



Exploiting Indoor Mobile Laser Scanner Trajectories for Interpretation of Indoor Scenes

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Promoter: Prof. Dr. Ir. George Vosselman

Supervisor: Michael Peter



Indoor 3D Model Reconstruction to Support Disaster Management in Large Buildings

Smart Indoor Models in 3D (SIMs3D)

SIMs3D Project partners

1. STW as a technology foundation

2. Academic partners:

University of Twente (UT), EOS Department

Delft University of Technology (TUD), GIS Technology

3. Companies:

Cyclomedia Technology B.V.

Leap3D

CGI Nederland B.V. as a software advisor

4. End Users:

iNowit Brandweer Nederland as an end user and advisor for user cases

Open Geospatial Consortium (OGC) as the user of the final IndoorGML



Enabling new technology



UNIVERSITY OF TWENTE.

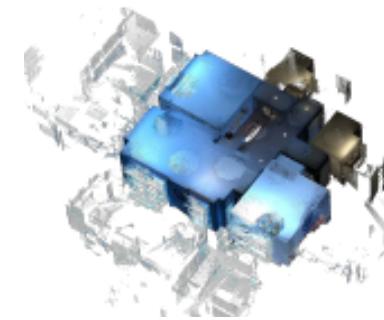


Indoor 3D Model Reconstruction to Support Disaster Management in Large Buildings

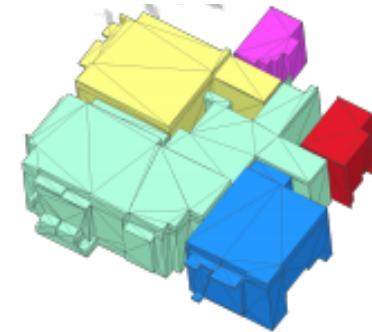
Smart Indoor Models in 3D (SIMs3D)

SIMs3D Project Goals

1. Indoor 3D reconstruction from point clouds (UT)
2. Emergency responses in public buildings (TUD)



Point clouds



3D model

Ikehata et al. 2015

Data:

Mobile Laser Scanner (MLS) point cloud

Terrestrial Laser Scanner (TLS)

Images

Microsoft Kinect



ZebRevo

Shayan Nikoohemat



Our backpack system



NavVis M3 Trolley

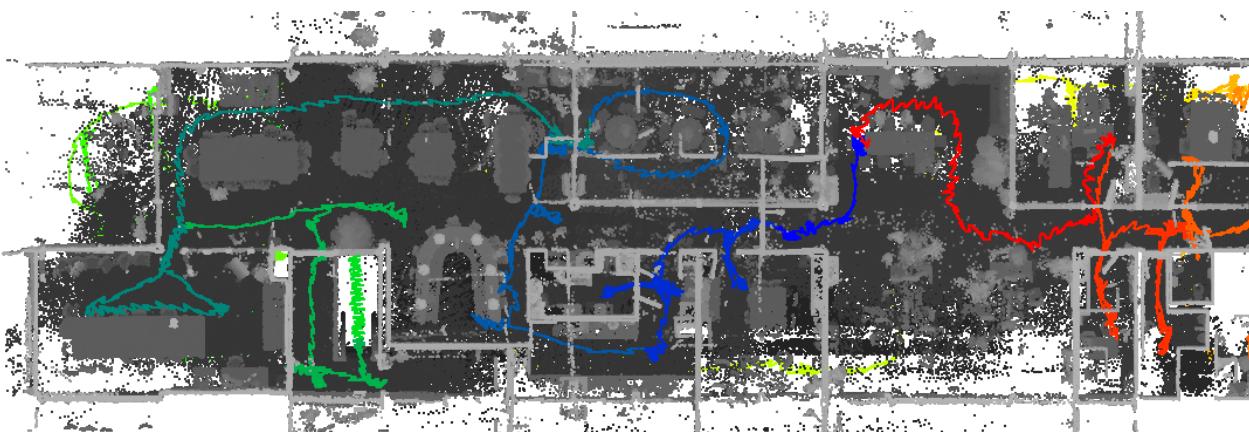
Exploiting MLS Trajectory for Interpretation of Indoor laser Scanner Data

Problem and Motivation:

- Permanent structure reconstruction, wall detection
- Room segmentation
- Opening detection from cluttered data: door, window
- Reflection from the glass surfaces



Zeb1



Shayan Nikoohemat

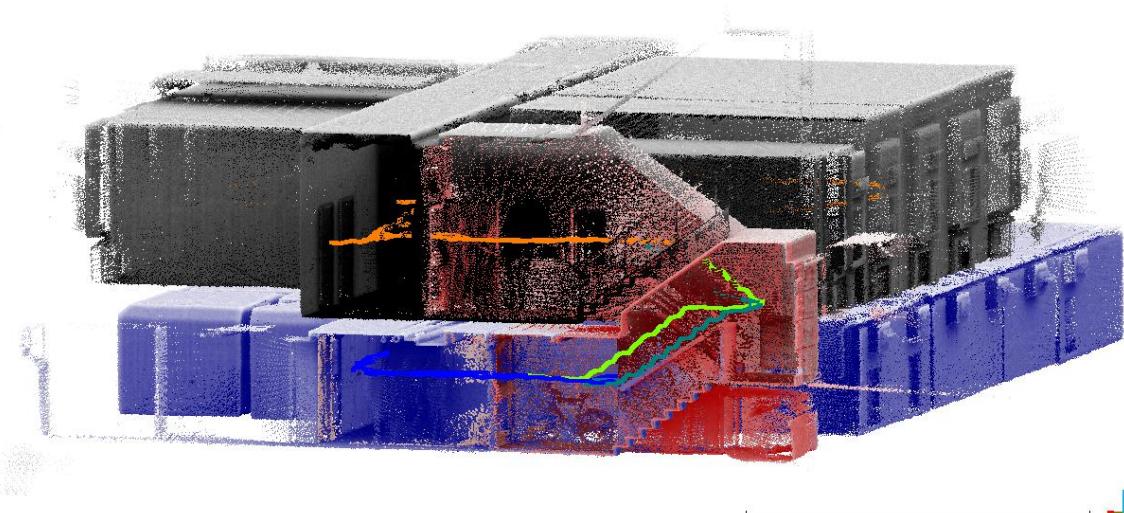


trajectory

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Problem and Motivation:

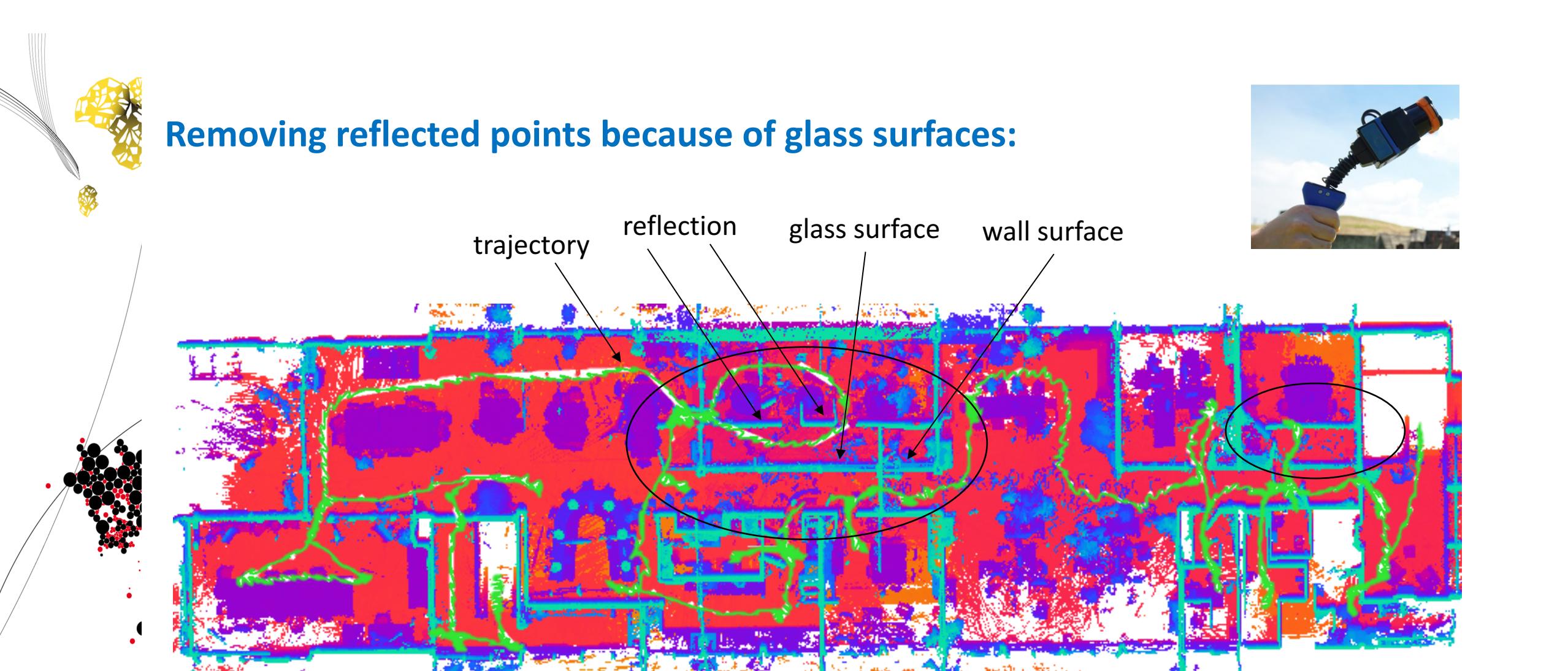
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Zeb1 point cloud
source: A. Elseicy thesis

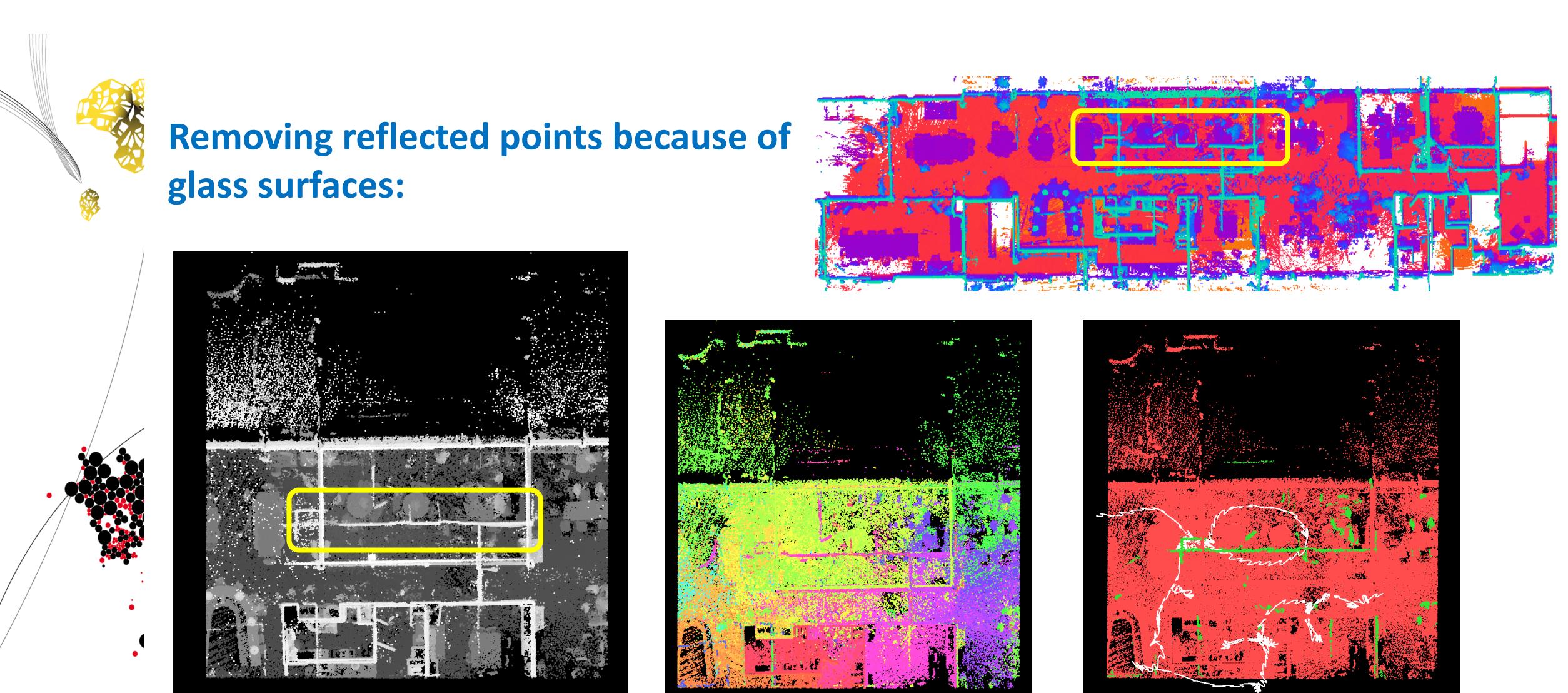


Zeb1

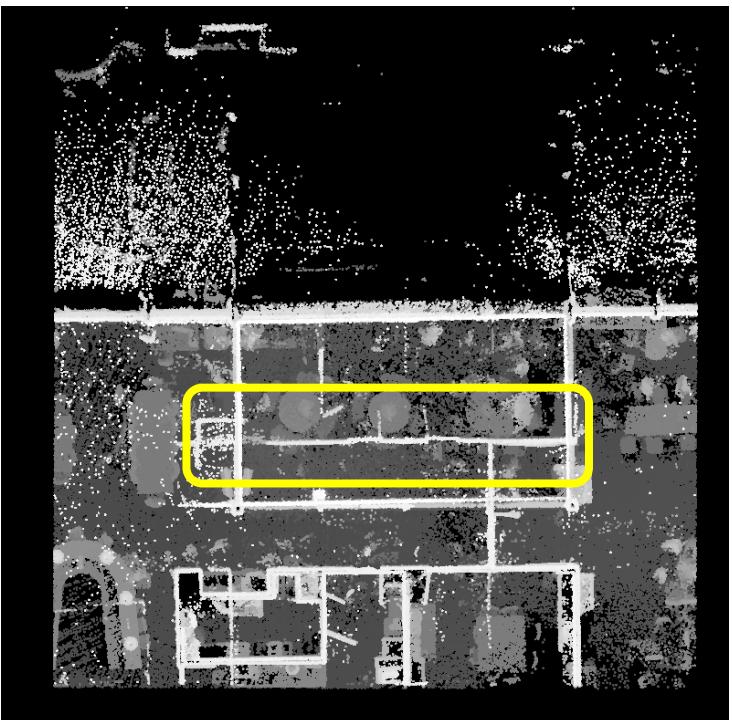


Top view of a room containing reflected surfaces.

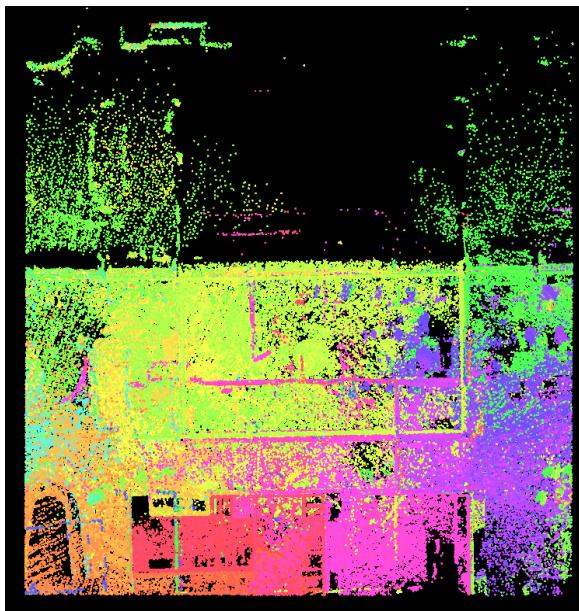
Points are colored by height.



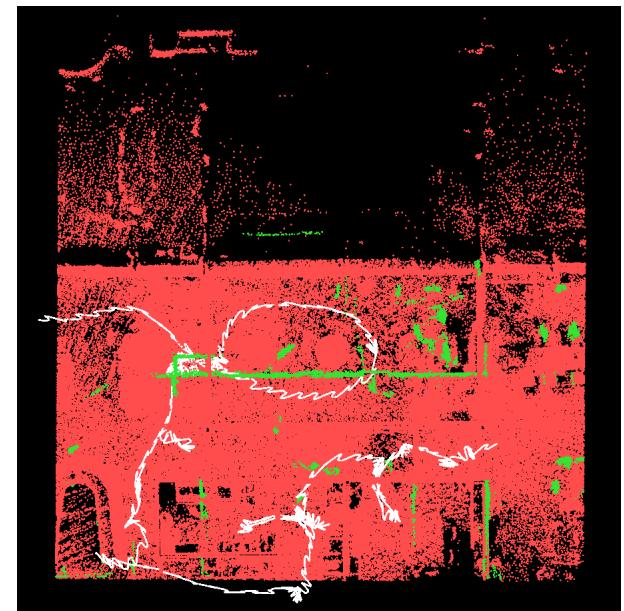
Removing reflected points because of glass surfaces:



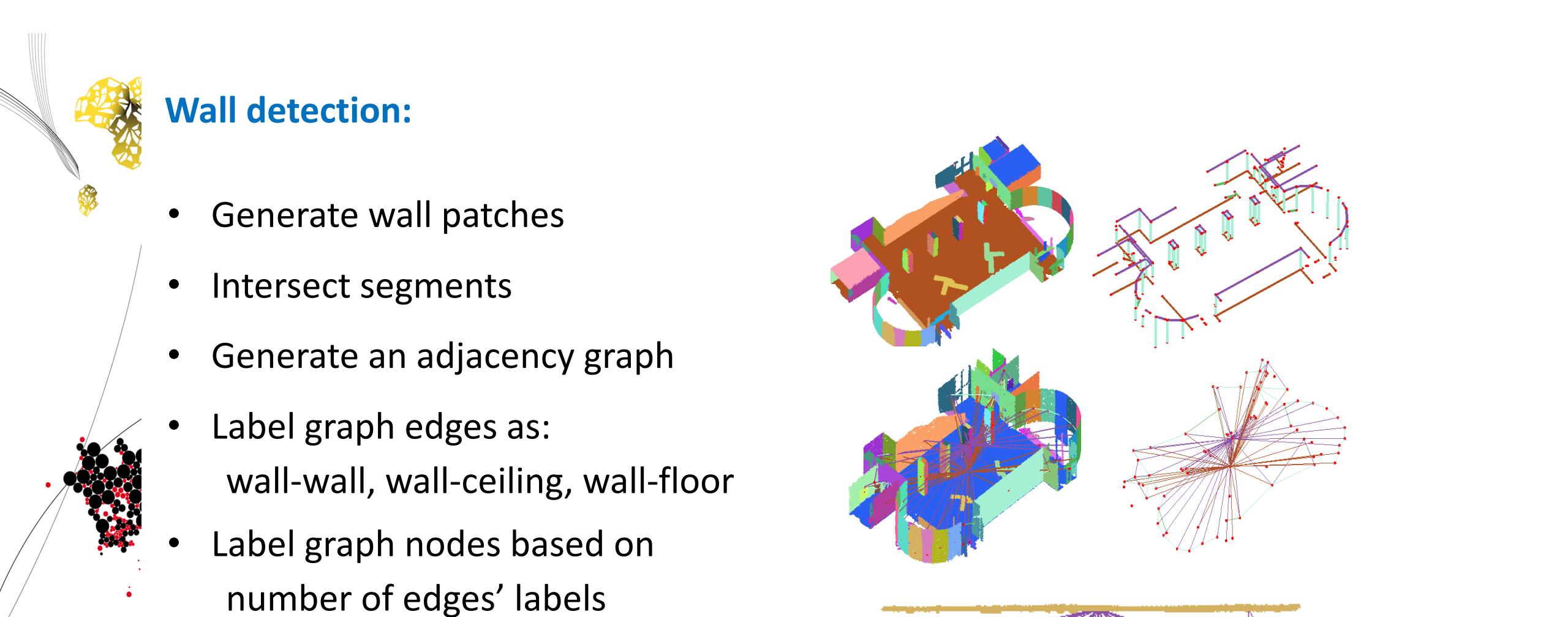
Top view of a room with reflected surfaces, yellow area.



Top view of the same area, colored by time.

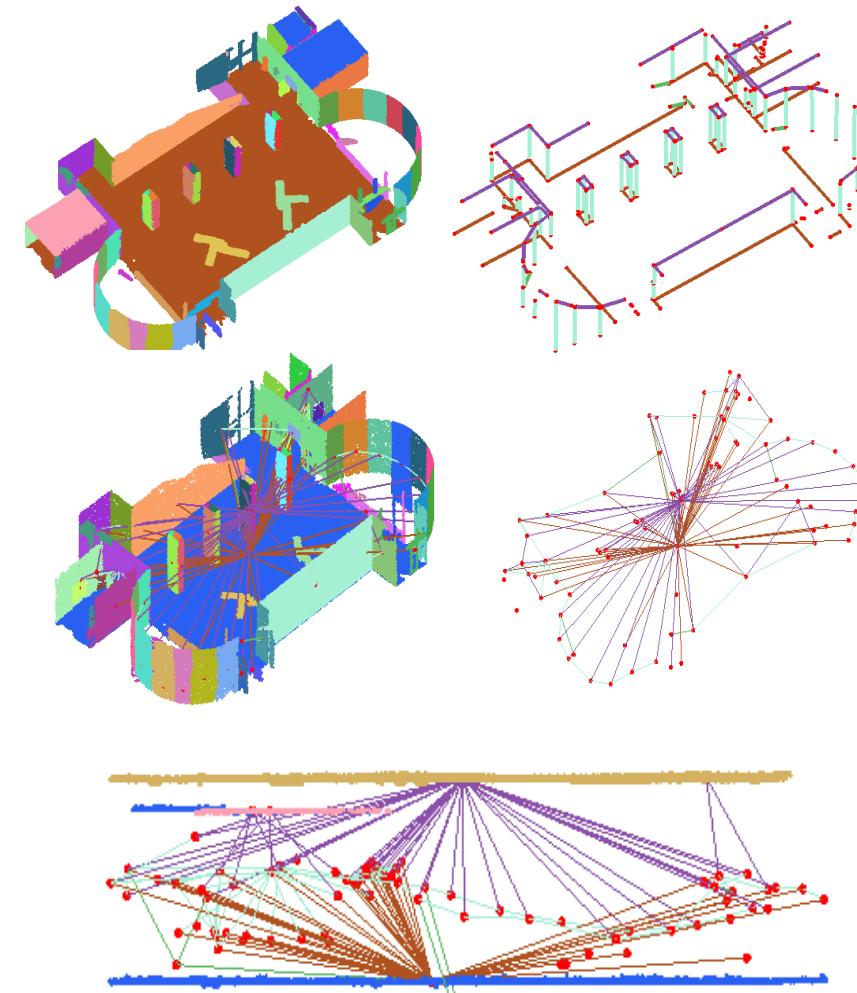


Top view. Reflected segments are green.



Wall detection:

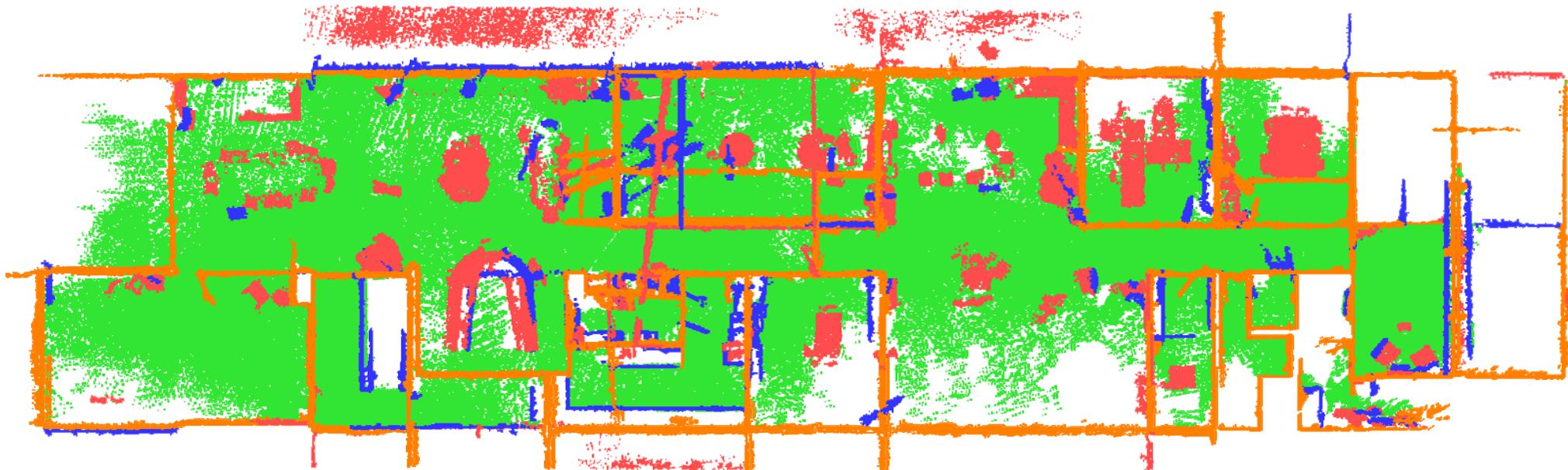
- Generate wall patches
- Intersect segments
- Generate an adjacency graph
- Label graph edges as:
wall-wall, wall-ceiling, wall-floor
- Label graph nodes based on
number of edges' labels



Side view of the graph

Exploiting MLS Trajectory for Interpretation of Indoor laser Scanner Data

Wall detection result:



Zeb1 data from Fire brigade building (top view)



Exploiting MLS Trajectory for Interpretation of Indoor laser Scanner Data

Wall detection result:



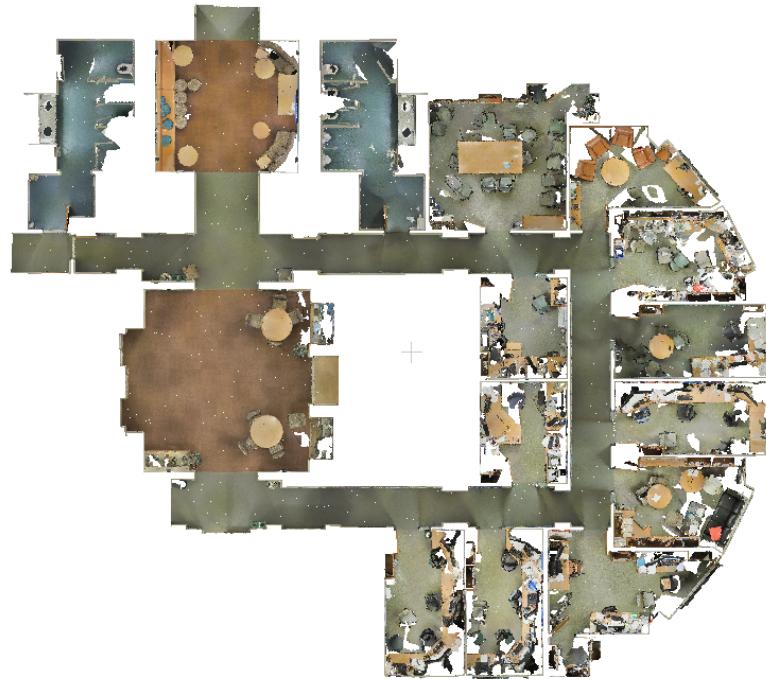
Detected walls from 1st iteration (top view)



Detected walls from 2nd iteration (top view)

Exploiting MLS Trajectory for Interpretation of Indoor laser Scanner Data

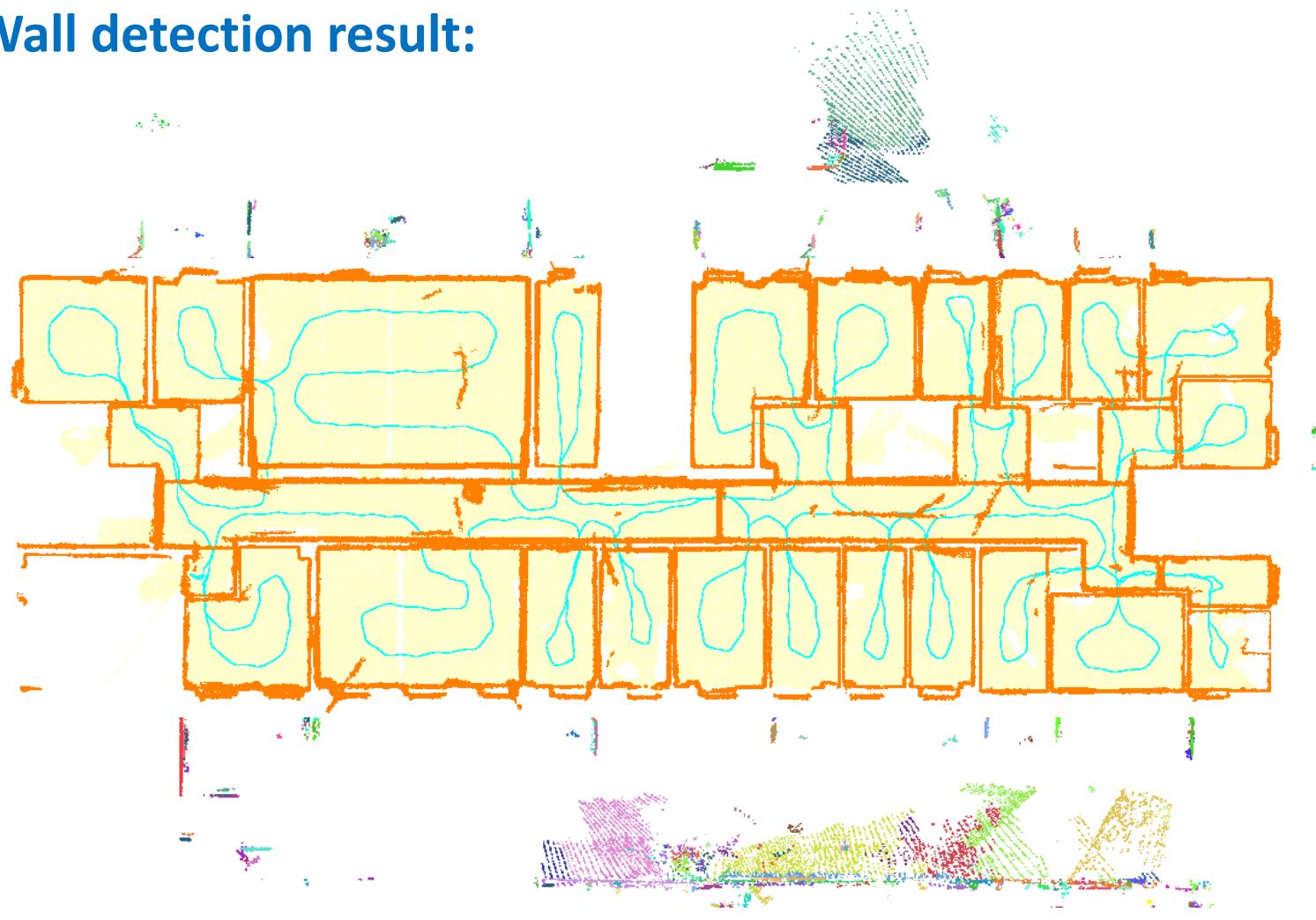
Wall detection result:



Matterport system

Matterport data from Stanford
Compared with ground truth (top view)
noise: 1 cm, 2 mil points

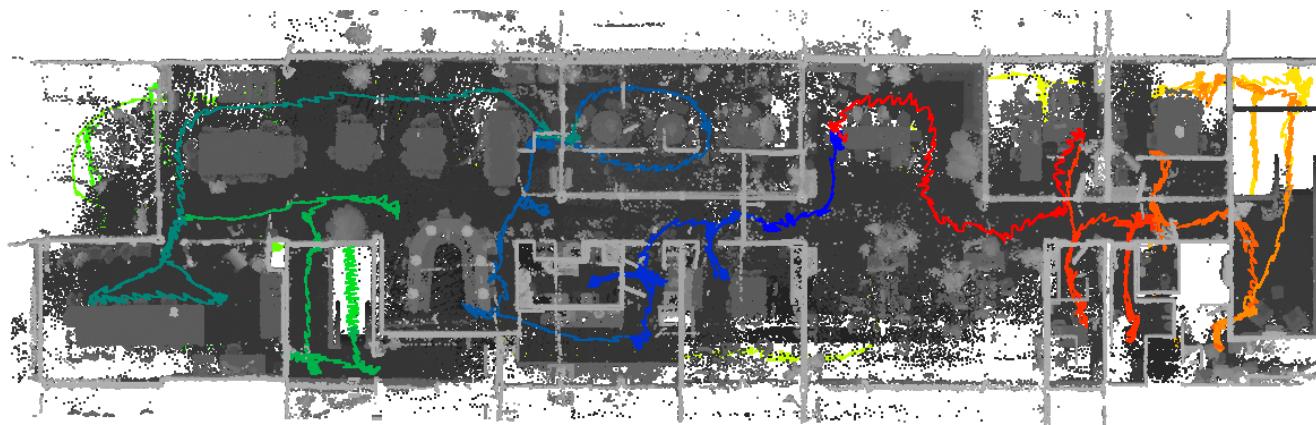
Wall detection result:



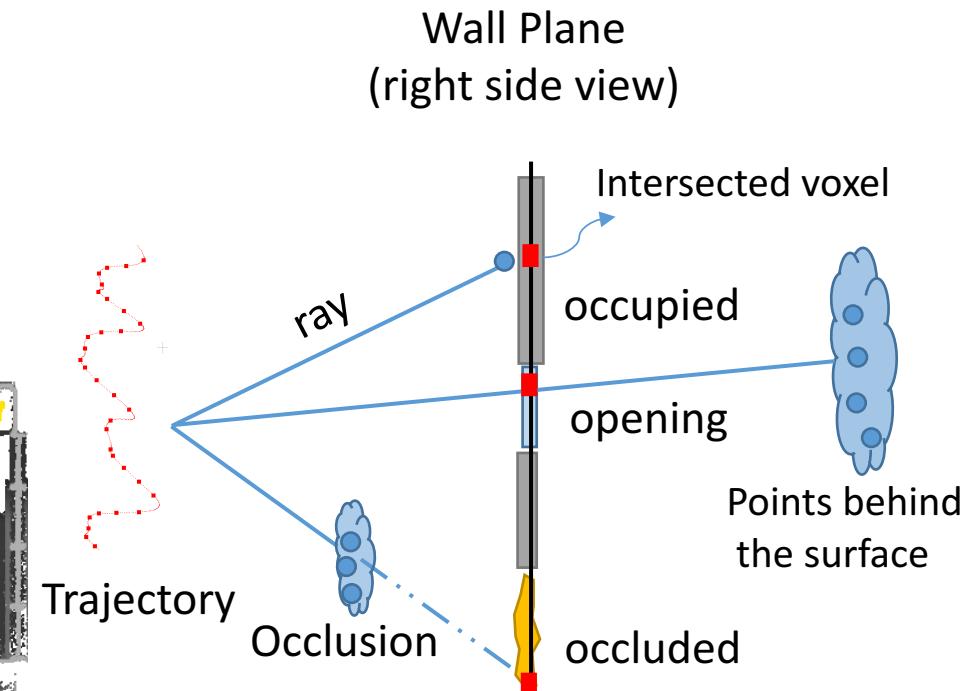
Exploiting MLS Trajectory for Interpretation of Indoor laser Scanner Data

Opening detection: Opening detection using occlusion test

- Point clouds from Zeb1
- MLS trajectory as sensor position

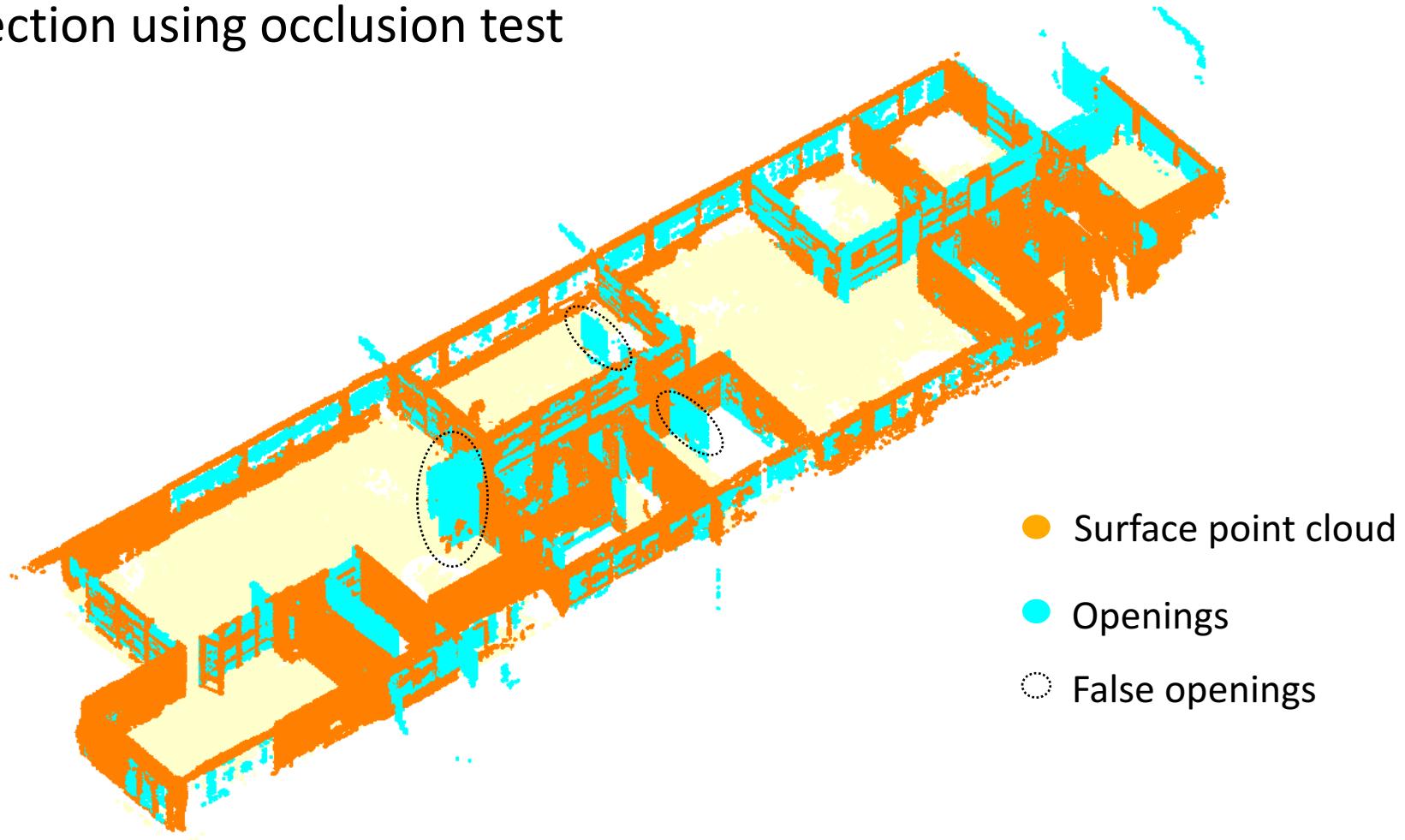


Zeb1 data from Fire brigade building (top view)



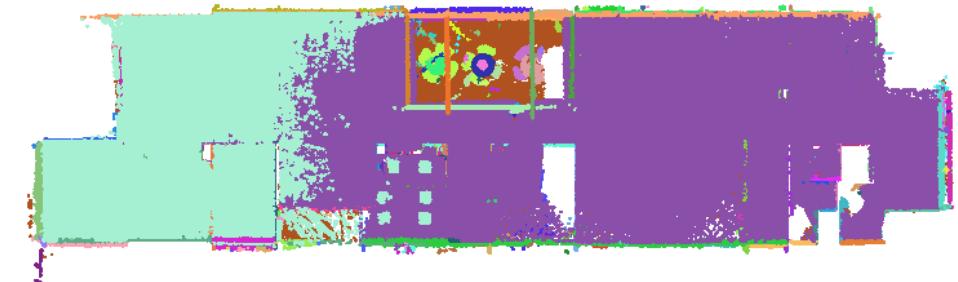
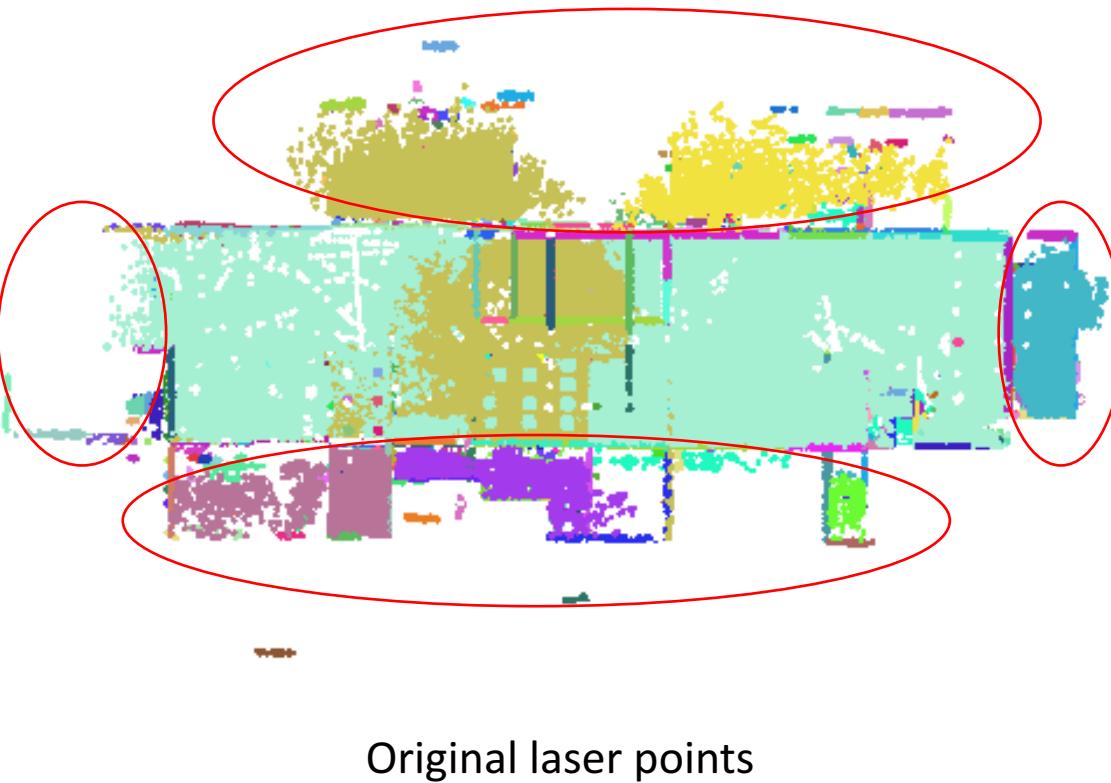
Opening detection :

Opening detection using occlusion test



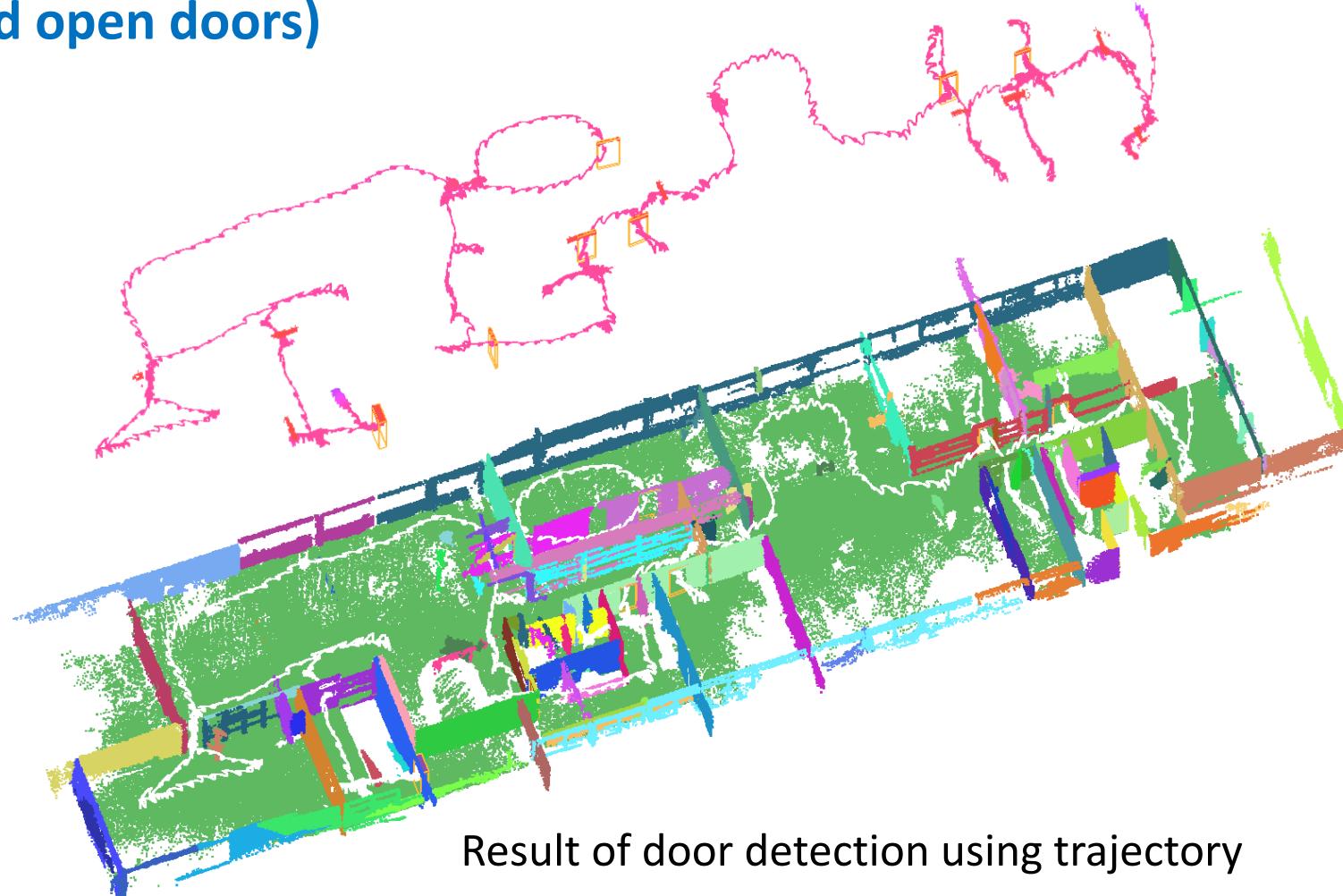
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Opening detection : modified laser points



Exploiting MLS Trajectory for Interpretation of Indoor laser Scanner Data

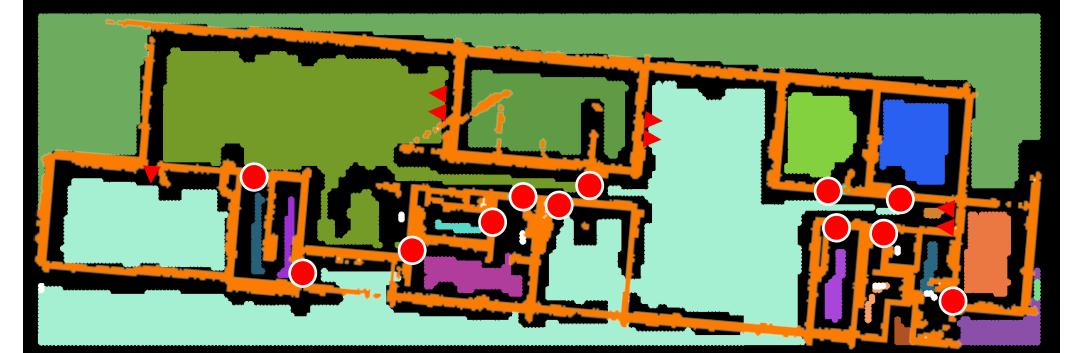
Door detection (closed and open doors)



Space partitioning and navigable space using voxels



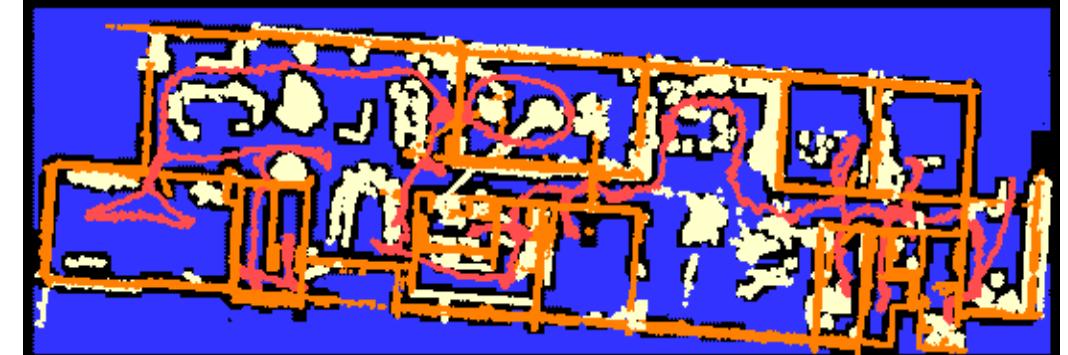
Space partitions



Space partitions, walls and doors

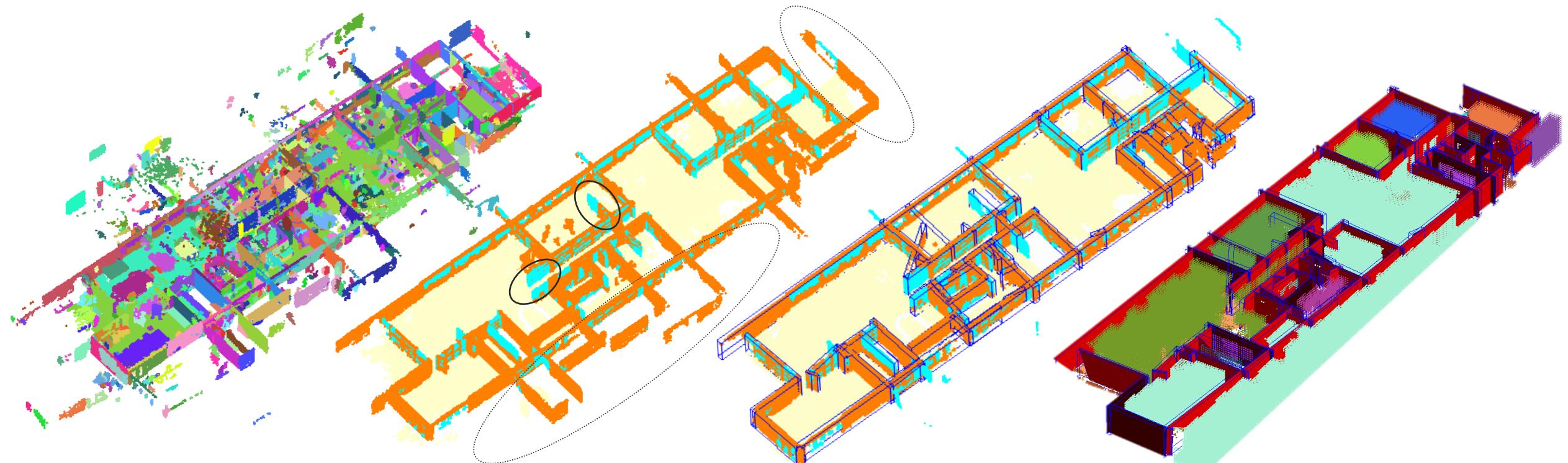


Space partitions and ground truth walls



Space partitions and navigable space

All results together



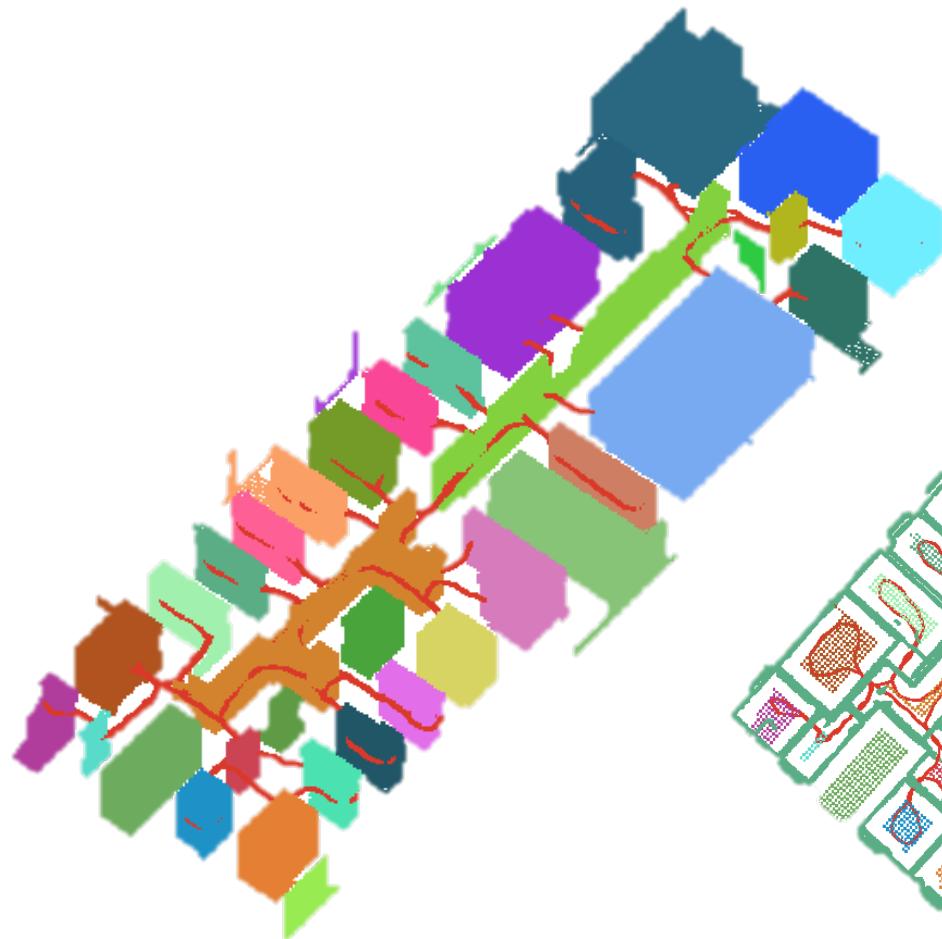
Point clouds

Openings (cyan color)

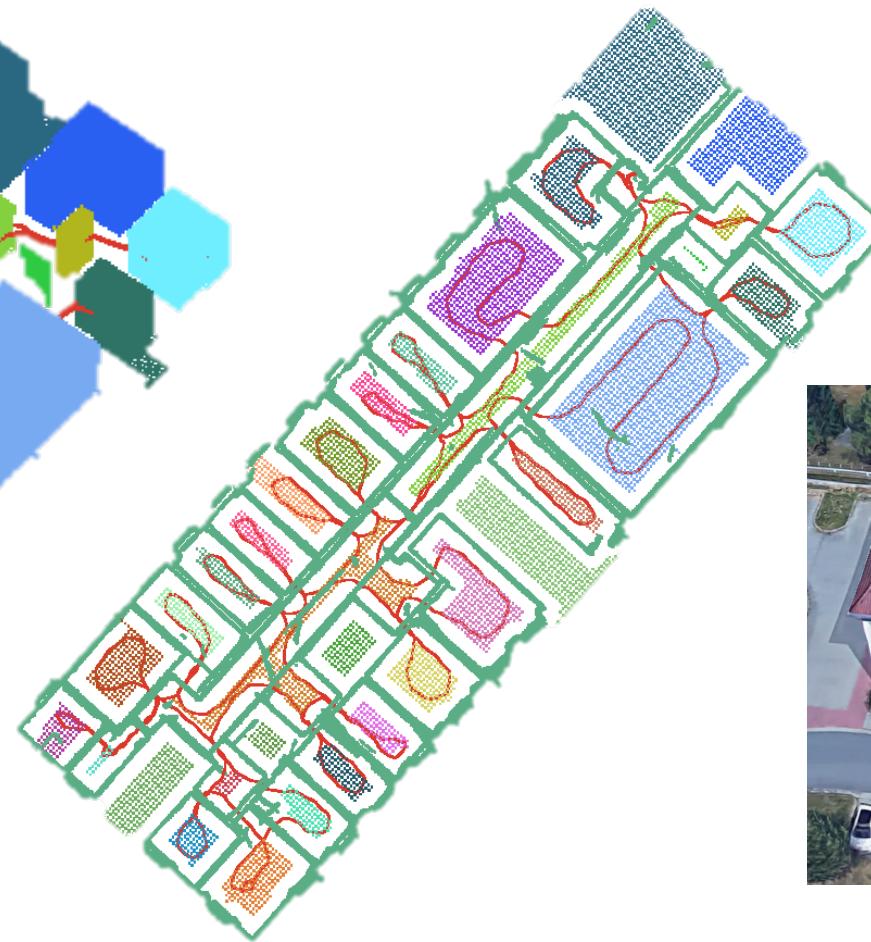
Wall Boxes

Space Partitions

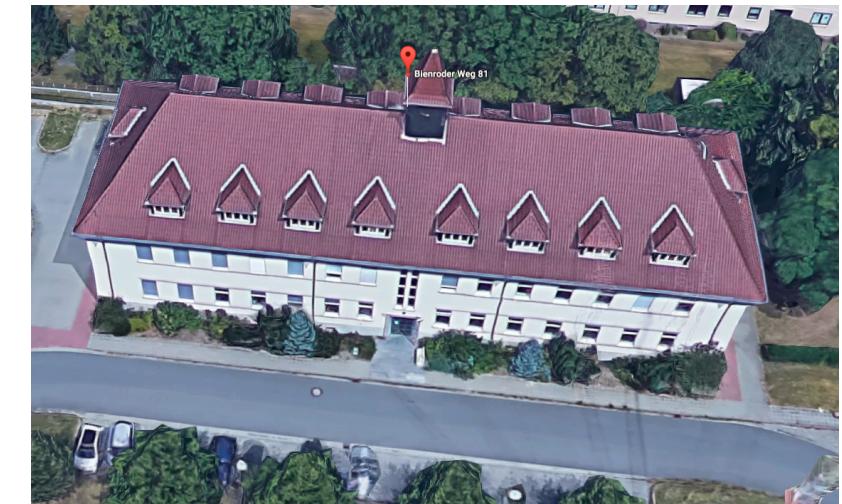
Backpack system data



Space Partitions and trajectory

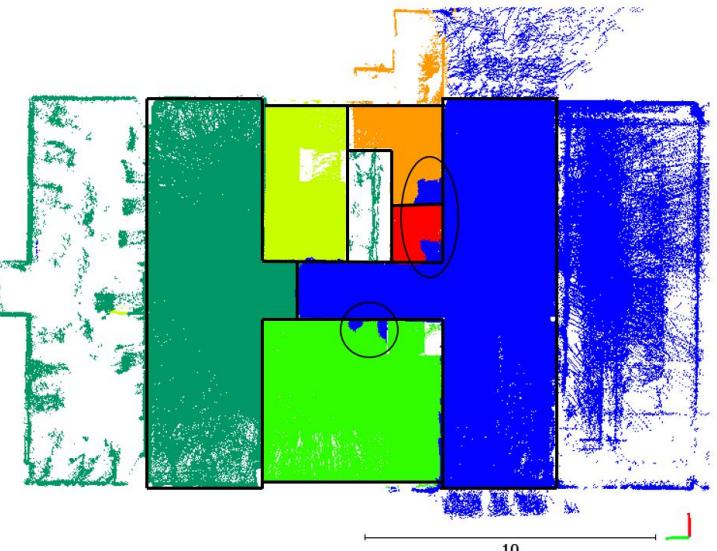
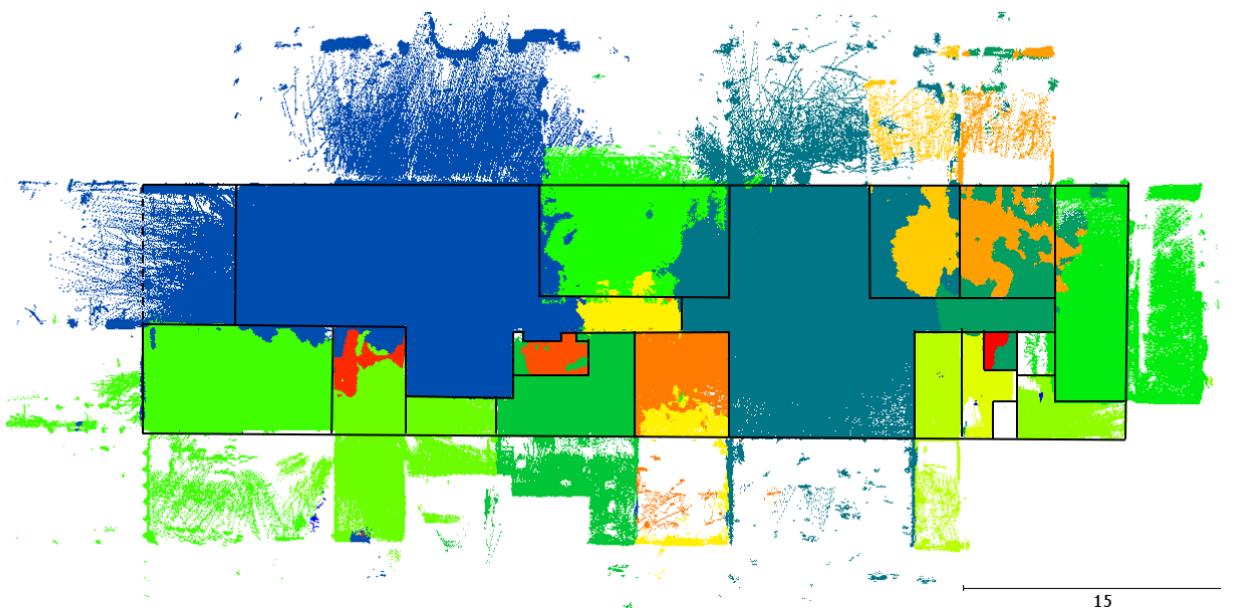
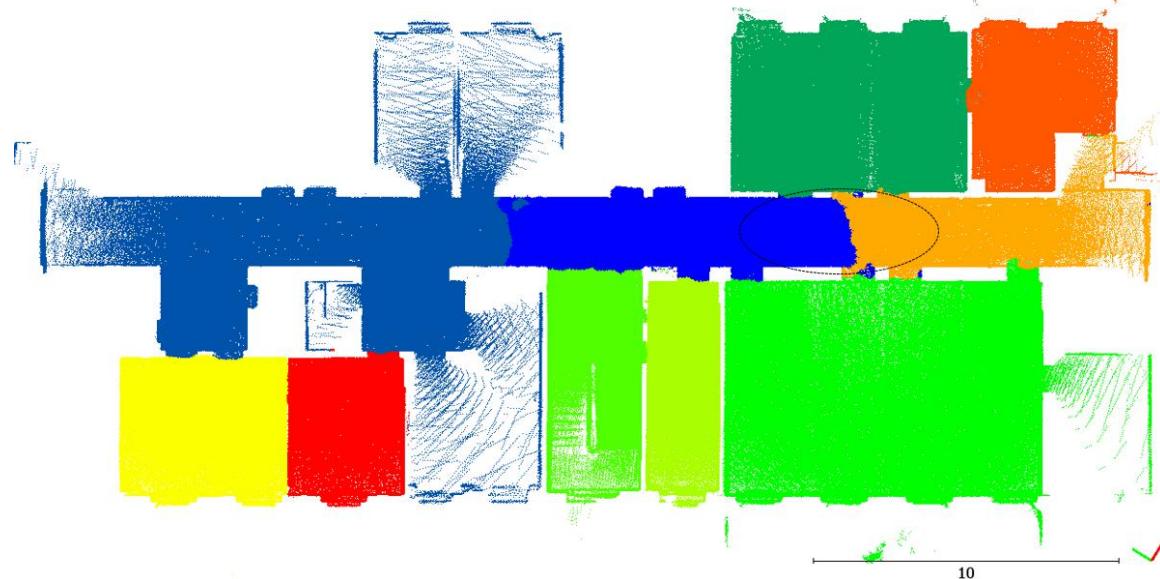
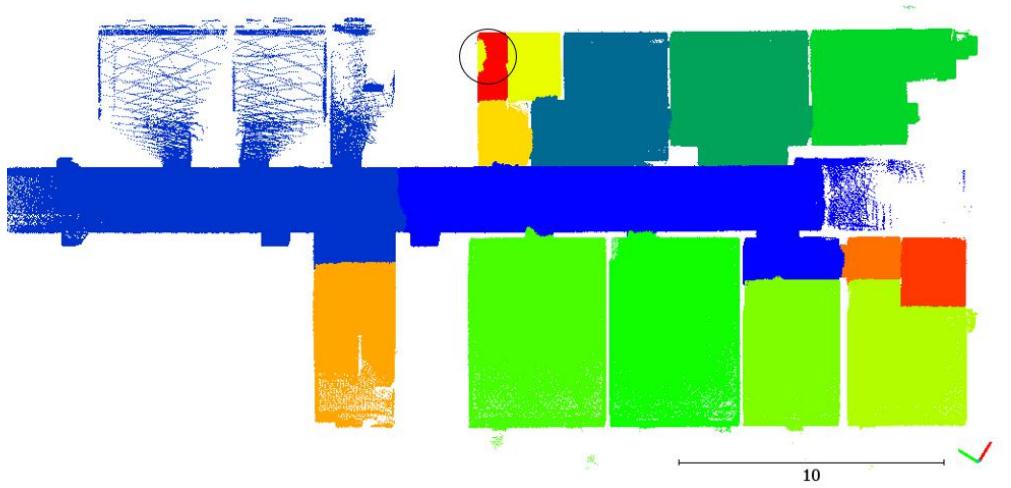


Navigable space and walls



Diemen Building in Braunschweig

Room Segmentation



source: A. Elseicy's Msc Thesis
Shayan Nikoohemat

Exploiting MLS Trajectory for Interpretation of Indoor laser Scanner Data

More details:

In The ISPRS GeoSpatial Week 2017, Wuhan, China

Exploiting Indoor Mobile Laser Scanner Trajectories for Semantic Interpretation of Point Clouds

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Commission IV, WG IV/5

KEY WORDS: Indoor Point Clouds, MLS Trajectory, Indoor Reconstruction, Opening Detection, Occlusion Reasoning, 3D Model

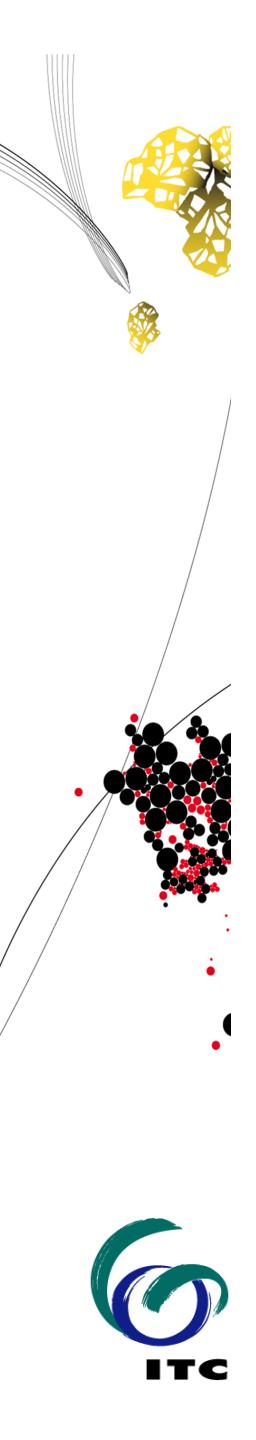
ABSTRACT:

The use of Indoor Mobile Laser Scanners (IMLS) for data collection in indoor environments has been increasing in the recent years. These systems, unlike Terrestrial Laser Scanners (TLS), collect data along a trajectory instead of at discrete scanner positions. In this research, we propose several methods to exploit the trajectories of IMLS systems for the interpretation of point clouds. By means of occlusion reasoning and use of trajectory as a set of scanner positions, we are capable of detecting openings in cluttered indoor environments. In order to provide information about both the partitioning of the space and the navigable space, we use the voxel concept for point clouds. Furthermore, to reconstruct walls, floor and ceiling we exploit the indoor topology and plane primitives. The results show that the trajectory is a valuable source of data for feature detection and understanding of indoor MLS point clouds.

1. INTRODUCTION

Indoor 3D models are required for navigation, building maintenance, disaster management and many other applications. However, manual creation of indoor 3D models is an expensive and cumbersome process for large buildings such as airports, shopping malls and office buildings. Therefore, automated methods for reconstruction of indoor environments have been developed.

reconstruction methods are often limited to Manhattan-World structure (Budroni and Boehm, 2010) or employ horizontal floor/ceiling and vertical wall assumption (Ochmann et al., 2016; Oesau et al., 2014; Xiao and Furukawa, 2014). Other methods generate 2.5D models (Oesau et al., 2014; Turner and Zakhor, 2014) or do not consider the detection of openings and addition of semantics (Mura et al., 2014a; Oesau et al., 2014; Xiao and



Thank you for your attention

Questions?



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Exploiting MLS Trajectory for Interpretation of Indoor laser Scanner Data

Parameters:

| Algorithm | Parameters | Value |
|---|--------------------------------------|-----------------|
| Surface Growing Segmentation | distance to surface | 0.10 m |
| | seed search radius | 1.0 m |
| Reflection Removal | time difference | 150 s |
| | # of reflected points in a segment | 70% |
| Surface Patch Generation | planes distance | 0.60 m |
| | segments distance | 0.40 m |
| | planes angle | 10 degree |
| Wall/Floor/Ceiling Detection | intersection threshold | 0.10 m |
| | surface angle threshold | 20 degree |
| | floor height estimation (optional) | - |
| | ceiling height estimation (optional) | - |
| | dist to floor, ceiling (optional) | 0.50 m |
| Prune Wall Detection | dist to floor, ceiling | 0.50 m |
| Occlusion Test (Opening detection) | voxel size | 0.10 m |
| | closenees dist to surface | 0.60 m |
| Space Partitioning | voxel size | 0.10 m |
| | search windows size | 5*voxel_size |
| Door Detection | voxel size | 0.10 |
| | door size (width, height) | 9*21*voxel_size |
| | search windows size | 5*voxel_size |
| | percentage of void_hood points | 70% |
| | trajectory search radius | 0.15 m |

| Class | Precision | Recall | F1-Score |
|---------|-----------|--------|----------|
| Wall | 0.88 | 0.95 | 0.91 |
| Floor | 0.93 | 0.98 | 0.95 |
| Ceiling | 0.93 | 0.98 | 0.95 |

Accuracy of results for wall,
Floor and ceiling

| Class | Precision |
|----------|-----------|
| Openings | 0.73 |
| Occluded | 0.57 |
| Occupied | 0.89 |

Accuracy of results for openings

Analyzing the methods:

- **Wall Detection:** relies on the segment generalization and connectivity of segments.
 - **Opening Detection:** relies on the wall detection results, challenge in occluded openings and reflection from glass.
 - **Door Detection:** relies on the trajectory and input door size parameter.
 - **Space Partitioning:** windows and gaps in the data are problematic for space partitioning.
-
- | | |
|--|---|
| <ul style="list-style-type: none">• Advantages of our method:<ul style="list-style-type: none">+ Applicable on non-Manhattan World+ Applicable on non-vertical walls+ Scalable to large datasets+ Improvable with iterations | <ul style="list-style-type: none">• Disadvantages of our method:<ul style="list-style-type: none">- Big gaps in the data challenge adjacency graph- Heavy clutter near the ceiling is problematic- Each methods relies on previous results |
|--|---|