

# Lesson 5: Morphology

1. Introduction
2. Expansion and contraction
3. Dilation and erosion
4. Opening and closing
5. Skeletons
6. Distance maps



# 1/6. Introduction

- **Morphology:** analysis of the **shape** of connected components.
- **Algebraic** operators applicable to binary images in order to extract components useful in representing shape
  - Contours
  - Convex hull
  - Skeletons
- Goals:
  - Image simplification
  - Elimination of irrelevances
  - Preservation of useful characteristics

Parallelizable techniques

- **Definitions:** Let  $A$  and  $B$  be binary images,  $p$  and  $q$  two pixels with indices  $[i, j]$  y  $[k, l]$  respectively, and  $\Omega$  the universal binary image.

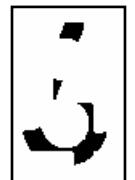
- **Union:**

$$A \cup B = \{p | p \in A \vee p \in B\}$$



- **Intersection:**

$$A \cap B = \{p | p \in A \wedge p \in B\}$$



- **Complement:**

$$\bar{A} = \{p | p \in \Omega \wedge p \notin A\}$$



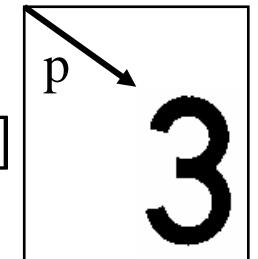
- **Difference:**

$$A - B = A \cap \bar{B}$$



- **Translation:**

$$A_p = \{a + p | a \in A\}$$



- **Vectorial sum:**

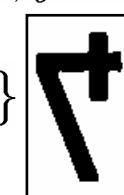
$$p + q = [i + k, j + l]$$

- **Vectorial difference:**

$$p - q = [i - k, j - l]$$

- **Reflex:**

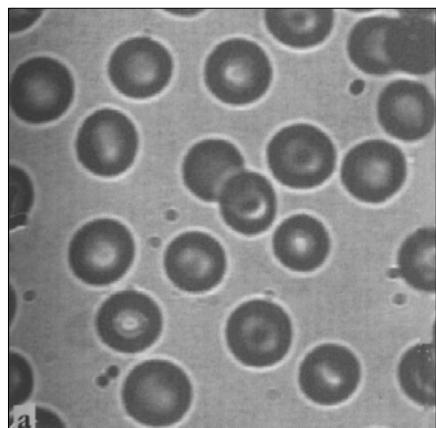
$$A' = \{-p | p \in A\}$$



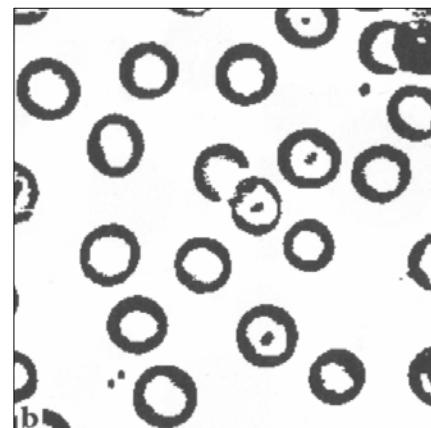
# Introduction

- Example: thresholding

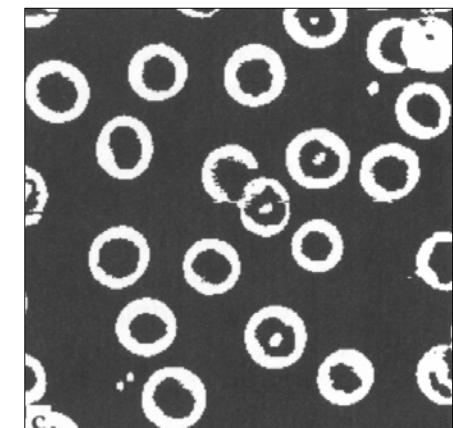
*A*



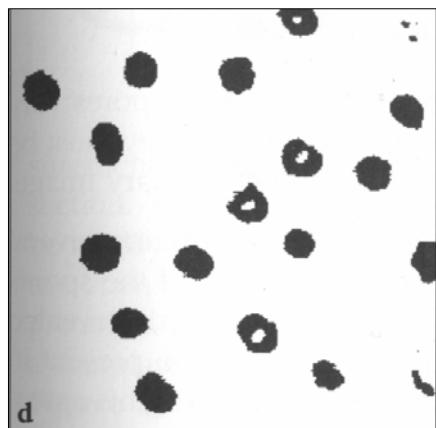
*B*



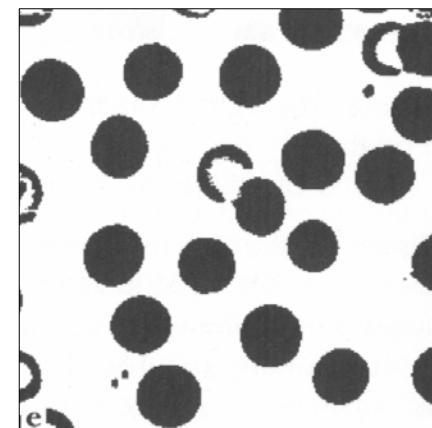
$\overline{B}$



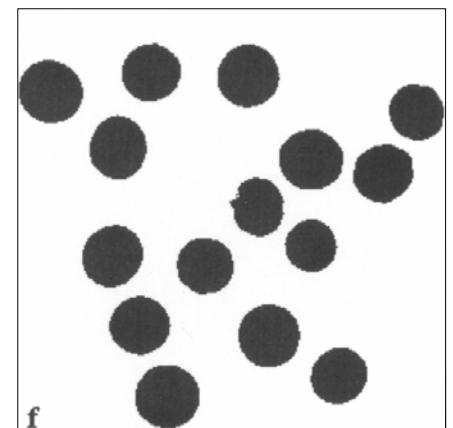
*C*



$B \cup C$



*D*



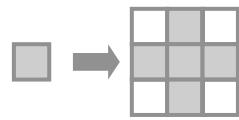
**Not touching borders**

**Size filter**

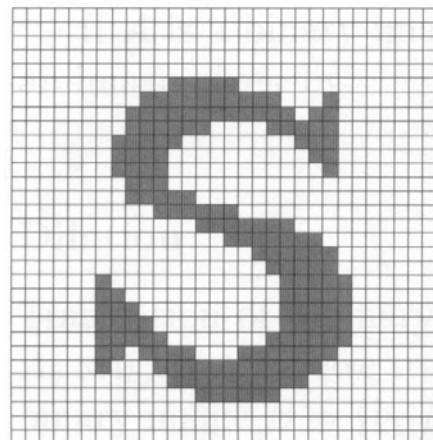
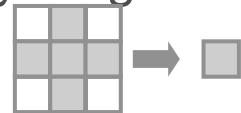


## 2/6. Expansion and contraction

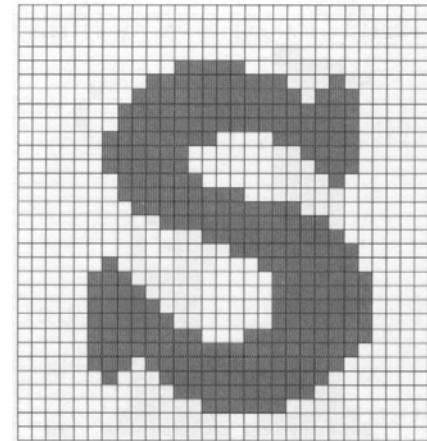
- Transformations to turn foreground pixels into background and vice-versa.
- Expansion:** turn a pixel from 0 to 1 if any neighbor is 1.



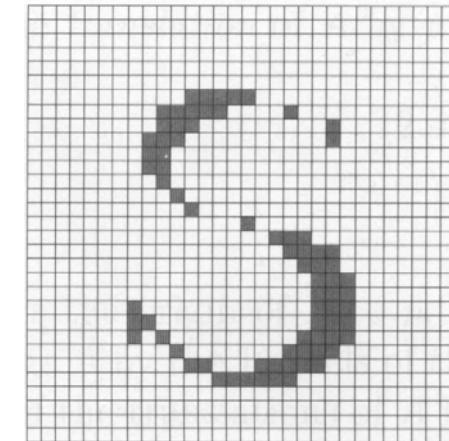
- Contraction:** turn a pixel from 1 to 0 if any neighbor is 0.



Expansion (4v)



Contraction (4v)



Expanding a blob is equivalent to contracting the background

$S^k$   $S$  expanded  $k$  times

- $S^{-k}$   $S$  contracted  $k$  times

$$(S^m)^{-n} \neq (S^{-n})^m$$

- Neither invertible:  
 $\neq S^{m-n}$

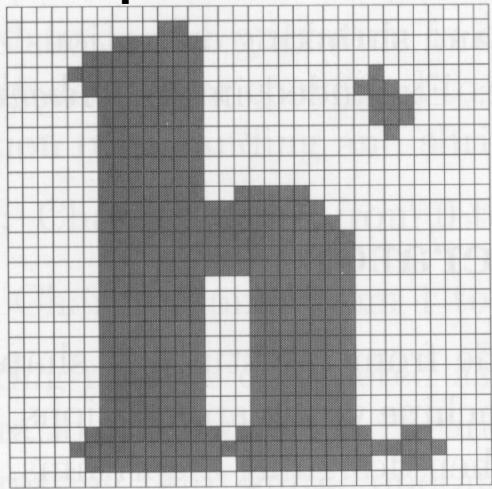
$$S \subset (S^k)^{-k}$$

$$S \supset (S^{-k})^k$$

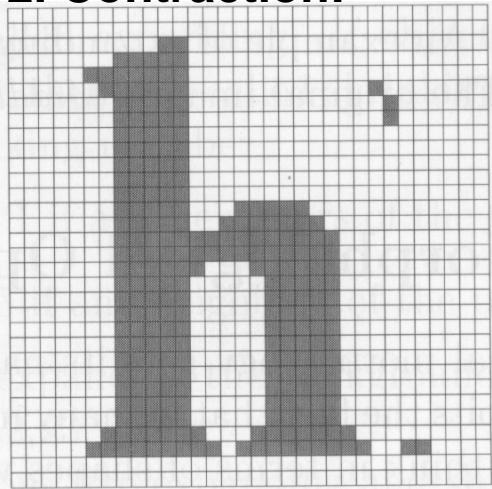


# Expansion y contracción

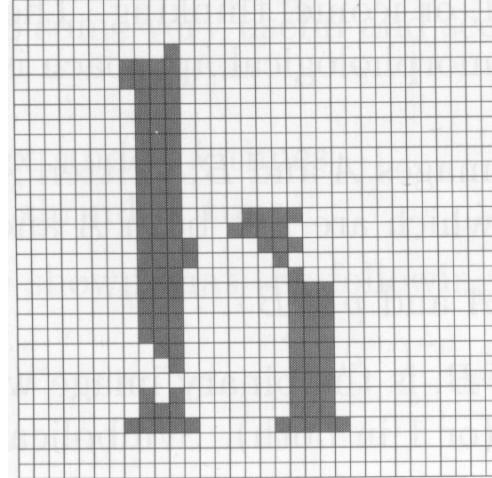
1. Expansion:



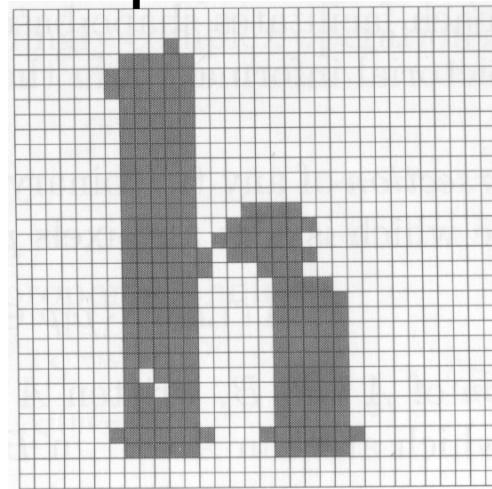
2. Contraction:



1. Contraction:



2. Expansion:



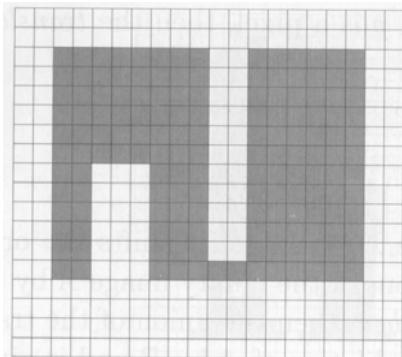
- **Expansion + Contraction:** elimination of undesired holes (salt noise).

- **Contraction + Expansion:** Elimination of noise (pepper noise).

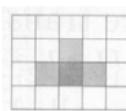


# 3/6. Dilation y erosion

A

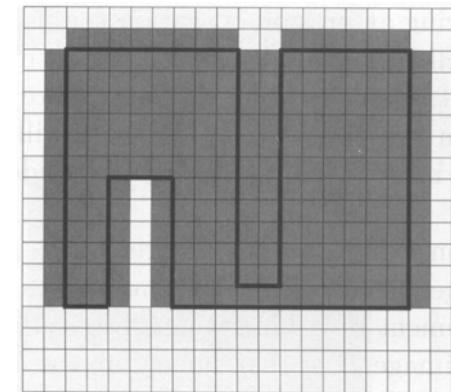


B

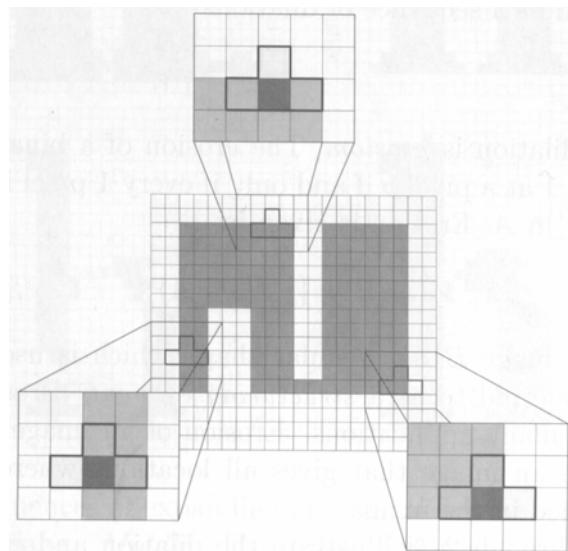


- **Dilation:** union of translations of an image  $A$  by each pixel of image  $B$ , called *structural element*.

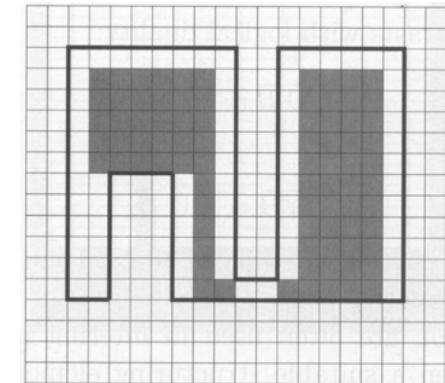
$$A \oplus B = \bigcup_{b_i \in B} A_{b_i}$$



- **Erosion** inverse operation



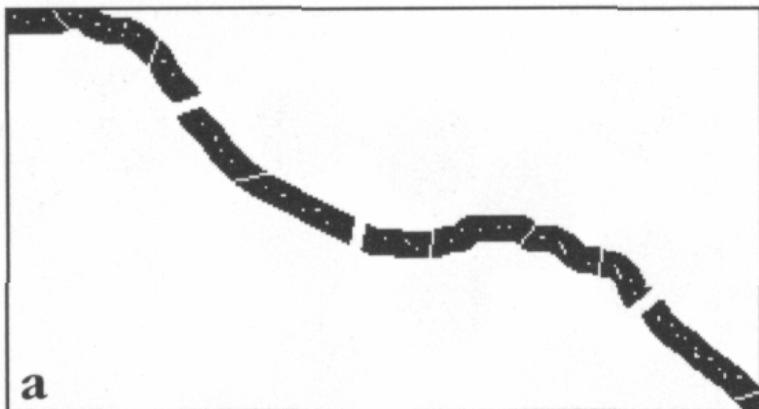
$$A \ominus B = \{p | B_p \subseteq A\}$$



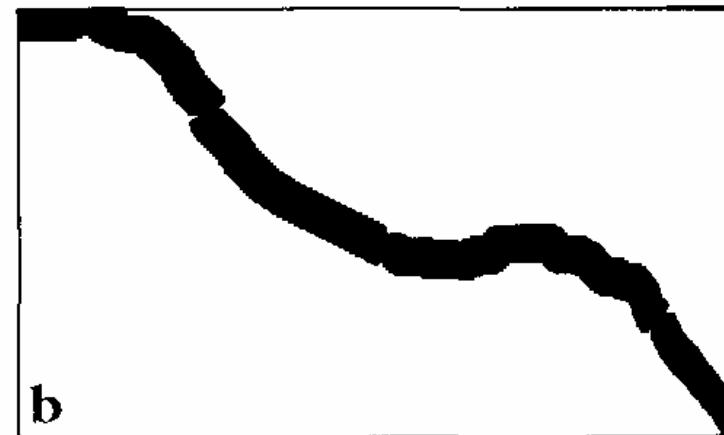
Asociative y commutative?



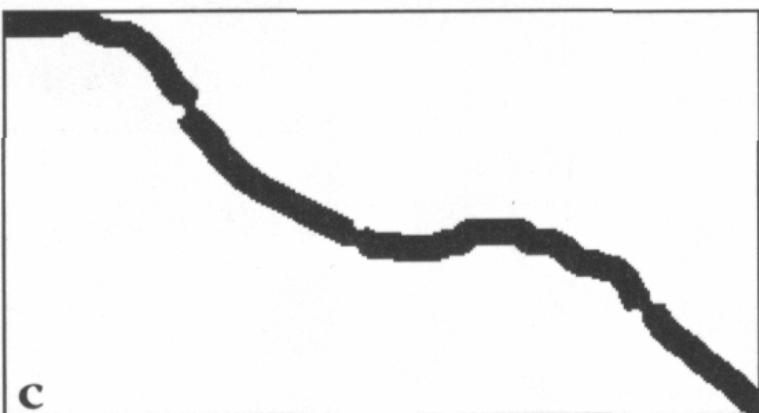
# Object connection



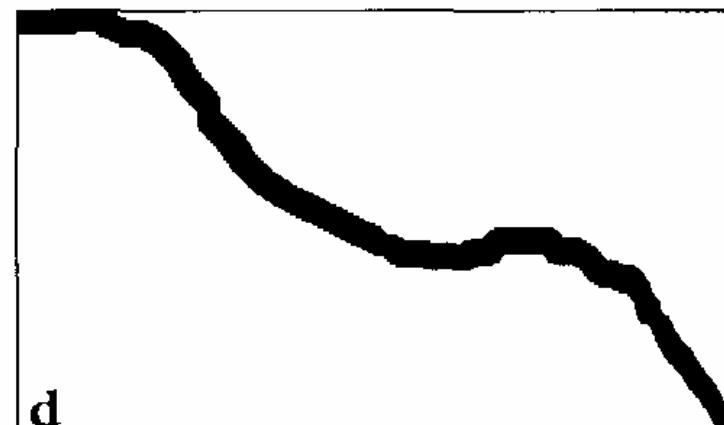
Original



Dilated twice



Eroded twice

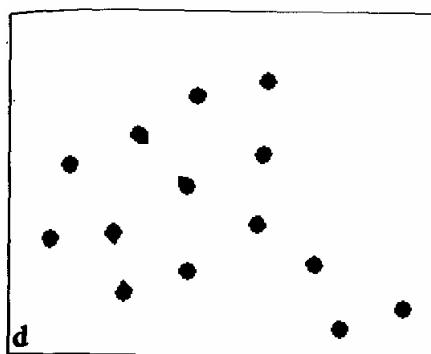
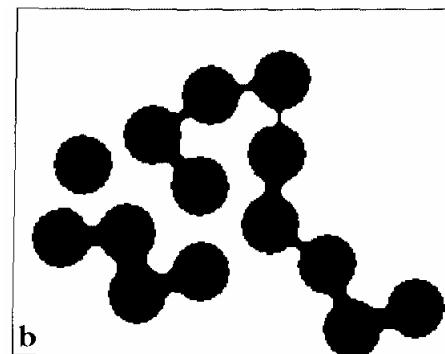
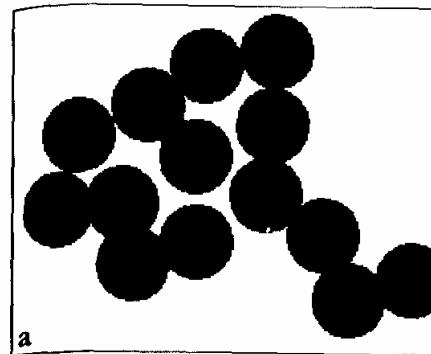


Dilated and eroded four times

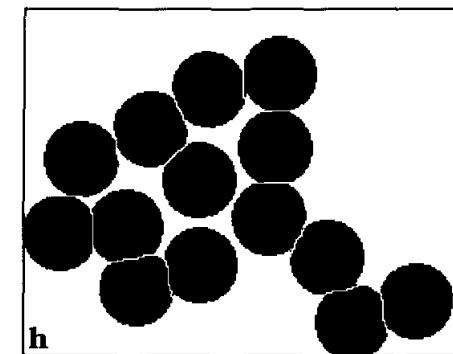
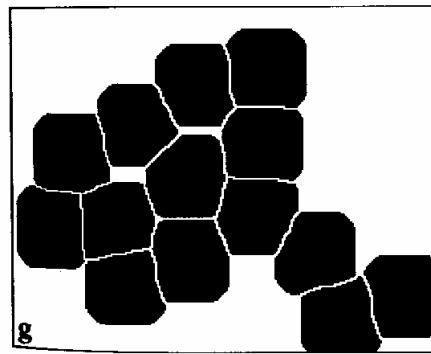
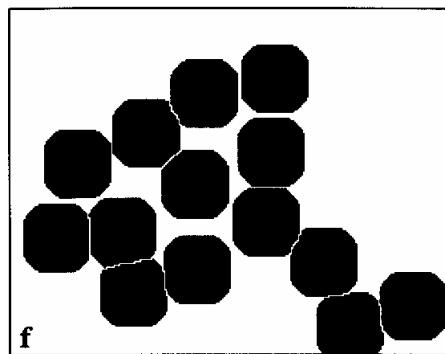
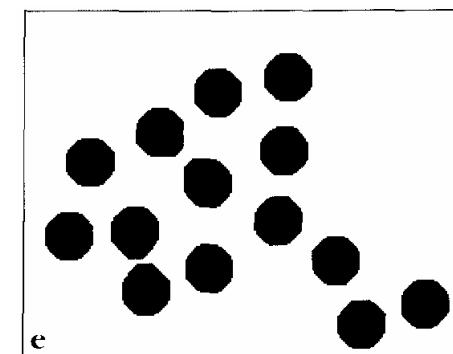


# Object separation

- a. Original
- b. Eroded twice
- d. Eroded 7 times

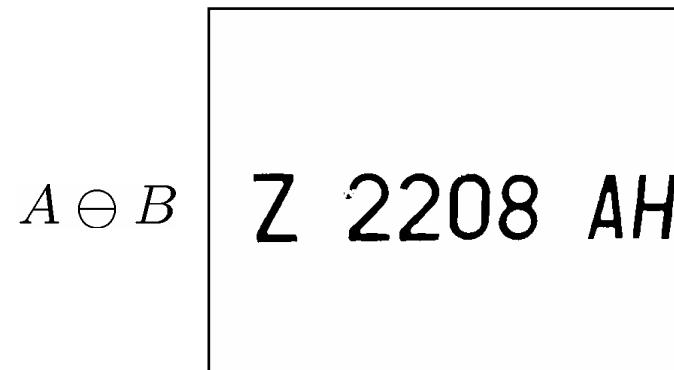
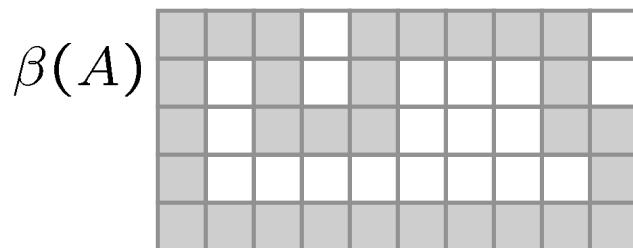
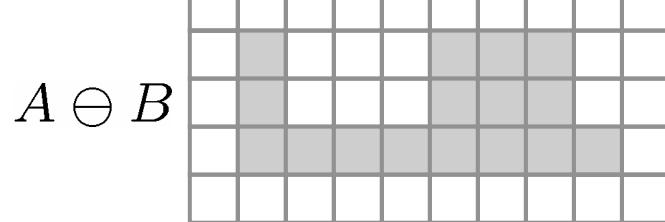
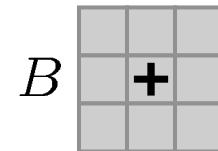
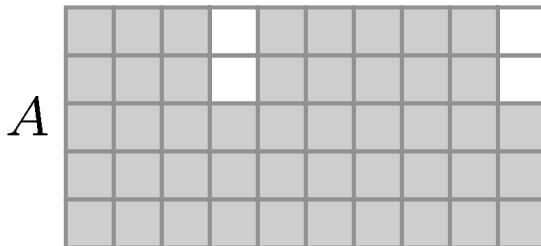


- e. Dilated four times with XOR
- f. Dilated seven times with XOR
- g. Dilated nine times with XOR
- h. AND with original image



# Contour extraction

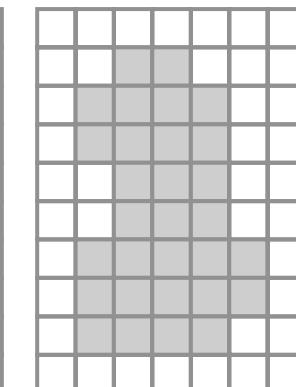
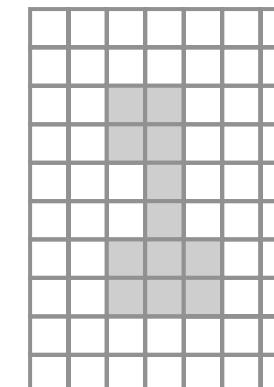
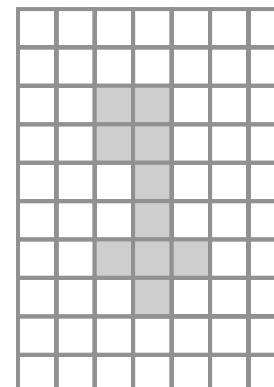
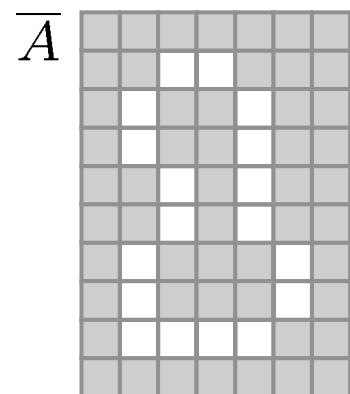
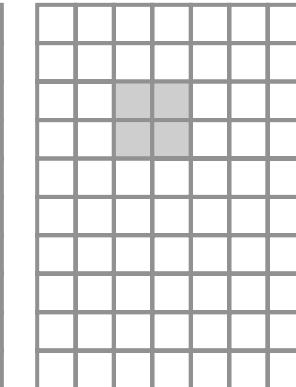
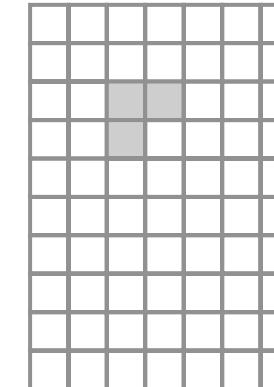
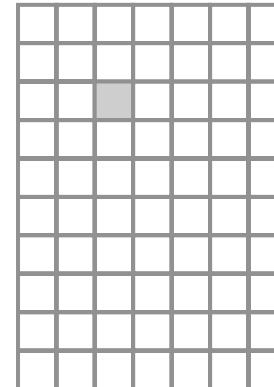
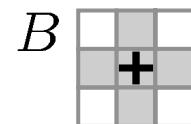
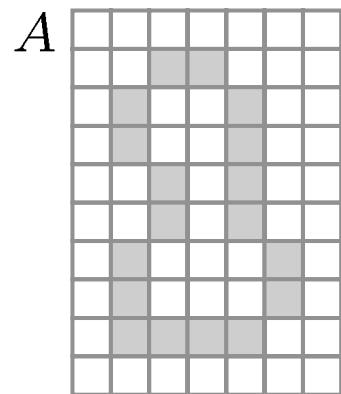
$$\beta(A) = A - (A \ominus B)$$



# Region filling

$$X_0 = p$$

$$X_k = (X_{k-1} \oplus B) \cap \overline{A}$$



$$X_k = X_{k-1}$$

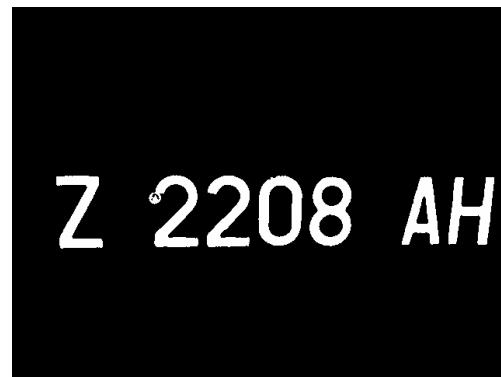


# Holes

$M$

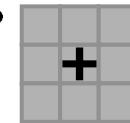


$A = \overline{M}$



The total area might be less sensitive to noise, although the Euler number might be more discriminating.

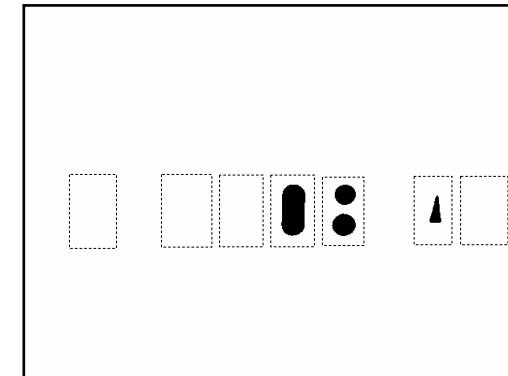
$B$



$X_0$



$X_n$



# Hit or Miss

- Selection of pixels with some particular properties (corners, isolated, border).

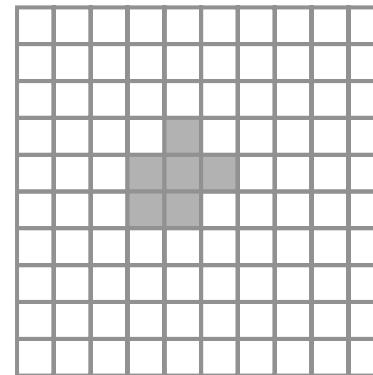
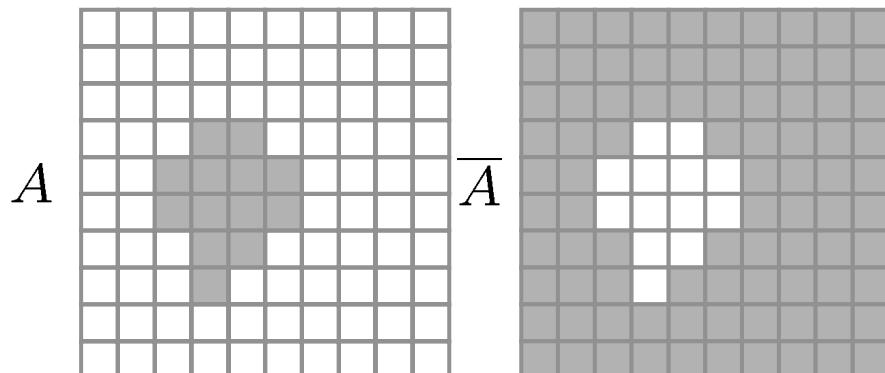
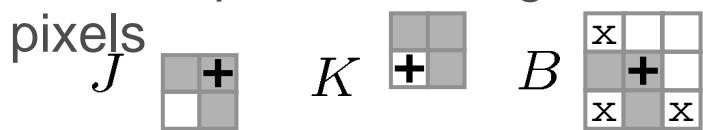
$$B = (J, K), J \cap K = \emptyset$$

$$A \otimes B = (A \ominus J) \cap (\overline{A} \ominus K)$$

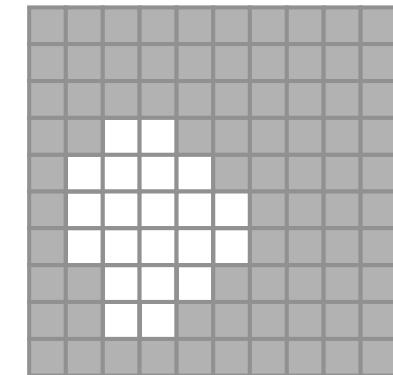
- Example: sup. rig. corners



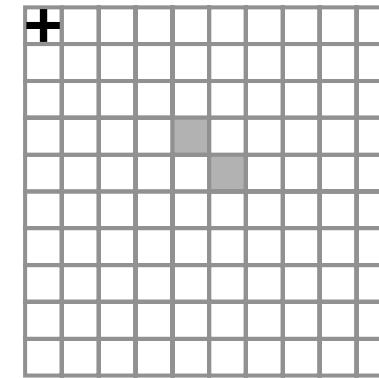
- $J$ : description of object pixels
- $K$ : description of background pixels



$A \ominus J$



$\overline{A} \ominus K$



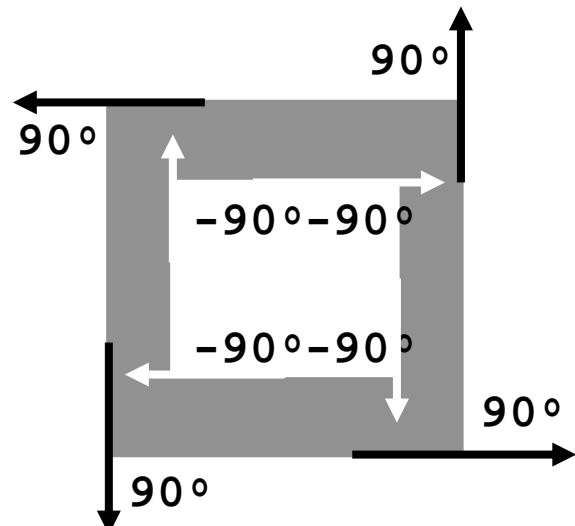
$(A \ominus J) \cap (\overline{A} \ominus K)$



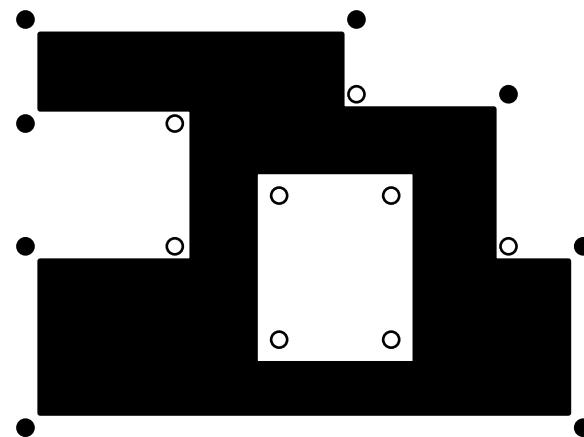
# Computation of Euler number

$$Eu = C - H$$

- Given a closed polygonal line, the sum of its angles should be  $+360^\circ$



- The Euler number is equal to the number of convex corners minus the number of concave corners, all divided by four:



$$\frac{N_{\bullet} - N_{\circ}}{4} = 0$$

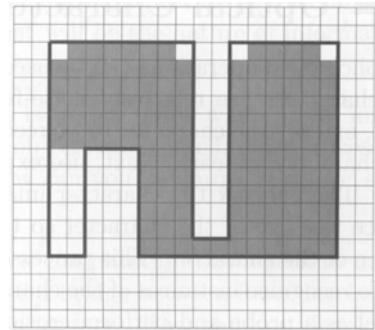
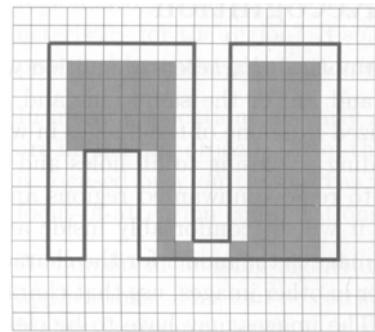
**C and H cannot be computed separately**



# 4/6. Opening and closing

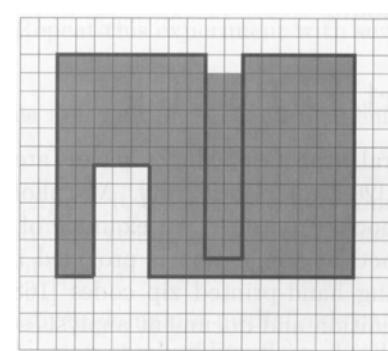
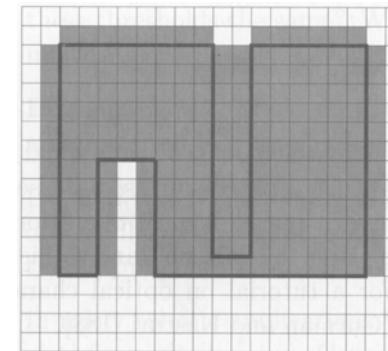
- **Opening:** erosion + dilation with the same element

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



- **Closing:** dilation + erosion with the same element

$$A \bullet K = (A \oplus K) \ominus K$$



- Eliminates all regions too small to contain the structural element

- Fills all holes and cavities smaller than the structural element

$$A \circ K \circ K = A \circ K$$

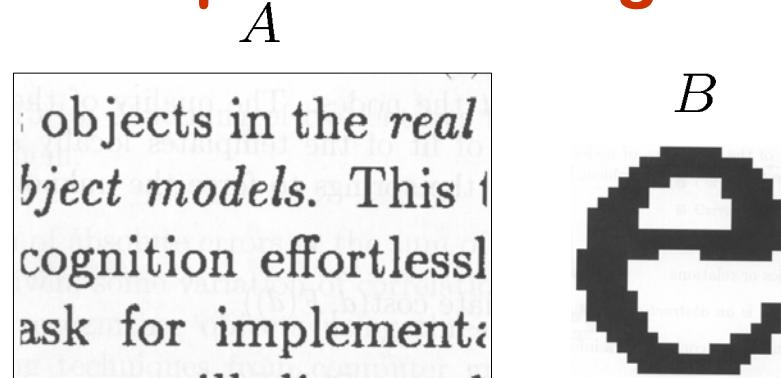
Idempotent

$$A \bullet K \bullet K = A \bullet K$$

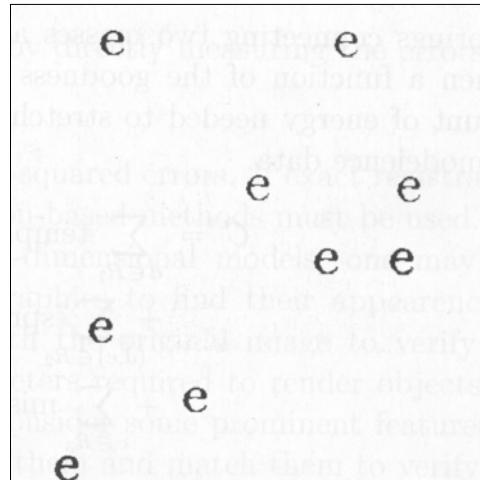


# Applications

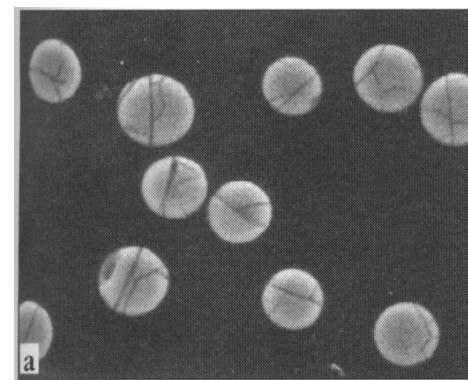
- Template matching



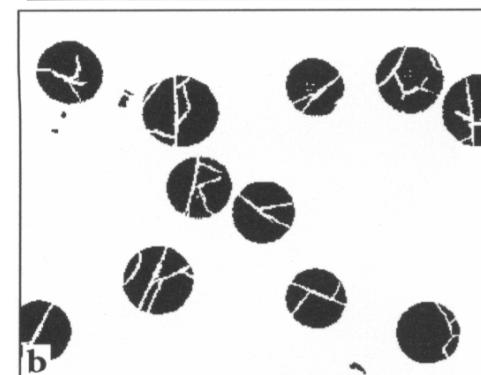
$A \circ B$



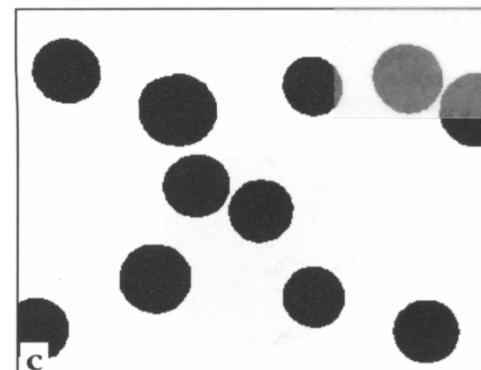
- Reconstruction



Original



Thresholded

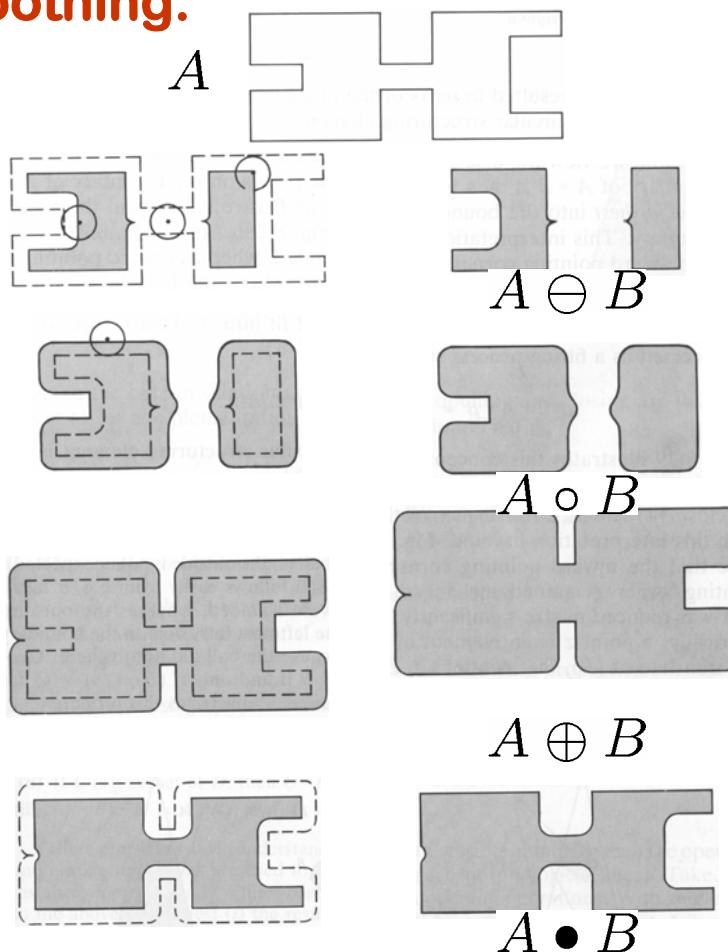


Closing



# Applications

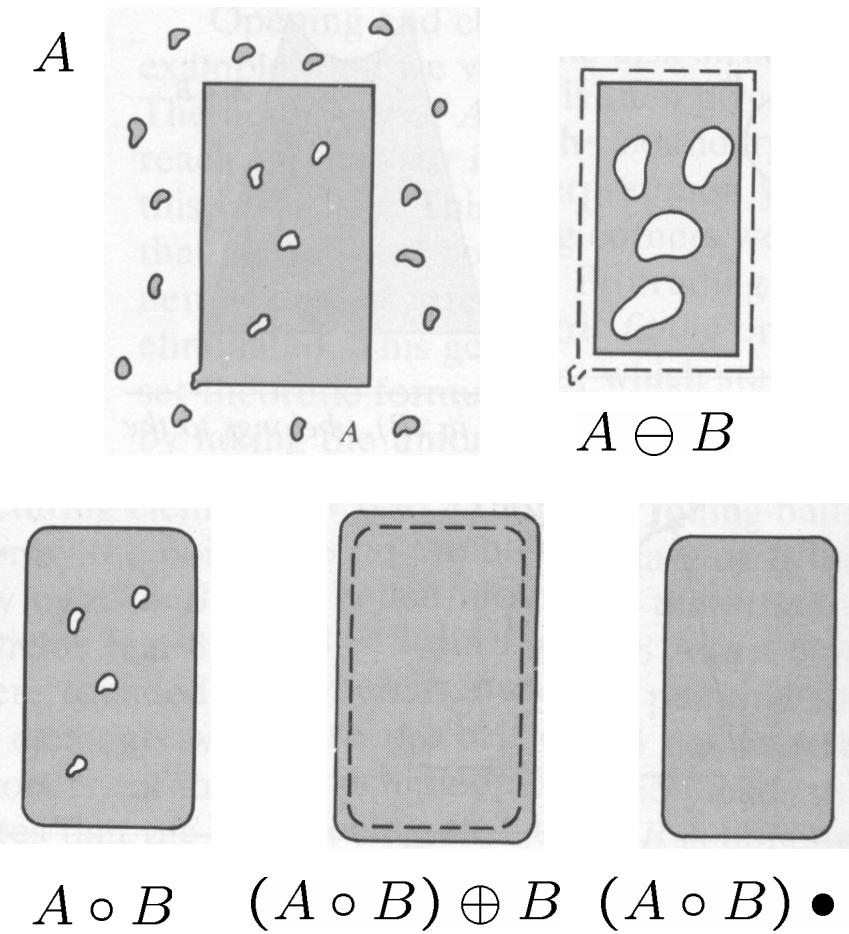
- **Smoothing:**



- **Opening:** smooth convex corners.
- **Closing:** smooth concave corners

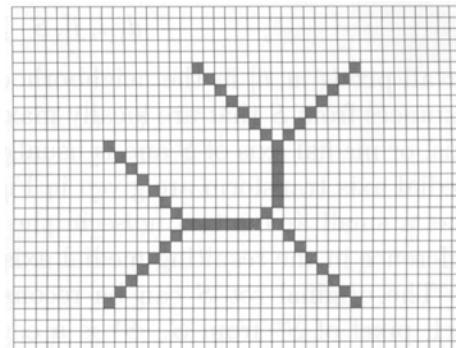
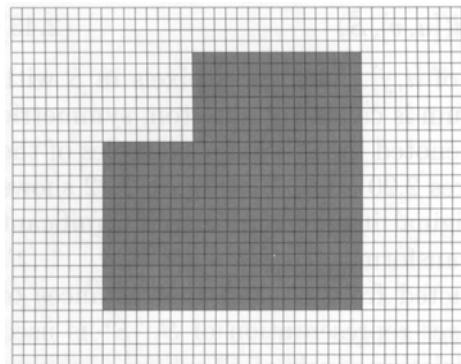
- **Morphologic filtering:**  $B$  is a disc of size  $\geq$  all noise components.

$$(A \circ B) \bullet B$$

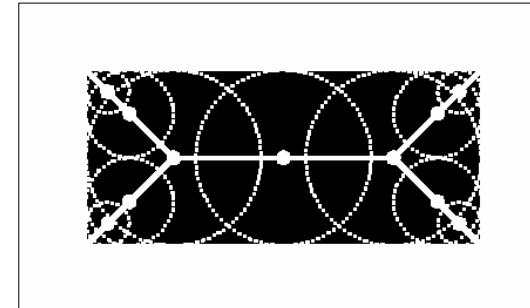


# 5/6. Skeletons

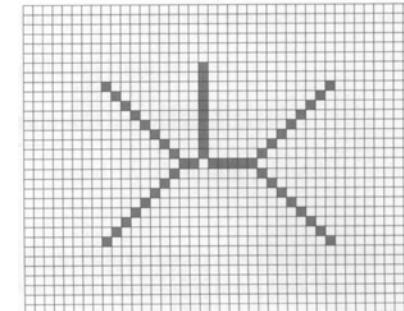
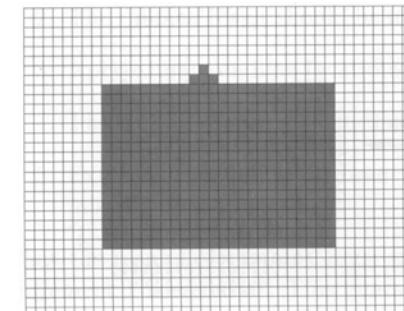
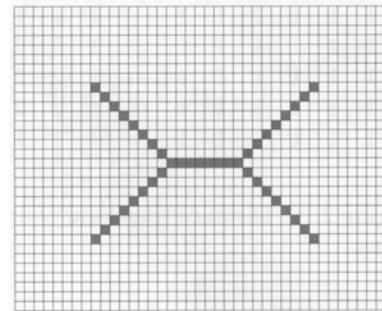
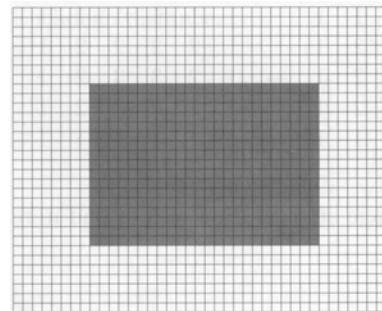
- **Skeletons, axis of symmetry  $S^*$ :** geometric place of the centers of all at least bi-tangent circles.



- $S^*$  is a compact representation of  $S$ ; it represents the *shape* of the region.



- Highly sensitive to noise.



The frontier is also a compact representation of shape.



# Thinning

$$B^1$$

$$B^2$$

$$B^3$$

$$B^4$$

$$B^5$$

$$B^6$$

$$B^7$$

$$B^8$$

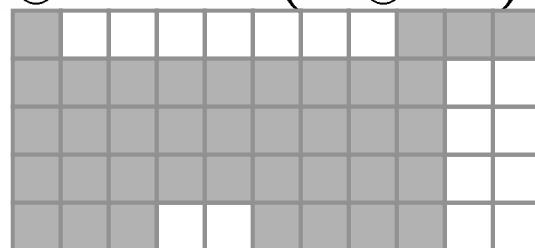
$$A \odot B^i = A - (A \otimes B^i)$$

$$A$$

$$A \otimes B^1$$

$$A \odot B^1$$

$$A \odot B^{1,2} = (A \odot B^1) \odot B^2$$

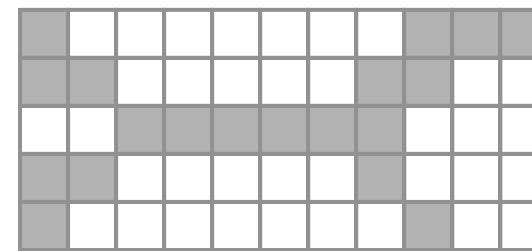


$$A \odot B^{1,2,3}$$

$$A \odot B^{1,2,3,4,5}$$

$$X_0 = A$$

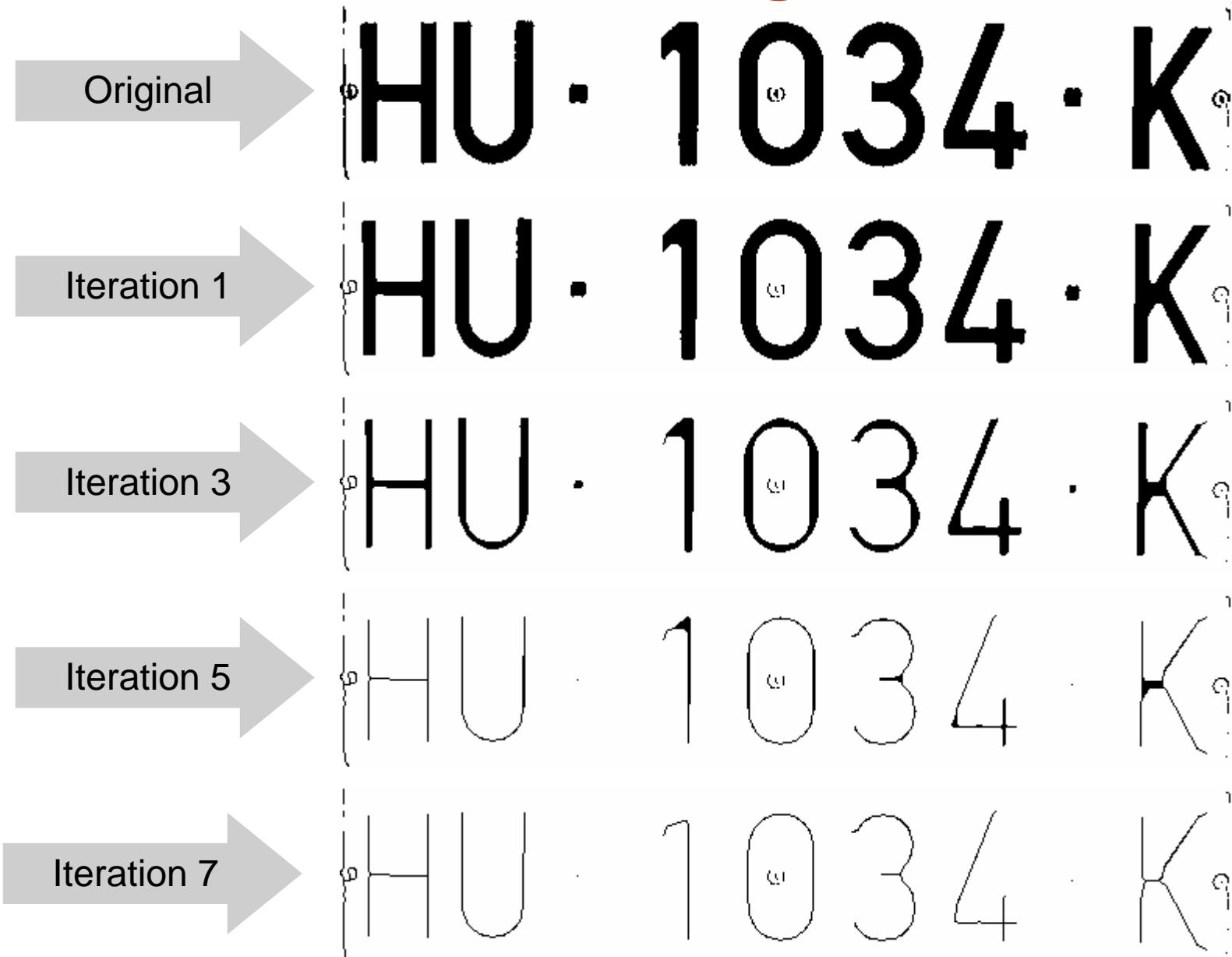
$$X_k = X_{k-1} \odot B^{\{1, \dots, n\}}$$



Final result



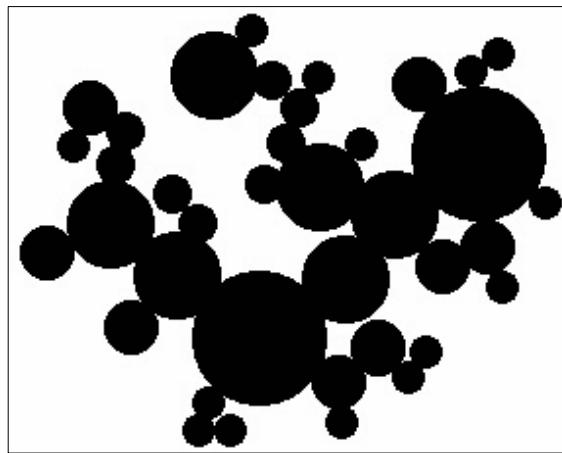
# Thinning



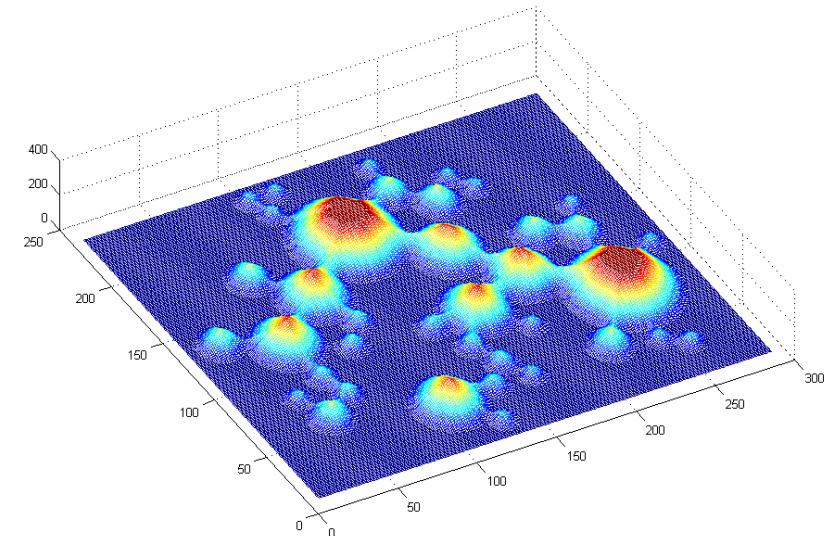
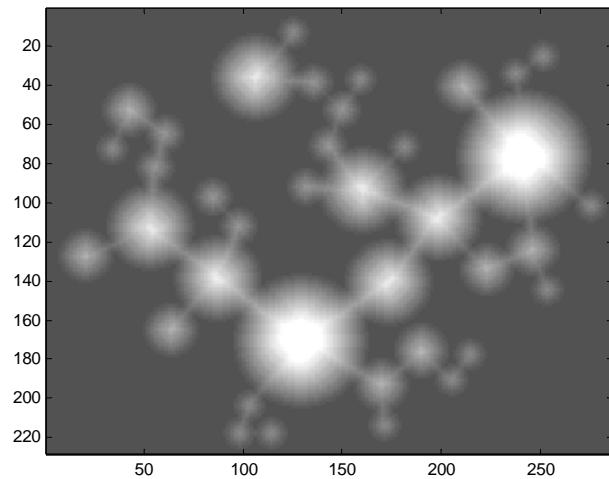
# 6/6. Euclidean distance maps (EDM)

- Image representing the smallest distance of e/pixel to the background.

*A*



*D*



*D(3d)*

There are several possible definitions for distance

# Distance measurements

- Fundamental properties:

$\forall p, q, r :$

- $d(p, q) \geq 0,$
- $d(p, q) = 0 \Leftrightarrow p = q$
- $d(p, q) = d(q, p)$
- $d(p, r) \leq d(p, q) + d(q, r)$

$$d([i_1, j_1], [i_2, j_2]) =$$

- Euclidean:

$$\sqrt{(i_1 - i_2)^2 + (j_1 - j_2)^2}$$

- Manhattan:

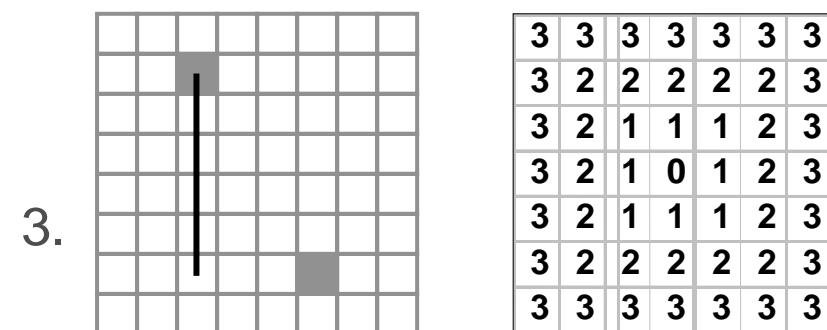
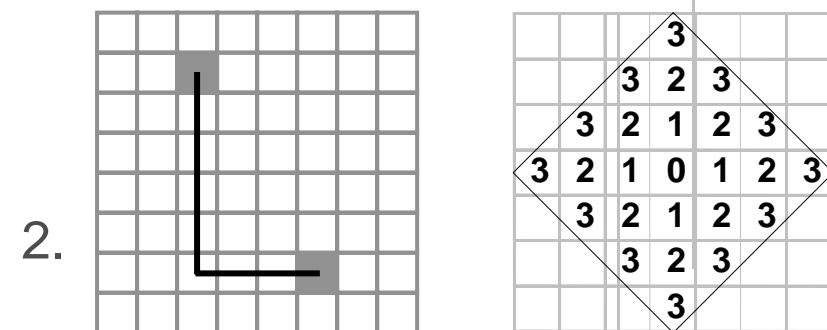
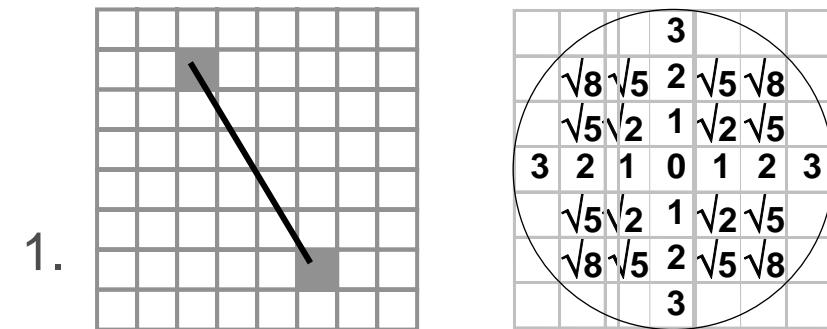
$$|i_1 - i_2| + |j_1 - j_2|$$

- Chess:

$$\max(|i_1 - i_2|, |j_1 - j_2|)$$

**Euclidean is closest to the real case;  
Most costly to compute**

**Discs:** pixels at distance  $\leq k$ .



# Obtaining distance maps

$$f^0[i, j] = B[i, j]$$

$$\begin{aligned} f^m[i, j] &= f^0[i, j] \\ &+ \min(f^{m-1}[u, v]) \end{aligned}$$

$$\forall [u, v] : d([u, v], [i, j]) = 1$$

1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1

1	1	1	1	1	1	1
1	2	2	2	2	2	1
1	2	2	2	2	2	1
1	2	2	2	2	2	1
1	2	2	2	2	2	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1

- **Iteration 0:** original image.
- **Iteration 1:** All pixels not adjacent to background change to 2.
- **Next iterations:** pixels farther from background change.
- No pixels changes when the distances to all have been computed.

1	1	1	1	1	1	1
1	2	2	2	2	2	1
1	2	3	3	2	1	
1	2	3	3	2	1	
1	2	2	2	2	2	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1

