

# Morphological Image Processing

# Outline

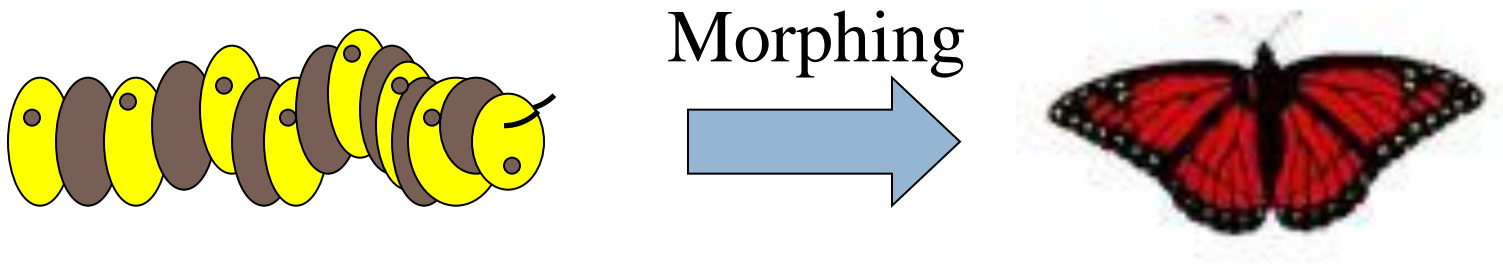
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- Introduction
- Erosion and Dilation
- Opening and Closing
- Boundary Extraction
- Hole Filling
- Hit-or-Miss Transformation

# Introduction

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- Morphological operations come from the word “morphing” in Biology which means “**changing a shape**”.

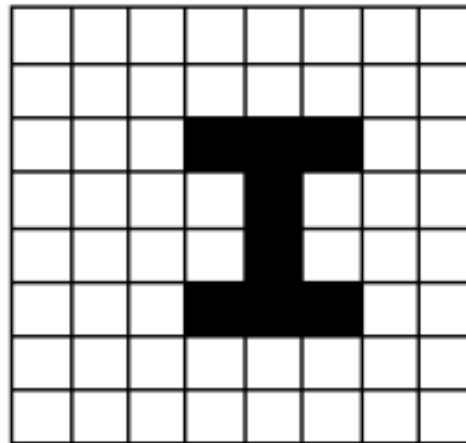


- In the context of image processing , it refers to a set of mathematical tools that can be used to
  - ▣ Extract useful description and representation of regions in images (boundaries, skeletons, convex hull)
  - ▣ Remove imperfections introduced during segmentation (thinning, regions filling)

# Introduction

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- A binary image can be considered as a set by considering “black” pixels (with image value “1”) as elements in the set and “white” pixels (with value “0”) as outside the set, or vice versa.

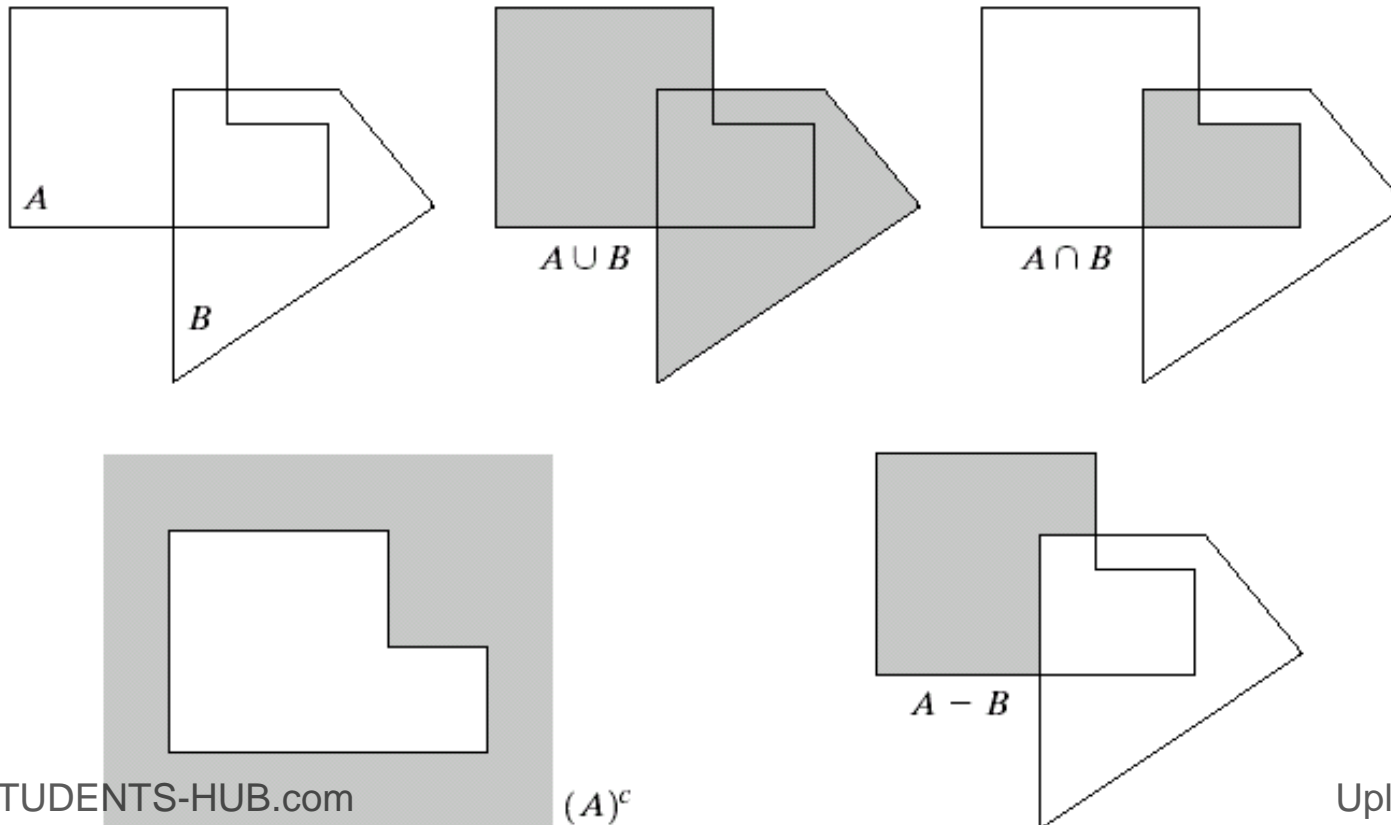


- Morphological filters are essentially **set operations**

# Basic Set Operations

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- Let  $x, y, z, \dots$  represent locations of 2D pixels, e.g.  $x = (x_1, x_2)$ ,  $S$  denote the complete set of all pixels in an image, let  $A, B, \dots$  represent subsets of  $S$ 
  - Each set may represent one object. Each pixel  $(x, y)$  has its status: **belong to a set** or **not belong to a set**.



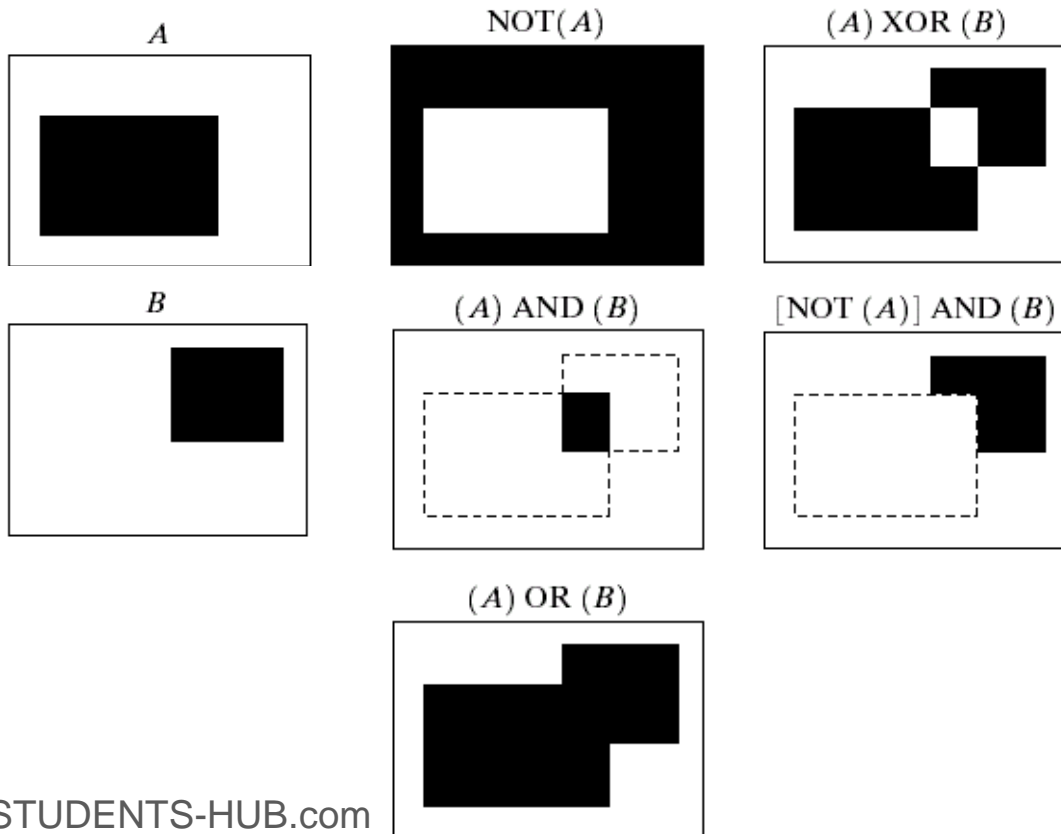
a	b	c
d	e	

**FIGURE 9.1**

(a) Two sets  $A$  and  $B$ . (b) The union of  $A$  and  $B$ . (c) The intersection of  $A$  and  $B$ . (d) The complement of  $A$ . (e) The difference between  $A$  and  $B$ .

# Basic Set Operations

$p$	$q$	$p \text{ AND } q$ (also $p \cdot q$ )	$p \text{ OR } q$ (also $p + q$ )	$\text{NOT } (p)$ (also $\bar{p}$ )
0	0	0	0	1
0	1	0	1	1
1	0	0	1	0
1	1	1	1	0



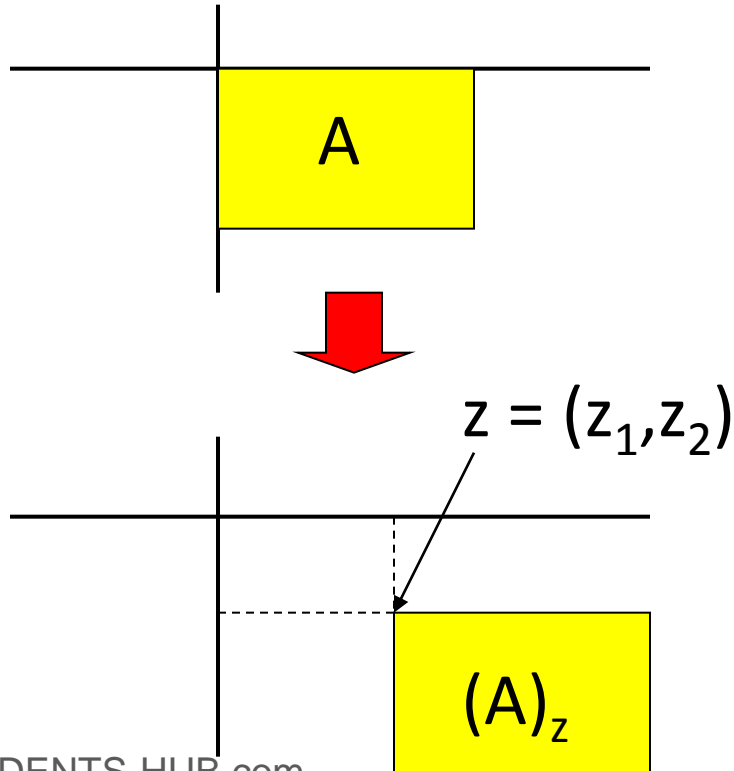
**\*For binary images only**

# Basic Set Operations

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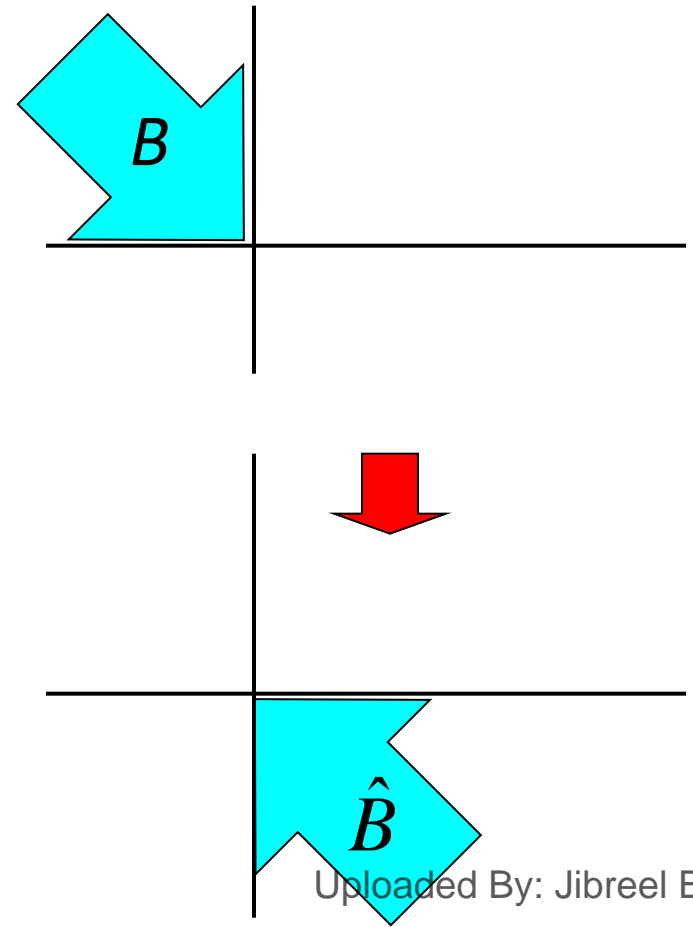
## Translation

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



## Reflection

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



# Structuring Elements

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- All morphological operations are based on using **structuring elements** (similar to filter masks)

1	1	1
1	1	1
1	1	1

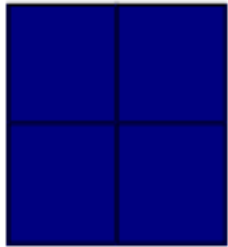
0	1	0
1	1	1
0	1	0

- Morphological operations are based moving the structuring element over the image pixels to check for a HIT or FIT
  - **FIT** - all **ON** pixels in the structuring element cover **ON** pixels in the image
  - **HIT** - any **ON** pixel in the structuring element covers an **ON** pixel in the image



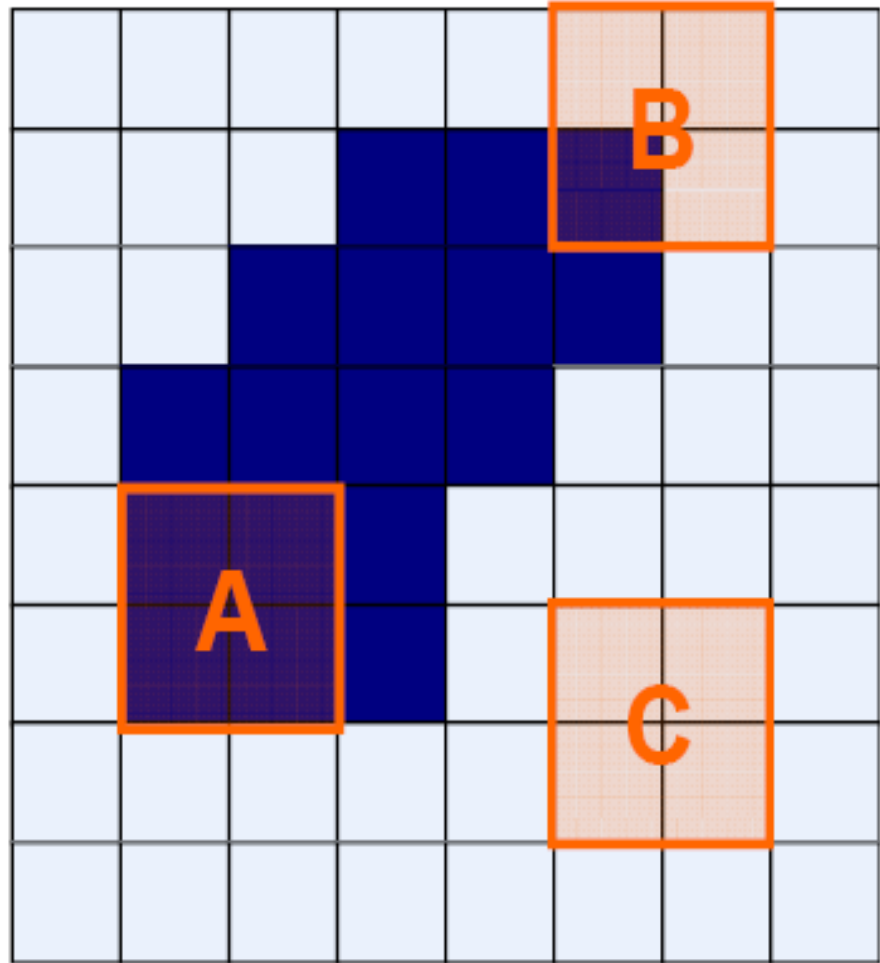
# Structuring Elements

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Structuring Element

- A – Fit
- B – Hit
- C – Miss



# Structuring Elements

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- Structuring elements can be of any size and make any shape
- Rectangular shapes are usually used with the origin being at the center pixel

1	1	1
1	1	1
1	1	1

0	1	0
1	1	1
0	1	0

0	0	1	0	0
0	1	1	1	0
1	1	1	1	1
0	1	1	1	0
0	0	1	0	0

- Fundamentally morphological image processing is very like spatial filtering
- The structuring element is moved across every pixel in the original image to give a pixel in a new processed image
- The value of this new pixel depends on the operation performed

# Structuring Elements

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0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	1	1	0	0	0	0	0	0	0
0	0	1	<b>B</b>	1	1	1	0	0	0	0	0
0	1	1	1	1	1	1	1	<b>C</b>	0	0	0
0	1	1	1	1	1	1	1	0	0	0	0
0	0	1	1	1	1	1	1	0	0	0	0
0	0	1	1	1	1	1	1	1	0	0	0
0	0	1	1	1	1	1	<b>A</b>	1	1	1	0
0	0	0	0	0	1	1	1	1	1	0	0
0	0	0	0	0	0	0	0	0	0	0	0

1	1	1
1	1	1
1	1	1

Structuring Element 1

0	1	0
1	1	1
0	1	0

Structuring Element 2

Location	Structuring Element 1	Structuring Element 2
A	FIT	FIT
B	HIT	FIT
C	HIT	HIT

# Dilation

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- Dilation of image  $A$  by structuring element  $B$  is given by  $A \oplus B$  such that

$$A \oplus B = \{z \mid \bigwedge_z (B)_z \cap A \neq \emptyset\}$$

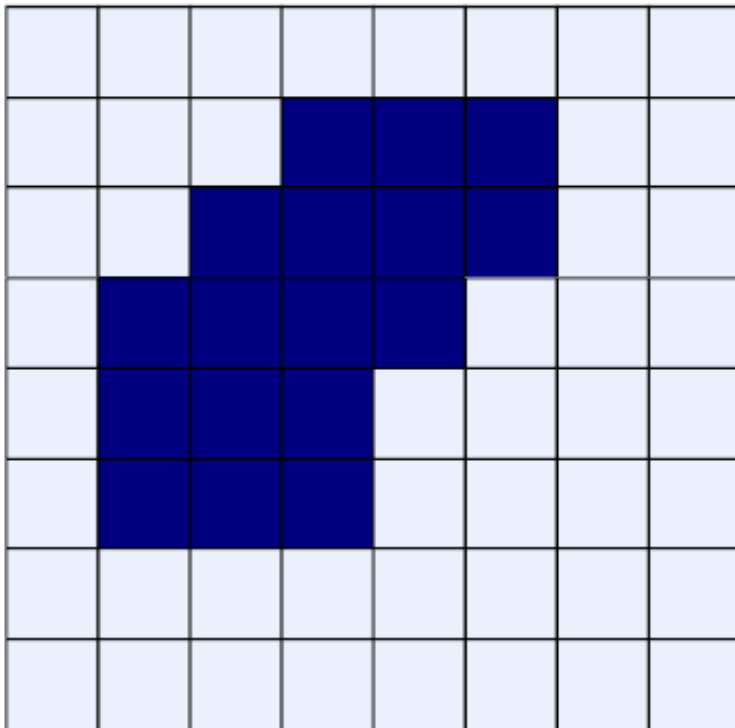
- The structuring element  $B$  is positioned with its origin at  $(x, y)$  and the new pixel value is determined using the rule:

$$g(x, y) = \begin{cases} 1 & , \text{ if } B \text{ hits } A \\ 0 & , \text{ otherwise} \end{cases}$$

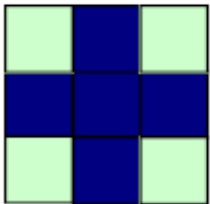
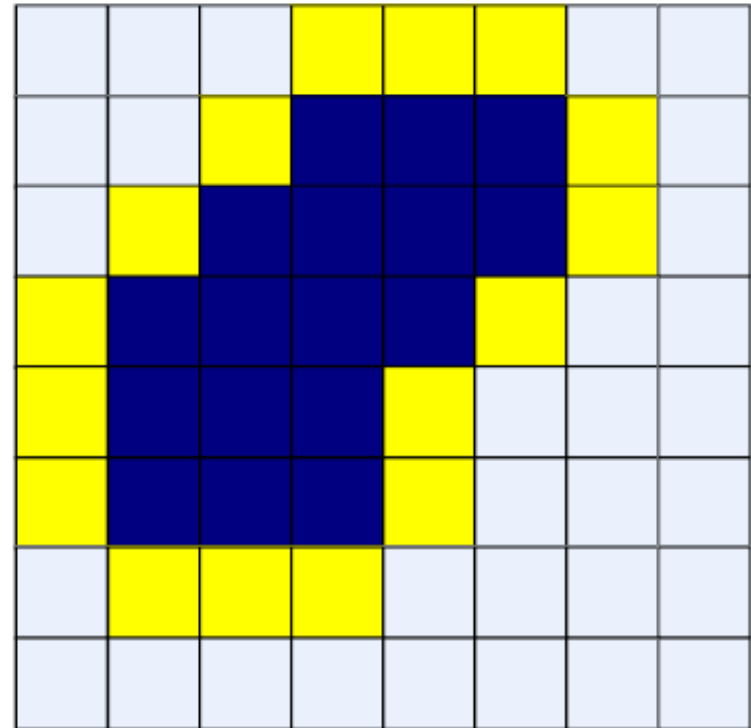
# Dilation

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Original Image



Processed Image



Structuring Element

- \* Keep pixels at which the structuring element hits at least one ON pixel
- \* Dilation enlarges objects



Pixels added to object



Kept Pixels

# Dilation

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Original image



Dilation by 3\*3  
square structuring  
element



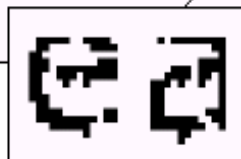
Dilation by 5\*5  
square structuring  
element

**Watch out:** In these examples a [1](#) refers to a black pixel!

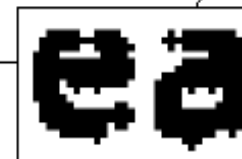
# Dilation

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Historically, certain computer programs were written using only two digits rather than four to define the applicable year. Accordingly, the company's software may recognize a date using "00" as 1900 rather than the year 2000.



**Historically, certain computer programs were written using only two digits rather than four to define the applicable year. Accordingly, the company's software may recognize a date using "00" as 1900 rather than the year 2000.**



0	1	0
1	1	1
0	1	0

a c  
b

**FIGURE 9.5**

(a) Sample text of poor resolution with broken characters (magnified view). (b) Structuring element. (c) Dilation of (a) by (b). Broken segments were joined.

## What Is Dilation For?

- Dilation can repair breaks



- Dilation can repair intrusions





# Erosion

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- Erosion of image  $A$  by structuring element  $B$  is given by  $A \ominus B$  such that

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

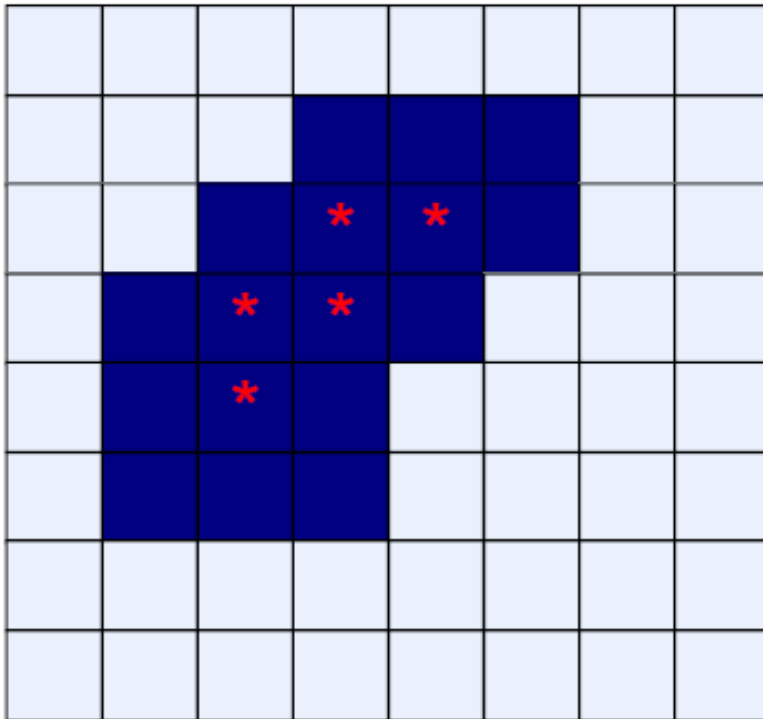
- The structuring element  $B$  is positioned with its origin at  $(x, y)$  and the new pixel value is determined using the rule:

$$g(x, y) = \begin{cases} 1 & , \text{ if } B \text{ fits } A \\ 0 & , \text{ otherwise} \end{cases}$$

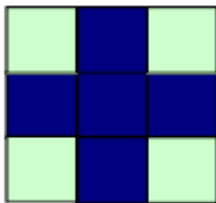
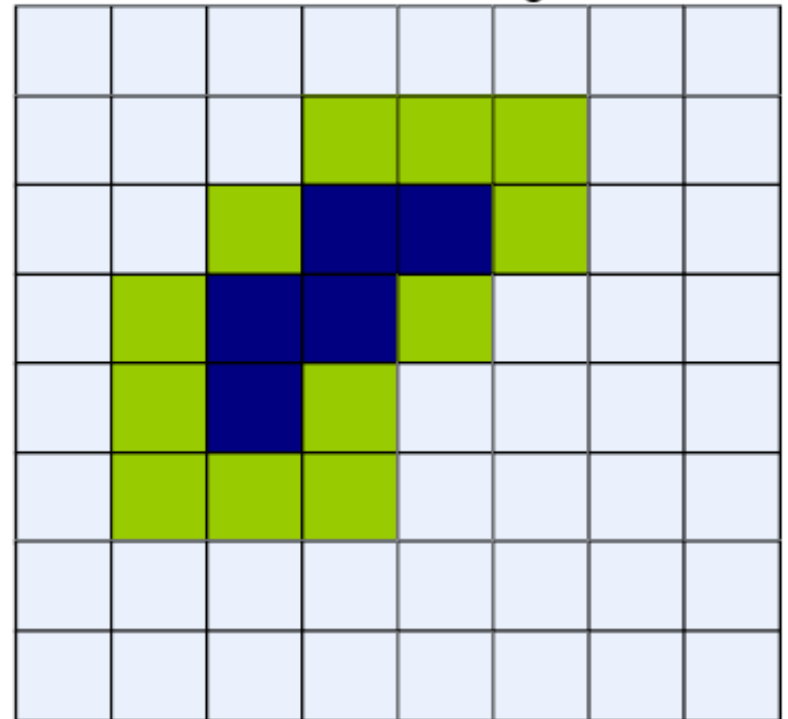
# Erosion

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Original Image



Processed Image



Structuring Element

- \* Keep pixels at which the structuring element scores a FIT
- \* Erosion shrinks objects



Removed  
Pixels

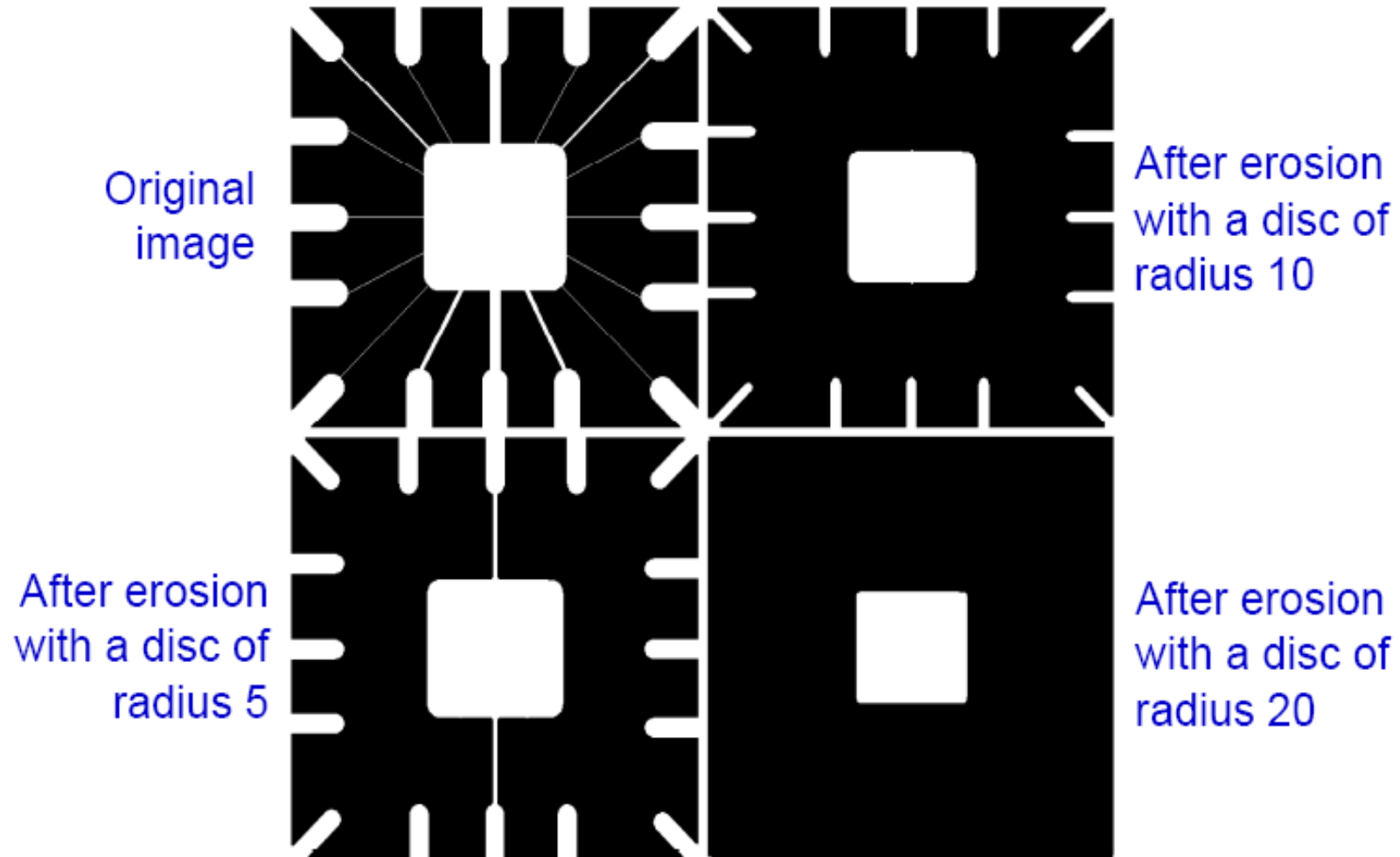


Kept Pixels

# Erosion

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## Using Erosion to remove image component



# Erosion

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Original image



Erosion by 3\*3  
square structuring  
element

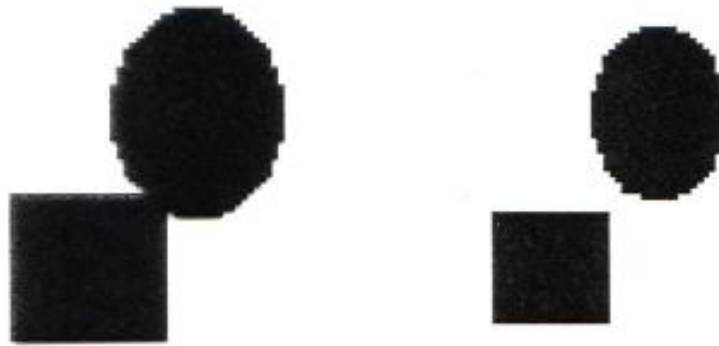


Erosion by 5\*5  
square structuring  
element

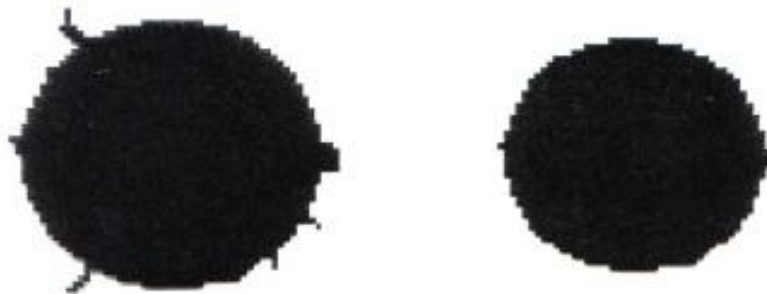
- **Watch out:** In these examples a 1 refers to a black pixel!

## What Is Erosion For?

- Erosion can split apart joined objects

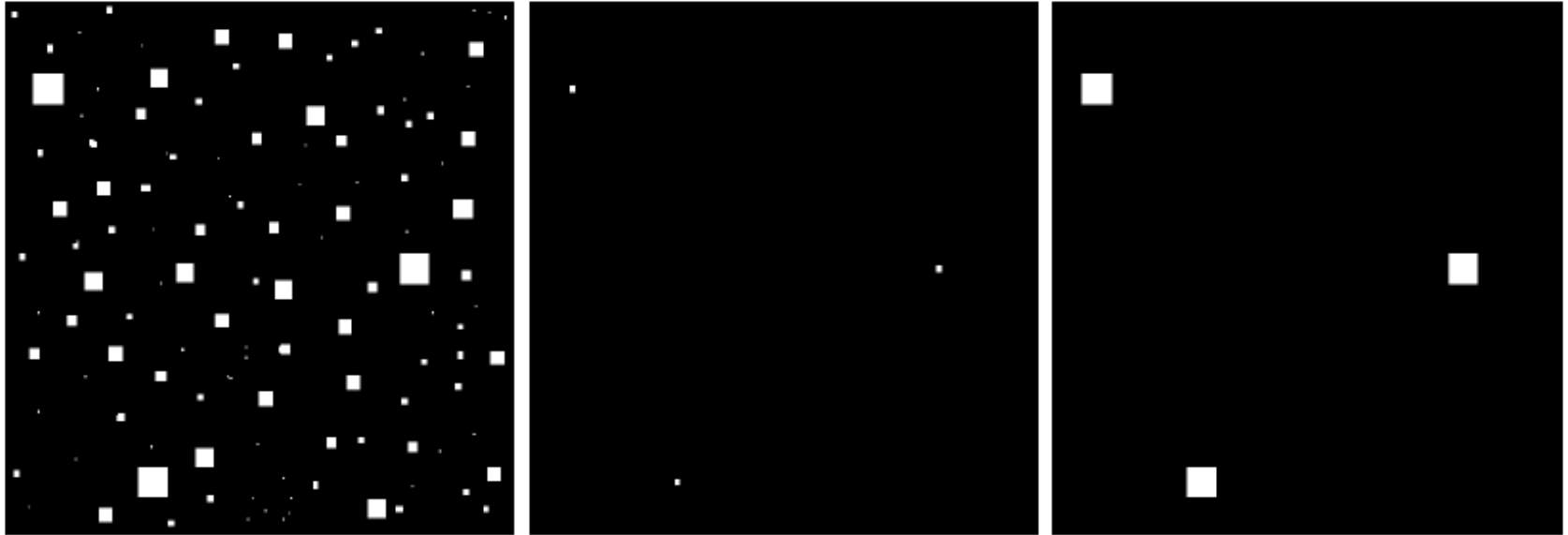


- Erosion can strip away extrusions



# Application of Dilation and Erosion

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a b c

**FIGURE 9.7** (a) Image of squares of size 1, 3, 5, 7, 9, and 15 pixels on the side. (b) Erosion of (a) with a square structuring element of 1's, 13 pixels on the side. (c) Dilation of (b) with the same structuring element.

Erosion thins objects in a binary image

- Image details smaller than SE are removed

Dilation grows object in a binary image

# Opening

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- The opening operation on image  $A$  by structuring element  $B$  is defined as

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

in other words, it is erosion followed by dilation

- Generally, opening is used to
  - Smooth the contour of an object
  - Break narrow paths between large objects
  - Eliminate thin protrusions

# Opening

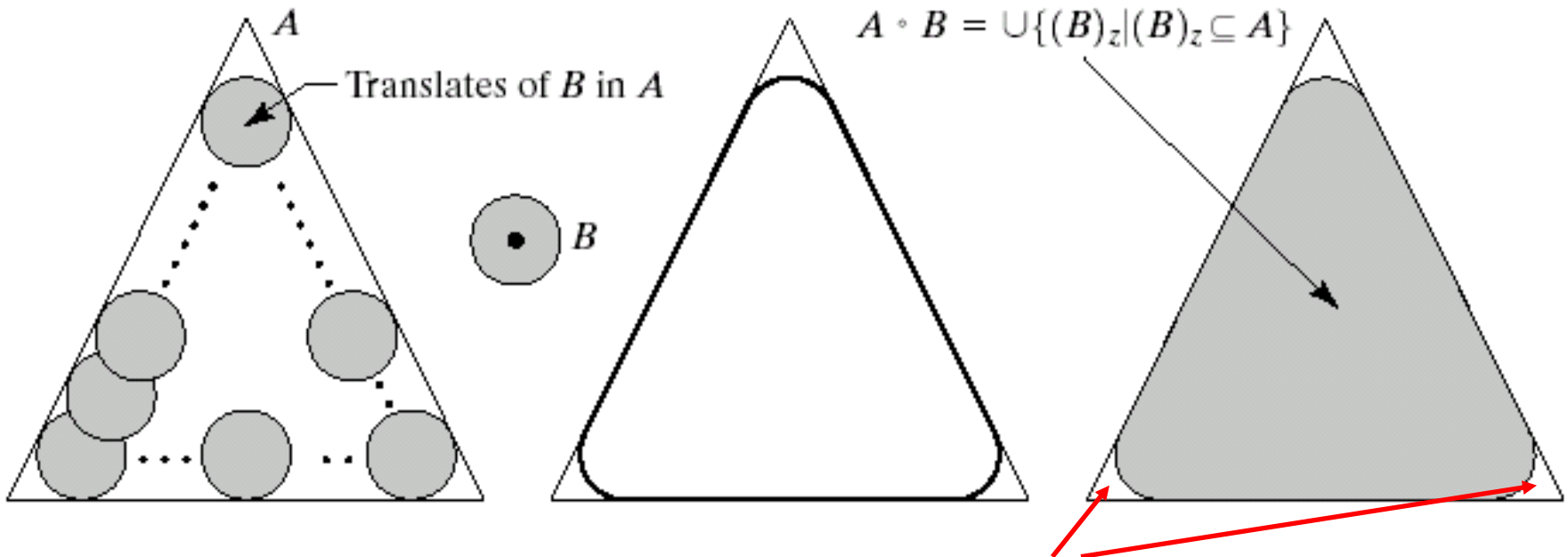
24

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

or

$$A \circ B = \cup \{(B)_z \mid (B)_z \subseteq A\}$$

= Combination of all parts of A that can completely contain B



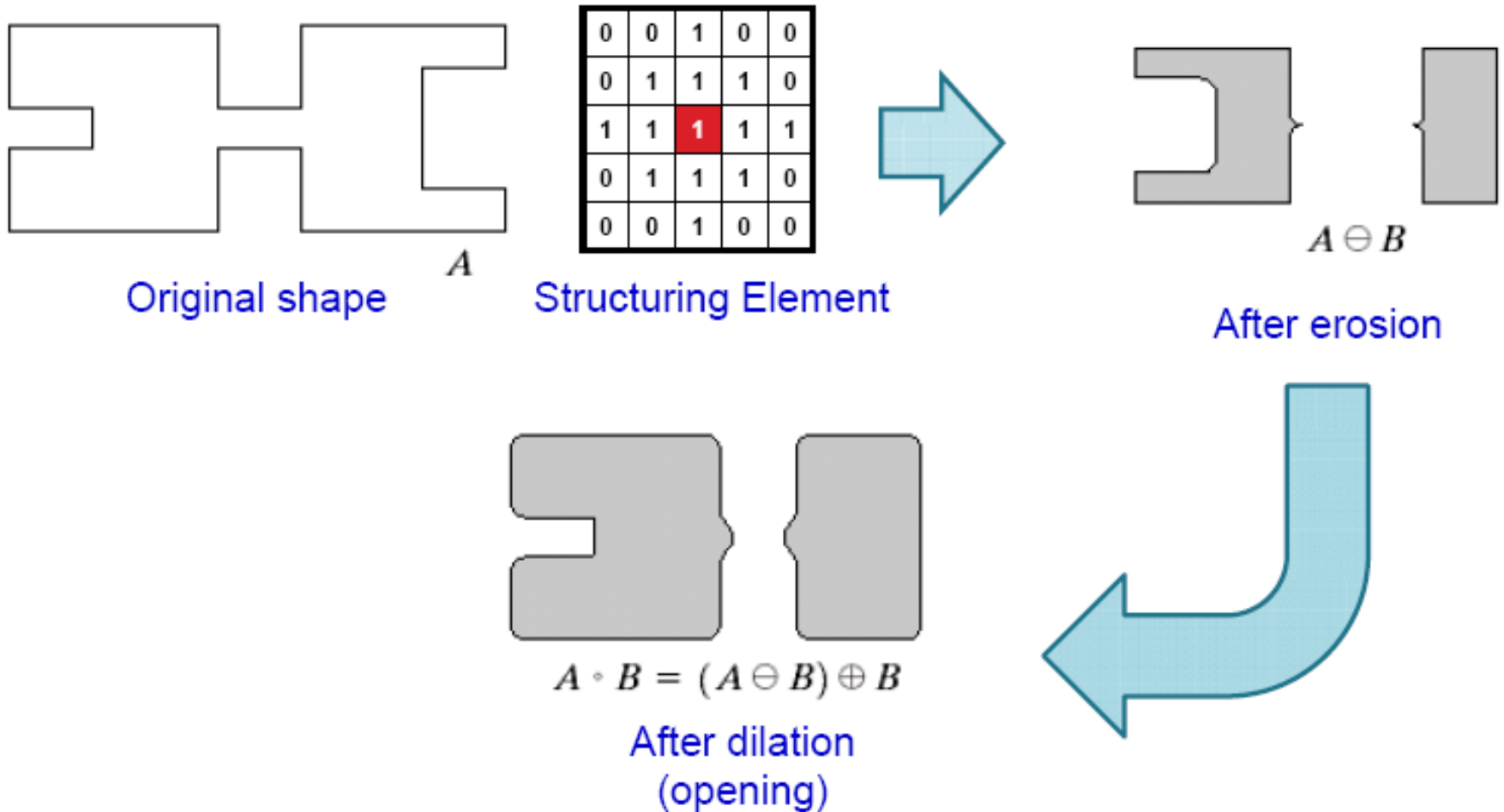
Opening eliminates narrow and small details and corners.



# Opening

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- **Example**



# Closing

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- The closing operation on image  $A$  by structuring element  $B$  is defined as

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

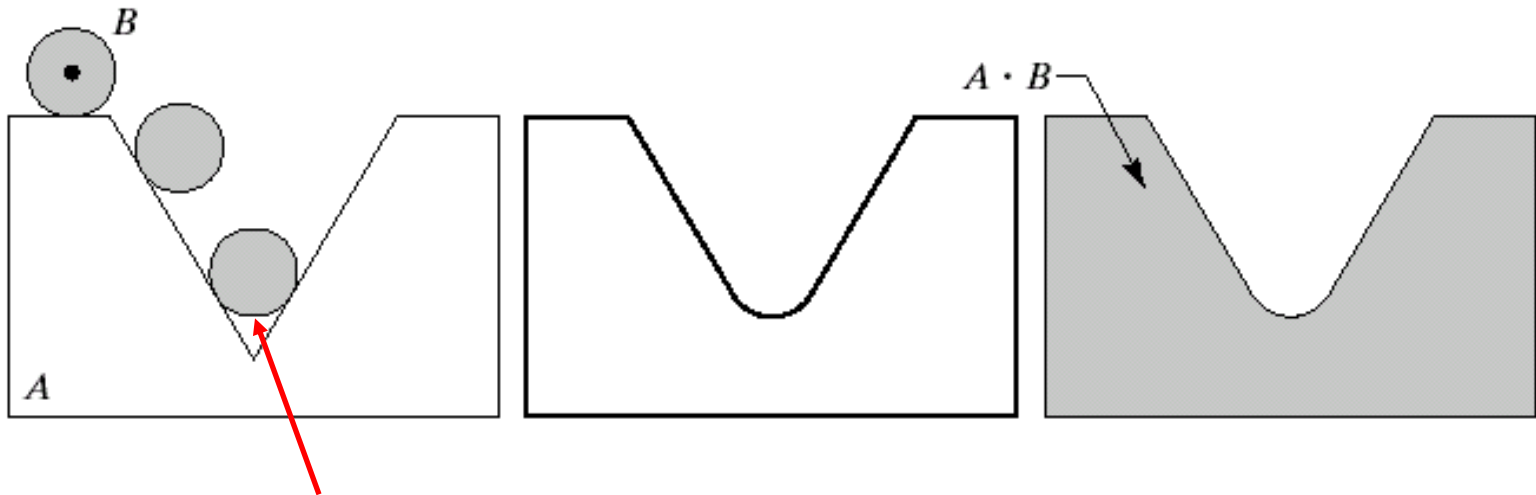
in other words, it is dilation followed by erosion

- Generally, closing is used to
  - Smooth the contour of an object
  - Fuse narrow paths between large objects
  - Eliminate thin small holes and fill gaps in the contour

# Closing

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

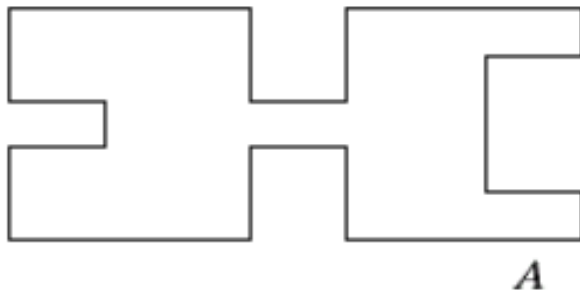


Closing fills narrow gaps and notches

# Closing

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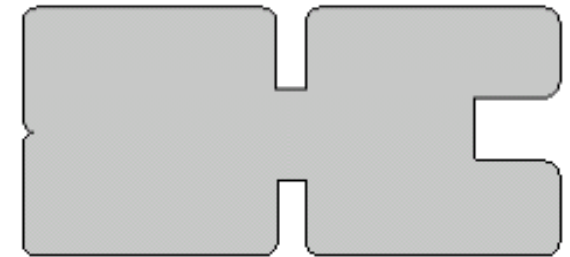
- **Example**



Original shape

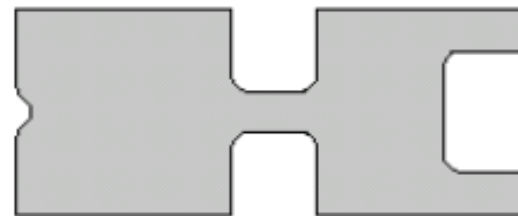
0	0	1	0	0
0	1	1	1	0
1	1	1	1	1
0	1	1	1	0
0	0	1	0	0

Structuring Element



$A \oplus B$

After dilation



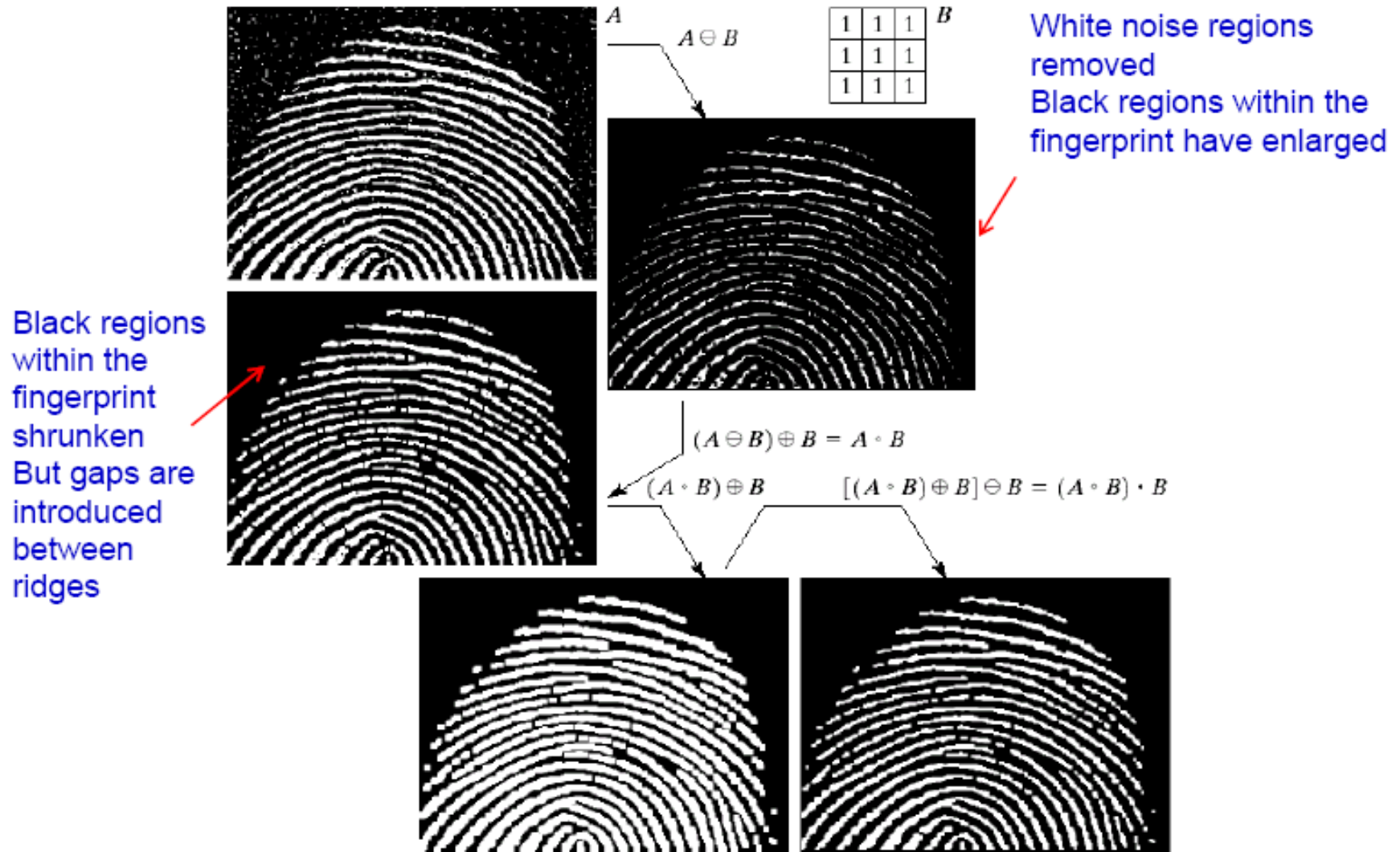
$A \cdot B = (A \oplus B) \ominus B$

After erosion  
(Closing)



# Example

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# Hit-or-Miss Transformation

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- A basic tool for shape detection
- Suppose we have an image  $A$  that consists of three objects  $C$ ,  $D$ , and  $E$  and we want to detect the presence of shape  $D$ ; assuming we know its shape

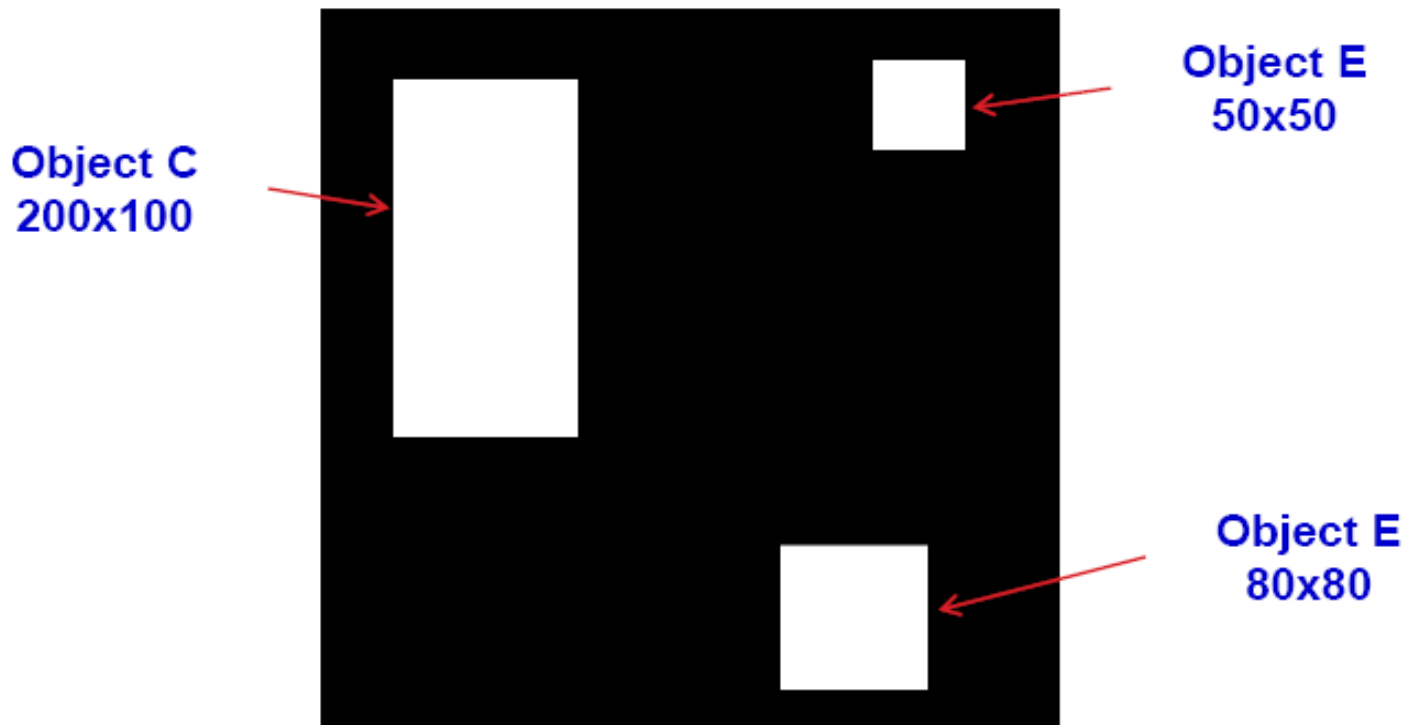
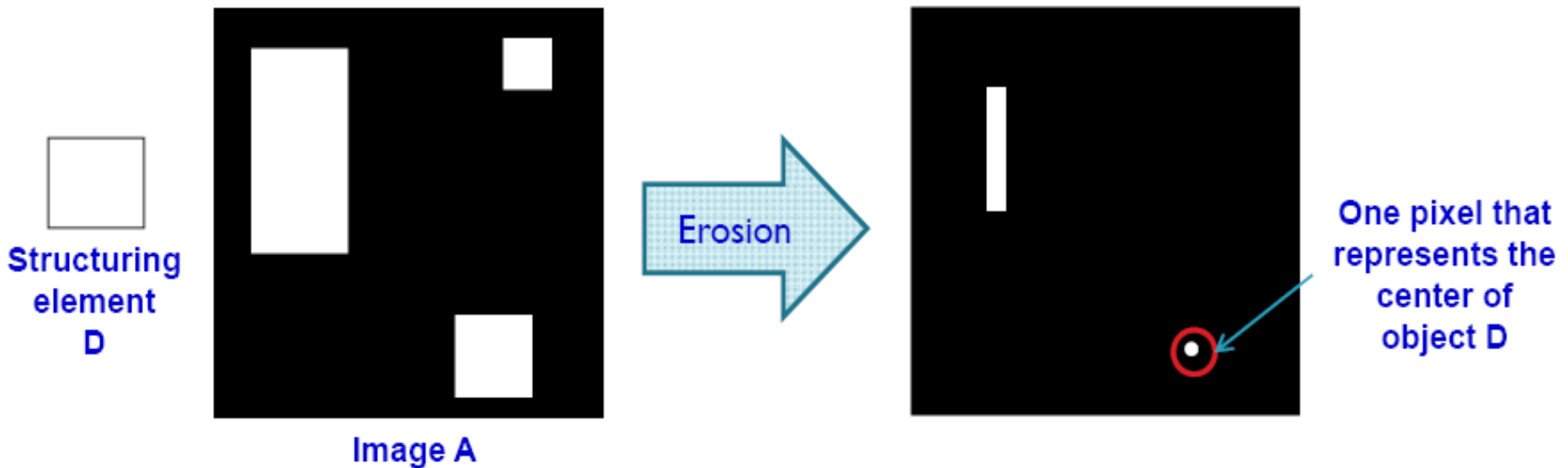


Image A

# Hit-or-Miss Transformation

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- We can start by eroding  $A$  with a structuring element that with the same shape as  $D$

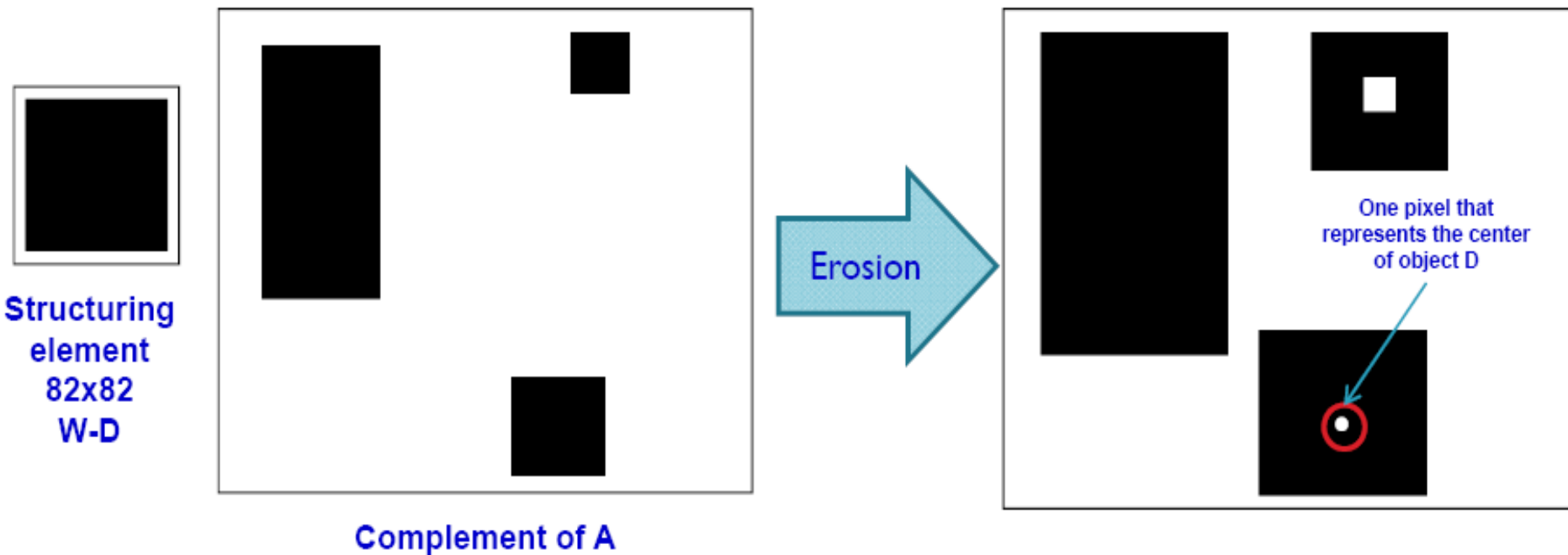


- However, the result may contain parts of larger objects

# Hit-or-Miss Transformation

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- Consider the complement of  $A$  and a new structuring element that is one pixel thicker than the object



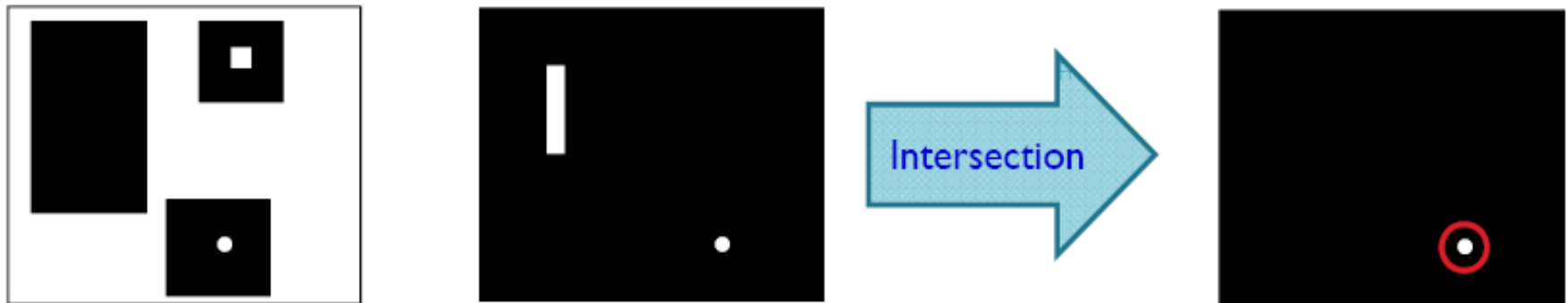
- Note how the center of object  $D$  was detected again !



# Hit-or-Miss Transformation

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- Consider the intersection of the erosion results from the previous two slides



- Object is successfully located !
- Formally, the hit-or-miss transformation on image  $A$  to detect some object  $X$ , is performed by

$$A \circledast X = (A \ominus X) \cap [A^c \ominus (W - X)]$$

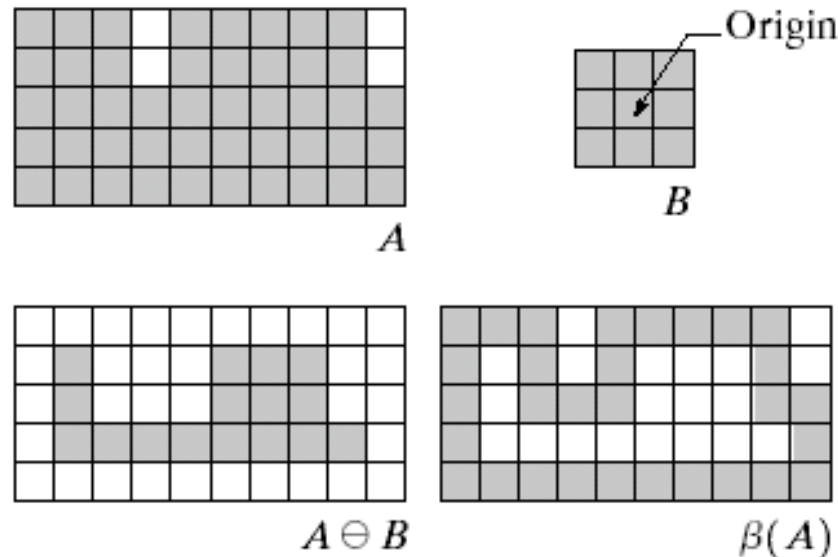
where  $W$  is a the object  $X$  thickened by one pixel

# Boundary Extraction

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- The boundary of set  $A$  denoted by  $\beta(A)$  can be obtained by first eroding  $A$  by a suitable structuring element  $B$  then perform the set difference between  $A$  and its erosion

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



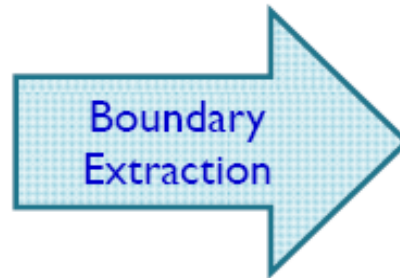
# Boundary Extraction

35

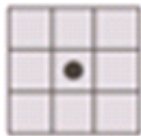
- **Example**



Original binary image



Boundary detected based on morphological operators



$B$

Structuring element

# Hole Filling

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- A hole may be defined as a background region surrounded by a connected border of foreground pixels
- The algorithm presented here assumes that we know one pixel for each hole in the image
- **Algorithm**
  - Form an array  $X_0$  of the same size as  $A$  and initialize it with zeros except locations that correspond to pixels inside the regions to be filled
  - Apply the following equation iteratively on array  $X$  until  $X_k = X_{k-1}$  to form the filled holes

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

- The original image with filled holes is found by

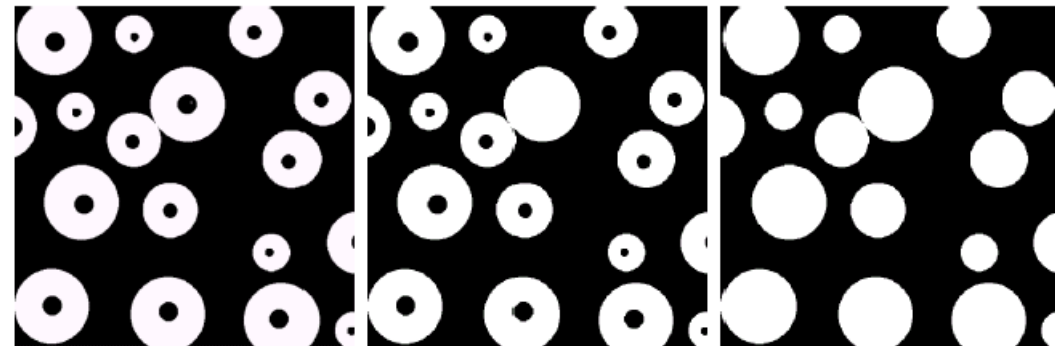
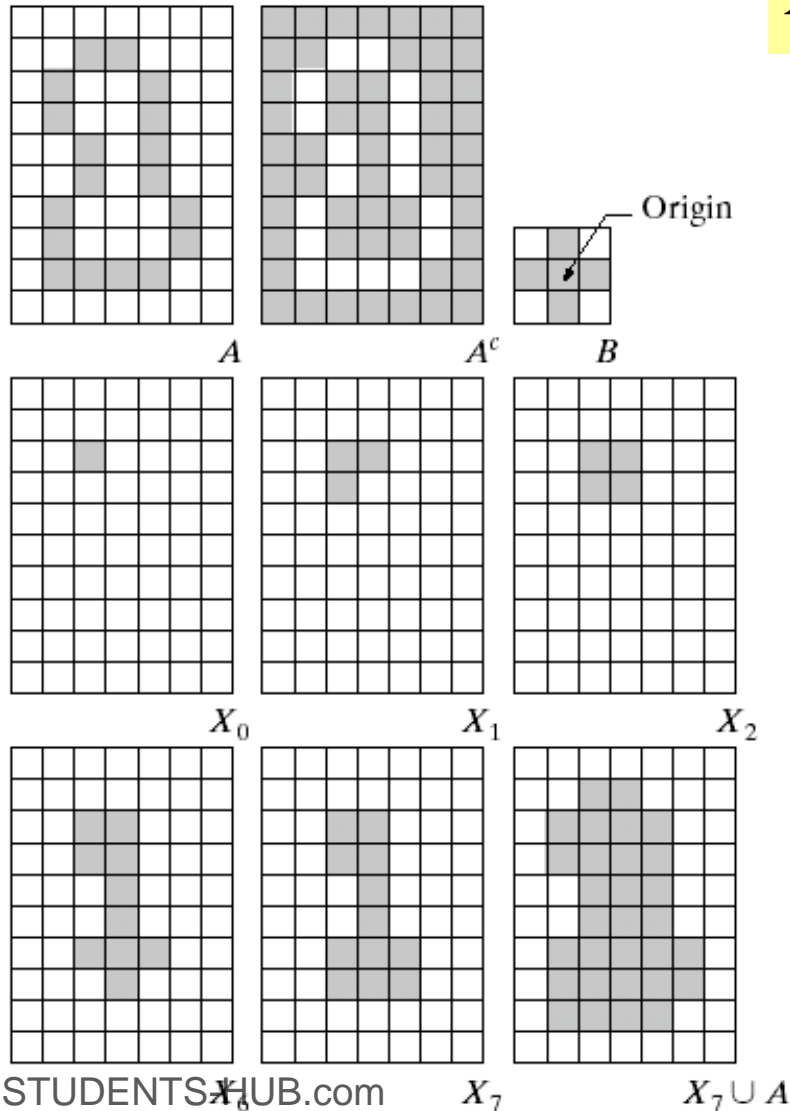
$$A_{filled} = X_k \cup A$$

# Hole Filling

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

where  $X_0 = \text{seed pixel } p$



Original  
image

Results of region filling

# Connected Components

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- Let  $Y$  represent a connected component contained in a set  $A$ 
  - ▣ Assume that a point  $p$  of  $Y$  is known
  - ▣ Then the following yields all the element of

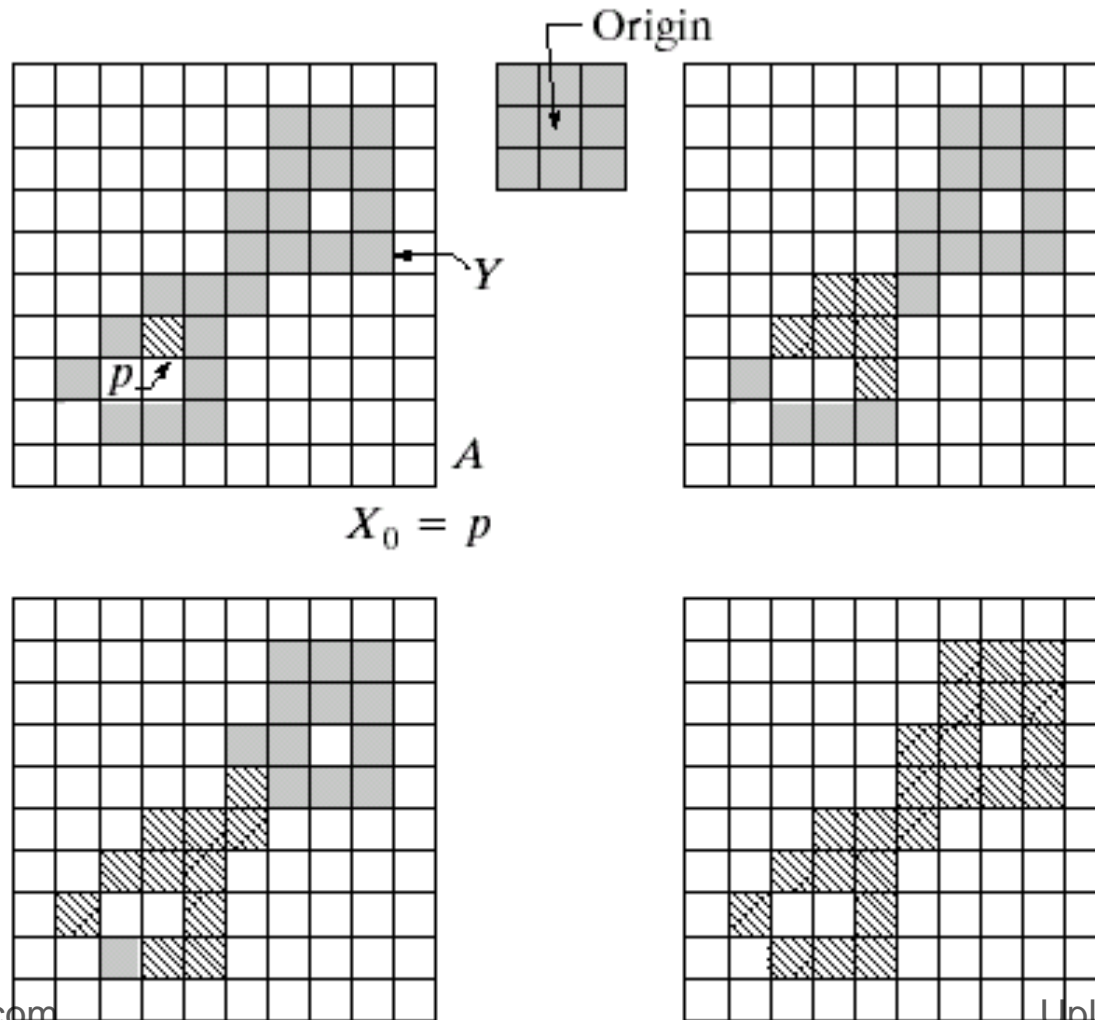
$$X_k = (X_{k-1} \oplus B) \cap A, \quad k = 1, 2, 3, \dots$$

- where  $X_0 = p$ ,  $B$  is the suitable structuring element
- The algorithm terminates if  $X_k = X_{k-1}$ , and let  $Y = X_k$

# Connected Components

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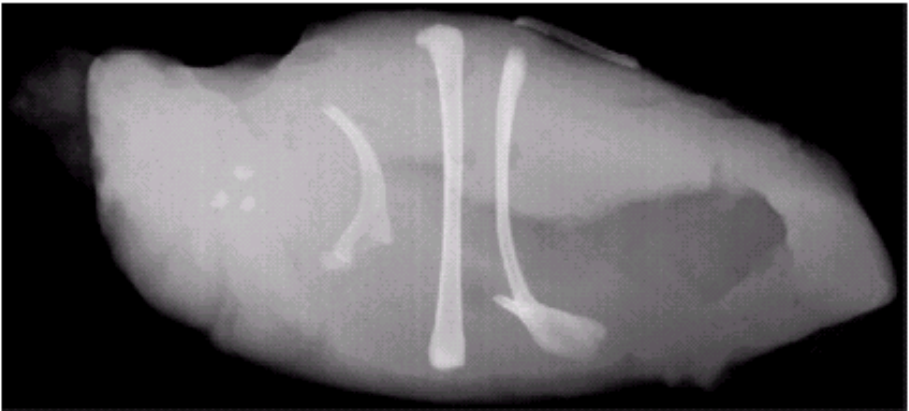
$$X_k = (X_{k-1} \oplus B) \cap A \quad \text{where } X_0 = \text{seed pixel } p$$



# Connected Components

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X-ray image  
of bones



Thresholded  
image



Connected  
components



Connected component	No. of pixels in connected comp
01	11
02	9
03	9
04	39
05	133
06	1
07	1
08	743
09	7
10	11
11	11
12	9
13	9
14	674



# Convex Hull

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- A set  $A$  is said to be convex
  - ▣ if the straight line segment joining any two points in  $A$  lies entirely within  $A$
- The convex hull  $H(=C(S))$  of an arbitrary set  $S$  is the smallest convex set containing  $S$
- The set difference  $H-S$  is called the convex deficiency of  $S$
- Morphological algorithm for obtaining the convex hull  $C(A)$  of a set  $A$

➡ Let  $B^i$ ,  $i=1,2,3,4$ , representing the four structuring elements

➡  $X_k^i = (X_{k-1}^i \circledast B^i) \cup A$ ,  $i = 1,2,3,4$  and  $k = 1,2,3,\dots$   
with  $X_0^i = A$

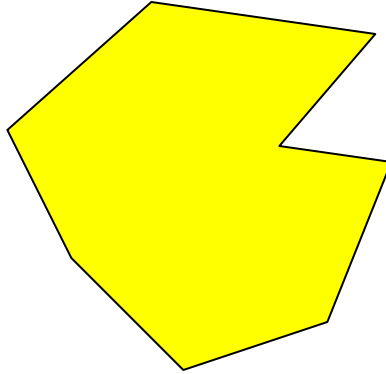
➡ When the results converge to  $D^i$ , the convex hull of  $A$  is given by

$$C(A) = \bigcup_{i=1}^4 D^i$$

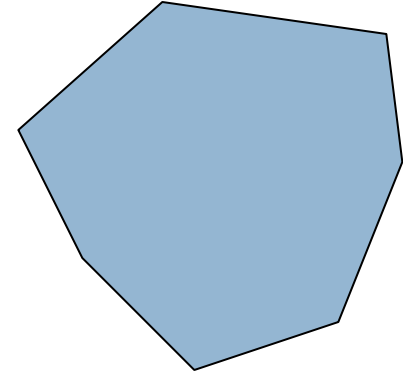
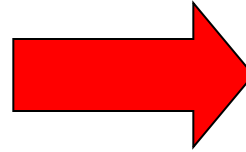
# Convex Hull

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Convex hull  
has no  
concave part.



Convex hull



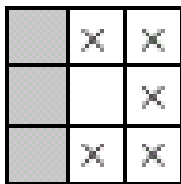
Algorithm:

$$C(A) = \bigcup_{i=1}^4 D^i$$

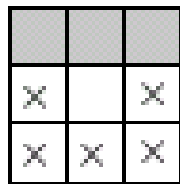
where

$$D^i = X_{conv}^i$$

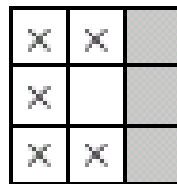
$$X_k^i = (X_{k-1} \circledast B^i) \cup A, \quad i = 1, 2, 3, 4$$



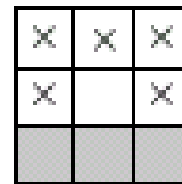
$B^1$



$B^2$



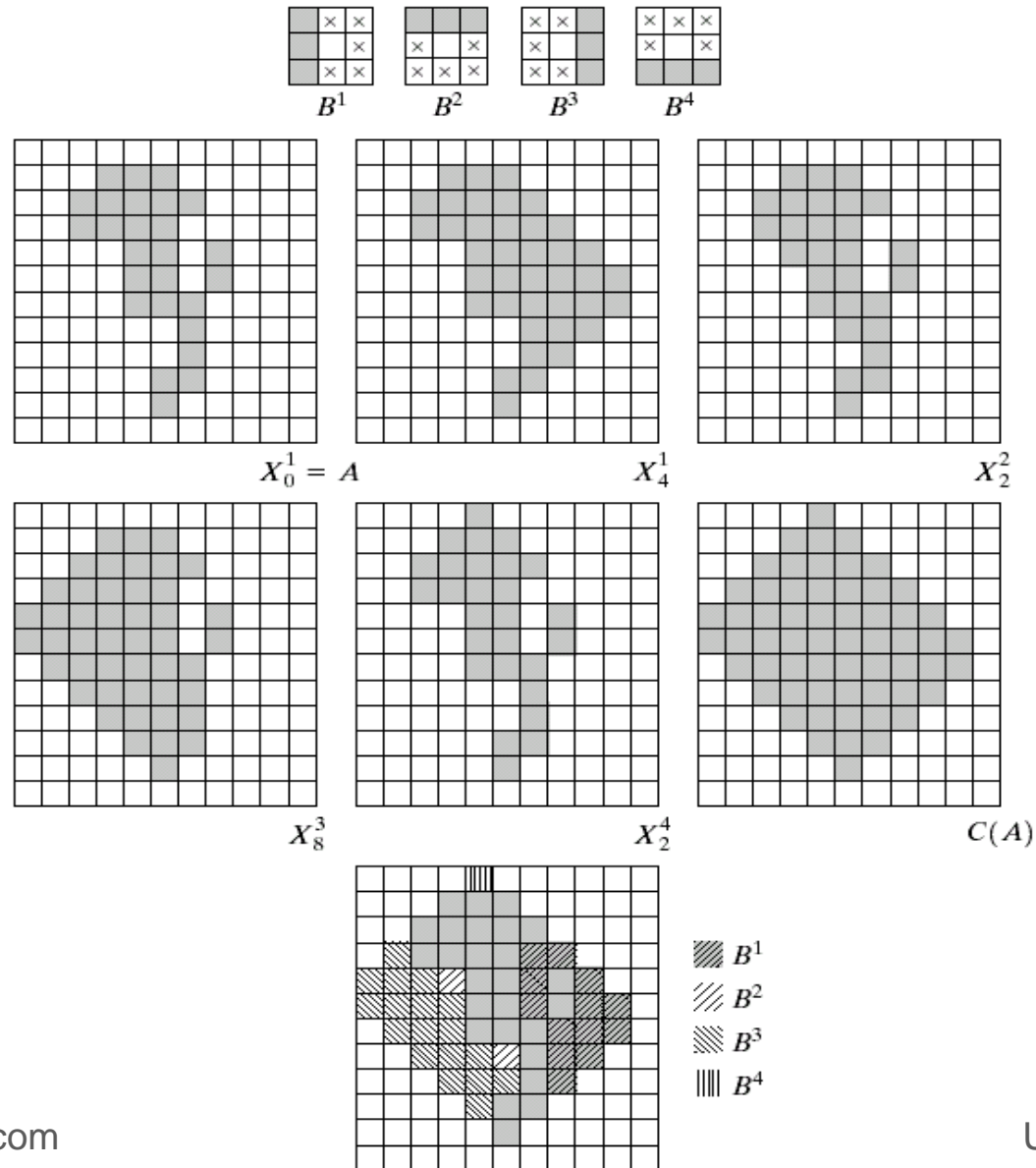
$B^3$



$B^4$

# Convex Hull

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# Thinning

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→ The thinning of a set  $A$  by a structuring element  $B$

$$\begin{aligned} A \otimes B &= A - (A * B) \\ &= A \cap (A * B)^c \end{aligned}$$

→ A more useful expression for thinning based on a sequence of structuring elements

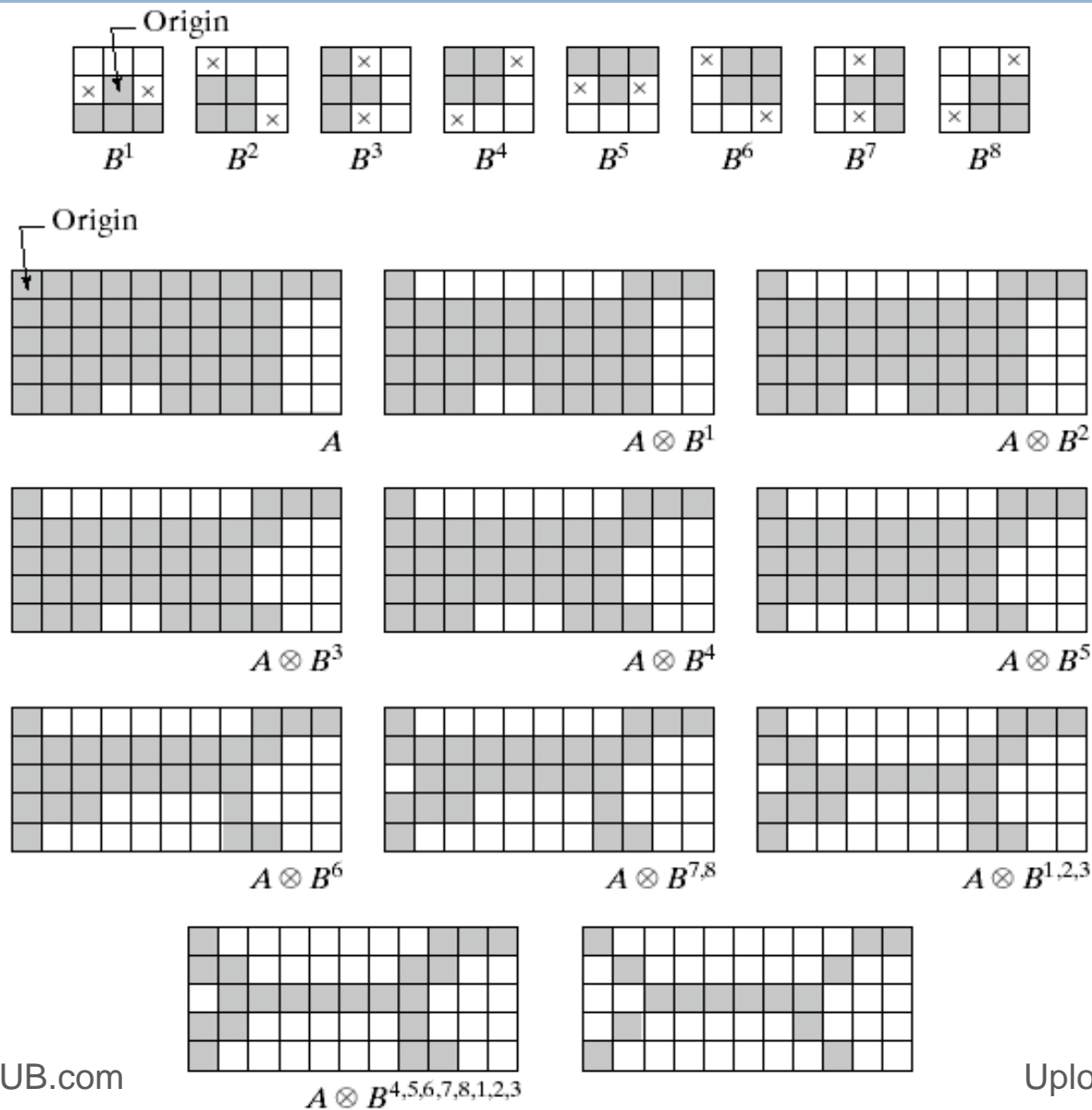
$$\{B\} = \{B^1, B^2, B^3, \dots, B^n\}$$

→ where  $B^i$  is a rotated version of  $B^{i-1}$

$$A \otimes \{B\} = (((...((A \otimes B^1) \otimes B^2)...) \otimes B^n)$$

# Thinning

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# Thickening

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- Thickening is the morphological dual of thinning

$$A \odot B = A \cup (A * B)$$

- As in thinning, thickening can be defined as a sequential operation

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

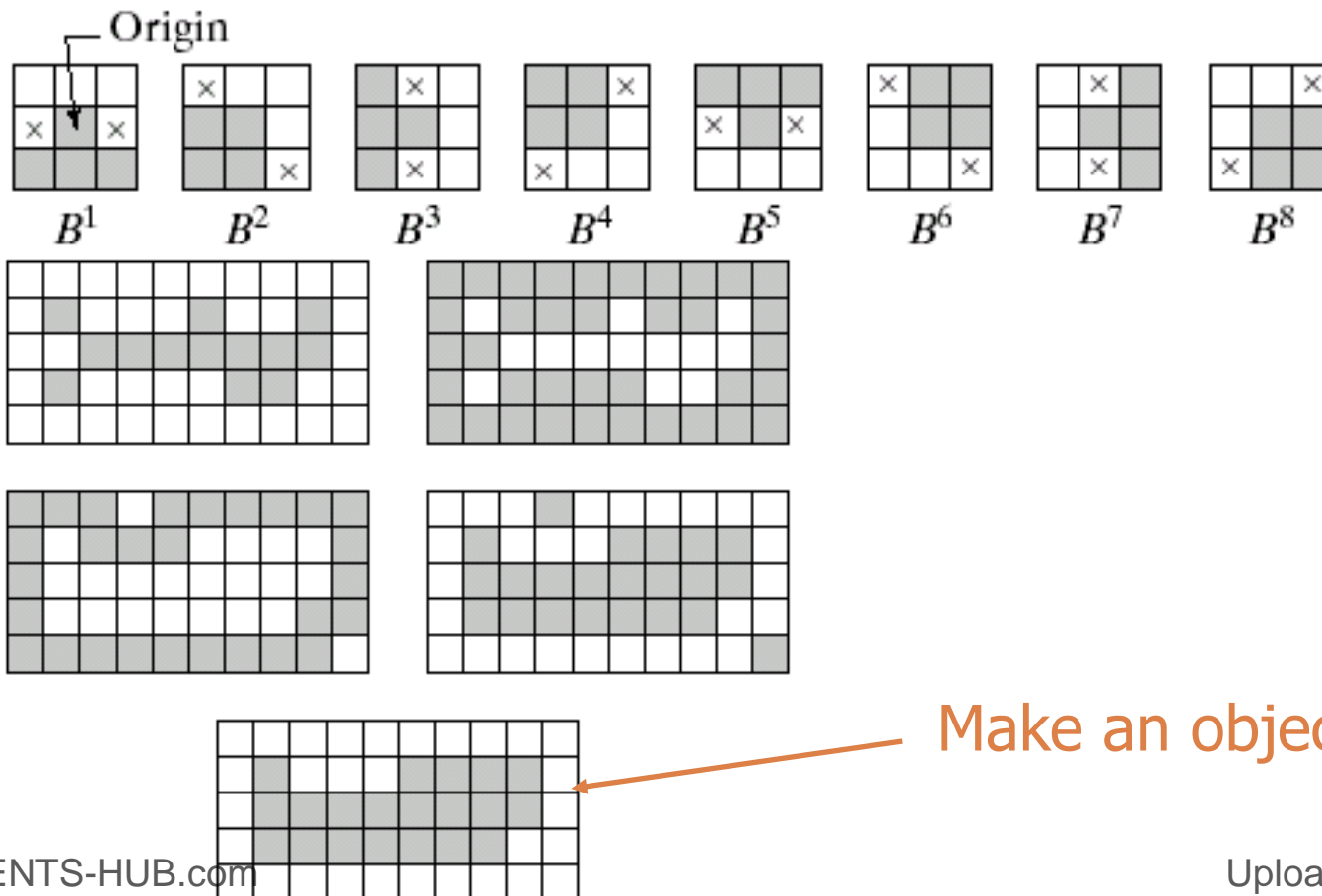
- The structuring elements used for thickening have the same form in thinning, but with all 1's and 0's interchanged
- In general, thickening is accomplished by thinning the background and then taking complement of the result
- The thinned background forms a boundary for the thickening process

# Thickening

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$$A \oplus B = A \cup (A * B)$$

$$A \oplus \{B\} = (((...((A \oplus B^1) \oplus B^2)...)) \oplus B^n)$$



# Skeletons

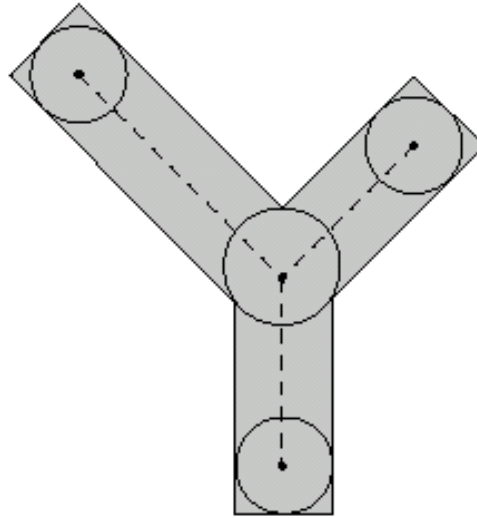
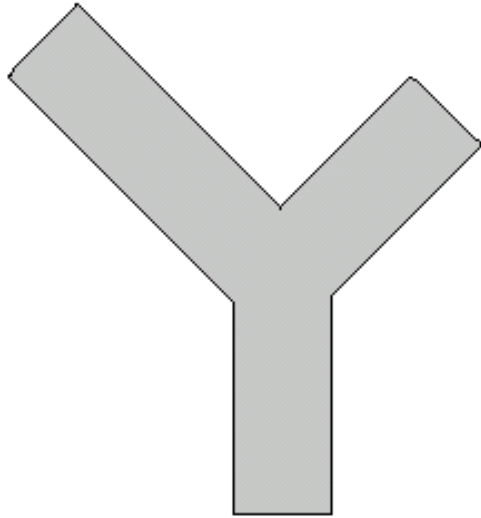
48

- Skeleton,  $S(A)$ , of a set  $A$ 
  1. If  $z$  is a point of  $S(A)$  and  $(D)z$  is the largest disk centered at  $z$  and contained in  $A$ , one cannot find a larger disk containing  $(D)z$  and included in  $A$ . The disk  $(D)z$  is called a maximum disk
  2. The disk  $(D)z$  touches the boundary of  $A$  at two or more different places
- An inner point belongs to the skeleton if it has at least two closest boundary points

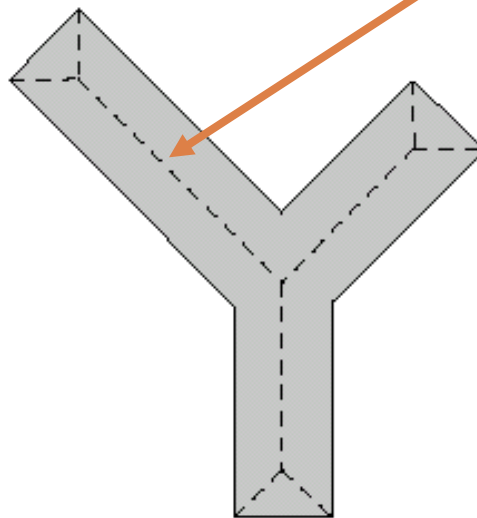
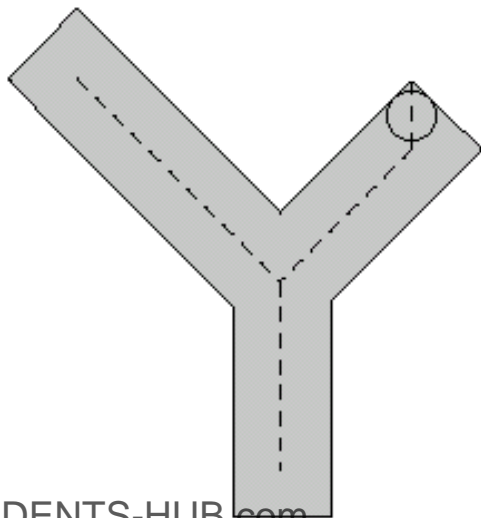


# Skeletons

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Dot lines are  
skeletons of this  
structure



# Skeletons

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

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

k successive erosions

- And K is the last iterative step before A erodes to an empty set
- The set A can be reconstructed by

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

where k successive dilations

$$(S_k(A) \oplus kB) = (((... (S_k(A) \oplus B) \oplus B) \oplus ...) \oplus B$$