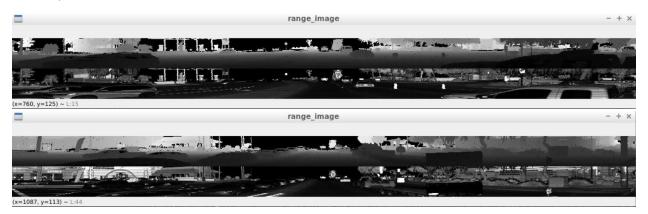
Name: Yousef Omar

Section 1: Compute Lidar Point-Cloud from Range Image

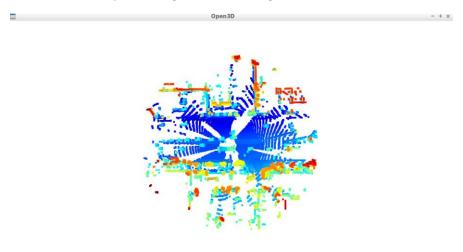
Visualize range image channels (ID_S1_EX1)

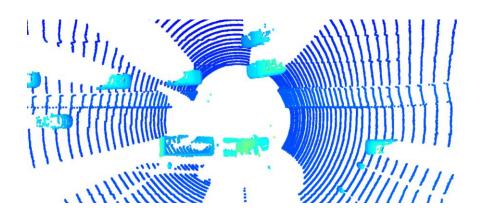
Below are a few examples from sequence #3 showing the distance information (Range) and the intensity values.

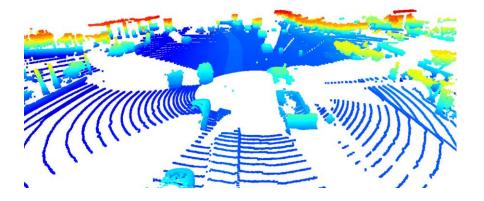


Visualize lidar point-cloud (ID_S1_EX2)

Below is an example looking at different angles:

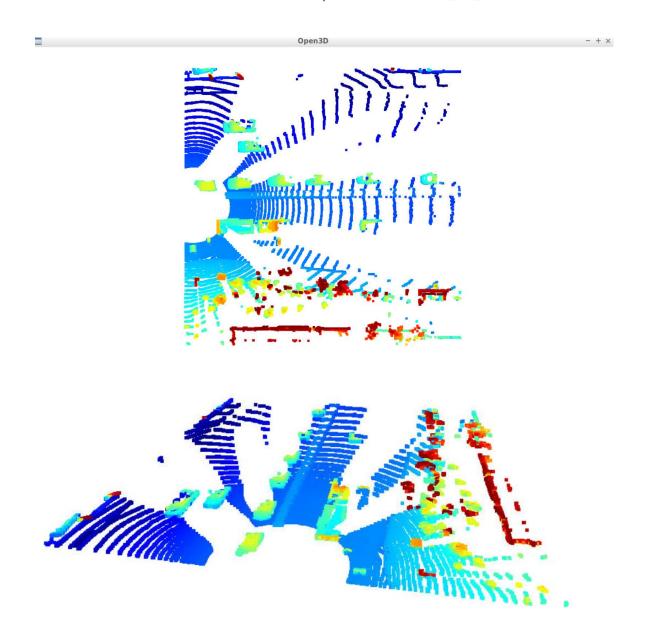






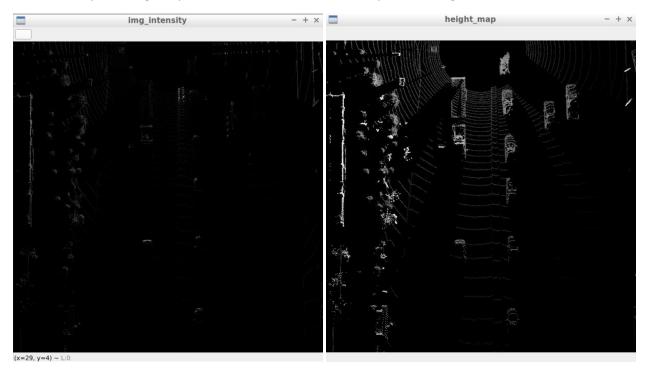
Section 2 : Create Birds-Eye View from Lidar PCL

Convert sensor coordinates to BEV-map coordinates (ID_S2_EX1)

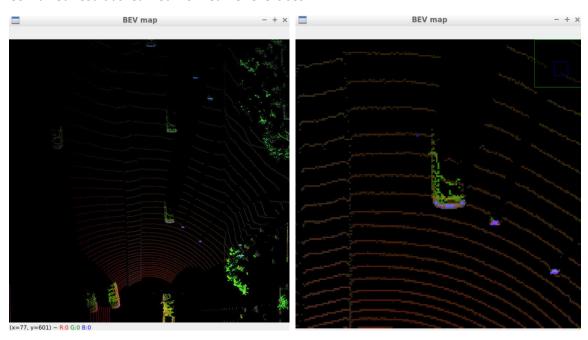


Compute layers of the BEV map (ID_S2_EX2)

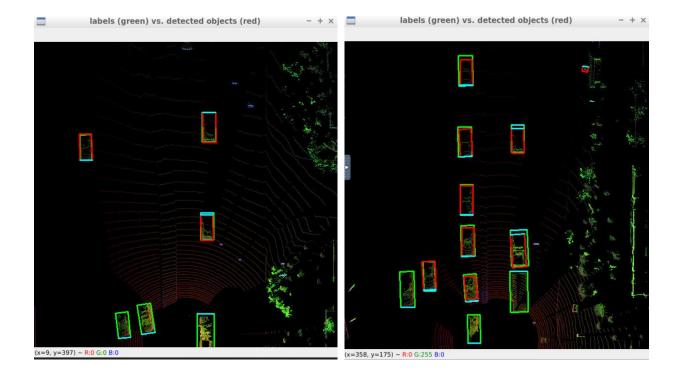
Both intensity and height layers were created for the BEV map, some images can be seen below:



Combined result as learned from earlier exercises:



Frame 44 and frame 0 from sequence #3 showing BEV with labels and detection results from darknet:



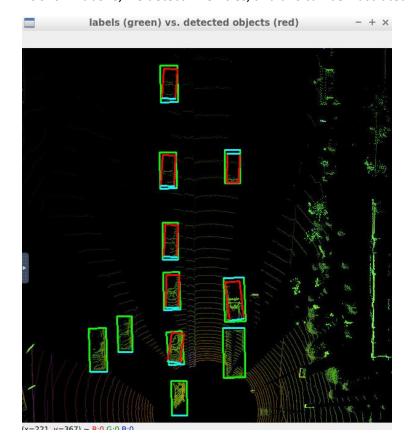
Section 3 : Model-based Object Detection in BEV Image

Add a second model from a GitHub repo (ID_S3_EX1)

Example of running the fpn-resnet:

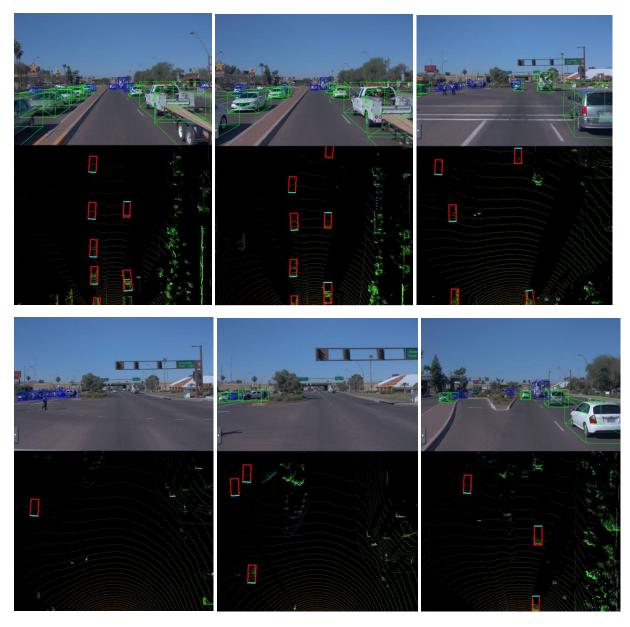
```
[[ 9.4790626e-01
                  3.6297336e+02
                                 2.9504230e+02
                                                 9.0972990e-01
   1.6079109e+00
                  2.1099014e+01
                                  5.2011196e+01
                                                 3.1189139e+00]
 [ 9.3985742e-01
                 2.6199463e+02
                                 4.1286606e+02
                                                 8.4745914e-01
   1.5074542e+00
                  2.0475266e+01
                                 4.7333508e+01 -1.4048363e-02]
 [ 8.8659835e-01
                  3.6055228e+02
                                  2.1289929e+02
                                                 9.0944433e-01
   1.5172868e+00
                  2.0398354e+01
                                 5.3292740e+01
                                                 3.1197257e+00]
                 3.6710925e+02
                                                 8.3207297e-01
 8.8226134e-01
                                 4.0901282e+02
                                                 3.0985608e+00]
   1.5089393e+00
                  2.1720228e+01
                                 5.0476284e+01
  7.9364973e-01
                  3.6469626e+02
                                  5.4978656e+02
                                                 7.8988516e-01
   1.6350408e+00
                  2.2307636e+01
                                 4.8100849e+01
                                                 3.0805218e+00]
 [ 7.3704052e-01
                  2.5970703e+02
                                  1.9777594e+02
                                                 9.9575454e-01
                                 6.1965141e+01 -3.0225177e+00]
   2.1285961e+00
                  2.3672895e+01
                                                 8.4734130e-01
  6.8496990e-01
                  3.5519940e+02
                                  1.2160492e+02
   1.5579846e+00
                                 4.7682808e+01
                  2.0121210e+01
                                                 2.9977283e+0011
```

As shown above, we detect 7 vehicles, and this can be illustrated from the red rectangles below:



Extract 3D bounding boxes from model response (ID_S3_EX2)

Below are several instances from Sequence #3 showing the BEV and 3D detection projected on the image:



Section 4: Performance Evaluation for Object Detection

Compute intersection-over-union, false-negatives and false-positives, precision and recall (ID_S4_EX1,2,3)

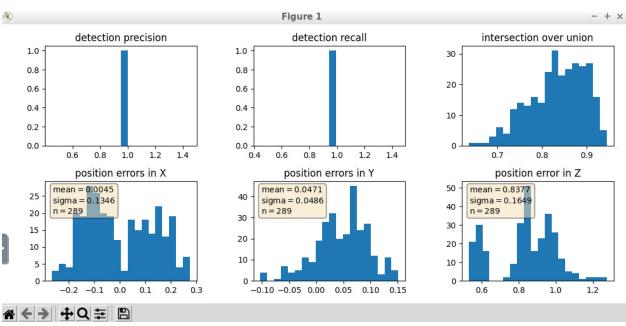
Using the following task preperations:

```
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 = [50, 150]|
| exec_data = ['pcl_from_rangeimage']|
| exec_data = ['pcl_from_pcl', 'detect_objects', 'validate_object_labels',
| 'measure_detection_performance']
| exec_tracking = []
| exec_visualization = ['show_detection_performance']
| configs_det = det.load_configs(model_name="darknet")
```

I have obtained the following results:



Extra test to make sure everything is done correctly:

