

Evaluation of Synthetic Video Data in Machine Learning Approaches for Parking Space Classification

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Examples of training data extracted from simulated environment footage. Weather and lighting conditions include sunny, cloudy and foggy samples.

Motivation

- Rising demand for annotated data in computer vision and machine learning tasks
 - Problematic for new applications and special cases
 - Huge effort of gathering data plus manual labeling makes many tasks unfeasible
 - Limits scientific progress to a relatively small number of applications
- Simulated data as an easy and controllable alternative to real-world image data
- But: Is the use of purely synthetic training data sufficient to build a recognition system with comparable performance?

Simulated Data Generation¹

- Simulated car park environment based on Unreal Engine 4
- Video data gathered from environment with artificial cameras
- Footage covers 5 different weather and lighting scenarios in total
- Exemplary task: parking space classification



Different weather and lighting scenarios within the simulated environment. Left: sunshine, Center: overcast sky, Right: fog

Automatic Data Extraction

- Data extraction on the fly
- Simultaneous extraction of training data and ground-truth labeling
- Training data: image crops of single parking spaces
- Ground truth: unique parking space ID, system time stamp, and occupancy status

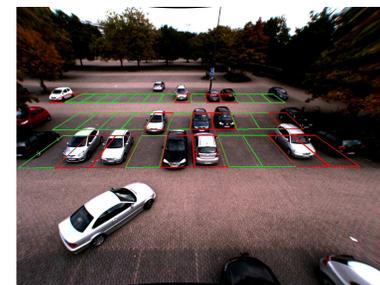
Experiments

- Training and evaluation of kNN and SVM classifiers
- DoG features of different filter sizes as input features
- Full replacement of image training data with synthetic images
- Training data comprises 3 different weather/lighting scenarios: sunny, cloudy, foggy
- Evaluation on real-world video sequences of two different car parks
- Measurement of classifier performance for individual parking spaces averaged per row

Sequence A



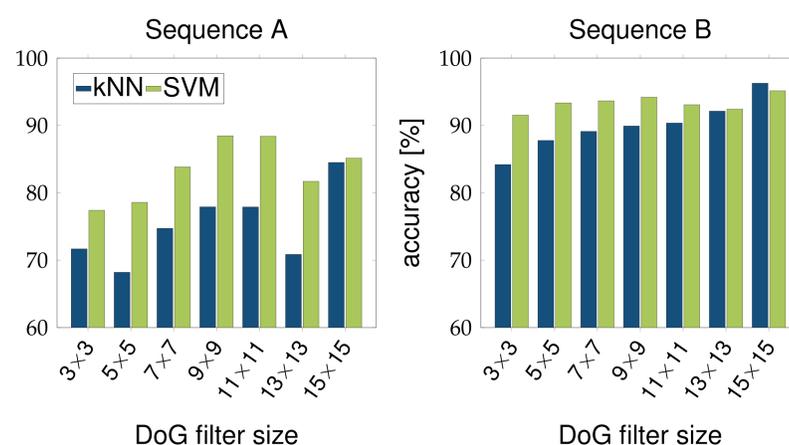
Sequence B



Real-time classification of two different parking areas. Red boxes are identified as occupied, green boxes denote available parking spaces.

Results

- SVM outperforms kNN classifier in most cases
 - Classification rate dependent on distance to camera
 - Slightly less accurate than previously evaluated classifiers trained on real-world data
- Parking space classification task can be solved without use of real video data



Outlook

- Create more diverse weather/lighting scenarios and additional parking areas
 - Add disruptive elements for more realism, such as wind simulation, camera movement, and pedestrians
 - Extend usability of simulated environment to other machine learning tasks, e.g. vehicle tracking and orientation classification
- Ultimately enable new applications for image processing algorithms

¹M. Tschentscher, B. Pruß, and D. Horn, "A Simulated Car-Park Environment for the Evaluation of Video-Based On-Site Parking Guidance Systems," in 2017 IEEE Intelligent Vehicles Symposium (IV), June 2017, pp. 1571–1576.