

## **CE676A Project**

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### Acknowledgements

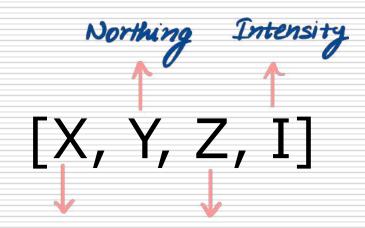
- I would like to express my sincere gratitude and appreciation to Prof. Bharat Lohani for his invaluable teaching.
- □Also, I thank to our TAs Moonis Ali & Kaustav Saha for their guidance throughout the lab work and project.

## Objective

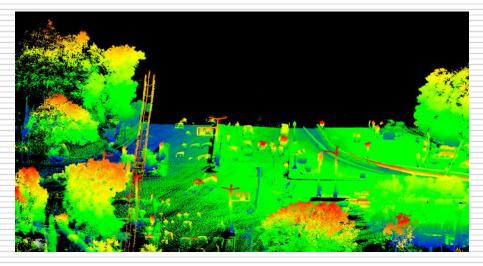
- Understand LiDAR point cloud data.
- Realize the importance of deep learning in segmenting LiDAR point cloud data.
- Adequate training with diverse labelled data crucial for optimal performance of deep learning model.
- To generate high-quality labelled data for training deep learning models that can accurately segment LiDAR point cloud data.

### **LiDAR Point Data**

### □ LiDAR point data is usually represented as:



Easting Elevation



Х	Y	Z	Intensity
670844.784	3399755.21	336.199	3
670844.773	3399755.21	336.191	7
670844.77	3399755.2	336.183	18
	•••	•••	•••

### **Segmentation & Labelling**

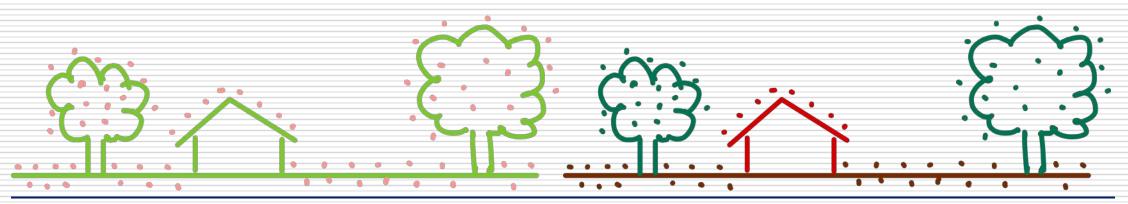
Point cloud data lacks object context/meaning.

Segmentation groups points into objects.

Labelling adds attributes and classification information.

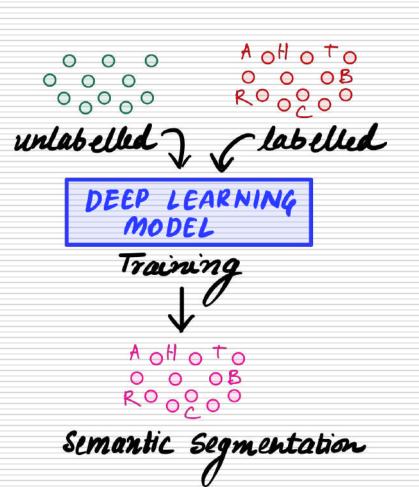
Segmentation and labelling improve LiDAR data processing.

Labelled point cloud data helps train deep learning models.



## **Deep Learning Model**

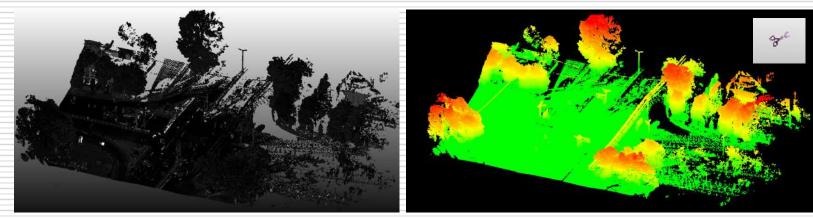
- DL models can segment point cloud data with high accuracy
- Optimal performance requires a large amount of labeled training data.
- A classifier model assigns points to predefined categories using labeled data.
- Large and diverse datasets are required for efficient training
- Labeled data generation is challenging and requires manual or automated annotation
- The quality and quantity of labeled data directly impact DL model performance.



### **SoP-Extraction of Instance Groups**

Extraction of Instance Groups:

- 1. Download raw data file from G-Drive.
- 2. Import structure.bin file and raw file, select RGB and intensity components only
- 3. Color point cloud based on height (Tools Projection Export coordinate to SF(s))
- 4. Segment out objects, merge objects of the same label class and save IG to HDD folder as .csv
- 5. Open setup wizard application file and assign label and sub-label values.
- 6. Merge classified output and upload to Drive.

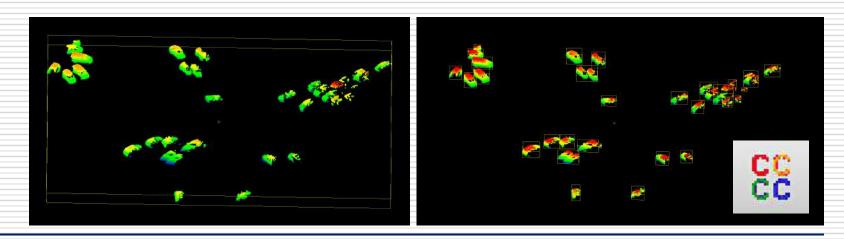


### **SoP-Extraction of Instance Objects**

Extraction of Instance Objects:

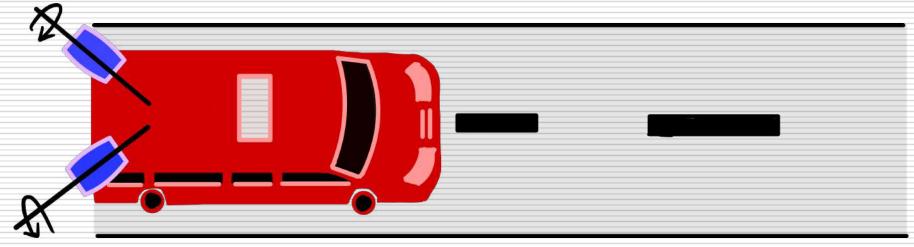
- 1. Download corrected classified data
- 2. Extract instance objects one IG at a time and save as .csv in corresponding folder
- 3. Save as "Place\_N\_Sensor\_XXZ\_n\_YYYYY.csv"
- 4. Upload the file to Instance Object folder.

The Label Connected Components tool in CloudCompare is used to group together adjacent points that belong to the same object or surface.



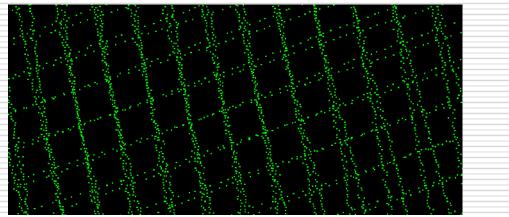
### Mobile Laser Scanner (Streetmapper)

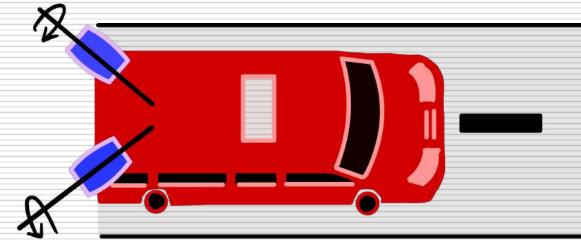
- Mobile laser scanners (MLS) are used in land surveying and can be mounted on vehicles.
- MLS reduces survey time and can quickly scan large areas.
- MLS may have lower data density than static methods.



### **Evidence of two scanners**

#### Two parallel scanning patterns at approx. 90-degree angles suggest that two scanners were used.





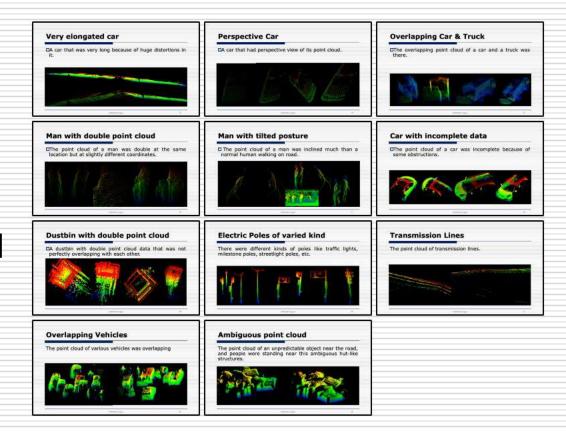
### Challenges

Challenges:

- 1. Difficulty in identifying points in a group leads to labeling inaccuracies.
- 2. Errors may occur if some points are not marked as unlabeled during object extraction.
- 3. Manual classification of ground points is time-consuming and prone to errors.
- 4. No auto-save feature can result in work loss due to technical issues.
- 5. Not having an undo-redo feature increase workload in case of errors during data segmentation.
- 6. The Setup Wizard crashes when processing huge point cloud data. Potential Solutions:
- 1. Use of filters like the CSF filter to filter out ground and non-ground points.
- 2. Training deep learning models to identify objects automatically segment point cloud data with improved labeling accuracy and efficiency.

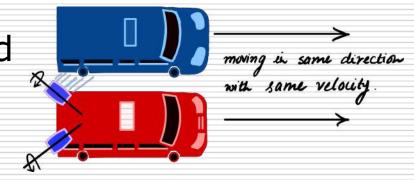
## **10 Interesting Instance Objects**

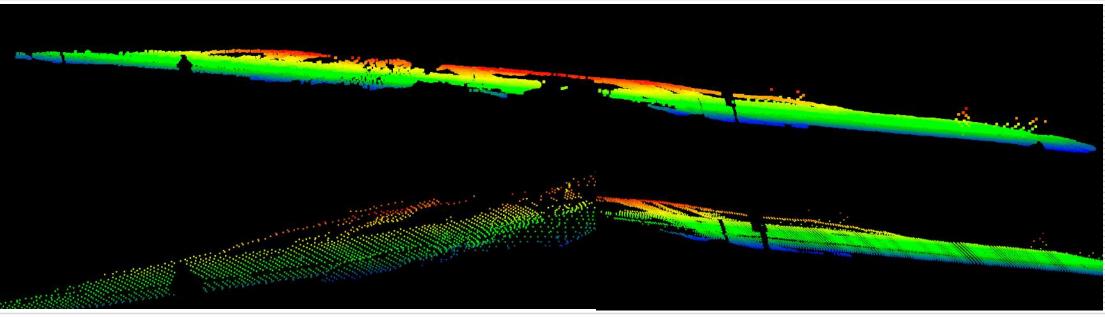
□Very Elongated Car Perspective Car Overlapping car and truck ■Man with double point cloud ■Man with tilted posture Car with incomplete data Dustbin with double point cloud Electric poles of varied kind Transmission lines Overlapping vehicles Ambiguous point cloud



### Very elongated car

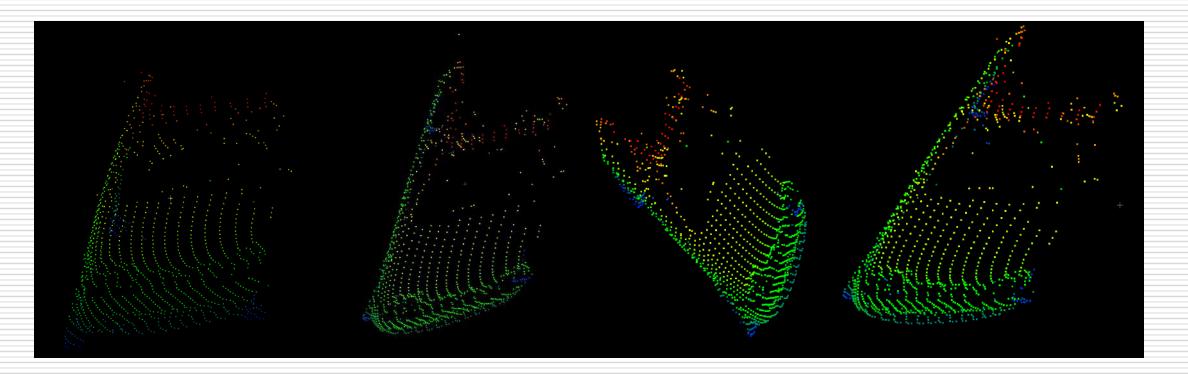
#### A car that was very long and had huge distortions in it.





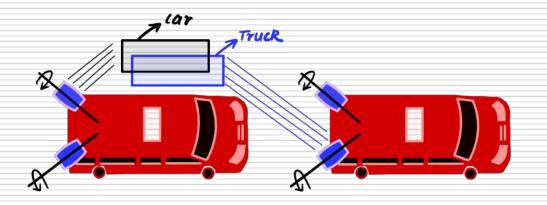
### **Perspective Car**

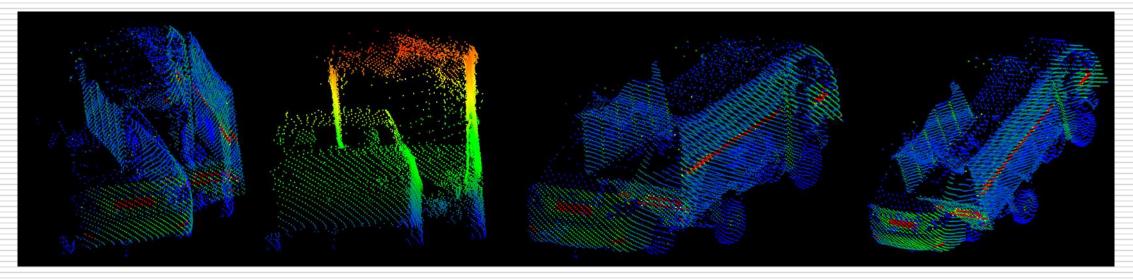
## A car that had perspective view of its point cloud.



### **Overlapping Car & Truck**

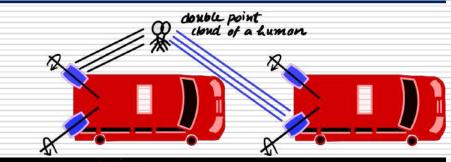
The overlapping point cloud of a car and a truck was there.

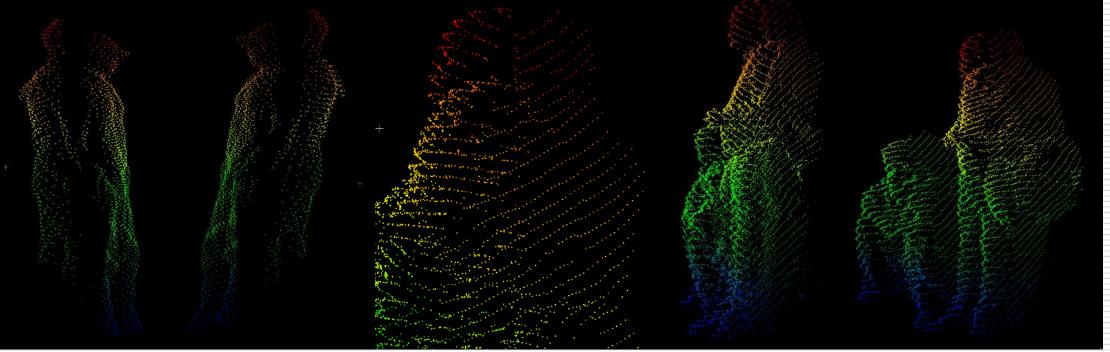




### Man with double point cloud

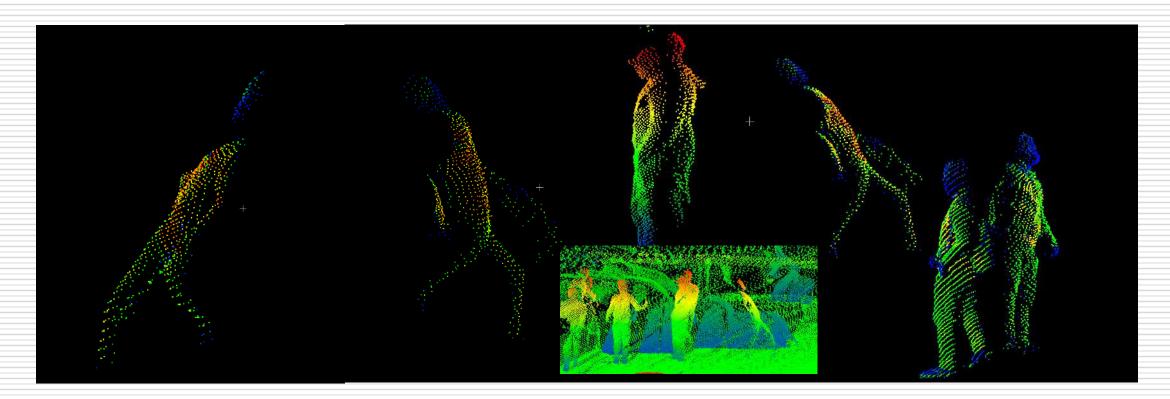
The point cloud of a man was double at the same location but at slightly different coordinates.





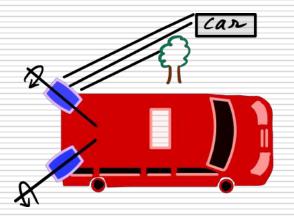
### Man with tilted posture

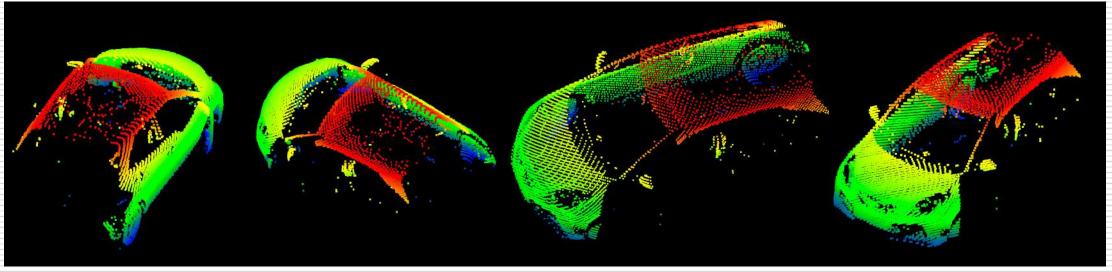
## The point cloud of a man was inclined much than a normal human walking on road.



### Car with incomplete data

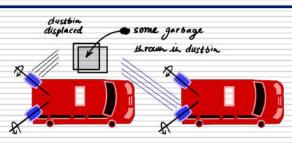
The point cloud of a car was incomplete because of some obstructions.

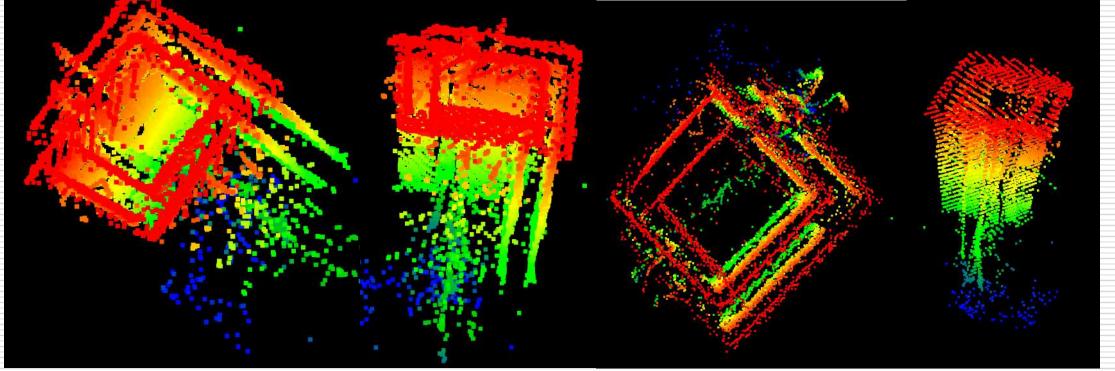




### Dustbin with double point cloud

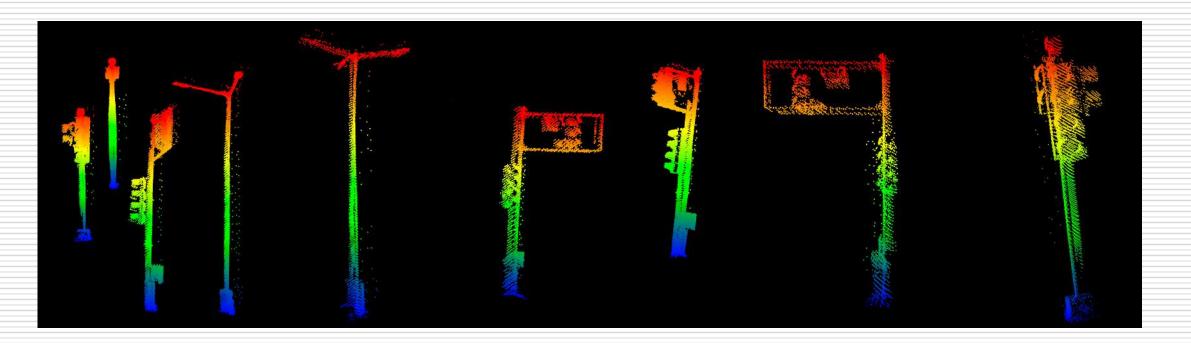
A dustbin with double point cloud data that was not perfectly overlapping with each other.





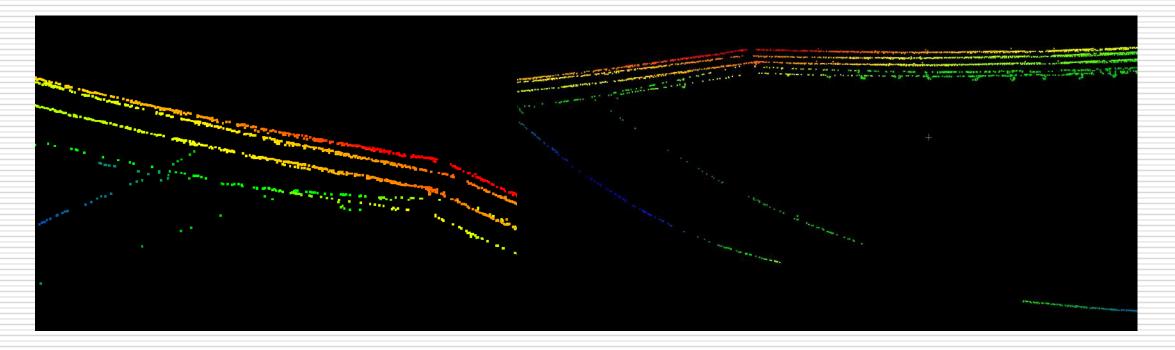
### **Electric Poles of varied kind**

There were different kinds of poles like traffic lights, milestone poles, streetlight poles, etc.



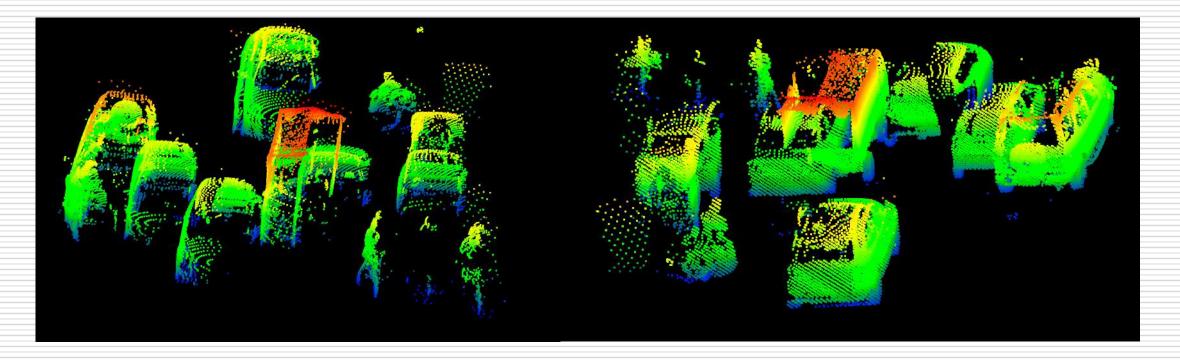
### **Transmission Lines**

### The point cloud of transmission lines.



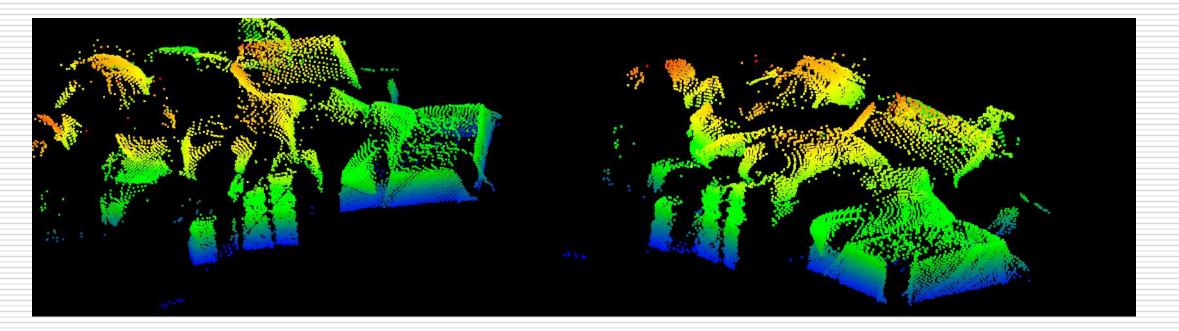
## **Overlapping Vehicles**

### The point cloud of various vehicles was overlapping

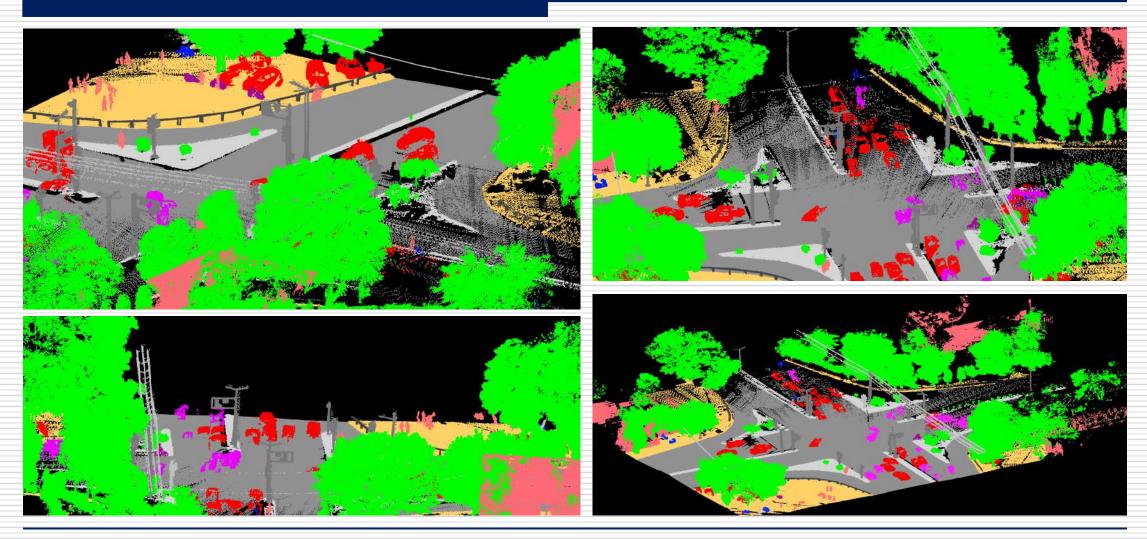


### **Ambiguous point cloud**

The point cloud of an unpredictable object near the road, and people were standing near this ambiguous hut-like structures.



### **LiDAR Point Cloud**



# Thank You!

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