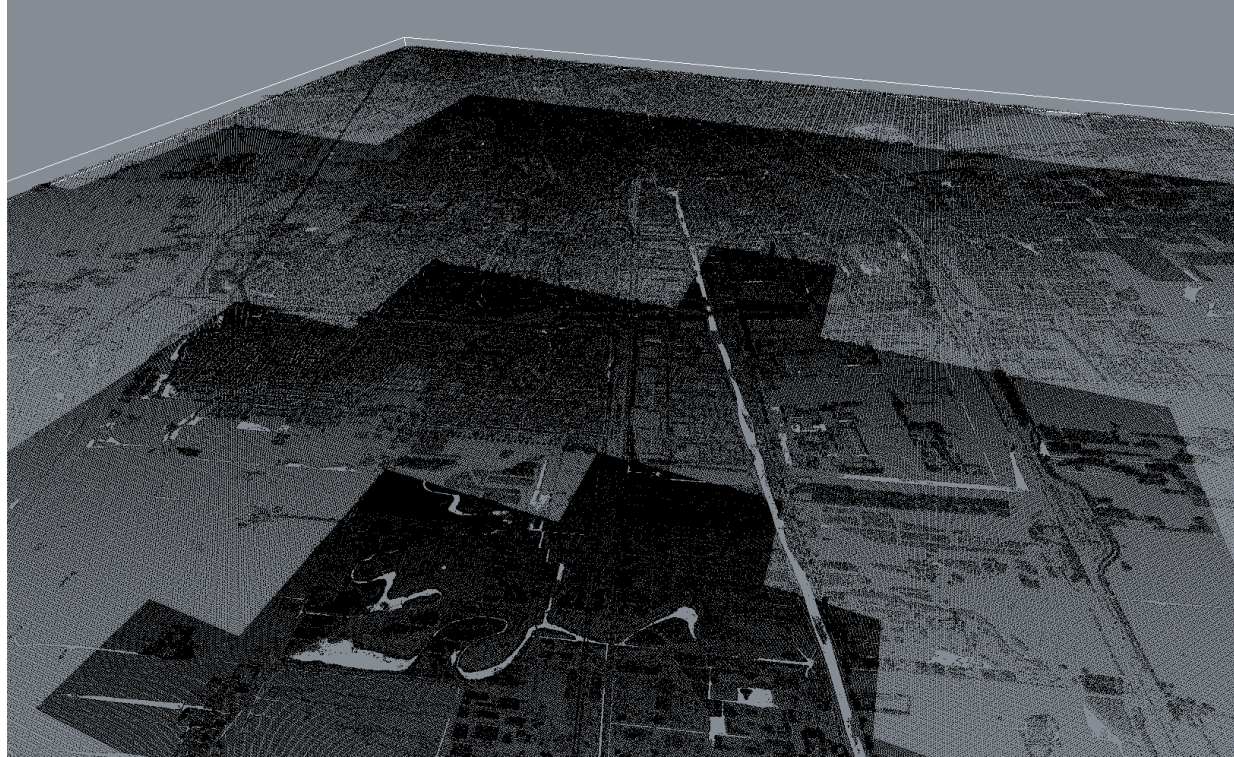
Results – analysis



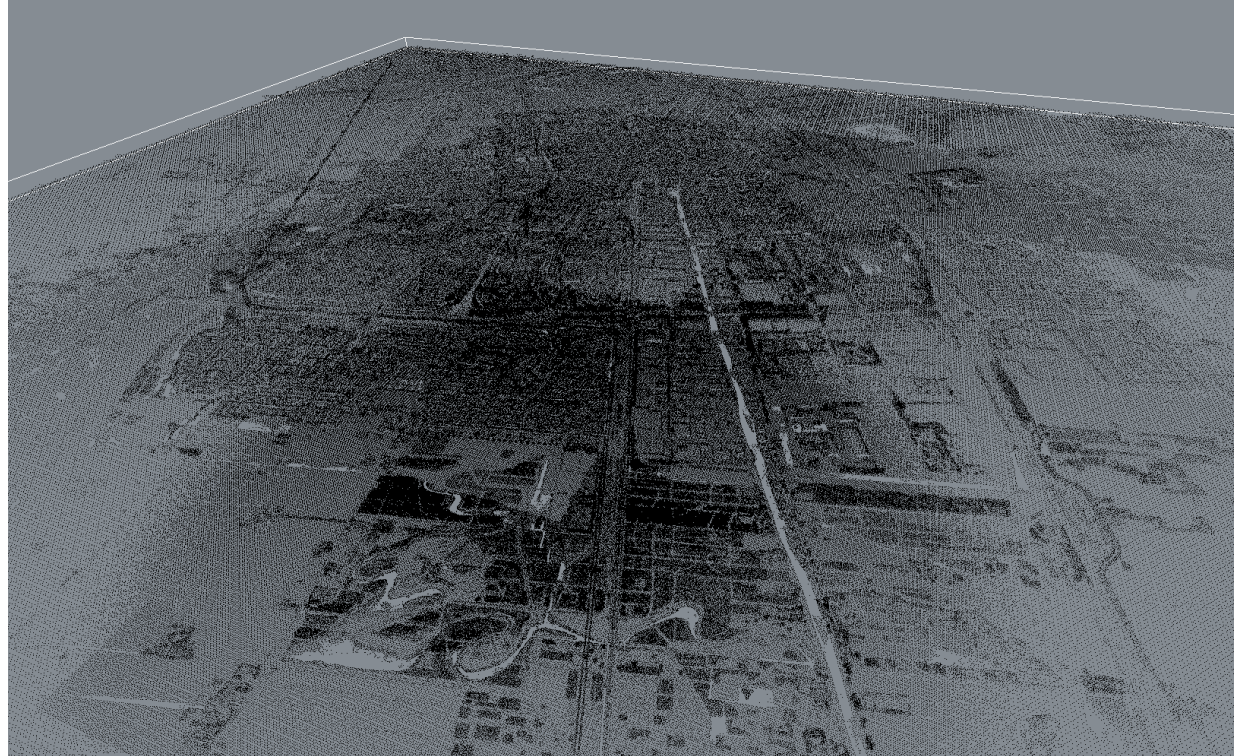
Results – analysis



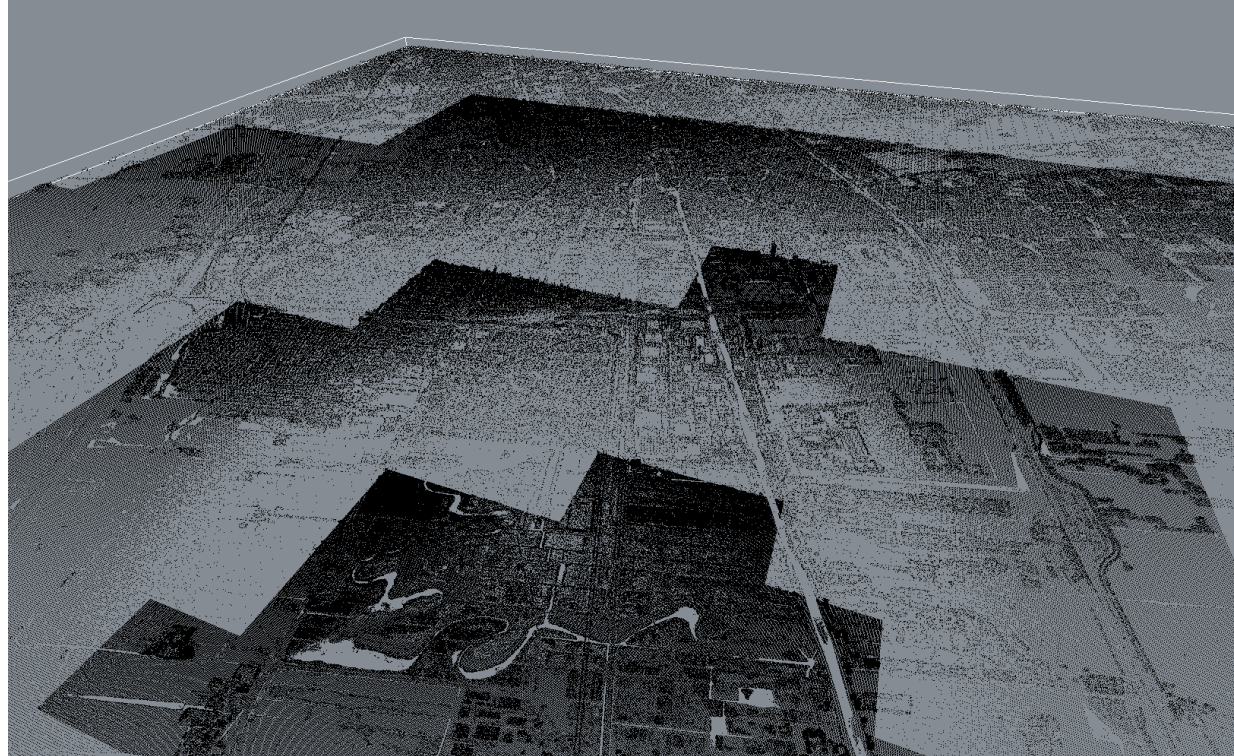
Results – perspective original



Results – perspective vario-scale



Results – removed points



Findings

- The theoretical importance of the perspective transformation matrix for vario-scale visualization

Findings

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- Circle packing method to set the upper limit for density in a data set, dependant on the distance from the camera origin

Findings

- The theoretical importance of the perspective transformation matrix for vario-scale visualization
- Circle packing method to set the upper limit for density in a data set, dependant on the distance from the camera origin
- **Improvements in processing speed are needed for 30fps support**

Future work

- 4D index implementation

Future work

- 4D index implementation
- Determine a global density formula relationship

Future work

- 4D index implementation
- Determine a global density formula relationship
- Practical implementation improvements
 - Implementation in Greyhound
 - GPU calculations
 - Addition in stead of subtraction
 - Process only affected octree blocks, not all

Conclusion

- Circle packing method is used to enforce the density formula for vario-scale visualization

Conclusion

- Circle packing method is used to enforce the density formula for vario-scale visualization
- It is possible to visualize the AHN2 data set in a vario-scale manner with existing web viewer architecture

Conclusion

- Circle packing method is used to enforce the density formula for vario-scale visualization
- It is possible to visualize the AHN2 data set in a vario-scale manner with existing web viewer architecture
- The current solution is not fast enough for 30+ fps performance

P5 presentation

Student: Jippe van der Maaden
Main mentor: Prof.dr.ir. P.J.M. van Oosterom
Second mentor: Dr.ir. B.M. Meijers
Examiner: Dr. H.M.H. van der Heijden