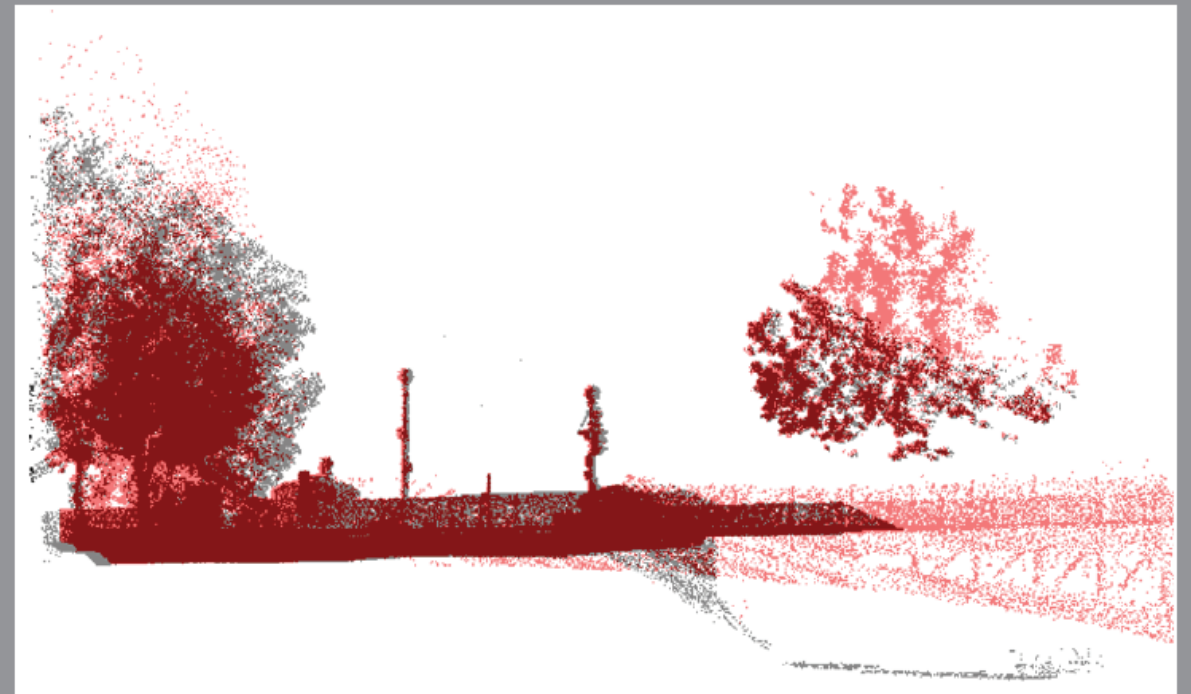
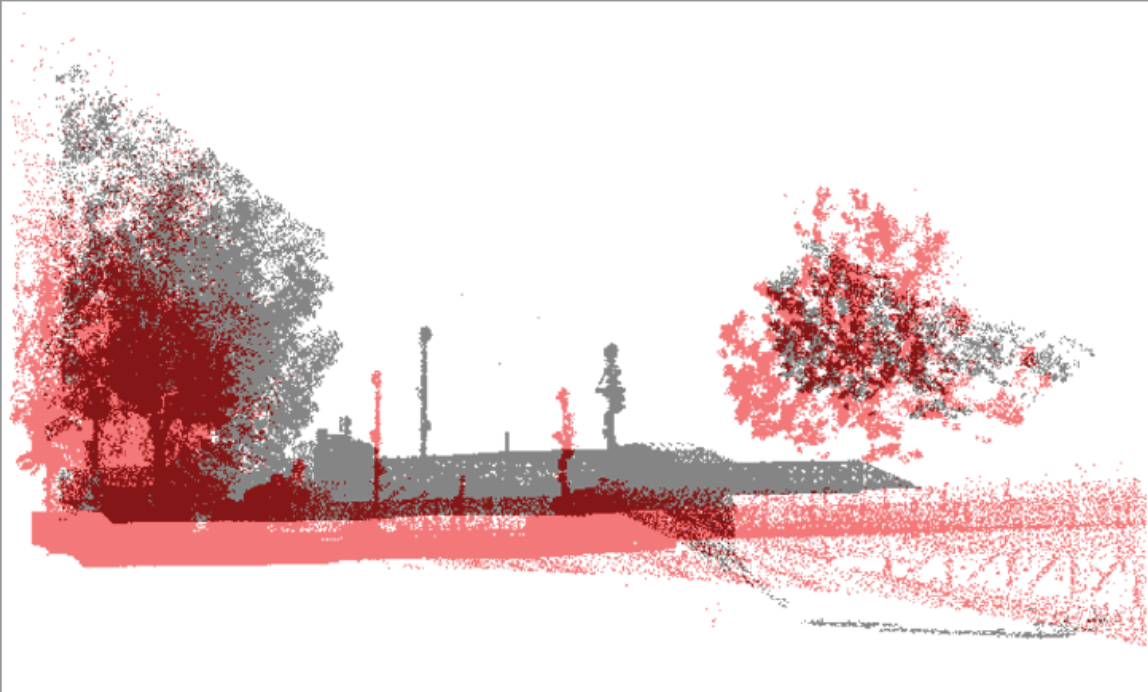


# IMAGE-BASED METHOD FOR THE PAIRWISE REGISTRATION OF MOBILE LASER SCANNING POINT CLOUDS

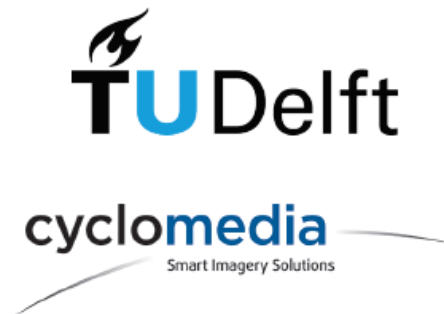


**ANTRIA  
CHRISTODOULOU**

**MSc Geomatics**

October 2018

Peter van Oosterom, Ravi Peters  
Peter Joosten, Berry van Someren



Paper: <https://bit.ly/2NIQuUx>

IMAGE-BASED METHOD  
FOR THE PAIRWISE REGISTRATION OF  
MOBILE LASER SCANNING POINT CLOUDS

# POINT CLOUDS: DATA-SETS OF 3D POINTS

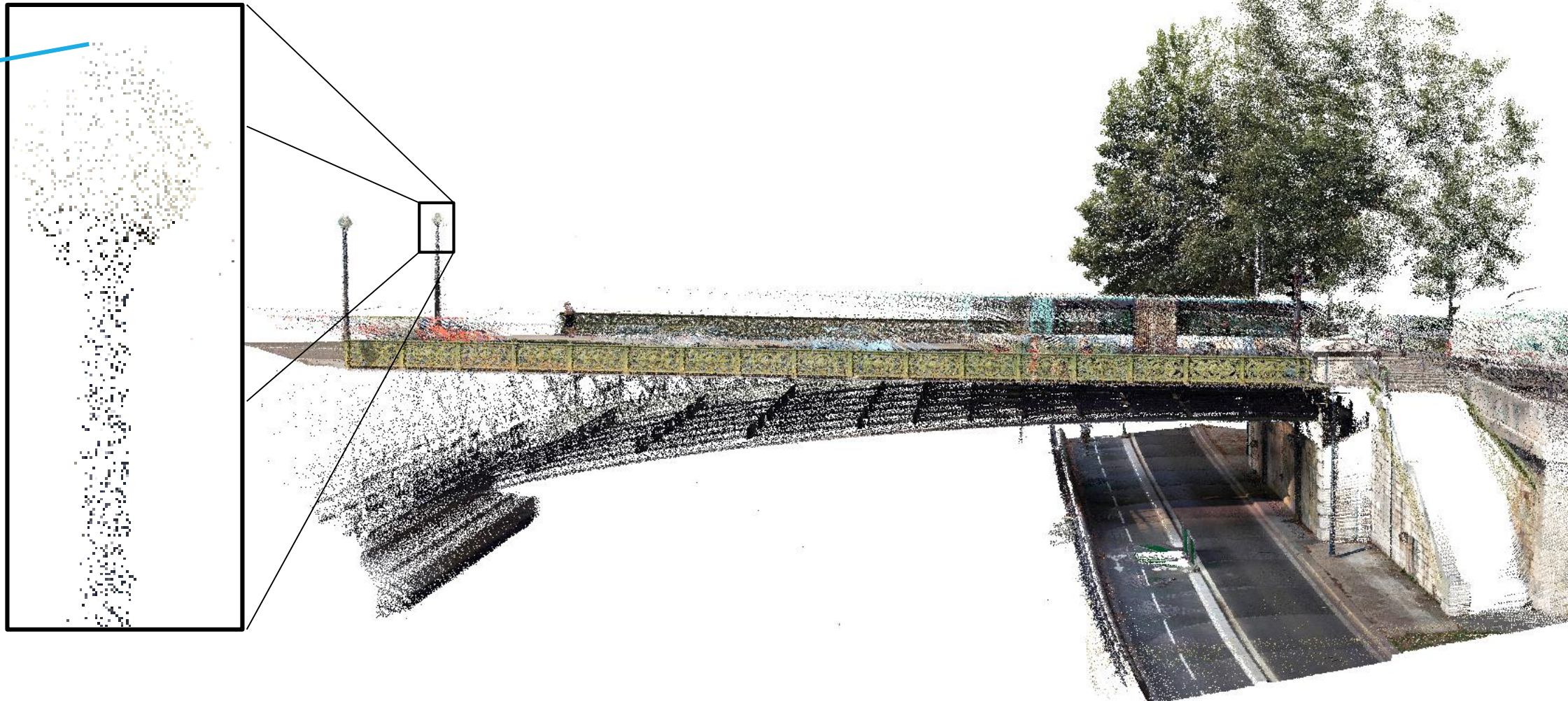
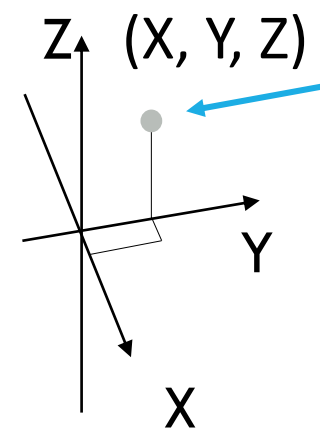
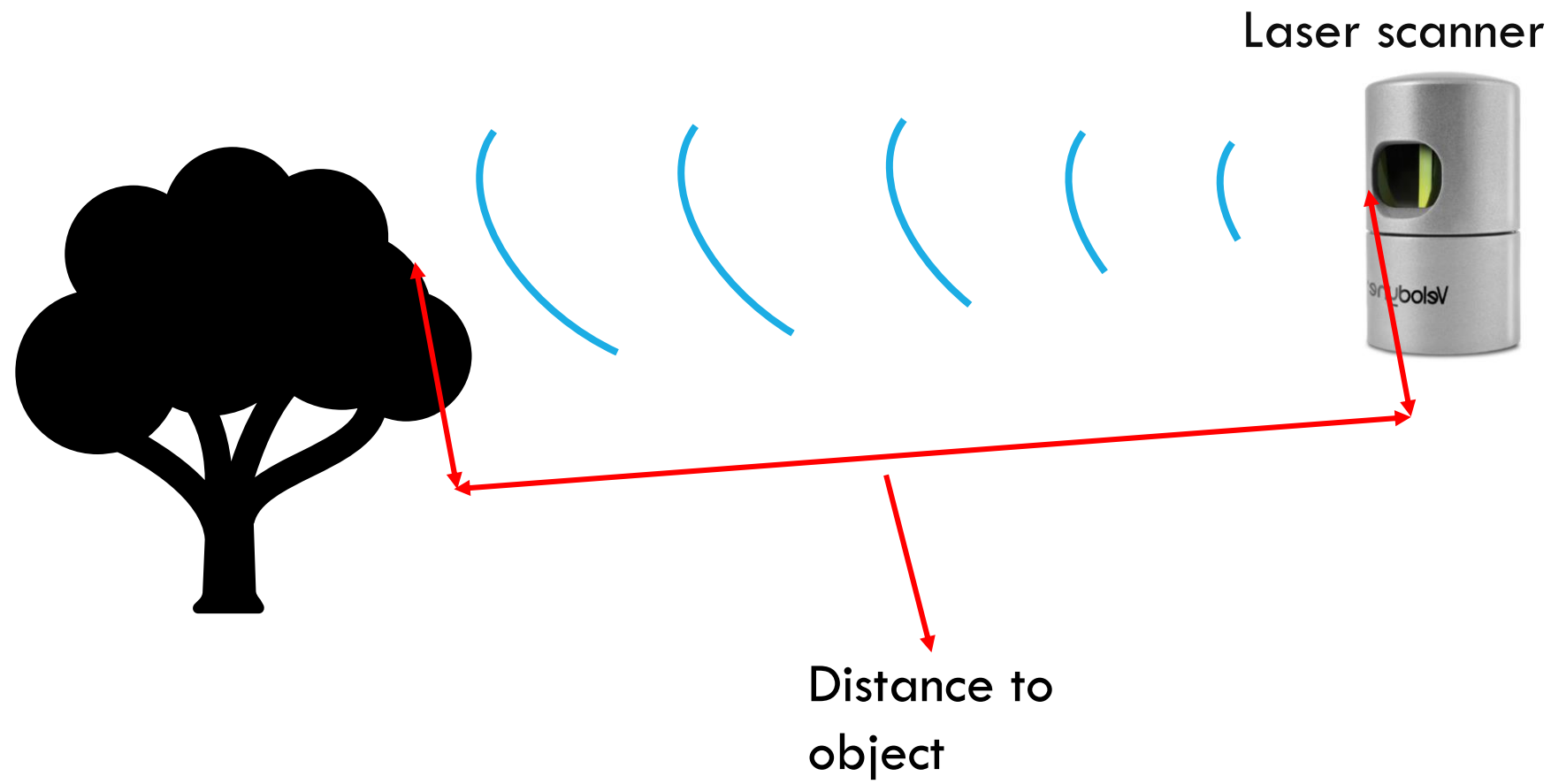


IMAGE-BASED METHOD  
FOR THE PAIRWISE REGISTRATION OF  
MOBILE LASER SCANNING POINT CLOUDS

# LASER SCANNING





# MOBILE LASER SCANNING

IMAGE-BASED METHOD  
FOR THE PAIRWISE REGISTRATION OF  
MOBILE LASER SCANNING POINT CLOUDS

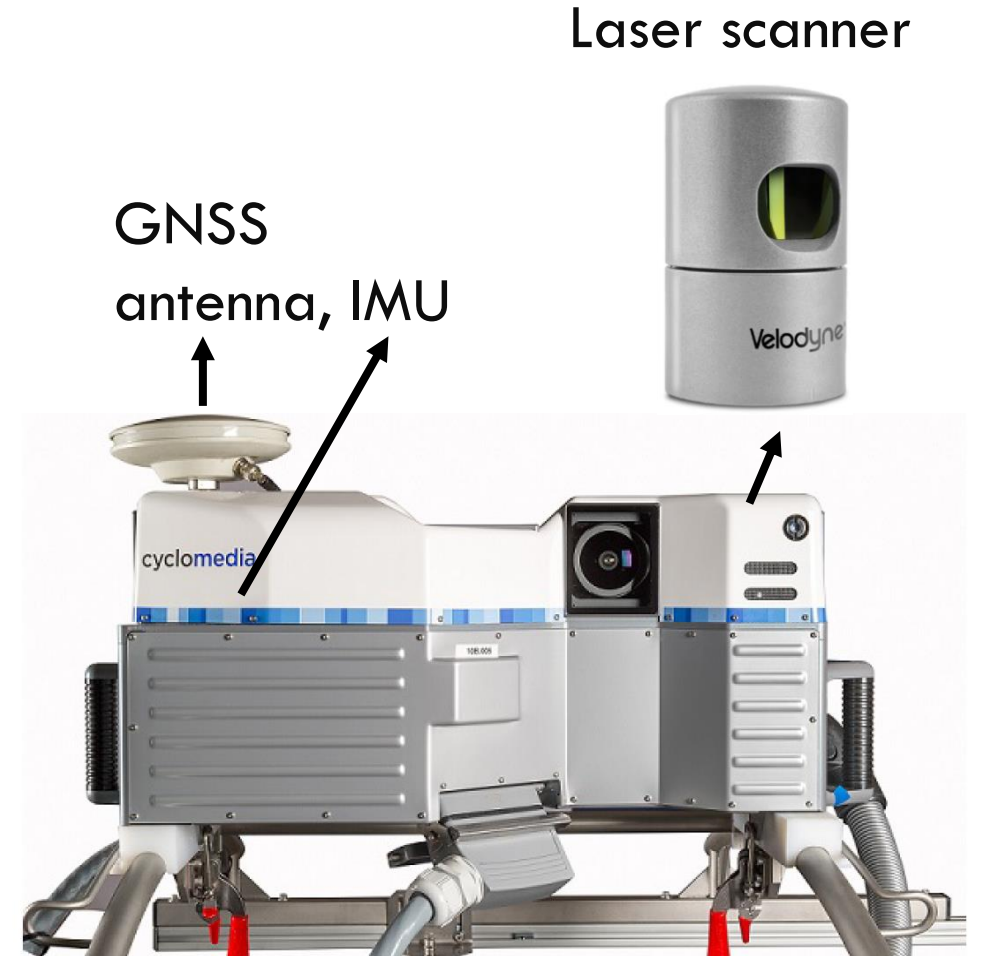
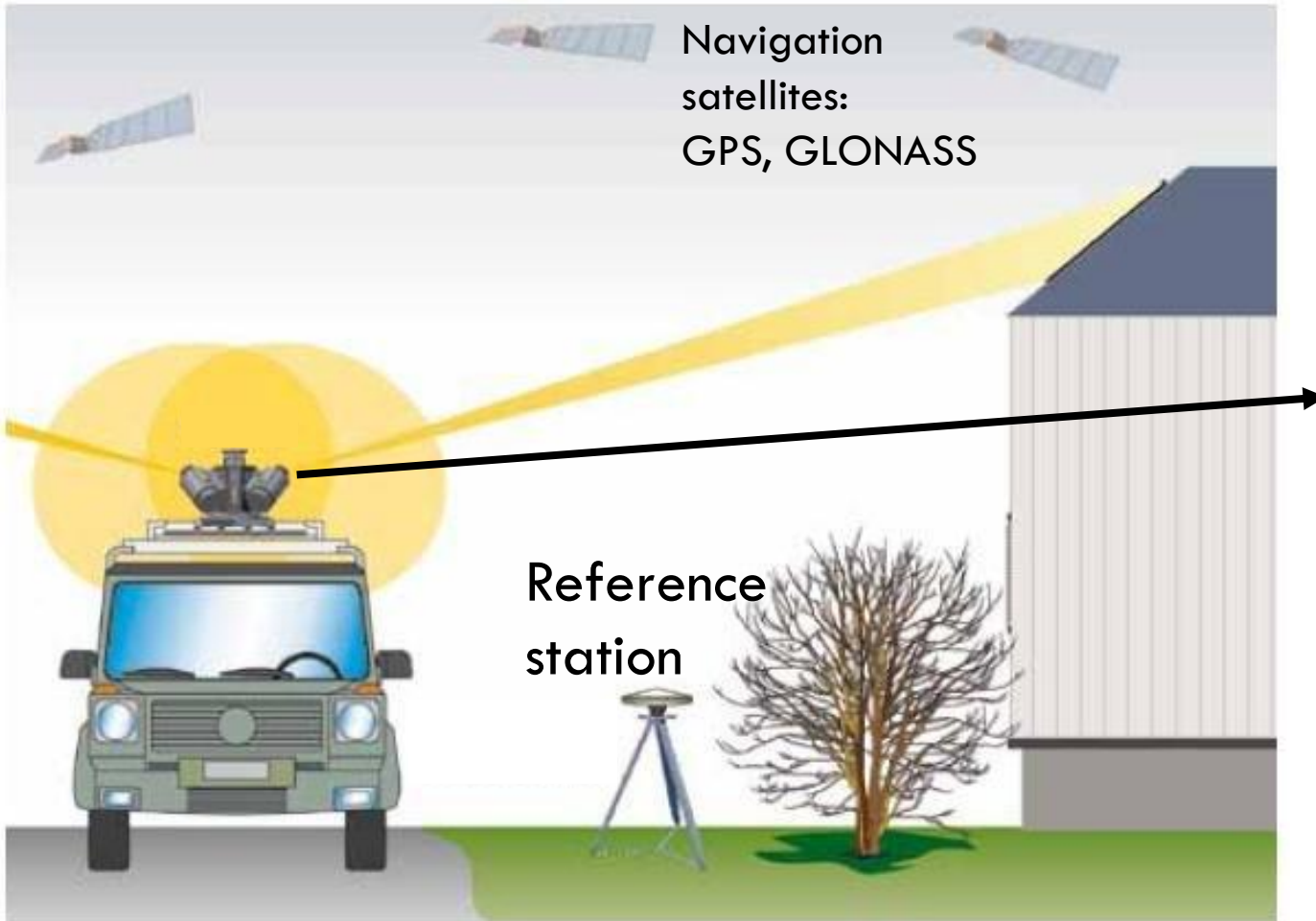
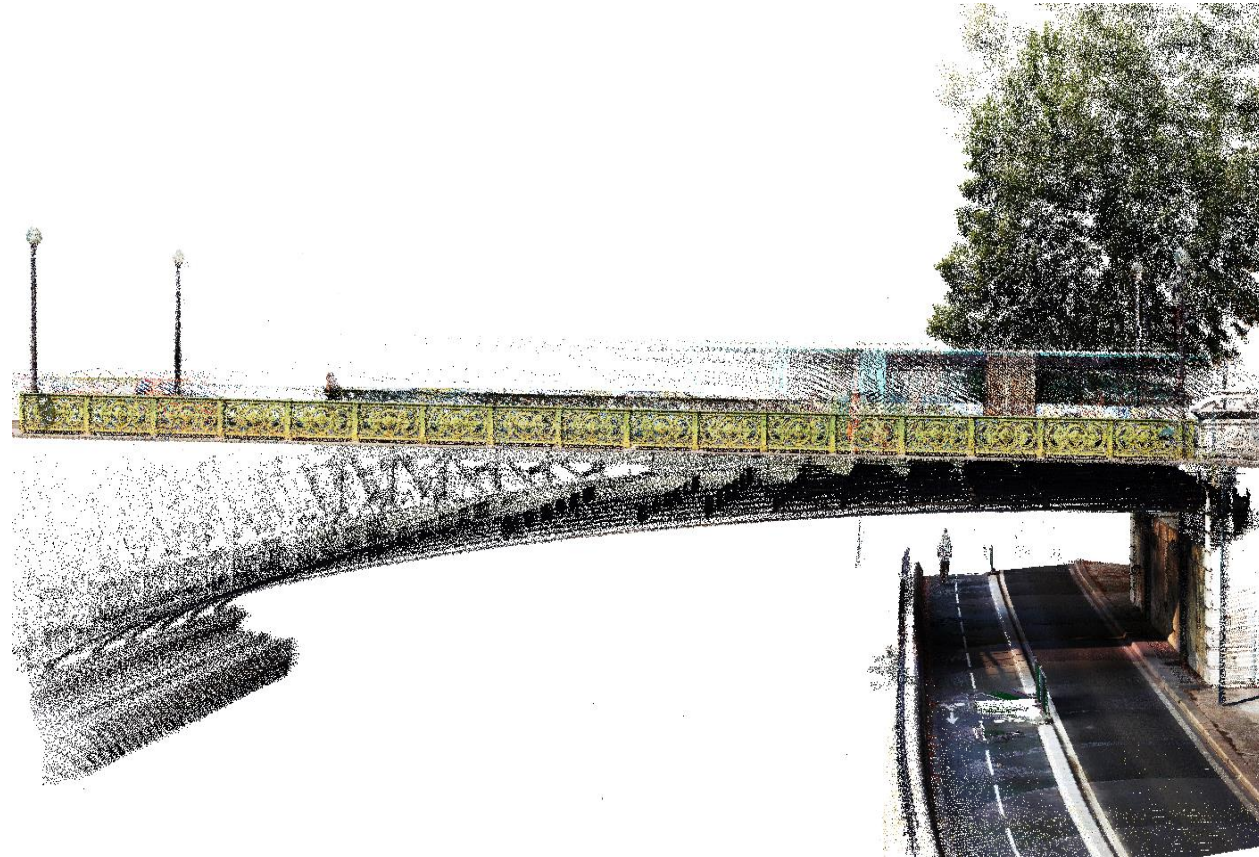


IMAGE-BASED METHOD  
FOR THE PAIRWISE REGISTRATION OF  
MOBILE LASER SCANNING POINT CLOUDS

# MOBILE LASER SCANNING POINT CLOUDS



Street-view image

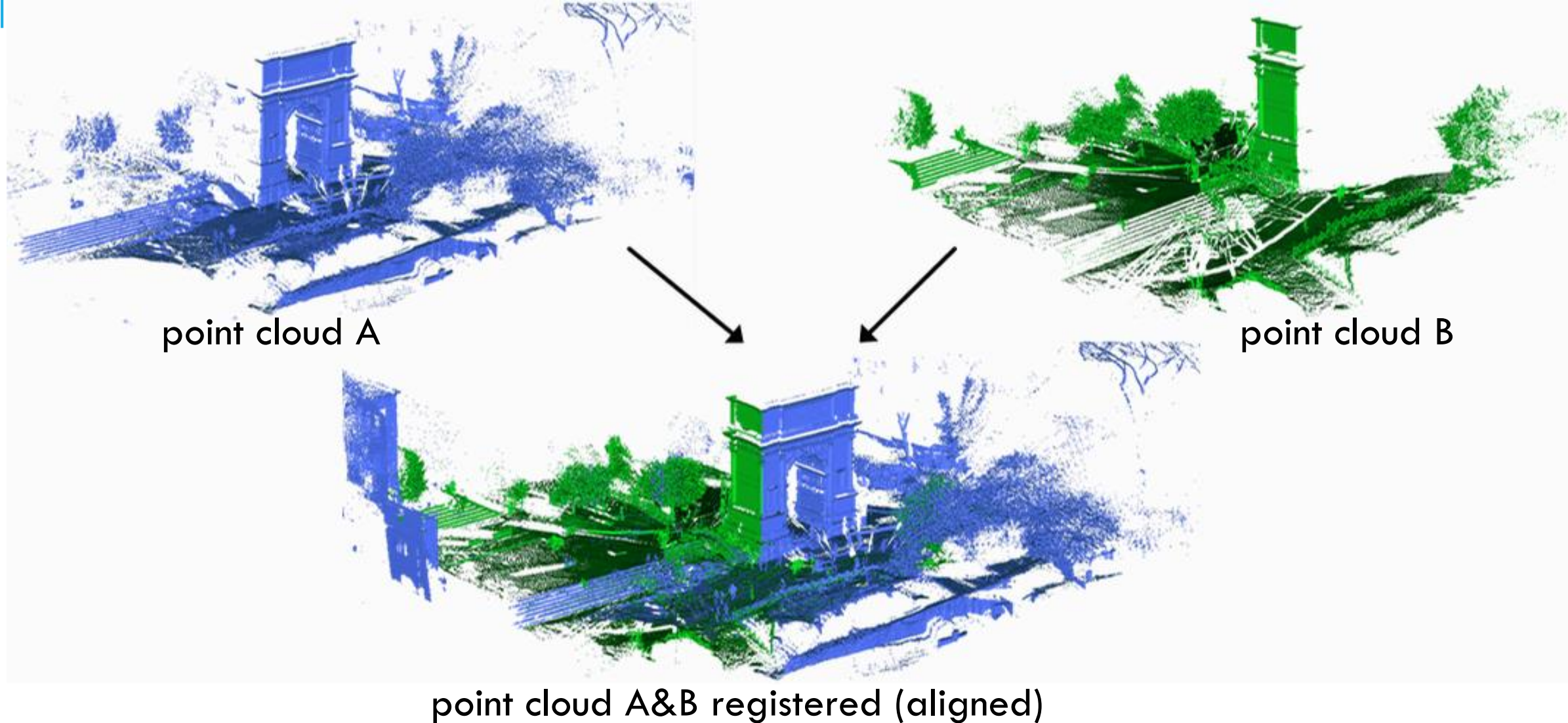


Street-view point cloud

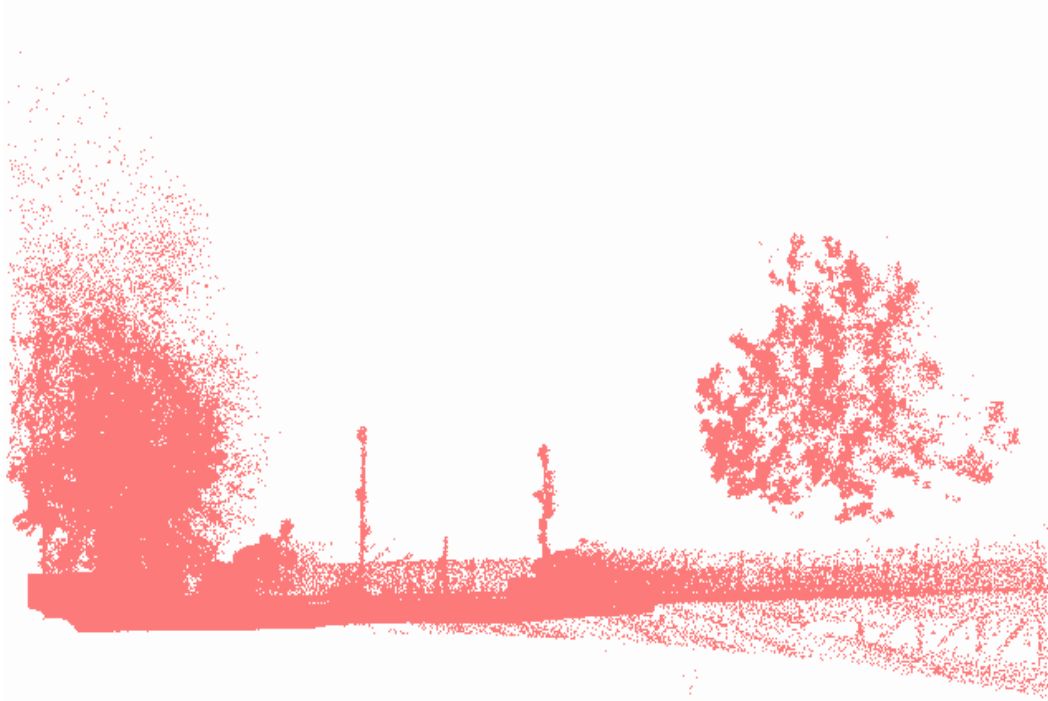


IMAGE-BASED METHOD  
FOR THE PAIRWISE REGISTRATION OF  
MOBILE LASER SCANNING POINT CLOUDS

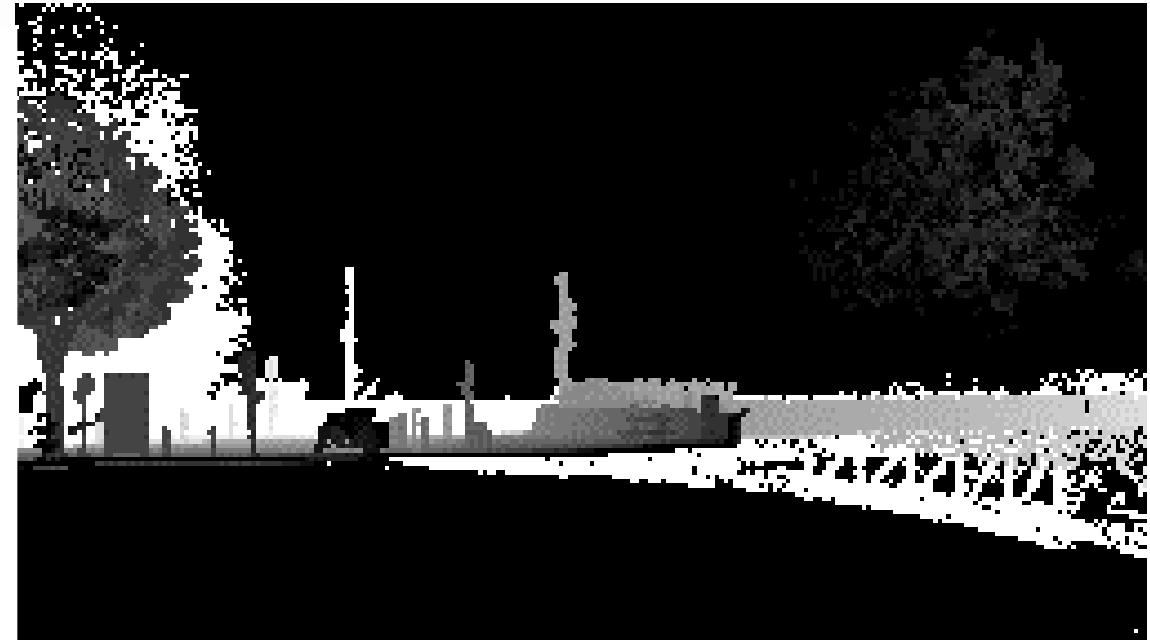
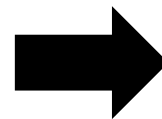
# PAIRWISE REGISTRATION



# IMAGE-BASED METHOD



3D



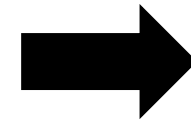
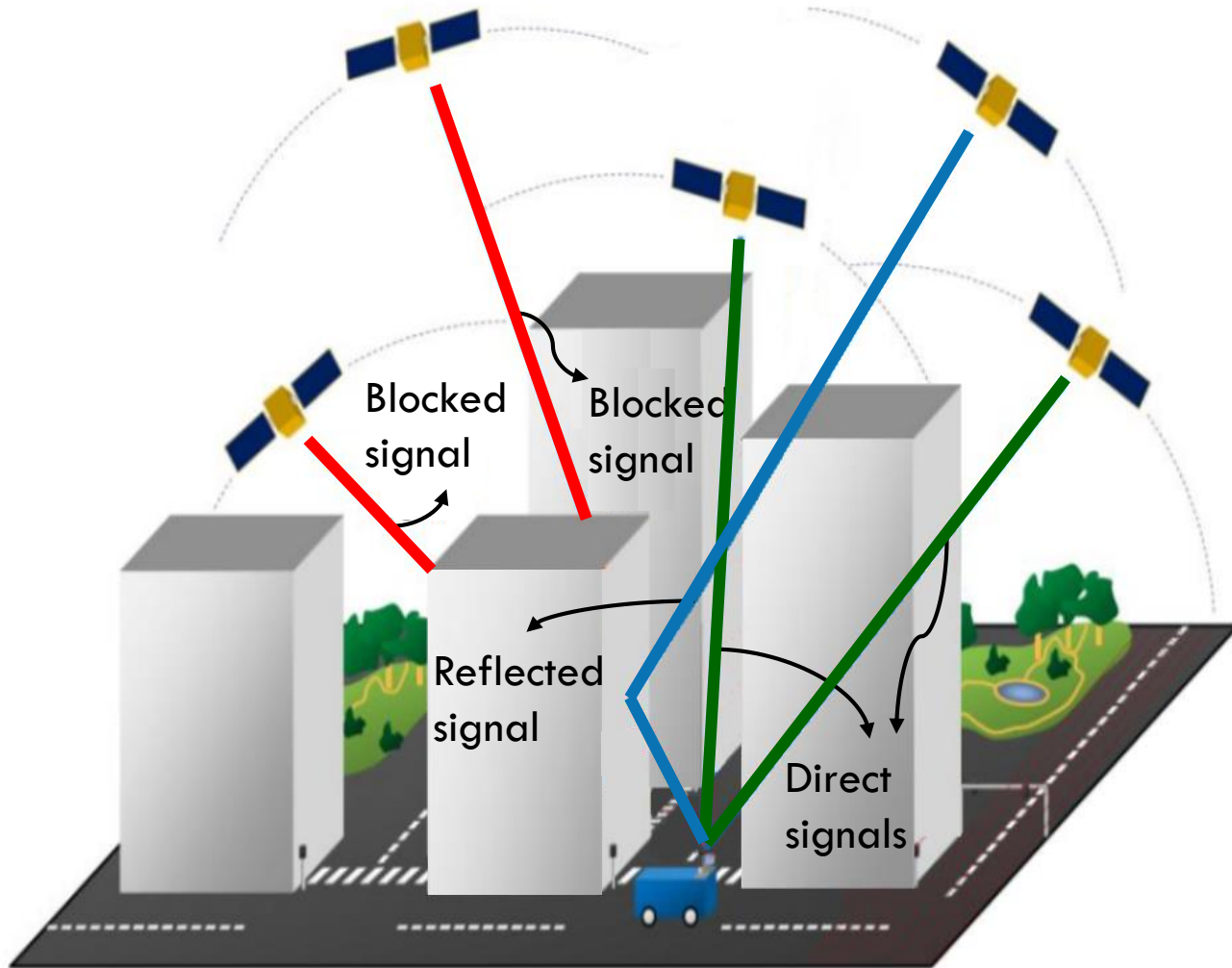
2D

# THE PROBLEM





# THE PROBLEM



Available satellites may not be sufficient for the determination of the positioning

# EFFECT OF THE PROBLEM



Noticeable at streets' crossings



# TO ACQUIRE A COMPLETE REPRESENTATION



Scanning at time t1

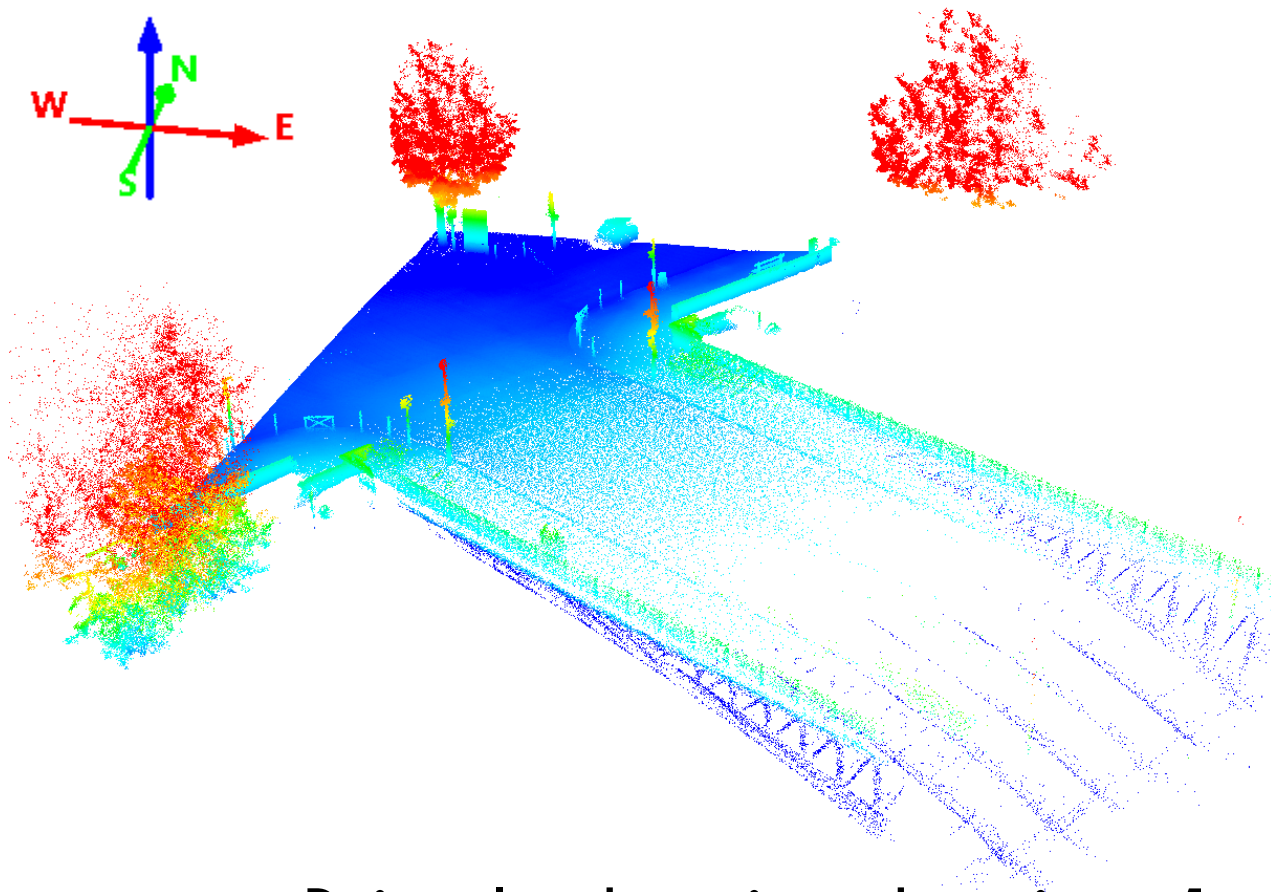


# TO ACQUIRE A COMPLETE REPRESENTATION

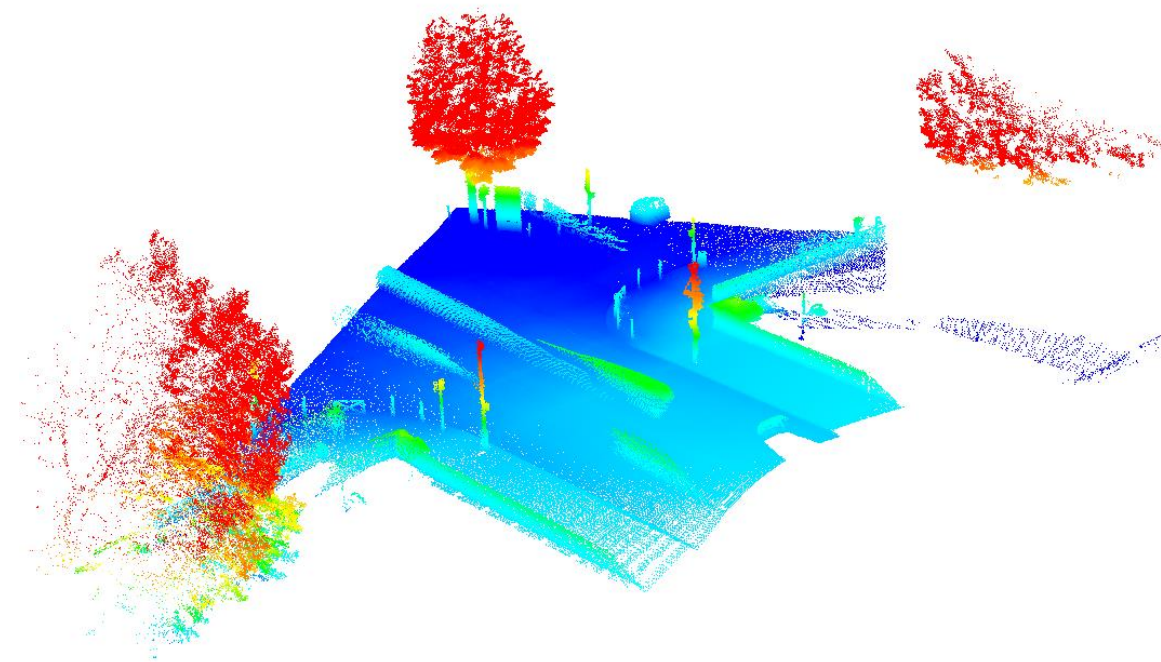


Scanning at time  $t_2$

# TWO OVERLAPPING POINT CLOUDS FROM THE SAME SCENE



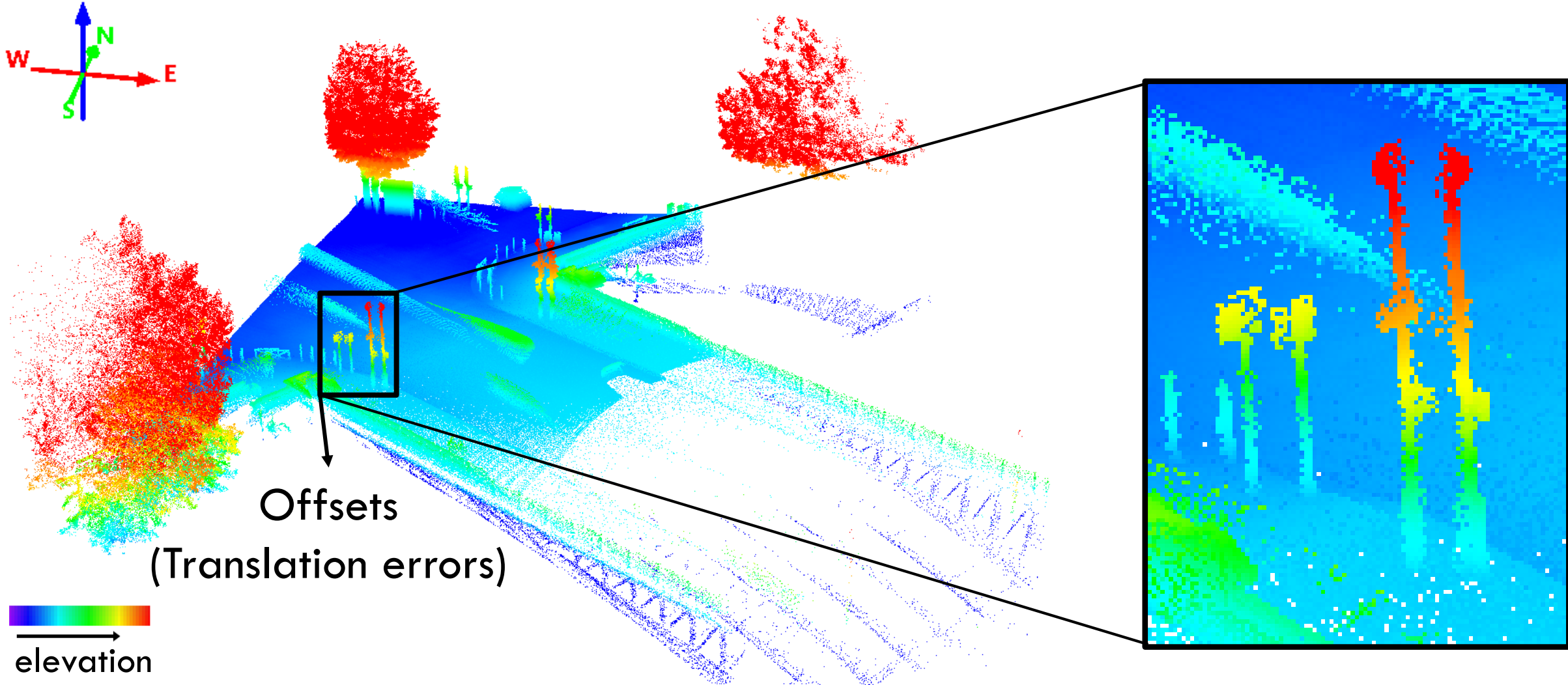
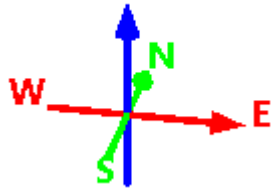
Point cloud retrieved at time t1



Point cloud retrieved at time t2



# SCANNING OF THE SAME SCENE AT TIMES T1 & T2

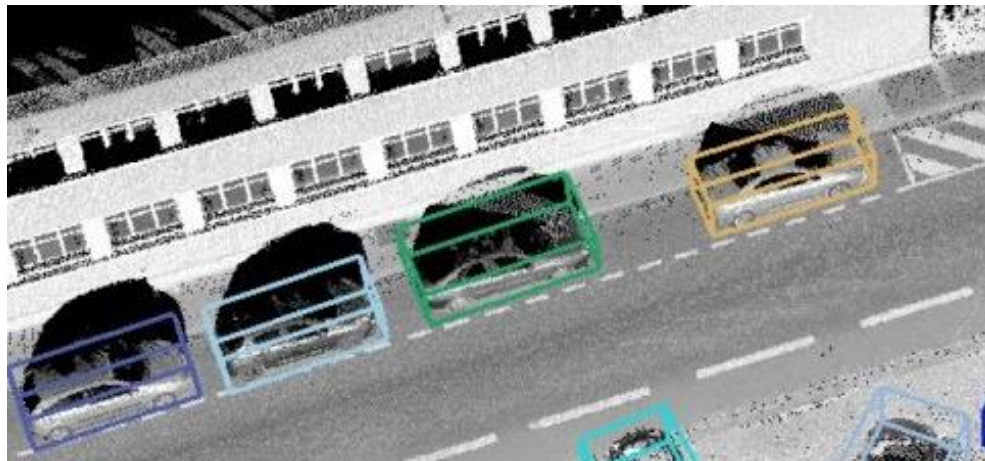




# WHY TO REGISTER POINT CLOUDS?



# MOTIVATION



Object recognition



UAV navigation-Disaster Management



Office measurements

# RELATED WORK: 3D LOCAL POINT CLOUD REGISTRATION

A 3D point cloud visualization of a landscape. The scene includes a road in the foreground, several trees of varying sizes, and utility poles in the background. The point cloud is rendered in a reddish-pink color. A semi-transparent white rectangular box is overlaid on the right side of the image, partially obscuring the background.



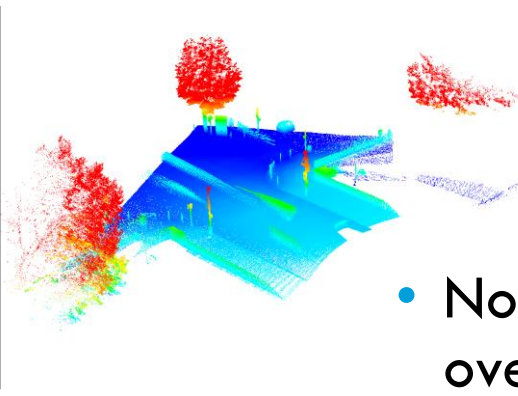
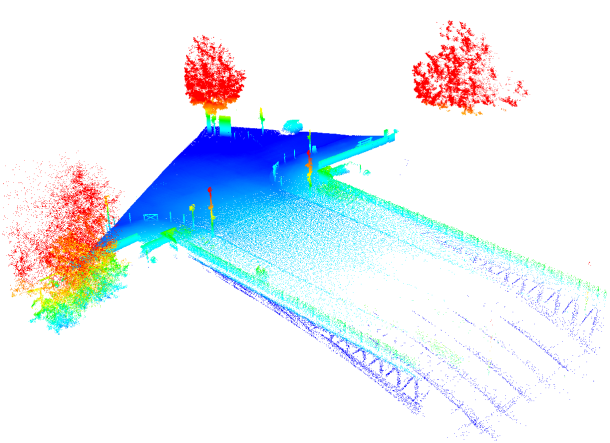
# COMMONLY USED: ITERATIVE CLOSEST POINT (ICP) ALGORITHM

- In general ICP-based algorithms perform good when:

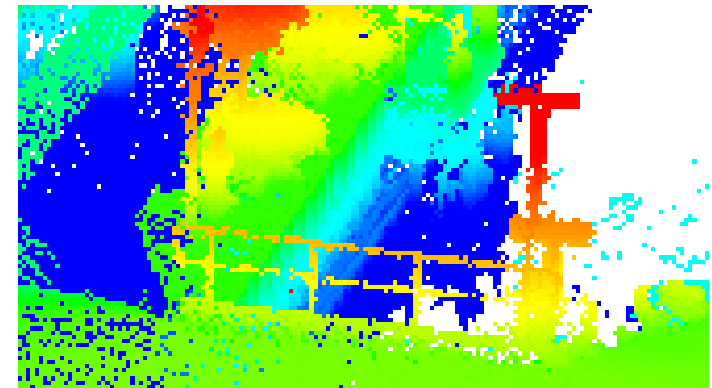


- Big overlap
- Initial positions have small offset

- However, in Mobile Laser Scanning point clouds:



- No big overlaps



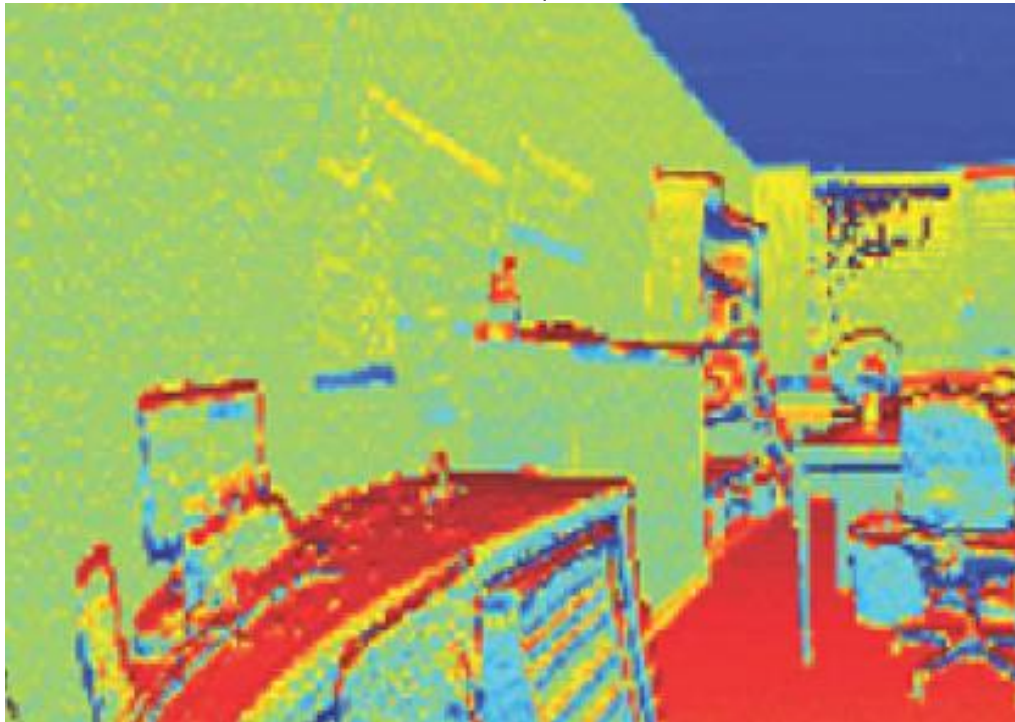
- Large offset (~1m)

# RELATED WORK: 2D LOCAL POINT CLOUD REGISTRATION

The image displays a 3D point cloud of a landscape. The scene includes several trees on the left and right sides, a road or path running horizontally across the middle, and a body of water in the foreground. The point cloud is rendered in a light blue and grey color scheme. A semi-transparent white rectangular box is positioned in the upper right quadrant of the image, containing the text 'RELATED WORK: 2D LOCAL POINT CLOUD REGISTRATION' in a bold, black, sans-serif font.

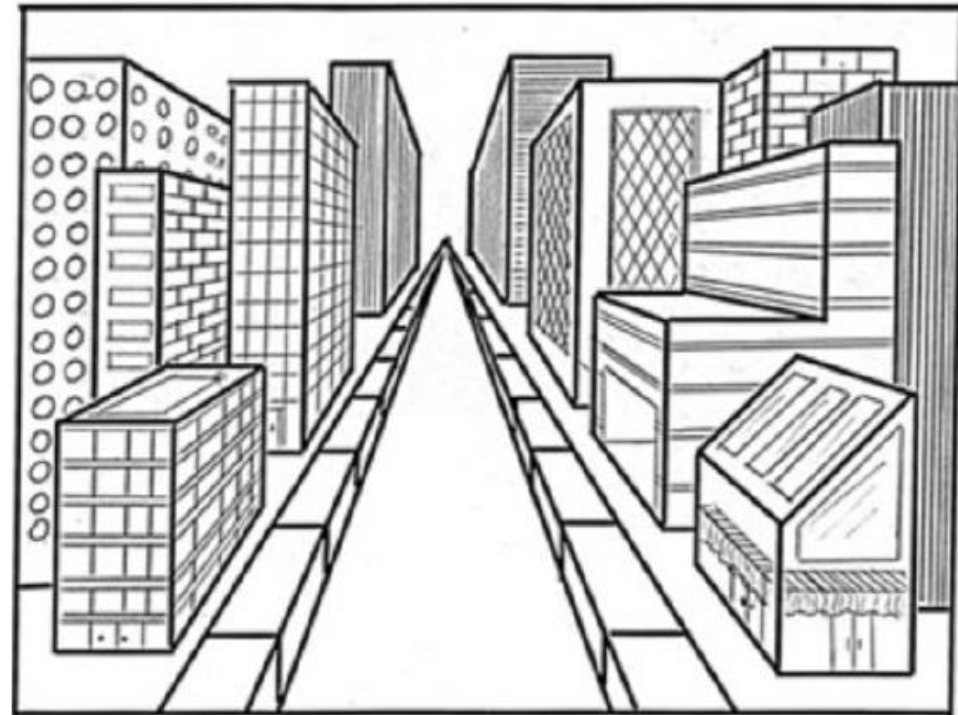
# 2D LOCAL POINT CLOUD REGISTRATION

Lin et al., 2017



- Bearing angle image
- Stress changes of directions

Liang et al., 2017



- Perspective image
- What a person can see

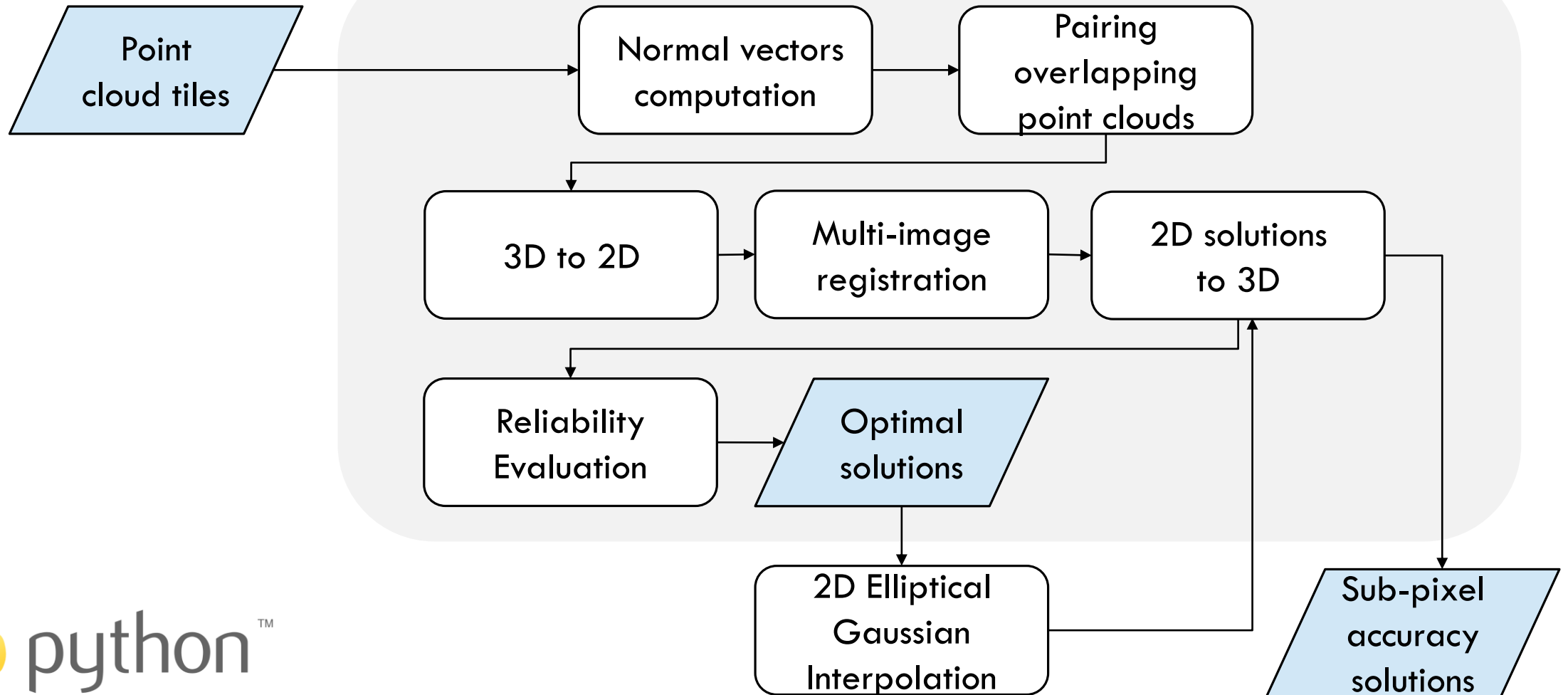


# DEVELOPED IMAGE-BASED METHOD

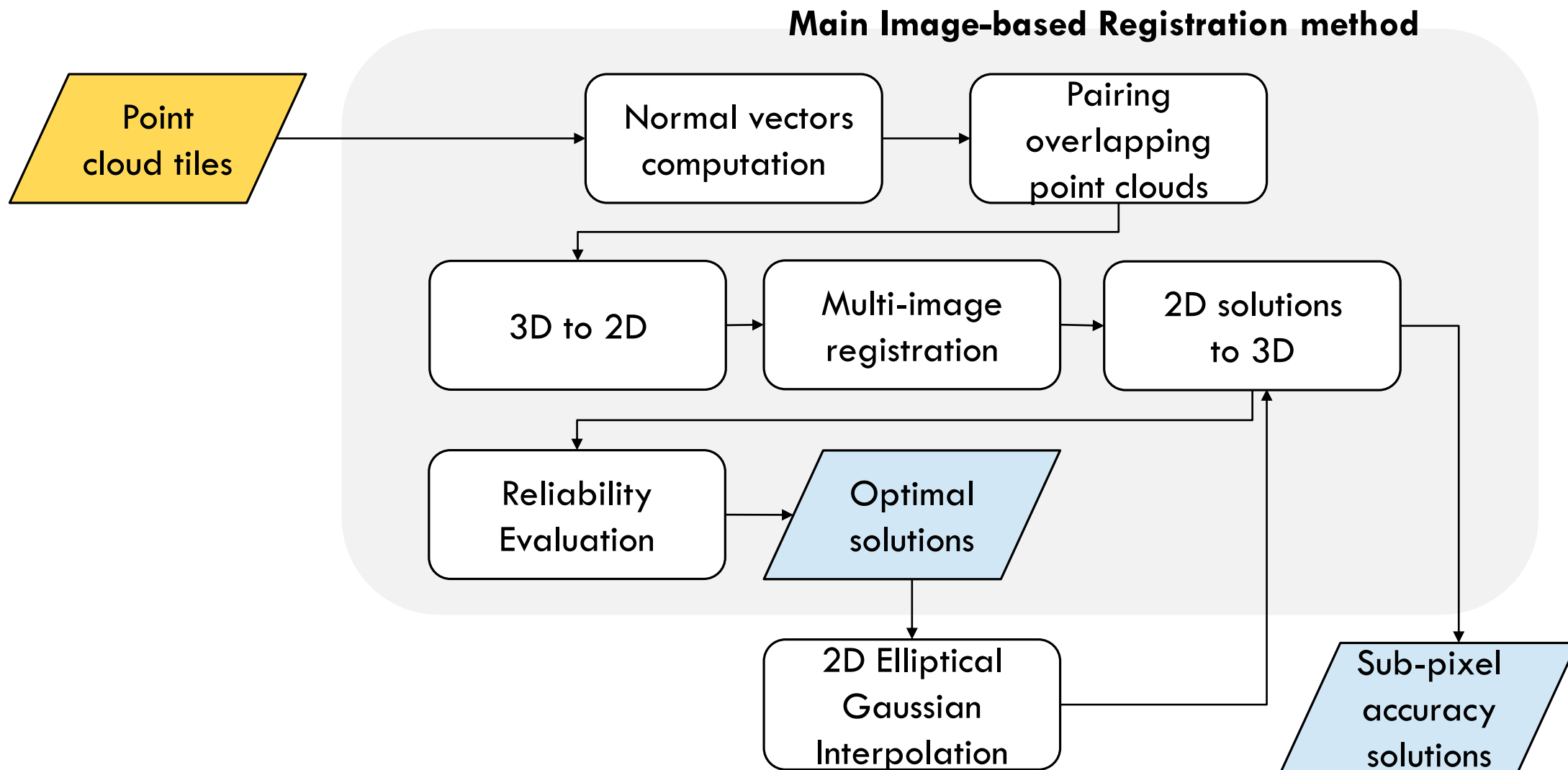


# DEVELOPED IMAGE-BASED METHOD

## Main Image-based Registration method

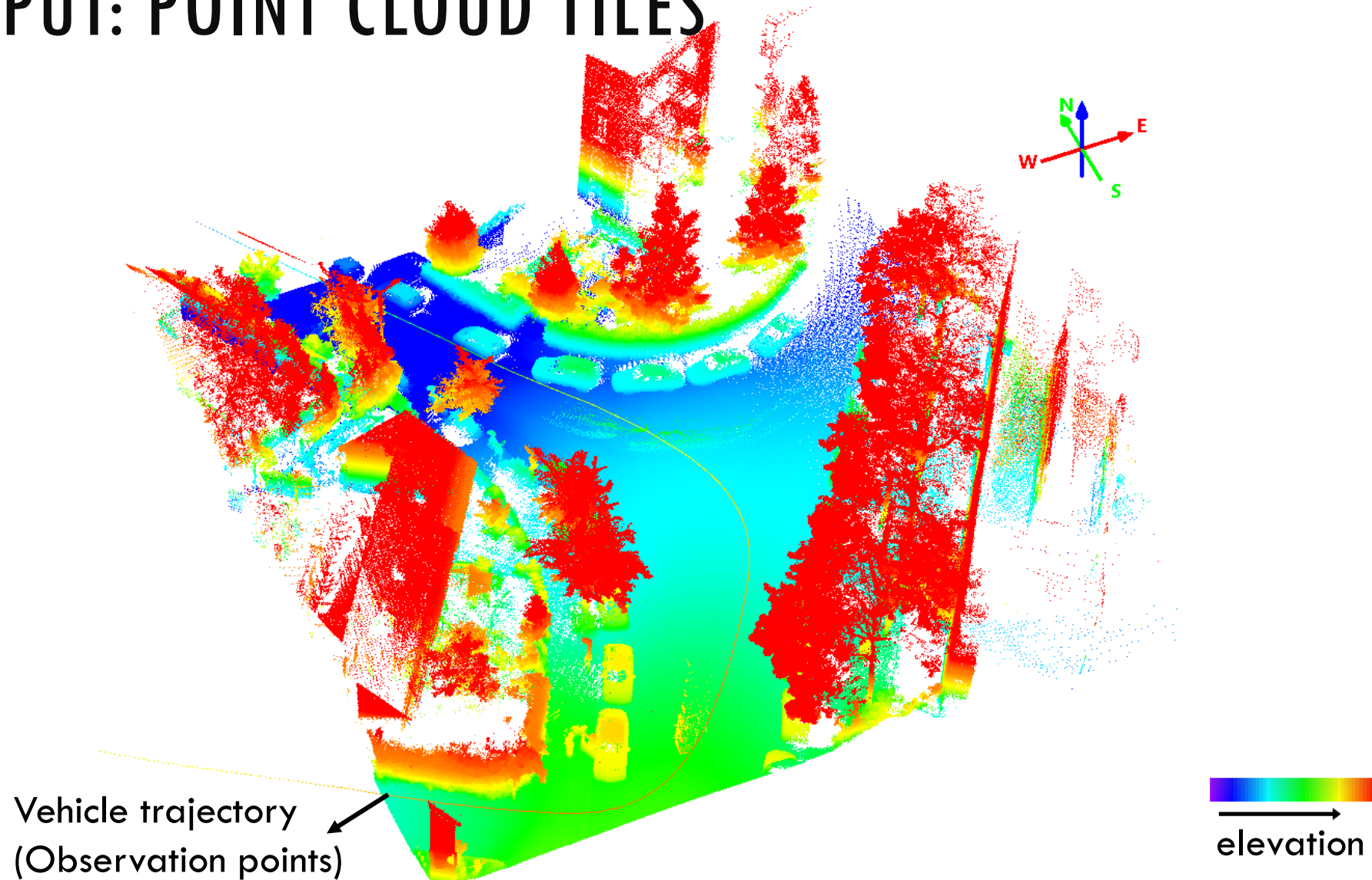


# INPUT: POINT CLOUD TILES

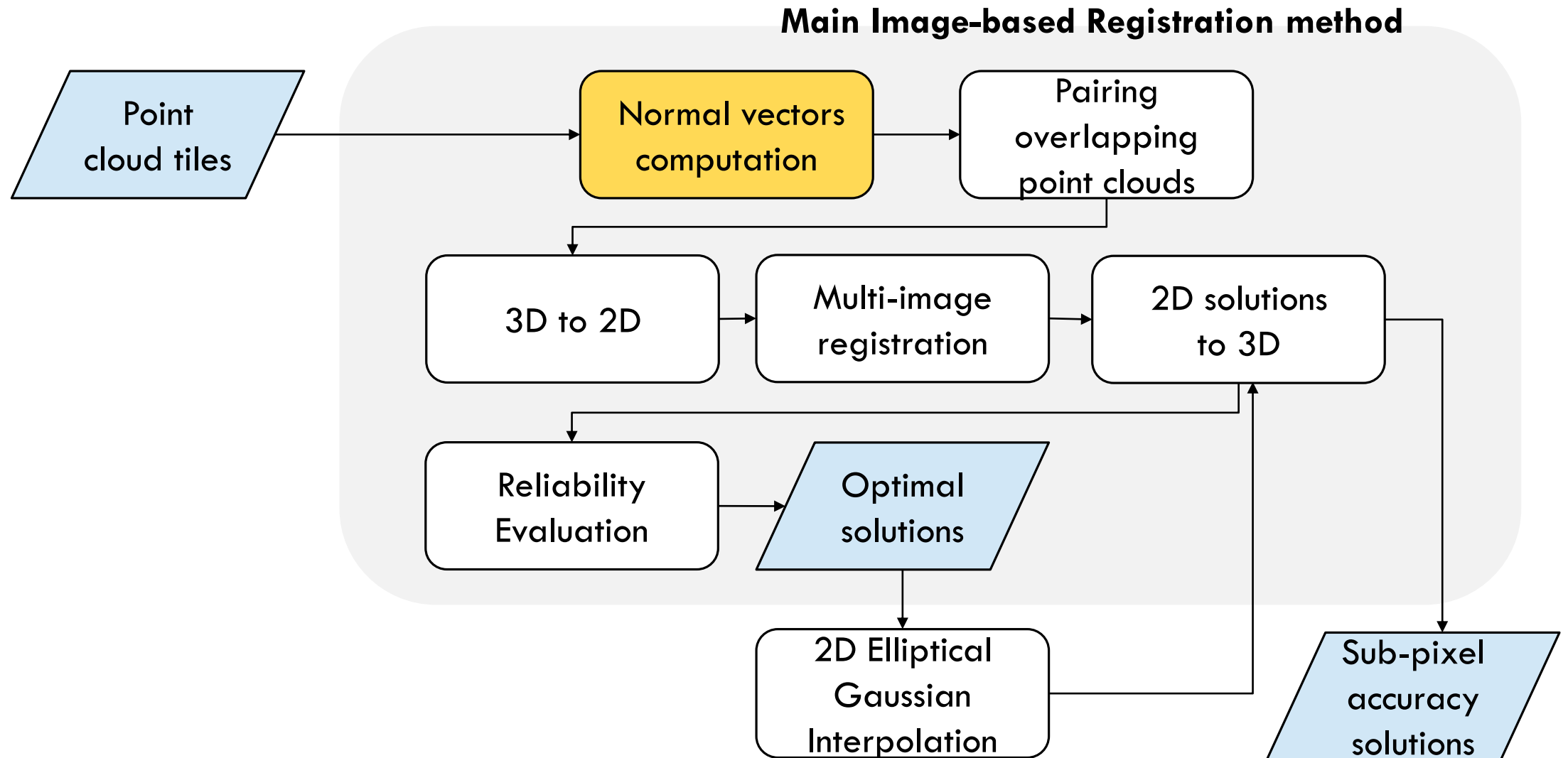




# INPUT: POINT CLOUD TILES

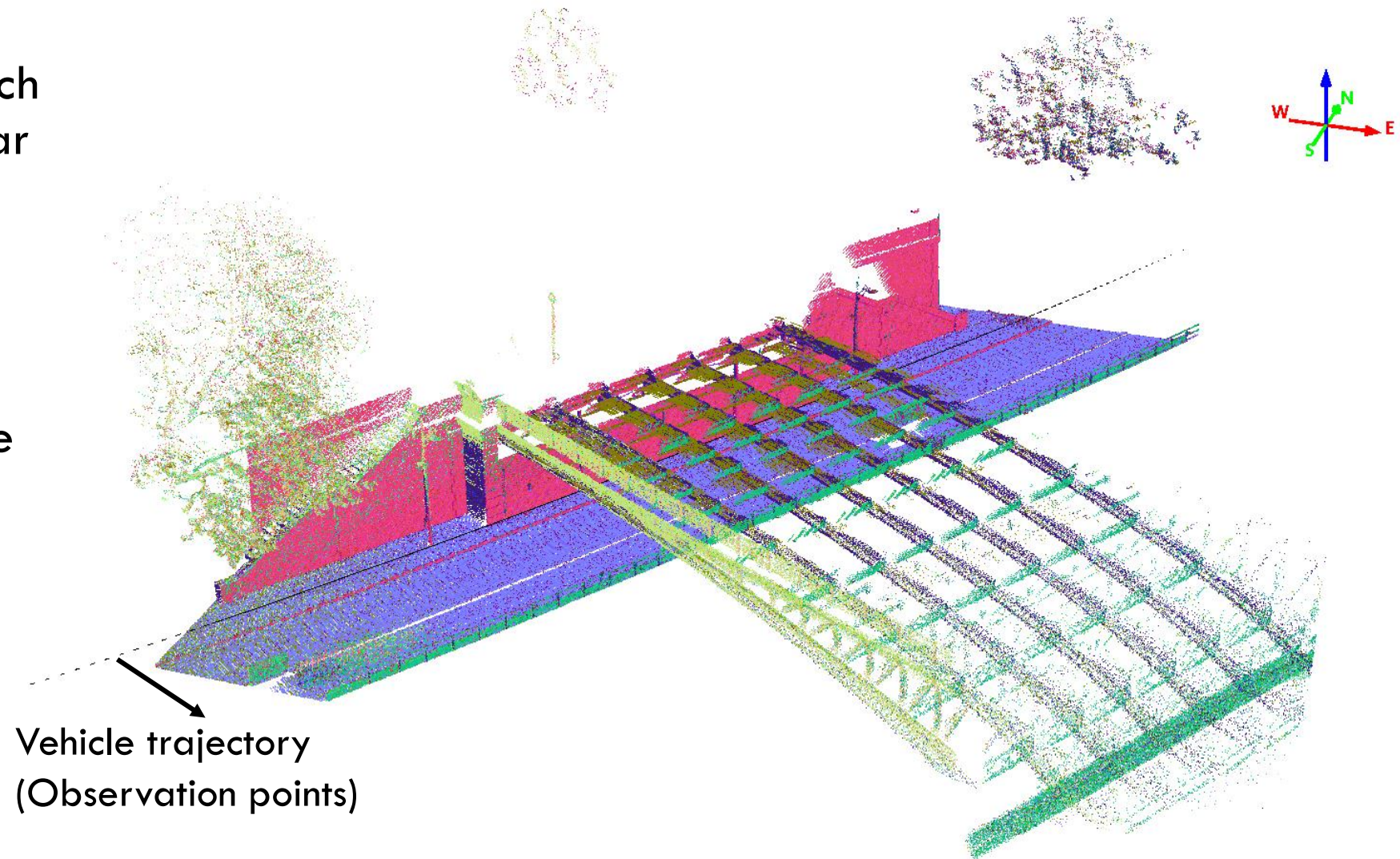


# 1. NORMAL VECTORS COMPUTATION



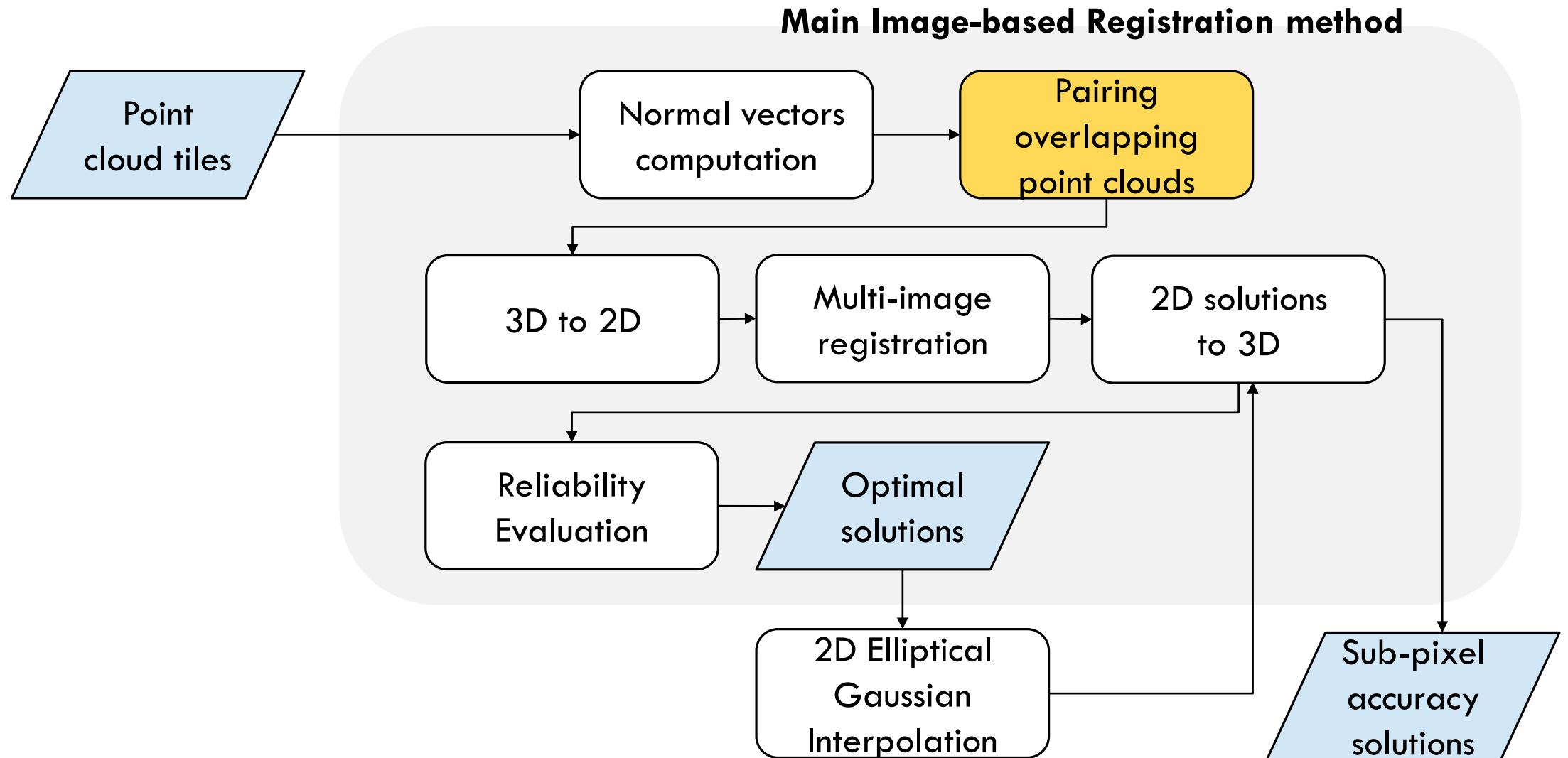
# NORMAL VECTORS COMPUTATION

- Normal vector of each point = perpendicular vector to a fitted plane around an observed point
- To make the normal vectors consistent: use of trajectory points
- Surfaces' orientation

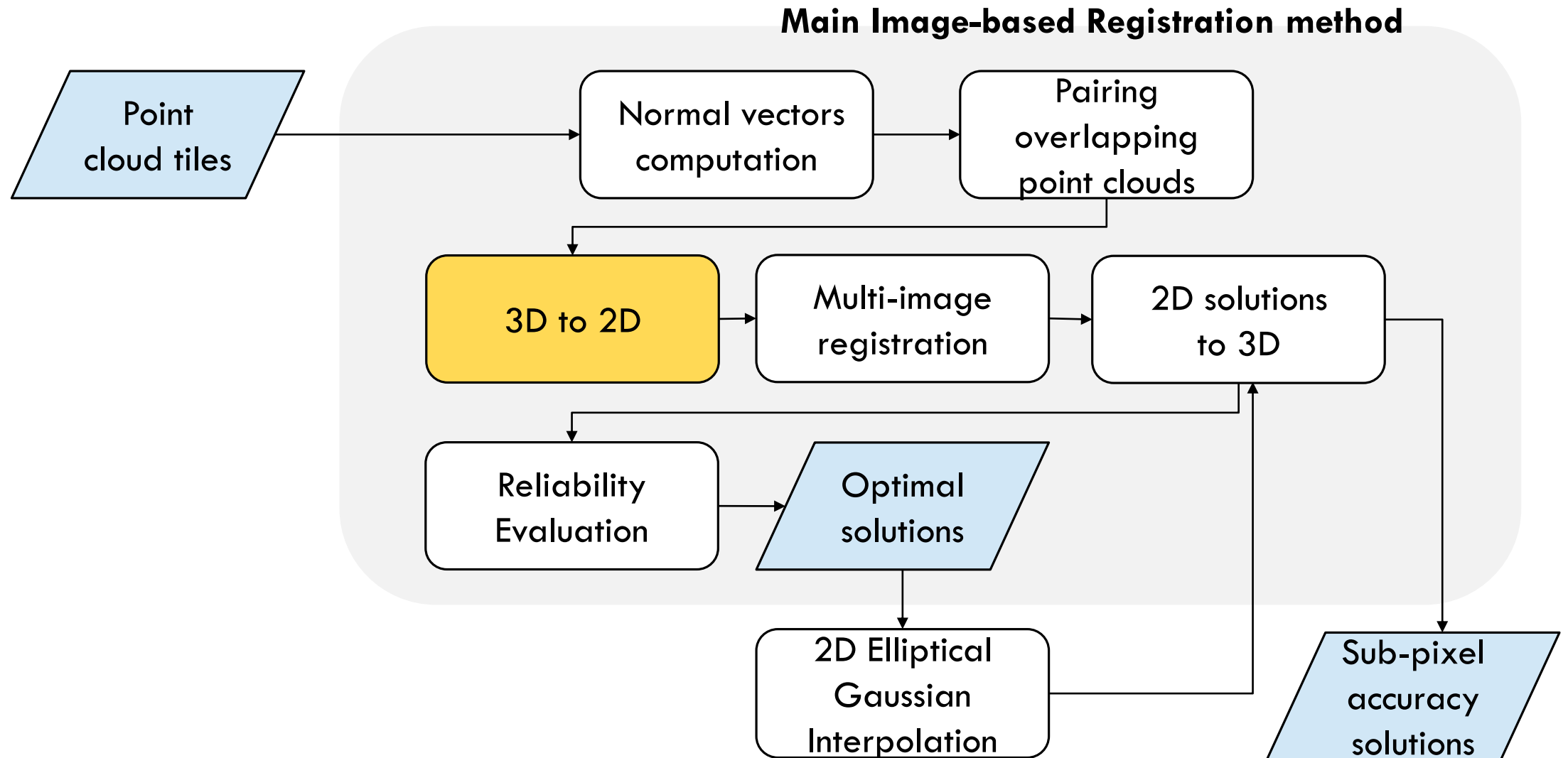




# PAIRING OVERLAPPING POINT CLOUDS

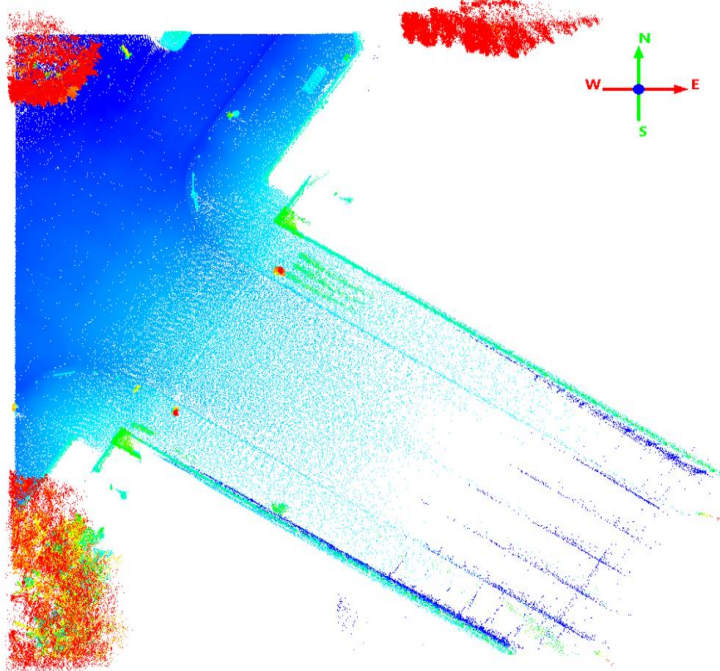


# 3D TO 2D

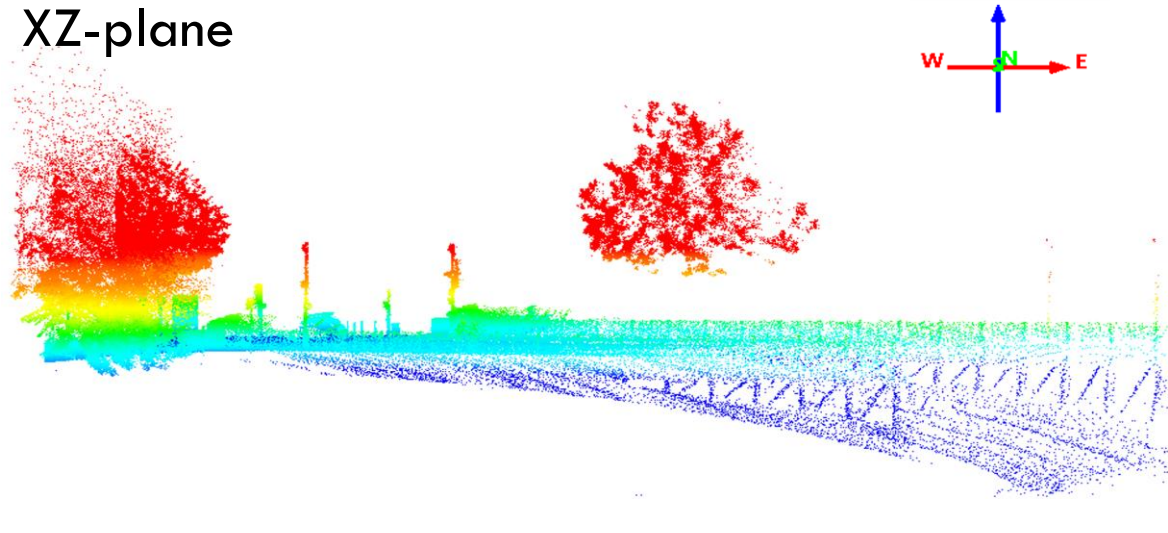


# 3D TO 2D: REDUCTION OF DIMENSIONS (1) : MAIN EXAMPLE

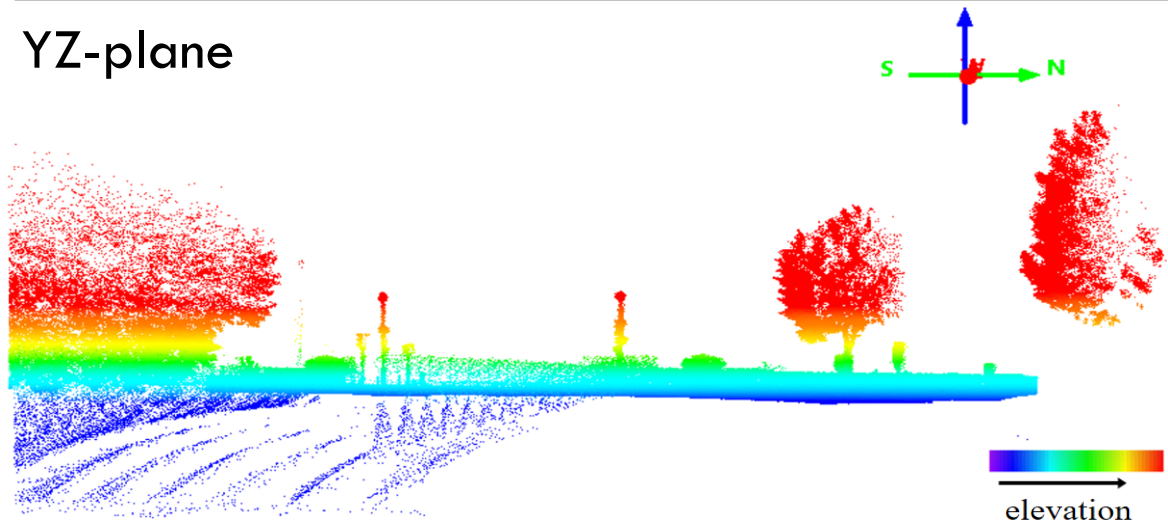
XY-plane



XZ-plane



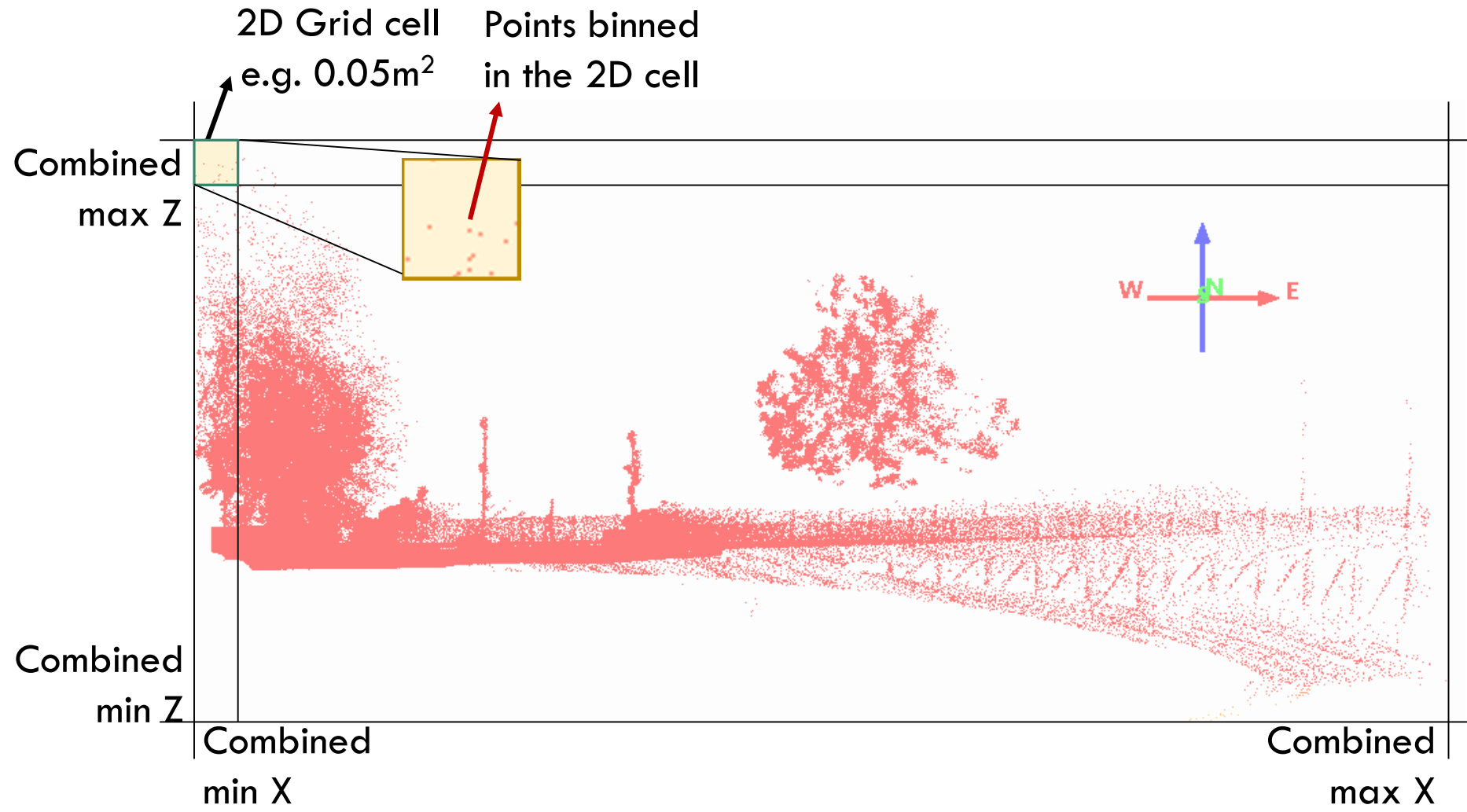
YZ-plane





# 3D TO 2D: REDUCTION OF DIMENSIONS (1)

- 2D grid creation **based** on the two dimensions of each **plane**
- **Spatial binning** of points into 2D grid cells

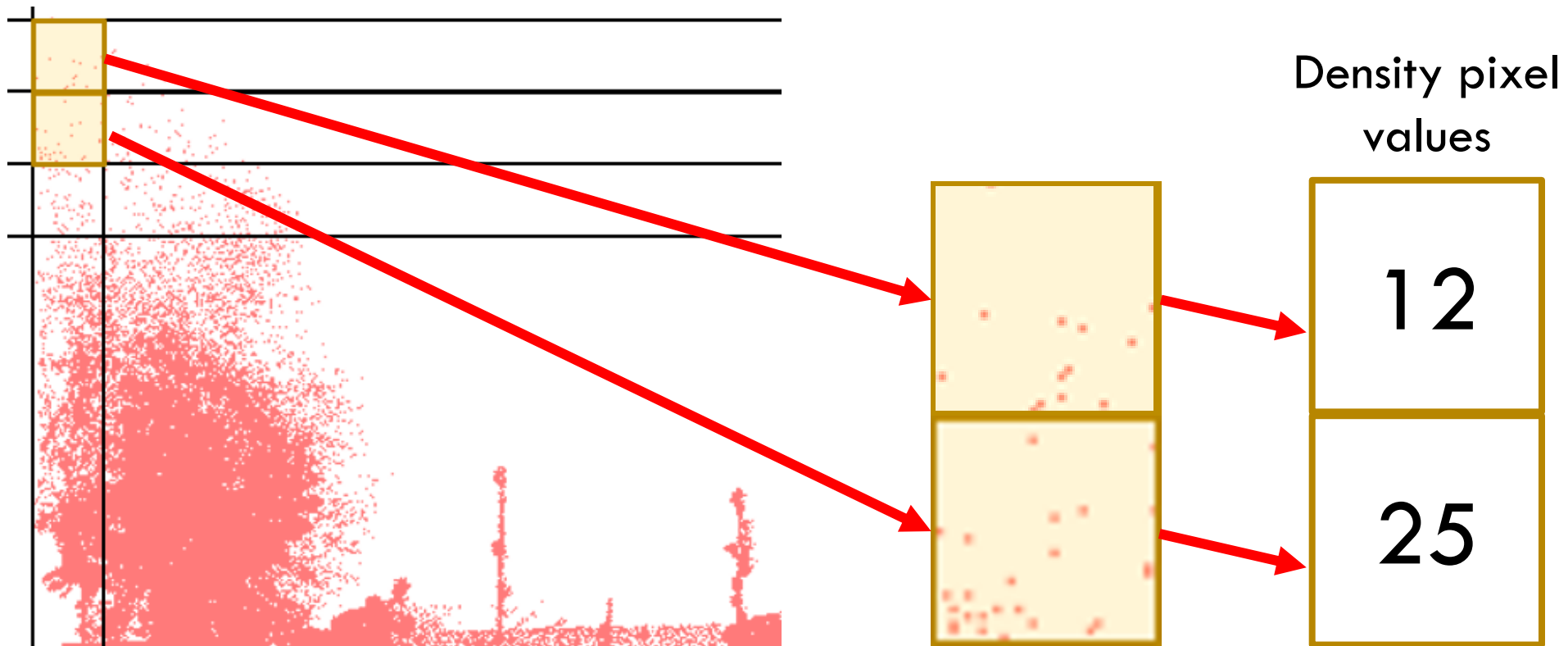


## 3D TO 2D: ASSIGNING VALUES TO THE 2D CELLS (2)

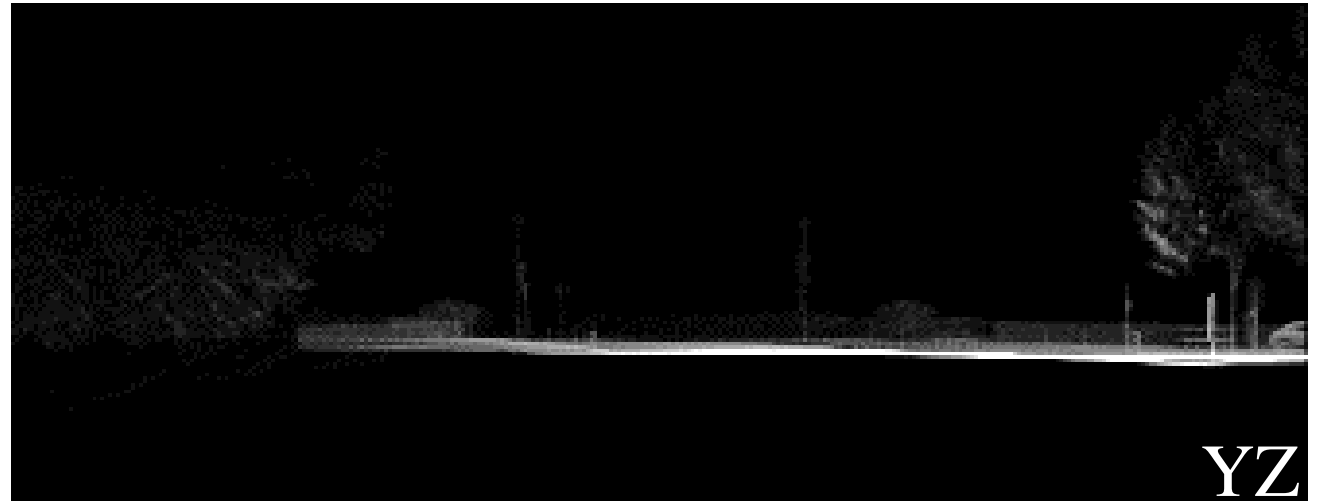
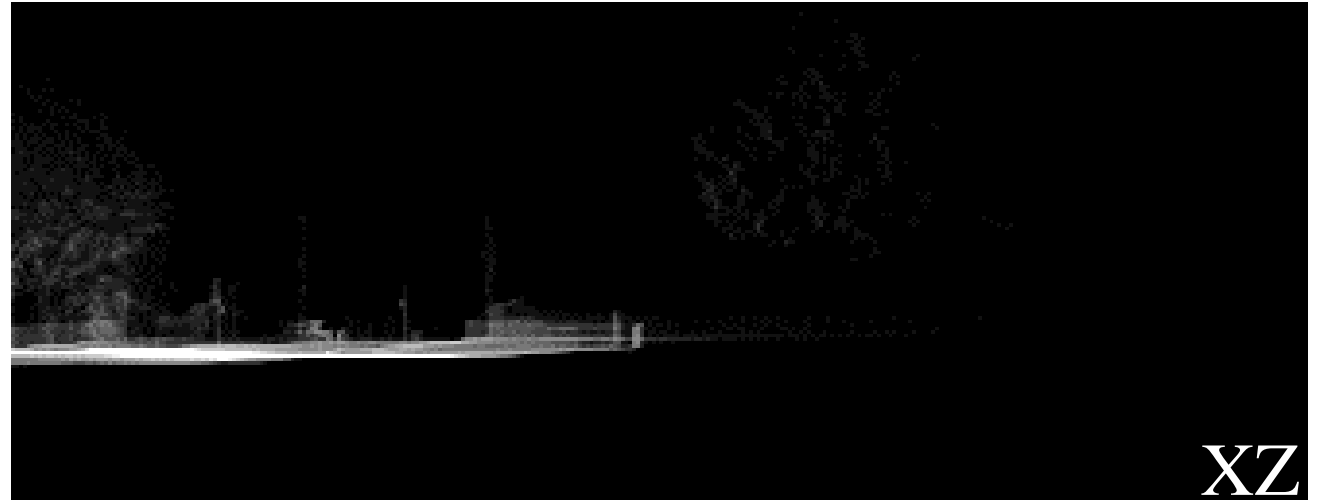
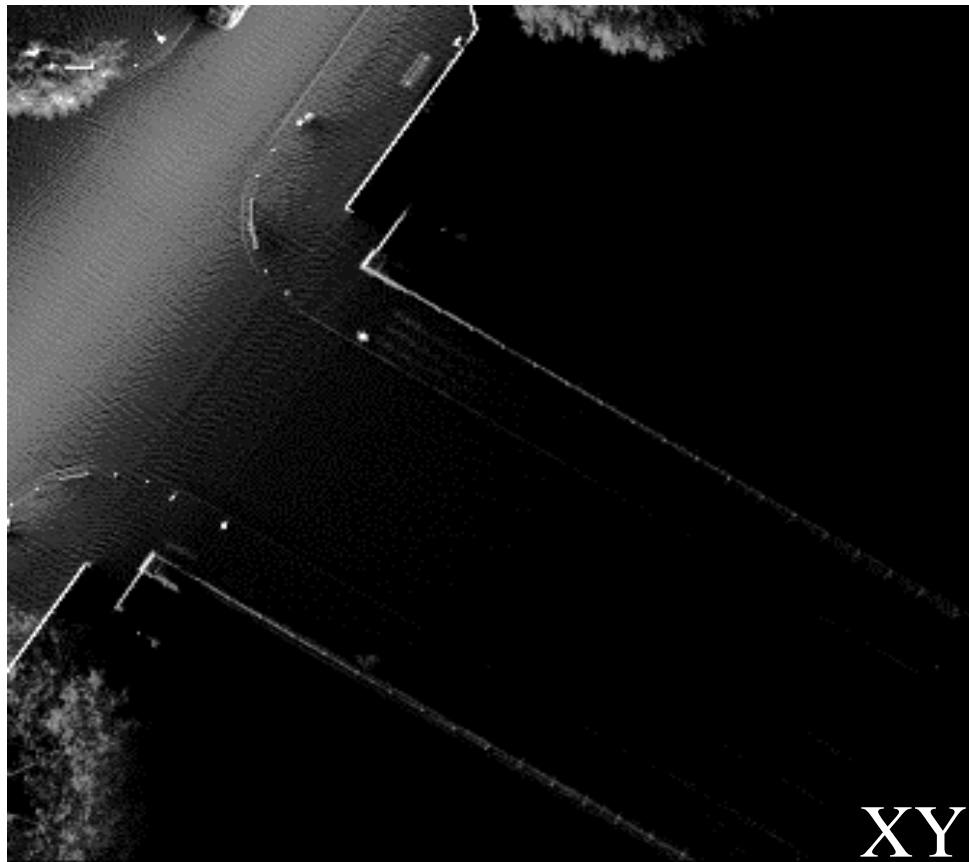
- Different attributes of points → Different image types

# 3D TO 2D: DENSITY

- DENSITY



# 3D TO 2D: DENSITY





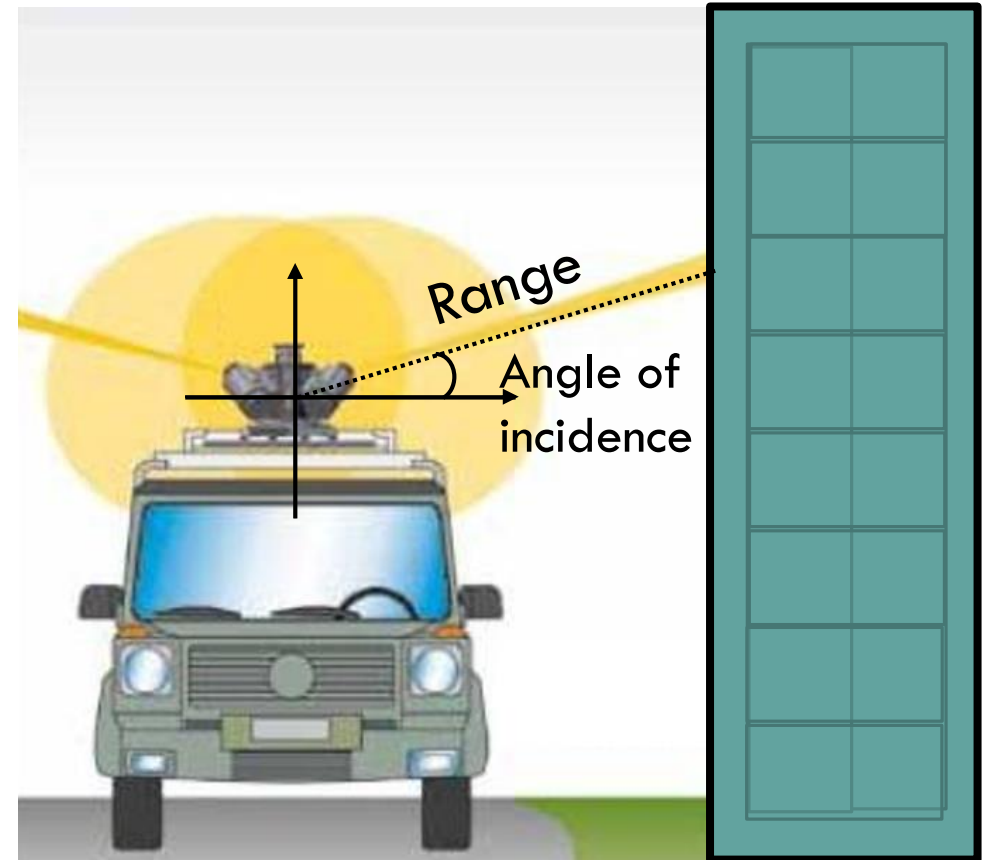
# 3D TO 2D: INTENSITY

- **INTENSITY: strength of the backscattered signal**

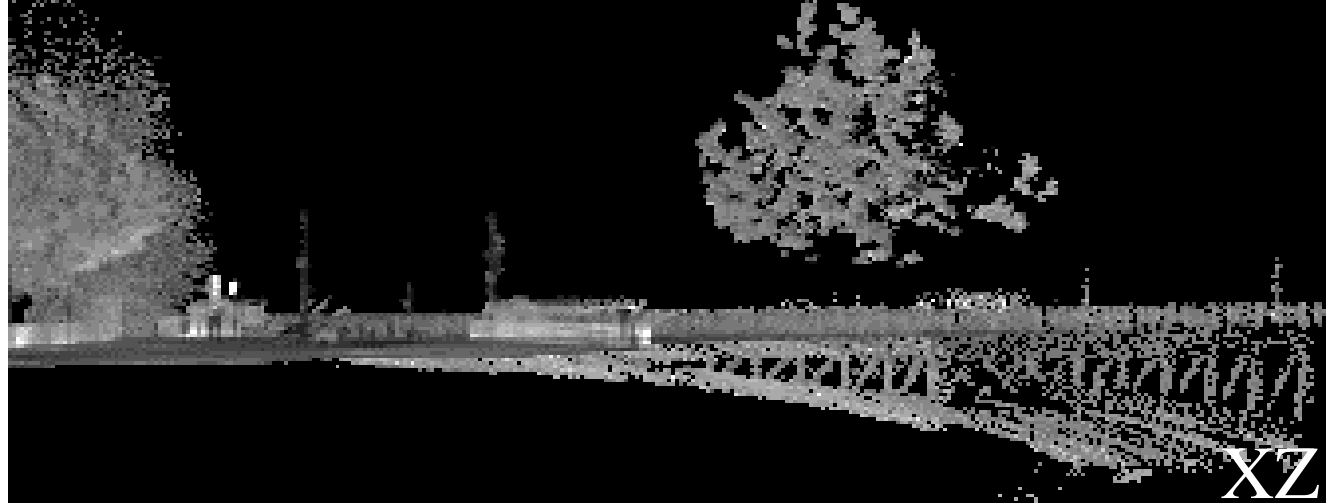
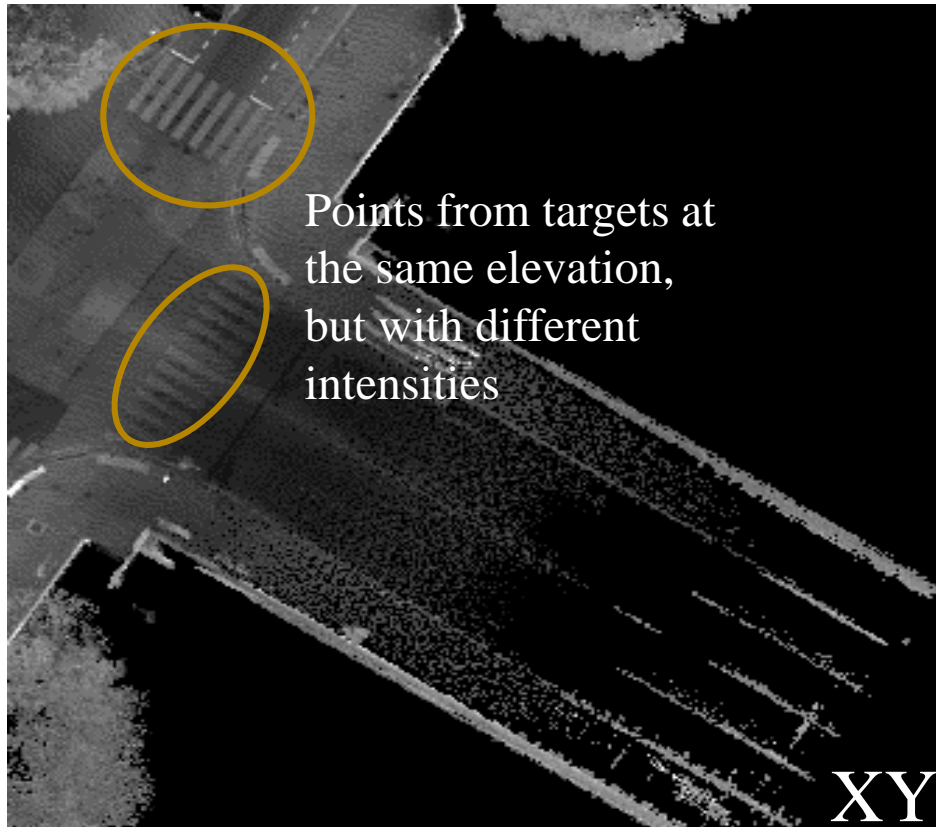
**Range:** scanning system - object

Object's **reflectivity**

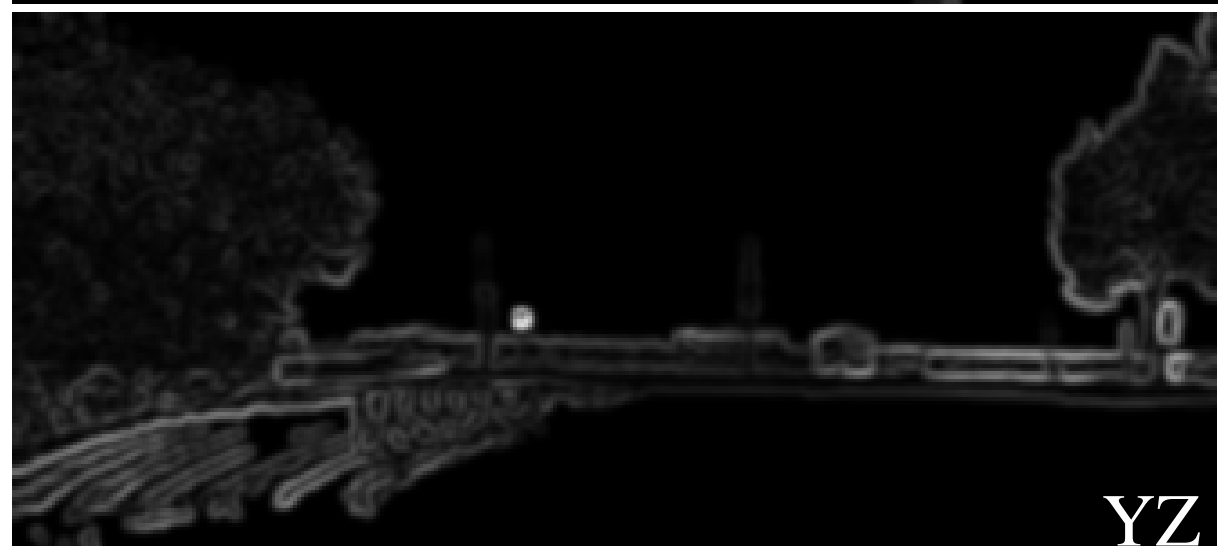
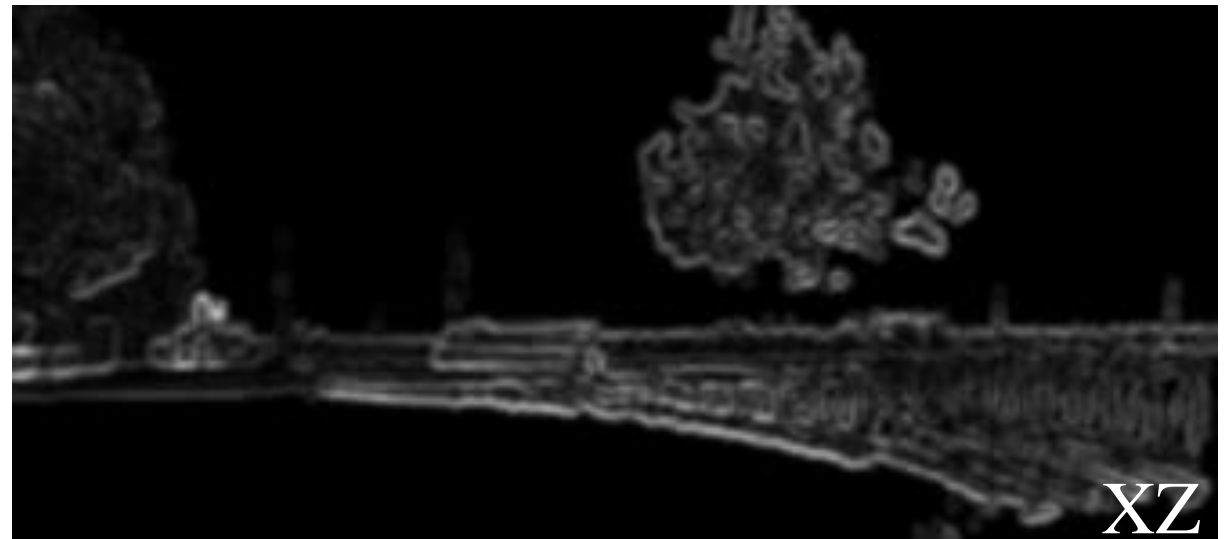
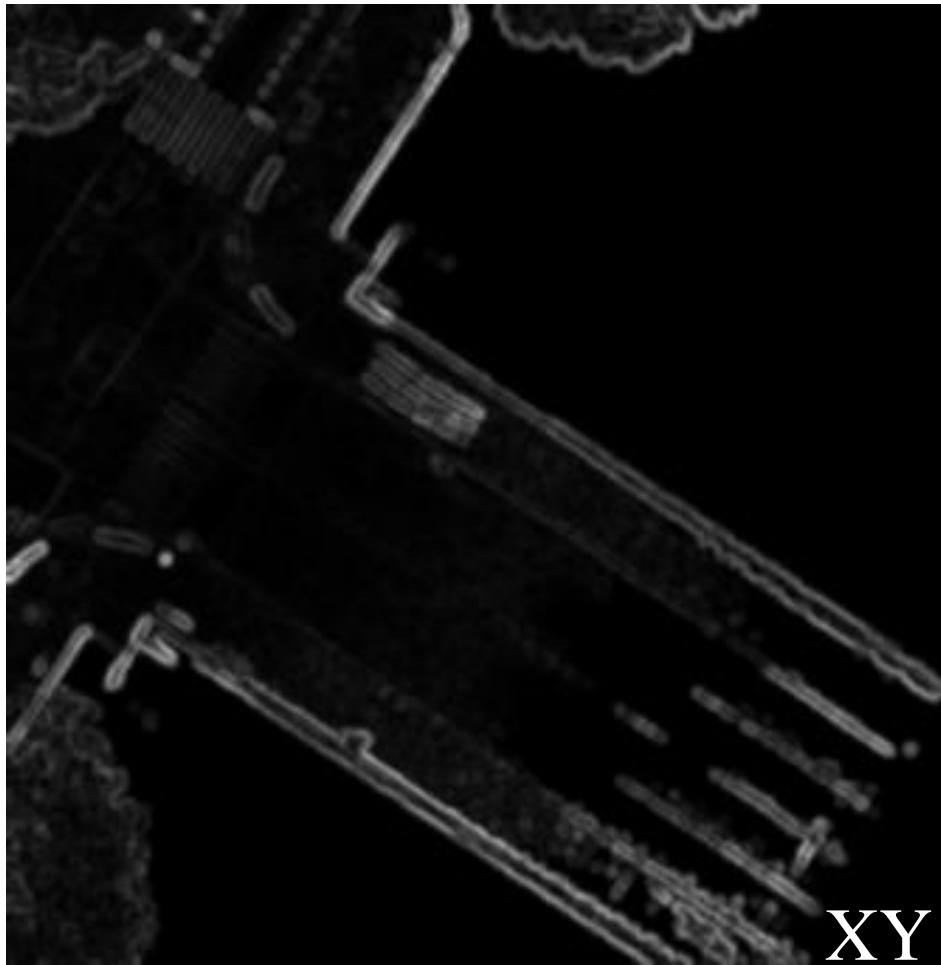
**Angle of incidence :** scanning system – object



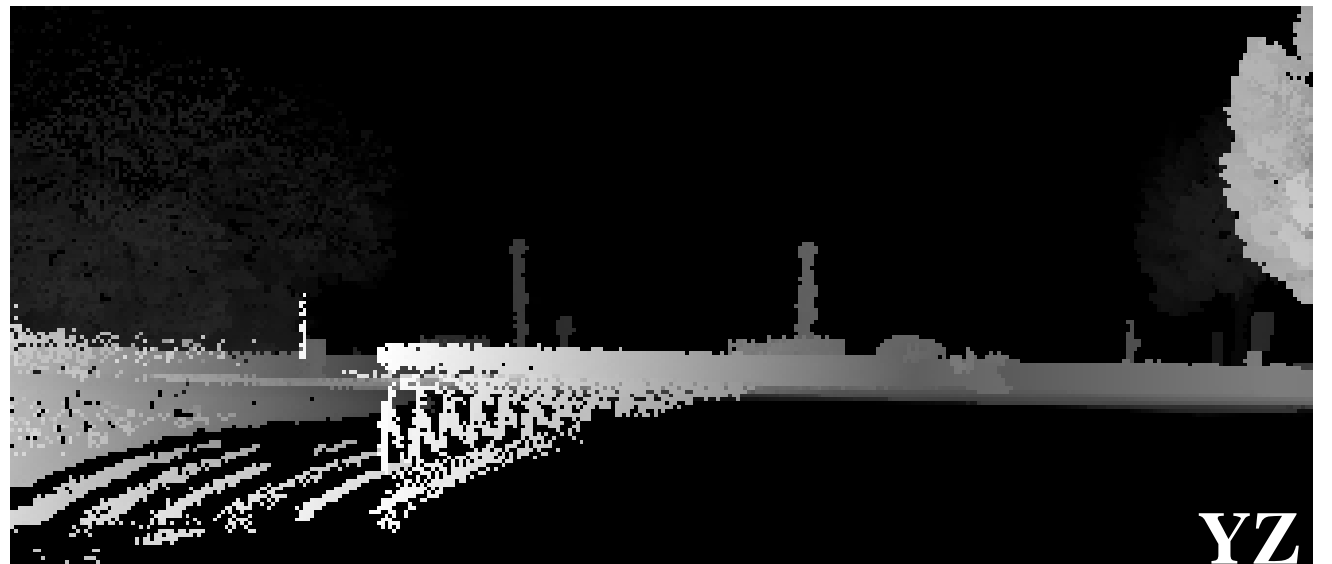
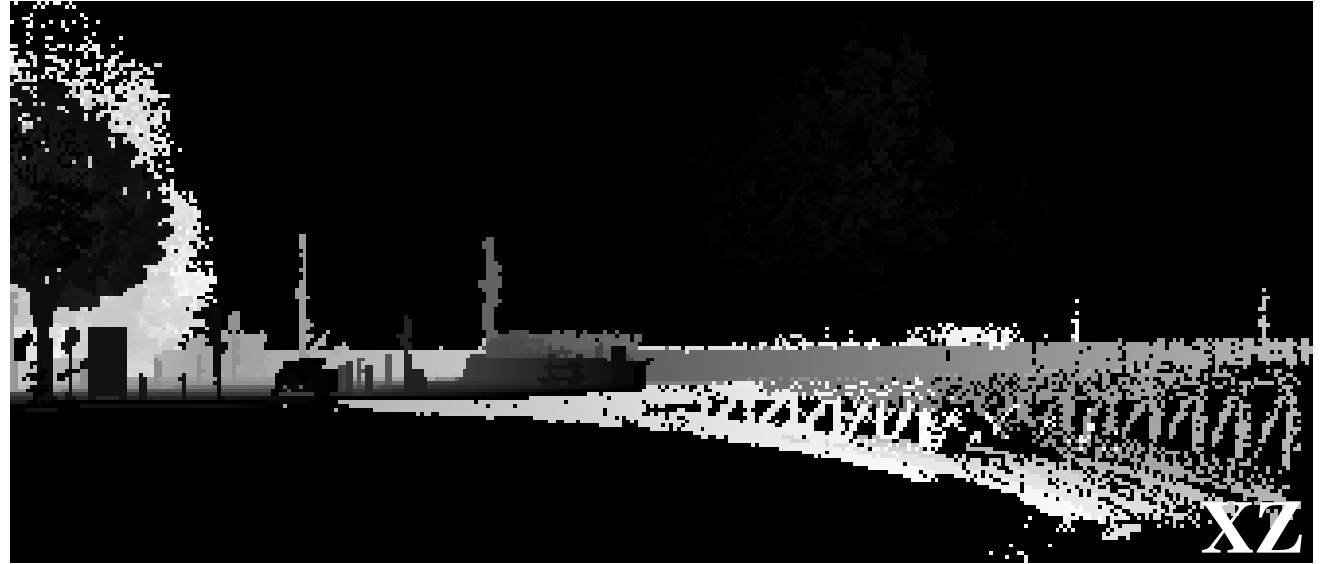
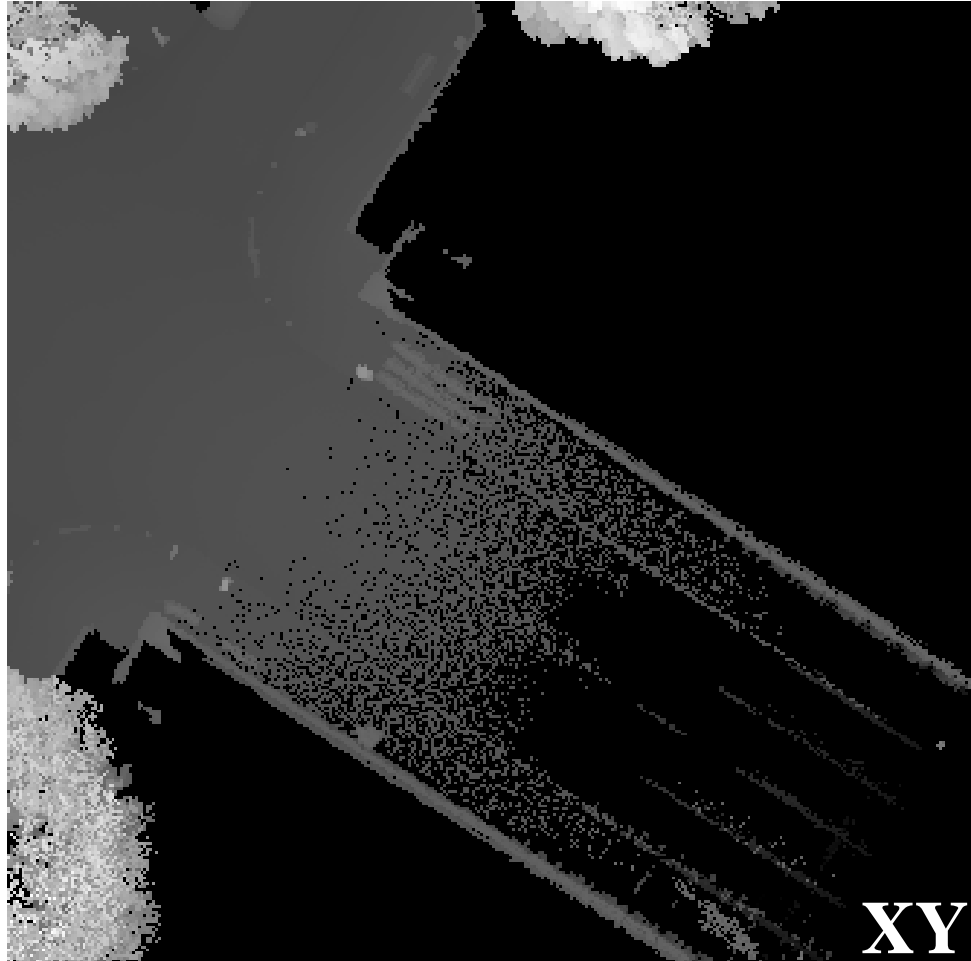
# 3D TO 2D: INTENSITY



# 3D TO 2D: INTENSITY GRADIENT: RATE OF INTENSITY VALUES CHANGE

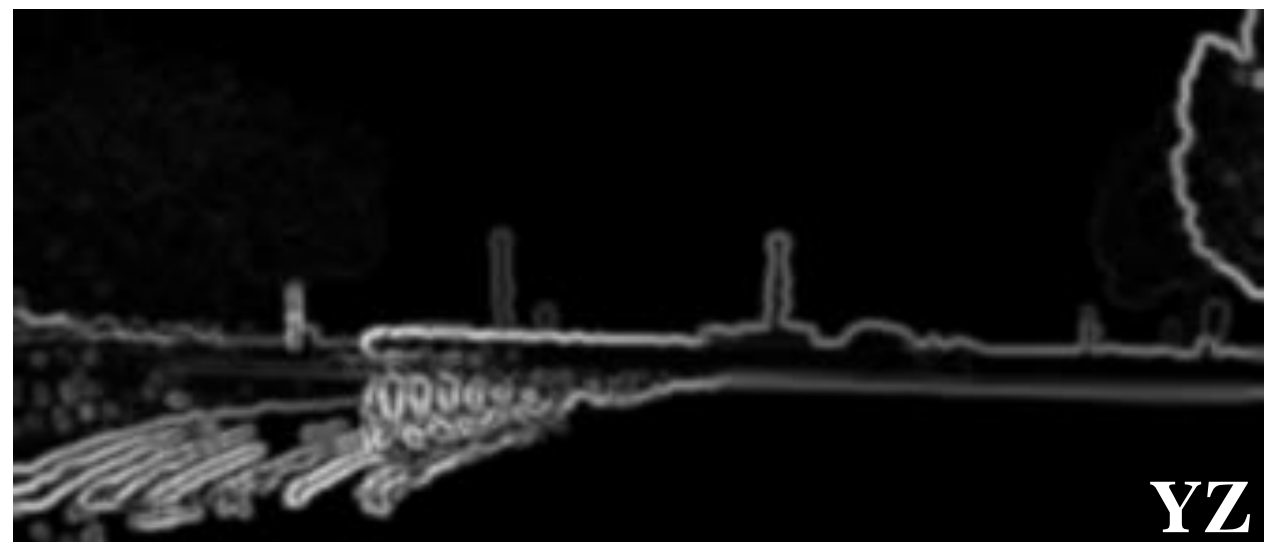
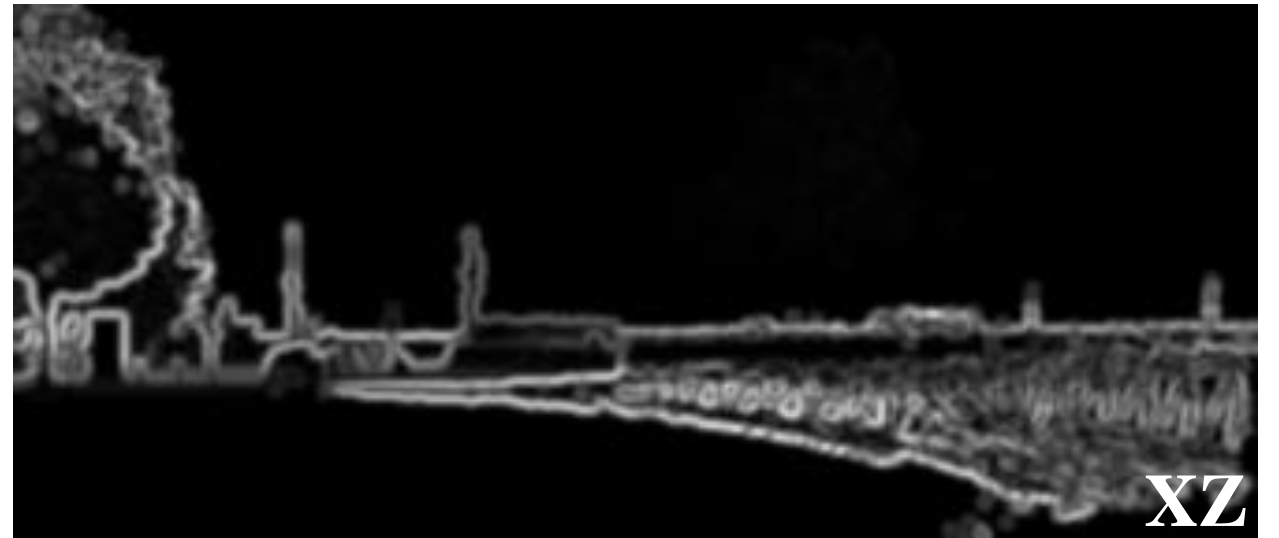
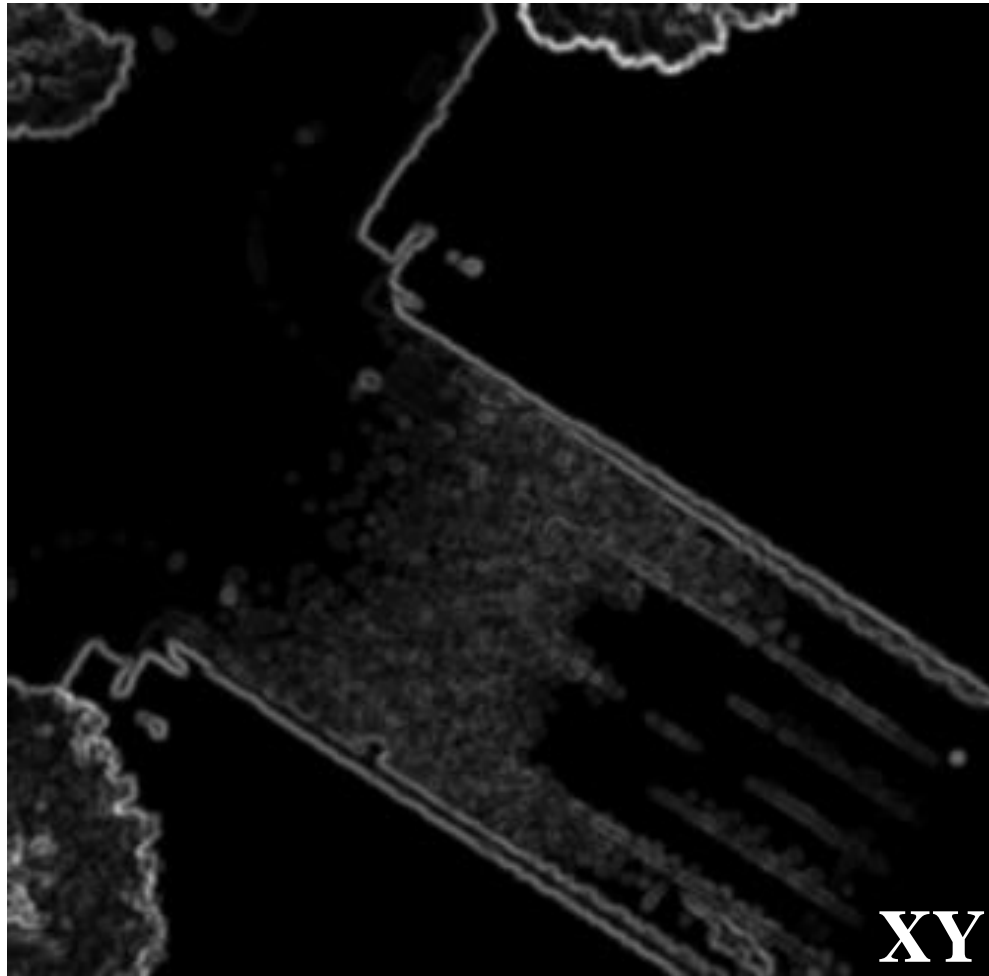


# 3D TO 2D: DEPTH

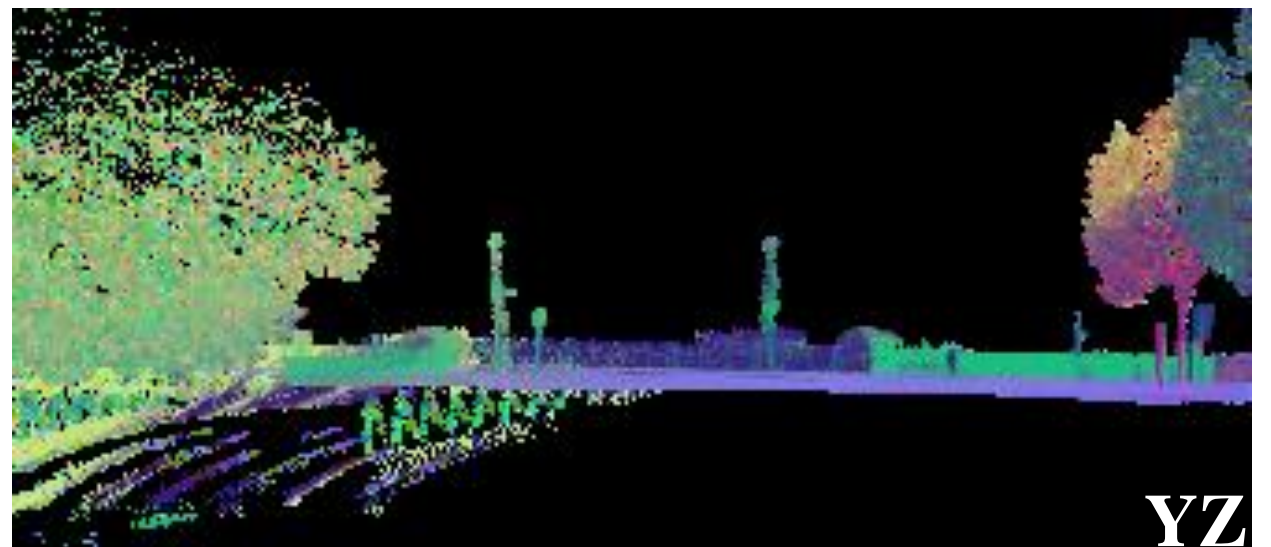
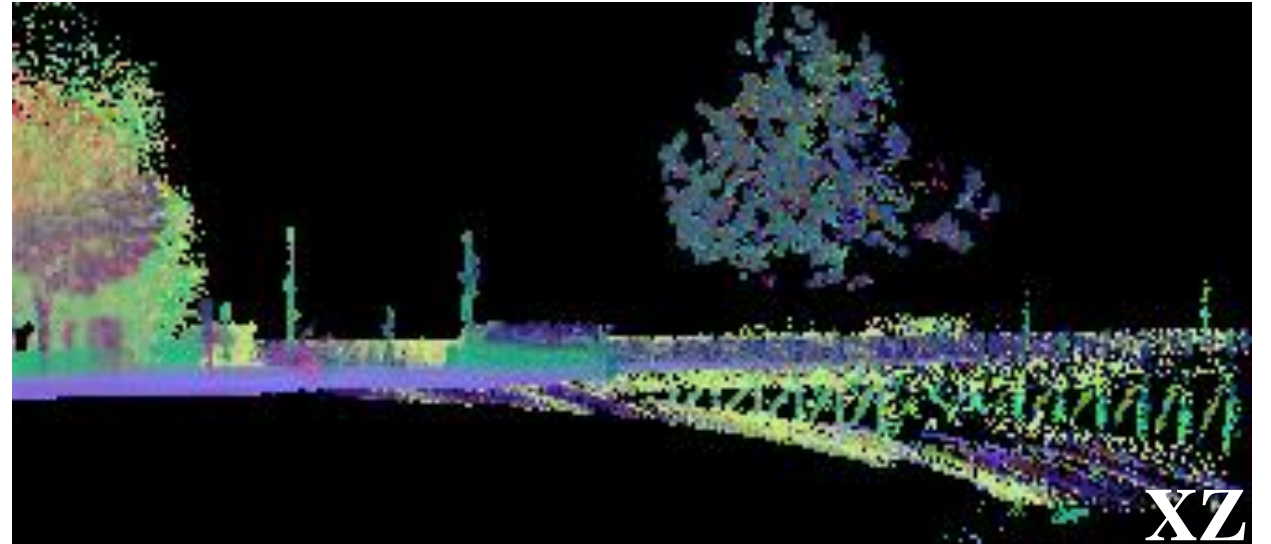
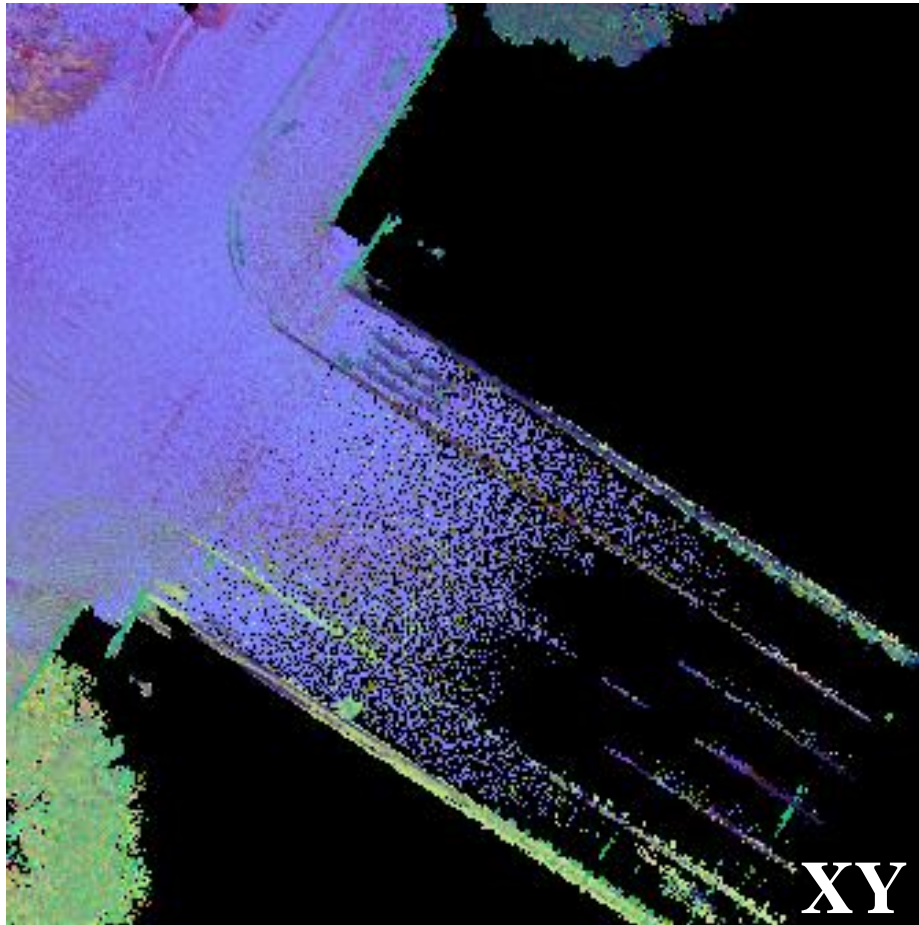




# 3D TO 2D: DEPTH GRADIENT: RATE OF DEPTH VALUES CHANGE

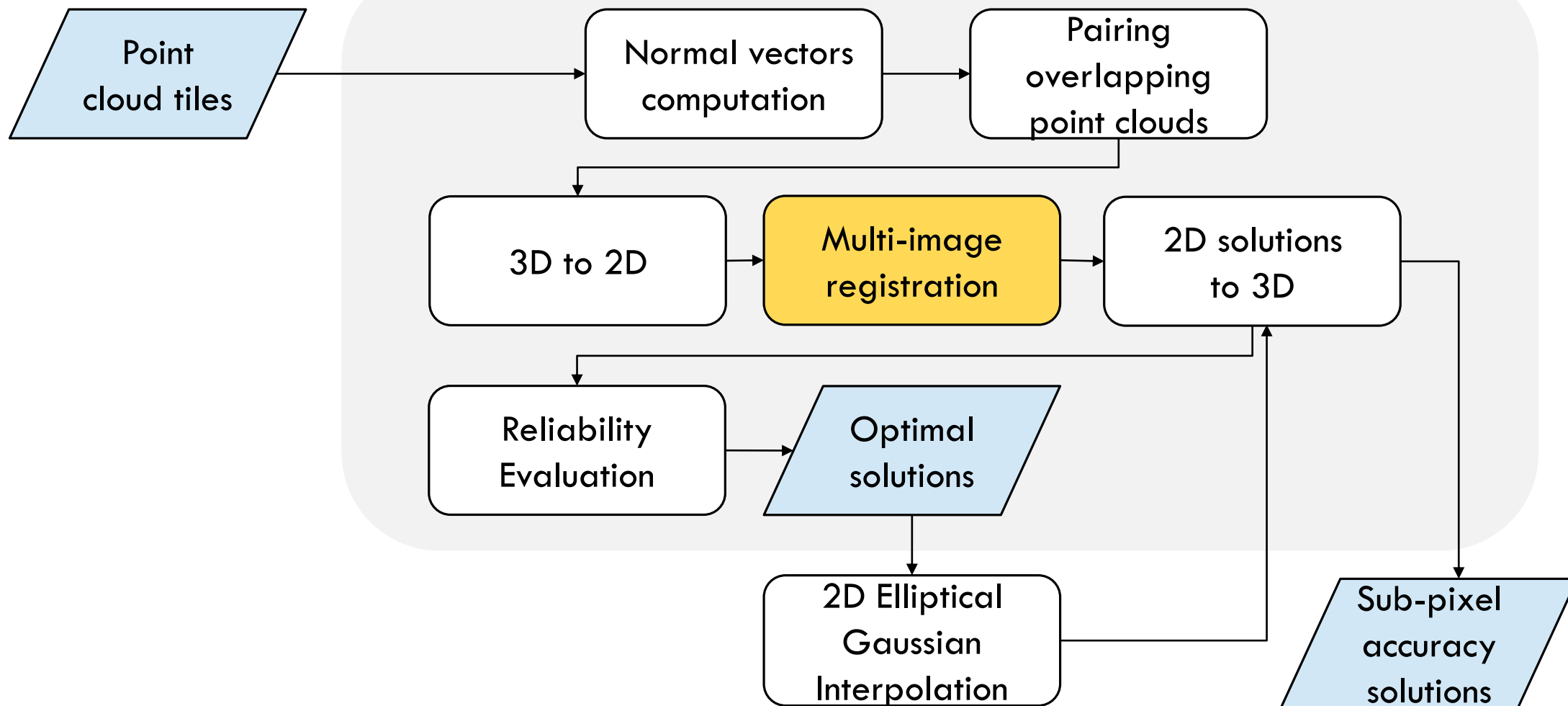


# 3D TO 2D: NORMAL VECTORS: ORIENTATION OF SURFACES

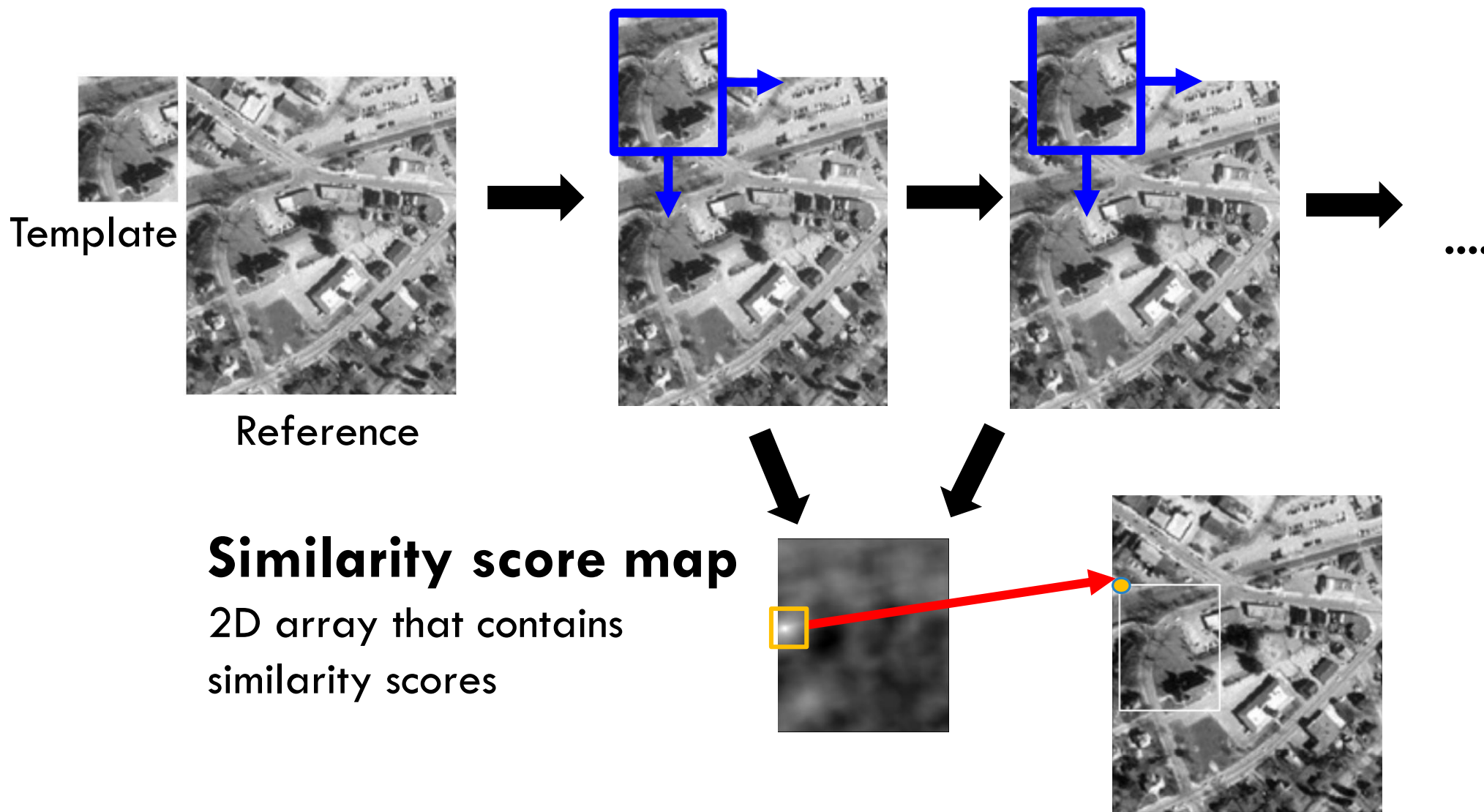


# IMAGE PAIRWISE REGISTRATION

## Main Image-based Registration method



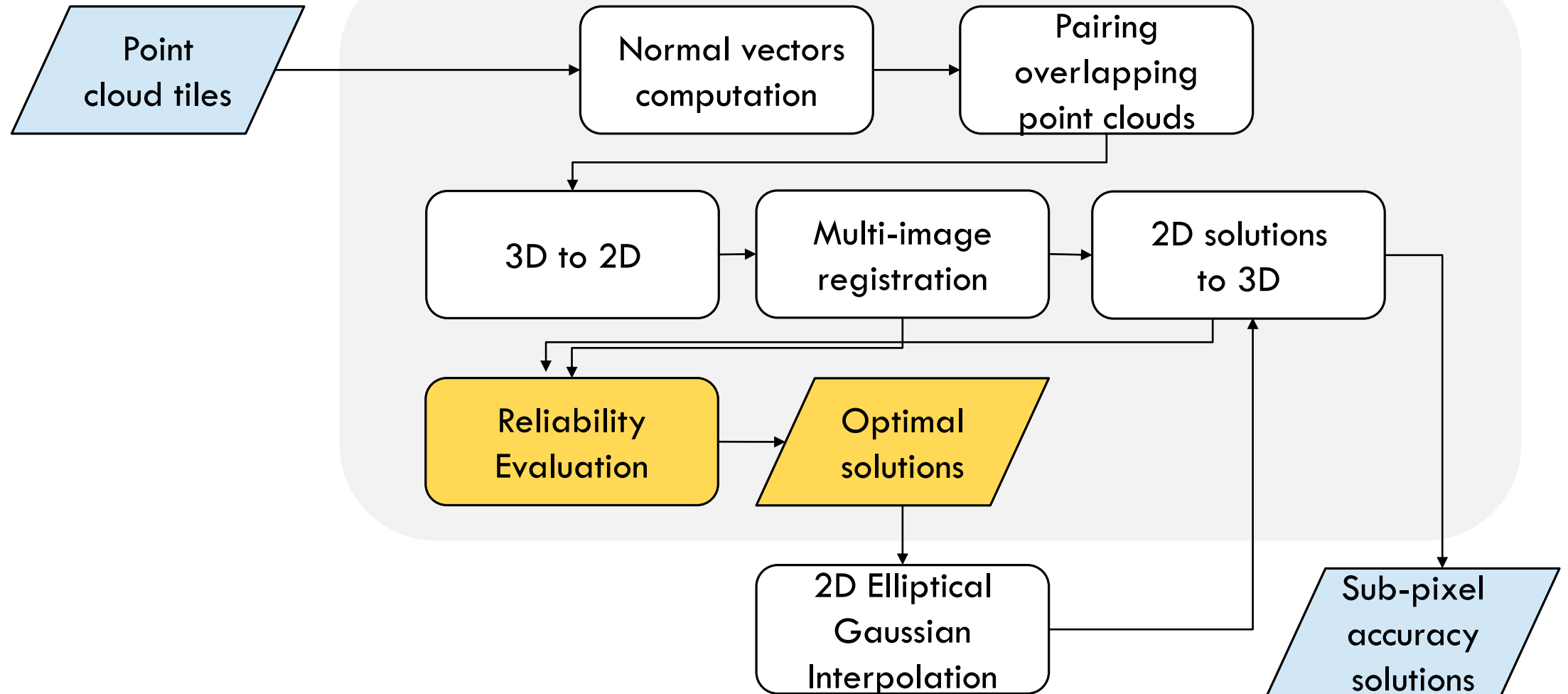
# IMAGE PAIRWISE REGISTRATION: TEMPLATES MATCHING



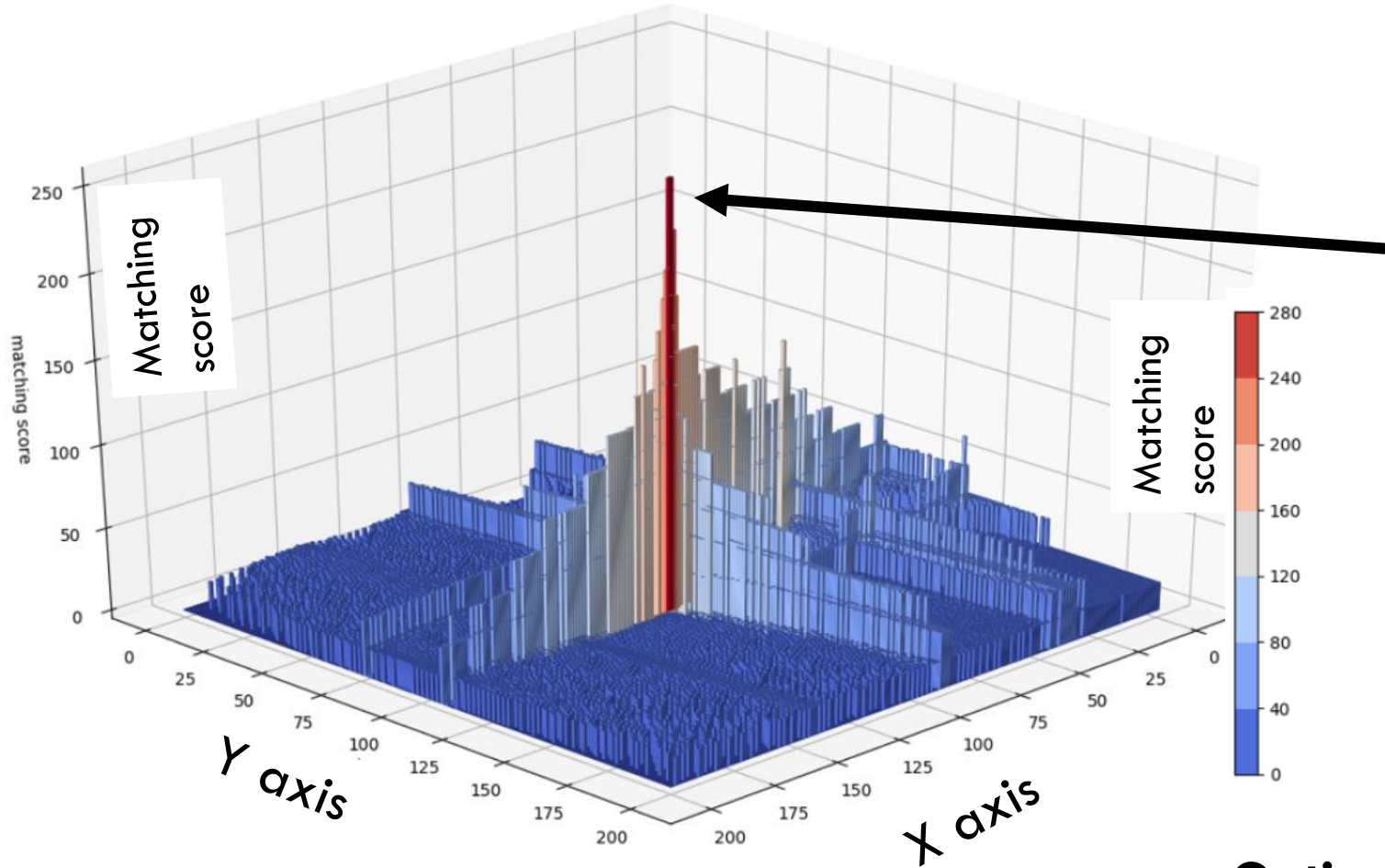


# RELIABILITY EVALUATION & OPTIMAL SOLUTIONS

## Main Image-based Registration method



# RELIABILITY EVALUATION & OPTIMAL SOLUTIONS



XY Similarity scores map visualized in 3D

Evaluation of strength and uniqueness of the highest similarity values

Score map accepted

Matching accepted

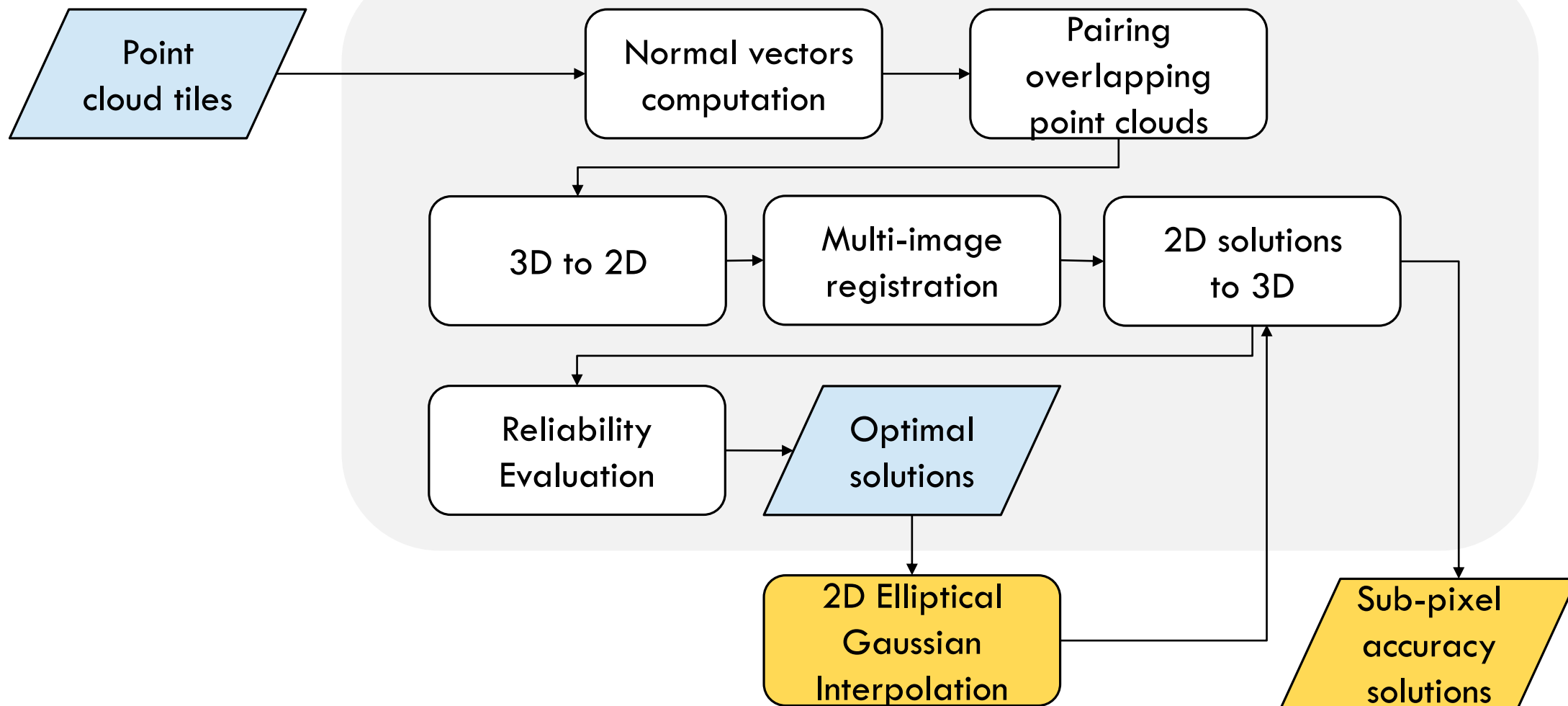
Score map rejected

Matching rejected

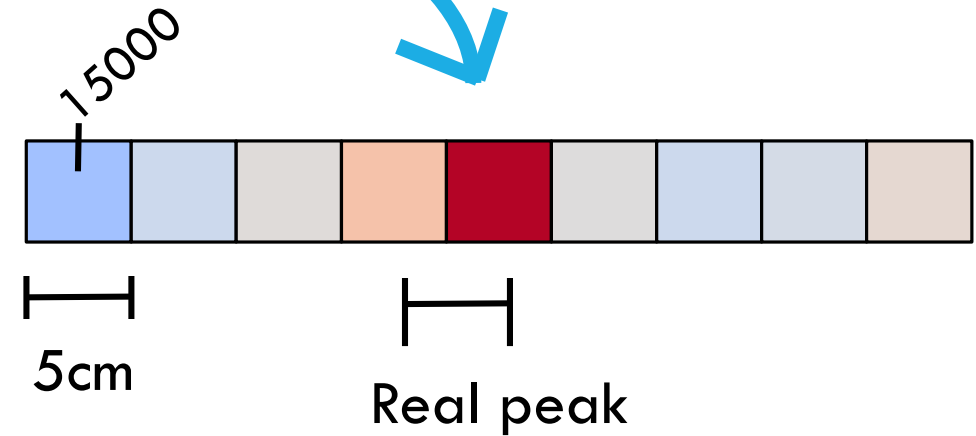
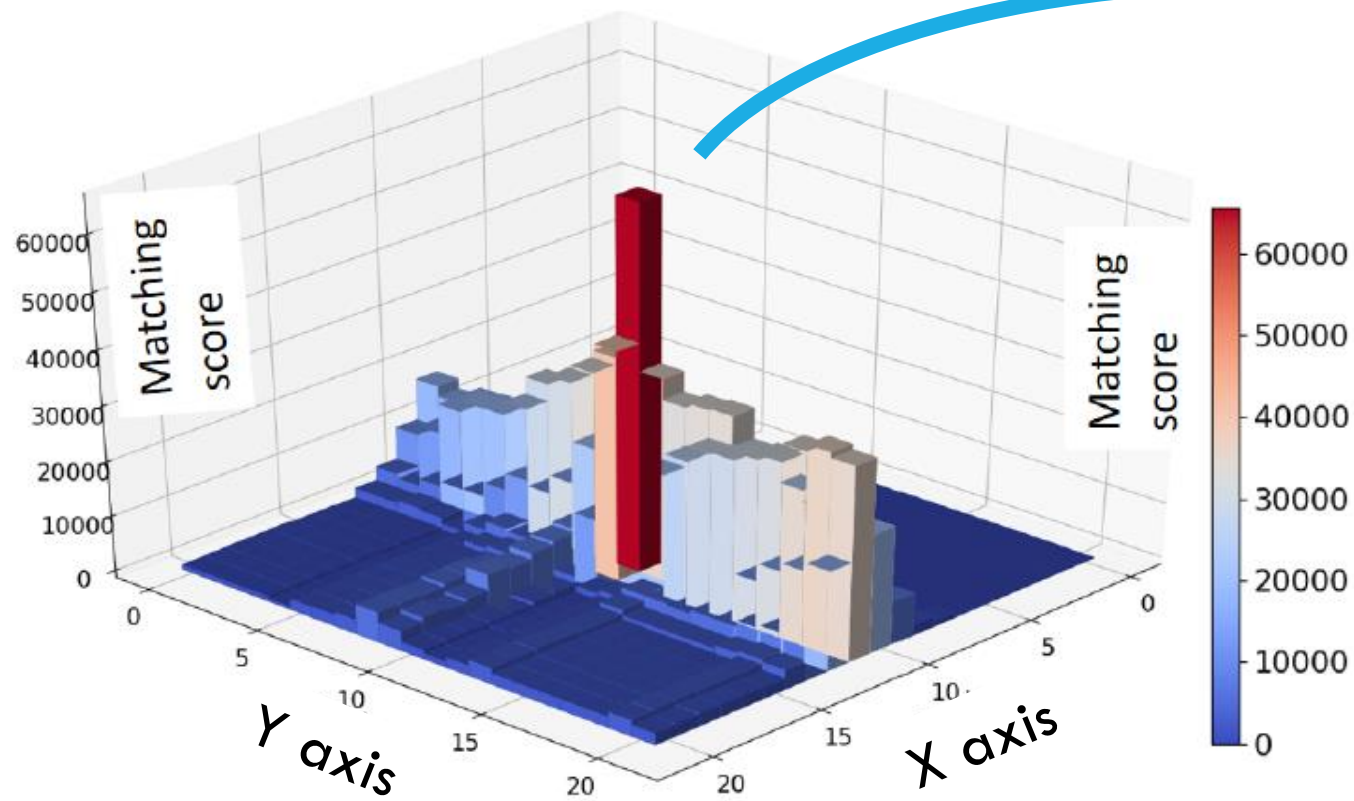
**Optimal X** = mean (accepted x values)

# IMAGE PAIRWISE REGISTRATION

## Main Image-based Registration method



# SUB-PIXEL ACCURACY



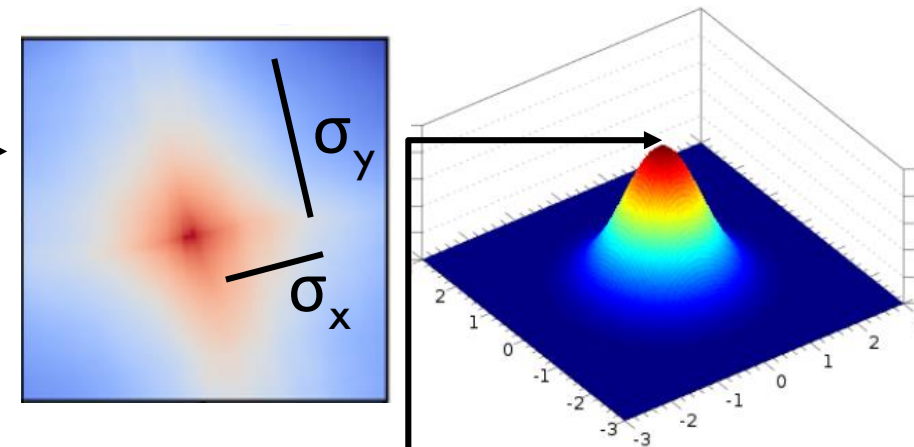
Every 5cm → 1 value



# 2D ELLIPTICAL GAUSSIAN INTERPOLATION

1. **Score maps** that **passed** the evaluation criteria
2. The **peak** computed earlier is obtained
3. The **neighboring region** to the peak pixel is found
4. Least squares adjustment to find the optimal 2D elliptical Gaussian fit
5. The **highest point** of the **fitted Gaussian surface** is the new **matching location** with **subpixel accuracy**

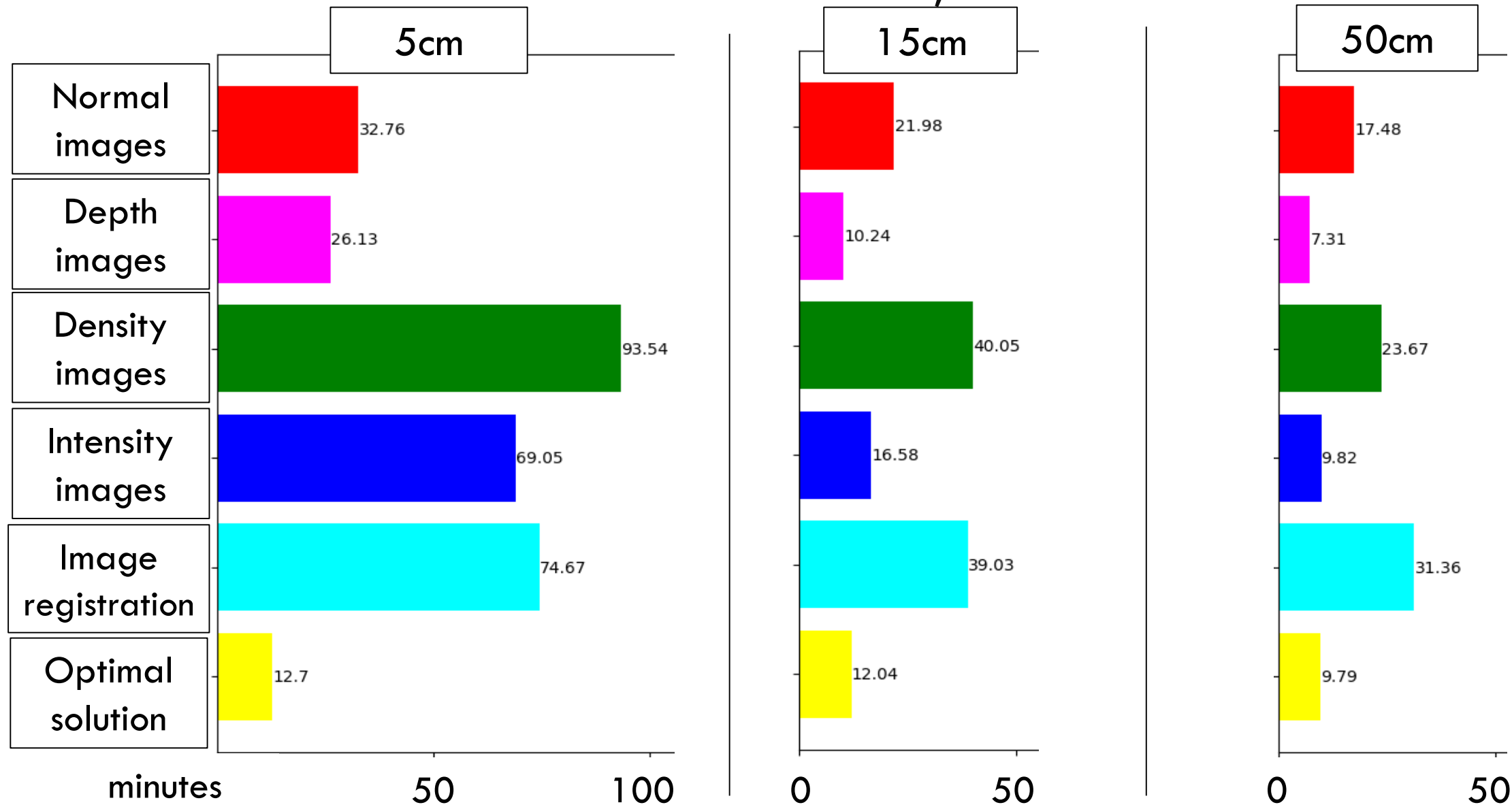
0	2362	7134	17489	28672	12523	5242	3195	2185
0	2521	5149	16805	30299	11648	4783	3780	2759
0	908	2956	17913	33446	15011	4952	3948	2903
0	593	5777	21010	37484	22585	9557	8919	8291
737	1470	7607	19113	65535	41452	12594	3818	2634
428	1392	2364	14868	36470	21889	2100	2223	1045
34	429	3847	13457	32818	22204	4622	2990	1891
0	0	3223	11988	33745	26056	6424	4960	3855
773	1002	4673	11008	34129	28899	8439	4935	2873



# METHOD'S PERFORMANCE

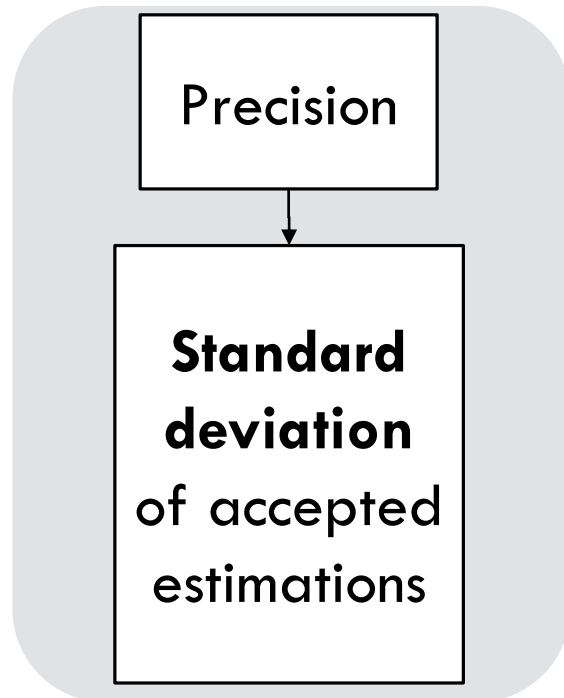


# EXECUTION TIMES: 5886 IMAGE PAIRS, 218 POINT CLOUD PAIRS

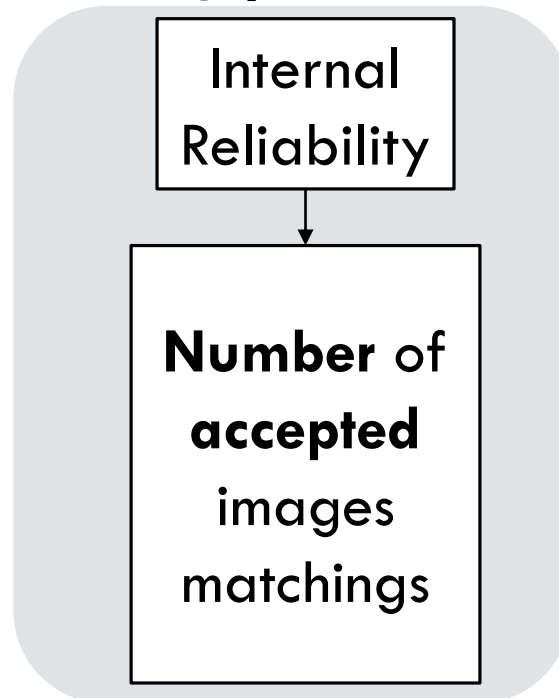


# OVERALL QUALITY EVALUATION

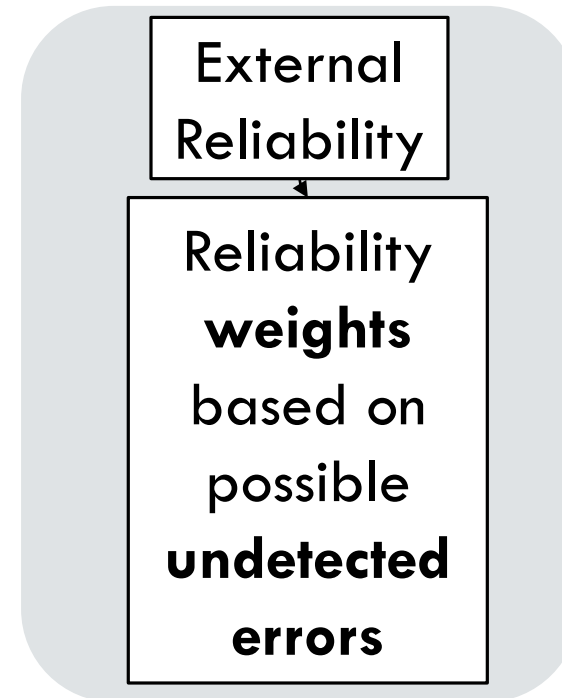
Is a translation estimation **precisely measured?**



Is a translation estimation **strongly checked?**



What is the **influence** from possible **undetected errors?**

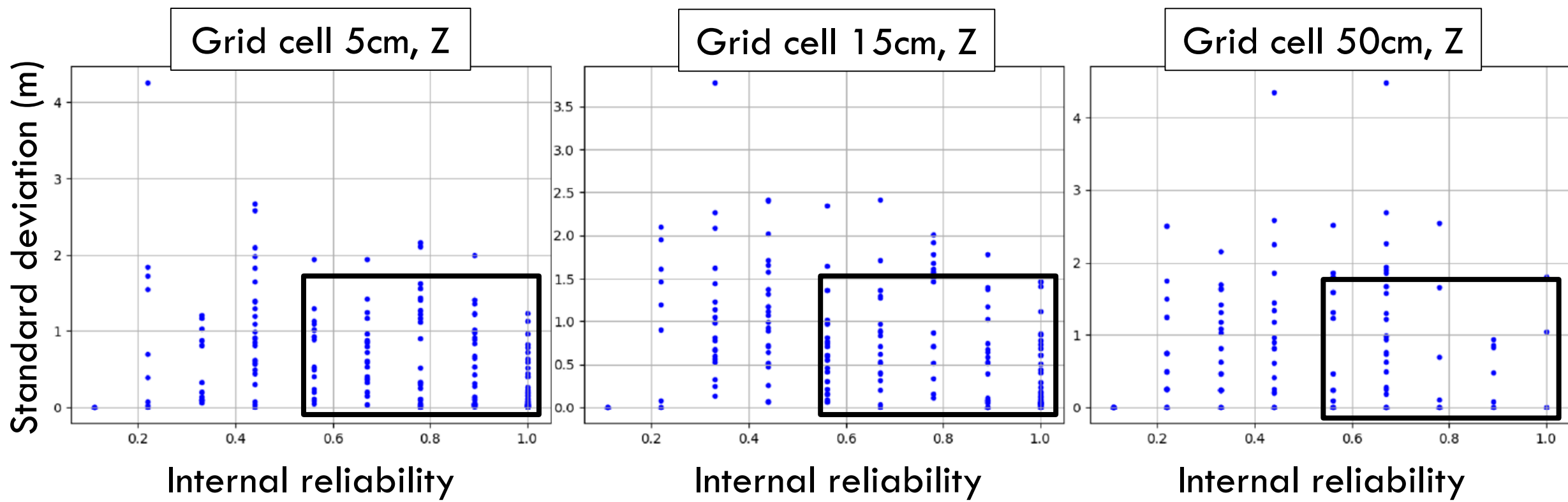




# INTERNAL RELIABILITY — PRECISION

**LOW** Standard deviation (MIN = 0)

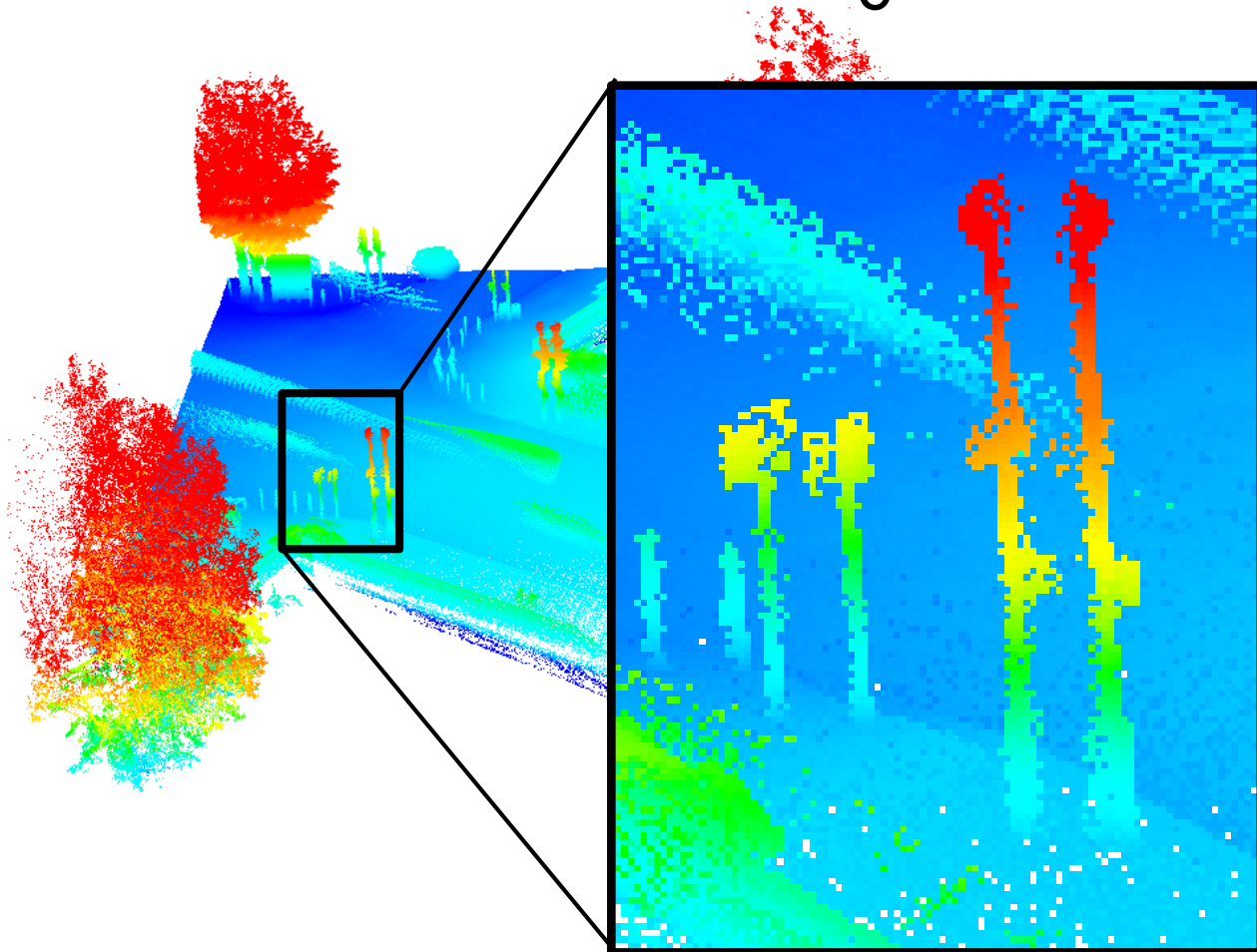
**HIGH** Internal reliability (MAX = 1)



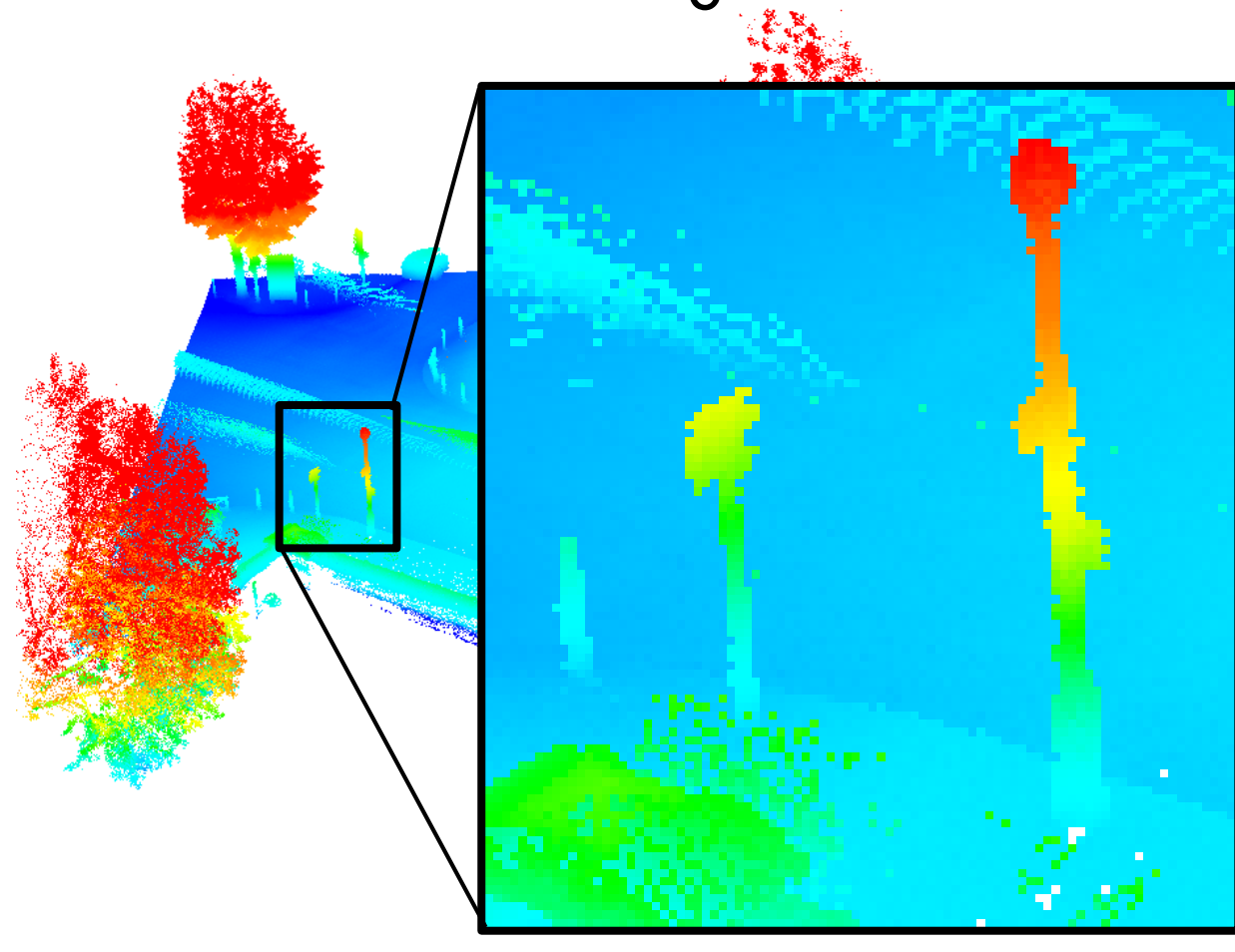
**Number of accepted image matchings**

# CASE 1: BIG OVERLAP, CHANGES IN THE SCENE, DISTINCT CORRESPONDING OBJECTS

Before registration

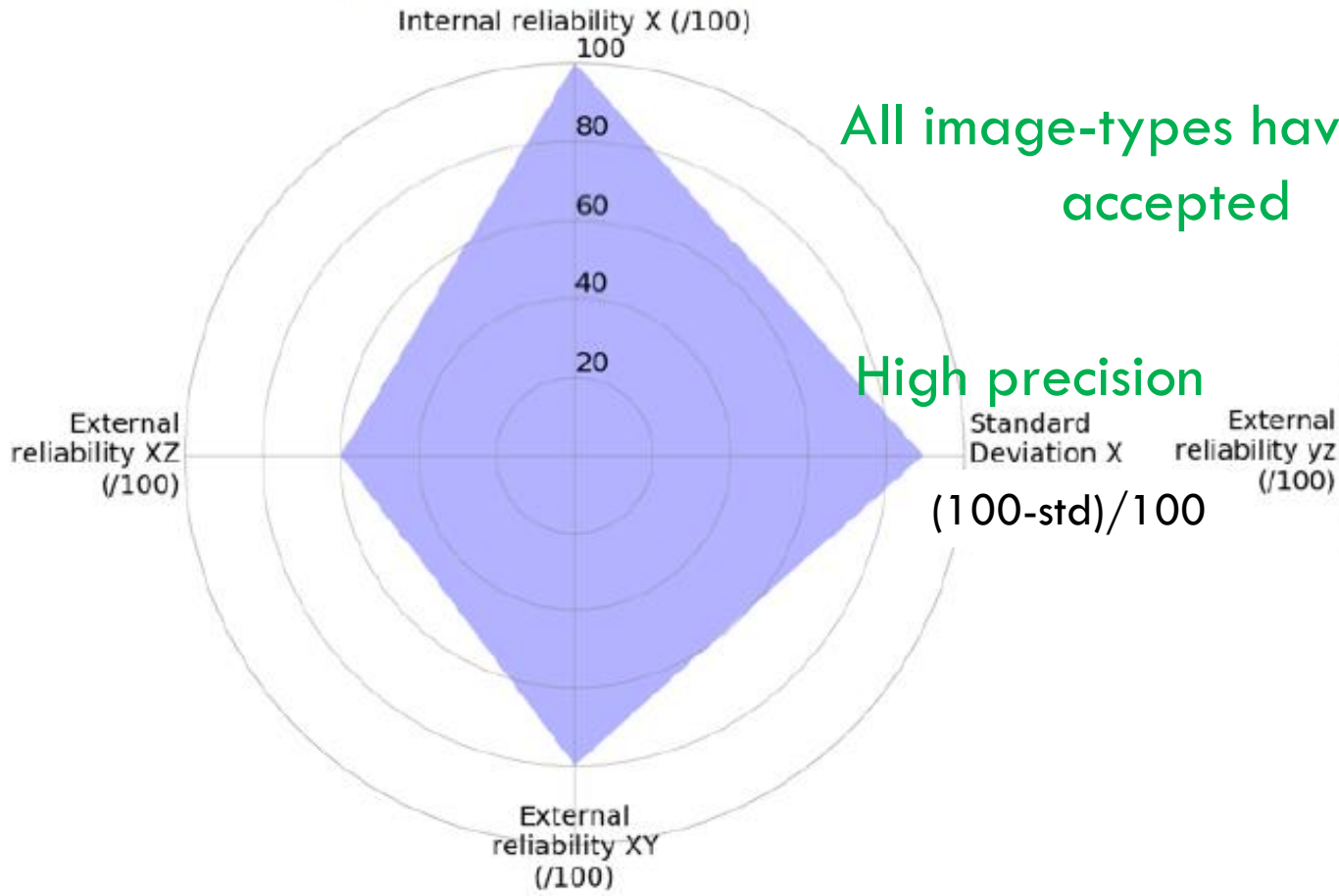


After registration



# CASE 1: BIG OVERLAP, CHANGES IN THE SCENE, DISTINCT CORRESPONDING OBJECTS

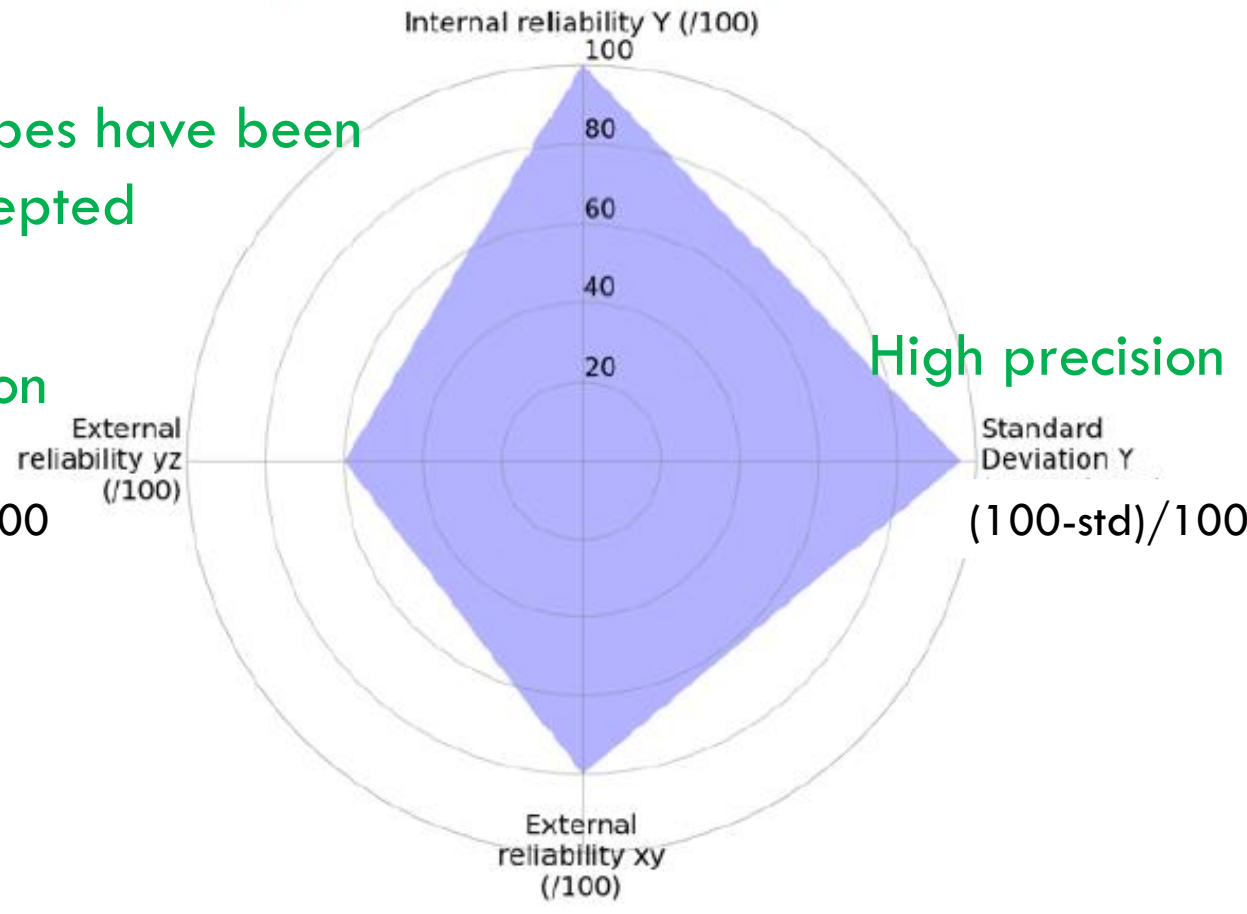
X translation = -0.67m



All image-types have been accepted

High precision

Y translation = -0.03m

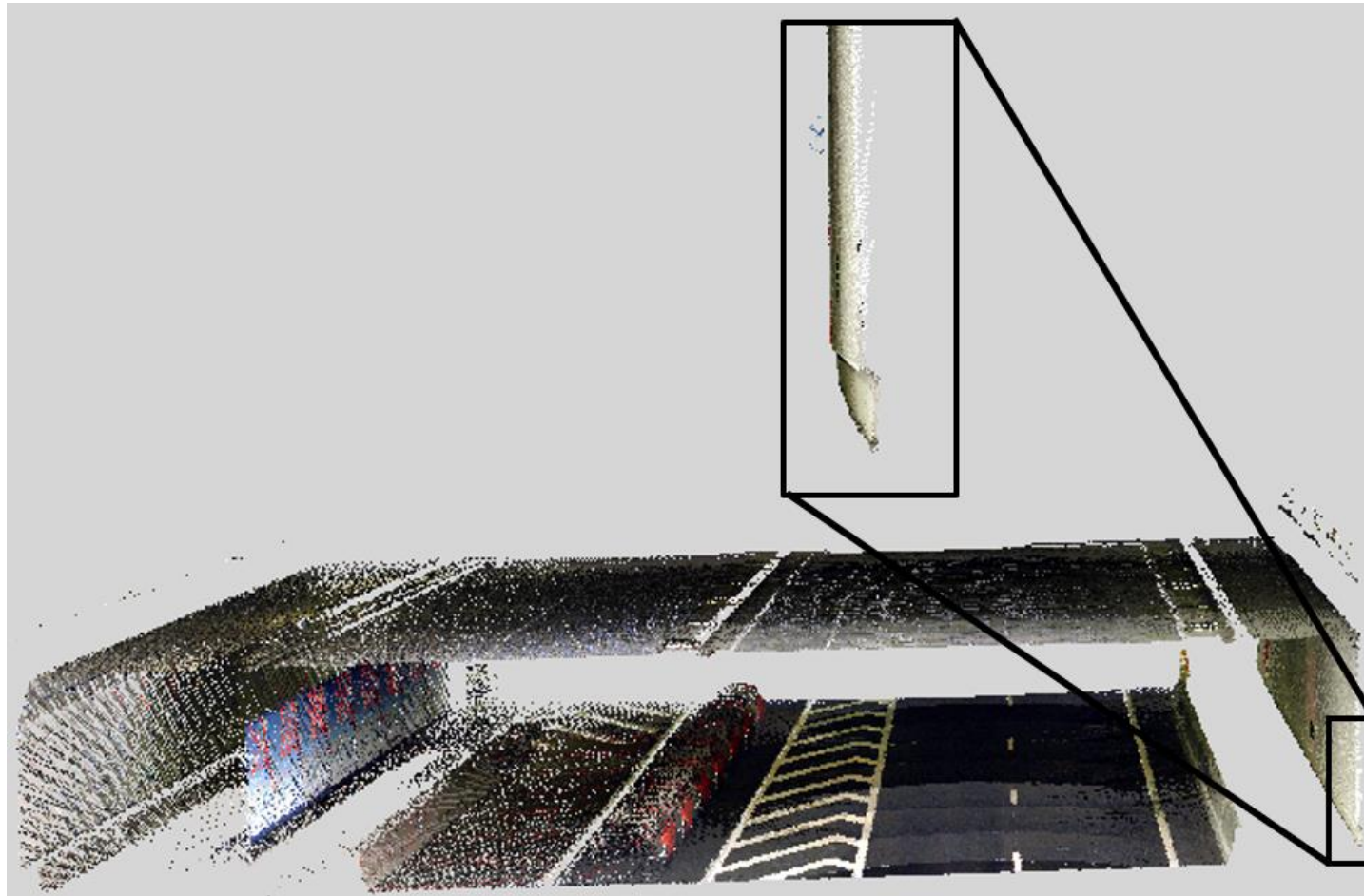


High precision

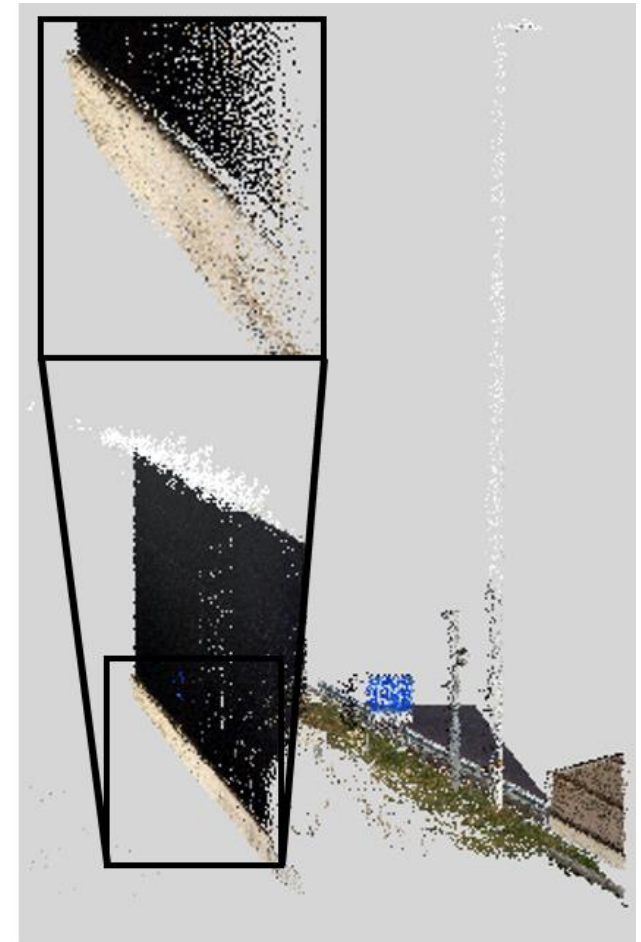
HIGH

# CASE 2: ABSENCE OF OVERLAP, TUNNEL

Point cloud 1



Point cloud 2



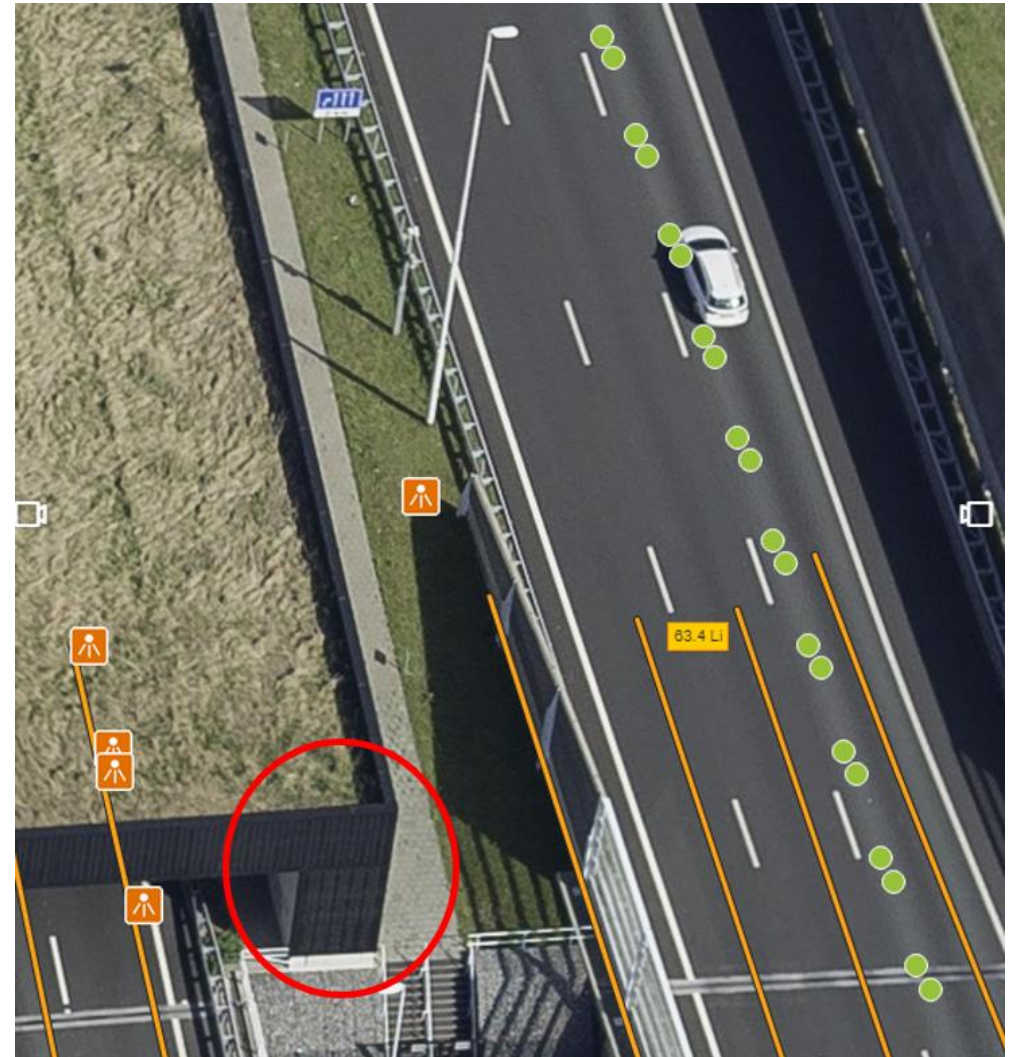
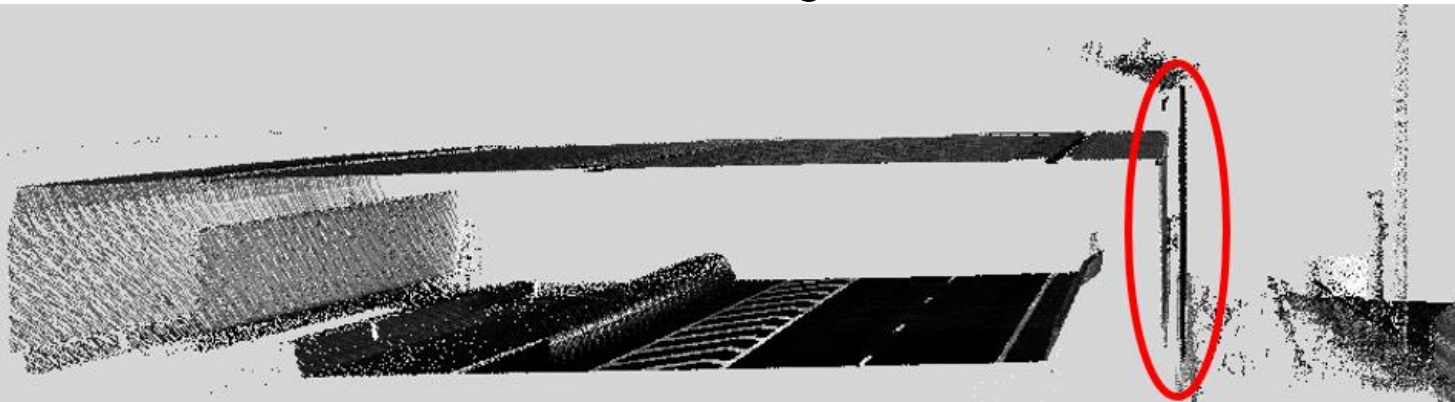


# CASE 2: ABSENCE OF OVERLAP, TUNNEL

Before registration

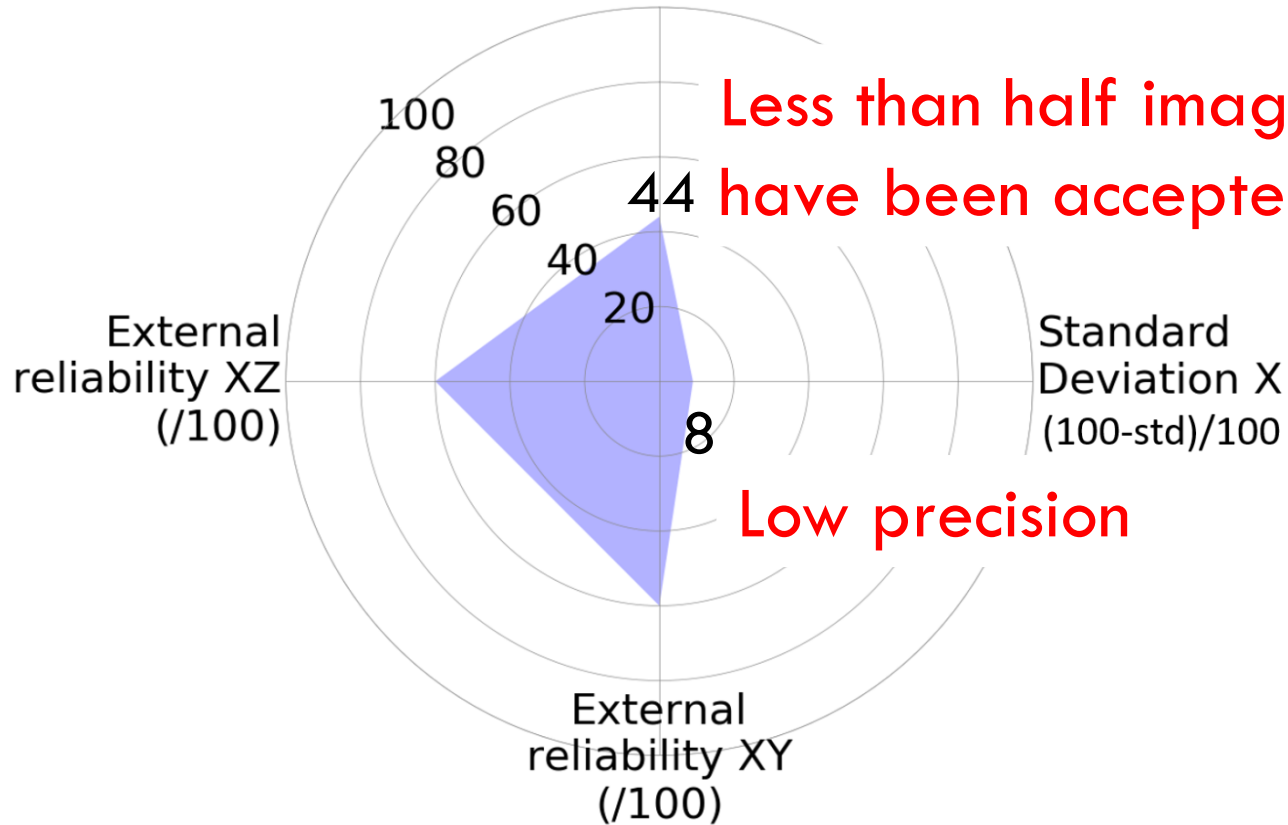


After registration



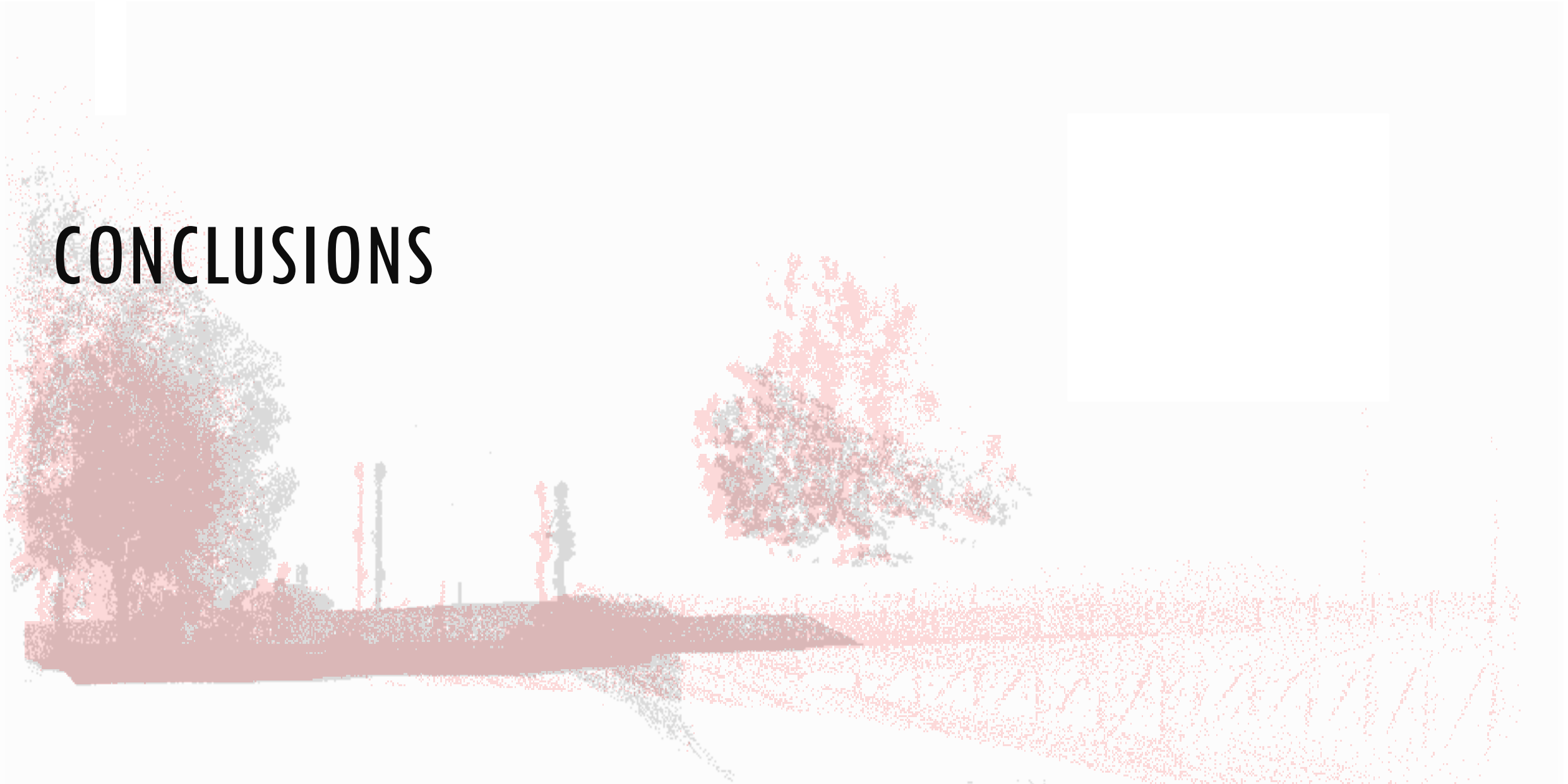
# CASE 2: ABSENCE OF OVERLAP, TUNNEL

X translation = 2.11m  
Internal reliability X (/100)



**IDEAL RESULT**

# CONCLUSIONS



The background features a semi-transparent point cloud of a building on the left and a 3D wireframe model of a building on the right, both rendered in shades of red and grey. The text is overlaid on this background.

**TO WHAT EXTENT IS IT POSSIBLE TO AUTOMATICALLY,  
RELIABLY,  
PRECISELY  
AND EFFICIENTLY**

**ALIGN MOBILE LASER SCANNING DATA  
RELATIVELY  
USING AN IMAGE-BASED TECHNIQUE?**



# CONCLUSIONS – MAIN METHOD

1. It is possible to use an image-based technique to align mobile laser scanning data relatively.
2. More reliable and precise with small grid cells.
3. More efficient with large grid cells.
4. Even if good or poor quality the method informs the user. → Simple template matching technique → registration is more depended on number of pixels → possibility for redundant computations

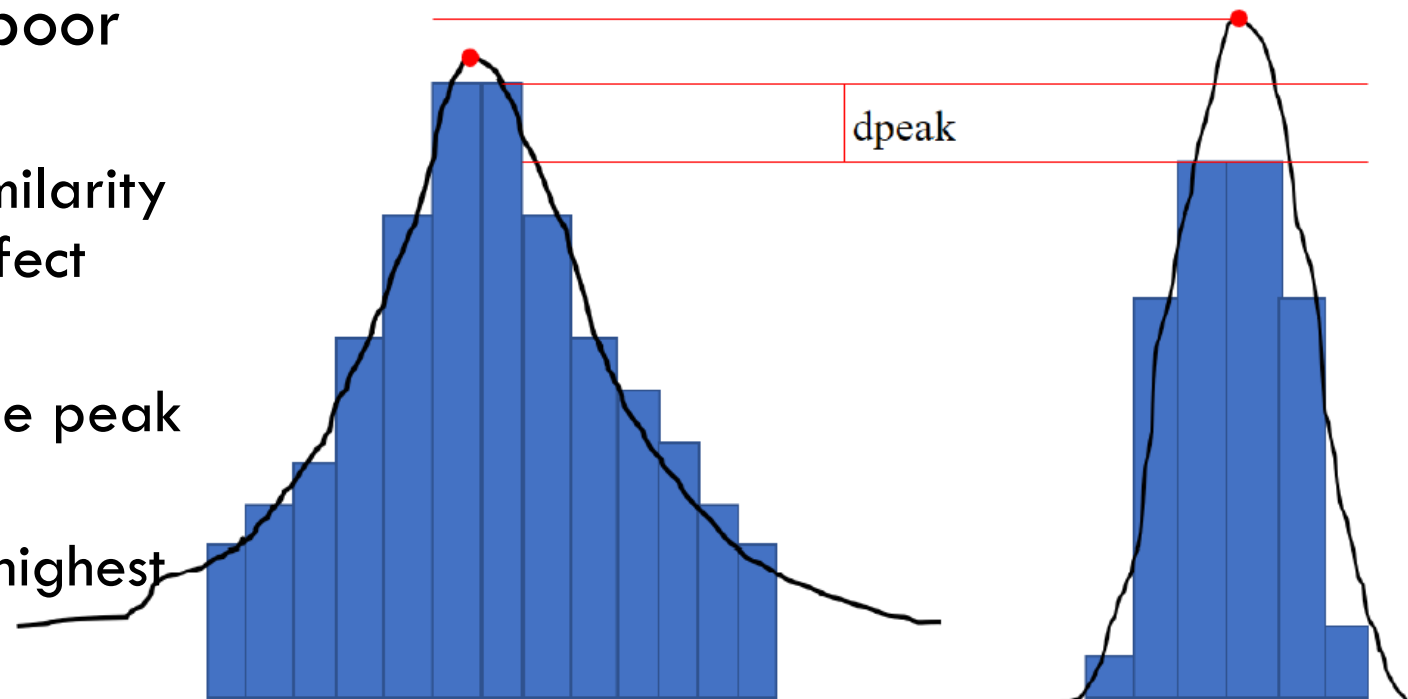
# CONCLUSIONS – MAIN METHOD

It is possible to have good registration results:

- When there is some overlap between two point clouds AND,
- corresponding objects between the point clouds are distinct in pairs of 2D projections.
- Even when the offset is large

# CONCLUSIONS – SUB-PIXEL ACCURACY

1. Shows potential
2. Needs improvement
3. Why there are results with poor quality:
  - Different distributions of the similarity values, not always and not perfect Gaussian
  - Distribution of values around the peak not the same ( $-\sigma_x \neq +\sigma_x$ )
  - Sub-pixel applied only on the highest pixel



# FUTURE WORK



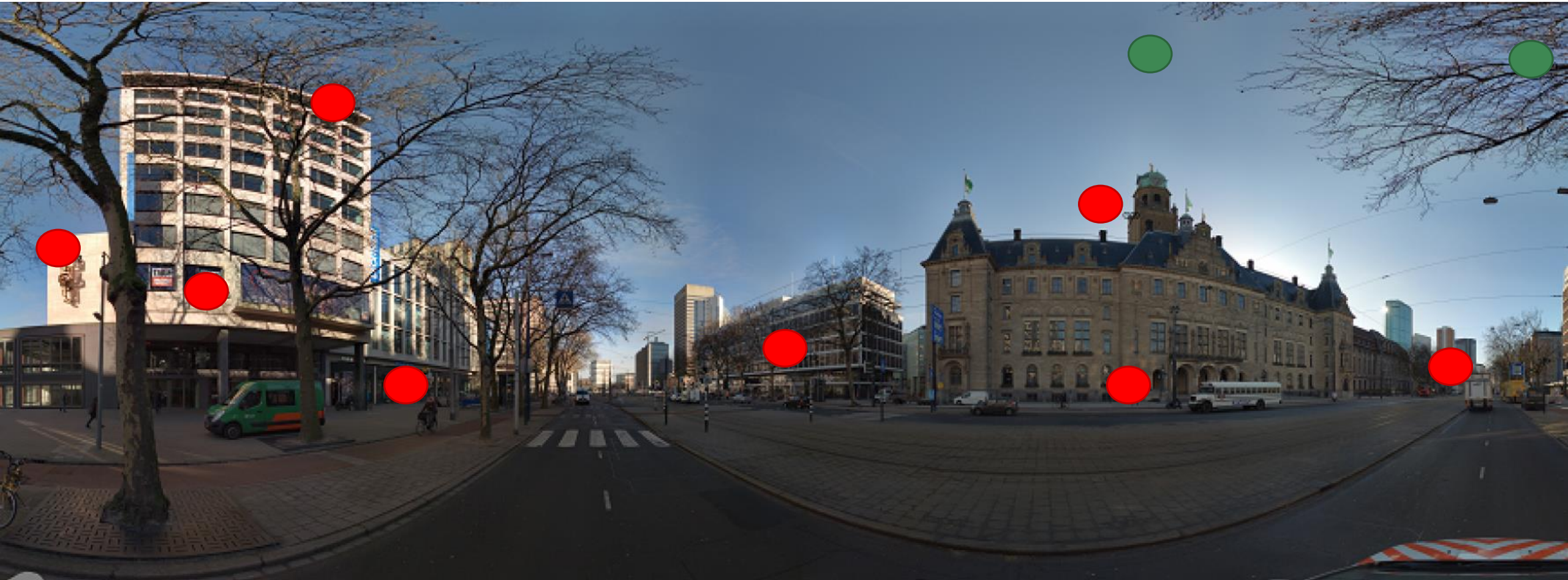


# FUTURE WORK

1. Lidar compensation for vehicle distance
2. Improvement of sub-pixel accuracy method
3. Comparison with ICP

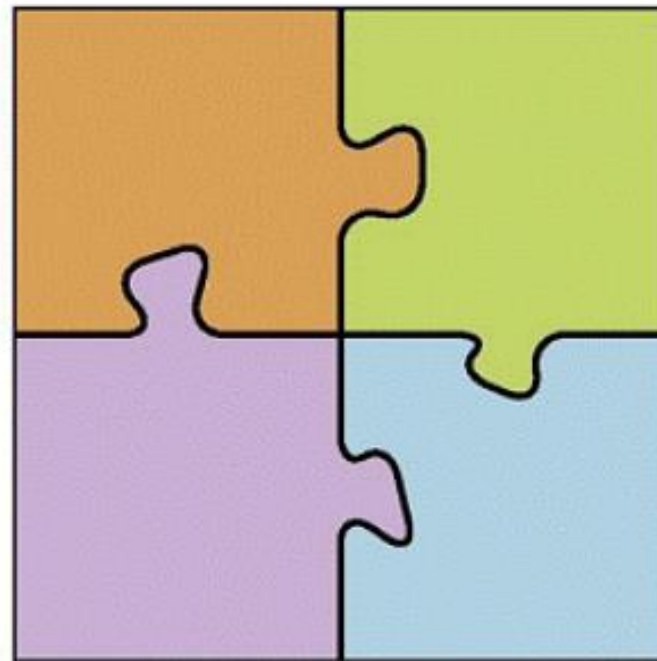
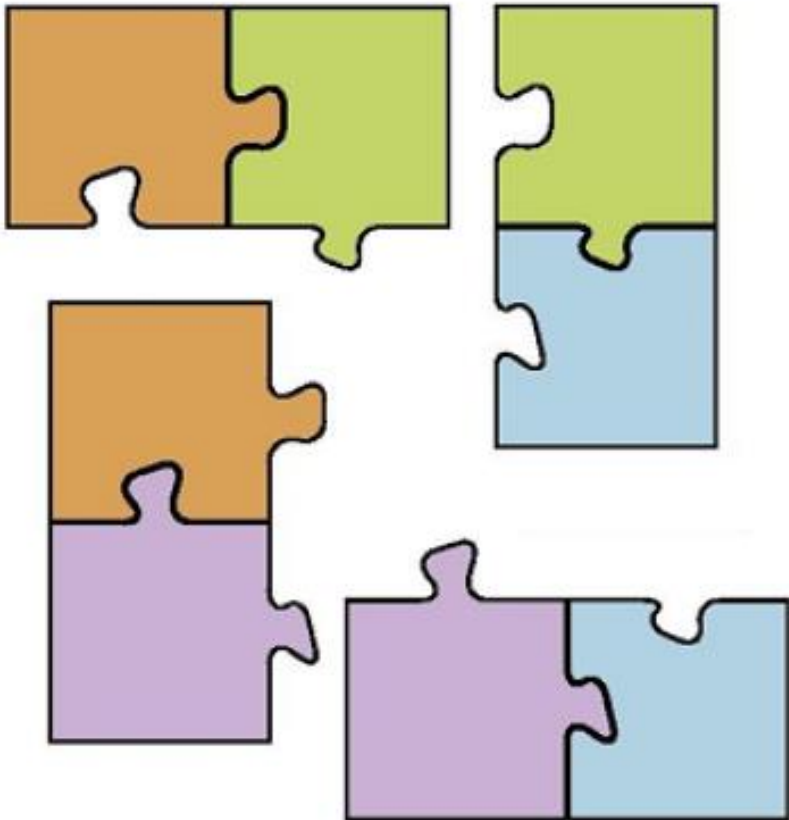
# FUTURE WORK

- Reliability weights based on the accuracy of the point clouds absolute position



# FUTURE WORK

- Global registration



# FUTURE WORK

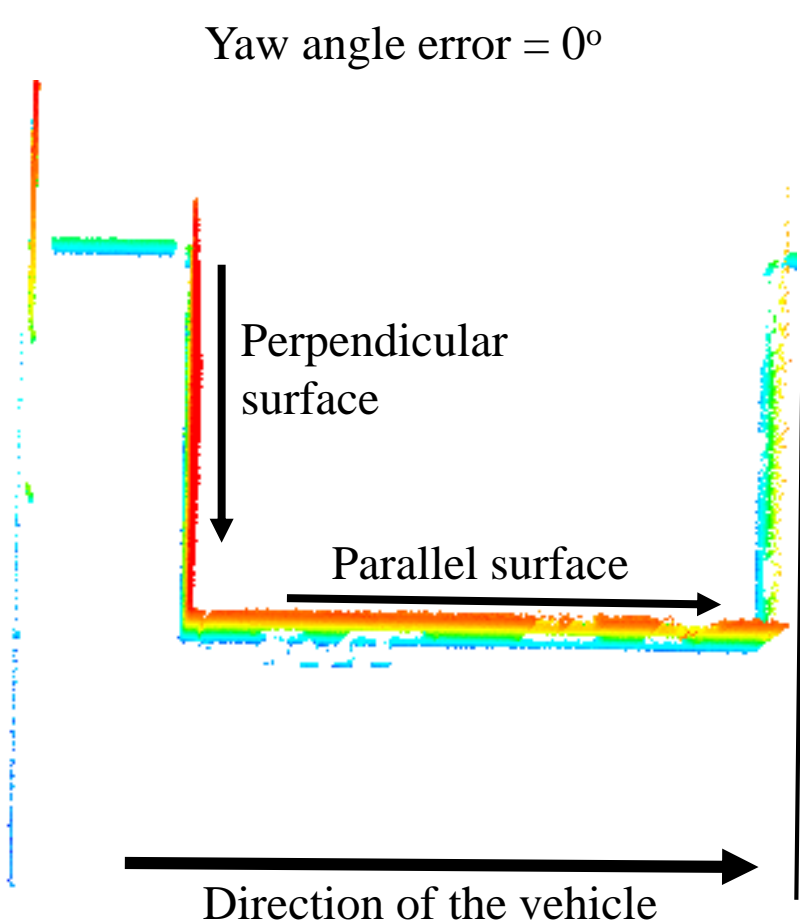
- Moving objects rejection



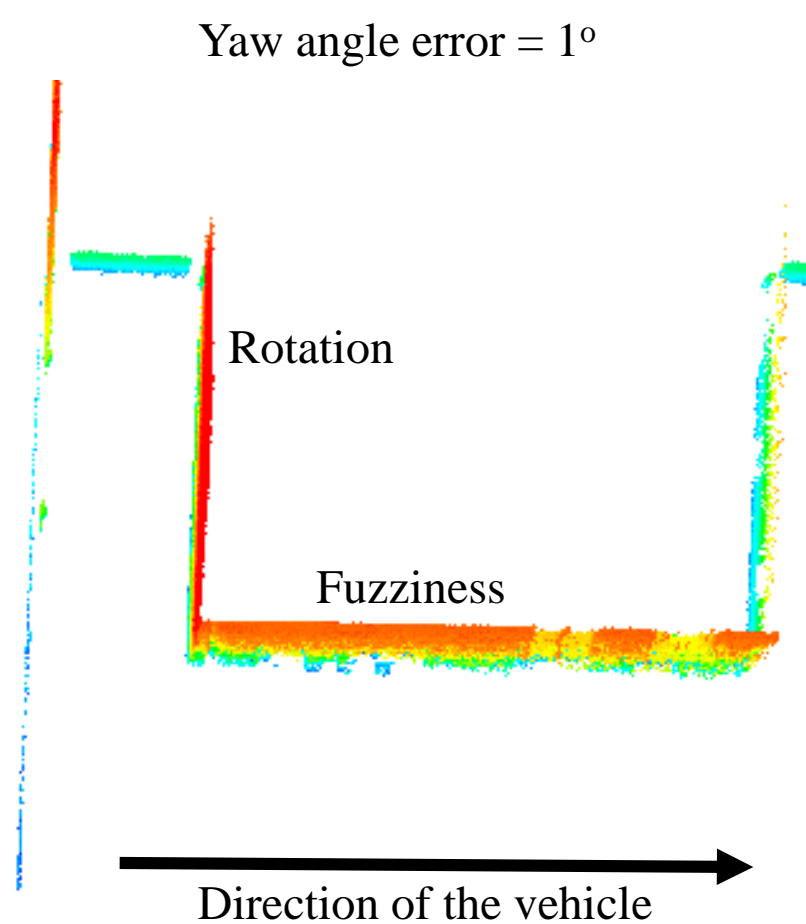
# FUTURE WORK

- Deal with the effects of orientation errors

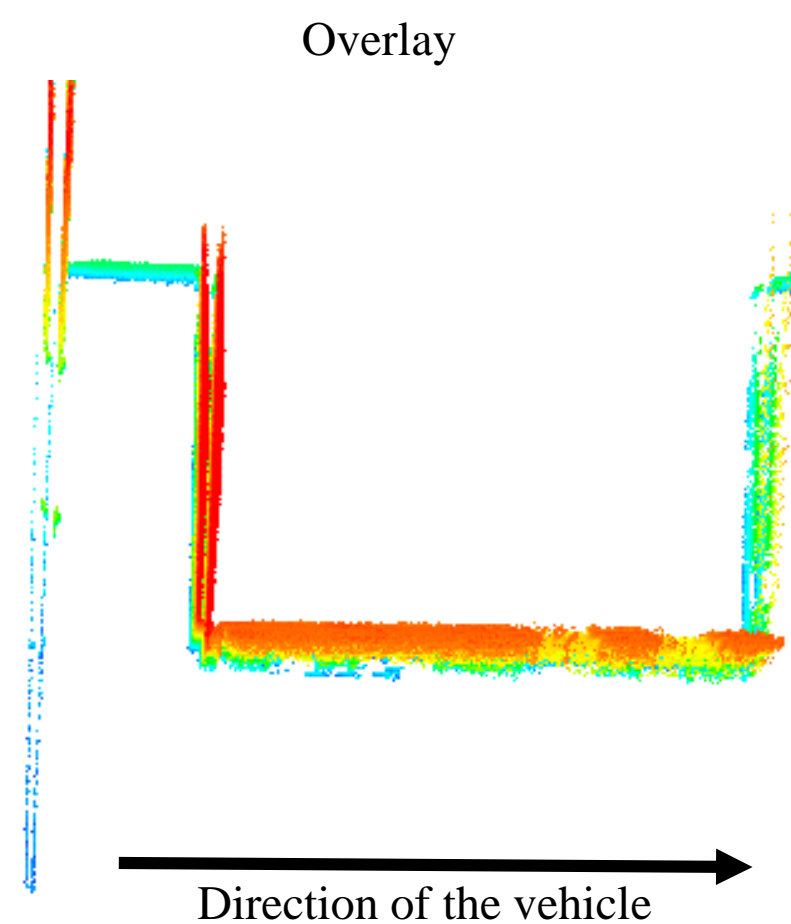
Yaw angle error =  $0^\circ$



Yaw angle error =  $1^\circ$



Overlay





THANK YOU!  
QUESTIONS?

**Image-based method for the pairwise registration of mobile laser scanning point clouds**

**Antria Christodoulou**

Supervisors: Peter van Oosterom, Peter Joosten,  
Berry van Someren, Ravi Y. Peters



Paper: <https://bit.ly/2NIQuUx>