

Learning People Detection Models from Few Training Samples

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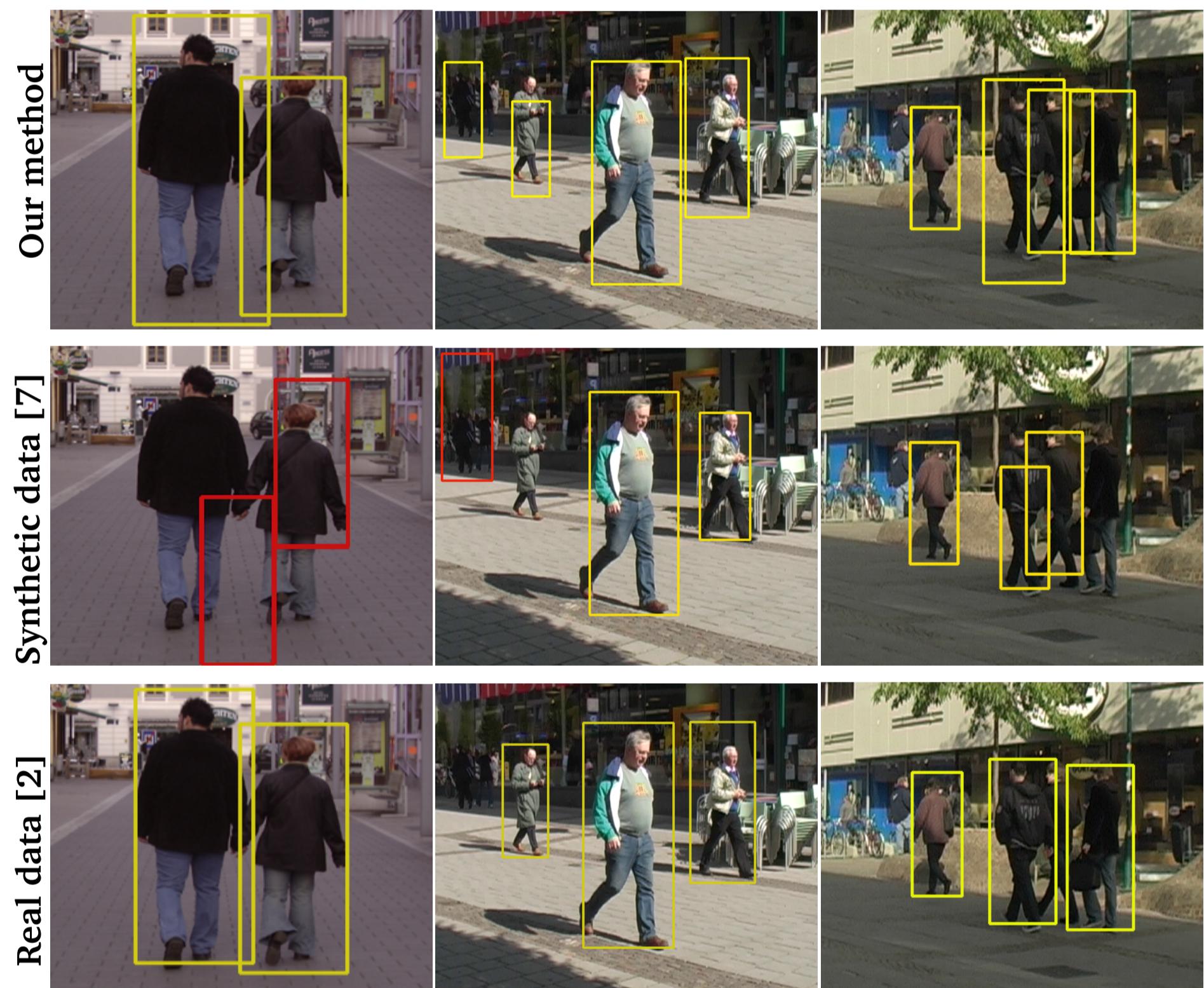


Goal

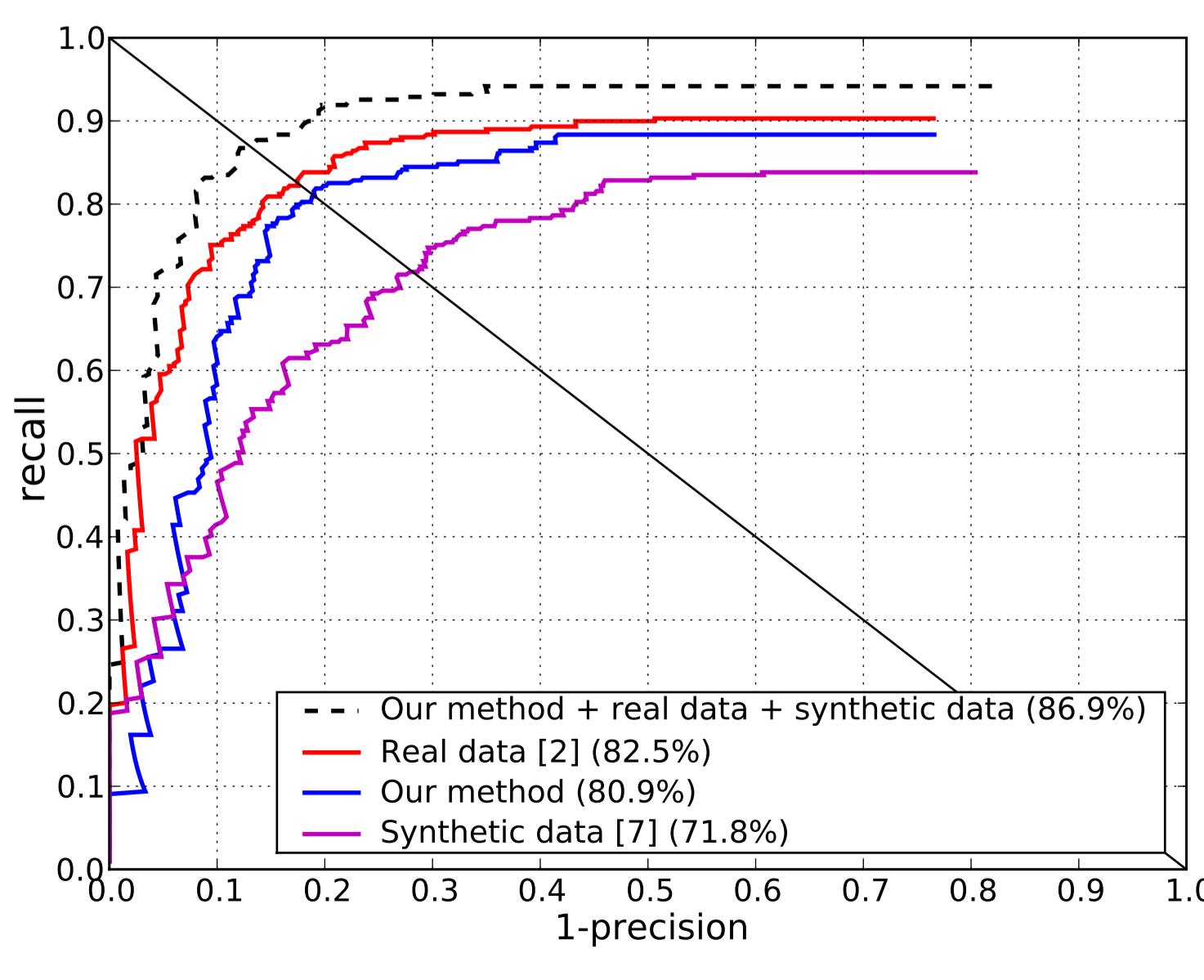
- Propose a novel technique to train people detectors from only a few observed training subjects
- Push the performance of current detection systems trained on hundreds of manually annotated pedestrians
- Approach the lack-of-training-data problem by automatically generating realistic training samples

Contributions

- Explore the applicability of state-of-the-art 3D human model to learn people detectors
- Compare the results to prior work (e.g. [2, 7])

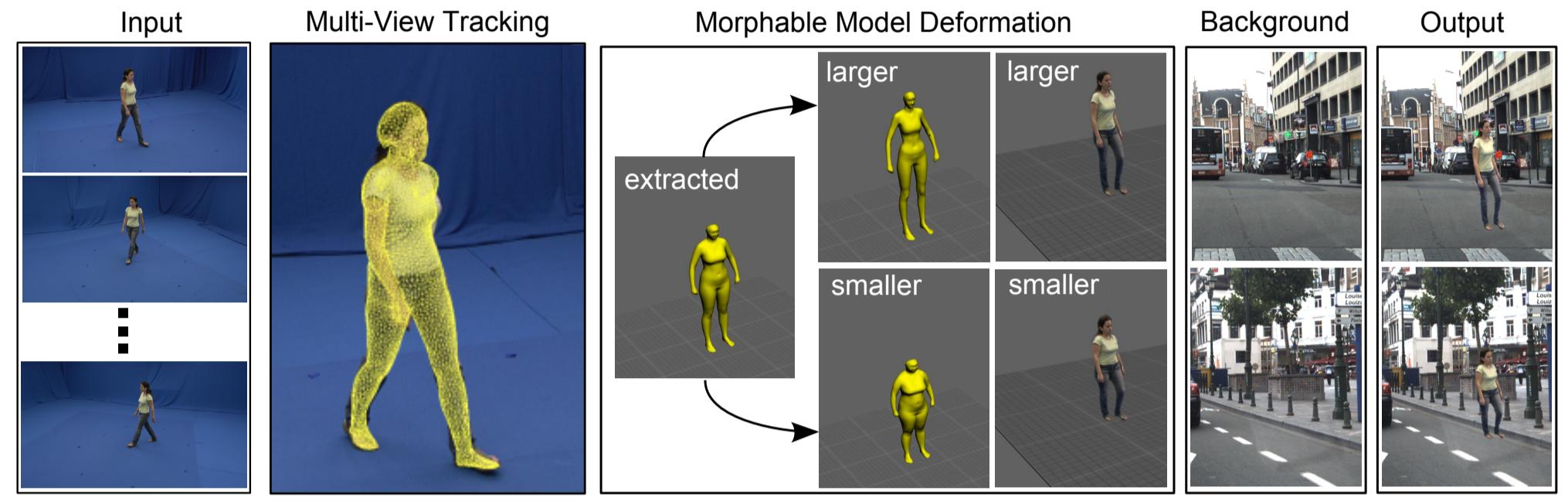


- Analyze various combinations of synthetic and real training data
 ⇒ outperform current methods which use real training data only



Proposed Approach

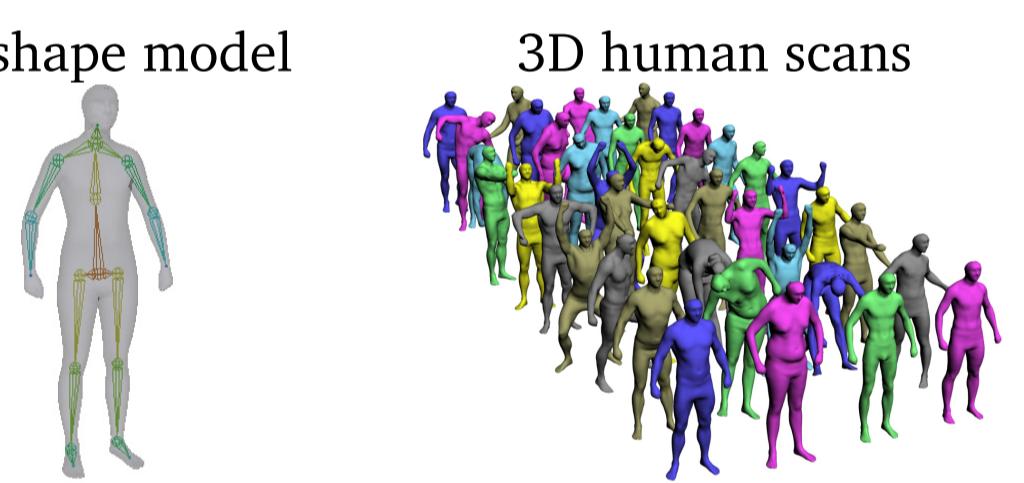
- Generate realistic synthetic data by MovieReshape [6]
- Combine reshaped humans with backgrounds
- Automatically obtain 2D part annotations from known 3D joint positions



⇒ Realistic distributions of human appearance and shape

Statistical 3D human shape model [5]

- Learn shape from 3D laser scans of humans
- Represent shape variations via PCA
- Embed kinematic skeleton with linear blend skinning



Automatic model fitting

- Fit the parameters of 3D body model to silhouettes
 ⇒ particle filter-based estimator

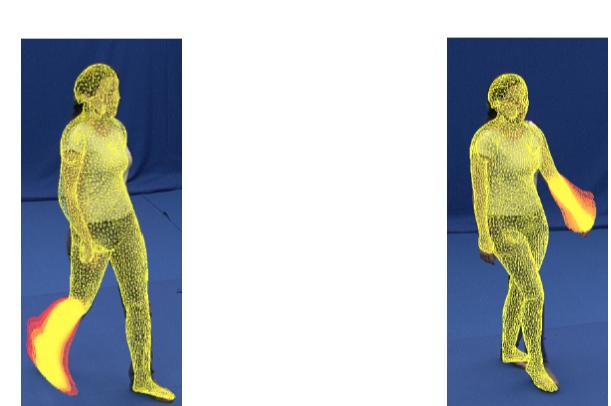
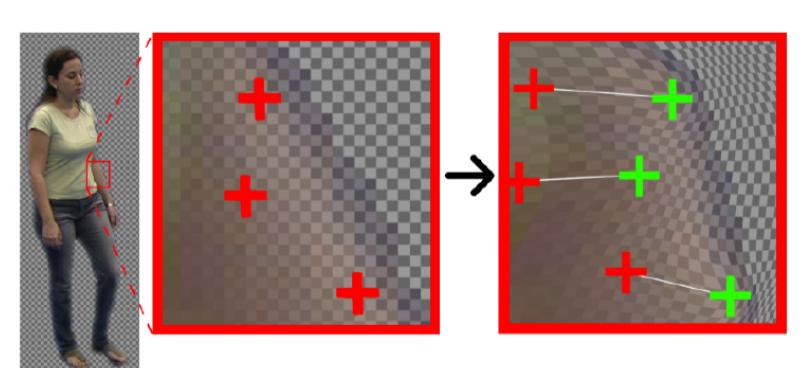


Image deformation

- Sample 3D shape parameters $\pm 3\sigma$ from the mean shape
- Use 3D offset vectors to drive 2D image warping



Composition with background

- Adjust color distribution of pedestrian w.r.t. background

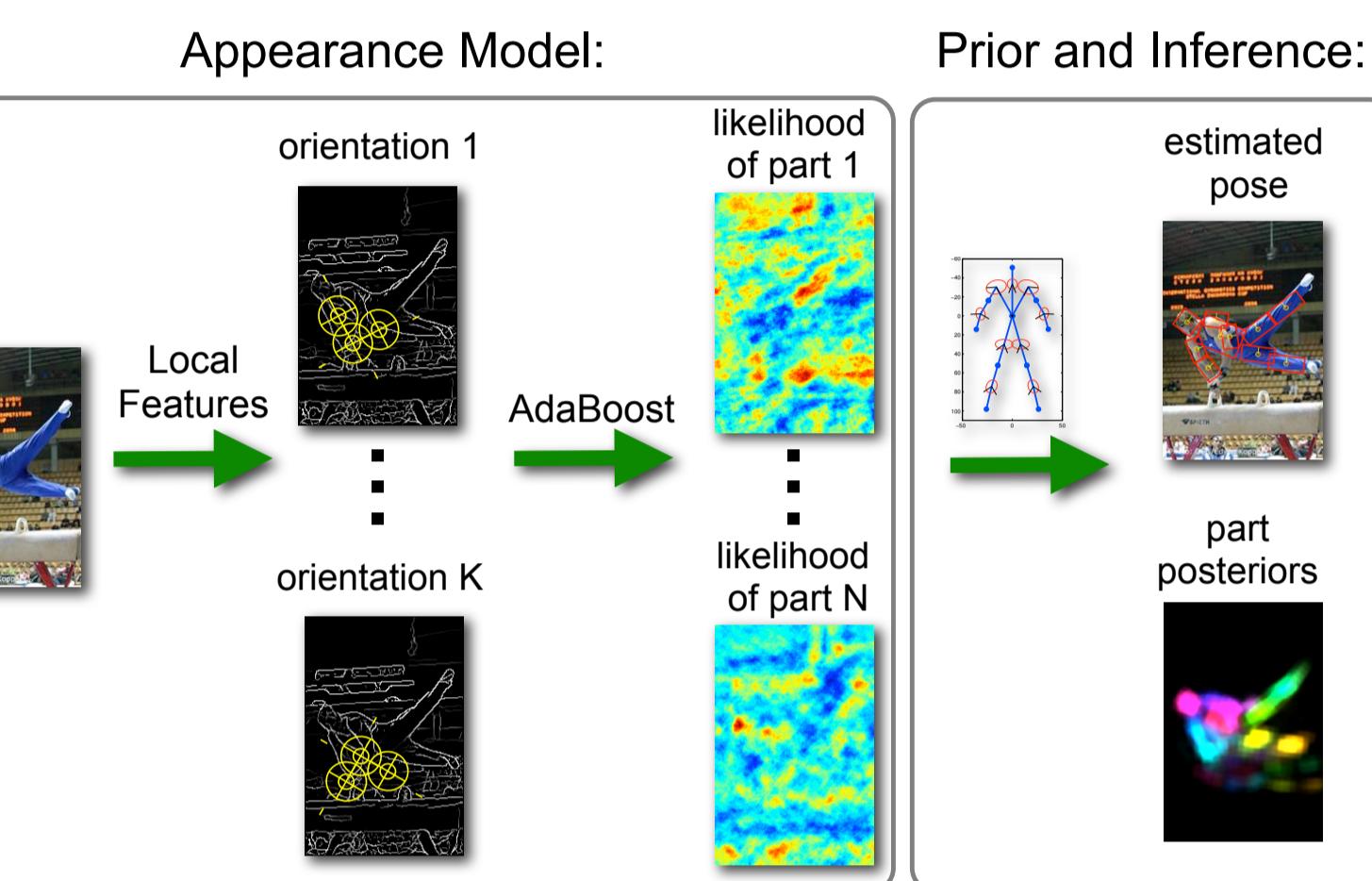
Sample output images with gradual height changes



People Detection Models

Pictorial structures (PS) [1, 4]

- Flexible configuration of body parts with pose prior
- AdaBoost part detectors learned from dense shape context descriptor
- Inference by sum-product belief propagation



Histograms of oriented gradients (HOG) [3]

- Sliding window detection
- Monolithic template based on HOG features
- Histogram intersection kernel SVM

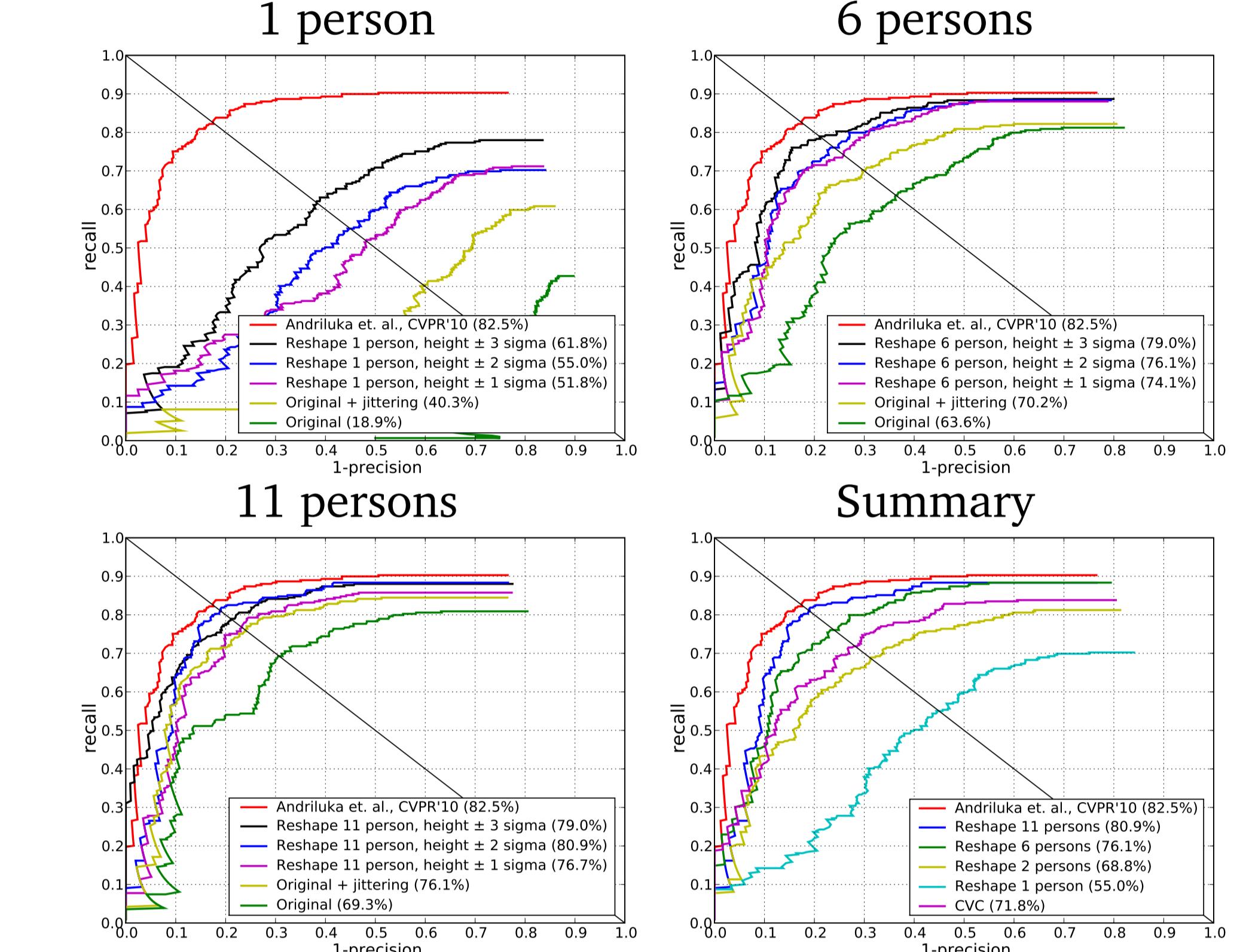
Datasets

- Reshape data (our method)*: 11 persons, ~ 2000 reshaped images per person
- CVC (virtual pedestrians)* [7]: 3432 images total
- Multi-viewpoint real data* [2]: 2972 train images, 248 test and 248 validation images

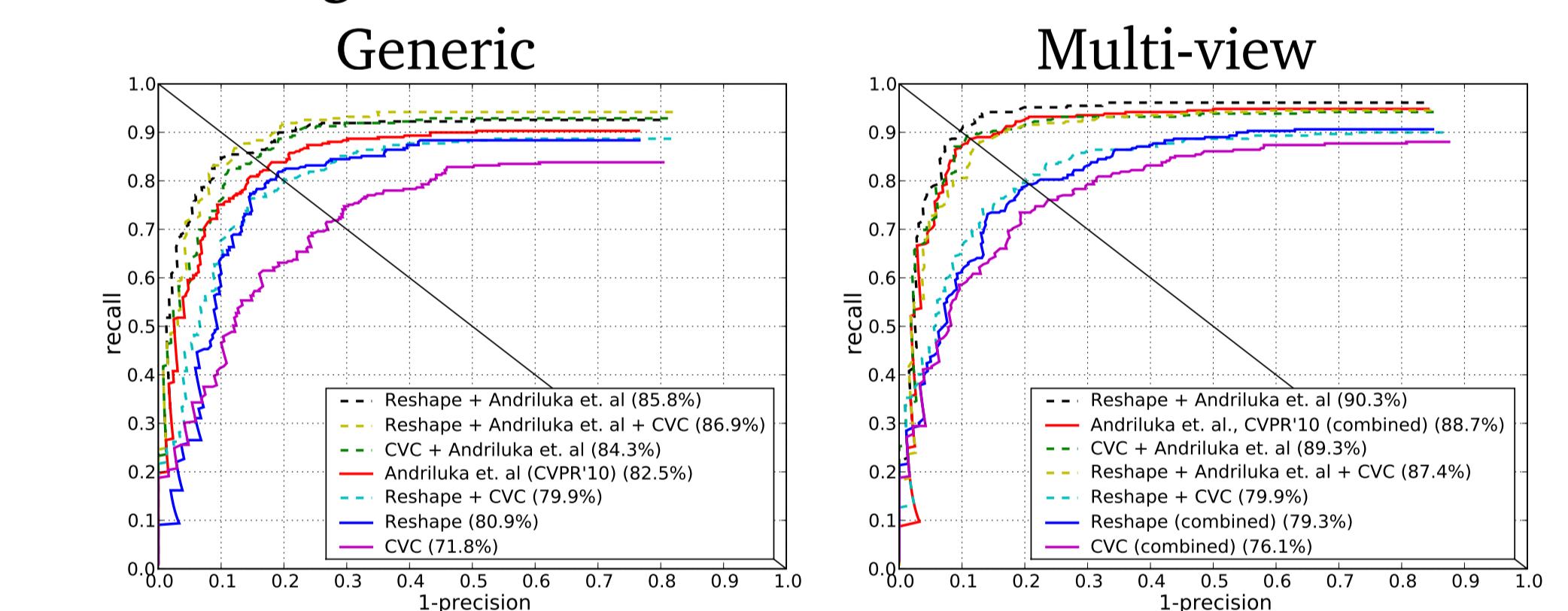


Results

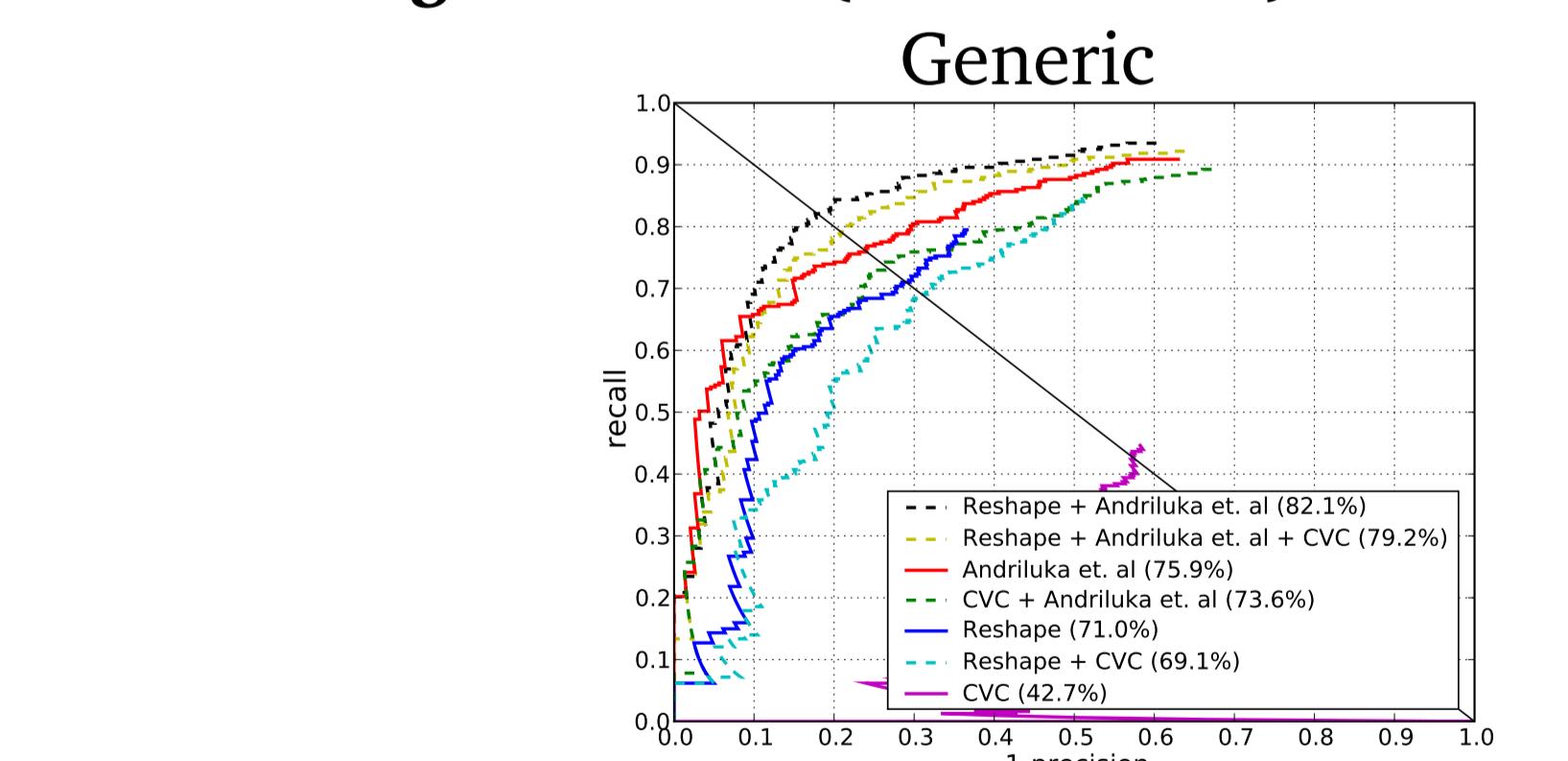
Using Reshape data (PS model)



Combining detectors (PS model)



Combining detectors (HOG model)



References

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