

Real-time Semantic Computer Vision for Co-robotics

Towards Realistic Predictors

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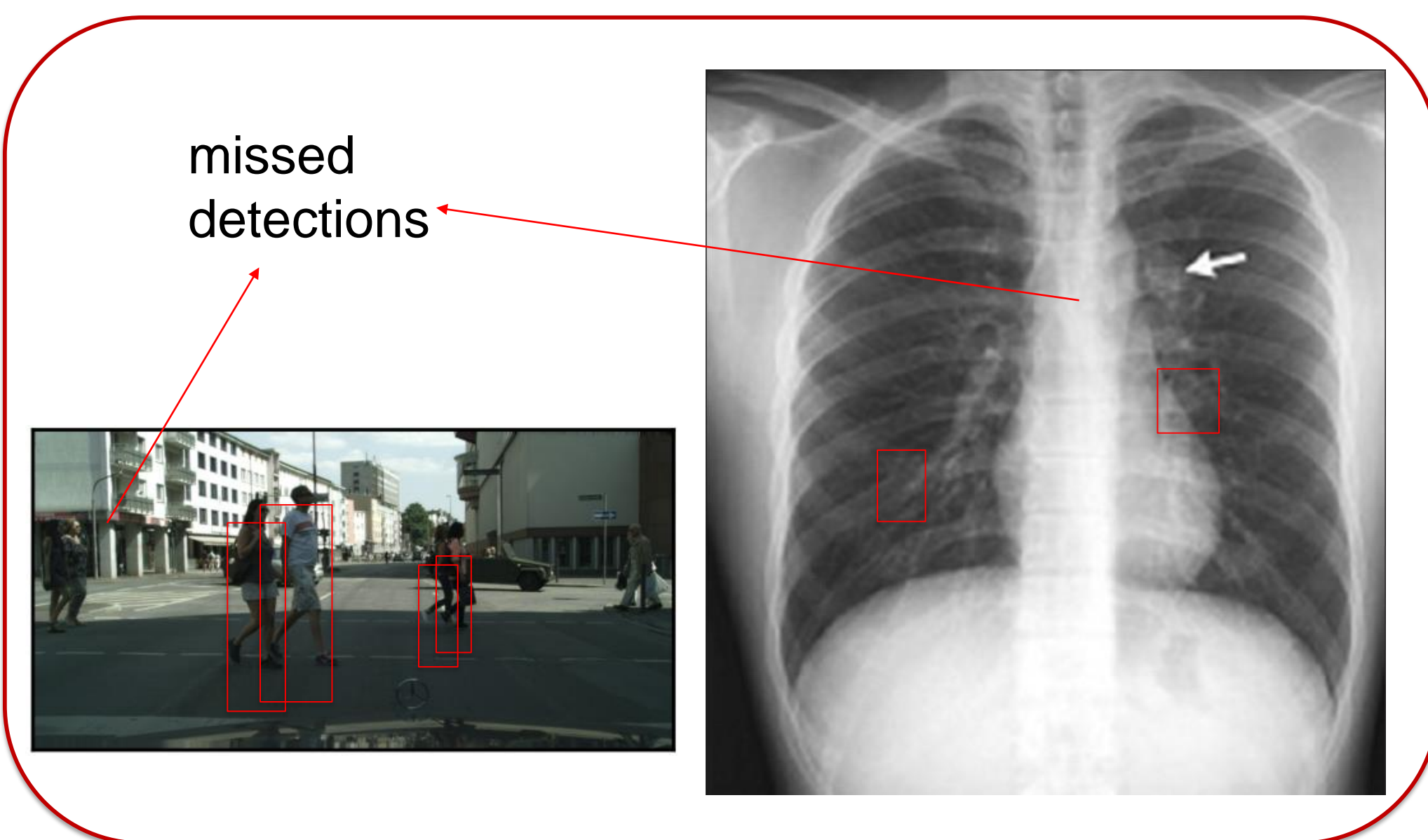


Introduction

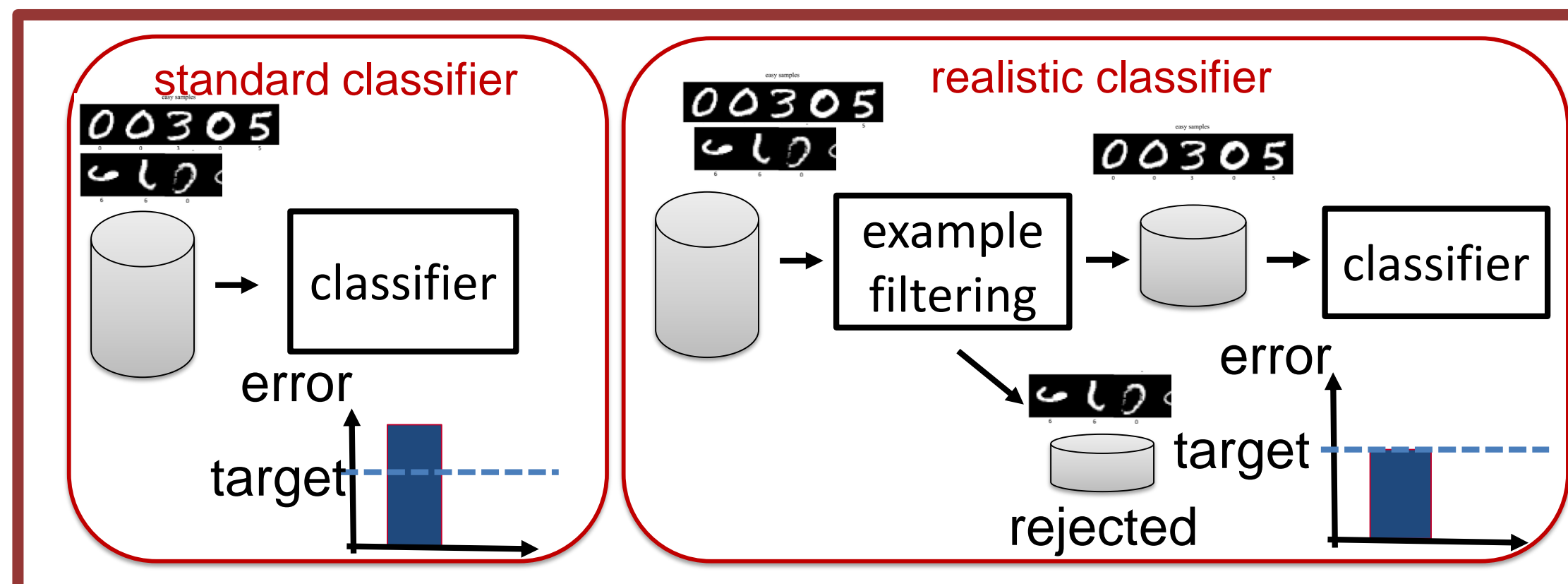
- Most machine learning and computer vision models are **over-optimistic**. They are trained to perform as well as possible on every example, without regard for how hard the task is.
- This is not what humans do! They analyze the difficulty of each task first, accept tasks that are doable and **refuse tasks that are too hard**. We refer this as **realism**.



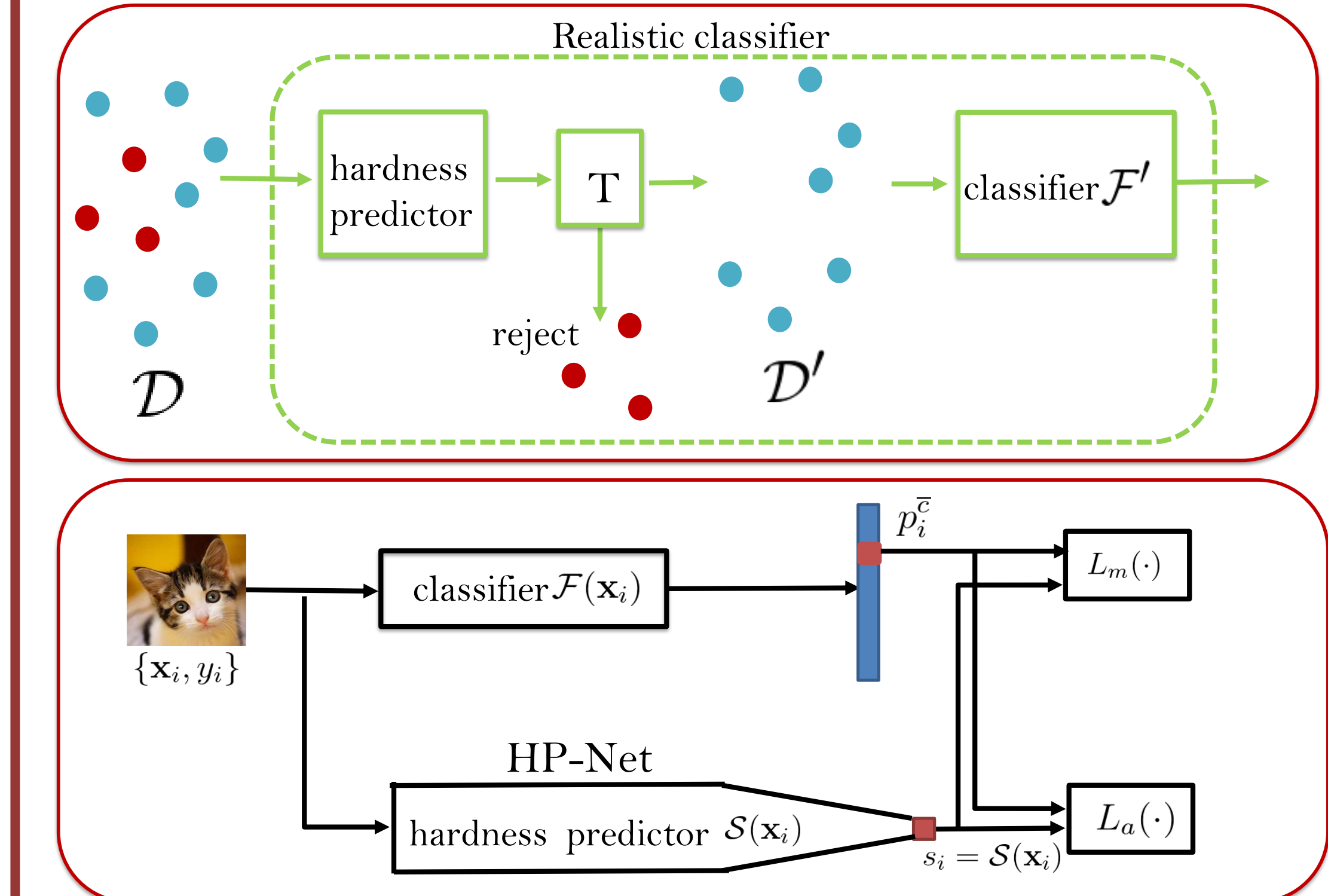
This trait is **critical** for applications where it is OK not to classify all examples so as to **guarantee performance rates on the ones that are classified**, like circumventing risk in autonomous driving and tumor detection and classification;



- Vision system should either
 1. Refuse to perform the hard task;
 2. Request additional information from other sensors.
- We define and propose a new type of classifiers, **realistic classifiers** that **reject examples deemed too hard** so as to **guarantee a target performance on the accepted examples**.



Architecture



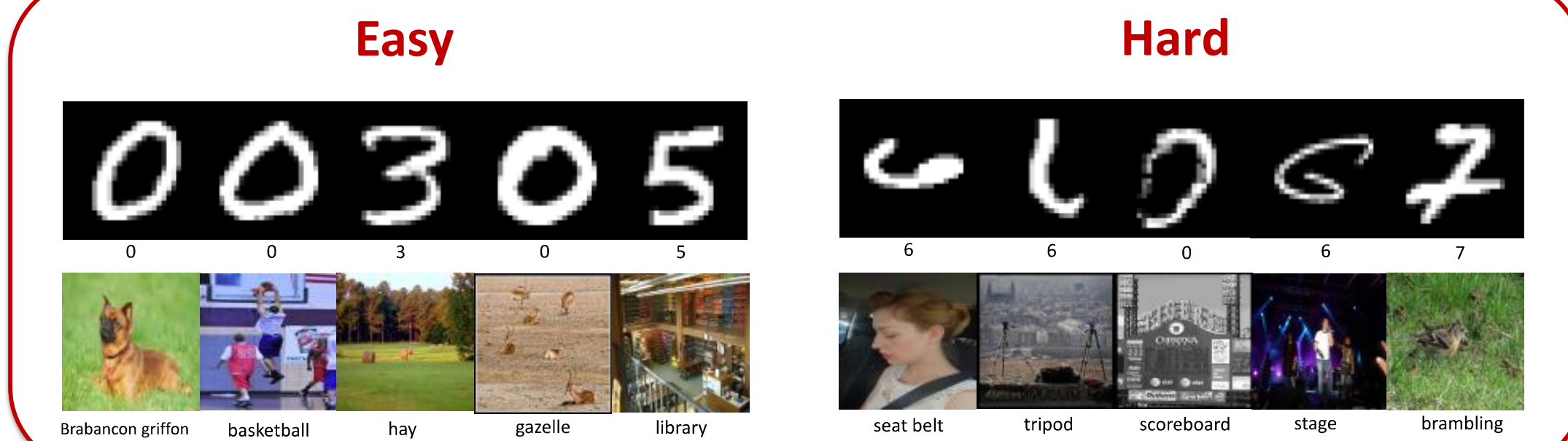
- Training**
 - Train classifier \mathcal{F} and HP-Net S jointly on training set D ;
 - Run S on D and eliminate hard examples, to create realistic training set D' ;
 - Learn realistic classifier \mathcal{F}' on D' , with S fixed;
 - Output pair S and \mathcal{F}' .
- Testing**
 - Run test examples x by S , reject hard examples, classify remaining with \mathcal{F}' . In all cases, x rejected if $S(x) > T$, for some threshold T .

Loss function

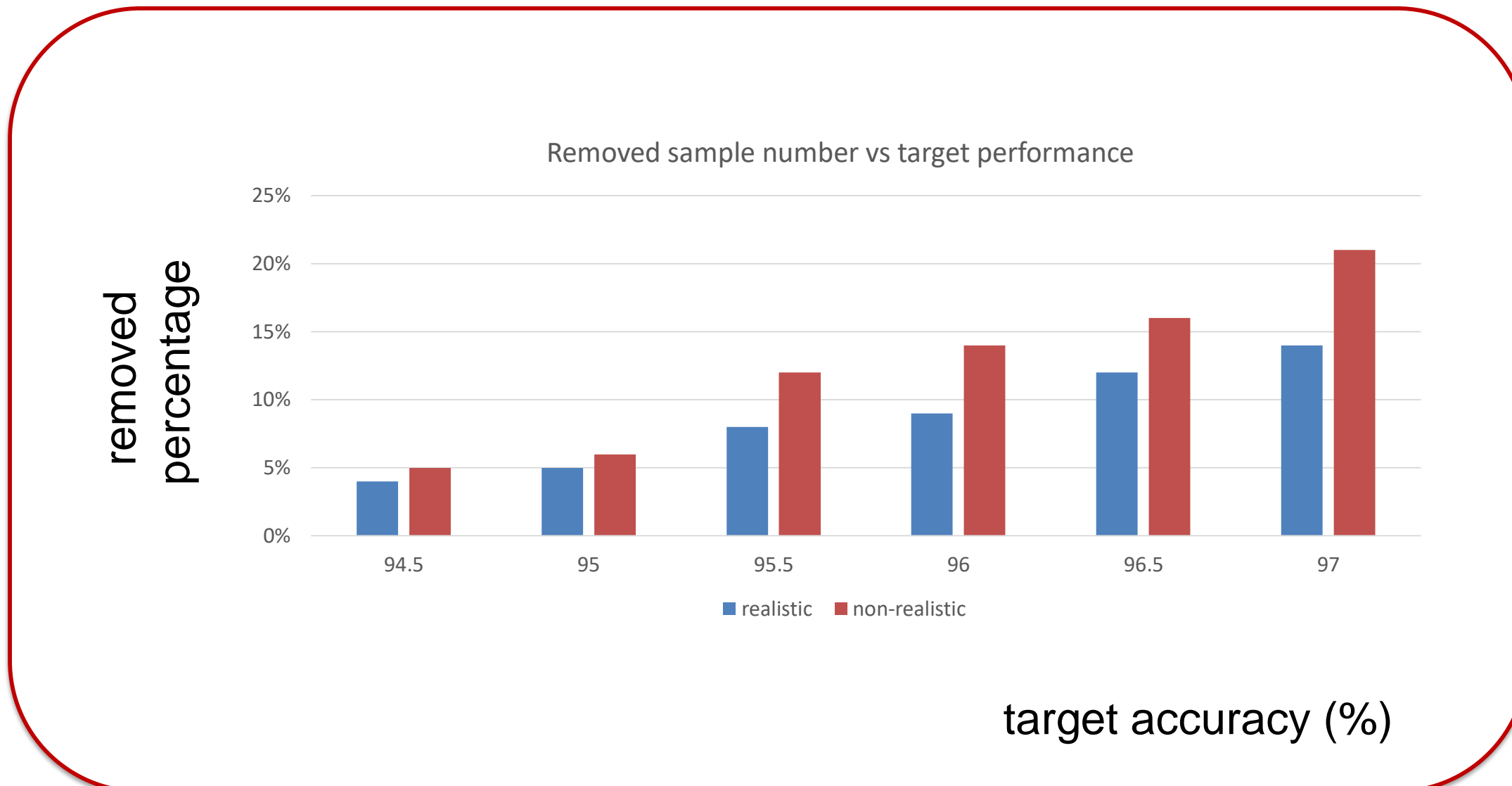
$$L_m(\{x_i, y_i, s_i\}) = \sum_i s_i l(x_i, y_i)$$

$$L_a(\{x_i, y_i, s_i\}) = - \sum_i \{p_i^c \log(1 - s_i) + (1 - p_i^c) \log s_i\}$$

Experiments



- Easy samples are **clearly written**, close to prototypical digits, while hard samples are "in between" digits.
- Evaluations of realistic predictors**



- Realistic predictor **guarantees a target performance while accepting and classifying more examples than non-realistic one**.



- VGG is a weaker model, but equals the ResNet when rejecting 5% of the examples, and beats it by 2% when rejecting 10%.