

EE 584 MACHINE VISION

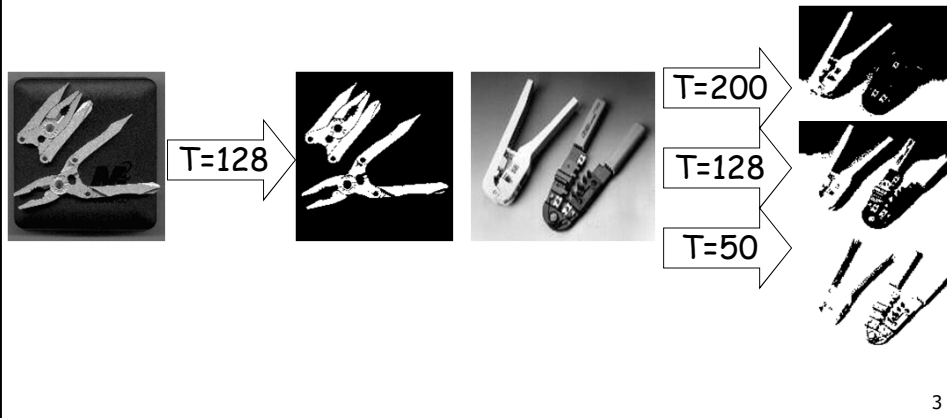
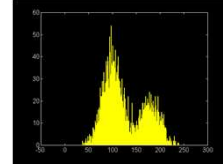
Binary Images Analysis
Geometrical & Topological Properties
Connectedness
Binary Algorithms
Morphology

Binary Images

- Binary (two-valued; black/white) images gives better efficiency in acquiring, storage, processing and transmission.
 - simplified description in terms of shape
- Obtained by *thresholding* or *segmentation* of a gray-level image
- One can compute some geometrical (e.g. area, orientation) and topological (connectedness) properties using binary images
- Remember that all we have is a "silhouette", but good recognition by humans is encouraging

Binarization

- Effect of different thresholds T on the image histogram



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Binary Images : Geometrical Properties (1/4)

- Assume *continuous* binary image, $b(x,y)$
(background 0, foreground 1)

$$\text{Area: } A = \iint b(x, y) dx dy$$

$$\text{Position: } \bar{x} = \frac{\iint x b(x, y) dx dy}{\iint b(x, y) dx dy}$$

$$\bar{y} = \frac{\iint y b(x, y) dx dy}{\iint b(x, y) dx dy}$$

Center of mass (1st moment along one of the axis)

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Binary Images : Geometrical Properties (2/4)

For discrete images,

- We denote the set of pixels in a region by R
- Assuming square pixels, we obtain

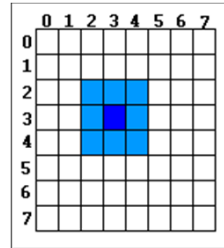
■ *Area:*

$$A = \sum_{(x,y) \in R} 1$$

■ *Centroid:*

$$\bar{x} = \frac{1}{A} \sum_{(x,y) \in R} x$$

$$\bar{y} = \frac{1}{A} \sum_{(x,y) \in R} y$$



Source: Shapiro & Stockman

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Binary Images : Geometrical Properties (3/4)

Orientation : Find the line for which the integral of the square of the distance to the object points is a minimum

$$E = \iint r^2 b(x, y) dx dy \quad r \text{ is a perpendicular distance from } (x,y) \text{ to orientation line}$$

The line passes thru *center of mass* with an angle given by :

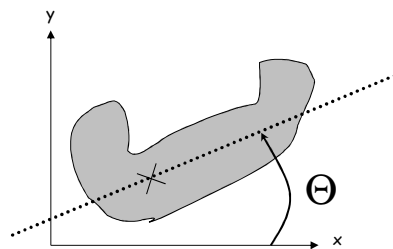
[See the derivations on the distributed notes]

$$\tan 2\Theta = \frac{b}{a - c}$$

$$a = \iint (x')^2 b(x', y') dx' dy'$$

$$b = 2 \iint (x'y') b(x', y') dx' dy'$$

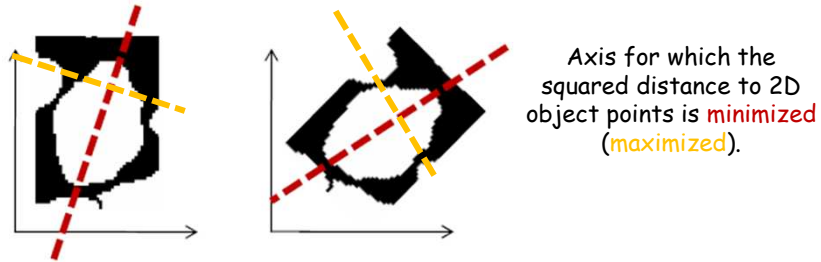
$$c = \iint (y')^2 b(x', y') dx' dy'$$



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Binary Images : Geometrical Properties (4/4)

- Another technique to calculate the axis of the second least moment



- Compute the eigenvectors of 2nd moment matrix

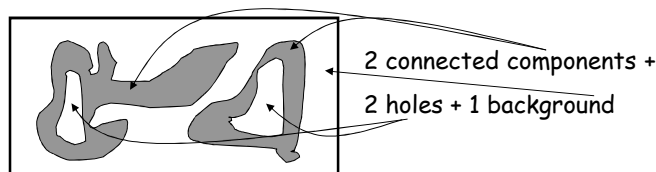
$$\begin{bmatrix} \mu_{20} & \mu_{11} \\ \mu_{11} & \mu_{02} \end{bmatrix} = VDV^T = \begin{bmatrix} v_{11} & v_{12} \\ v_{22} & v_{22} \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} \begin{bmatrix} v_{11} & v_{12} \\ v_{21} & v_{22} \end{bmatrix}^T$$

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Binary Images : Topological Properties

In case of more than one object is in the field of view:

- Area, center of mass or orientation produce an *average* value
- Different components should be separated by defining the meaning of *connectedness* between two binary cells
- Two points on an image are connected, if a path can be found along which $b(x,y)$ is constant
- A connected component is the maximal set of connected points



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Topological Properties: Discrete Connectedness

- **Neighbors** : pixels having boundaries
 - 4-neighbors : only edge-adjacent pixels
 - 8- neighbors : edge- & corner-adjacent pixel

B	C	B
C	H	C
B	C	B

4-connected

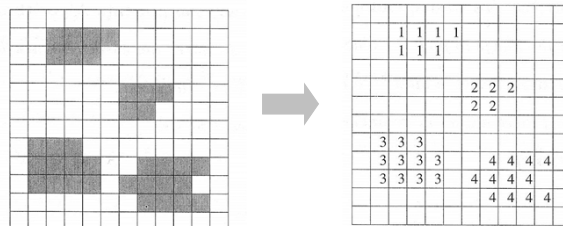
C	C	C
C	H	C
C	C	C

8-connected

- **Path** : a sequence of neighboring pixels

Topological Properties : Connected Components

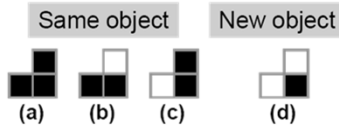
- Dividing a binary image into connected components:
Connected component labeling (CCL)



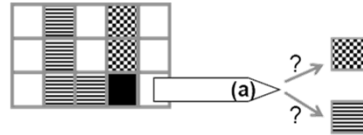
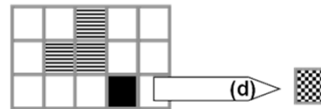
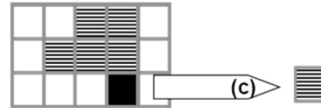
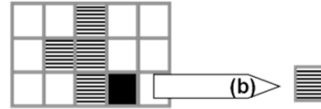
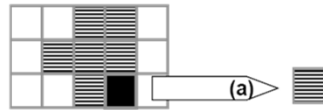
- Two approaches for CCL
 - Recursive
 - Sequential (2-pass)

Sequential Connected Components (1/2)

- Labeling a pixel only requires to consider its prior and superior neighbors.
- It depends on the type of connectivity used for foreground (4-connectivity here).



What happens in these cases?

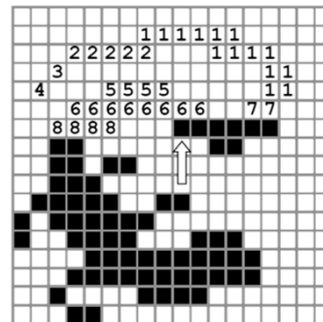


Equivalence table

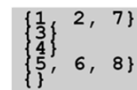
Sequential Connected Components (2/2)

- Process the image from left to right, top to bottom:

- 1.) If the next pixel to process is 1
 - i.) If only one of its neighbors (top or left) is 1, copy its label.
 - ii.) If both are 1 and have the same label, copy it.
 - iii.) If they have different labels
 - Copy the label from the left.
 - Update the equivalence table.
 - iv.) Otherwise, assign a new label.



- Re-label with the smallest of equivalent labels



Binary Properties:

- Some properties obtained using connected components analysis
 - Area : Usually used in "size filters"
 - Perimeter : Boundary following algorithms
 - Compactness : $(\text{Perimeter})^2 / \text{Area}$
 - *circle* : the most compact
 - *line* : the least compact
 - Circularity:

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Circularity

- Measure the deviation from a perfect circle

- *Circularity*:
$$C = \frac{\mu_R}{\sigma_R}$$

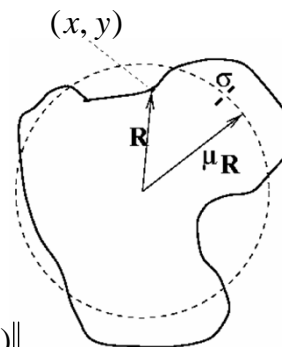
where μ_R and σ_R^2 are the mean and variance of the distance from the centroid of the shape to the boundary pixels (x_k, y_k) .

- *Mean radial distance*:

$$\mu_R = \frac{1}{K} \sum_{k=0}^{K-1} \|(x_k, y_k) - (\bar{x}, \bar{y})\|$$

- *Variance of radial distance*:

$$\sigma_R^2 = \frac{1}{K} \sum_{k=0}^{K-1} \left[\|(x_k, y_k) - (\bar{x}, \bar{y})\| - \mu_R \right]^2$$



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Invariant Descriptors

- S is a subset of pixels (region).
- Central (j,k) th moment defined as:

$$\mu_{jk} = \sum_{(x,y) \in S} (x - \bar{x})^j (y - \bar{y})^k$$

- μ_{jk} is invariant to translation of S .
- Interpretation:
 - μ_{00} : 0th central moment: *area*
 - $\mu_{02,20}$: 2nd central moment: *variance*
 - $\mu_{03,30}$: 3rd central moment: *skewness*
 - $\mu_{04,40}$: 4th central moment: *kurtosis*

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Moment Invariants

- Normalized central moments

$$\eta_{pq} = \frac{\mu_{pq}}{\mu_{00}^\gamma}, \quad \gamma = \frac{p+q}{2} + 1$$

- A set of *invariant moments* can be defined for object description:

$$\phi_1 = \eta_{20} + \eta_{02}$$

$$\phi_2 = (\eta_{20} - \eta_{02})^2 + 4\eta_{11}^2$$

$$\phi_3 = (\eta_{30} - 3\eta_{12})^2 + (3\eta_{21} - \eta_{03})^2$$

$$\phi_4 = (\eta_{30} + \eta_{12})^2 + (\eta_{21} + \eta_{03})^2$$

(Additional invariant moments ϕ_5, ϕ_6, ϕ_7 can be found in the literature).

- Invariant to rotation, scaling & translation (RST)

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Binary Algorithms : Distance Transform

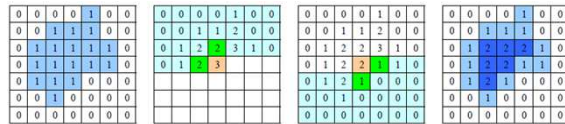
- A measure is necessary to find the distance between two pixels on a binary image, $b(i,j)$

- City-block, chessboard
- Euclidean

$$d_1(k,l) = |k| + |l|$$

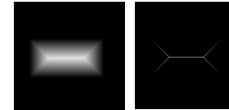
$$d_2(k,l) = \sqrt{k^2 + l^2}$$

- Distance Transform, $D(i,j)$: Minimum distance between an object and all background points** $D(i,j) = \min_{k,l:b(k,l)=0} d(i-k, j-l)$,



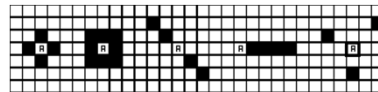
City-block distance transform computation:
Minimum of N&W (S&E) neighbor + 1 and merge 2 passes

- Points that give the locally maximum (ridges) of such distances are called *medial axis (skeleton)*



Binary Algorithms : Morphological Operations

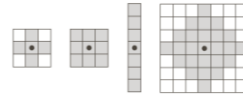
- Mathematical morphology is a set-theoretical approach to image analysis, based on shape.
- Signals are locally compared with *structuring elements S* of arbitrary shape with a reference point R



- Aim** : Transforming signals into simpler ones by removing irrelevant information
- Usually used for binary images, but gray-level extensions also exist

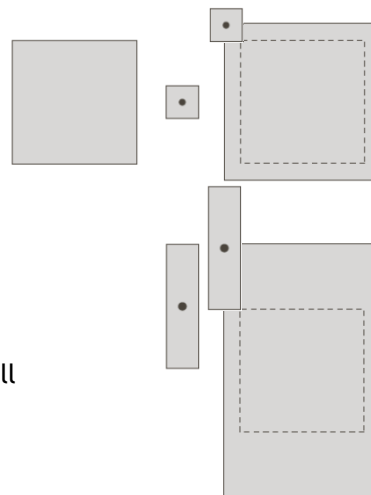
Morphological Operators

- **Basic idea**
 - Scan the image with a structuring element
 - Perform set operations (intersection, union) of image content with structuring element
- **Two basic operations**
 - Dilation
 - Erosion
- **Several important combinations**
 - Opening
 - Closing
 - Boundary extraction



Dilation

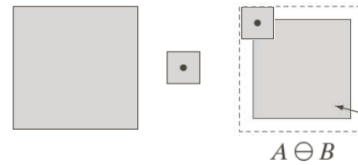
- **Definition**
 - “The dilation of A by B is the set of all displacements z , such that $(\hat{B})_z$ and A overlap by at least one element”.
 - $(\hat{B})_z$ is the mirrored version of B , shifted by z
- **Effects**
 - If current pixel z is foreground, set all pixels under $(B)_z$ to foreground.
 - ⇒ Expand connected components
 - ⇒ Grow features
 - ⇒ Fill holes



Erosion

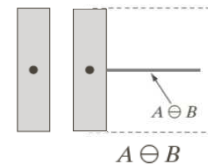
- Definition

- “The erosion of A by B is the set of all displacements z , such that $(B)_z$ is entirely contained in A ”.

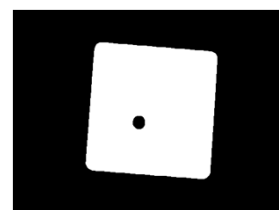
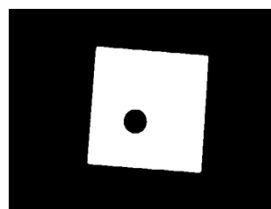


- Effects

- If not every pixel under $(B)_z$ is foreground, set the current pixel z to background.
 - ⇒ Erode connected components
 - ⇒ Shrink features
 - ⇒ Remove bridges, branches, noise



Effects of Dilation and Erosion



Opening

- **Definition**

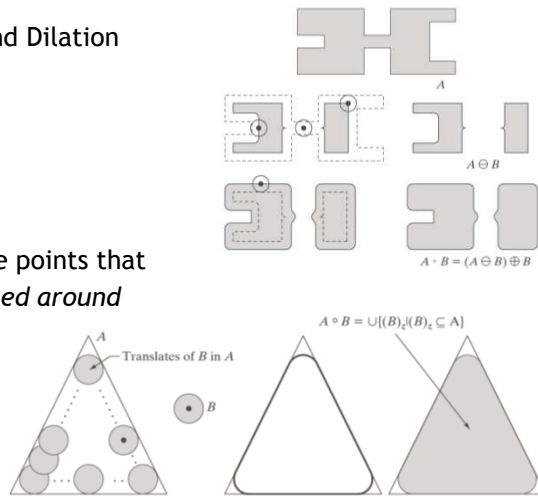
- Sequence of Erosion and Dilation

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

- **Effect**

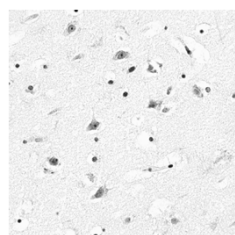
- $A \circ B$ is defined by the points that are reached if B is rolled around inside A .

⇒ Remove small objects, keep original shape.

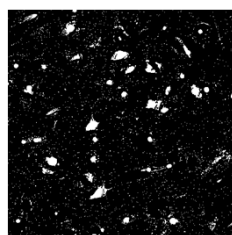


Effect of Opening

- Feature selection through *size* of structuring element



Original image



Thresholded



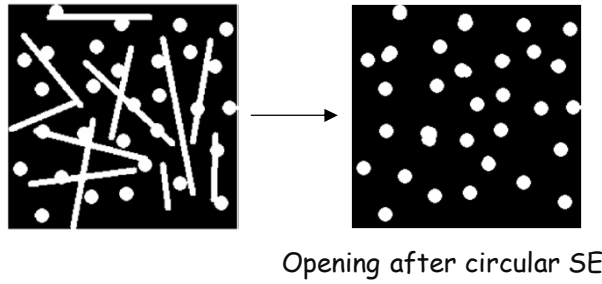
Opening with smaller SE



Opening with larger SE

Effect of Opening

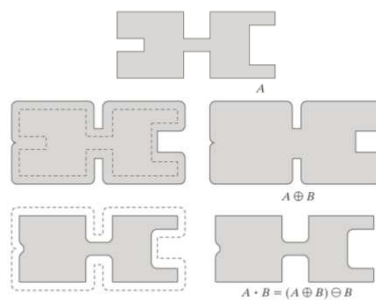
- Feature selection through *shape* of structuring element (SE)



Closing

- Definition
 - > Sequence of Dilation and Erosion
$$A \cdot B = (A \oplus B) \ominus B$$

- Effect
 - > $A \cdot B$ is defined by the points that are reached if B is rolled around on the outside of A .
 - ⇒ Fill holes, keep original shape.

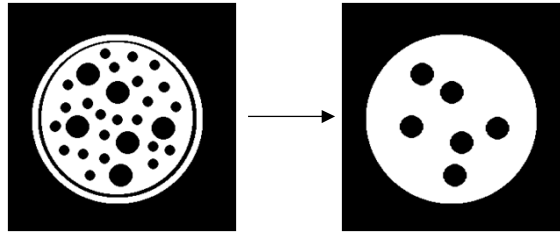


Effect of Closing

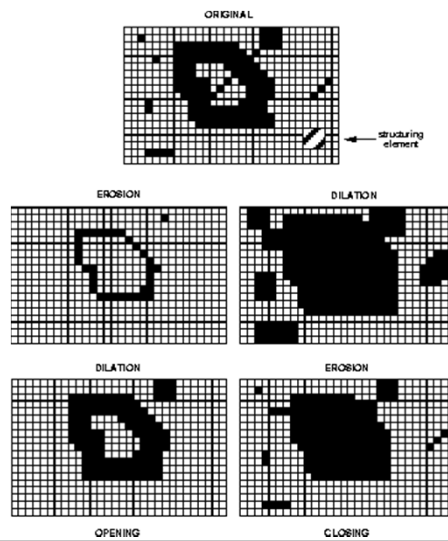
- Fill holes in thresholded image (e.g. due to specularities)



Size of structuring elements determines which structures are selectively filled



Binary Algorithms : Morphological Operations



Example Application: Opening + Closing



Morphological Boundary Extraction

- Definition

- First erode A by B , then subtract the result from the original A .

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



- Effects

- If a 3x3 structuring element is used, this results in a boundary that is exactly 1 pixel thick.

Morphology Operators on Grayscale Images

- Dilation and erosion typically performed on binary images.
- If image is grayscale:
 - for dilation take the neighborhood max,
 - for erosion take the min.

