



Detection of Temporary Objects in Mobile Lidar Point Clouds

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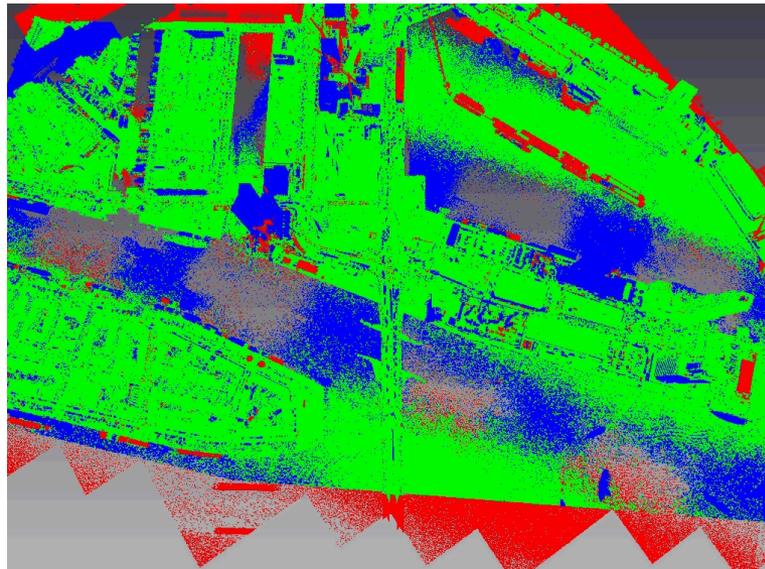
Kourosh Khoshelham

Sander Oude Elberink



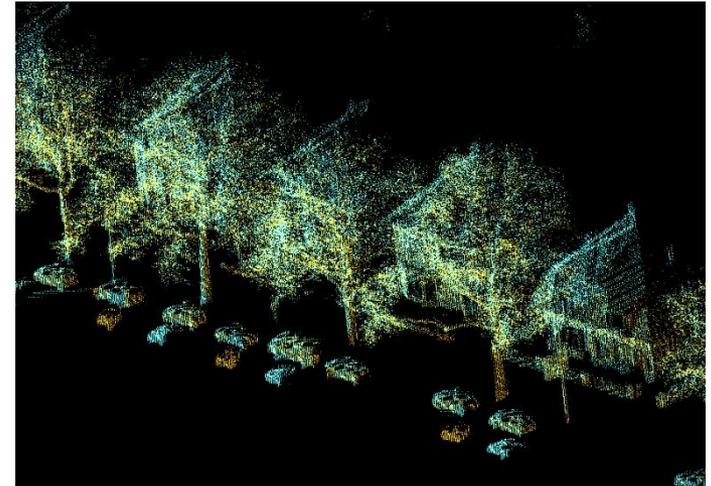
BACKGROUND

- Mobile laser scanning provides an accurate recording of road environments useful for many applications;
- Usually only permanent objects are of interest;
- Temporary objects can hamper the analysis of permanent ones;
- Example: change detection using multi-epoch data where temporary objects appear as (false) change signals.



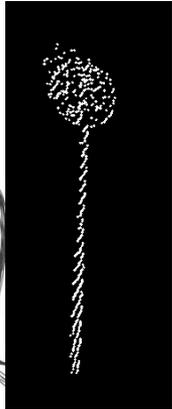
APPROACH

- Temporary objects: $\left\{ \begin{array}{l} \text{Static} \\ \text{Moving} \end{array} \right.$
- Static temporary objects:
→ segmentation + classification based on shape features
- Moving temporary objects:
→ segmentation + closest point analysis in two sensor data

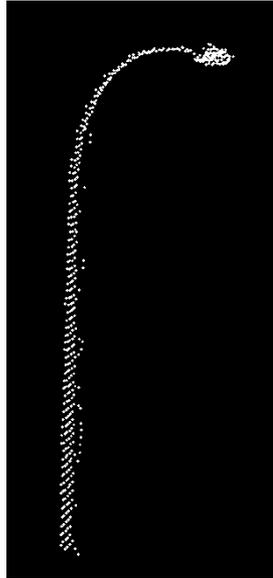


STATIC TEMPORARY OBJECTS

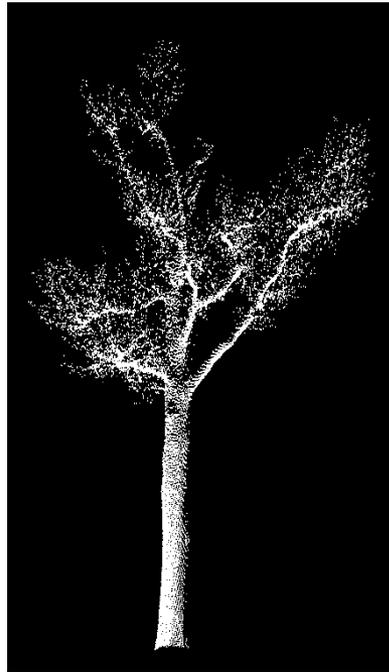
- Classification based on shape features:



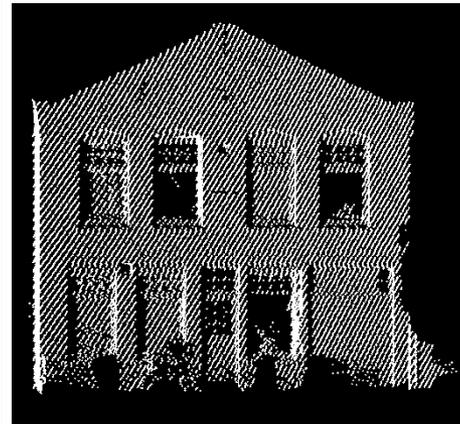
Linear
& short



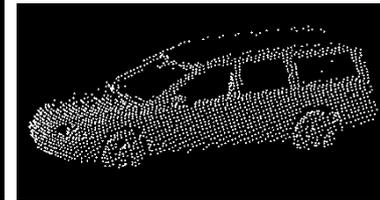
Linear
& tall



Volumetric
& tall



Planar



Volumetric
& short

MOVING TEMPORARY OBJECTS

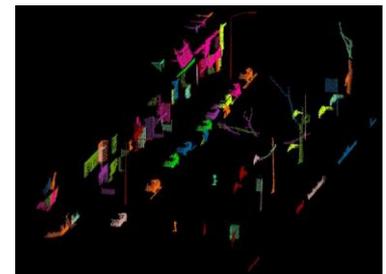
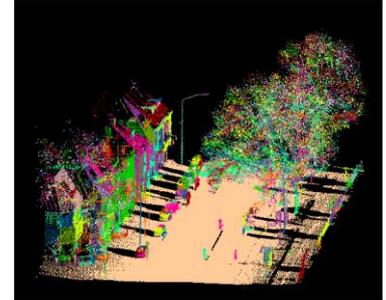
- Detection based on closest point analysis with two sensor data





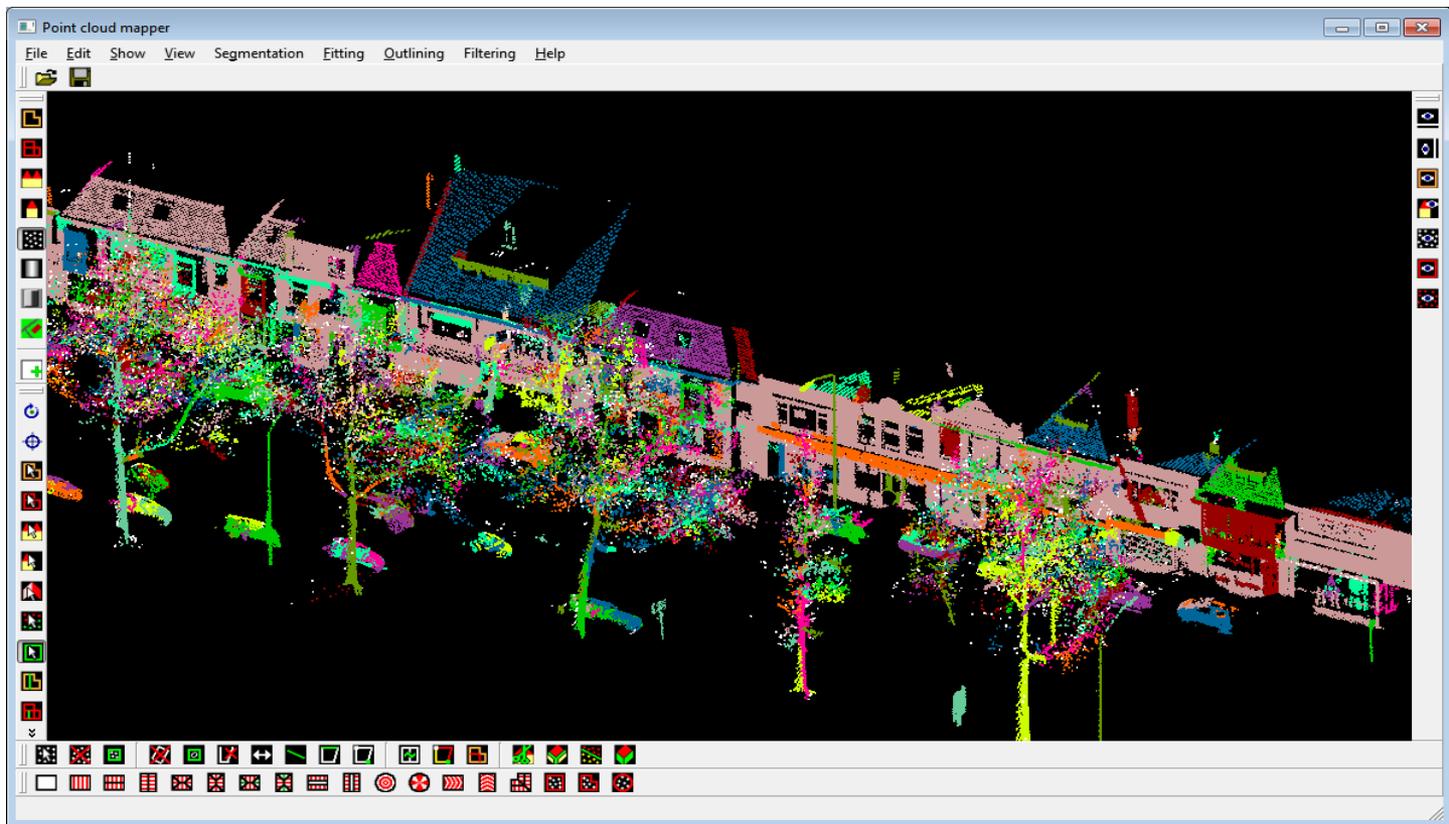
METHODS

1. Ground removal
2. Connected-component segmentation
3. Static temporary objects:
→ feature extraction + classification
4. Moving temporary objects:
→ closest point analysis + classification



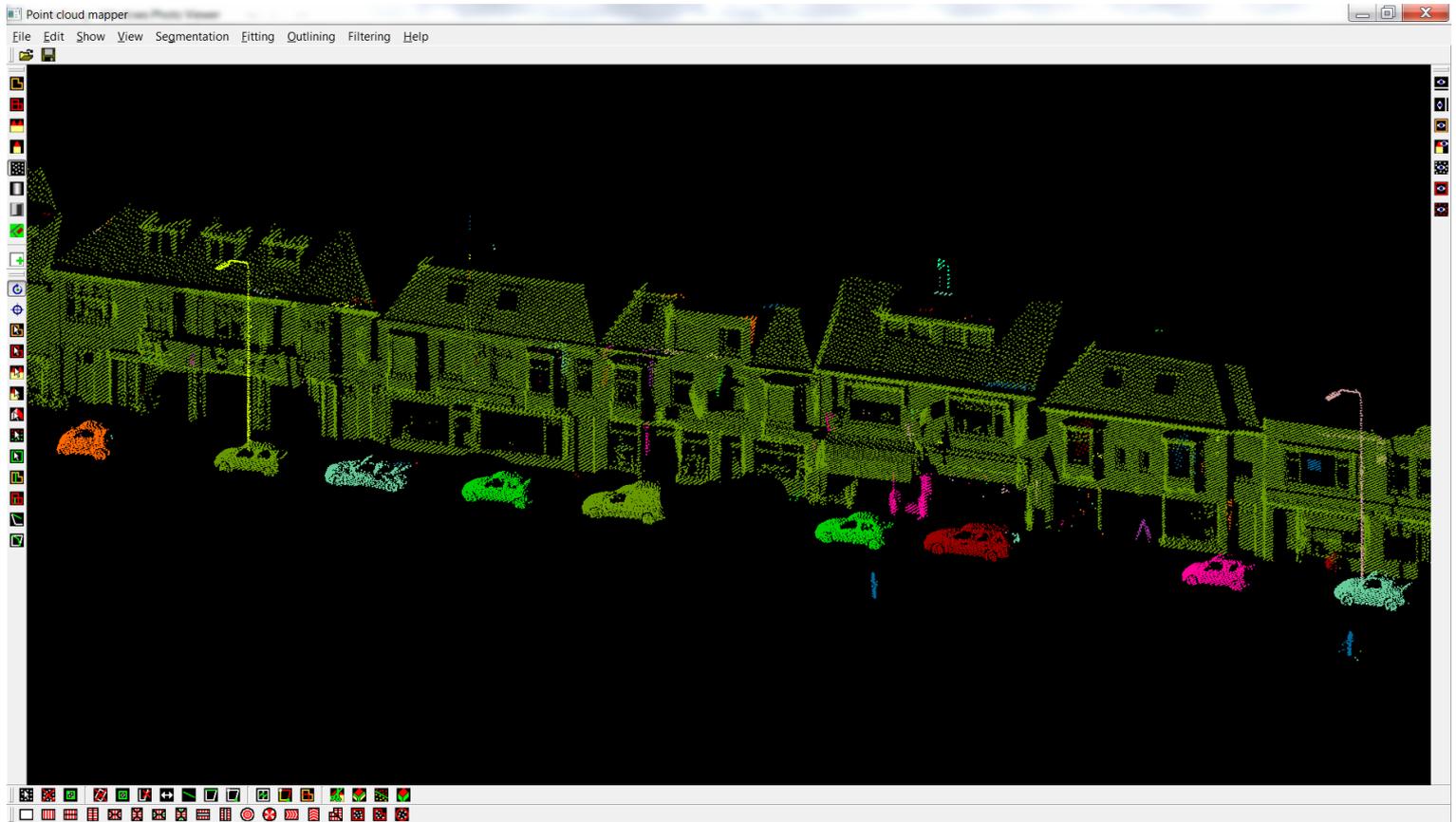
GROUND REMOVAL

- Plane-based segmentation
- Ground = large & low segment



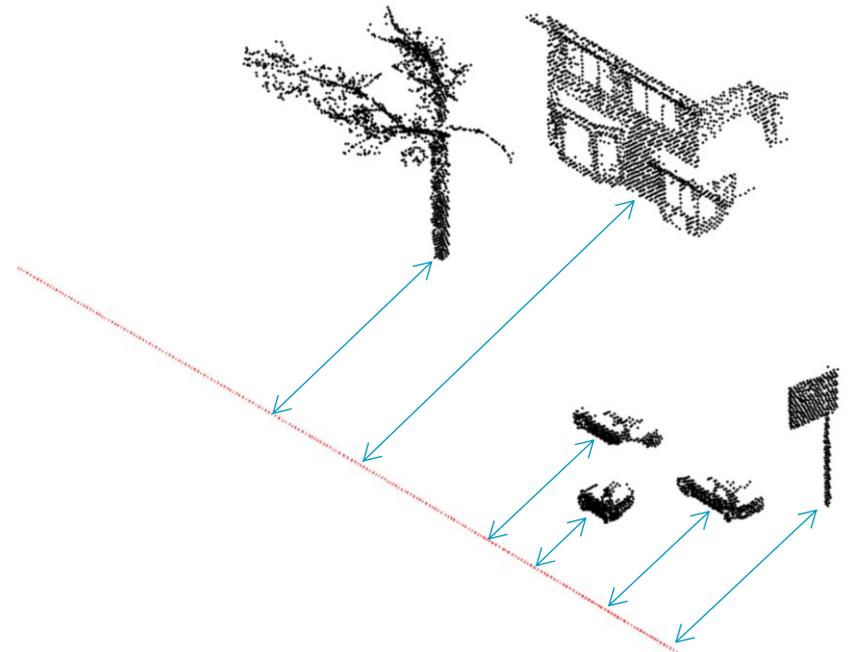
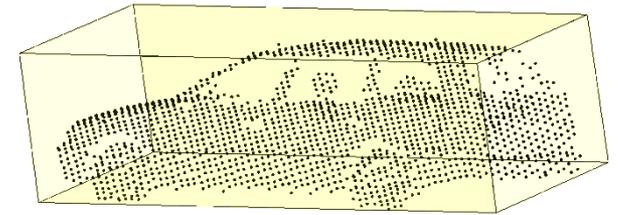
CONNECTED COMPONENT SEGMENTATION

- Points that are closer than a certain distance (e.g. 30 cm) belong to one connected component;



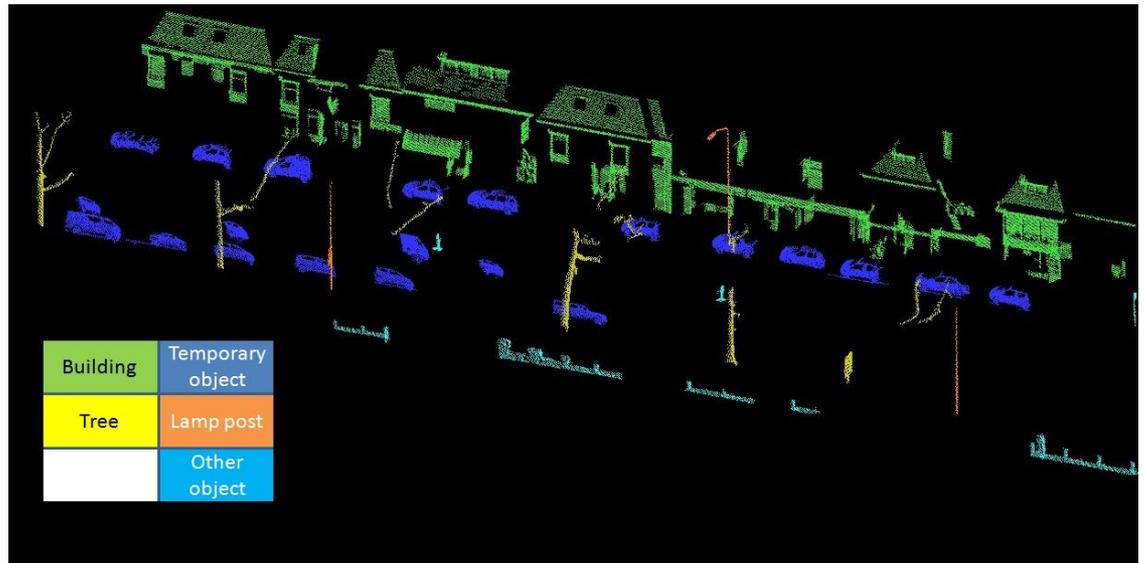
FEATURE EXTRACTION

- Shape features
 - Size (number of points)
 - Area on horizontal plane
 - Mean density
 - Height of the lowest point
 - Height
- Contextual feature
 - Distance to trajectory
- Eigen-based features
 - Anisotropy
 - Planarity
 - Sphericity
 - Linearity



CLASSIFICATION

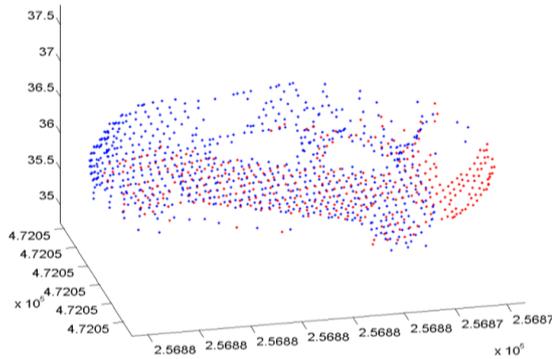
- Ground truth for training and evaluation: manual labeling (115 samples)
- Feature selection
 - Forward Selection (FS)
 - Backward Elimination (BE)
- Classification
 - Linear Discriminant Classifier (LDC)
 - Support Vector Machine (SVM)



CLOSEST POINT ANALYSIS

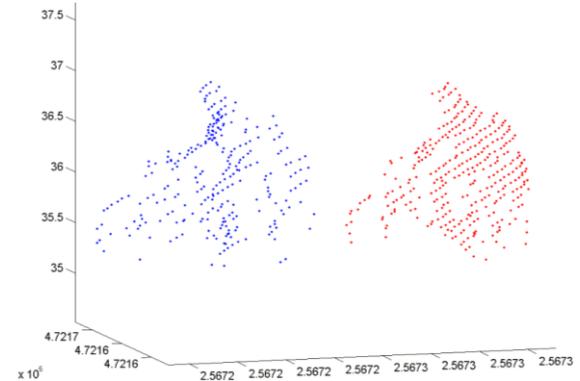


108 pairs have been done , 133 still to go

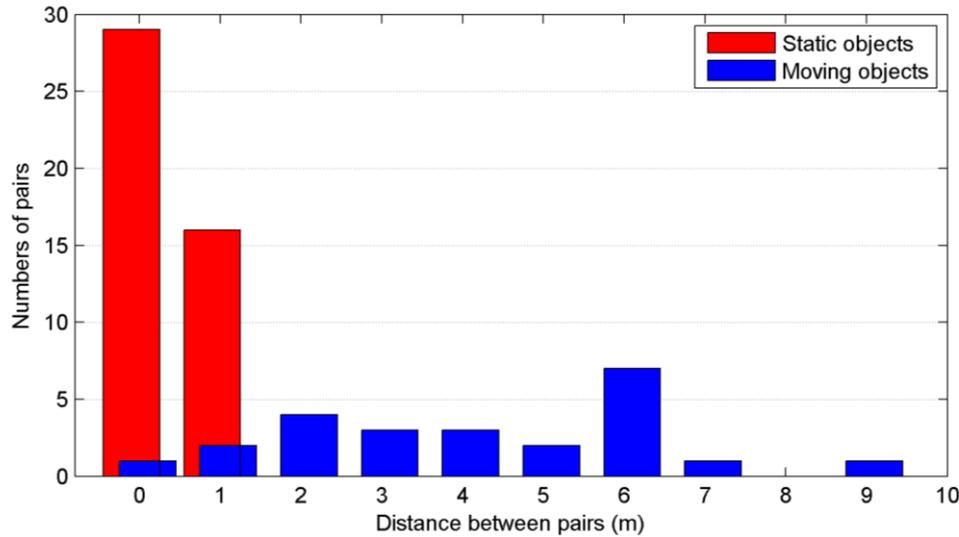


Static

73 pairs have been done , 168 still to go



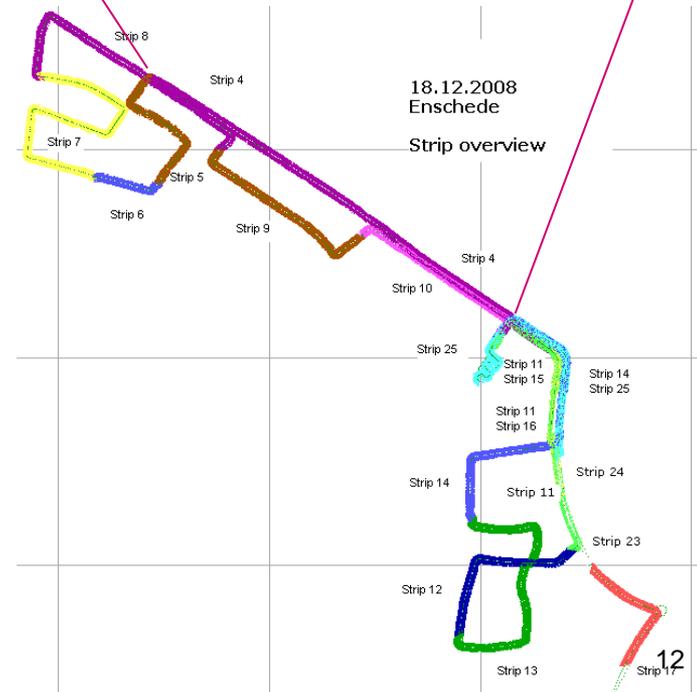
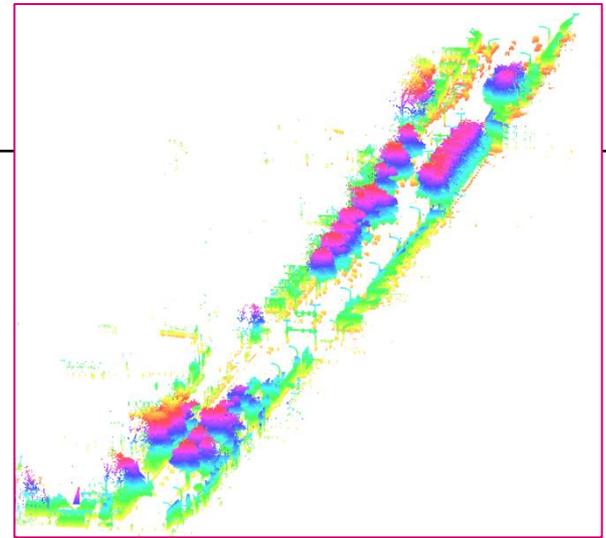
Moving



	Static objects	Moving objects
Number	45	24

EXPERIMENTS

- Study area
→ Enschede, urban
- Data
→ TopScan MLS
→ One strip: ~20 mio points





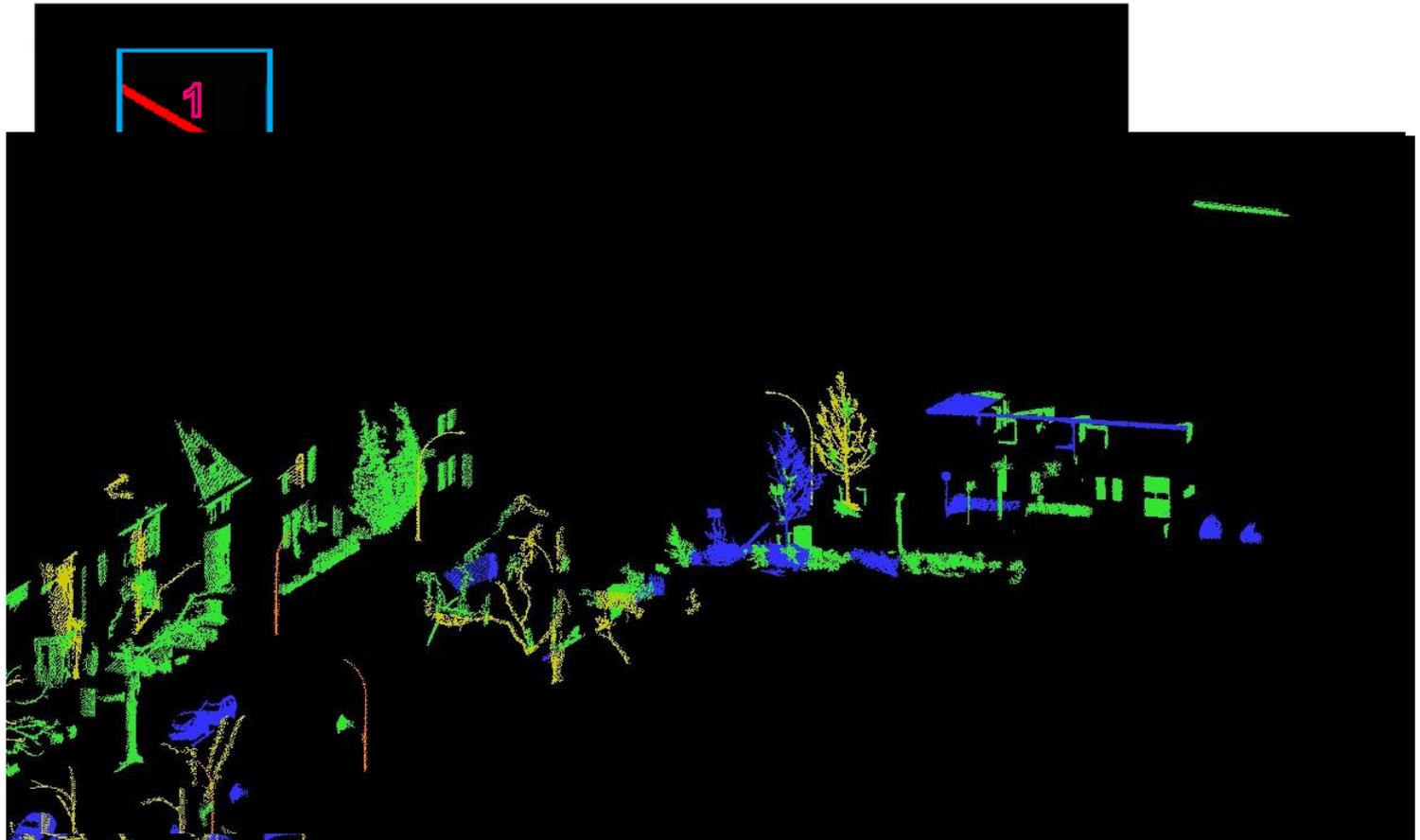
RESULTS

- Detection results for static temporary objects (test set):

Classifier	Completeness	Correctness
LDC - all features	0.90	0.86
LDC - FS (8 features)	0.90	0.79
LDC - BE (8 features)	0.95	0.91
SVM - all features	1.00	0.91
SVM - FS (9 features)	1.00	1.00
SVM - BE (11 features)	1.00	0.95

RESULTS

- Classification results for the whole strip :





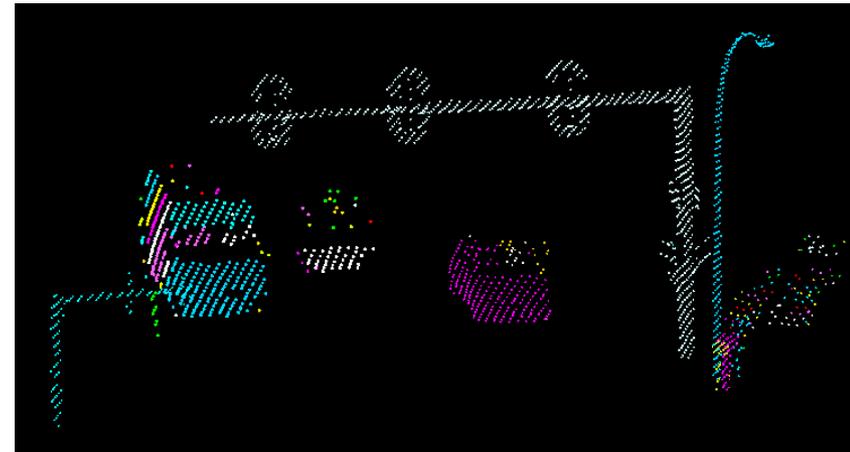
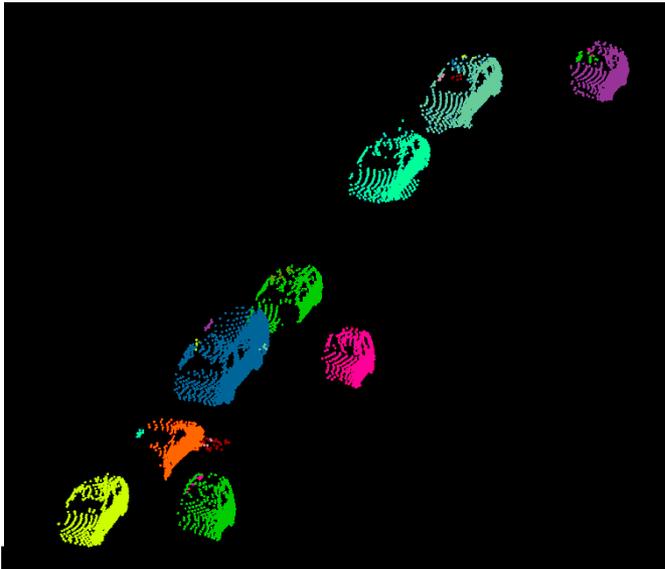
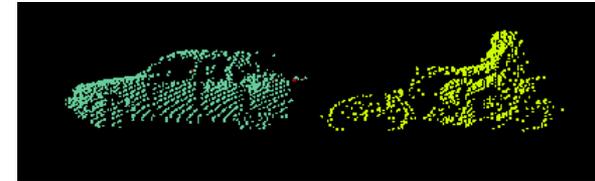
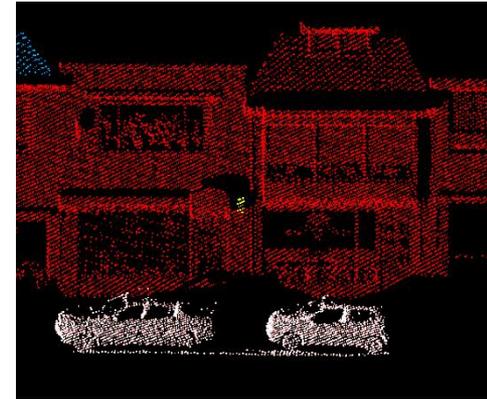
RESULTS

- Detection results for moving objects:
 - Total 64 samples
 - 4 false positives
 - 6 false negatives

	Sensor 1	Sensor 2	Sensor1 + Sensor2
Completeness	0.93	0.86	0.90
Correctness	0.90	0.96	0.93
Overall Accuracy	0.84	0.83	0.84

LIMITATIONS

- Occlusion;
- Overgrown and undergrown segments;
- Shrunk or elongated shapes due to movement.
- Variable shape and size of vehicles.





SUMMARY

- Object-based approach: obvious choice as we are dealing with objects not points;
- Connected component segmentation of objects works well (but not perfect!);
- Shape features are suitable for classification of static temporary objects; Accuracy > 90%.
- Distance between closest points is a useful measure for detecting moving objects; Accuracy > 80%
- Occlusion: objects seen by one sensor but not the other = problem for moving objects.



THANK YOU!

