



國立高雄科技大學

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# Haar-Like Feature

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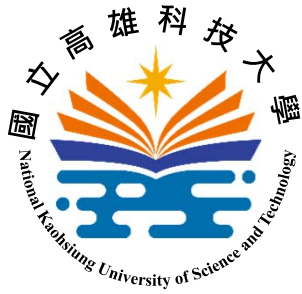
May 6, 2019





# Outline

- Introduction
- Response Definition
- Fast Computation



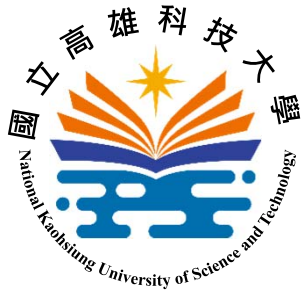
# Introduction

- Origin of Haar-Like Feature
  - Haar-like features are derived from Haar wavelet functions proposed by Alfred Haar in 1909.
  - They are firstly used in the computer vision area for object detection in 1998.
  - They became well-known since its use in real-time face detection by Viola and Jones in 2001.

C. Papageorgiou, et. al., “A General Framework for Object Detection”, *IEEE Intl. Conf. on Computer Vision*, 1998.

P. Viola and M. Jones, “Rapid Object Detection using a Boosted Cascade of Simple Features,” *IEEE Intl. Conf. on CVPR*, 2001.



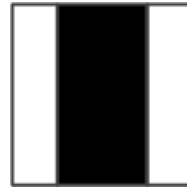


# Introduction

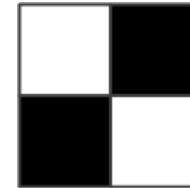
- About Haar-Like Feature
  - A Haar-Like feature is a **response** from a rectangle in the form of Haar-like prototype.
  - A prototype is composed of black and white rectangular regions
  - The size of black and white regions are equal.



Edge Prototype



Line Prototype



Diagonal Prototype

# Response Definition

- Definition  $\phi_t(I, \theta)$ 
  - $I$ : gray-level image
  - $\theta = (l, t, w, h)$ : rectangle configuration
  - $t$ : selected prototype



$I$ : image



$\theta$ : configuration

Haar-like  
Rectangle

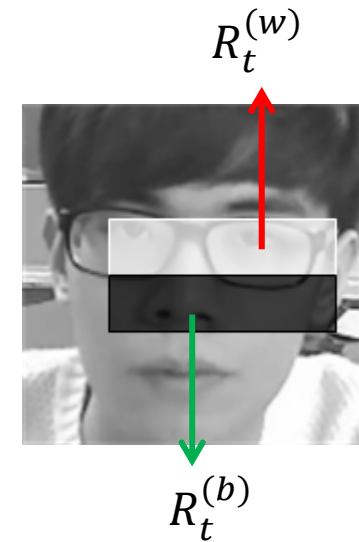


$t$ : prototype

# Response Definition

- Definition  $\phi_t(I, \theta)$

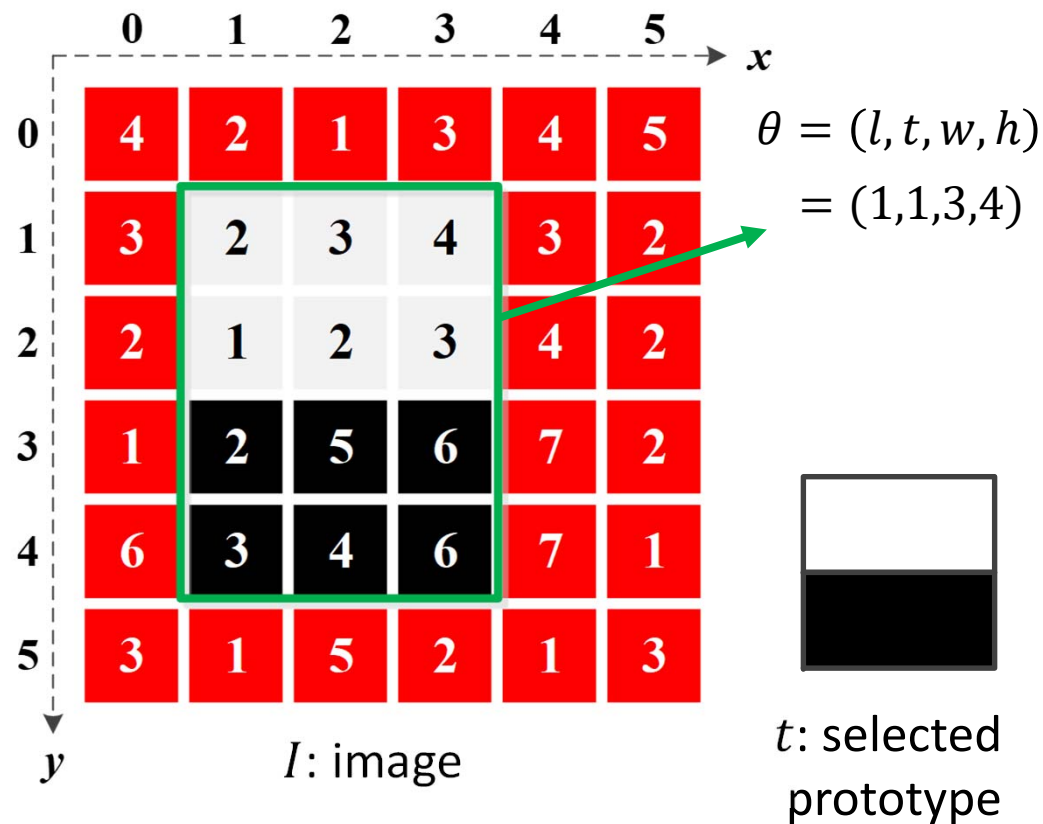
$$\phi_t(I, \theta) = \sum_{(x,y) \in R_t^{(w)}} I(x, y) - \sum_{(x,y) \in R_t^{(b)}} I(x, y)$$



- sum up point intensities in  $R_t^{(w)}$  (white)
- sum up point intensities in  $R_t^{(b)}$  (black)
- calculate the different between them.

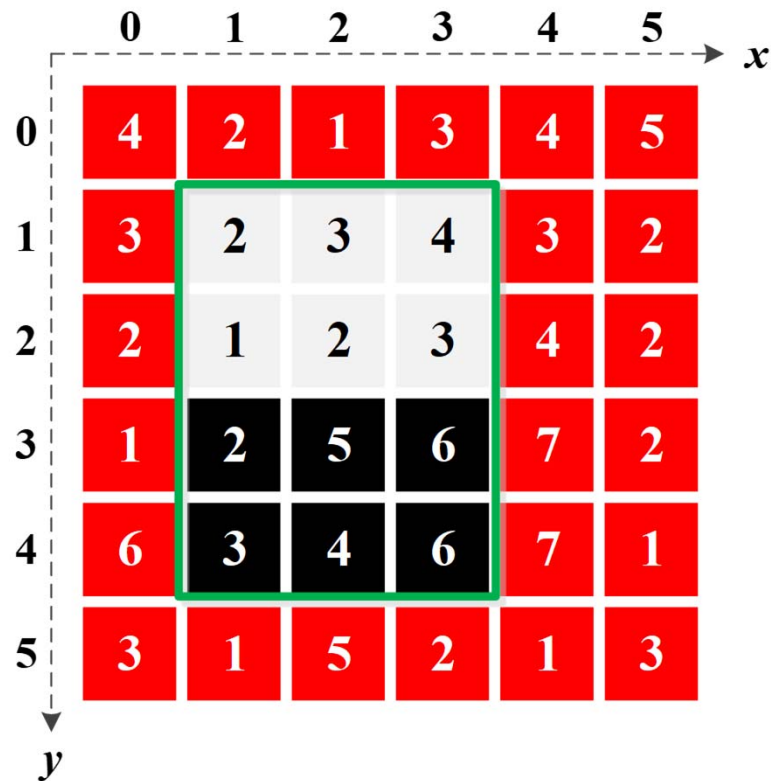
# Response Definition

- Definition  $\phi_t(I, \theta)$



# Response Definition

- Definition  $\phi_t(I, \theta)$



$$\phi_t(I, \theta)$$

$$= \sum \begin{pmatrix} 2 + 3 + 4 + \\ 1 + 2 + 3 \end{pmatrix}$$

$$- \sum \begin{pmatrix} 2 + 5 + 6 + \\ 3 + 4 + 6 \end{pmatrix}$$

$$\phi_t(I, \theta) = 15 - 26 = -11$$





# Response Definition

- Computation Issue
  - Computing Haar-like features in this direct way is computational expensive.
    - Time complexity for computing a Haar-like feature is  $O(w \times h)$
    - An effective descriptor is generally composed of thousands of Haar-like features.
  - Integral image is used to address this issue

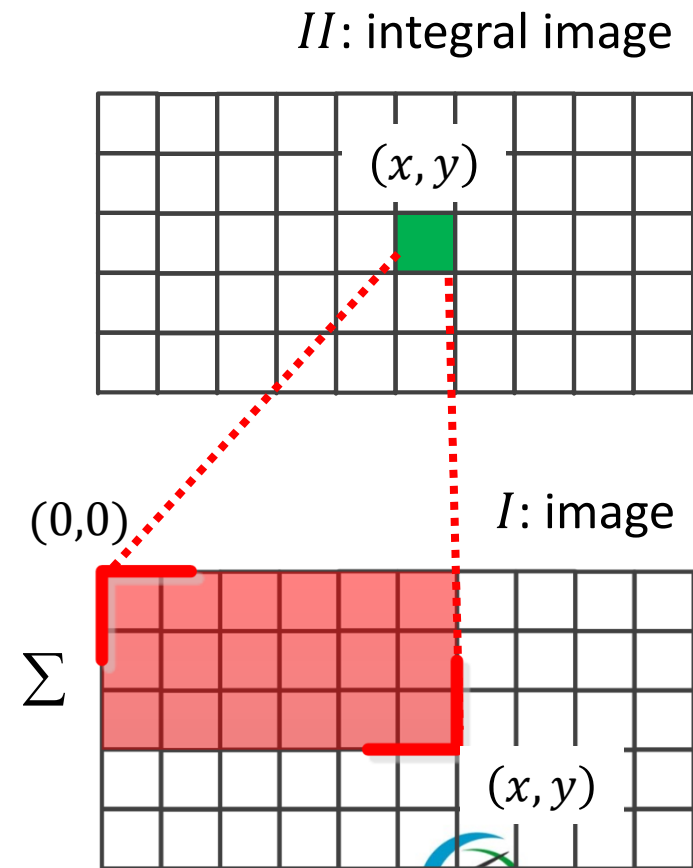
$$O(w \times h) \rightarrow O(1)$$



# Fast Computation

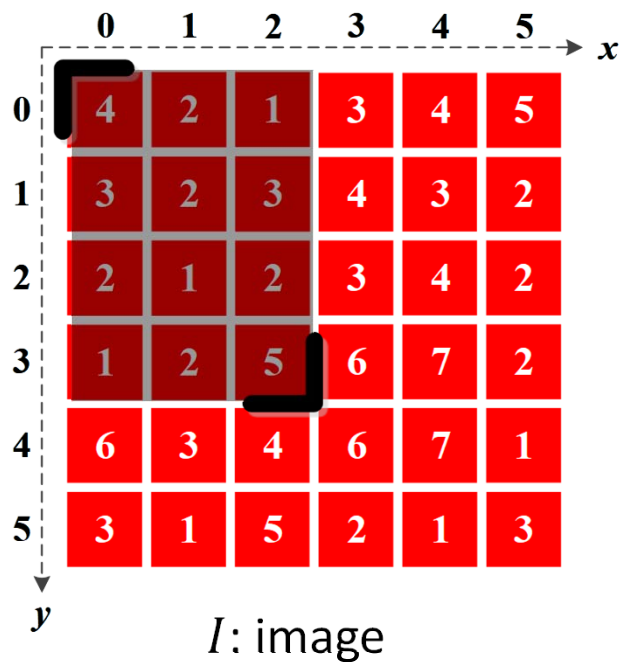
- Integral Image  $I I(\cdot)$ 
  - $I I(x, y)$  is defined as intensity sum of points in rectangle  $(0,0, x, y)$ 
    - $(0,0)$ :left-top point
    - $(x, y)$ :right-bottom point

$$I I(x, y) = \sum_{u=0}^{u=x} \sum_{v=0}^{v=y} I(u, v)$$



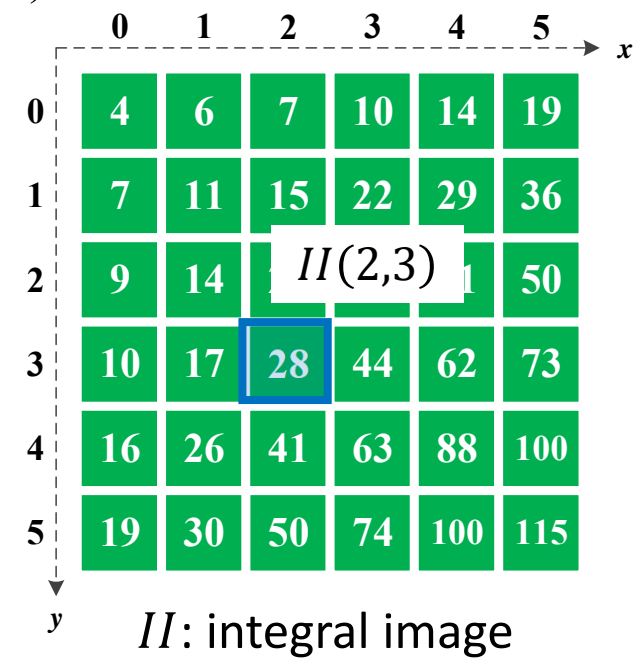
# Fast Computation

- Integral Image  $II(.)$



$$II(2,3) = \sum_{u=0}^2 \sum_{v=0}^3 I(u,v)$$

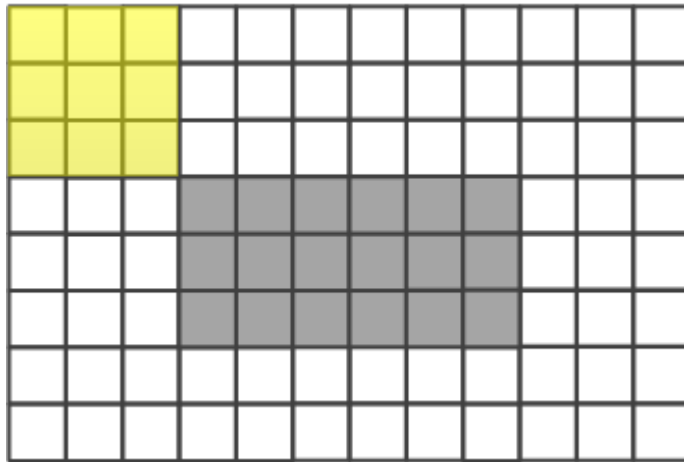
$$\begin{aligned}
 &= (4 + 2 + 1) \\
 &+ (3 + 2 + 3) \\
 &+ (2 + 1 + 2) \\
 &+ (1 + 2 + 5) \\
 &= 7 + 8 + 5 + 8 \\
 &= 28
 \end{aligned}$$



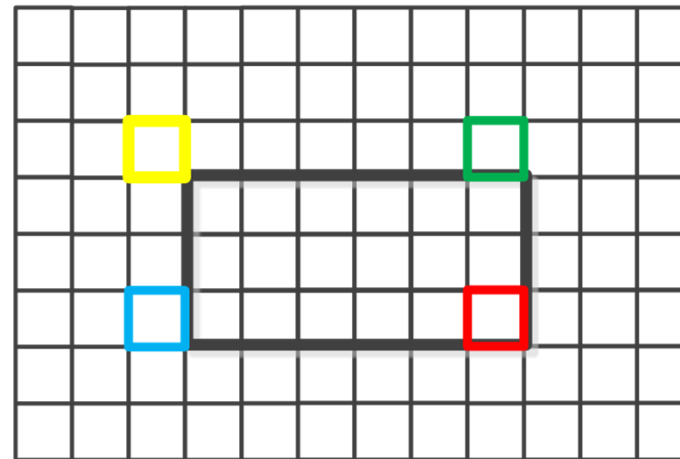
# Fast Computation

- Speed-Up Strategy
  - We can compute sum of points in a rectangle in **constant time** using  $II(.)$

$I$ : image



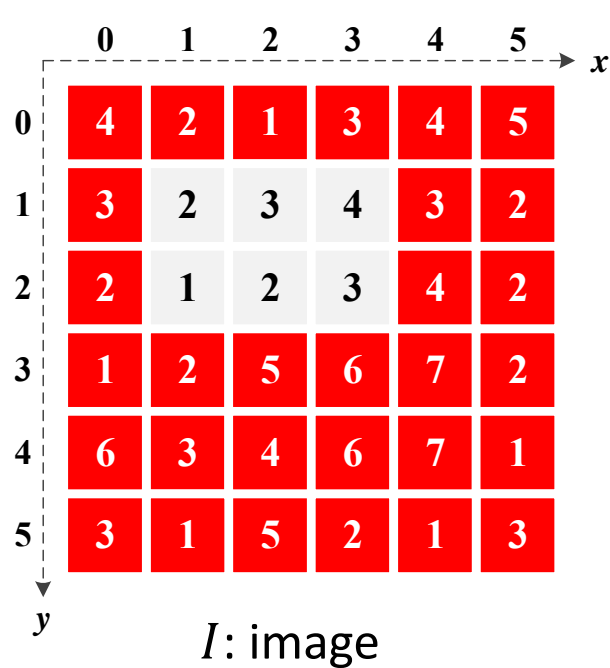
$II$ : integral image



$$\text{Gray} = \text{Red} - \text{Green} - \text{Blue} + \text{Yellow} = \square_{\text{red}} - \square_{\text{green}} - \square_{\text{blue}} + \square_{\text{yellow}}$$

# Fast Computation

- Speed-Up Strategy: Example

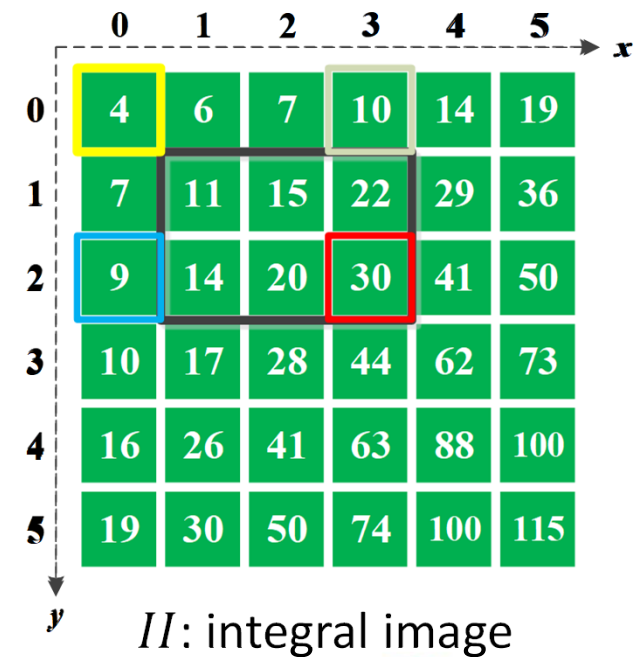


$$\sum_{(x,y) \in R(w)} I(x,y)$$

$$= \square - \square - \square + \square$$

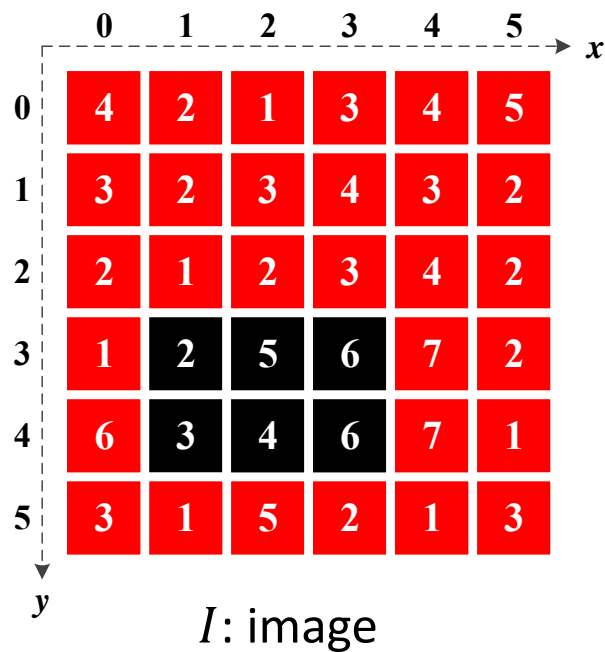
$$= 30 - 10 - 9 + 4$$

$$= 15$$



# Fast Computation

- Speed-Up Strategy: Example

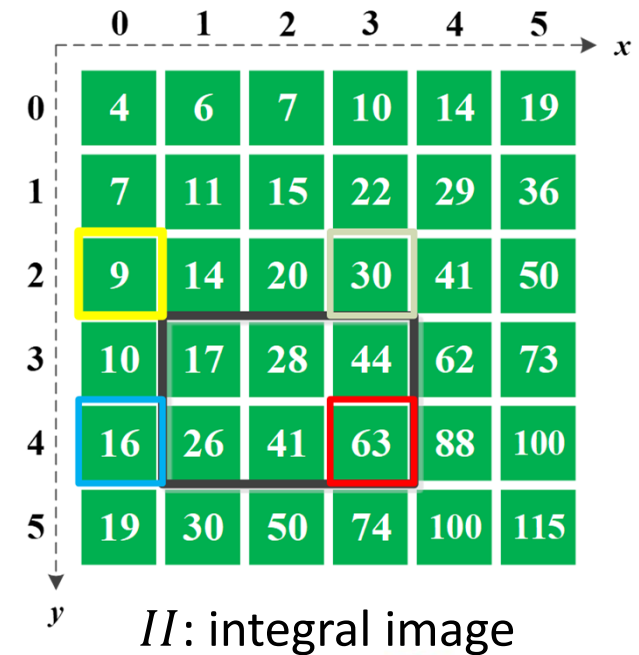


$$\sum_{(x,y) \in R(b)} I(x,y)$$

$$= \square - \square - \square + \square$$

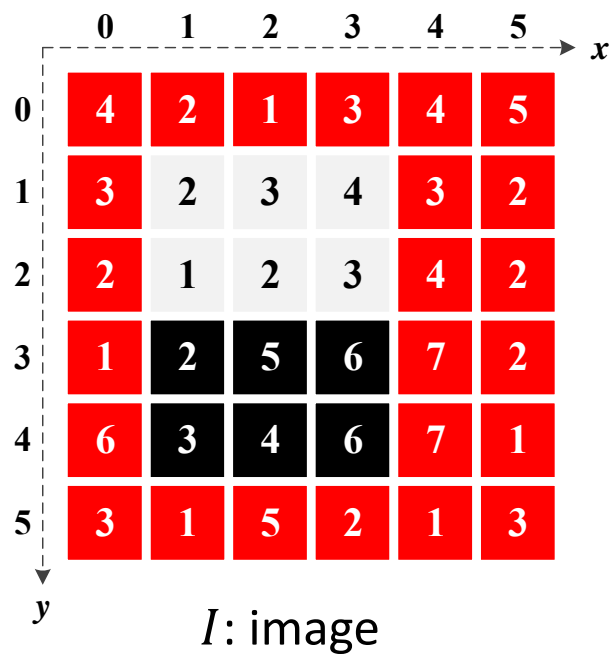
$$= 63 - 30 - 16 + 9$$

$$= 26$$



# Fast Computation

- Speed-Up Strategy: Example



$$\phi_t(I, \theta) = \sum_{(x,y) \in R^{(w)}} I(x, y) = 15$$

Constant Time

$$- \sum_{(x,y) \in R^{(b)}} I(x, y) = 26$$

Constant Time

$$\phi_t(I, \theta) = 15 - 26 = -11$$

Constant Time

