

# Computer Vision

CSE/ECE 344/544

# Message

- If you do not pass the quiz but still want to do the course, then: (i) this is your research theme/area, (ii) you meet me by Monday (in between 2:30 - 3:30)
- I am ok to teach 10, 20, 50, 100 students if they are interested genuinely but I do not want 200 students just because CV/ML/IR is hot area and it looks good to have these courses in their resume.

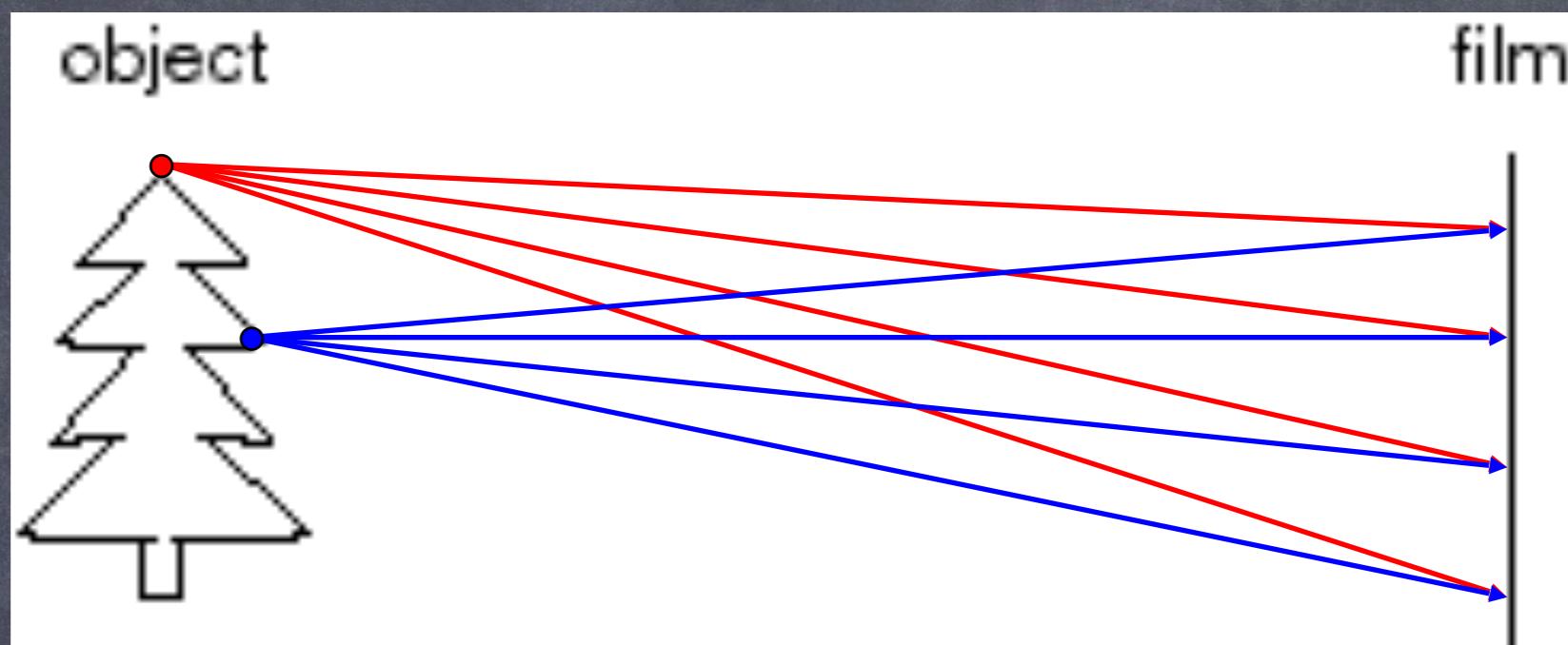


# Computer Vision

Image from: <http://kirkh.deviantart.com/art/BioMech-Eye-168367549>

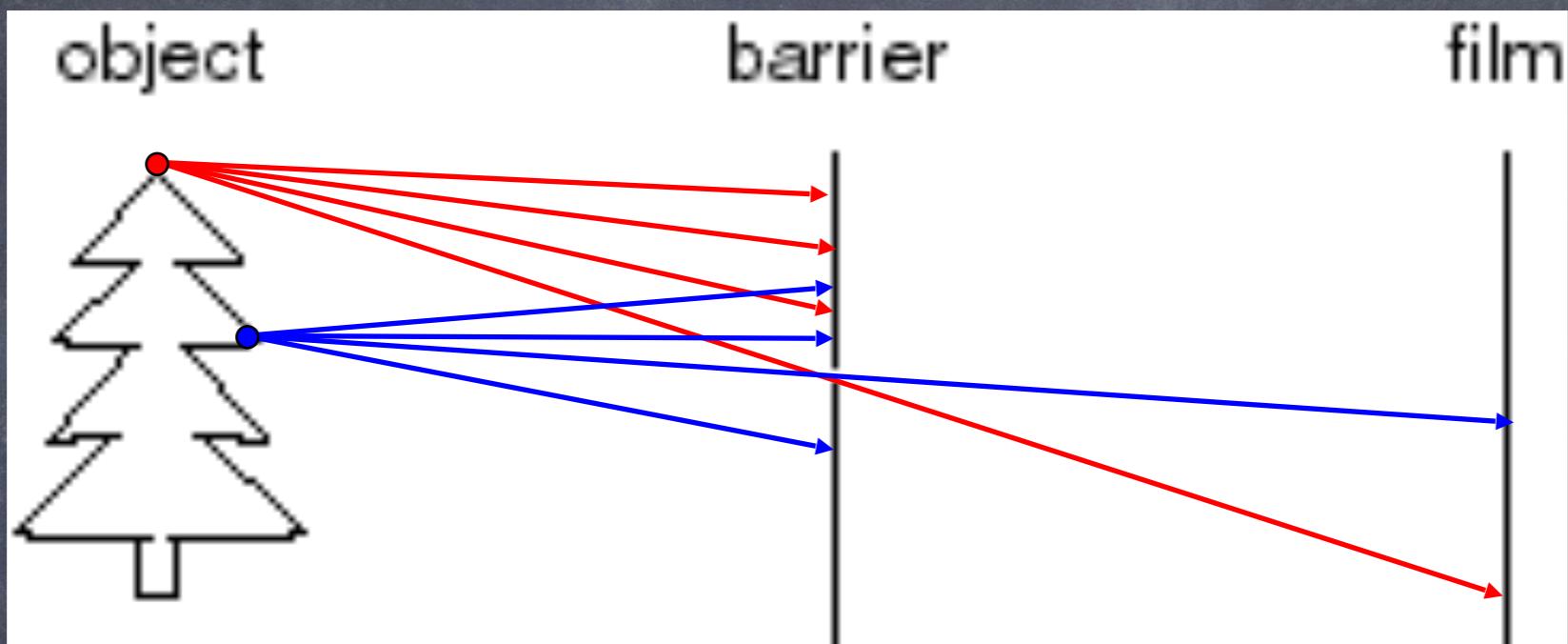
What is the first step in Image Processing and Computer Vision?

# Image formation



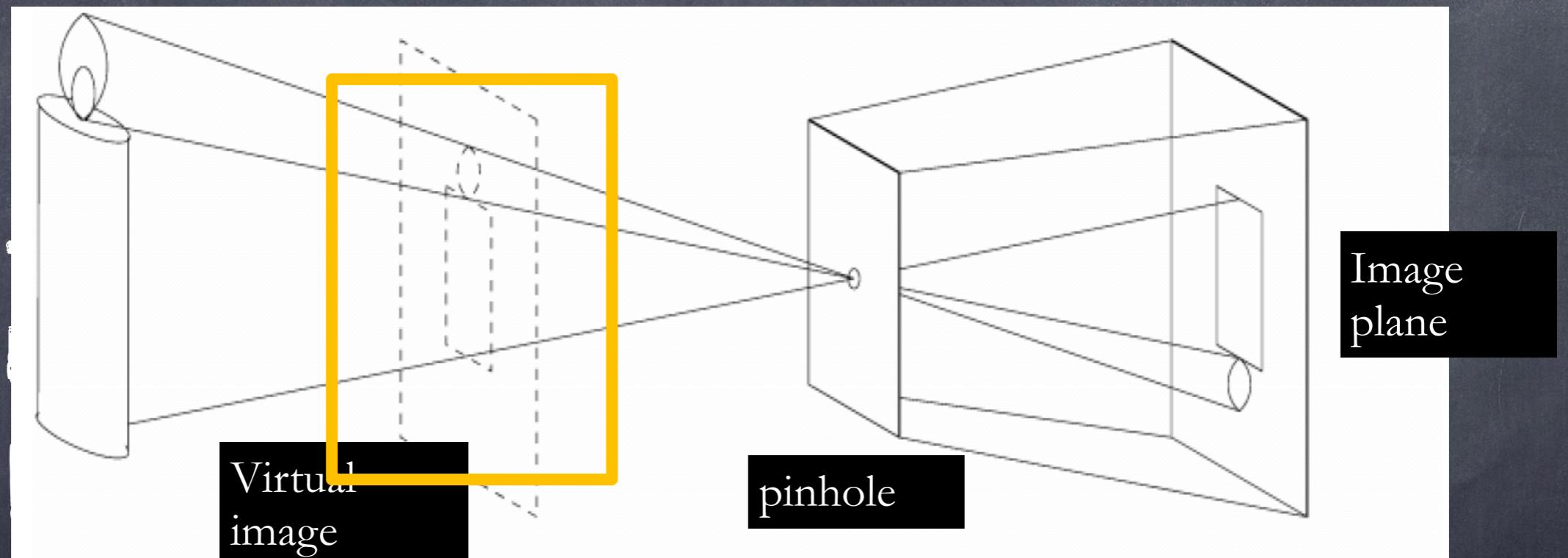
- Let's design a camera
  - Idea 1: put a piece of film in front of an object
  - Do we get a reasonable image?

# Pinhole camera



- Add a barrier to block off most of the rays
  - This reduces blurring
  - The opening is known as the **aperture**
  - How does this transform the image?

# Pinhole camera



If we treat pinhole as a point, only one ray from any given point can enter the camera.

# Perspective effects



# Perspective effects



# Physical parameters of image formation

## Geometric

- Type of projection
- Camera pose

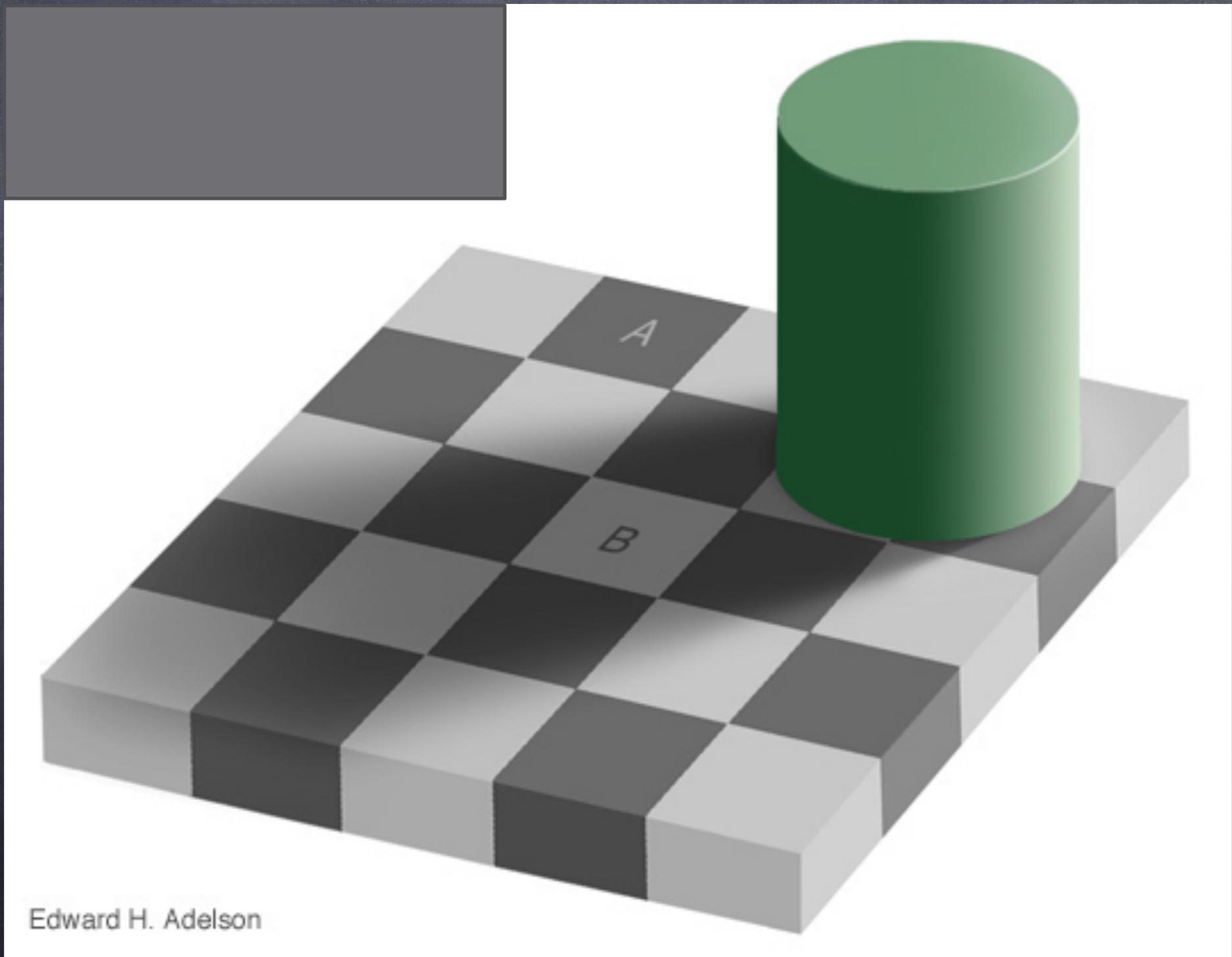
## Optical

- Sensor's lens type
- focal length, field of view, aperture

## Photometric

- Type, direction, intensity of light reaching sensor
- Surfaces' reflectance properties

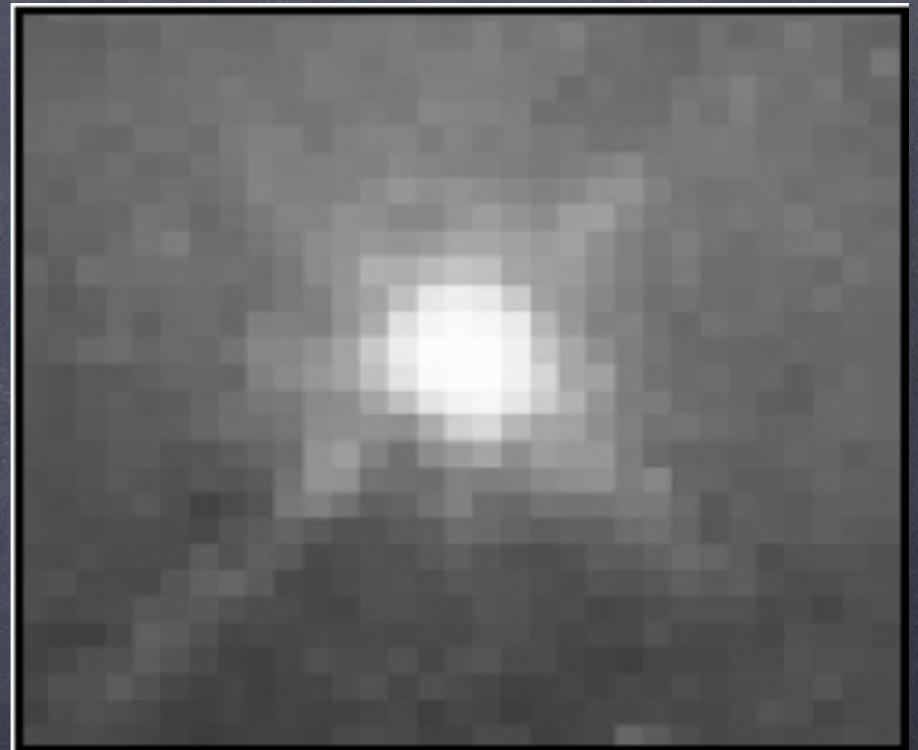
# Why is Vision Hard?



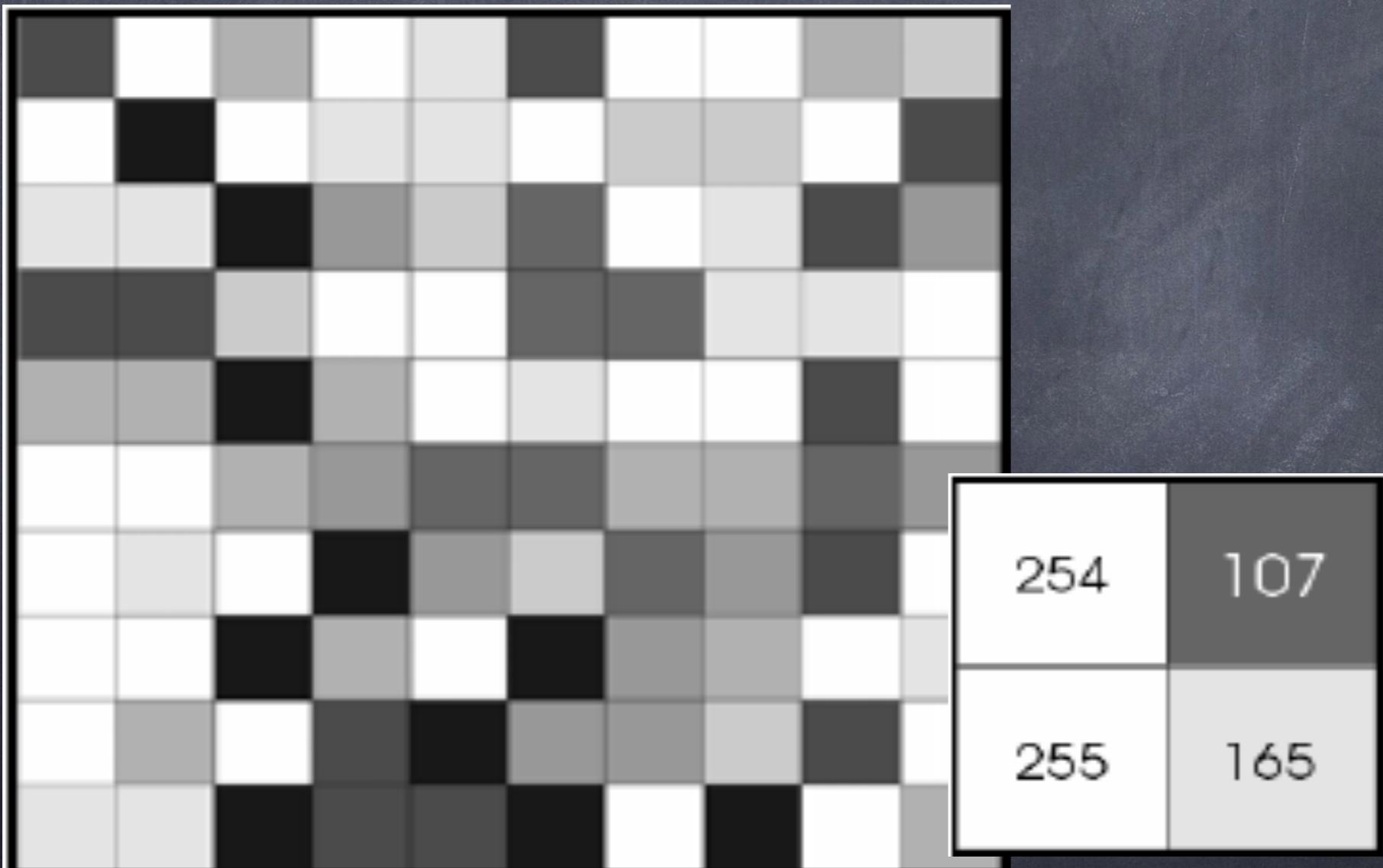
Edward H. Adelson

# What is an image?

- An image is two dimensional function,  $f(x,y)$ , where  $x$  and  $y$  are spatial coordinates, and the amplitude of  $f$  at any pair of coordinates  $(x,y)$  is called the intensity or grey level of the image at that point.



# Digital Image?



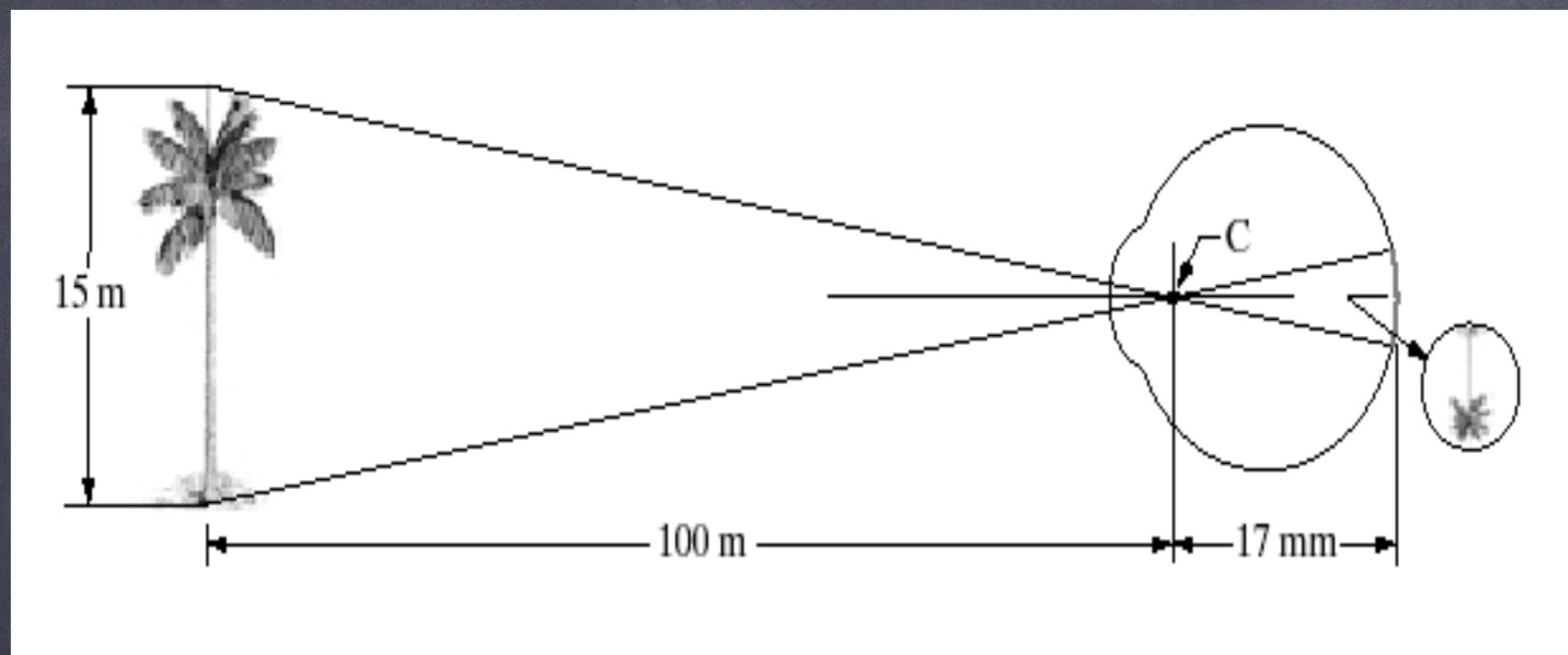
# Fundamental Steps

- Fundamental steps are
  - Image/video Acquisition
  - Enhancement, Restoration and Reconstruction
  - Segmentation
  - Representation and Description
  - Applications: e.g. Object Recognition

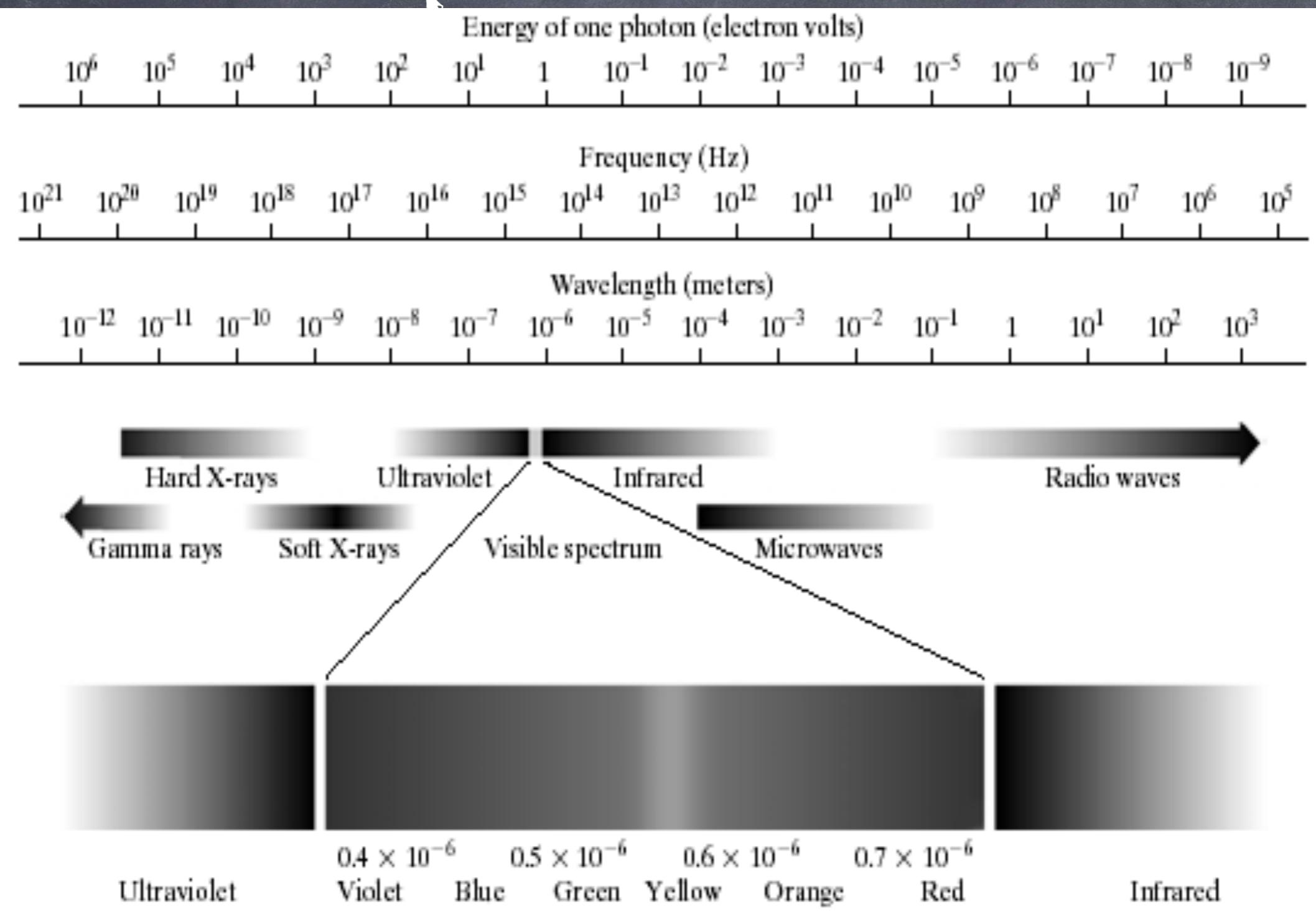
# Acquisition Tools

- Human Eye
- Ordinary Camera
- X-Ray Machine
- Positron Emission Tomography, MRI, etc.
- Infrared Imaging
- Geophysical Imaging

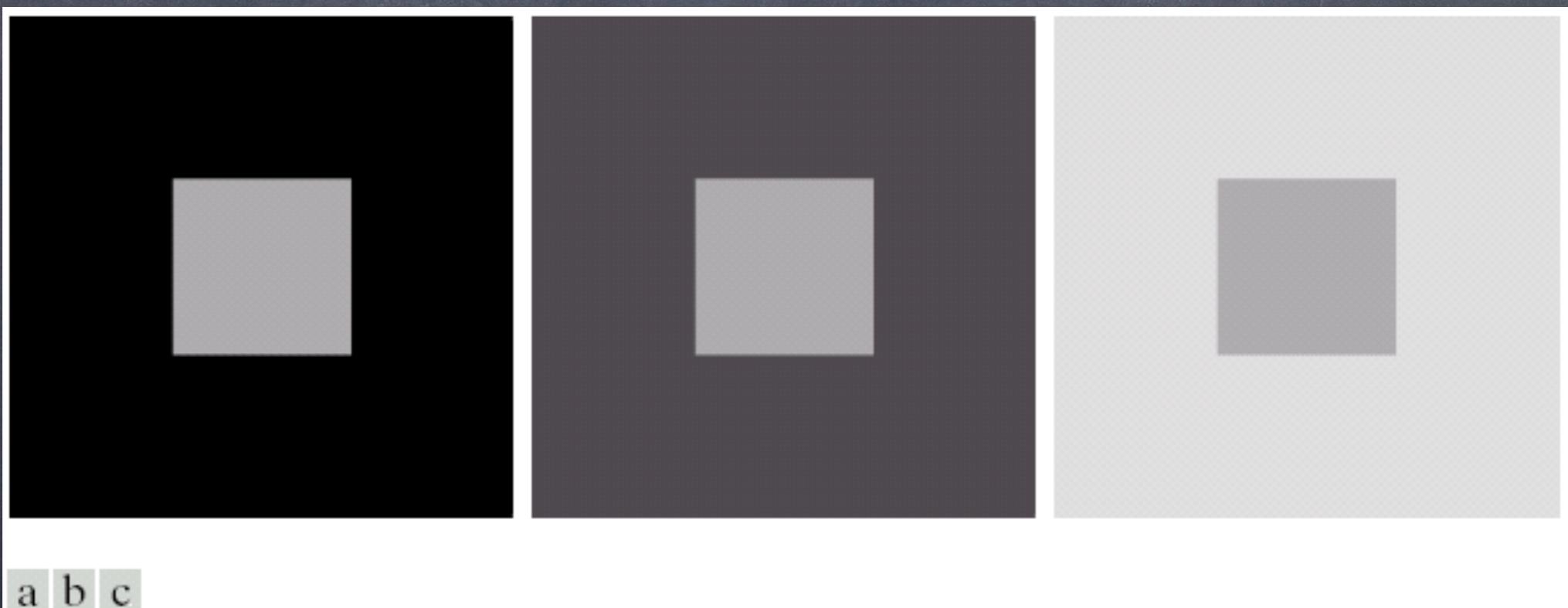
# Image Formation in the Eye



# Electromagnetic spectrum

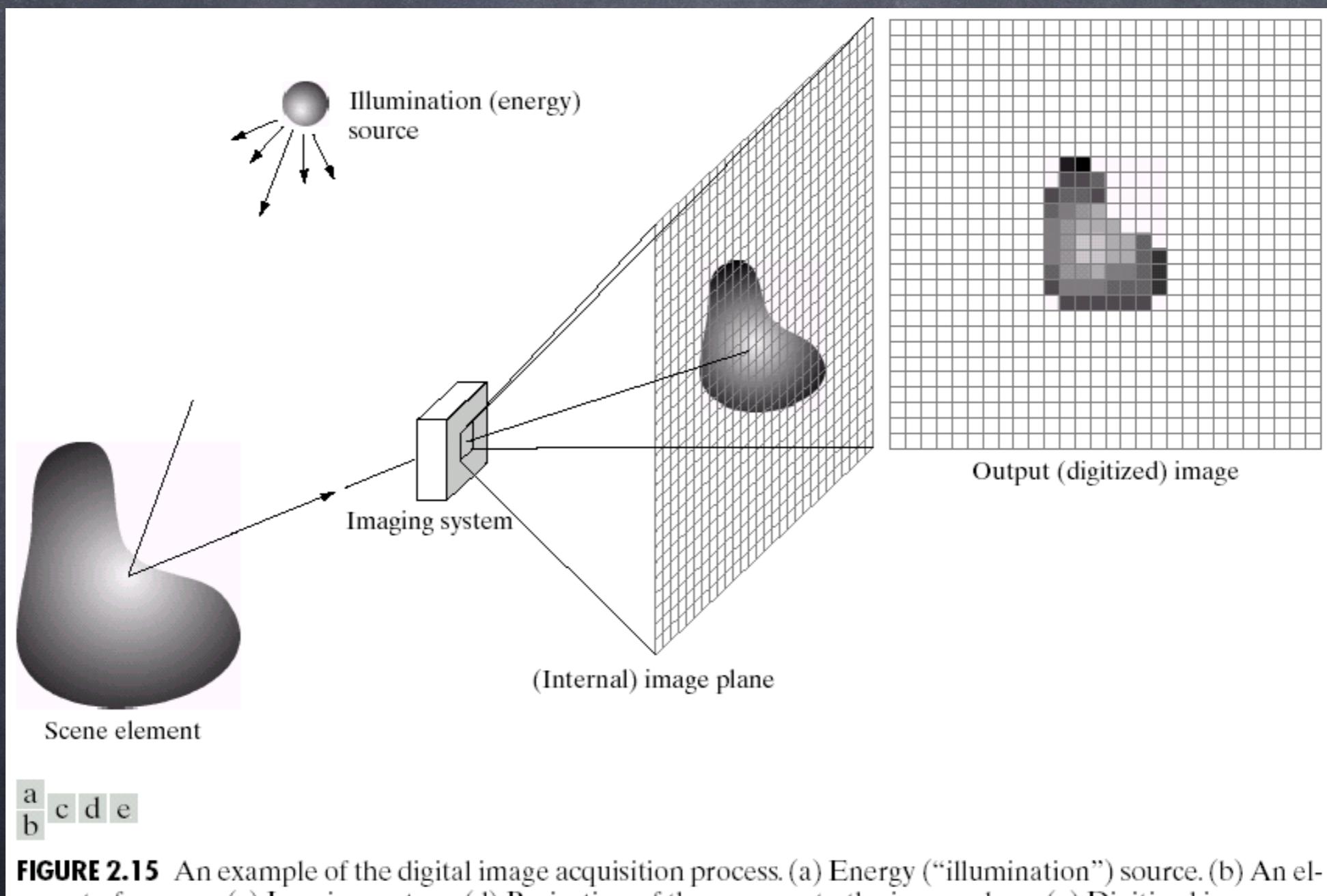


# Image Acquisition ... Depends on background



**FIGURE 2.8** Examples of simultaneous contrast. All the inner squares have the same intensity, but they appear progressively darker as the background becomes lighter.

# Image Formation



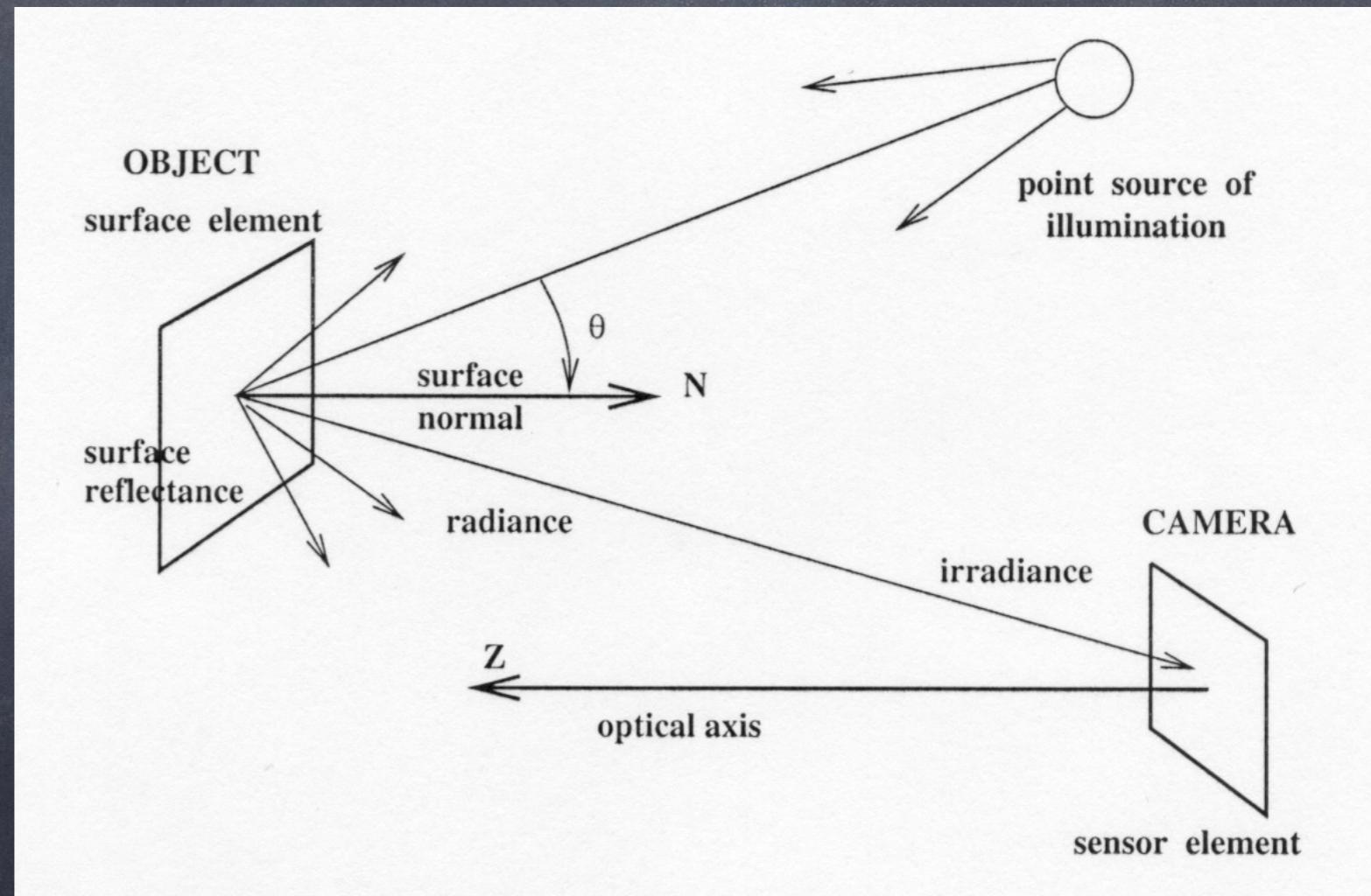
$$f(x,y) = \text{reflectance}(x,y) * \text{illumination}(x,y)$$

# A Simple model of image formation

The scene is illuminated by a single source.

The scene reflects radiation towards the camera.

The camera senses it via chemicals on film (or via sensors).

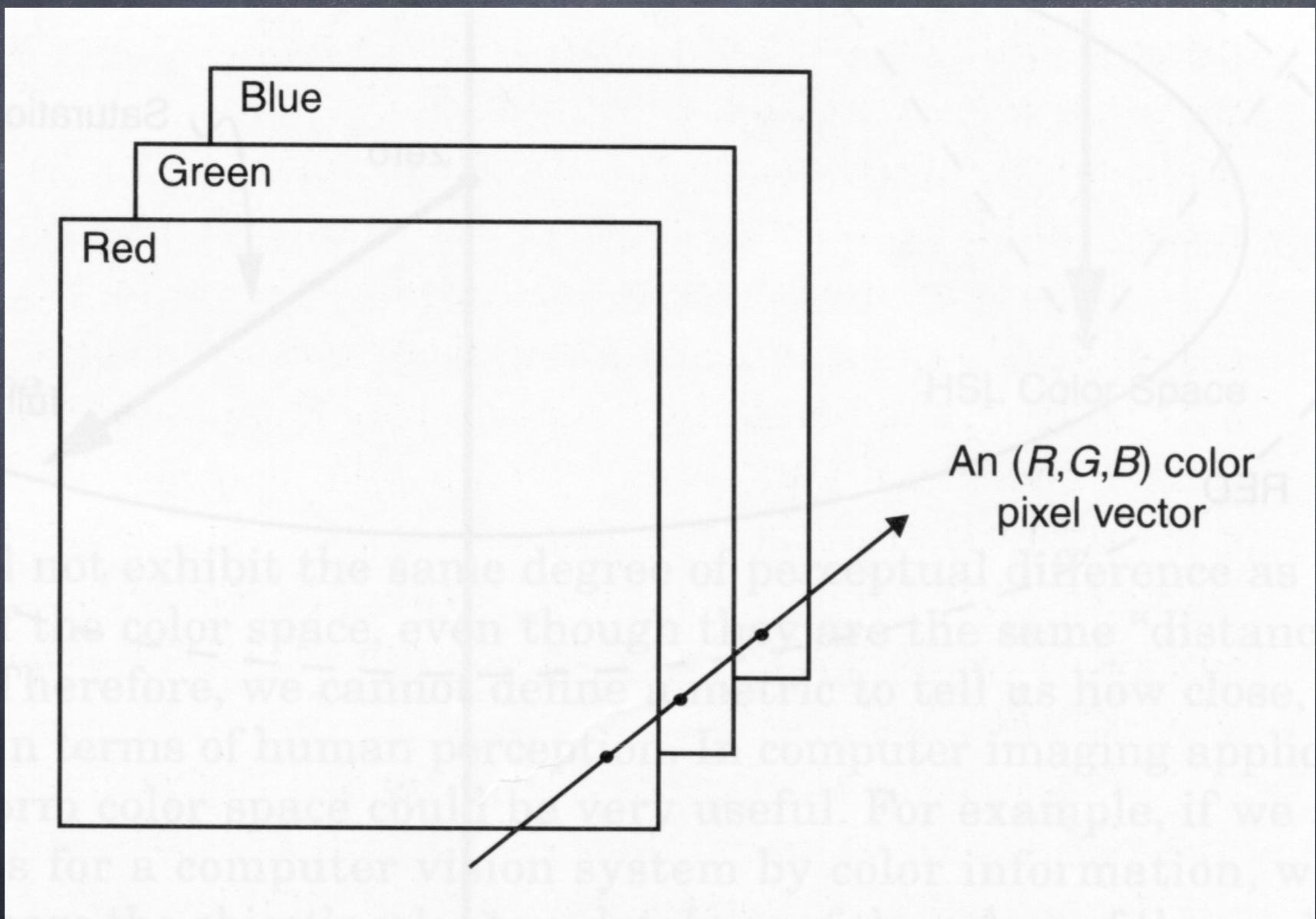


# What is an image?

- We can think of an image as a function,  $f$ 
  - $f(x, y)$  gives the intensity at position  $(x, y)$
  - Realistically, we expect the image only to be defined over a rectangle, with a finite range
- A color image is just three functions pasted together. We can write this as a "vector-valued" function:

$$f(x, y) = \begin{bmatrix} r(x, y) \\ g(x, y) \\ b(x, y) \end{bmatrix}$$

# Color image



# Sampling and Quantization

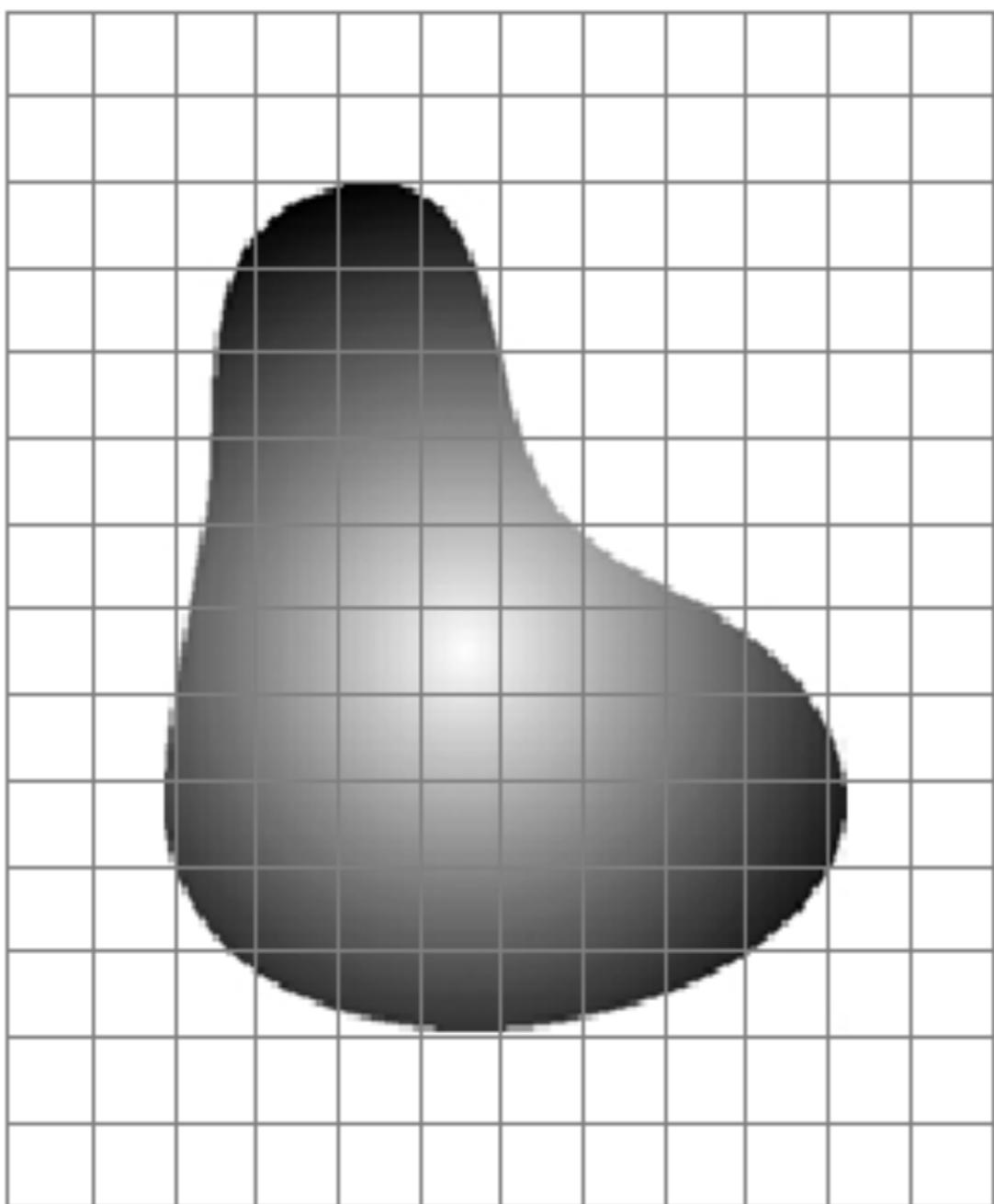
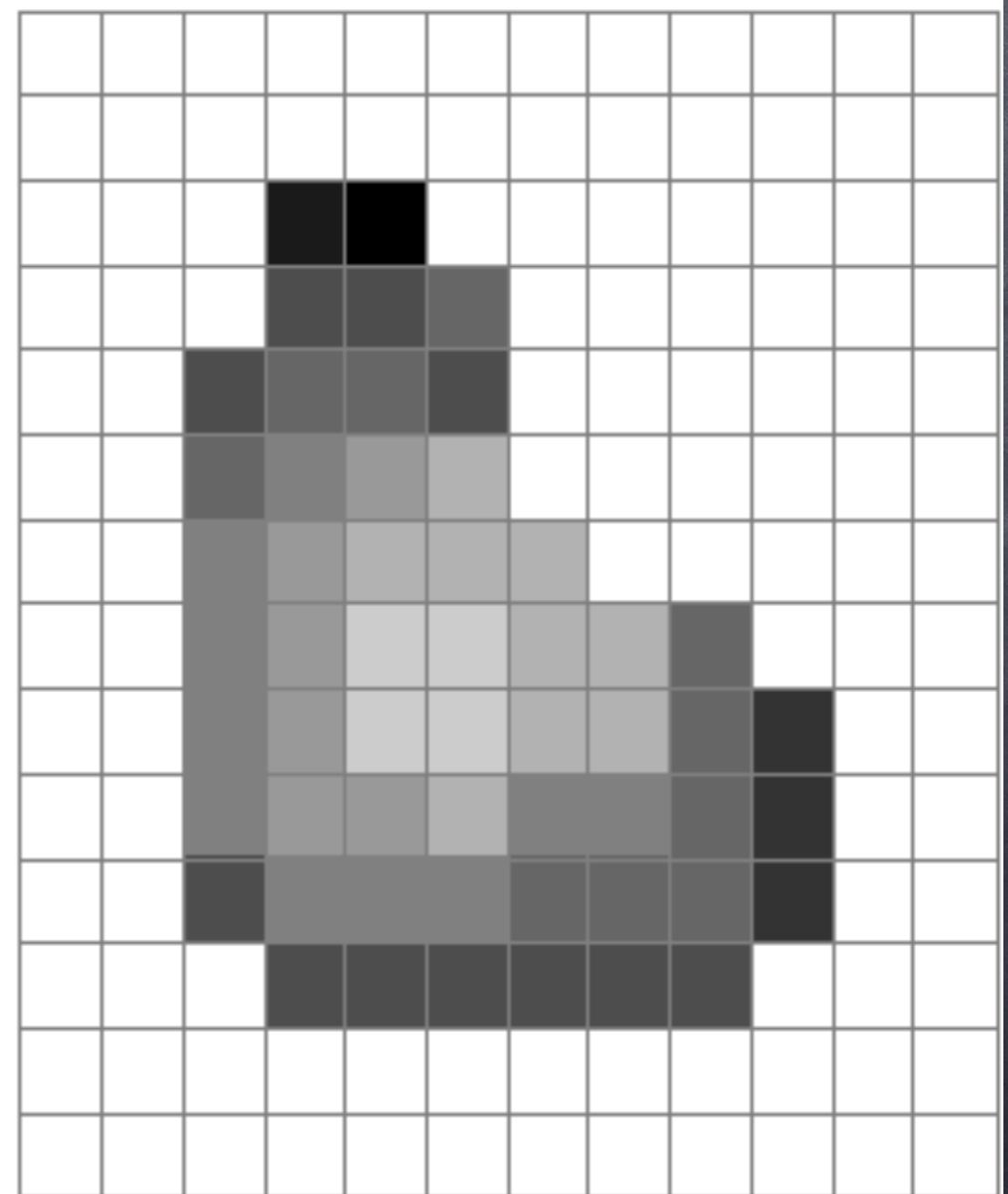


Image before sampling and quantization

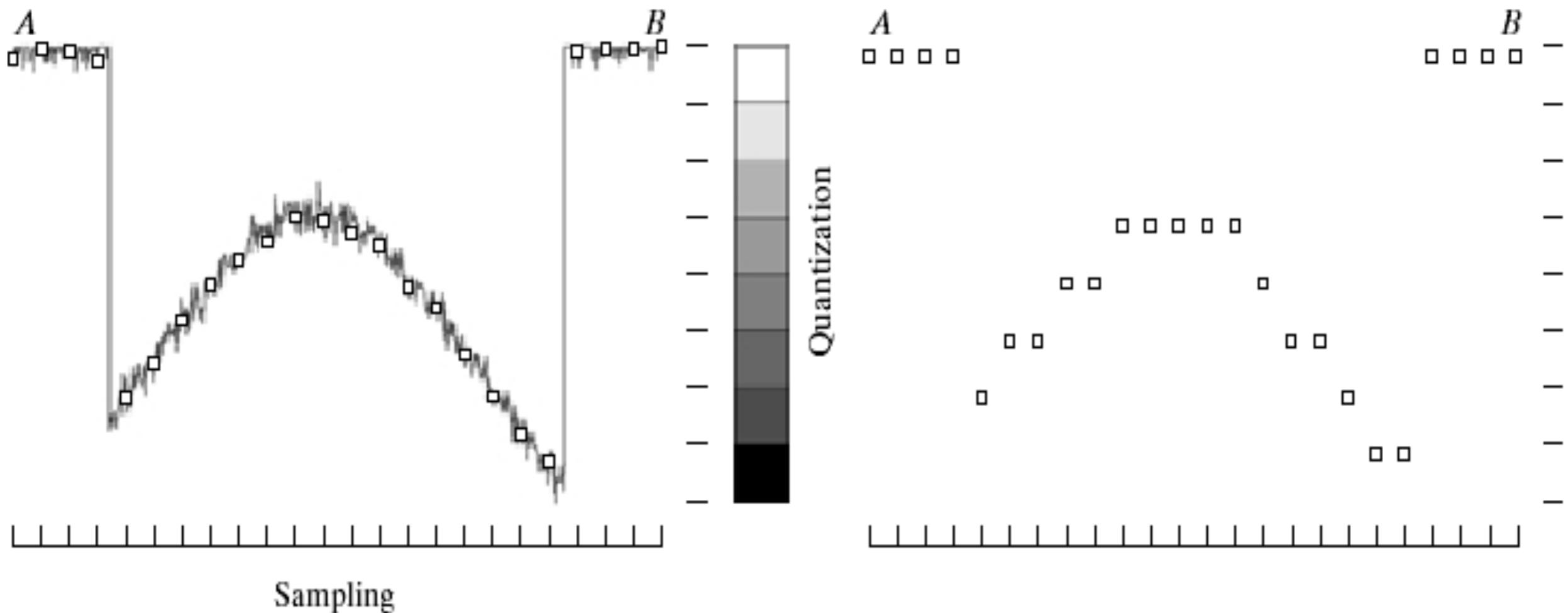


Result of sampling and quantization

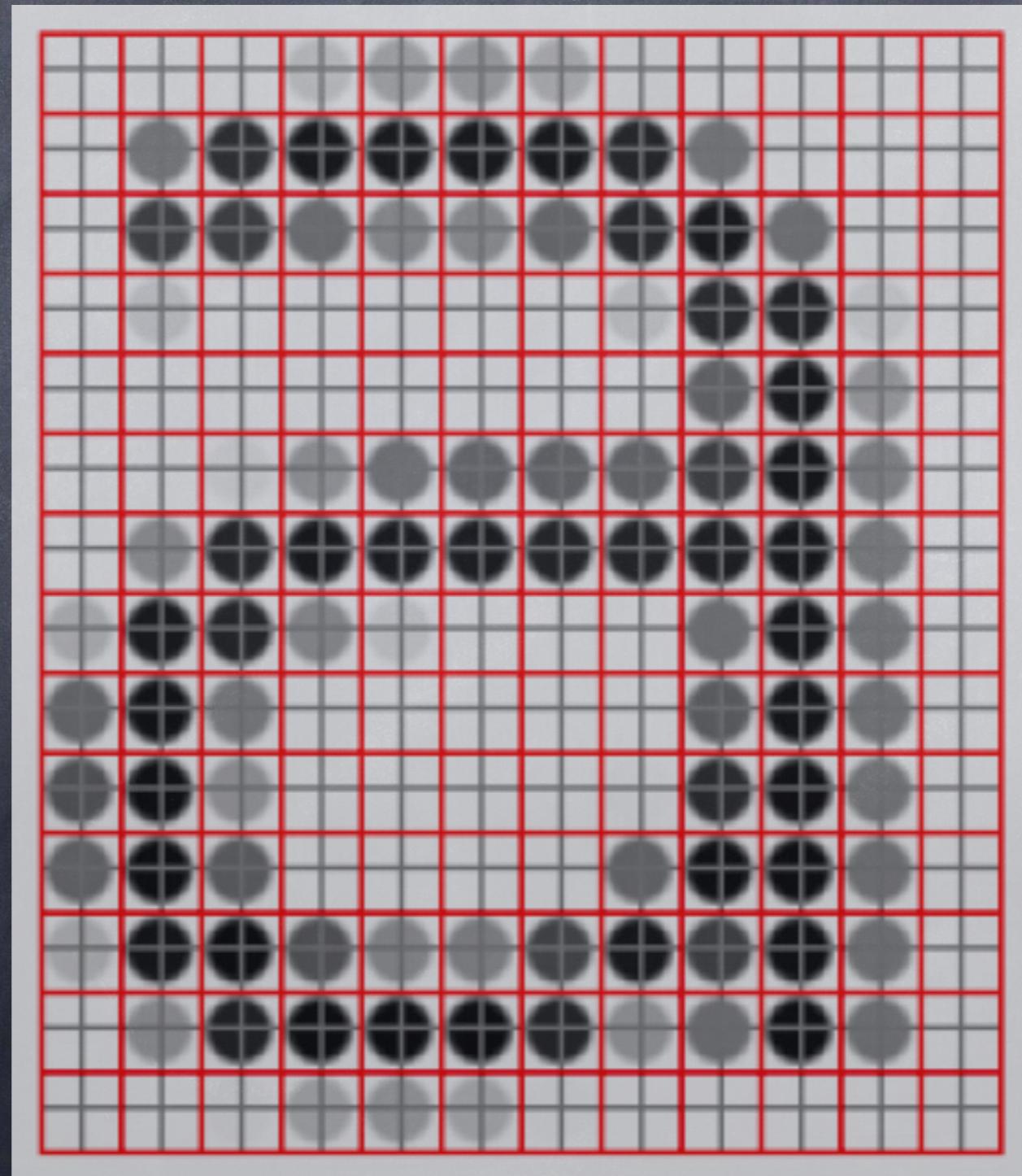
# Image Sampling and Quantization

- The output of most sensors is a continuous voltage waveform whose amplitude and spatial behavior are related to the physical phenomenon being sensed.
- To create a digital image, we need to convert the continuous sensed data into digital form. This involves two processes: *sampling and quantization*.
- Digitizing the coordinate values is called *sampling*.
- Digitizing the amplitude values is called *quantization*.

# Sampling and Quantization



# Image sampling



# Quantization

- Quantization is the procedure of constraining the value of a function at a sampling point to a predetermined finite set of discrete values.
- If we want to specify the temperature of Delhi, ranging from  $0^{\circ}\text{C}$  to  $50^{\circ}\text{C}$ , up to a precision of  $0.1^{\circ}\text{C}$
- we must be able to represent 501 possible values, which require 9 bits to represent one sample.
- On the other hand, if we only need a precision of  $1^{\circ}\text{C}$ , we only have 51 possible values requiring 6 bits for the representation.
- For image processing, higher precision give higher image quality but requires more bits in the representation of the samples.

# Sampling and Quantization

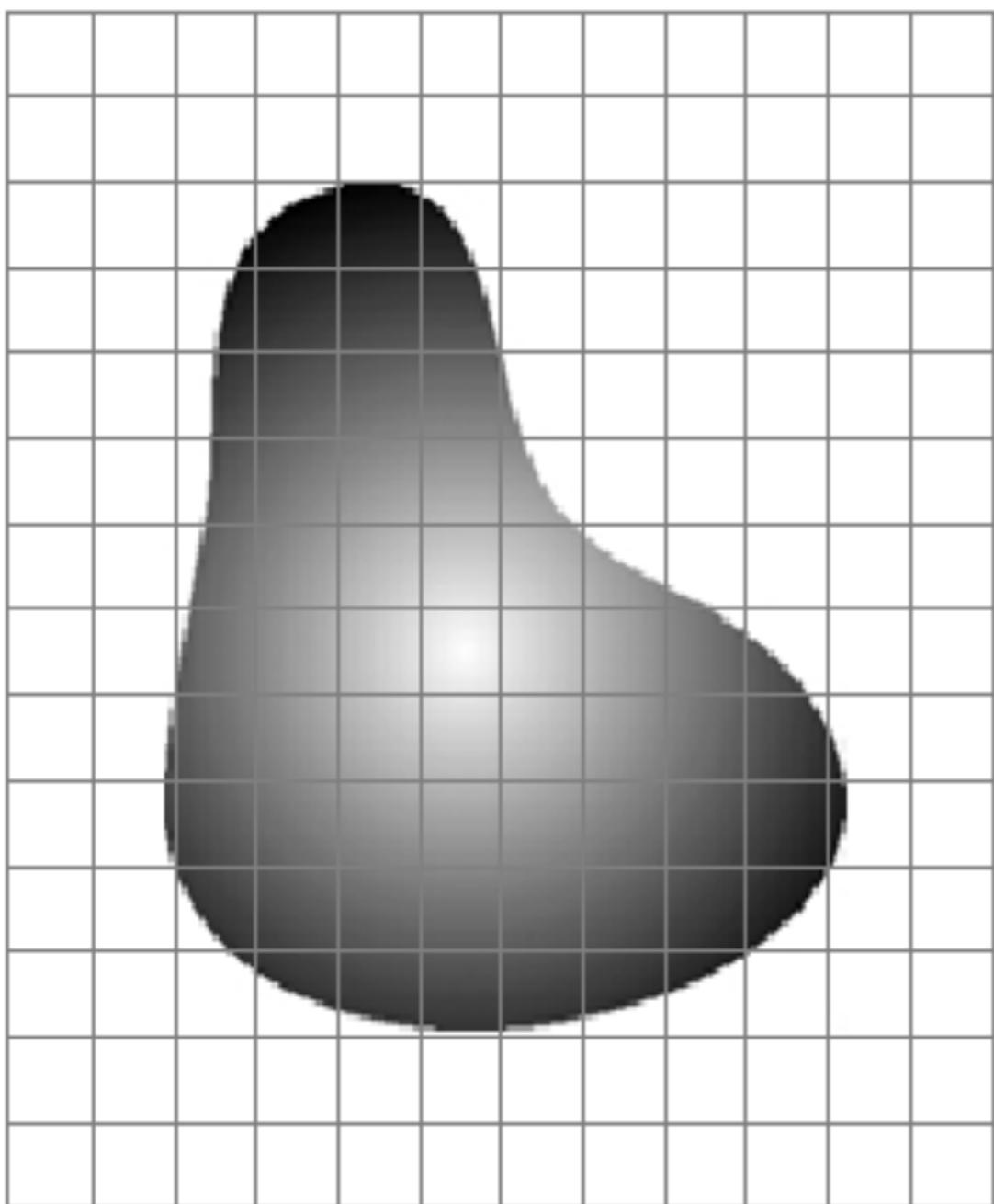
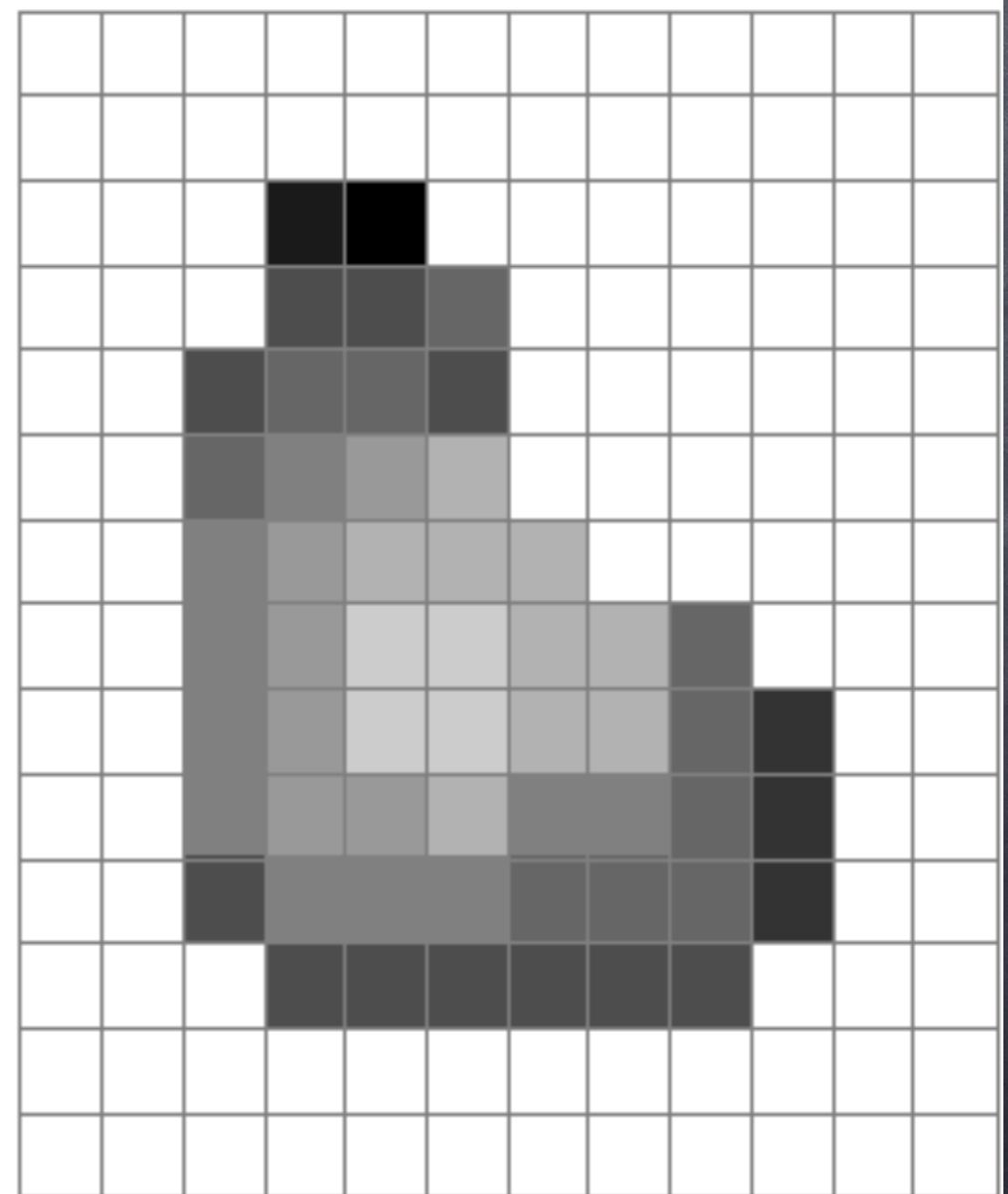


Image before sampling and quantization



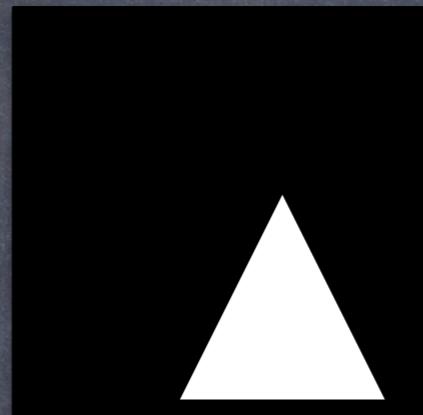
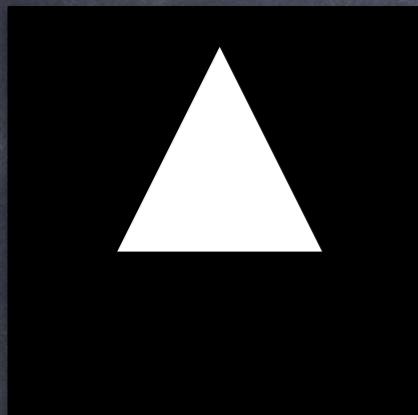
Result of sampling and quantization

# Image Pixel Values

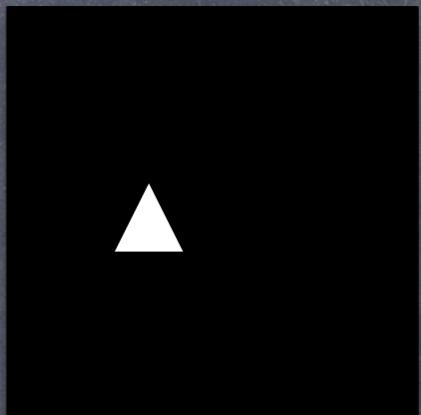
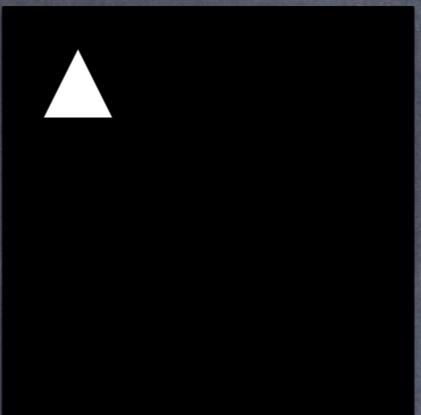


0	3	2	5	4	7	6	9	8
3	0	1	2	3	4	5	6	7
2	1	0	3	2	5	4	7	6
5	2	3	0	1	2	3	4	5
4	3	2	1	0	3	2	5	4
7	4	5	2	3	0	1	2	3
6	5	4	3	2	1	0	3	2
9	6	7	4	5	2	3	0	1
8	7	6	5	4	3	2	1	0

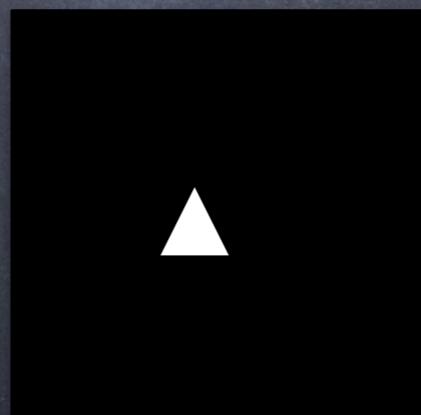
# Example: Big triangles vs. Little triangles



Big triangles

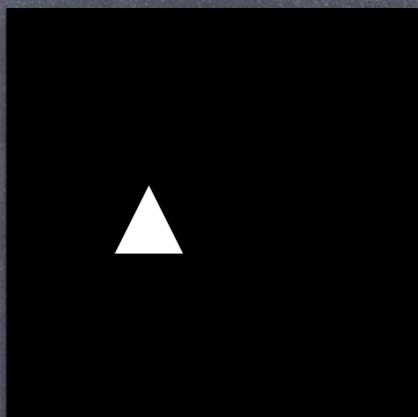
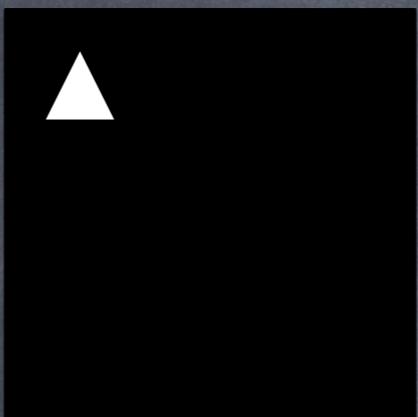
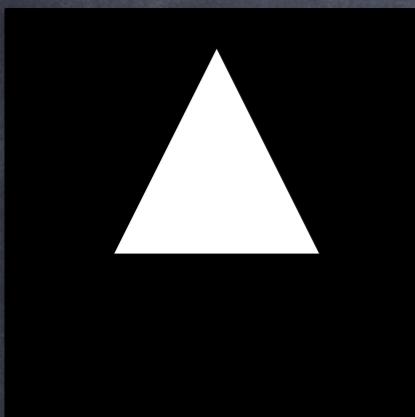


Little triangles



?

Example: Big triangles  
vs. Little triangles



# Example: Big triangles vs. Little triangles

0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	1	0	0	0	0	0	0	0
0	0	0	1	1	1	0	0	0	0	0	0
0	0	0	1	1	1	1	1	0	0	0	0
0	0	1	1	1	1	1	1	1	0	0	0
0	0	0	0	0	0	0	0	0	1	0	0
0	0	0	0	0	0	0	0	0	0	1	0
0	0	0	0	0	0	0	0	0	0	0	1
0	0	0	0	0	0	0	0	0	0	0	0

Sum = 16

0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0

Sum = 13

0	0	0	0	0	0	0	0	0	0	0	0
0	1	1	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0

Sum = 2

0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0

Sum = 3

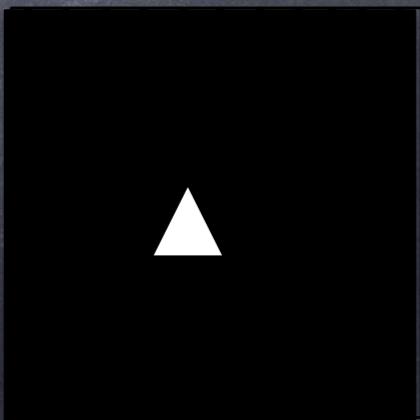
## Rule

If sum > 10

Answer = Big triangle

Else

Answer = Little triangle



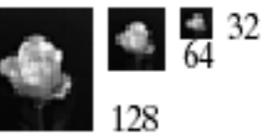
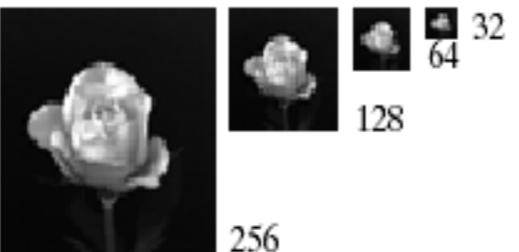
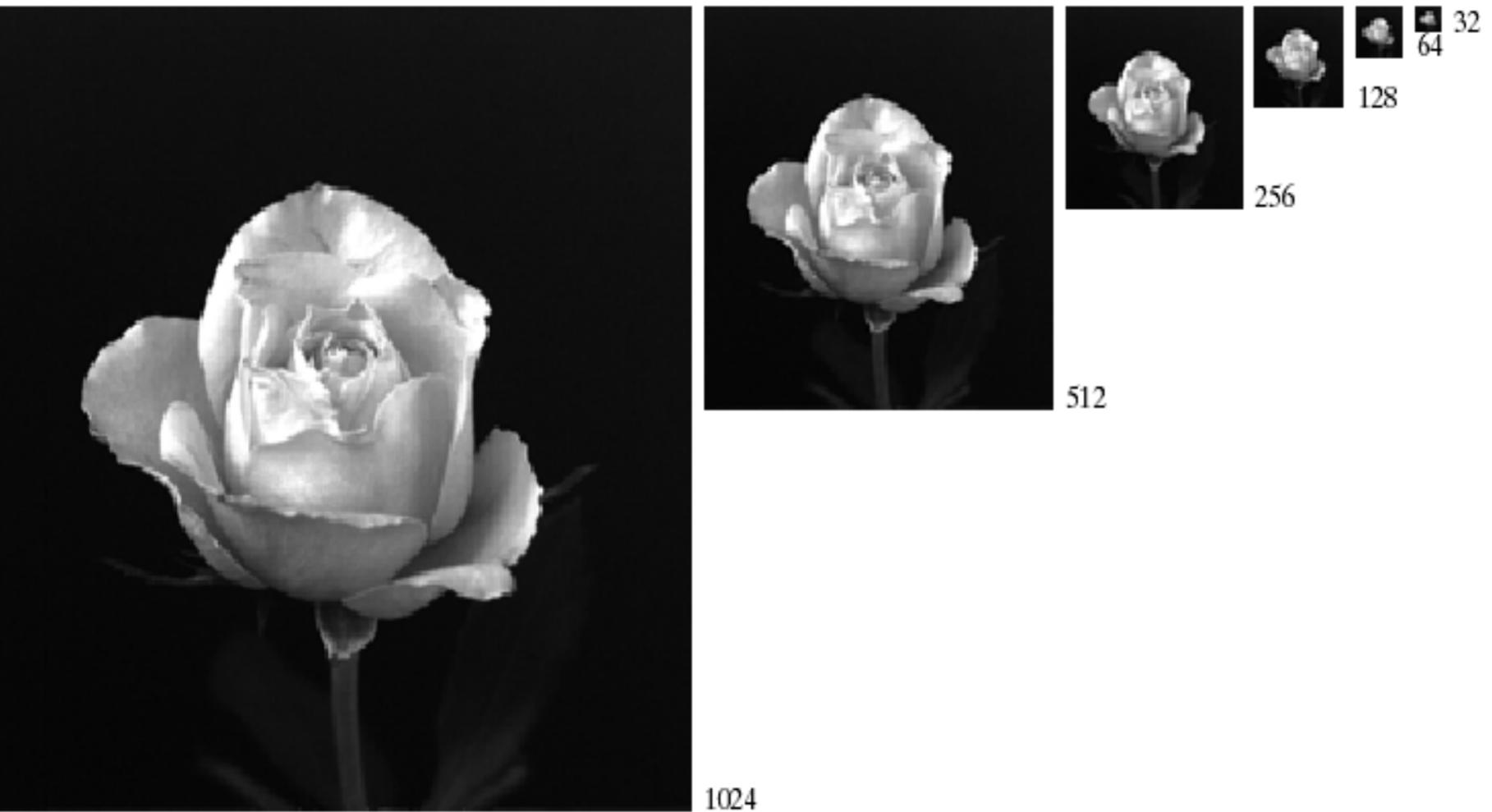
Little triangle

# Image Size

- Due to processing, storage, and sampling hardware considerations, the number of gray levels typically is an integer power of 2:  
$$L = 2^k$$
- Where  $k$  is number of bits required to represent a grey value
- The discrete levels should be equally spaced and that they are integers in the interval  $[0, L-1]$ .
- The number of bits,  $b$ , required to store a digitized image is  
$$b=M*N*k.$$
- Where  $M, N$  represent size of image

# Resolution

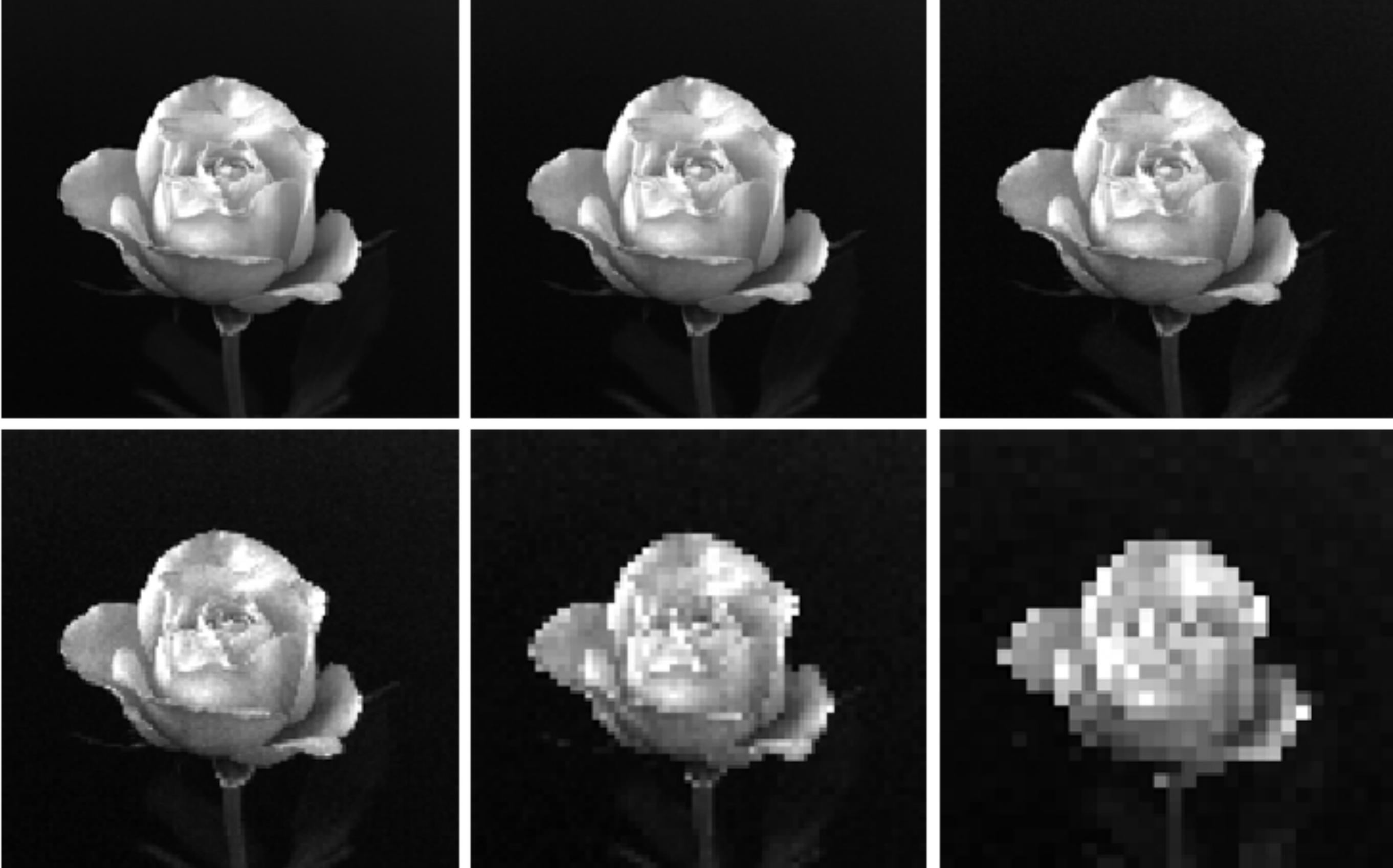
- Spatial Resolution: Spatial resolution is the smallest detectable detail in an image.
- Grey level Resolution: *Gray-level resolution* similarly refers to the smallest detectable change in gray level.



32

128

256



(a) 1024\*1024, 8-bit image. (b) 512\*512 image resampled into 1024\*1024 pixels by row and column duplication. (c) through (f) 256\*256, 128\*128, 64\*64, and 32\*32 images resampled into 1024\*1024 pixels.

# MP, PPI, DPI ...

- Megapixel:

- $2048 \times 1536 = 3,145,728$  pixels

- 3.1 megapixels

- PPI/DPI:

- 15 inch (38 cm) display

- 12 inches (30.48 cm) wide by 9 inches (22.86 cm) high,

# IP CV

- An image processing and computer vision operation typically defines a new image  $g$  in terms of an existing image  $f$ .

- We can transform either the range of  $f$ .

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

- Or the domain of  $f$ :

$$g(x, y) = f(t_x(x, y), t_y(x, y))$$

# IP - CV

- In reality, image processing and CV is nothing but playing with matrices operation with 256 grayscale images-
- Absolute difference of two images
- Add two images or add constant to image
- Complement image
- Divide one image into another or divide image by constant
- Linear combination of images
- Multiply two images or multiply image by constant
- Subtract one image from another or subtract constant from image

# Absolute difference of two images

- $X = [255 \ 10 \ 75; 44 \ 225 \ 100];$
- $Y = [50 \ 50 \ 50; 50 \ 50 \ 50];$
- $Z = \text{adi}(X, Y) \rightarrow$  we need to write a function  
 $\text{adi}$
- $Z = [205 \quad 40 \quad 25; 6 \quad 175 \quad 50]$

# Absolute difference of two images



Add two images or add constant to  
image

- $X = [255 \ 0 \ 75; 44 \ 225 \ 100];$
- $Y = [50 \ 50 \ 50; 50 \ 50 \ 50];$
- $Z = \text{add2im}(X, Y)$
- $Z =$
- $[255 \ 50 \ 125; 94 \ 255 \ 150]$

Subtract two images or add constant to  
image

•  $X = [255 \ 10 \ 75; 44 \ 225 \ 100];$

•  $Y = [50 \ 50 \ 50; 50 \ 50 \ 50];$

•  $Z = \text{sub2im}(X, Y)$

•  $Z =$

•  $[205 \ 0 \ 25; 0 \ 175 \ 50]$

# Complement image

- ④  $X = [255 \ 10 \ 75; 44 \ 225 \ 100];$
- ④  $X_2 = \text{comp}(X)$
- ④  $X_2 =$
- ④  $[0 \ 245 \ 180; 211 \ 30 \ 155]$

Divide one image into another or divide  
image by constant

•  $X = [255 \ 10 \ 75; 44 \ 225 \ 100];$

•  $Y = [50 \ 20 \ 50; 50 \ 50 \ 50];$

•  $Z = \text{div}(X, Y)$

•  $Z =$

•  $[5 \ 1 \ 2; 1 \ 5 \ 2]$

# Transpose

SampleArray = 1 3 5

2 4 6

SampleArray'

= 1 2  
3 4  
5 6



# Nested Operations

- $X = [ 255 \ 10 \ 75; 44 \ 225 \ 100];$
- $Y = [ 50 \ 20 \ 50; 50 \ 50 \ 50 ];$
- $Z = \text{div}(\text{add2im}(X,Y), 2)$
- $Z =$
  
- $[128 \quad 15 \quad 63; \quad 47 \quad 128 \quad 75]$

# IP CV

- An image processing - computer vision operation typically defines a new image  $g$  in terms of an existing image  $f$ .
- We can transform either the range of  $f$ .

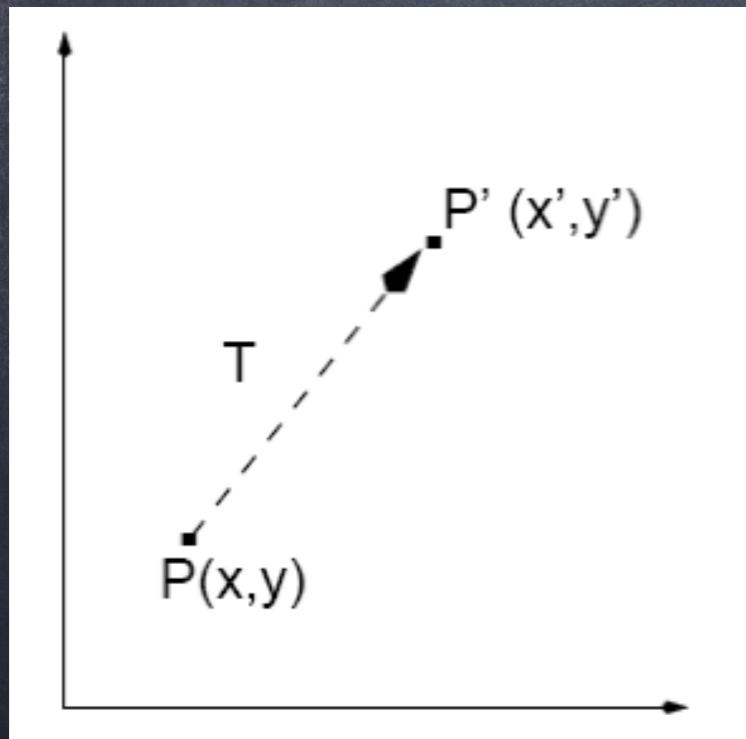
$$g(x, y) = t(f(x, y))$$

- Or the domain of  $f$ :

$$g(x, y) = f(t_x(x, y), t_y(x, y))$$

# 2D Translation

- Moves a point to a new location by adding translation amounts to the coordinates of the point.



$$x' = x + dx, \quad y' = y + dy$$

or

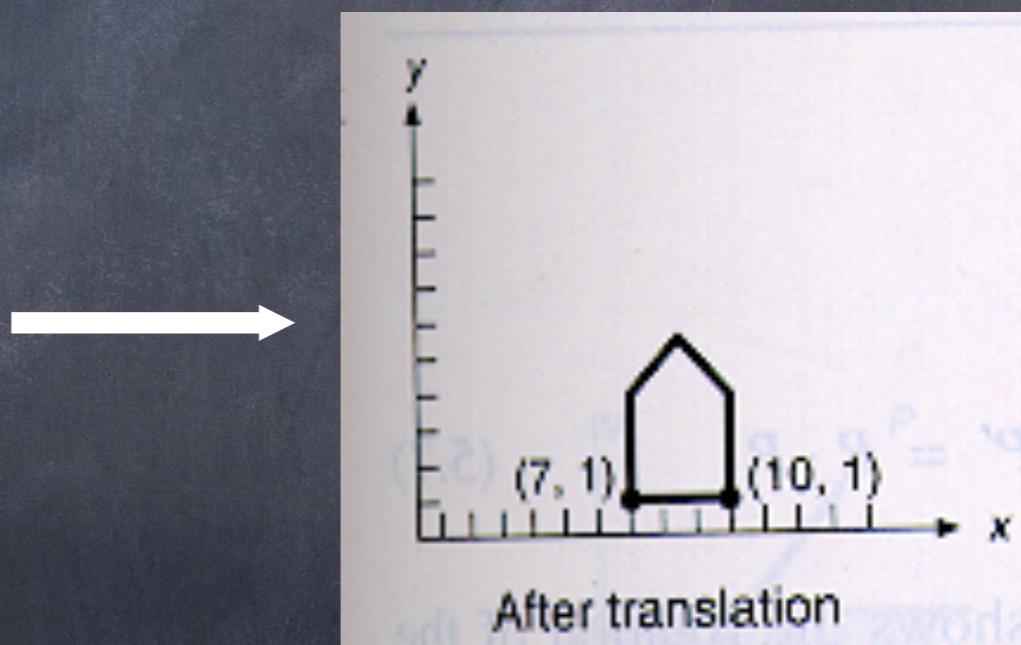
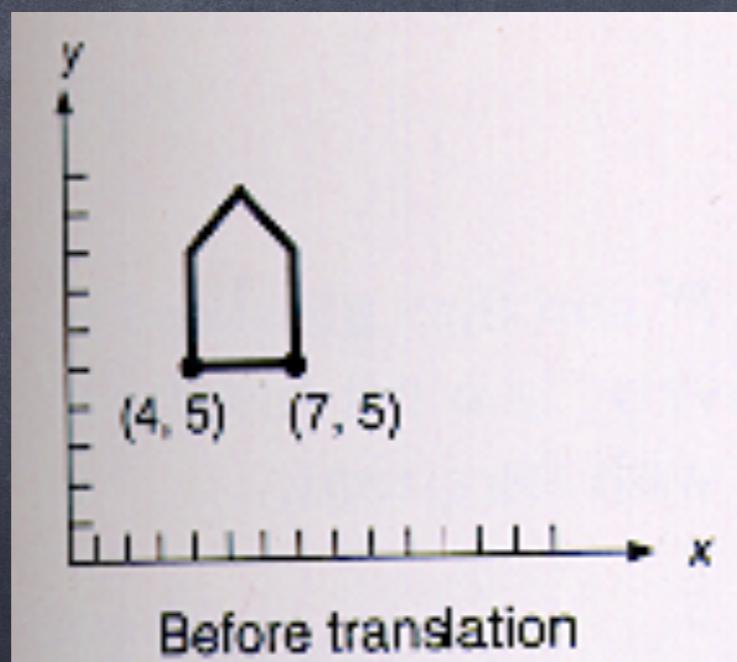
$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} dx \\ dy \end{bmatrix}$$

or

$$\underline{P' = P + T}$$

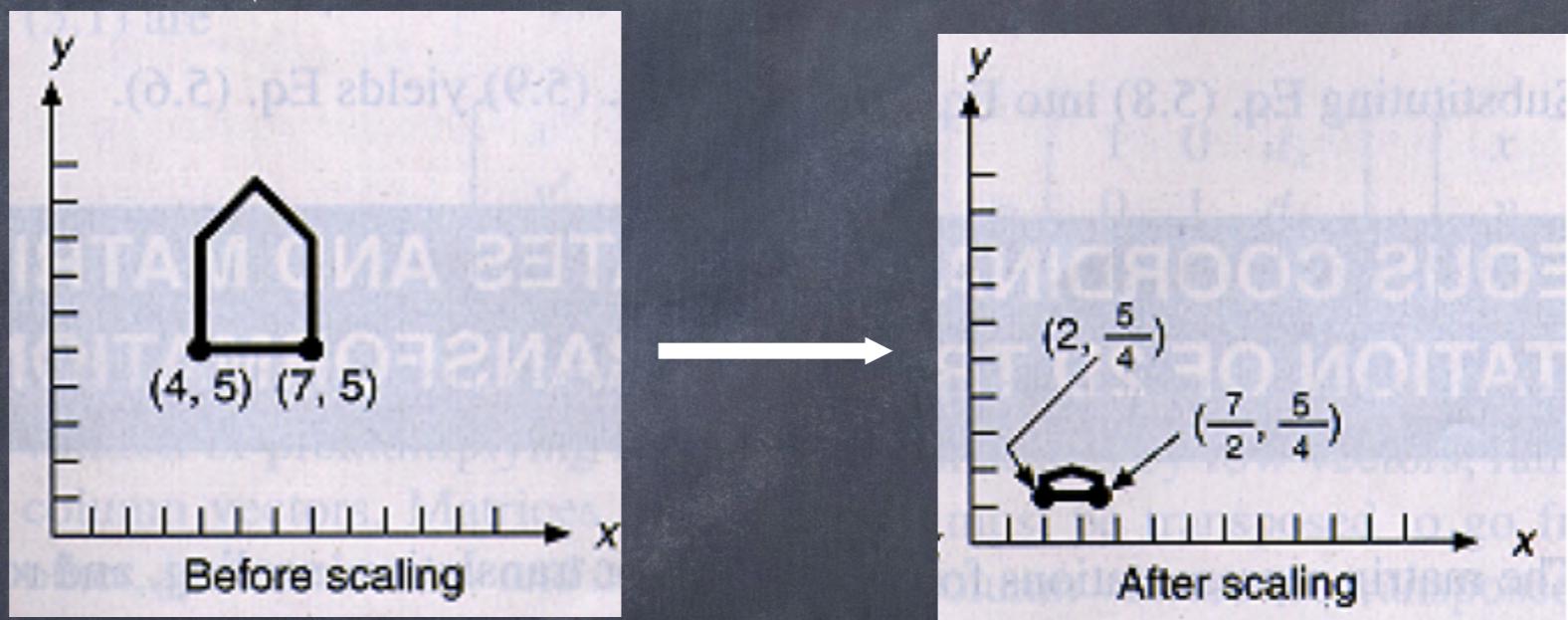
# 2D Translation (cont'd)

- To translate an object, translate every point of the object by the same amount.



# 2D Scaling

- Changes the size of the object by multiplying the coordinates of the points by scaling factors.



$$x' = x s_x, \quad y' = y s_y$$

or

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}, \quad s_x, s_y > 0$$

or

$$P' = S P$$

# 2D Scaling (cont'd)

If  $s_x = s_y$  uniform scaling

If  $s_x \neq s_y$  nonuniform scaling

If  $s_x, s_y < 1$ , size is reduced, object moves closer to origin

If  $s_x, s_y > 1$ , size is increased, object moves further from origin

If  $s_x = s_y = 1$ , size does not change

# 2D Rotation

- Rotates points by an angle  $\theta$  about origin
- ( $\theta > 0$ : counterclockwise rotation)

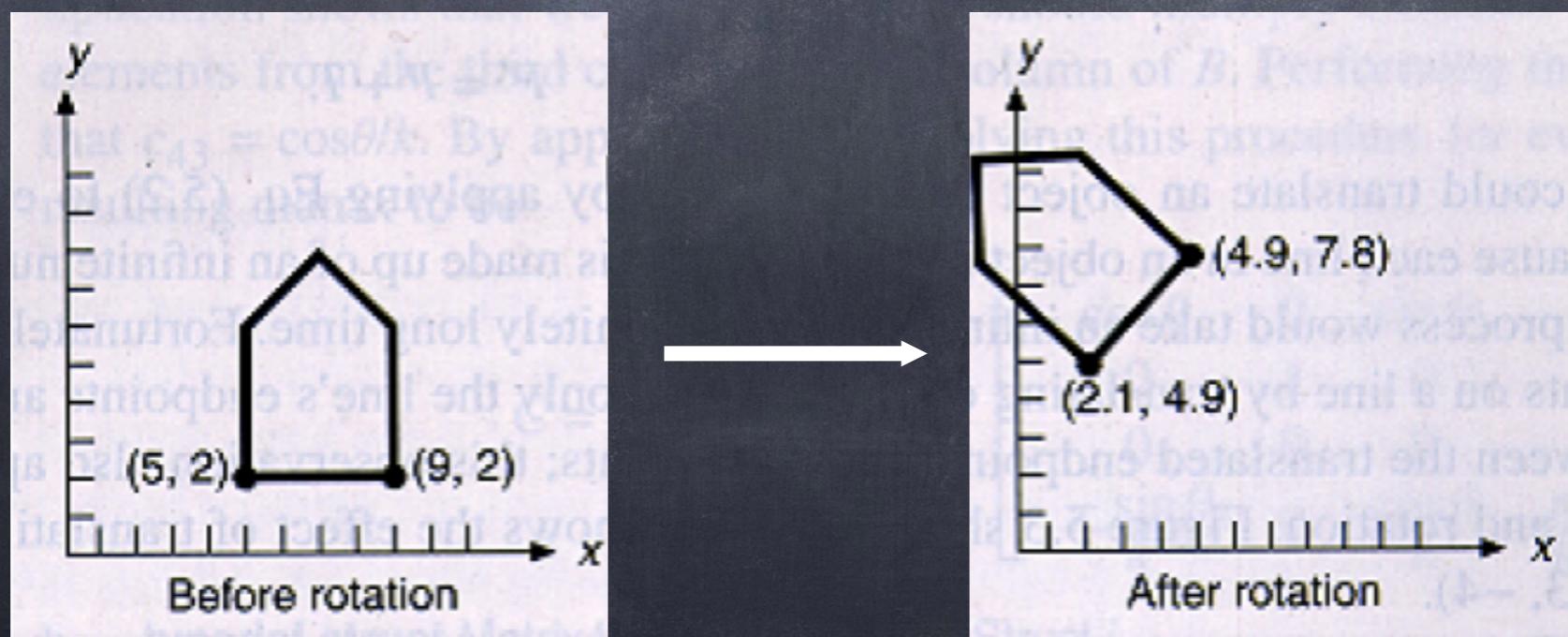
# 2D Rotation (cont'd)

- We have:

$$x' = x\cos(\theta) - y\sin(\theta), \quad y' = x\sin(\theta) + y\cos(\theta) \quad \text{or}$$

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} \quad \text{or}$$

$$\underline{P' = R P}$$



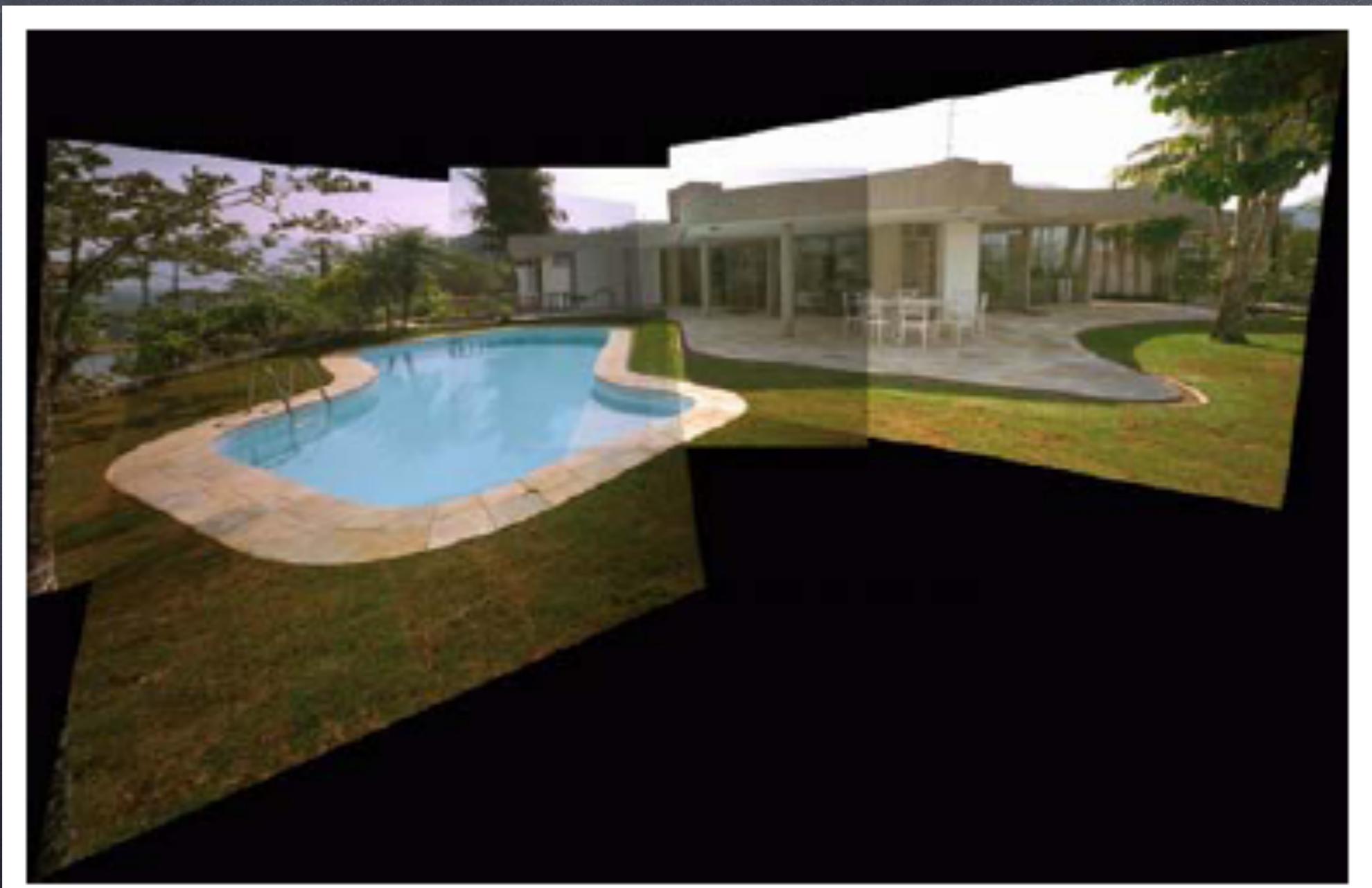
# Summary of 2D transformations

Translation:  $P' = P + T$

Scale:  $P' = S P$

Rotation:  $P' = R P$

Food for thought: How can we create an image mosaic?



# Affine Transform

• Next lecture