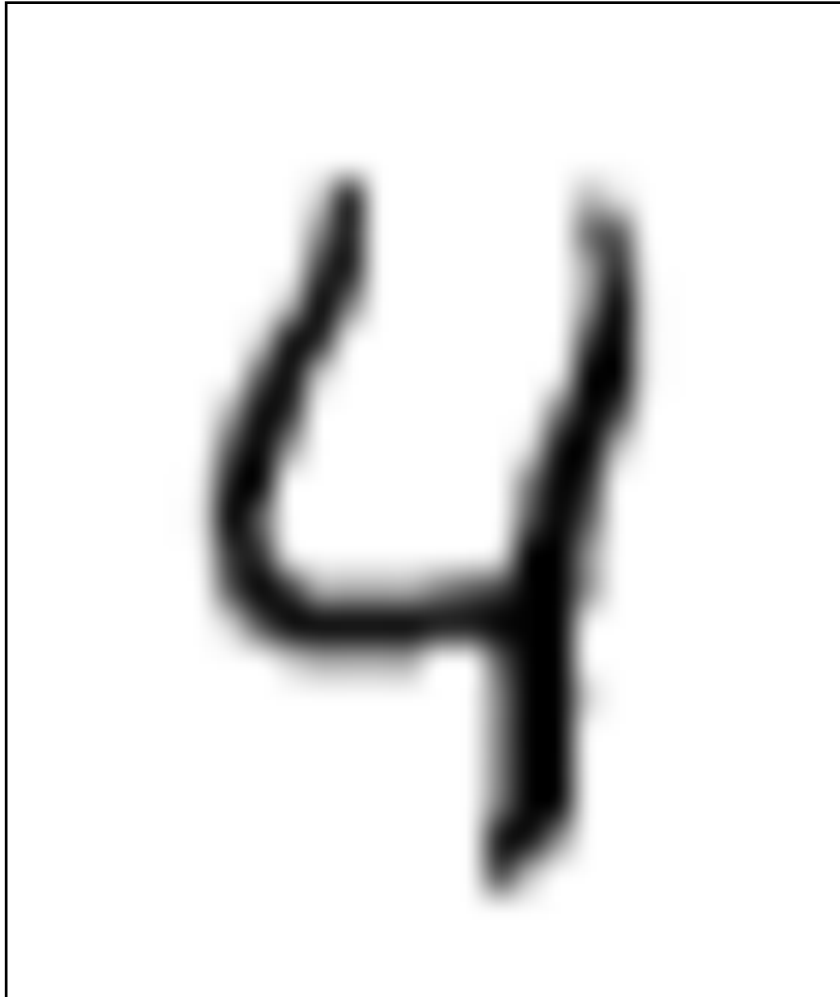


# HAAR-like features for images

# Images

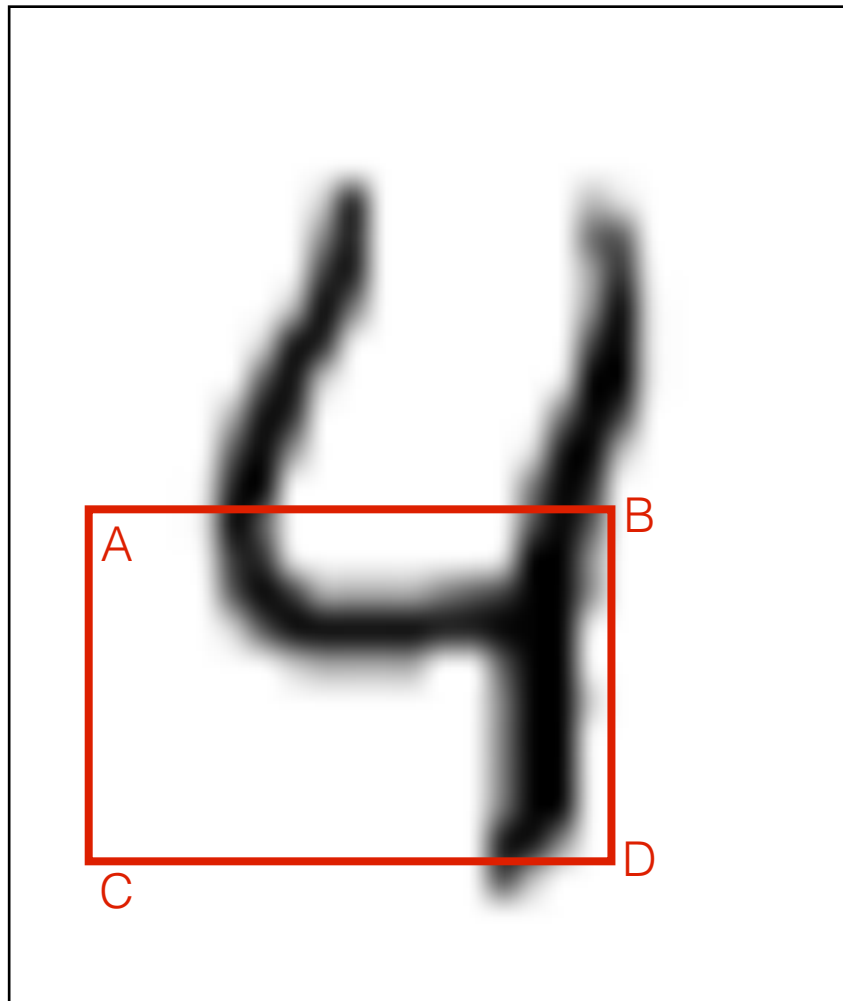


- digit images are scanned hand written digits

# Digit scan dataset

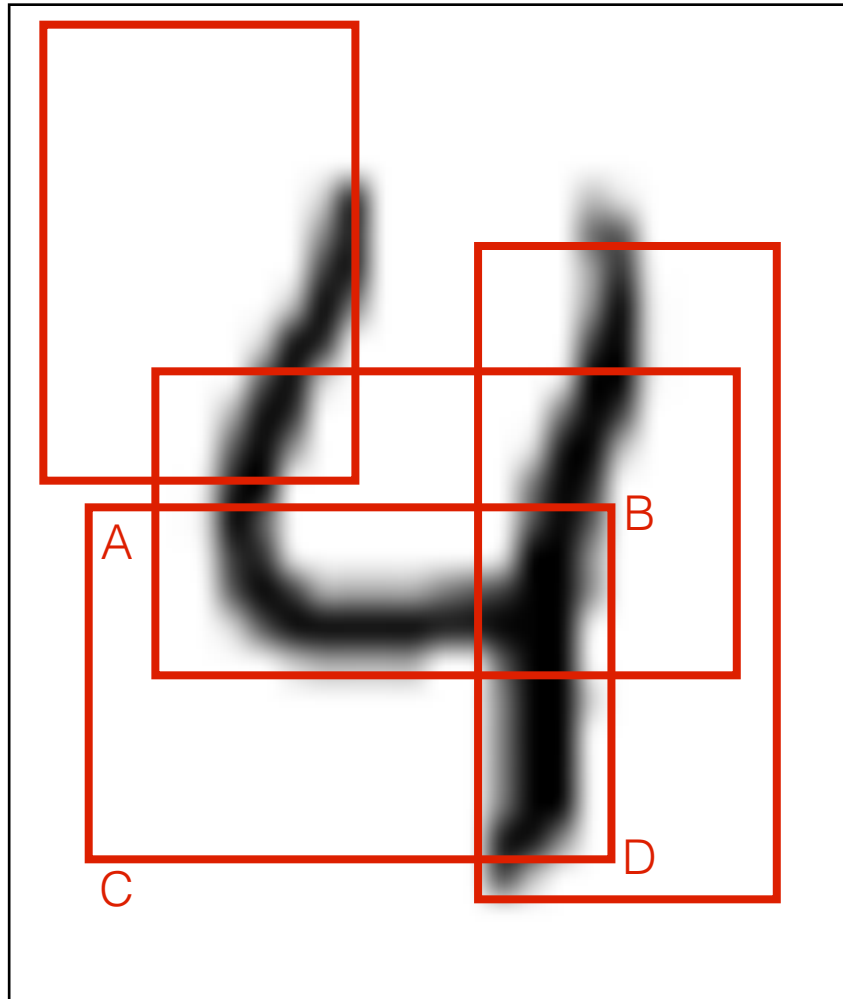
- 60,000 scans
- 10 classes : 0,1,2,...,9
  - roughly uniform distributed
- each scanned image 28x28 pixels square
- comes split into (train, test)
  - no cross validation
- very learnable: most algorithms score 5% or less error
- <http://yann.lecun.com/exdb/mnist/>

# Rectangle black level



- rectangle ABCD can act like an image “mask” : it selects/cuts that rectangle out of an image
  - or of any image

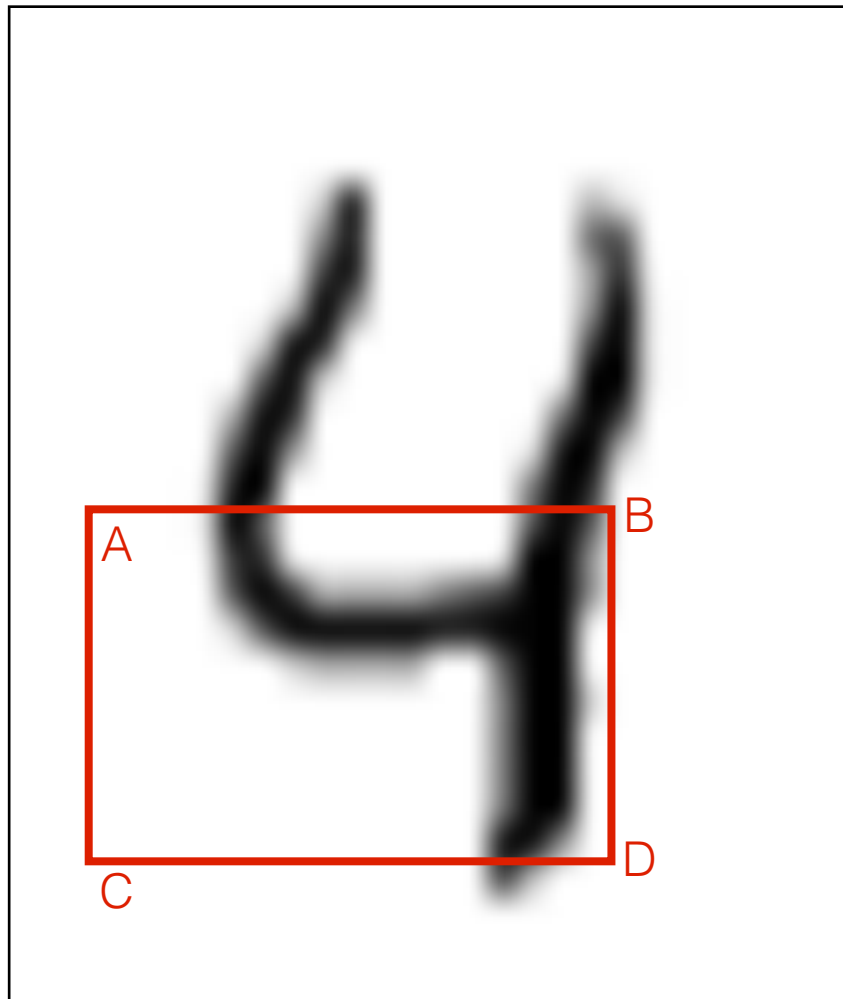
# Rectangle black level



- a given set  $S$  of rectangles cuts  $S$  different masks for an image

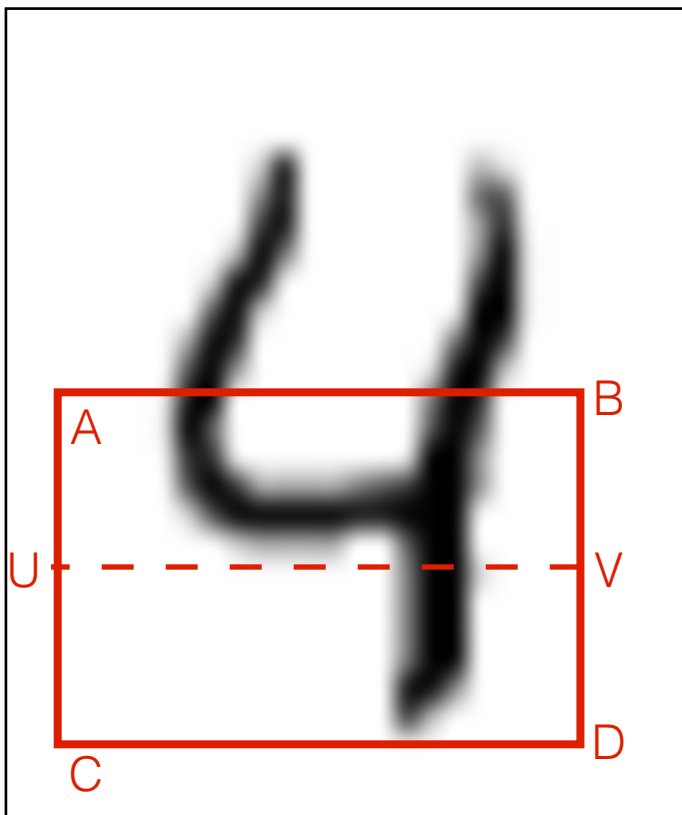
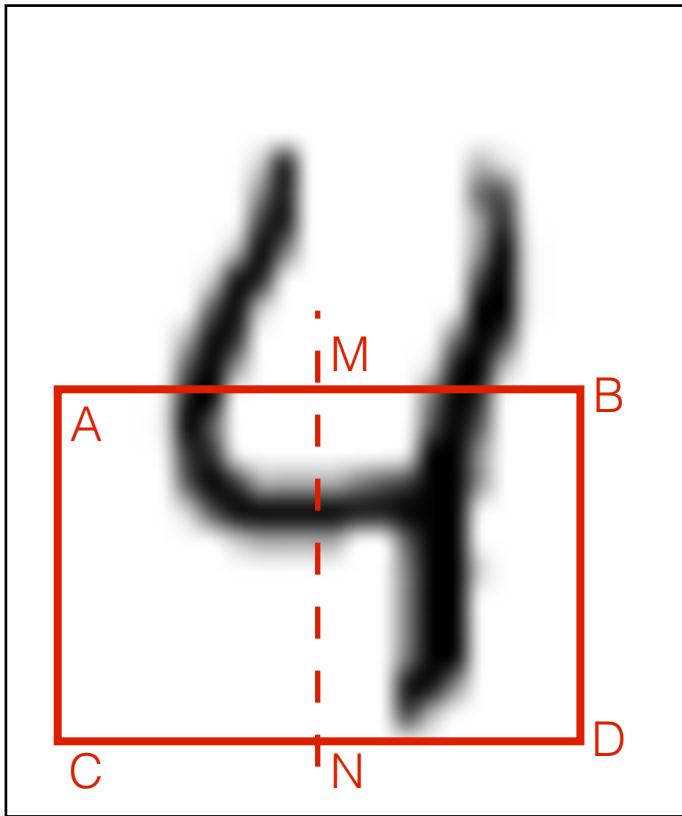
# Rectangle black level

o



- for each rectangle  $r=ABCD$  on image  $X$  we can compute a “black value”
  - $\text{black}_r(X)$  = number of black pixels in the mask cut by  $r$  in image  $X$
- we can compute  $\text{black}_r(X)$  efficiently, if we compute in the right order!
  - dynamic programming

# Vertical, horizontal features for a rectangle



- horizontal feature

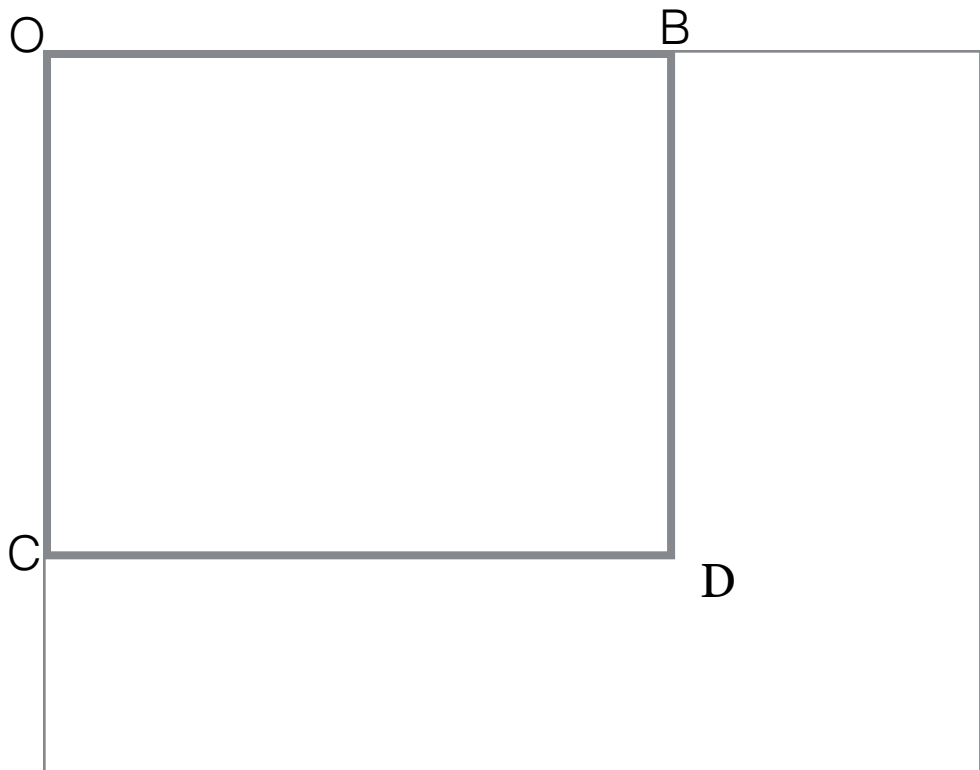
$$\begin{aligned}\Delta_{hr}(X) &= \text{black}_{r\text{-left}}() - \text{black}_{r\text{-right}}(X) \\ &= \text{black}_{AMCN}(X) - \text{black}_{MBND}(X)\end{aligned}$$

- vertical feature

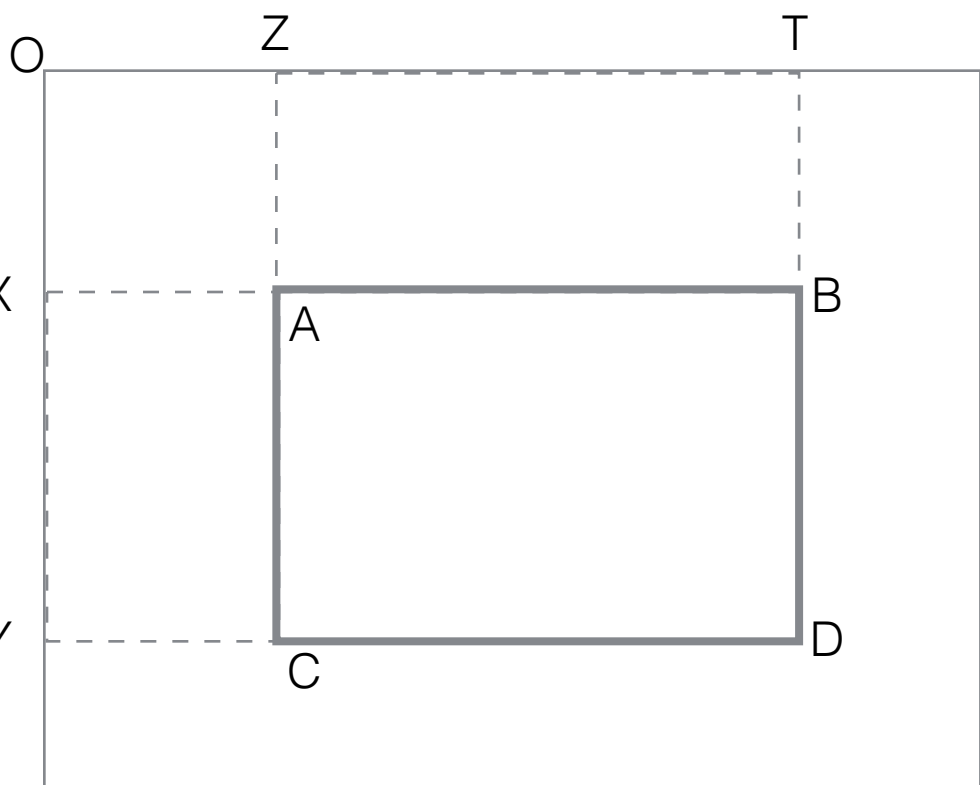
$$\begin{aligned}\Delta_{hv}(X) &= \text{black}_{r\text{-top}}() - \text{black}_{r\text{-bottom}}(X) \\ &= \text{black}_{ABUV}(X) - \text{black}_{UVCD}(X)\end{aligned}$$

- $|S|$  rectangles, 2 features each  $\Rightarrow 2|S|$  features extracted (from each image)
  - if we also store the  $\text{black}_r(X)$  value, that's 3 features/rectangle ( $\text{black}_r(X)$ ,  $\Delta_{hr}(X)$ ,  $\Delta_{hv}(X)$ ) for  $3|S|$  features extracted.

# How to compute $\text{black}_r(X)$ efficiently



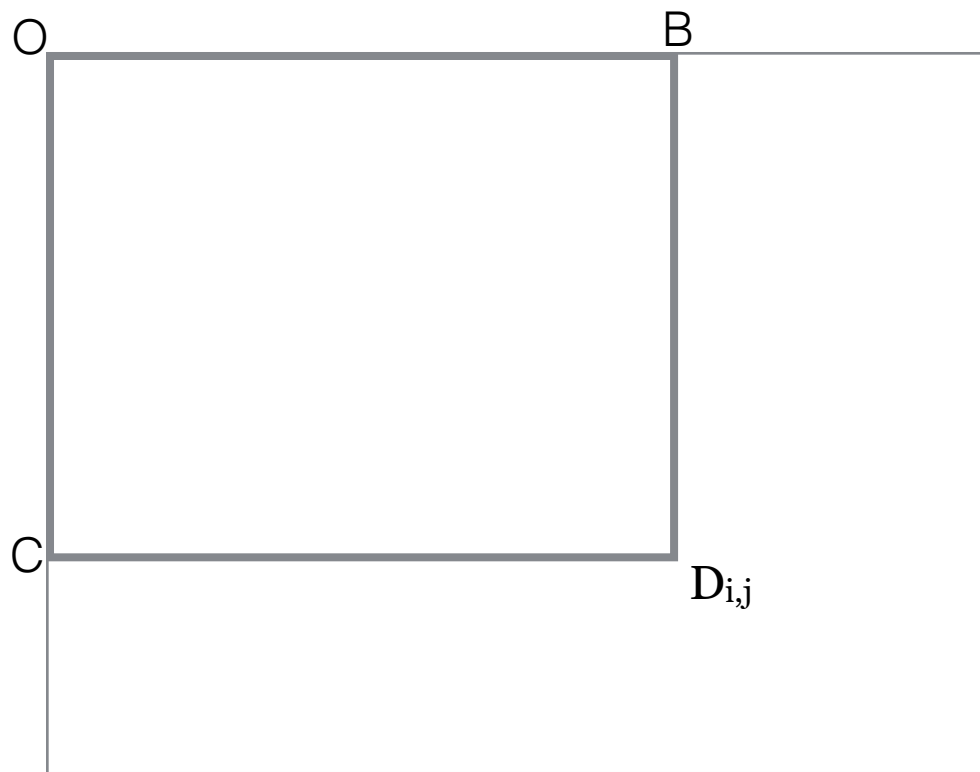
- first compute it for all rectangles cornered in O (A=O) fix image corner.
  - That is compute  $\text{black}_r(X)$  for each pixel D



- then every rectangle  $r=ABCD$  can be computed in constant time from O-cornered rectangles
- $\text{black}(\text{rectangle } ABCD) = \text{black}(OTYD) - \text{black}(OTXB) - \text{black}(OZXC) + \text{black}(OZXA)$

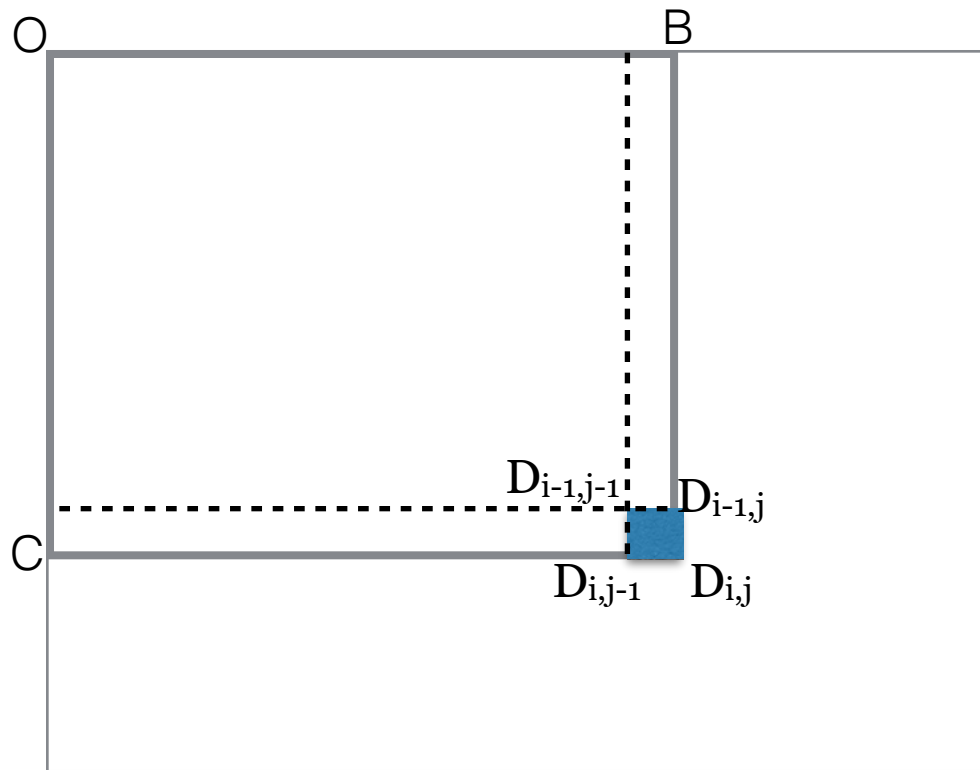


# O-corner rectangles computation



- $r=OBCD$  determined by  $D$
- naively one can compute all  $\text{black}_r(X) = \text{black}_D(X)$  for all rectangles as
  - for  $i=1:n$
  - for  $j=1:n$ 
    - $D=D_{ij}$  pixel
    - $\text{black}_{D_{ij}}(X) = \text{count of black pixels in } OBCD$
- total  $O(n^4)$  running time
  - $n = \text{size of the square image}$

# O-corner rectangles : dynamic programming



- $r=OBCD$  determined by  $D$
- dynamic programming computes a rectangle from the rectangle computed already

- for  $i=1:n$

- for  $j=1:n$

- $D=D_{ij}$  pixel

$$\begin{aligned} \text{black\_}D_{ij}(X) = & \\ & \text{black\_}D_{i,j-1}(X) + \\ & \text{black\_}D_{i-1,j}(X) - \\ & \text{black\_}D_{i-1,j-1}(X) + \\ & \text{black}(\text{pixel\_}D_{ij}, X) \end{aligned}$$

- total  $O(n^2)$  running time
  - much better