Department of Computer Science University of Bristol

## COMS30121 - Image Processing and Computer Vision

www.cs.bris.ac.uk/Teaching/Resources/COMS30121



Lecture 06

# Object Detection Basics I

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# What is 'Object Recognition'?

- Object Recognition aims at bridging the 'semantic gap' between...
  - given pixel values, and
  - meaningful objects (grouping of pixels + classification of groups)
- > image regions need to be found and assigned with semantic labels from a space of object classes
- Why do shape detection and segmentation rarely work for real-world object detection?

Variable visual appearance

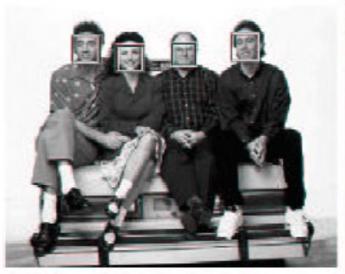
- high intra-class, low inter-class variance
- classes are rarely well defined
- change of illumination, scale, pose + deformation, occlusion...
- → object recognition is a difficult task

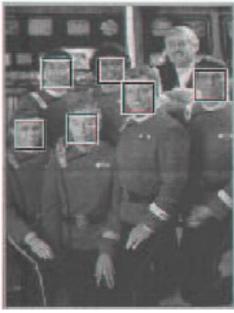














First Real-time Detection Method: Viola & Jones' (2001) (base line standard for off-the-shelf method for almost a decade)

#### **Selected Example Algorithm:** Viola & Jones' Real-time Method (2001)

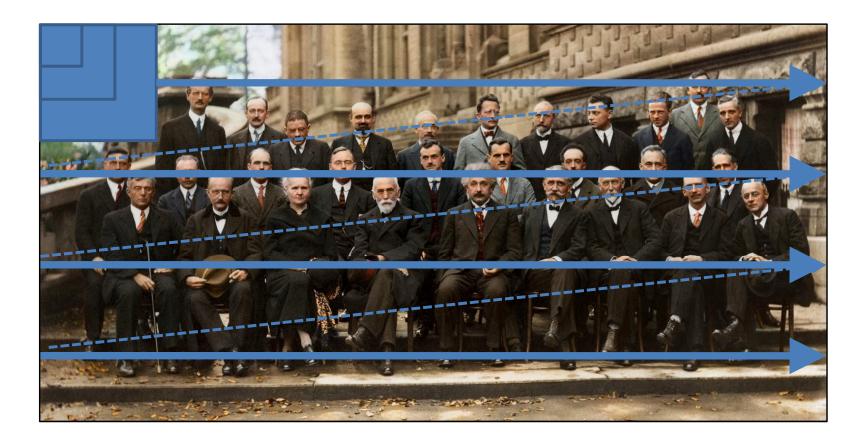
#### **Our Agenda:**

- Viola Jones technique overview
- Sliding Window Detectors
- Haar-like Features
- Feature Extraction and Integral Images
- Weak Classifiers
- Boosting and Classifier Evaluation
- Cascades of Boosted Classifiers

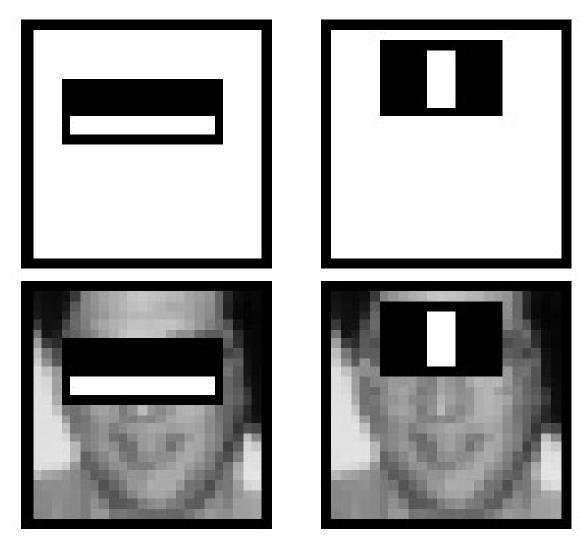
Best description of full details available in consolidated paper by Viola and Jones, International Journal of Computer Vision, 2004

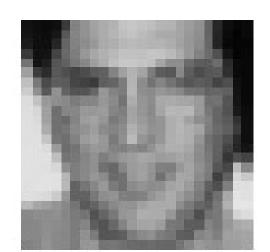
## Shift and Scale Invariance: Sliding Window Detectors

- image is tested for object presence window-by-window
- the window is `slided' and `scaled' throughout the image
- each resulting window is judged w.r.t. an object model giving a response indicating object presence or absence



#### **Basic Object Model Idea:** Characteristic Set of Block Features



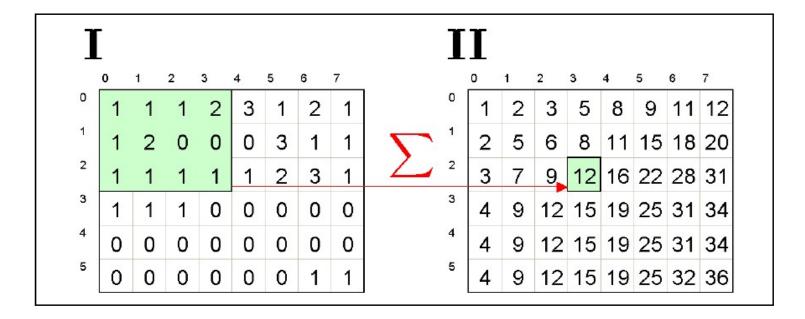


Viola & Jones' (2001)

## Integral Images & Integration Rule

(INTEGRATION RULE OF CONVOLUTION)

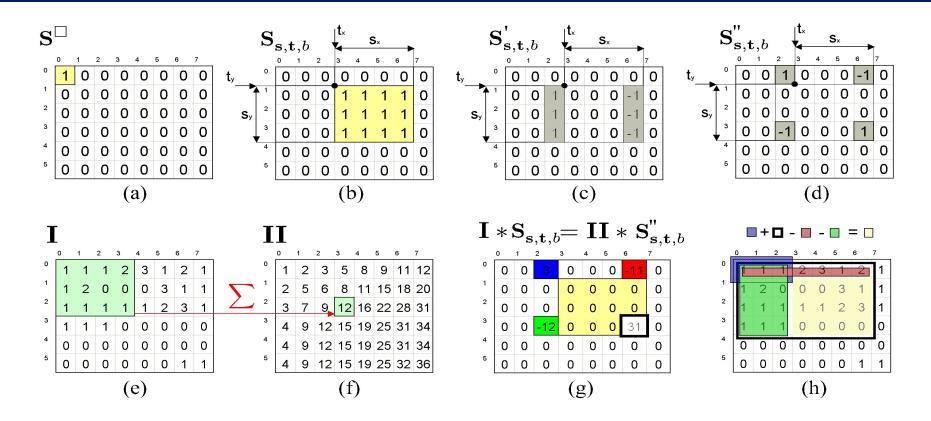
$$(\mathbf{S_k} * \mathbf{I})^{[n]} = \mathbf{S_k}^{[q]} * \mathbf{I}^{[p]}$$
 given  $n = p + q$ 



(IMAGE INTEGRATION) 
$$\mathbf{II}(-1,y) = 0; \qquad \mathbf{II}(x,y) = \mathbf{II}(x-1,y) + A(x,y);$$

$$A(x,-1) = 0;$$
  $A(x,y) = A(x,y-1) + \mathbf{I}(x,y).$ 

## Fast 'BlockImage' Convolution



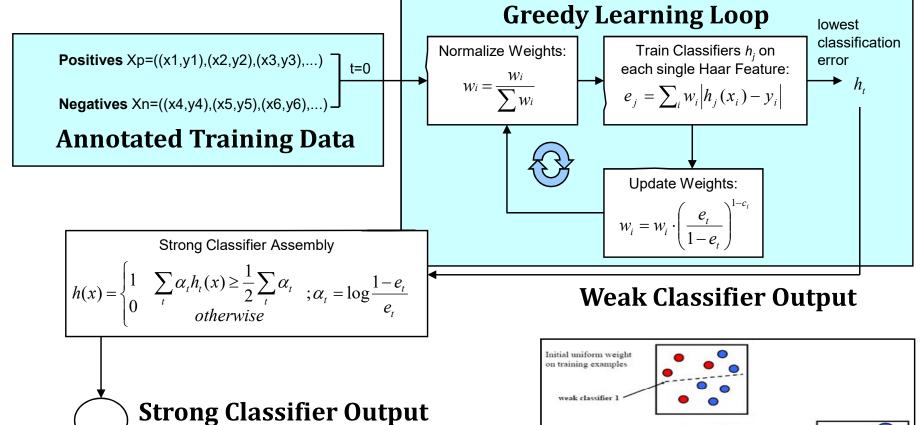
(FAST BLOCK IMAGE CONVOLUTION)

$$I * S_k = II(t_1 - 1, t_2 - 1) + II(s_1 + t_1 - 1, s_2 + t_2 - 1)$$

$$- II(s_1 + t_1 - 1, t_2 - 1) - II(t_1 - 1, s_2 + t_2 - 1)$$

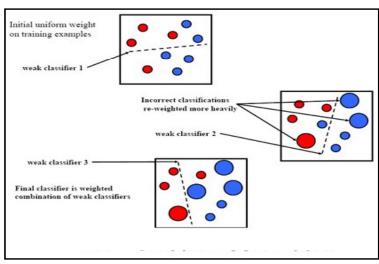
where  $\mathbf{k} = ((\mathbf{s}_1, \mathbf{s}_2), (\mathbf{t}_1, \mathbf{t}_2), b)$  holds the scale and translation parameters

## Modelling Objects by Boosting

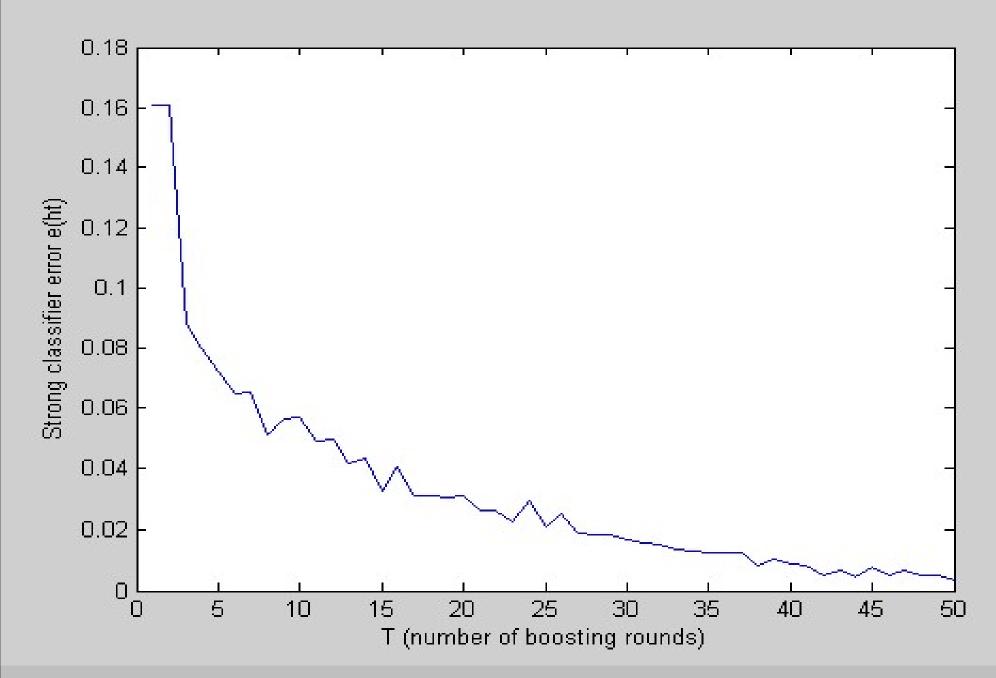


**AdaBoost Algorithm** 

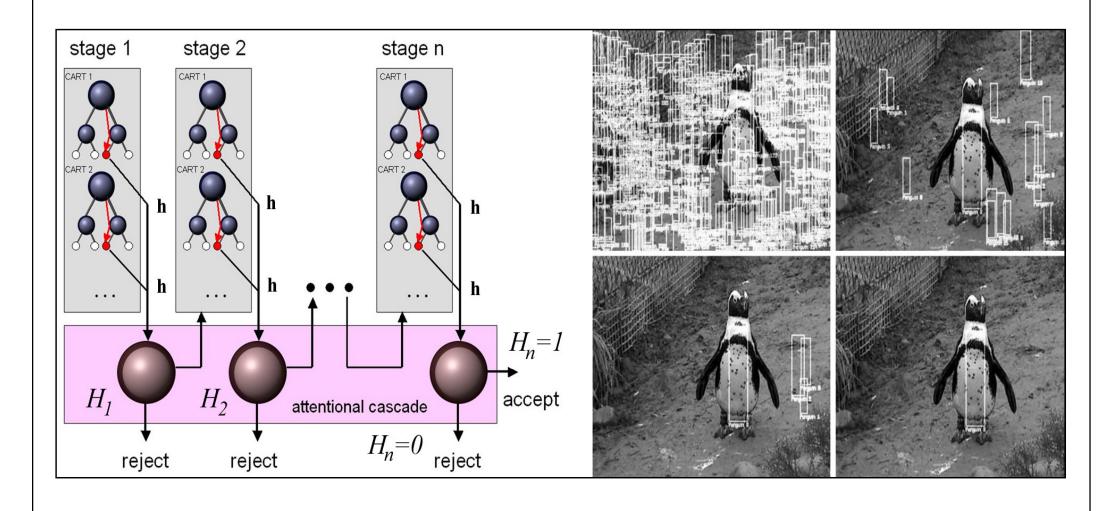
(see paper by Viola and Jones 2004)



## Error Reduction as Boosting adds Classifiers



# **Concept of Attentional Cascading**



#### On Window Resolution

