





# Knowing a Good HOG Filter When You See It: Efficient Selection of Filters for Detection Ejaz Ahmed<sup>1</sup>, Gregory Shakhnarovich<sup>2</sup> and Subhransu Maji<sup>3</sup>

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### **Problem** :

- Fast automatic filter selection method.
- Selected filters should be discriminative and diverse.
- Learn universal model of filter "goodness".
- Beneficial for large number of methods which rely on collection of filters.

#### **Visual Categories as Collection of Filters :**



Poselets

Candidate

Generation





**Discriminative Patches** 

**Common Architecture :** 



Expensive

Evaluation







Exemplar SVMs

Run as detector

Selected Filters (n)

(n << N)

















**ESVMs** 



```
from f_{g,j}
```

**Selected Filters** (w , λ) **Test Category** 

selected



- > Automatically selects discriminative and non redundant filters.
- evaluation fast

By passes explicit

**Category Independent Model:** 



Pool of Filters (N)



Rank (lda)

Rank (svm)

Poselet

31.5

31.6

 $\succ \Delta_{g,i,j} = y_{g,i} - y_{g,j}$ , for i > j measures how much better  $f_{g,i}$  is  $\min_{\mathbf{w}} \frac{1}{2} \|\mathbf{w}\|^2 + C \sum_{g=1}^{G} \sum_{i=1}^{N-1} \sum_{j=i+1}^{N} \left[ 1 - \left\langle \mathbf{w}, \delta \phi_{g,i,j} \right\rangle \right]_+ \Delta_{g,i,j}$ 

### Selecting a Diverse Set of Filters:

 $\succ$   $x_i \in \{0,1\}, i \in \{1, \dots, N\}$  indicator variable  $\succ$   $A_{ii}$  similarity between filter i and j

$$\max_{\mathbf{x}\in\{0,1\}^N, \sum_i x_i=n} \sum_i \widehat{y}_i x_i - \lambda \sum_i \max_{j\neq i} A_{ij} x_i x_j.$$



Results :						
ion		/OC 2007 test		Training sppedup		
	ſ	MAP	δΜΑΡ	Initial	Selection	Overall
	2	9.03				
26		6.66	-2.37	8x	8x	8x
27.		7.78	-1.25	1x	4.4x	2.4x
2		7.38	-1.65	1x	8x	3x
2		8.34	-0.69	1x	8x	3x
2		7.53	-1.50	1x	8x	3x
	2	8.51	-0.52	1x	8x	3x
	2	7.81	-1.22	1x	8x	3x
	29.04		+0.01	1x	8x	3x
	28.19		-0.84	8x	8x	8x
29.46		9.46	+0.43	8x	8x	8x
anking			ESVM	ESVM Detection		
100	150	200	Method			
52.8	68.6	80.0	Oracle	Oracle		
52.0	00.0	80.0	Random			18.53
52.2	67.9	79.7	Freq			16.23
54.3	70.2	80.2	Rank(lda	Rank(Ida)		
	74.0	81.1	Rank(lda) + Freq			18.75
55.4	71.2		Rank(Ida	Rank(Ida) + Freq + Div		



## **Take Home Message:**

Fast automatic filter selection method using intrinsic properties of filters