

# Computer Graphics and Image Processing

## Part 3: Image Processing 6 – Morphological Operations

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# Morphological operations

- Main concepts
- Structuring element
- Erosion
- Dilation
- Morphological gradients
- Opening
- Closing

Additional resources and sources of some images:

<http://www.inf.u-szeged.hu/ssip/1996/morpho/morphology.html>  
<http://www.cs.ru.nl/~ths/rt2/col/h11/11morphENG.html>

# Basic concepts

**Morphology:** a study of structure or form.

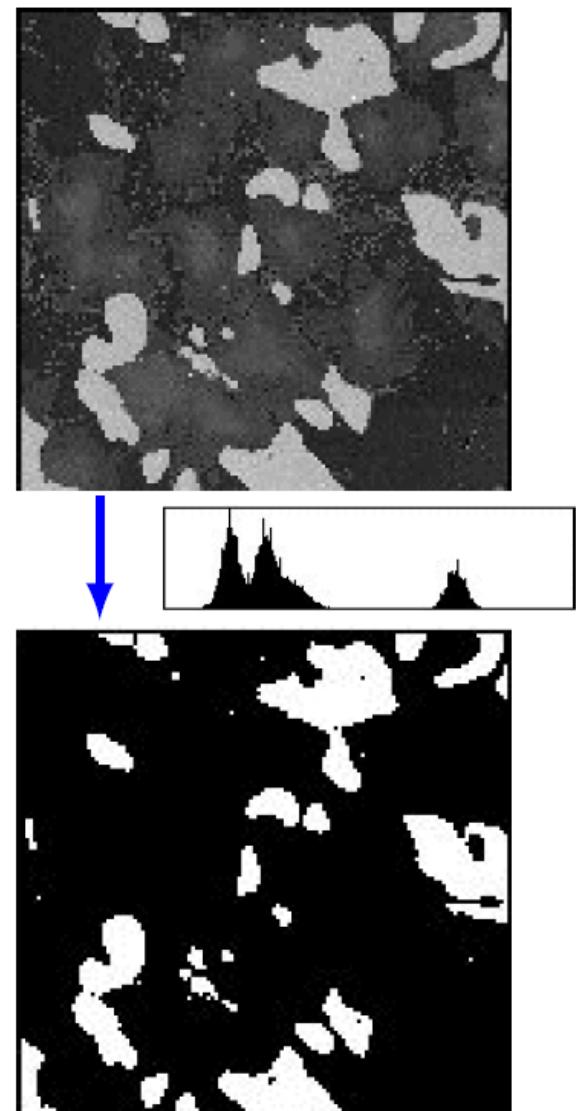
(<http://www.merriam-webster.com/dictionary/morphology>)

**Morphological image processing** may remove imperfections of a binary image.

- Regions in binary images produced by simple thresholding are typically distorted by noise.

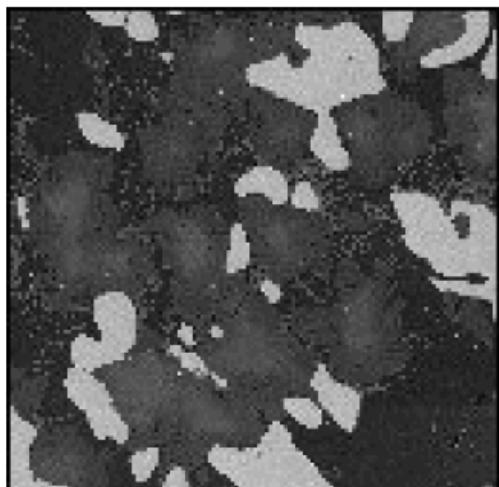
**Morphological operations** are non-linear and account for structure and forms of regions (objects) to improve an image.

- These operations can be extended to greyscale images.



# Binary images: object/background

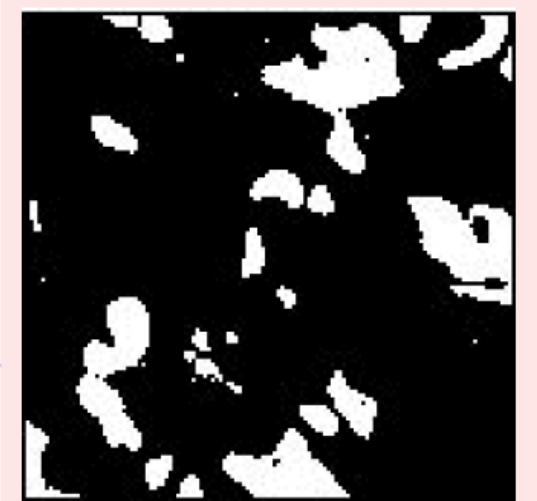
White  $\underbrace{\text{background}}_{0}$  – black foreground  $\underbrace{\text{objects}}_{1} \Rightarrow$



*For morphological processing, the foreground (i.e. object) and background pixels are treated as 1s and 0s, respectively, independently of their visual black-white coding.*

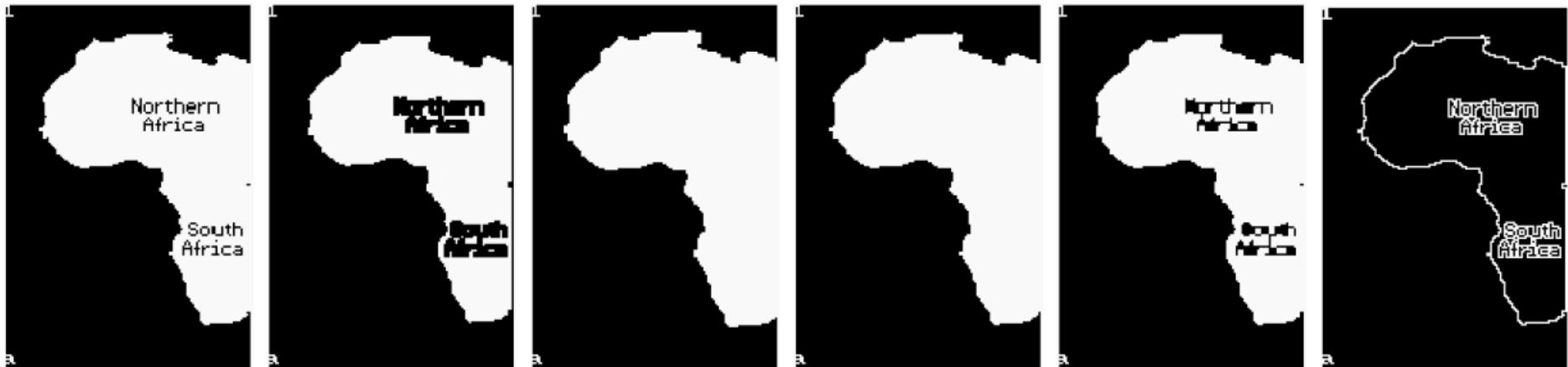


Black  $\underbrace{\text{background}}_{0}$  – white foreground  $\underbrace{\text{objects}}_{1} \Rightarrow$



# Morphological operations: examples

White objects on black background: “1” / “0” as white / black



Image

Erosion

Dilation

Closing

Opening

Gradient



Black objects on white background: “1” / “0” as black / white

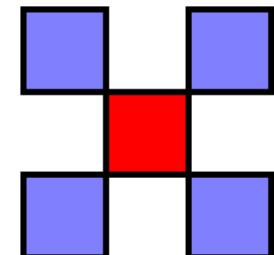
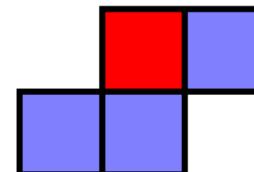
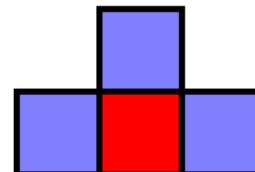
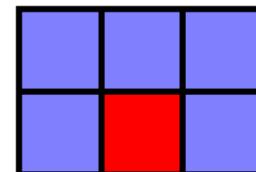
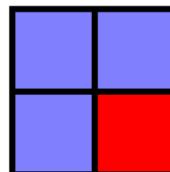
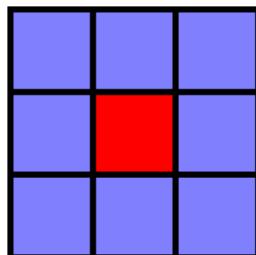
# Basic properties

Morphological operations rely on relative ordering of pixel values, not on their numerical values.

- Thus the operations are especially suited to binary image processing.
- These operations can be applied also to greyscale images such that their absolute pixel values are of no or minor interest.

Morphological operations probe an image with a small shape or template called a **structuring**, or structure **element** (SE).

- SE resembles a convolution kernel in linear filtering.



# Structuring element

A small binary image, i.e. a small matrix of pixels, each with a value of zero (0) or one (1).

- Zero-valued pixels of the SE are ignored.
- **Size** of the SE: the matrix dimensions.
- **Shape** of the SE: the pattern of ones and zeros.
- **Origin** of the SE: usually, one of its pixels.
  - Generally, the origin can be also outside the matrix.

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

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

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

Origin

1	1	1
1	1	1
1	1	1

Square 5x5 element

Diamond-shaped 5x5 element

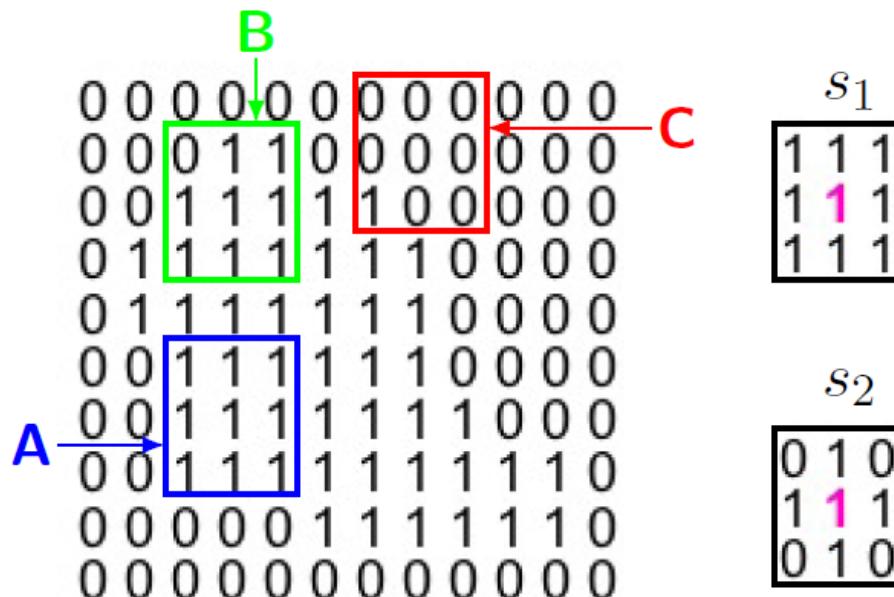
Cross-shaped 5x5 element

Square 3x3 element

# Structuring element

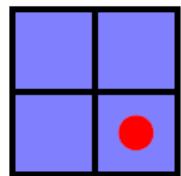
When an SE is placed in a binary image, each its pixel is associated with the pixel of the area under the SE:

- The SE **fits** the image if **for each** of its pixels set to 1 the corresponding image pixel is also 1.
- The SE **hits** (intersects) the image if **at least for one** of its pixels set to 1 the corresponding image pixel is also 1.



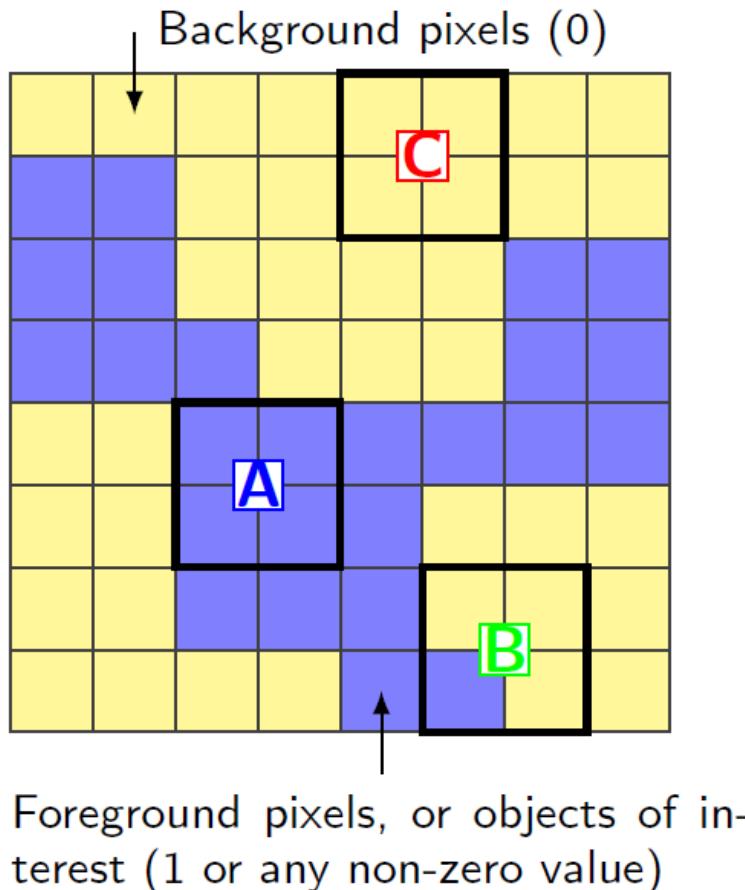
	A	B	C
fit	$s_1$	yes	no
	$s_2$	yes	yes
hit	$s_1$	yes	yes
	$s_2$	yes	yes

# Probing with an SE



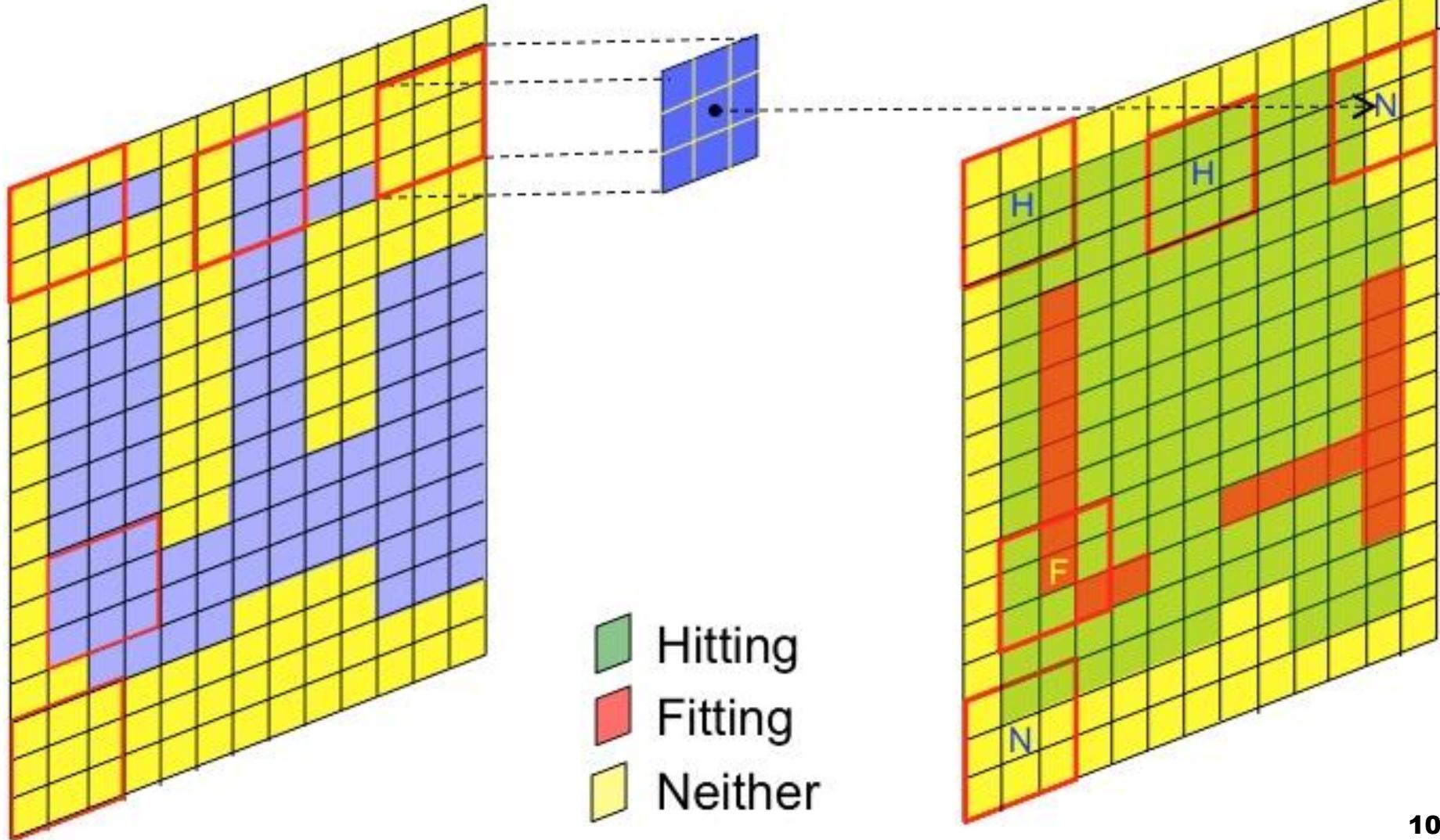
SE

- A:** SE **fits** the foreground (object);
- B:** SE **hits** (intersects with) the object;
- C:** SE neither fits, nor hits the object.



- SE is positioned at all possible locations in an image and compared with the pixel neighbourhood at each location.
- Operations test whether the SE “fits” within the neighbourhood or “hits” the neighbourhood.
- An operation on a binary image creates a new binary image in which the pixel has a foreground value only if the test at that place in the input image is successful.

# Probing with an SE



# Boundary handling

- Mainly two strategies for padding

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

Extend image boundary with zeros  
(Coderunner quiz *strategy 0*)

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

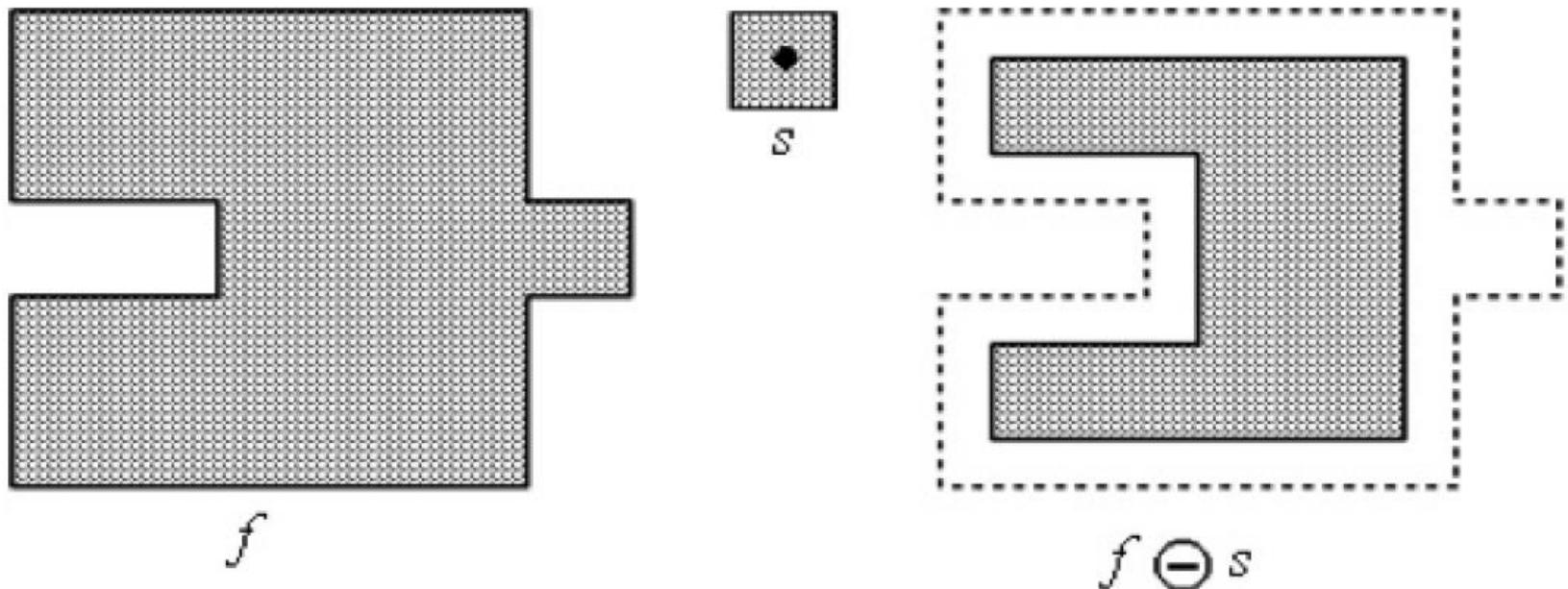
Extend image boundary with boundary values  
(Coderunner quiz *strategy 1*)

# Fundamental operation 1: Erosion

Erosion  $f \ominus s$  of a binary image  $f$  by an SE  $s$  produces a new binary image  $g = f \ominus s$ .

Eroded image  $g$  has ones in all locations  $(x, y)$  of an origin of the SE  $s$  at which  $s$  **fits** the input image  $f$ .

For all pixel coordinates  $(x, y)$ ,  $g(x, y) = 1$  if  $s$  fits  $f$  and 0 otherwise.



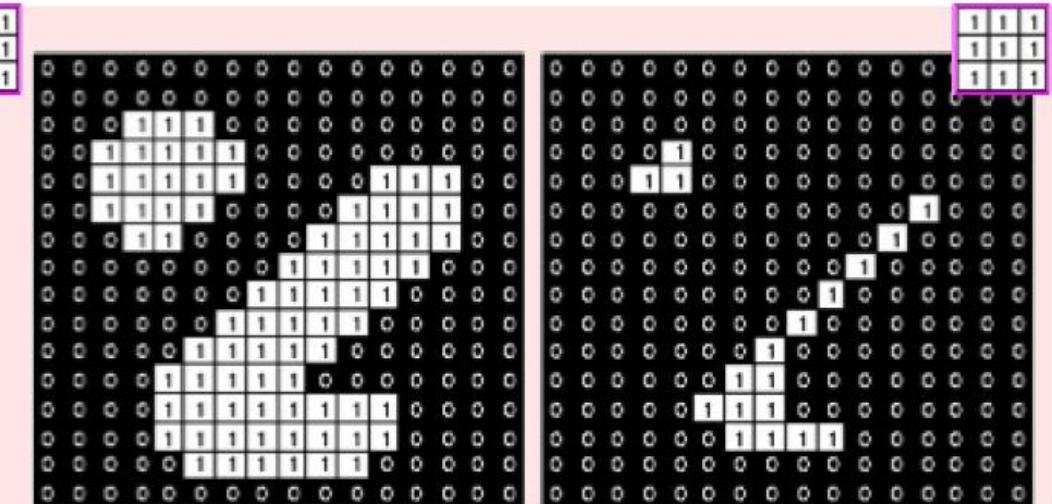
# Erosion examples



Binary image



Eroded image



- Erosion with small (e.g.  $2 \times 2 - 5 \times 5$ ) square SEs shrinks an image by stripping away a layer of pixels from both inner and outer region boundaries.
- Holes in and gaps between foreground objects become larger, and small details are eliminated.



# Erosion of a binary image

**Larger SEs have a more pronounced effect.**

- Erosion with a large SE is similar to an iterated erosion with a smaller SE of the same shape.
- If a pair of SEs  $s_1$  and  $s_2$  are identical in shape, with  $s_2$  twice the size of  $s_1$ , then  $f \ominus s_2 \approx (f \ominus s_1) \ominus s_1$ .

**Erosion removes fine (small-scale) details, as well as noise from a binary image.**

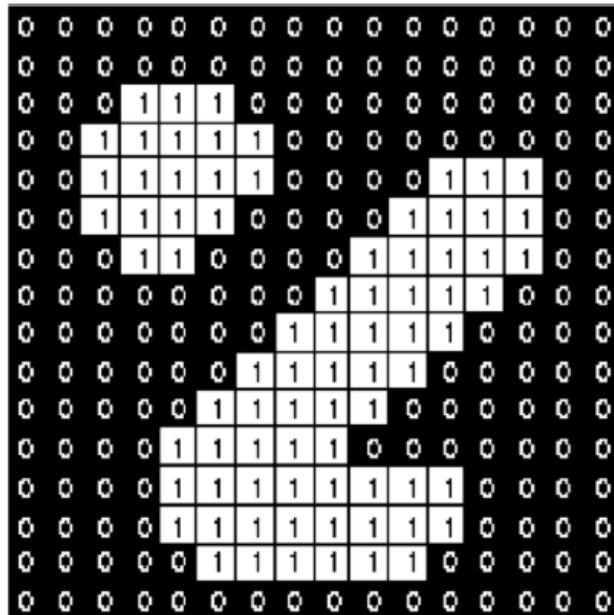
- Simultaneously, erosion reduces the size of regions of interest (objects), too.
- Background areas are growing, i.e. an image with the black or white background becomes blacker or whiter, respectively.

# Boundary detection by erosion

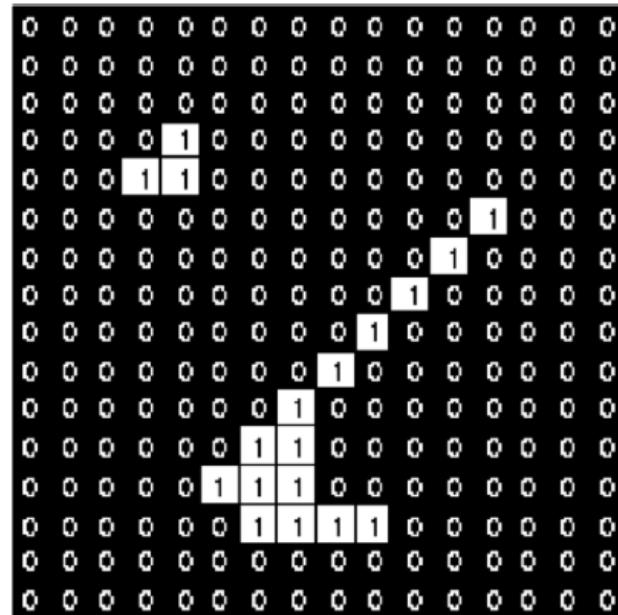
**Internal gradient**  $b = f - (f \ominus s)$  of each region:

by subtracting the eroded image from an original image.

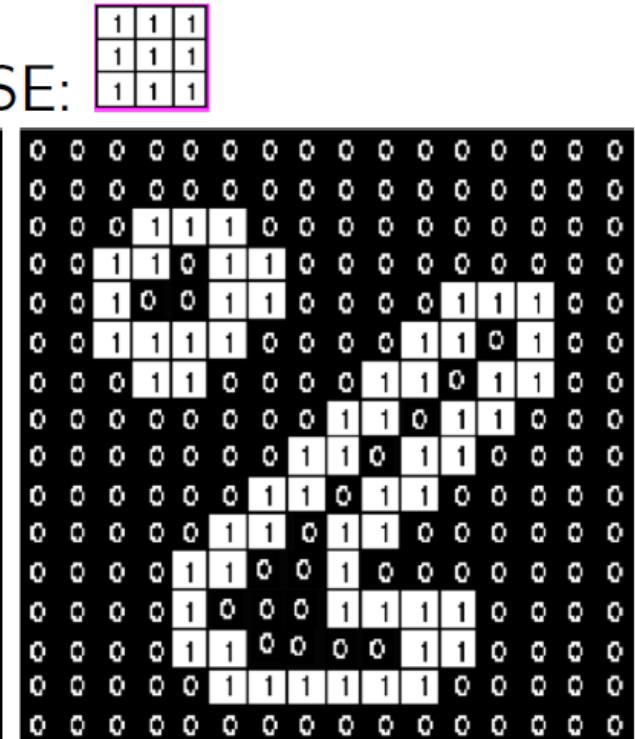
Gradient is computed with the  $3 \times 3$  square SE:



Binary image  $f$



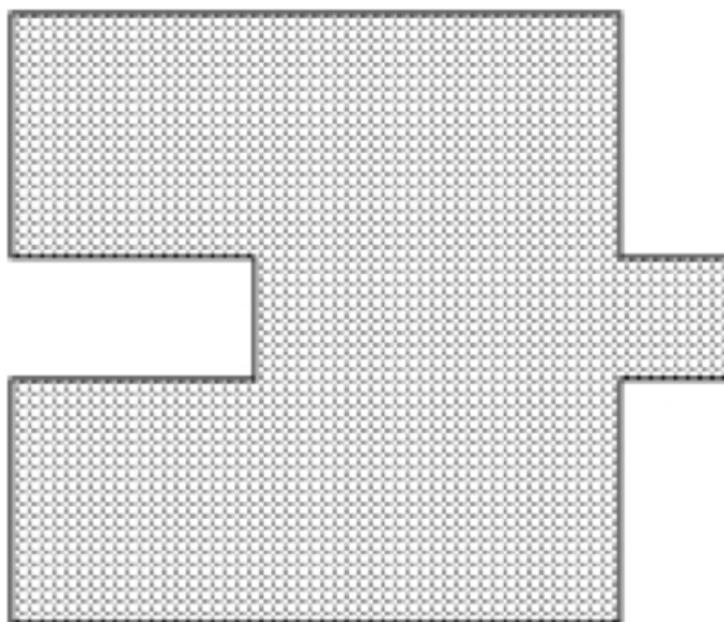
Eroded image  $f \ominus s$



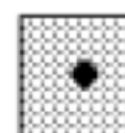
Boundary  $b = f - (f \ominus s)$

# Boundary detection by erosion

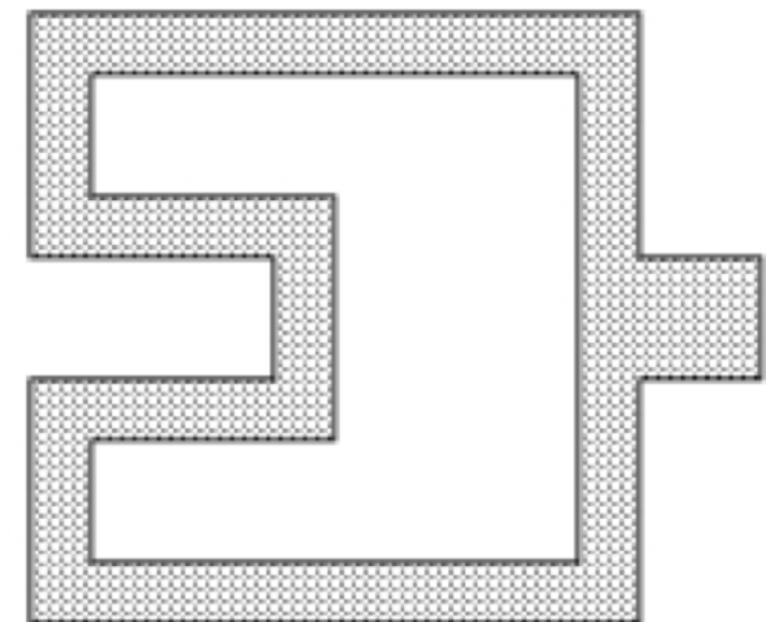
**Internal gradient** of each region:



$f$



$s$



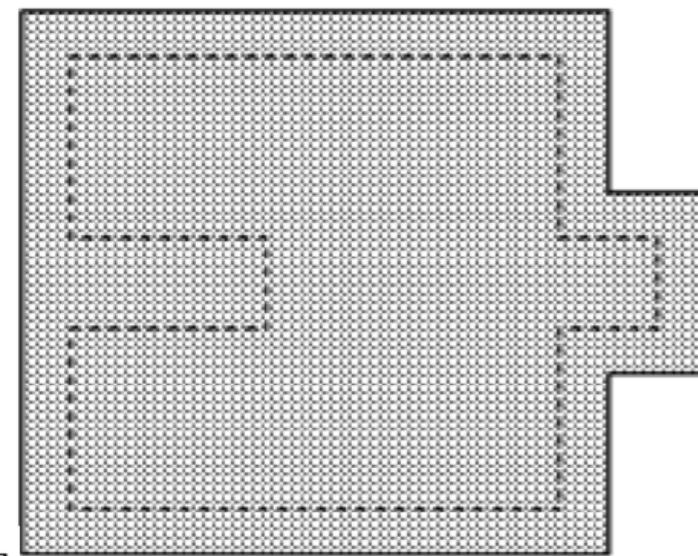
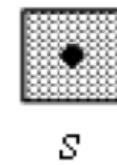
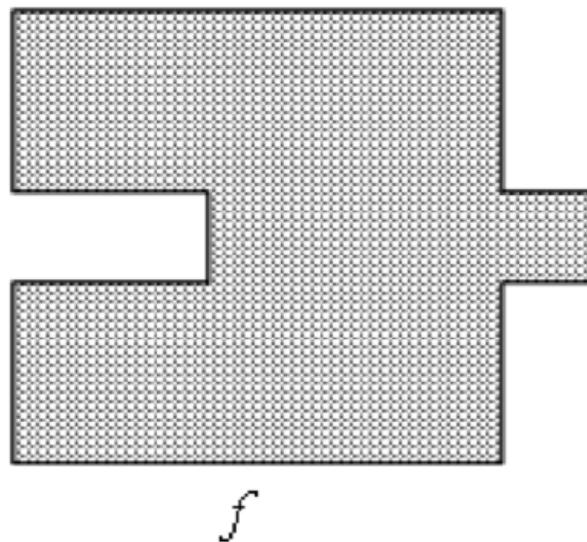
$f - (f \ominus s)$

# Fundamental operation 2: Dilation

Dilation  $f \oplus s$  of a binary image  $f$  by an SE  $s$  produces a new binary image  $g = f \oplus s$ .

Dilated image  $g$  has ones in all locations  $(x, y)$  of an origin of the SE  $s$  at which  $s$  hits the input image  $f$ .

For all pixel coordinates  $(x, y)$ ,  $g(x, y) = 1$  if  $s$  hits  $f$  and 0 otherwise.



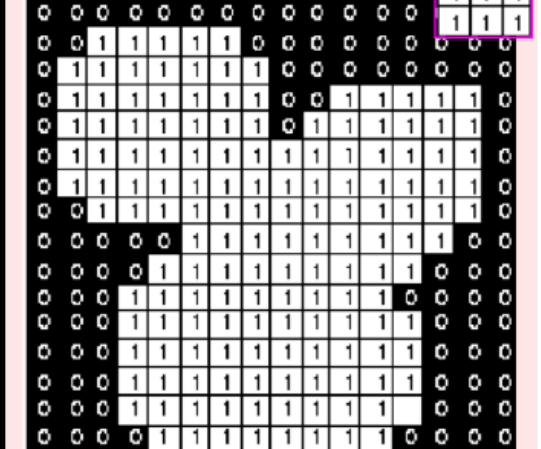
# Dilation examples



## Binary image



## Dilated image



- Dilation is opposite to erosion: it adds a layer of pixels to the inner and outer boundaries of regions.
  - Holes enclosed by a single region and gaps between different regions become smaller; small intrusions into boundaries of a region are filled in.



# Set-theoretic binary operations

- Many morphological operations are combinations of erosion, dilation, and simple set-theoretic operations.
  - Set-theoretic **complement** of a binary image:

$$f^c(x, y) = \begin{cases} 1 & \text{if } f(x, y) = 0 \\ 0 & \text{if } f(x, y) = 1 \end{cases}$$

**complement**

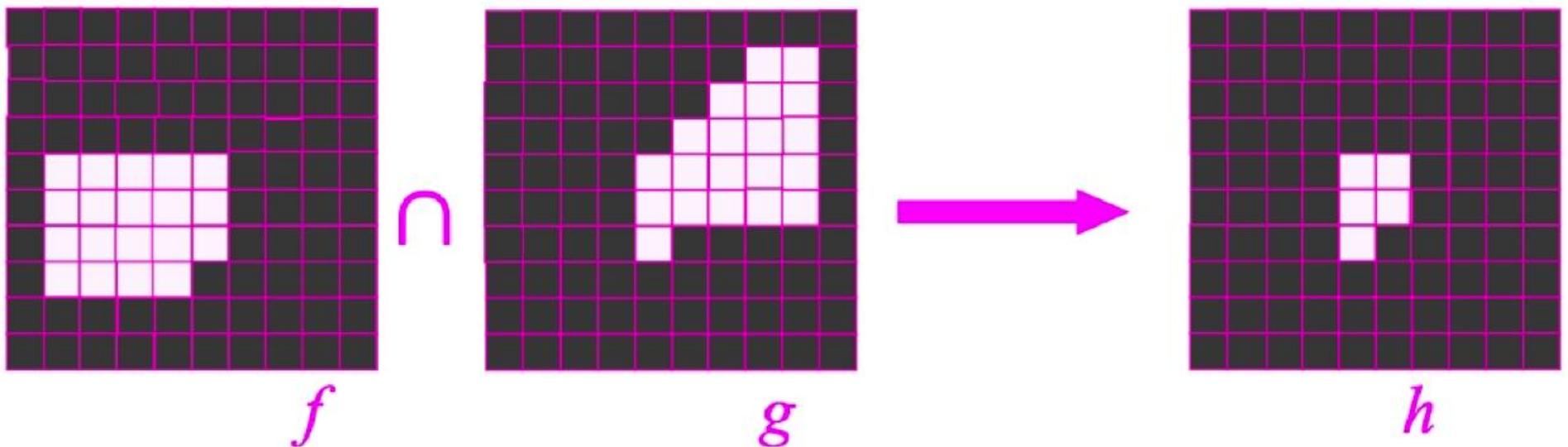
f

f<sup>c</sup>

# Set-theoretic operations: intersection

**Intersection**  $h = f \cap g$  of two binary images  $f$  and  $g$ :

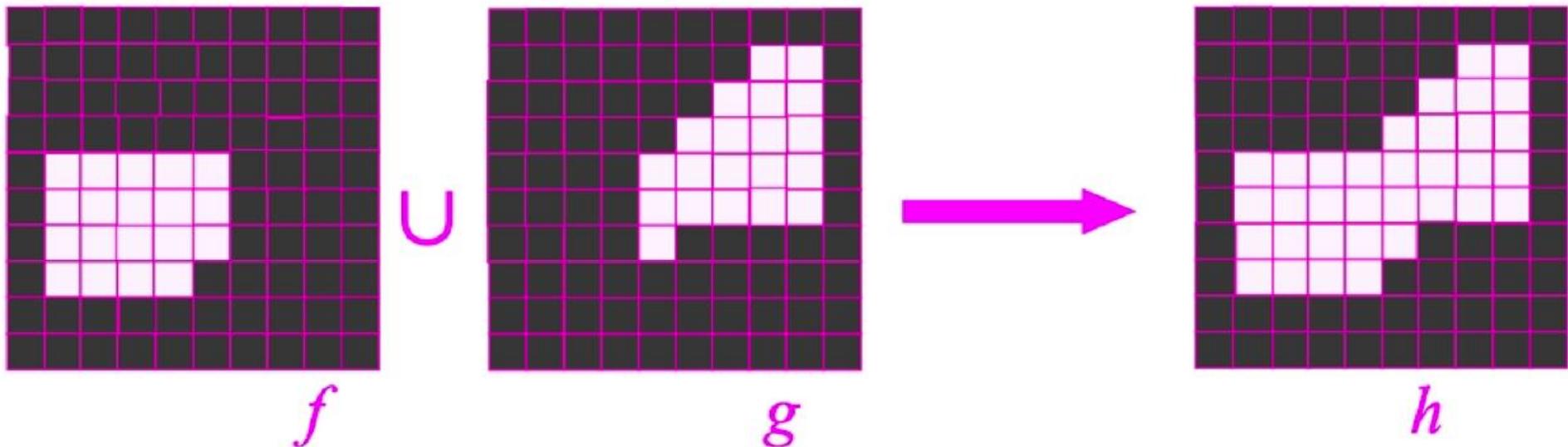
$$h(x, y) = \begin{cases} 1 & \text{if } f(x, y) = 1 \text{ AND } g(x, y) = 1 \\ 0 & \text{otherwise} \end{cases}$$



# Set-theoretic operations: union

**Union**  $h = f \cup g$  of two binary images  $f$  and  $g$ :

$$h(x, y) = \begin{cases} 1 & \text{if } f(x, y) = 1 \text{ OR } g(x, y) = 1 \\ 0 & \text{otherwise} \end{cases}$$



# Duality of dilation and erosion

Dilation and erosion are **dual operations** in that they have opposite effects:  $f \oplus s = f^c \ominus s_{\text{rot}}$ .

- $f^c$  – the complement of  $f$  produced by replacing “1”s / “0”s with “0”s / “1”s.
- $s_{\text{rot}}$  – the SE  $s$  rotated by  $180^\circ$ .
  - If a SE is symmetric with respect to rotation, then  $s_{\text{rot}}$  does not differ from  $s$ .

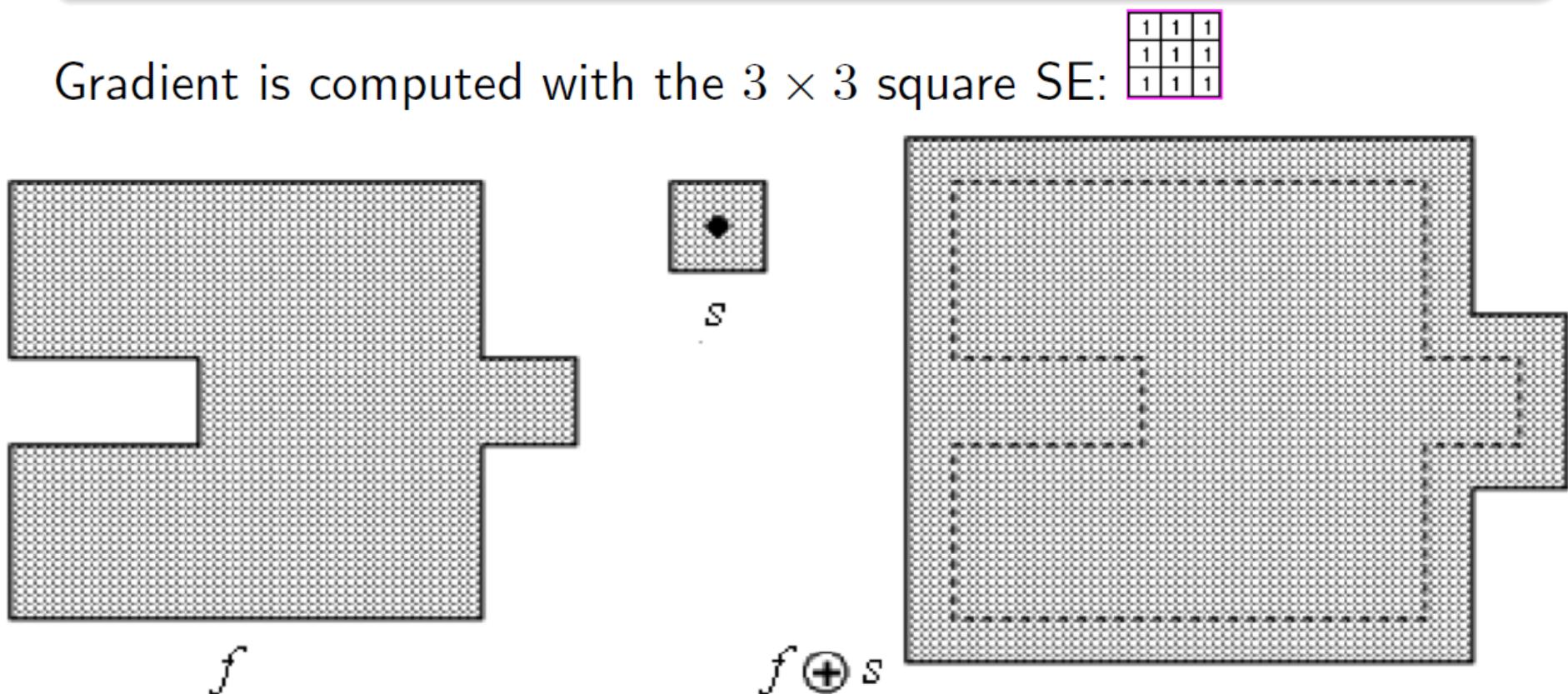
If a binary image is considered as a collection of connected regions of “1”s on background of “0”s:

- **Erosion** is the fitting of the SE to these regions.
- **Dilation** is the fitting of the SE, rotated if necessary, into the background, followed by inversion of the result.

# Boundary detection by dilation

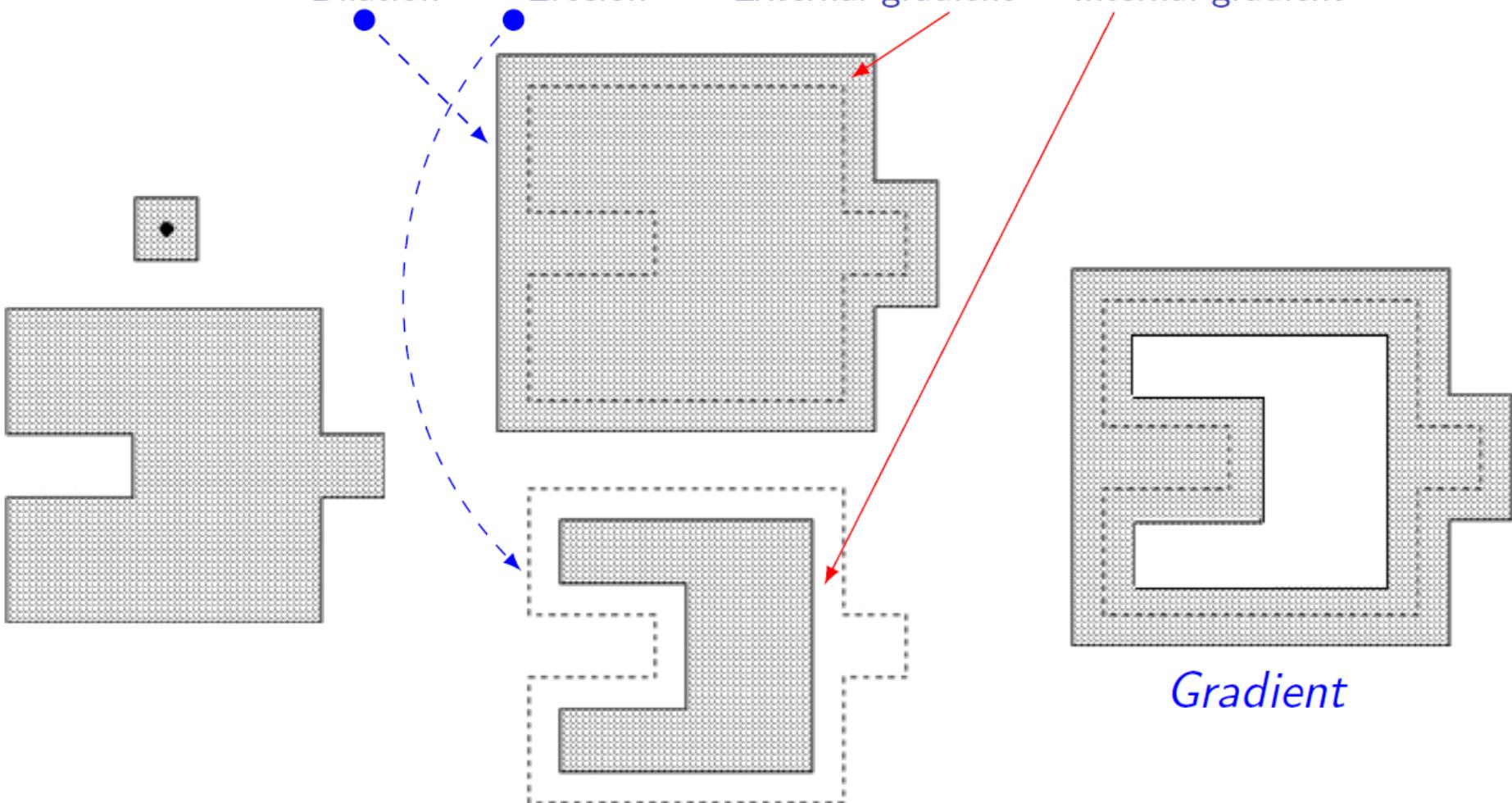
**External gradient**  $b = (f \oplus s) - f$  of each region:

by subtracting an original image from the dilated image.



# Morphological gradient = Dilatation - Erosion

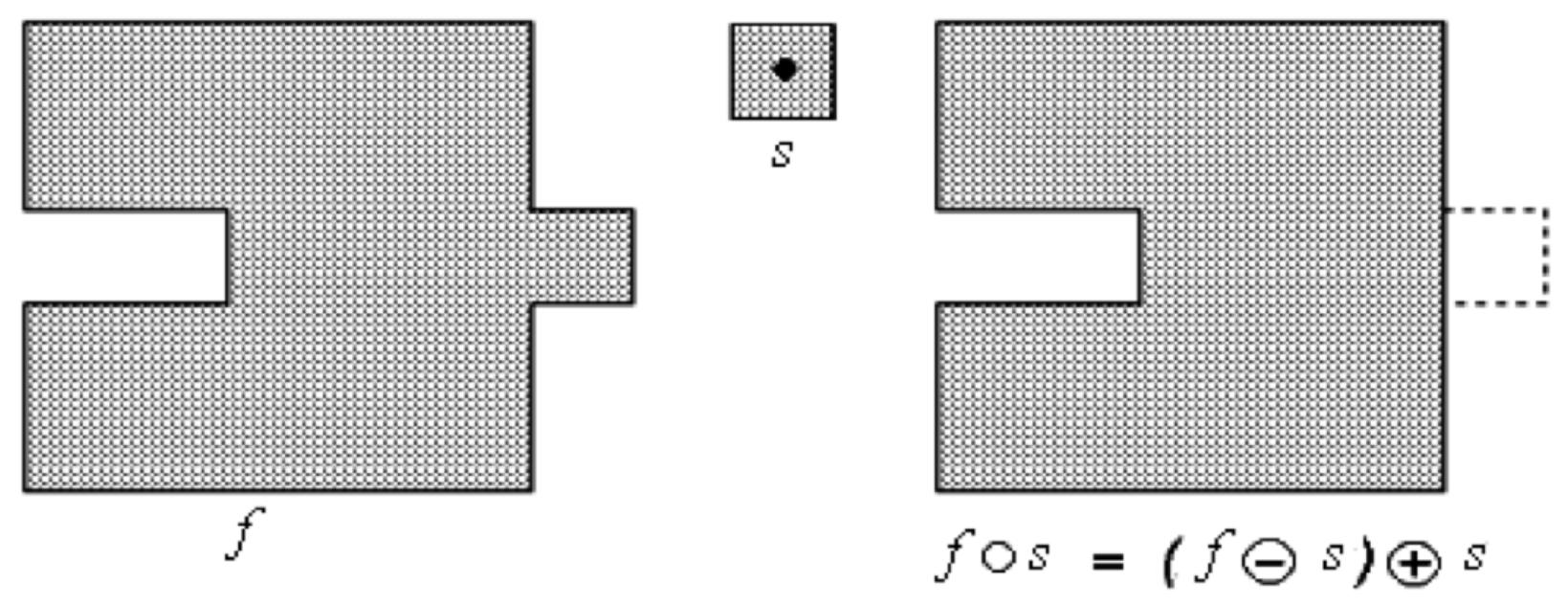
$$\nabla f = \underbrace{(f \oplus s)}_{\text{Dilation}} - \underbrace{(f \ominus s)}_{\text{Erosion}} \equiv \underbrace{[(f \oplus s) - f]}_{\text{External gradient}} + \underbrace{[f - (f \ominus s)]}_{\text{Internal gradient}}$$



# Opening

Opening  $f \circ s$  of an image  $f$  by a structuring element  $s$  is an **erosion** followed by a **dilation**:

$$f \circ s = (f \ominus s) \oplus s$$



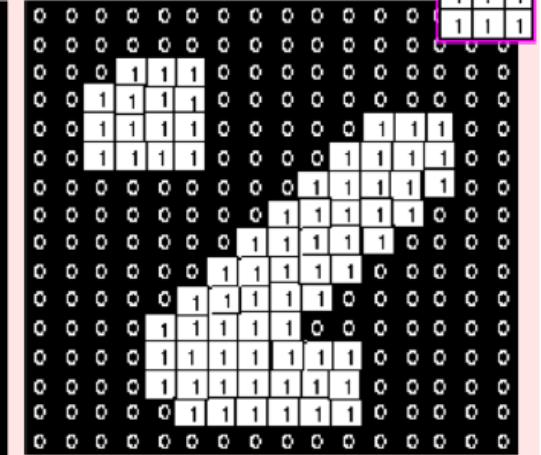
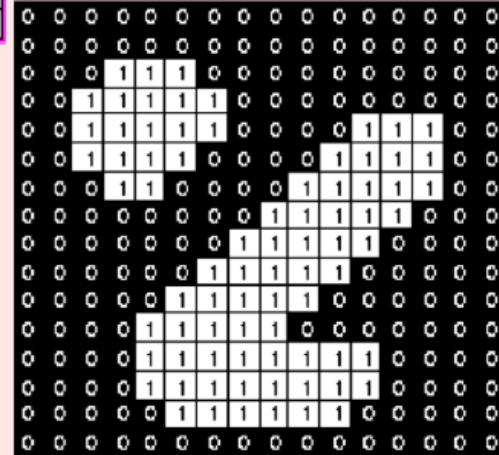
# Opening examples



## Binary image

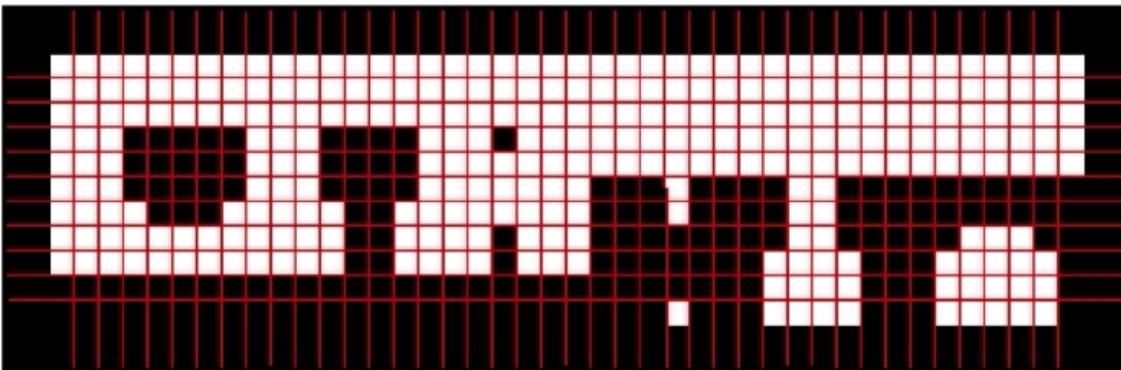


# Opened image

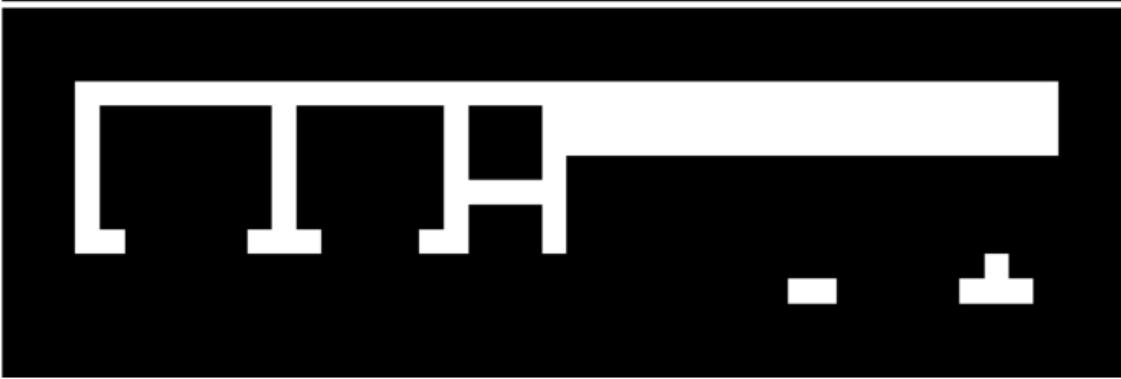


Opening is so called because it can open up a gap between objects connected by a thin bridge of pixels.

# Opening examples



Initial image  $f$ :  
white "1"s and black "0"s



SE  $s$  

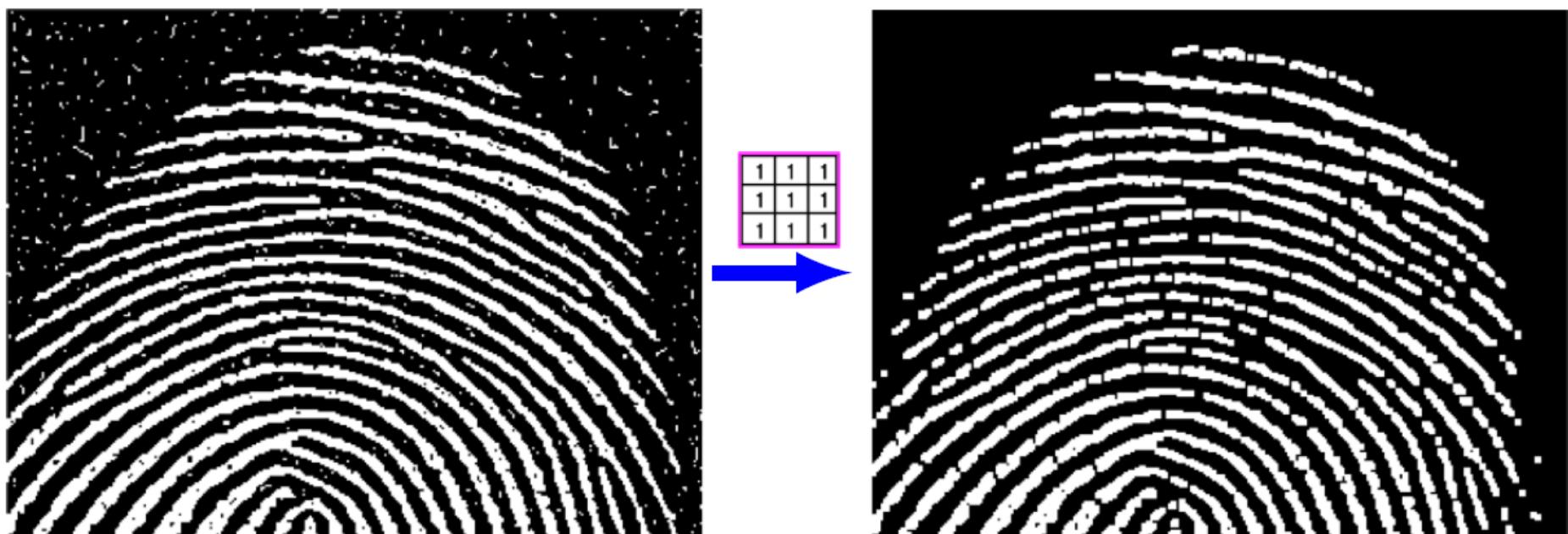
Erosion  $f \ominus s$



Opening  $f \circ s = (f \ominus s) \oplus s$

# Opening a binary image

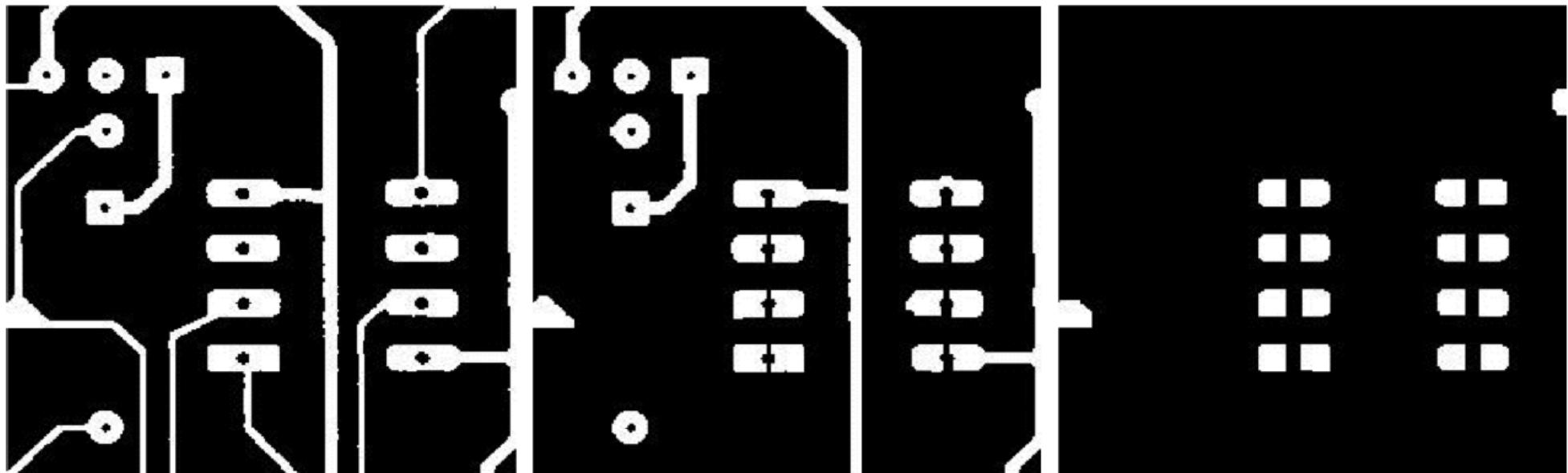
Any regions that survived the erosion are restored to their original size by the dilation:



**Idempotent operation:**  $(f \circ s) \circ s = f \circ s$

- Once an image is opened, next openings with the same structuring element have no further effect.

# Opening a binary image



Binary image

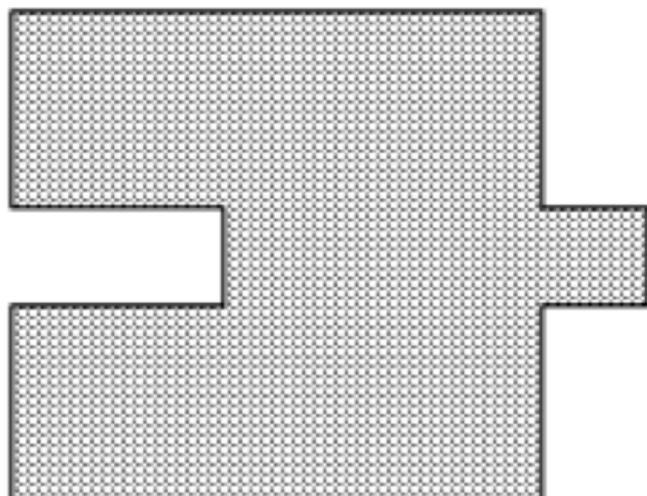
Opening with  
a  $5 \times 5$  SE

Opening with  
a  $9 \times 9$  SE

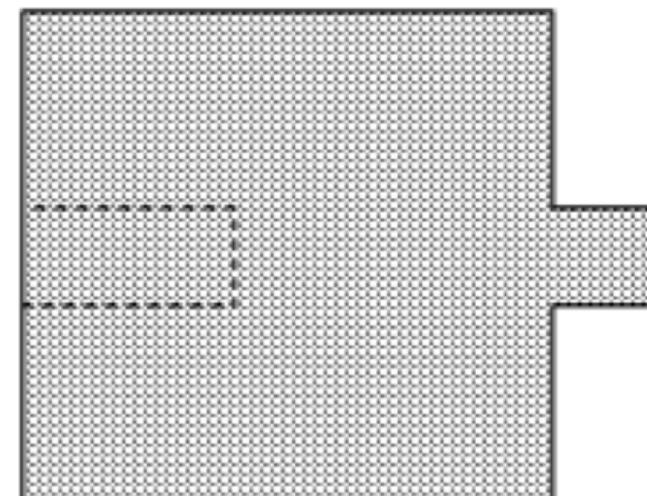
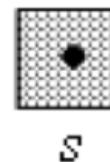
# Closing

Closing  $f \bullet s$  of an image  $f$  by a structuring element  $s$  is a **dilation** followed by an **erosion**:

$$f \bullet s = (f \oplus s) \ominus s$$



$f$



$f \bullet s = (f \oplus s) \ominus s$

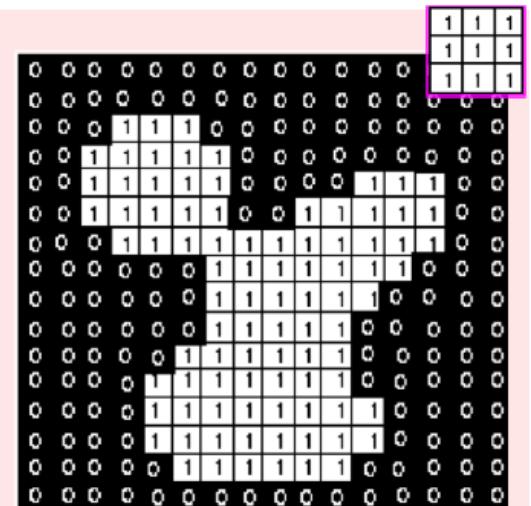
# Closing examples



## Binary image

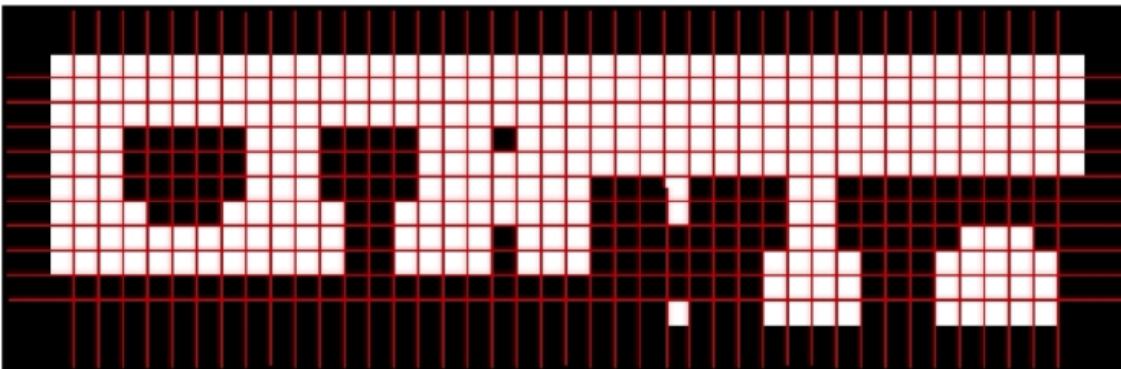


## Closed image



Closing is so called because it can fill holes in the regions while keeping the initial region sizes.

# Closing examples



Initial image  $f$ :  
white "1"s and black "0"s



SE  $s$  

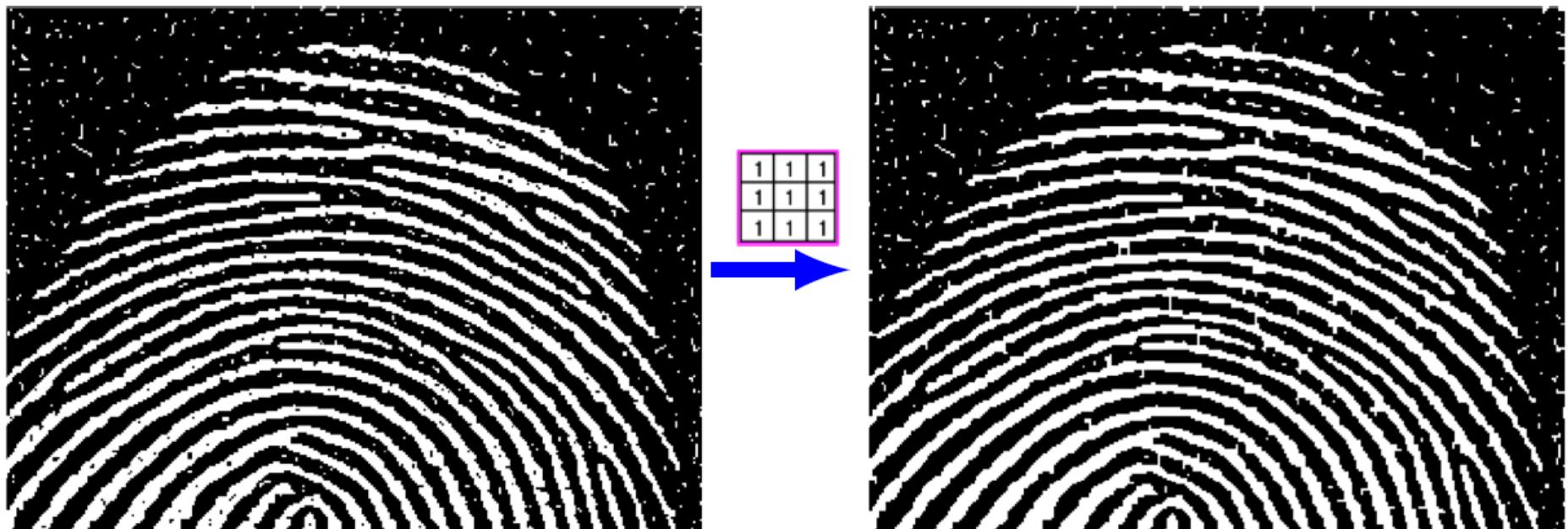
Dilation  $f \oplus s$



Closing  $f \circ s = (f \oplus s) \ominus s$

# Closing a binary image

Any holes or channels smaller than the structuring element are filled:



**Idempotent operation:**  $(f \bullet s) \bullet s = f \bullet s$

- Once an image is closed, next closings with the same structuring element have no further effect.

# Closing vs. opening

Closing is the **dual operation** of opening.

- Just as opening is the dual operation of closing.

**Closing** of a binary image (*dual implementation*):

- Take the **complement** of that image (“1 / 0”  $\Rightarrow$  “0 / 1”).
- Perform **opening** with the structuring element.
- Take the **complement** of the result.

**Opening** of a binary image (*dual implementation*):

- Take the **complement** of that image (“1 / 0”  $\Rightarrow$  “0 / 1”).
- Perform **closing** with the structuring element.
- Take the **complement** of the result.

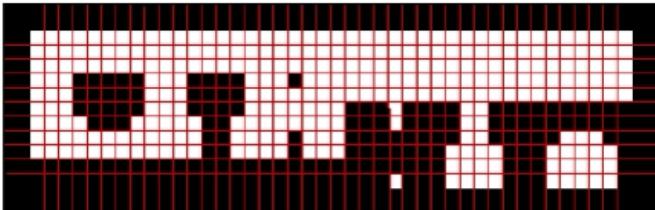
# Closing vs. opening

## **Closing with a square or disk SE:**

- Fills thin connections within an object.
- Eliminates small holes and fills dents in contours.
- Fills small gaps in parts of an object.

## **Opening with a square or disk SE:**

- Breaks thin connections within an object.
- Eliminates small islands and sharp protrusions.



# Binary image: closing and opening



Initial image  $f$



Closed, then opened image  
 $g = (f \bullet s) \circ s$

# Binary image: opening and closing



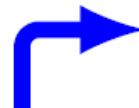
Initial image  $f$



Opened, then closed image  
 $g = (f \circ s) \bullet s$

# Opening + closing vs. closing + opening

$\underbrace{\text{erode} - \text{dilate}}_{\text{open}} - \underbrace{\text{dilate} - \text{erode}}_{\text{close}}$





Opened



Closed

$\underbrace{\text{dilate} - \text{erode}}_{\text{close}} - \underbrace{\text{erode} - \text{dilate}}_{\text{open}}$



Opened + Closed



Closed + Opened