

# Digital Image Processing

## Lecture 7

### Segmentation and labeling

- Region growing
- Region splitting and merging
- Labeling
- Watersheds

**smart**

### More morphological algorithms from G&W

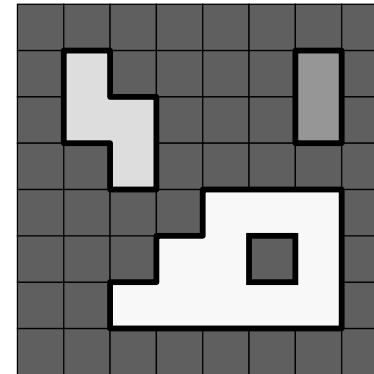
### Gonzales & Woods:

- Chapter 10 pp. 763-778
- Chapter 9 pp. 642-664

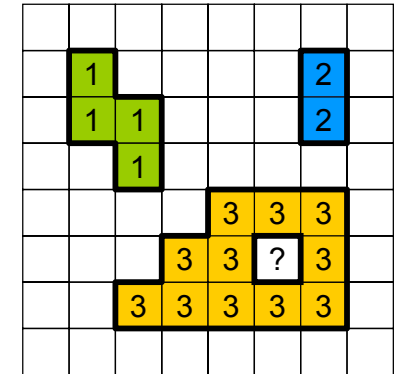
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# Segmentation and labeling of objects

Gray scale image



Segmented and labeled image



Segmentation subdivides an image into its constituent regions or objects.

## Methods for segmentation

- Point, line and edge detection ☺ ➡
- Thresholding ☺
- Region-based segmentation
  - Region growing ☀
  - Region splitting and merging ☀
- Watershed segmentation ☀
- Labeling
  - The RT-algorithm (Run-track) ☺
  - The RB-algorithm (Raster-scan Border-follow) ☀

Lab 3

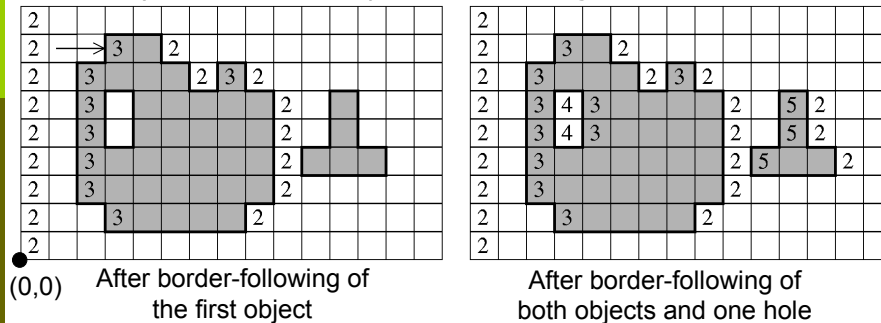
☺ know already    ☀ today    ➡ later

## Labeling, 2 different algorithms

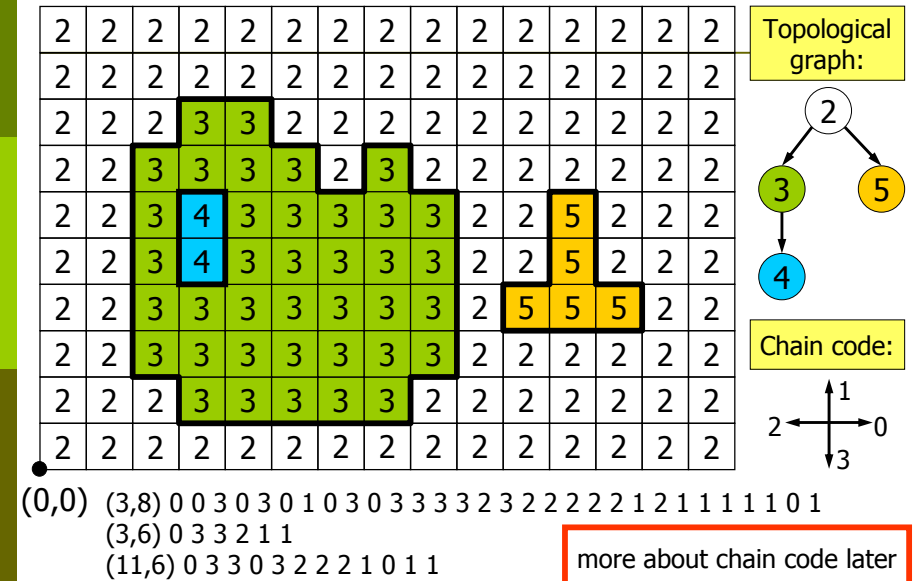
- The RT-algorithm (Run-track)
  - Fast
- The RB-algorithm (Raster-scan Border-follow)
  - Not as fast as the RT-algorithm
  - Gives more information
- There are also other methods, for example:
  - Flood fill from computer graphics...
  - Gonzalez & Woods: Extraction of connected components, see figure 9.17 and 9.18.

## The RB-algorithm (Raster-scan Border-follow)

- Raster-scan the image from left to right and from top to bottom.
- When the scanning reach a border (0->1 or 1->0), stop the scanning and start border-follow instead.
- During the border-following process, labels are set to the right of the border.
- Finally all labels are expanded to the right, see next slide.



## The RB-algorithm, final result



## Region growing

- Start with a set of seed points and expand them to neighboring pixels if they satisfy a certain criterion.
- Examples of criteria are:
  - Right intensity
  - Right color
  - Right texture
- For a simple example see next slide. It is from G&W 2:nd edition, in the 3:rd edition they have made it a bit too complicated, I think.
  - a) Original X-ray image of a defective weld with cracks. The challenge is to determine the size and location of the cracks.
  - b) The seed points are obtained using a high threshold, T1.
  - c) Result of region growing to intensity pixels >T2. (T2<T1.)
  - d) Boundaries of segmented defective weld (black) overlaid on original image.
  - There is a similarity with thresholding with hysteresis! Lab 6

## Region growing, cont.

≈ Fig. 10.51

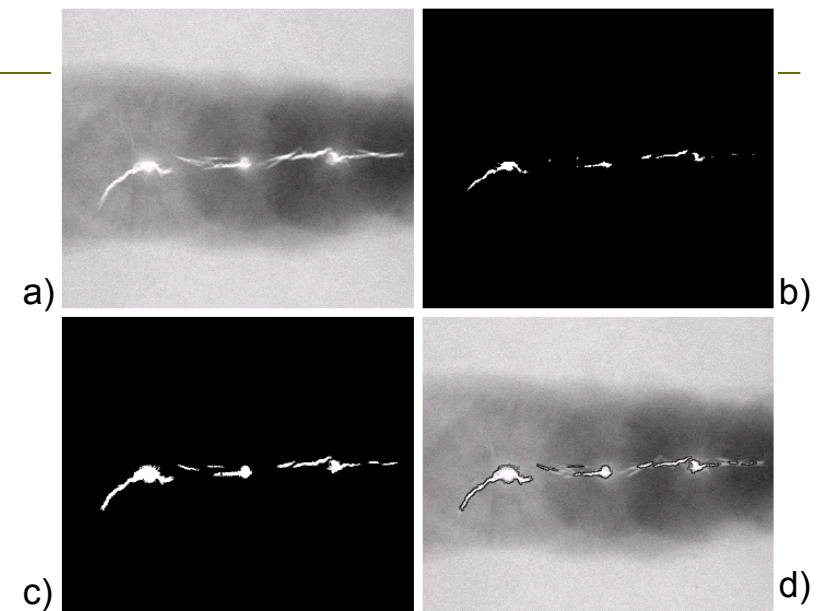
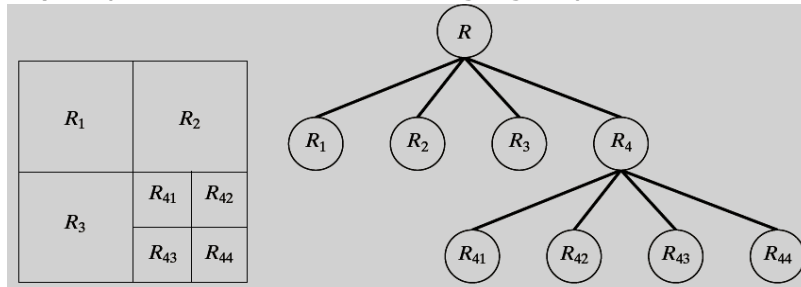


Fig. 10.52

## Region splitting and merging

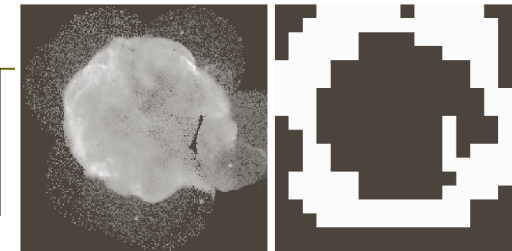
- 1) Split into four disjoint quadrants any region  $R_i$  for which  $Q(R_i)=\text{FALSE}$ .
- 2) When no further splitting is possible, merge any adjacent region  $R_j$  and  $R_i$  for which  $Q(R_j \cup R_k)=\text{TRUE}$ . (Alternative simplification:  $Q(R_j)=\text{TRUE}$  and  $Q(R_k)=\text{TRUE}$ )
- 3) Stop when no further merging is possible



## Region splitting and merging

Fig. 10.53

Original image of the Cygnus Loop supernova



Smallest Quad-Region: 32x32

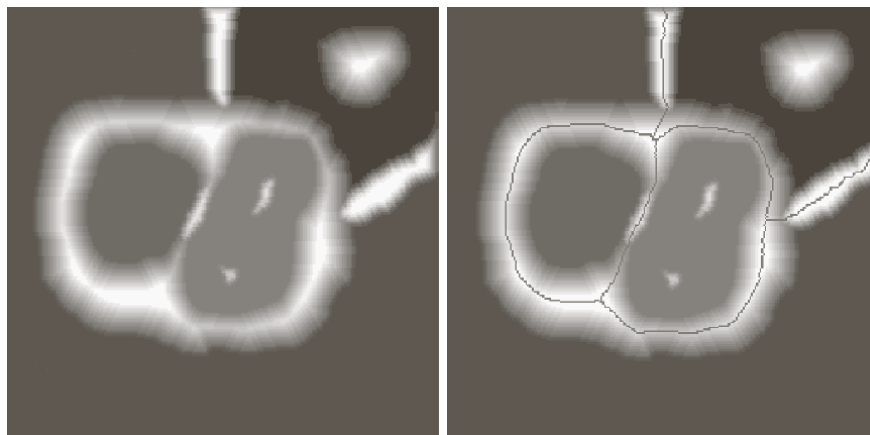
Smallest Quad-Region: 16x16



Smallest Quad-Region: 8x8

$$Q = \begin{cases} \text{TRUE}, & \text{if } \sigma(std) > a \text{ AND } 0 < m(\text{mean}) < b \\ \text{FALSE}, & \text{otherwise} \end{cases}$$

## Watershed segmentation, introduction



a)

h)

Fig. 10.54

## Watershed segmentation, introduction

- Fig. 10.54 a) The watershed algorithm regards a gray scale image as a landscape, where the lakes are preferably dark pixels and the ridges and mountains are preferably bright pixels.
- A watershed = a ridge between individual lakes and rivers.
- A catchments basin = a geographic area whose water flows to a certain lake.
- Fig. 10.54 h) The watershed algorithm can find the watersheds in the image. They form closed curves and the area inside each closed curve, the catchments basin, can be regarded as a segmented, individual object.

## Watershed segmentation, description

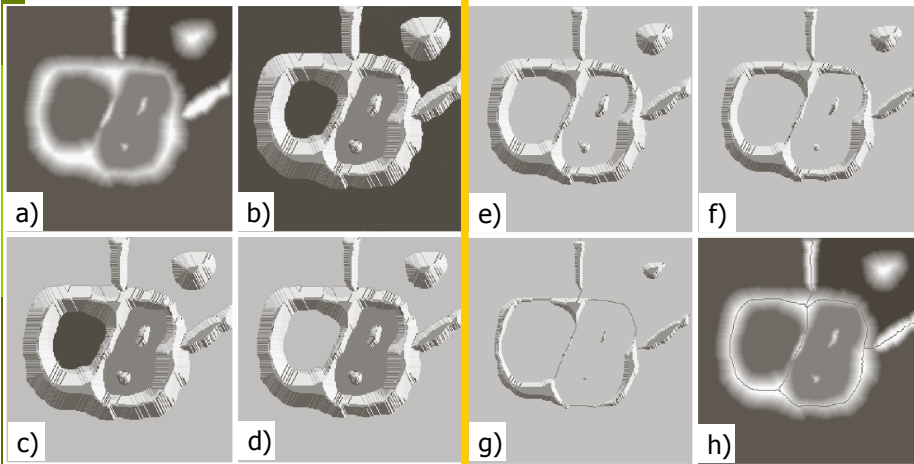


Fig. 10.54

## Watershed segmentation, description

- Fig. 10.54 b) is a topographic view of 10.54 a). It is intended to give a 3D-feeling.
- In the beginning, the landscape in a) and b) is totally dry. Local minima are detected and holes are punched in each of them.
- Then the entire landscape is flooded from below by letting water rise through the holes at a uniform rate.
- In c), there is water (light gray) in the background.
- In d), there is also water in the left lake.
- In e), there is water in both lakes.
- In f), there is the first sign on overflow between two catchments basins. Therefore, a short dam has been built between them.
- In g), there are rather long dams.
- Finally, the whole landscape is over flooded with water, separated by high dams, watersheds.
- In h), these watersheds are shown overlaid on the original image.

## Watershed segmentation, dam construction

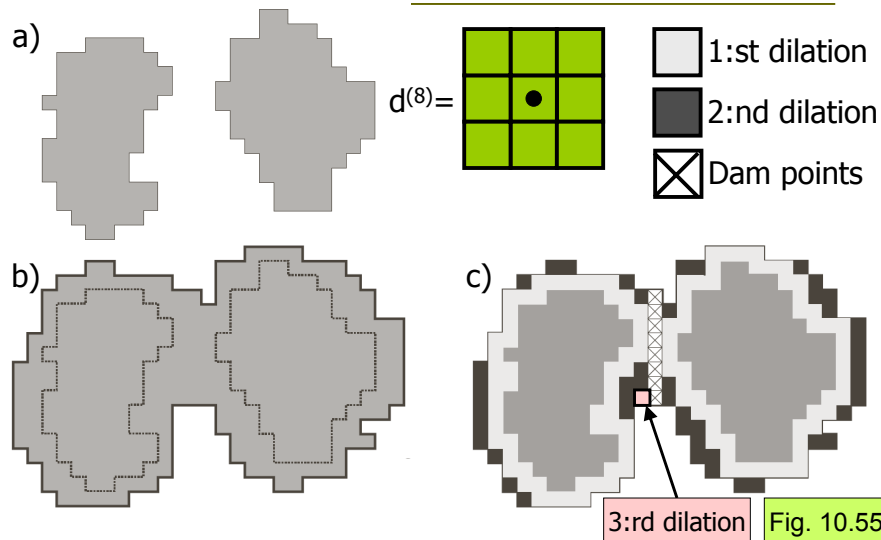


Fig. 10.55

## Watershed segmentation, dam construction

- Let  $C_{n-1}(M_1)$  denote the pixels in lake  $M_1$  at flooding step  $n-1$ . These are shown to the left in a).
- Let  $C_{n-1}(M_2)$  denote the pixels in lake  $M_2$  at flooding step  $n-1$ . These are shown to the right in a).
- Flooding step  $n$  is shown in b).
- Let  $C(n-1)$  denote the union of  $C_{n-1}(M_1)$  and  $C_{n-1}(M_2)$ . There are 2 objects in  $C(n-1)$  in a) and 1 object in  $C(n)$  in b). This indicates that something has happened in step  $n$  and that some action must be performed.
- Dilate the pixels in a) with the  $d^{(8)}$ -structuring element considering the following constraints:
  - Dilation is not allowed outside b).
  - Dilation is not allowed so that the pixels in  $M_1$  and  $M_2$  are united.
  - Instead, if the pixels in  $M_1$  and  $M_2$  are going to be united, denote: dam pixel.
- The first dilation in c) gives the light gray pixels.
- The second dilation in c) gives the dark gray pixels and the X-marked dam pixels.

## Watershed segmentation, Ex 1)

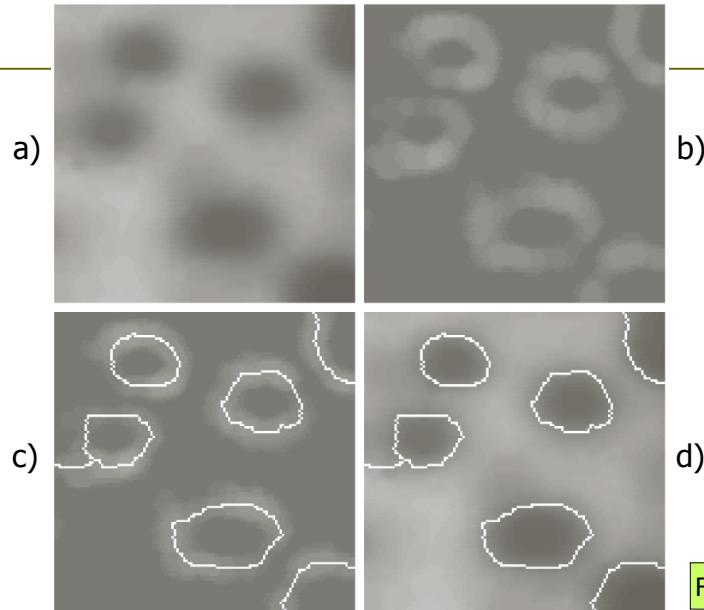


Fig. 10.56

## Watershed segmentation, Ex 1)

- a) Original image with dark objects that are going to be segmented.
- b) A usual pre-processing step is to calculate the magnitude of the gradient.
- c) The result of the watershed algorithm overlaid on the gradient image.
- d) The result of the watershed algorithm overlaid on the original image.

## Watershed segmentation, Ex 2) oversegmentation

Advanced topic!

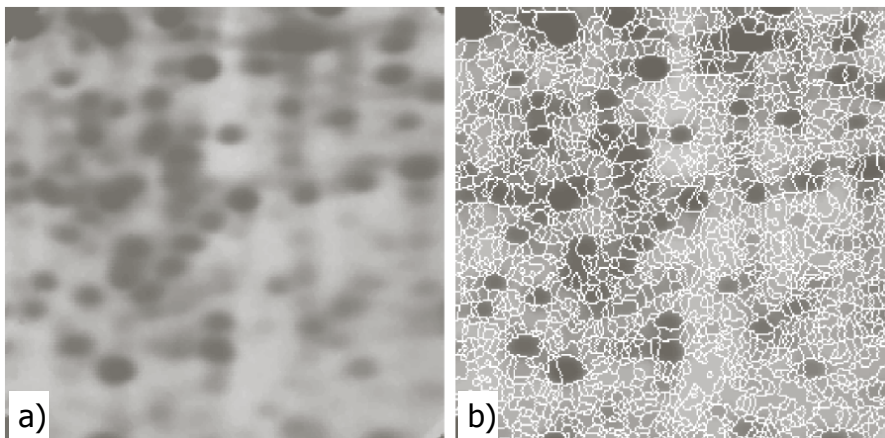


Fig. 10.57

## Watershed segmentation, Ex 2) oversegmentation

Advanced topic!

- a) Original image with dark object to be segmented.
- Pre-processing by computation of the magnitude of the gradient.
- Result from the watershed algorithm overlaid on the original image. Oversegmentation!
- The oversegmentation was probably caused by too many local minima.



## Watershed segmentation, Ex 2) oversegmentation cured

Advanced topic!

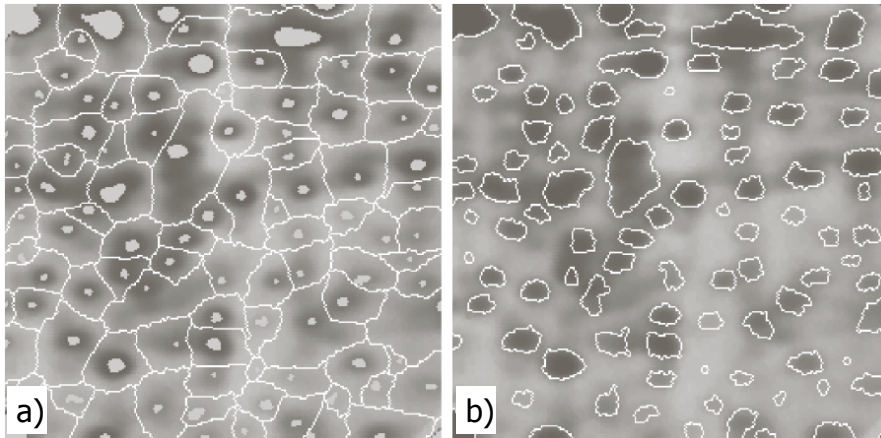


Fig. 10.58

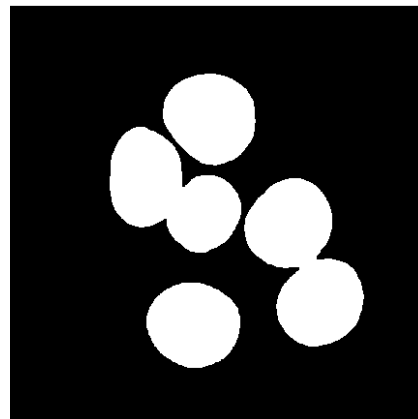
## Watershed segmentation, Ex 2) oversegmentation cured

Advanced topic!

- Fig. 10.58 a) The light gray **internal markers** are regions received by applying a low-pass filter on the original image followed by consideration of the following criteria:
  - The region must be surrounded by pixels of higher values.
  - The region must consist of a number of connected pixels.
  - The points in the region must have approximately the same gray scale value.
- The **external markers** are the watersheds received when the watershed-algorithm was applied to the low-pass filtered image. The internal markers were then used as local minima.
- The external markers in a) divides the original image into regions. Each region is processed individually. The magnitude-of-the-gradient image was calculated, the internal marker was used as local minimum and the watershed-algorithm was applied. The result is shown in Fig. 10.58 b)!

## Watershed segmentation, Ex 3) for overlapping binary objects

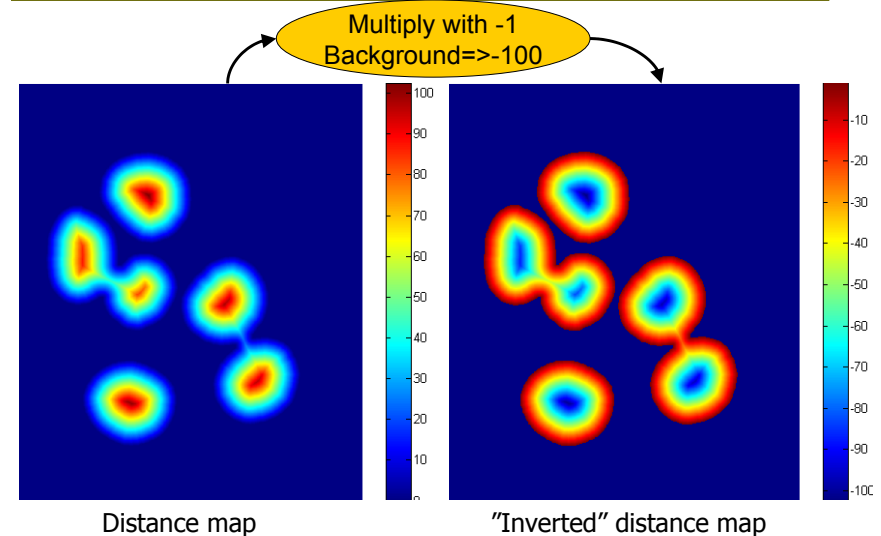
- The watershed algorithm can also be used to segment overlapping binary objects.
- Distance mapping is used to construct a landscape with lakes.
- Computation: See this and the following 2 slides.



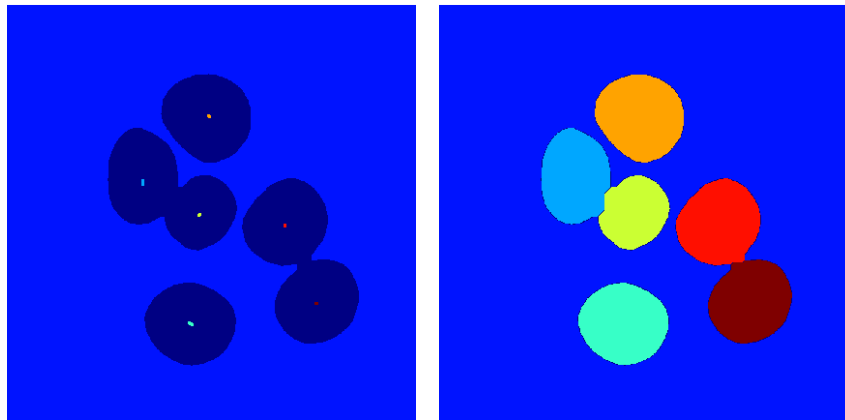
Binary input image

Lab 5!

## Ex3) Generate a distance map of the objects and “invert” it



### Ex3) Detect local min & apply the watershed algorithm on the "inverted" distance map



Labeled local min

Result of the watershed algorithm

### Some basic morphological algorithms: 1) Boundary extraction

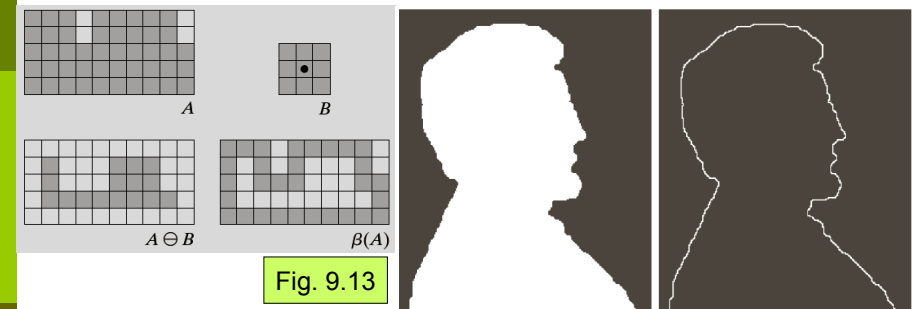


Fig. 9.13

Fig. 9.14

$$\beta(A) = A - (A \ominus B) \quad (9.5-1)$$

The object A eroded by B

### 2) Finding connected components

The image  $X_0$  contains the initial seed, which is iteratively dilated by B and multiplied by A.

Extra object does not matter

Binary multiplication

The object  $X_{k-1}$  dilated by B

$$X_k = (X_{k-1} \oplus B) \cap A, \quad k = 1, 2, 3, \dots \quad (9.5-3)$$

Fig. 9.17

### 2) Finding connected components

Original image



Thresholded image



Eroded image



The eroded image can be raster-scanned until a pixel is found. This pixel is labeled by  $i$  and the connected component is reconstructed. Its area (number of pixels) is noted. Then the image is scanned again until the next unlabeled pixel is found and labeled by  $i+1$ , etc.

Connected component	No. of pixels in connected comp
01	11
02	9
03	9
04	39
05	133
06	1
07	1
08	743
09	7
10	11
11	11
12	9
13	9
14	674
15	85

Fig. 9.18

### 3) Convex hull

- An set A is said to be convex if the straight line segment joining any two point in A lies entirely within A.
- The convex hull of S is the smallest convex set containing S.

The element  $B^i$  found a match in the image  $X_{k-1}$

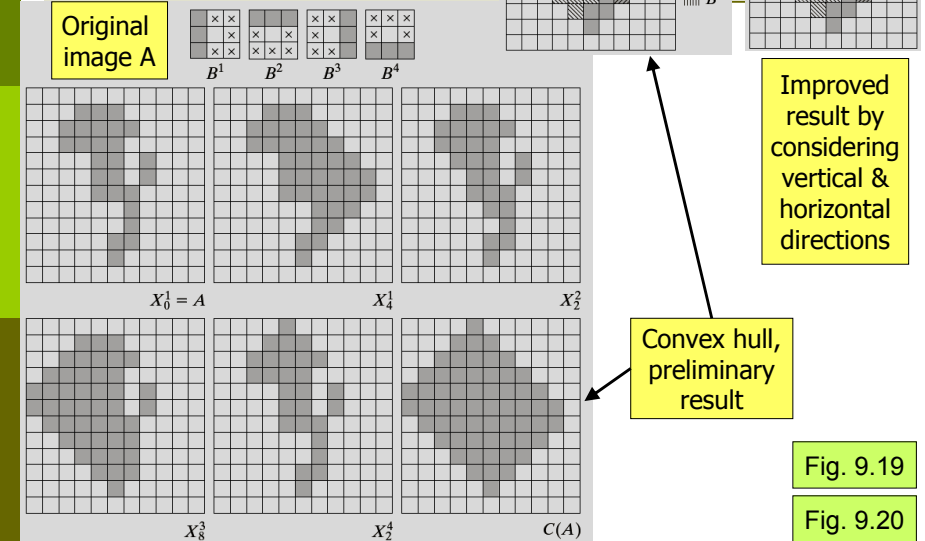
union by A

$$X_0^i = A$$

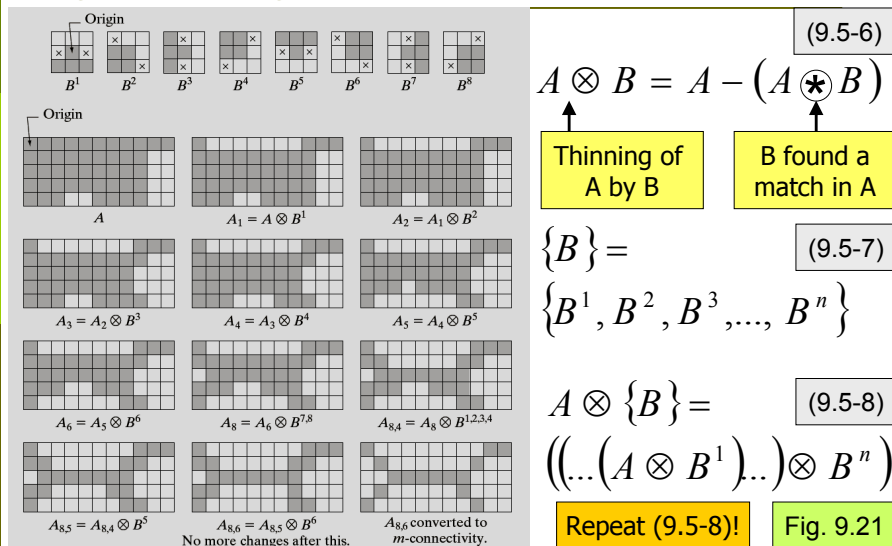
$$X_k^i = (X_{k-1} \star B^i) \cup A, \quad i = 1, 2, 3, \dots \quad k = 1, 2, 3, \dots \quad (9.5-4)$$

$$C(A) = \bigcup_{i=1}^4 D^i, \text{ where } D^i = X_k^i \text{ for } k \text{ converging} \quad (9.5-5)$$

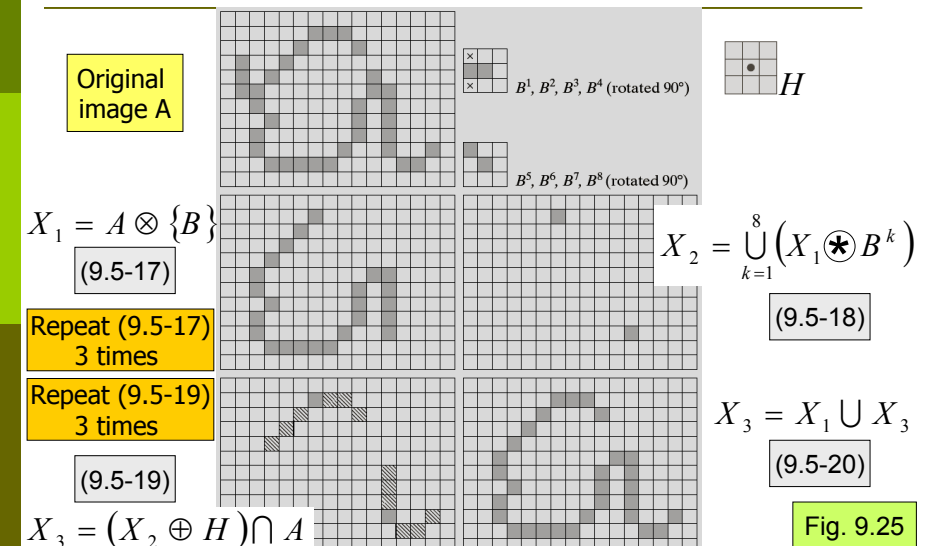
### 3) Convex hull



### 4) G&W's thinning, connectivity preserving shrinking (to 8-connective skeleton)

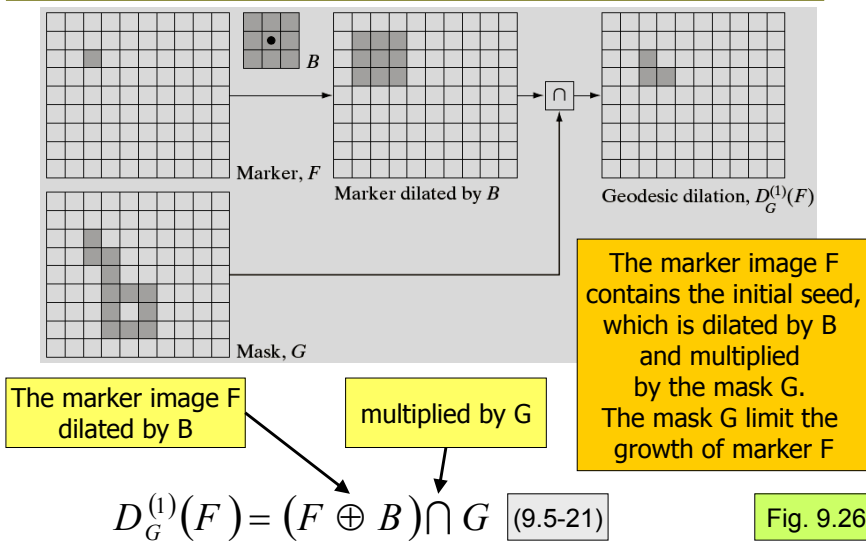


### 5) Pruning





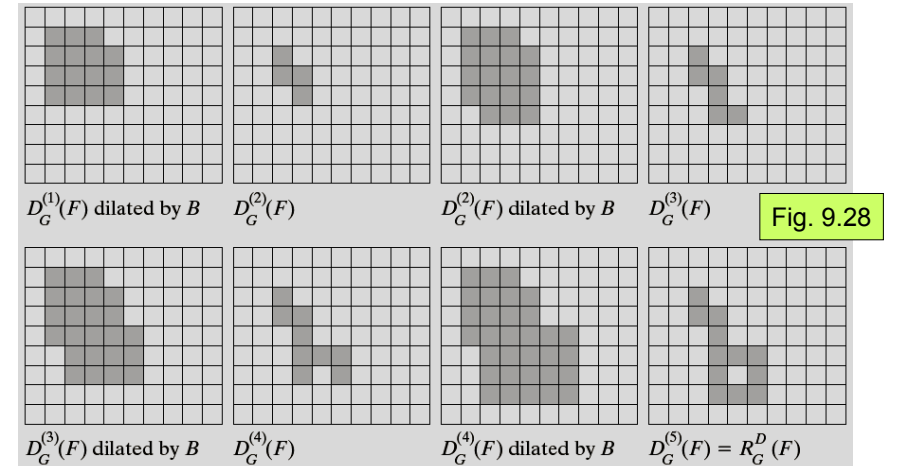
## 6a) Geodesic dilation of size 1



## 6b) Morphological reconstruction by geodesic dilation iterated until stability

Similar to (9.5-3)

when  $D_G^{(k)}(F) = D_G^{(k+1)}(F)$ :  $R_G^D(F) = D_G^{(k)}(F)$  (9.5-25)



## 7) Opening by reconstruction

Ex) Extracting characters with long vertical strokes

$$(9.5-27) \quad O_R^{(n)}(F) = R_F^D[F \ominus nB] \quad \text{n erosions by } B$$

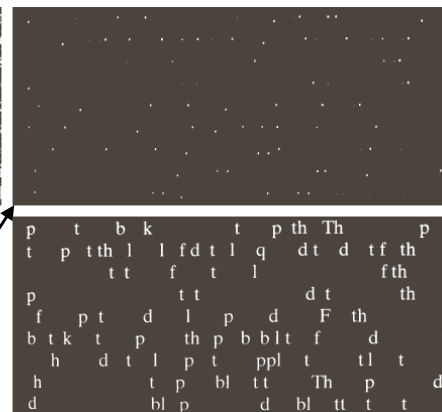
ponents or broken connection paths. There is no point past the level of detail required to identify those. Segmentation of nontrivial images is one of the most difficult problems in computerized analysis procedures. For this reason, it is often necessary to be taken to improve the probability of rugged segmentation, such as industrial inspection applications, at least some of the time. The experienced designer invariably pays considerable attention to such

Original image  $F$

1 erosion by a 51X1 element

Resulted image

Fig. 9.29



## 8) Filling holes

$$F(x, y) = \begin{cases} 1 - I(x, y), & \text{if } (x, y) \text{ is on the border of } I \\ 0, & \text{otherwise} \end{cases} \quad (9.5-28)$$

$$H = [R_{I^c}^D(F)]^c \quad (9.5-29)$$

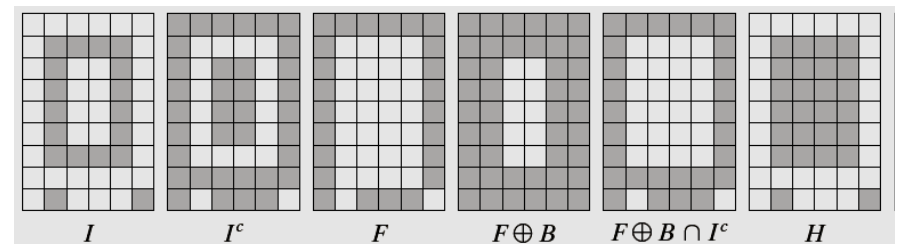


Fig. 9.30

## 8) Filling holes

Original image I

Complement image  $I^c$  used as a mask

ponents or broken connection paths. There is no position past the level of detail required to identify those. Segmentation of nontrivial images is one of the most difficult processing. Segmentation accuracy determines the evolution of computerized analysis procedures. For this reason, care must be taken to improve the probability of rugged segmentation, such as industrial inspection applications, at least some of the environment is possible at times. The experienced designer invariably pays considerable attention to such

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Marker image, F

white frame

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Resulted image

Fig. 9.31

## 9) Border clearing

$$F(x, y) = \begin{cases} I(x, y), & \text{if } (x, y) \text{ is on the border of } I \\ 0, & \text{otherwise} \end{cases} \quad (9.5-30)$$

$$X = I - R_I^D(F) \quad (9.5-31)$$



Image with objects touching the the border  $R_I^D(F)$

error in G&W

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Resulted image, X

Fig. 9.32