

# Advanced Digital Image Processing

## Chapter 2: Digital Image Fundamental

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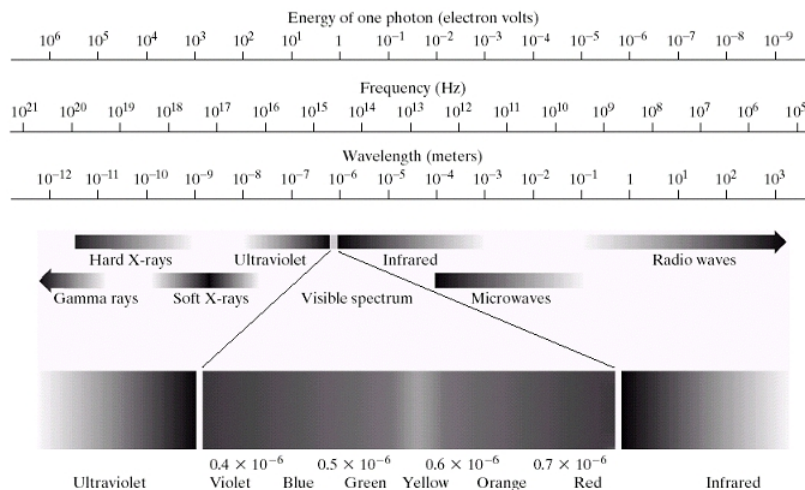


## Outline

- 1 Light: Spectrum and Attribute
- 2 Eyes: Elements of Visual Perception
  - Structure of Eyes
  - Characteristics of HVS
- 3 Image Sensing and Acquisition
- 4 Image Sampling and Quantization
  - Monochromatic Image Model
  - Resolution
  - Zooming and Shrinking
- 5 Basic Relationships between Pixels
  - Neighborhood
  - Adjacency
  - Path
  - Region and Boundary
  - Distance



## Electro-magnetic Spectrum



## Light Attribute

### Mono-chromatic (Achromatic) light

Described by **Intensity** or **Gray level**.

### Chromatic light

Described by three quantities,

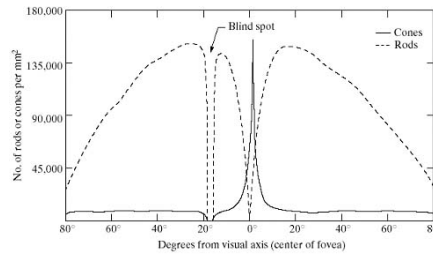
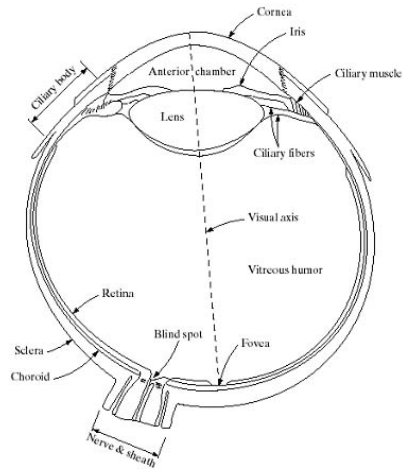
- : total amount of energy that flows from the light source
- : Measured in lumens(lm), a measure of the amount of energy an observer *perceives* from a light source.
- : a *subjective* descriptor of light perception that is practically impossible to measure.

### Example

Far infrared light: Large radiance but almost zero luminance

## Structure of Eyes

## Structures of the Human Eye



## Structure of Eyes

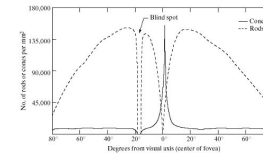
## Two Classes of Receptors

\_\_\_\_\_ : 6 to 7 million.

- Located primarily in the center of the retina (**fovea**)
- resolve **fine detail**
- highly sensitive to **color**.
- **cone vision, photopic, bright-light vision**

\_\_\_\_\_ : 75 to 150 million (larger)

- Radially symmetric about the fovea (except **blind spot**)
- give a general, overall picture of **field of view**.
- **not** involved in color vision but sensitive to **low illumination**.
- **rod vision, scotopic, dim-light vision**



## Structure of Eyes

## Example of Receptors

## Example

## Sensitivity:

Why objects are in bright color in daylight, but colorless in moonlight?

**Only rods are stimulated.**

## Example

## Density:

- 337,000 cones on fovea (**1.5mm x 1.5mm**)  
ie. cone density: 150,000 elements/mm<sup>2</sup>
- CCD have this number of elements in a receptor array no larger than **5mm x 5mm**.

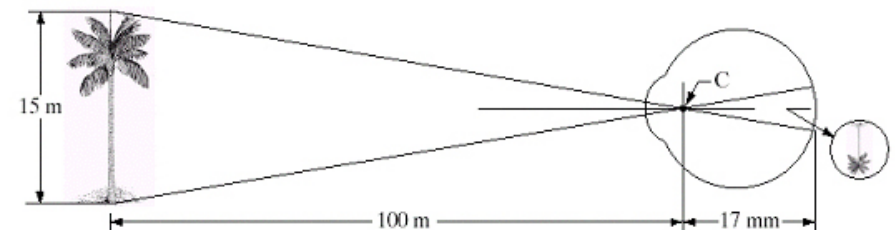


## Structure of Eyes

## Image Formation in the Eye

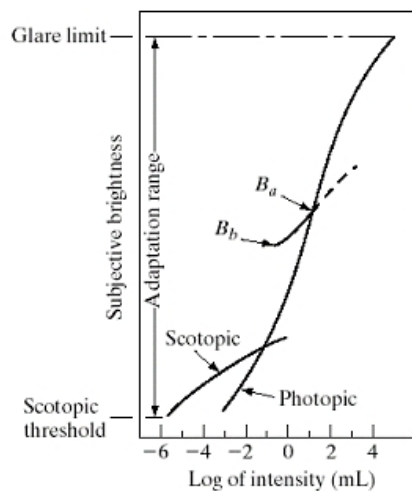
## Definition

\_\_\_\_\_ is the distance between the center of the lens and the retina.



## Characteristics of HVS

### Brightness Adaptation



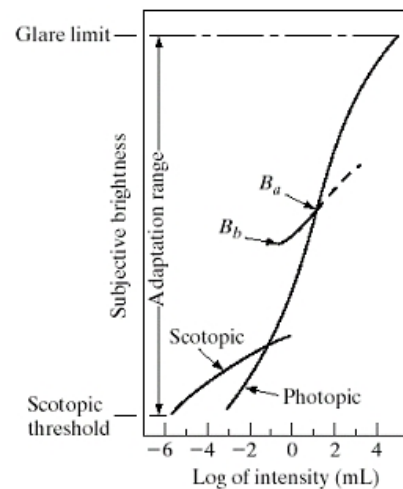
- Subjective brightness of HVS is a **log** function of light intensity
- HVS can adapt enormous range around  $10^{10}$  from **scotopic threshold** to **glare limit** ( $10^6$  for photopic)

#### Definition

\_\_\_\_\_ : HVS cannot operate over the range simultaneously. Rather, it accomplishes this large variation by changes in its overall sensitivity.

## Characteristics of HVS

### Brightness Adaptation



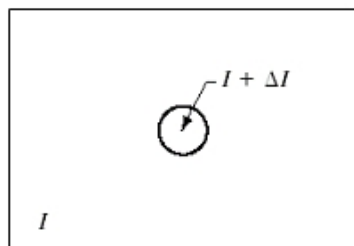
#### Example

Let \_\_\_\_\_ be at  $B_a$

- The range below  $B_b$  is indistinguishable black.
- The (dashed) range above  $B_a$  is not restricted, but too high may change to new adaptation level.
- The adaptation range is small compared with the total range

## Characteristics of HVS

### Bright Discrimination



- Changing light at **specific adaptation level**
- $\Delta I$  in the form of a short-duration flash.

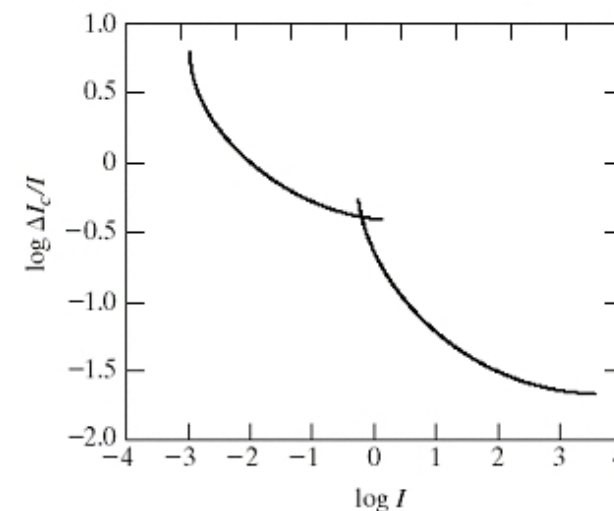
#### Definition

\_\_\_\_\_ is the quantify  $\Delta I/I$ , where  $\Delta I$  is the increment of illumination discriminable **50%** of the time with background illumination  $I$ .

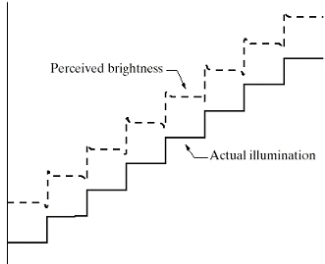
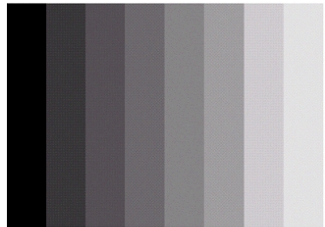
- Small **Weber ratio**  $\Rightarrow$  Good

## Characteristics of HVS

### Chart of Weber Ratio



## Mach Bands



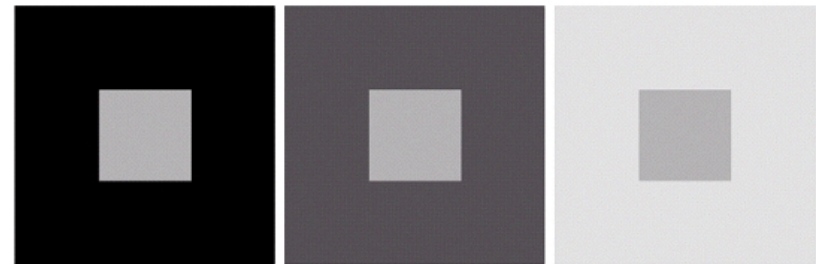
- The perceived brightness is not a simple function of intensity
- HVS **undershoots** or **overshoots** around the boundary of regions of different intensities.

### Definition

The seemingly scalloped bands are called \_\_\_\_\_ (Ernst Mach, 1865)



## Simultaneous Contrast

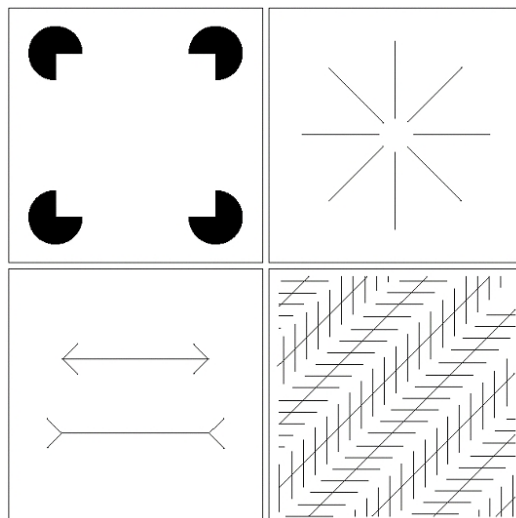


- The perceived brightness is not a simple function of intensity

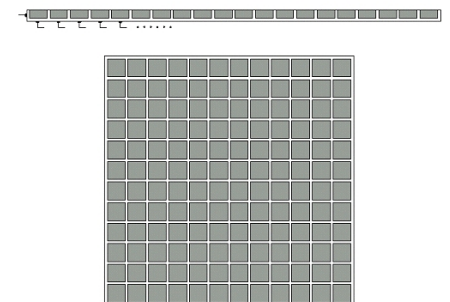
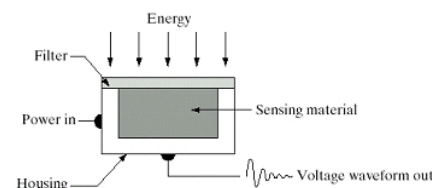
### Definition

\_\_\_\_\_ : All the inner squares have the same intensity, but they appear progressively darker as the background becomes lighter.

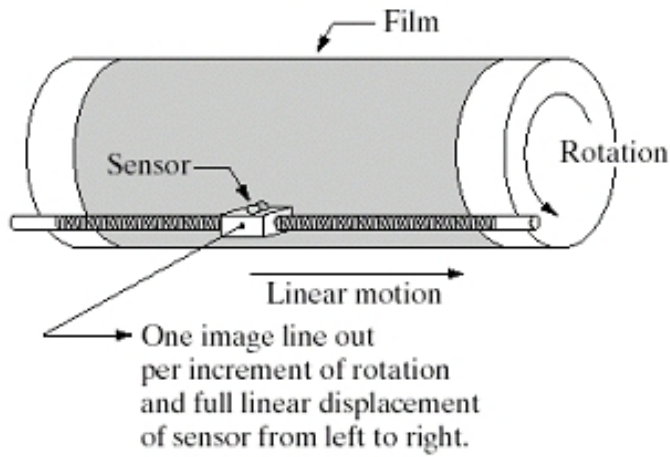
## Optical Illusions



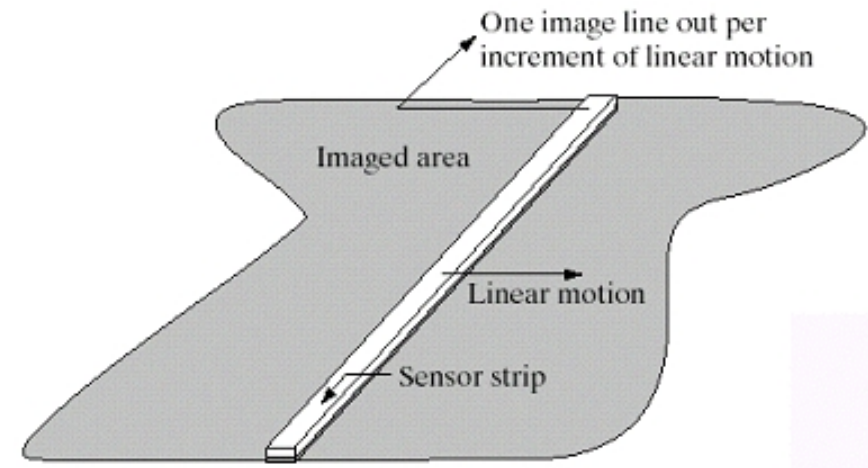
## Line sensor and Array Sensor



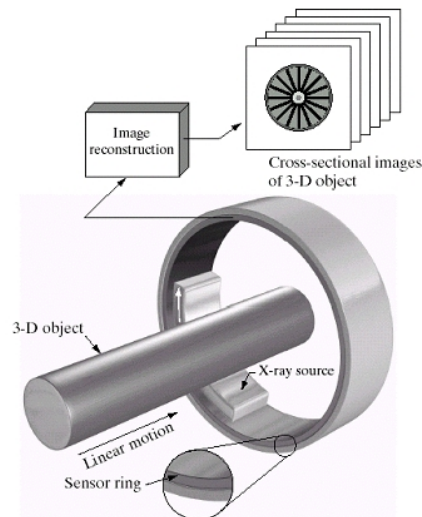
## Single sensor



## Line sensor strip

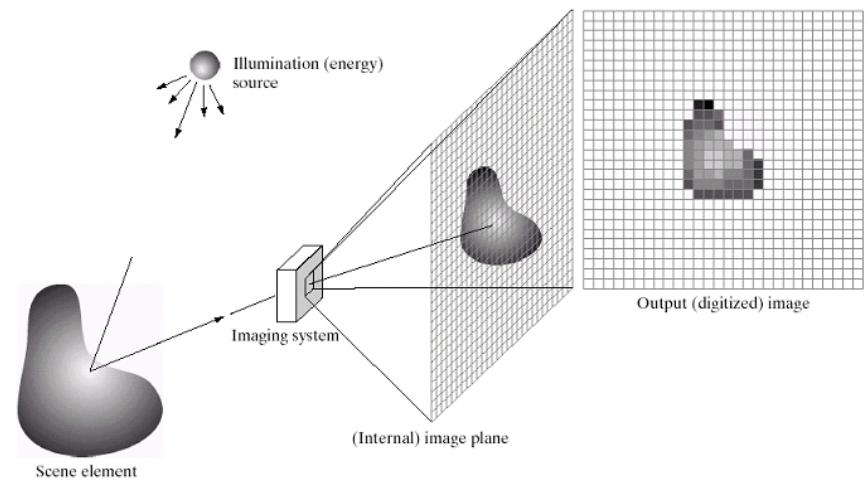


## Circular sensor strip



## Monochromatic Image Model

## Image Acquisition by Array sensor





## Monochromatic Image

- The values of image are proportional to **energy radiated** by a physical source (e.g., EM waves).
- Therefore,  $0 < f(x, y) < \infty$
- $f(x, y) = i(x, y)r(x, y)$ 
  - $0 < i(x, y) < \infty$   
: Amount of source illumination incident on the scene being viewed.  
(sun:  $90,000 \text{ lm/m}^2$ , cloud:  $10,000 \text{ lm/m}^2$ , full moon:  $0.1 \text{ lm/m}^2$ , office:  $1000 \text{ lm/m}^2$ )
  - $0 < r(x, y) < 1$   
: Amount of illumination reflected by the objects in the scene.  
(black velvet: 0.01, stainless steel: 0.65, snow: 0.93)



## Intensity (Gray Level)

Given the **intensity**  $I$  of an monochromatic image at  $(x_0, y_0)$ ,

$$I = f(x_0, y_0), \\ L_{min} \leq I \leq L_{max}$$

### Definition

The interval  $[L_{min}, L_{max}]$  is called the \_\_\_\_\_

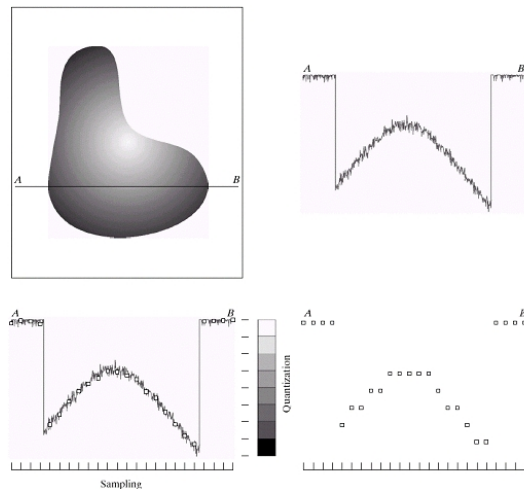
### Example

Common practice:

- $[0, L - 1]$ : (black  $I = 0$ , white  $I = L - 1$ )
- Ex: For  $[0, 255]$ , gray scale  $L = 256$



## Sampling and Quantization

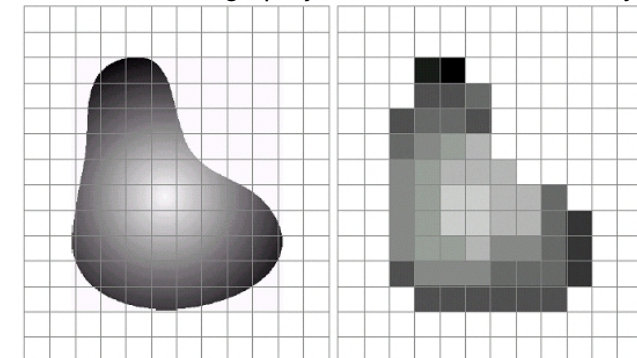


- \_\_\_\_\_ : Digitizing the coordinate values of  $f(x, y)$
- \_\_\_\_\_ : Digitizing the amplitude values of  $f(x, y)$

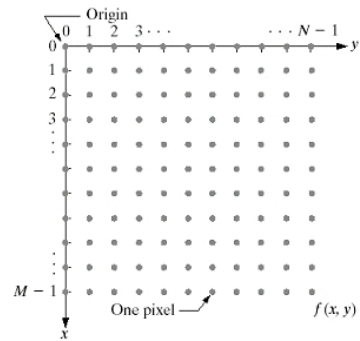


## Sampling and Quantization

Continuous image projected onto a sensor array.



## Digital Image Notation



- Assume an image  $f(x, y)$  is sampled so that the resulting digital image has  $M$  rows and  $N$  columns ( $M \times N$  digital image).
- Origin of coordinate  $(x, y) = (0, 0)$ . ( $x$ :row,  $y$ :column)
- Note:  $(x, y)$  is not the actual values of physical coordinates when the image was sampled.



## Storage bits for $N \times N$ Digital Image

Number of storage bits for various values of  $N$  and  $k$ .

$N/k$	1 ( $L = 2$ )	2 ( $L = 4$ )	3 ( $L = 8$ )	4 ( $L = 16$ )	5 ( $L = 32$ )	6 ( $L = 64$ )	7 ( $L = 128$ )	8 ( $L = 256$ )
32	1,024	2,048	3,072	4,096	5,120	6,144	7,168	8,192
64	4,096	8,192	12,288	16,384	20,480	24,576	28,672	32,768
128	16,384	32,768	49,152	65,536	81,920	98,304	114,688	131,072
256	65,536	131,072	196,608	262,144	327,680	393,216	458,752	524,288
512	262,144	524,288	786,432	1,048,576	1,310,720	1,572,864	1,835,008	2,097,152
1024	1,048,576	2,097,152	3,145,728	4,194,304	5,242,880	6,291,456	7,340,032	8,388,608
2048	4,194,304	8,388,608	12,582,912	16,777,216	20,971,520	25,165,824	29,360,128	33,554,432
4096	16,777,216	33,554,432	50,331,648	67,108,864	83,886,080	100,663,296	117,440,512	134,217,728
8192	67,108,864	134,217,728	201,326,592	268,435,456	335,544,320	402,653,184	469,762,048	536,870,912



## $M \times N$ Digital Image

- $M \times N$  Digital Image =  $M \times N$  Matrix

$$f(x, y) = \begin{pmatrix} f(0, 0) & f(0, 1) & \cdots & f(0, N-1) \\ f(1, 0) & f(1, 1) & \cdots & f(1, N-1) \\ \vdots & \vdots & \ddots & \vdots \\ f(M-1, 0) & f(M-1, 1) & \cdots & f(M-1, N-1) \end{pmatrix}$$

- \_\_\_\_\_ : image can have  $2^k$  gray level. ( $L = 2^k$ )

### Example

- \_\_\_\_\_ : 256 possible gray-level.
- \_\_\_\_\_ : Black and white
- The number of bits required to store a digital image:  $M \times N \times k$  bits.



## Spatial Resolution

### Definition

\_\_\_\_\_ (width  $2W$ ): consists of one line (width  $W$ ) and its adjacent space (width  $W$ ).

### Definition

\_\_\_\_\_ is the smallest number of discernible **line pairs** per unit distance;

### Example

100 line pairs per millimeter, dpi., 1024x768 pixels.



## Spatial Resolution

### Subsampling

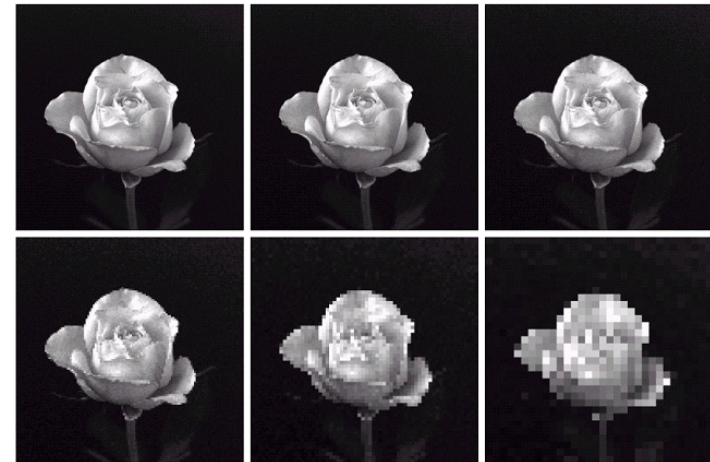
A  $1024 \times 1024$ , 8-bits image subsampled down to size  $32 \times 32$  pixels.



## Spatial Resolution

### Sampling Checkerboard

Resampled image: **Sampling checkerboard**



## Gray-Level Resolution

### Definition

\_\_\_\_\_ is the smallest discernible change in gray level.  
(subjective).

### Example

8 bits/10 bits/12 bits/16 bits images.

### Example

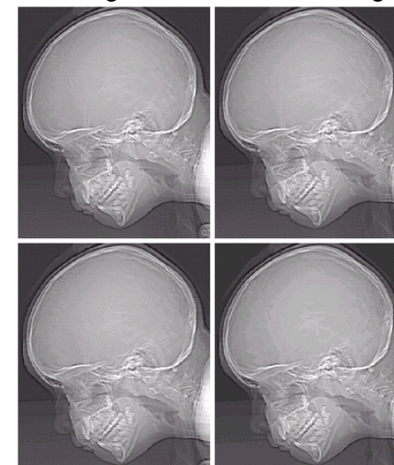
$L$ -level digital image of size  $M \times N$ :

- Spatial resolution:  $M \times N$  pixels.
- Gray-level resolution:  $L$  levels.
- Reasonable smallest image free of objectionable sampling checkerboards/false contouring:  $M, N = 256$  with  $L = 64$ .

## Gray-Level Resolution

### False Contouring

$452 \times 374$  image with 256/128/64/32 gray level

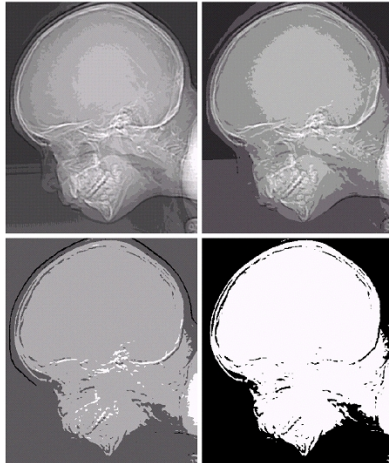




## False Contouring

### False Contouring

452 × 374 image with 16/8/4/2 gray level: **False contouring**



## Isopreference curve

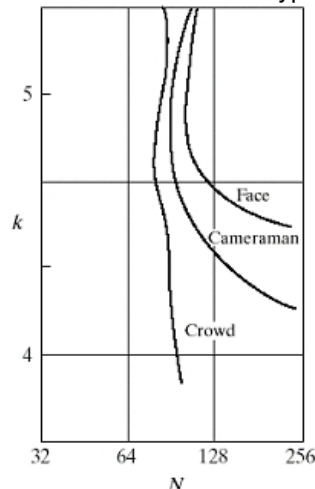
### Image Details

Image with low/medium/high level of detail



## Isopreference curve

Isopreference curves for the three types of images



## Aliasing

### Definition

\_\_\_\_\_ is the number of samples taken per unit distance. (in both spatial directions)

### Definition

\_\_\_\_\_ : twice the highest frequency of **band-limited** signal to avoid **aliasing**.

In practical, image is finite size, i.e., not band-limited signal, therefore,

### Tip

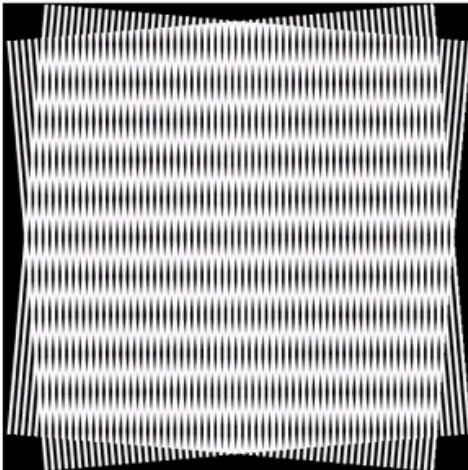
To sample an image, the principal approach for \_\_\_\_\_ is to reduce its high-frequency components by **blurring the image prior to sampling**.

## Resolution

### Aliasing

#### Moire Patterns

Moire pattern effect



- **Periodic function** is one special case in which a function of infinite duration **can** be sampled over a finite interval **without violating** the sampling theorem.
- **Aliasing**: caused by a **break-up of the periodicity**, and produce a 2-D sinusoidal waveform(**aliasing frequency**) running in vertical direction.



## Zooming and Shrinking

### Zooming Digital Image

- Two steps:
  - The creation of new pixel **locations**.
  - The assignment of **gray levels** to those new locations.
- Two popular approaches:
  - \_\_\_\_\_
  - \_\_\_\_\_



## Zooming and Shrinking

### Zooming

#### Nearest neighbor interpolation

Suppose that we want to enlarge a  $500 \times 500$  image 1.5 times to  $750 \times 750$ .

- **Shrinking and Laying** an imaginary  $750 \times 750$  grid over the original image.
- Looking for the **closest pixel** in the original image and assign its gray level to the new pixel in the grid
- **Expanding** the grid to the original specified size.

#### Example

\_\_\_\_\_ is a special case of nearest neighbor interpolation **when the zooming factor is an integer**.

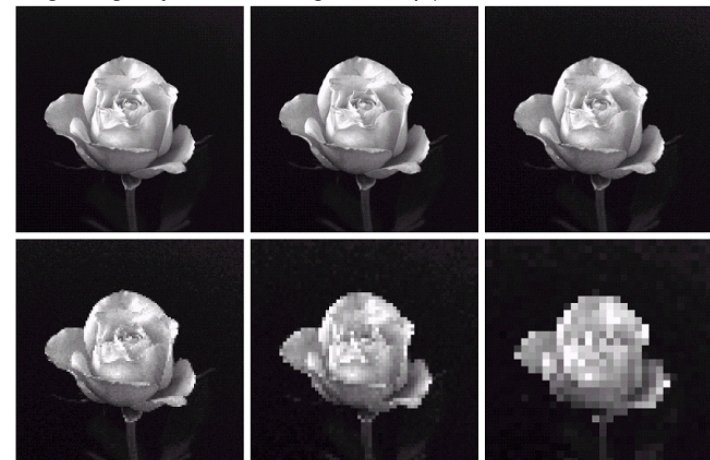


## Zooming and Shrinking

### Zooming

#### Nearest neighbor interpolation

zooming image by nearest neighbor intp(**fast but checkerboard effect**)



## Zooming

### Bilinear interpolation

Using **four nearest neighbors** of a point  $(x', y')$  to interpolate  $v(x', y')$ .

$$v(x', y') = ax' + by' + cx'y' + d$$

where the  $a, b, c, d$  are determined from **the four equations in four unknowns** that can be written using the four nearest neighbors of point  $(x', y')$



## Zooming

### Bilinear interpolation

Zooming image by bilinear interpolation



## Shrinking Digital Image

Similar to Zooming

- **Expand** the grid to fit the original image
- Do gray-level nearest neighbor or bilinear **interpolation**.
- **Shrink** the grid back to its original size.

### Example

\_\_\_\_\_ is a special case of nearest neighbor interpolation when the shrinking factor is an **integer**.

### Tip

To reduce possible aliasing effects, it is a good idea to **blur image slightly before shrinking it**.



## Neighbors of a Pixel

### Definition

A pixel  $p$  at coordinates  $(x, y)$  has **neighbors**,

- $N_4(p): (x + 1, y), (x - 1, y), (x, y + 1), (x, y - 1)$
- $N_D(p): (x + 1, y + 1), (x + 1, y - 1), (x - 1, y + 1), (x - 1, y - 1)$
- $N_8(p): \{N_4(p), N_D(p)\}$

$N_4(p), N_D(p), N_8(p)$  may fall **outside** the image if  $(x, y)$  is on the border of the image.



## Adjacency

### Definition

Let  $V$  be the set of gray-level values used to define adjacency.

- binary image:  $V \subset \{0, 1\}$ .
- 8-bit image  $V \subset \{0, 1, 2, 3, \dots, 255\}$ .

### Example

For example,  $V = \{1\}$  if referring to adjacency of pixels with value 1.



## Adjacency of a pixel

### Definition

\_\_\_\_\_ is the **neighborhood with similar gray-level**:

- \_\_\_\_\_ Two pixels  $p$  and  $q$  with values from  $V$  are 4-adjacent if  $q$  is in  $N_4(p)$ .
- \_\_\_\_\_ Two pixels  $p$  and  $q$  with values from  $V$  are 8-adjacent if  $q$  is in  $N_8(p)$ .
- \_\_\_\_\_ Two pixels  $p$  and  $q$  with values from  $V$  are m-adjacent
  - if  $q$  is in  $N_4(p)$ , **or**
  - if  $q$  is in  $N_D(p)$  **and**  $N_4(p) \cap N_4(q)$  has **no** pixels whose value are from  $V$ .



## Adjacency of a pixel

### Example

			Adjacency		
0	1	1	0	1	1
0	1	0	0	1	0
0	0	1	0	0	1



## Adjacency of an image

### Definition

\_\_\_\_\_ if some pixel in  $S_1$  is (4-,8-,m-)adjacent to some pixel in  $S_2$ .

### Example

TBA



## Digital Path

### Definition

\_\_\_\_\_ from pixel  $p$  with coordinates  $(x, y)$  to pixel  $q$  with coordinates  $(s, t)$  is a **sequence of distinct pixels** with coordinates

$$p = (x, y) = (x_0, y_0), (x_1, y_1), \dots, (x_n, y_n) = (s, t) = q$$

- \_\_\_\_\_ :  $n$  (not  $n + 1$ )
- \_\_\_\_\_ : if  $(x_0, y_0) = (x_n, y_n)$
- \_\_\_\_\_ : depending on the **type of adjacency** specified.
- No ambiguity in the **m-path**.



## Connectivity

### Definition

Let  $S$  be a subset of pixels in an image.

- \_\_\_\_\_ :  
Two pixels \_\_\_\_\_ if there **exists a path between them** consisting **entirely** of pixels **in  $S$** .
- \_\_\_\_\_ :  
For any pixel  $p$  in  $S$ , the **set of pixels that are connected to it in  $S$**  is called a \_\_\_\_\_.
- \_\_\_\_\_ :  
If  $S$  **only has one connected component**, then  $S$  is called \_\_\_\_\_.



## Region and Boundary

### Definition

\_\_\_\_\_ : Let  $R$  be a **subset** of pixels in an image. We call  $R$  a \_\_\_\_\_ of the image if  $R$  is a **connected set**.

### Definition

\_\_\_\_\_ : The boundary of a region  $R$  is the set of pixels in the region that **have one or more neighbors that are not in  $R$** .

### Example

If  $R$  happens to be an entire image, the **boundary** of  $R$  is the pixels in the **first and last rows and columns** of the image.



## Boundary versus Edge

### Definition

- \_\_\_\_\_ :  
  - Pixels with **derivative values** that **exceed a preset threshold**.
  - Based on a measure of **gray-level discontinuity** at a point.
- Boundary:
  - **Closed** path
  - **Global** concept
- Edge:
  - **May not** be a closed path
  - **Local** concept





## Distance

### Euclidean Distance

For pixel  $p, q$  with coordinates  $(x, y), (s, t)$ .

#### Definition

$$D_e(p, q) = \sqrt{(x - s)^2 + (y - t)^2}$$

#### Example

Equal distance: **Disk**



## Distance

### $D_4$ Distance

For pixel  $p, q$  with coordinates  $(x, y), (s, t)$ .

#### Definition

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

#### Example

Equal distance: **Diamond**

```

      2
    2 1 2
  2 1 0 1 2
    2 1 2
      2
    
```

## Distance

### $D_8$ Distance

For pixel  $p, q$  with coordinates  $(x, y), (s, t)$ .

#### Definition

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

#### Example

Equal distance: **Square**

```

    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
    
```

## Distance

### $D_m$ Distance

#### Definition

- $D_m(p, q)$  the **shortest m-path** between the points
- The distance between two pixels will depend on **the values of the pixels along the path** and **the values of their neighbors**.

#### Example

Let  $V = \{1\}$

```

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

$D_m(p, q) = 2$      
  $D_m(p, q) = 3$      
  $D_m(p, q) = 3$      
  $D_m(p, q) = 4$