

TRAFFIC-SIGN-RECOGNITION WITH MACHINE LEARNING

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AGENDA

Introduction to YOLO (You Only Look Once) and its advantages for object detection

Description of the dataset used for training and testing the model

Explanation of the training parameters

Results and performance evaluation of the model on the traffic sign dataset

Discussion of potential applications and future work for traffic sign recognition using YOLO and other object detection models



CNN (Convolutional Neural Network)

Feature extraction of input image

- VGG16
- ImageNet
- RetineNet
- Resnet50

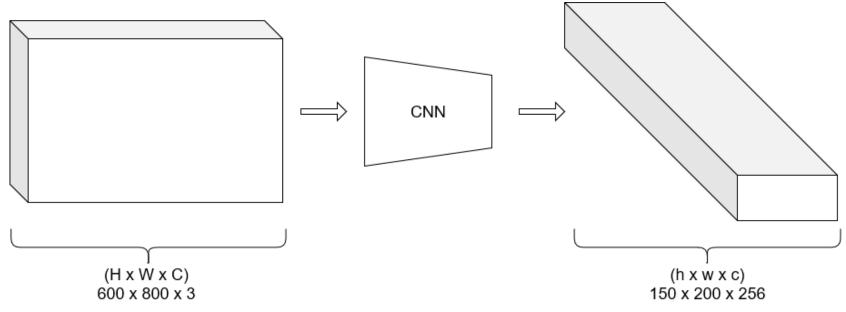
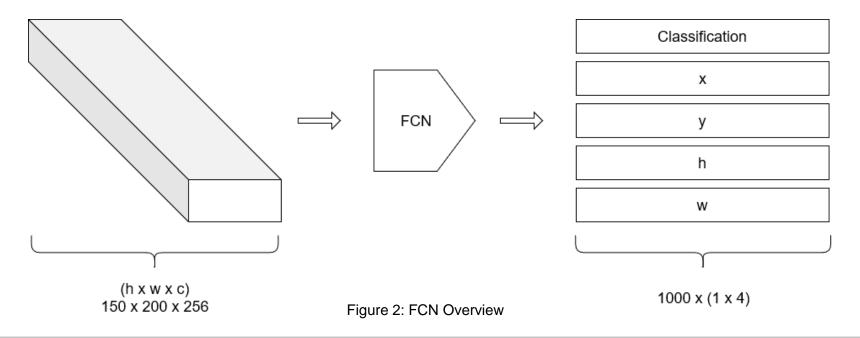


Figure 1: CNN Overview



FCN (Fully Connected Network)

Computes probability for each anchor box to contain an object and estimate the coordinates of the searched object. Example with 1000 anchor boxes:





NMS (Non-Maximum Suppression)

When a proposal is selected, all proposals that have an intersection over union (IoU) with the selected proposal of more than a certain threshold (usually 0.5) are removed.

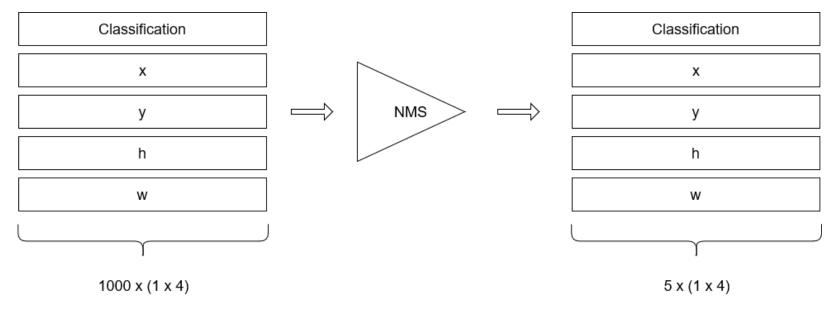


Figure 3: NMS Overview



DESCRIPTION OF THE DATASET USED FOR TRAINING AND TESTING THE MODEL

- Using 47 different traffic signs
- Each sign gets a class id for labeling
- Label-files define:
 - Class id
 - X-position (normalized)
 - Y-position (normalized)
 - Box width (normalized)
 - Box height (normalized)









Figure 4: Example of image labeling

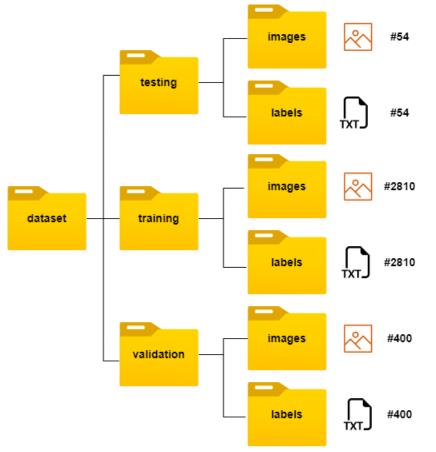


Figure 5: Folder structure of data

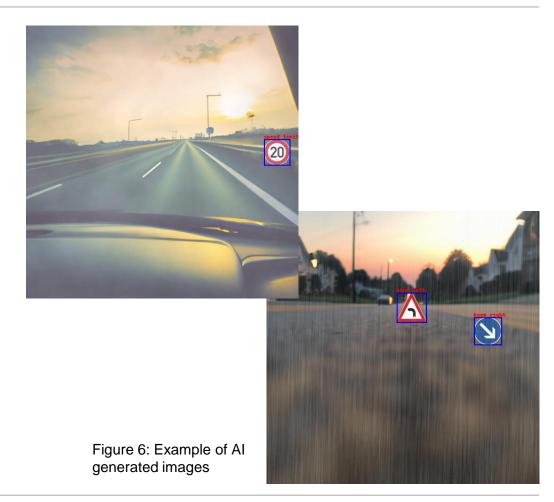


DESCRIPTION OF THE DATASET USED FOR TRAINING AND TESTING THE MODEL

Using AI to enhance our dataset

- Generate background images with various traffic environments with DALL

 E 2
- Label traffic sign images
- Paste traffic signs in random positions and sizes on background image
- Add augmentations like weather conditions
- Create corresponding label files automatically





EXPLANATION OF THE TRAINING PARAMETERS

- ≥300 Epochs
- Image size: 640x640
- Batch-Size: 4
- Transfer learning approach with pretrained weigths:
 - YoloV7 trained on Coco-Dataset

That makes no sense in this application! E.g. The directions of the arrows within the signs must point in the right direction.

- Data Augmentation:
 - hsv_h: 0.015 [image HSV-Hue augmentation (fraction)]
 - hsv_s: 0.7 [image HSV-Saturation augmentation (fraction)]
 - hsv_v: 0.4 [image HSV-Value augmentation (fraction)]
 - degrees: 0.15 [image rotation (+/- deg)]
 - translate: 0.2 [image translation (+/- fraction)]
 - scale: 0.5 [image scale (+/- gain)]
 - shear: 0.2 [image shear (+/- deg)]
 - perspective: 0.0 [image perspective (+/- fraction), range 0-0.001]
 - flipud: 0.0 [image flip up-down (probability)]
 - fliplr: 0.0 [image flip left-right (probability)]



Precision:
$$\frac{TP}{TP+FP} = \frac{TP}{Positive \ results \ total}$$

Recall:
$$\frac{TP}{TP+FN} = \frac{TP}{To \ detect \ total}$$

Intersection over union (IoU): $\frac{area\ of\ overlap}{area\ of\ union}$

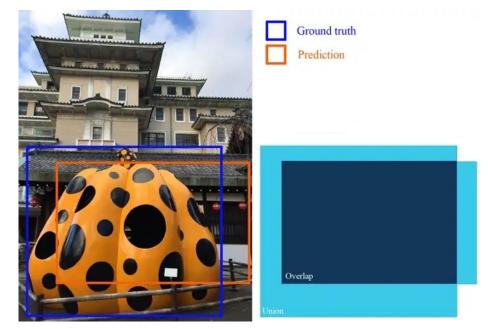


Figure 7: Example for IoU

Mean average precision (mAP): $\frac{1}{\#classes} \sum_{c \ \epsilon \ classes} \frac{\#TP(c)}{\#TP(c) + \#FP(c)}$



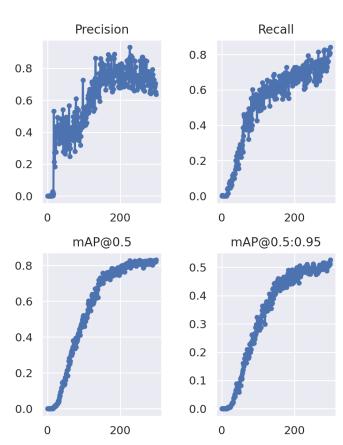


Figure 8: Example of performance evaluation

Precision:
$$\frac{TP}{TP+FP} = \frac{TP}{Positive \ results \ total}$$

Recall:
$$\frac{TP}{TP+FN} = \frac{TP}{To \ detect \ total}$$

Mean Average Precision (mAP):
$$\frac{1}{\#classes} \sum_{c \ \epsilon \ classes} \frac{\#TP(c)}{\#TP(c) + \#FP(c)}$$

mAP@0.5: IoU threshold \geq 0.5

mAP@0.5:0.95: IoU threshold ≥0.5:0.05:0.95



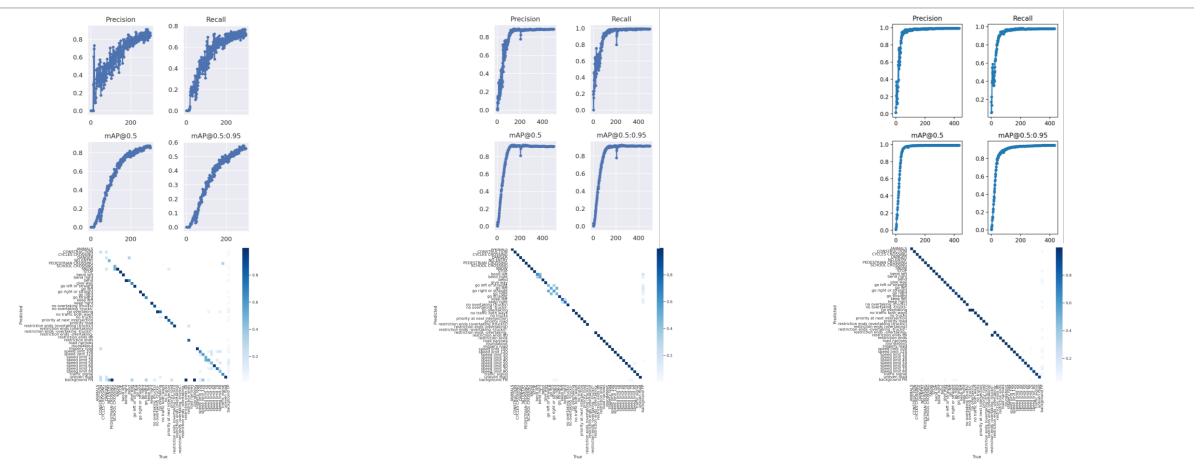


Figure 9: Performance of real-world data only dataset

Figure 10: Performance of Al-generated data only dataset

Figure 11: Performance of mixed dataset







DISCUSSION OF POTENTIAL APPLICATIONS AND FUTURE WORK FOR TRAFFIC SIGN RECOGNITION USING YOLO AND OTHER OBJECT DETECTION MODELS

Improve detection model

- Improve hyperparameters
- Enhance dataset
- Train different object detection models and compare to YOLOv7
- Improve AI generated images
- Add the traffic signs that are not included in the dataset (more classes)



DISCUSSION OF POTENTIAL APPLICATIONS AND FUTURE WORK FOR TRAFFIC SIGN RECOGNITION USING YOLO AND OTHER OBJECT DETECTION MODELS

Potential Applications

- Modell compression with TensorRT [https://developer.nvidia.com/tensorrt]
- App development (Android)
 [https://developer.android.com/studio/]
- Model deployment on smartphone
 [https://www.tensorflow.org/lite/android]
- → Improve accuracy, compress model, build an app and deploy the trained model into app.

Result: traffic-sign detecting camera within smartphone!



ANY QUESTIONS?

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CNN

Feature extraction of input image

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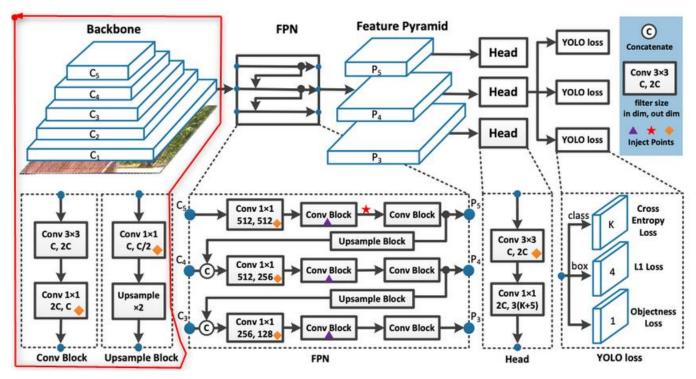


Figure 12: Detailed overview of CNN



Image features are combined and fused in the Neck

Feature Pyramid Backbone Conv 3×3 C, 2C filter size dim, out dir YOLO loss Conv 3×3 Conv 1×1 Conv 3×3 C, 2C C, C/2 C, 2C 512, 256 Conv 1×1 Conv 1×1 2C, 3(K+5) 2C, C Conv 1×1 Conv Block Conv Block Upsample Block **YOLO loss** Head

Figure 13: Detailed overview of Neck

Final layers are passed to Head, which makes prediction

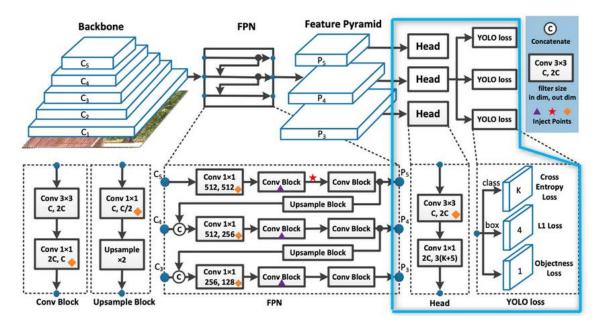
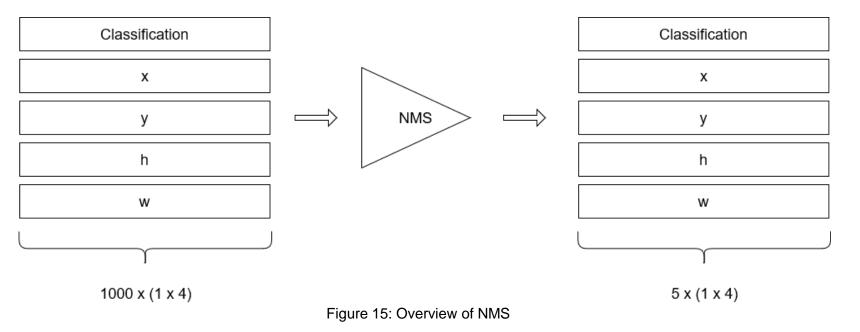


Figure 14: Detailed overview of Head



NMS

When a proposal is selected, all proposals that have an intersection over union (IoU) with the selected proposal of more than a certain threshold (usually 0.5) are removed.





DESCRIPTION OF THE DATASET USED FOR TRAINING AND TESTING THE MODEL

Real world data from GTSDB database: (GTSDB - German Traffic Sign Detection Benchmark)

Roboflow shares pictures of this category. These images are already labeled.

Training: 1100 images

Validation: 108 images

Testing: 54 images

Same data folder structure as shown as in figure 5.

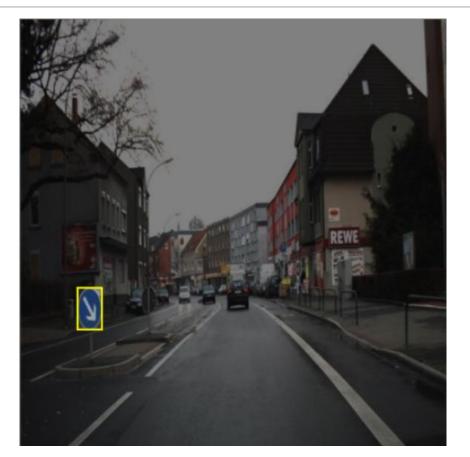


Figure 16: Example of GTSDB image