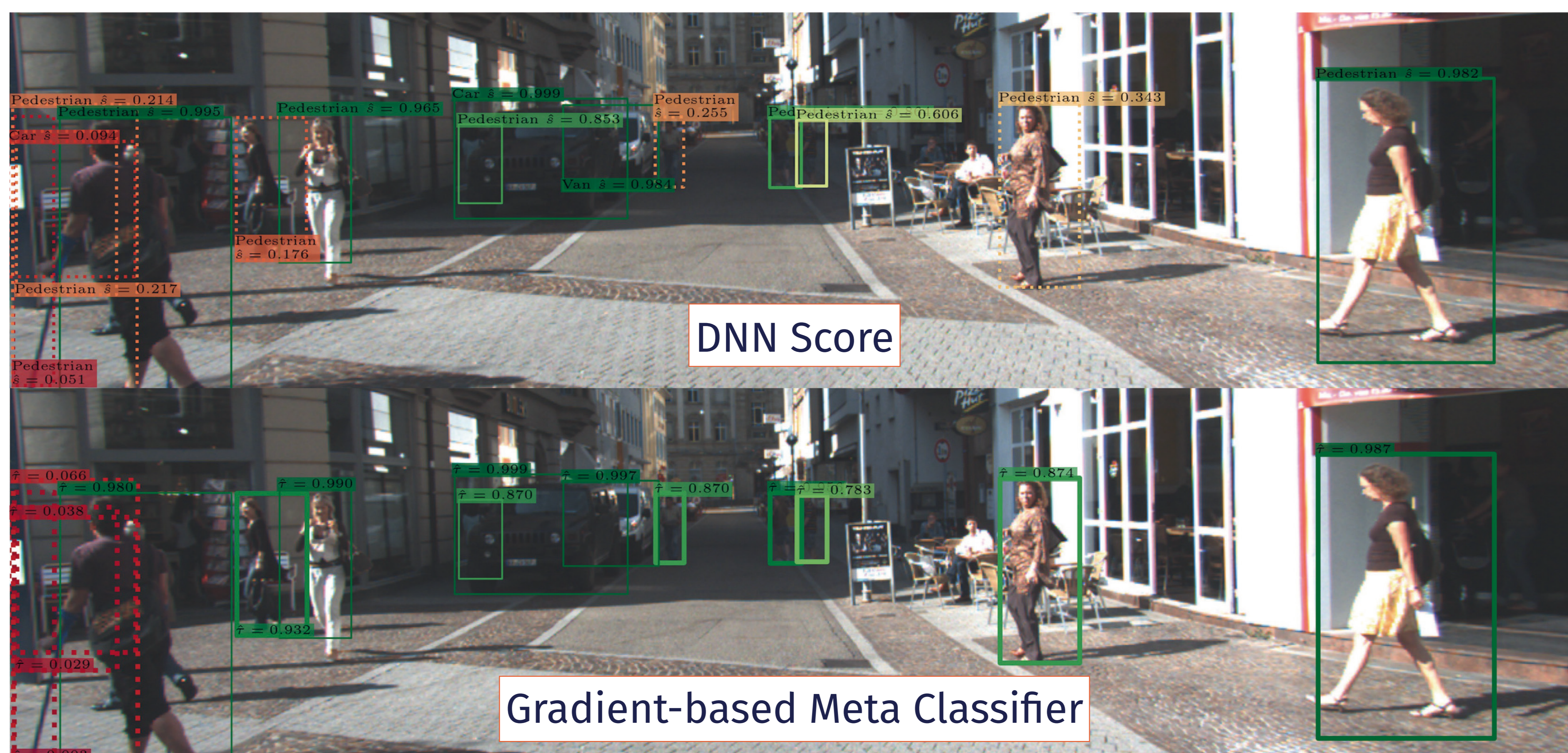


## Gradient-based Uncertainty Measures

Given an input to a DNN, self-learning gradients reflect the expected learning stress exerted on the model parameters. It is related to classification uncertainty metrics like softmax entropy and can be used to estimate prediction uncertainty. A generalization of this concept for object detectors has been developed in AP3.4 and further investigated in EWS1.

## Safety Hypothesis:

This method addresses the safety concern of Unreliable Confidence Information. It allows for statistically improved and reliable (calibrated) confidence estimation without requiring architectural changes or re-training of the object detector.



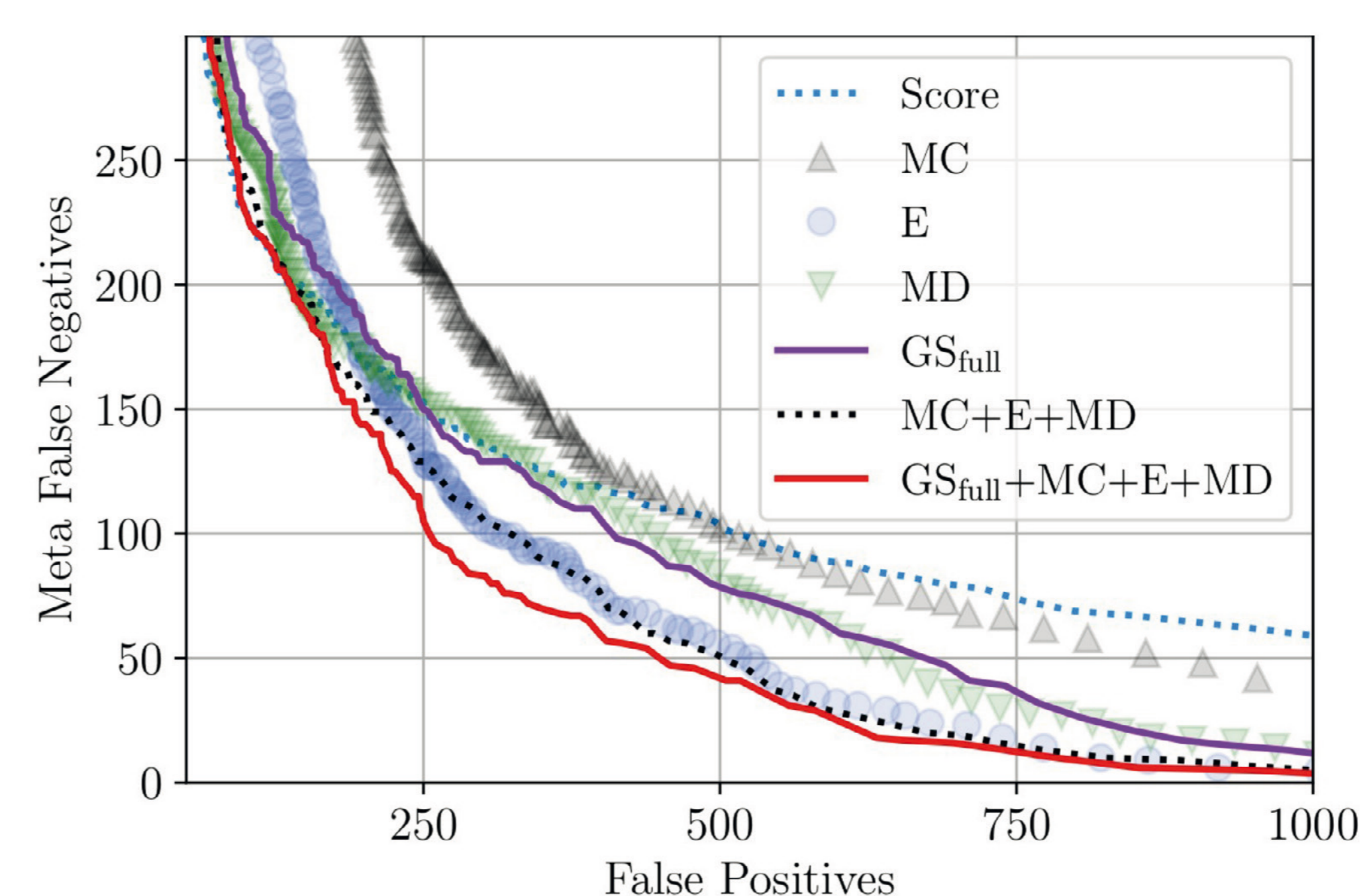
## Instance-wise Learning Gradient

Estimating uncertainty for each predicted instance requires the adaptation of the learning gradient. We derive instance-wise self-learning gradients by restricting to proposed candidate boxes for an output box and regarding:

$$g^{\text{cand}}(\mathbf{x}, \mathbf{w}, \hat{y}^j) := \nabla_{\mathbf{w}} \mathcal{L}(\text{cand}[\hat{y}^j](\mathbf{x}, \mathbf{w}), \bar{y}^j)$$

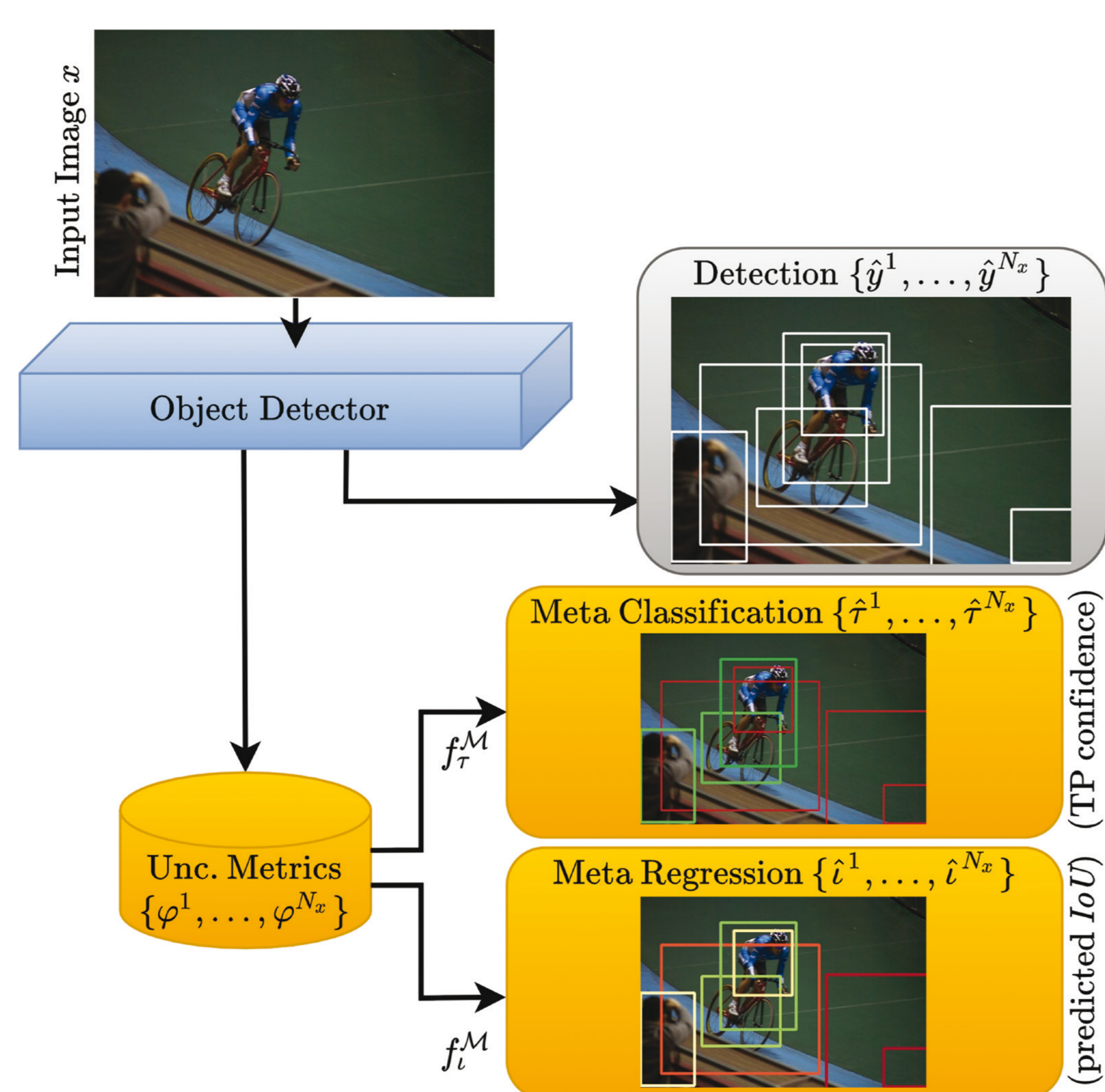
This amounts to the learning stress on prior boxes responsible for predicting the regarded instance.

## Decisions based on Meta Classification

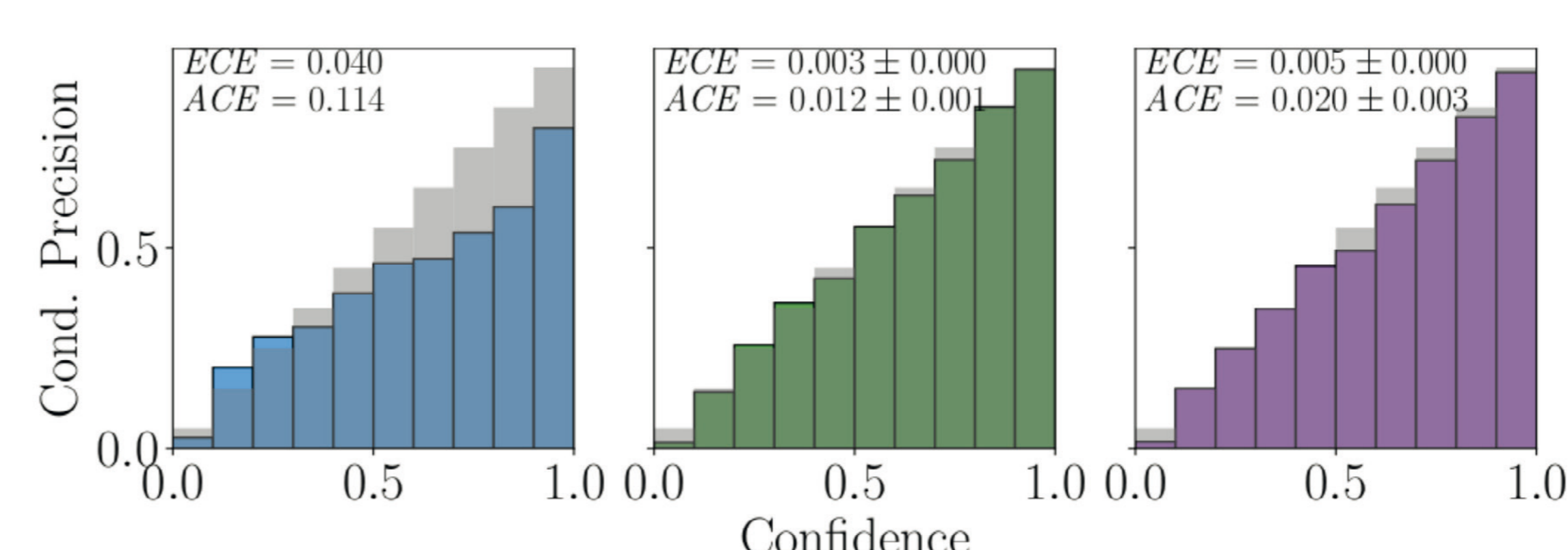


Significant improvement in terms of False Positive error count for fixed False Negatives

## Meta Classification and Meta Regression



## Confidence Reliability



Calibrated confidences from meta classification (center and right) compared with the native DNN score (left).

## References:

T. Riedlinger, et al. „Gradient-Based Quantification of Epistemic Uncertainty for Deep Object Detectors.“ arXiv preprint arXiv:2107.04517 (2021)