



Can you still see me?

The effect of image compression on state-of-the-art object detection neural networks

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17.02.2020









Motivation

Problem:

- High quality data generated by **front facing cameras** on cars (~9GB / minute)
- $10 \text{ cars} \times 8 \text{ hours} \times 60 \text{ mins} \times 9\text{GB} = 43 \text{ TB / day}$
- ~1PB/month
- Storage of 1PB is costly for a company (~5000\$ / month).



Can you still see me? - Motivation



Question: In the object detection framework, how much can image be **compressed** and **preserve accuracy** of object detection?



100% jpg



20% jpg



5% jpg



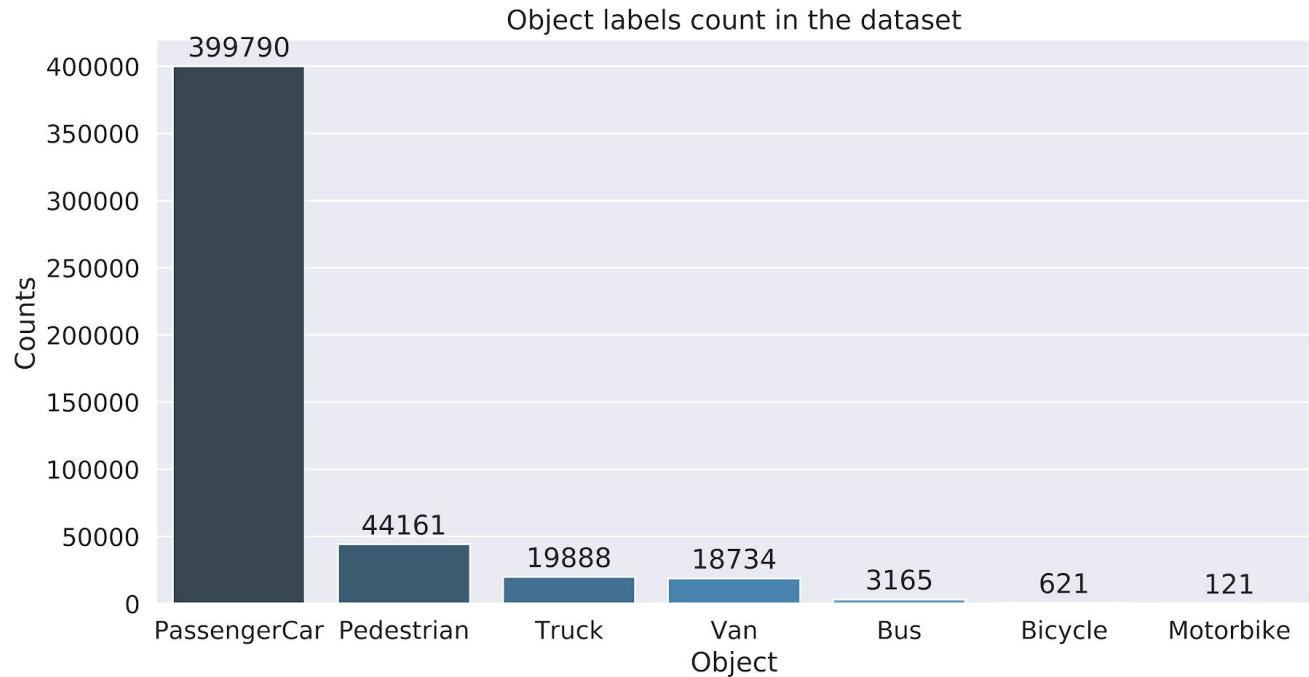
Data

- **57935 labelled images** in full resolution (.bmp format)
 - ~800 broken images among them
- **7 labelled classes**
- **Fisheye camera** (120°)
- **3 frames / minute**



Data Labels

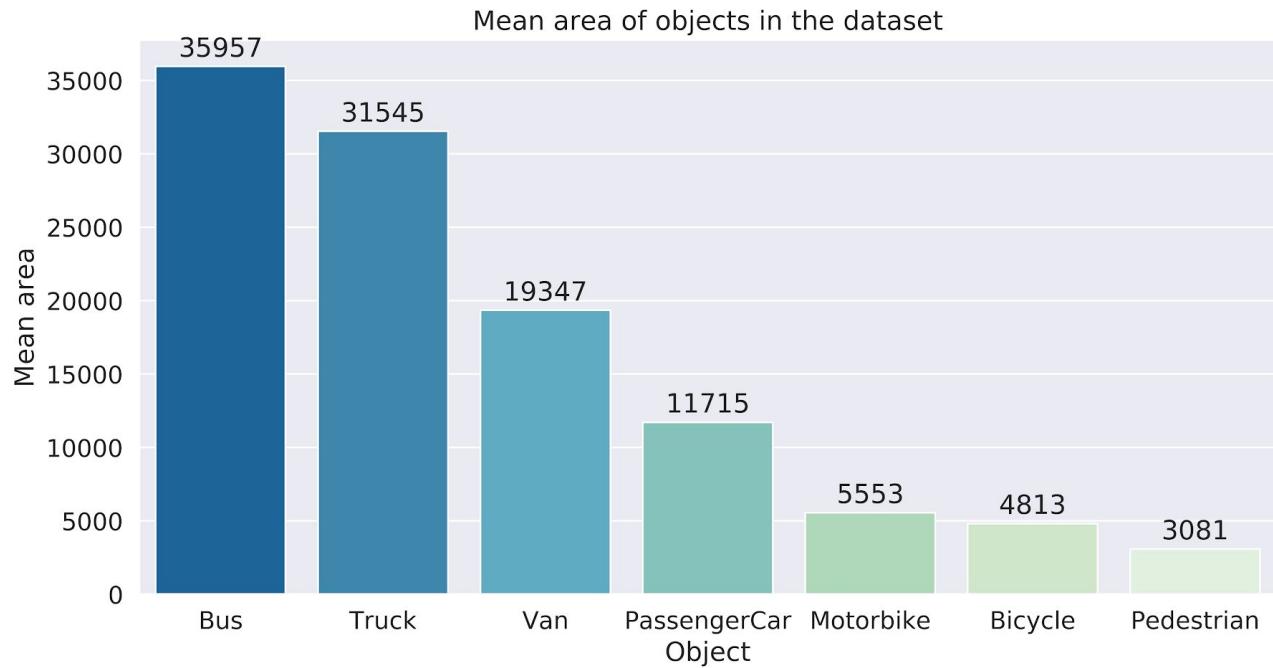
- PassengerCar
- Truck
- Van
- Bus
- Pedestrian
- Bicycle
- Motorbike





Object Localization

Bounding boxes



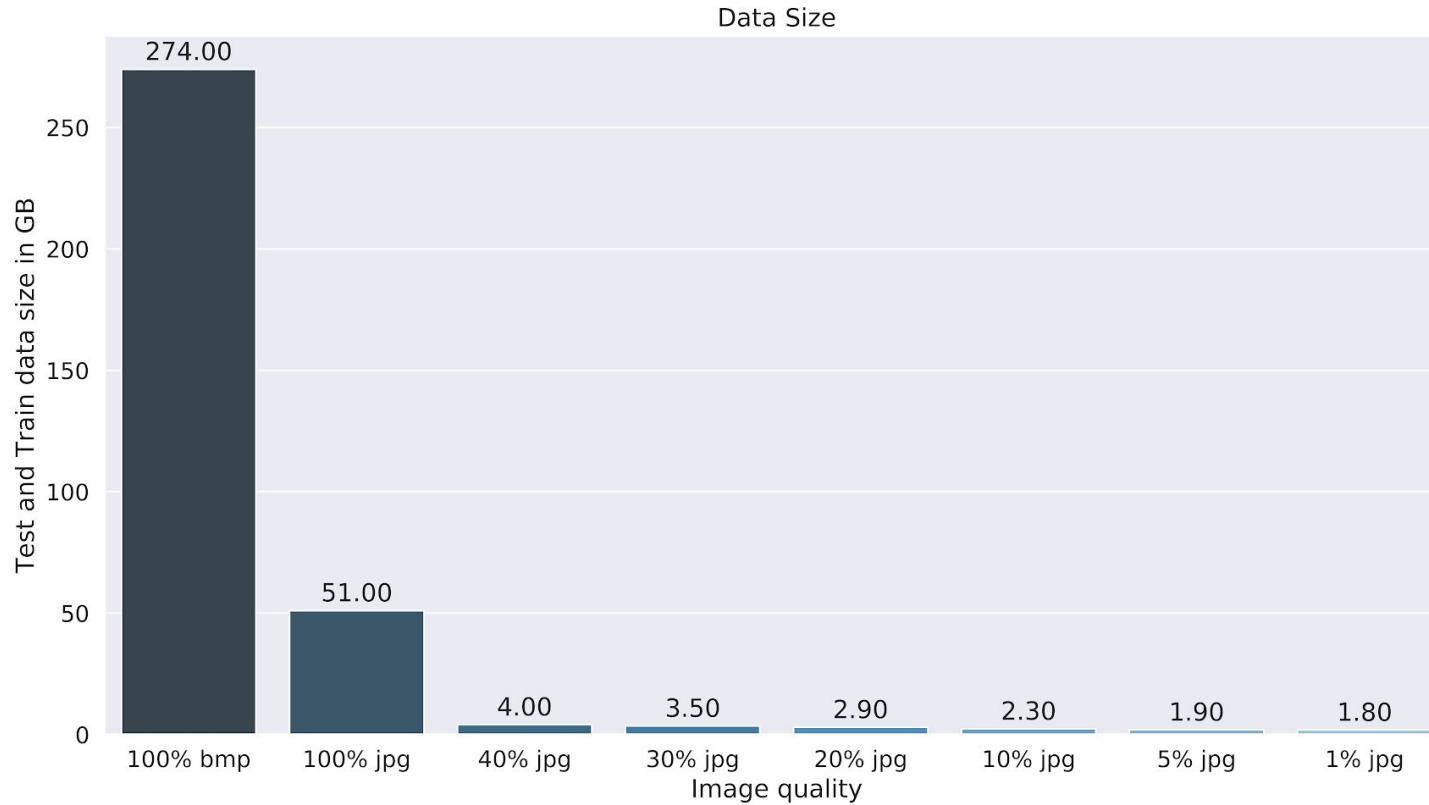


JPEG Compression

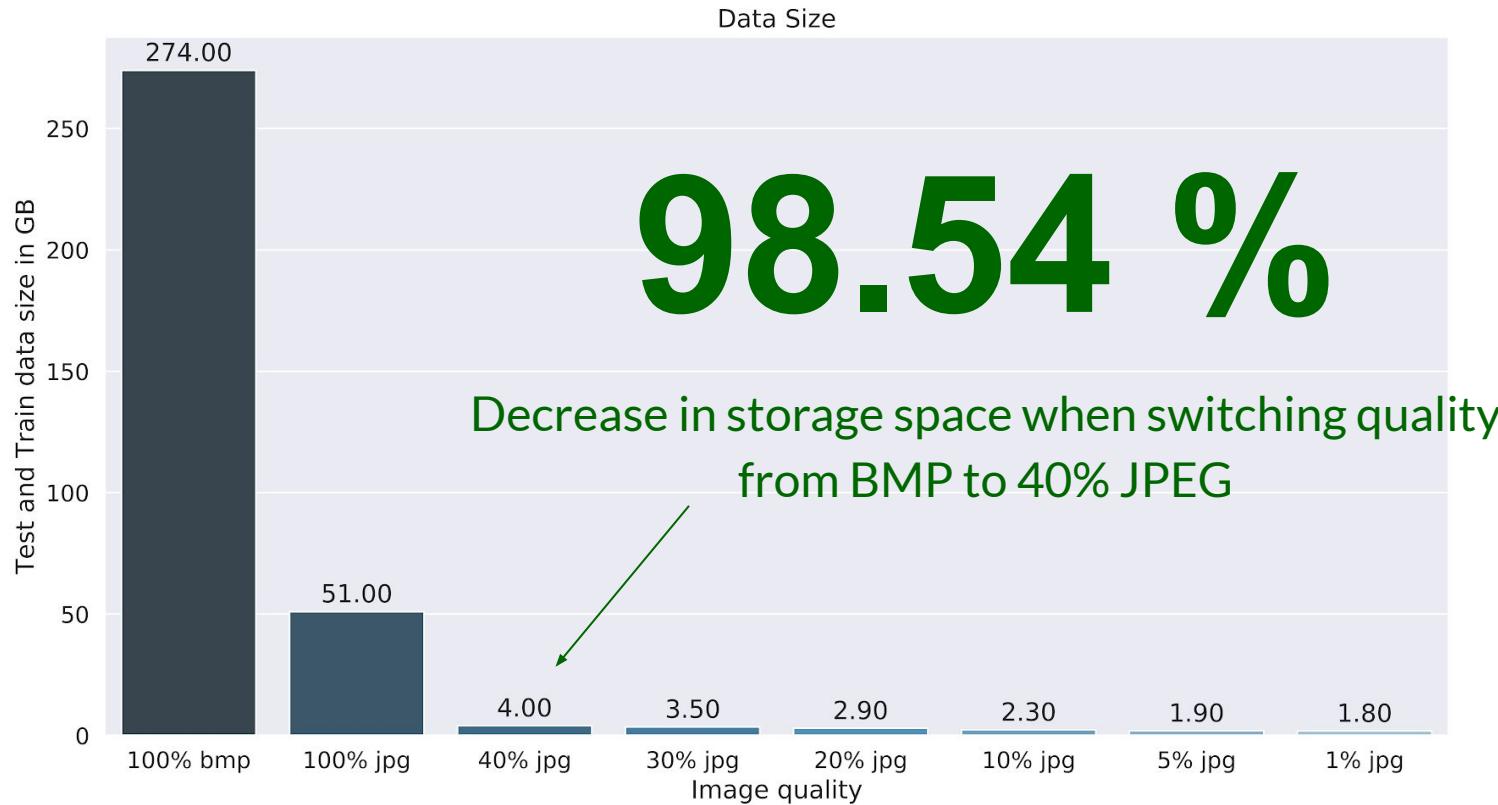
- **the most common** method of compression for digital images;
- uses a **lossy** form of compression based on the discrete cosine transform;
- “**lossy**” means that some original image information is lost and cannot be restored, affecting image quality.



Can you still see me? - Compression



Can you still see me? - Compression





Literature overview

- **Low resolution [Koziarski and Cyganek, 2018]**
 - **Networks:** AlexNet, VGGNet, and ResNet
 - **Results:**
 1. Final accuracy decreases with picture size
 2. Applying super-resolution allows achieving classification accuracy close to that with original size
- **Image distortion [Dodge and Karam, 2016]**
 - **Image transformation:** blur, noise and contrast
 - **Results:**
 1. NN accuracy dropped dramatically
 2. NN rely heavily on textures in object classification, which get erased after transformations

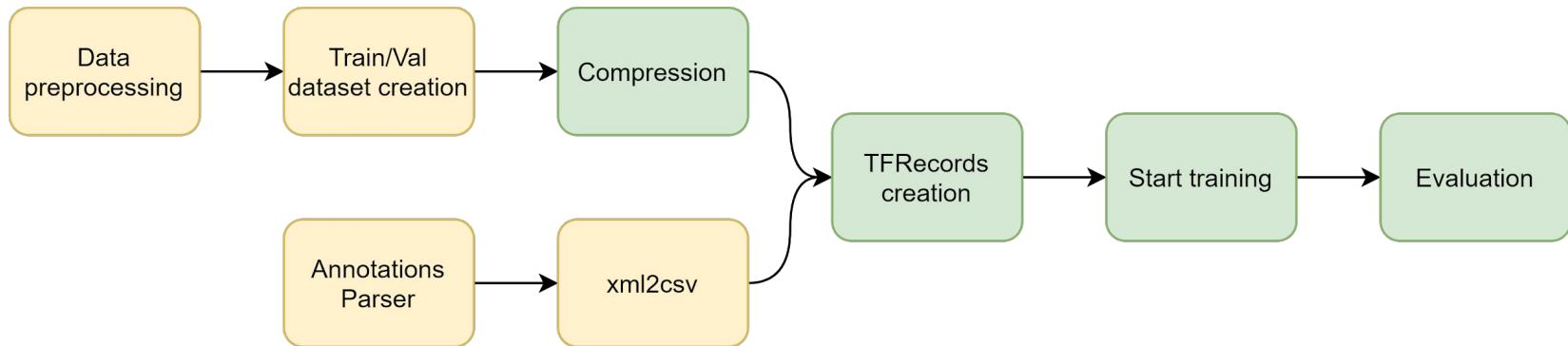


Literature overview

- **Image compression**
[Zanjani et al., 2019], [Dodge and Karam, 2016], [Dejean-Servieres et al., 2017]
 - **Compression:** JPEG
 - **Networks:** different variations of VGGNet, AlexNet
 - **Results:** pictures with similar-to-original quality have close values of target metrics
 - **[Zanjani et al., 2019] experiments:**
 - Training on uncompressed data, testing on compressed
 - Training and testing on compressed data -> better results



Solution Pipeline



Main tools: Python 3.7, TensorFlow 1.12, VCS Github



Transfer Learning

Used pre-trained tensorflow models:

- SSD inception v2 coco
 - SSD refers to the architectures that use a single feed-forward CNN to directly predict classes and anchor offsets without requiring a second stage per-proposal classification operation
- Faster RCNN
 - Detection happens in 2 stages:
 - Region Proposal Network: Predicts class-agnostic box proposals
 - Predict a class and class-specific box refinement for each proposal
 - Resnet-101
 - Inception V2



Model settings

- **SSD inception v2 (COCO Dataset)**
 - 300x300, 600x600
 - 12-14 hours per training
 - 128 batch size
 - 15.000 steps
- **Faster RCNN Inception v2 (COCO Dataset)**
 - 1820x940
 - 6 hours per training
 - 16 batch size
 - 20.000 steps
- **Faster RCNN ResNet101 (Kitti Dataset)**
 - 1820x940
 - 3 hours per training
 - 4 batch size
 - 20.000 steps

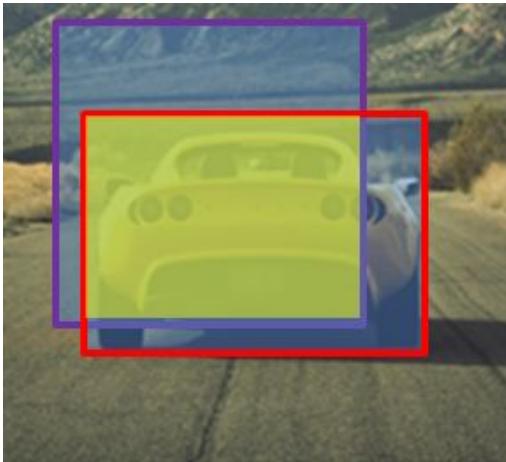
Evaluation metrics

- **Classification**
 - Precision, Recall, F1
- **Object Detection**
 - Mean Average Precision (mAP)
 - Weighted mAP for imbalanced classes





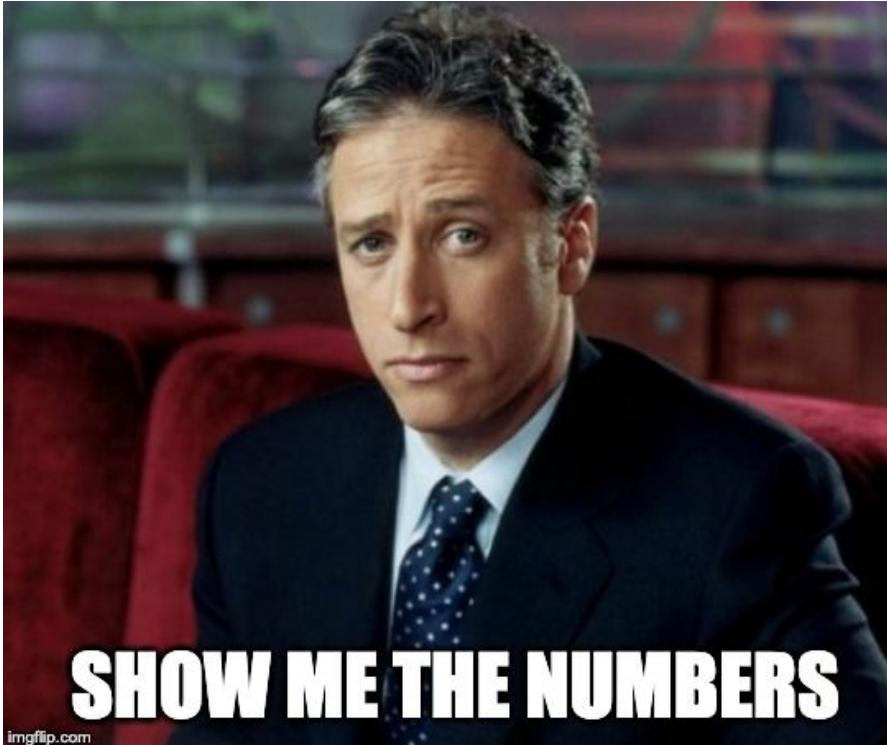
Intersection over Union



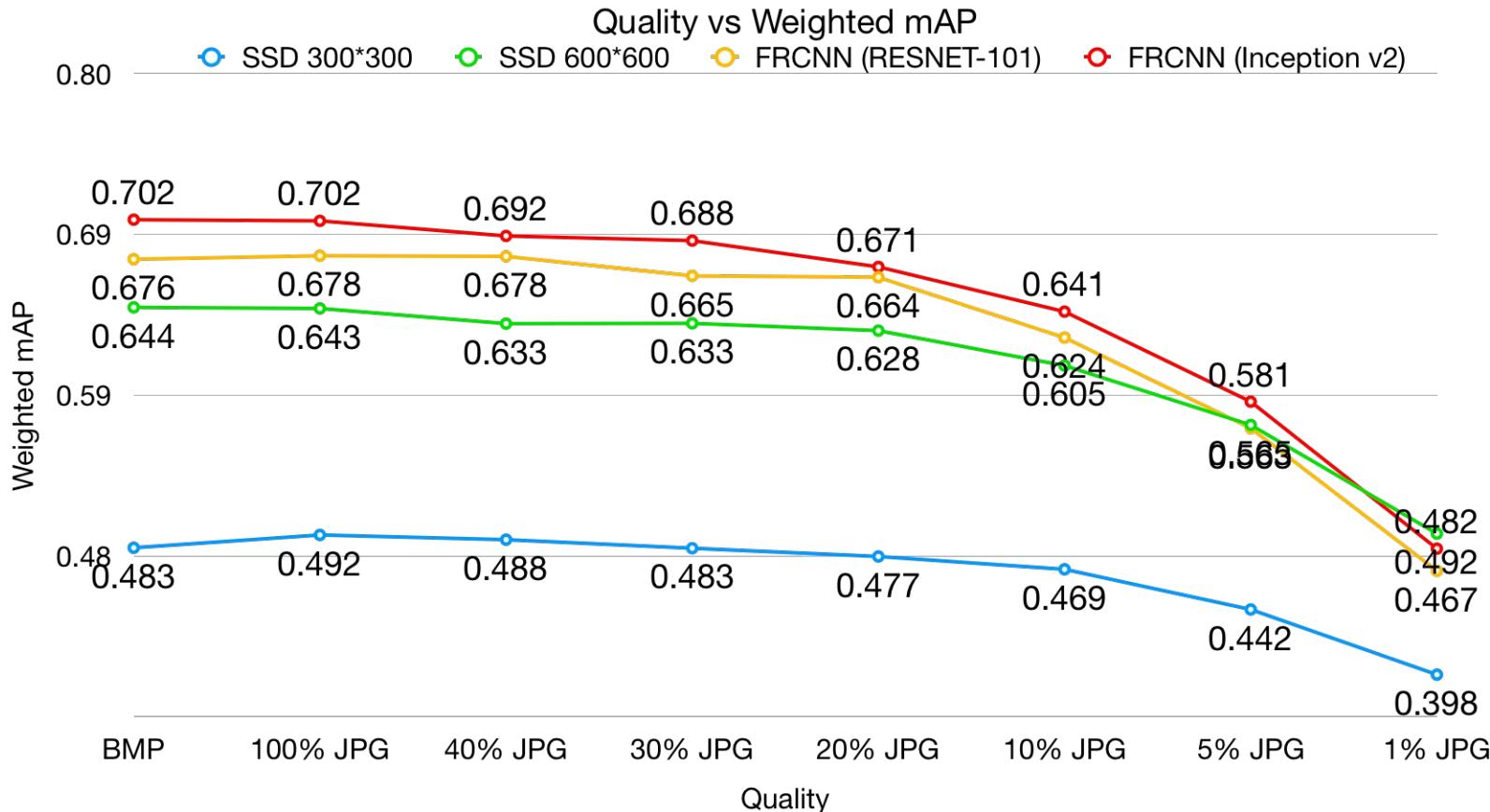
$$IoU = \frac{\text{Area of Intersection}}{\text{Area of Union}}$$

A prediction is **correct** if $IoU \geq 50\%$

Can you still see me? - Results



Can you still see me? - Results





Compression of grayscale images

- **Idea:** compression of grayscale image will better preserve some edge features for lower qualities
- Experiments on 40% and 5% quality data with SSD 600x600

Can you still see me? - Results



40%



Can you still see me? - Results



5%





Did grayscale help?

| SSD at 600*600 | Color image | Grayscale image |
|----------------|-------------------|--------------------|
| 40% | 0.632856 (4 GB) | 0.635504 (3.1 GB) |
| 5% | 0.565084 (1.9 GB) | 0.559135 (0.75 GB) |

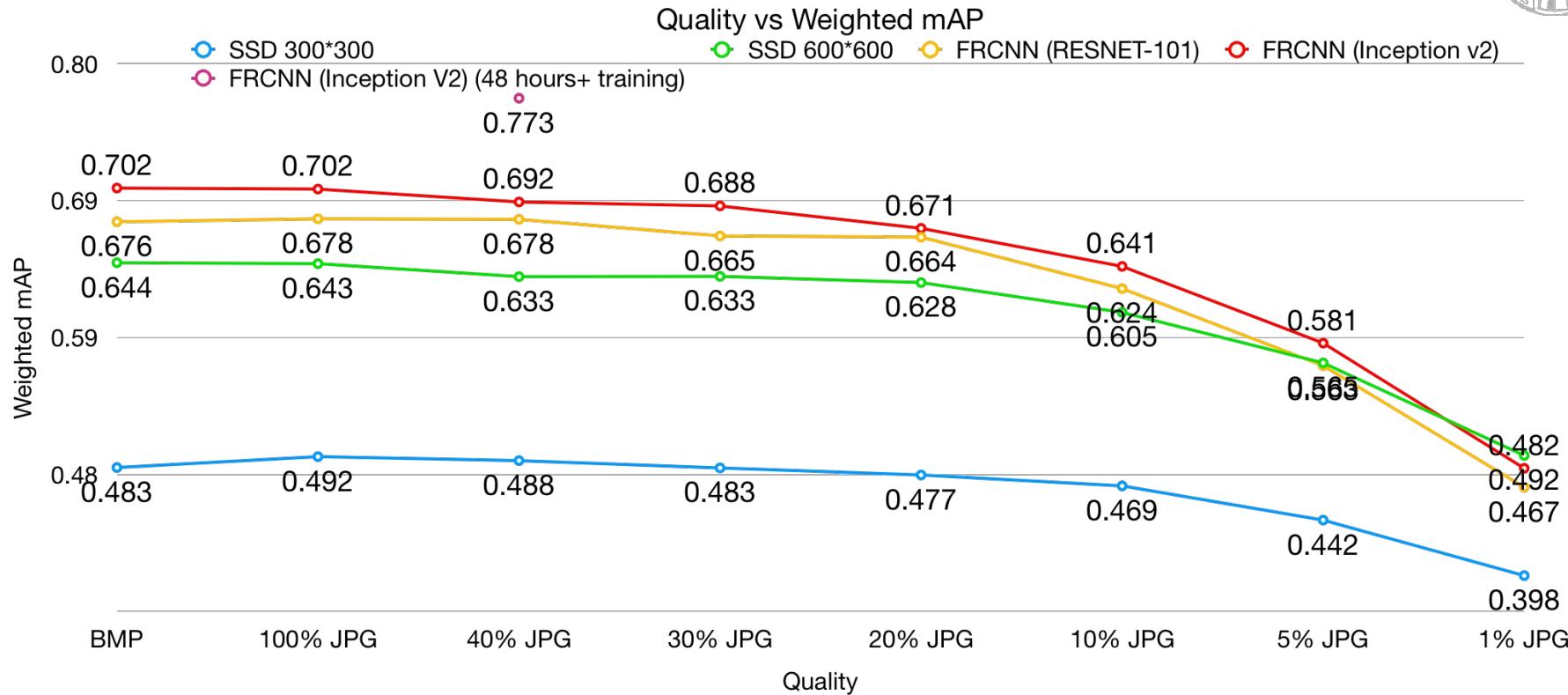


Did increasing the steps help?

| | | |
|-----------------------------|---------------|----------------|
| FRCNN (Inception V2) | 20.000 | 200.000 |
| Time taken | ca 6 hours | ca 48 hours + |
| Weighted mAP | 0.6915 | 0.772963 |

an increase of 8%

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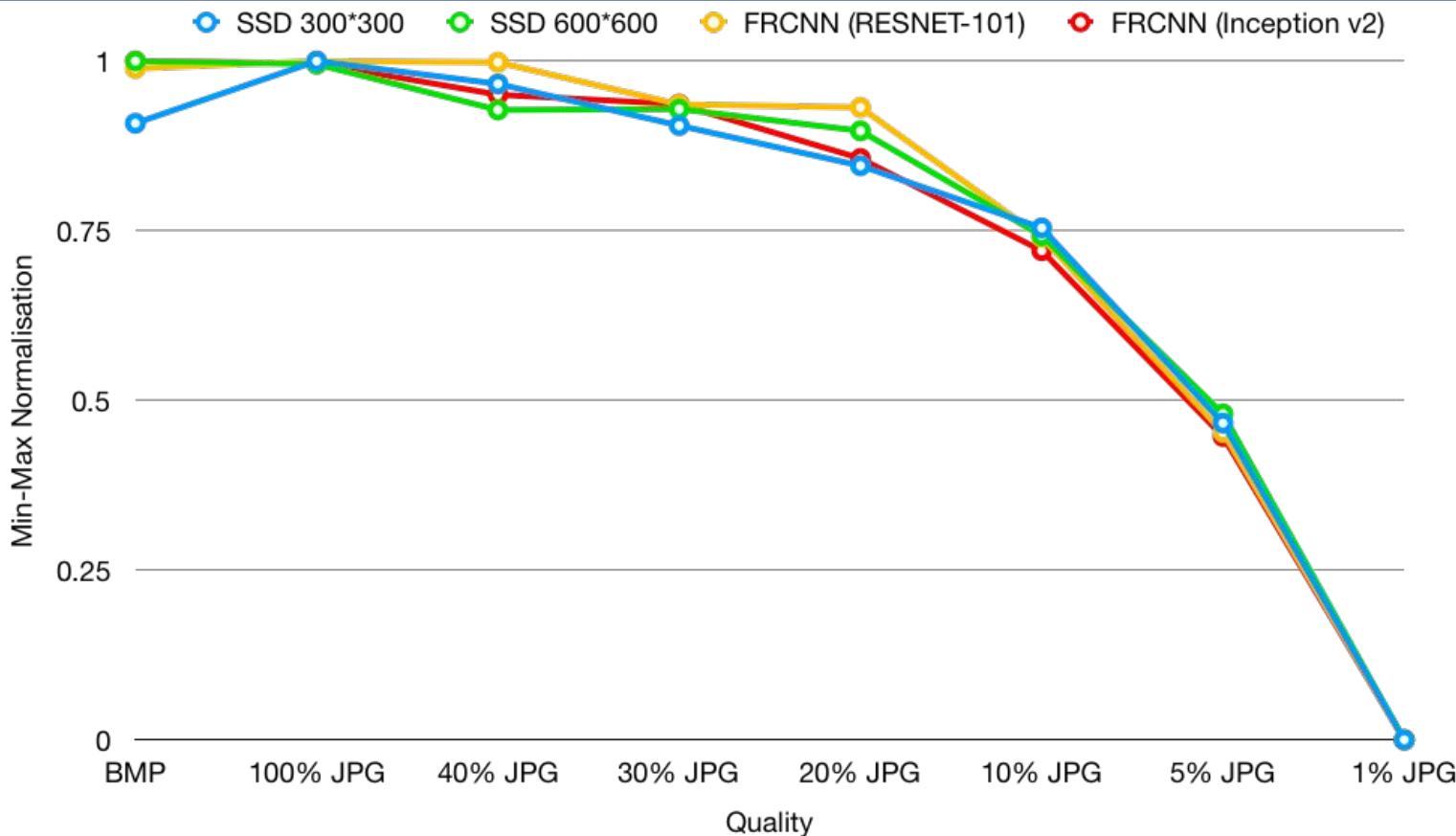




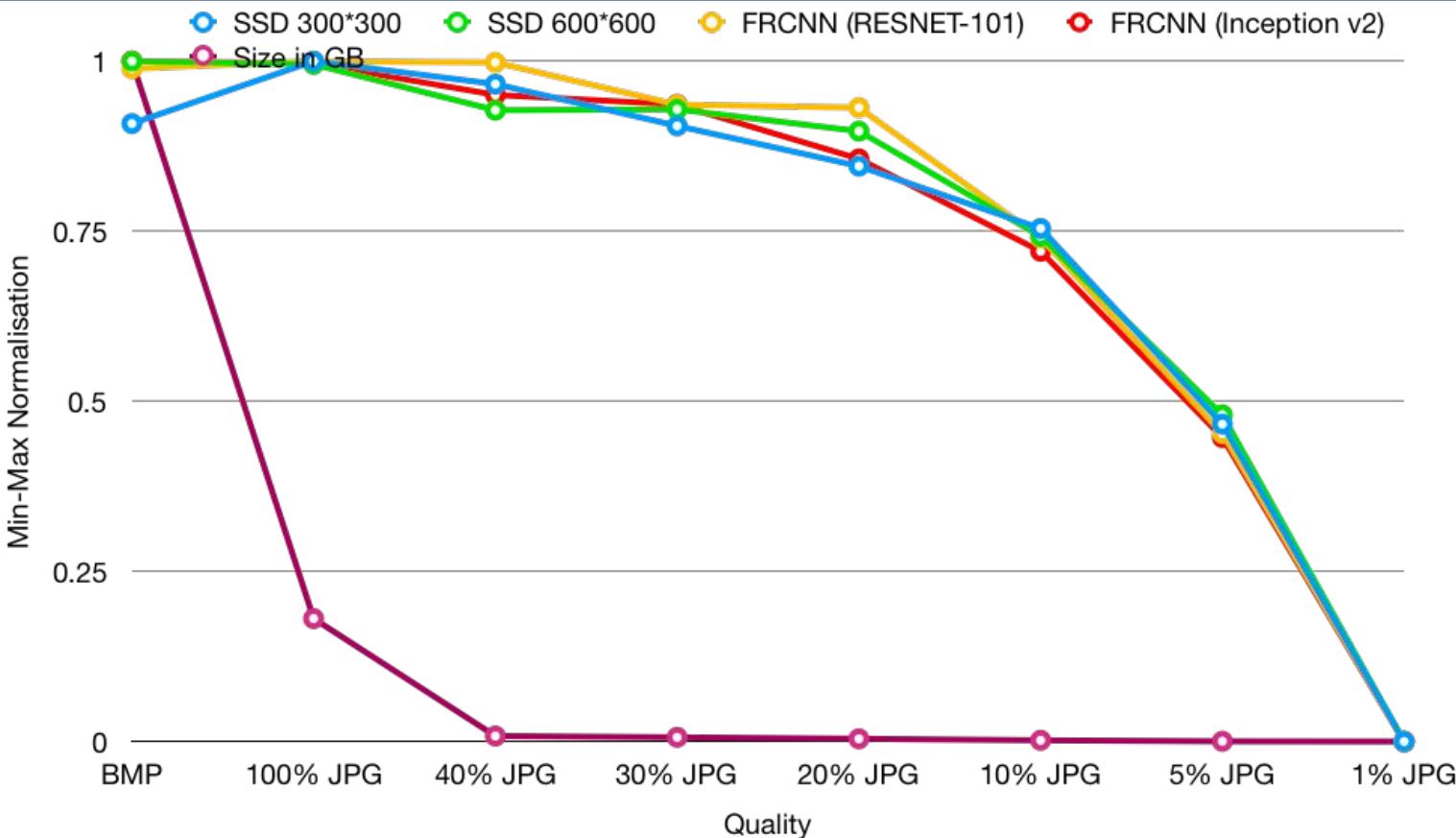
Comparing performance

We performed min-max scaling on each result, in order to normalise the data on 0-1 scale

Min-max Normalised Results



Min-max Normalised Results



Can you still see me? - Results



Ground truth



Can you still see me? - Results



100% FRCNN

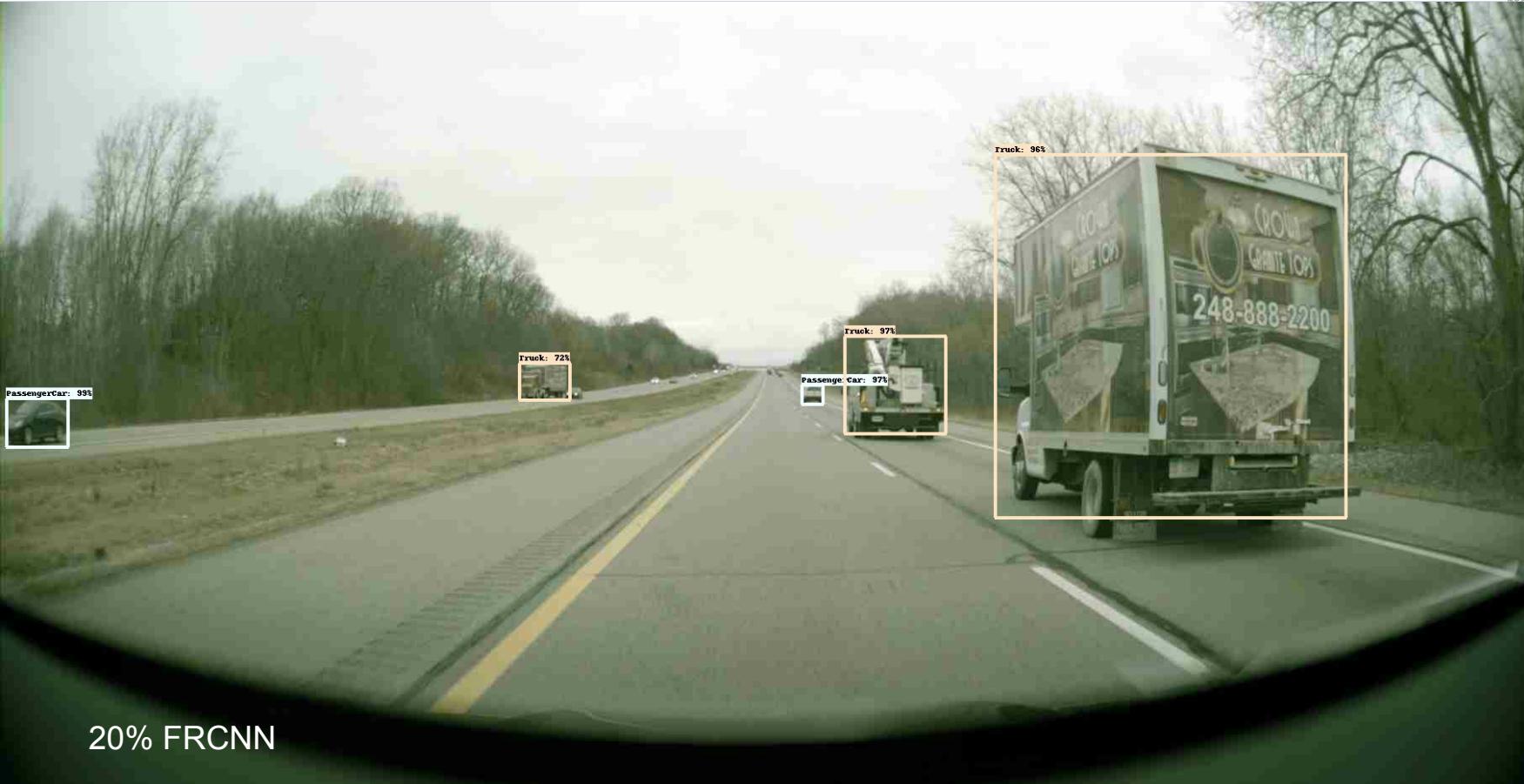
Can you still see me? - Results



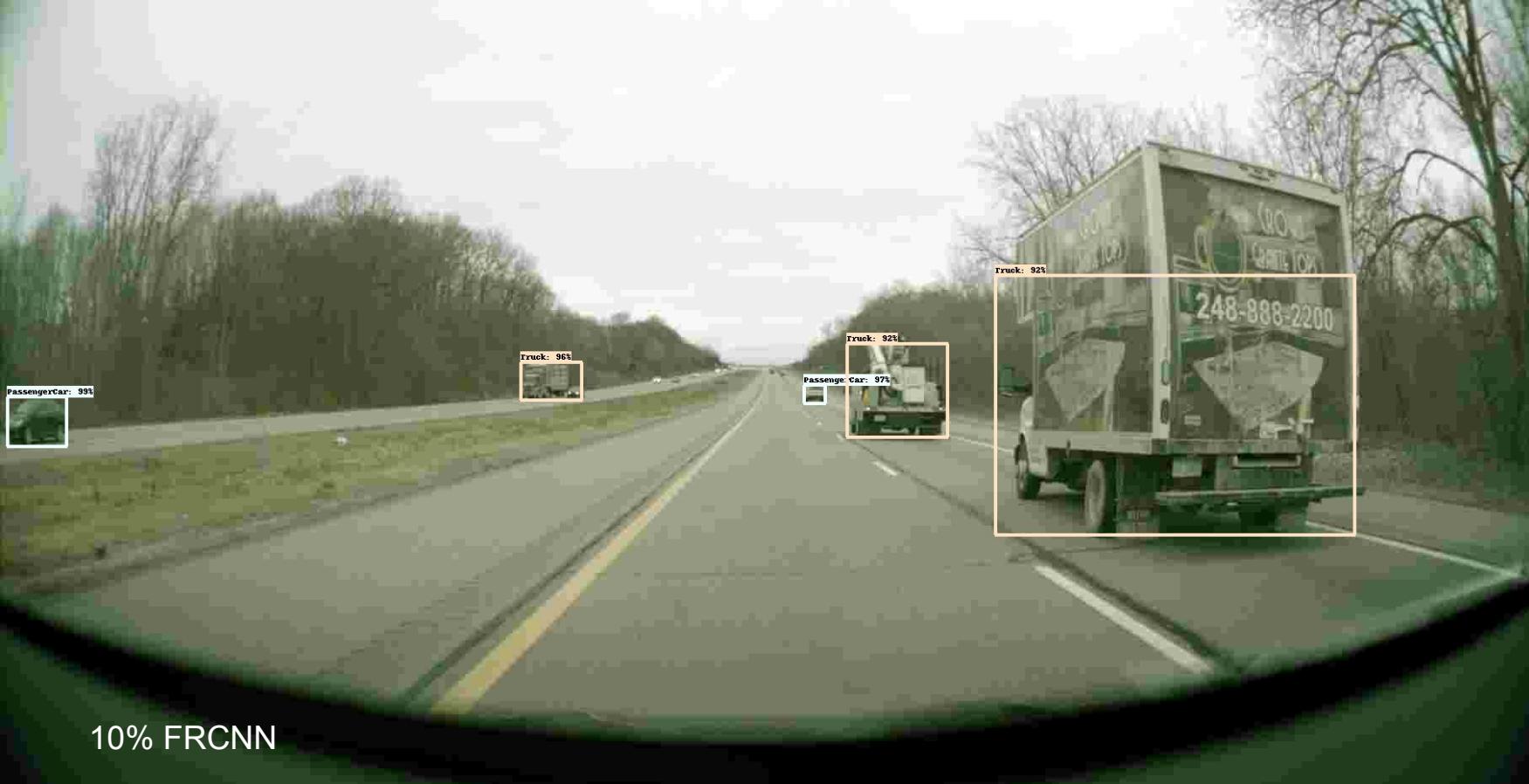
Can you still see me? - Results



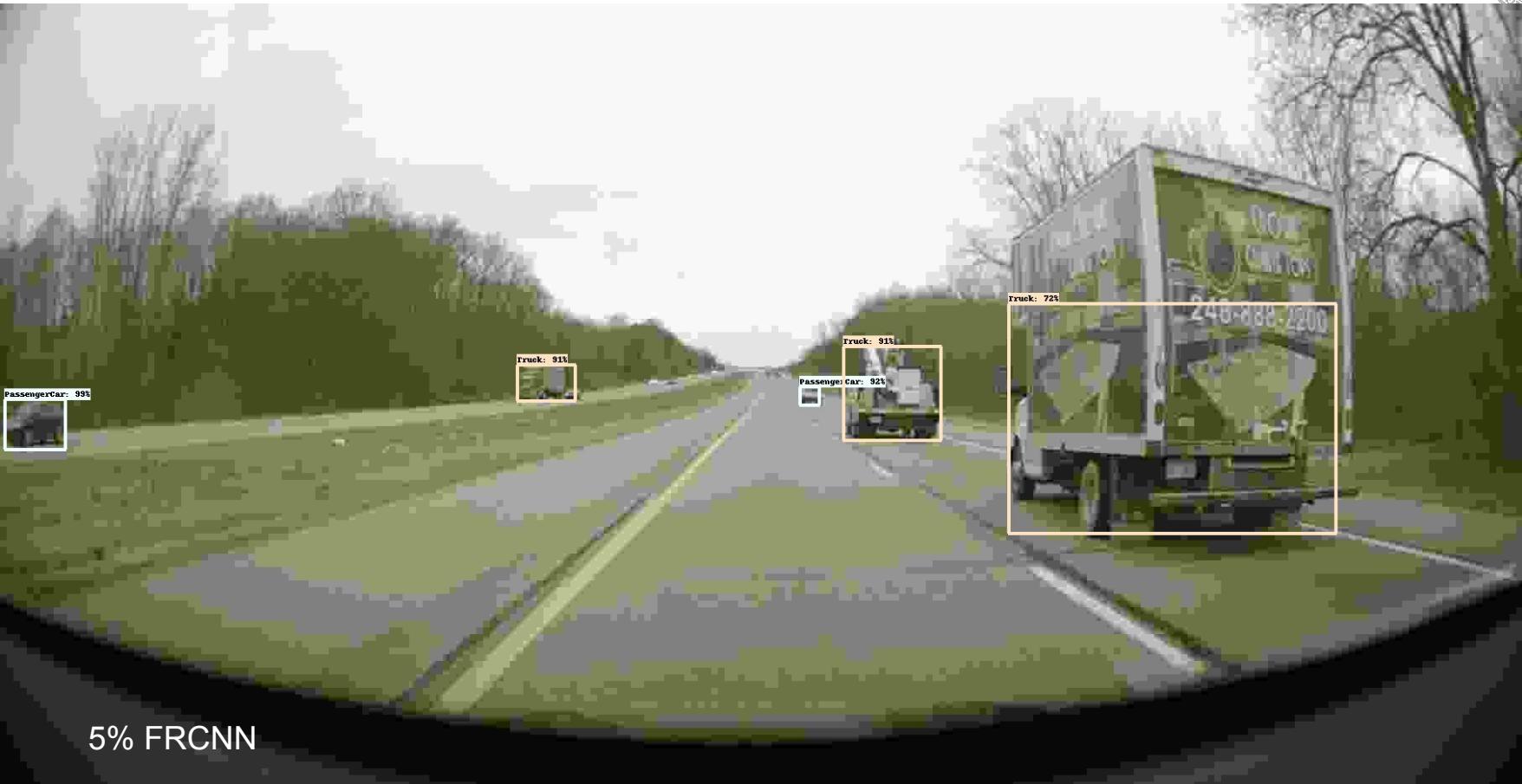
Can you still see me? - Results



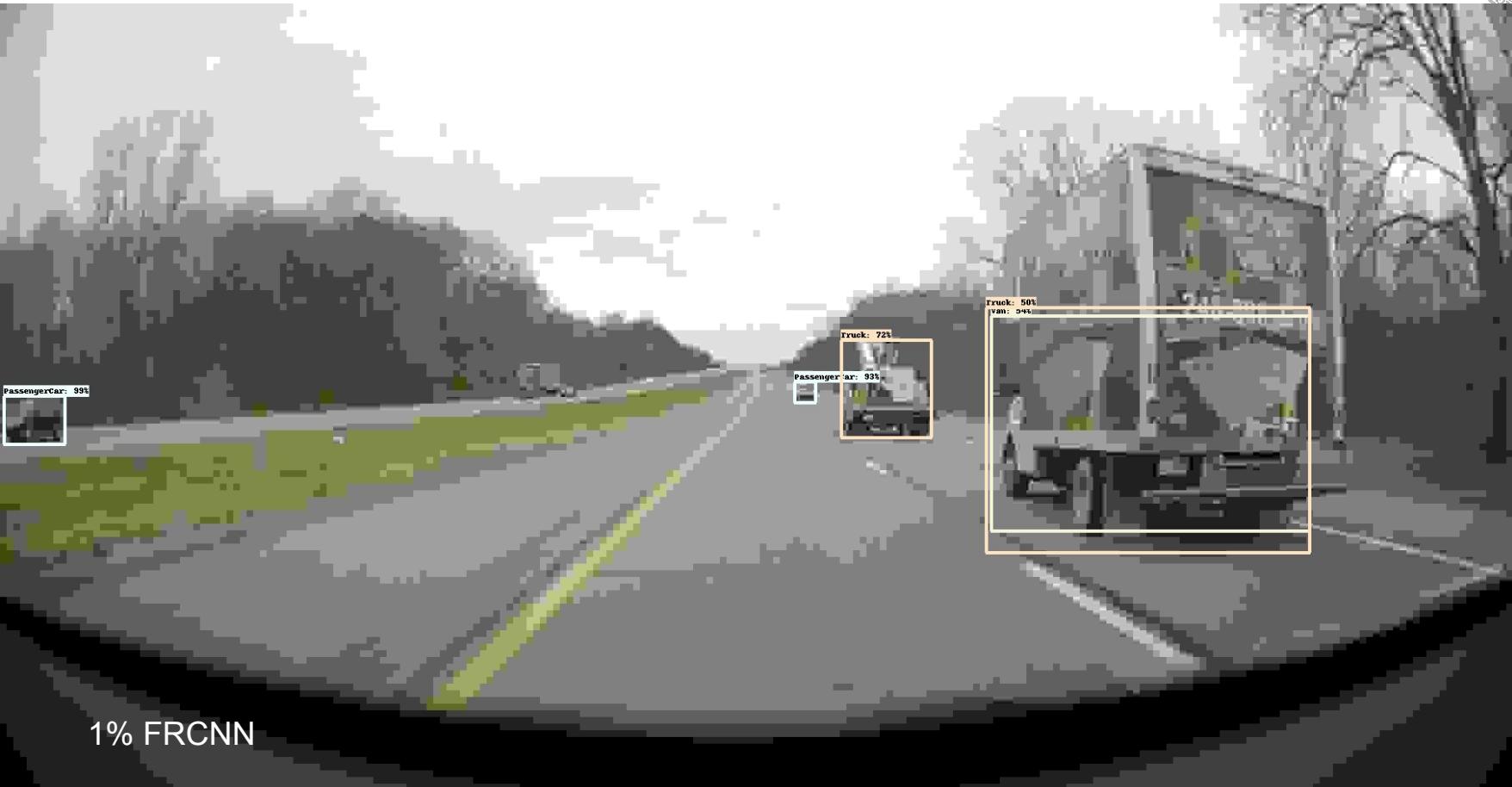
Can you still see me? - Results



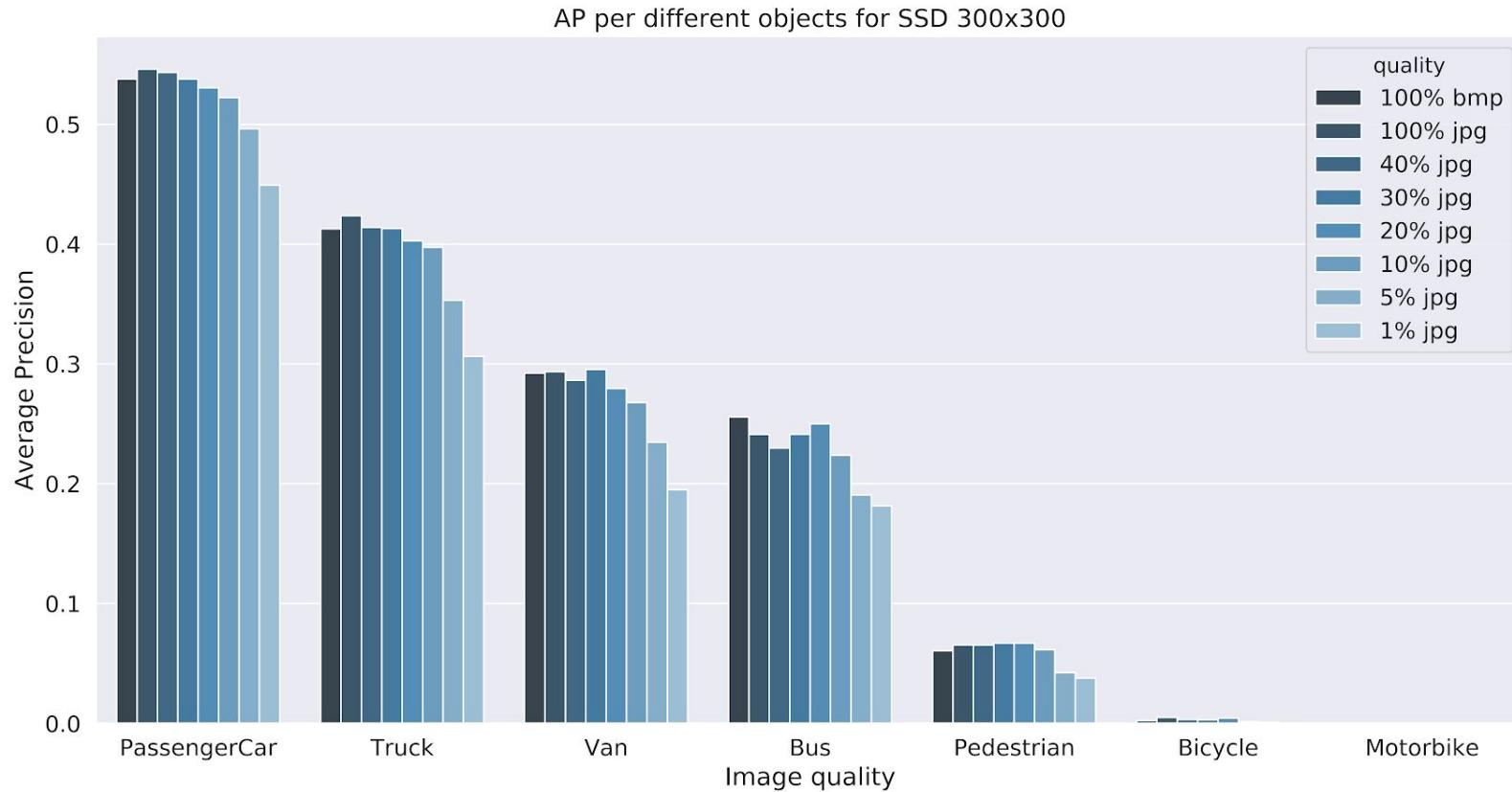
Can you still see me? - Results



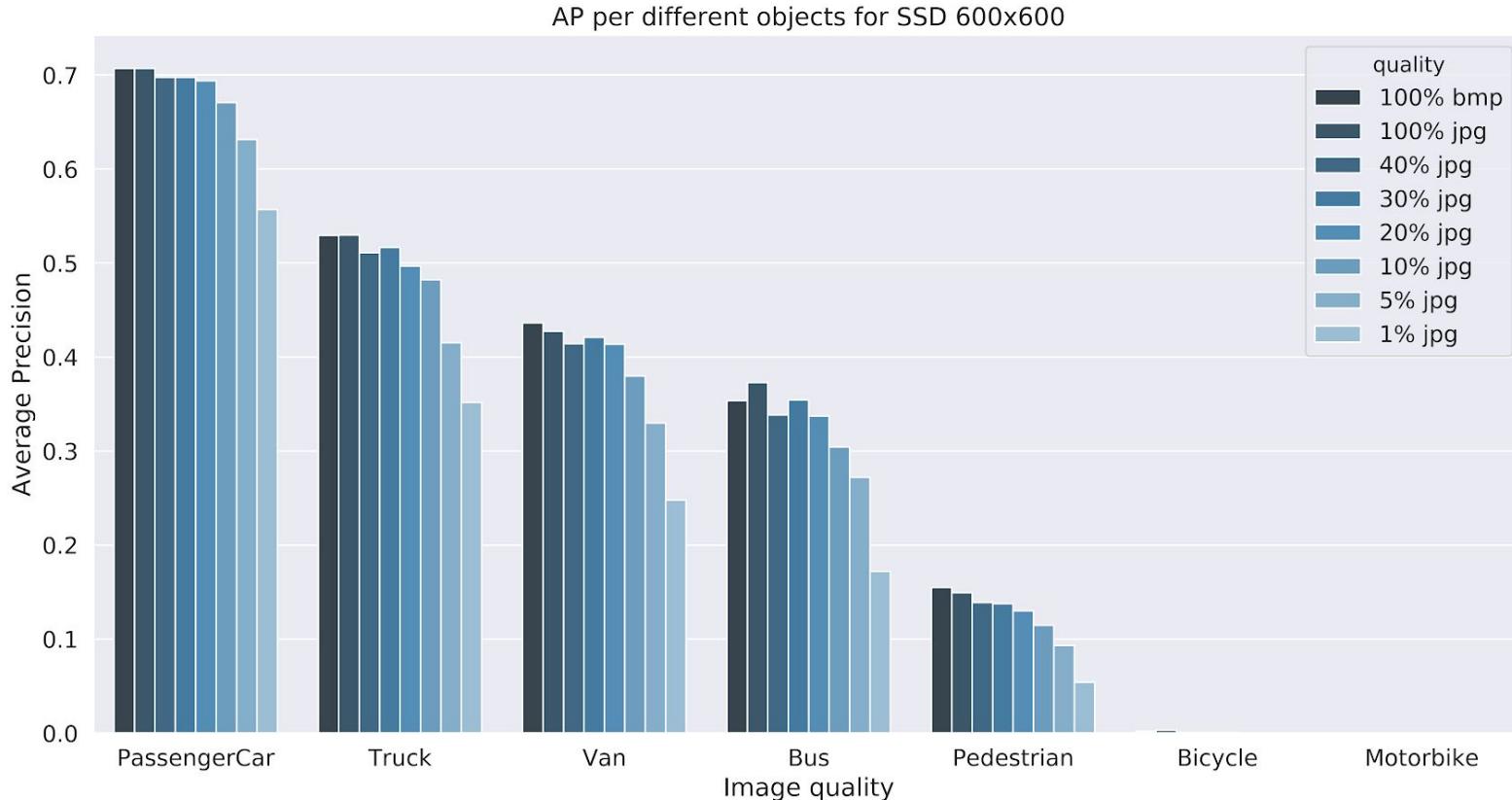
Can you still see me? - Results



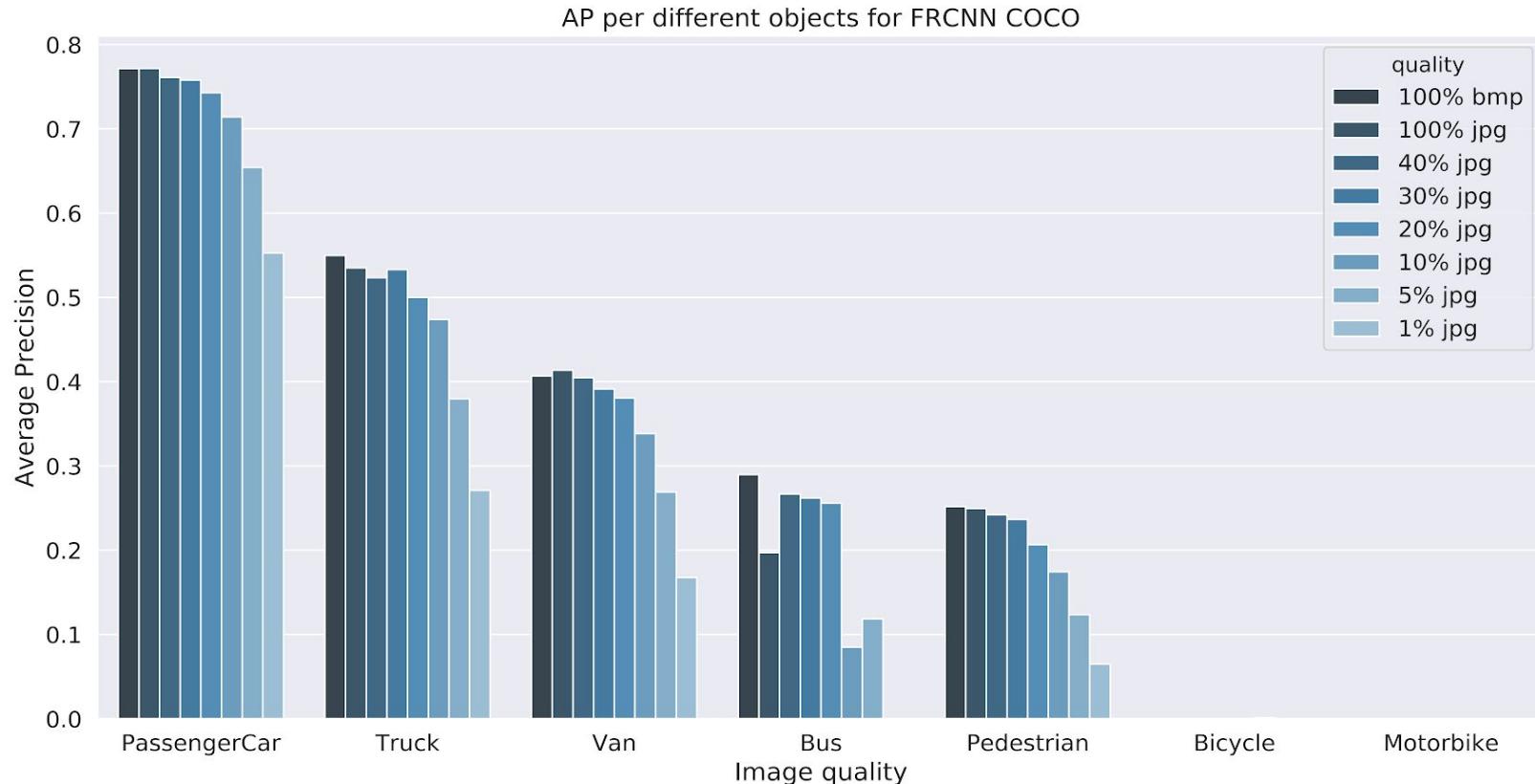
Can you still see me? - Results



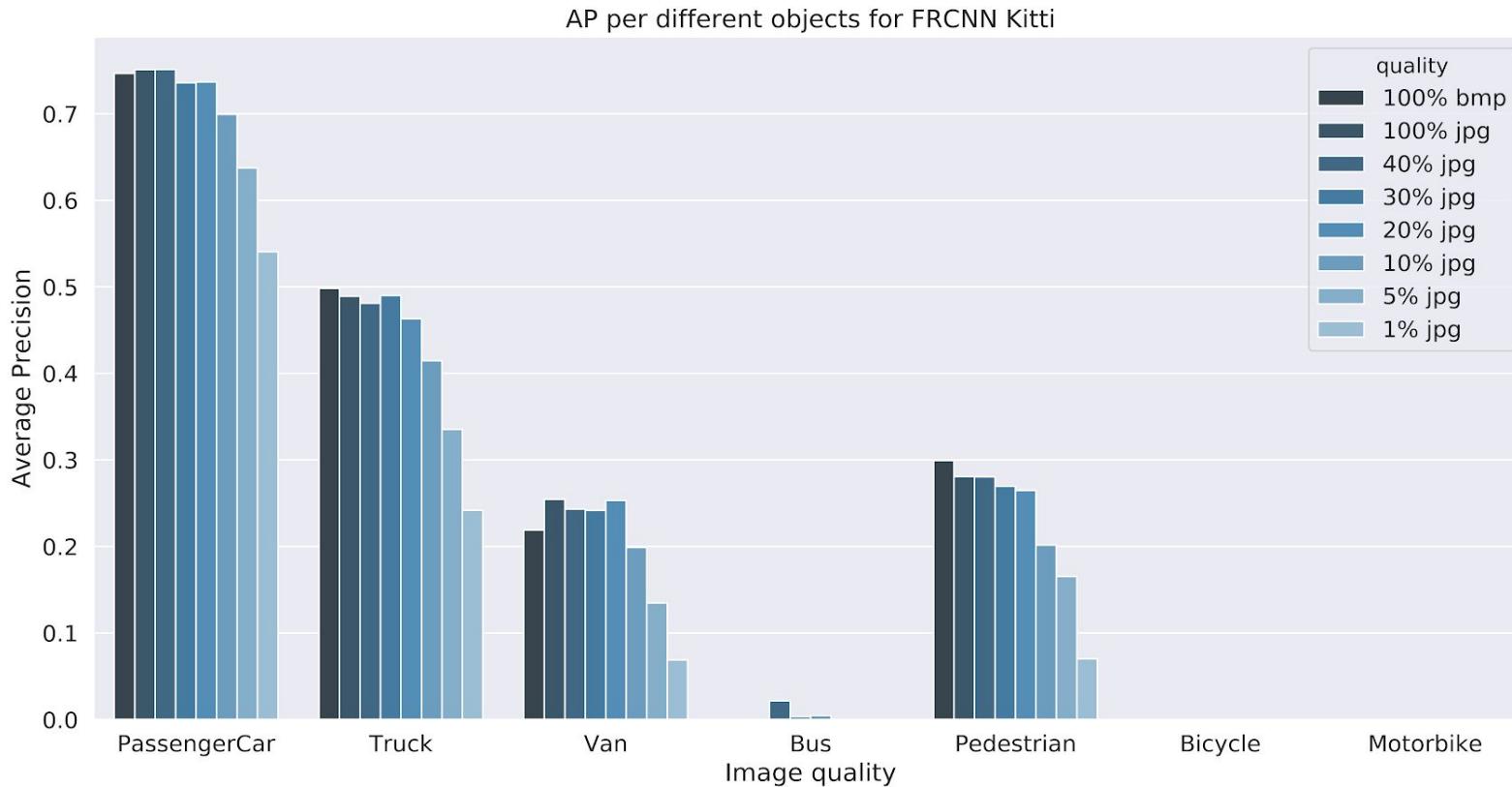
Can you still see me? - Results



Can you still see me? - Results



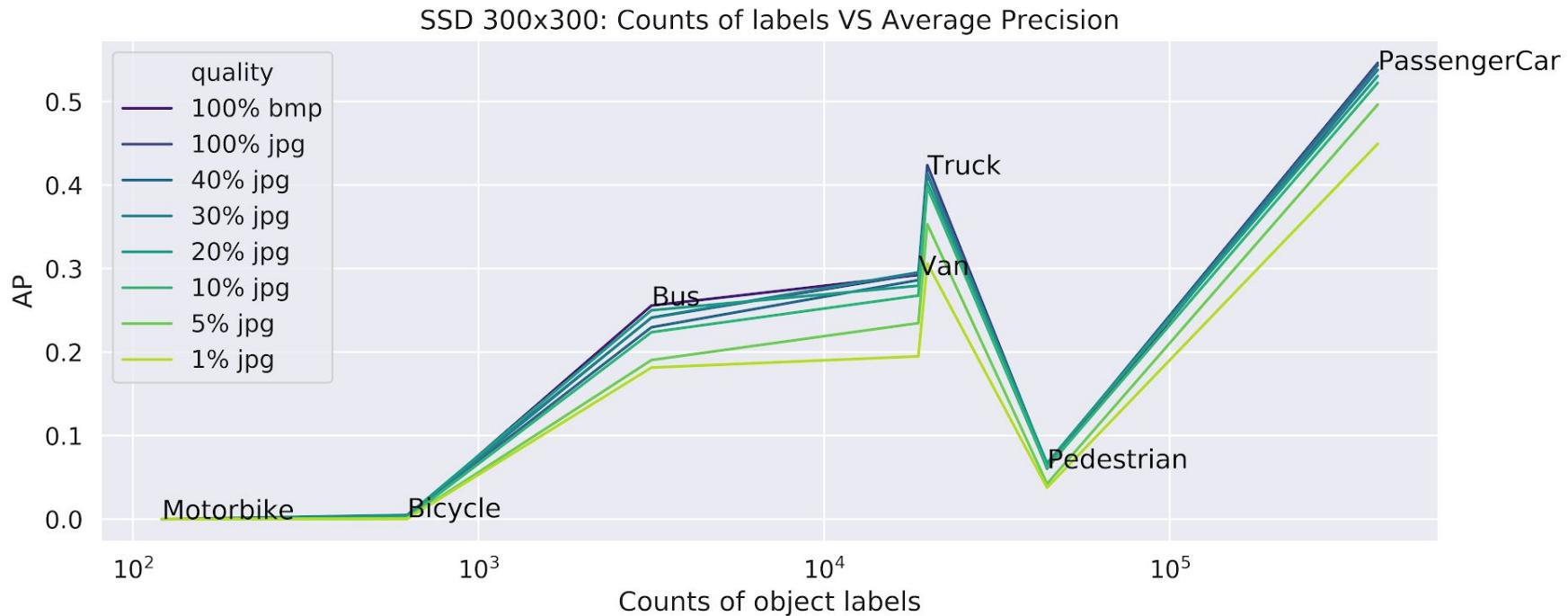
Can you still see me? - Results



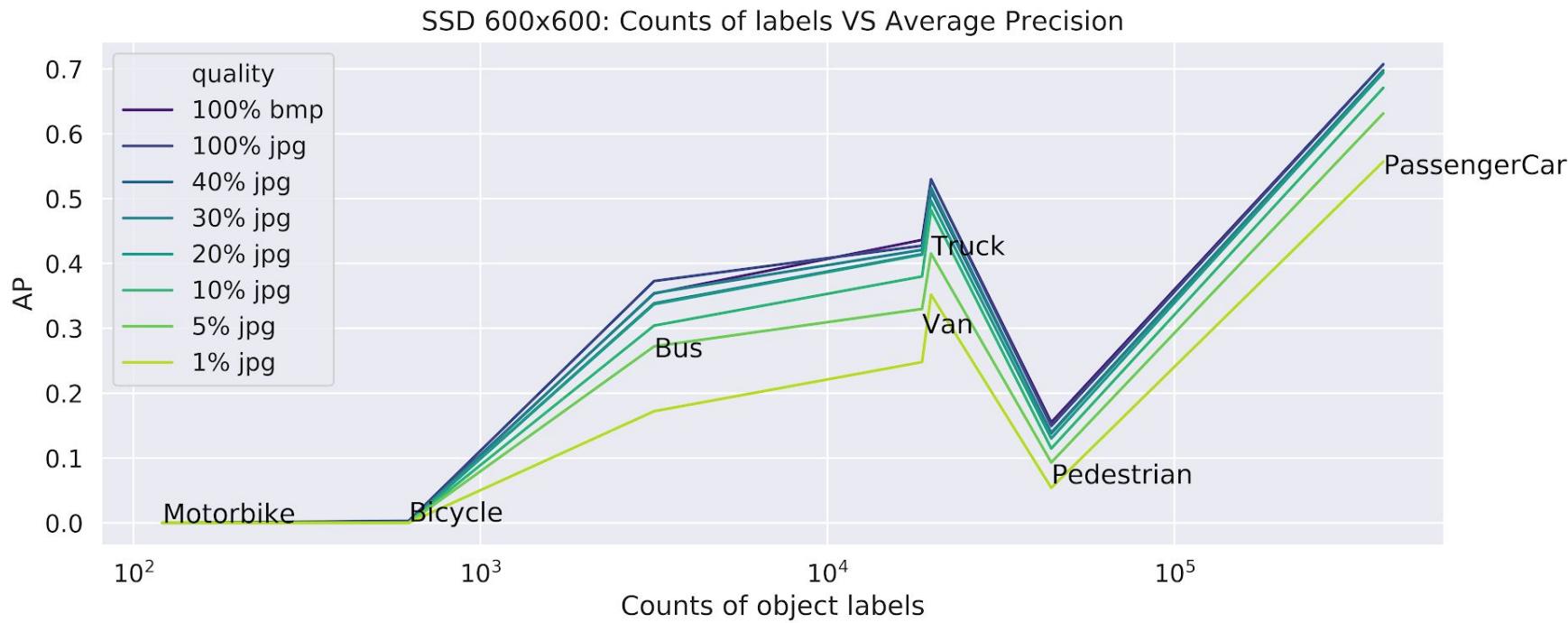


Performance insights

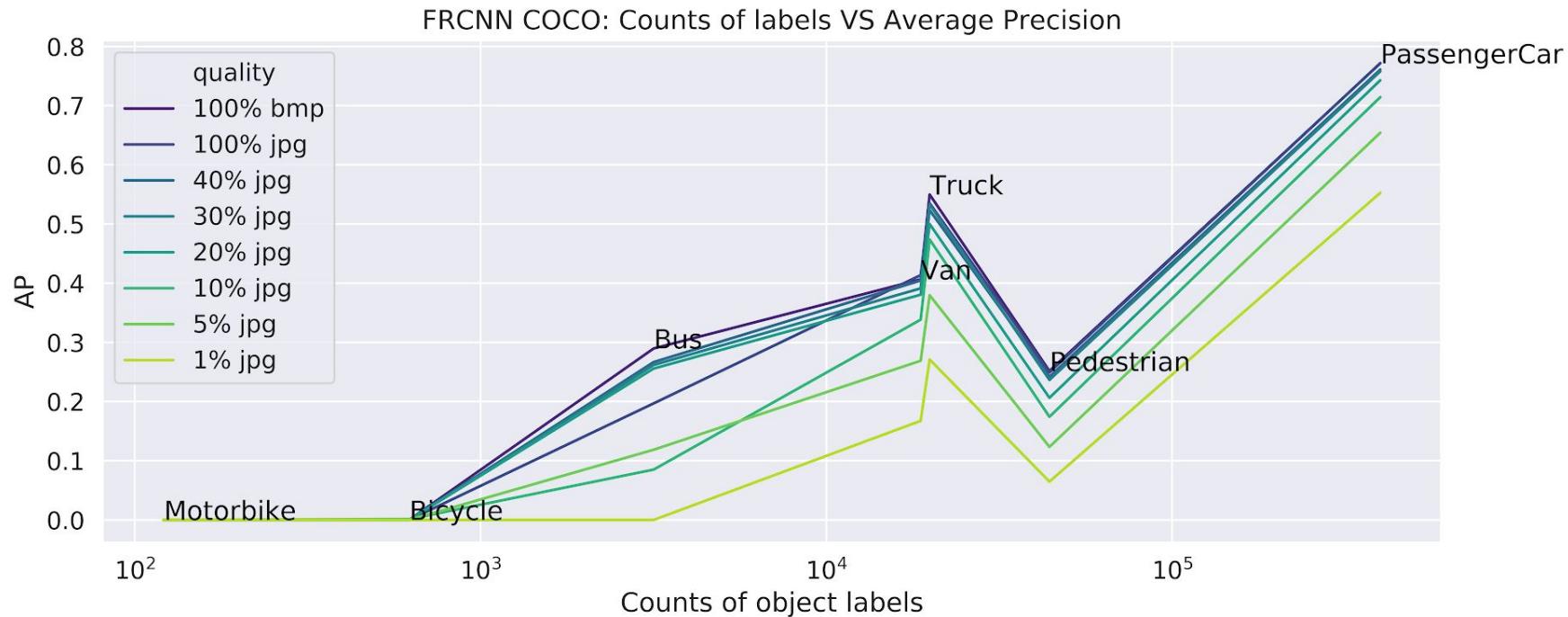
Can you still see me? - Results



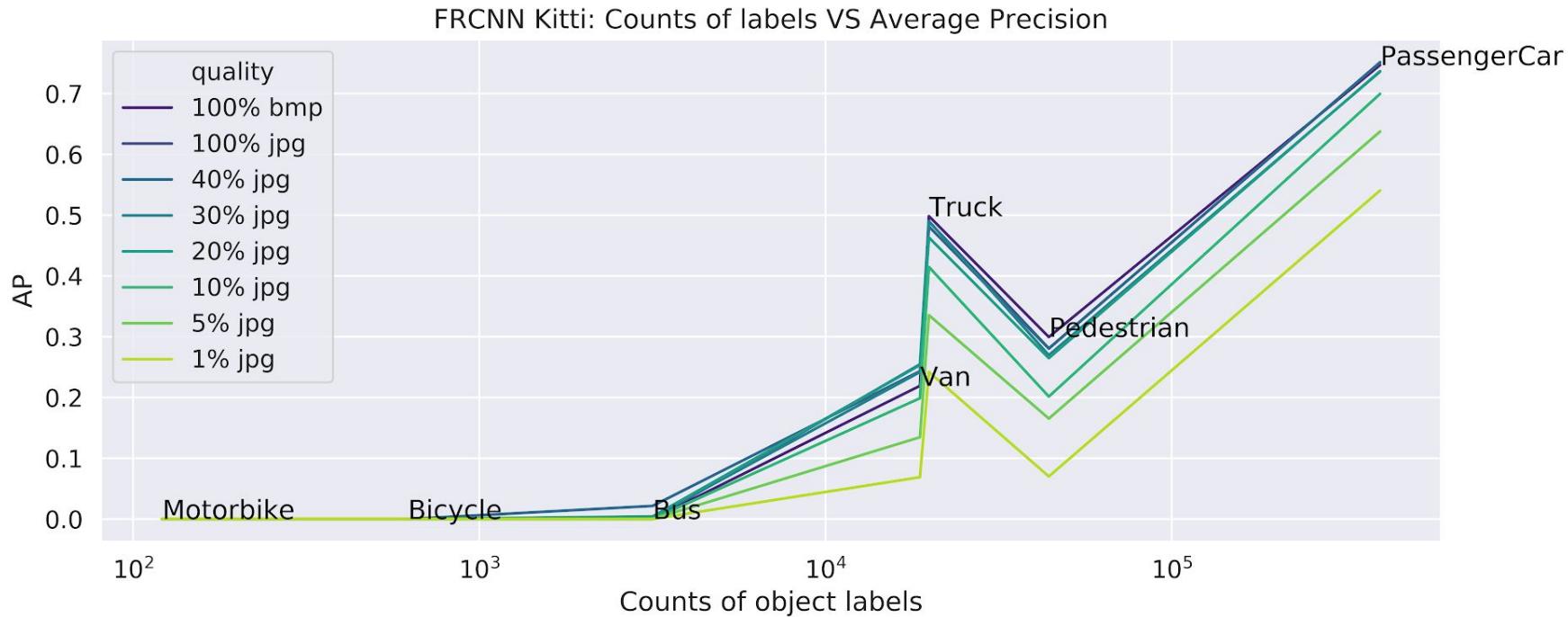
Can you still see me? - Results



Can you still see me? - Results

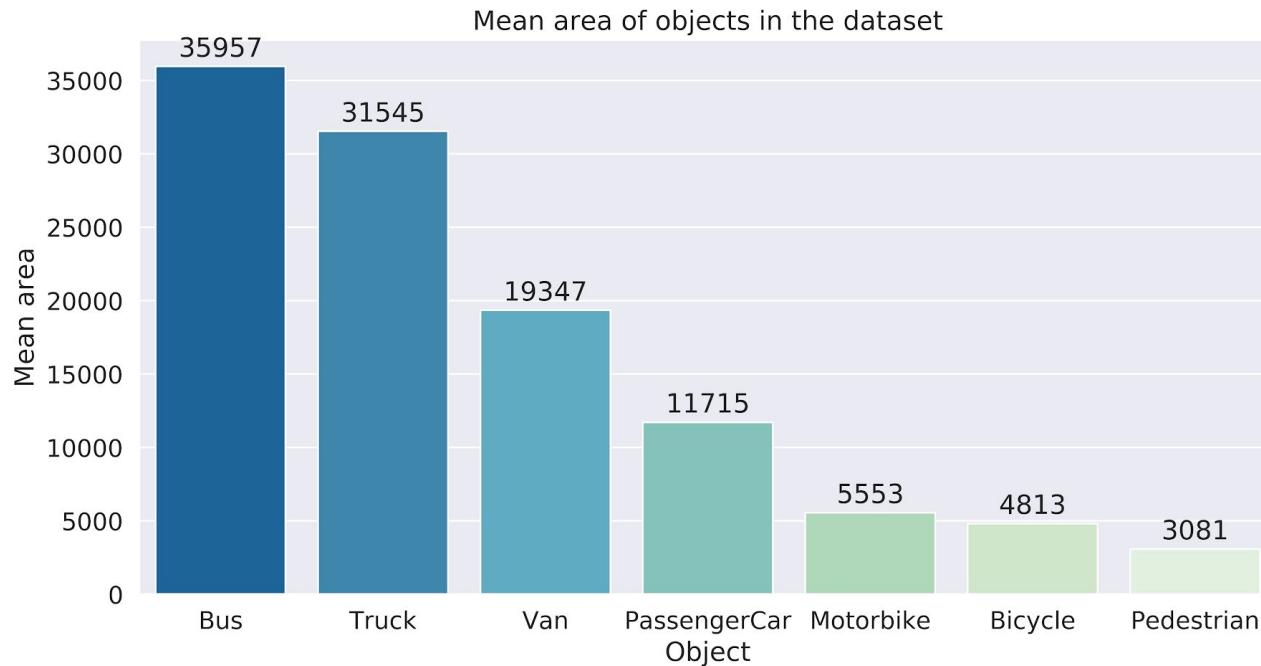


Can you still see me? - Results



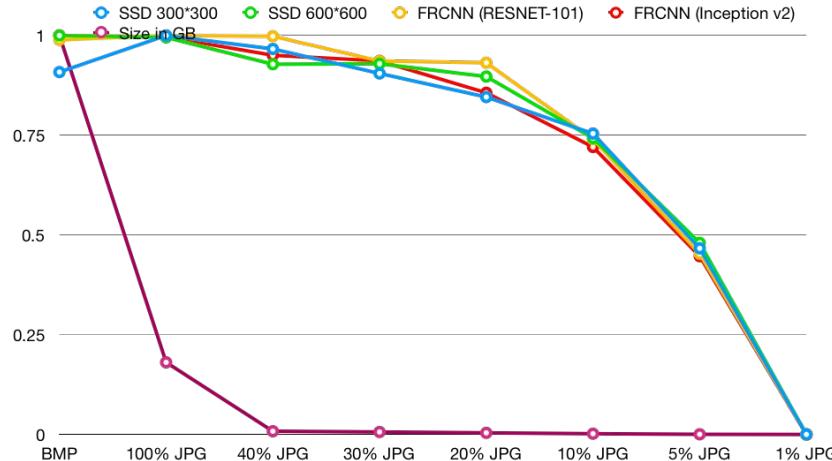
Even though pedestrians are more in number, they are small, so detected worse, than truck.

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While 40% and 30% jpg images show no significant difference in **precision** as compared to 100%, they bring huge drop in **storage size**.





- Research of other compression techniques, especially compression by neural nets, that preserve better features in the context of object detection;
- More experiments and more fine-tuning of the models
- Cross-validation
- Developing an NN architecture and training from scratch with more data
- Bigger batch size
- Uniform batch creation / Obtaining more data for classes like “Motorbike” and “Bus”



So that was it.
Thank you for coming.

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Can you still see me? - Reference









