

## What we discussed last time

- Image acquisition, formation and sensing
- Sampling, quantisation
- Image representation
- Spatial, Intensity and temporal resolution

Q1. Consider an image with 100 lines and 1000 pixels per line. Each pixel can take 256 different values. The total amount of bits needed to store that image is \_\_\_\_ ?

Q2. A color video has frame rate of 30 frames per second. Considering that spatial resolution of each frame (images) is  $1000 \times 1000$  pixels and every color component is stored using 8 bits then a minute of (uncompressed) video will occupy ?

Q3. An imaging system and object surface has illumination and reflectance range as:

$$1 \leq i(x, y) \leq 100 \text{ and } 0.5 \leq r(x, y) \leq 1$$

The acquired image is stored as grayscale with 8 bits per pixel. An intensity value of say 25 at any location will get stored as \_\_\_\_?

This lecture will cover:

- Neighbourhood and Connectivity
- Connected component labeling
- Distance measures
- Basic Image Operations

# Neighbors of a Pixel

A pixel  $p$  at coordinates  $(x,y)$  has four *horizontal* and *vertical* neighbors whose coordinates are given by:

$(x+1,y)$ ,  $(x-1, y)$ ,  $(x, y+1)$ ,  $(x,y-1)$

	$(x, y-1)$	
$(x-1, y)$	$P(x,y)$	$(x+1, y)$
	$(x, y+1)$	

This set of pixels, called the *4-neighbors* of  $p$ , is denoted by  $N_4(p)$ .

# Neighbors of a Pixel

The four *diagonal* neighbors of  $p$  have coordinates:

$(x+1, y+1), (x+1, y-1), (x-1, y+1), (x-1, y-1)$

and are denoted by  $N_D(p)$ .

$(x-1, y+1)$		$(x+1, y-1)$
	$P(x, y)$	
$(x-1, y-1)$		$(x+1, y+1)$

These points, together with the 4-neighbors, are called the 8-neighbors of  $p$ , denoted by  $N_8(p)$ .

$(x-1, y+1)$	$(x, y-1)$	$(x+1, y-1)$
$(x-1, y)$	$P(x, y)$	$(x+1, y)$
$(x-1, y-1)$	$(x, y+1)$	$(x+1, y+1)$

Connectivity between pixels is important:

- Because it is used to determine components/regions in an image, trace contours and establish boundaries of objects

Two pixels are connected if:

- They are “neighbors” in some sense (i.e.  $N_4(p)$ ,  $N_8(p)$ , ...)
- Their gray levels satisfy a specified criterion of similarity (e.g. equality, ...)

Let  $V$ : a set of intensity values used to define adjacency.

In a binary image,  $V = \{1\}$ , if we are referring to adjacency of pixels with value 1.

In a gray-scale image, the idea is the same, but  $V$  typically contains more elements, for example,  $V = \{180, 181, 182, \dots, 200\}$

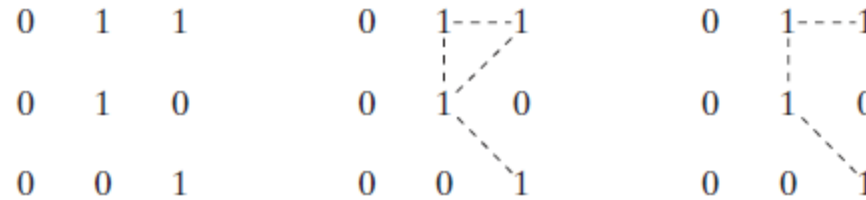


1. **4-adjacency:** Two pixels  $p$  and  $q$  with values from  $V$  are 4-adjacent if  $q$  is in the set  $N_4(p)$ .
2. **8-adjacency:** Two pixels  $p$  and  $q$  with values from  $V$  are 8-adjacent if  $q$  is in the set  $N_8(p)$ .
3. **m-adjacency (mixed):** Two pixels  $p$  and  $q$  with values from  $V$  are m-adjacent if :
  - $q$  is in  $N_4(p)$  **or**
  - $q$  is in  $N_D(p)$  **and** the set  $N_4(p) \cap N_4(q)$  has no pixel whose values are from  $V$  (no intersection)

# Types of Adjacency

Mixed adjacency eliminates the ambiguities that often arise when 8-adjacency is used.

Eg.



a b c

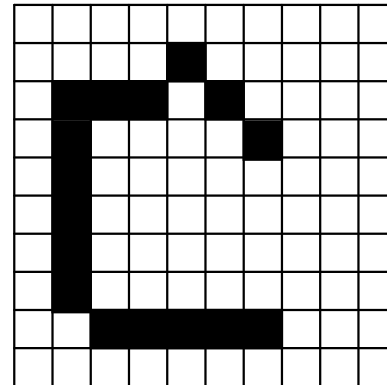
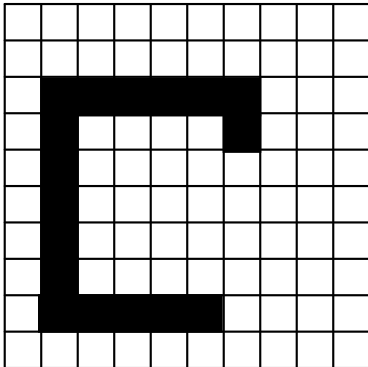
**FIGURE 2.26** (a) Arrangement of pixels; (b) pixels that are 8-adjacent (shown dashed) to the center pixel; (c) *m*-adjacency.

A digital path (or curve) from pixel  $p$  with coordinate  $(x,y)$  to pixel  $q$  with coordinate  $(s,t)$  is a sequence of distinct pixels:

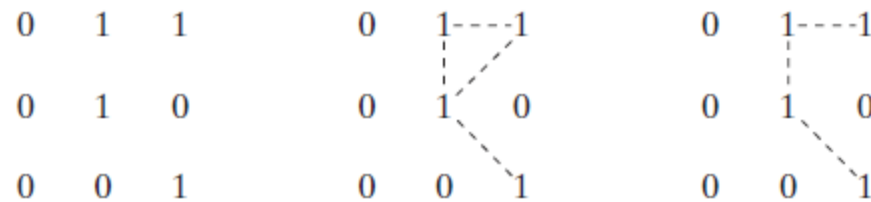
$$(x_0, y_0), (x_1, y_1), \dots, (x_n, y_n),$$

where  $(x_0, y_0) = (x, y)$  and  $(x_n, y_n) = (s, t)$  and

$(x_i, y_i)$  and  $(x_{i-1}, y_{i-1})$  are adjacent for  $1 \leq i \leq n$



Return to the previous example:



a b c

**FIGURE 2.26** (a) Arrangement of pixels; (b) pixels that are 8-adjacent (shown dashed) to the center pixel; (c) *m*-adjacency.

In figure (b) the paths between the top right and bottom right pixels are 8-paths. And the path between the same 2 pixels in figure (c) is m-path

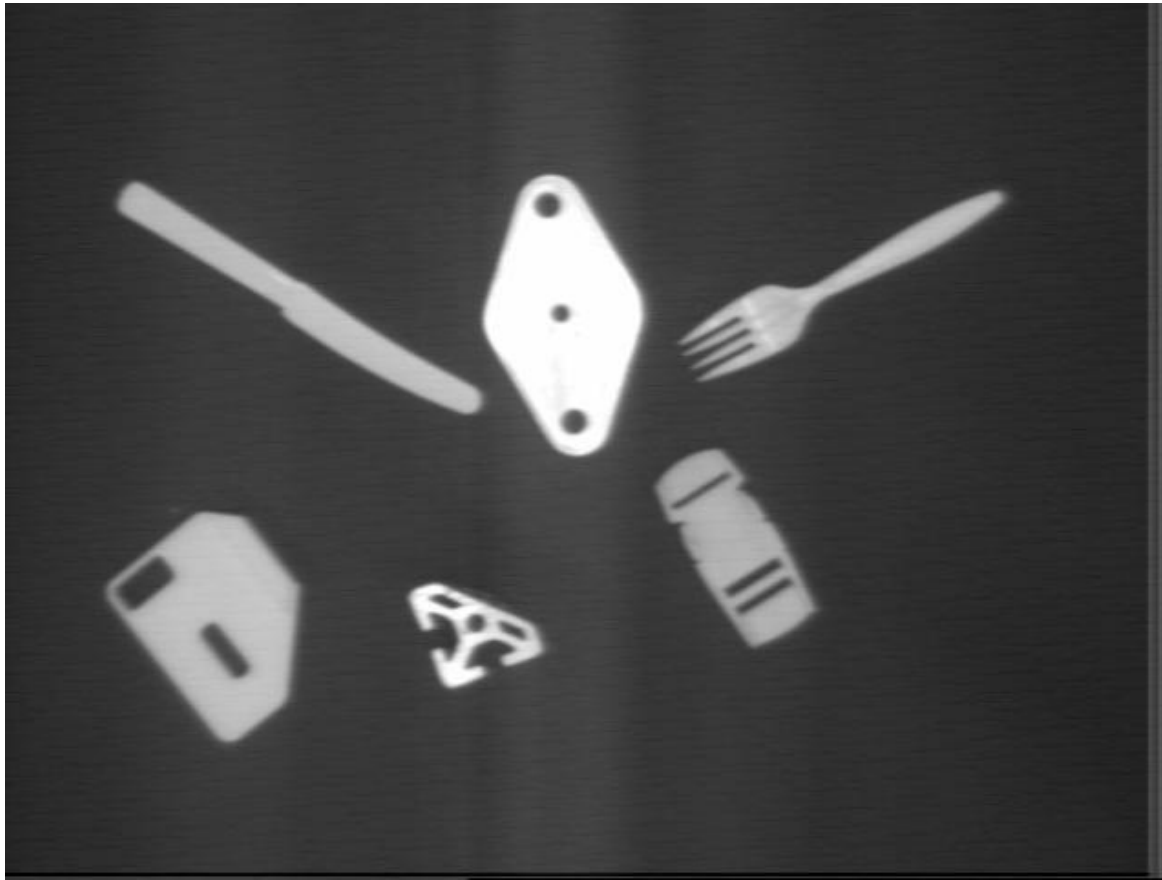
- Let  $R$  represent a subset of pixels in an image.
- A pixel pair  $p$  and  $q$  are said to be connected in  $R$  if there exists a path between them consisting entirely of pixels in  $R$ .
- A **connected component** is a maximal set of pixels in  $R$  that is connected.
- There can be more than one such set within a given  $R$ .
- Set  $R$  is a **connected set** if it has only one connected component.

## Region

Let  $R$  be a subset of pixels, we call  $R$  a region of the image if  $R$  is a connected set.

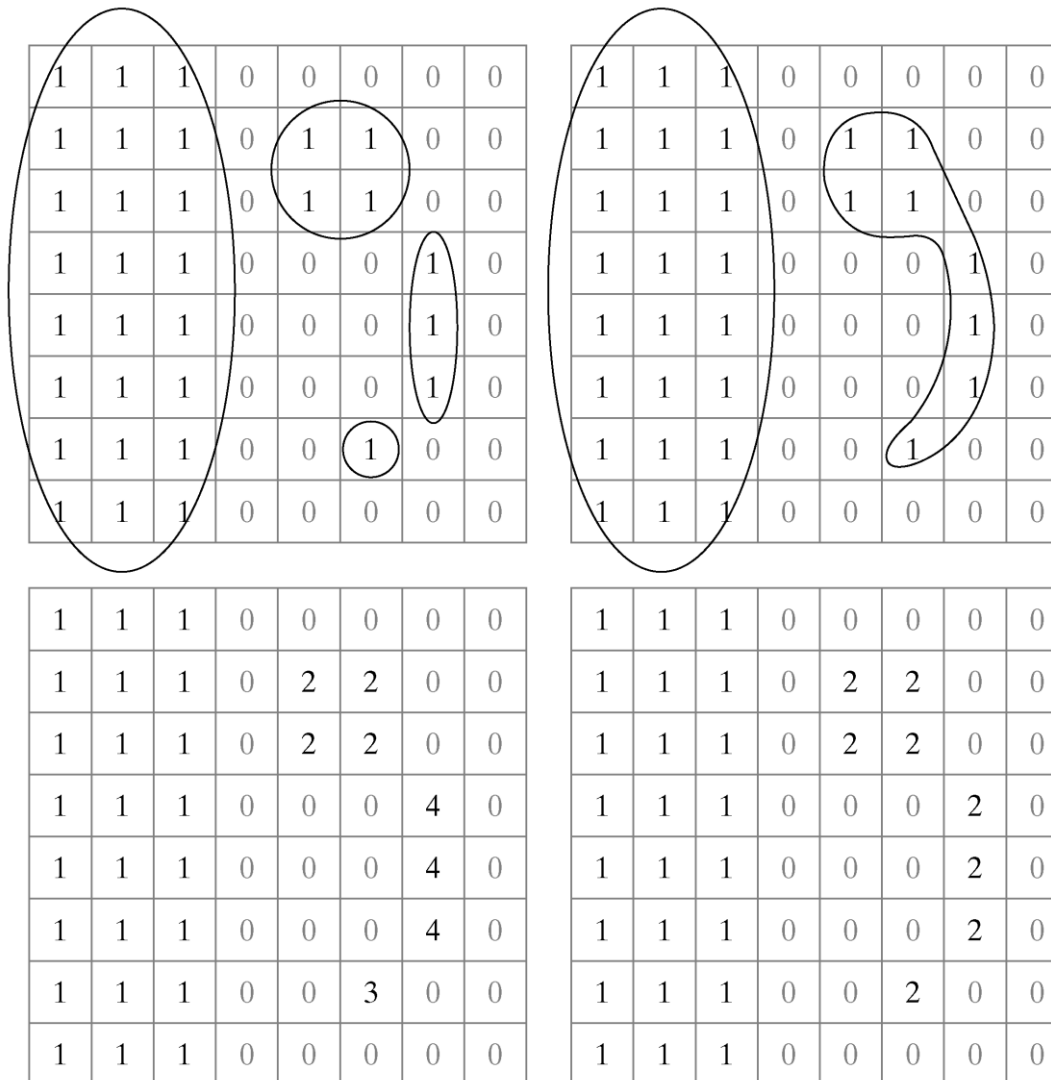
## Boundary

The *boundary* (also called *border* or *contour*) of a region  $R$  is the set of pixels in the region that have one or more neighbors that are not in  $R$ .



Need to **SEGMENT** image into separate **COMPONENTS**(regions)

# Connected Components



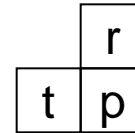
a b  
c d

**FIGURE 9.19**  
Connected components  
(a) Four 4-connected components.  
(b) Two 8-connected components.  
(c) Label matrix obtained using 4-connectivity  
(d) Label matrix obtained using 8-connectivity.



# Connected Components Labeling

## First Pass



- Examine pixels in scanline order
- When we visit point p, points r and t have been visited and labeled.
- If  $p=0$ : no action;
- If  $p=1$ : check r and t.
  - both r and t = 0; assign new label to p;
  - only one of r and t is a 1. assign its label to p;
  - both r and t are 1:
    - same label => assign it to p;
    - different label => assign lowest no. to p and

	1					2	2				3	3	3	3	3	3	3
4	1	1				5	2	2	2		3						
	1				6	5			2	2	3		7	7	7	7	7
						5	5	5	2		3		7				7
8		9					5	5			3		7		10		7
8		9									3		7	7	10		7
8	8	9									3						7
	8										3	3	3	3	3	3	7

(a)

Mark labels in neighbors as equivalent (they are the same.)

# Connected Components Labeling

## Resolve Equivalences

- **Union-Find** Structure and operations
- Obtain transitive closure using **Floyd-Warshall** (F-W) algorithm

$$4 \equiv 1$$

$$6 \equiv 5, 2$$

$$9 \equiv 8$$

$$10 \equiv 7, 3$$

## Second Pass

- examine the pixels in scanline order
- Replace the label with equivalent component label

	1					2	2				3	3	3	3	3	3	3
1	1	1				2	2	2	2		3						
	1				2	2			2	2	3		3	3	3	3	3
					2	2	2	2			3		3				3
8		8				2	2				3		3		3		3
8		8									3		3	3	3		3
8	8	8									3						3
	8										3	3	3	3	3	3	3

(d)

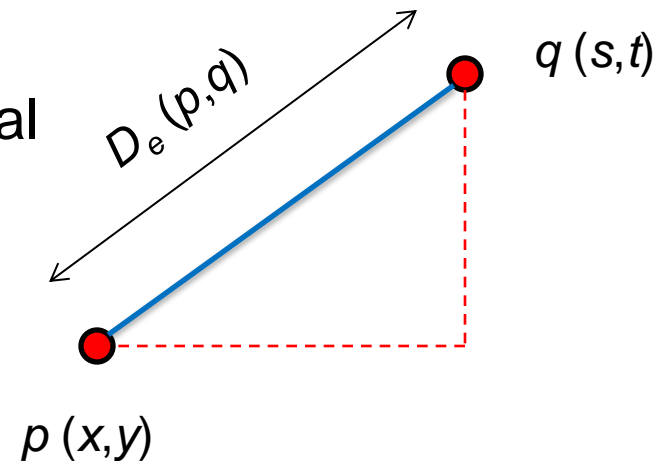
For pixels  $p$ ,  $q$  and  $z$ , with coordinates  $(x,y)$ ,  $(s,t)$  and  $(v,w)$ , respectively,  $D$  is a distance function if:

- (a)  $D(p,q) \geq 0$  ( $D(p,q) = 0$  iff  $p = q$ ),
- (b)  $D(p,q) = D(q,p)$ , and
- (c)  $D(p,z) \leq D(p,q) + D(q,z)$ .

The ***Euclidean Distance*** between  $p$  and  $q$  is defined as:

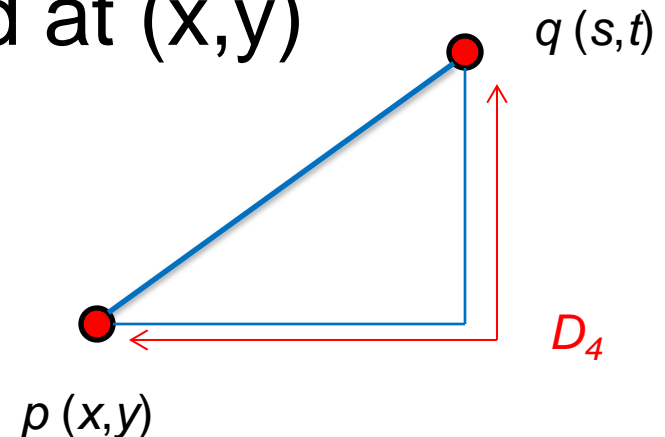
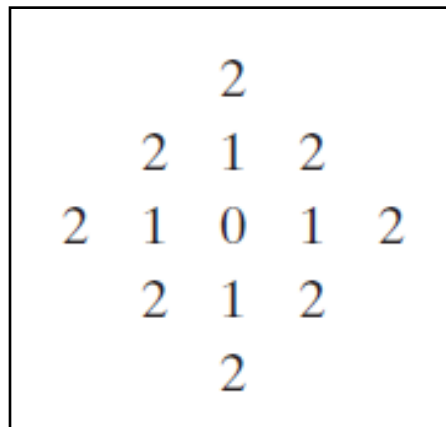
$$D_e(p, q) = [(x - s)^2 + (y - t)^2]^{1/2}$$

Pixels having a distance less than or equal to some value  $r$  from  $(x, y)$  are the points contained in a disk of radius  $r$  centered at  $(x, y)$



The  **$D_4$  distance** (also called **city-block distance**) between  $p$  and  $q$  is defined as:

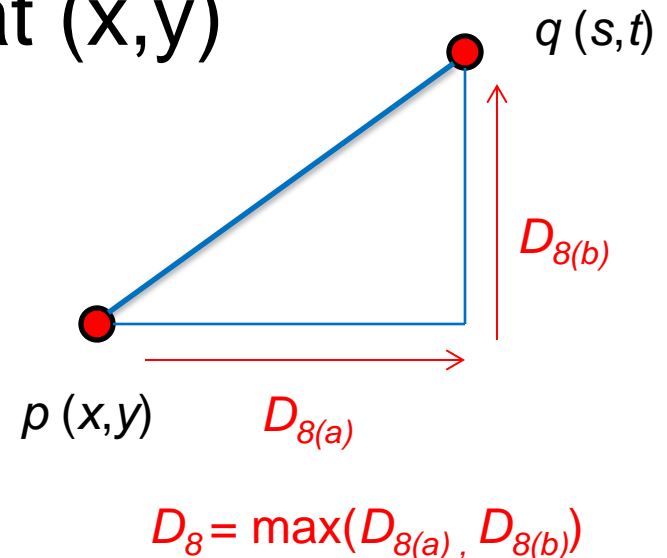
- $D_4(p, q) = |x - s| + |y - t|$
- forms a Diamond centered at  $(x, y)$
- e.g. pixels with  $D_4 \leq 2$  from  $p$

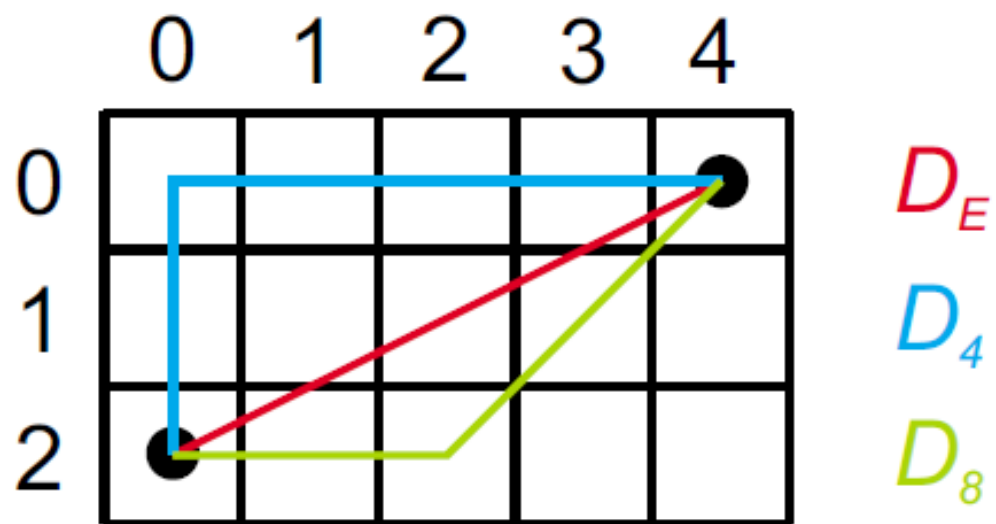


The  **$D_8$  distance** (also called **chessboard distance**) between  $p$  and  $q$  is defined as:

- $D_8(p, q) = \max(|x - s|, |y - t|)$
- forms a square centered at  $(x, y)$
- e.g. pixels with  $D_8 \leq 2$  from  $p$

2	2	2	2	2
2	1	1	1	2
2	1	0	1	2
2	1	1	1	2
2	2	2	2	2



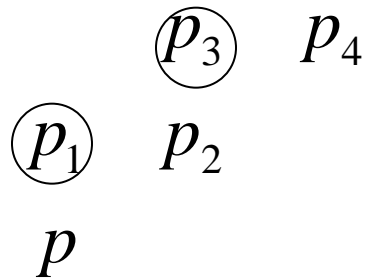


$D_4$  and  $D_8$  distances between  $p$  and  $q$  are independent of any paths that exist between the points

**$D_m$  distance:** is defined as the shortest  $m$ -path between the points. In this case, the distance between two pixels will depend on the values of the pixels along the path, as well as the values of their neighbors.



e.g. assume  $p, p_2, p_4 = 1$   
 $p_1, p_3 =$  can have either 0 or 1

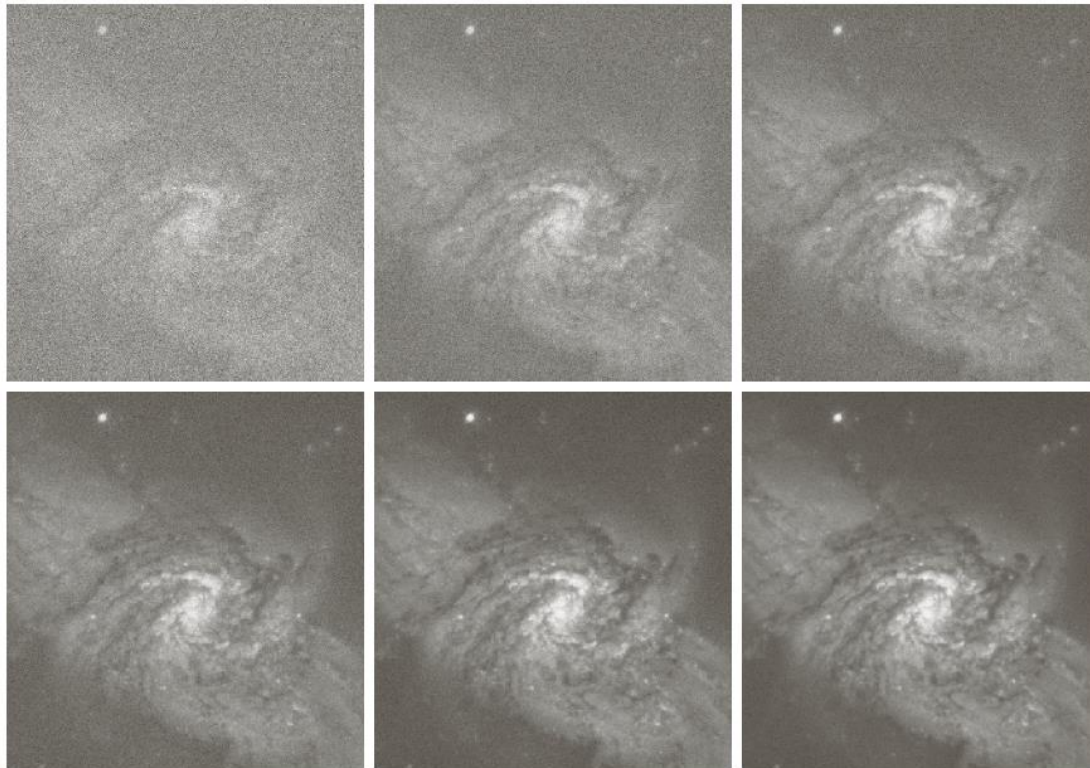


If only connectivity of pixels valued 1 is allowed, and  $p_1$  and  $p_3$  are 0, the m-distance between  $p$  and  $p_4$  is 2.

If either  $p_1$  or  $p_3$  is 1, the distance is 3.

If both  $p_1$  and  $p_3$  are 1, the distance is 4  
 ( $pp_1p_2p_3p_4$ )

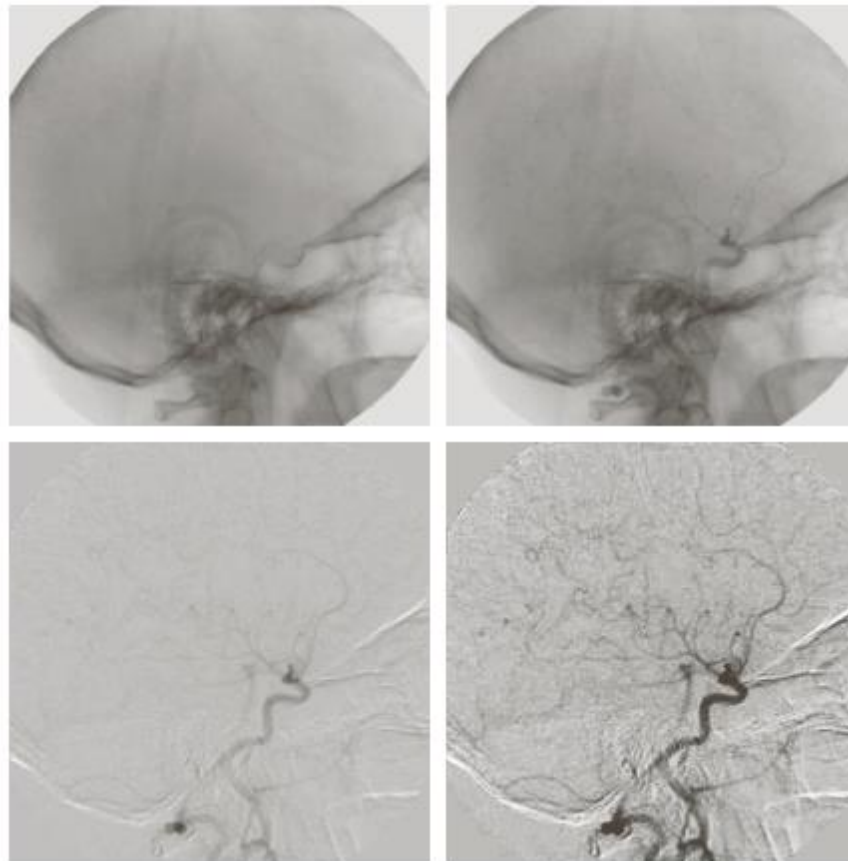
## Arithmetic Operations - Addition



a	b	c
d	e	f

**FIGURE 2.26** (a) Image of Galaxy Pair NGC 3314 corrupted by additive Gaussian noise. (b)–(f) Results of averaging 5, 10, 20, 50, and 100 noisy images, respectively. (Original image courtesy of NASA.)

## Arithmetic Operations - Subtraction



a	b
c	d

**FIGURE 2.28**

Digital subtraction angiography. (a) Mask image. (b) A live image. (c) Difference between (a) and (b). (d) Enhanced difference image. (Figures (a) and (b) courtesy of The Image Sciences Institute, University Medical Center, Utrecht, The Netherlands.)

## Arithmetic Operations - Multiplication

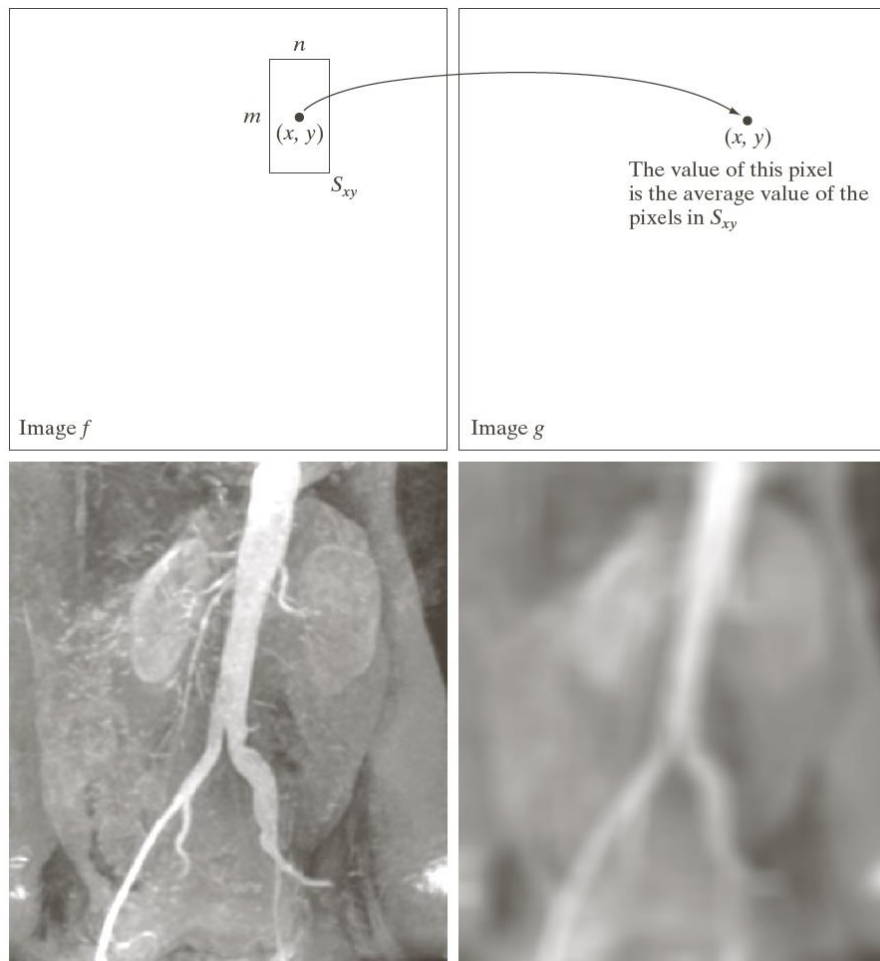


a b c

**FIGURE 2.30** (a) Digital dental X-ray image. (b) ROI mask for isolating teeth with fillings (white corresponds to 1 and black corresponds to 0). (c) Product of (a) and (b).

# Basic Image Operations

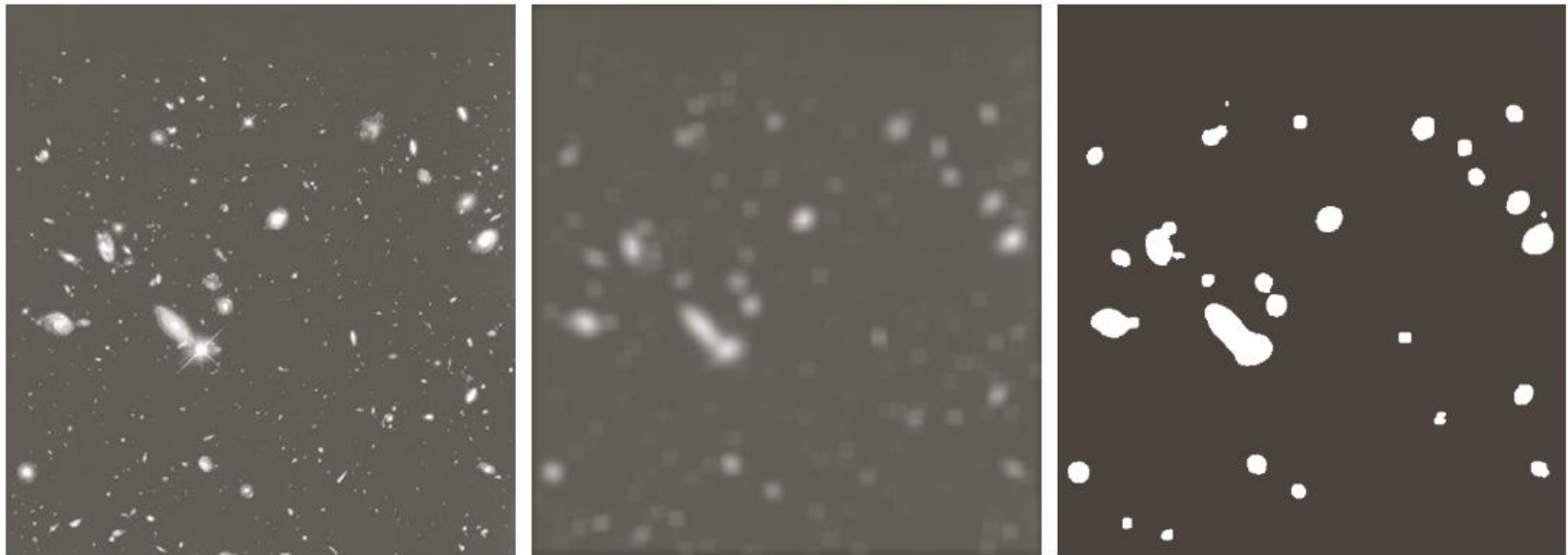
## Spatial Operations – Neighborhood operations



**FIGURE 2.35**

Local averaging using neighborhood processing. The procedure is illustrated in (a) and (b) for a rectangular neighborhood. (c) The aortic angiogram discussed in Section 1.3.2. (d) The result of using Eq. (2.6-21) with  $m = n = 41$ . The images are of size  $790 \times 686$  pixels.

## Spatial Operations – Neighborhood operations



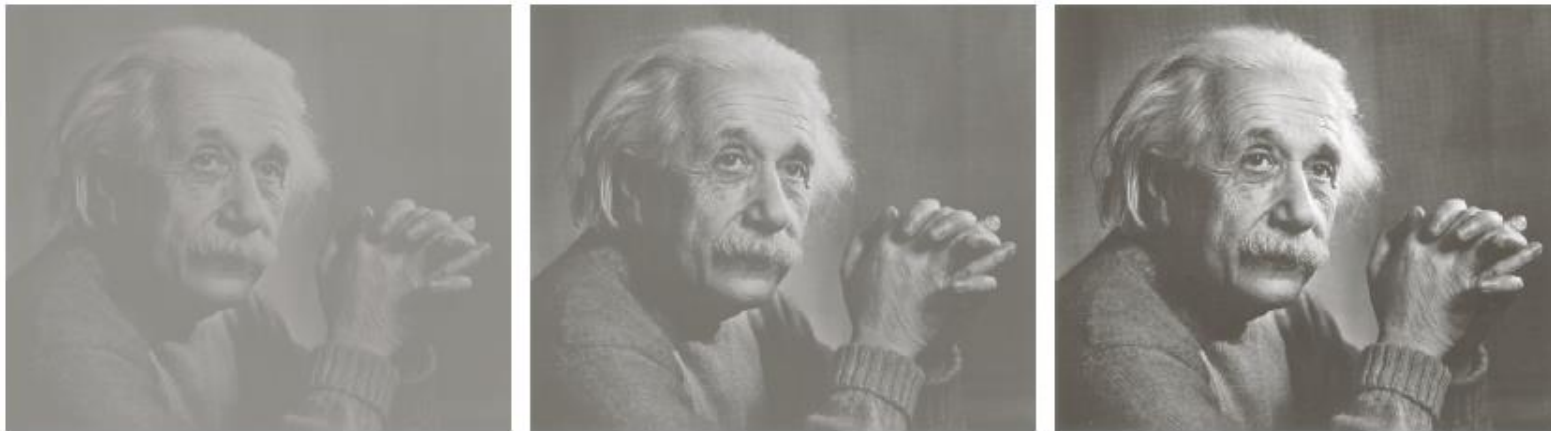
a b c

**FIGURE 3.34** (a) Image of size  $528 \times 485$  pixels from the Hubble Space Telescope. (b) Image filtered with a  $15 \times 15$  averaging mask. (c) Result of thresholding (b). (Original image courtesy of NASA.)

# Basic Image Operations

**Probabilistic methods** – used in many ways.

Ex.- Treat intensity values as random variable, then the mean and variance can be useful measure of image characteristics.



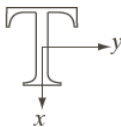
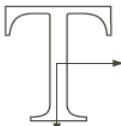

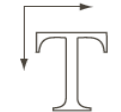


a b c

**FIGURE 2.41**  
Images exhibiting  
(a) low contrast,  
(b) medium  
contrast, and  
(c) high contrast.

## Spatial Operations – Geometric transformations

**TABLE 2.2**

Affine transformations based on Eq. (2.6.–23).

Transformation Name	Affine Matrix, T	Coordinate Equations	Example
Identity	$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$x = v$ $y = w$	
Scaling	$\begin{bmatrix} c_x & 0 & 0 \\ 0 & c_y & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$x = c_x v$ $y = c_y w$	
Rotation	$\begin{bmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$x = v \cos \theta - w \sin \theta$ $y = v \sin \theta + w \cos \theta$	
Translation	$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ t_x & t_y & 1 \end{bmatrix}$	$x = v + t_x$ $y = w + t_y$	
Shear (vertical)	$\begin{bmatrix} 1 & 0 & 0 \\ s_v & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$x = v + s_v w$ $y = w$	
Shear (horizontal)	$\begin{bmatrix} 1 & s_h & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$x = v$ $y = s_h v + w$	



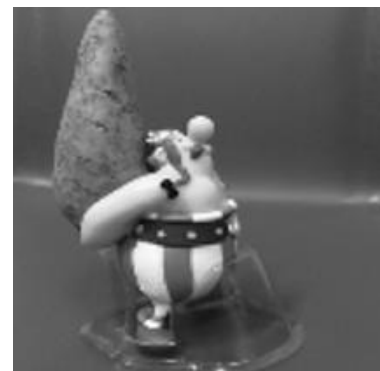
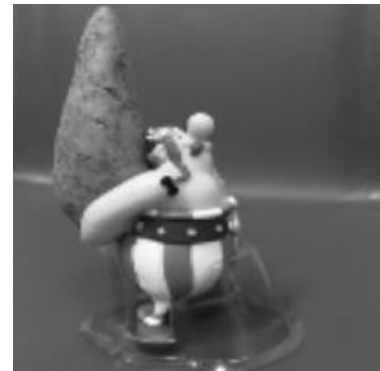
# Basic Image Operations

Image Resizing:

Down-sampling and  
Up-sampling

Interpolation methods:

1. Nearest neighbour technique
2. Bilinear technique
3. Bicubic technique



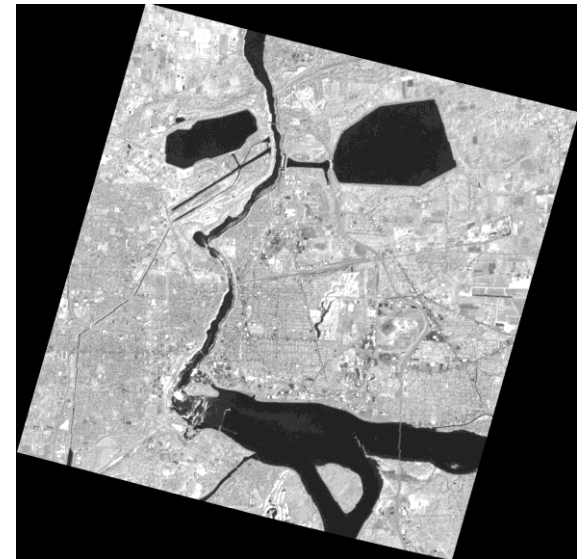
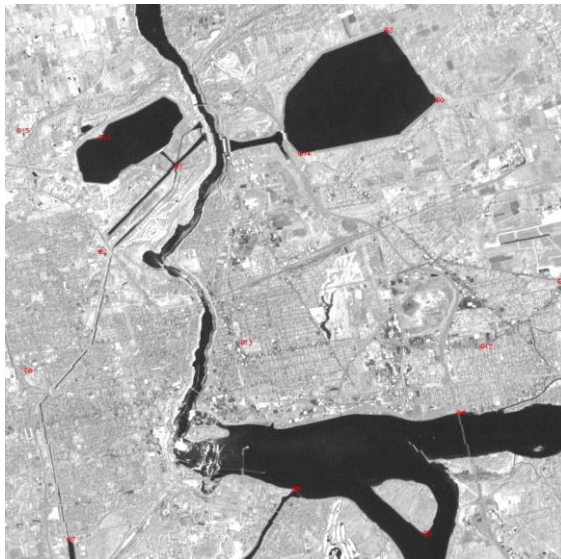
128 → 1024

64 → 1024

# Basic Image Operations

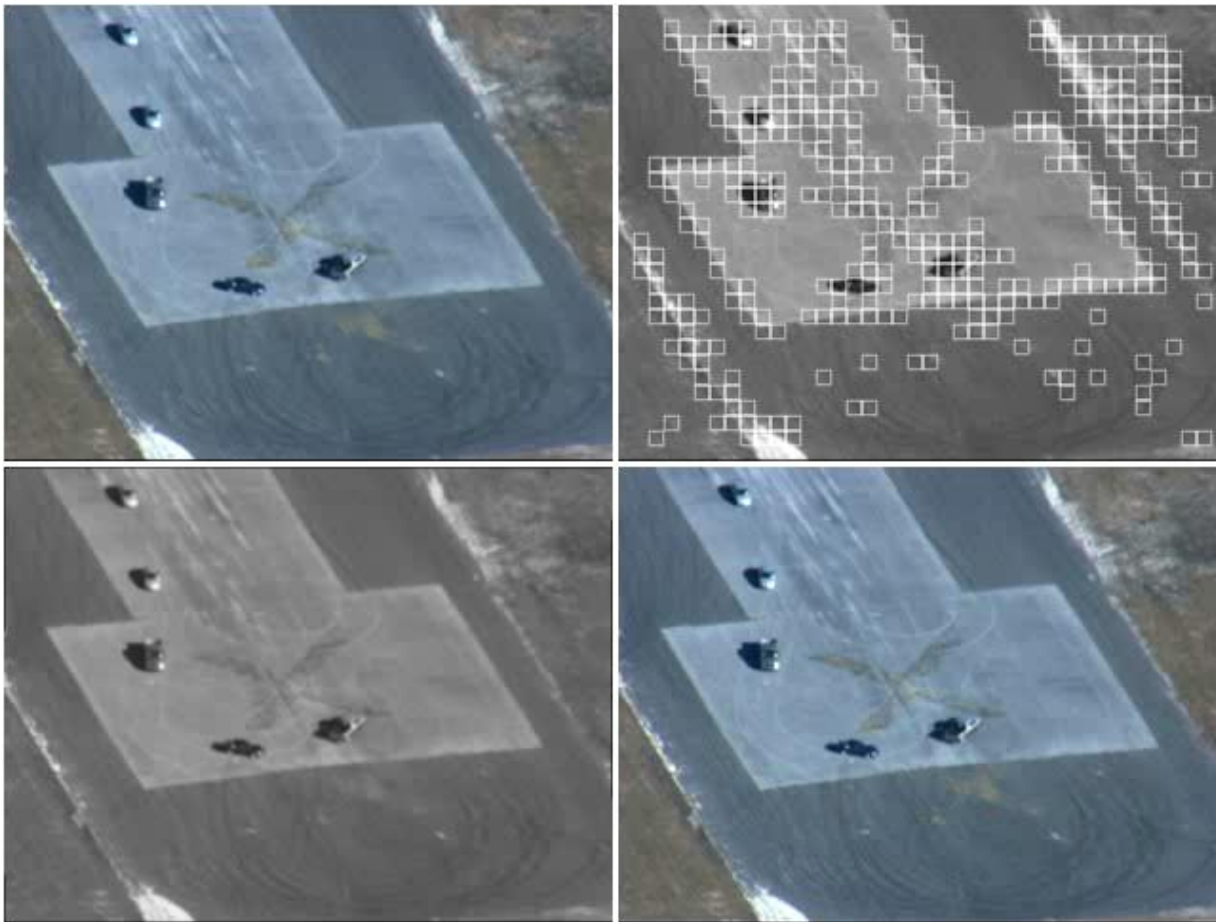
## Application – Image Registration

Image Registration is a process of estimating optimal transformation between two images which need to be aligned



# Basic Image Operations

## Application— Image Registration



We have looked at:

- Pixel neighbourhood and connectivity
- Distance measures
- Connected component labeling
- Basic Image Operations

Next time we start to look at techniques for image enhancement