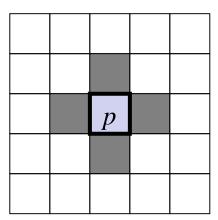
Binary Images Processing and Mathematical Morphology

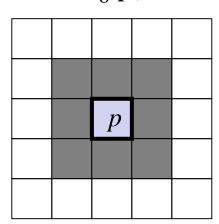
Relations Between Pixels

Neighbors and Neighborhood:

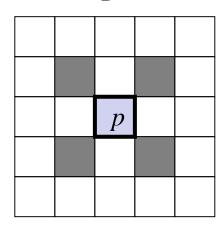




 $N_8(p)$:



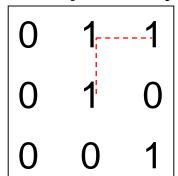
 $N_D(p)$:



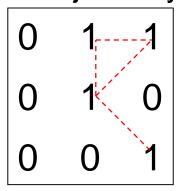
Adjacency (for samevalue pixels):

Examples for "1" pixels:

4-adjacency



8-adjacency

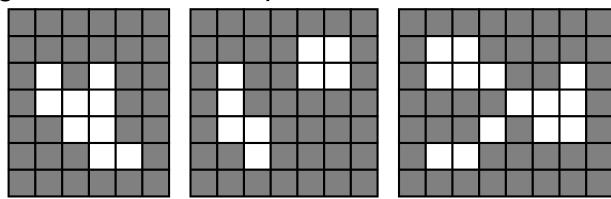


Connected Components (Regions)

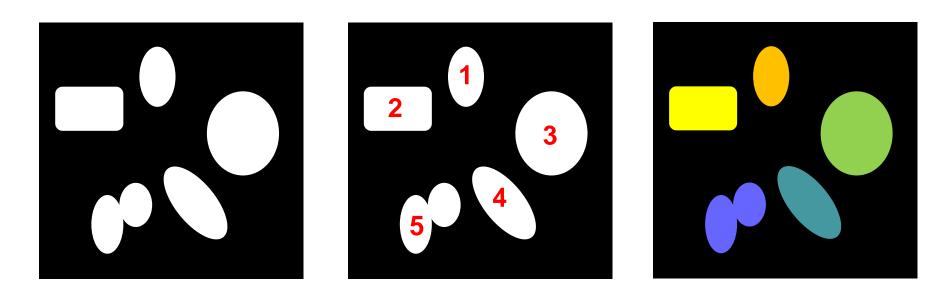
Definitions (the simple version):

- Path: A list of pixels, with each pixel <u>adjacent</u> to the next one.
- Two pixels p and q are connected if there exists a path connecting p and q.
- A connected component is a subset of pixels such that all the pixels in it are connected to one another.

How many connected components of white pixels are in these images? Note: This depends on the connectivity.



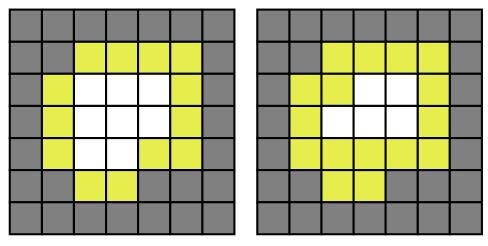
Connected Component Labeling



This is the process of identify separate foreground objects in a binary image. Each connected component is given an index or label.

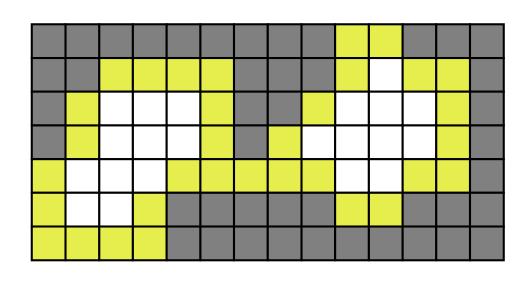
Region Boundary

The **boundary** of a region R is the subset of pixels that (1) belong to R, and (2) have neighbors that are not in R.



These two are based on different adjacencies. Can you tell?

Pixels of a region R that are located on the image border are automatically considered part of the boundary of R.



Morphological Image Processing

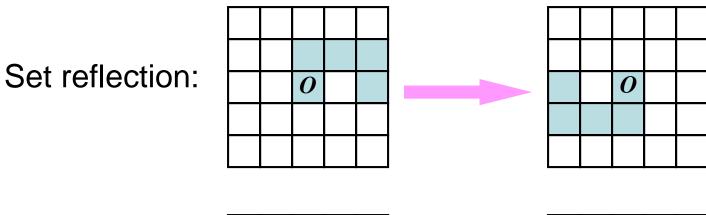
- Morphology is the study of forms and structures, such as in biology.
- Mathematical morphology: How we represent and operate on "shapes". We do this by treating regions of pixels as <u>sets</u>.
- The main practical applications of morphological operations in image processing are mainly in preprocessing and earlystage segmentation.
- The main discussions are focused on binary images, with a little extensions to gray-scale images.

Set Operations

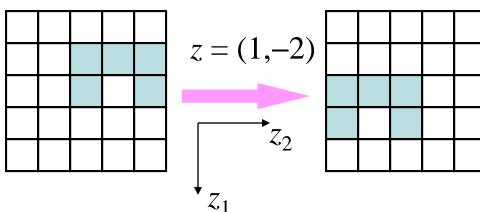
- Basic set relations and operations:
 - Empty (null) set
 - Subset
 - Intersection and Union
 - Complement
 - Set difference
- Specific set operations (for elements on a grid) that we need:
 - Set reflection: $\hat{B} = \{ w \mid w = -b, b \in B \}$
 - Set translation: $(A)_z = \{c \mid c = a + z, a \in A\}$

Set Operations

When we consider the set to contain the <u>shaded</u> pixels in an image:

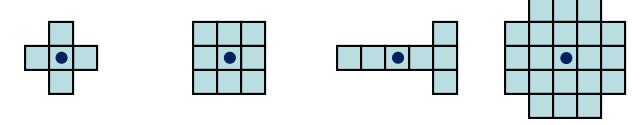


Set translation:

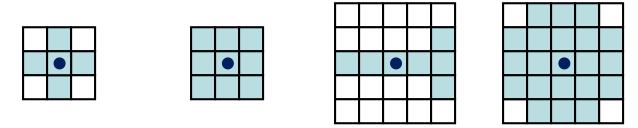


Structuring Element (SE)

Many morphological operations involve the interaction between an image and a **structuring element**, which is usually a small set of points that define a shape. A structuring element can be symmetric or non-symmetric. Examples:



They are often represented as part of a rectangular sub-image which looks like a spatial filter. But the operations are different.



(The ● dots represent the "centers" of the structuring elements.)

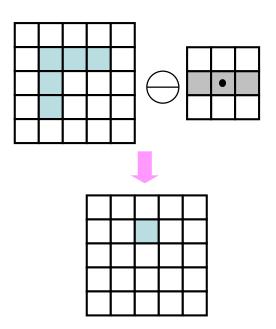
Erosion

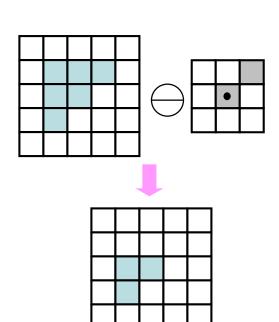
Mathematical definition of erosion:

$$A \ominus B = \{ z \mid (B)_z \subseteq A \} = \{ z \mid (B)_z \cap A^c = \emptyset \}$$

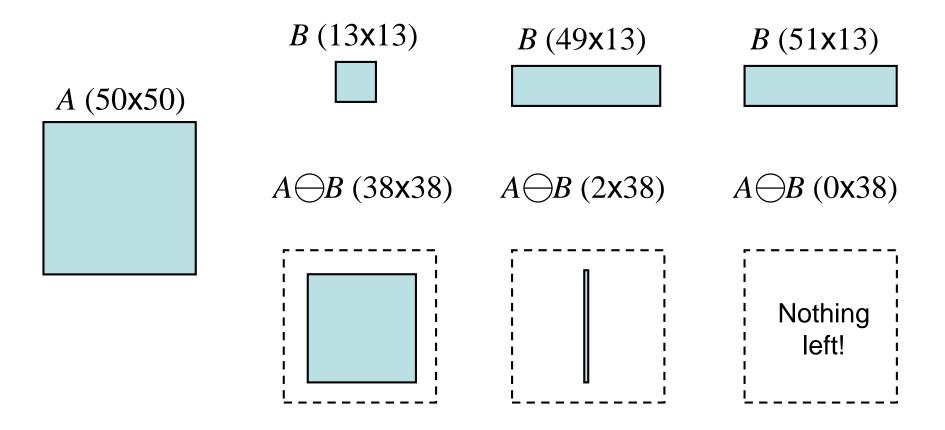
Definition in plain English: The centers of all the translated B (the SE) that are contained in A.

Examples:





Erosion



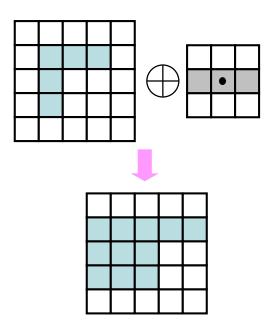
Dilation

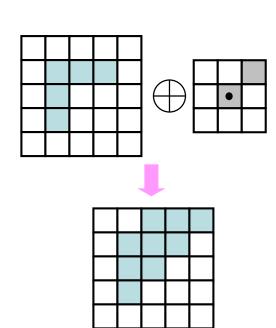
Mathematical definition of dilation:

$$A \oplus B = \{a+b \mid a \in A, b \in B\} = \bigcup_{z \in A} (B)_z$$

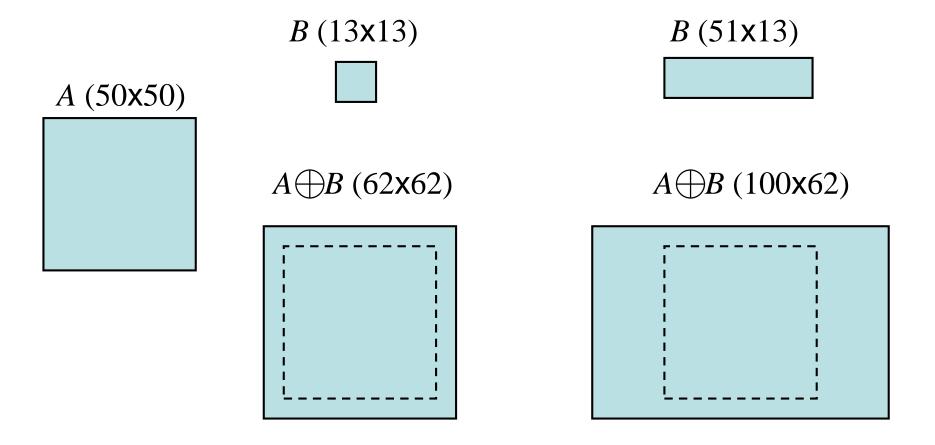
Definition in plain English: The union of all the translated B (the SE) whose centers are contained in A.

Examples:





Dilation



Dilation

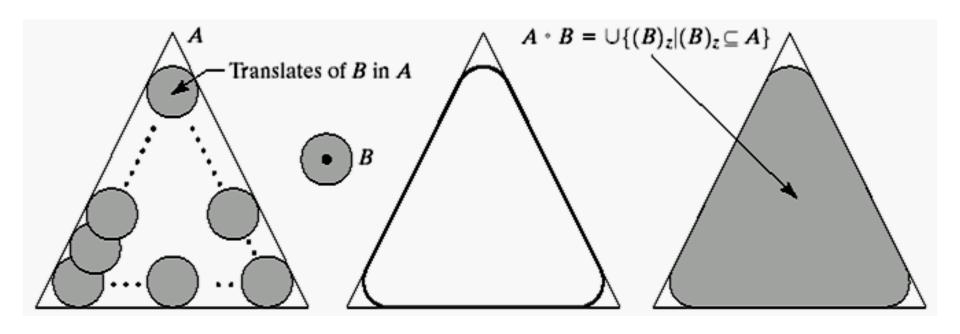
Historically, certain computer programs were written using only two digits rather than four to define the applicable year. Accordingly, the company's software may recognize a date using "00" as 1900 rather than the year 2000.

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

Opening

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

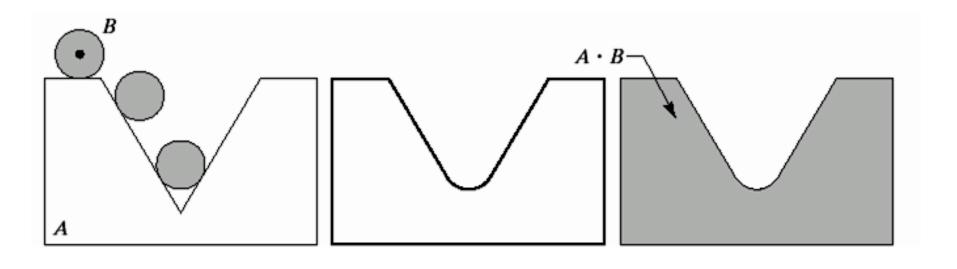


Effect: Preserving the part of *A* where *B* can be fitted in:

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

Closing

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

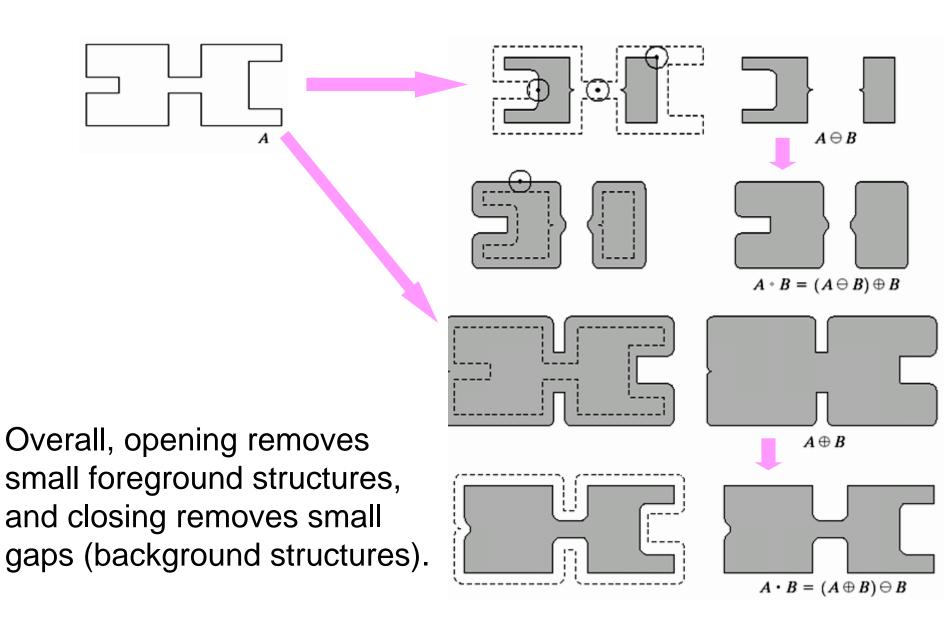


Effect: Preserving the part of A^c where the reflection of B can

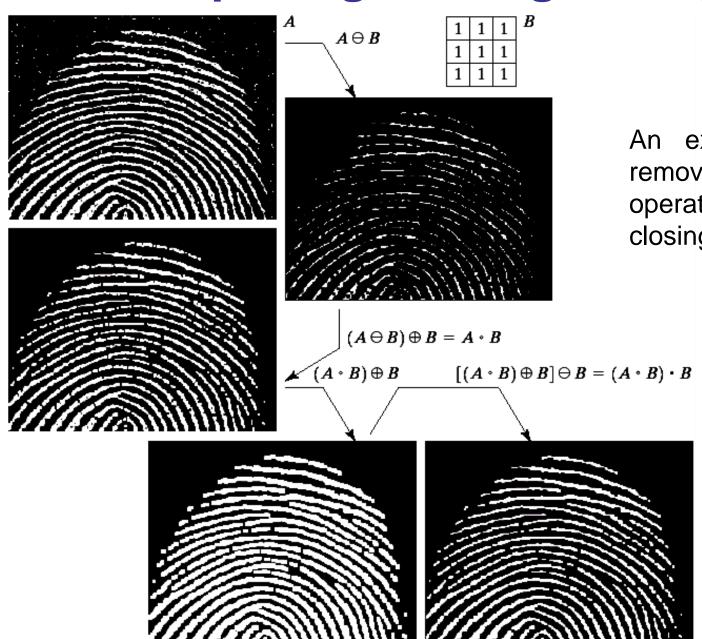
be fitted in:

$$A \bullet B = \left[\bigcup_{(\hat{B})_z \not\subset A} (\hat{B})_z \right]^c$$

Opening/Closing Example



Opening/Closing Example



An example of noise removal with an opening operation followed by a closing operation.

More Properties of Operators

A few additional properties:

$$A \circ B \subseteq A \subseteq A \bullet B$$
 $C \subseteq D \Rightarrow C \circ B \subseteq D \circ B \text{ and } C \bullet B \subseteq D \bullet B$
 $(A \circ B) \circ B = A \circ B$
 $(A \bullet B) \bullet B = A \bullet B$

De Morgan's Laws (Duality):

The erosion of A corresponds to the dilation of A^C , and vice versa.

$$(A \ominus B)^{c} = A^{c} \oplus \hat{B}$$
$$(A \oplus B)^{c} = A^{c} \ominus \hat{B}$$

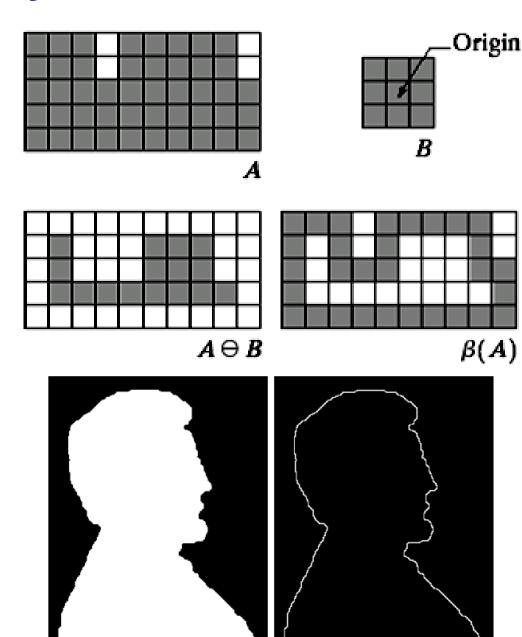
The opening of A corresponds to the closing of A^C , and vice versa.

$$(A \bullet B)^{c} = A^{c} \circ \hat{B}$$
$$(A \circ B)^{c} = A^{c} \bullet \hat{B}$$

Boundary Extraction

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

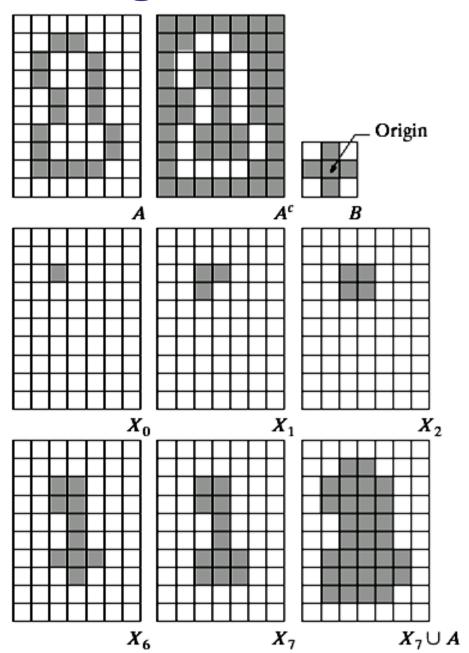
The exact boundary extracted depends on B.



Region Filling

$$X_k = (X_{k-1} \oplus B) \cap A^c$$
$$(X_0 = \{p\})$$

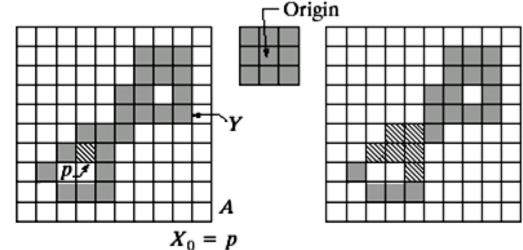
This is an iterative method to fill a region (set to foreground) from a seed point p in A^c . The exact region filled depends on B.

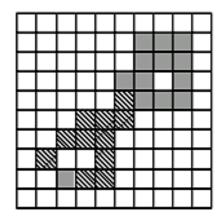


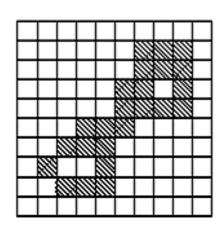
Connected Component Extraction

$$X_k = (X_{k-1} \oplus B) \cap A$$
$$(X_0 = \{p\})$$

This is an iterative method to find a connected component of A that contains a known seed point p in A. The exact connected component that is extracted depends on B.







Morphological Reconstruction

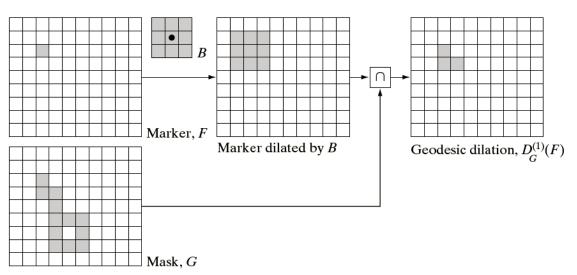
- Two images are involved:
 - Marker image: Where morphological operations are applied.
 - Mask image: This constraints the result of morphological operations.
- The structuring element is only used to define connectivity and is typically isotropic.
- The "shape" information is in the mask image.
- "Region filling" and "connected component extraction" are special cases of morphological reconstruction.

Geodesic Dilation and Erosion

Geodesic Dilation:

$$D_G^{(1)}(F) = (F \oplus B) \cap G$$

$$D_G^{(n)}(F) = D_G^{(1)}[D_G^{(n-1)}(F)]$$

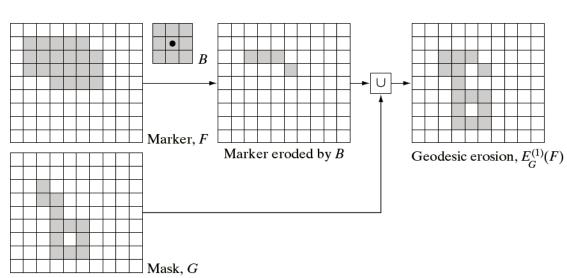


When the result converges, this is called "morphological reconstruction by dilation".

Geodesic Erosion:

$$E_G^{(1)}(F) = (F \ominus B) \cup G$$

$$E_G^{(n)}(F) = E_G^{(1)}[E_G^{(n-1)}(F)]$$



When the result converges, this is called "morphological reconstruction by erosion".

Example – Opening by Reconstruction

Goal: To find characters with long vertical strokes.

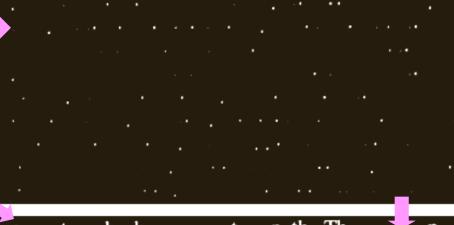
Original (also used as the mask image)

ponents or broken connection paths. There is no point tion past the level of detail required to identify those

Segmentation of nontrivial images is one of the most processing. Segmentation accuracy determines the evon of computerized analysis procedures. For this reason, to be taken to improve the probability of rugged segments such as industrial inspection applications, at least some the environment is possible at times. The experienced in designer invariably pays considerable attention to such

Result of regular opening

Erosion (marker image)

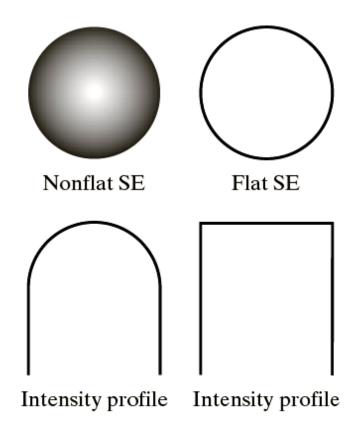


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ptblish tpth The pth light lig
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Morphological reconstruction

Gray-Scale Morphology

- The structuring element becomes 3-D:
 - Two spatial dimensions.
 - One dimension in intensity.

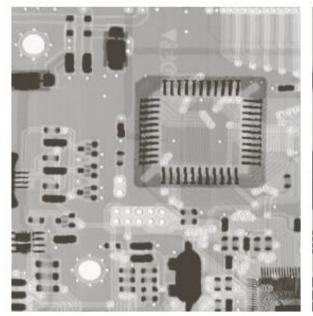


The flat one is used more often.

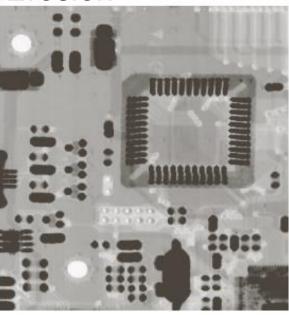
Gray-Scale Morphology with Flat SE

These are actually max and min filters.

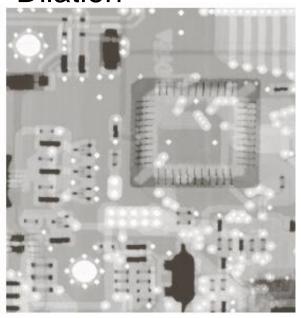
- Gray-scale dilation: $[f \oplus b](x, y) = \max_{(x,y) \in b} \{f(x-s, y-t)\}$
- Gray-scale erosion: $[f \ominus b](x, y) = \min_{(x,y) \in b} \{f(x+s, y+t)\}$



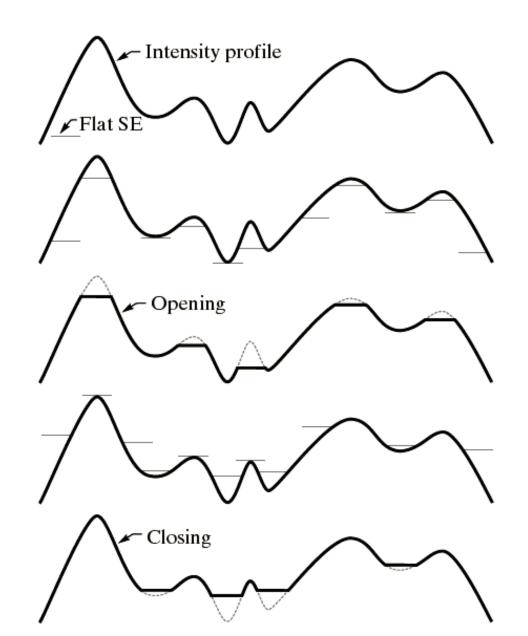




Dilation

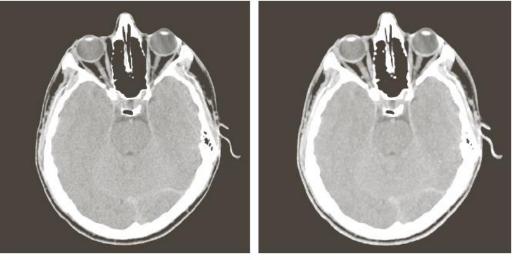


Gray-Scale Opening and Closing



Morphological Gradient

$$g = (f \oplus b) - (f \ominus b)$$



Dilation





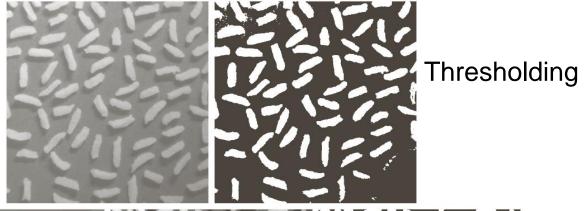


Gradient

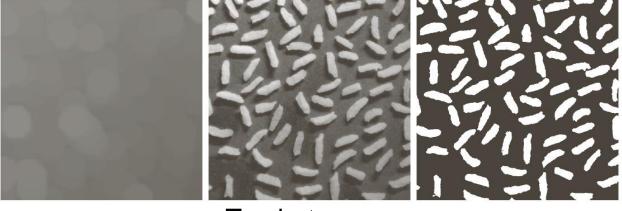
Top-Hat and Bottom-Hat

To keep only small objects (those smaller than the SE):

- Top-hat: Keep light objects: $T_{hat}(f) = f (f \circ b)$
- Bottom-hat: Keep dark objects: $B_{hat}(f) = (f \bullet b) f$



Opening



Top-hat

Top-hat and thresholding