

## Visualize range image channels (ID\_S1\_EX1)

### Task preparations

In file `loop_over_dataset.py`, set the attributes for code execution in the following way:

```
data_filename = 'training_segment-  
1005081002024129653_5313_150_5333_150_with_camera_labels.tfrecord' #  
show_only_frames = [0, 1]  
model = "darknet"  
sequence = "1"  
configs_det = det.load_configs(model_name='darknet')  
## Selective execution and visualization  
exec_detection = []  
exec_tracking = []  
exec_visualization = ['show_range_image']  
exec_list = make_exec_list(exec_detection, exec_tracking,  
exec_visualization)
```

The resultant image is shown as



Fig 1: Range\_Image

## Visualize lidar point-cloud (ID\_S1\_EX2)

### Task preparations

In file `loop_over_dataset.py`, set the attributes for code execution in the following way:

```
data_filename = data_filename = 'training_segment-  
10963653239323173269_1924_000_1944_000_with_camera_labels.tfrecord'  
show_only_frames = [0, 200]  
model = "darknet"  
sequence = "3"  
configs_det = det.load_configs(model_name='darknet')  
## Selective execution and visualization  
exec_detection = []  
exec_tracking = []  
exec_visualization = ['show_pcl']  
exec_list = make_exec_list(exec_detection, exec_tracking,  
exec_visualization)
```

An example frame is shown below

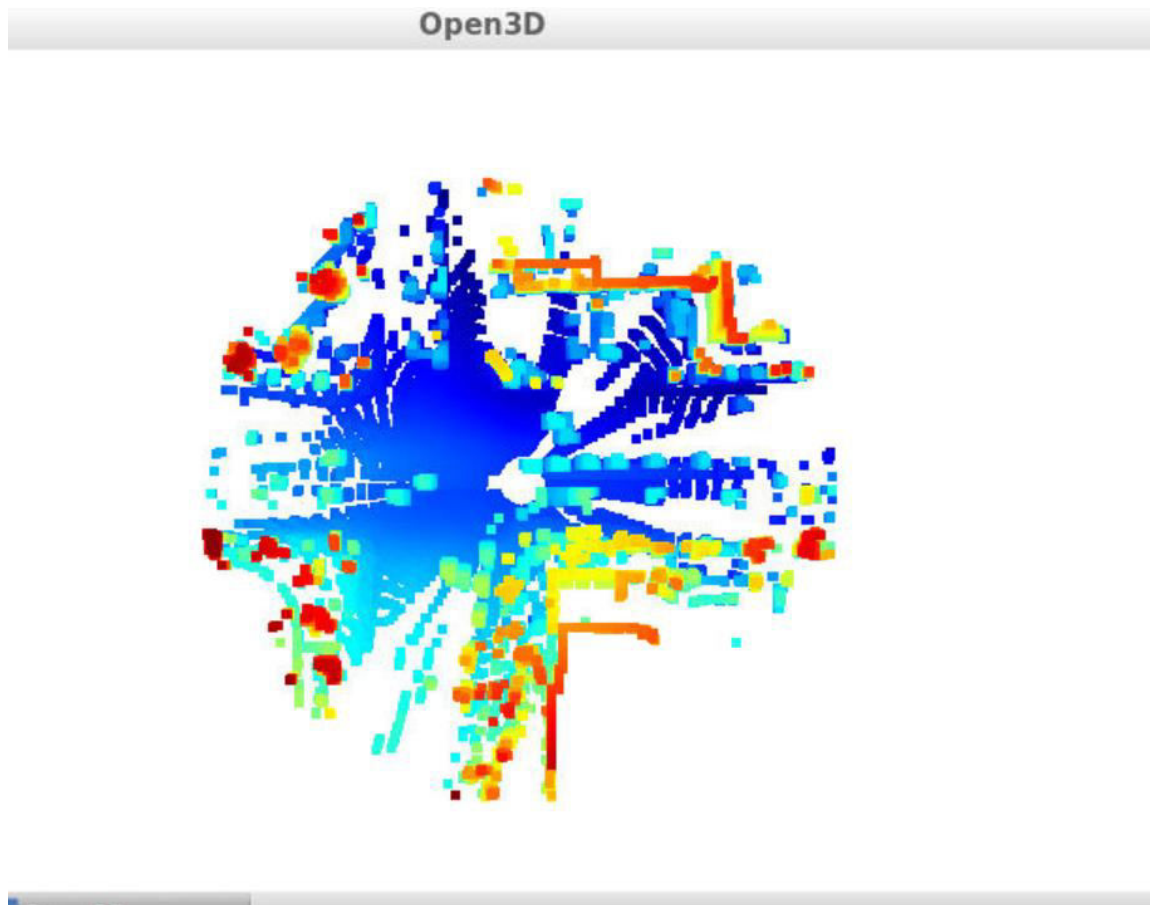


Fig 2. Visualization of LIDAR point cloud

Find 10 examples of vehicles with varying degrees of visibility in the point-cloud

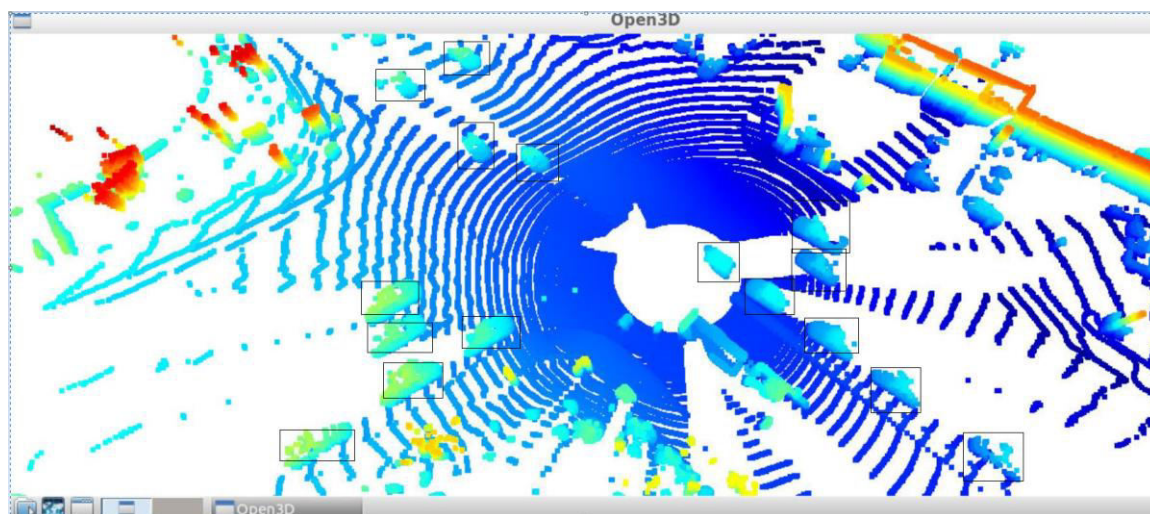
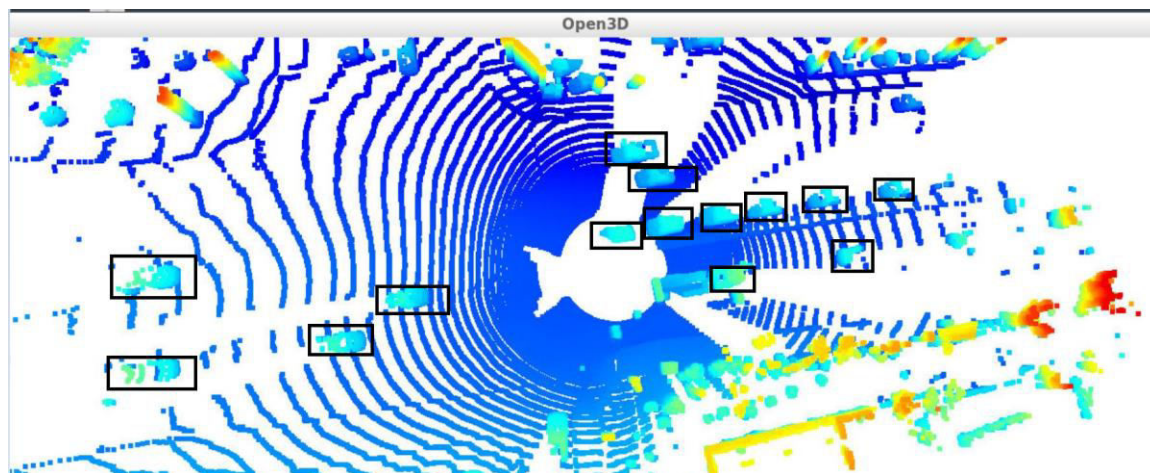
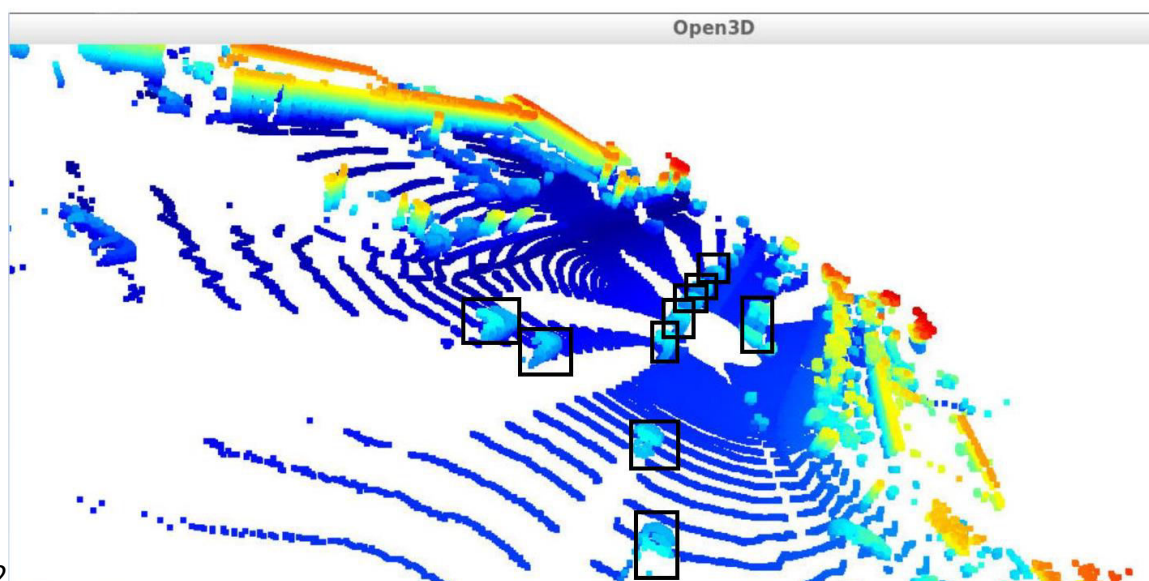


Fig 3.1



Fig



3.2

Fig 3.3



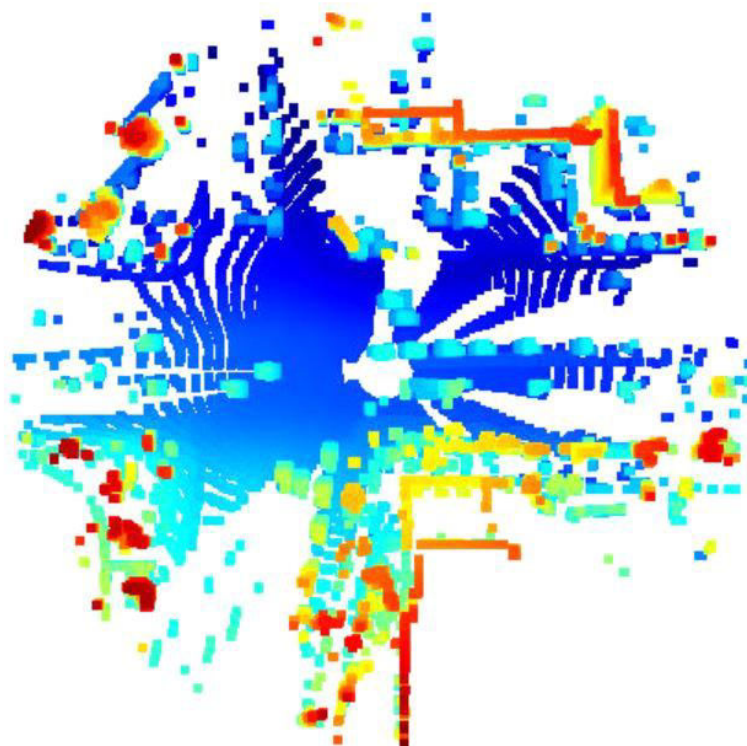


Fig 3.4

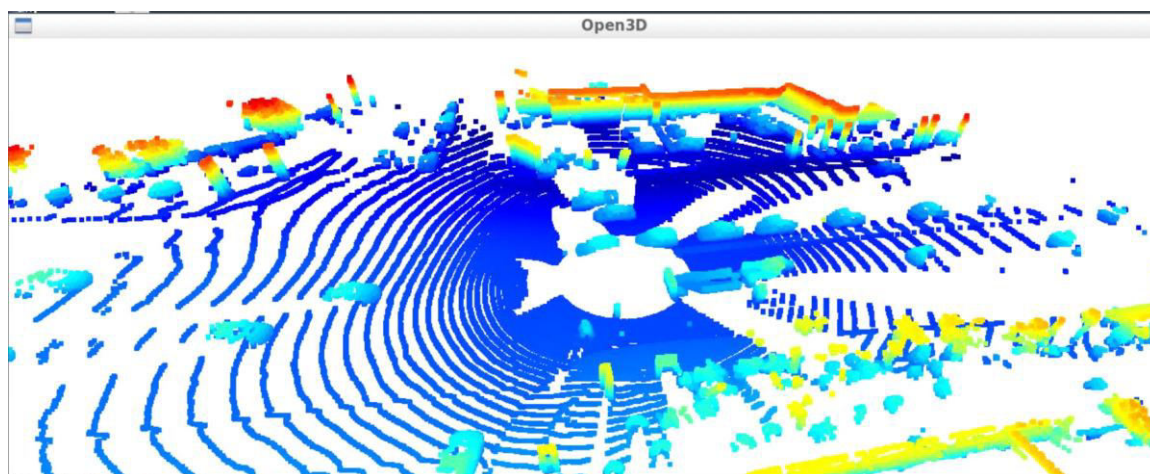


Fig 3.5

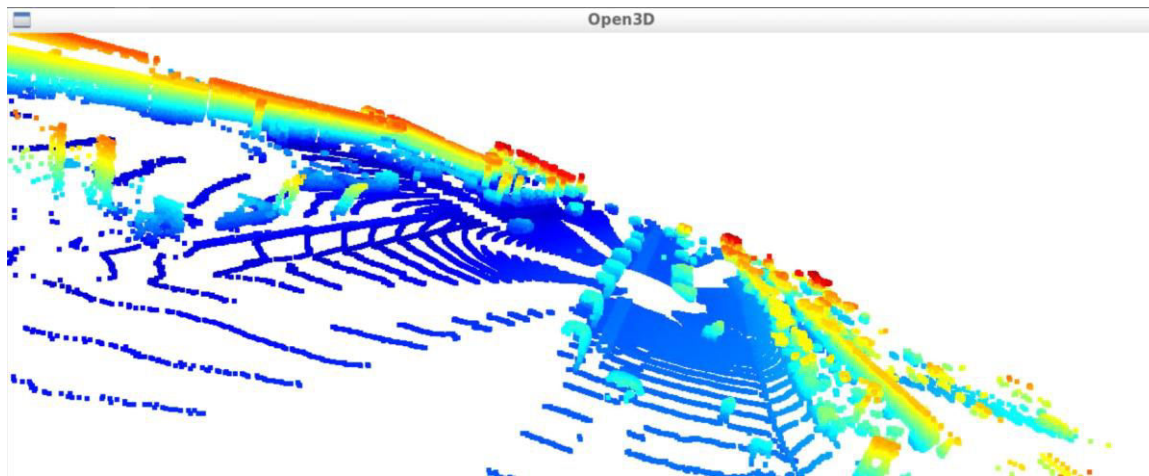


Fig 3.6

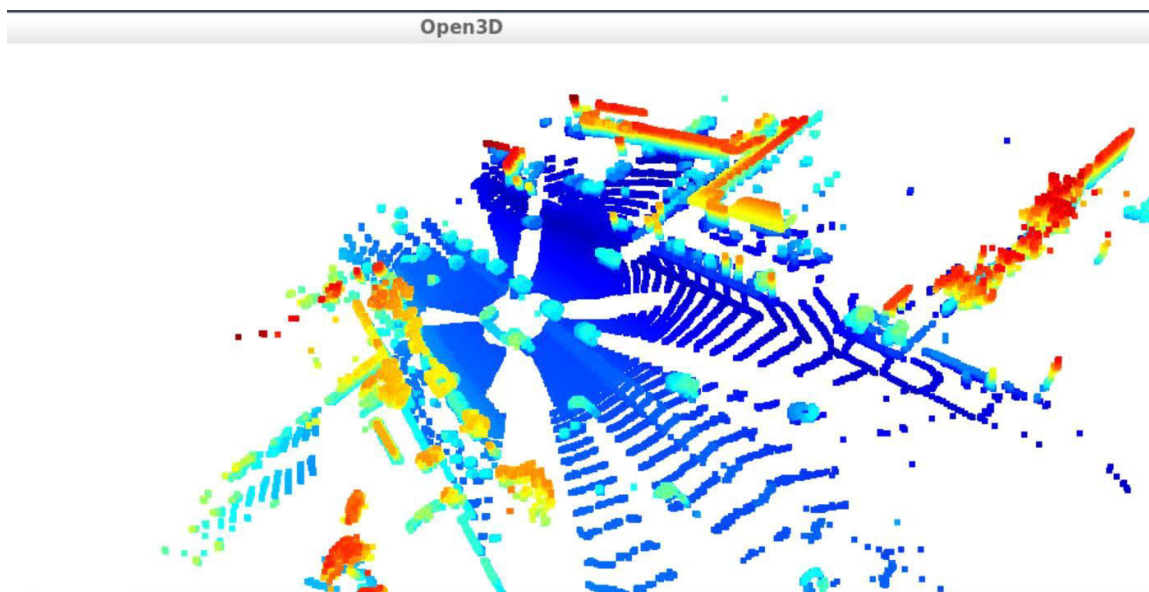


Fig 3.7

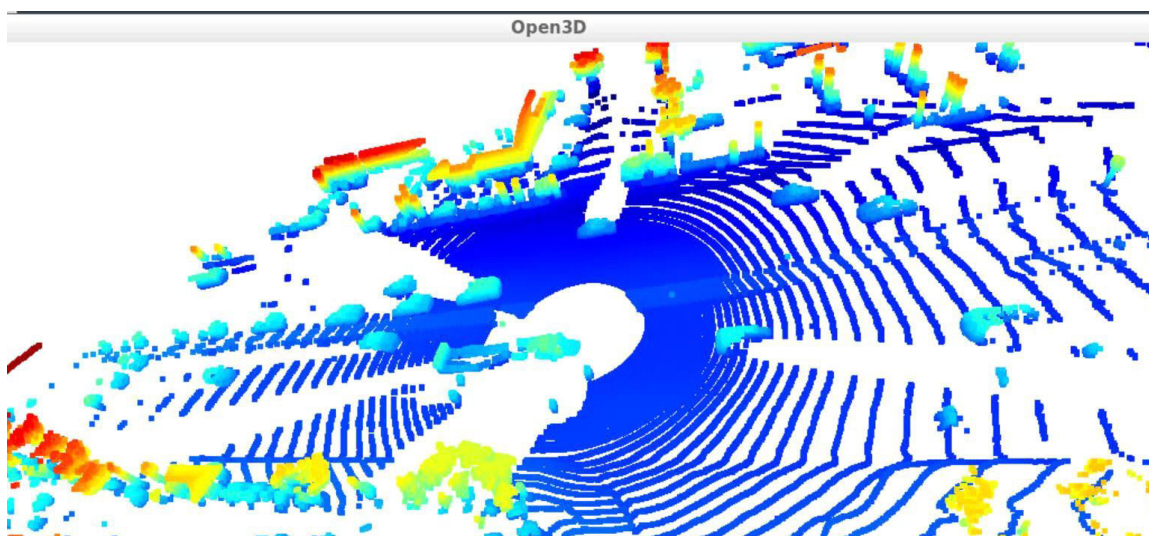


Fig 3.8

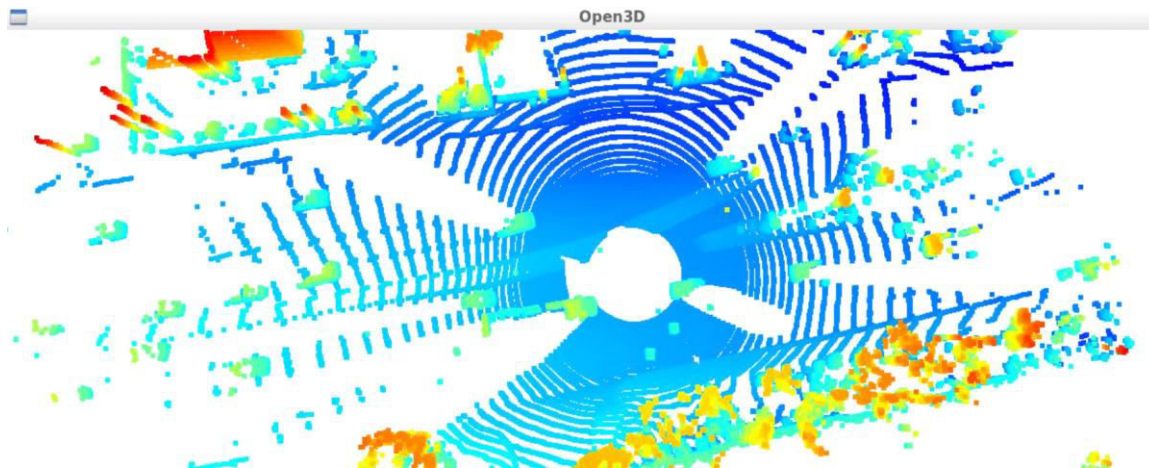


Fig 3.9

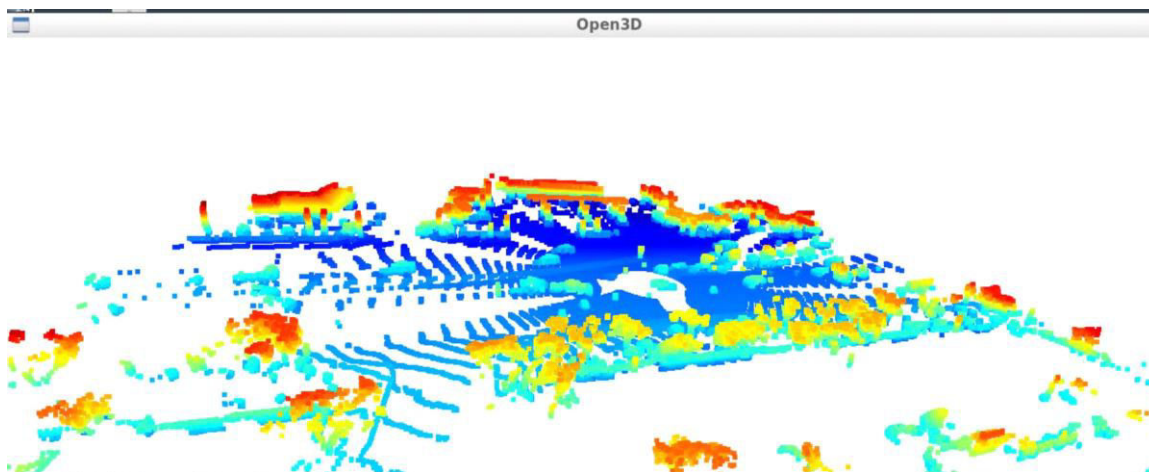


Fig 3.10

Try to identify vehicle features that appear stable in most of the inspected examples and describe them

The most prominent vehicle feature visible is the front bumper of the car. In many cases the roof as well as the sides are also visible.

## Section 2 : Create Birds-Eye View from Lidar PCL

**Convert sensor coordinates to BEV-map coordinates (ID\_S2\_EX1)**

**Compute intensity layer of the BEV map (ID\_S2\_EX2)**

**Compute height layer of the BEV map (ID\_S2\_EX3)**

In file `loop_over_dataset.py`, set the attributes for code execution in the following way:



```

1 data_filename = 'training_segment-
  1005081002024129653_5313_150_5333_150_with_camera_labels.tfrecord'
2 show_only_frames = [0, 1]
3 model = "darknet"
4 sequence = "1"
5 configs_det = det.load_configs(model_name='darknet')
6 ## Selective execution and visualization
7 exec_detection = ['bev_from_pcl']
8 exec_tracking = []
9 exec_visualization = []
10 exec_list = make_exec_list(exec_detection, exec_tracking,
    exec_visualization)

```

The result images are

Open3D

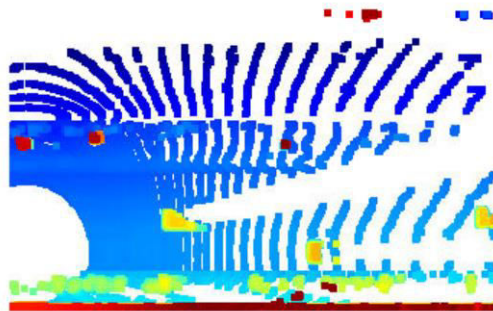


Fig 4: An example visualization to BEV map coordinates

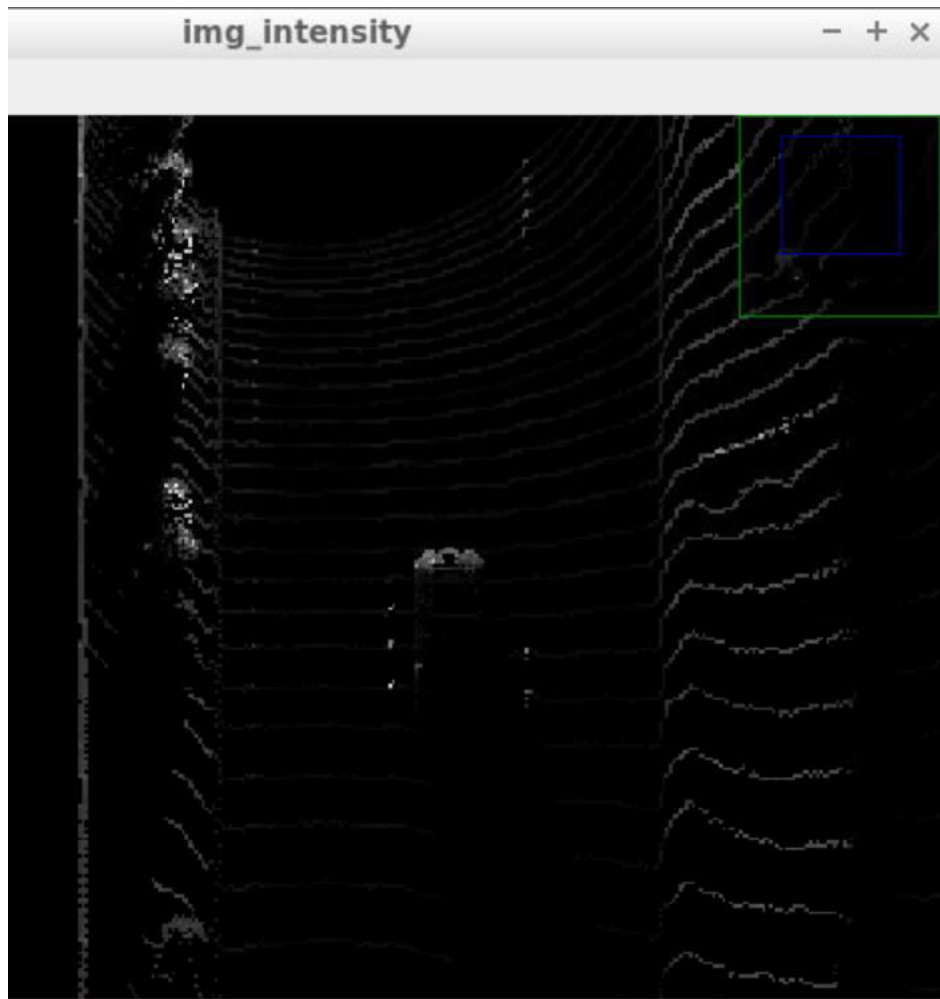


Fig 5: An example intensity layer from the BEV map



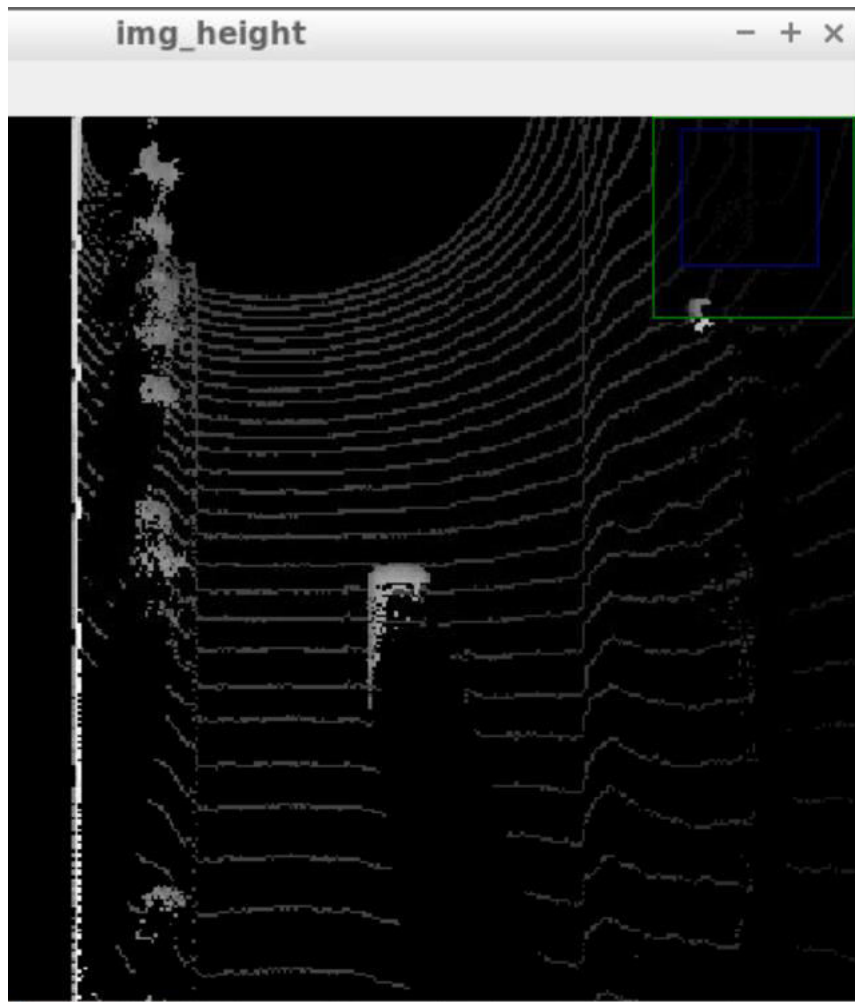


Fig 6 : An example height layer from the BEV map

### Section 3 : Model-based Object Detection in BEV Image

**Add a second model from a GitHub repo (ID\_S3\_EX1)**

**Extract 3D bounding boxes from model response (ID\_S3\_EX2)**

Task preparations

In file `loop_over_dataset.py`, set the attributes for code execution in the following way:

```

1 data_filename = 'training_segment-
1005081002024129653_5313_150_5333_150_with_camera_labels.tfrecord'
2 show_only_frames = [50, 51]
3 model = "resnet"
4 sequence = "1"
5 configs_det = det.load_configs(model_name='fpn_resnet')
6 ## Selective execution and visualization
7 exec_detection = ['bev_from_pcl', 'detect_objects']
8 exec_tracking = []
9 exec_visualization = ['show_objects_in_bev_labels_in_camera']
10 exec_list = make_exec_list(exec_detection, exec_tracking,
exec_visualization)

```

In the file objdet\_tools.py in 'misc' the following changes were made

```

# show combined view
out_img = cv2.resize( out_img,( 500, 500))
cv2.imshow('labels vs. detected objects', out_img)

```

The result image is



Fig 7 : 3D bounding boxes added to the images

#### Section 4 : Performance Evaluation for Object Detection

Compute intersection-over-union between labels and detections (ID\_S4\_EX1)

Compute precision and recall (ID\_S4\_EX3)

```
1 data_filename = 'training_segment-  
1005081002024129653_5313_150_5333_150_with_camera_labels.tfrecord'  
2 show_only_frames = [50, 150]  
3 model = "darknet"  
4 sequence = "1"  
5 configs_det = det.load_configs(model_name='darknet')  
6 ## Selective execution and visualization  
7 exec_detection = ['bev_from_pcl', 'detect_objects',  
8 'validate_object_labels', 'measure_detection_performance']  
9 exec_tracking = []  
10 exec_visualization = ['show_detection_performance']  
exec_list = make_exec_list(exec_detection, exec_tracking,  
exec_visualization)
```

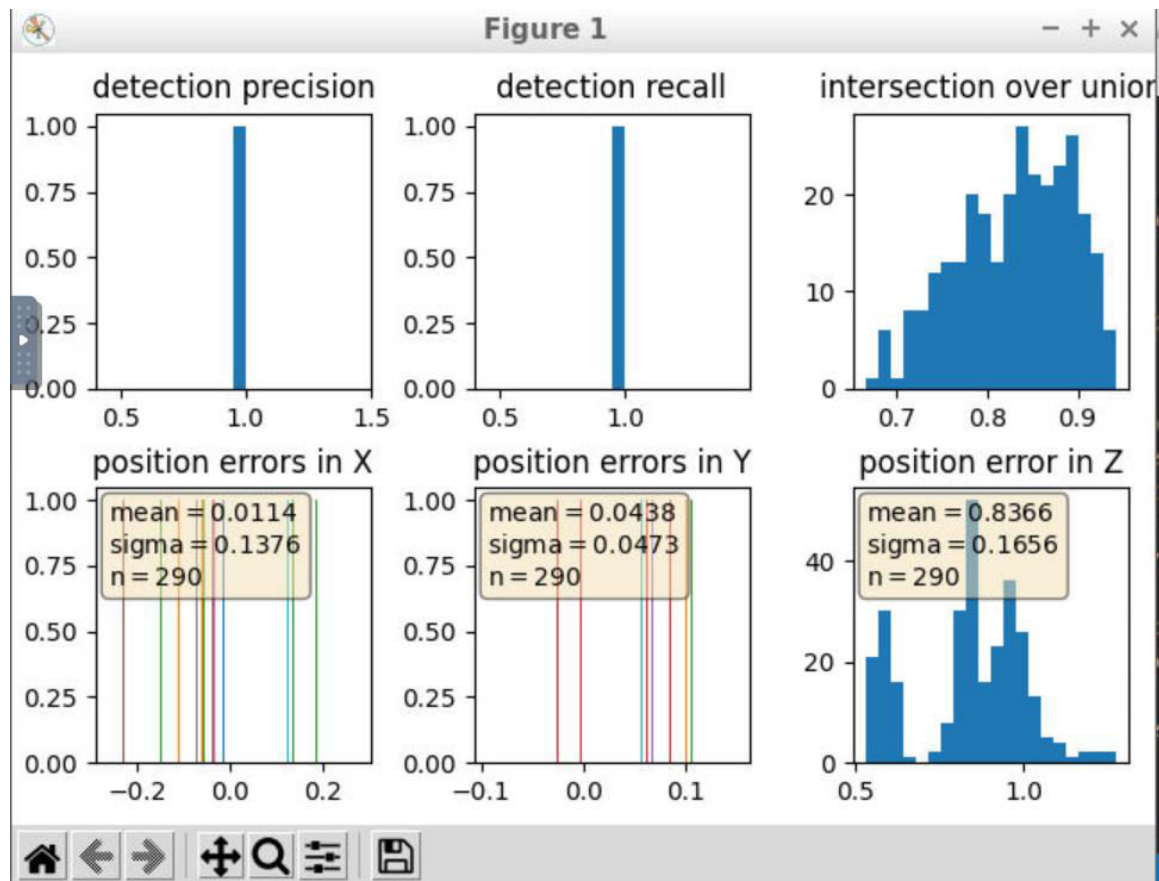


Fig 8. Graphing performance metrics

