

# Introduction to Computer Vision

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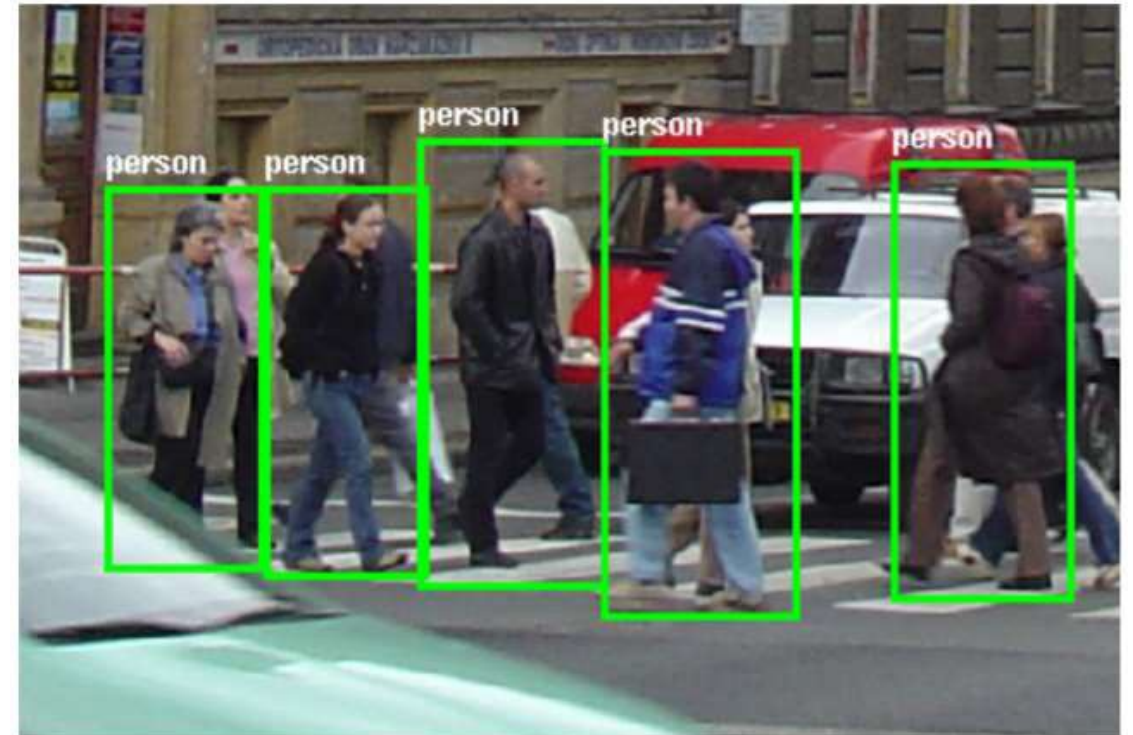
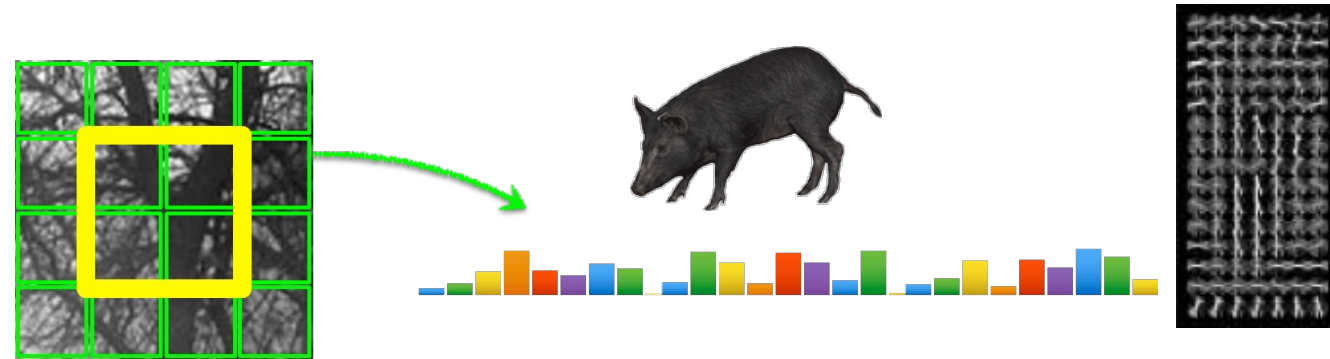
**Kaveh Fathian**

Assistant Professor  
Computer Science Department  
Colorado School of Mines

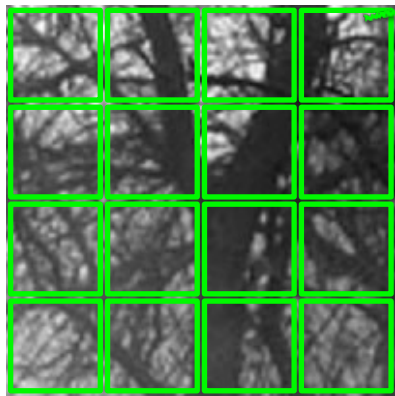
**Lecture 16**

# Learning Objectives

- HOG Descriptors
- HOG Pedestrian Detection



# HOG Descriptors



8x8 cell

30	20	10	...				
40	60	70					
50	70	60					
⋮			⋮				

	20		...				
40	60	70					
	70						
⋮			⋮				

Compute gradient:

Outperforms

-1	0	1
----	---	---

centered

-1	1
----	---

uncentered

0	1
-1	0

diagonal

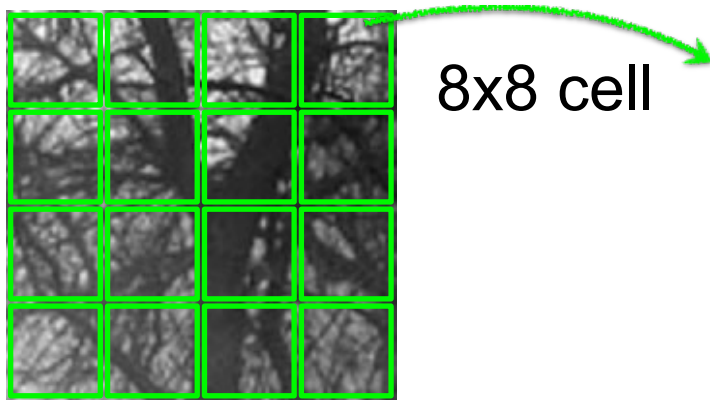
1	-8	0	8	-1
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cubic-corrected

-1	0	1
-2	0	2
-1	0	1

Sobel

# HOG Descriptors

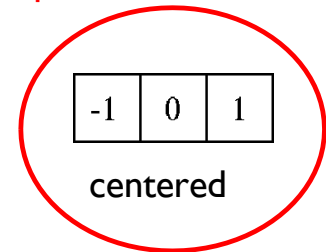


30	20	10	...				
40	60	70					
50	70	60					
⋮			⋮				

	20		...				
40	60	70					
	70						
⋮			⋮				

Compute  
gradient

Outperforms



$$X \text{ dir: } -40 + 70 = 30$$

$$Y \text{ dir: } -20 + 70 = 50$$

$$\text{Grad mag: } \sqrt{30^2 + 50^2} = 58$$

$$\text{Grad deg: } \tan^{-1}(50/30) = 60^\circ$$

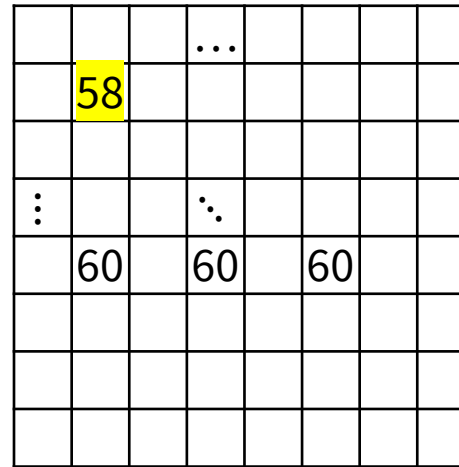
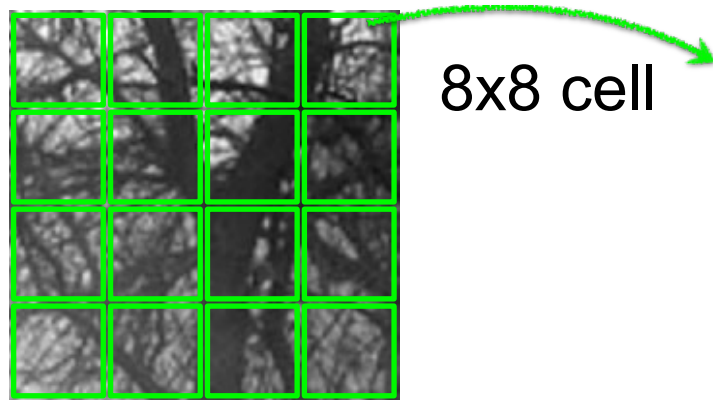
			...				
	58						
⋮			⋮				

Grad magnitude

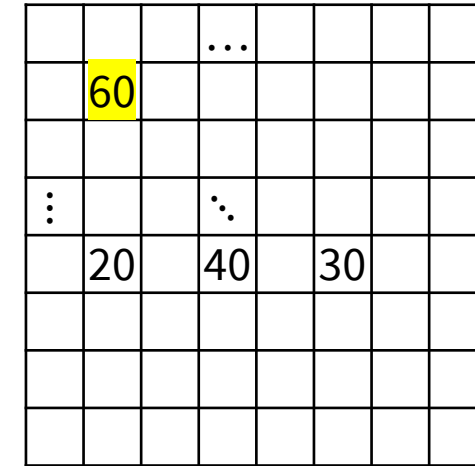
			...				
	60						
⋮			⋮				

Grad angle

# HOG Descriptors

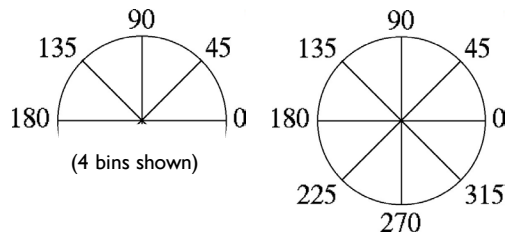


Grad magnitude



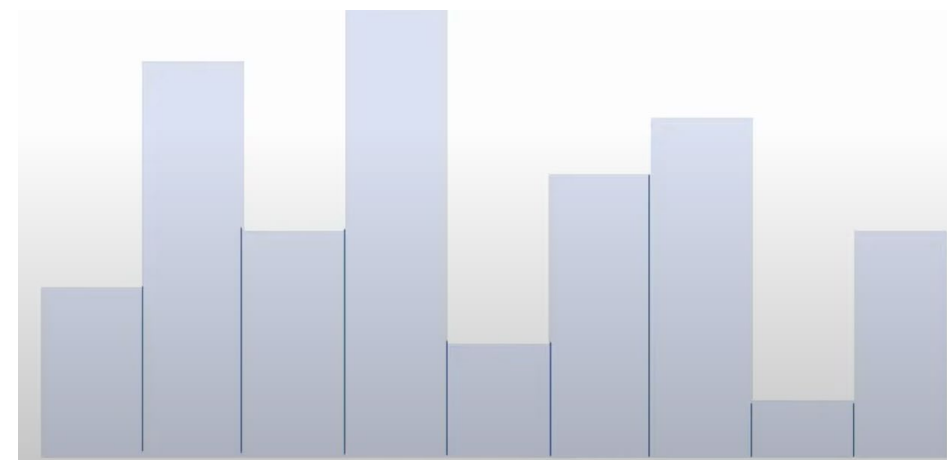
Grad angle

Orientation: 9 bins  
(for unsigned angles 0-180)



Outperforms

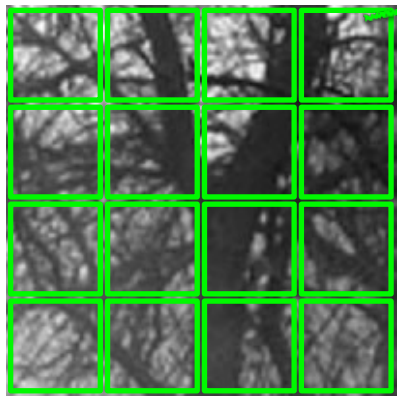
Unsigned angles:  
Signed angles:



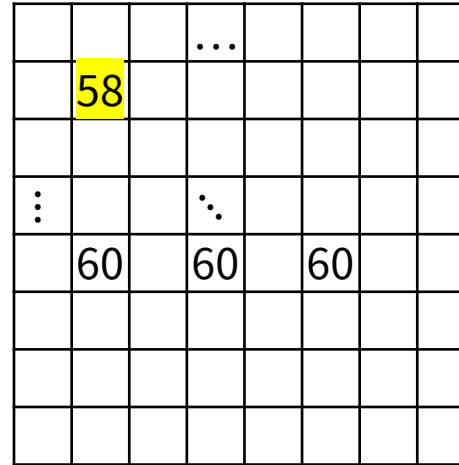
0	20	40	60	80	100	120	140	160
0	40	80	160	200	240	280	320	360



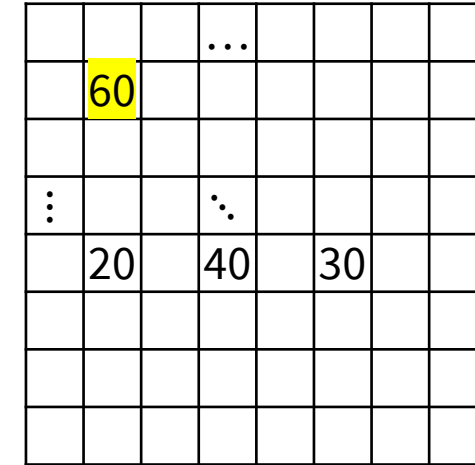
# HOG Descriptors



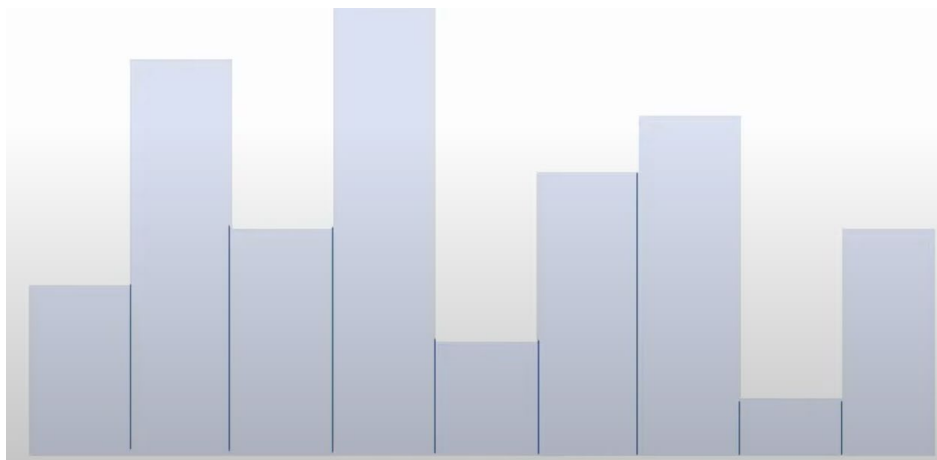
8x8 cell



Grad magnitude

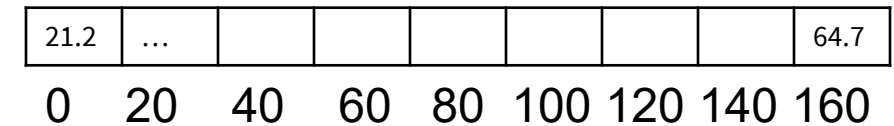


Grad angle

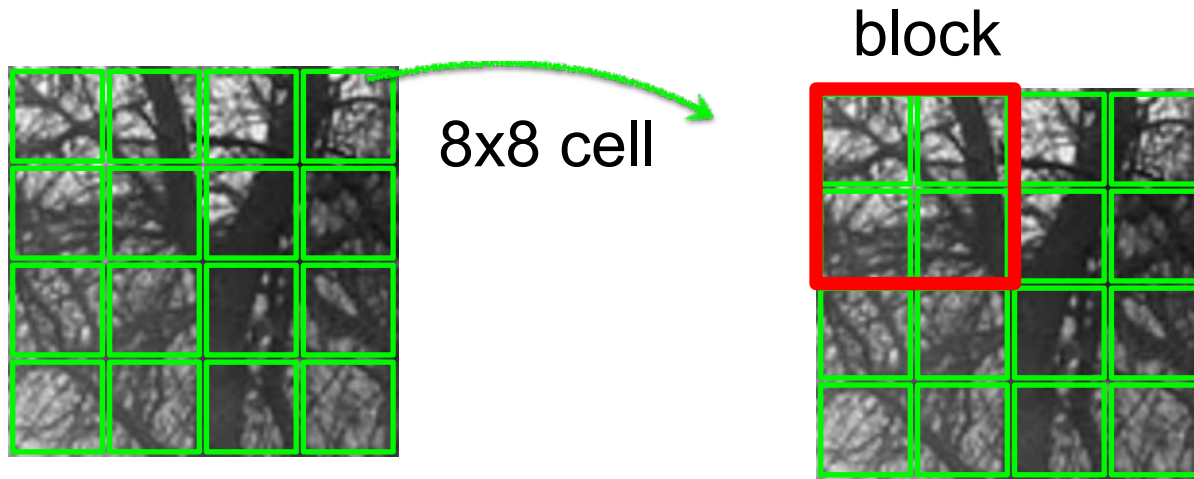


Unsigned angles: 0 20 40 60 80 100 120 140 160  
Signed angles: 0 40 80 160 200 240 280 320 360

Descriptor for cell



# HOG Descriptors



Group 2x2 cells into blocks and normalize their descriptor:

Each block produces a vector:  
 $(4 \text{ cells}) \times (9 \text{ bins}) = 36\text{-dimensional block descriptor}$

21.2	...							64.7
21.2	...							64.7
21.2	...							64.7
21.2	...							64.7
0    20    40    60    80    100    120    140    160								

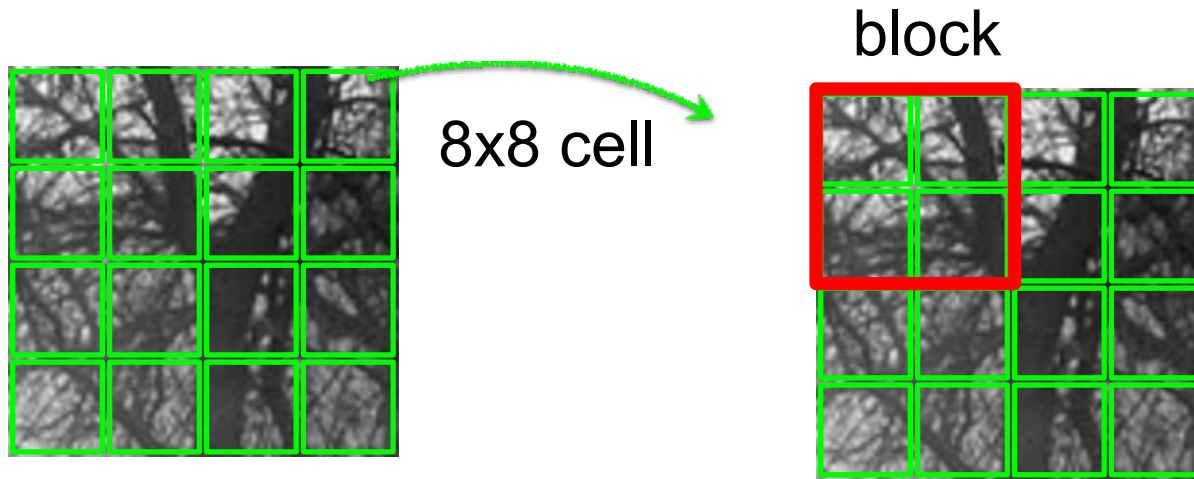
Group & normalize (L2 norm)

0.1	...							0.3
-----	-----	--	--	--	--	--	--	-----

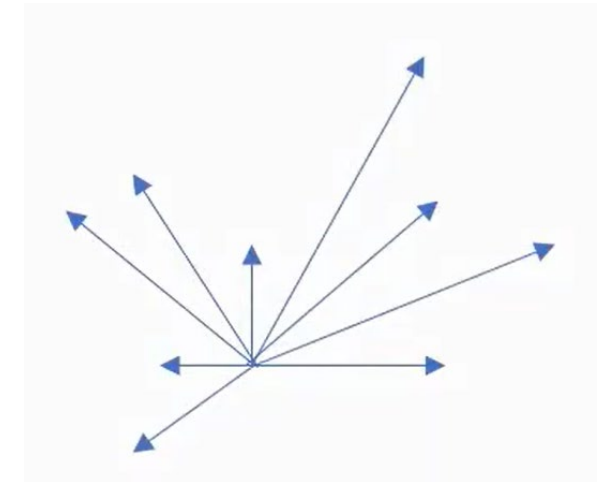
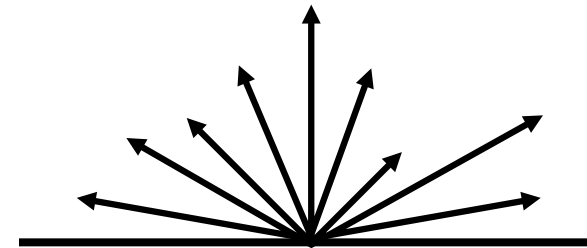
HOG descriptor



# HOG Descriptors



HOG descriptor  
for block

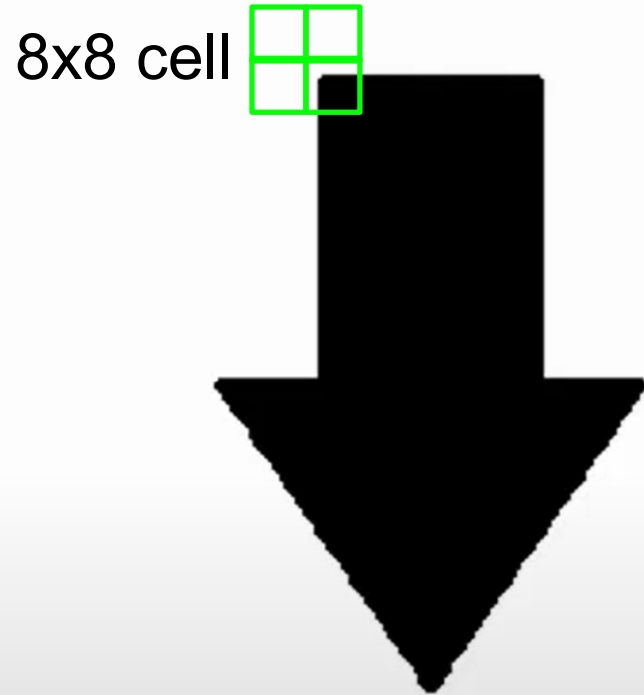


Visualization



# HOG Example

Input Image



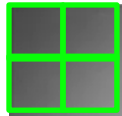
HOG Descriptors



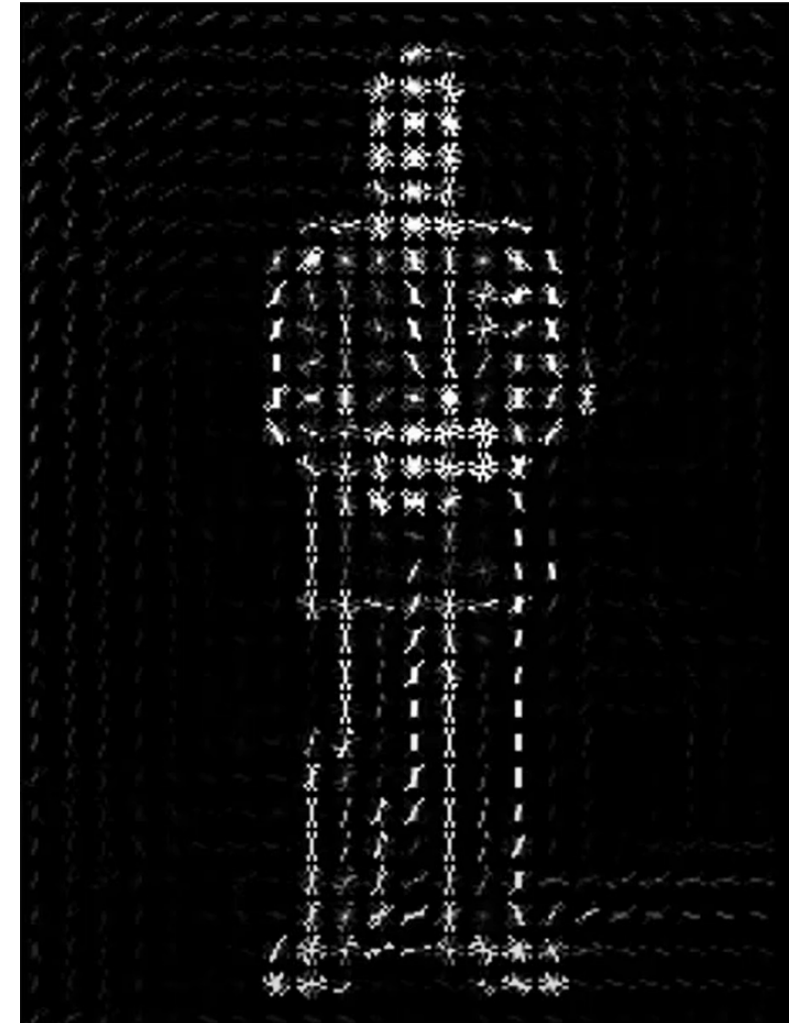
# HOG Example

Input Image

8x8 cell



HOG Descriptors



# Dalal-Triggs Pedestrian Detector

## Algorithm

1. Extract fixed-sized (64x128 pixel) window at each position and scale
2. Compute HOG (histogram of oriented gradient) features within each window
3. Score the window with a linear SVM classifier
4. Perform non-maxima suppression to remove overlapping detections with lower scores



Navneet Dalal and Bill Triggs, Histograms of Oriented Gradients for Human Detection, CVPR05

# Extracting HOG Features



150x300



64x128



8x16

8x8 cell

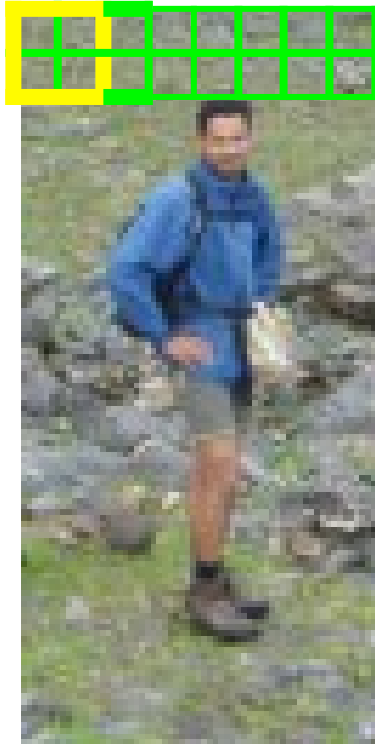
30	20	10	...				
40	60	70					
50	70	60					
⋮			⋮				

# HOG Pedestrian Detection

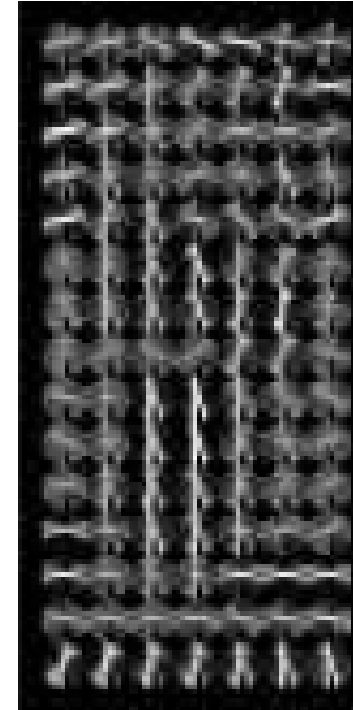
1 cell step size

visualization

128 pixels  
16 cells  
15 blocks



$$15 \times 7 \times 4 \times 9 = 3780$$



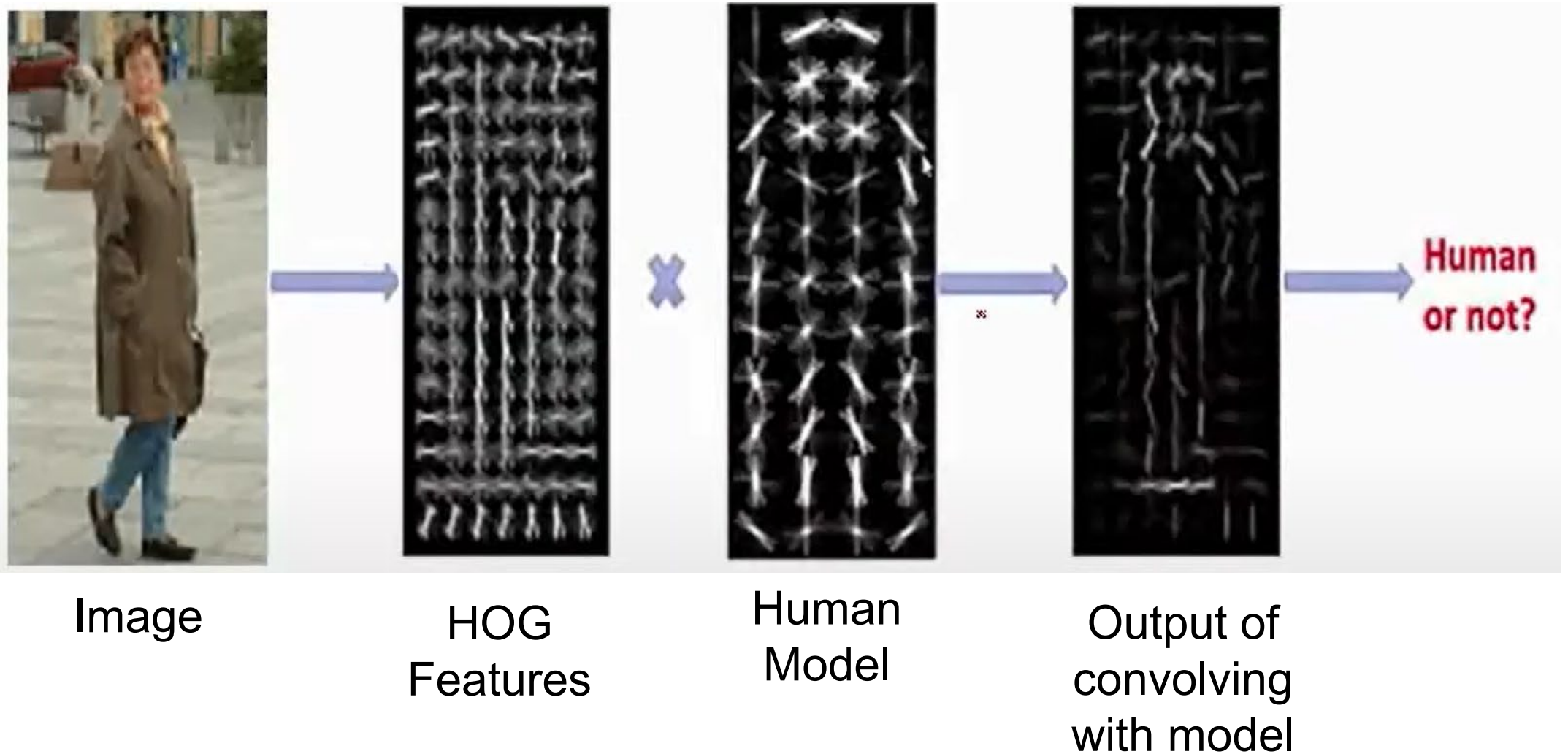
64 pixels  
8 cells  
7 blocks

Redundant representation due to overlapping blocks

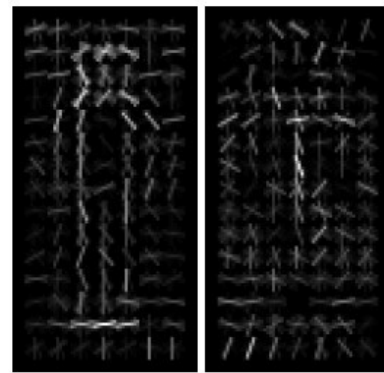
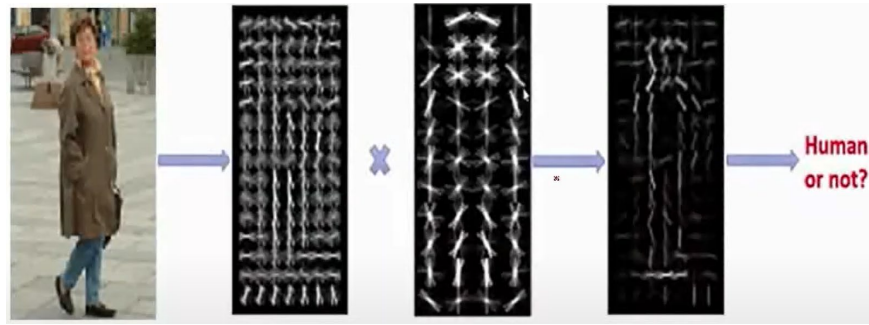


<http://chrisjmccormick.wordpress.com/2013/05/09/hog-person-detector-tutorial/>

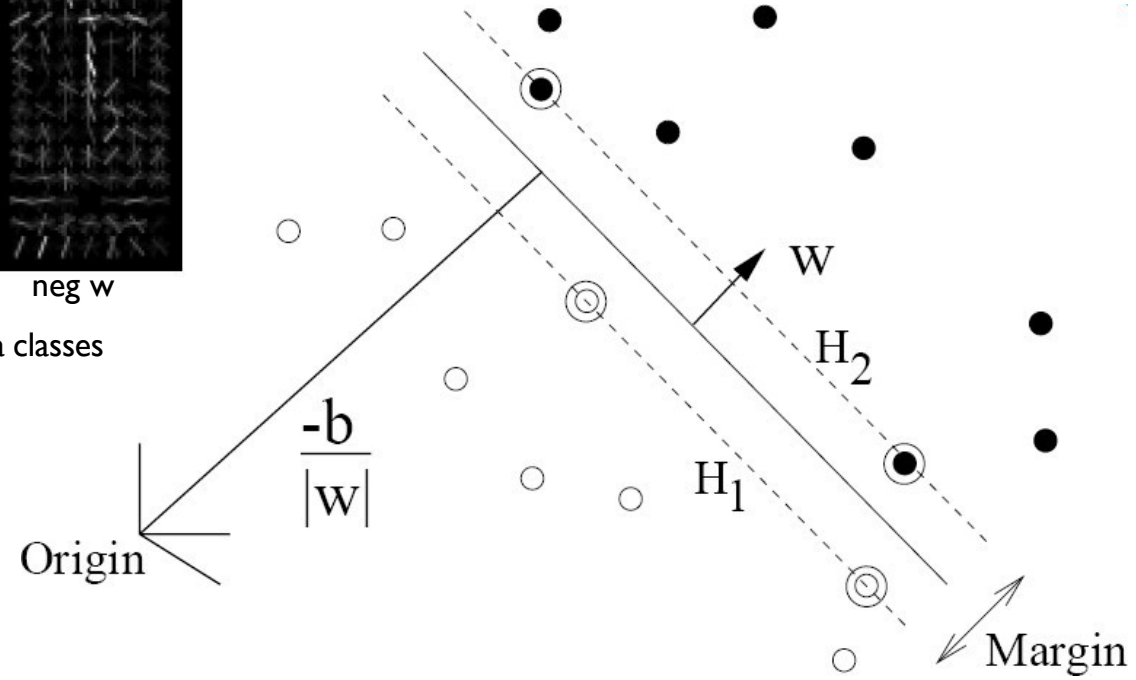
# HOG Pedestrian Detection

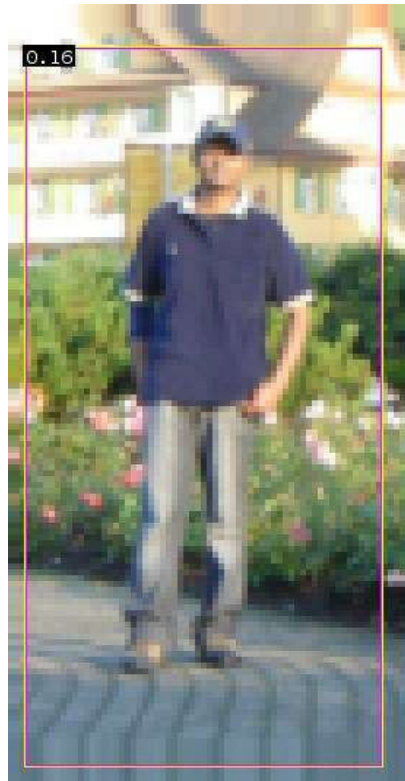
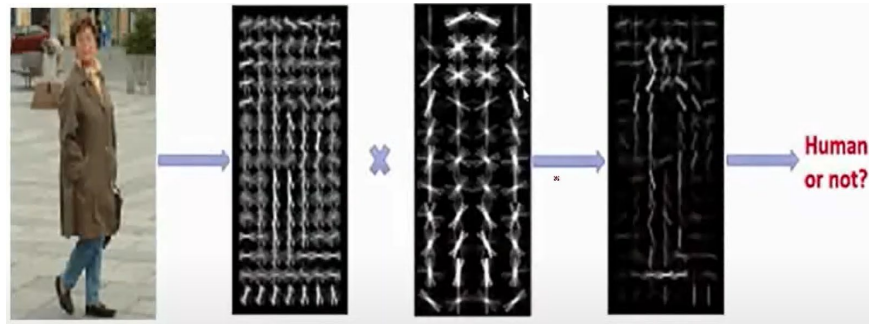






Training data classes





$$0.16 = w^T x - b$$

$$\text{sign}(0.16) = 1$$

$\Rightarrow$  pedestrian

# Multi-Scale Detection

