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<td>MOCO</td>
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Caltech Pedestrian Dataset:
Evaluated Algorithms

- ACF: 18 channels AdaBoost INRIA. Evolution of ChnFtrs [source code].
- ACF++: channels AdaBoost Caltech.
- ACF-Caltech: channels AdaBoost Caltech.
- ACF+SDt: channels AdaBoost Caltech. SDt = Stabilized Dt (motion features).
- AFS: multiple linear SVM INRIA. Accelerated version of FeatSynth.
- AFS+Geo: multiple linear SVM INRIA. Variant of AFS with geometry constraints.
- CCF: deep AdaBoost Caltech.
- CCF+CF: deep+channels AdaBoost Caltech. Checkerboards + flow-based features from [42].
- Checkerboards+: channels AdaBoost Caltech.
- ChnFtrs: 17 channels AdaBoost INRIA.
- CompACT-Deep: multiple boosting Caltech.
- ConvNet: pixels DeepNet INRIA.
- Crosstalk: channels AdaBoost INRIA.
- DBN-Isol: HOG DeepNet INRIA.
- DeepParts: pixels DeepNet Caltech+.
- FastCF: channels AdaBoost INRIA/Caltech. 100 fps on a CPU.
- F-DNN: pixels DeepNet Caltech+.
- F-DNN+SS: pixels DeepNet Caltech+.
- FeatSynth: multiple linear SVM INRIA.
- FisherBoost: HOG+COV FisherBoost INRIA.
- FPDW: channels AdaBoost INRIA.
- FtrMine: channels AdaBoost INRIA.
- Franken: channels AdaBoost INRIA.
- HikSvm: HOG HIK SVM INRIA. Multiple occlusion specific models.
- HOG: HOG linear SVM INRIA. Boundary effect fixed since publication.
- HOG-LBP: HOG+LBP linear SVM INRIA.
- InformedHaar: channels AdaBoost INRIA/Caltech.
- JointDeep: color+gradient deep net INRIA/Caltech.
- Katamari: channels AdaBoost INRIA/Caltech.
- LatSvm-V1: HOG latent SVM PASCAL.
- LatSvm-V2: HOG latent SVM INRIA.
- LDCF: channels AdaBoost Caltech.
- LDCF+: channels AdaBoost Caltech.
- LFOV: pixels DeepNet Caltech.
- MLS: HOG AdaBoost INRIA.
- MOCO: HOG+LBP latent SVM Caltech.
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<td>MT-DPM + Context [54]</td>
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