

Deep Learning Strong Parts for Pedestrian Detection

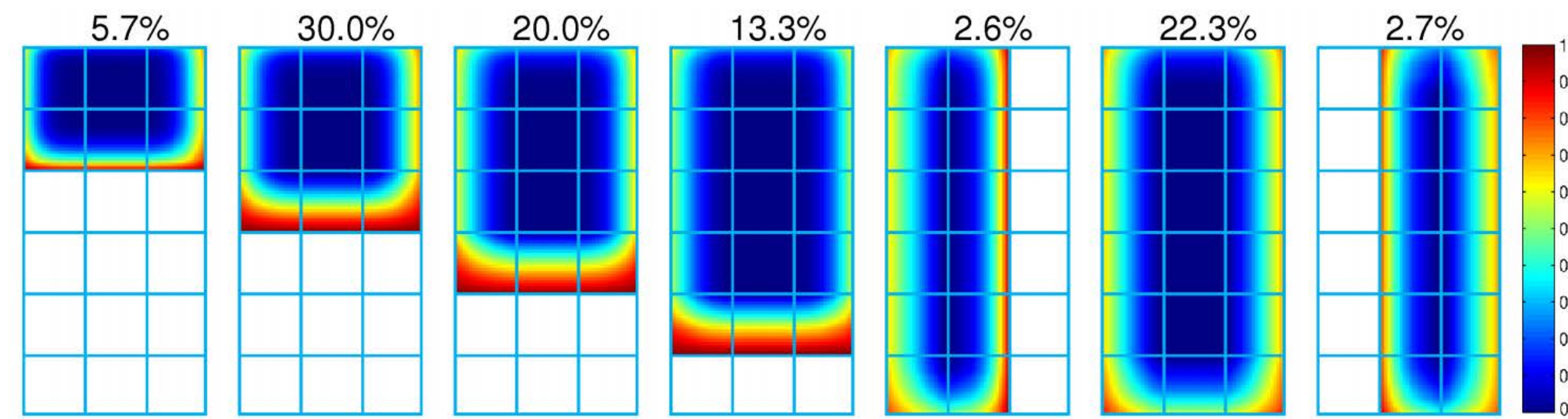
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1. Motivation

Background

- **Occlusion distribution of Caltech Pedestrian dataset.** Over 97% of occluded pedestrians belong to a small subset of hundreds of possible occlusion types.

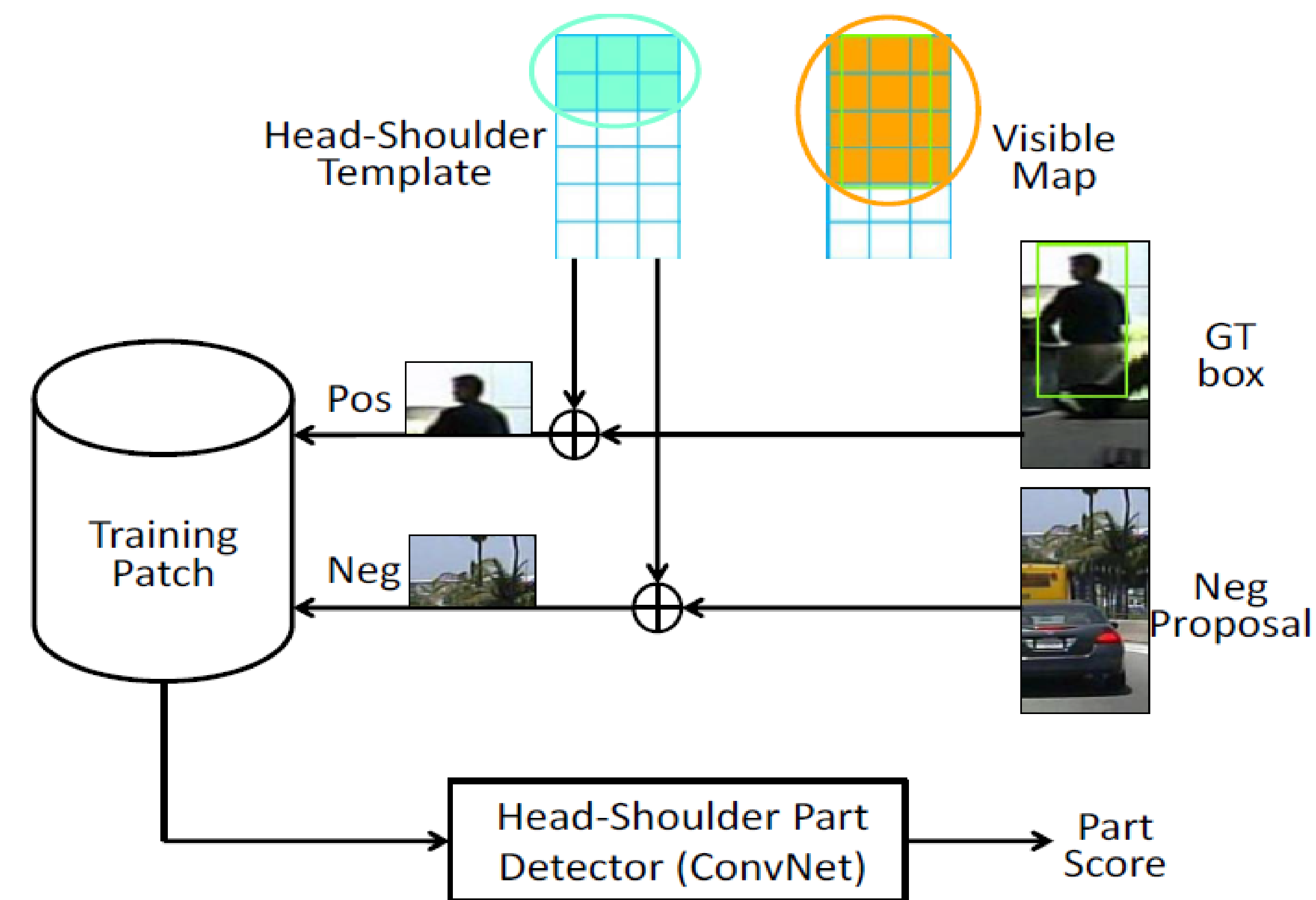


- Occlusion types may vary in different scenes.

Idea

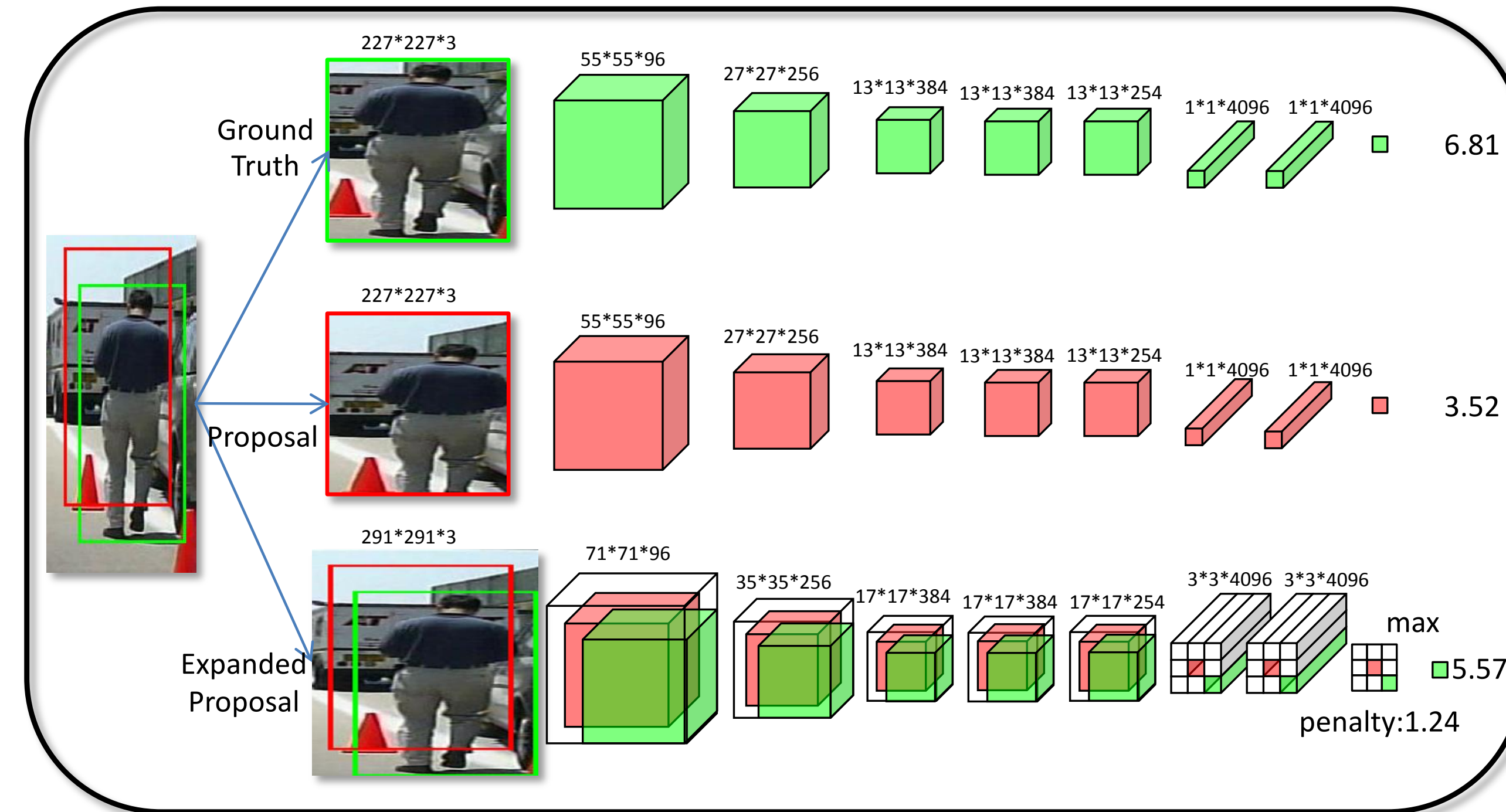
- **Part Detector Pool.** We construct a detector pool which includes part detectors of all sizes and positions.
- **Part Selecting.** We select significant and complement part detectors in different scenes or datasets.

2.1 Training Part Detectors



- **LDCF Proposals.** Use LDCF detector to propose candidates
- **Part Patches.** Crop positive and negative part patches from positive and negative candidates, respectively.

2.2. Handling Shifted Proposals



- **Fully Convolutional Neural Network.** Reformulate the fully connected layers as convolutional layers.
- **Expanded Proposal.** Given a proposal (x_{min}, y_{min}, w, h) , crop it as $(x'_{min}, y'_{min}, w', h')$, where

$$x'_{min} = x_{min} - \frac{16n}{227} \times w, \quad y'_{min} = y_{min} - \frac{16n}{227} \times h,$$

$$w' = (1 + \frac{32n}{227}) \times w, \quad h' = (1 + \frac{32n}{227}) \times h.$$

- **Score Adjustment.** The final score of this proposal is adjusted by

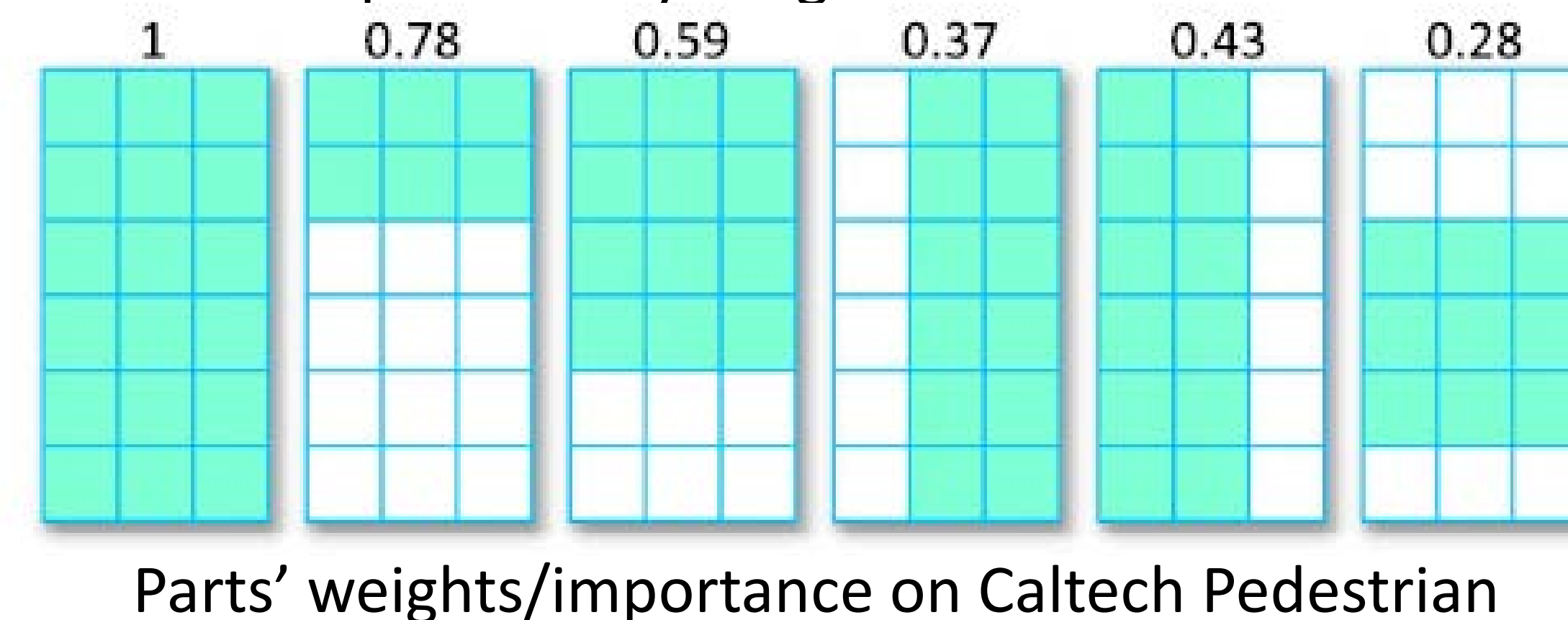
$$s = \max_{1 \leq i, j \leq n+1} \{S_{i,j} - P_{i,j}\}$$

$$P_{i,j} = a \times (|i - \frac{n+2}{2}| + |j - \frac{n+2}{2}|) \times \frac{32}{227}$$

$$+ b \times (|i - \frac{n+2}{2}|^2 + |j - \frac{n+2}{2}|^2) \times (\frac{32}{227})^2$$

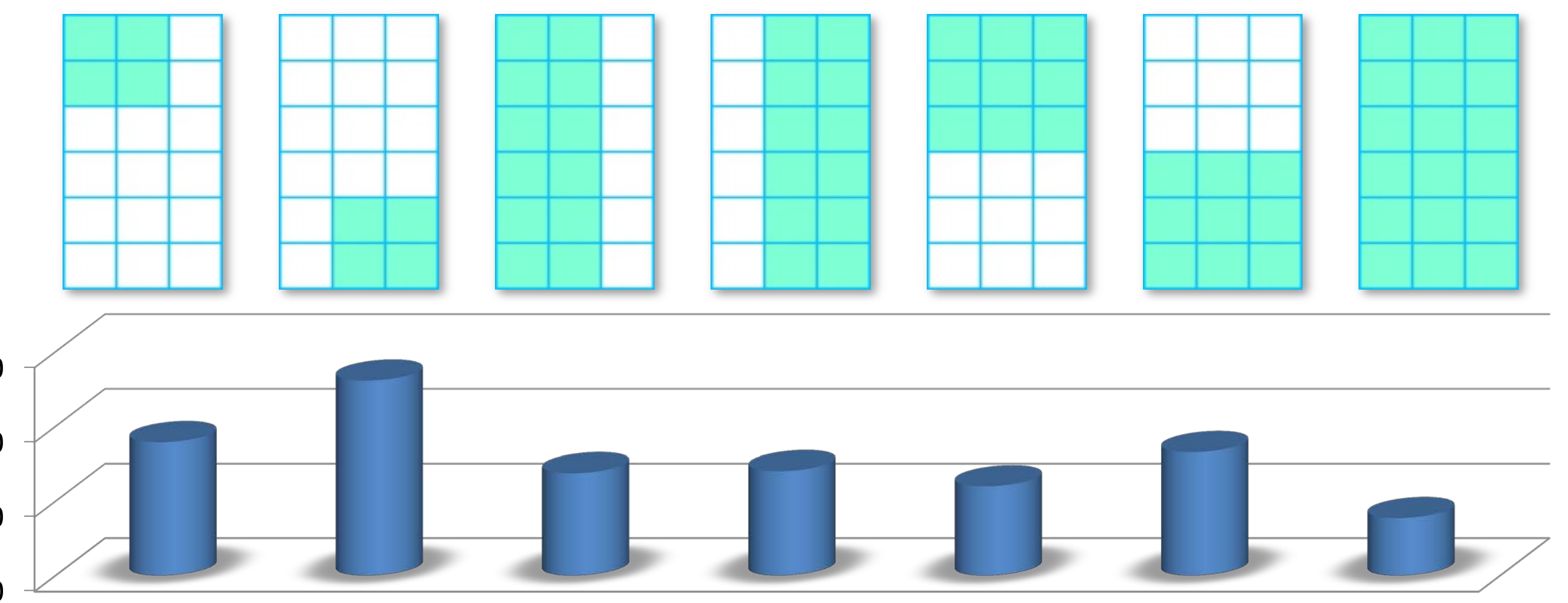
2.3. Complementary Parts

- **Complementary Parts.** Combine part scores; Select significant parts according to combinational weights; Re-learn the complementary weights.

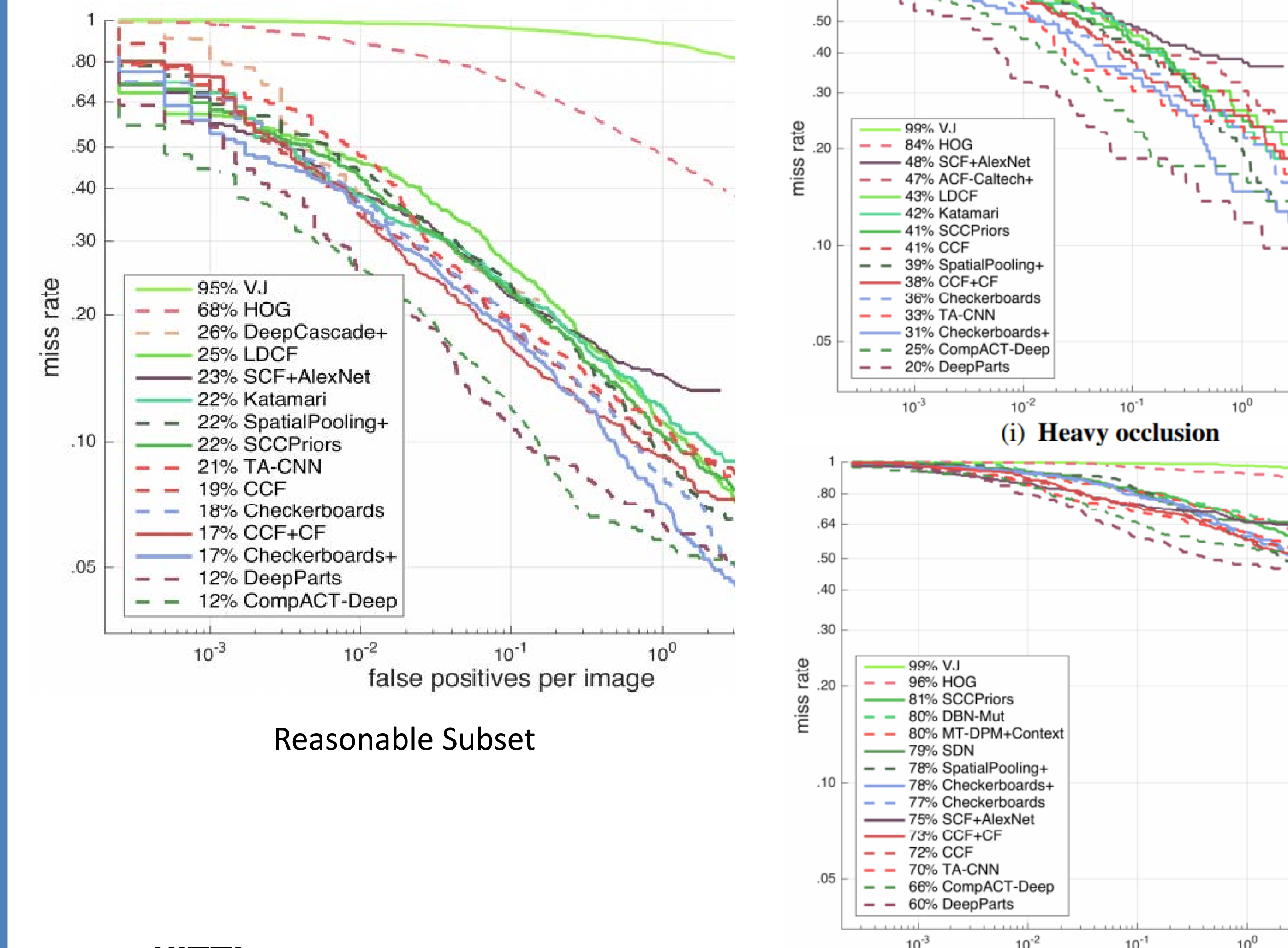


3 Results

Effectiveness of Part Detectors



Overall Results



KITTI

