Poselets: Body part detectors trained using 3d human pose annotations (ICCV 2009)

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Person detection results on the PASCAL challenge data

	Poselets	Second-highest score
VOC 2010	48.5	47.5 ***
VOC 2009	48.6	47.9 ***
VOC 2008	54.1	43.1 **
VOC 2007	46.9	43.2 *

* P. Felzenszwalb, R. Girshick, D. McAllester, D. Ramanan <u>Object Detection with</u> <u>Discriminatively Trained Part Based Models</u>, (Release 4, 2010)
** P. Felzenszwalb, R. Girshick, D. McAllester, D. Ramanan <u>Object Detection with</u> <u>Discriminatively Trained Part Based Models</u>, PAMI (preprint, 2009)
*** P. Felzenszwalb, R. Girshick, D. McAllester, D. Ramanan, PASCAL VOC 2010 competition

What are poselets?

Poselets are discriminative parts - not necessarily semantic.

- Requirements:
 - Should tell us about the 3D pose
 - Should be easy to find from a 2D input image

What poselets look like



Slide credit: [Bourdev & Malik, ICCV09]

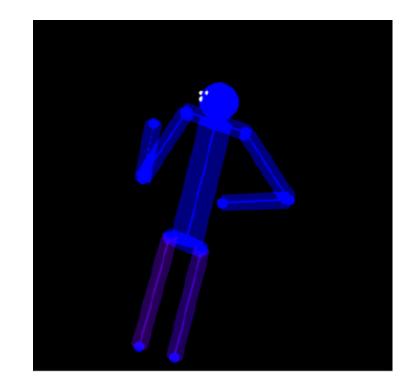
What are poselets?

- Poselets are discriminative "parts", not necessarily semantic that describes parts of a pose.
- Requirements:
 - Should tell us about the 3D pose
 - Should be easy to find from a 2D input image

How do we enforce these requirements?

Configuration space

- Lost in transfer to 2D
- Fixed number of degrees of freedom
- Specified completely by positions of joints in a 3D coordinate space registered to the camera



"Should tell us about the 3D pose" = Tightly clustered in configuration space

Appearance space

- Pixel values i.e. the image itself
- Clothing, illumination, occlusion, background clutter etc.



"Should be easy to find from a 2D input image" = Should be tightly clustered in appearance space

Poselets



Poselets capture part of the pose from a given viewpoint

Slide credit: [Bourdev & Malik, ICCV09]

Poselets



Examples may differ visually but have common semantics

Slide credit: [Bourdev & Malik, ICCV09]

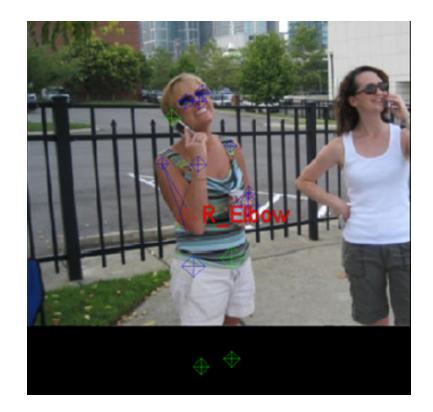
Poselets



But how are we going to create training examples of poselets? Slide credit: [Bourdev & Malik, ICCV09]

H3D dataset – humans in 3D

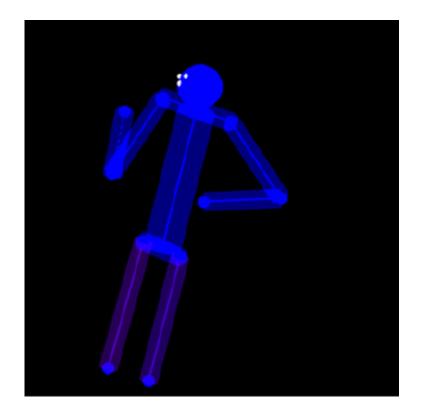
- 2000 annotated people
- 19 keypoint annotations



Appearance space annotation

H3D dataset – humans in 3D

- 2000 annotated people
- 19 keypoint annotations



Configuration space annotation

H3D dataset – humans in 3D

- 2000 annotated people
- 19 keypoint annotations
- 15 regions –
 "face", "hair" etc.



Region label annotation

How do we train a poselet for a given



Slide credit: [Bourdev & Malik, ICCV09]

nces at training time

How do we find a similar

pose configuration in the

Given part of a human pose

KUNA

Slide credit: [Bourdev & Malik, ICCV09]

training set?

Finding correspondences at training time



Slide credit: [Bourdev & Malik, ICCV09]

Finding correspondences at training time



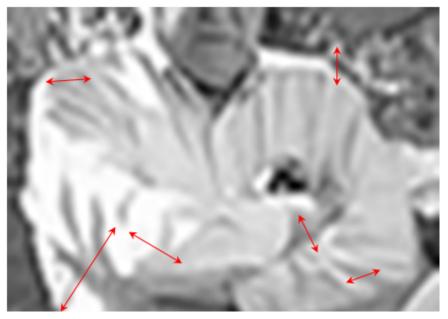






Finding correspondences at training time

Residual Error





Sum over keypoints

$$\begin{aligned} d_s(r) &= \sum_i w_s(i) || \mathbf{x_s}(i) - \mathbf{x_r}(i) ||_2^2 (1 + h_{s,r}(i)) \\ exp(-\mathbf{x_s}(i)^2 / (2\sigma^2)) & \text{O if visible or invisible in both} \\ \sigma \text{ otherwise} \end{aligned}$$

Registering candidate matches







Similarity transformation (4DOF):

- X translation: t_x
- Y translation: t_y
- Rotation: α
- Scaling: s

Registering candidate matches



$$d_{s}^{*}(r) = \min_{t_{x}, t_{y}, \alpha, s} d_{s} \left(r_{t_{x}, t_{y}, \alpha, s} \right)$$
$$x_{r_{t_{x}, t_{y}, \alpha, s}} = transform(x|t_{x}, t_{y}, \alpha, s)$$



Seed s

Example r

$$d_s(r) = \sum_i w_s(i) ||\mathbf{x}_s(i) - \mathbf{x}_r(i)||_2^2 (1 + h_{s,r}(i))$$

Matching done in 3D => clustered in configuration space

Training poselet classifiers











Residual
Error:0.150.200.100.850.150.35

- 1. Given a seed patch
- 2. Find the closest patch for every other person
- 3. Sort them by residual error
- 4. Threshold them

Slide credit: [Bourdev & Malik, ICCV09]

Training poselet classifiers

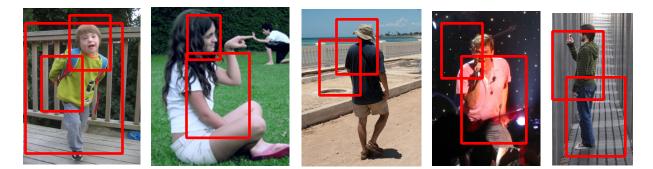


- ^{1.} Given a seed patch
- 2. Find the closest patch for every other person
- 3. Sort them by residual error
- 4. Threshold them
- ^{5.} Use them as positive training examples to train a **linear SVM with HOG features**

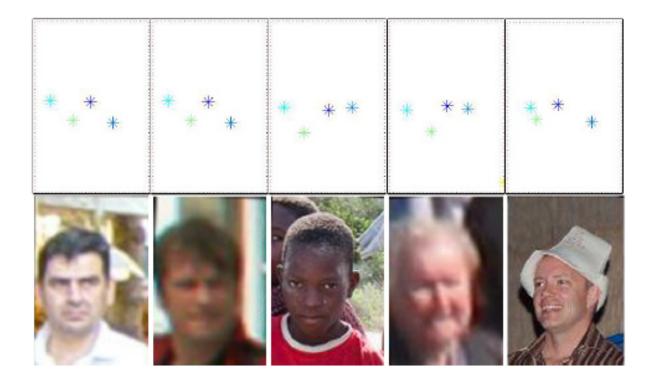
Slide credit: [Bourdev & Malik, ICCV09]

Which poselets should we train?

- 96x64 scanning window over all scales and orientations on 1500 training images.
- 120K poselet candidates generated.
- Strict 2-stage pruning (300 poselets):
 - Individually effective (good cross-validation)
 - Complementary (large pairwise distances)



Selected poselets

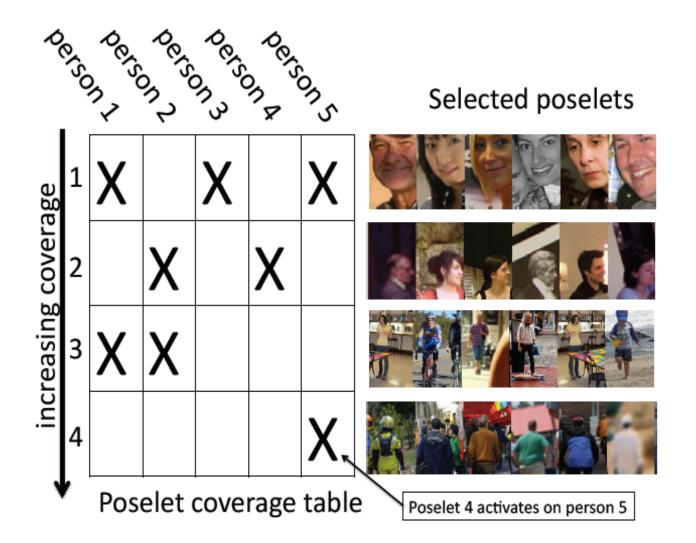


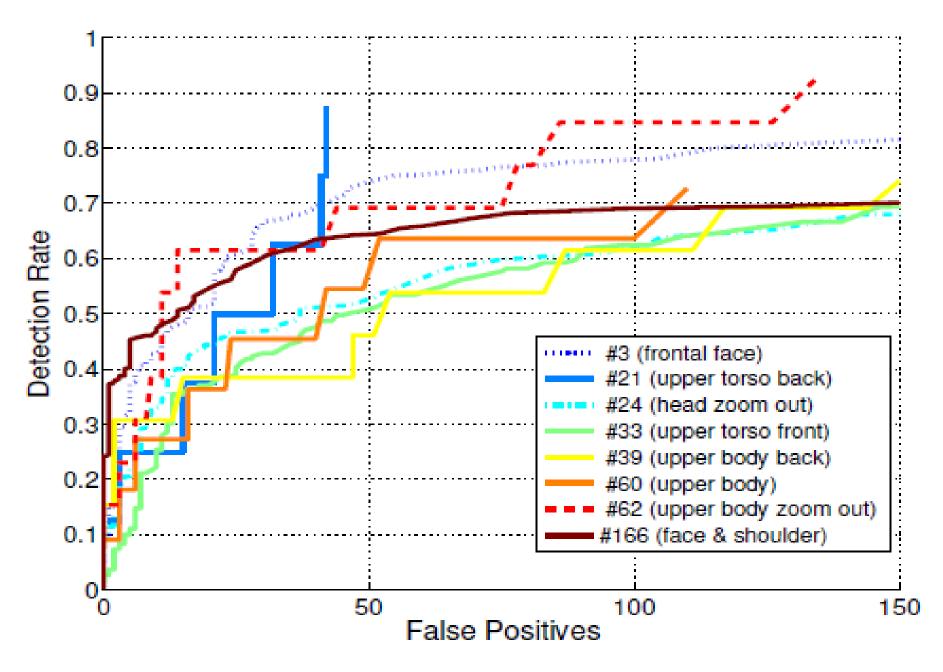
Configuration space

Examples

A frontal face poselet

Selected poselets





Poselets in isolation – "individually effective"

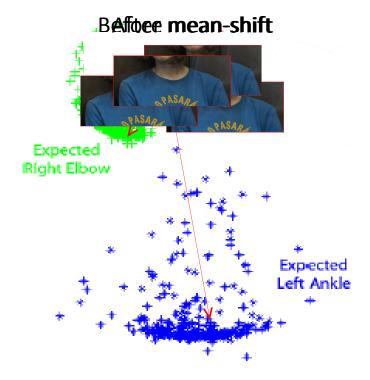
Discussion

• Why do upper body poselets do better?

• Why do we want complementarity?

Detecting torsos and localizing keypoints – The Hough transform

- H3D can find expected keypoint location given poselet detection.
- Poselet detector in scanning window, mean-shift for nonmaximum suppression



Discussion

• What is the dimension of the voting space?

Does a poselet predict *all* keypoints equally well?

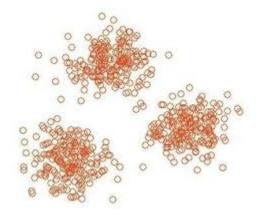
• Are all poselets equal?

Combining poselet predictions

 $P(O|x) \propto \sum w_i a_i(x)$

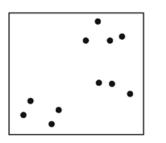
Poselet index

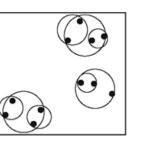
Poselet i vote for O being at x

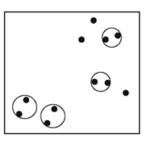


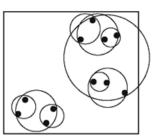
2D Hough space (keypoint)

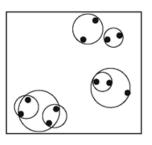
Agglomerative clustering in Hough space

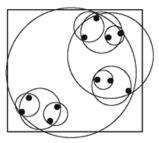












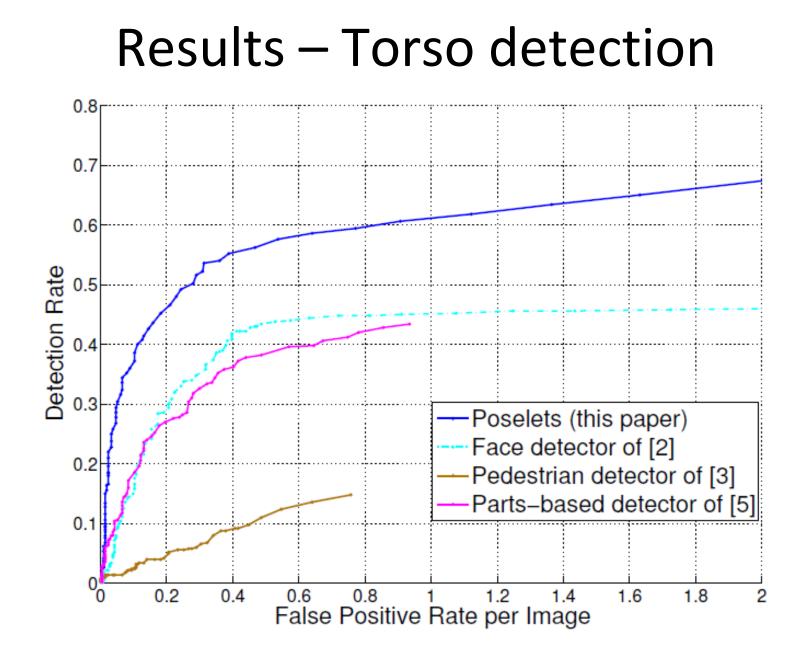
Learning task-specific weights: Max-Margin Hough transform

 SVM-like formulation that classifies truepositive clusters versus false-positive clusters using poselet votes as descriptors:

$$\min_{w,b,\xi} \frac{1}{2}w^Tw + C\sum_{i=1}^T \xi_i$$

s.t. $y_i(w^TA_i + b) \ge 1 - \xi_i$
 $w \ge 0, \xi_i \ge 0, \forall i = 1, 2, ..., N$

1 if true positive peak, 0 else Score of poselet j in Hough peak i



Key strengths

- Discussion: Why does this method do so much better than the next best?
 - Tightly controlled poselet training and selection process enabled by H3D
 - Large number of poselets guaranteed to be semantically same vs. stick-figure approach
 - Lots of occlusions in test set => Hough transform based voting is advantageous

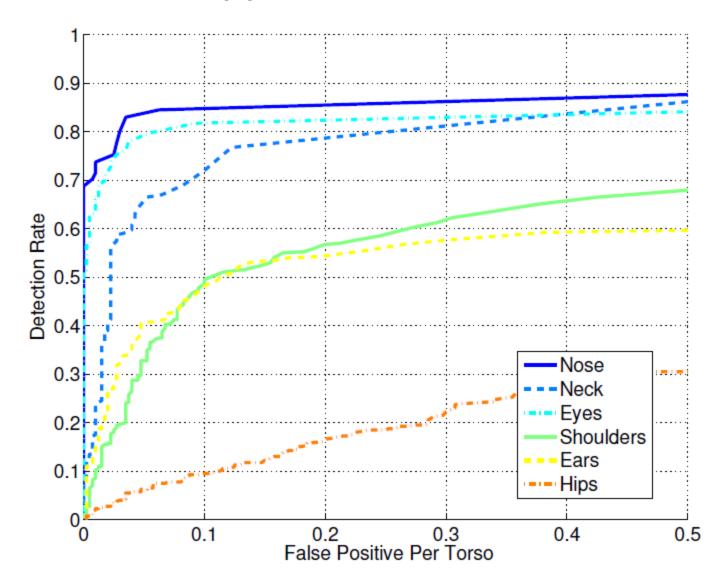
Results - Torso detection



Pascal VOC person detection

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Keypoint detection



Discussion

- Why are the nose, eyes and neck predicted best?
- A 2D projection can represent more than one 3D configuration - metamers. Are metamers really handled at all by the poselets method?
- Did we really collect 3D data?
- What advantage did 3D really give us?
- Can we do without 3D annotations?

Poselets website

http://eecs.berkeley.edu/~lbourdev/poselets

The set of published poselet papers H3D data set + Matlab tools Java3D annotation tool + video tutorial Matlab code to detect people using poselets Latest trained poselets: <u>http://</u> <u>www.cs.berkeley.edu/~lbourdev/poselets/</u> <u>poselets_person.html</u>

Slide credit: [Bourdev &