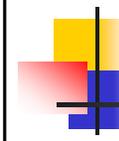


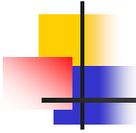
Digital Image Processing

Chapter 2: Digital Image Fundamentals



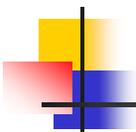
Human and Computer Vision

- We can't think of image processing without considering the human vision system.
- We observe and evaluate the images that we process with our visual system.



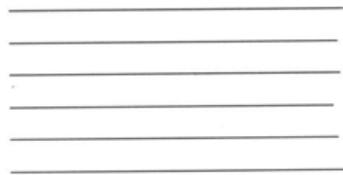
Simple questions

- What intensity differences can we distinguish?
- What is the spatial resolution of our eye?
- How accurately we estimate and compare distances and areas?
- How do we sense colors?
- By which features can we detect and distinguish objects?

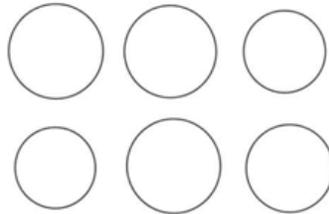


Test images

a



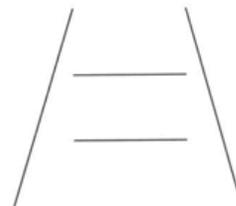
b



c



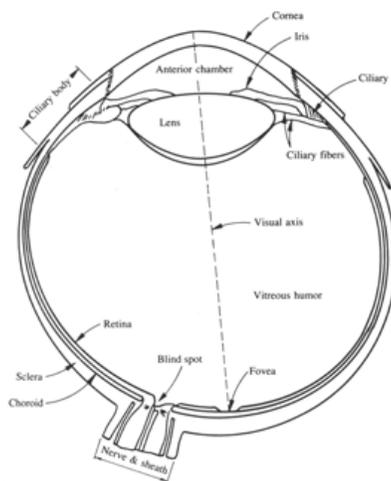
d



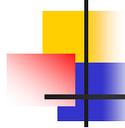
Test images

- Test images for distances and area estimation:
 - a) Parallel lines with up to 5% difference in length.
 - b) Circles with up to 10% difference in radius.
 - c) The vertical line appears longer but actually has the same length as the horizontal line.
 - d) Deception by perspective: the upper line appears longer than the lower one but actually have the same length.

Structure of the Human Eye



- Shape is nearly a sphere.
- Average diameter = 20 mm.



Lens & Retina

- Lens

both infrared and ultraviolet light are absorbed, in excessive amounts, can cause damage to the eye.

- Retina

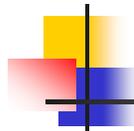
Innermost membrane of the eye. When the eye is properly focused, light from an object outside the eye is imaged on the retina.



Receptors

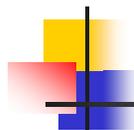
- Receptors are divided into 2 classes:

- Cones
- Rods



Cones

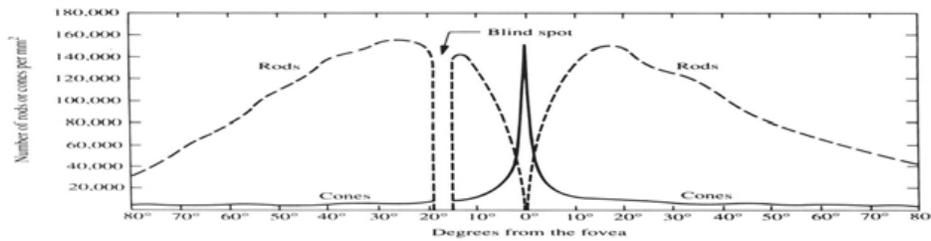
- 6-7 million, located primarily in the central portion of the retina (muscles controlling the eye rotate the eyeball until the image falls on the fovea).
- Highly sensitive to color.
- Each is connected to its own nerve end thus human can resolve fine details.
- Cone vision is called photopic or bright-light vision.



Rods

- 75-150 million, distributed over the retina surface.
- Several rods are connected to a single nerve end reduce the amount of detail discernible.
- Serve to give a general, overall picture of the field of view.
- Sensitive to low levels of illumination.
- Rod vision is called scotopic or dim-light vision.

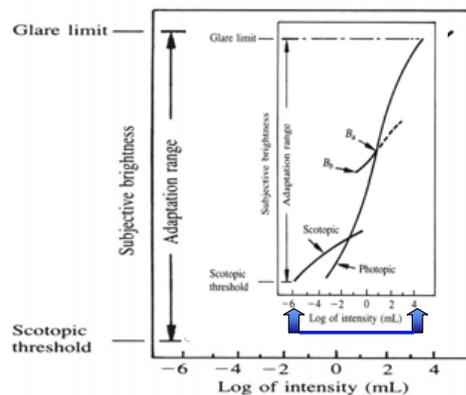
Cross section of the eye



- Blind spot ⇒ the absence of receptors area.
- Cones are most dense in the center of the retina (in the area of the fovea).

Brightness adaptation and discrimination

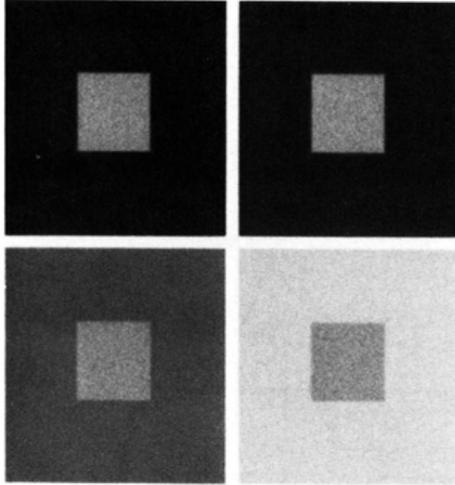
- The total range of intensity levels it can discriminate simultaneously is rather small compared with the total adaptation range.



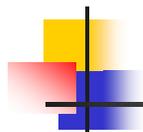


Simultaneous contrast

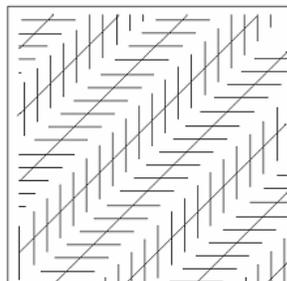
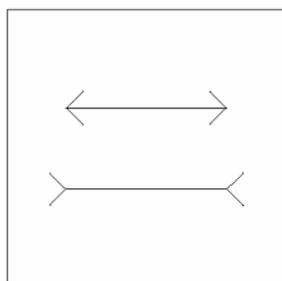
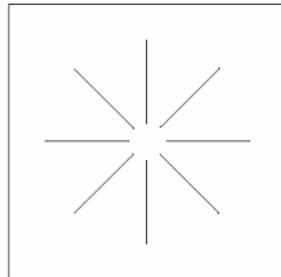
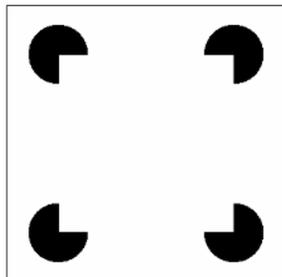
Which small square is the darkest one ?

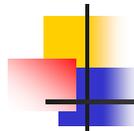


- All the small squares have exactly the same intensity, but they appear to the eye progressively darker as the background becomes brighter.



Human Perception Phenomena



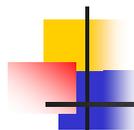


Signals

- a signal is a function that carries information.
- usually content of the signal changes over some set of spatiotemporal dimensions.

Vocabulary:

Spatiotemporal: existing in both space and time;
having both spatial extension and temporal duration



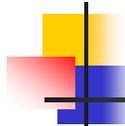
Time-Varying Signals

- Some signals vary over time:

$f(t)$

for example: audio signal

may be thought at one level as a collection various tones of differing audible frequencies that vary over time.



Spatially-Varying Signals

- Signals can vary over space as well.
- An image can be thought of as being a function of 2 spatial dimensions:

$$f(x,y)$$

- for monochromatic images, the value of the function is the amount of light at that point.
- medical CAT and MRI scanners produce images that are functions of 3 spatial dimensions:

$$f(x,y,z)$$



Spatiotemporal Signals

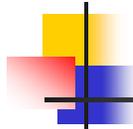
What do you think a signal of this form is?

$$f(x,y,t)$$

x and y are spatial dimensions;

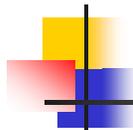
t is time.

Perhaps, it is a video signal, animation,
or other time-varying picture sequence.



Types of Signals

- most naturally-occurring signals are functions having a continuous domain.
- however, signals in a computer are discrete samples of the continuous domain.
- in other words, signals manipulated by computer have discrete domains.



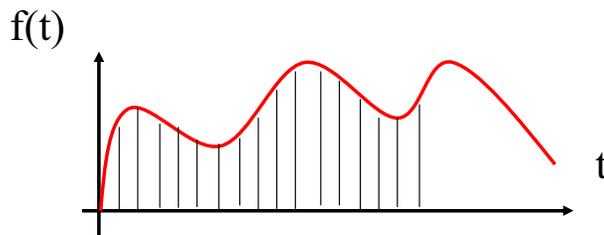
Analog & Digital

- most naturally-occurring signals also have a real-valued range in which values occur with infinite precision.
- to store and manipulate signals by computer we need to store these numbers with finite precision. thus, these signals have a discrete range.

signal has continuous domain and range = analog
signal has discrete domain and range = digital

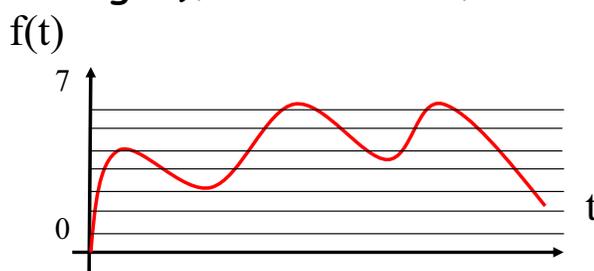
Sampling

- sampling = the spacing of discrete values in the domain of a signal.
- sampling-rate = how many samples are taken per unit of each dimension. e.g., samples per second, frames per second, etc.

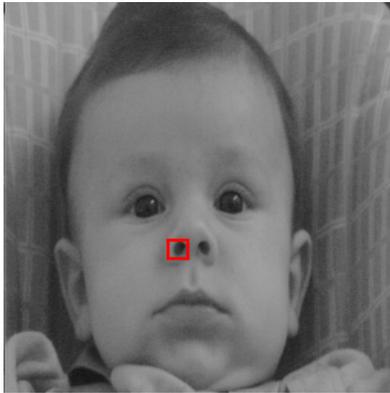


Quantization

- Quantization = spacing of discrete values in the range of a signal.
- usually thought of as the number of bits per sample of the signal. e.g., 1 bit per pixel (b/w images), 16-bit audio, 24-bit color images, etc.



Digital Image Representation



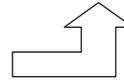
Pixel values in highlighted region

| | | | | | | | | | |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 99 | 71 | 61 | 51 | 49 | 40 | 35 | 53 | 86 | 99 |
| 93 | 74 | 53 | 56 | 48 | 46 | 48 | 72 | 85 | 102 |
| 101 | 69 | 57 | 53 | 54 | 52 | 64 | 82 | 88 | 101 |
| 107 | 82 | 64 | 63 | 59 | 60 | 81 | 90 | 93 | 100 |
| 114 | 93 | 76 | 69 | 72 | 85 | 94 | 99 | 95 | 99 |
| 117 | 108 | 94 | 92 | 97 | 101 | 100 | 108 | 105 | 99 |
| 116 | 114 | 109 | 106 | 105 | 108 | 108 | 102 | 107 | 110 |
| 115 | 113 | 109 | 114 | 111 | 111 | 113 | 108 | 111 | 115 |
| 110 | 113 | 111 | 109 | 106 | 108 | 110 | 115 | 120 | 122 |
| 103 | 107 | 106 | 108 | 109 | 114 | 120 | 124 | 124 | 132 |

CAMERA



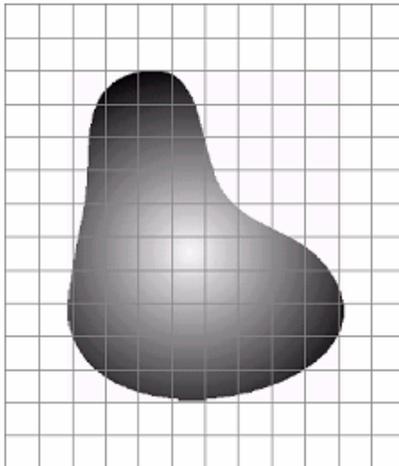
DIGITIZER



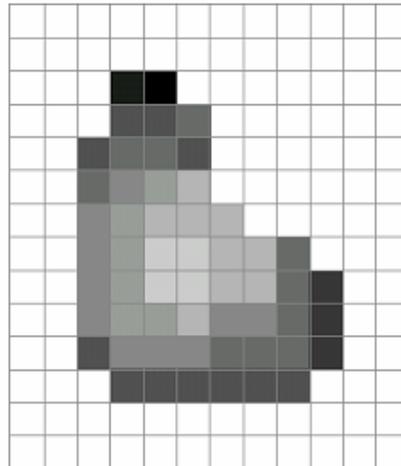
A set of number
in 2D grid

Samples the analog data and digitizes it.

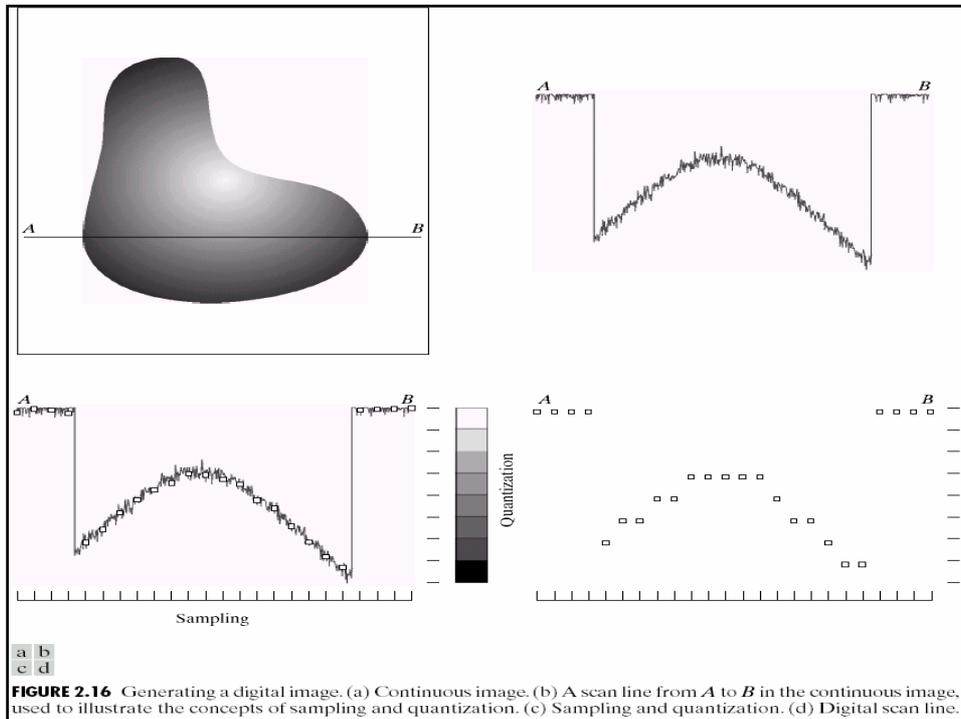
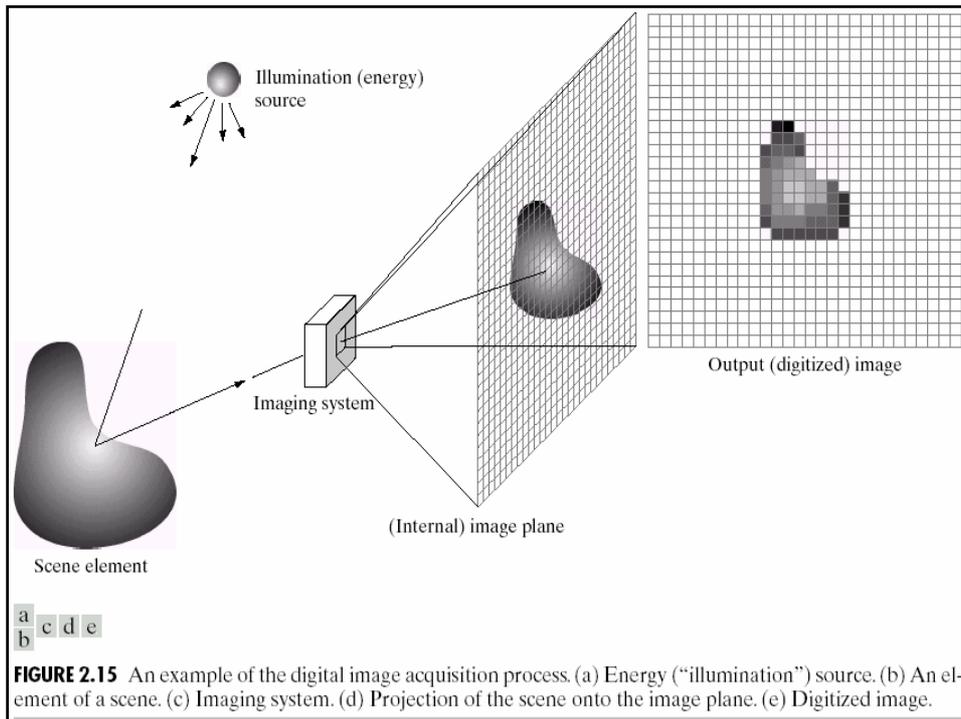
Example of Digital Image

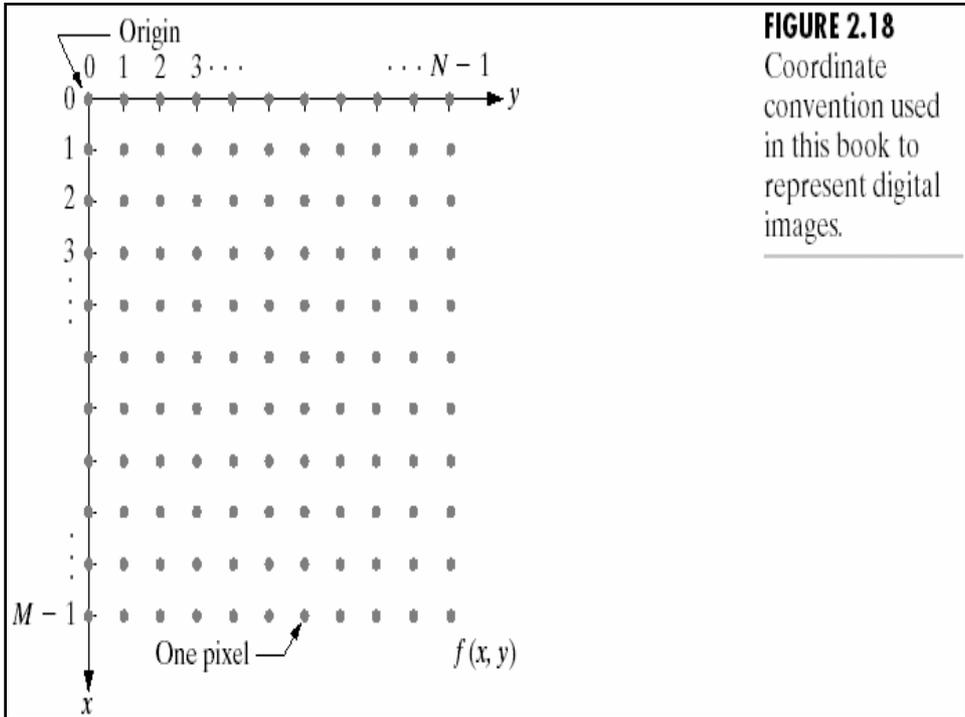
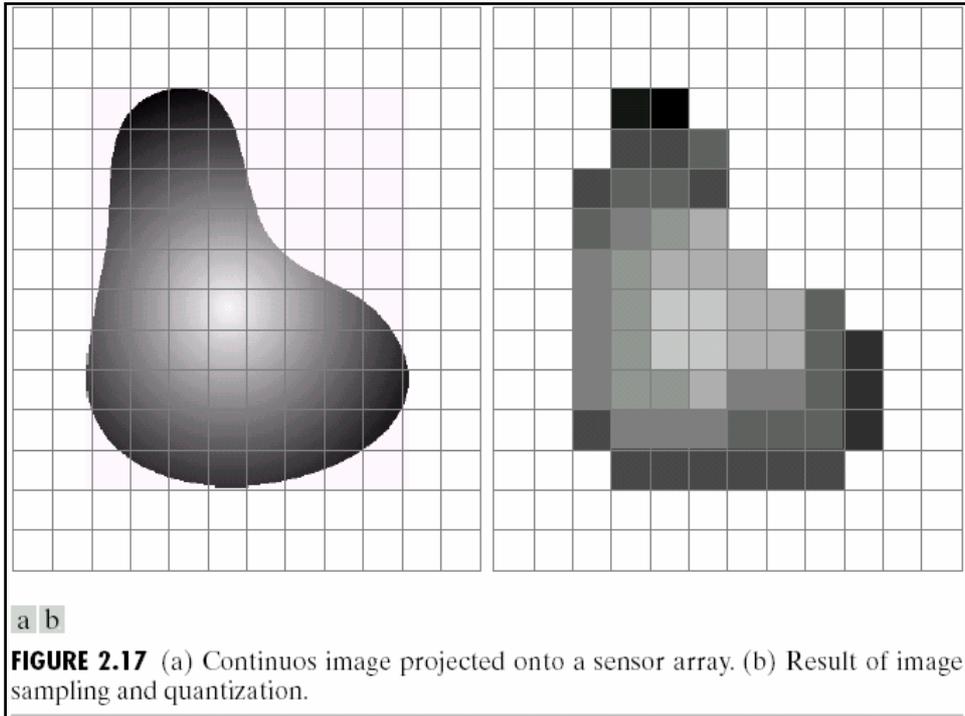


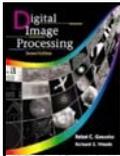
Continuous image
projected onto a
sensor array



Result of image
sampling and
quantization







Chapter 2: Digital Image Fundamentals

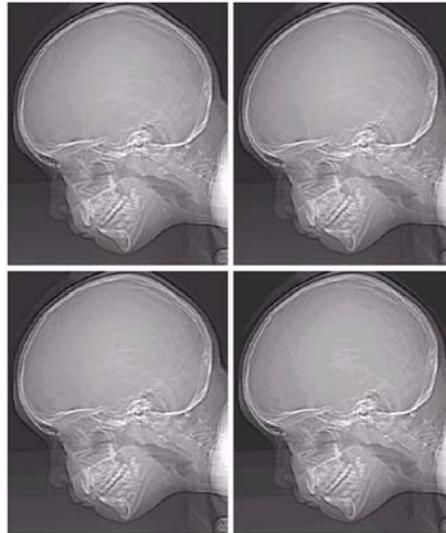
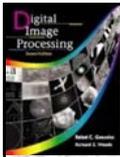


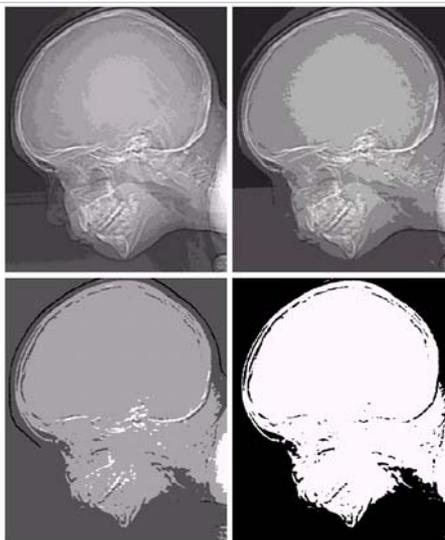
FIGURE 2.21
(a) 452×374 , 256-level image. (b)–(d) Image displayed in 128, 64, and 32 gray levels, while keeping the spatial resolution constant.

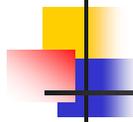


Chapter 2: Digital Image Fundamentals

c f
g h

FIGURE 2.21
(Continued)
(e)–(h) Image displayed in 16, 8, 4, and 2 gray levels. (Original courtesy of Dr. David R. Pickens, Department of Radiology & Radiological Sciences, Vanderbilt University Medical Center.)

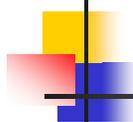




Light-intensity function

- image refers to a 2D light-intensity function, $f(x,y)$
- the amplitude of f at spatial coordinates (x,y) gives the intensity (brightness) of the image at that point.
- light is a form of energy thus $f(x,y)$ must be nonzero and finite.

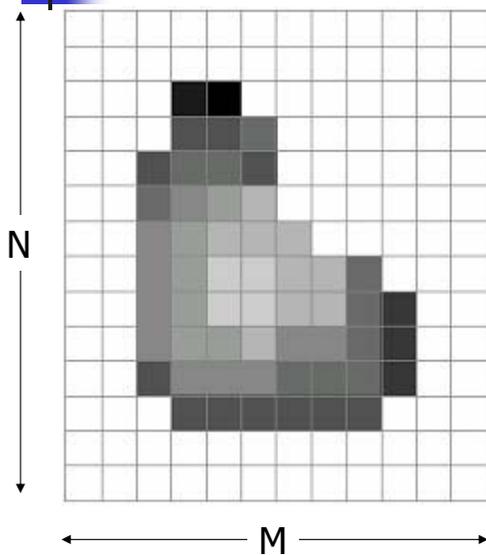
$$0 < f(x, y) < \infty$$



Gray Level

- We call the intensity of a monochrome image f at coordinate (x,y) the gray level (I) of the image at that point.
 - I lies in the range $L_{min} \leq I \leq L_{max}$
 - L_{min} is positive and L_{max} is finite
 - gray scale = $[L_{min}, L_{max}]$
 - 0 = Black, L = White

Number of bits



- The number of gray levels typically is an integer power of 2

$$L = 2^k$$

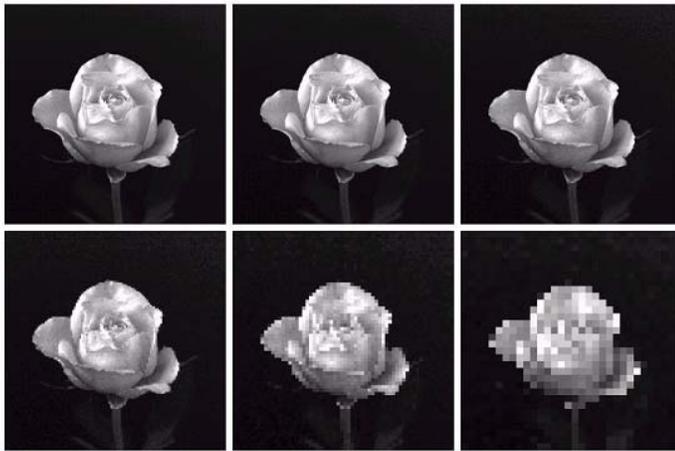
- Number of bits required to store a digitized image

$$b = M \times N \times k$$

Resolution

- Resolution (how much you can see the detail of the image) depends on sampling and gray levels.
- the bigger the sampling rate (n) and the gray scale (g), the better the approximation of the digitized image from the original.
- the more the quantization scale becomes, the bigger the size of the digitized image.

Checkerboard effect



| | | |
|---|---|---|
| a | b | c |
| d | e | f |

(a) 1024x1024

(b) 512x512

(c) 256x256

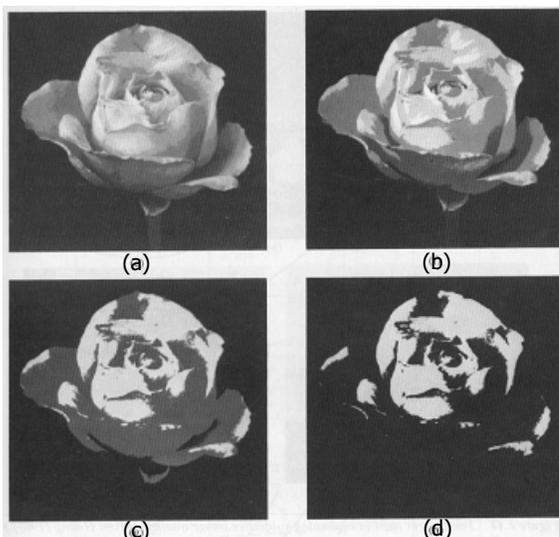
(d) 128x128

(e) 64x64

(f) 32x32

- if the resolution is decreased too much, the checkerboard effect can occur.

False contouring



(a) Gray level = 16

(b) Gray level = 8

(c) Gray level = 4

(d) Gray level = 2

- if the gray scale is not enough, the smooth area will be affected.
- False contouring can occur on the smooth area which has fine gray scales.

Nonuniform sampling

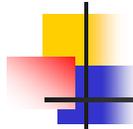
- for a fixed value of spatial resolution, the appearance of the image can be improved by using adaptive sampling rates.
 - fine sampling \Rightarrow required in the neighborhood of sharp gray-level transitions.
 - coarse sampling \Rightarrow utilized in relatively smooth regions.

Example



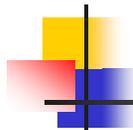
a b c

FIGURE 2.22 (a) Image with a low level of detail. (b) Image with a medium level of detail. (c) Image with a relatively large amount of detail. (Image (b) courtesy of the Massachusetts Institute of Technology.)



Example

- an image with a face superimposed on a uniform background.
 - background ⇒ little detailed information
⇒ coarse sampling is enough.
 - face ⇒ more detail ⇒ fine sampling.
- if we can use adaptive sampling, the quality of the image is improved.
- Moreover, we should care more around the boundary of the object ⇒ sharp gray-level transmission from object to background.



Nonuniform quantization

- unequally spaced levels in quantization process influences on the decreasing the number of gray level.
 - use few gray levels in the neighborhood of boundaries. Why? eye is relatively poor at estimate shades of gray near abrupt level changes.
 - use more gray levels on smooth area in order to avoid the “false contouring”.



Definitions

- Monochromatic (achromatic) light: Light that is void of color
 - Attribute: Intensity (amount) .
 - Gray level is used to describe monochromatic intensity.

- Chromatic light: To describe it, three quantities are used:
 - Radiance: The total amount of energy that flows from the light source (measured in Watts).
 - Luminance: The amount of energy an observer perceives from a light source (measured in lumens).
 - Brightness: A subjective descriptor of light perception that is impossible to measure (key factor in describing color sensation)

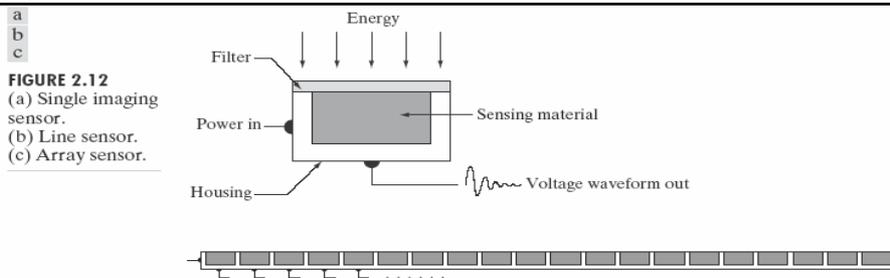
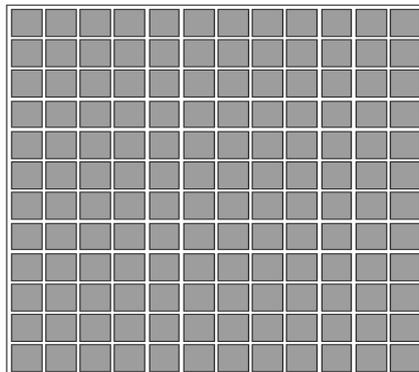


FIGURE 2.12
 (a) Single imaging sensor.
 (b) Line sensor.
 (c) Array sensor.

- Incoming energy (reflected or transmitted through) is transformed into a voltage.
- Sensor material is responsive to a particular type of energy being detected.



Single Sensor

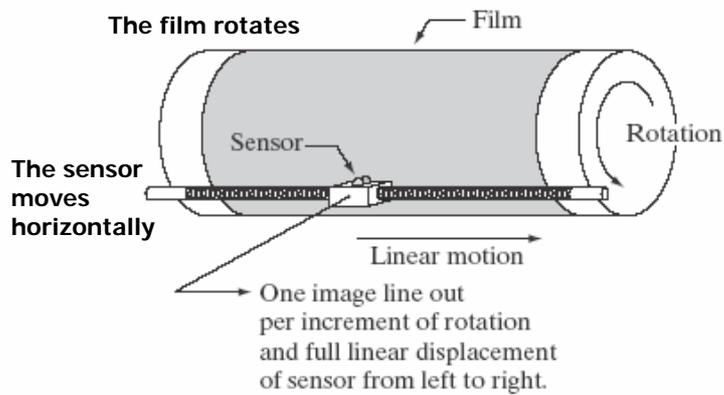


FIGURE 2.13 Combining a single sensor with motion to generate a 2-D image.

Spatial And Gray Level Resolution

- An L -level image of size $M \times N$
- Spatial resolution:
 - # of samples per unit length or area.
 - Dots Per Inch (DPI): specifies the size of an individual pixel.
- Gray level resolution:
 - Number of bits per pixel.
 - Usually 8 bits.
 - Color image has 3 image planes to yield $8 \times 3 = 24$ bits/pixel.
 - Too few levels may cause false contour.

Basic Relationship b/w pixels

- Neighbors of a pixel
- Connectivity
- Labeling of Connected Components
- Relations, Equivalences, and Transitive Closure
- Distance Measures
- Arithmetic/Logic Operations

Neighbors of a pixel

- a pixel **p** at coordinate (x,y) has

- $N_4(p)$: 4-neighbors of p
 $(x+1, y), (x-1,y), (x,y+1), (x,y-1)$

| | | | |
|---|---|---|--|
| | | x | |
| x | p | x | |
| | | x | |
- $N_D(p)$: 4-diagonal neighbors of p
 $(x+1, y+1), (x+1,y-1), (x-1,y+1), (x-1,y-1)$

| | | | | |
|--|--|---|---|---|
| | | x | | x |
| | | | p | |
| | | x | | x |
- $N_8(p)$: 8-neighbors of p :
a combination of $N_4(p)$ and $N_D(p)$

| | | | |
|---|---|---|---|
| | x | x | x |
| x | p | x | |
| | x | x | x |

Connectivity

- Let V be the set of gray-level values used to defined connectivity
 - 4-connectivity :
 - 2 pixels p and q with values from V are 4-connected if q is in the set $N_4(p)$
 - 8-connectivity :
 - 2 pixels p and q with values from V are 8-connected if q is in the set $N_8(p)$
 - m-connectivity (mixed connectivity):
 - 2 pixels p and q with values from V are m-connected if
 - q is in the set $N_4(p)$ **or**
 - q is in the set $N_D(p)$ and the set $N_4(p) \cap N_4(q)$ is empty.

Example

```

0 1 1
0 1 0
0 0 1
  
```

arrangement
of pixels

```

0 1-1
0 1 0
0 0 1
  
```

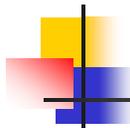
8-neighbors of the
center pixel

```

0 1-1
0 1 0
0 0 1
  
```

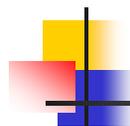
m-neighbors of
the center pixel

- m-connectivity eliminates the multiple path connections that arise in 8-connectivity.



Adjacent

- a pixel p is adjacent to a pixel q if they are connected.
- two image area subsets S_1 and S_2 are adjacent if some pixel in S_1 is adjacent to some pixel in S_2 .



Exercise

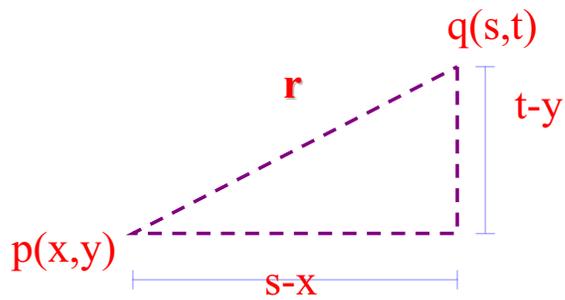
- Consider the two image subsets S_1 and S_2 :

| | S_1 | | | | | S_2 | | | | |
|---|-------|---|---|---|---|-------|---|---|---|---|
| 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 1 | 0 |
| 1 | 0 | 0 | 1 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |
| 1 | 0 | 0 | 1 | 0 | 1 | 1 | 0 | 0 | 0 | 0 |
| 0 | 0 | 1 | 1 | 1 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0 | 0 | 1 | 1 | 1 | 0 | 0 | 1 | 1 | 1 | 1 |

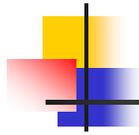
- For $V=\{1\}$, determine whether S_1 and S_2 are
 - 4-connected
 - 8-connected
 - m-connected

Euclidean distance between p and q

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



City-block distance:



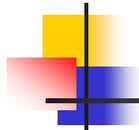
D_4 distance

$$D_4(p, q) = |x - s| + |y - t|$$

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

- diamond centered at (x, y)
- $D_4 = 1$ are 4-neighbors of (x, y)

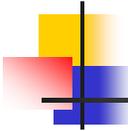
Chessboard distance: D_8 distance



$$D_8(p, q) = \max(|x - s|, |y - t|)$$

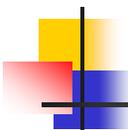
| | | | | |
|---|---|---|---|---|
| 2 | 2 | 2 | 2 | 2 |
| 2 | 1 | 1 | 1 | 2 |
| 2 | 1 | 0 | 1 | 2 |
| 2 | 1 | 1 | 1 | 2 |
| 2 | 2 | 2 | 2 | 2 |

square centered at (x, y)



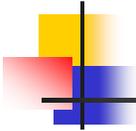
Arithmetic Operators

- used extensively in most branches of image processing.
- Arithmetic operations for 2 pixels p and q :
 - Addition : $p+q$ used in image average to reduce noise.
 - Subtraction : $p-q$ basic tool in medical imaging.
 - Multiplication : $p \times q$
 - to correct gray-level shading result from non-uniformities in illumination or in the sensor used to acquire the image.
 - Division : $p \div q$
- Arithmetic Operation entire images are carried out pixel by pixel.



Logic operations

- AND : $p \text{ AND } q$ ($p \cdot q$)
- OR : $p \text{ OR } q$ ($p + q$)
- COMPLEMENT : NOT q (\bar{q})
- logic operations apply only to binary images.
- arithmetic operations apply to multivalued pixels.
- logic operations used for tasks such as masking, feature detection, and shape analysis.
- logic operations perform pixel by pixel.



Mask Operation

- Besides pixel-by-pixel processing on entire images, arithmetic and Logical operations are used in neighborhood oriented operations.

| | | | | |
|-----|-------|-------|-------|-----|
| | | ⋮ | | |
| | Z_1 | Z_2 | Z_3 | |
| ... | Z_4 | Z_5 | Z_6 | ... |
| | Z_7 | Z_8 | Z_9 | |
| | | ⋮ | | |



Mask Operation

- Let the value assigned to a pixel be a function of its gray level and the gray level of its neighbors.
- e.g., replace the gray value of pixel Z_5 with the average gray values of it's neighborhood within a 3x3 mask.

$$Z = \frac{1}{9}(Z_1 + Z_2 + Z_3 + \dots + Z_9)$$

Mask operator

- In general term:

$$\begin{aligned} Z &= \frac{1}{9} Z_1 + \frac{1}{9} Z_2 + \frac{1}{9} Z_3 + \dots + \frac{1}{9} Z_9 \\ &= w_1 Z_1 + w_2 Z_2 + w_3 Z_3 + \dots + w_9 Z_9 \\ &= \sum_{i=1}^9 w_i Z_i \end{aligned}$$

| | | |
|-------|-------|-------|
| w_1 | w_2 | w_3 |
| w_4 | w_5 | w_6 |
| w_7 | w_8 | w_9 |



| | | |
|-----|-----|-----|
| 1/9 | 1/9 | 1/9 |
| 1/9 | 1/9 | 1/9 |
| 1/9 | 1/9 | 1/9 |

Mask coefficient

- Proper selection of the coefficients and application of the mask at each pixel position in an image makes possible a variety of useful image operations
 - noise reduction
 - region thinning
 - edge detection
- Applying a mask at each pixel location in an image is a computationally expensive task.