

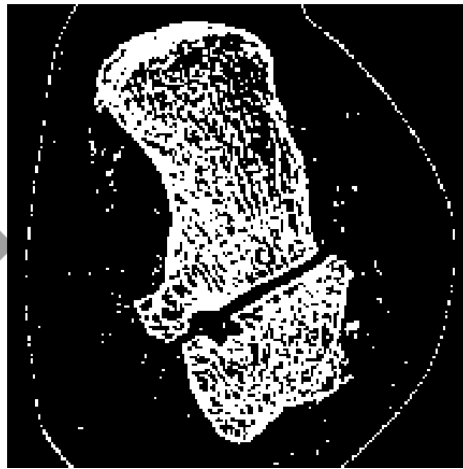
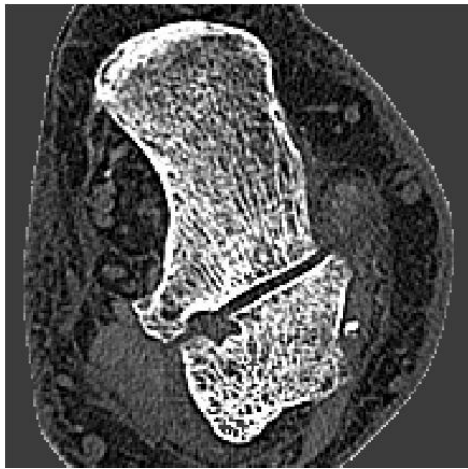


# >>> IMAGE PROCESSING AND COMPUTATIONAL PHOTOGRAPHY

SESSION 4b: BINARY IMAGE ANALYSIS.

Oriol Pujol & Simone Balocco

# BINARY IMAGES

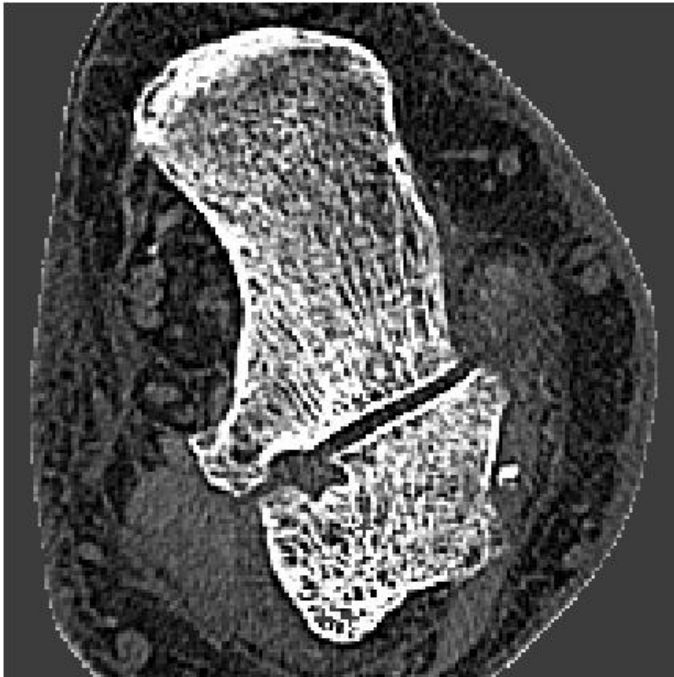


Creation

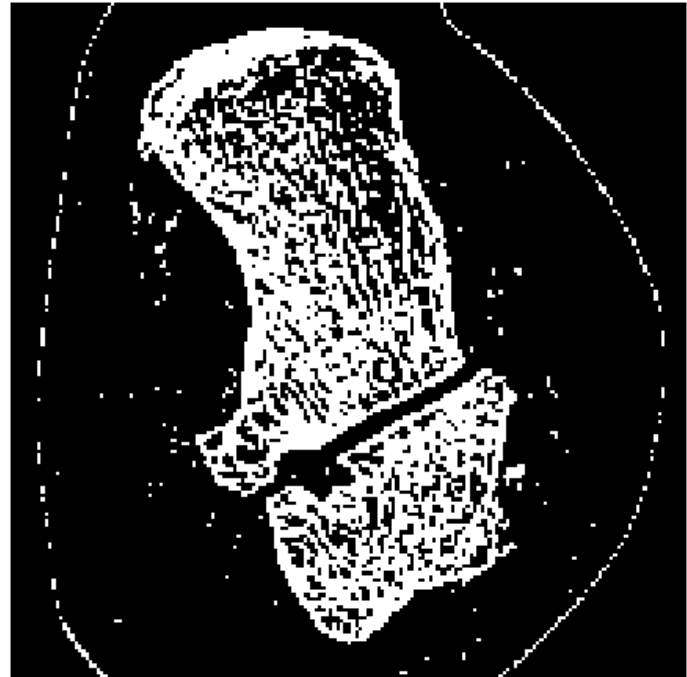
Processing

# SEGMENTATION

- Separating object from background in a grayscale picture
  - A simple method: thresholding by pixel (voxel) color
    - All pixels (voxels) with color above a threshold are set to 1



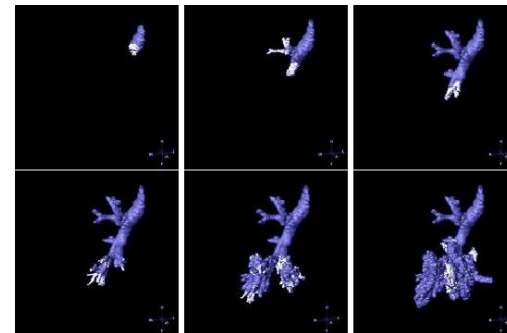
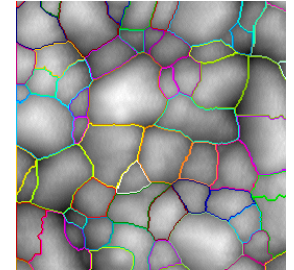
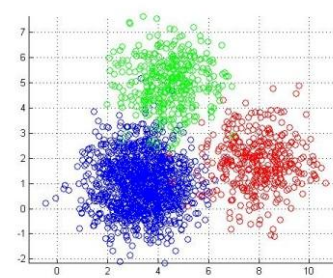
Grayscale picture



Thresholded binary picture

# SEGMENTATION

- Separating object from background in a grayscale picture
  - A simple method: thresholding by pixel (voxel) color
  - Other methods:
    - K-means clustering
    - Watershed
    - Region growing
    - Snakes and Level set
    - Graph cut
    - ...



# OTSU'S METHOD

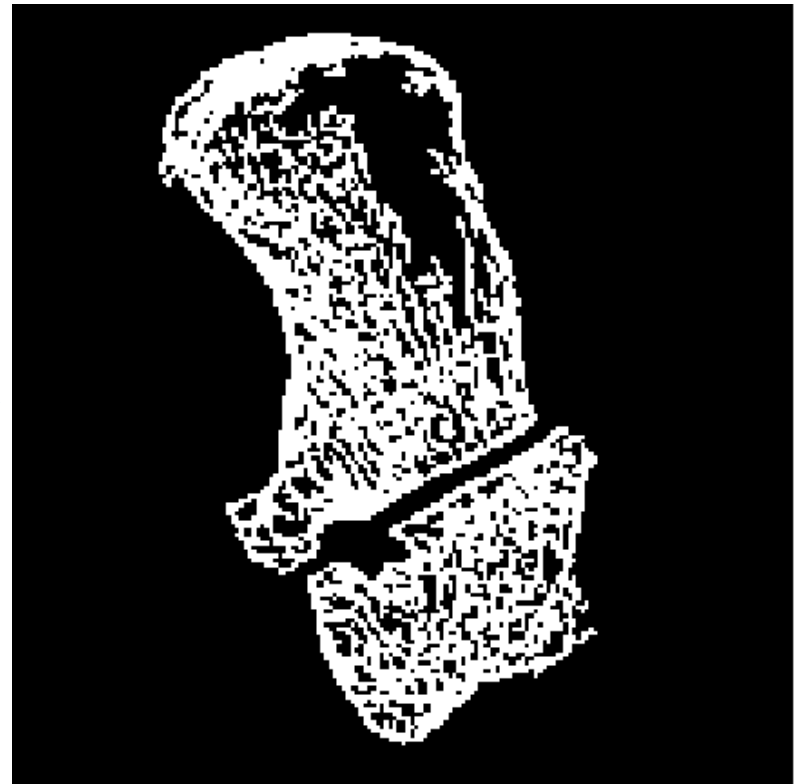
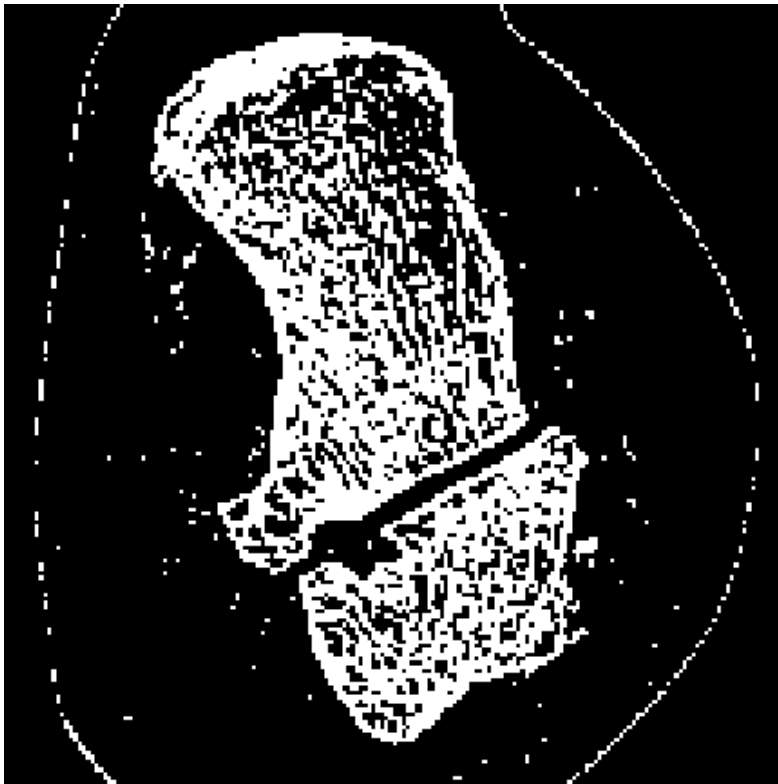
$$\text{minimize}_t \quad p_+(t)\sigma_+^2(t) + p_-(t)\sigma_-^2(t)$$

**Goal:** Reduce the intra-class variance. The best thresholding should divide the image evenly such that a similar amount of pixels goes to each region. We can do that by weighting each variance by the amount of pixels considered.

$P_+$  is the number of pixels in the positive region.

$\sigma_+$  is the standard deviation of the pixels of the positive region

# Connected Components



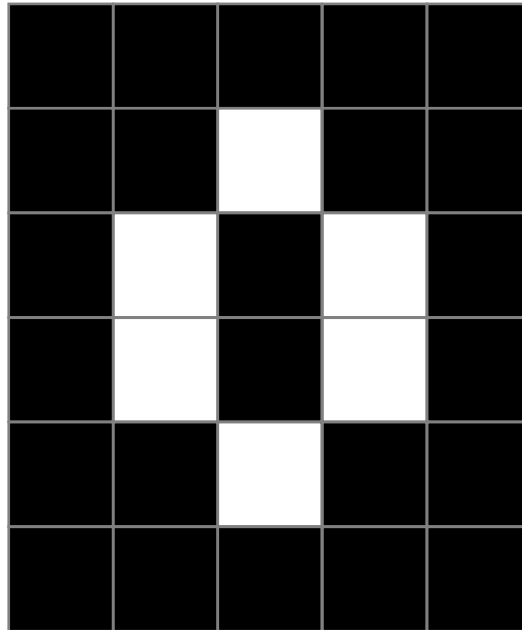
Take the largest 2 components of the object

# CONNECTIVITY

We talk about connectivity referring to the definition of the neighbors of a Pixel. We may find 2 kinds of connectivity, either 4 or 8.

This allows to define the blob.

# Connected Components



How many connected components are there in the object?



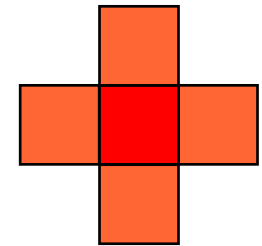
# Connectivity (2D)

- Two pixels are connected if their squares share:
  - A common edge
    - 4-connectivity
  - A common vertex
    - 8-connectivity

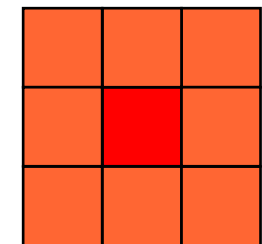
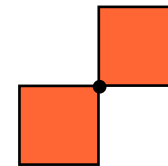
Two connected pixels



All pixels connected to x



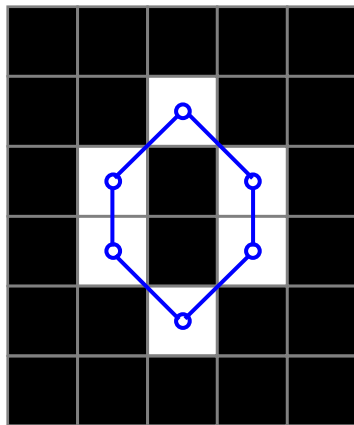
4-connectivity



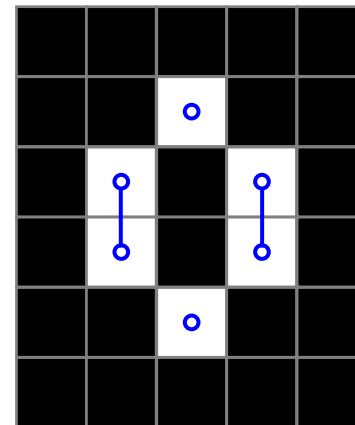
8-connectivity

# Connectivity (2D)

- Connected component
  - A maximum set of pixels (voxels) in the object or background, such that any two pixels (voxels) are connected via a path of connected pixels (voxels)



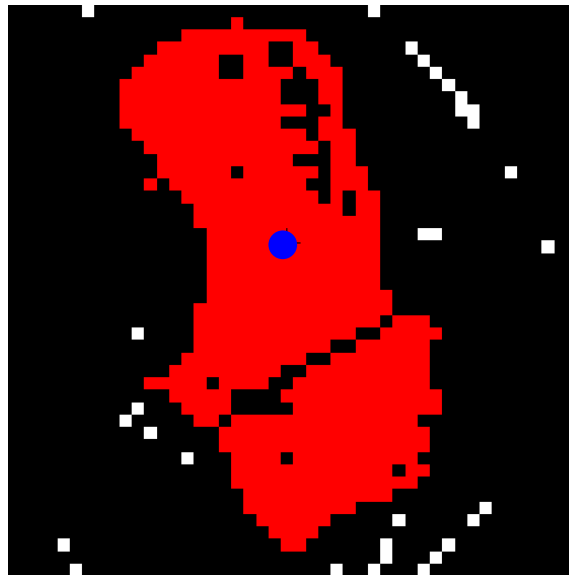
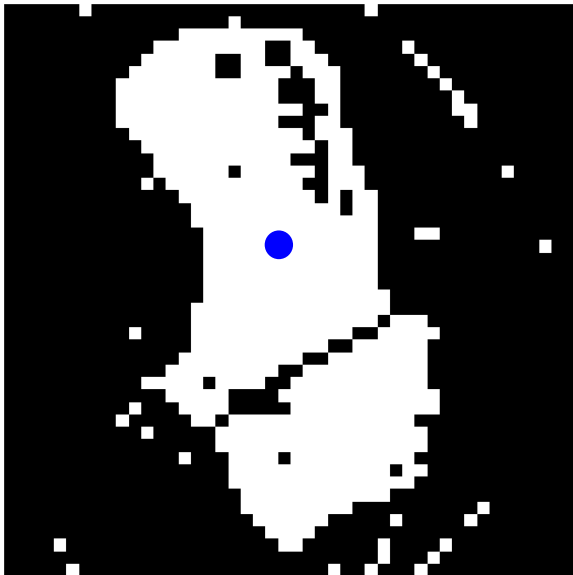
8-connected object  
(1 component)



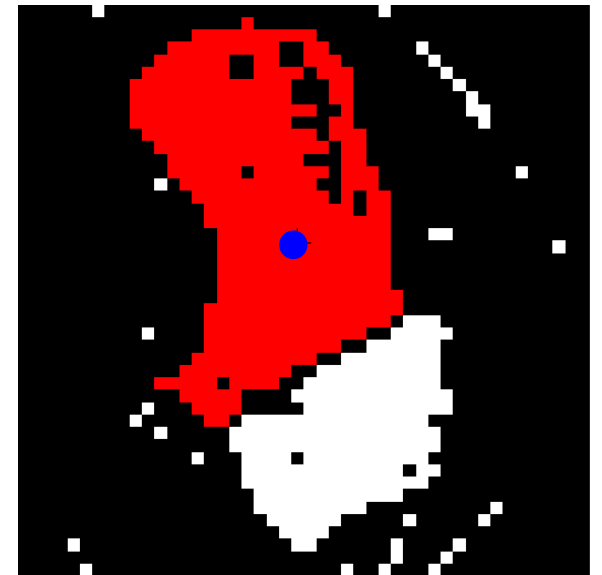
4-connected object  
(4 components)

# Connectivity (2D)

- What is the component connected to the blue pixel?



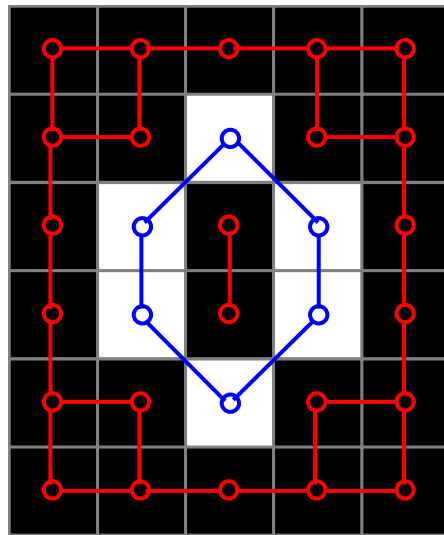
8-connectivity



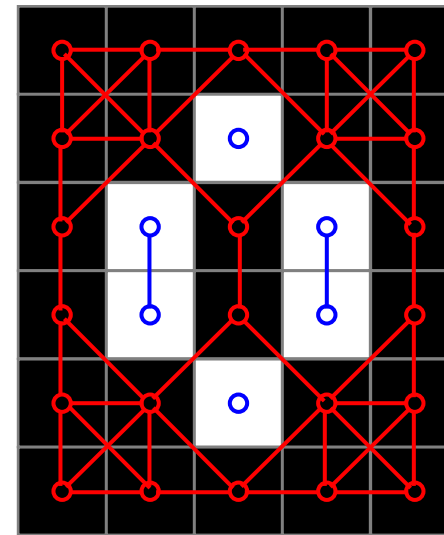
4-connectivity

# Connectivity (2D)

- Different connectivity for object (O) and background (B)
  - 2D pixels: 4- and 8-connectivity respectively for O and B (or B and O)



Object: 8-connectivity (1 comp)  
Background: 4-connectivity (2 comp)



Object: 4-connectivity (4 comp)  
Background: 8-connectivity (1 comp)

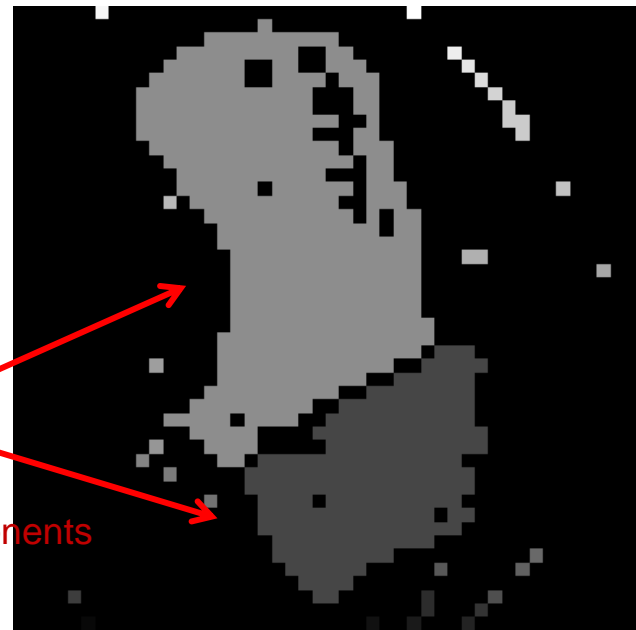
# Finding Connected Components

- Labeling all components in an image:
  - Loop through each pixel (voxel). If it is not labeled, find its connected component, then **label** all pixels (voxels) in the component.



One component

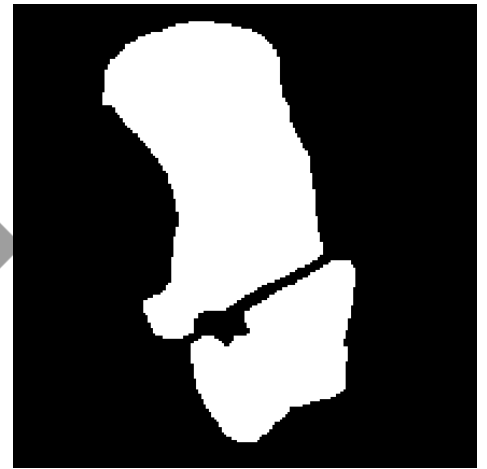
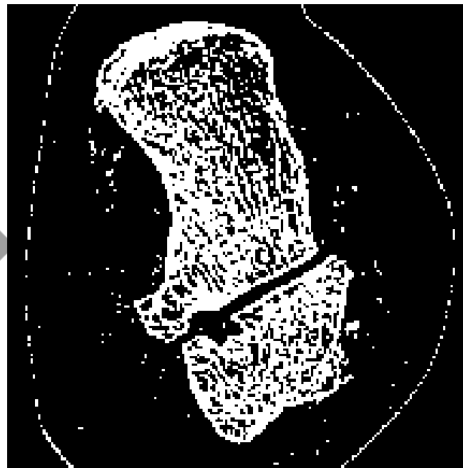
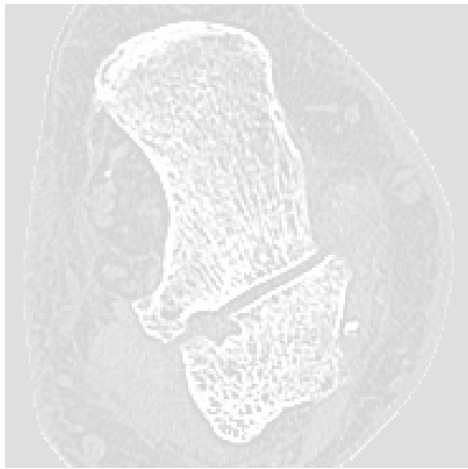
8-connected object



Two components

4-connected object

# Binary Pictures

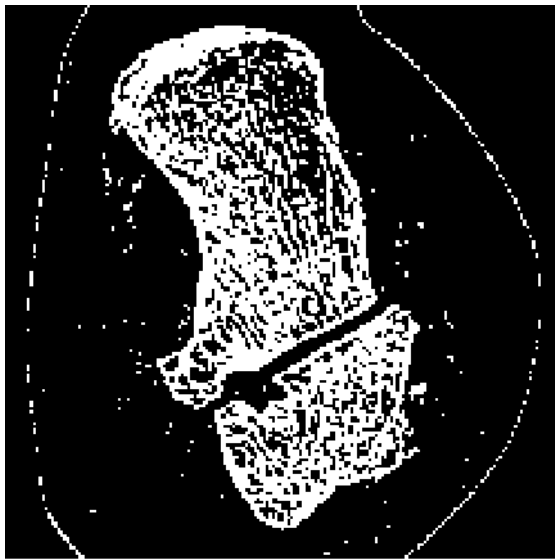


Creation

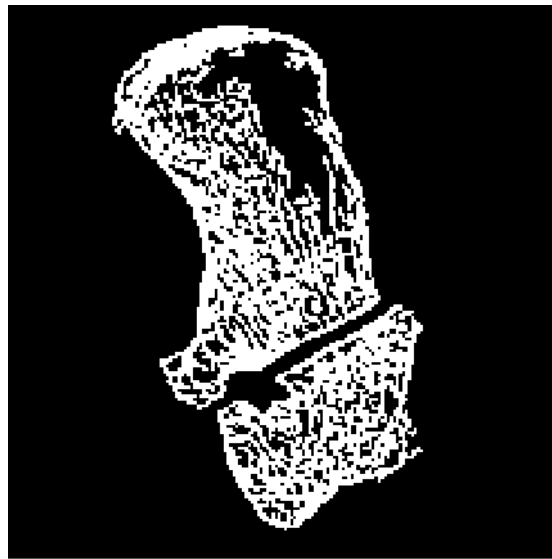
Processing

# Using Connected Components

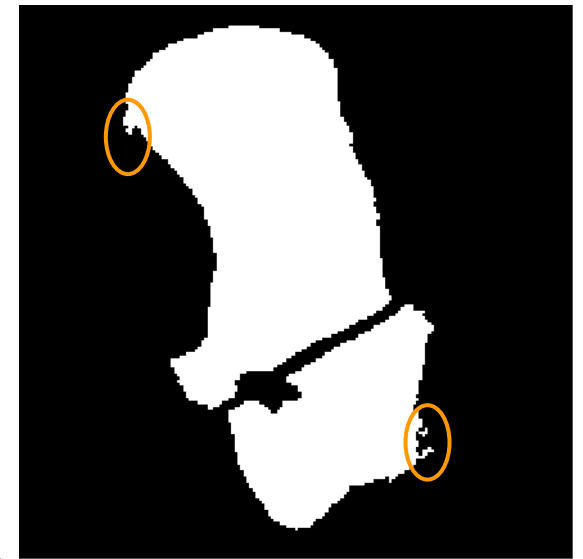
- Pruning isolated islands from the main object
- Filling interior holes of the object



Take the largest 2 components of the object

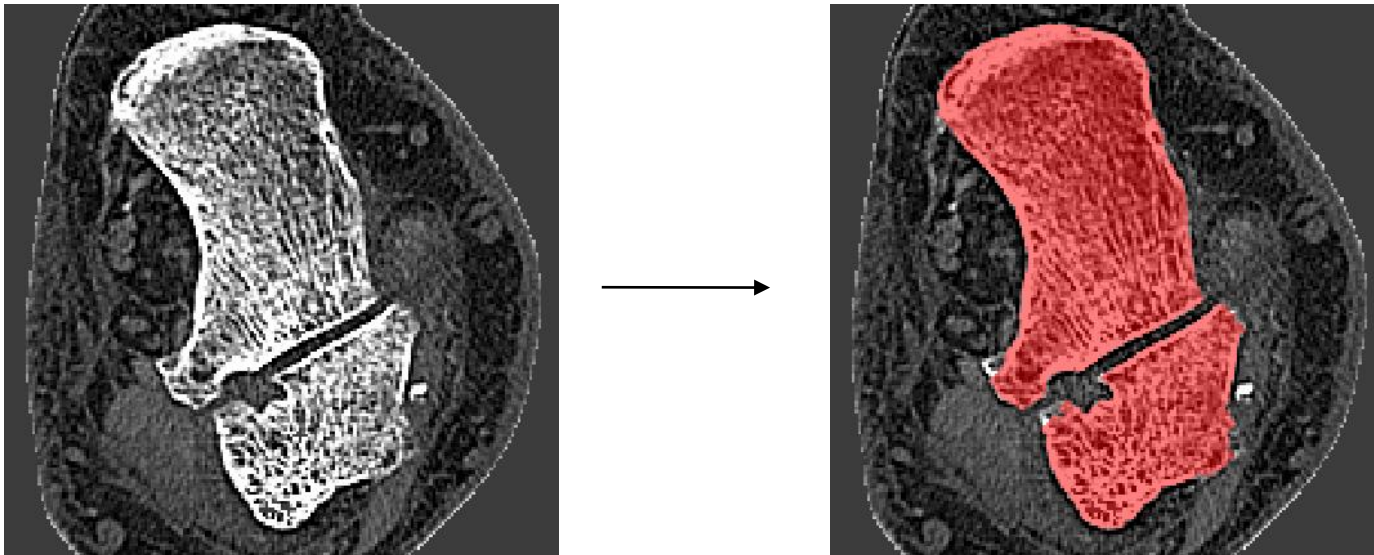


Invert the largest component of the background



# EXAMPLE

- A simple 2D segmentation routine
  - Initial segmentation using thresholding
  - Using connected components and opening/closing to “clean up” the segmentation.

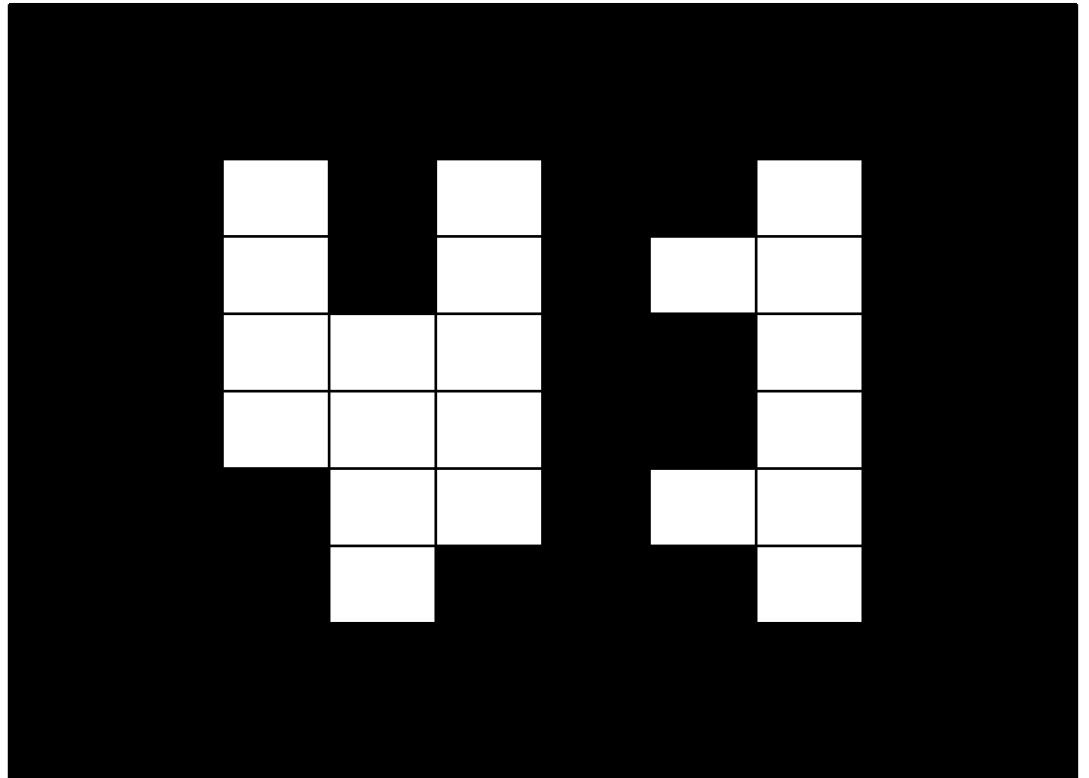
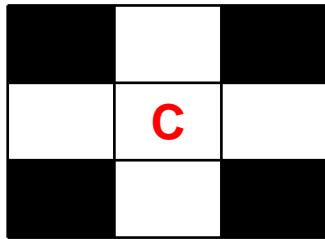




# Mathematical morphology operations

Operations are defined by an structuring element and the image we want to process

Structuring element is defined by a neighborhood and a center

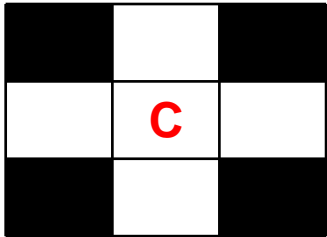


# Basic operations: Dilation

Let  $p=(p_x,p_y)$  the coordinates of all white pixels in image  $I$  and  $s=(s_x,s_y)$  the coordinates of a white pixel in the structuring element with respect to the center:

The dilation is defined as the union of the translations  $p + s$  for all  $s$  in the structuring element.

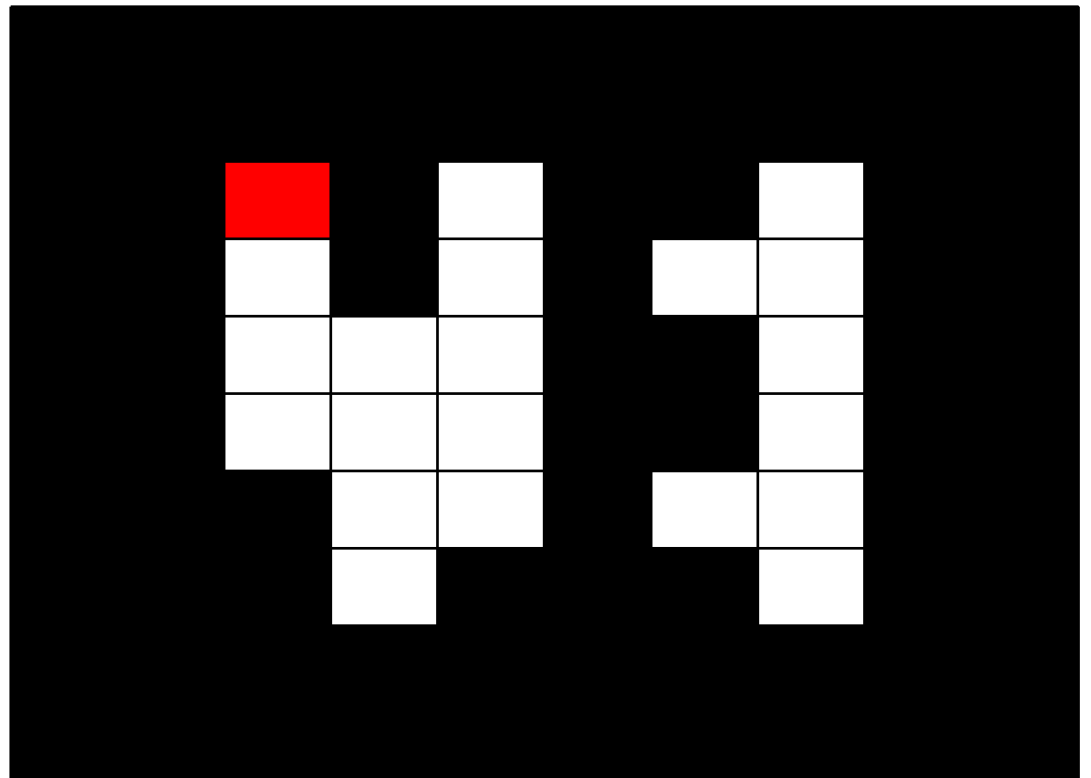
$$D = \{p + s, \forall s \in S, \forall p \in P\}$$



Example:

$S = \{(0,-1),(0,0),(-1,0), (0,+1),(+1,0)\}$

$p = (3,3)$

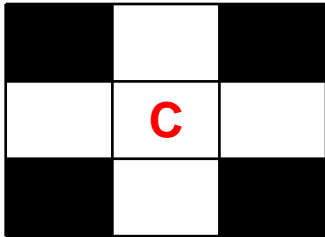


# Basic operations: Dilation

Let  $p=(p_x,p_y)$  the coordinates of all white pixels in image  $I$  and  $s=(s_x,s_y)$  the coordinates of a white pixel in the structuring element with respect to the center:

The dilation is defined as the union of the translations  $p + s$  for all  $s$  in the structuring element.

$$D = \{p + s, \forall s \in S, \forall p \in P\}$$

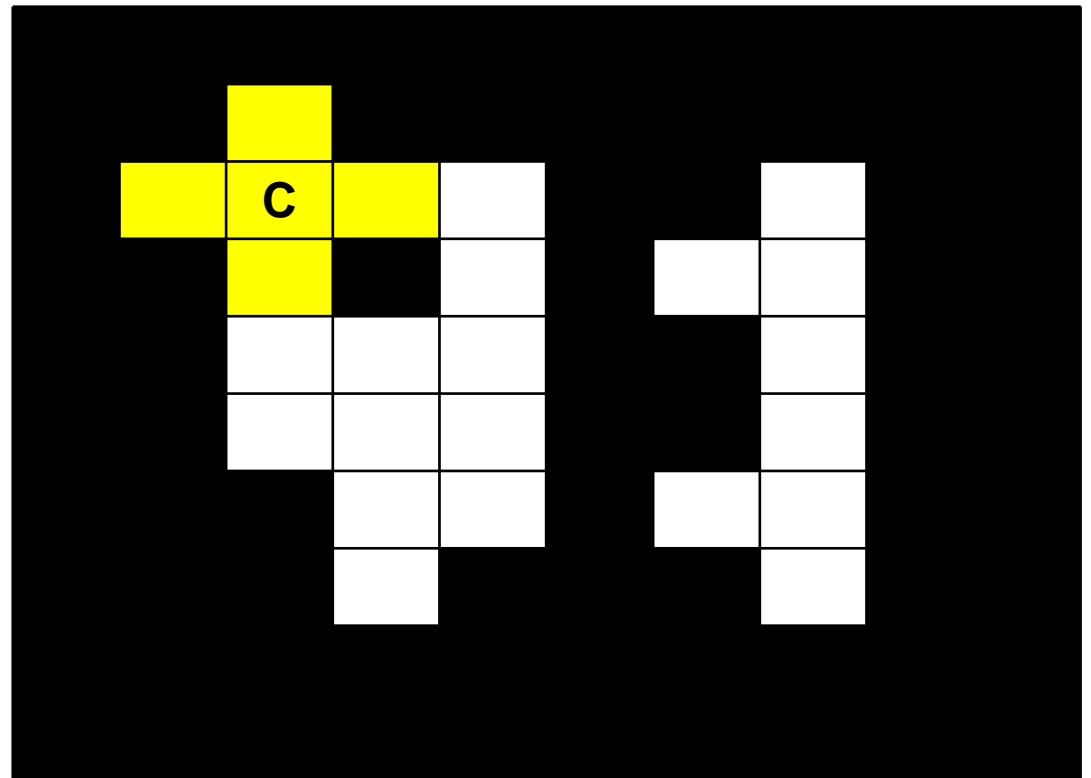


Example:

$$S = \{(0,-1),(0,0),(-1,0),(0,+1),(+1,0)\}$$

$$p = (3,3)$$

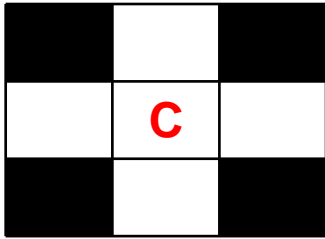
$$p+s = \{(3,2),(3,3),(2,3),(3,4),(4,3)\}$$



# Basic operations: Dilation in practice

Put the SE over all white pixels of the image

$$D = \{p + s, \forall s \in S, \forall p \in P\}$$

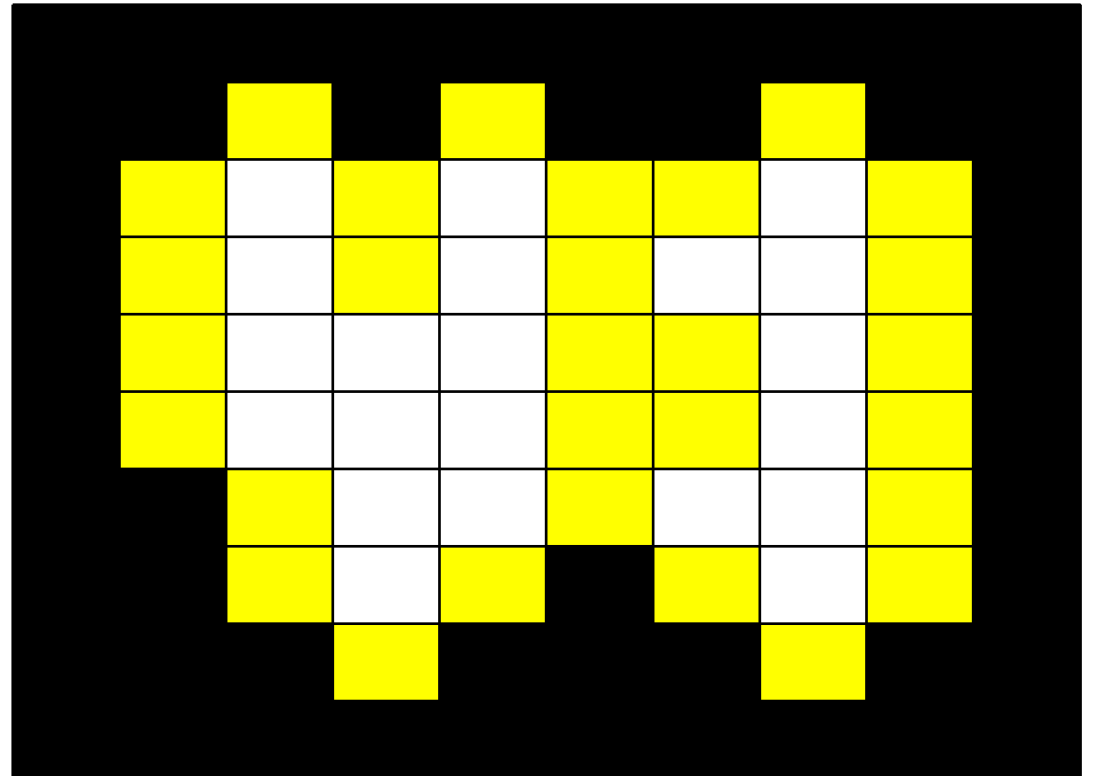


Example:

$S = \{(0,-1), (0,0), (-1,0), (0,+1), (+1,0)\}$

$p = (3,3)$

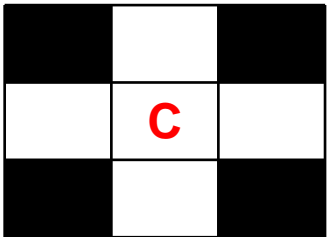
$p+s = \{(3,2), (3,3), (2,3), (3,4), (4,3)\}$



# Basic operations: Erosion

Let  $p=(p_x,p_y)$  the coordinates of a pixel in the image and  $P$  the set of the coordinates of all white pixels in image  $I$ . Let  $s=(s_x,s_y)$  the coordinates of pixel in the structuring element and  $S$  the set of the coordinates of all white pixel in the structuring element with respect to the center:

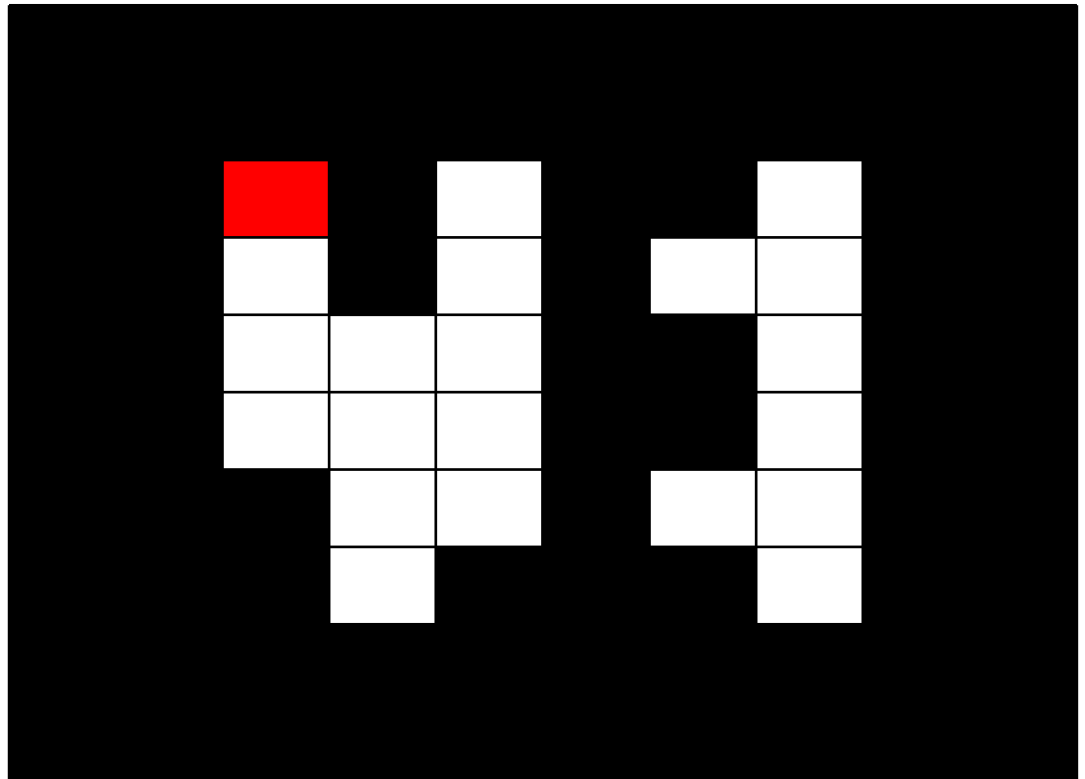
The erosion is defined as the elements  $E = \{p \mid p + s \in P, \forall s \in S\}$



Example:

$S = \{(0,-1), (0,0), (-1,0), (0,+1), (+1,0)\}$

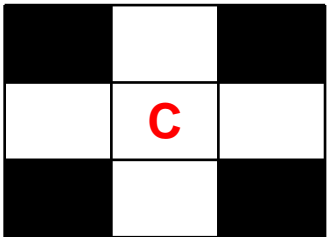
$p = (3,3)$



# Basic operations: Erosion

Let  $p=(p_x,p_y)$  the coordinates of a pixel in the image and  $P$  the set of the coordinates of all white pixels in image  $I$ . Let  $s=(s_x,s_y)$  the coordinates of pixel in the structuring element and  $S$  the set of the coordinates of all white pixel in the structuring element with respect to the center:

The erosion is defined as the elements  $E = \{p \mid p + s \in P, \forall s \in S\}$

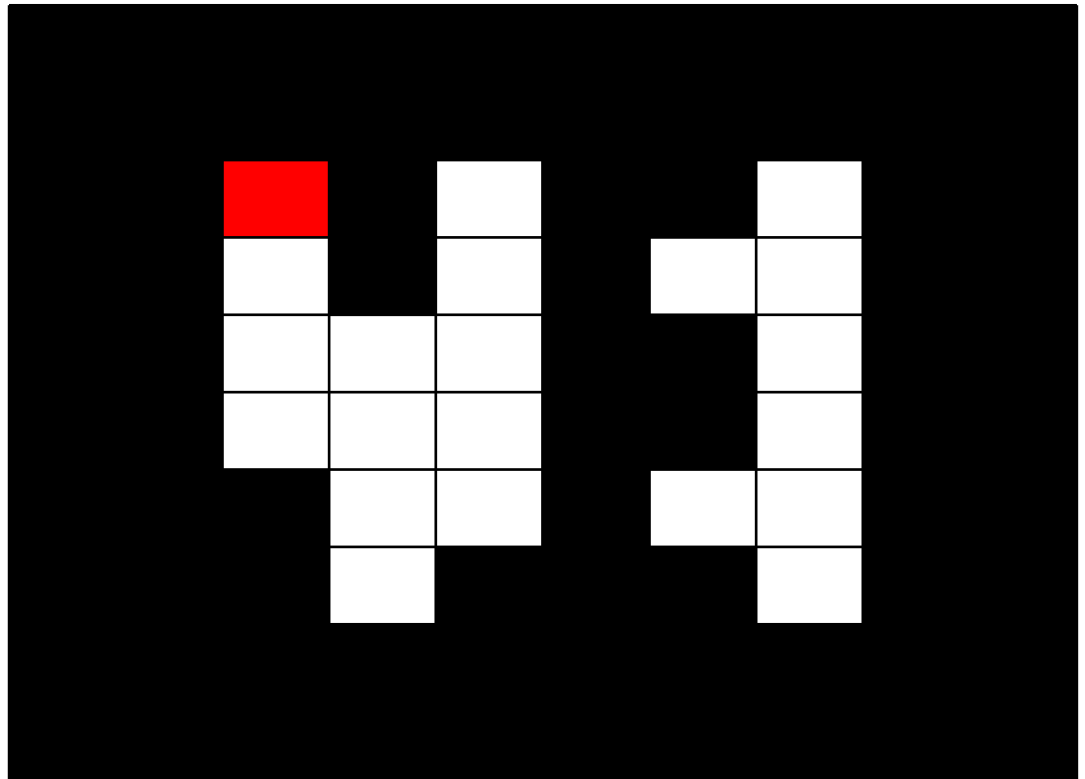


Example:

$S = \{(0,-1),(0,0),(-1,0),(0,+1),(+1,0)\}$

$p = (3,3)$

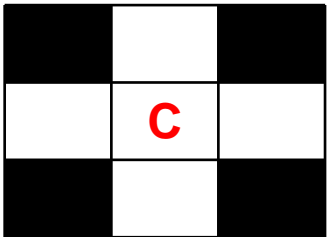
$p+s = \{(3,2),(3,3),(2,3),(3,4),(4,3)\}$  pertanyen a  $P$ ?



# Basic operations: Erosion

Let  $p=(p_x,p_y)$  the coordinates of a pixel in the image and  $P$  the set of the coordinates of all white pixels in image  $I$ . Let  $s=(s_x,s_y)$  the coordinates of pixel in the structuring element and  $S$  the set of the coordinates of all white pixel in the structuring element with respect to the center:

The erosion is defined as the elements  $E = \{p \mid p + s \in P, \forall s \in S\}$

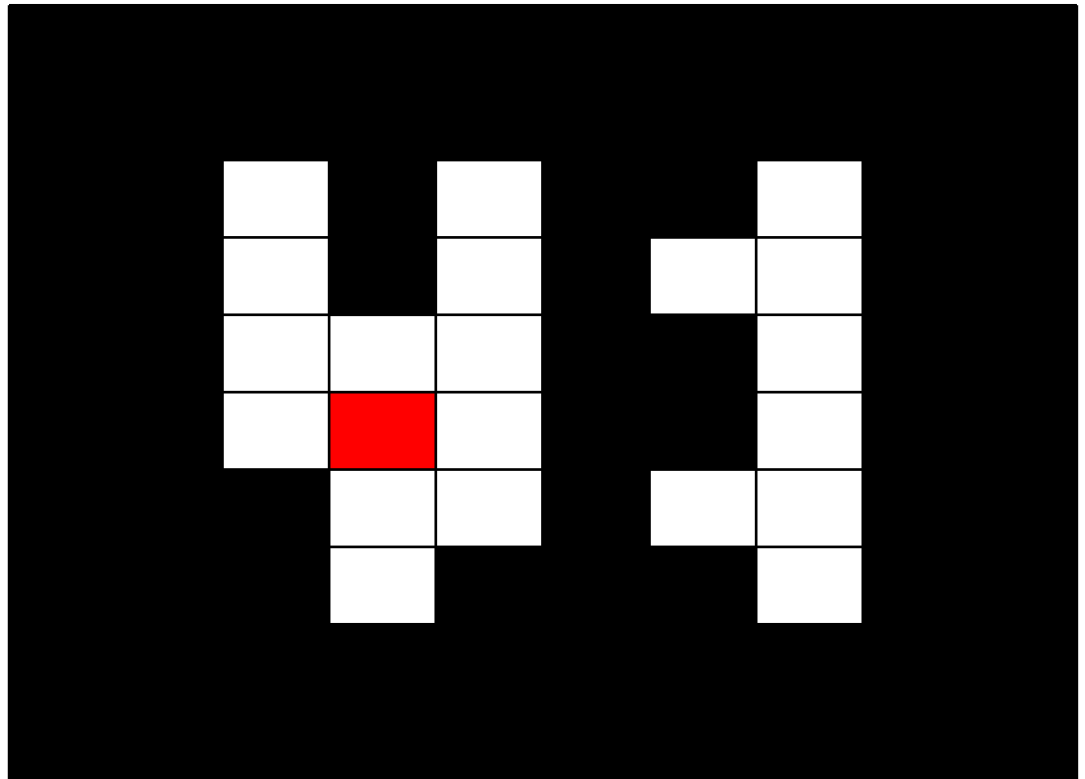


Example:

$S = \{(0,-1),(0,0),(-1,0),(0,+1),(+1,0)\}$

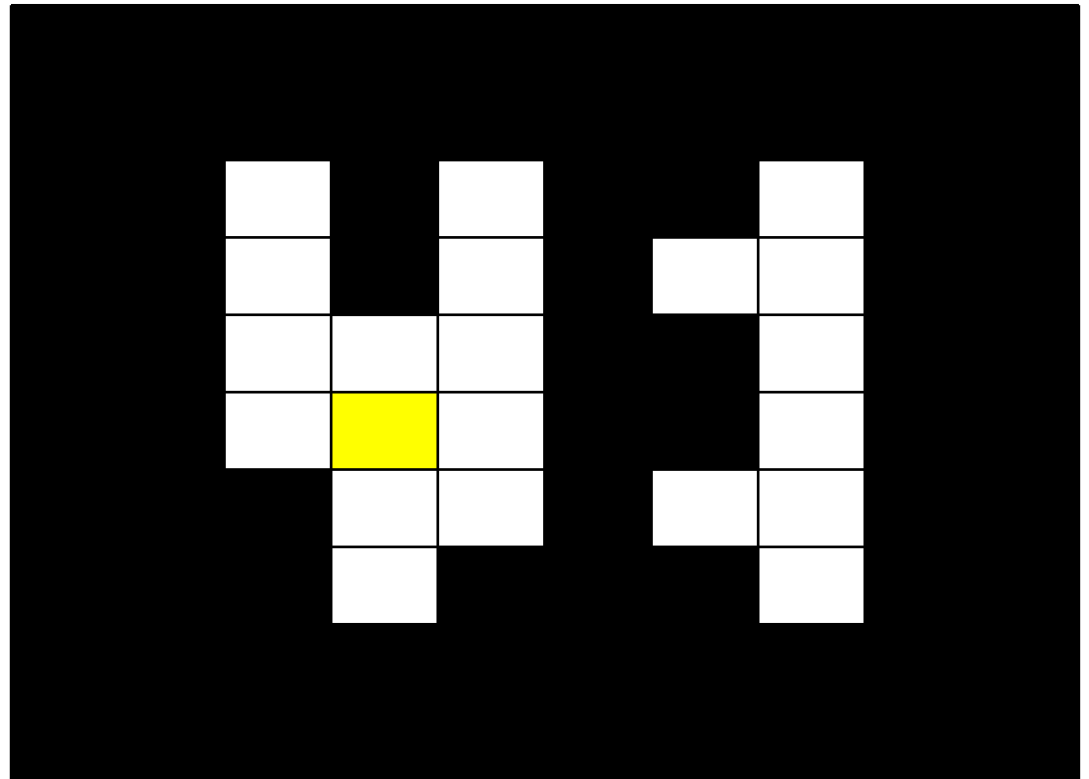
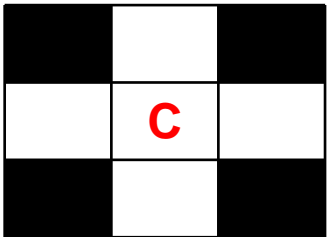
$p = (6,4)$

$p+s = \{(6,3),(6,4),(5,4),(6,5),(7,4)\}$  pertanyen a  $P$ ?



# Basic operations: Erosion in practice

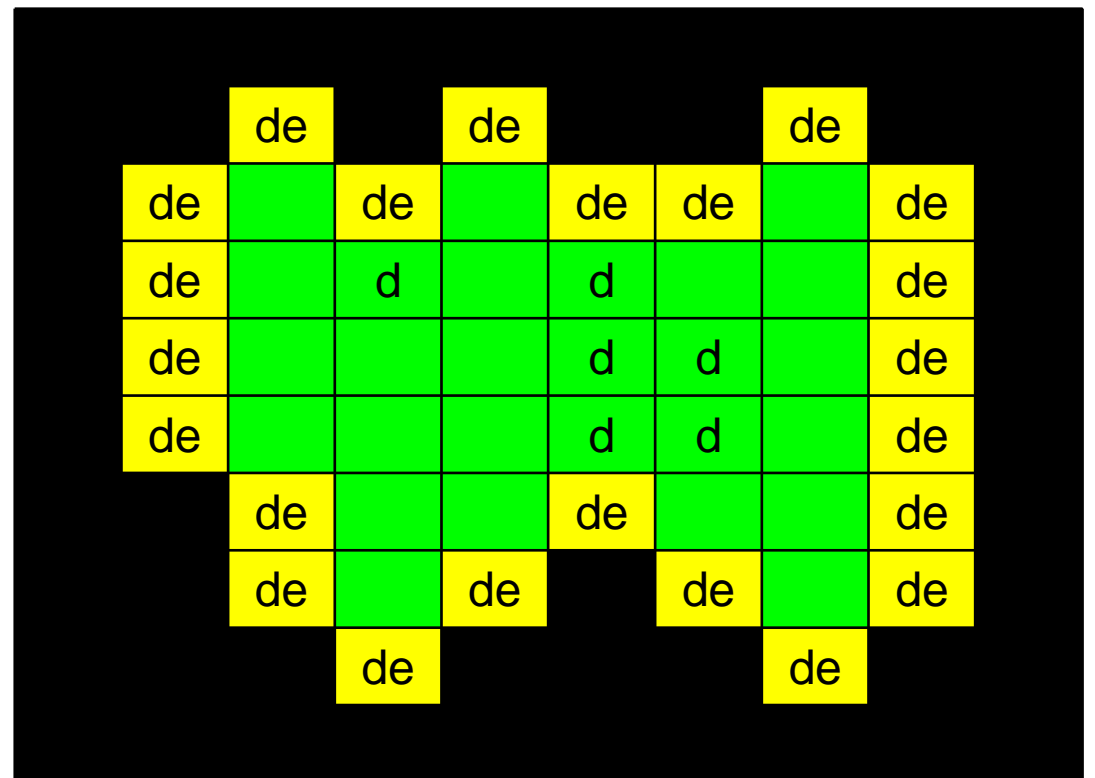
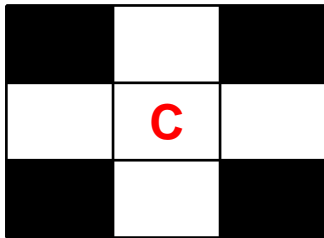
“Keep all the points in which the structuring element fits in the white region”  
(only true if the center belongs to the inside of the structuring element)





# Composing operations: Closing = Dilation + Erosion

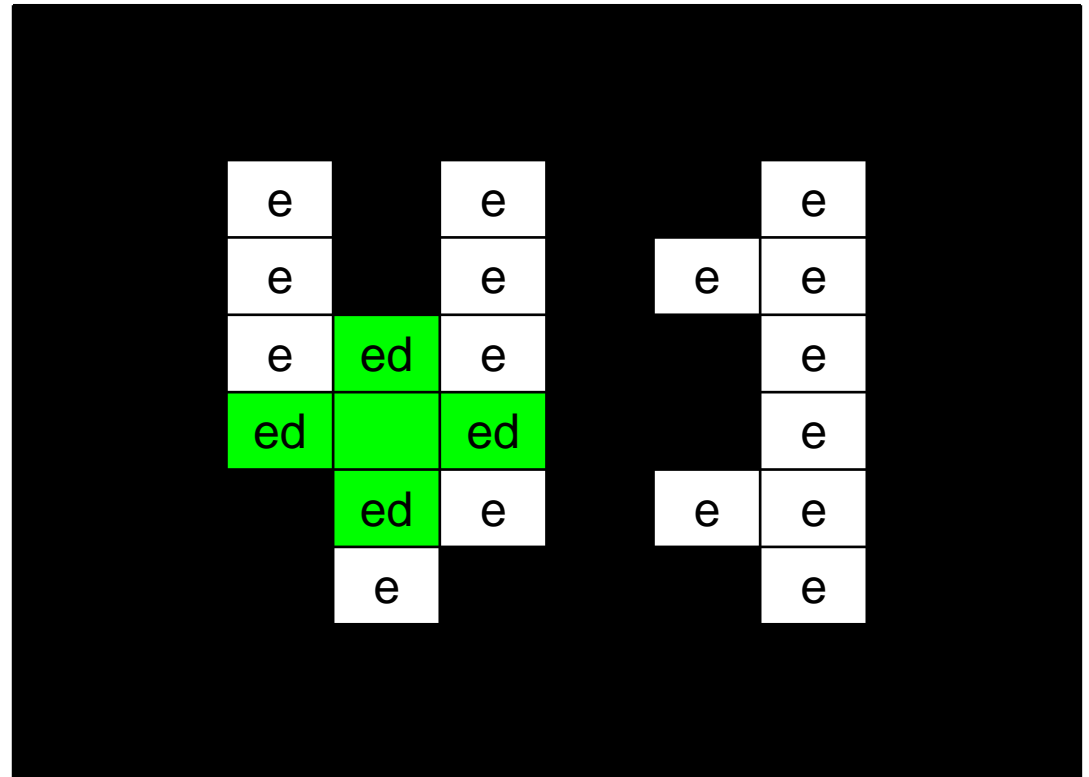
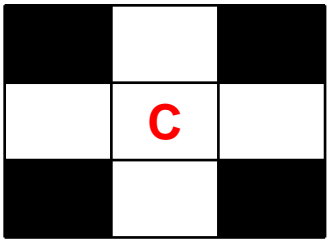
Recovers the original objects closing holes, and joining structures smaller than the SE



# Composing operations:

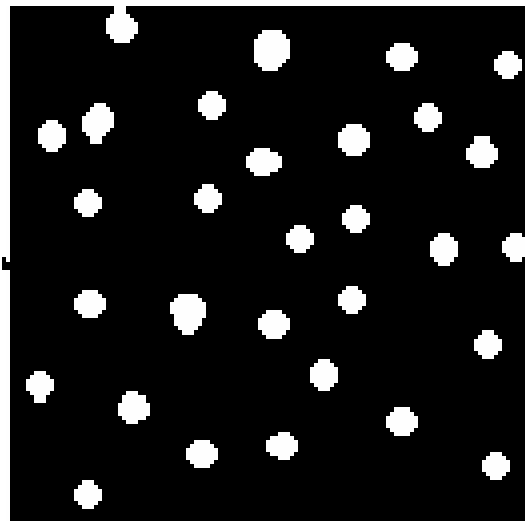
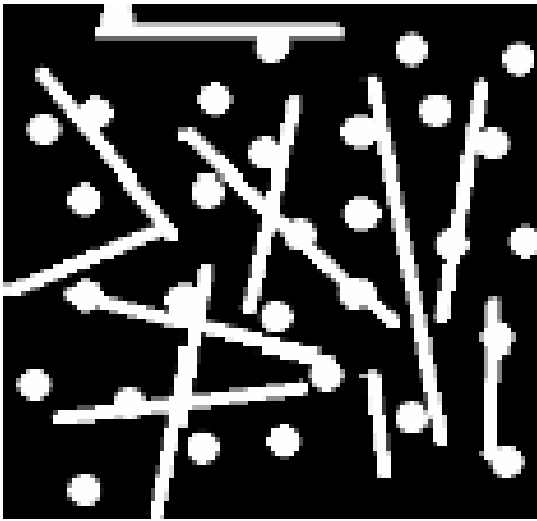
## Opening = Erosion + Dilation

Removes parts of the elements not conforming the SE shape



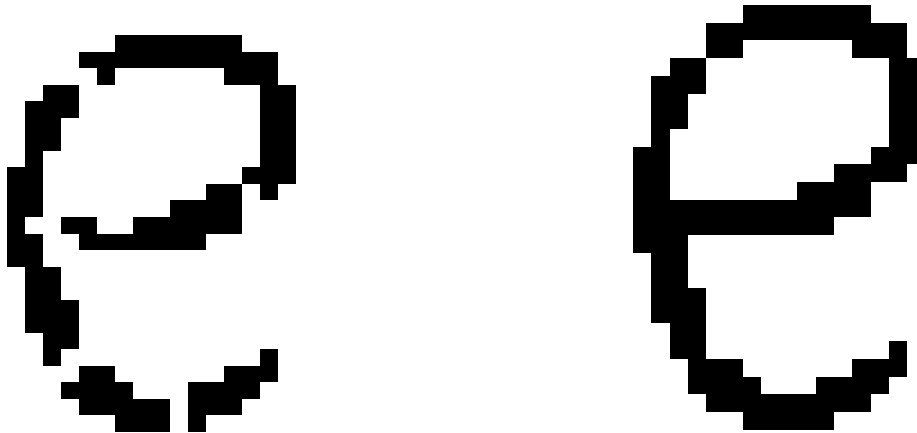
# Aplicaciones

Separation of structures: If we know that we are looking for disk structures of 7 pixels of diameter, which operation would you use?



# Aplicacions

Restauration of binary scanned letters.



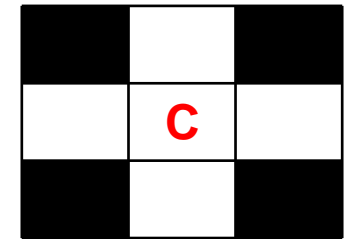
Which operation did I use?

# Morphological Gradient: Contour detection

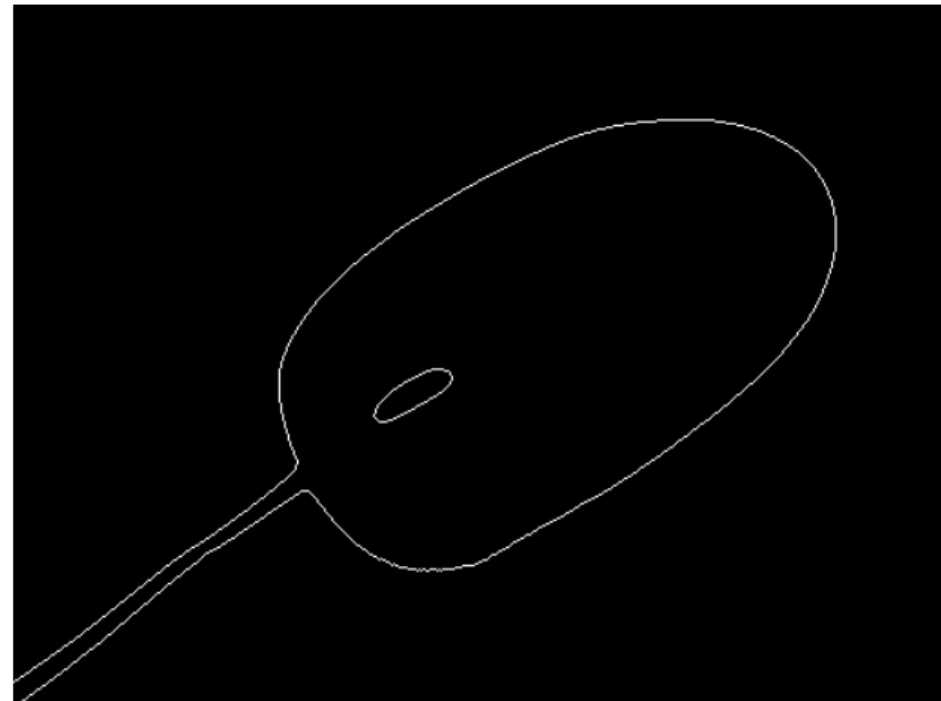
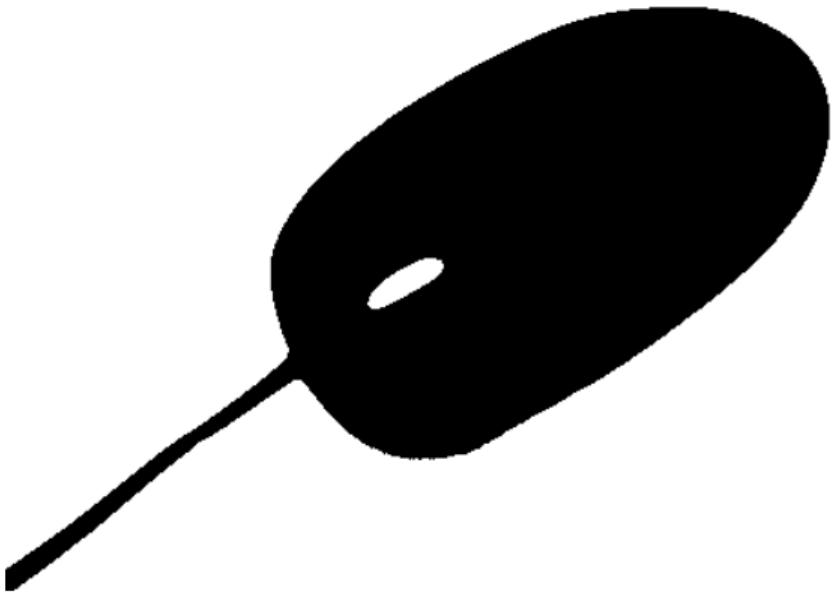
Dilation(I)-I External contour

I – erosion(I) Internal contour

Dilation(i)-Erosion(i) Both contours



Inversa de I



# Morphological Gradient: Contour detection

Dilation(I)-I External contour

I – erosion(I) Internal contour

Dilation(i)-Erosion(i) Both contours

