

# **Motivation:**

- Current algorithms are developed and evaluated on manually cropped dataset. In practice, images come from a detector which introduces larger inter-camera variations.
- $\succ$  Current algorithms optimize components of the reidentification pipeline individually.

Pedestrian detector



Feature extraction

Photometric 💊 Geometric 🔶 transforms



# **Contribution:**

- $\succ$  A new dataset providing bounding boxes from a pedestrian detector<sup>[1]</sup> and manually labeling.
- $\succ$  A new deep model for jointly optimizing all components in the re-identification pipeline. It is designed to be robust to imperfect detection.

# **A New Dataset:**

- $\succ$  We build a new dataset in more realistic settings.
  - Pedestrians images are automatically detected by detectors
  - 5 pairs of camera settings in uncontrolled environment
  - 1360 identities, each identity has ~4.8 images per view
  - Manually labeled bounding boxes are also provided



### DeepReID: Deep Filter Pairing Neural Network for Person Re-Identification Wei Li Rui Zhao Tong Xiao Xiaogang Wang

# **Deep Neural Network for ReID:**







 $\blacktriangleright$  Jointly optimizing key components of re-identification pipeline: *feature extraction*, *photometric transformations*, geometric transformations.

- **Feature Extraction** is modeled by two parallel convolutional layers and multiple filter pairs.
- **Photometric Transformation** is modeled by height factoring and maxout grouping layers, and the difference detected by paired filters.









- displacement matrices.
- connected layer.

### **Experiments:**



## Detected



### **Reference:**

[1] P. Felzenszwalb, R. Girshick, D. McAllester, D. Ramanan,



**Geometric Transformation** is modeled by convolution on

> A **Mixture of Transformations** are modeled by fully

### Labeled

 $\blacktriangleright$  Our model works best on both detected and labeled samples. Our model drops 0.76% (relatively -3.7%), while the second best (KISSME) drops 2.47%(-17.4%) when automatically detected bounding boxes are used.

<ul> <li>7.87%)</li> <li>0.52%)</li> <li>7.10%)</li> <li>1.17%)</li> <li>0.61%)</li> <li>6.45%)</li> <li>9.90%)</li> </ul>	<ul> <li>Our model performs reasonably well on a much smaller dataset:</li> <li>971 identities</li> <li>2 images per identity per view</li> </ul>
0.61%) 6.45%) 9.90%) 2.83%) 9.40%)	perview
100	