Vehicle Detection and Classification from Images

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Introduction

Reminder of Our Project Goals



Detection

Progress in the Detection Task



Detection

- A vehicle/non-vehicle SVM classifier is trained on HOG features
- Detection is done with sliding windows

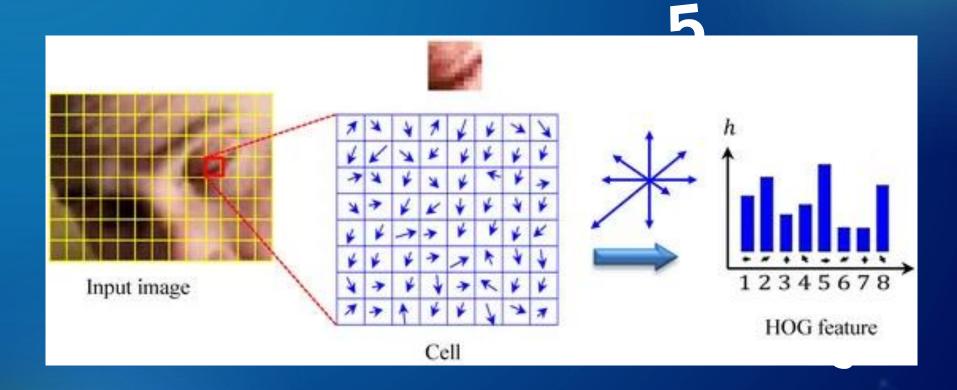
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HOG Features

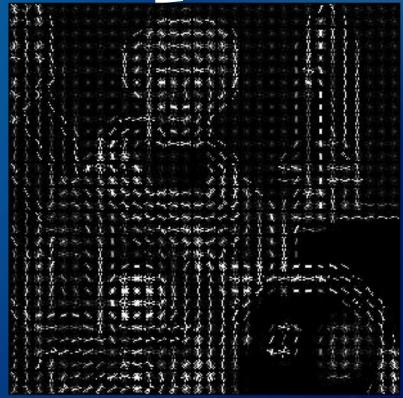


HOG Features

Input image



Histogram of Oriented Gradients



Dataset

Non-Vehicles















Vehicles















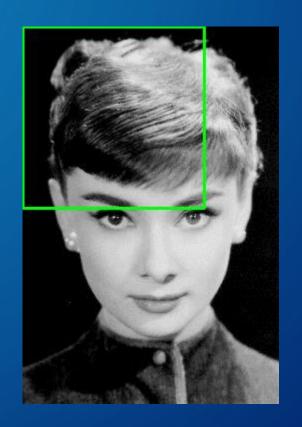
- 64x64 images
- 56201 vehicles
- 181788 non-vehicles

Model training

- An SVM with rbf kernel and 576 parameters is trained.
- Training takes ~10 minutes

	precision	recall	f1-score	support
vehicle	0.89	0.73	0.81	39253
non-vehicle	0.92	0.97	0.95	127340
accuracy			0.92	166593
macro avg	0.91	0.85	0.88	166593
weighted avg	0.92	0.92	0.91	166593

Sliding Window



Sliding Window



Classification

Progress in the Classification Task



- Dataset from TAU Vehicle Type Recognition
 Competition on Kaggle
 (https://www.kaggle.com/competitions/vehicle/data)
- Normally, it consists of 17 classes, but only 6 are used (Motorcycle, Car, Bicycle, Van, Bus, Truck)
- Dataset can be widened by combining other datasets or data augmentation

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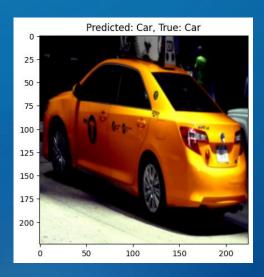
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- ResNet50 is chosen an experimental training has been done
- The model seems to be trainable with default parameters
- Next Steps:
 - Data augmentation
 - HP tuning for real training
 - Obtaining the metrics (Precision, recall, F1)

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Thanks!

Q&A

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