# FROM BLACK-BOX TO WHITE-BOX: EXAMINING CONFIDENCE CALIBRATION UNDER DIFFERENT CONDITIONS

# FRANZISKA SCHWAIGER, MAXIMILIAN HENNE, FABIAN KUEPPERS, FELIPPE SCHMOELLER ROZA ,KARSTEN ROSCHER , ANSELM HASELHOFF





HOCHSCHULE RUHR WEST UNIVERSITY OF APPLIED SCIENCES



## THE PERCEPTION CHALLENGE



Source: Towards a Framework to Manage Perceptual Uncertainty for Safe Automated Driving, Krzysztof Czarnecki and Rick Salay







# BACKGROUND: CONFIDENCE CALIBRATION FOR CLASSIFICATION

Modern neural networks are overconfident [1]

# **Measurement of Miscalibration**: *Expected*

*Calibration Error* (ECE):

$$ECE = \sum_{n=1}^{N} \frac{|I(n)|}{|D|} \cdot |acc(n) - conf(n)|$$



0.0 0.2 0.4 0.6 0.8 1.0 Confidence

Fig. 1: Reliability Diagram of a 110layer ResNet on CIFAR-100 [1]

### **OBJECT DETECTION PIPELINE**



A Bochkovskiy et. al, Yolov4: Optimal speed and accuracy of object detection



# BACKGROUND: CONFIDENCE CALIBRATION FOR OBJECT DETECTION

Modern object detectors are also not well calibrated [2]

### **Measurement of Miscalibration**:

Detection Expected Calibration Error (D-ECE):

$$D - ECE_K = \sum_{n=1}^{N_{total}} \frac{|I(n)|}{|D|} \cdot |prec(n) - conf(n)|$$

Histogram Binning for Object Detection Histogram Binning for Classification



Going from 1D Histogram Binning to multi-dimensional Histogram Binning



## **BACKGROUND: NON-MAXIMUM-SUPPRESSION (NMS)**

NMS

**Method**: selects a single bounding box with the highest confidence score out of many overlapping bounding boxes

**Intersection over Union (IoU):** controls how aggressively overlapping boxes are discarded





# WHITE-BOX VS. BLACK-BOX

Miscalibration of object detectors is measured before (white-box) and after (black-box) NMS

Training and evaluation is done on the COCO2017 validation dataset

TP:  $IoU(\Box, \Box) \ge threshold$ FP:  $IoU(\Box, \Box) < threshold$ 





# WHITE BOX CALIBRATION RESULTS



### RetinaNet



Calibrated D-ECE: 0.981%

0.6

0.8

1.0

### 1.0 --- Avg. Precision 0.8 --- Avg. Confidence 0.8 Relative Amount of Samples 0.6 O f 0.4 04 0.2 0.0 0.0 0.2 0.4 0.6 0.8 1.0 1.0 -1.0 --- Perfect Calibration 0.8 -Actual Gap Gap 0.6 0.6 0.4 0.4 0.2 0.2 0.0 0.2 0.4 0.6 0.8 1.0 0.0 Before

### Faster R-CNN



Uncalibrated D-ECE: 4.20%

Calibrated D-ECE: 0.861%

# **QUALITATIVE RESULTS**

### NMS potentially degrades initially well-calibrated predictors



### After white-box calibration

1.0

0.8

0.4

0.2

0.0

1.0 -

0.8 -

0.6

0.4

0.2

0.0

0.0

10

0.75

0.5

0.25

0.0

0.0

0.0

--- Avg. Precision

--- Avg. Confidence

Relative Amount of Samples

0.2

0.2

0.25

--- Perfect Calibration

Actual

Gan Gan

0.4

0.4

0.6

0.6

0.75

0.5

0.8

0.8

1.0

0.4

0.3

0.2

0.1

1.0

D-ECE

Position-dependent Heatmap of RetinaNet

### CONCLUSION

- Focal loss decreases average confidence
- $\circ$  White Box Calibration works
- NMS making predictions overconfident
- Detections on border of image tend to be worse calibrated than near the center
- Future work should focus on other box aggregation methods than NMS (e.g. average boxes)

### REFERENCES

All quantitave calibration results are given in our paper in more detail.

### References

[1] Chuan Guo, Geoff Pleiss, Yu Sun, and Kilian Q. Weinberger.

### On Calibration of Modern Neural Networks.

In Proceedings of the 34th International Conference on Machine Learning, volume 70 of Proceedings of Machine Learning Research, pages 1321–1330. PMLR, August 2017.

[2] Küppers, F.; Kronenberger, J.; Shantia, A.; and Haselhoff, A. "2020.

### Multivariate Confidence Calibration for Object Detection.

In Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition Workshops, 326–327.

