Object Detection in an Urban Environment

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- Project overview
- This section should contain a brief description of the project and what we are trying to achieve. Why is object detection such an important component of self driving car systems?
- Dataset
- Dataset analysis
- This section should contain a quantitative and qualitative description of the dataset. It should include images, charts and other visualizations.
- Cross validation
- ▶ This section should detail the cross validation strategy and justify your approach.
- Training
- Reference experiment
- This section should detail the results of the reference experiment. It should includes training metrics and a detailed explanation of the algorithm's performances.
- Improve on the reference
- This section should highlight the different strategies you adopted to improve your model. It should contain relevant figures and details of your findings.

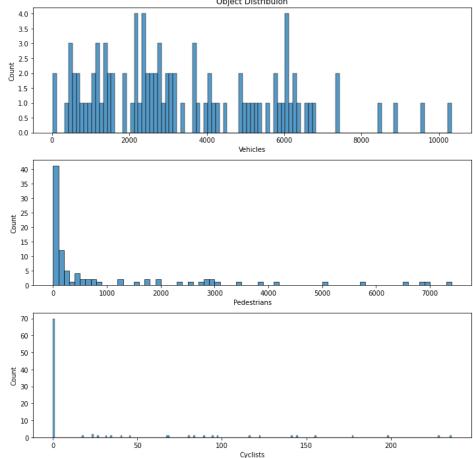
Project Overview

- ► The project goal is to use transfer learning techniques to accurately predict
 - vehicles, pedestrians and cyclists and
 - their position in the image

in video scenes that have been recorded.

This task is crucial in self-driving vehicles to ensure the correct and safe trajectory planning of the ego vehicle

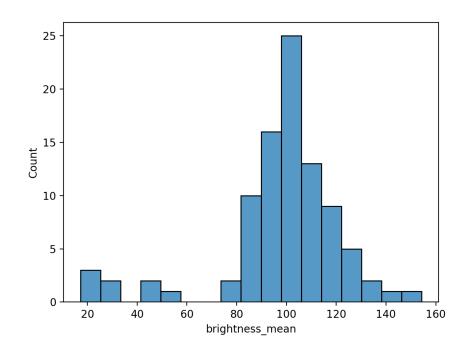
Exploratory Data Analysis Object distribution over the TFRecords



- The figure shows the distribution of the objects that can be detected for all TFRecords available
- It can clearly be observed that the #vehicles is widely distributed, whereas many frames do not have any cyclists or pedestrians
- The resulting split should try to cover these distributions as well

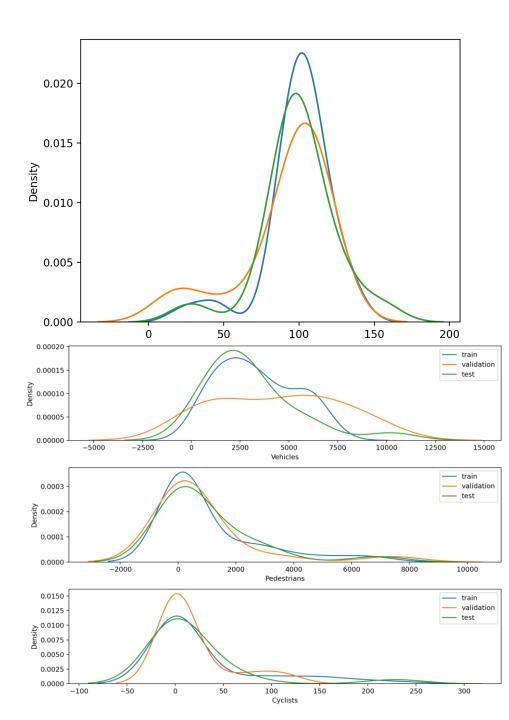
Exploratory Data Analysis Image Brightness (Day/Night scenes)

- From the calculated perceived brightness, we can clearly identify what images have been taken during the day or night.
- We want to take this into consideration when training the model. Otherwise, there is the risk of underperforming in night scenarios

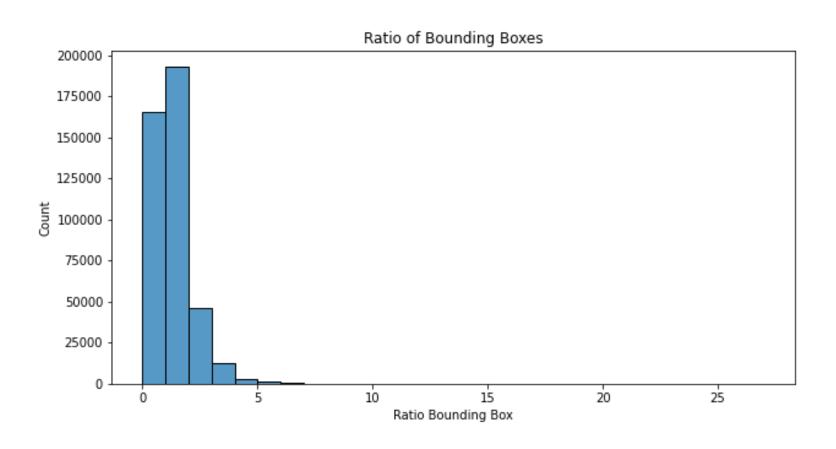


Split Creation

- For the creation of the different splits, we will take the classical splitting approach of 60/20/20 and evaluate whether we get similar probability densities for the classes that we evaluated before. This split can be chosen as we have a fairly large number of measurements
- From the probability distributions it can be observed that the shuffle procedure create similar datasets for training, validation and test. It will be necessary to evaluate later, whether it is sufficient or if we must make use of data augmentation to increase the features that are currently under-represented

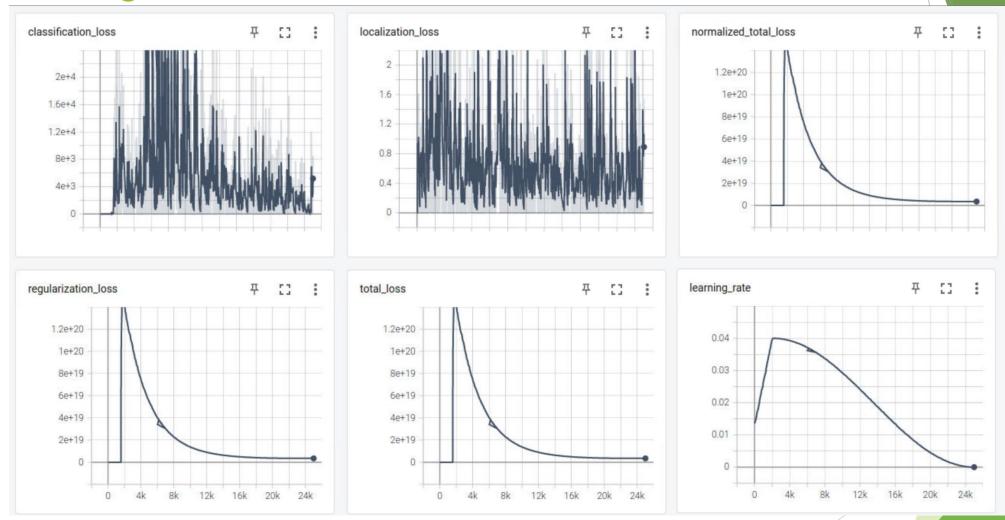


Bounding Boxes Ratio



```
anchor_generator {
  multiscale_anchor_generator {
    min_level: 3
    max_level: 7
    anchor_scale: 4.0
    aspect_ratios: 1.0
    aspect_ratios: 2.0
    aspect_ratios: 0.5
    scales_per_octave: 2
}
```

Training of Reference Model: Evolution of Loss

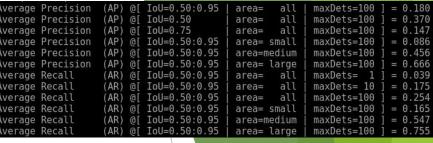


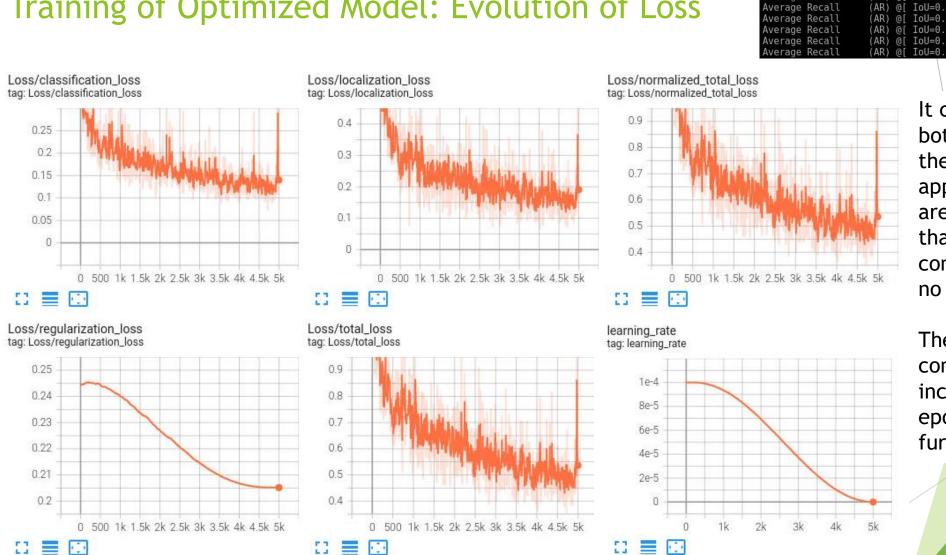
The huge losses are an indication that the learning value is too high! Model is overfitting!

Training of Optimized Model: Changes to the configuration

- ▶ The following changes were done to improve the performance of the model:
 - Reduced the learning rate to avoid wrong learning of the model \rightarrow large losses
 - Change of the optimization algorithm to ADAM
 - Include augmentations
 - random_rgb_to_gray
 - random_adjust_brightness
 - random_adjust_contrast
 - random_adjust_saturation

Training of Optimized Model: Evolution of Loss



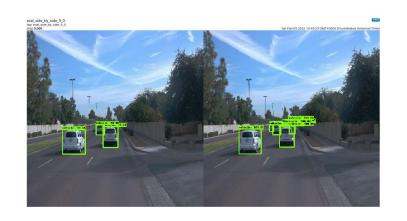


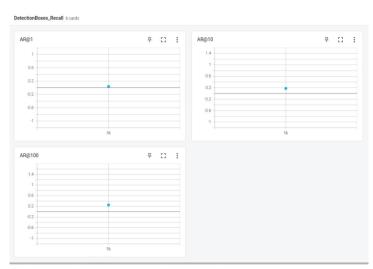
It can be observed that both the losses (here shown the training but the same applies for the validation) are decreasing. This means that the model is converging and that there is no overfitting of the model.

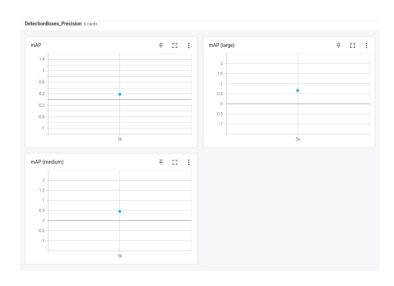
The losses have not converged yet. We can increase the number of epochs to even improve further the model

Further Possible Improvements

- Due to GPU time restrictions I could not perform the training for more than 5k epochs. Increase the values as originally configured would improve the performance of the model
- In the generated animations, it can be observed that the non-max suppression value was set too low (1e-8). Increasing this value to ~0.2 would reduce the number of windows significantly







Optimized Model Evaluation -Tensorboard

Evaluation was only possible at the end of the training as there were memory issues with the udacity workspace. Nevertheless the the initial value of the different metrics (AP50, mAP) were 0 at the beginning and that they are increasing