

CSE 554

Lecture 1: Binary Pictures

Fall 2018

Geometric Forms

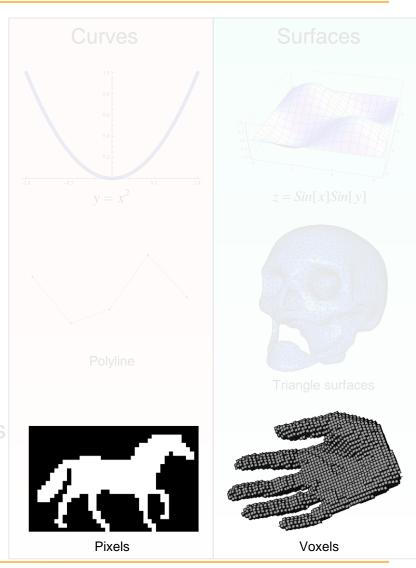


Continuous forms

- Defined by mathematical functions
- E.g.: parabolas, splines, subdivision surfaces

Discrete forms

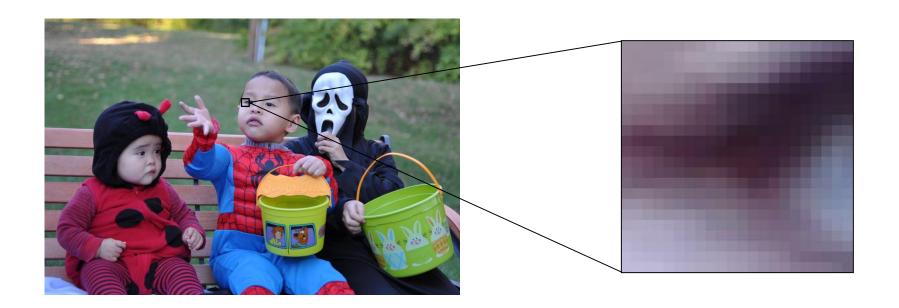
- Disjoint elements with connectivity relations
- E.g.: polylines, triangle surfaces, pixels and voxels



Digital Pictures



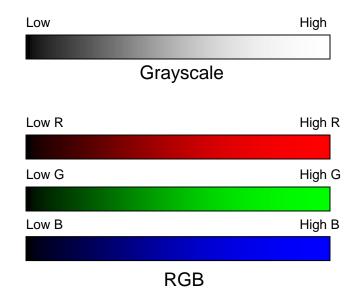
- Made up of discrete points associated with colors
 - Image: 2D array of pixels

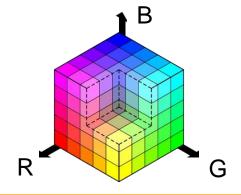


Digital Pictures



- Color representations
 - Grayscale: 1 value representing grays from black (lowest value) to white (highest value)
 - 8-bit (0-255), 16-bit, etc.
 - RGB: 3 values each representing colors from black (lowest value) to pure red, green, or blue (highest value).
 - 24-bit (0-255 in each color)
 - XYZ, HSL/HSV, CMYK, etc.

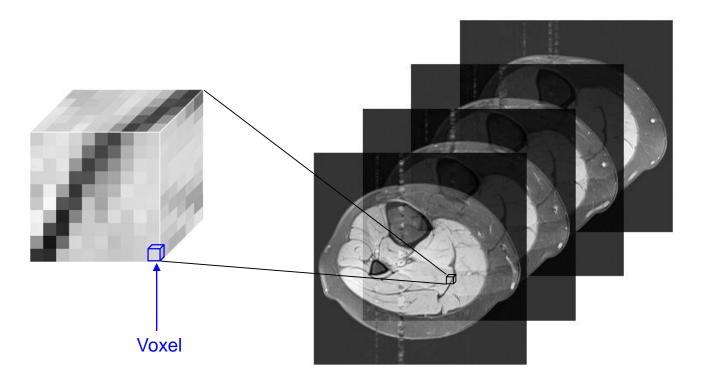




Digital Pictures



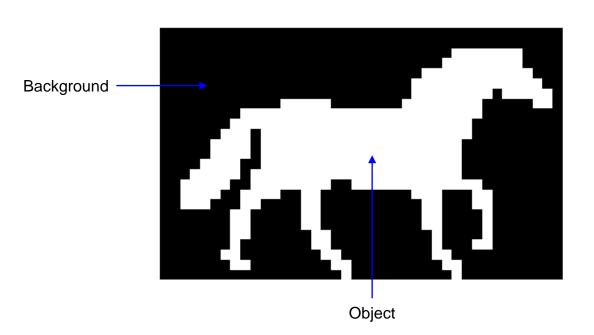
- Made up of discrete points associated with colors
 - Volume: 3D array of voxels



Binary Pictures



- A grayscale picture with 2 colors: black (0) and white (1)
 - The set of 1 or 0 pixels (voxels) is called object or background
 - A "blocky" geometry



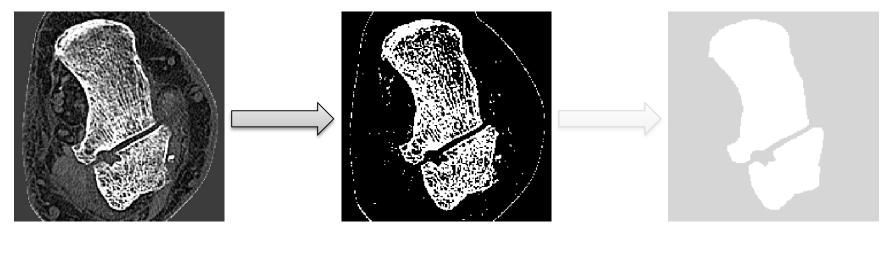




Analogy: Lego, Minecraft

Binary Pictures



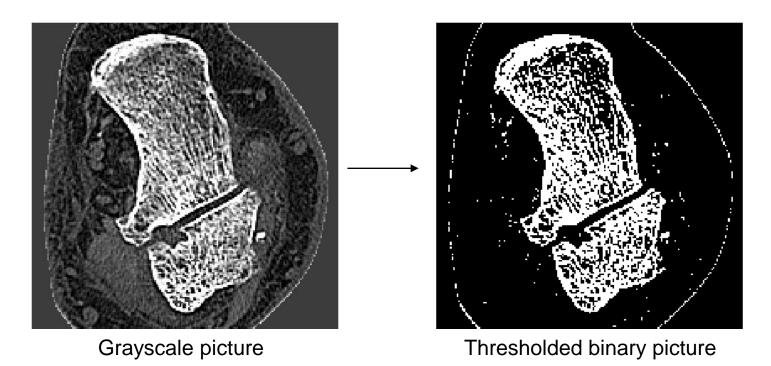


Creation Processing

Segmentation



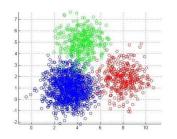
- Separating object from background in a grayscale picture
 - A simple method: thresholding by pixel (voxel) color
 - All pixels (voxels) with color above a threshold is set to 1

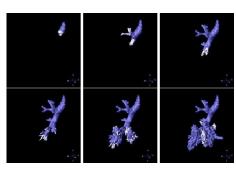


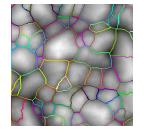
Segmentation



- Separating object from background in a grayscale picture
 - A simple method: thresholding by pixel (voxel) color
 - Other methods:
 - K-means clustering
 - Watershed
 - Region growing
 - Snakes and Level set
 - Graph cut
 - •







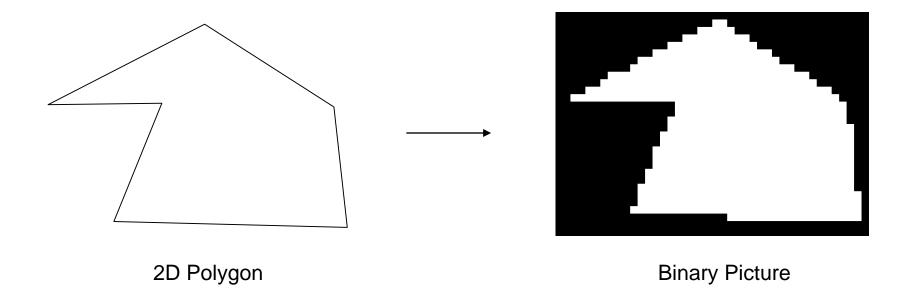


More details covered in Computer Vision course

Rasterization

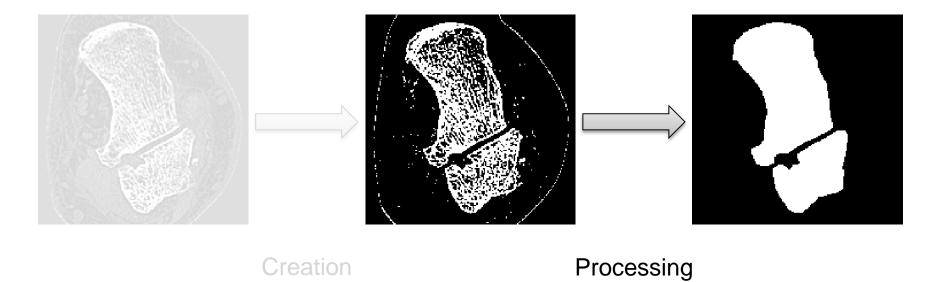


- Filling the interior of a shape by pixels or voxels
 - Known as "scan-conversion", or "pixelization / voxelization"
 - More details covered in Computer Graphics course



Binary Pictures





Binary Pictures





Removing islands and filling holes

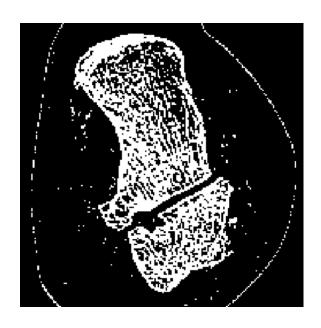
Smoothing boundaries

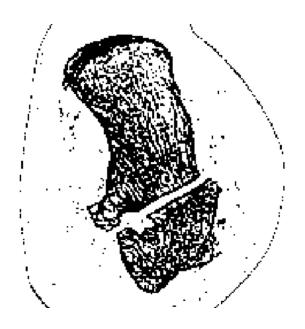
Islands & Holes



Observations:

- Islands (holes) are not as "big" as the object (background).
- Islands (holes) of the object are holes (islands) of the background



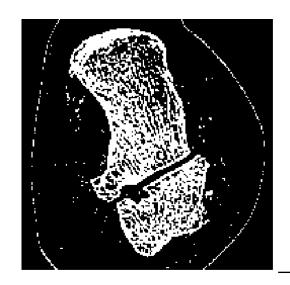


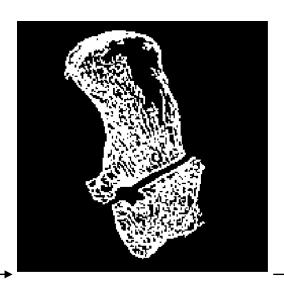
Islands & Holes



Observations:

- Islands (holes) are not as "big" as the object (background).
- Islands (holes) of the object are holes (islands) of the background







Take the largest 2 connected components of the object

Invert the image, take the largest connected component of the object, invert again

Connected Components

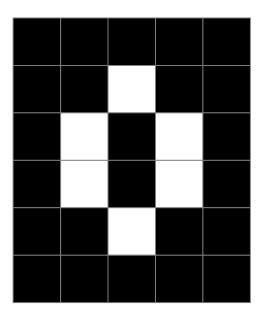


Definition

 A maximum set of pixels (voxels) in the object or background, such that any two pixels (voxels) in the set are connected by a path of connected pixels (voxels)

Connected Components

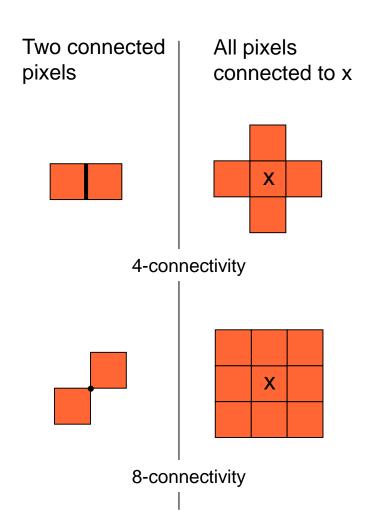




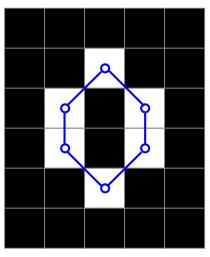
How many connected components are there in the object? What about background?



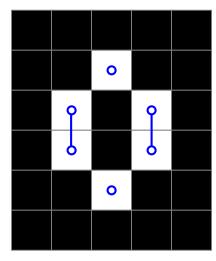
- Two pixels are connected if their squares share:
 - A common edge
 - 4-connectivity
 - A common vertex
 - 8-connectivity







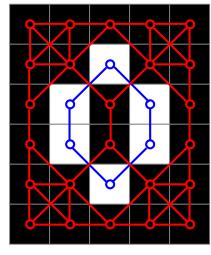
Object: 8-connectivity (1 comp)



Object: 4-connectivity (4 comp)

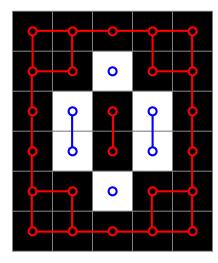


What connectivity should be used for the background?



Object: 8-connectivity (1 comp)

Background: 8-connectivity (1 comp)



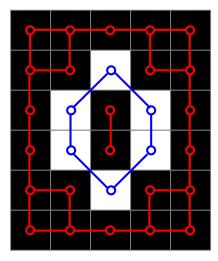
Object: 4-connectivity (4 comp)

Background: 4-connectivity (2 comp)

Paradox: a closed curve does not disconnect the background, while an open curve does.

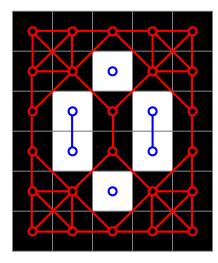


Different connectivity for object (O) and background (B)



Object: 8-connectivity (1 comp)

Background: 4-connectivity (2 comp)

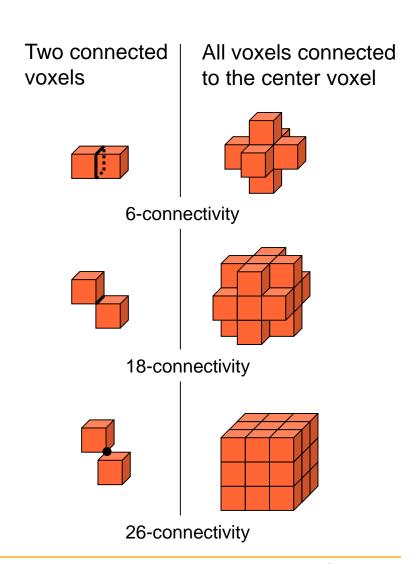


Object: 4-connectivity (4 comp)

Background: 8-connectivity (1 comp)



- Two voxels are connected if their cubes share:
 - A common face
 - 6-connectivity
 - A common edge
 - 18-connectivity
 - A common vertex
 - 26-connectivity
- Use 6- and 26-connectivity respectively for O and B (or B and O)



Finding Connected Components



- The "flooding" algorithm
 - Start from a seed pixel/voxel, expand the connected component
 - Either do depth-first or breadth-first search (a LIFO stack or FIFO queue)

```
// Finding the connected component containing an object pixel p
```

- 1. Initialize
 - 1. Create a result set S that contains only p
 - 2. Create a Visited flag at each pixel, and set it to be False except for p
 - 3. Initialize a queue (or stack) Q that contains only p.
- 2. Repeat until Q is empty:
 - 1. Pop a pixel x from Q.
 - 2. For each unvisited object pixel y connected to x, add y to S, set its flag to be visited, and push y to Q.
- 3. Output S

Finding Connected Components



- Why using a "visited" flag?
 - Otherwise, the program will not terminate
- Why not checking to see if y is in S?
 - Checking the visited flag is much faster (O(1) vs. O(log n))

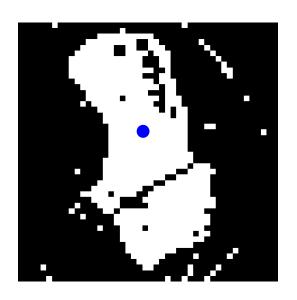
```
1. ...
2. Repeat until Q is empty:

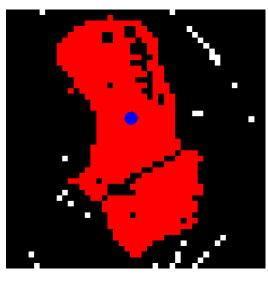
        Pop a pixel x from Q.
         For each unvisited object pixel y connected to x, add y to S, set its flag to be visited, and push y to Q.

3. Output S
```

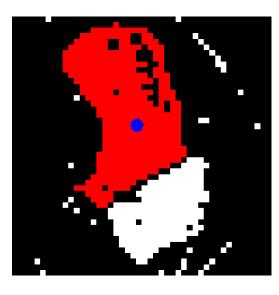


Connected components containing the blue pixel:







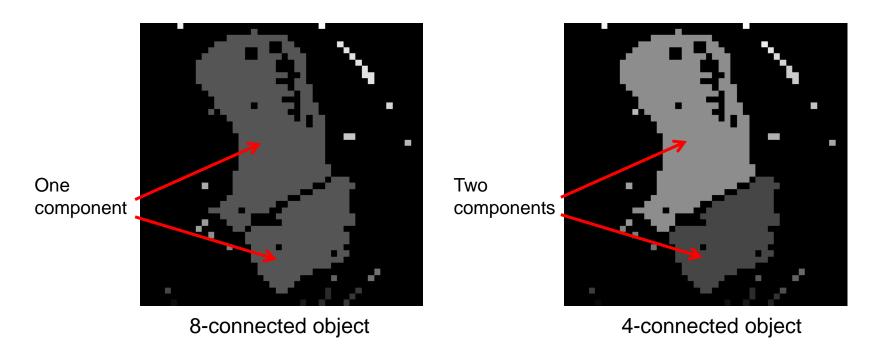


4-connectivity

Finding Connected Components



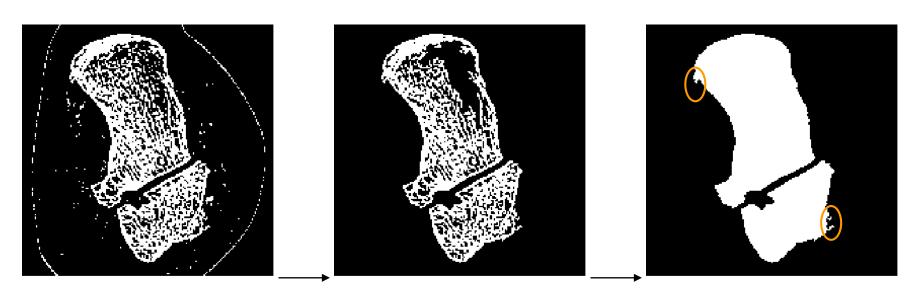
- Labeling all components in an image:
 - Loop through each pixel (voxel). If it is not labeled, use it as a seed to find a connected component, then label all pixels (voxels) in the component.



Using Connected Components



- Pruning isolated islands from the main object
- Filling interior holes of the object



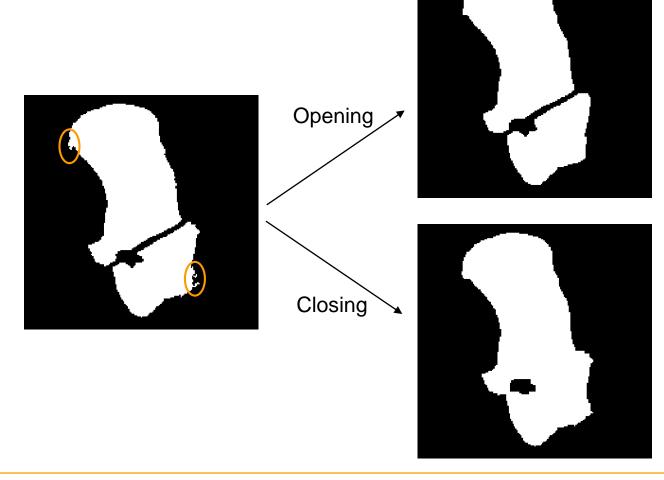
Take the largest components of the object

Invert the largest component of the background

Morphological Operators



Smoothing out object boundary



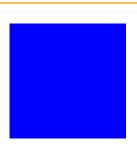
Morphological Operators



- Operations to change shapes
 - Erosion
 - Dilation
 - Opening: first erode, then dilate.
 - Closing: first dilate, then erode.



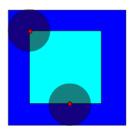
Input:



x

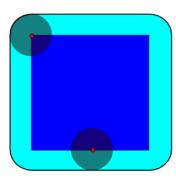
Object (A)

Structure element at $x (B_x)$



Erosion

$$A \ominus B = \{x \in A | B_x \subseteq A\}$$



Dilation

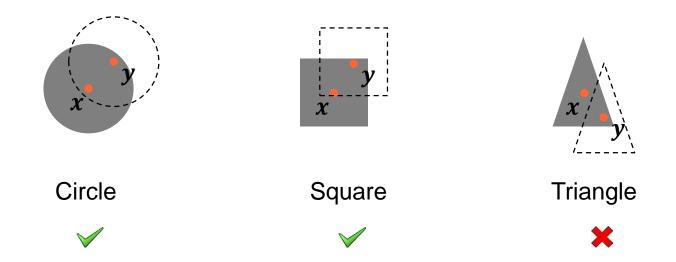
$$A \oplus B = \bigcup_{x \in A} B_x$$



Structure element B is symmetric if:

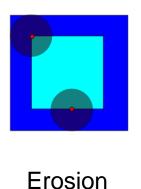
$$x \in B_y \iff y \in B_x$$

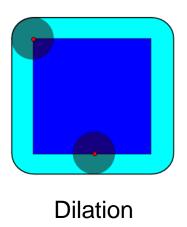
• Examples:





- Duality (for symmetric structuring elements)
 - Erosion (dilation) is equivalent to dilation (erosion) of the background



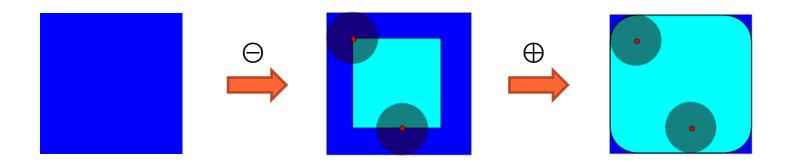


$$A \ominus B = \overline{\overline{A} \oplus B}$$

$$A \oplus B = \overline{\overline{A} \ominus B}$$



Opening (erode, then dilate)

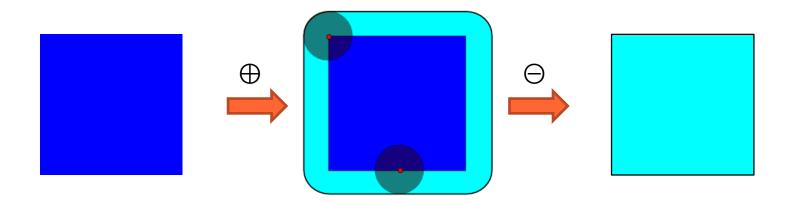


$$A \circ B = (A \ominus B) \oplus B$$

- Union of all structure elements B that can fit inside A
 - Shaves off convex corners and thin spikes



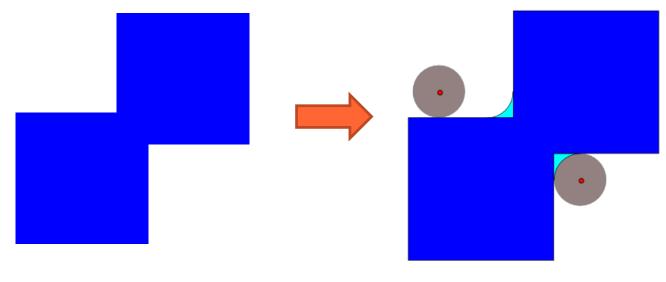
Closing (dilate, then erode)



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



Closing (dilate, then erode)

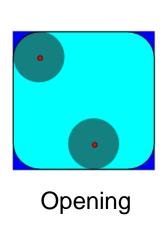


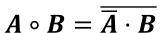
$$A \cdot B = (A \oplus B) \ominus B$$

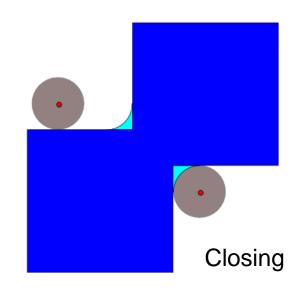
- Complement of union of all B that can fit in the complement of A
 - Fills concave corners and thin tunnels



- Duality, again! (for symmetric structuring elements)
 - Opening (closing) object is equivalent to closing (opening) background



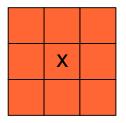


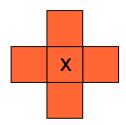


$$A \cdot B = \overline{\overline{A} \circ B}$$

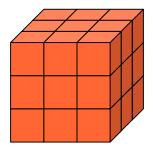


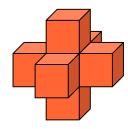
- Structuring elements (symmetric)
 - 2D pixels: square or cross





3D voxels: cube or cross



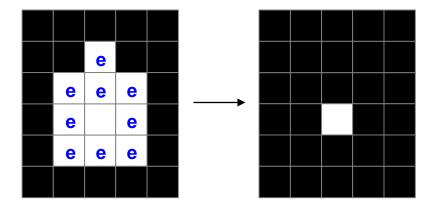




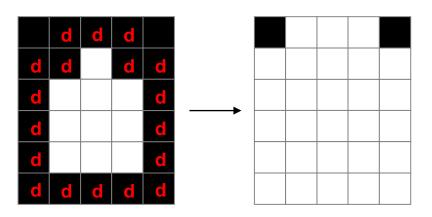
Structuring element:



- Erosion
 - e: an object pixel with some background pixel in its square neighborhood



- Dilation
 - d: a background pixel with some object pixel in its square neighborhood



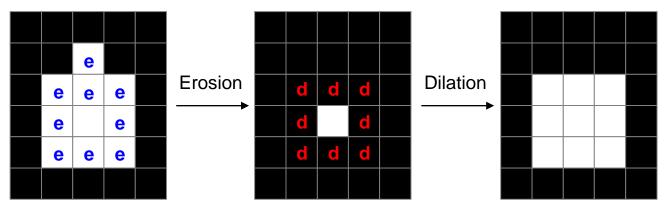


Structuring element:



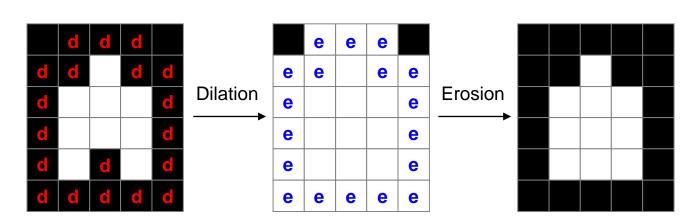
Opening

Union of 3x3 white squares that fit inside object



Closing

Union of 3x3 black squares that fit outside object





- Increasing the size of the structuring element
 - Leads to more growing/shrinking and more significant smoothing



Original



Opening by 3x3 square



Opening by 5x5 square

- Equivalent to repeated applications with a small structuring element
 - E.g.: k erosions (dilations) followed by k dilation (erosions) with a 3x3 square is equivalent to opening (closing) with a (2k+1)x(2k+1) square.



- Implementation tips
 - Using duality of erosion and dilation, you only need to implement one function to do both morphological operations (for symmetric structure elements).
 - Dilation is same as erosion of the background
 - When performing multiple-round opening, make sure you first do k times erosion then k times dilation
 - What happens if you alternate erosion and dilation for k times?
 - Handle image boundary in a graceful way (not crashing the program...)
 - For example, treat outside of the image as background

Lab Module 1



- A simple 2D segmentation routine
 - Initial segmentation using thresholding (using your code from Lab 0)
 - Using connected components and opening/closing to "clean up" the segmentation.

