



# Object Segmentation

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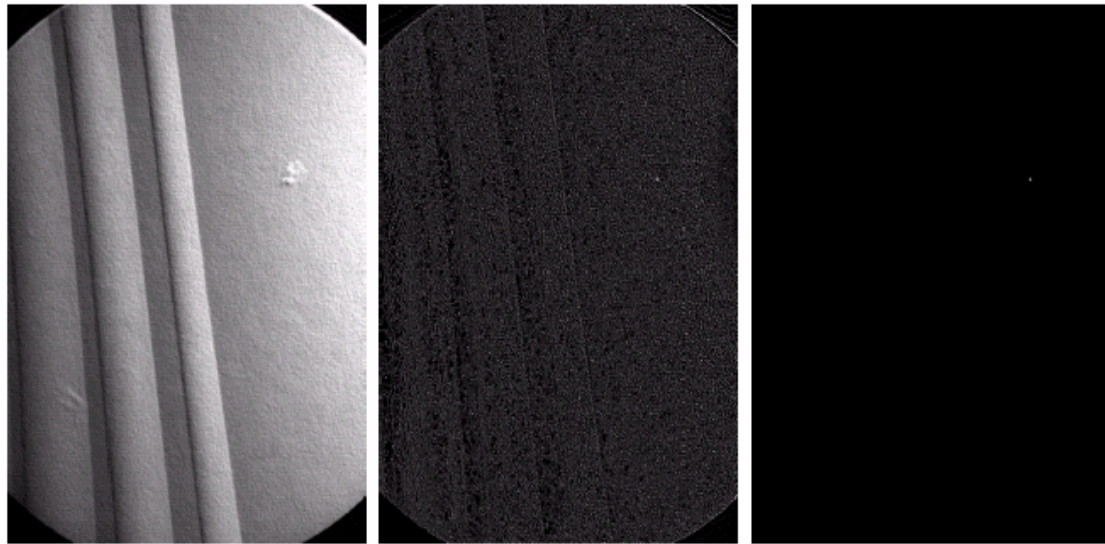
- Segmentation divides an image into regions or objects (segments)
- The degree of segmentation is highly application dependent
- Segmentation typically based on either *similarity* or *discontinuity*
- All images from Digital Image Processing, Gonzalez and Woods

- Assume grey level discontinuities
- Other discontinuities are possible
  - texture
  - curvature
  - any measurement of the image...
- A discontinuity is merely where the image or measurement changes quickly

# Points

- An isolated discontinuity in which a single pixel differs greatly from its neighbors
- Use a “point mask”
- Measure the response and threshold
- Basic pattern matching

# A Hole in a Jet Turbine



-1	-1	-1
-1	8	-1
-1	-1	-1

**FIGURE 10.2**  
 (a) Point detection mask.  
 (b) X-ray image of a turbine blade with a porosity.  
 (c) Result of point detection.  
 (d) Result of using Eq. (10.1-2).  
 (Original image courtesy of X-TEK Systems Ltd.)

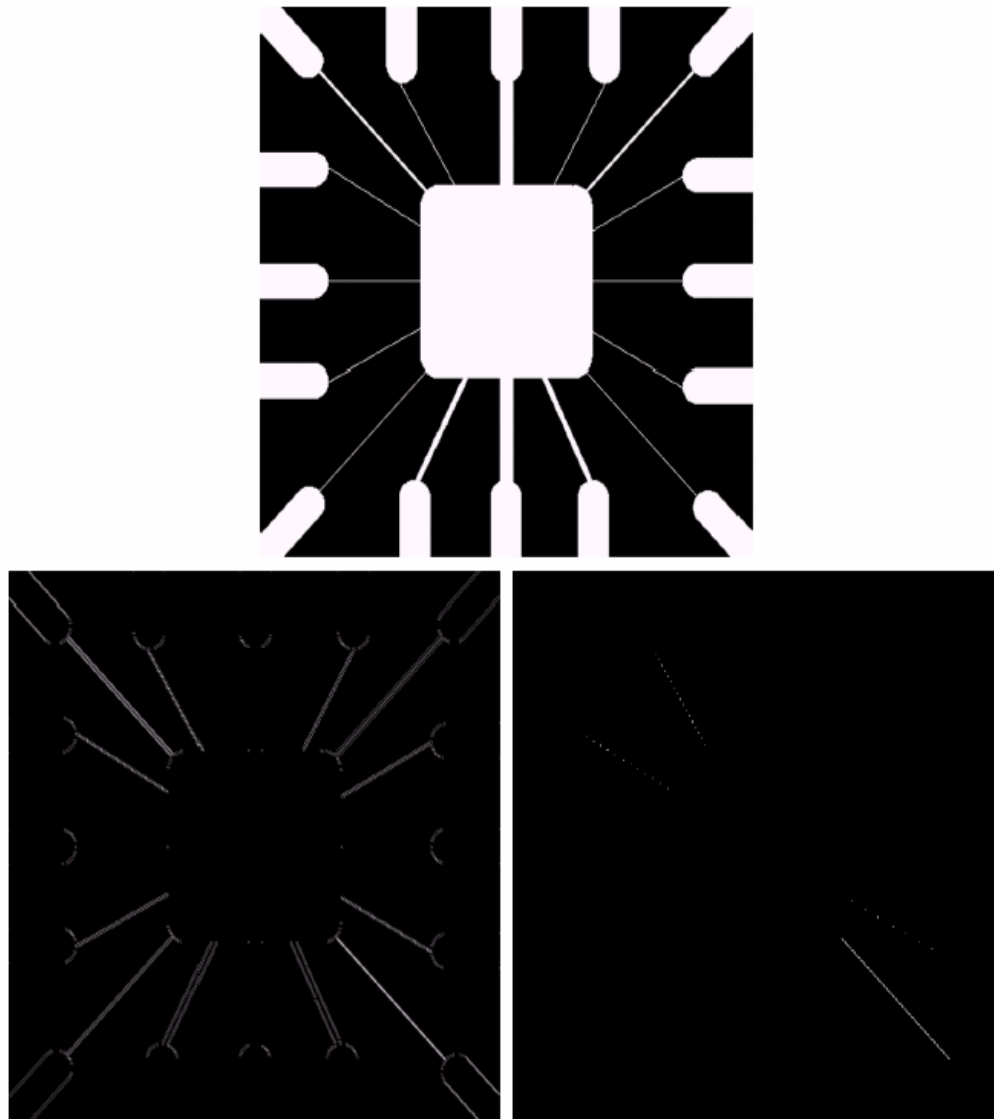
- Lines are connected points
- Points on a line are similar to neighbors along the line, but different than neighbors off the line
- Use a “line mask”
- Need a separate mask for each direction

# 4 Line Masks

**FIGURE 10.3** Line masks.

-1	-1	-1	-1	-1	2	-1	2	-1	2	-1	-1
2	2	2	-1	2	-1	-1	2	-1	-1	2	-1
-1	-1	-1	2	-1	-1	-1	2	-1	-1	-1	2
Horizontal			+45°			Vertical			-45°		

# Binary Wirebond Mask



a  
b c

**FIGURE 10.4**

Illustration of line detection.

(a) Binary wirebond mask.

(b) Absolute value of result after processing with  $-45^\circ$  line detector.

(c) Result of thresholding image (b).



- Edge represents the discontinuity between two regions
- Typically between an “object” and its “background” although objects can also share an edge
- Realistically, an edge represents a gradual discontinuity

# Edge Masks

a
b c
d e
f g

**FIGURE 10.8**  
A  $3 \times 3$  region of an image (the  $z$ 's are gray-level values) and various masks used to compute the gradient at point labeled  $z_5$ .

$z_1$	$z_2$	$z_3$
$z_4$	$z_5$	$z_6$
$z_7$	$z_8$	$z_9$

-1	0	0	-1
0	1	1	0

Roberts

-1	-1	-1	-1	0	1
0	0	0	-1	0	1
1	1	1	-1	0	1

Prewitt

-1	-2	-1	-1	0	1
0	0	0	-2	0	2
1	2	1	-1	0	1

Sobel

# Diagonal Masks

0	1	1	-1	-1	0
-1	0	1	-1	0	1
-1	-1	0	0	1	1

Prewitt

0	1	2	-2	-1	0
-1	0	1	-1	0	1
-2	-1	0	0	1	2

Sobel

a	b
c	d

**FIGURE 10.9** Prewitt and Sobel masks for detecting diagonal edges.

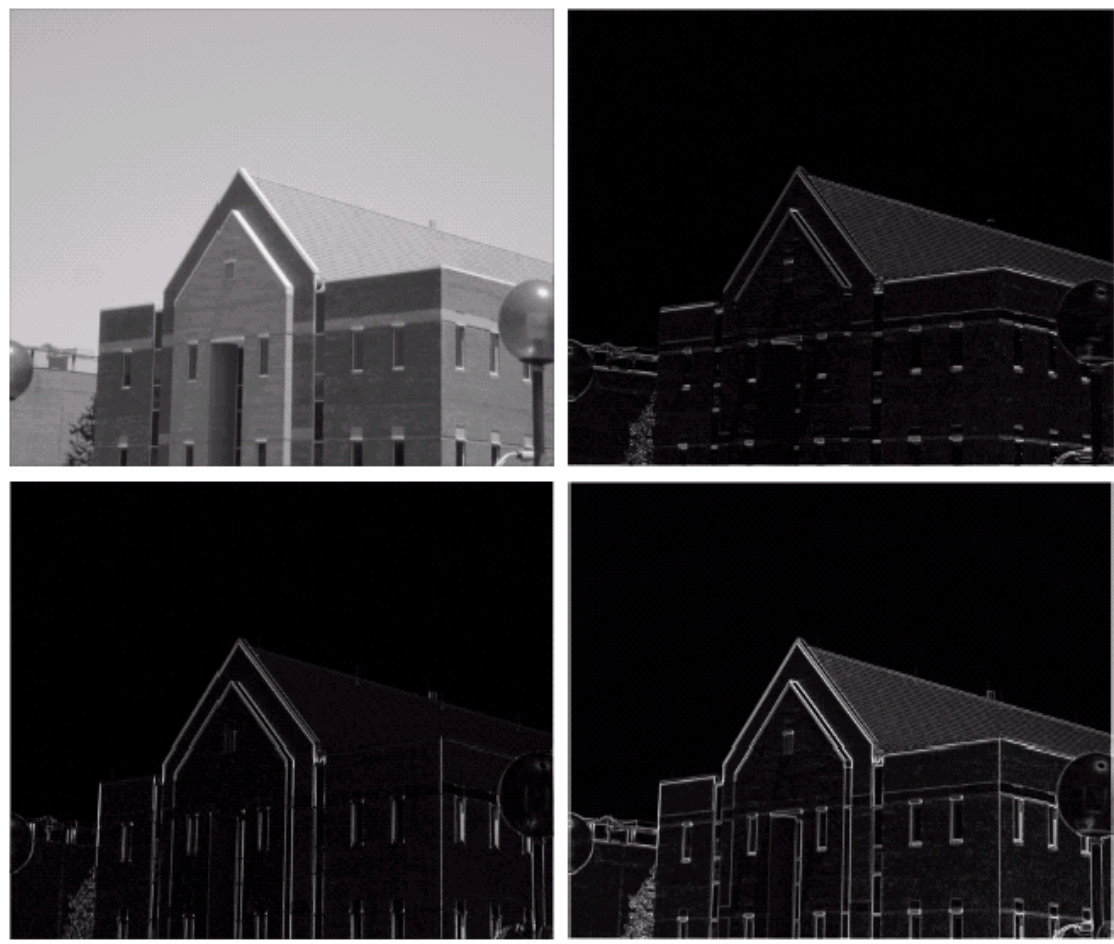
# Application of Gradient

a b  
 c d

**FIGURE 10.10**  
 (a) Original image.  
 (b)  $|G_x|$ , component of the gradient in the  $x$ -direction.  
 (c)  $|G_y|$ , component in the  $y$ -direction.  
 (d) Gradient image,  $|G_x| + |G_y|$ .



# Smoothed Gradient



a b  
c d

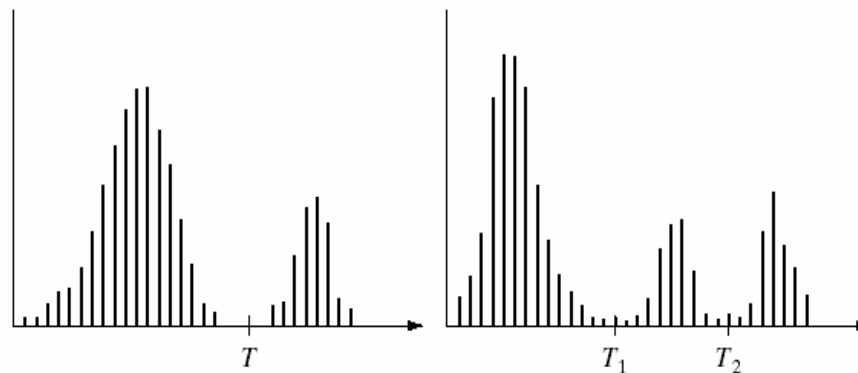
**FIGURE 10.11**  
Same sequence as in Fig. 10.10, but with the original image smoothed with a  $5 \times 5$  averaging filter.

- Look at each edge pixel
  - strength of gradient response
  - direction of gradient
- If the difference for both these criteria is small in neighboring edge pixels, the pixels are part of the same edge
- Keep track of all linked edges

- Simple and intuitive to apply
- May be binary (object and background) or multilevel
- May be
  - global (depends only on intensity)
  - local (also depends on other quantities)
  - adaptive (also depends on pixel position)

# Histogram

- Histograms provide a good visual idea of why thresholding works
- Separate the image based on intensities that fall in the valley(s) of the histogram



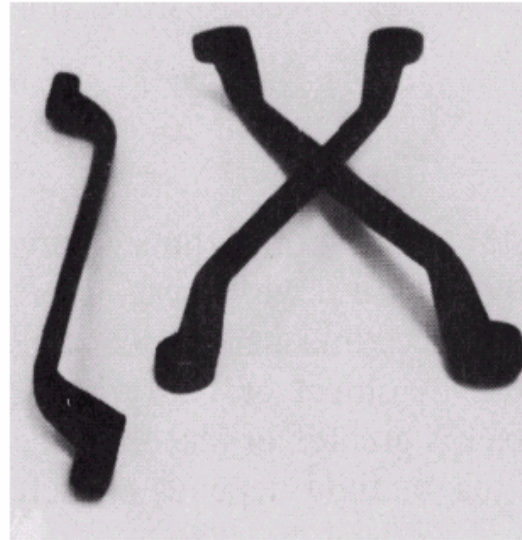
a b

**FIGURE 10.26** (a) Gray-level histograms that can be partitioned by (a) a single threshold, and (b) multiple thresholds.



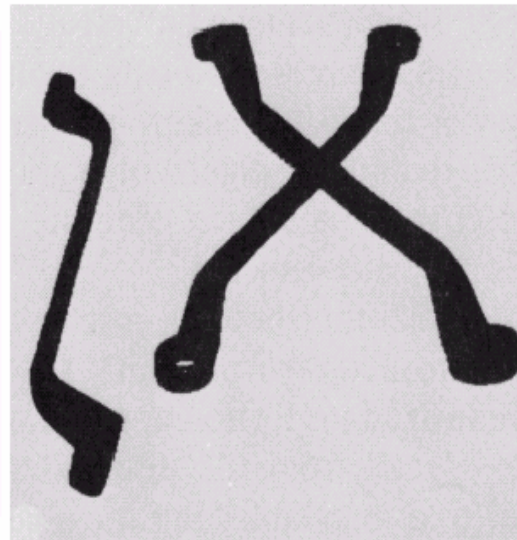
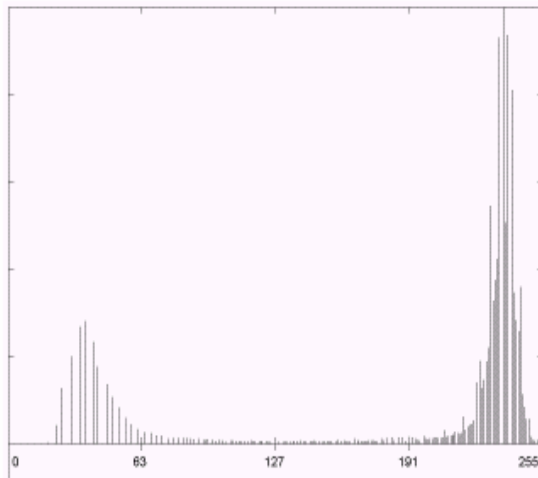
- Uses a single global threshold to separate object from background
- Threshold may be known *a priori* or estimated
  - average grey level of image (when object and background are the same size)
  - midway between minimum and maximum grey level
  - iteratively calculated from an initial guess

# Basic Segmentation



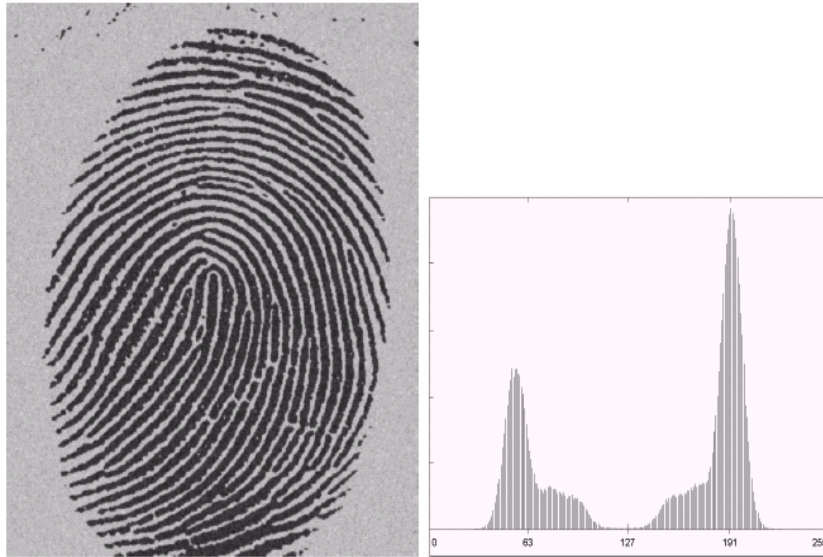
a  
b c

**FIGURE 10.28**  
 (a) Original image. (b) Image histogram.  
 (c) Result of global thresholding with  $T$  midway between the maximum and minimum gray levels.



- Start with an initial guess
- Segment the image
- Compute the average grey levels of the segmented regions
- Compute a new threshold as the average of the average grey levels
- Segment the image...
- Stop when the threshold stops changing

# Iterated Threshold



a b  
c

**FIGURE 10.29**  
(a) Original image. (b) Image histogram. (c) Result of segmentation with the threshold estimated by iteration. (Original courtesy of the National Institute of Standards and Technology.)



- If more than one variable for each pixel is available, threshold on multiple values
  - MRI
  - Color (RGB, HIS)
  - Color and IR
- Use a cluster analysis to find “clumps” in the multidimensional histogram

# RGB Thresholding



a b c

**FIGURE 10.39** (a) Original color image shown as a monochrome picture. (b) Segmentation of pixels with colors close to facial tones. (c) Segmentation of red components.

- Use notions of logic and set theory to break the image into regions
  - All pixels must be classified
  - Regions must be connected
  - Regions must be disjoint
  - All pixels in the same region share a property that pixels in different regions do not share
- Grow or split regions until these conditions are satisfied

- Start with seed pixels
- Grow each seed into a region of “similar” pixels, keeping connectivity
- Keep growing regions until no more pixels can be added
- Requires a good selection of seed points to identify all regions
- Regions may be merged

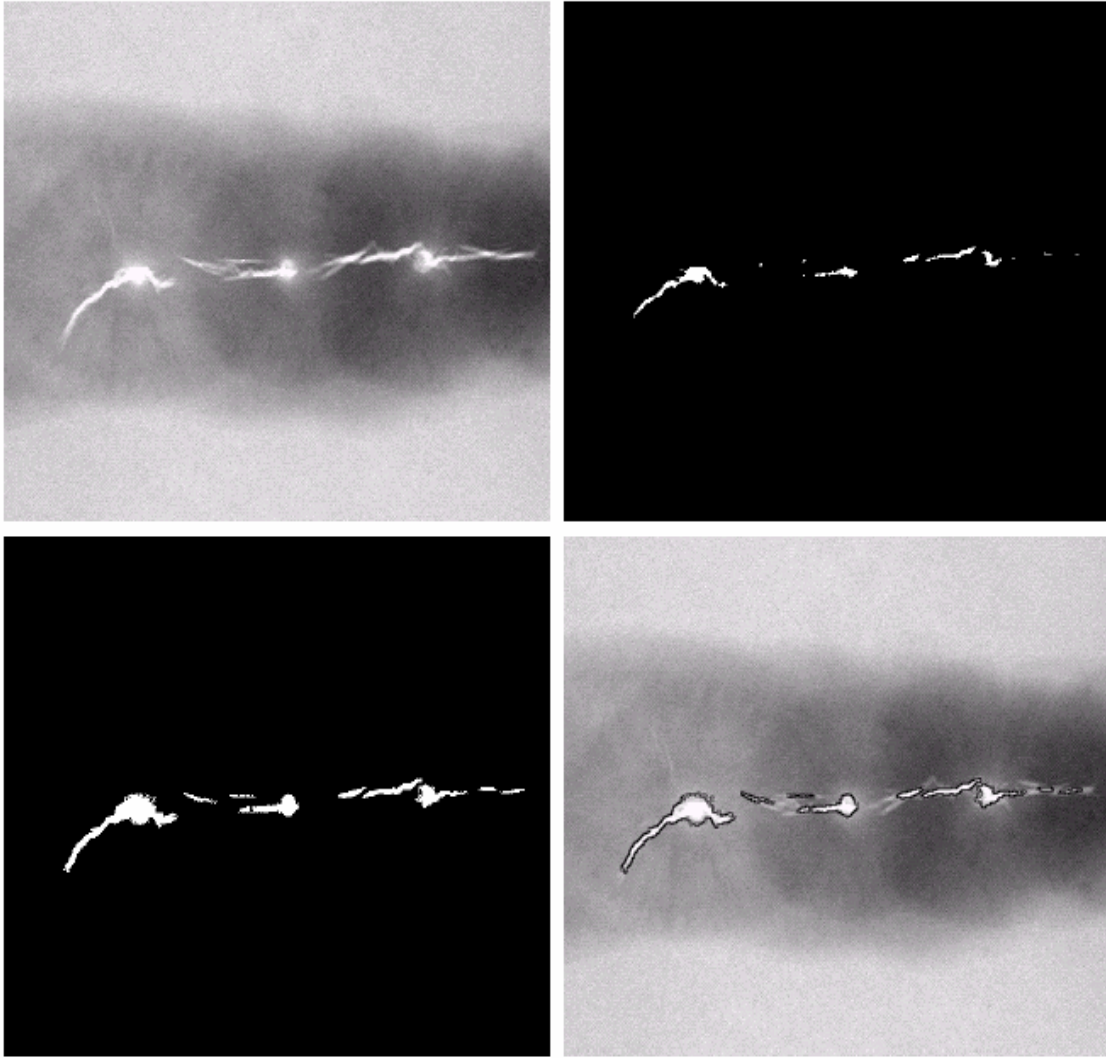


# Defective Welds

a b  
 c d

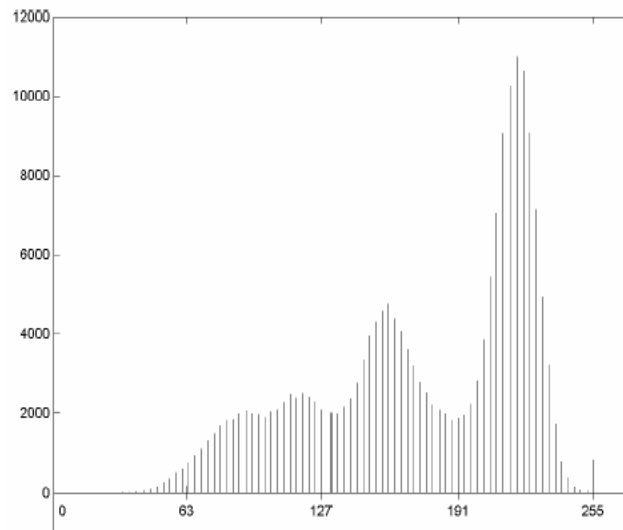
**FIGURE 10.40**

(a) Image showing defective welds. (b) Seed points. (c) Result of region growing. (d) Boundaries of segmented defective welds (in black). (Original image courtesy of X-TEK Systems, Ltd.).



# Welding Histogram

- Histogram used to determine the similarity criterion
- Connectivity provided the needed improvement over simple thresholding



**FIGURE 10.41**  
 Histogram of  
 Fig. 10.40(a).

- Subdivide the image into smaller and smaller regions until the regions are all “self-similar”, then join any regions that share the same “similarity”
- Use a quadtree data structure to manage the blocks
- “Predefined” square regions on which to investigate similarity

# Segmented Leaf

a b c

**FIGURE 10.43**

(a) Original image. (b) Result of split and merge procedure. (c) Result of thresholding (a).

