

Detecting Plastic Soup automatically

Using pre-trained Convolutional Neural Networks and Support Vector Machines

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Plastic Soup

- Large amounts of plastic end up in the world ocean [1]
- Automate the clean-up process
- Develop plastic soup recognition

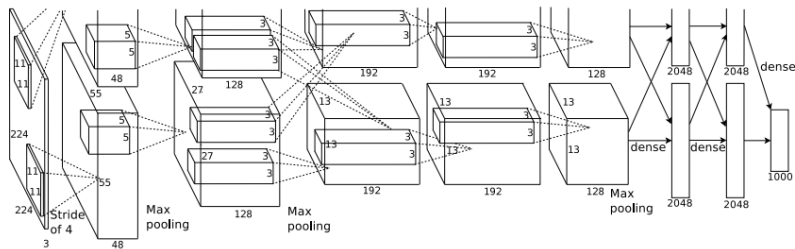


Figure 1: Boat in plastic soup

Current state of the art image techniques I

Convolutional Neural Networks

- Alexnet implementation to train large amounts of data [2]
- Current CNNs very high accuracy [3]

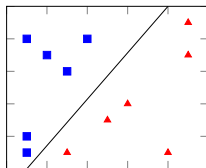


Current state of the art image techniques II

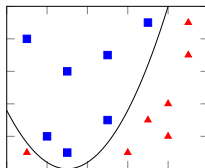
Support Vector Machine

- Classification algorithm widely used in CV
- Fit a hyperplane in the feature-space
- Side of the hyperplane on which a data-point lies is the classifier

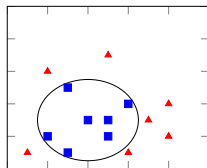
SVM with linear hyperplane



SVM with polynomial hyperplane



SVM with RBF hyperplane

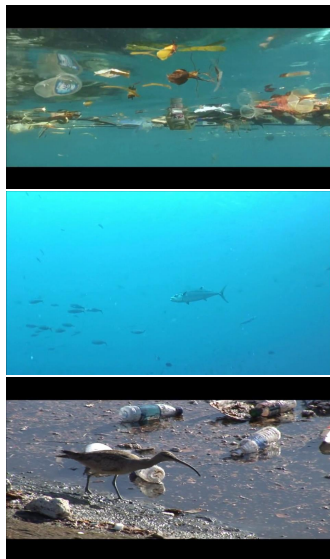


Research question

How does a pre-trained CNN in combination with an SVM perform when used for other classifications without being trained on a large amount of domain-specific data?

Dataset

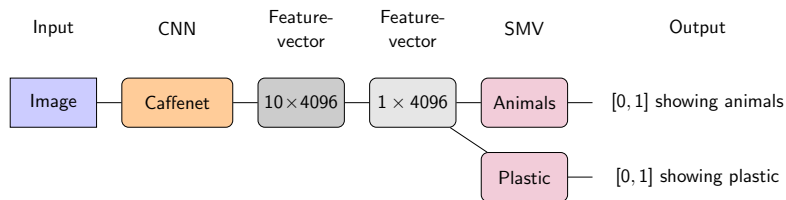
- 37165 images from short films
- annotated by my hand
- 16553 images above and 20612 images below water
- 20635 show plastic only
- 6972 show animals only
- 8502 show both



Method and evaluation

Approach

- Use a pre-trained Convolutional Neural Network as feature extractor
- Train an SVM on the second-to-last layer for this specific domain



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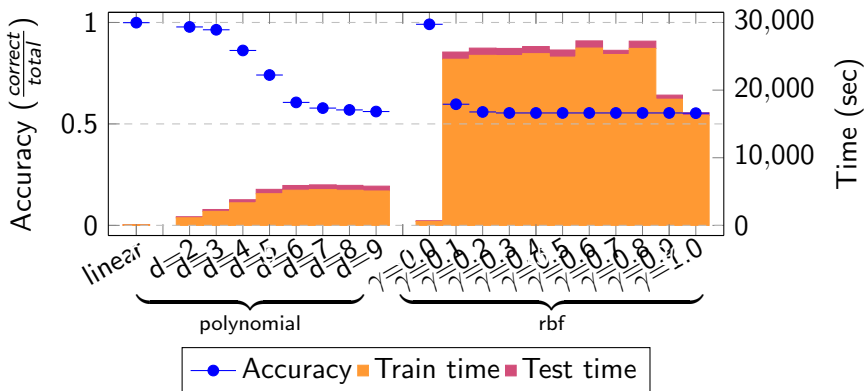
Evaluation

- split dataset in train, validate and test
- score the results on the annotated data

$$\frac{\#(Outcome_{True} \text{ and } Label_{True}) + \#(Outcome_{False} \text{ and } Label_{False})}{\#tested \text{ images}}$$

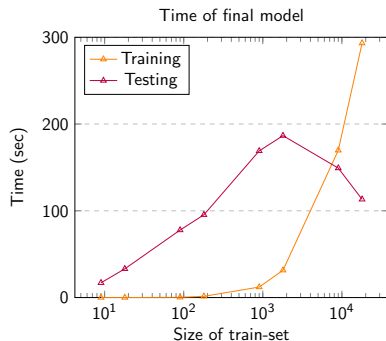
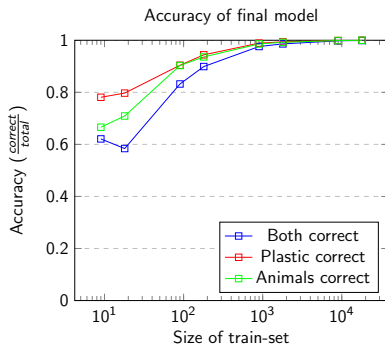
Results of the pipeline I

- linear SVM model works best
- 99.9% accuracy on the test-set



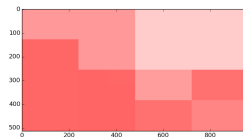
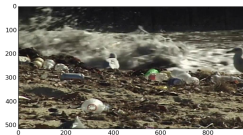
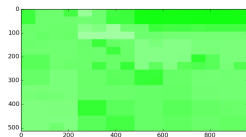
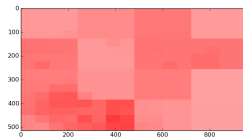
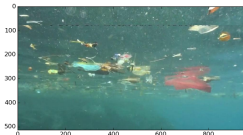
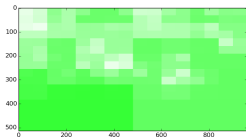
Results of the pipeline II

- chance of overfitting
- using small amounts of train-data
- high accuracy with small amounts of train-data



Localisation of plastic

- segment image, run each part through the pipeline
- results in heat-map that shows confidence of detecting plastic or animals



Conclusion

Research question

How does a pre-trained CNN in combination with an SVM perform when used for other classifications without being trained on a large amount of domain-specific data?

Answer

A Linear SVM trained on the second-to-last layer of a pre-trained CNN results in a high accuracy for the task of this project.

Discussion

Improvements for further research

- construct a better dataset to train and test the CNN and SVM
- use more cross-validation on finding parameters
- improve localisation of plastic
- improve time performance of the pipeline

A step closer to clean up the oceans of Plastic Soup

References I



D. Barnes and P. Milner.

Drifting plastic and its consequences for sessile organism dispersal in the atlantic ocean.

Marine Biology, 146(4):815–825, 2005.



A. Krizhevsky, I. Sutskever, and G. E. Hinton.

Imagenet classification with deep convolutional neural networks.

In *Advances in neural information processing systems*, pages 1097–1105, 2012.



A. S. Razavian, H. Azizpour, J. Sullivan, and S. Carlsson.

Cnn features off-the-shelf: an astounding baseline for recognition.

In *Computer Vision and Pattern Recognition Workshops (CVPRW), 2014 IEEE Conference on*, pages 512–519. IEEE, 2014.