



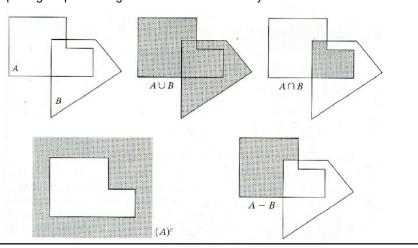
# Morphology

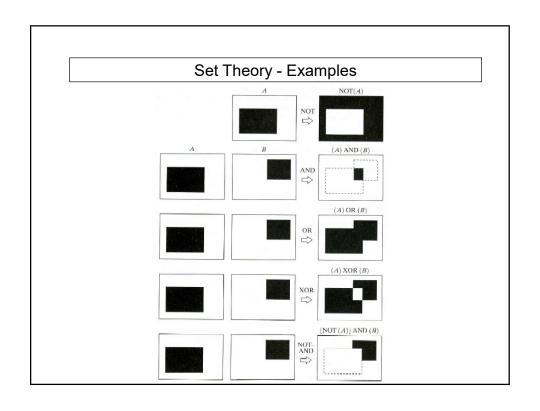
Material de clase: http://www.robotica-up.org/CV2018.zip

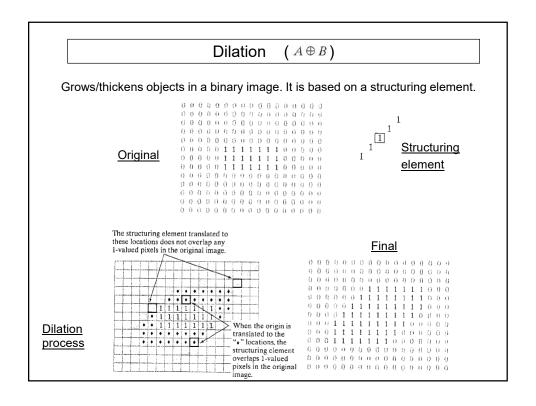
# Introduction

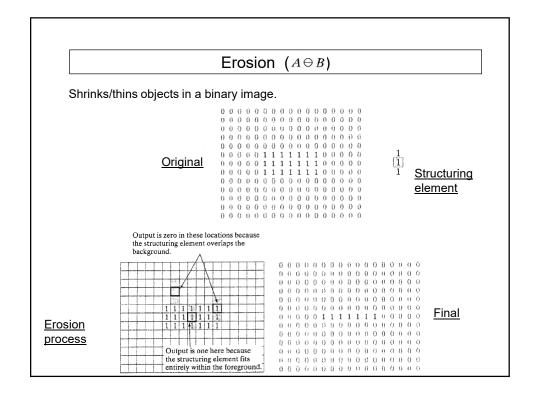
### What is morphology?

It is a tool for extracting components of an image (such as regions, contours, skeletons,...) which are useful in shape representation and description. Morphological processing is based on the set theory.









# Implementation – Structuring Element

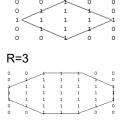
s=strel('option',R)

'diamond' - creates a flat diamond-shaped structuring element with the specified size, R. R is the distance from the structuring element origin to the points of the diamond. R must be a nonnegative integer scalar.

'disk' - creates a flat disk-shaped structuring element with the specified radius, R. R must be a nonnegative integer.

'line' - creates a flat linear structuring element

'octagon' - creates a flat octagonal structuring element



**Others** 

'pair', 'periodicline', 'rectangle', square'

Help strel .... for details

R=1

# Opening and closing

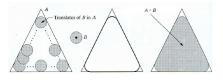
**Opening:** Removes completely regions of an object that cannot contain the structuring element, smooths object contours, breaks thin connections and removes thin protrusions

$$A \circ B = (A \ominus B) \oplus B$$

**Closing:** Also tends to smooth object contours. Joins narrow breaks and fills holes smaller than the structuring element.

$$A \cdot B = (A \oplus B) \ominus B$$

#### Geometric interpretation



Opening



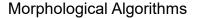
Closing

## Hit-or-Miss Transformation ( $A \otimes B$ )

Objective: To detect pixel configurations in an image upon the use of the background information.  $A \otimes B = (A \ominus B_1) \cap (A^c \ominus B_2)$ 

 $\begin{array}{c} B_1 \\ 1 \\ \hline 1 \\ 1 \end{array}$ 

The final image consists of all locations that match the pixels in B1 (a "hit") and none of the pixels in B2 (a "miss")



Border extraction 
$$\beta(A) = A - (A \ominus B)$$

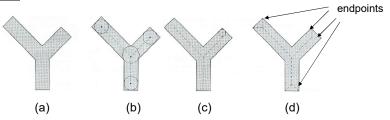
Region filling 
$$X_k = (X_{k-1} \oplus B) \cap A^c \quad k = 1, 2, 3, ...$$

A: imagen to be filled (M,N) B: structuring element

 $X_k$ : image (M,N) of "0" with a starting point "1" The algorithm ends at interaction k if  $X_k = X_{k+1}$ 

Skeletons

1 1 1 0 0 0 0 0

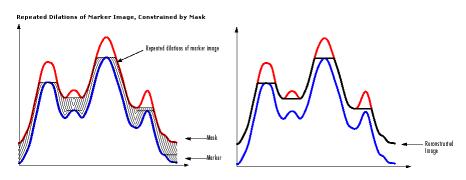


#### Morphological Algorithms - Connected Elements Objective: Object detection [L,num]=bwlabel(f,conn) f - Input binary image Conn - desired connectivity (default 8) 1 1 1 0 2 2 1 0 2 2 L – Label matrix 1 0 2 2 1 1 Num - total number of connected 1 1 1 0 0 0 1 1 1 0 0 0 elements 1 1 1 0 0 0 4 0 1 0 0 0 2 1 1 1 0 0 0 4 0 1 1 1 0 0 0 2 0 1 1 0 0 3 0 0 1 1 1 0 0 2 0 0

1 1 1 0 0 0 0 0

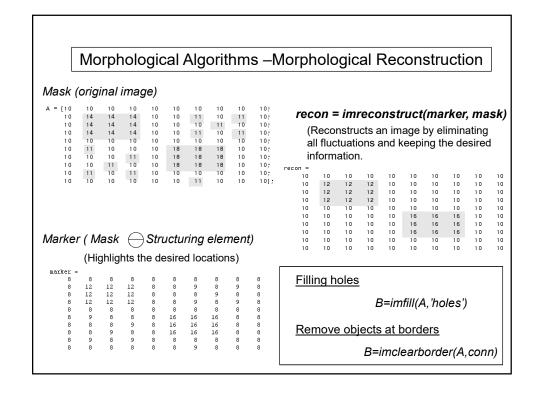
Labeling





Morphological reconstruction is based on an erosion with the following properties:

- It requires 2 images and one structuring element (instead of one image and one structuring element)
- Image 1 (f) is the *marker* (the starting point of the transformation)
- Image 2 (g) is the *mask* (the original image, the end of the transformation)



# Morphology in gray-level Images

Dilation
Erosion
Opening
Closing
Same as binary

Morphological Gradient (change of gray-level in an image

$$A \oplus B - A \ominus B$$

Top-hat transformation: Background equalization

$$h = f - (f \circ b)$$

Reconstruction: Better binarize the image first and work in binary

# Binarization by Threshold Selection

