



Determining the accessibility of parking spaces in unfavorable conditions

FINAL PROJECT OF INTRO2CV



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- 1. Introduction
- 2. Problem description
- 3. Dataset description
- 4. Computer vision pipeline
- 5. Method
- 6. Result
- 7. Conclusion



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INTRODUCTION

- Looking for a parking spot is a *common* and often stressful activity; it takes a large amount of time.
- Have to pay a high amount in parking fines.
- Imbalance between parking demand and availability because of the growth of automobile ownership.
- Consumes around one million barrels of oil every day and contributes to air pollution.
- ⇒ PROJECT: locate vacant parking spaces in parking lots based on bird's-eye view photographs acquired by cameras in unfavorable conditions.





What is the expected value?

- Graphical display of available parking spaces
- Center coordinates
- Occupation status

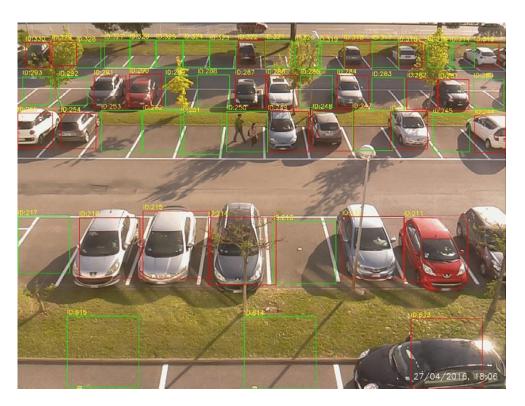






PROBLEM DESCRIPTION





Ideal case:

- There are parking markings
- The images of the cars are not distorted in any way



But what about real life?





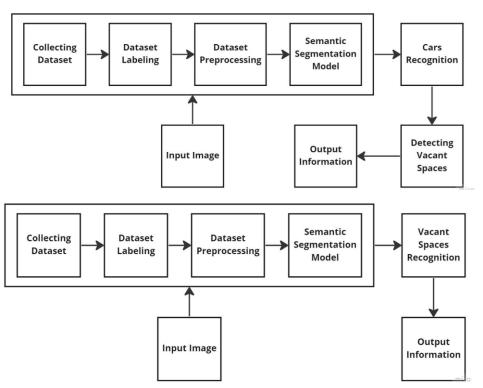
Real case:

- There are no parking markings
- Cars are covered with snow



Proposed methods





First method



Second method



Dataset - 1 DESCRIPTION





Dataset - 2 DESCRIPTION







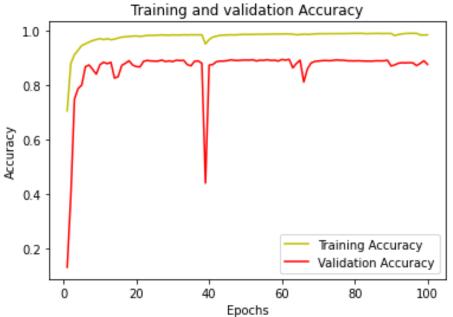


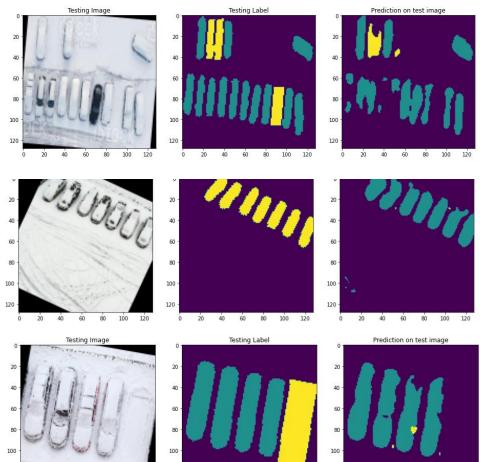


- 1. Manually collect images for train dataset (22 images)
- 2. Train resnet34 model for semantic segmentation (cars, parking, road, mud)
- 3. Apply morphological operations to labeled image
- 4. Using the contours, draw a rectangle for each "car area" that borders it
- 5. Retreating from the car at equal distances to the side, check whether the space is free or occupied

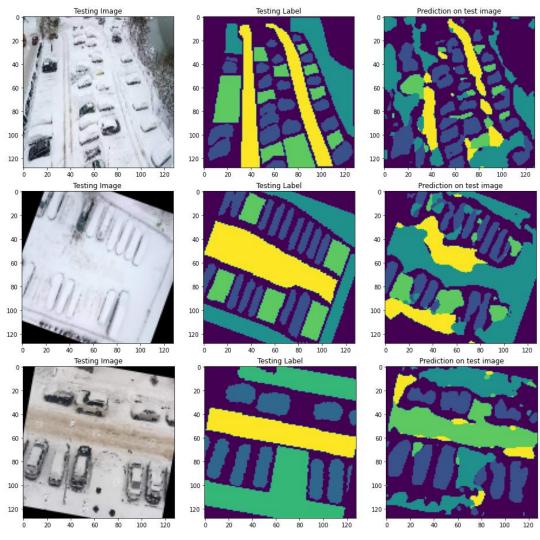


METHOD on Dataset-2

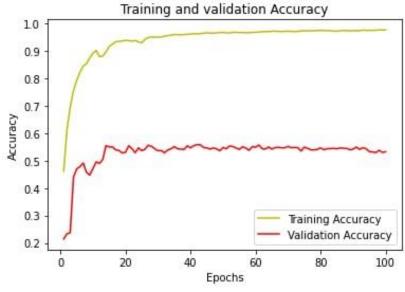








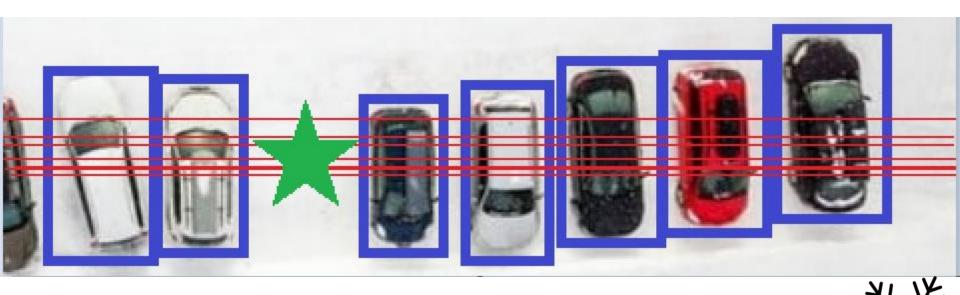
METHOD on Dataset-1







How does algorithm look?



RESULTS



To test the algorithm, we used a test dataset consisting of 6 images.



TPR	FPR	Mean error
0,54	0,14	0,6

Estimate quality metrics



RESULTS: GOOD EXAMPLES









RESULTS: BAD EXAMPLES





False negative



False positive







- 1. Using **semantic segmentation** to detect **parking spaces** as a class is **unreasonable**, because in winter conditions in most cases a free parking space (covered with snow) is no different from a road (covered with snow) or an environment (covered with snow).
- 2. A common sense approach is to use **semantic segmentation** to detect **snow-covered cars** and work with that information.
- 3. In such non-trivial tasks, the **data** set must be collected a manually, and its size must be large and cover all commonality.





References



Semantic segmentation model



Dataset Labeling