

Basic Algorithms for Digital Image Analysis

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Mathematical morphology

- **Morphology**
 - study of structure and form of animals, plants, or words, phrases
 - express structures and forms in terms of *structuring elements*
- **Mathematical morphology**
 - study of structure and form of images or other spatial structures
 - compare them to sliding *structuring element*
 - ⇒ hit objects with structuring element
 - ⇒ transform them to more revealing shapes



- 1 Basic notions of morphological processing
 - Erosion and dilation
 - Opening and closing
 - Hit-Miss
- 2 Other useful operations
 - Morphological medial axis
 - Morphological thinning
 - Pruning
- 3 Summary of morphological processing



Two basic morphological operations

- Most morphological operations are defined in terms of two basic operations
 - **erosion**
 - **dilation**
- Basic definitions
 - B : **structuring element** with origin c
 - ⇒ discrete set of points of specific configuration aimed at particular operations
 - X : input digital image
 - B_x : translation of B so as to have origin c in point x of X
 - ⇒ result of morphological operation is assigned to point x of output image
- Figure (object) and ground
 - figure: nonzero (black in illustrations)
 - ground: zero (white in illustrations)



Erosion of X by B

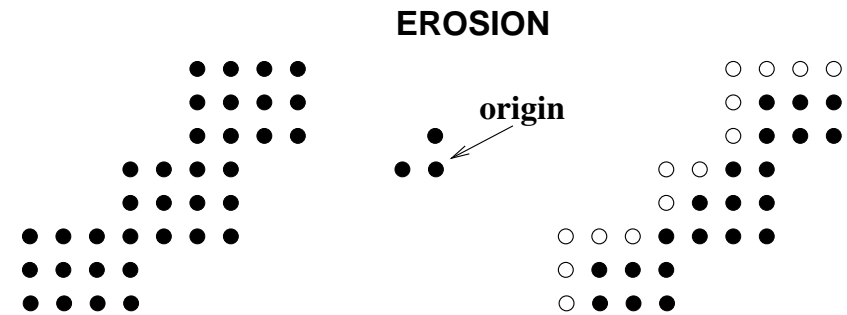
- Set of all points x such that B_x is included in X

$$X \ominus B = \{x : B_x \subset X\}$$

- Nonzero output when
 - **all points** of structuring element coincide with nonzero points of image
- Zero output otherwise



Example of erosion



- Structural element is shown in centre
- Empty circle is **deleted point**
⇒ point x where B_x is not included in X



Dilation of X by B

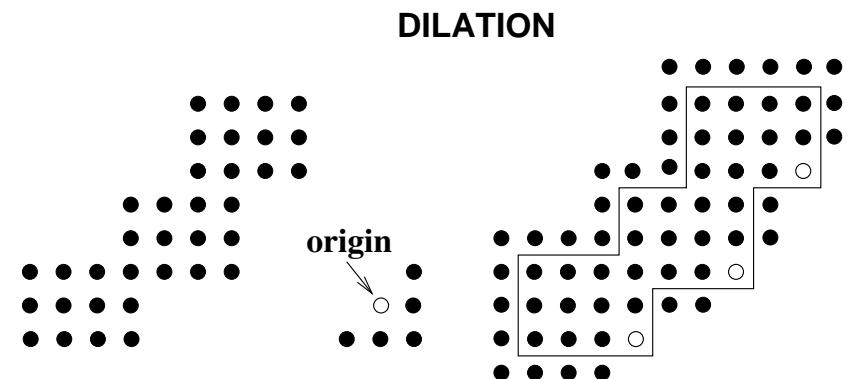
- Set of all points x such that intersection of B_x and X is nonempty

$$X \oplus B = \{x : B_x \cap X \neq \emptyset\}$$

- Nonzero output when
 - **at least one point** of structuring element coincides with nonzero points of image
⇒ B_x hits X
- Zero output otherwise



Example of dilation



- Contour shows original object
- Empty circles are deleted points
- Despite deletion, resulting object is larger
⇒ but shifted due to shift of origin



- **Translation invariance**

- shift of object leads to same shift in result

- Dilation is **commutative**

$$X \oplus B = B \oplus X$$

- $B \oplus X$: B is treated as image, X as structuring element
- ⇒ similar to commutativity in linear filtering

- Erosion is **not commutative with image**

$$X \ominus B \neq B \ominus X$$

- ⇒ large object, small structuring element
- ⇒ empty $B \ominus X$, nonempty $X \ominus B$



- But order of erosions is arbitrary

$$(X \ominus A) \ominus B = (X \ominus B) \ominus A$$

- Erosion and dilation are **not inverses of each other**

- ⇒ erosion to empty set cannot be reversed



- **Distributivity in structuring element**

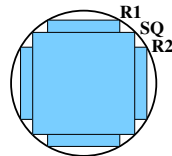
$$X \oplus (B \cup B') = (X \oplus B) \cup (X \oplus B')$$

$$X \ominus (B \cup B') = (X \ominus B) \cap (X \ominus B')$$

- Usage

- structuring element can be represented/approximated by **union** of two or more simple elements
- ⇒ disc approximated by union of square and rectangles to speed up erosion
- ⇒ erosion by rectangle implemented like fast running filter

Disc approximated by square SQ and rectangles R1, R2



- **Distributivity in image**

$$(X \cap Y) \ominus B = (X \ominus B) \cap (Y \ominus B)$$

$$(X \cup Y) \oplus B = (X \oplus B) \cup (Y \oplus B)$$

- **Iteration**

$$(X \ominus B) \ominus B' = X \ominus (B \oplus B')$$

$$(X \oplus B) \oplus B' = X \oplus (B \oplus B')$$

- dilation is **associative**, erosion is not
- Use iteration to **speed up erosion and dilation**
 - when structuring element can be represented as dilation of two or more smaller elements
 - ⇒ erosion by 5×5 square is two erosions by 3×3 square



$$\begin{array}{ccccc}
 1 & 1 & 1 & 1 & 1 \\
 1 & 1 & 1 & 1 & 1 \\
 1 & 1 & 1 & 1 & 1 \\
 1 & 1 & 1 & 1 & 1 \\
 1 & 1 & 1 & 1 & 1
 \end{array}
 =
 \begin{array}{ccc}
 1 & 1 & 1 \\
 1 & 1 & 1 \\
 1 & 1 & 1
 \end{array}
 \oplus
 \begin{array}{ccc}
 1 & 1 & 1 \\
 1 & 1 & 1 \\
 1 & 1 & 1
 \end{array}$$

- **Opening** of set X by structuring element B

$$X_B = (X \ominus B) \oplus B$$

- smooths contours, removes small spots and sharp caps
- can be used for object size distribution study
- ⇒ use openings with growing B

- **Closing** of set X by structuring element B

$$X^B = (X \oplus B) \ominus B$$

- fills up narrow channels and thin lakes
- can be used for inter-object distance study
- ⇒ use closings with growing B

- **Increasing**

- if $X \subset X'$, then

$$X \ominus B \subset X' \ominus B$$

$$X \oplus B \subset X' \oplus B$$

- if $B \subset B'$, then

$$X \ominus B' \subset X \ominus B$$

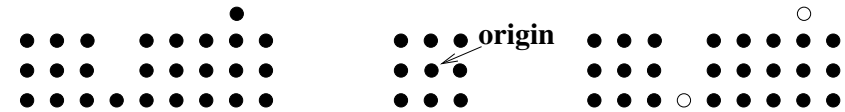
$$X \oplus B \subset X \oplus B'$$

- **Duality** with respect to complement operation

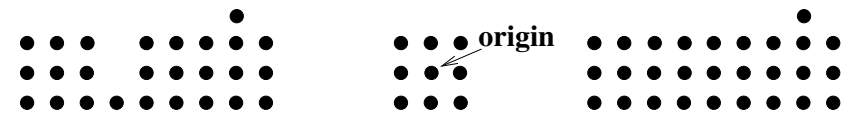
$$\underline{X} \oplus B^{\sim} = \underline{X \ominus B}$$

- \underline{X} is complement of X
- B^{\sim} is reflection of B with respect to its origin

OPENING



CLOSING



Properties of opening and closing

- Opening and closing are **idempotent** operations

$$(X_B)_B = X_B$$

$$(X^B)^B = X^B$$

⇒ Do not apply twice with *same* B

- Opening and closing are **duals of each other**

$$\underline{(X_B)} = (\underline{X})^B$$

$$\underline{(X^B)} = (\underline{X})_B$$

⇒ Opening is (complement of) closing of complement



Hit-Miss: Matching binary patterns

- Search for match or specific configuration
 - of **object and ground** pixels
- Specify both object and ground points of B
 - $B_{ob} \subset B$: object part of structuring element
 - $B_{gr} \subset B$: ground part of structuring element
- Hit-miss operator outputs object pixel when
 - B_{ob} matches object pixels **and**
 - B_{gr} matches ground pixels



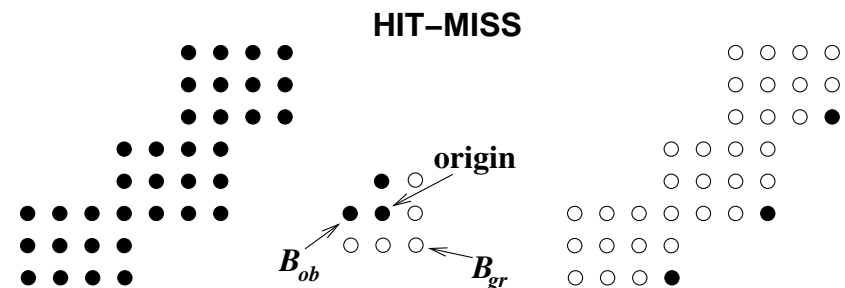
Definition of hit-miss

$$X \otimes B \doteq \{x : B_{ob} \in X \text{ and } B_{gr} \in \underline{X}\} = (X \ominus B_{ob}) \cap (\underline{X} \ominus B_{gr})$$

- B_{gr} is treated as consisting of **object** pixels
 - they have to match **complement** \underline{X}
- Hit-miss in morphologic processing corresponds to pattern matching in greyscale images
- Other definitions of hit-miss also exist
 - equivalent** to our definition
- Symbol \odot is also used to denote the hit-miss



Finding corners with hit-miss



- Both figure and ground pixels of B must match
- Only finds corners of specific orientation and shape
 - ⇒ rotate to find other corners of same shape



Boundary extraction

- **Boundary** of set X

$$\partial X = X \setminus (X \ominus G) \quad \text{with structuring element } G = \begin{matrix} & 1 & 1 & 1 \\ 1 & 1 & 1 & \\ 1 & 1 & 1 & \end{matrix}$$

- **Action**

- interior points are obtained by erosion
- they are subtracted from original image
- ⇒ difference is boundary
- 3×3 square G approximates small *digital disc*
 - often used in morphological processing
 - larger 'discs' (squares) obtained using *iteration* property



Medial axis

- **Medial axis** $S(X)$ of set X

$$S(X) = \bigcup_{n=0}^{n_{max}} (X \ominus nG) \setminus (X \ominus nG)_G \doteq \bigcup_{n=0}^{n_{max}} s_n(X)$$

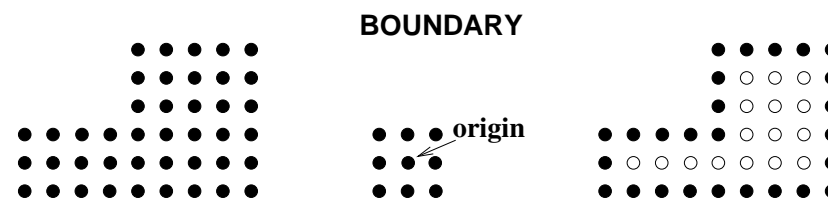
- MA obtained as union of its parts $s_n(X)$

- **Notation**

- structuring element G : 3×3 square ('disc')
- n_{max} : maximum size after which X erodes down to zero
- $(X \ominus nG)$: n^{th} iteration $(X \ominus G) \ominus G \ominus \dots \ominus G$
- $(X \ominus nG)_G$: opening of $(X \ominus nG)$ by G



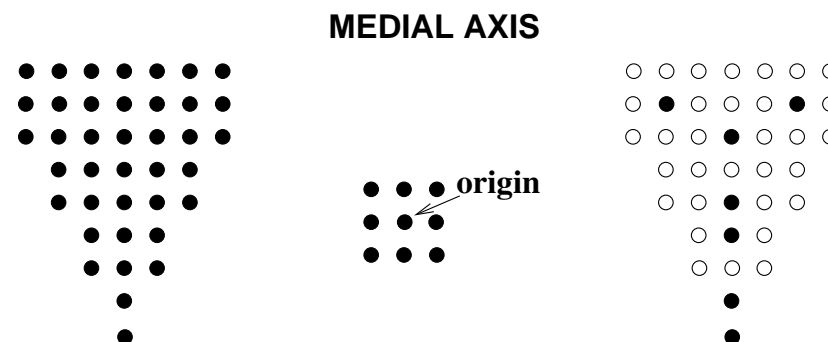
Example of boundary extraction



- Interior points are obtained by erosion
- Then subtracted from original image
- Difference is boundary



Example of morphological medial axis



- Morphological MA is obtained as union of parts
- Parts may be separate from each other
 - ⇒ contrary to continuous medial axis, morphological MA is not always connected



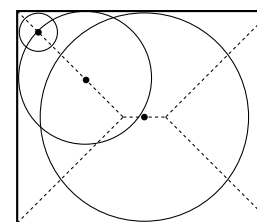
Restoring set X from its medial axis

$$X = \bigcup_{n=0}^{n_{max}} s_n(X) \oplus nG$$

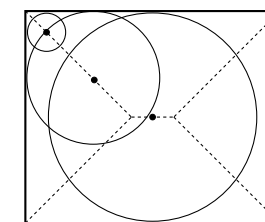
- $Y \oplus nG$ is n^{th} iteration of \oplus using G
- MA is detected as set of centres of maximal discs that
 - are contained in X **and**
 - touch boundary of X in two or more points
 ⇒ iterative *erosions* used
- To restore object from its MA take union of discs
 - centred on MA points **and**
 - having radii equal to contour distances
 ⇒ iterative *dilations* used



Meaning of morphological MAT and its inversion



obtain MA

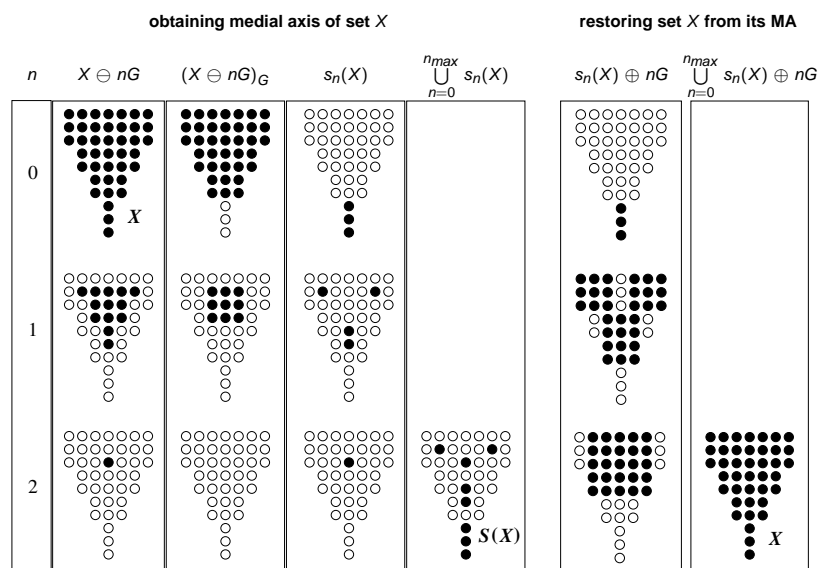


restore shape from MA

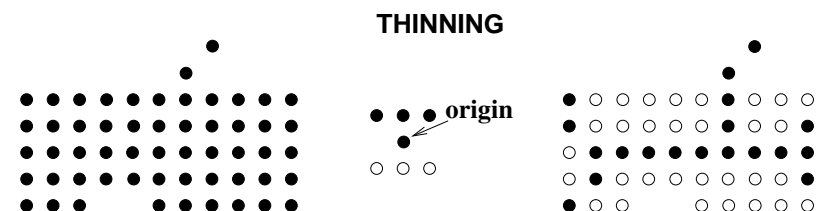
- Obtain medial axis as locus of inscribed circles
 - **erosions by discs**
- Restore shape from MA
 - **dilations by discs**



Example of morphological medial axis



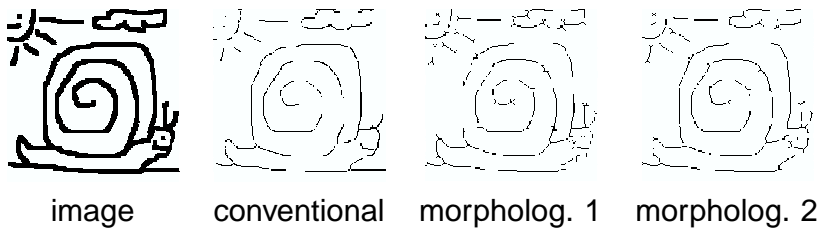
Thinning



- Uses hit-miss to reduce object to its skeleton
 - ⇒ set of branches
- Iterative operation
 - at each iteration, *eight rotated versions* of structural element are used
 - results may depend on order in which rotated versions are applied



Comparison different thinnings



- Result of conventional thinning is better
- Morphological result
 - depends on order of rotations (1,2)
 - ⇒ sensitive to object orientation
 - contains small parasite branches

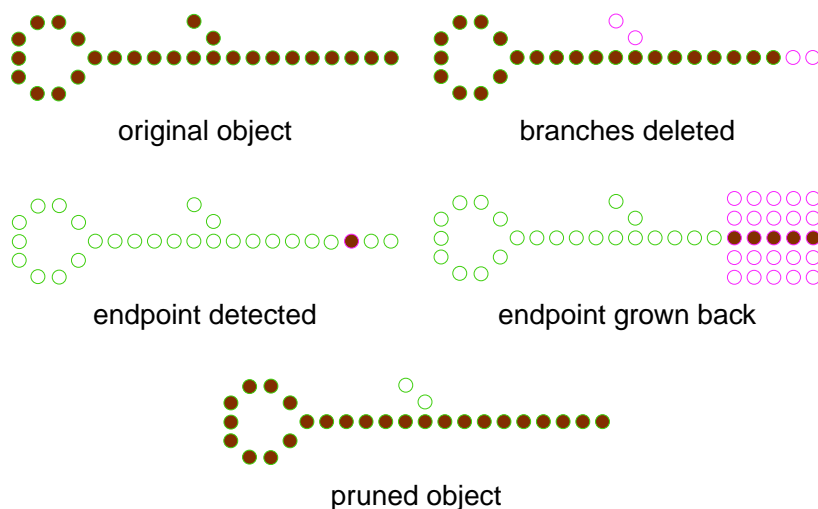


Morphological pruning of small branches

- Pruning eliminates small parasite branches of object
 - often applied after morphological MAT or thinning
- Morphological pruning involves several operations
 - recursively remove branches with rotating structural element
 - ⇒ number of recursions defines maximum length of branches to be removed
 - detect endpoints of remaining branches
 - recursively 'grow back' remaining branches by *geodesic dilation* using another element
 - ⇒ geodesic dilation is dilation limited to original set



Example of morphological pruning



Summary 1/2

- Morphological operations can be **extended to greyscale images**
- Morphology is used for
 - image pre-processing
 - ⇒ noise filtering, shape simplification
 - enhancing object structure
 - ⇒ skeletonisation, thickening
 - segmenting objects from ground
 - quantitative description of objects
 - ⇒ area, perimeter, projections, holes



- Morphological operations are best suitable
 - for processing images containing **many small objects**
 - for obtaining **statistical descriptions** of such images
 - ⇒ distributions of object sizes, perimeters, etc.
 - ⇒ fast hardware implementation available
- Not suitable for precise description of **large**, complex shapes
- Drawbacks
 - sensitivity to image orientation
 - approximate character of many results
 - ⇒ approximate MA, perimeter, etc.