## Latent Hough Transform for Object Detection

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#### **Detection with the Hough Transform**

#### Accumulation of inconsistent votes



Hough Space (position and scale)



How to enforce consistency of votes?

Voting for viewpoint?





• Voting for viewpoint?



#### votes inconsistent in viewpoint are not accumulated



• Voting for type?





#### What about color, aspect ratio, etc.?

### **Previous Works**

#### Voting for other attributes:

- pose (Seemann'07)
- viewpoint (Thomas'06, Razavi'10)
- depth (Sun'10)
- shapes (Marszalek'08)
- etc.

#### • But ....

- What attribute to choose?
- How to quantize it?
- There is also a cost of annotations
- We cannot use all attributes together
  - HT does not work well on high dimensions (Stephens'91)

# Can we learn the attributes to be consistent over?

### Hough Transform

Hough Space



### Hough Transform



#### Latent Hough Transform



• Every vote is a patch in a training image (Leibe'08)

#### Training Image



• Every vote is a patch in a training image (Leibe'08)

#### Latent Space



• Every vote is a patch in a training image (Leibe'08)

#### Latent Space



- Every vote is a patch in a training image (Leibe'08)
- A latent grouping can be represented as a matrix



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**Disjoint Groups** 

- Every vote is a patch in a training image (Leibe'08)
- A latent grouping can be represented as a matrix



• The number of votes is very large (~1 M)

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  → we can pre-group them together (~1000)





Training Image 2

- The number of votes is very large (~1 M)
- Votes from the same training image are all consistent

 $\rightarrow$  we can pre-group them together (~1000)



#### Interesting Special Cases of our Model

- Single Row
  - Hough transform with weighted training examples
    Related to Max Margin HT (Maji'09, Zhang'10)



- Clustering/Annotations
  - Disjoint grouping with a {0,1} matrix
  - Related to Latent SVMs: (Felzenszwalb et al.'10)



- One training image per group:
  - Related to Exemplar-SVMs (Malisiewicz et al.'11)



- Uniform weights
  - Equals a single group



### **Discriminative Learning of W**

$$\hat{W} = \operatorname*{arg\,max}_{W} O(W, R).$$

#### O(W,R) Objective: average precision on the validation set R

Our objective is non-convex and not even continuous

We do global optimization with a variation of simulated annealing

### Experiments

- Setup
  - Two datasets:
    - ETHZ cars dataset (Leibe et al.'06)
    - PASCAL VOC 2007 (Everingham et al.'07)
  - Pre-train a codebook per category only once
    - Using Hough Forests (Gall and Lempitsky'09)
    - The codebook and the offset stay identical
  - Learning W using the validation set

### Learning or Annotation?

- ~3000 training images, annotated for 14 views
- Testing on Leuven video sequence (Leibe'07)





















### **Disjoint or Shared Groups?**





#### **Disjoint Groups**

Shared Groups

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#### **Disjoint Groups**

Shared Groups

### **Overall Results**

#### PASCAL VOC 2007



### Contributions

 Introduced Latent Hough Transform to enforce consistency of the votes

 Discriminative learning of the latent space for object detection

State-of-the-art performance for voting based methods

### Visualization of Groups



#### Ignored Examples













#### 1<sup>st</sup> Group

















#### 2<sup>nd</sup> Group









#### Ignored Examples





#### Ignored Examples















### Thank You!