

EXTRACTION OF TRAFFIC SIGNS FROM POINT CLOUDS

Master Thesis 2014/2015

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- **Introduction**
- **Data**
- **Objectives**
- **Related Literature**
- **Methodology**
- **Implementation and results**
- **Discussion**

- A point cloud is a set of data points in some coordinate system.
- Mobile and Terrestrial laser scanners are widely used for scanning of urban areas.
- Problems of point cloud processing: discrimination and classification of objects from point clouds; machine learning.
- Challenge: extraction and classification of traffic signs by their types.

Area of investigation

- Recorded in 2009 and 2015 around HFT Stuttgart.
- Dozens of traffic signs.
- Lamp posts; poles; traffic lights; trees; wires etc.

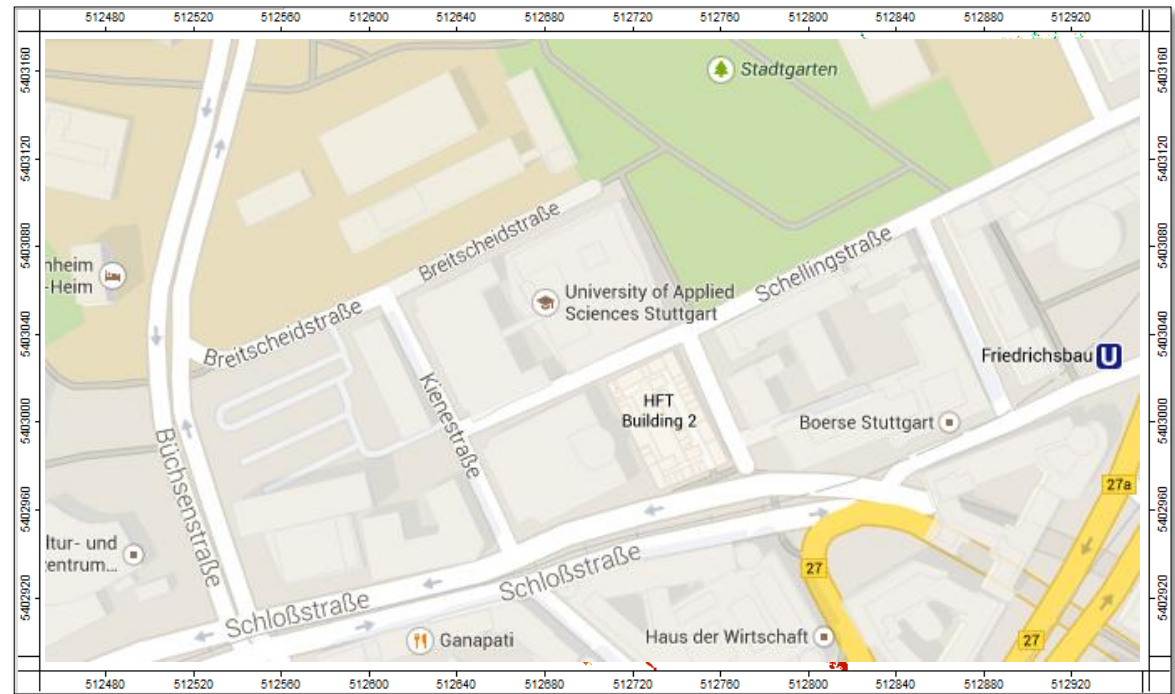


Figure 3. Area of investigation (from: Google Maps).

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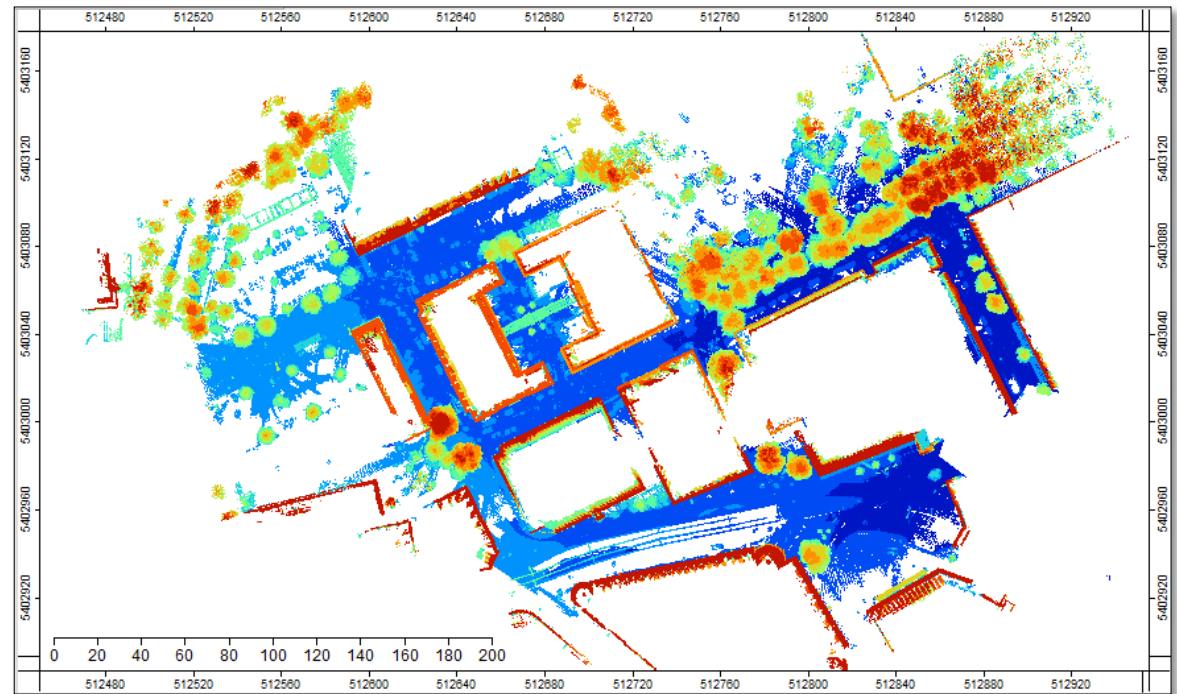
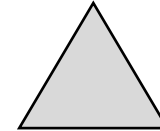


Figure 3. Area of investigation.

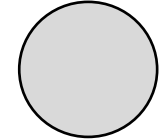
- 1. Eliminate redundant data**
- 2. Extract pole-like objects**
- 3. Extract traffic signs**
- 4. Extract traffic sign's plates**
- 5. Develop shape descriptors**

The main goal: Signs classification

1. Warning signs



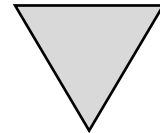
2. Regulatory signs



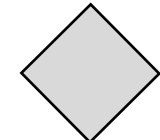
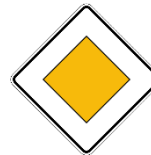
3. Information signs



4. Vorfahrt gewähren



5. Vorfahrtstraße



Gross, H., Thoennessen, U., 2006.

- Investigated eigenvalues $\lambda_1, \lambda_2, \lambda_3$.
- Defined eigenvalues combinations for some typical situations.
- Successfully extracted lines, edges and planes.

Eigenvalues and Eigenvectors

Moments:

$$\tilde{m}_{ijk} = \frac{\sum_{l=1}^N (x_l - \bar{x})^i (y_l - \bar{y})^j (z_l - \bar{z})^k}{R^{i+j+k} N}$$

Covariance matrix:

$$M = \begin{pmatrix} \tilde{m}_{200} & \tilde{m}_{110} & \tilde{m}_{101} \\ \tilde{m}_{110} & \tilde{m}_{020} & \tilde{m}_{011} \\ \tilde{m}_{101} & \tilde{m}_{011} & \tilde{m}_{002} \end{pmatrix}$$

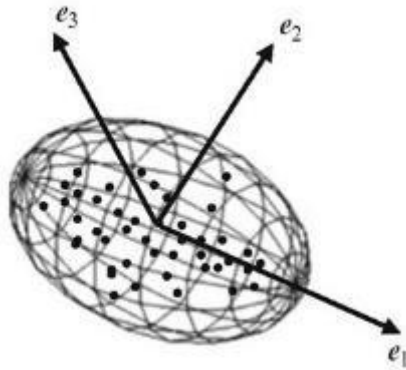


Figure 4. Eigenvectors and eigenvalues (source: www.what-when-how.com/advanced-methods-in-computer-graphics/collision-detection-advanced-methods-in-computer-graphics-part-2).

Kazhdan M., Funkhouser T., Rusinkiewicz S., 2003.

- 3D shape matching.
- Rotation invariant descriptors.
- Spherical Extent Function.

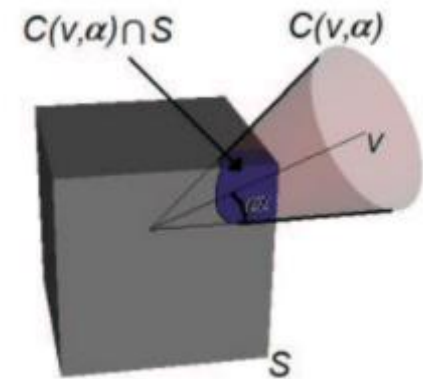
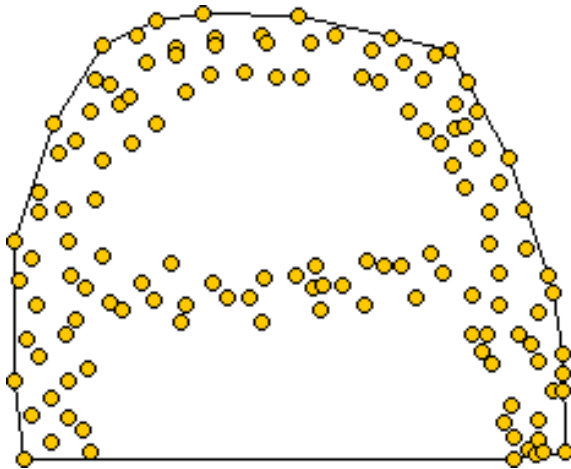


Figure 5. Visualization of the Radial Variance for a cube. (Source: Kazhdan M., Funkhouser T., Rusinkiewicz S., 2003. Rotation invariant spherical harmonic representation of 3D shape descriptors.).

Convex Hull



Alpha Shape

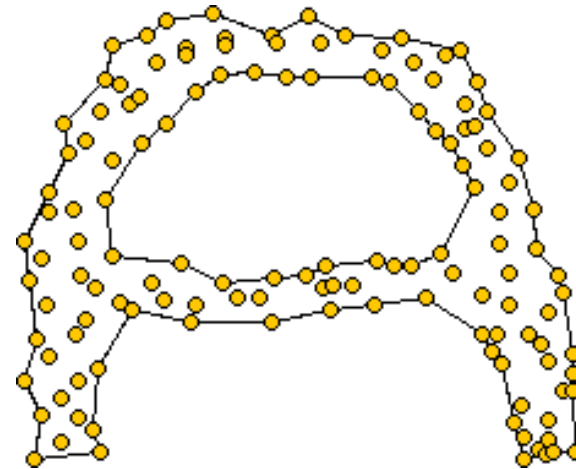


Figure 6. Convex Hull and Alpha Shape
(source: <http://cgm.cs.mcgill.ca/~godfried/teaching/projects97/belair/alpha.html>).

Alpha Shapes algorithm

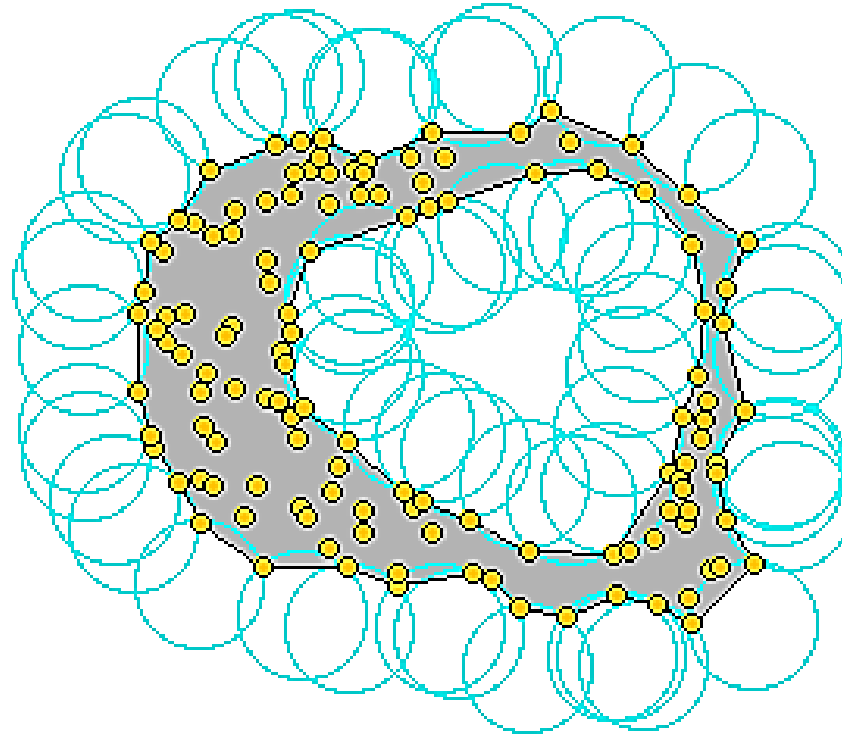


Figure 7. *Alpha shapes extracting principle.*

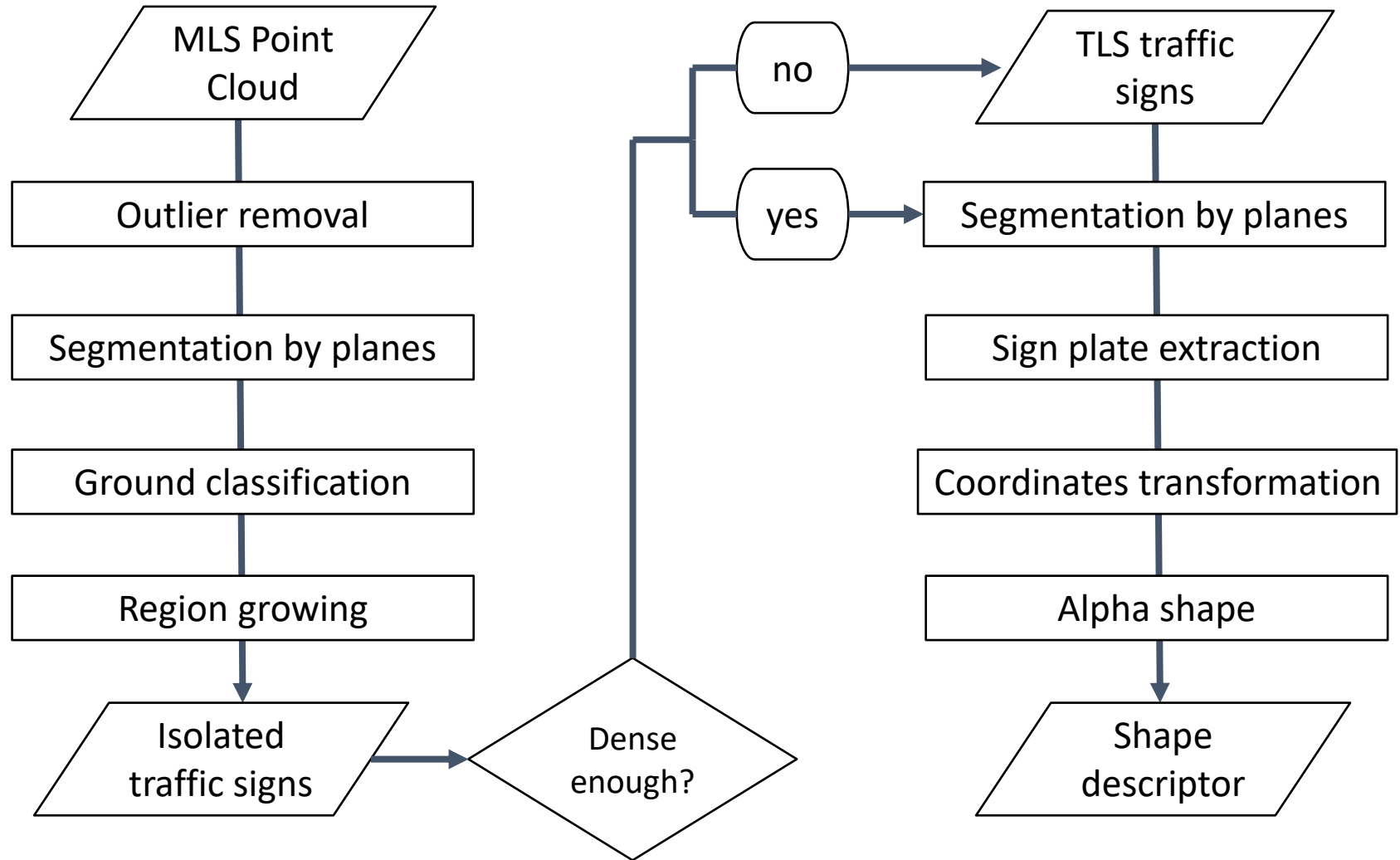
(Source: Shen Wei, 2008. *Building Boundary Extraction Based on LiDAR Point Clouds Data.*)

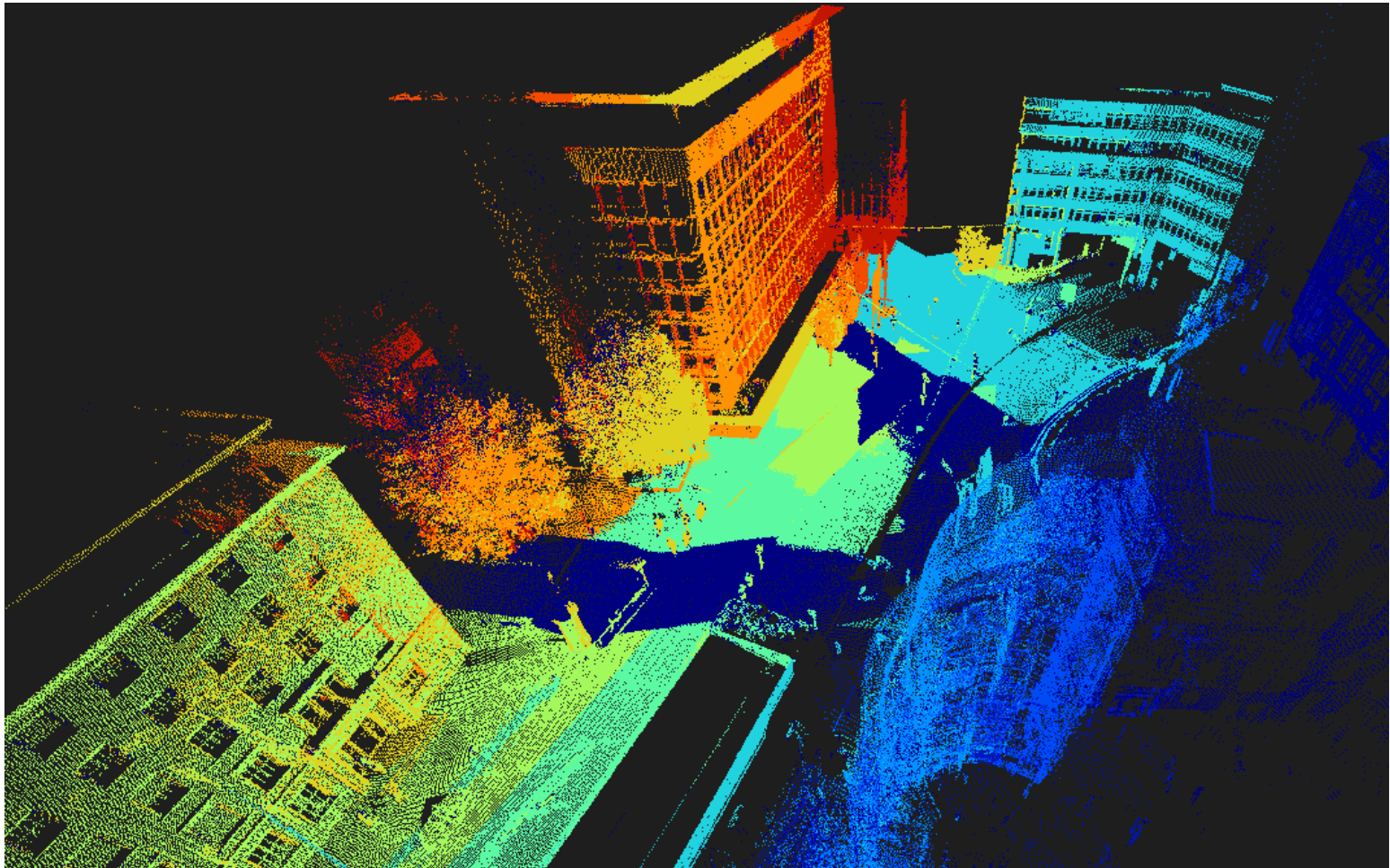
- **Remove ground points**
- **Extract pole-like objects**

MLS

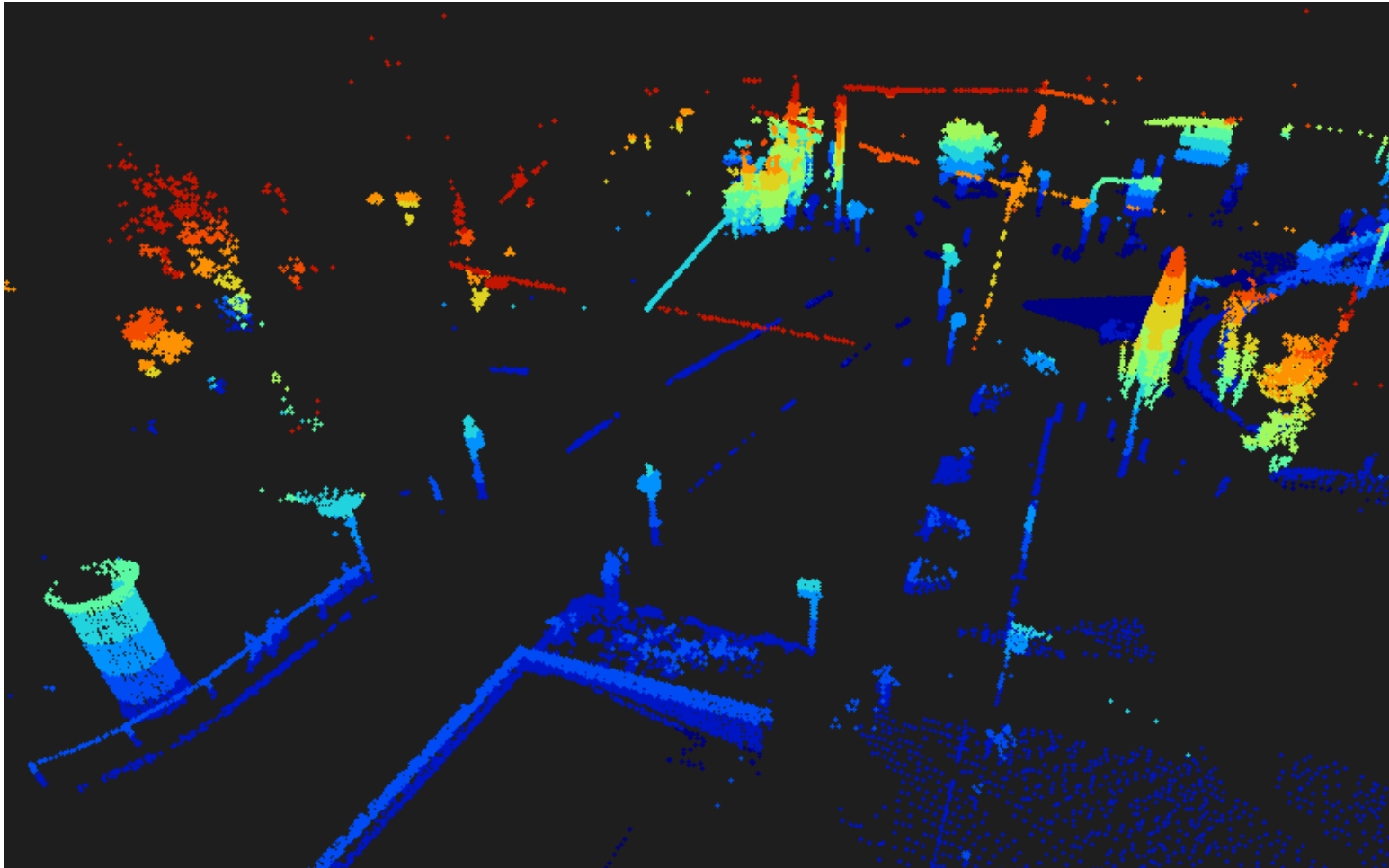
- **Separate plates from poles**
- **Purge redundant points**
- **Develop Shape Descriptor**

TLS





Segmentation by planes

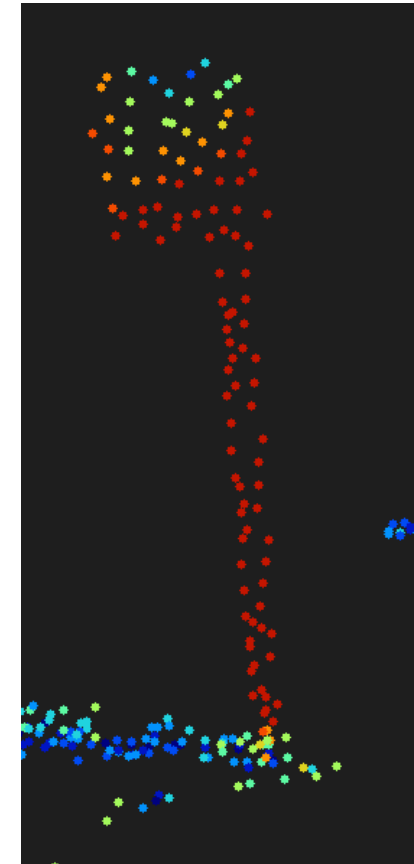


Z Variance < 4m

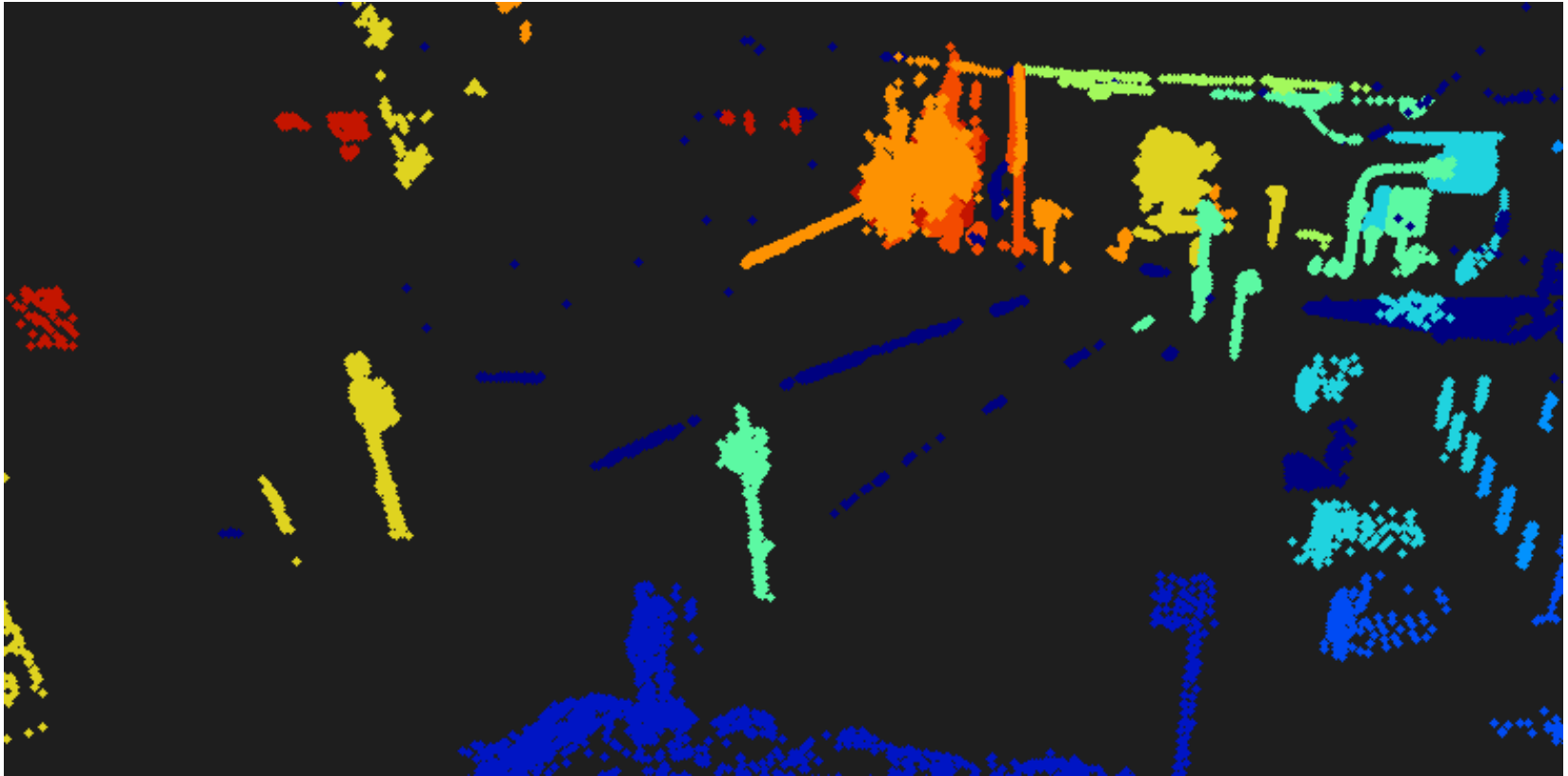
Region Growing

- Seed Point features:
High Linearity + Vertical Angle

Linearity: 80 ... 100 %
Vertical Angle: 80° ... 90°
Seed Point = Lin + VA > 160
- Search Radius:
1.8 ... 2.0 m



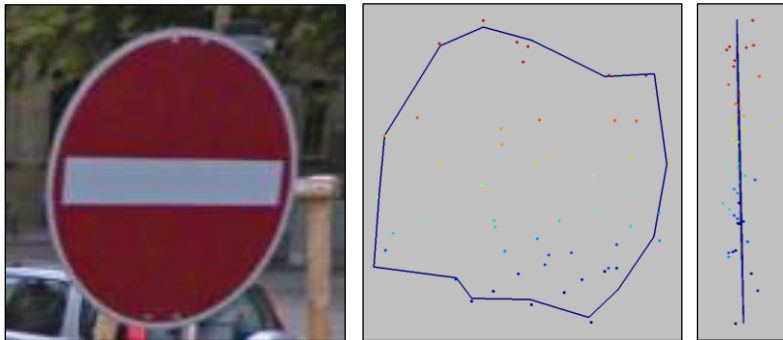
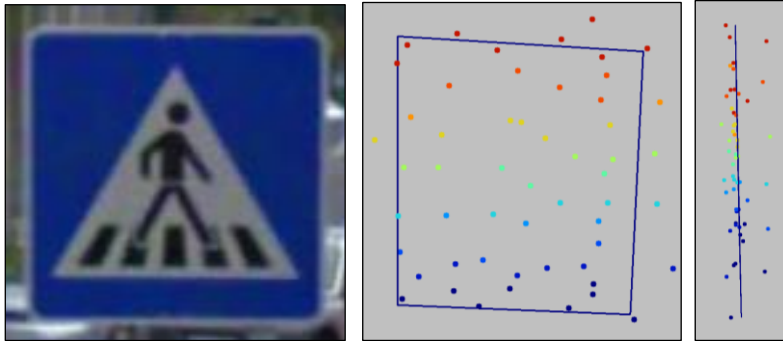
**Seed Points: vertical
elongated objects**

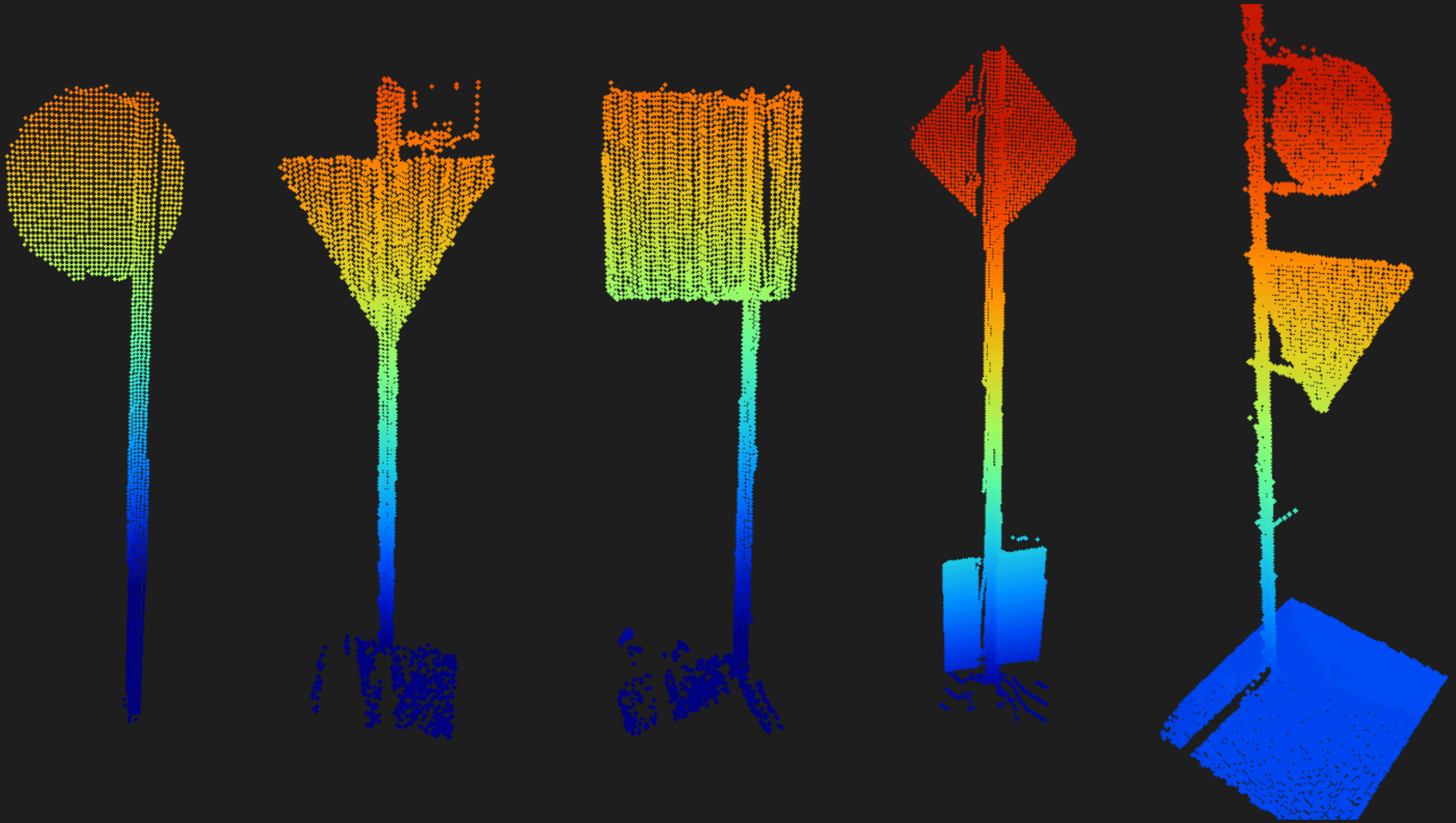


Detection of pole-like objects

Traffic signs' shapes from MLS Point Cloud

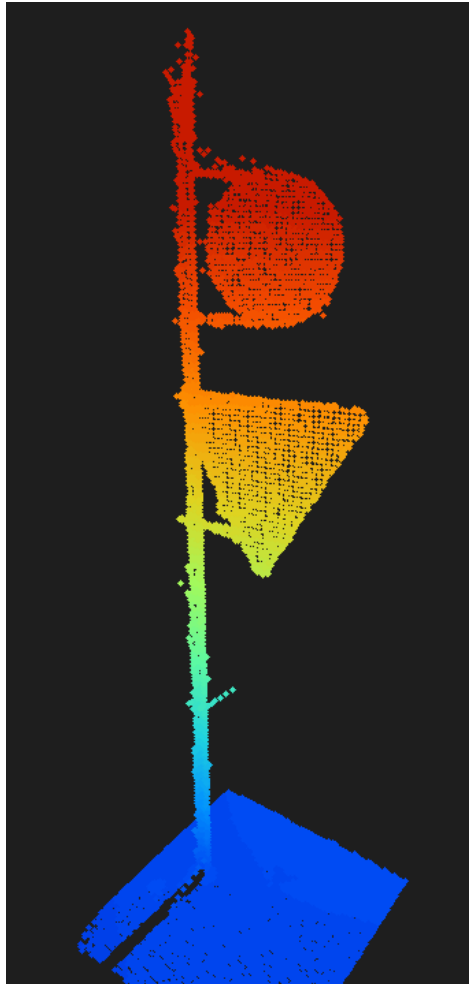
- Not sufficient point density !



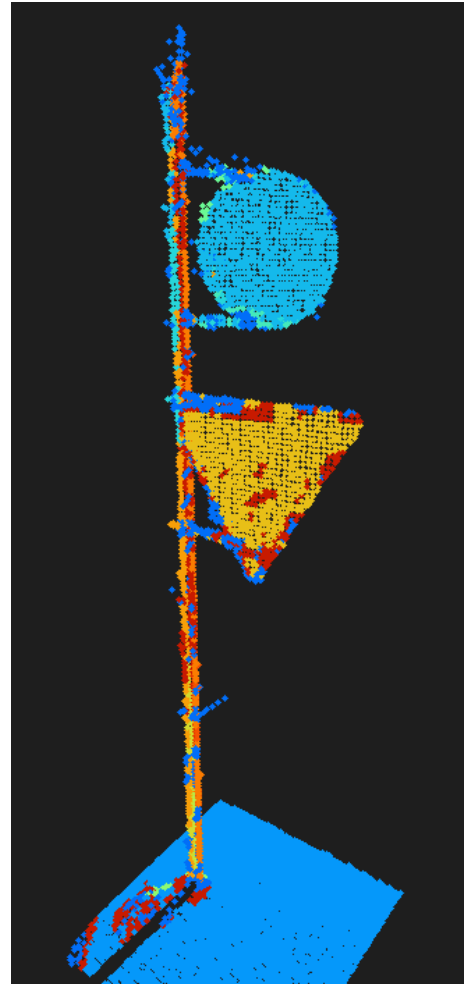


Signs taken with Terrestrial Laser Scanner (TLS)

Segmentation by planes

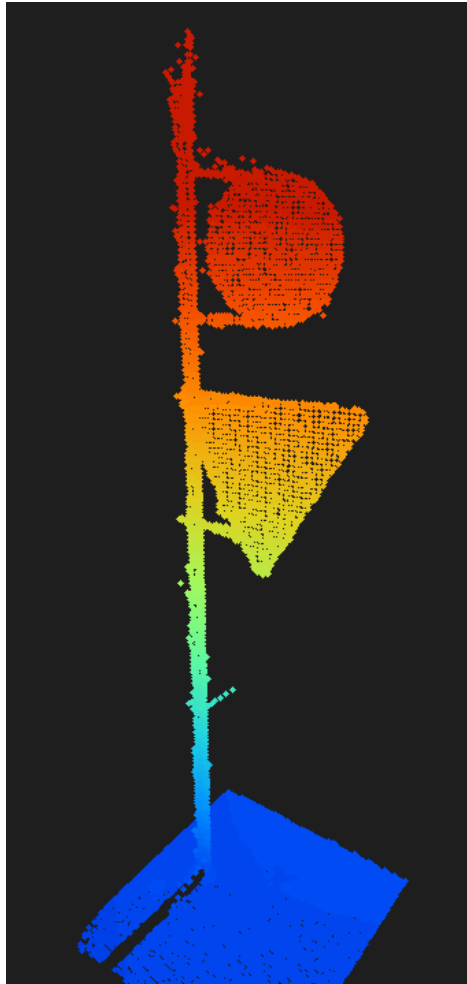


Z - coded

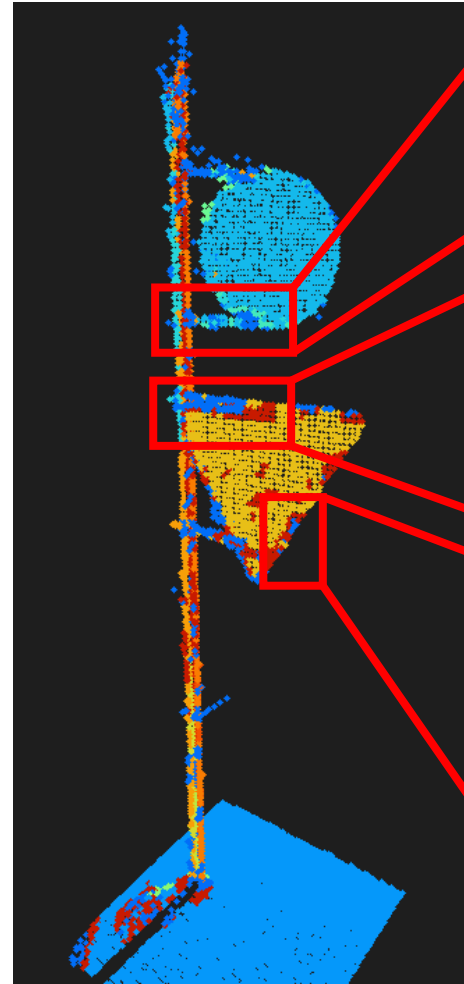


Segment ID

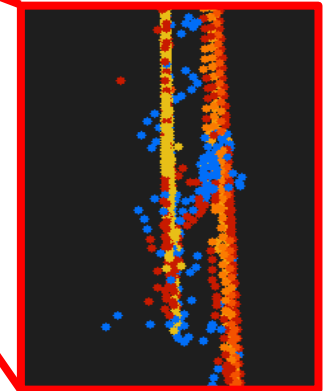
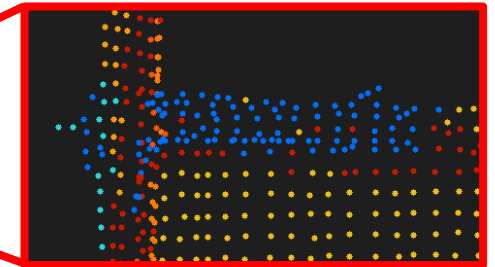
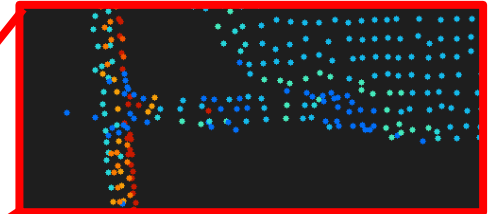
Segmentation by planes



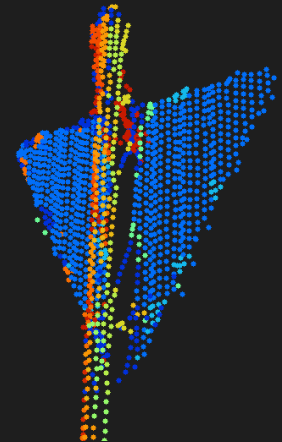
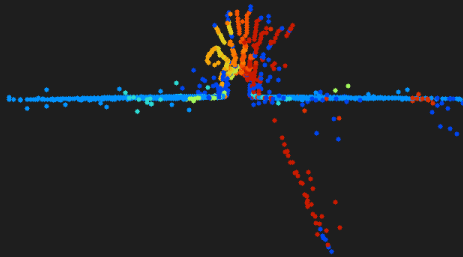
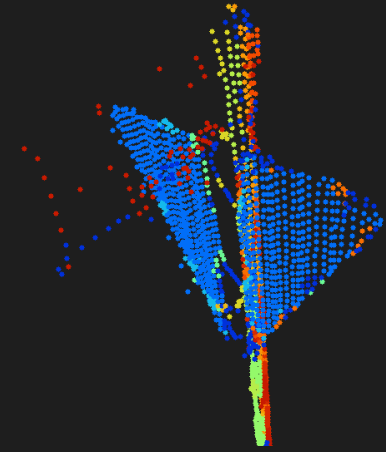
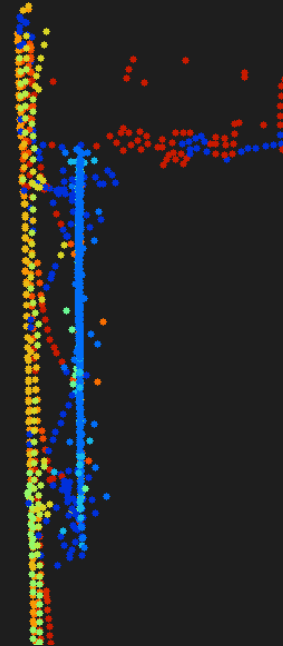
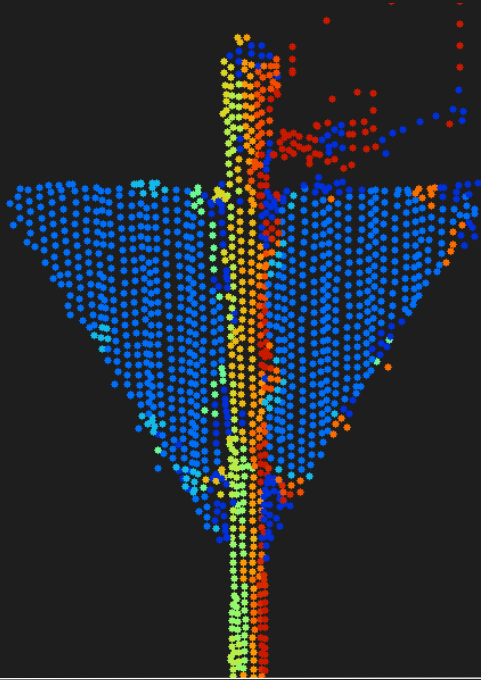
Z - coded



Segment ID

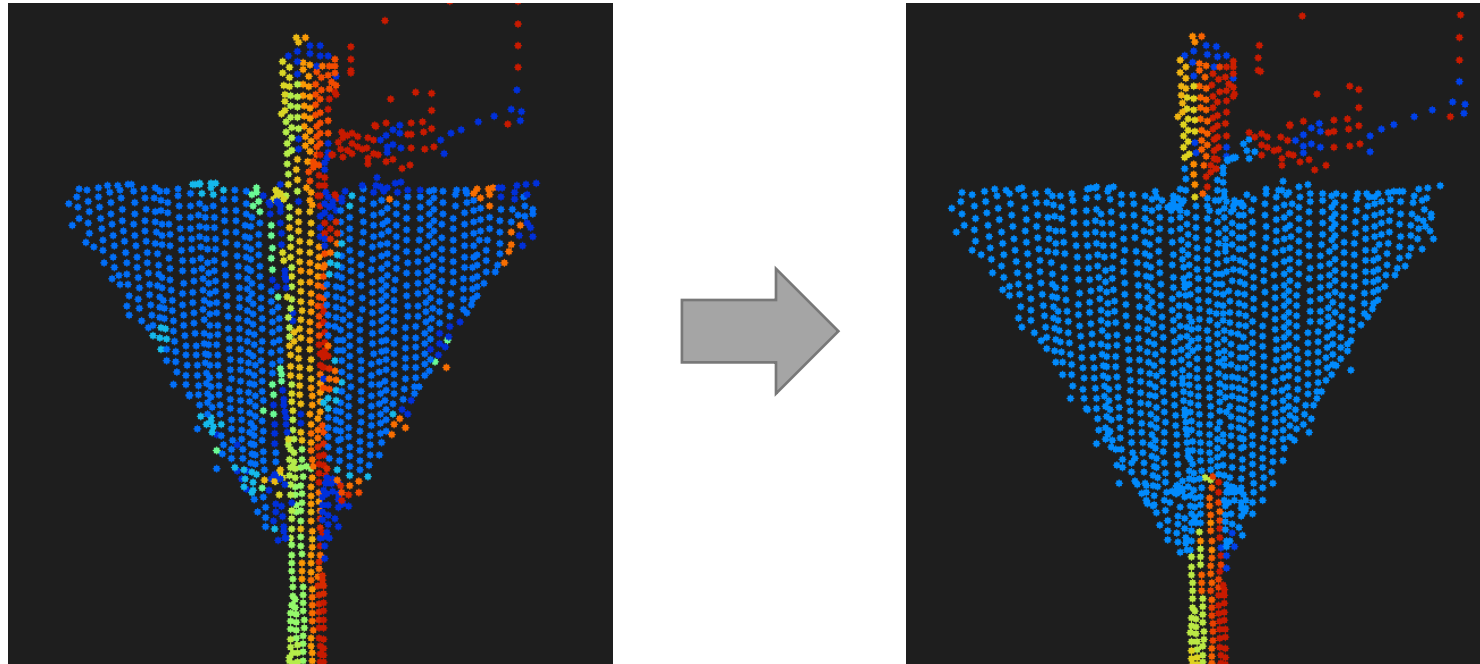


Implementation and results



Segment ID

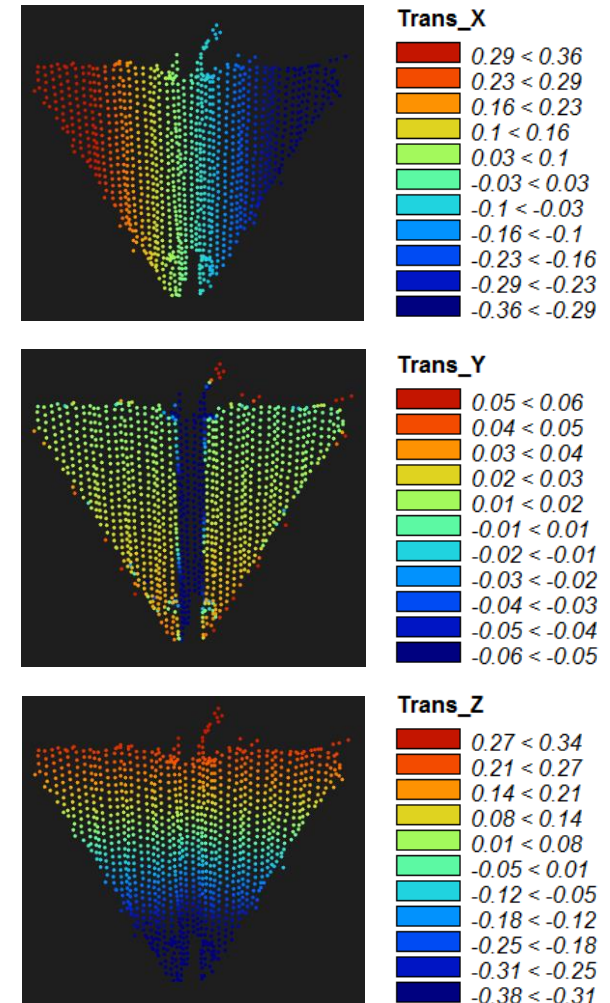
Assigning adjacent points to the segment



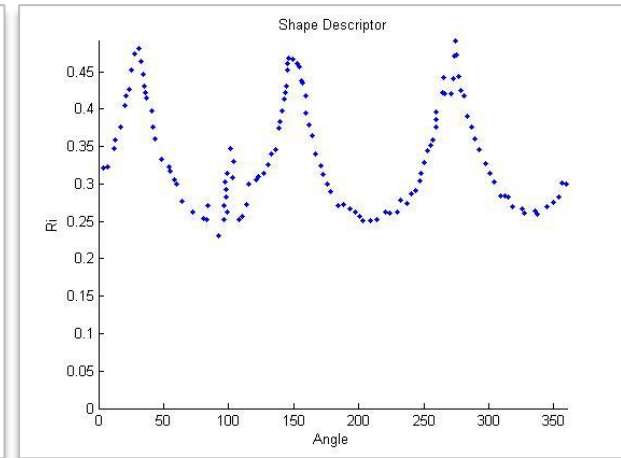
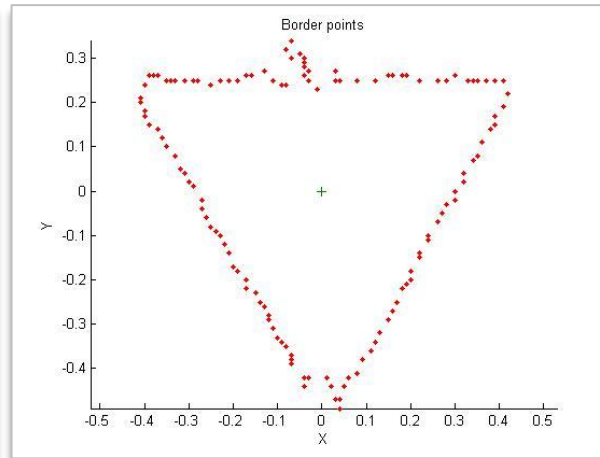
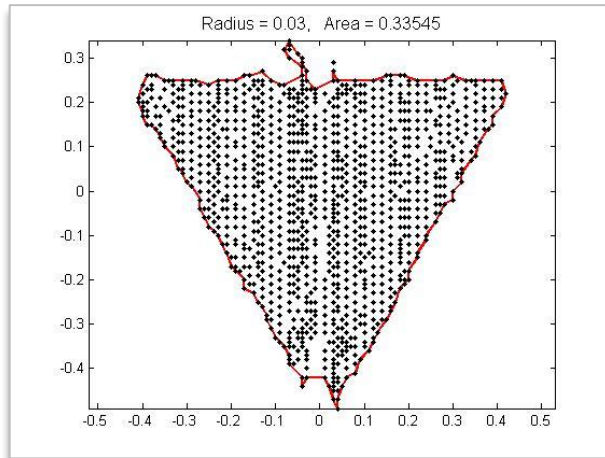
- Search in neighborhood around segment points
- Within radius R take the points into the segment
- Assign segment features to the captured points

Local coordinates transformation

- Extract the class
- Considering [X Y Z] and [Nx Ny Nz] transform to [Trans_X Trans_Y Trans_Z] with a local coordinate system
- Read [Trans_X Trans_Z]

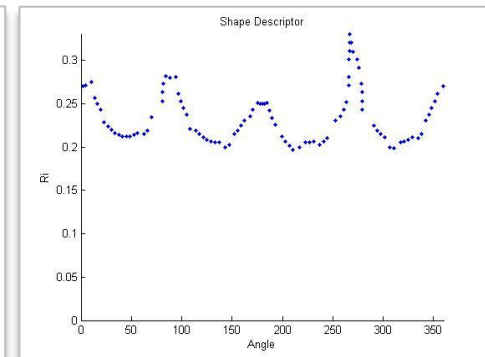
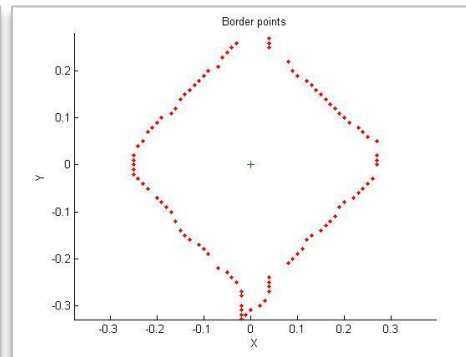
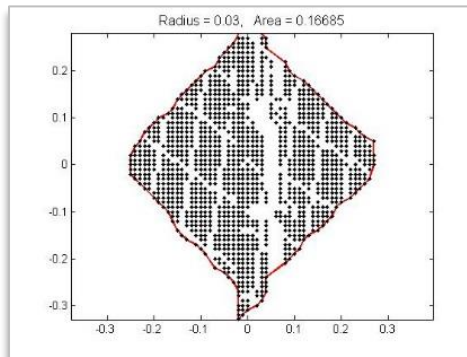
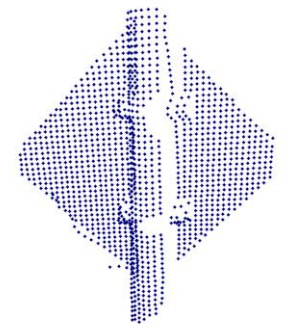
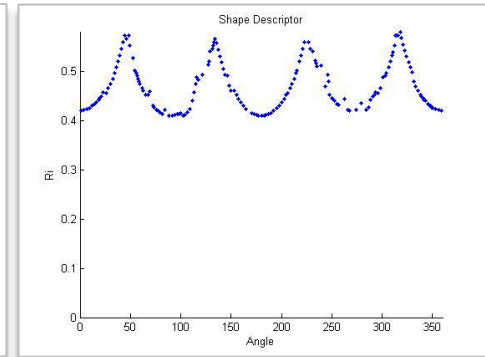
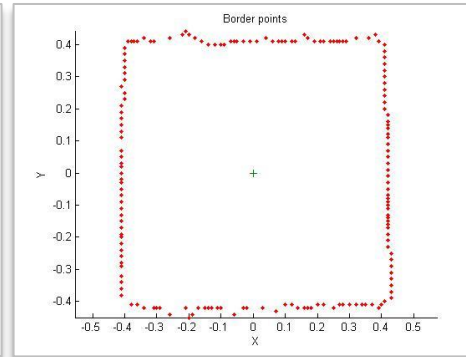
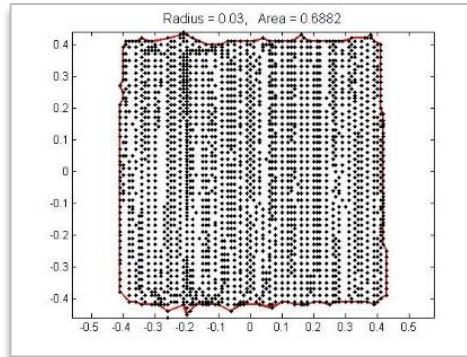
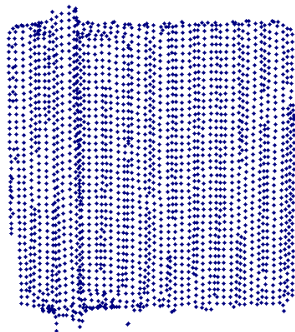
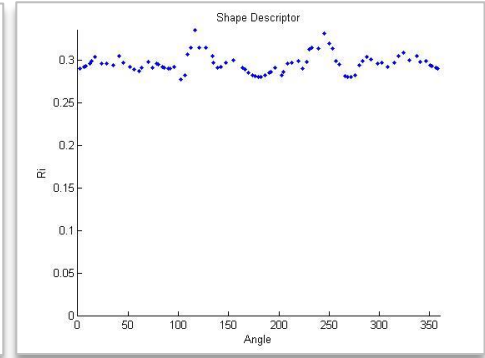
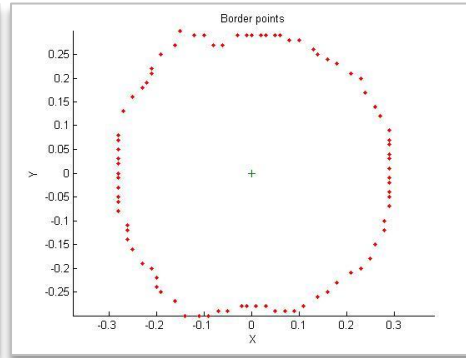
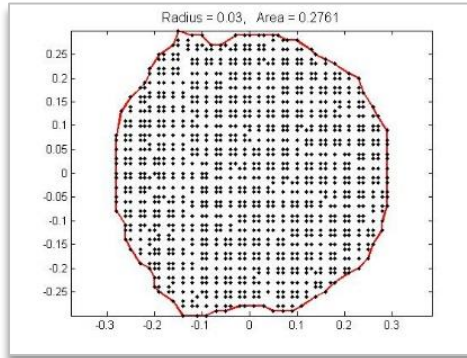
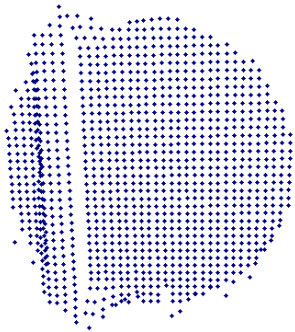


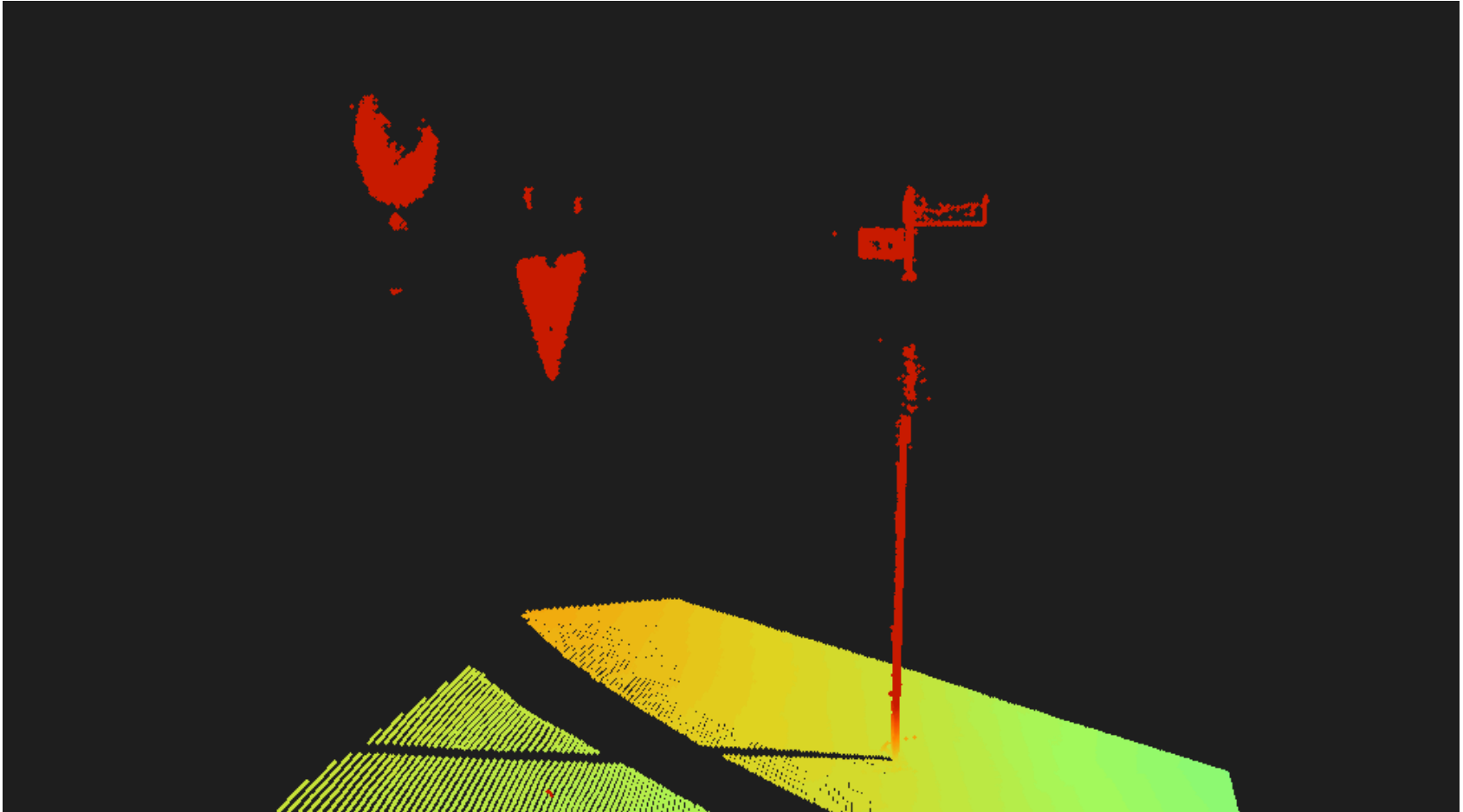
Alpha Shape and Shape Descriptor



- Alpha shape
- Extract border points
- Find center of mass (x_0, y_0)
- Find spherical coordinates of P_i
- Shape descriptor (r_i, α_i)

Implementation and results





Range errors in capturing of a traffic sign

Capturing of retro-reflective materials

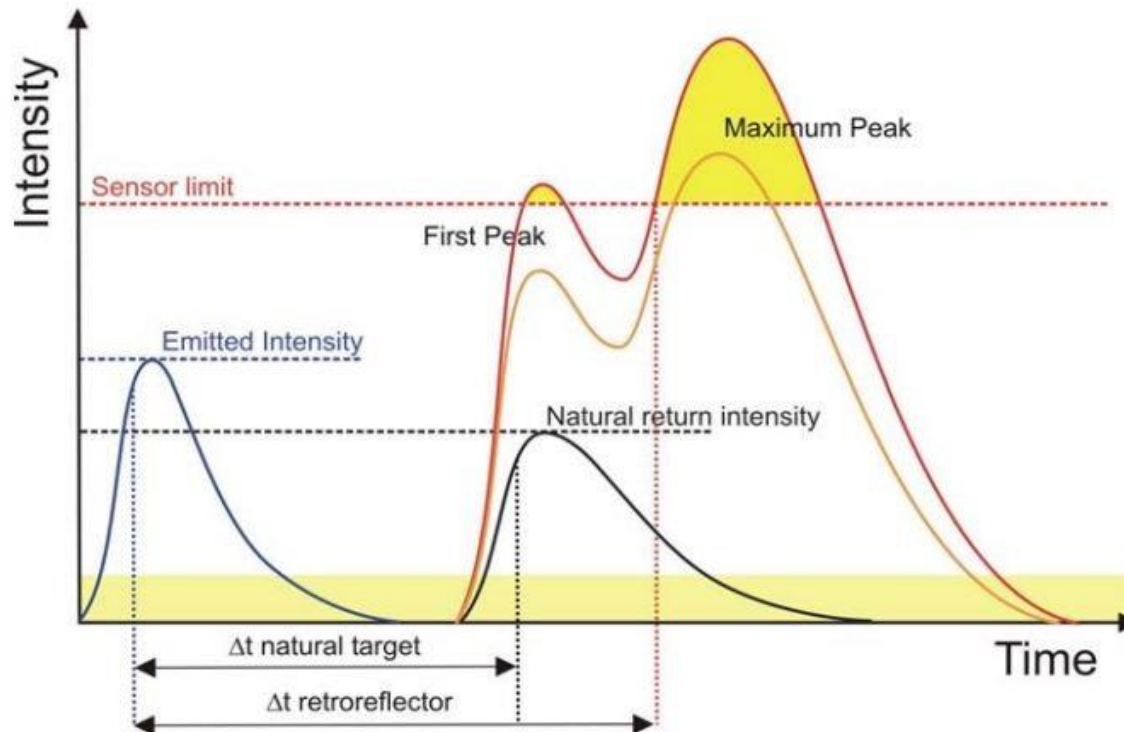


Figure 4. Hypothesis for a modified pulse waveform reflected from retro-reflective materials. (Source: Pesci, A.; Teza, G., 2008. Terrestrial laser scanner and retro-reflective targets: An experiment for anomalous effects investigation.)

Shape Descriptor for complex signs

