

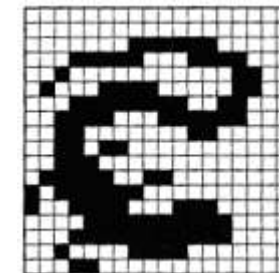
PRELUCRARI PE IMAGINI BINARE (ALB/NEGRU)

Imagine binara?

2 nuante:

alb ("0") – pixelii de fond ($I(x,y) = 255$ pt. imagini indexate cu 8 biti/pixel)

negru ("1") – pixelii apartinand obiectelor ($I(x,y) = 0$ pt. imagini indezate cu 8 biti/pixel)

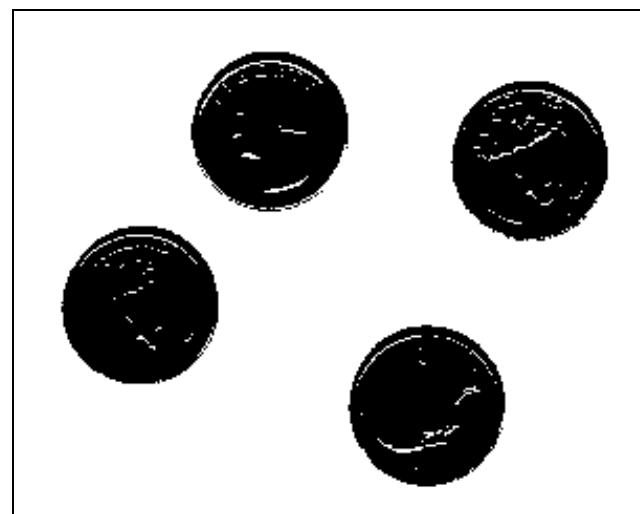


Notatii:

$$b(x, y) = \begin{cases} 1 & \text{pixel_obiect} \\ 0 & \text{pixel_fond} \end{cases} \Rightarrow \text{eticheta pixelului de la locatia (x,y)}$$



Grayscale



Alb/negru (binara)

Transformarea grayscale \Rightarrow alb/negru: **binarizare (thresholding)**

$$Dst(x, y) = \begin{cases} 0 & (\text{black} = '1') \quad , \quad \text{if } Src(x, y) < \text{threshold} \\ 255 & (\text{white} = '0') \quad , \quad \text{if } Src(x, y) \geq \text{threshold} \end{cases}$$

OPERATII MORFOLOGICE

Morfologie := [moprphos = forma] forma si structura organismelor vii

Morfologie matematica \Rightarrow unelte pentru modificarea formei sau extragere de componente, reprezentarea si descrierea formei unei regiuni / obiect (contur, skeleton).

Teorie mulțimilor (set-urilor) \Rightarrow Limbajul folosit in morfologia matematica

Fie A o **mulțime** din Z^2 . Dacă $a = (a_1, a_2)$ este un element din A :

$$a \in A.$$

Similar, daca a **nu** este un element din A :

$$a \notin A.$$

Mulțimea fără nici un element: \emptyset .

Notăție: $\{ \dots \}$

Elementele mulțimilor pe care le consideram: pixeli $b(x,y)$ ai obiectelor imagini binare

Relații / operații pe mulțimi

1. Incluziunea

$$A \subseteq B$$

2. Reuniunea

$$C = A \cup B$$

3. Intersecția

$$D = A \cap B$$

4. Mulțimi disjuncte (mutual exclusive)

$$A \cap B = \emptyset.$$

5. Complementul

$$A^c = \{w \mid w \notin A\}$$

6. Diferența

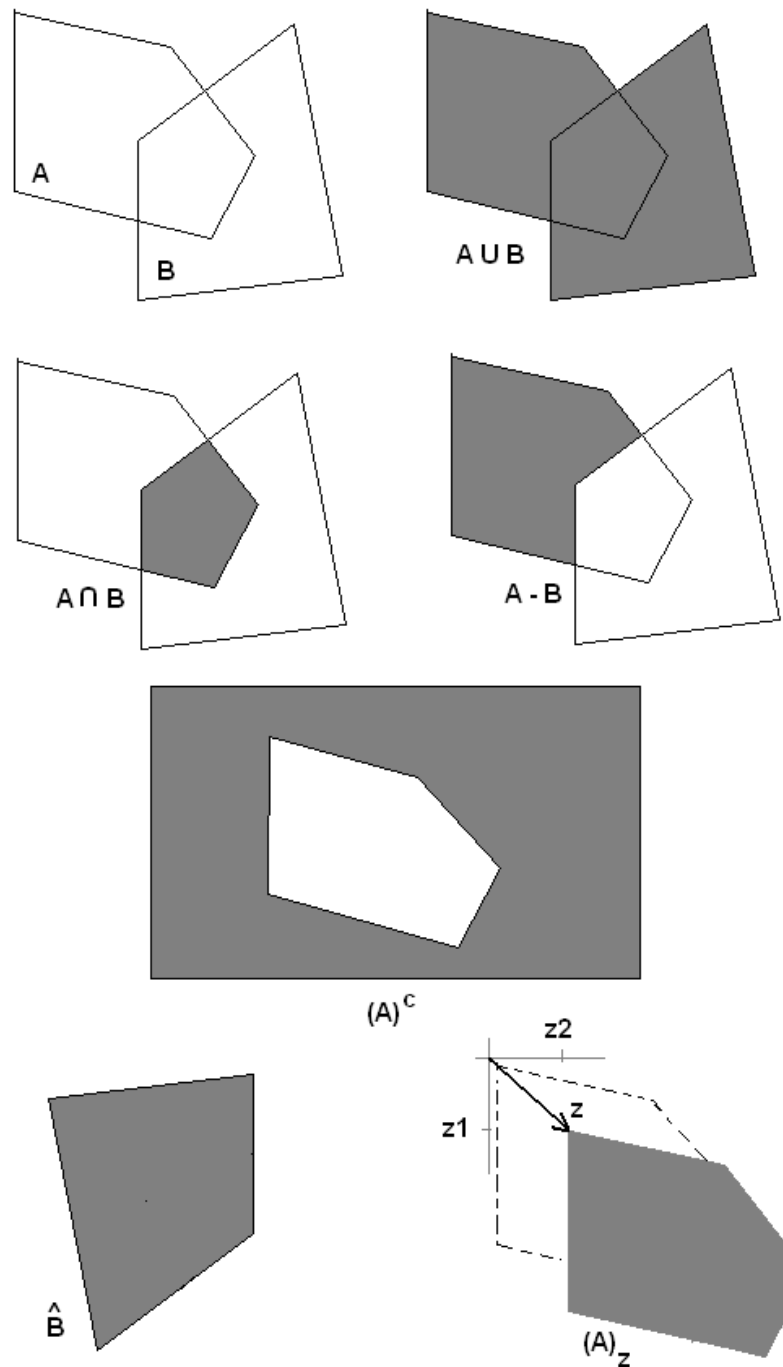
$$A - B = \{w \mid w \in A, w \notin B\} = A \cap B^c$$

7. Reflexia (flip orizontal + vertical)

$$\hat{B} = \{w \mid w = -b, \text{ for } b \in B\}$$

8. Translația (setului A cu $z = (z_1, z_2)$)

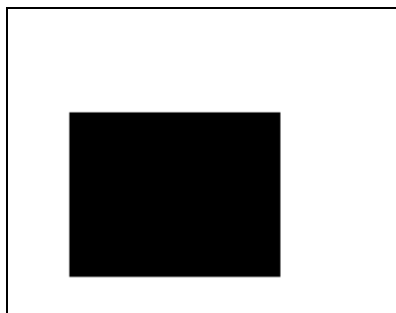
$$(A)_z = \{c \mid c = a + z, \text{ for } a \in A\}$$



Operații logice / aritmetice aplicate pe imagini binare

- **Unare:** imagine **op** operand_scalar
- **Binare:** imagine1 **op** imagine2
- Realizate la nivel de pixel

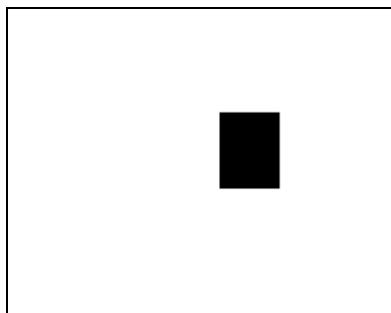
Operații logice: AND, OR, and NOT (COMPLEMENT) + orice alte combinații



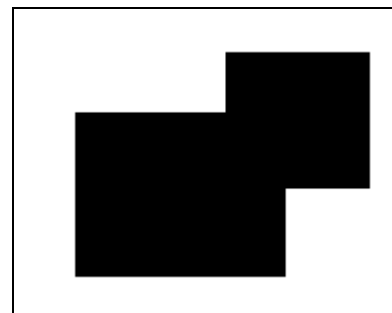
A



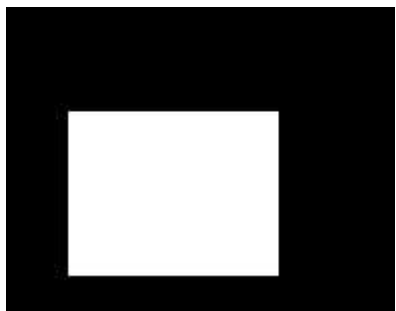
B



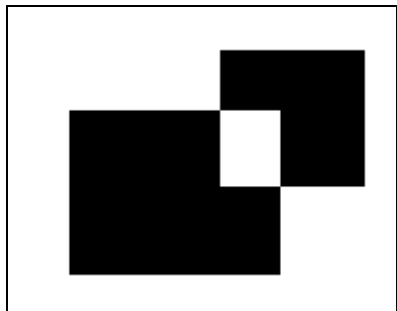
A and B



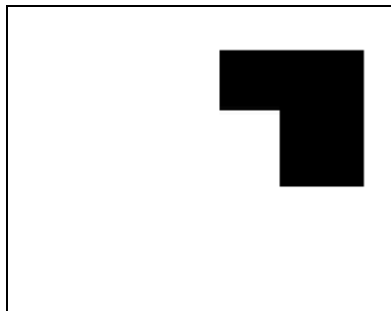
A or B



not (A) = A^C



A xor B



not(A) and B = B-A

DILATAREA SI EROZIUNEA

Dilatarea si eroziunea - cele doua primitive de baza ale operatiilor morfologice!

$A, B \subset Z^2$

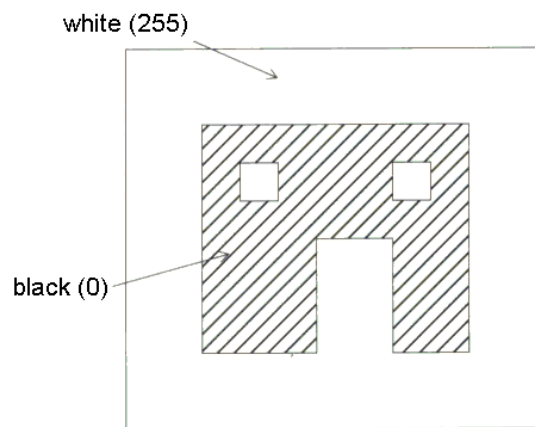
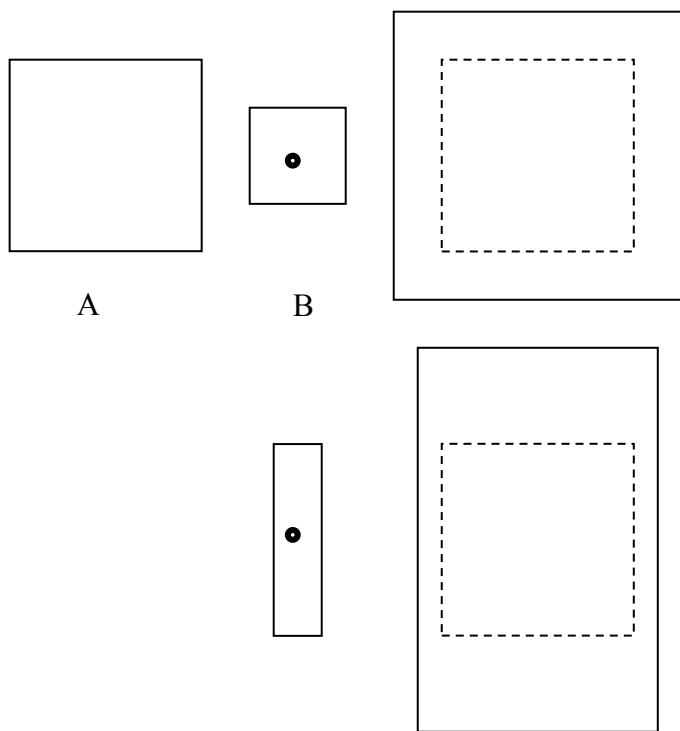
DILATAREA

Dilatarea A cu B

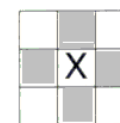
$$A \oplus B = \{z | (\hat{B})_z \cap A \neq \emptyset\} \quad \text{sau} \quad A \oplus B = \{z | [(\hat{B})_z \cap A] \subseteq A\}$$

B – element structural

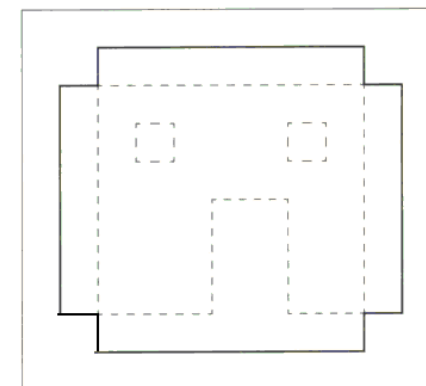
:



a. Original image.



b. Structural element; $x = \text{origin}$.



c. Image after dilation; original in dashes.

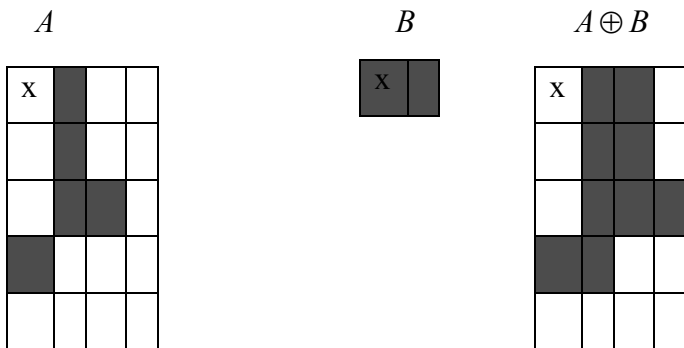
Alta definiție pt. dilatare

$a=(a_1, a_2, \dots, a_N)$ și $b=(b_1, b_2, \dots, b_M)$.

$$A \oplus B = \{z \in Z^2 \mid z = a + b \text{ pt. un } a \in A \text{ și } b \in B\}$$

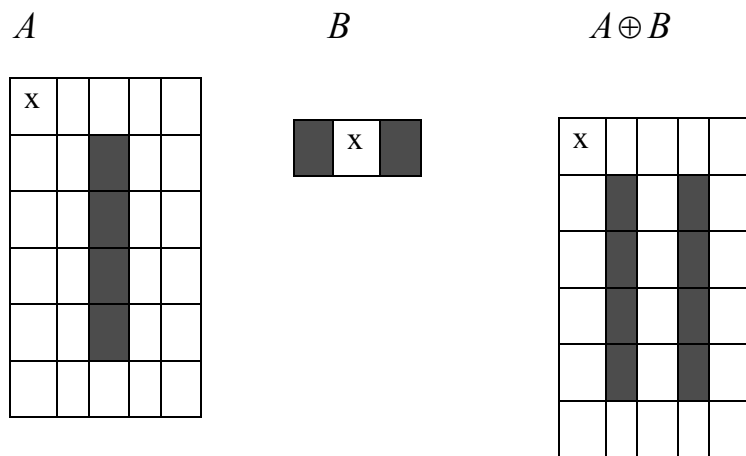
Ex: $A = \{(0,1), (1,1), (2,1), (2,2), (3,0)\}$;

$B = \{(0,0), (0,1)\}$



$$A \oplus B = \{(0,1), (1,1), (2,1), (2,2), (3,0), (0,2), (1,2), (2,2), (2,3), (3,1)\}$$

Ex: $A = \{(1,2), (2,2), (3,2), (4,2)\}$; $B = \{(0,-1), (0,1)\}$

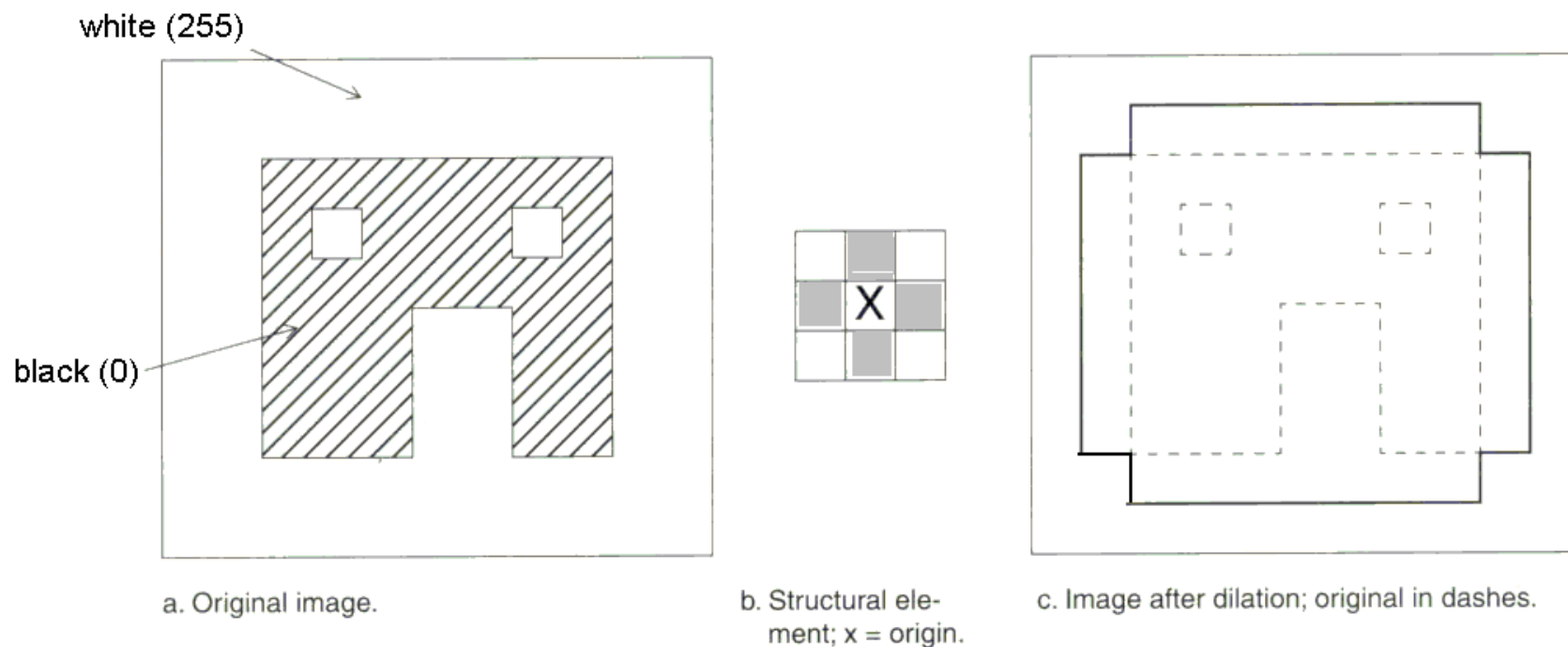


$$A \oplus B = \{(1,1), (2,1), (3,1), (4,1), (1,3), (2,3), (3,3), (4,3)\}$$

Modalitate practica de aplicare (laborator)

Se aplica elementul structural peste imaginea sursa

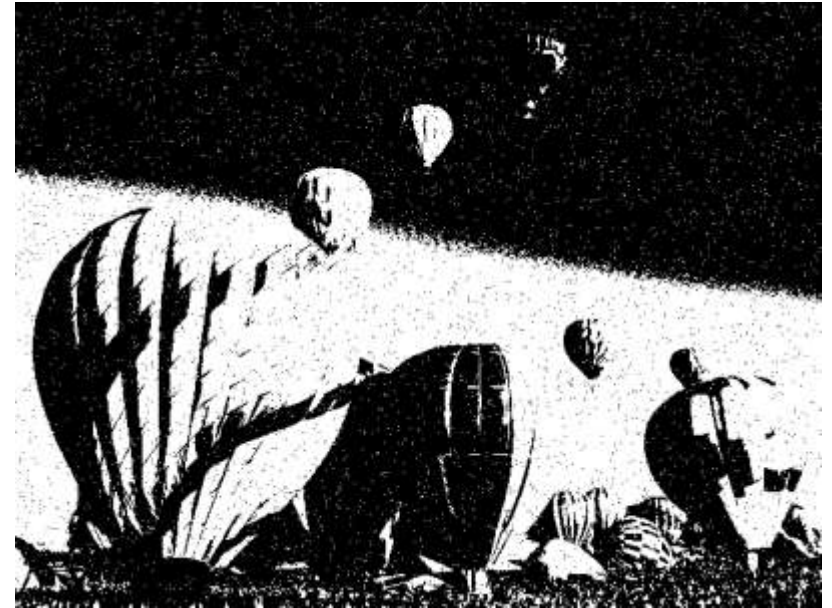
1. Daca originea elementului structural coincide cu un pixel ,0' (alb) din imaginea sursa nu fac nimic (trec la pixelul următor)
2. Daca originea elementului structural coincide cu un pixel ,1' negru (din imaginea sursa) realizez OR logic intre pixelii corespunzatori elementului structural si pixelii corespunzatori din imaginea sursa (asemănător convoluției – operația este OR logic si nu înmulțire aritmetica !!!)



Aplicații: Umplere goluri, unire legături slabe între obiecte (istmuri)



binarizare \Rightarrow



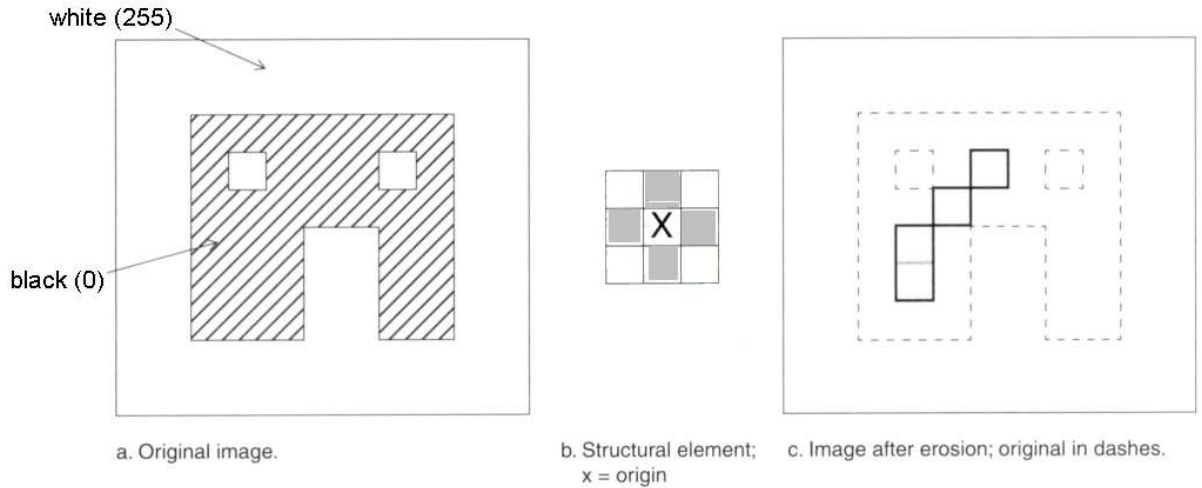
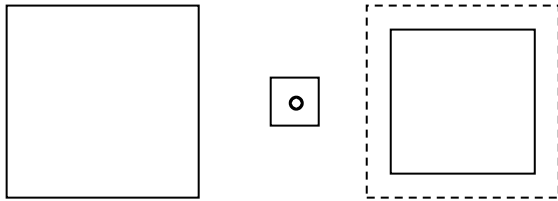
dilatare (obiect = negru) \Rightarrow



EROZIUNEA

Eroziunea A cu B

$$A \ominus B = \{z | (\hat{B})_z \subseteq A\}$$



Alta definiție pt. eroziune

Eroziunea \leftrightarrow dilatare (duale / complementare)

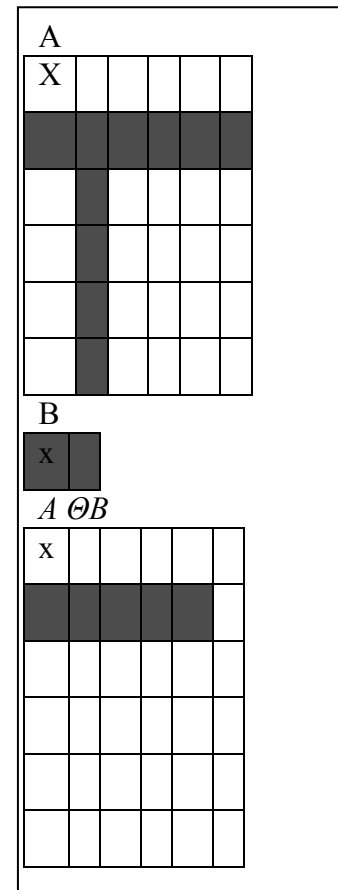
$a = (a_1, a_2, \dots, a_N)$ și $b = (b_1, b_2, \dots, b_N)$.

$$A \ominus B = \{x \in Z^2 | x + b \in A \text{ pt. orice } b \in B\}$$

Ex1: $A = \{(1,0), (1,1), (1,2), (1,3), (1,4), (1,5), (2,1), (3,1), (4,1), (5,1)\}$

$B = \{(0,0), (0,1)\}$

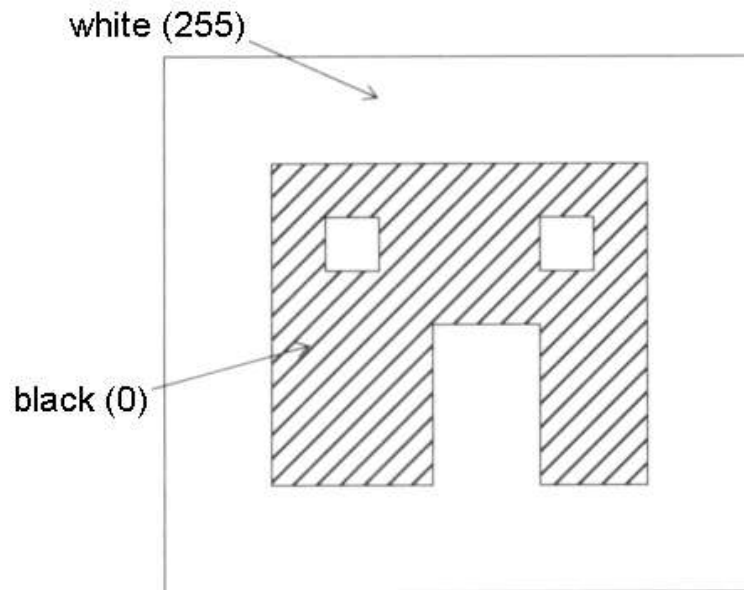
$A \ominus B = \{(1,0), (1,1), (1,2), (1,3), (1,4)\}$



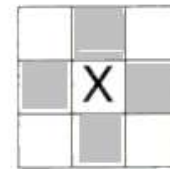
Modalitate practica de aplicare (laborator)

Se aplica elementul structural peste imaginea sursa

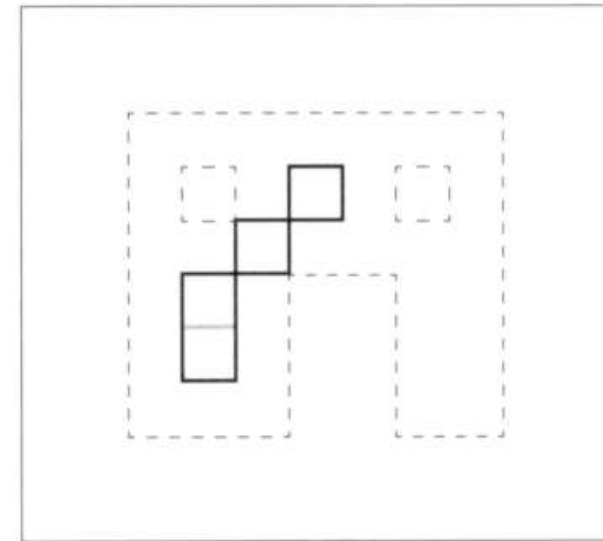
1. Daca originea elementului structural (B) coincide cu un pixel 0 din imaginea sursa nu fac nimic (trec la pixelul următor)
2. Daca originea elementului structural (B) coincide cu un pixel 1 din imaginea sursa (pixel curent) si oricare (exista cel puțin unul) dintre pixelii de 1 din elementul structural se extinde in afara obiectului (A) atunci schimb valoarea pixelului curent din imaginea sursa in 0.



a. Original image.

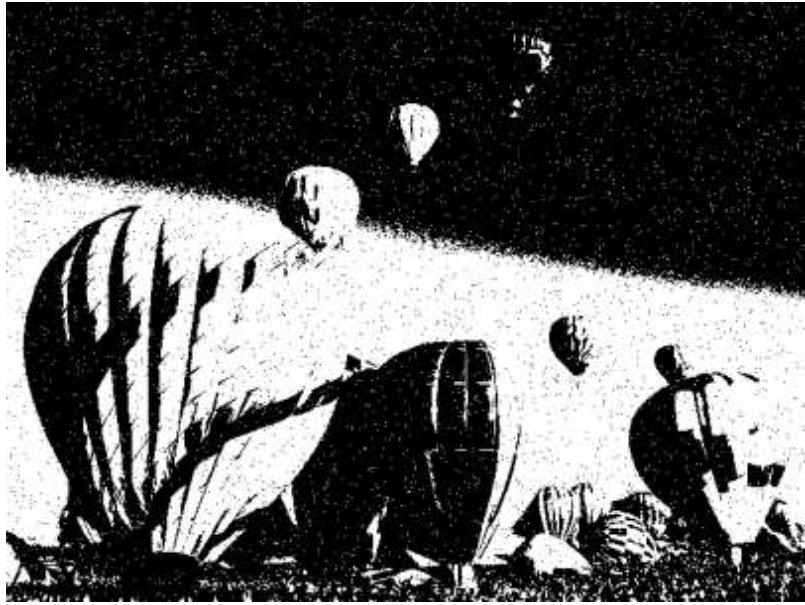


b. Structural element;
x = origin



c. Image after erosion; original in dashes.

Aplicații: Eliminarea obiectelor mici (zgomote), spargerea legăturilor slabe între obiecte (istmuri)



Imagine originală

eroziune (negru) ⇒

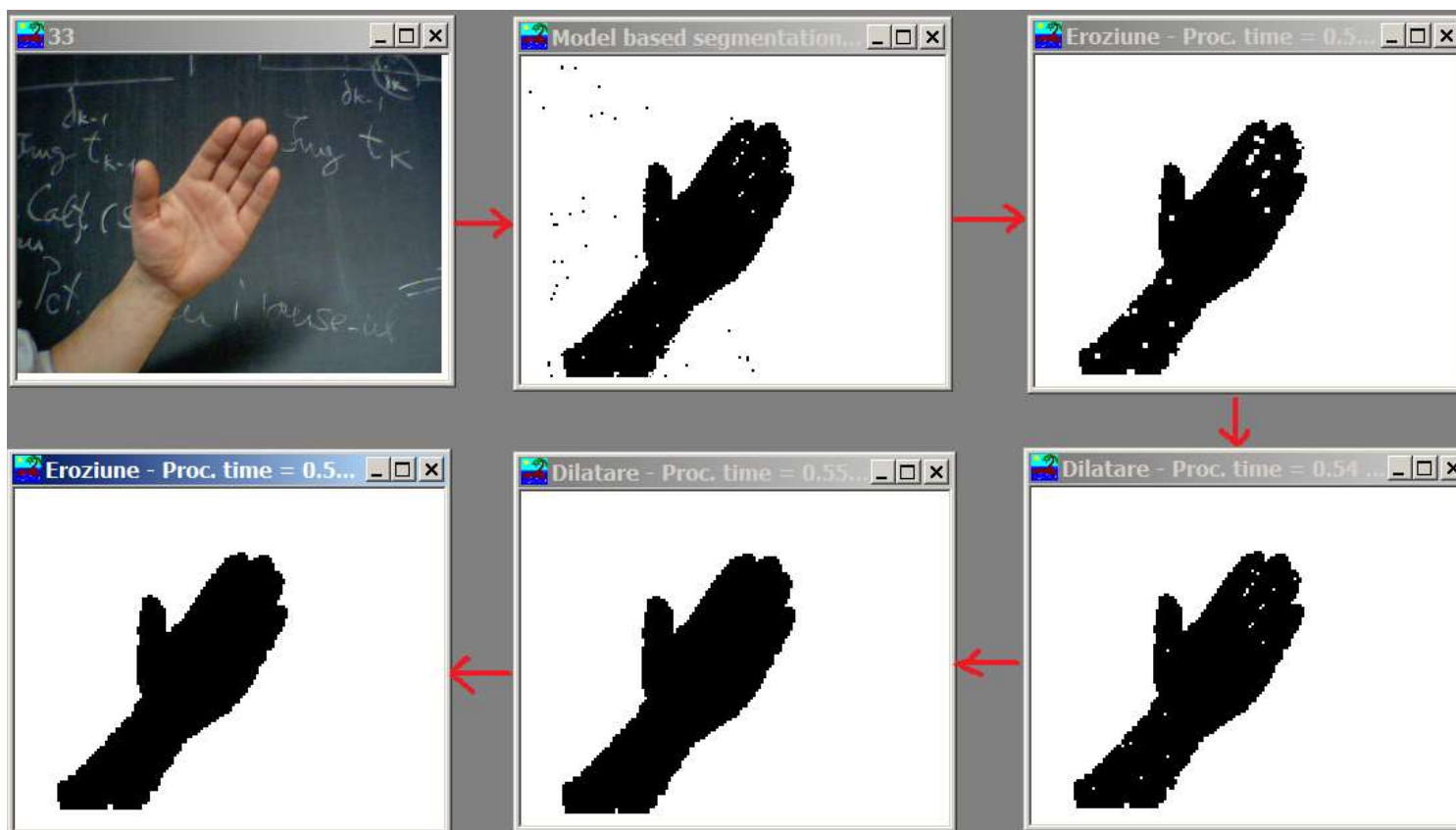


Imagine rezultat după eroziune

Exemplu de aplicare tipica a operatiilor morfologice in segmentarea imaginilor:

Dorim sa segmentam mana care apare in imaginea de mai jos. Realizam o clasificare a pixelilor bazata pe un model de culoare a pielii (invatat anterior) \Rightarrow o imagine cu „defecte/zgomote” (erori de clasificare a pixelilor)”: pixeli albi in zona obiect si pixeli negri in zona de fond.

- Aplicam o eroziune pt. a scapa de pixelii de „zgomot” din zona de fond (dreapta sus)
- Aplicam 2x dilatarii pt. a scapa de „golurile” din zona obiect (dreapta jos)
- Aplicam o eroziune pt. a reduce obiectul la aria initiala / reala (stanga jos)



DESCHIDERE SI INCHIDERE

Deschidere

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

Aplicații: netezire contur, eliminare obiecte mici, spargere legaturi slabe intre obiecte (istmuri)

Închidere

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

Aplicații: netezire contur, eliminare goluri mici din obiecte, unire legaturi slabe intre obiecte (istmuri)

Deschidere



Imagine rezultat după deschidere (obiect = negru)

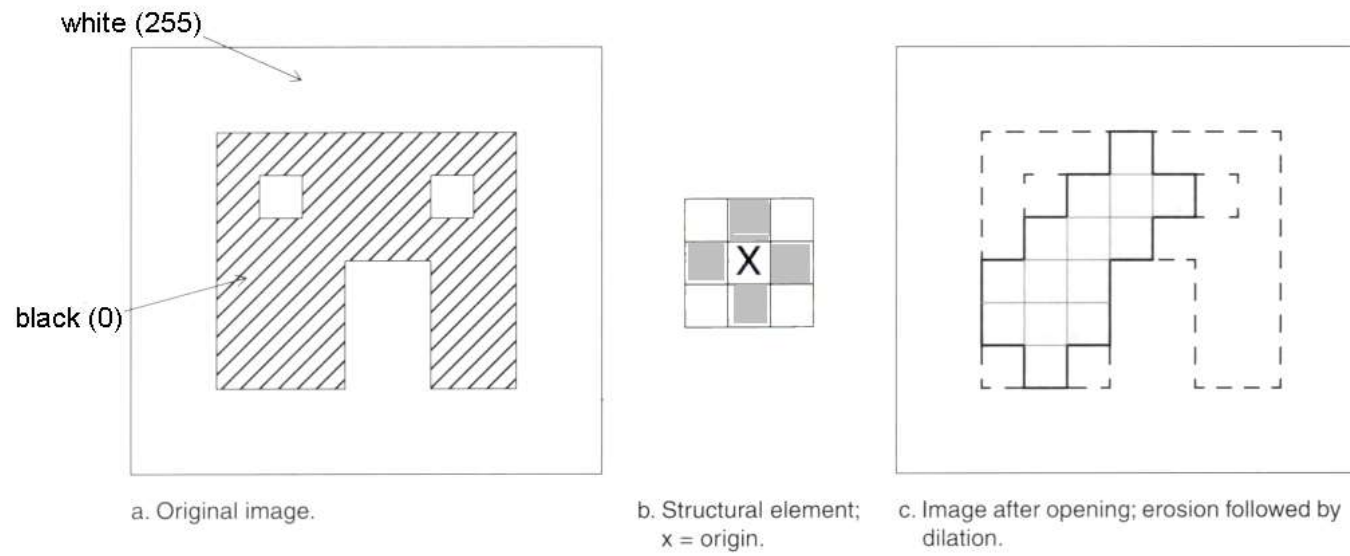
Închidere



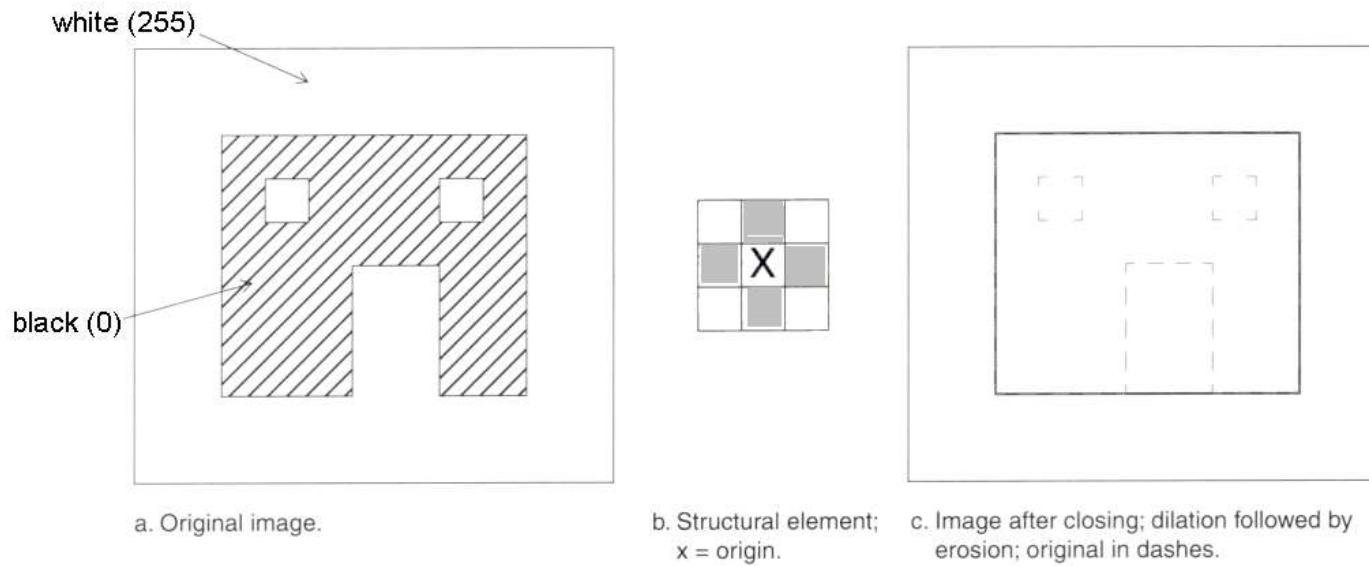
Imagine rezultat după închidere (obiect = negru)

Exemple:

Deschidere



Închidere



Proprietati (ale operatorilor morfologici)

1. $A \oplus B = B \oplus A$

2. $(A \ominus B)^C = A^C \oplus B$

3. $A \circ B \subseteq A$

4. $C \subseteq D \Rightarrow C \circ B \subseteq D \circ B$

5. $(A \circ B) \circ B = A \circ B$ (IDEMPOTENȚĂ)

6. $A \subseteq A \bullet B$

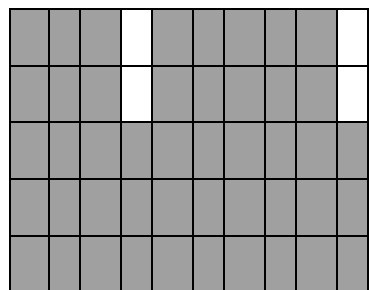
7. $C \subseteq D \Rightarrow C \bullet B \subseteq D \bullet B$

8. $(A \bullet B) \bullet B = A \bullet B$ (IDEMPOTENȚĂ)

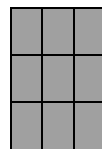
Aplicatii ale operatiilor morfologice de baza

EXTRAGERE CONTUR

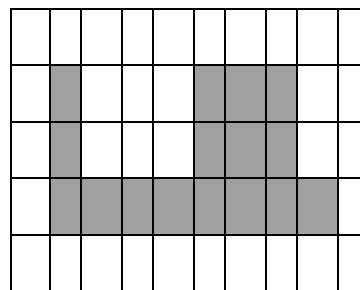
$$\beta^i(A) = A - (A \ominus B) \text{ (contur interior)}$$



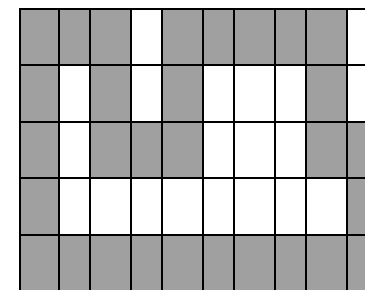
A



B



$A \ominus B$



$\beta^i(A)$

$$\beta^e(A) = (A \oplus B) - A \text{ (contur exterior)}$$

UMPLERE REGIUNI

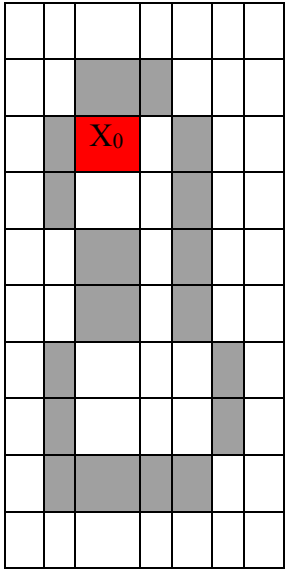
- p in interiorul conturului A (care se dorește a fi umplut in interior cu „1”)

1. $X_0 = p$, ($p = '1'$)

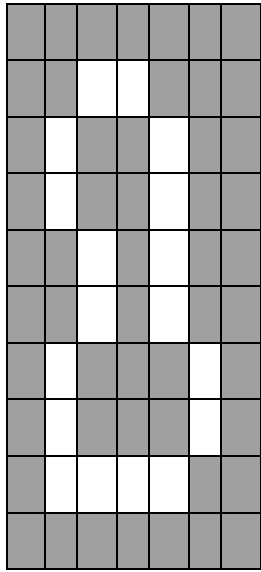
2. $X_k = (X_{k-1} \oplus B) \cap A^c$ $k=1,2,3$,

3. *Daca* $X_k = X_{k-1} \Rightarrow$ stop. Altfel repeta 2.

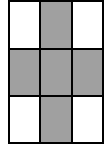
Obiectul final (umplut): $A \cup X_k$



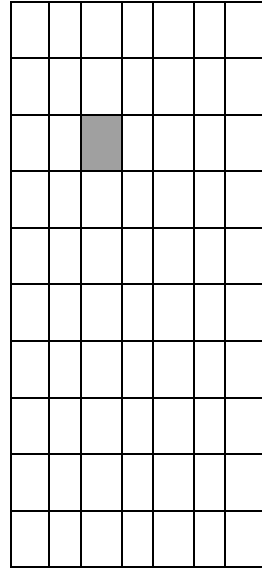
A



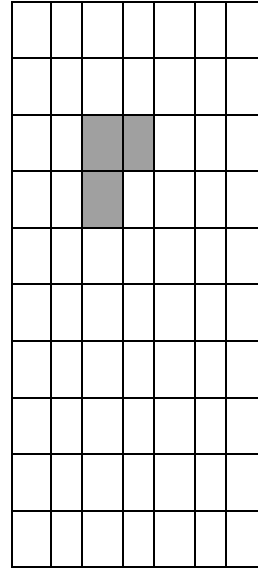
A^c



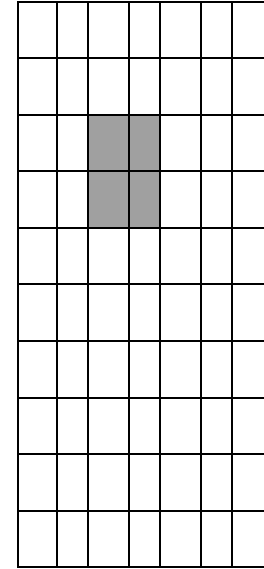
B



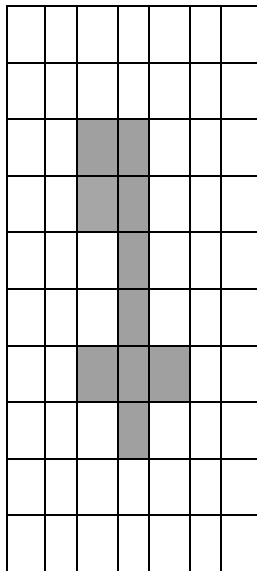
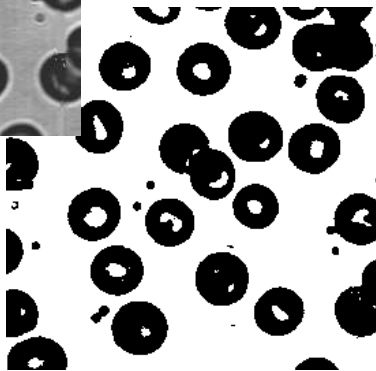
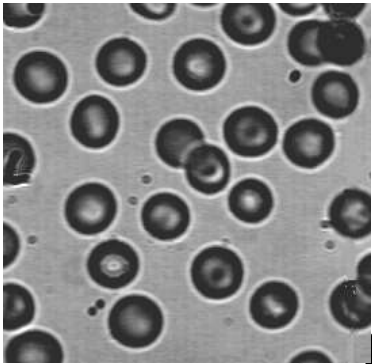
X₀



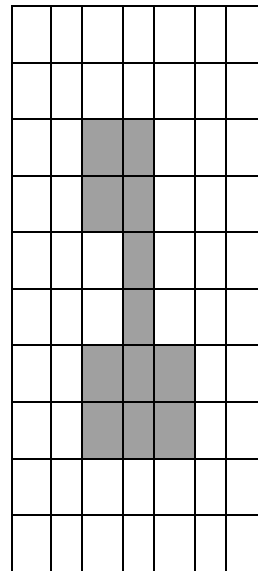
X₁



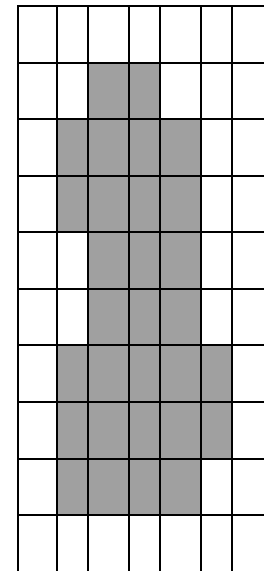
X₂



X₆



X₇



X₇ ∪ A

EXTRAGERE COMPONENTE CONEXE (ETICHETARE)

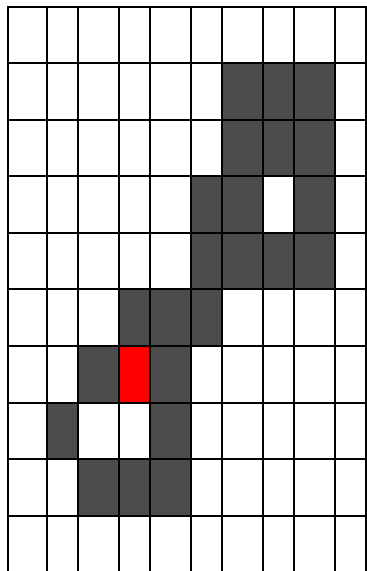
$A = \{ Y_1, Y_2, \dots, Y_n \}$, Y_i – componente conexe

$Y \subseteq A$

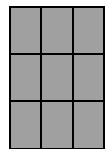
1. $p \in Y$. $X_0 = p$

2. $X_k = (X_{k-1} \oplus B) \cap A$ $k=1,2,3,\dots$

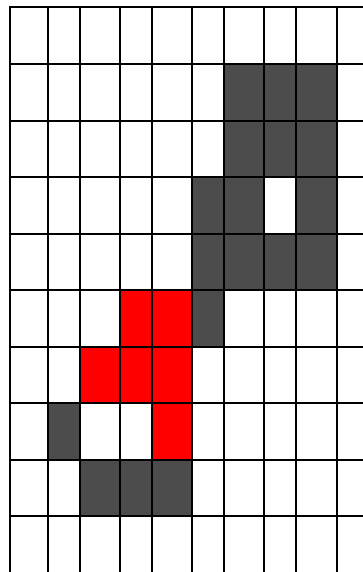
3. Dc. $X_k = X_{k-1} \Rightarrow$ stop ($Y = X_k$). Altfel repeta 2.



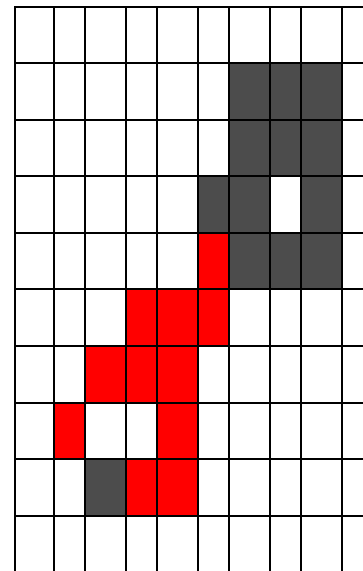
Y, p



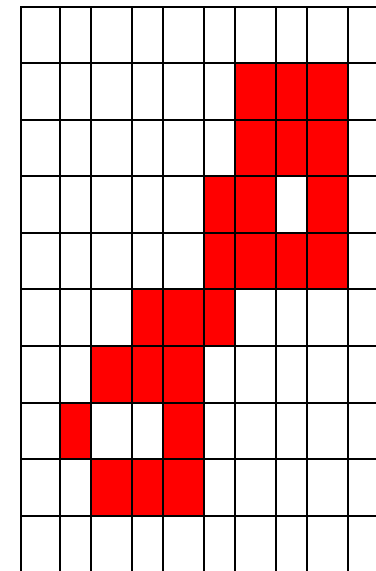
B



X_1



X_2



$X_3 = Y$

TRANSFORMATA „HIT-AND-MISS”

Se folosește la selecția unor seturi de pixeli cu proprietăți geometrice specifice: colturi, puncte izolate, puncte de contur, *template matching* (obiecte cu o anumită formă), subțiere, îngroșare etc.

Transformat hit & miss a unui set A cu elementele structurale (J,K):

$$A \otimes (J,K) = (A \ominus J) \cap (A^c \ominus K).$$

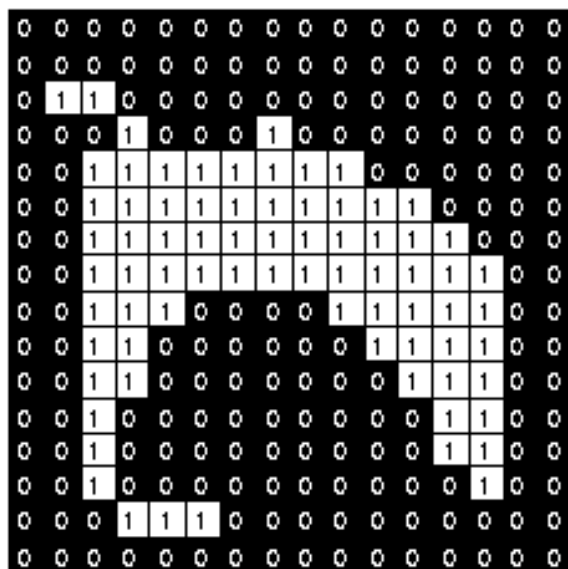
Ex. 1: Thinning (subțiere)

0	0	0
	1	
1	1	1

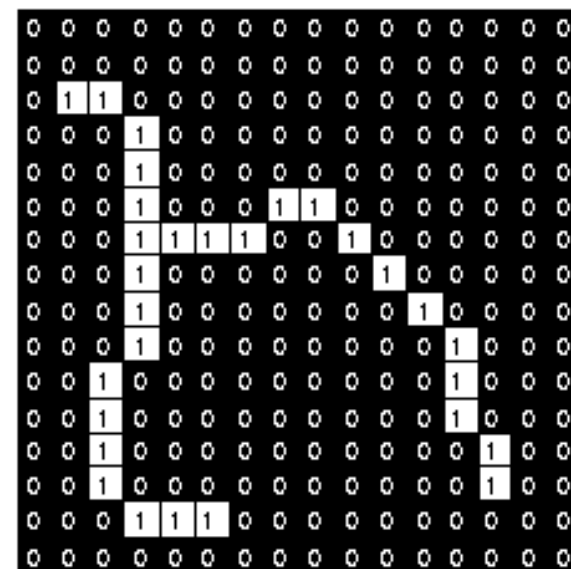
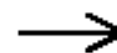
J

	0	0
1	1	0
	1	

K

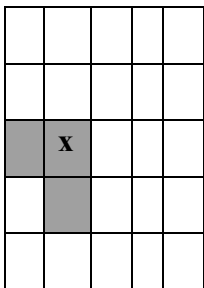


A

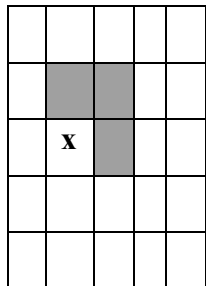


$A \otimes (J,K)$

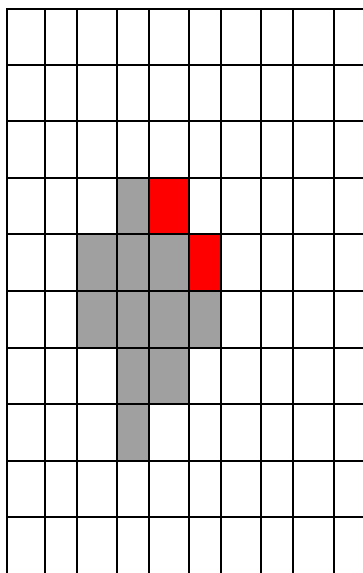
Ex.2: Detecție colțuri



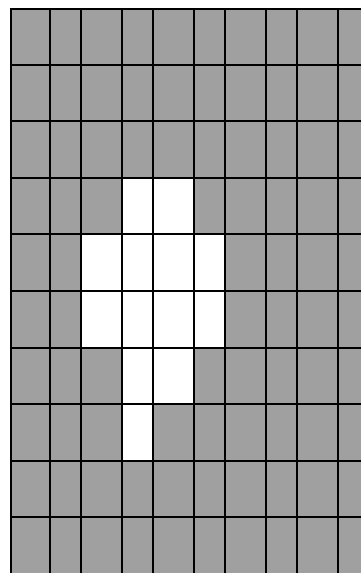
J



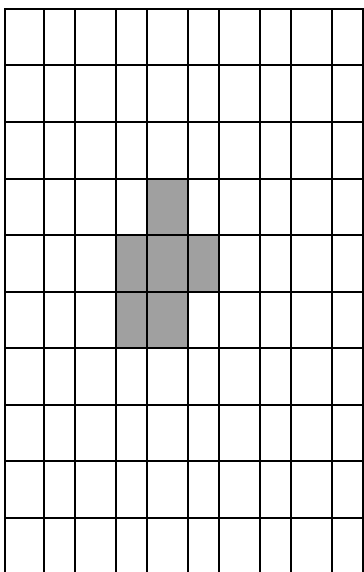
K



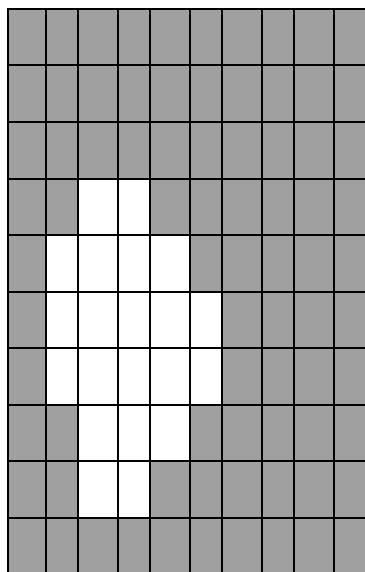
A



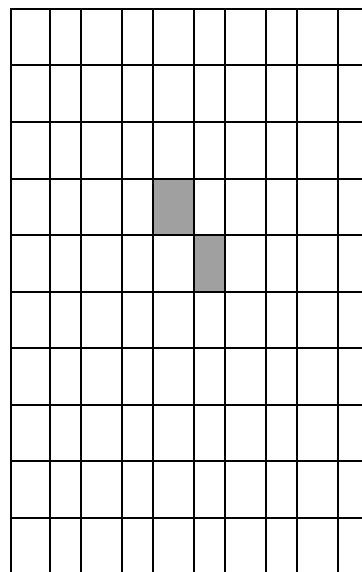
A^c



(A ⊖ J)

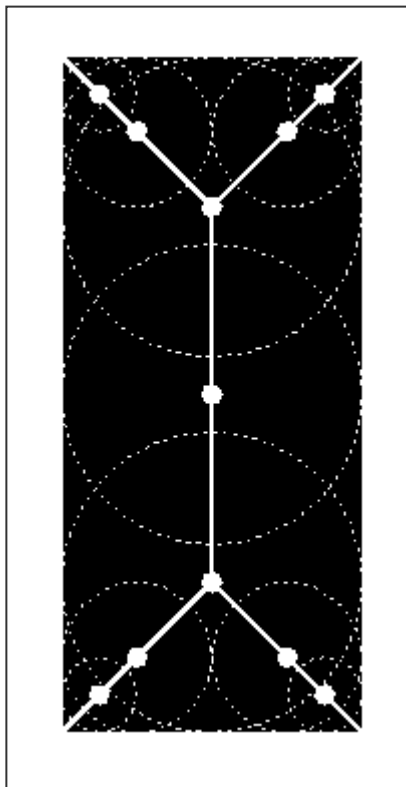


(A^c ⊖ K)



A ⊗ (J,K) = (A ⊖ J) ∩ (A^c ⊖ K)

SKELETIZAREA



„Skeletonul” setului A:

$$S(A) = \bigcup_{k=0}^K S_k(A)$$

$$S_k(A) = (A \ominus kB) - (A \ominus kB) \circ B$$

$$A \ominus kB = (\dots(A \ominus B) \ominus B) \dots \ominus B$$

$$K = \max\{k \mid (A \ominus kB) \neq \Phi\}$$

Reconstrucția setului A:

$$A = \bigcup_{k=0}^K (S_k(A) \oplus kB)$$

K – trebuie sa fie cunoscut

Referințe:

[1] Robert M. Haralick, Linda G. Shapiro, *Computer and Robot Vision*, Addison-Wesley Publishing Company, 1993

[2] Rafael C. Gonzalez, *Digital Image Processing*, Prentice-Hall, 2002