

Binary Images (Part I)

- Threshold
- Binary Image - Definition
- Connected Components
- Chain Code
- Edge Following

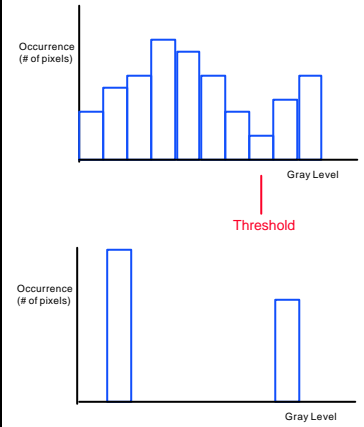
Grayscale Image



Binary Image



Thresholding



Thresholding a Grayscale Image

Original Image



Binary Image



Threshold too low

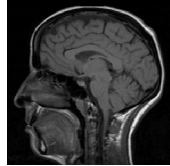


Threshold too high



FMRI - Example

Original Image



Threshold = 80



Threshold = 71



Threshold = 88

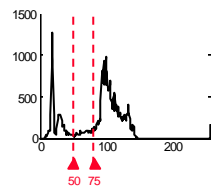


Segmentation using Thresholding

Original



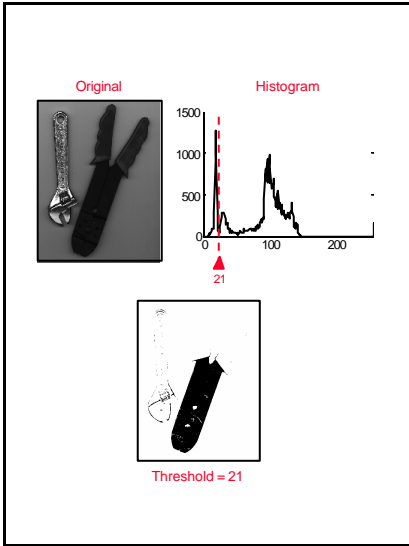
Histogram

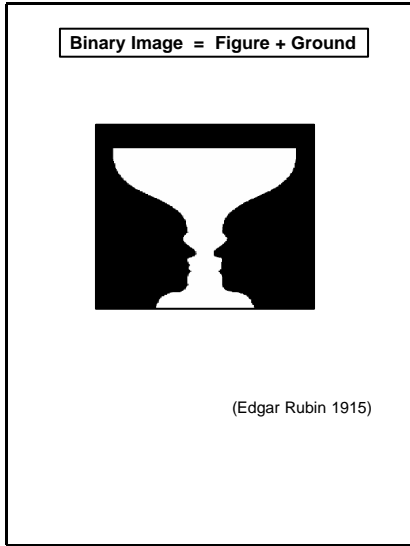


Threshold = 50



Threshold = 75





Connected Components

Neighborhoods:

4-neighbor metric

8-neighbor metric

Connected Components:

S = the set of object pixels

S is a **Connected Component** if for each pixel pair $(x_1, y_1) \in S$ and $(x_2, y_2) \in S$ there is a path passing through X-neighbors in S . (X = 4,8).

S may contain several connected components.

1 connected component-8
3 connected components-4

8-neighborhood:
1 object connected component
1 background connected component

4-neighborhood:
2 background connected components
4 object connected components

Always choose different neighborhood metrics for objects and backgrounds.

Chain Code

Each direction is assigned a code:

4-neighbor

8-neighbor

4-neighbor:

0303333232312121011011

8-neighbor:

076666553321212

Marking the Connected Components

Connected Component Algorithm: Two passes over the image.

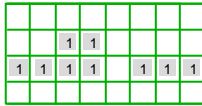
Pass 1:
Scan the image pixels from **left to right** and from **top to bottom**.
For every pixel P of value 1 (an object pixel), test top and left neighbors (4-neighbor metric).

- If 2 of the neighbors are 0: assign a new mark to P.
- If 1 of the neighbors isn't 0: assign the neighbor's mark to P.
- If 2 of the neighbors are not 0: assign the left neighbor's mark to P.

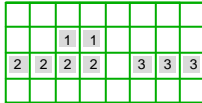
Pass 2:
Divide all marks to equivalence classes (marks of neighboring pixels are considered equivalent).
Replace each mark with the number of its equivalence class.

Connected Components - Example

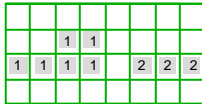
Original Binary image



Pass 1:



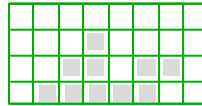
Pass 2:



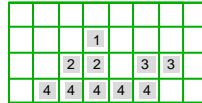
Equivalence Class number	Original mark
1	1,2
2	3

Connected Components - Example II

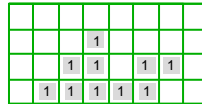
Original Binary image



Pass 1:



Pass 2:



Equivalence Class number	Original mark
1	1,2,3,4

Edges

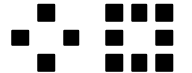
C = connected component of object S.
 D = connected component of \bar{S} .

The D-Edge of C = the set of all pixels in C that have a neighboring pixel in D.
 (neighboring-8 if C is 4-connected neighboring-4 if C is 8-connected).

Example:



The Edge of C for background D.



(4-neighbor) (8-neighbor)
 The Edge of C for hole D.

Distances

Two grid point: $P = (x,y)$ and $Q = (u,v)$

Euclidean Distance

$$d_e(P,Q) = \sqrt{(x-u)^2 + (y-v)^2}$$

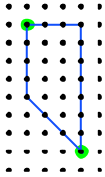
City Block Distance

$$d_b(P,Q) = |x-u| + |y-v|$$

Chessboard Distance

$$d_c(P,Q) = \max(|x-u|, |y-v|)$$

$d_e = 7.6$
 $d_b = 7$
 $d_c = 10$



d_e , d_b , d_c are all **metrics**:

1. Distance metric: $d(P,Q) \geq 0$
2. Positive: $d(P,Q) = 0$ iff $P=Q$
3. Symmetric: $d(P,Q) = d(Q,P)$
4. Triangular inequality: $d(P,Q) \leq d(P,R) + d(R,Q)$

All pixels at equal d_c distance form a "diamond" :



All pixels at equal d_b distance form a "square" :



All pixels at equal d_e distance form a "circle" :



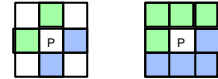
2-Pass Distance Algorithm

For each pixel calculate the d_c or d_b distance from a pixel in set S .

2 passes:

- Pass 1: scan image left-to-right and top-to-bottom
- Pass 2: scan image right-to-left and bottom-to-top.

For each pixel P mark as follows:



Pass 1: consider all neighbors of P that have been scanned $N_1 = \square$

$$d'(P,S) = \begin{cases} 0 & \text{if } P \in S \\ \min_{Q \in N_1} (d'(Q,S)) + 1 & \text{if } P \notin S \end{cases}$$

Pass 2: consider all neighbors of P that have been scanned $N_2 = \square$

$$d''(P,S) = \min_{Q \in N_2} (d'(P,S), d'(Q,S) + 1)$$

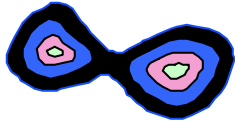
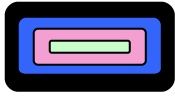
Example measuring d_c :

1	0	0	0	0	1	2	3	0	1	2	1
0	0	0	1	1	2	3	0	1	2	1	0
0	0	0	0	2	3	4	1	2	3	2	1

S is marked as 1 Pass 1: $d'(P,S)$ Pass 2: $d''(P,S)$

Skeletons

Consider all edge pixels of an object as the group S .

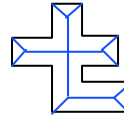
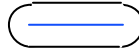
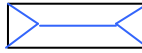


The pixels whose distance is a local maxima are the **Skeleton** of the object.

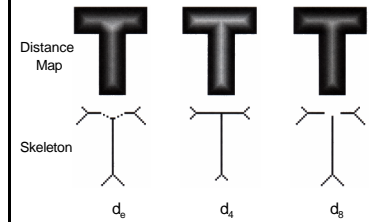
The **Skeleton** can be used as a shape descriptor.

MAT = Medial Axis Transform

Grass fire technique (Blum, 1993)



Skeletons - Example



Sensitivity to contour changes:

