

# Automatic semantic and thematic classification of 3D models to create CityGML

Merwin Rook

P5 Presentation

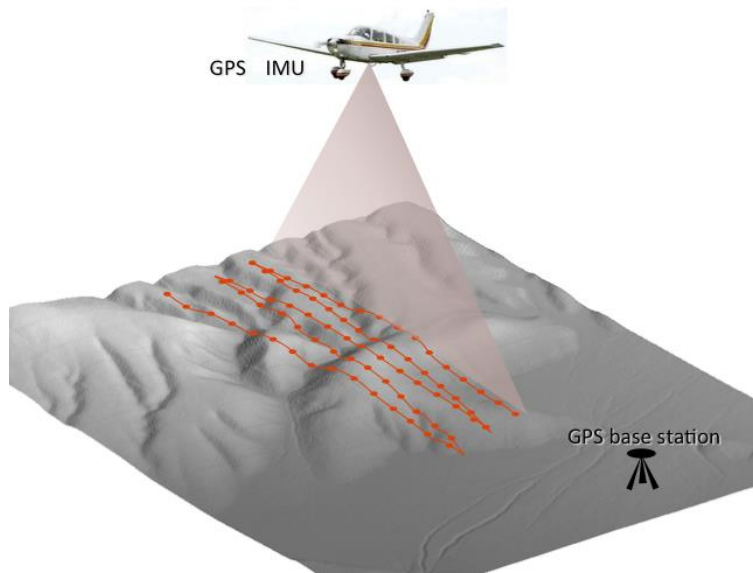
Master of Science in Geomatics for the Built  
Environment

November 4 2016

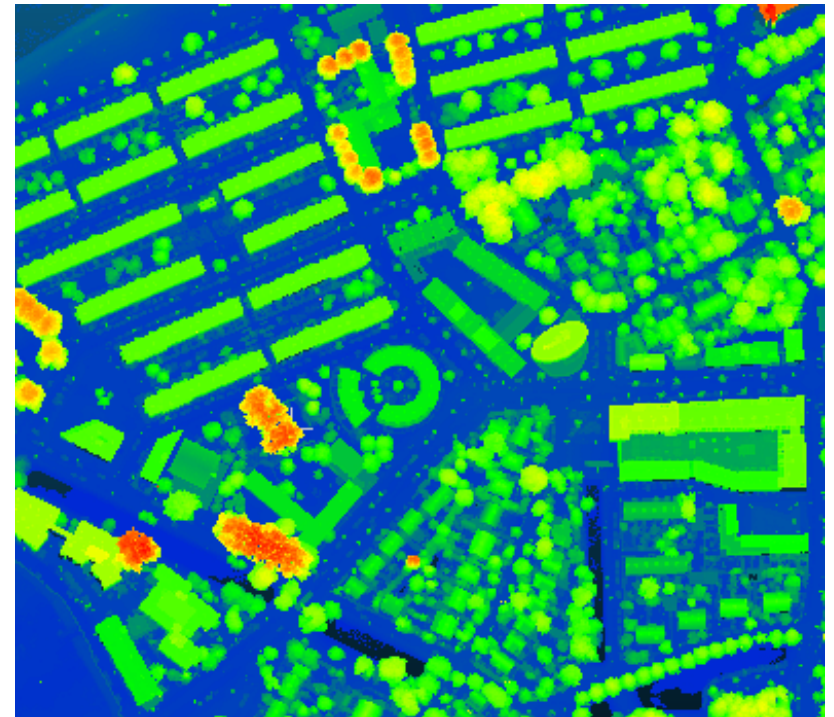
# Presentation overview

- Introduction
- Problem statement and research scope
- Methodology
  - Indexing
  - Region growing
  - Thematic classification
  - Semantic classification
  - LoD detection
- Results
- Conclusions
- Future work

# Advancing developments in remote sensing

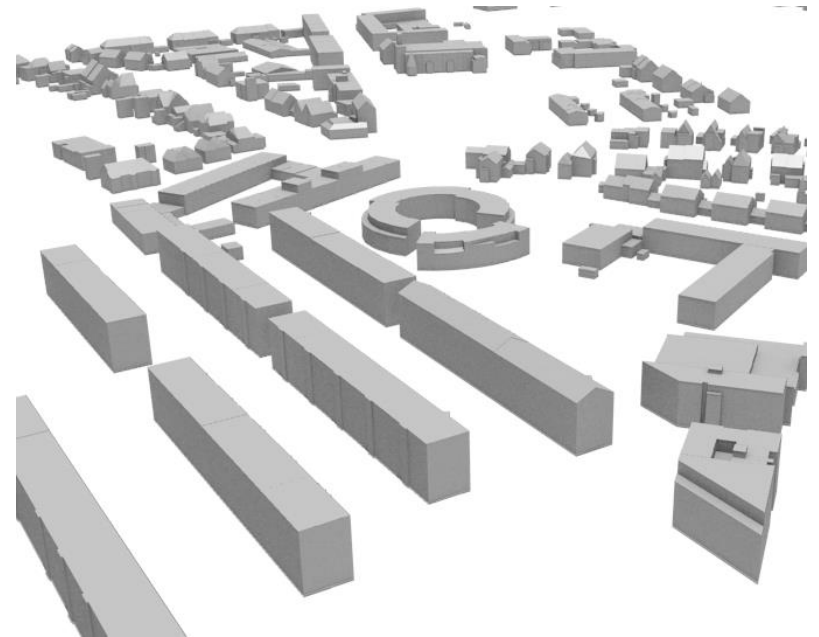
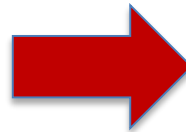
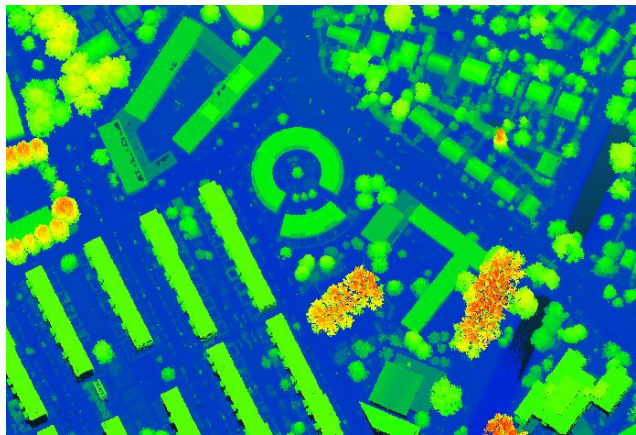


Source: [serc.carleton.edu](http://serc.carleton.edu)



Source: AHN3

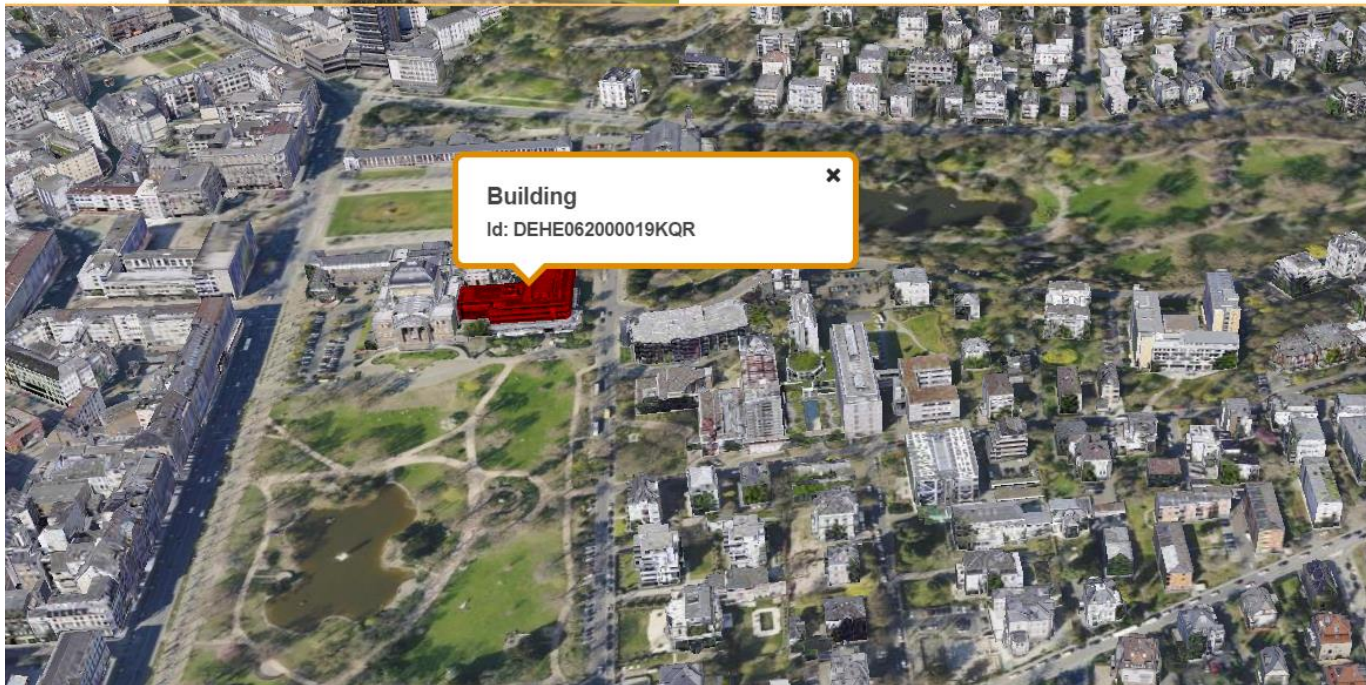
# 3D GIS: the creation of 3D city models



# Applications 3D GIS



Luftbilder und 3D-Stadtmodell Wiesbaden

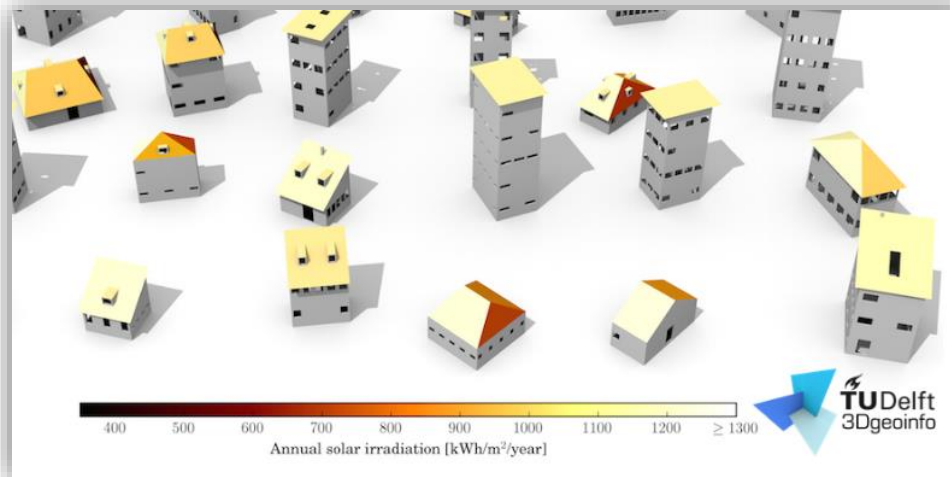


# Applications of 3D GIS



Flood modelling

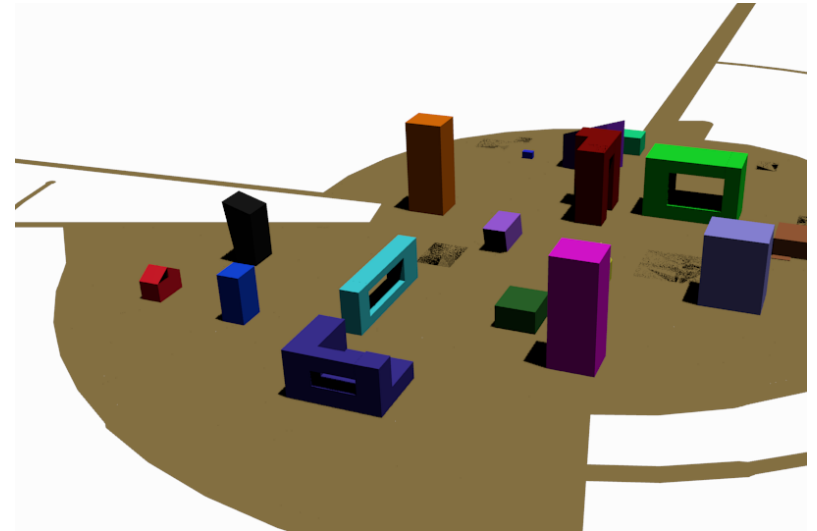
Solar irradiation analysis



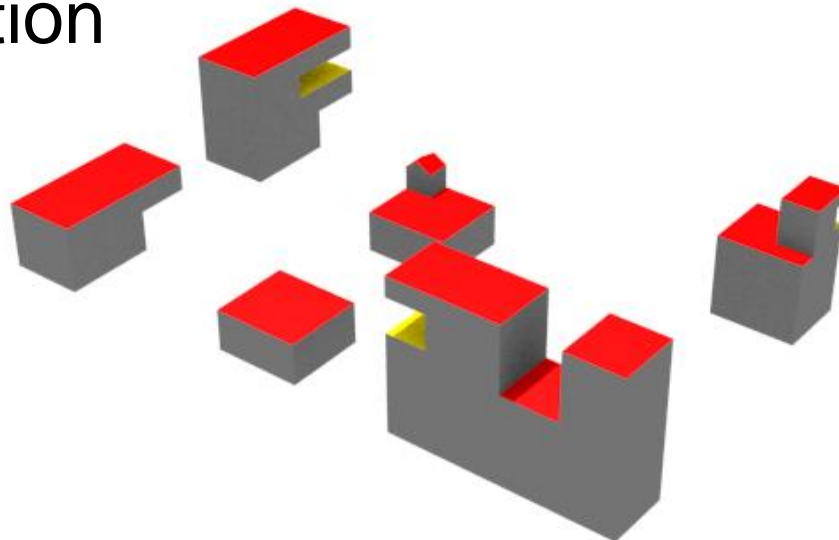
Source: Biljecki et al.  
(2016)

# Semantic and thematic information

- Thematic information



- Semantic information

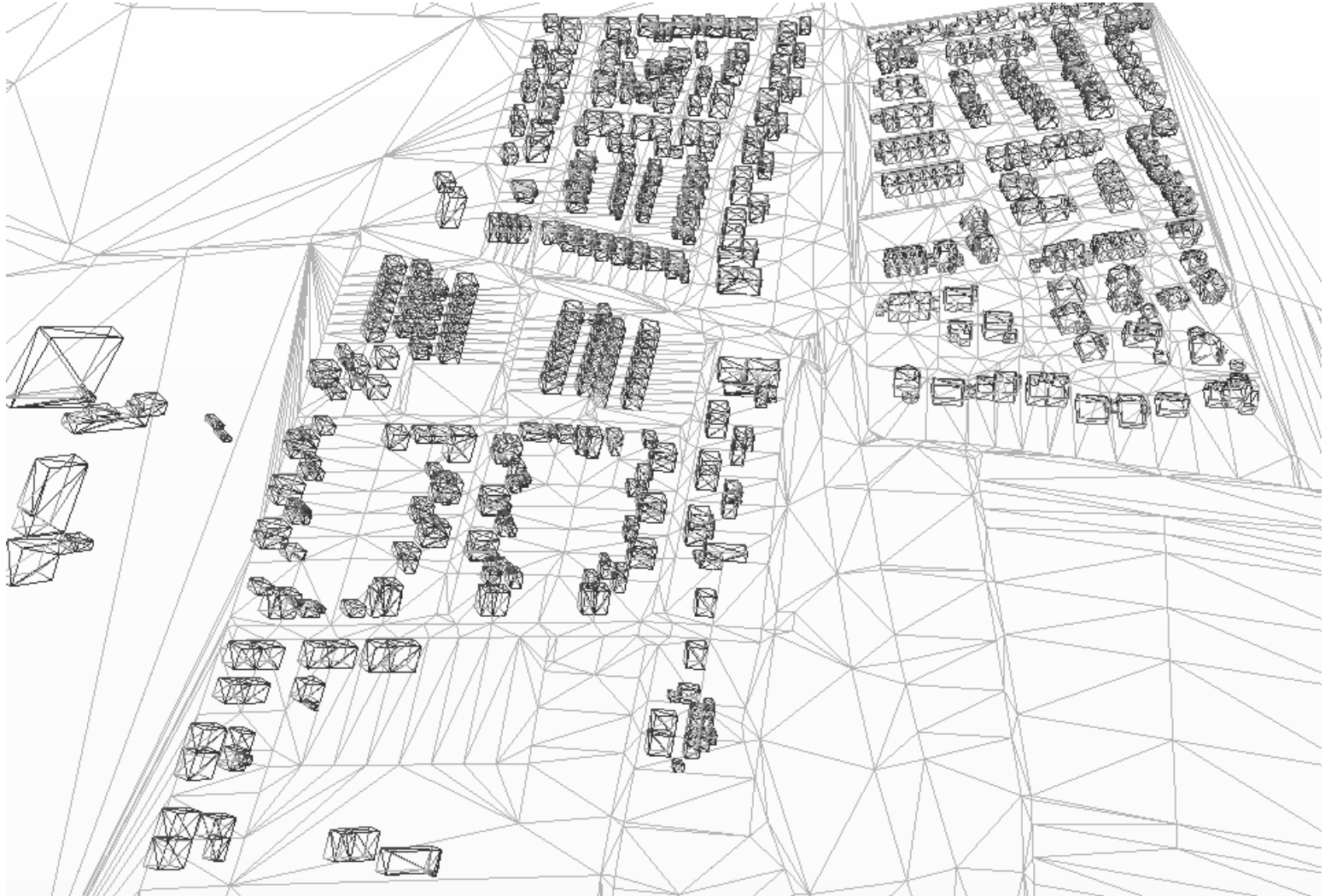


# Semantic and thematic information

- Use of semantic and thematic information:
  - Query: select surfaces or objects of interest
  - Spatial analysis & real world simulations
  - Interoperability



# Problem statement



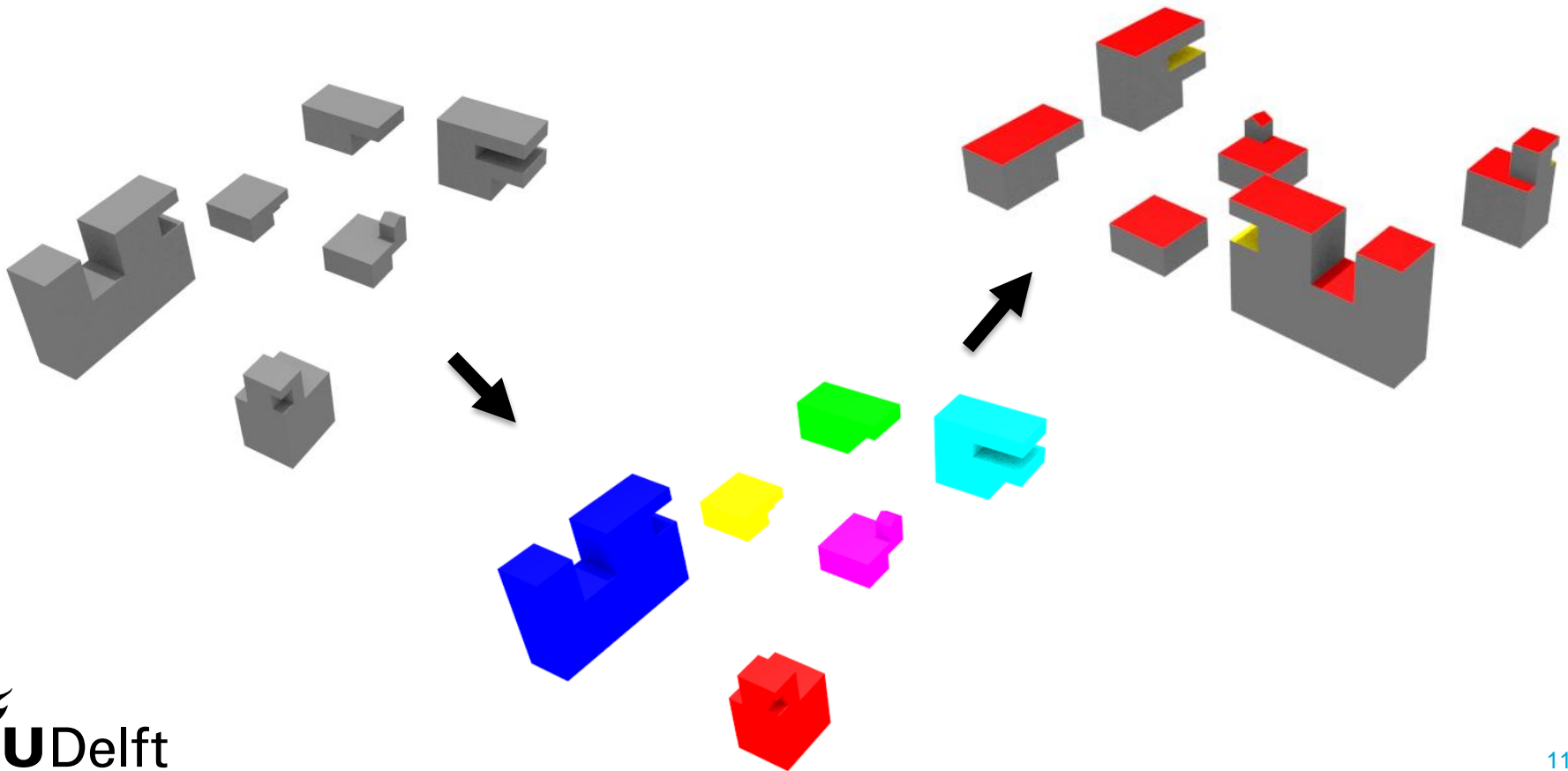
Data source: Waldbrücke model

# Semantic and thematic classification

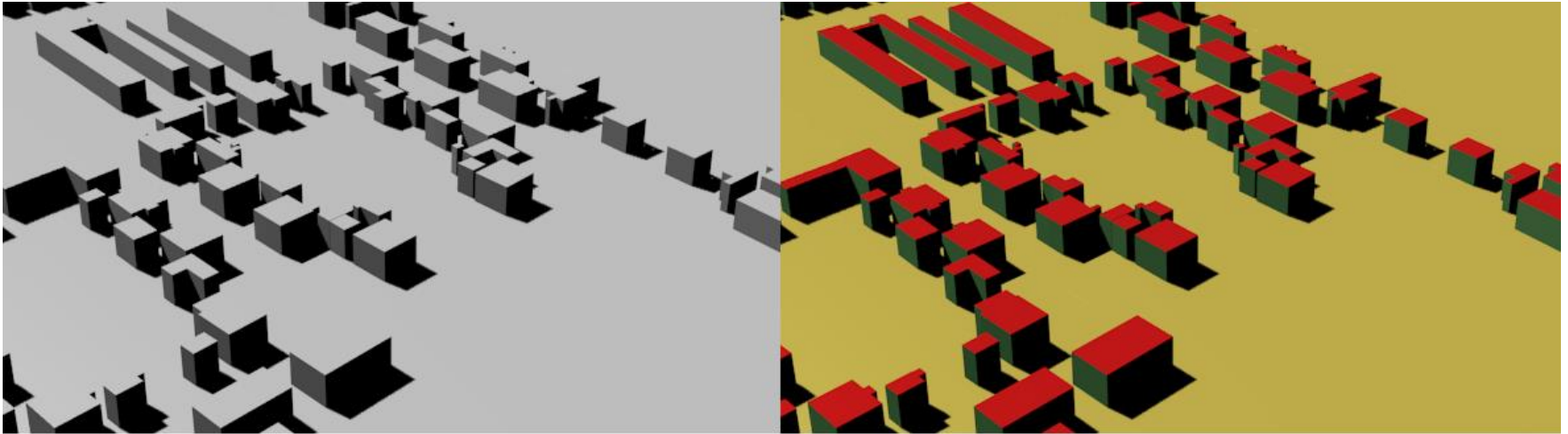
- Mostly a manual process: time and cost expensive
- Limitations of current research:
  - Thematic aspect is ignored
  - Not all semantic classes that are required to create CityGML are covered

# Research question

*How to automatically enrich a 3D city model with thematic and semantic information as defined in CityGML, by only utilising the models geometry?*



# Research goal



Automate the process of thematic and semantic classification, in accordance with the CityGML standard

# CityGML

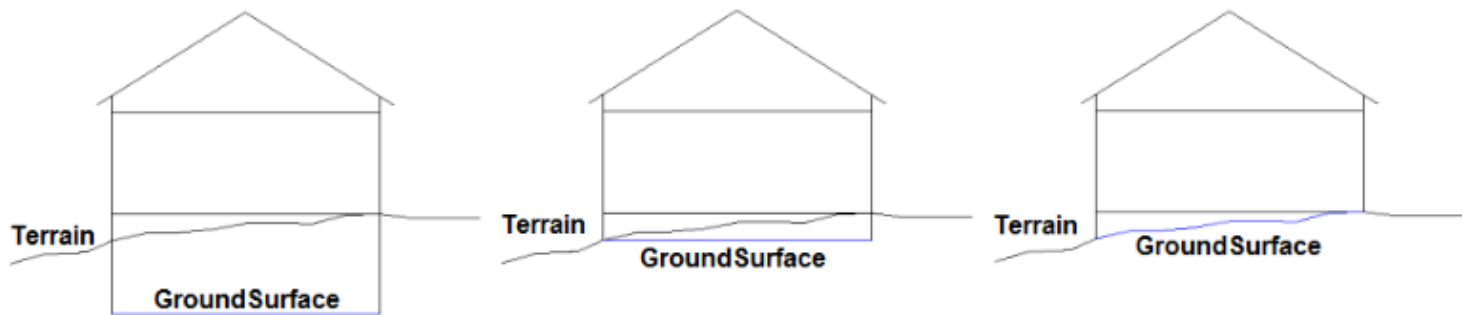
- OGC standard
- 3D city model with attached its thematic and semantic information.
  
- Most important: defines the thematic and semantic structure used in this research

# Research challenges

- 5 Challenges are identified

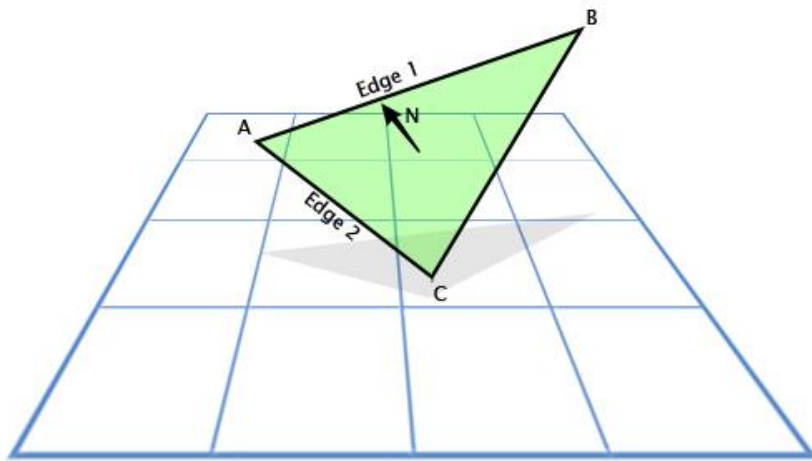
# Research challenges (1)

- Complexity of the semantic classes and the lack of definition: CityGML only defines the surface normal.

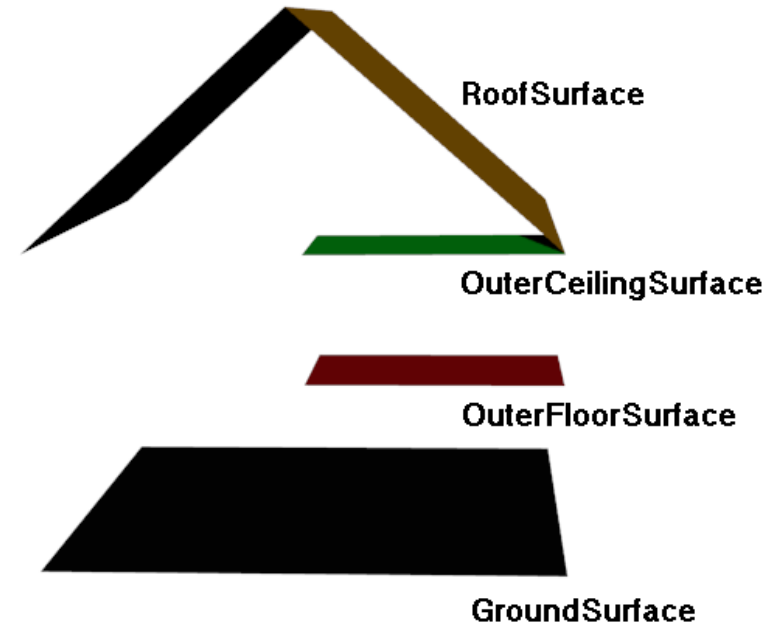


# Research challenges (2)

- Limited applicability of utilisation the surface normal



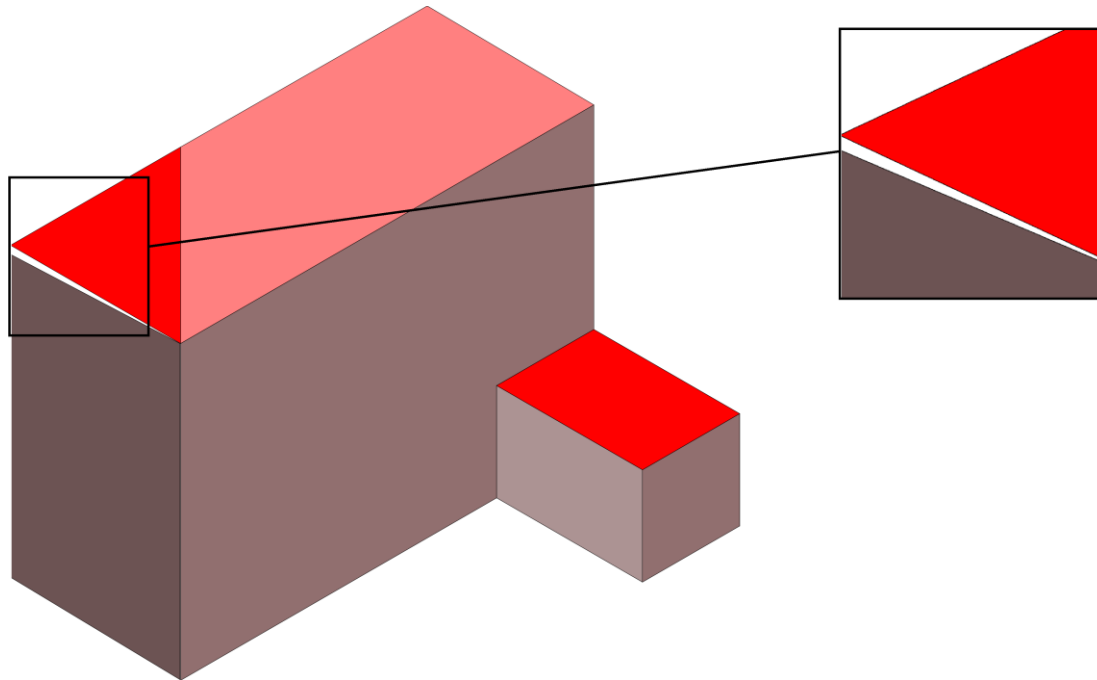
Source: [blog.wolfire.com](http://blog.wolfire.com)



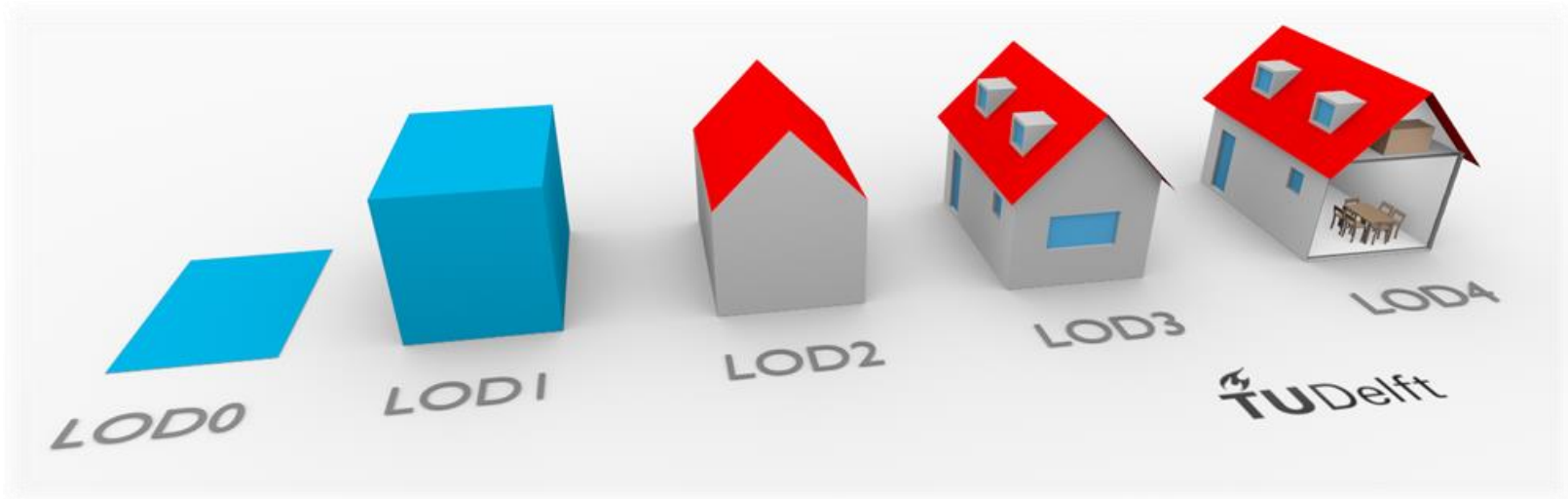


# Research challenges (3)

- Topology and spatial index
  - Reconstruction of the topology is hindered by slivers and double stored vertices.



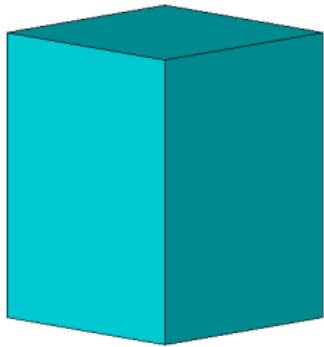
# Research challenges (4): Level of Detail (LoD)



Source: Biljecki et al., 2016

- Only the classification of LoD 1 and 2 models is researched

# Research challenges (4): LoD

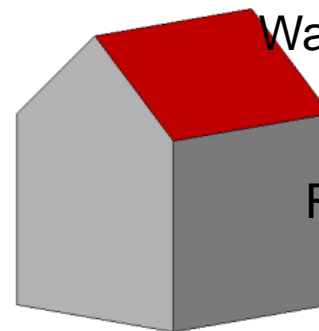


LoD1:

Only thematic information

LoD2:

Thematic and  
Semantic information

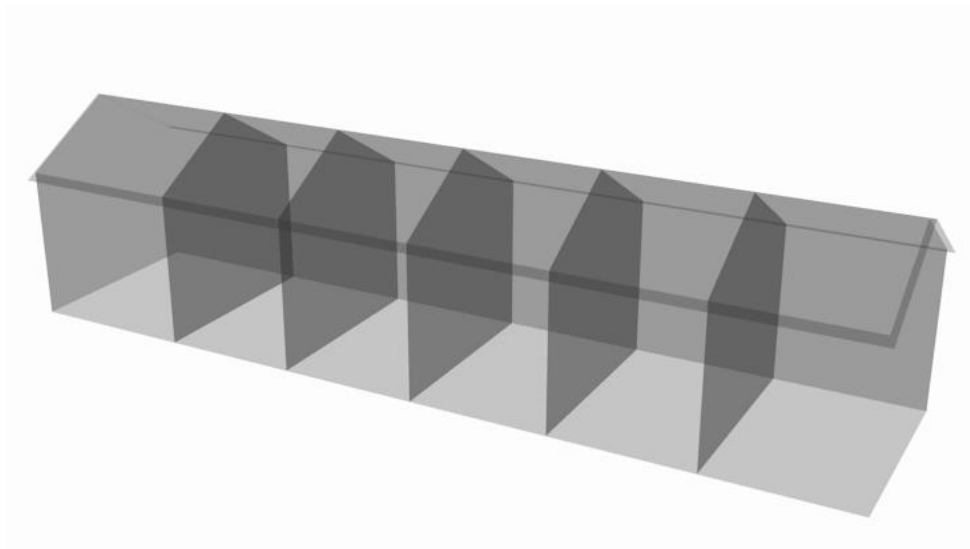


WallSurface

RoofSurface

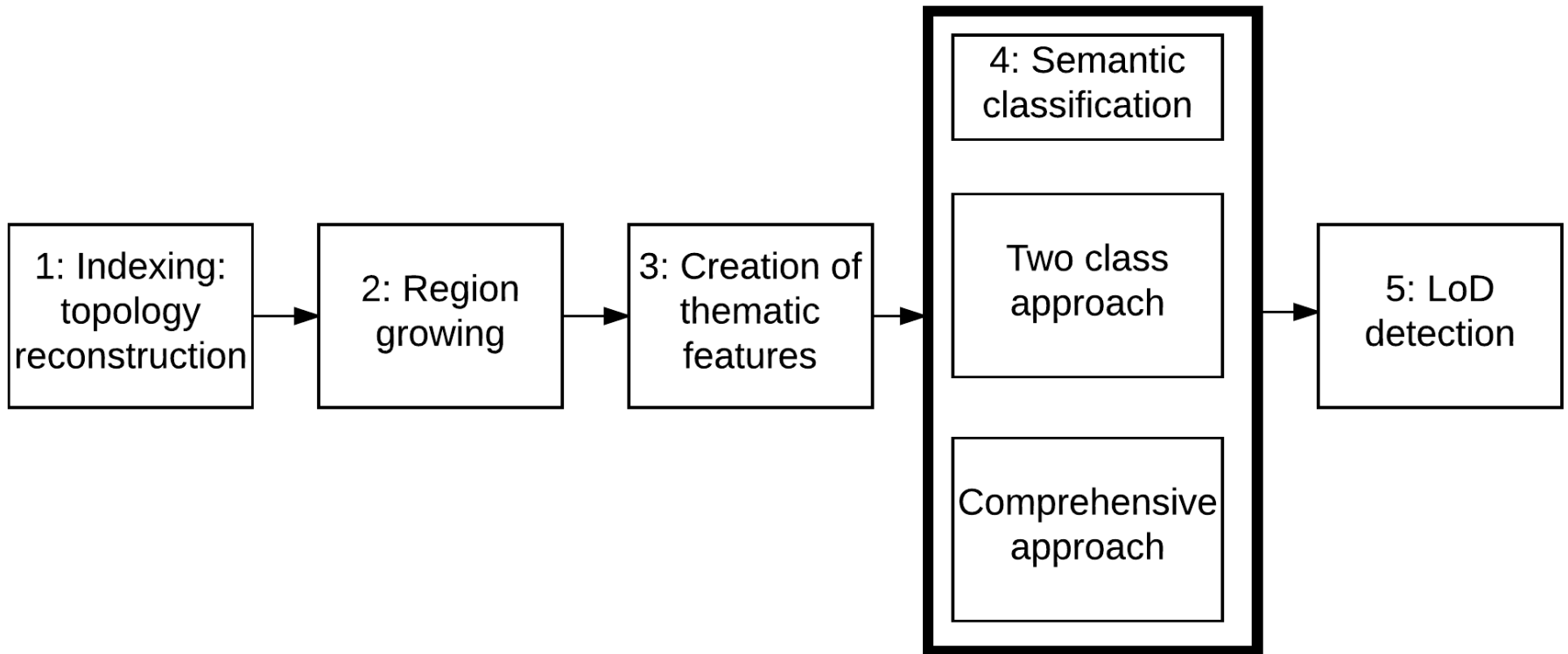
# Research challenges (5)

- Lack of definition for the thematic features

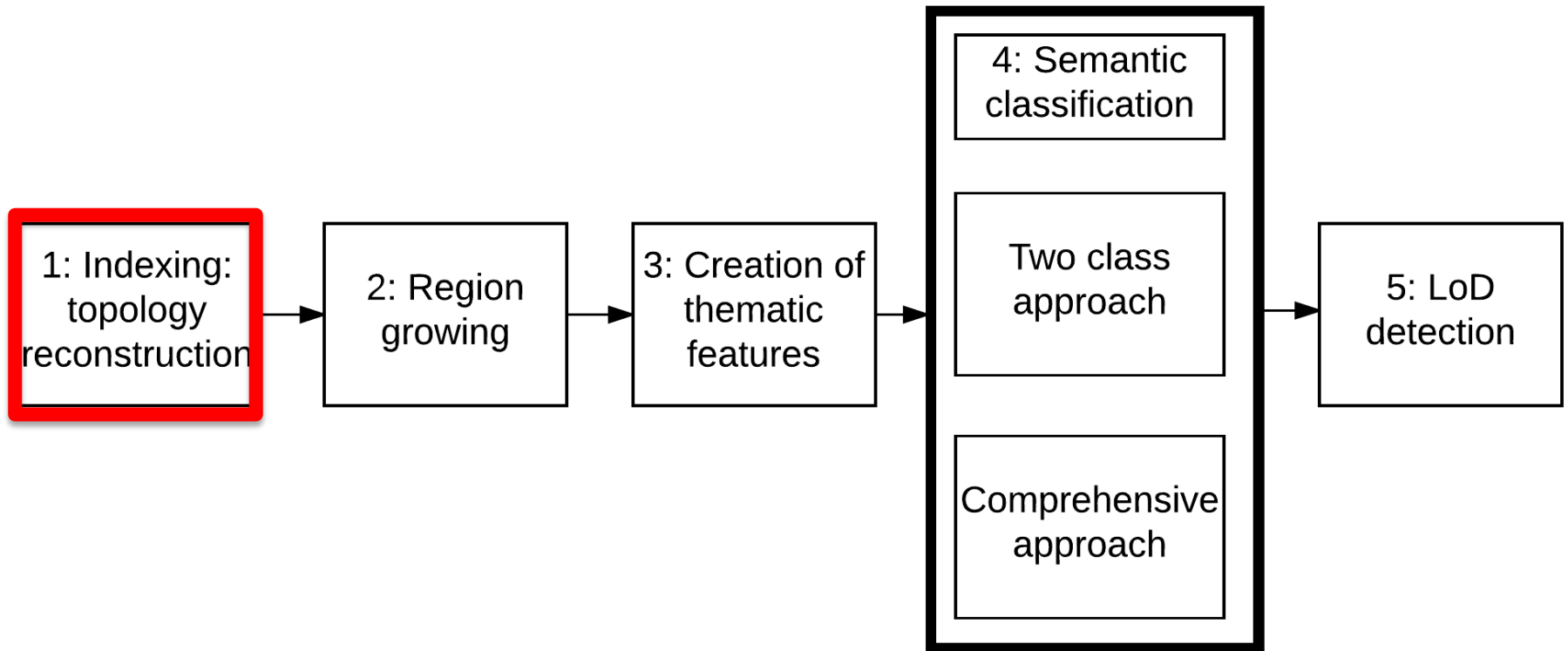


Data source: Rotterdam 3D

# Methodology

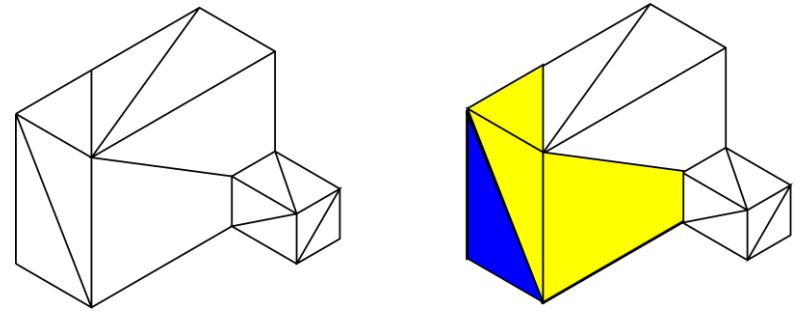


# Methodology: Indexing

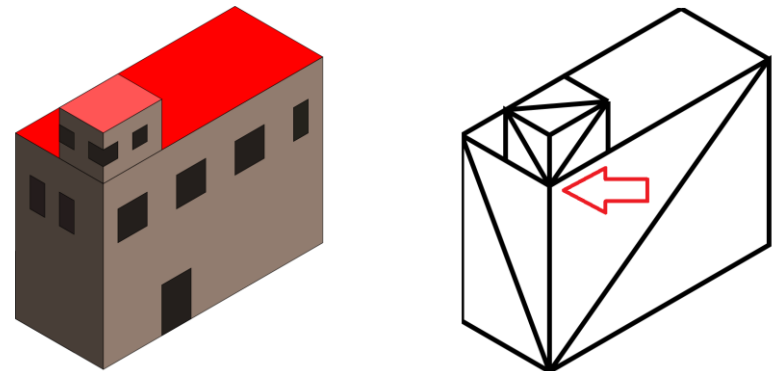
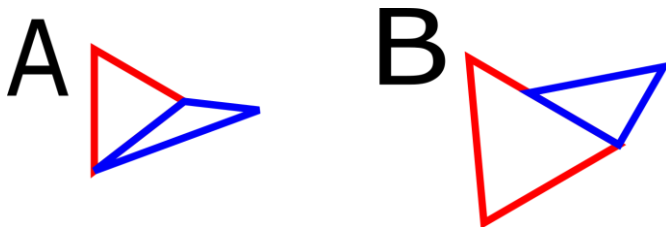


# Indexing

- Index: stores the adjacency relations between triangles

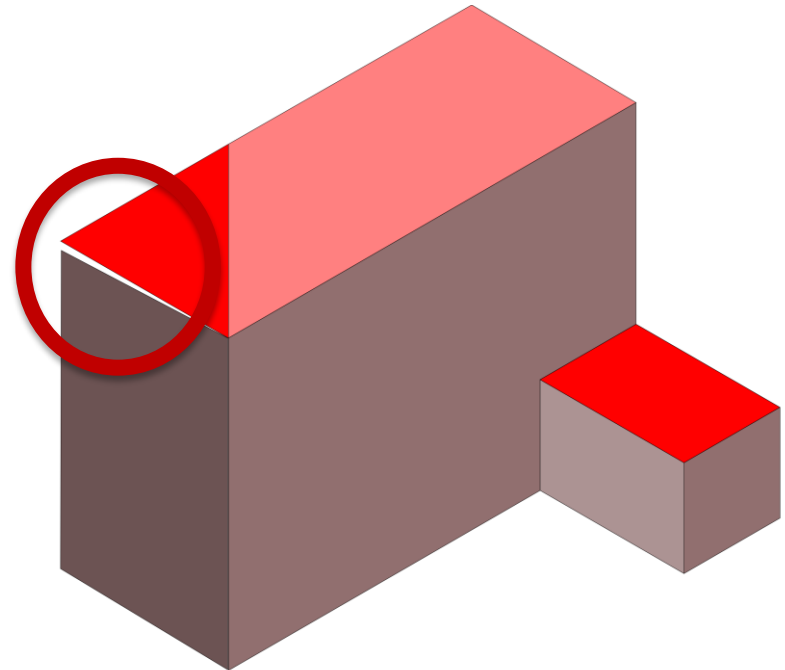
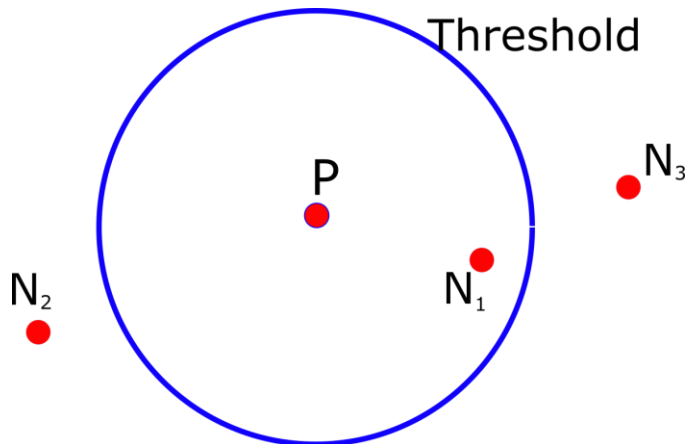


- Relationship is adjacent if triangles share a vertex



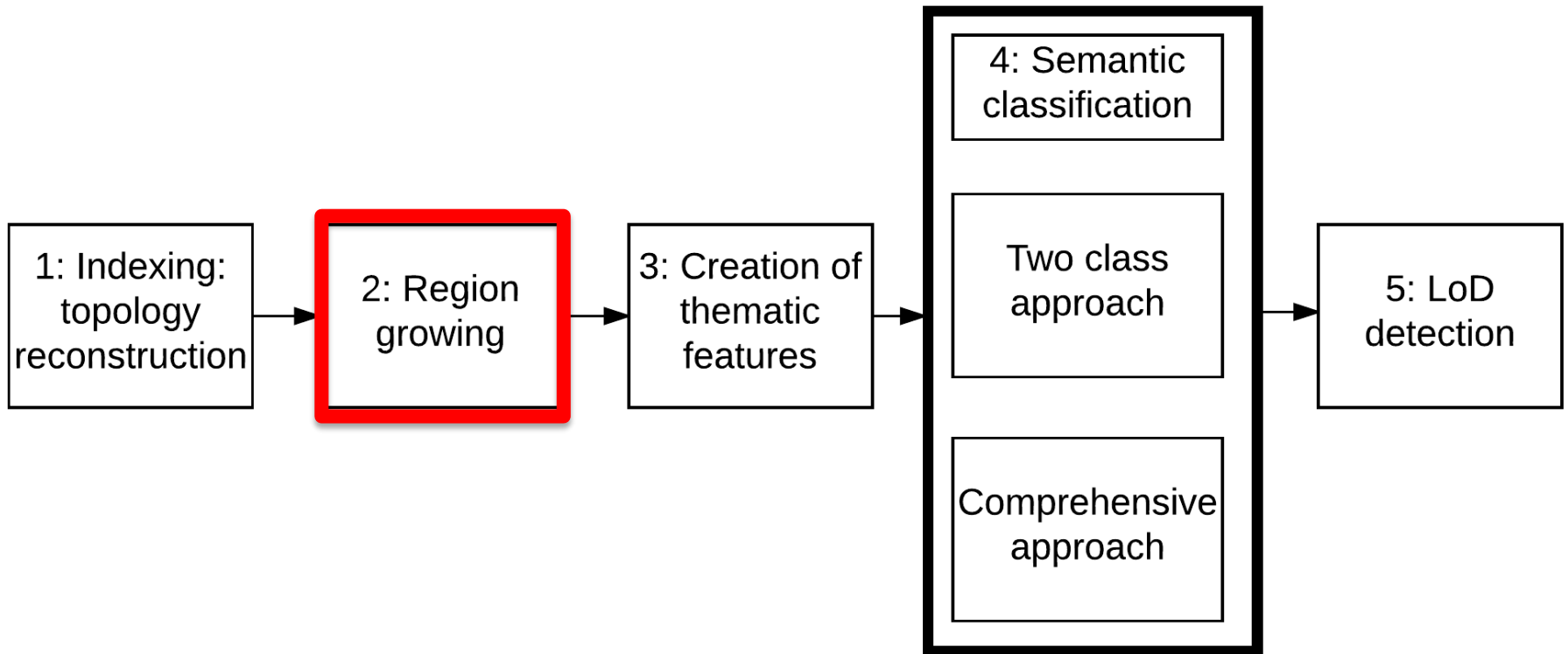
# Indexing (2): KD-tree, used to reconstruct the topologic relationships

- Errors are repaired by setting a threshold and the distance to the k nearest neighbours



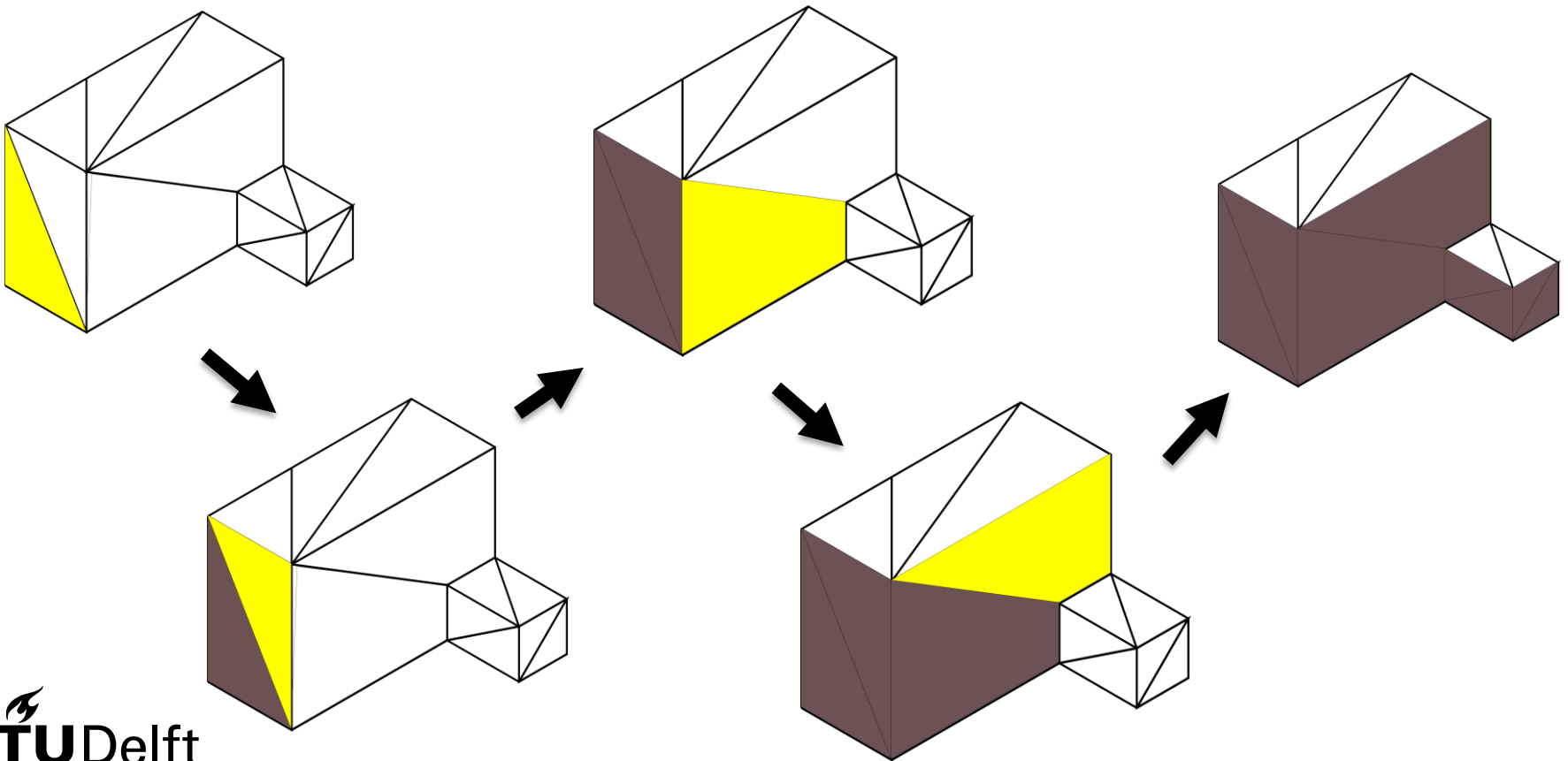


# Region Growing



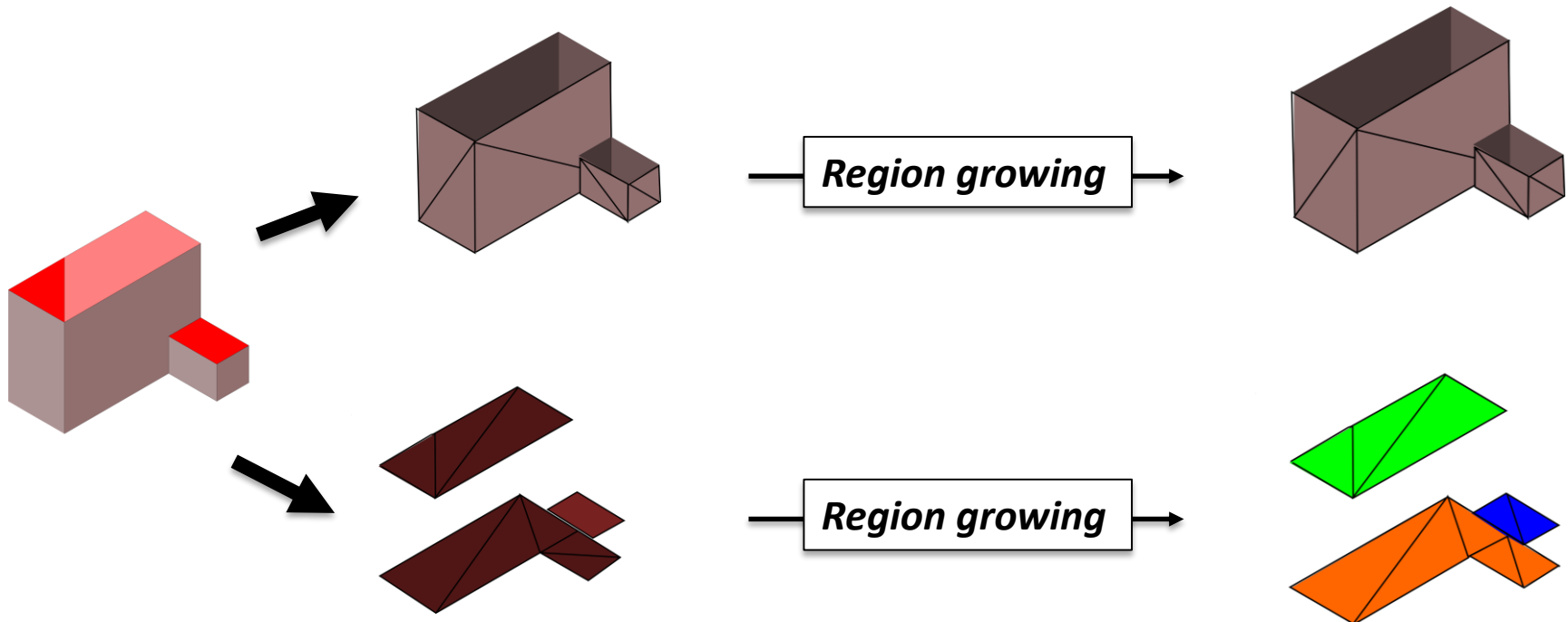
# Region Growing(1)

- Clusters adjacent spatial features together if they have similar geometrical properties.



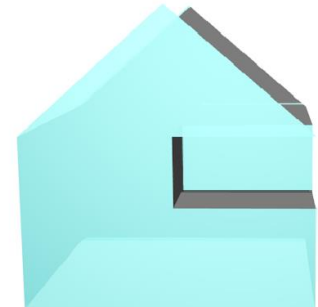
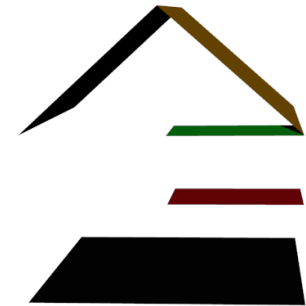
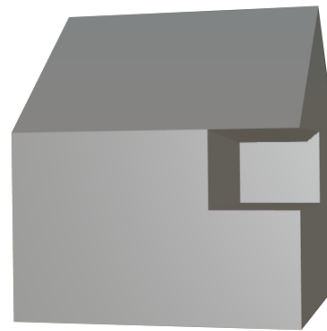
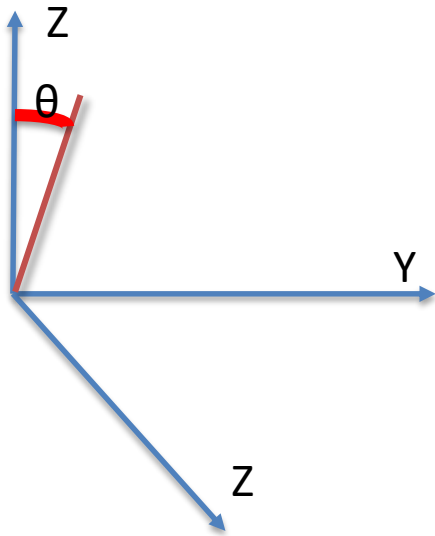
# Region Growing (2); Selecting Wall and non-Wall triangles

- Sets are individually processed in the region growing algorithm.

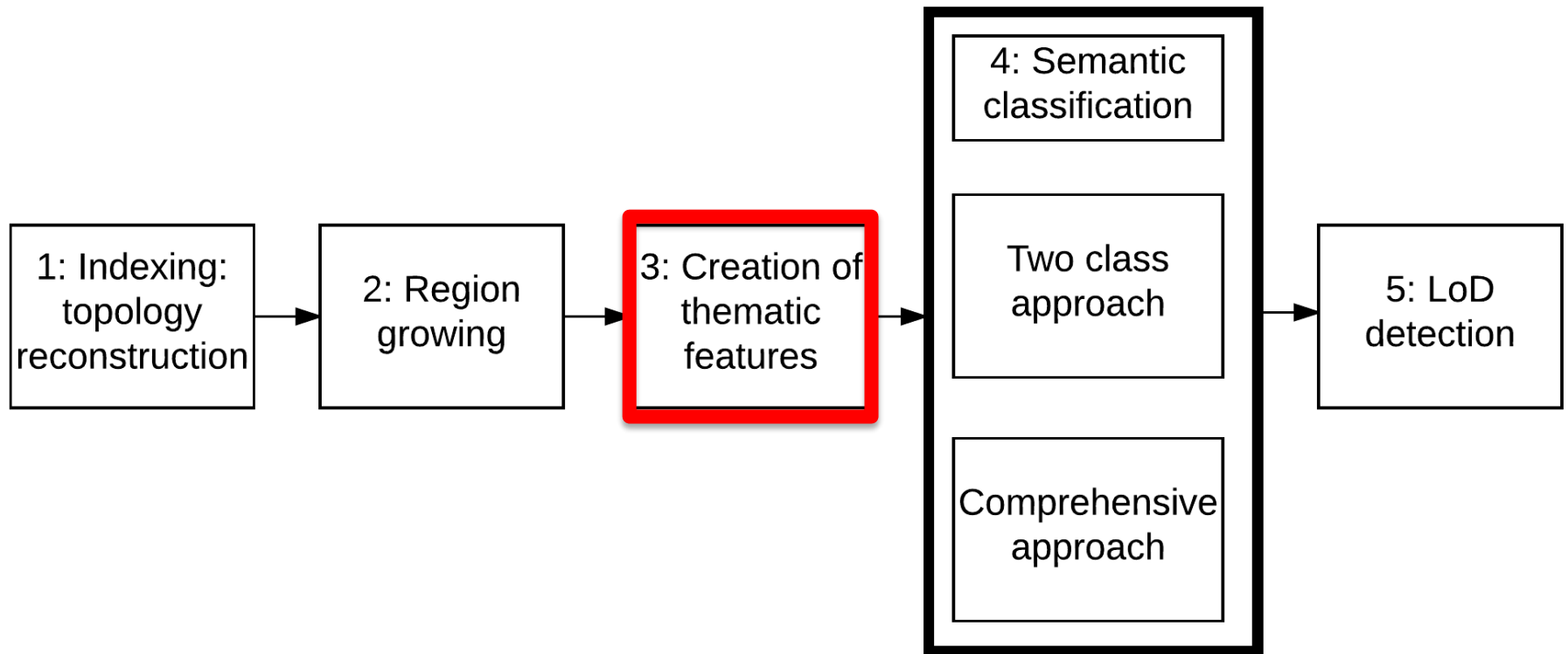


## 2 Sets, based on the semantic class: WallSurfaces

- Threshold to detect WallSurfaces is used in both approaches: implemented before the region growing
  - Threshold is based on the pitch angle of the triangle: defines verticality

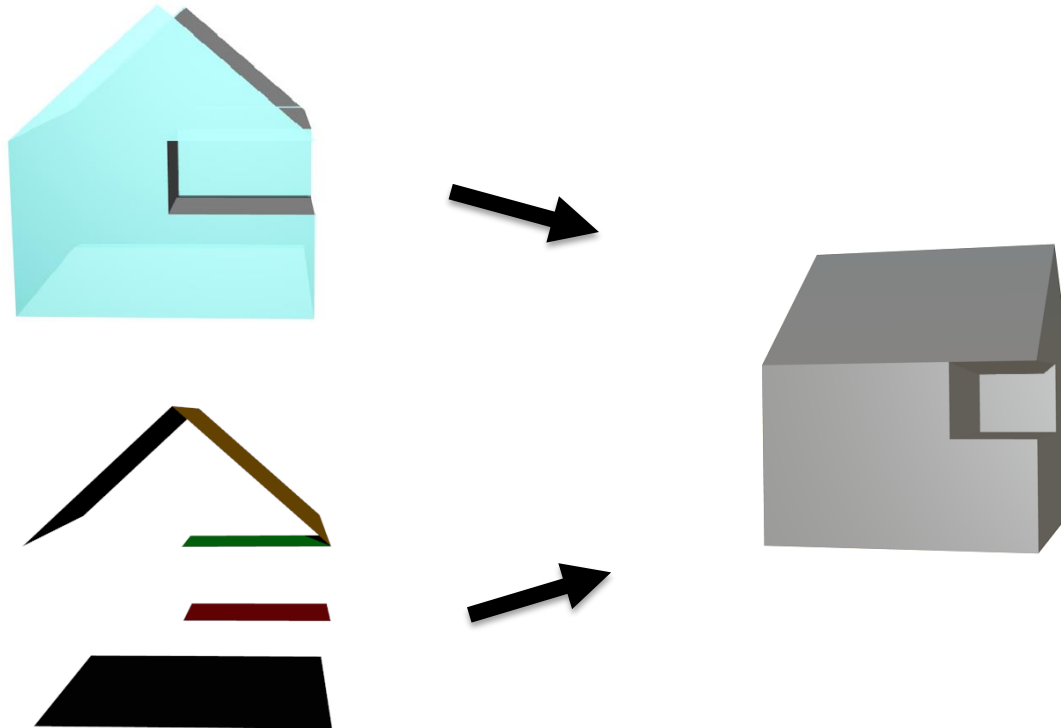


# Creation of thematic features



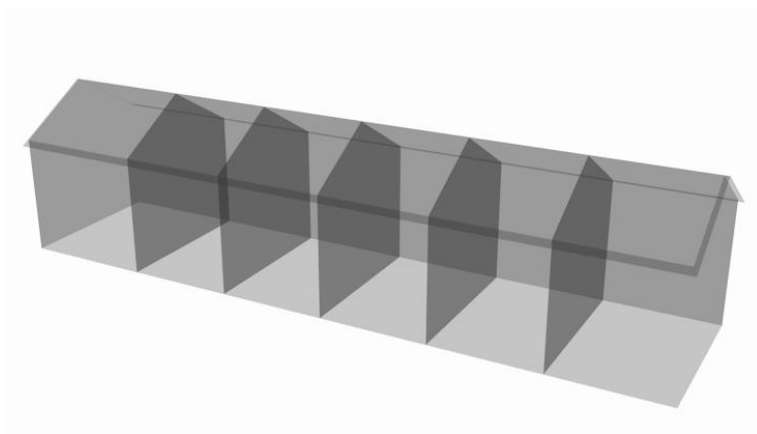
# Creation of thematic features: building reconstruction

- Adjacent regions are clustered to recompose the single building entities.

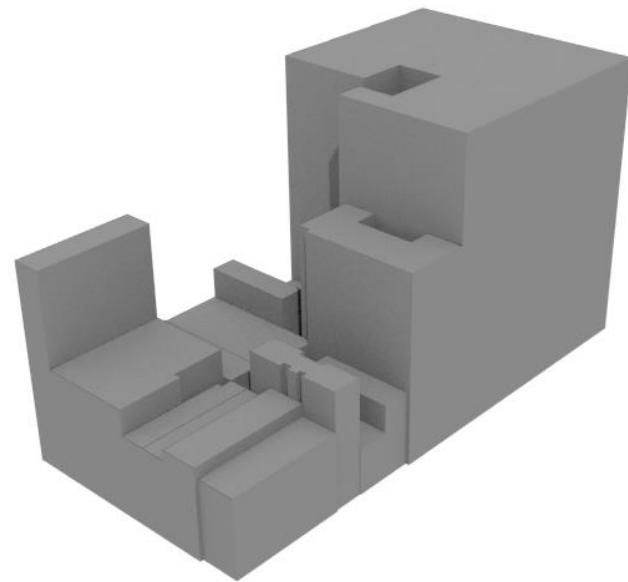


# Creation of thematic features: building reconstruction (3)

- Single building consists out of a set of connected Wall and non-Wall regions.



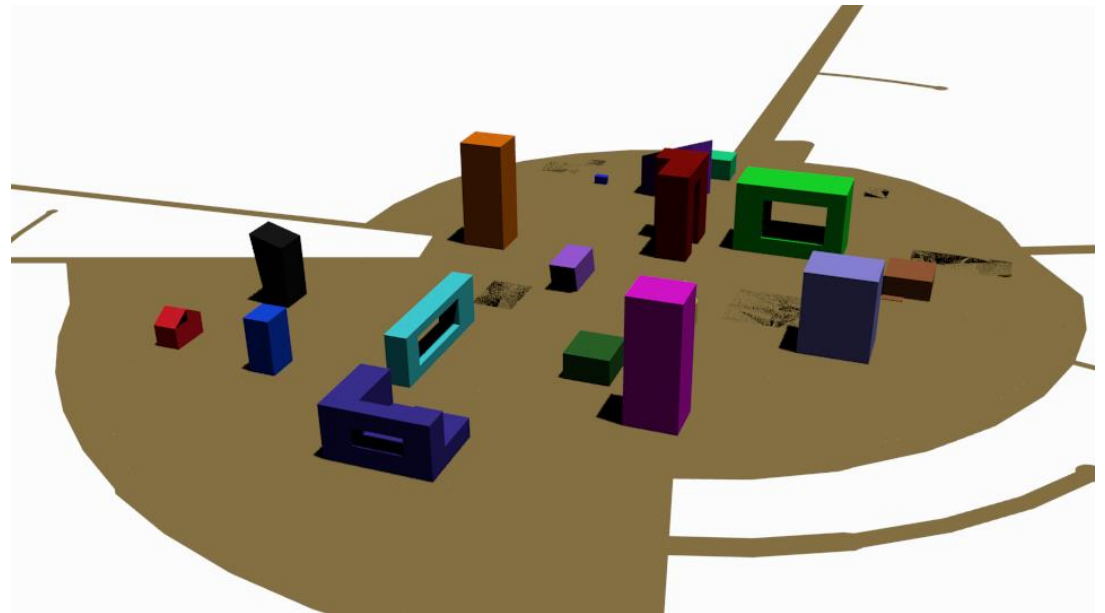
Data source: Rotterdam 3D



Data source: New York 3D

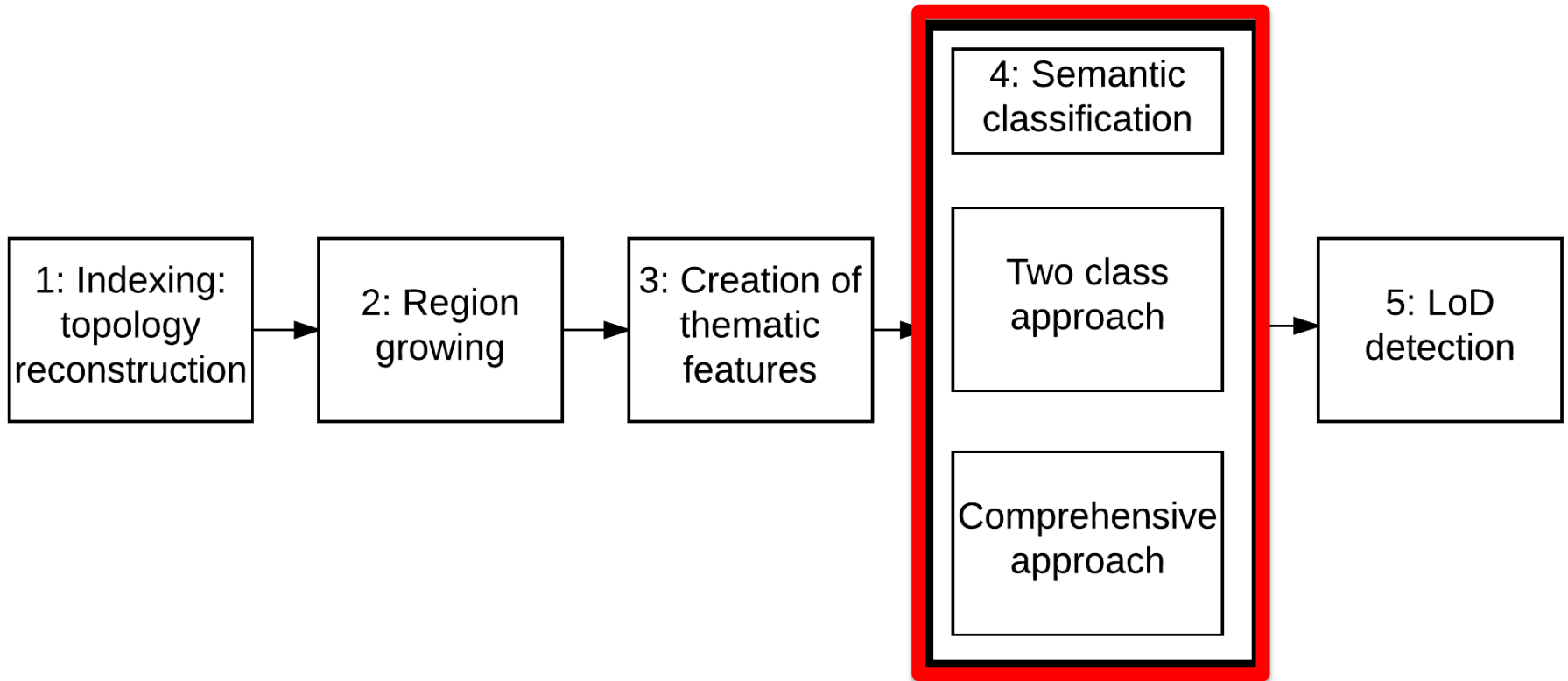
# Creation of thematic features: the terrain

- Recognition of regions which contain terrain polygons.
  - 25% of all triangles
  - Connected to more than 4 WallSurfaces



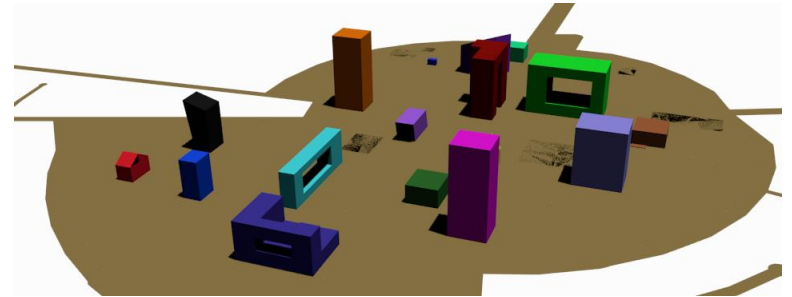


# Semantic classification

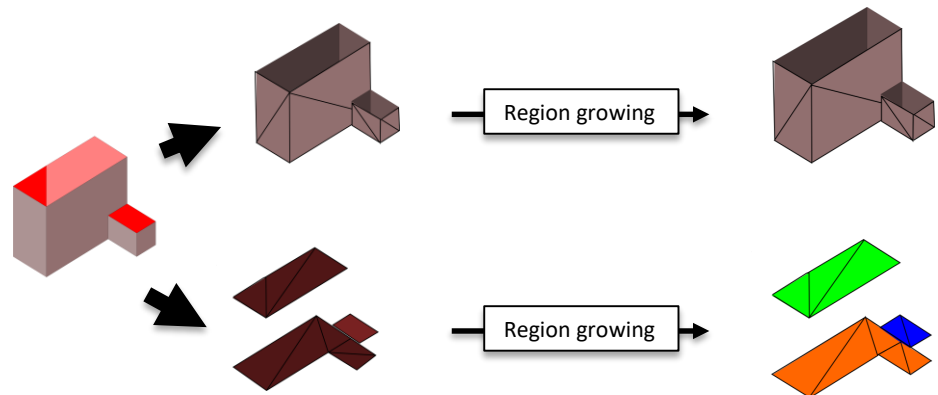


# Semantic classification

- Both approaches take the reconstructed buildings from the previous step as input

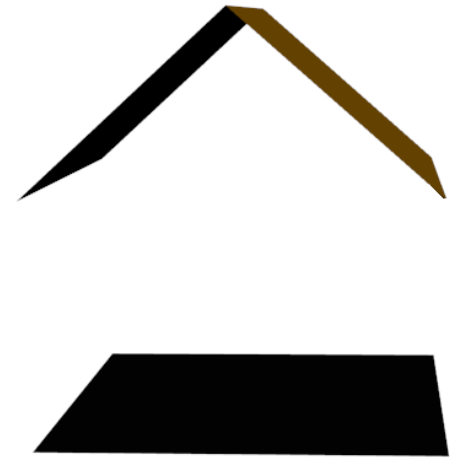


- Recognition of the WallSurfaces is used in the region growing



# Semantic classification

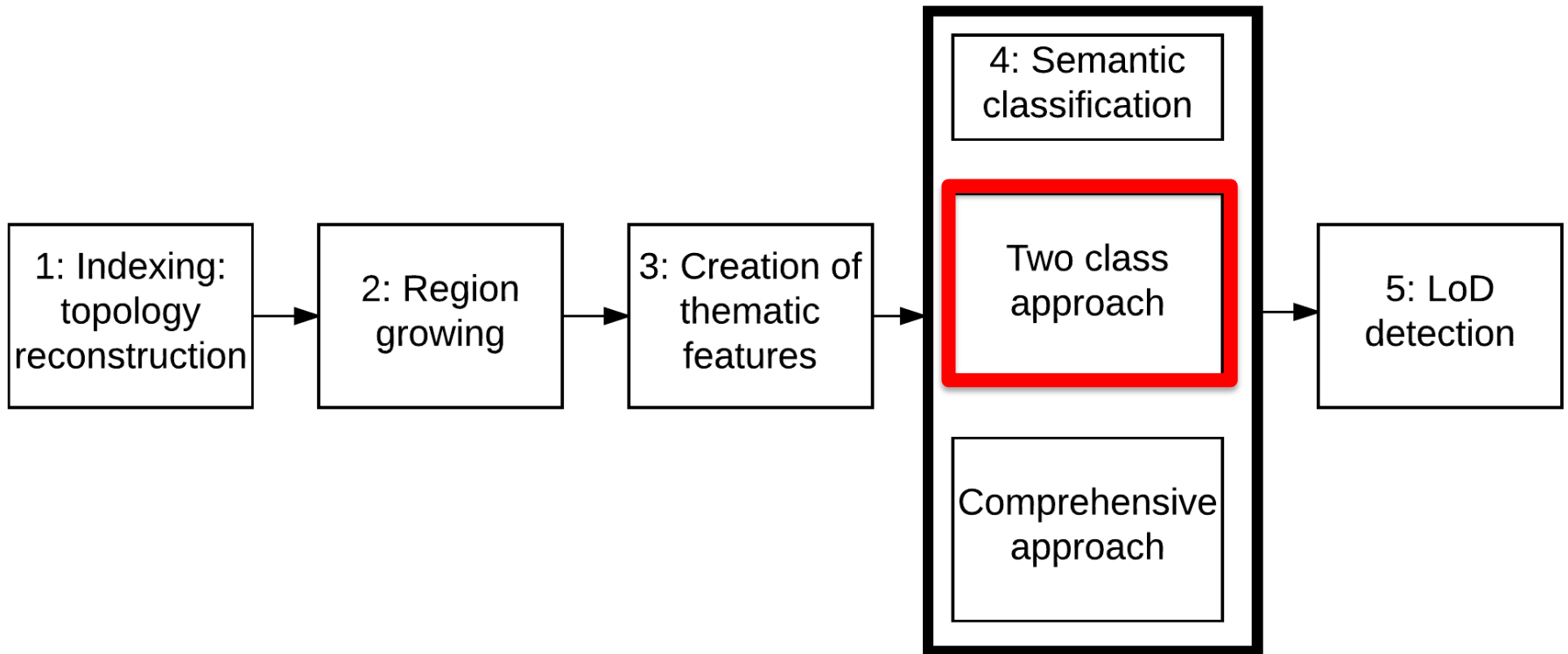
- Two approaches are implemented:
  - **Two class approach:** Limited to Roof- and GroundSurface



- **Comprehensive approach:** Additionally the classes OuterFloor- and OuterRoofSurface

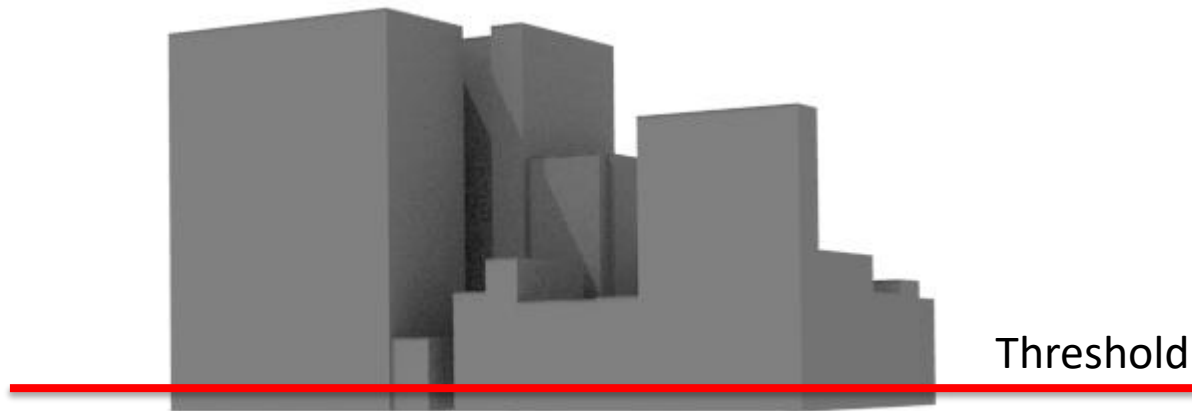


# Semantic classification



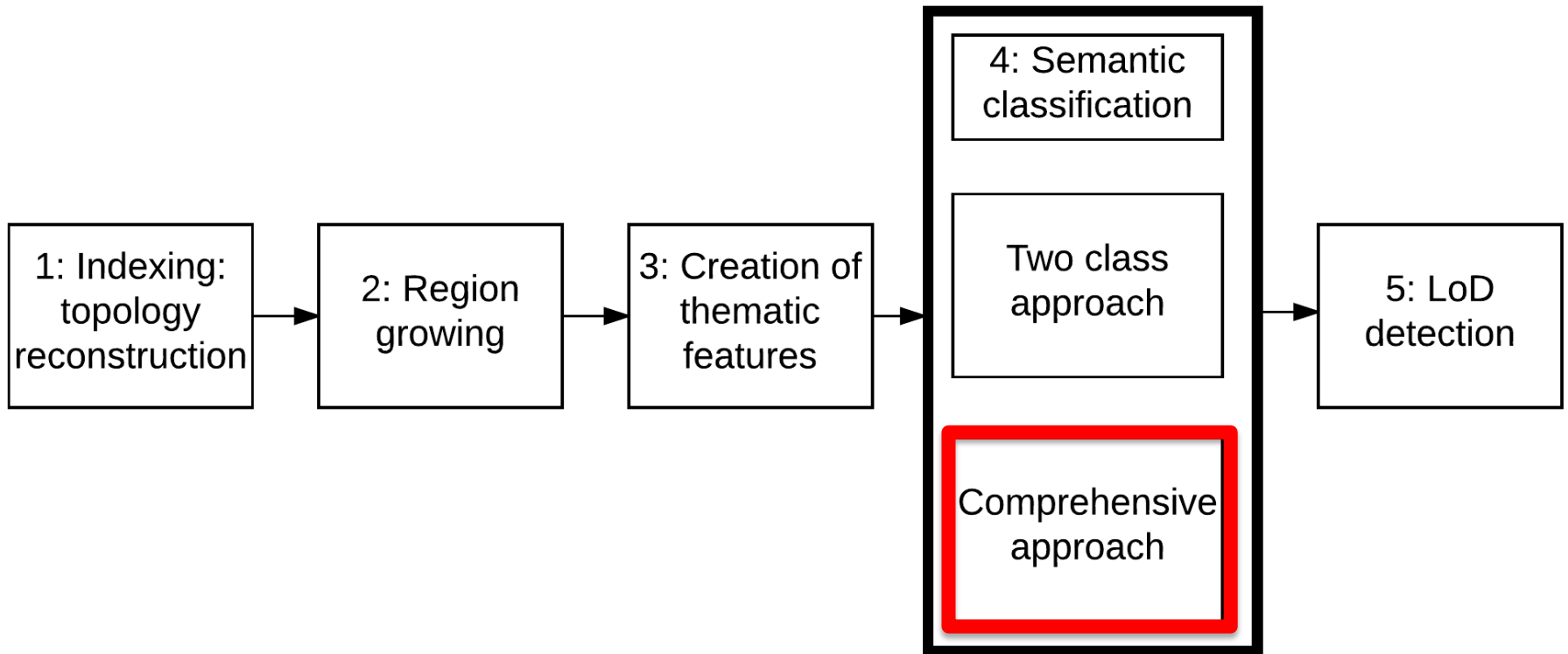
# Semantic classification: Two class approach

- Compute a threshold which is situated slightly above the GroundSurface.



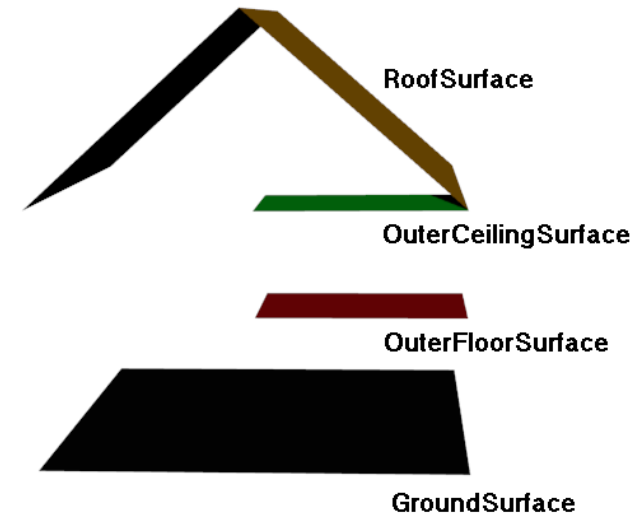
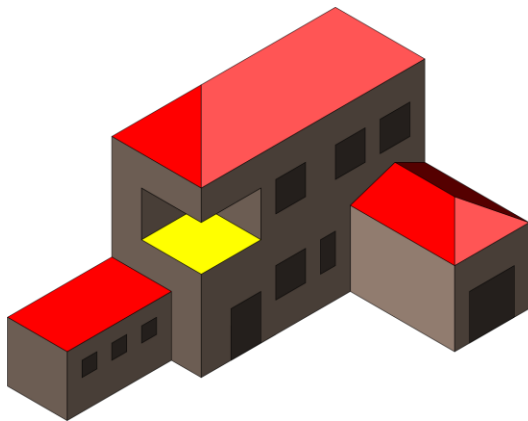
- GroundSurface is situated below the threshold, RoofSurfaces above the threshold.

# Semantic classification



# Semantic classification: Comprehensive approach

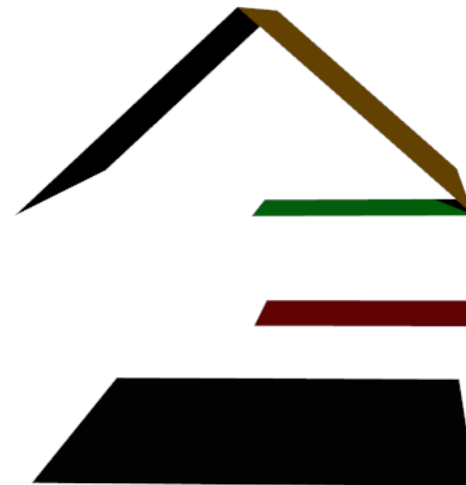
- Additional classes can not be classified based on height or surface normal



- An extension of the definitions for the semantic classes is needed

# Semantic classification: Comprehensive approach (1)

- Extensions of the definitions for the semantic classes in CityGML:
  - **RoofSurface**: Encloses a building from above.
  - **GroundSurface**: The GroundSurface encloses the building from below

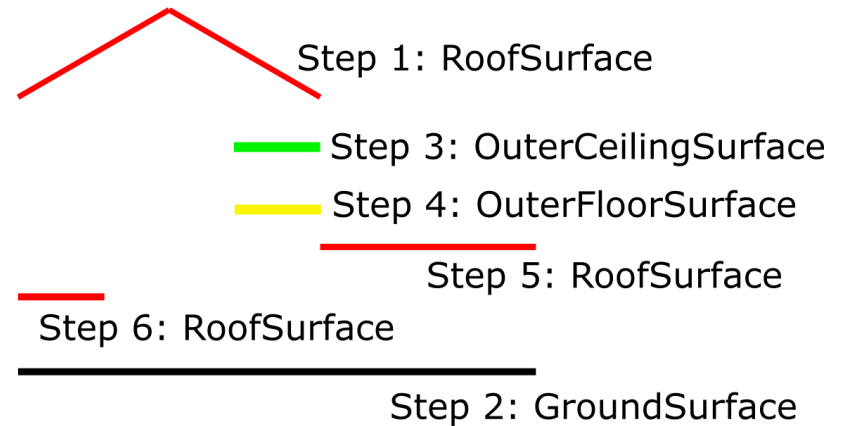
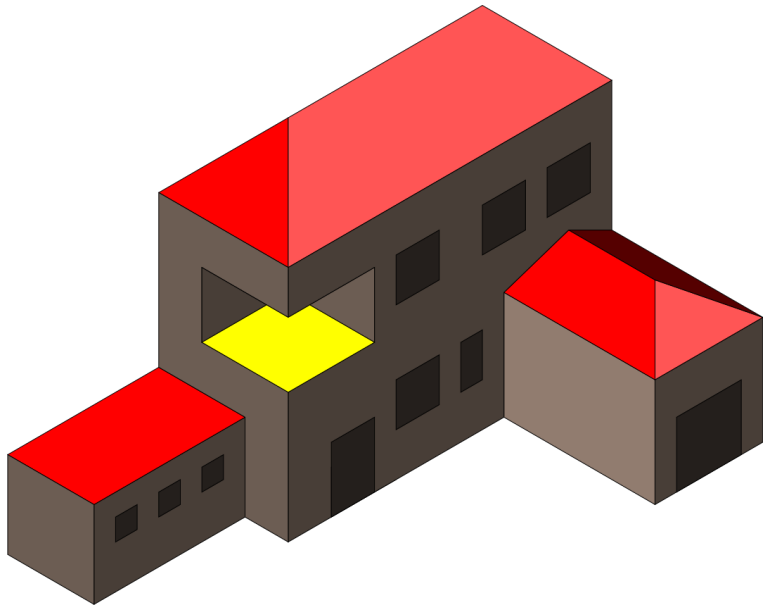




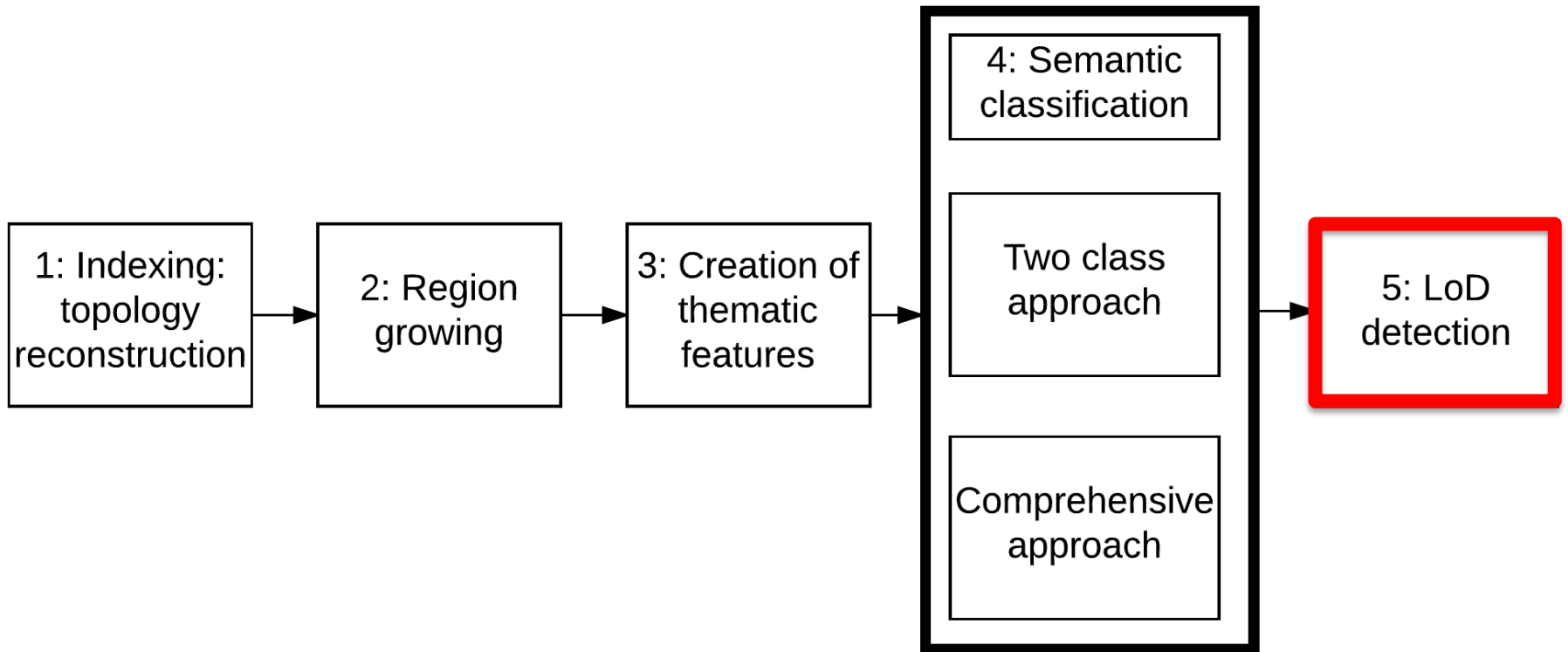
# Semantic classification: Comprehensive approach (2)

- These definitions allow for a rule based classification, based on arguments:
  1. RoofSurfaces cannot overlap
  2. GroundSurfaces cannot overlap
  3. OuterCeilingSurface and OuterFloorSurface cannot enclose a building from below or above
  4. An OuterFloorSurface cannot be present without an OuterCeilingSurface

# Semantic classification: Comprehensive approach (5)

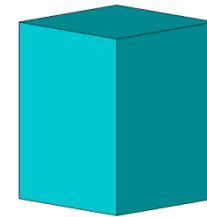


# Methodology: LoD detection

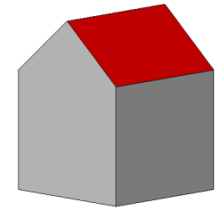


# LoD detection

- Two properties are used:
  1. The slope of the RoofSurfaces



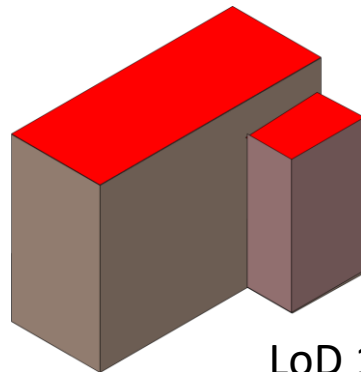
LoD 1



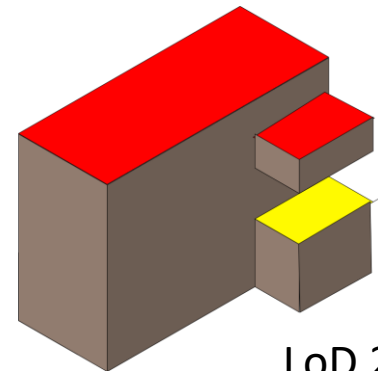
LoD 2

Source: Häfele, K. (2011)

2. The presence of the semantic classes OuterCeilingSurface and OuterFloorSurface



LoD 1



LoD 2

# Results

1. LoD detection
2. Thematic classification
3. Semantic classification:
  1. Two class approach: 5 test datasets
  2. Comprehensive approach: 3 test datasets

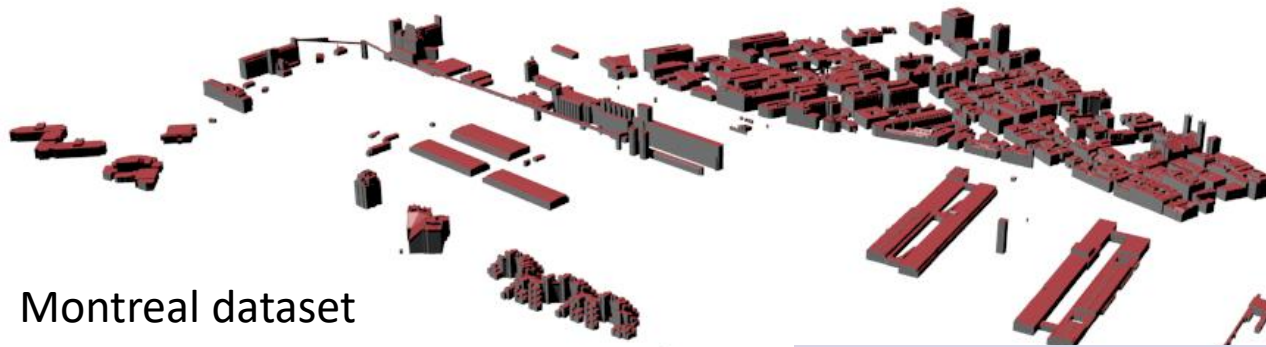
# Results: LoD detection and thematic classification

- For all models, the correct LoD is detected
- Thematic labelling:

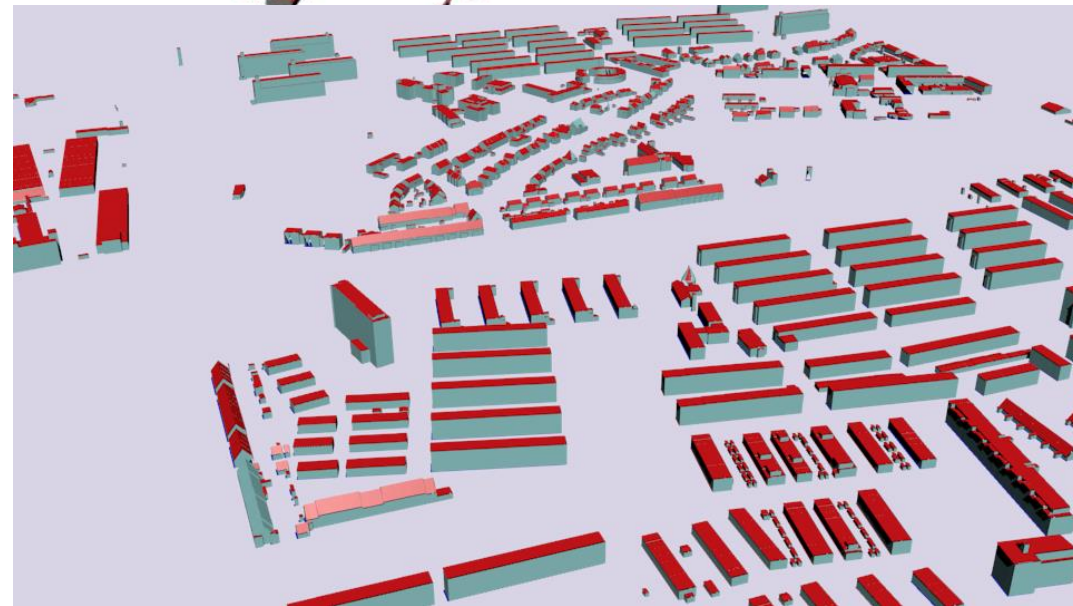
City model	Buildings in original dataset	LoD	Reconstructed buildings
Rotterdam	1544	2	507
Montreal	384	2	191
Switzerland	3151	2	2218
Waldbruke	606	2	273
New York	Unknown	1	276
CityEngine 1	6	2	6
CityEngine 2	7	2	7

# Semantic labelling, two class approach

99% accurate

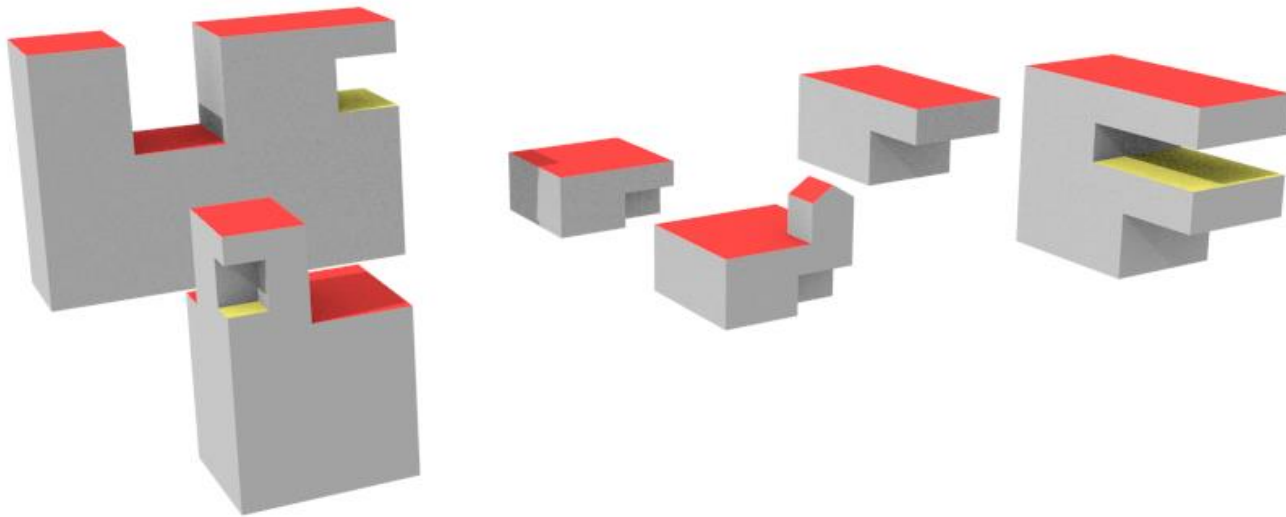


99% accurate



# Semantic labelling, comprehensive approach

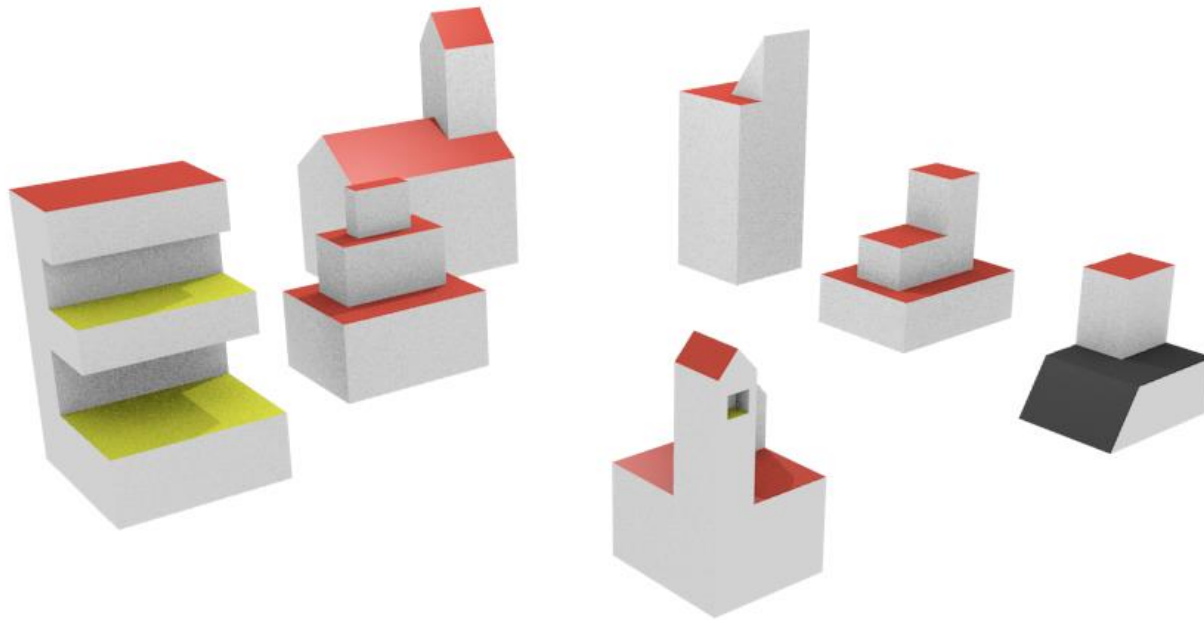
- CityEngine model 1: 100% accurate





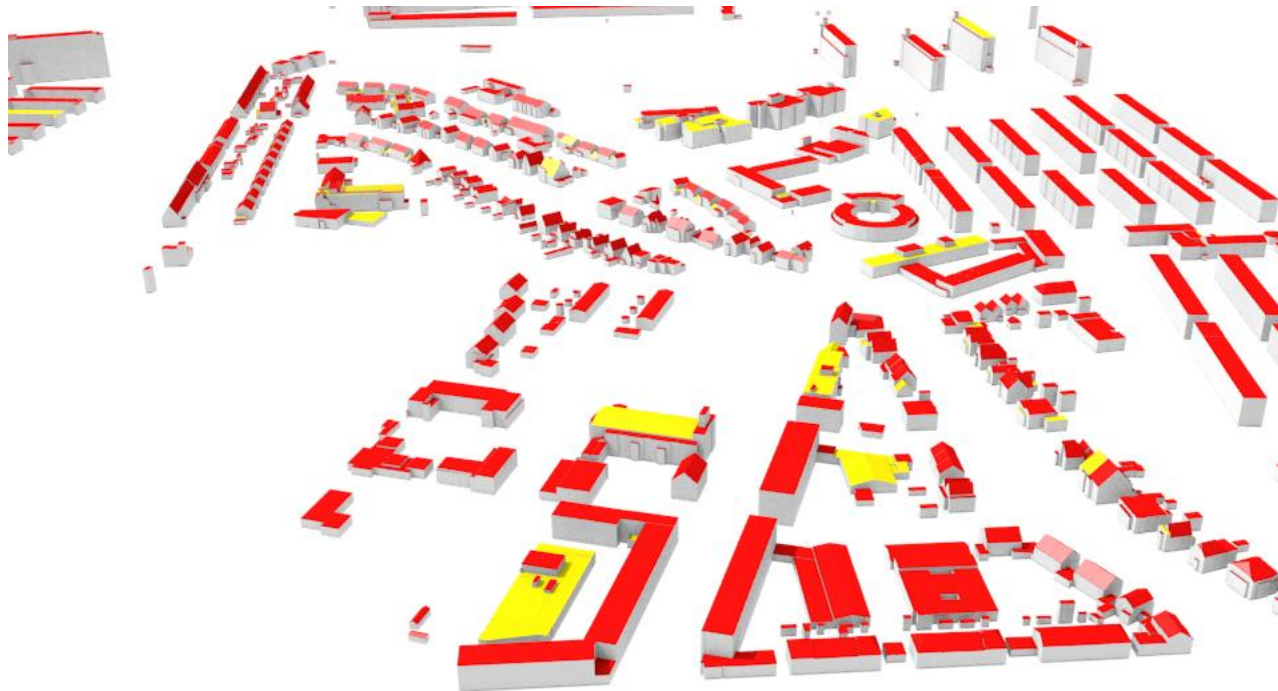
# Semantic labelling, comprehensive approach

- CityEngine model 2: 97,1% accurate



# Semantic labelling, comprehensive approach

- Rotterdam model: 97,1% accurate



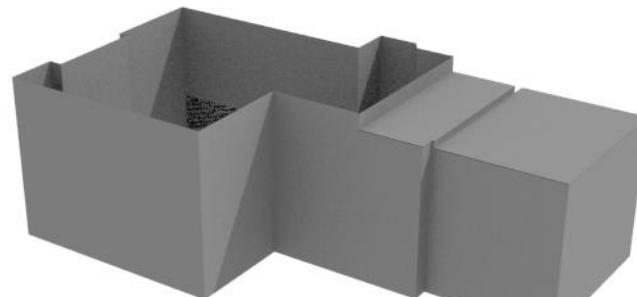
# Limitations of both approaches

- Presence of other objects



Source: Ettenheim dataset

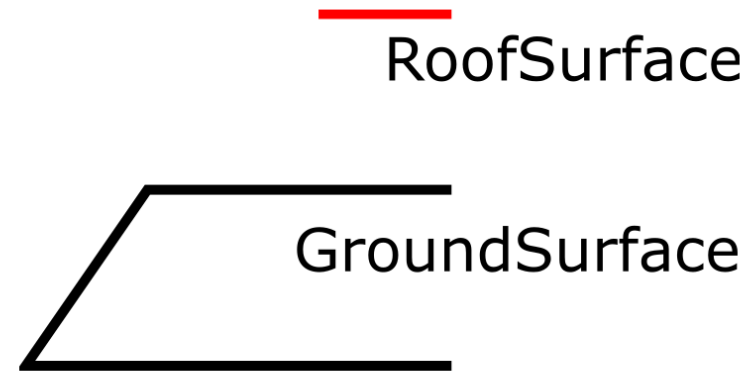
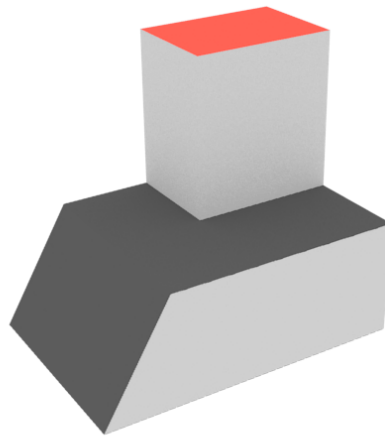
- Building reconstruction



Source: NY dataset

# Limitations of both approaches (1)

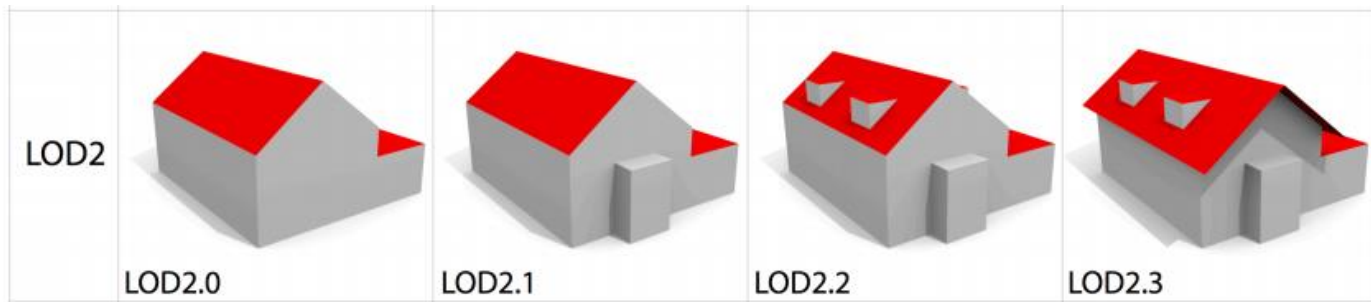
- One region can represent multiple classes



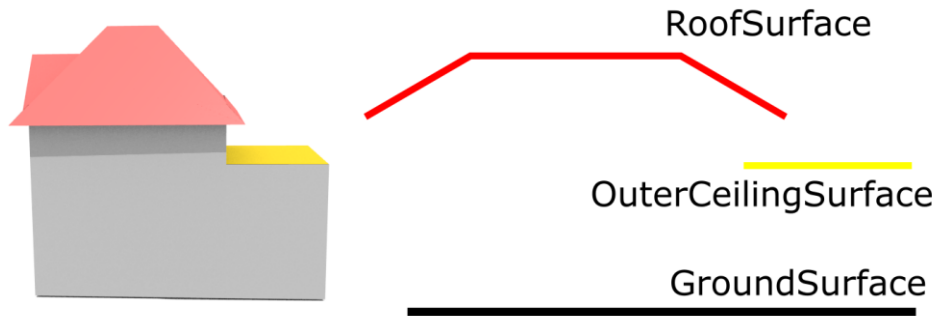
- Recognition of the terrain
  - Data dependent

# Limitations of the comprehensive approach

- Float precision errors
- Methodology works on models till LoD 2.3



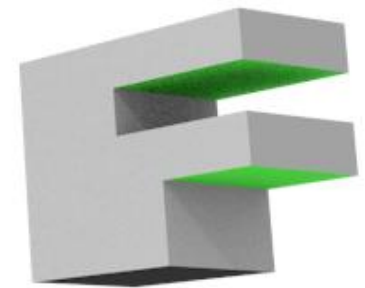
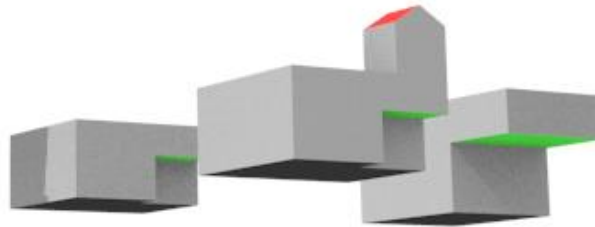
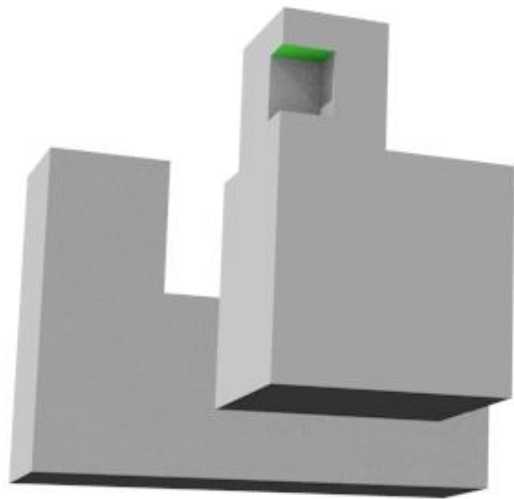
Source: Biljecki et al., 2016



# Conclusions

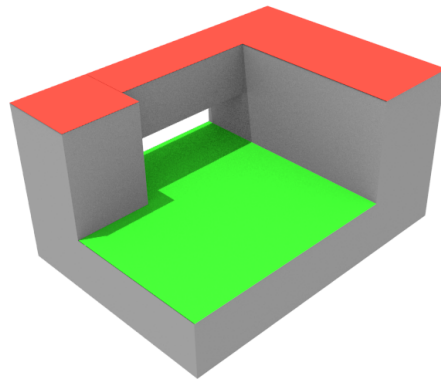
*How to automatically enrich a 3D city model with thematic and semantic information as defined in CityGML, by only utilising the models geometry?*

1. LoD detection
2. Semantic classes
3. Geometric properties
4. Remote sensing techniques
5. Thematic features
6. Accuracy

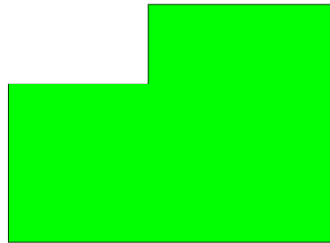


# Limitation and future work(1)

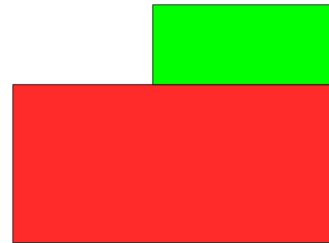
- Complete and partly overlapping surfaces



OuterFloorSurface



OuterFloorSurface

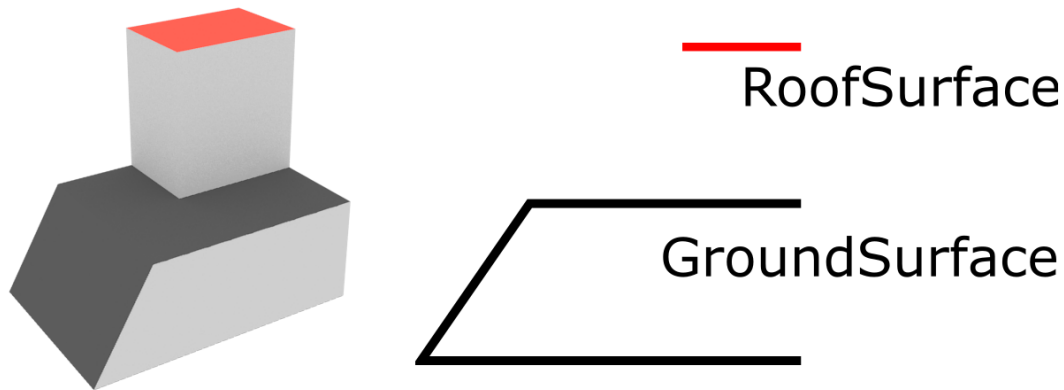


RoofSurface



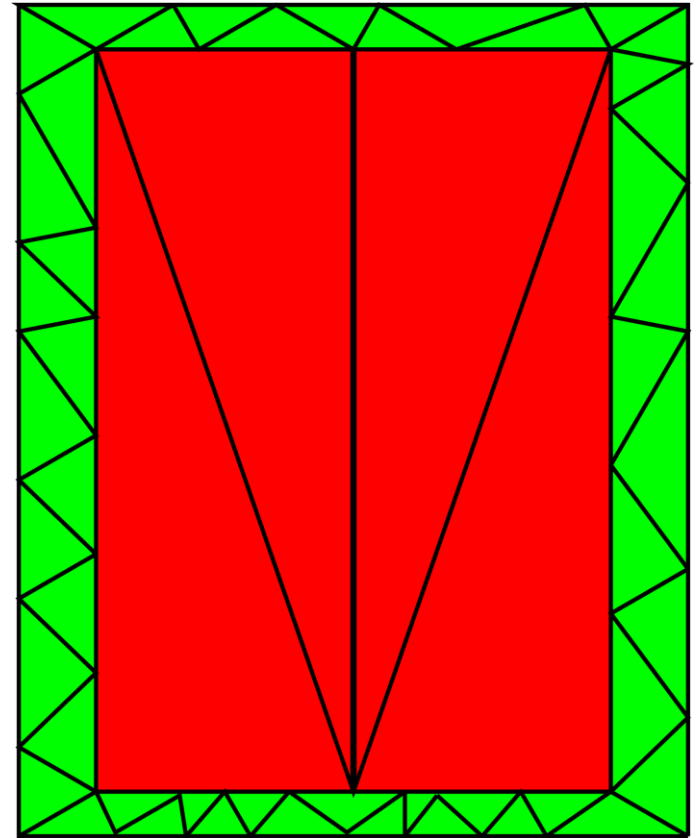
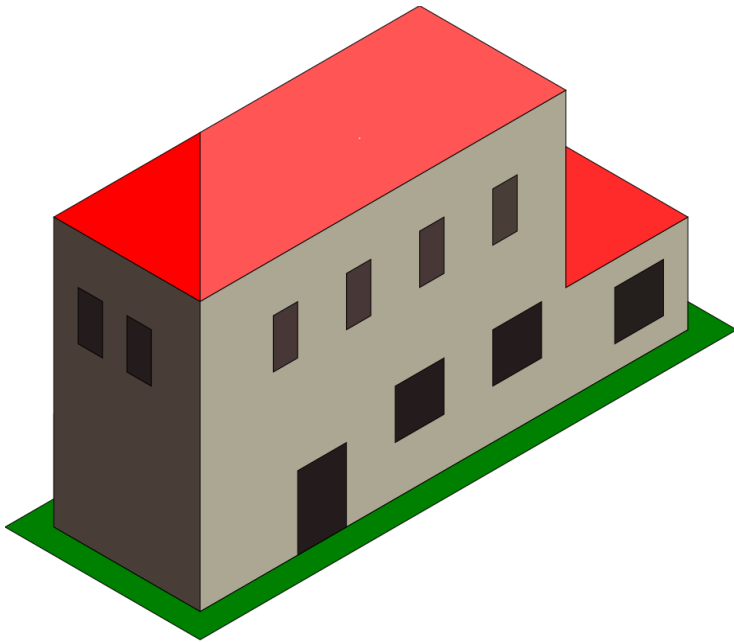
# Future work(2)

- Extend definitions of CityGML
- Improve the region growing algorithm



Source: Switzerland dataset

# Seperating the terrain from the GroundSurface



# References

- Biljecki, F., Ledoux, H., and Stoter, J. (2016). An improved lod specification for 3d building models. *Computers, Environment and Urban Systems*, 59:25–37.
- Getsi (2016) What is geodesy. [online]  
<http://serc.carleton.edu/gets/geodesy/index.html> (visited: November 1 2016)
- Häfele, K. (2011). CityGML Model of the BIEN-ZENKER Jasmin-Sun.
- SIG3D (2015). Modeling guide for 3d objects - part 2: Modeling of buildings (lod1, lod2, lod3).

# Methodology should work on all cases

