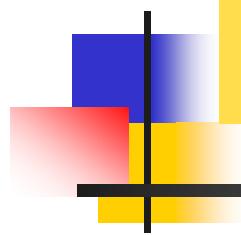


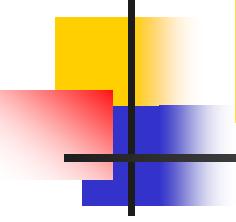
[2016 1학기 확률 및 통계]

# Image Processing



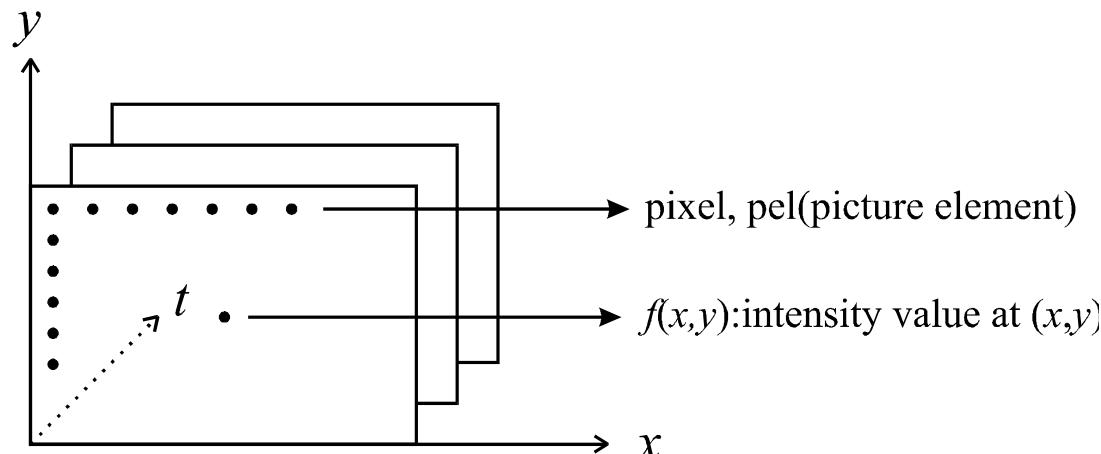
이상화

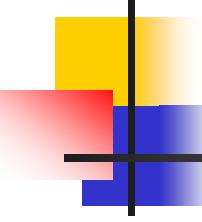
2016년 4월 19일



# What is Image?

- Image is a 2-D spatially sampled signal of 3-D world.
  - Sampling grid: pixel
  - $(x,y)$  : spatial coordinate
  - $t$  : temporal coordinate (video sequence)
- Image signal is a function of  $(x,y)$ .
  - 3-D surface on the  $x$ - $y$  plane
  - $0 \leq f(x,y) \leq L (=255)$ : gray - 8bit/pixel, color level- 24bit/pixel
  - Video signal is a function of  $(x,y,t)$ .

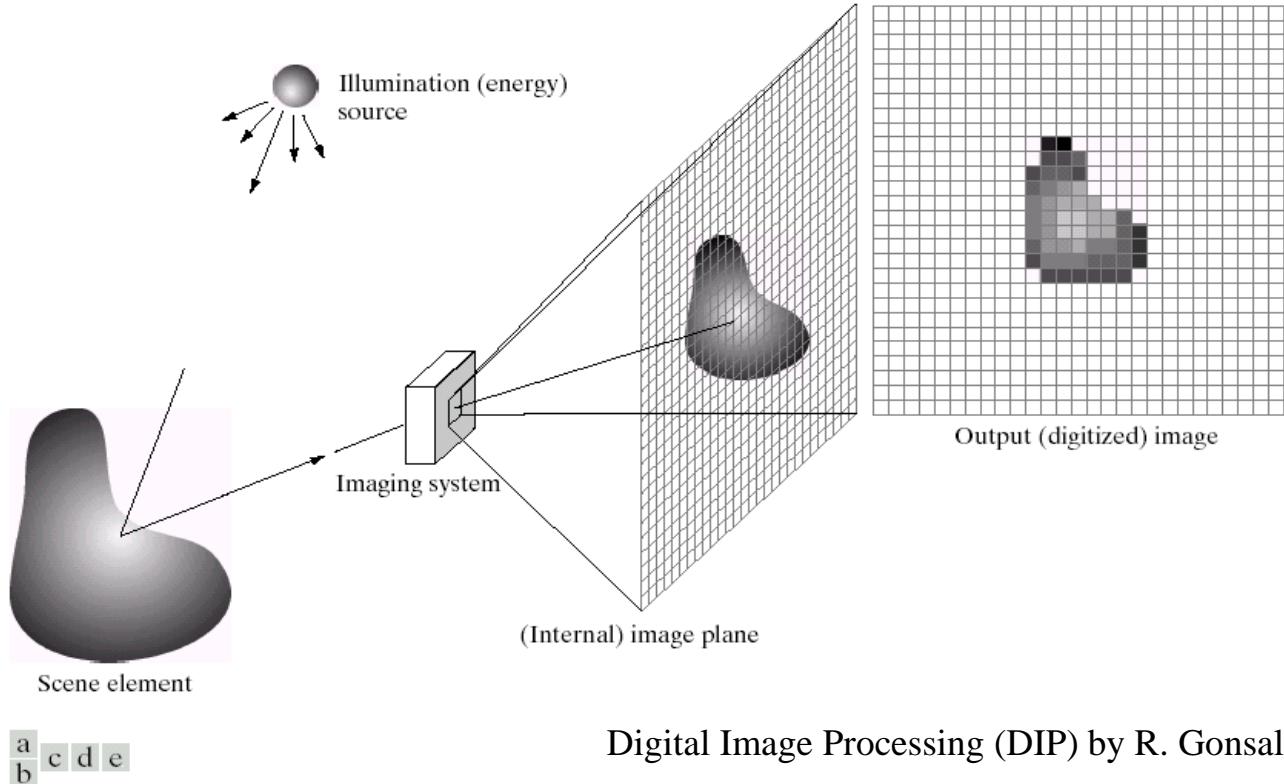




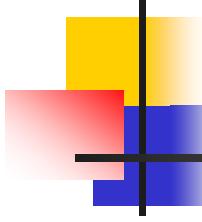
# Image Generation (1)

## ◆ Classical imaging system

- Reflected light from object surface is accumulated in the electronic sensor.
- The 3-D points are spatially sampled on the pixels.

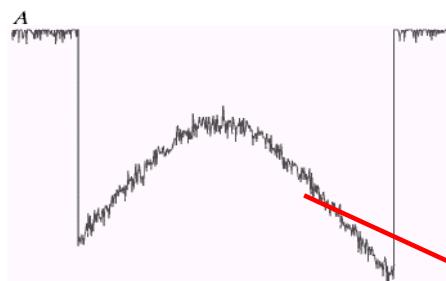
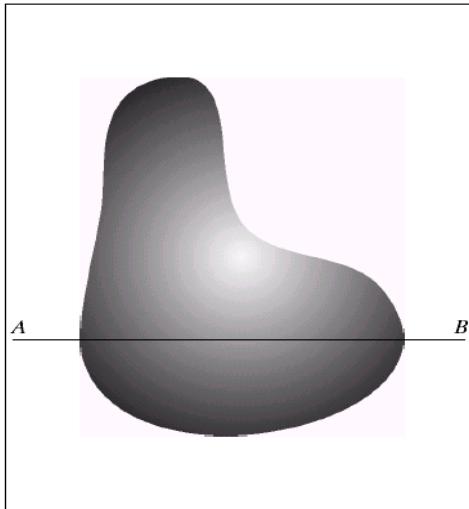


**FIGURE 2.15** An example of the digital image acquisition process. (a) Energy (“illumination”) source. (b) An element of a scene. (c) Imaging system. (d) Projection of the scene onto the image plane. (e) Digitized image.



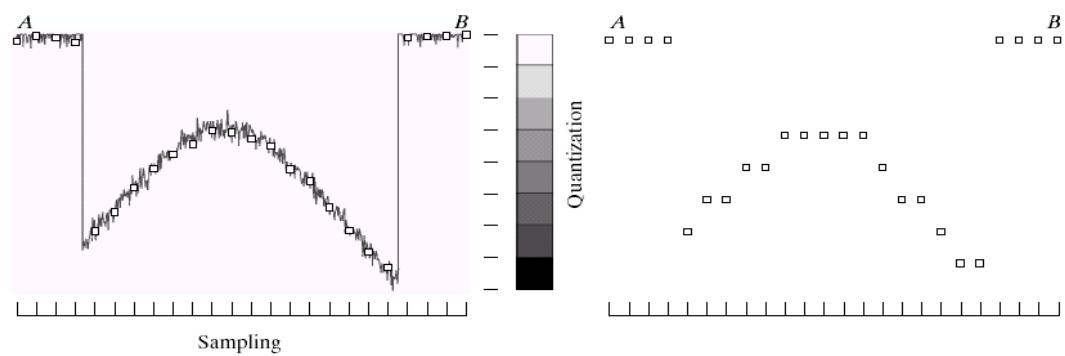
# Image Generation (2)

- ◆ Image signal is very noisy unlike the visual appearance.



Profile of a scan line

Random noise

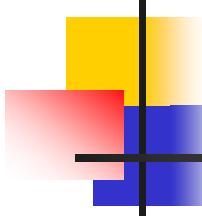


Profile on the pixel grid

DIP by R. Gonzalez

a  
b  
c  
d

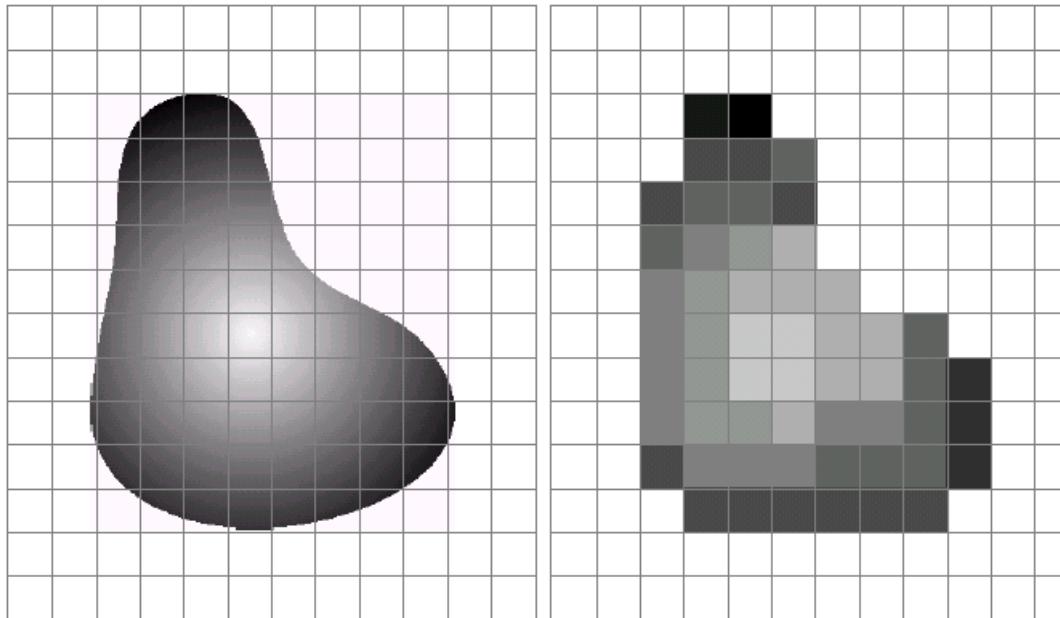
FIGURE 2.16 Generating a digital image. (a) Continuous image. (b) A scan line from A to B in the continuous image, used to illustrate the concepts of sampling and quantization. (c) Sampling and quantization. (d) Digital scan line.



# Image Generation (3)

## ◆ Image spatial sampling and quantization

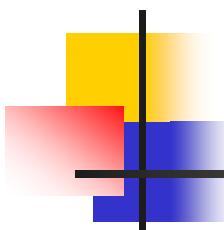
- By sampling on the sensor grid: loss of detailed visual information
- By quantization: loss of continuous values of original intensities



a b

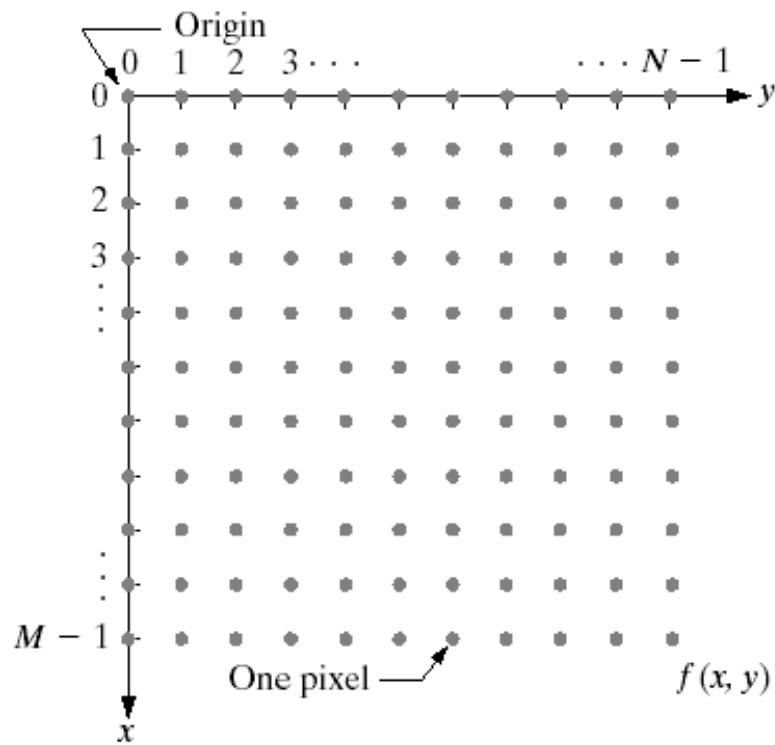
DIP by R. Gonzalez

**FIGURE 2.17** (a) Continuous image projected onto a sensor array. (b) Result of image sampling and quantization.



# Image Coordinates

- ◆ Image representation for digital processing: 2D array



DIP by R. Gonzalez

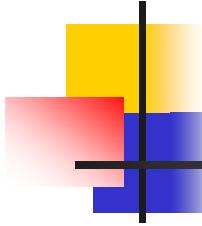
**FIGURE 2.18**

Coordinate convention used in this book to represent digital images.

$M \times N$  matrix or array

$$\mathbf{A} = \begin{bmatrix} a_{00} & a_{01} & \cdots & a_{0,N-1} \\ \vdots & \vdots & & \vdots \\ a_{M-1,0} & \cdots & \cdots & a_{M-1,N-1} \end{bmatrix}$$

$$a_{ij} = f(x=i, y=j)$$



# Image Resolution (# of Pixels)

- ◆ Image resolution is usually better when pixels are increased.



1024



512



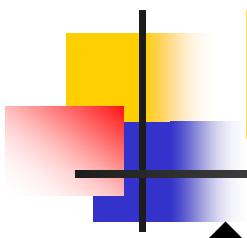
256



64

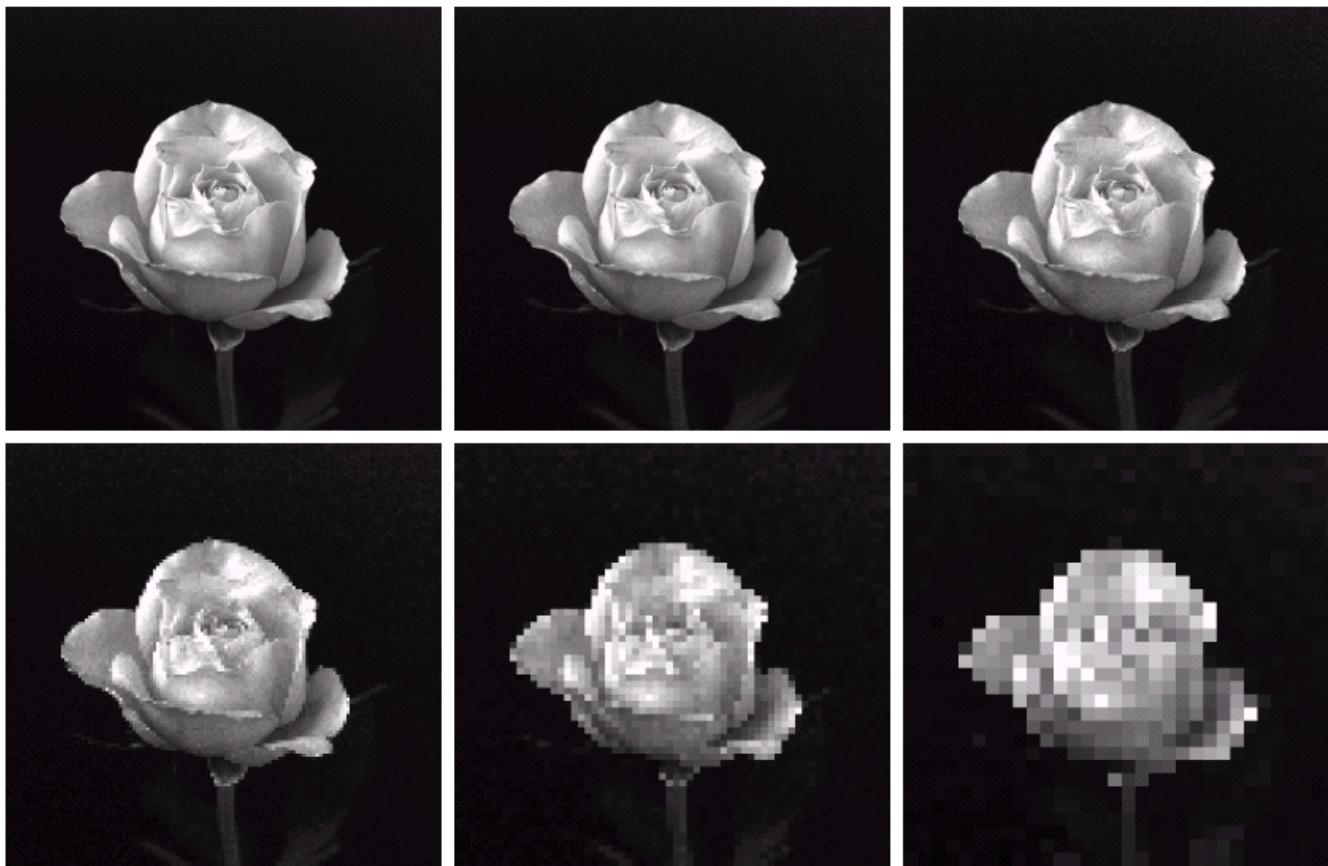
DIP by R. Gonzalez

**FIGURE 2.19** A  $1024 \times 1024$ , 8-bit image subsampled down to size  $32 \times 32$  pixels. The number of allowable gray levels was kept at 256.



# Image Resolution (Precision)

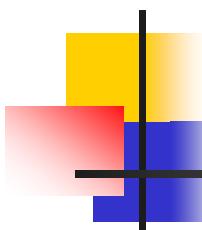
- ◆ Image resolution is better when spatial sampling is finer.



a b c  
d e f

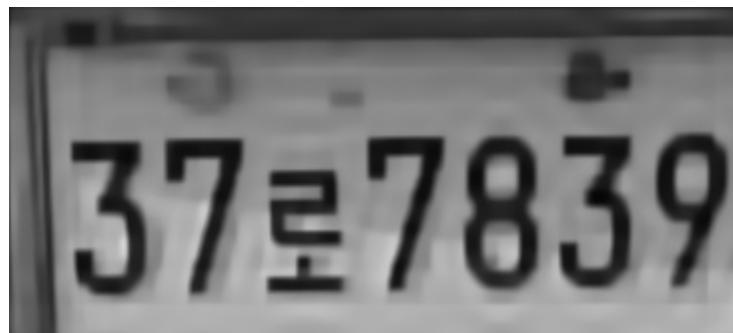
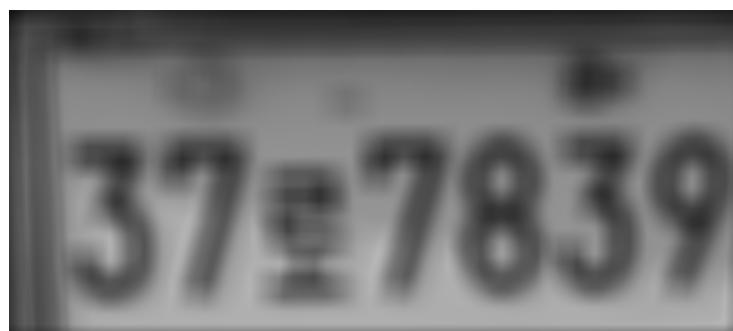
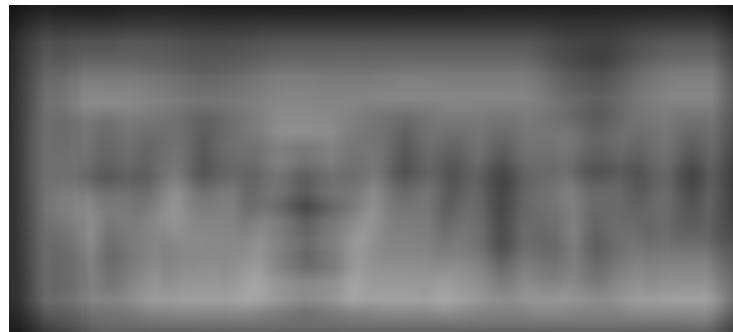
DIP by R. Gonzalez

**FIGURE 2.20** (a)  $1024 \times 1024$ , 8-bit image. (b)  $512 \times 512$  image resampled into  $1024 \times 1024$  pixels by row and column duplication. (c) through (f)  $256 \times 256$ ,  $128 \times 128$ ,  $64 \times 64$ , and  $32 \times 32$  images resampled into  $1024 \times 1024$  pixels.

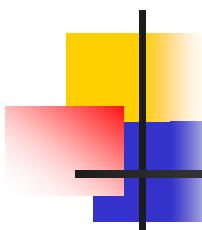


# Image Resolution (Blurring)

- ◆ Image resolution is better when blurring is less.

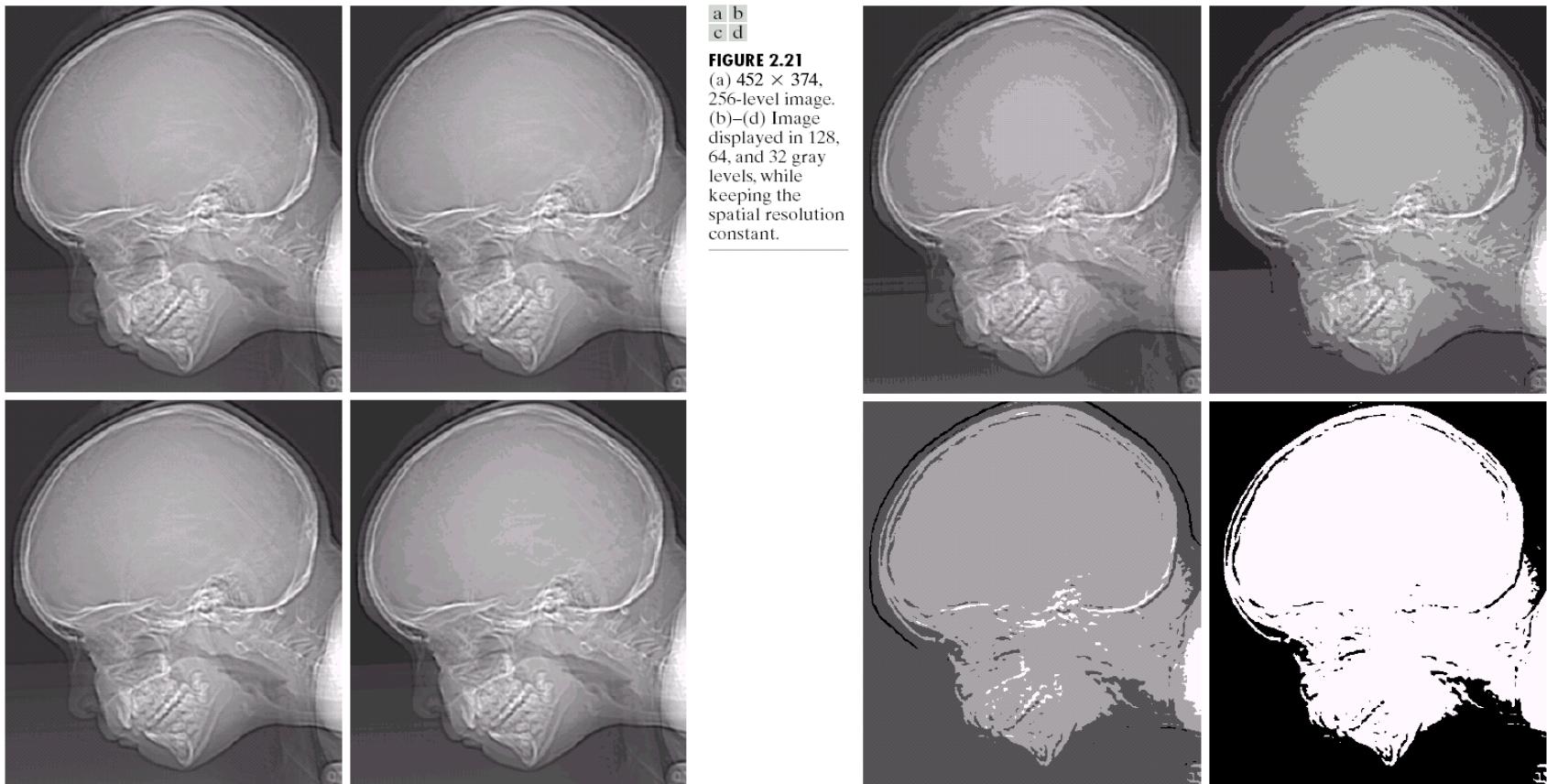


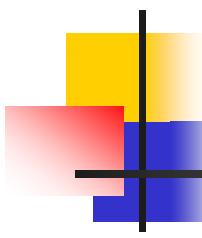
해상도 개선  
(deblurring)



# Image Levels (Quantization Levels)

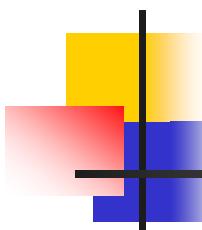
- ◆ The more intensity levels, the better the detailed contents.
- ◆ Quantization error – information loss – **random noise**





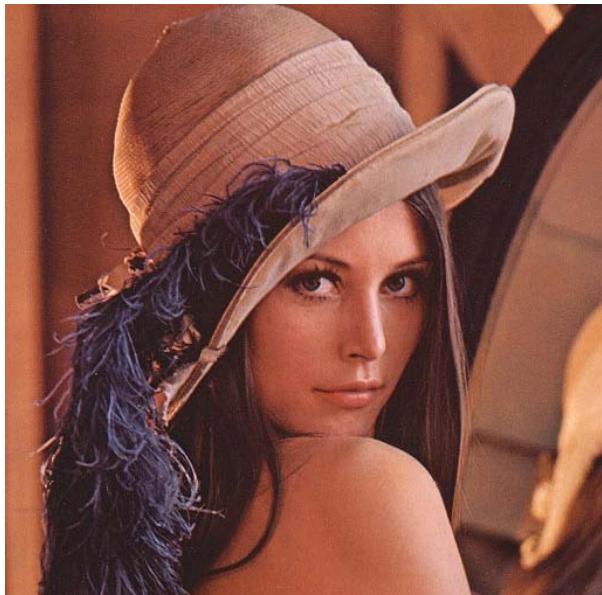
# How to Obtain Images?

- Functional forms
  - $f(x,y)$  : 2-D image
  - $f(x,y,t), f(x,y,z)$  : video sequence, 3D model
- Methods
  - Brightness(luminance) or color reflected from object surface
  - Absorption or Penetration characteristics of objects
    - X-ray imaging, Ultrasonic imaging, CT
  - Distance between objects and measuring instrument
    - sonar imaging, radar imaging, range camera, Lidar imaging
  - Temperature of an object
    - Thermal IR camera
  - Everything to generate 2-D signals: MRI, ....

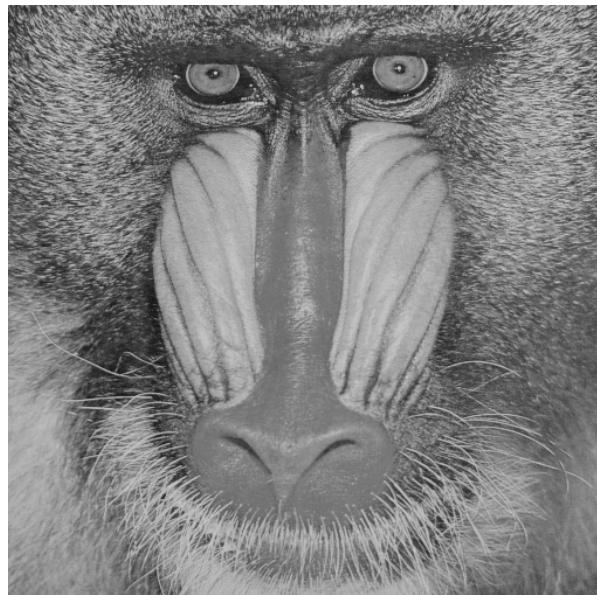


# Famous Still Images

- ◆ For image processing and image compression (JPEG)



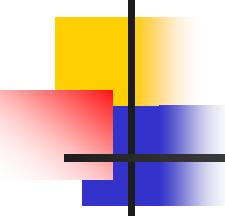
Lena



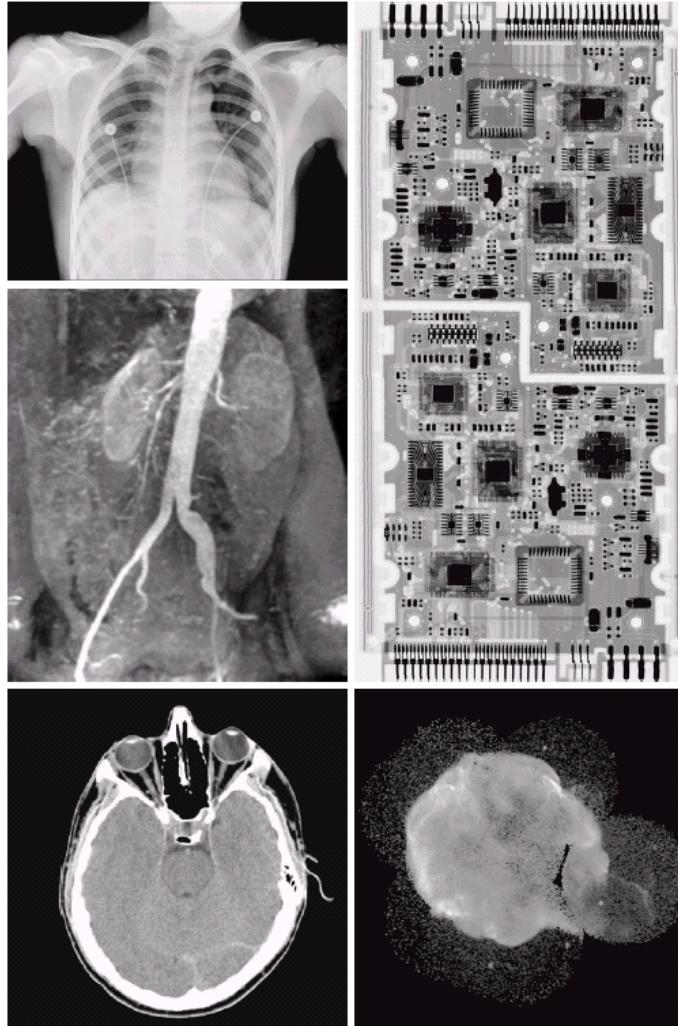
Barboon



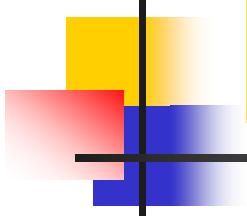
Boat



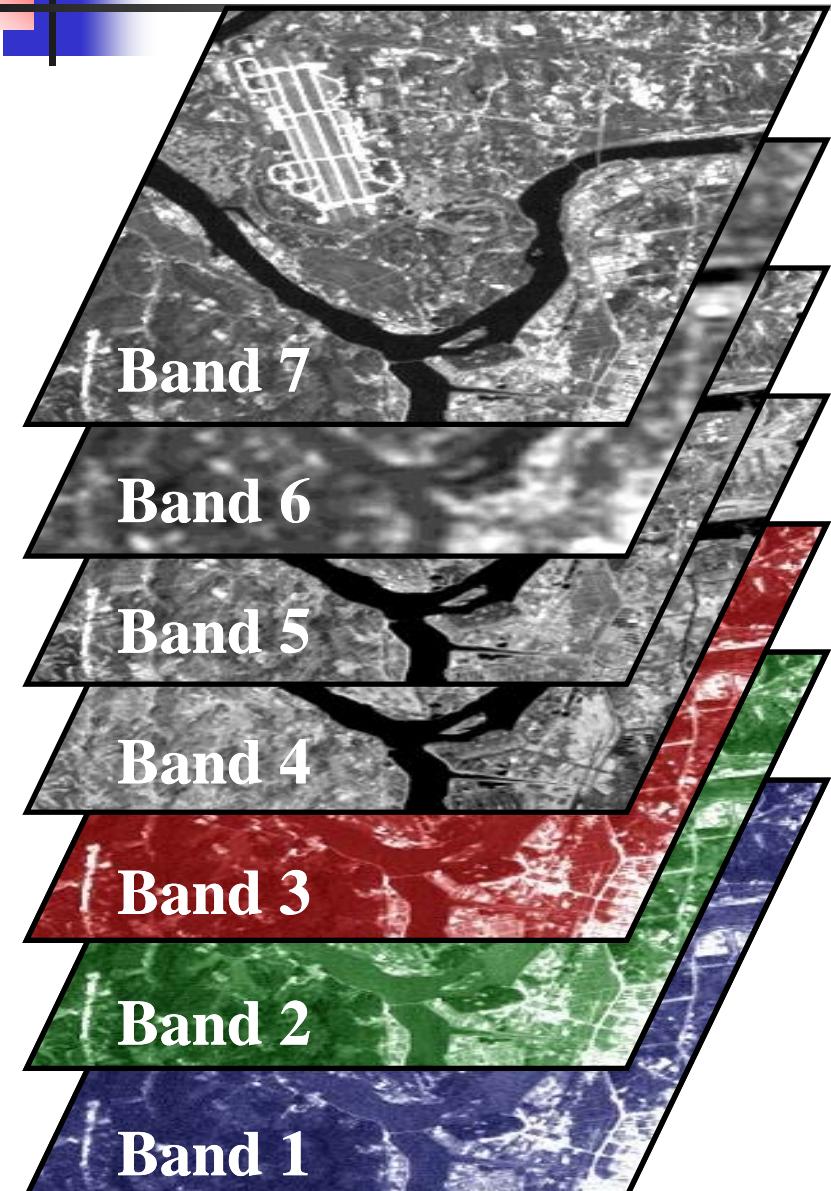
# X-ray Images



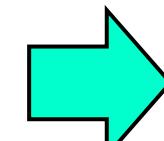
**FIGURE 1.7** Examples of X-ray imaging. (a) Chest X-ray. (b) Aortic angiogram. (c) Head CT. (d) Circuit boards. (e) Cygnus Loop. (Images courtesy of (a) and (c) Dr. David R. Pickens, Dept. of Radiology & Radiological Sciences, Vanderbilt University Medical Center, (b) Dr. Thomas R. Gest, Division of Anatomical Sciences, University of Michigan Medical School, (d) Mr. Joseph E. Pascente, Lixi, Inc., and (e) NASA.)



# Multi-spectral Images (1)



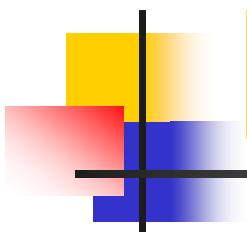
Band No.	Name	Wavelength ( $\mu\text{m}$ )	Characteristics and Uses
1	Visible blue	0.45–0.52	Maximum water penetration
2	Visible green	0.52–0.60	Good for measuring plant vigor
3	Visible red	0.63–0.69	Vegetation discrimination
4	Near infrared	0.76–0.90	Biomass and shoreline mapping
5	Middle infrared	1.55–1.75	Moisture content of soil and vegetation
6	Thermal infrared	10.4–12.5	Soil moisture; thermal mapping
7	Middle infrared	2.08–2.35	Mineral mapping



3,2,1  
fusion



Natural Color Fusion



# Multi-spectral Images (2)



Painting  
reconstruction

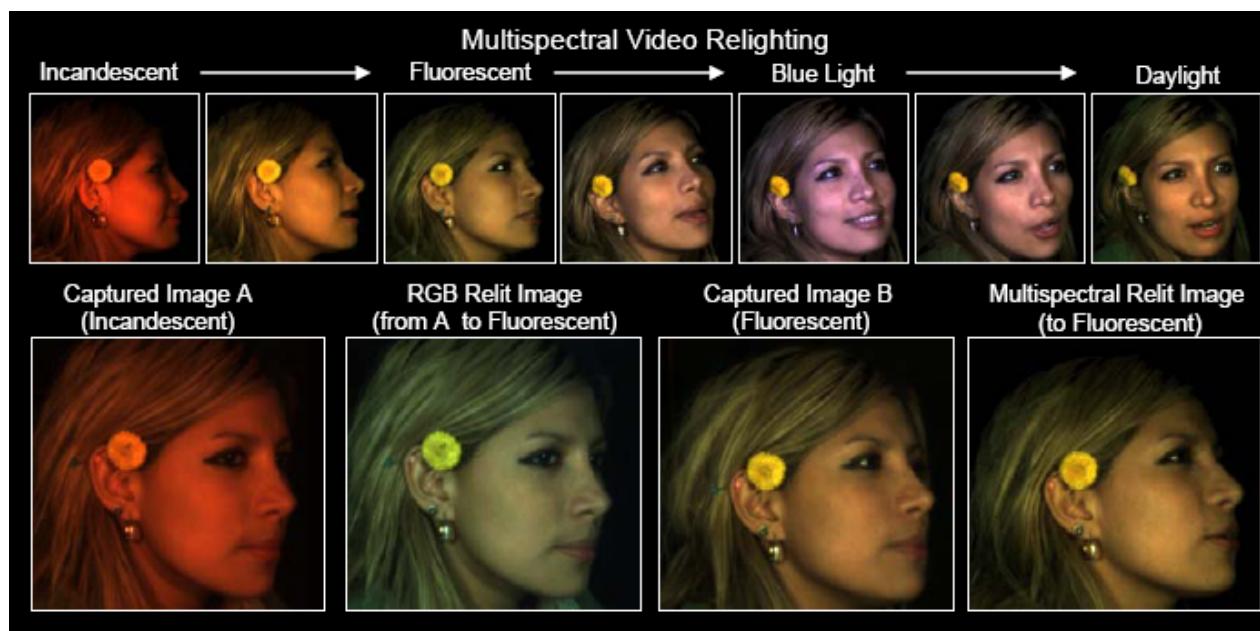
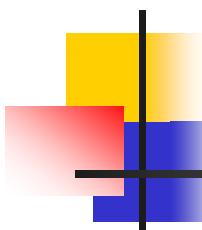
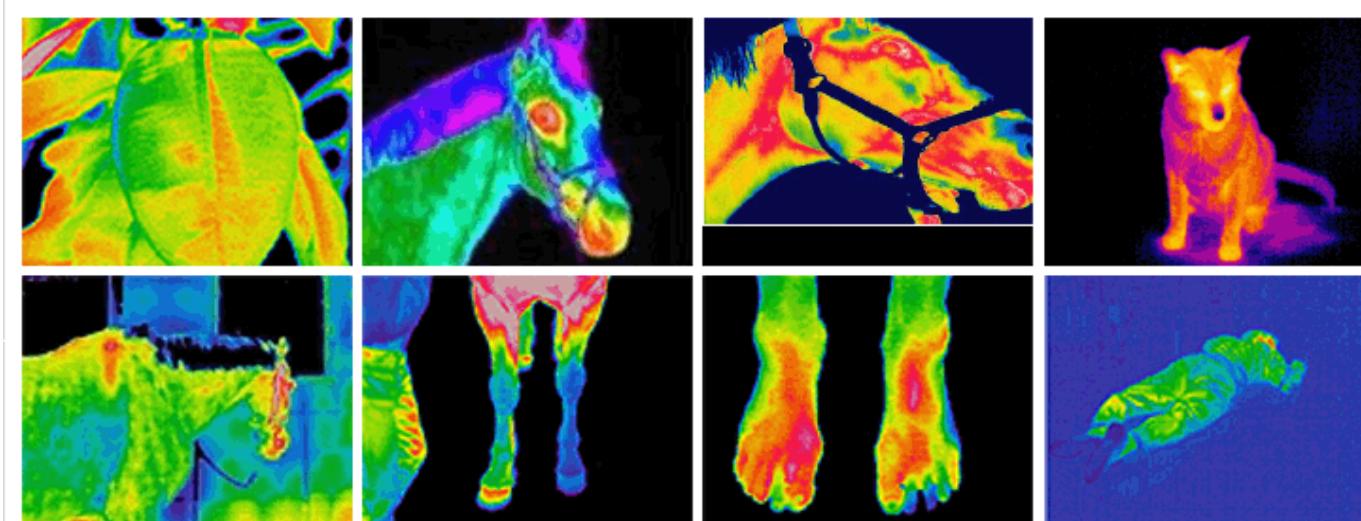


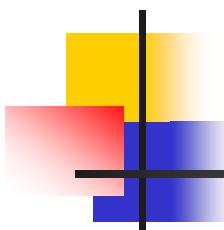
Image  
relighting



# Infrared Images (1)

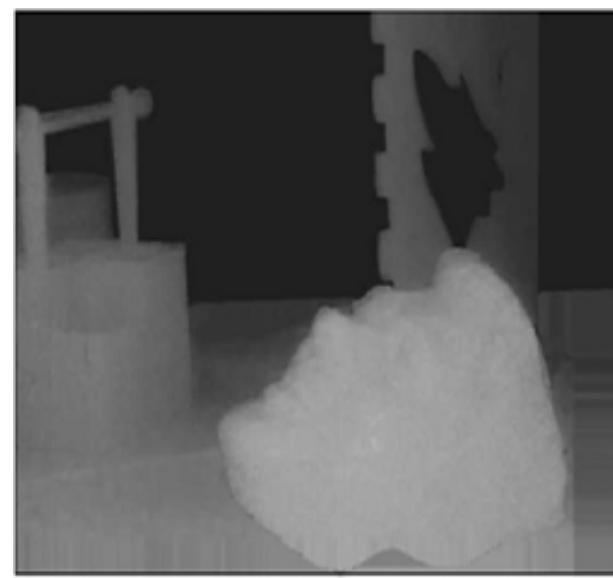
- ◆ Thermal IR: Temperature sensing



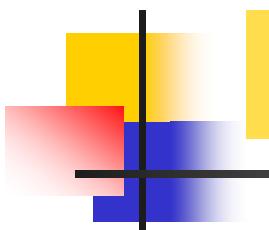


## Infrared Images (2)

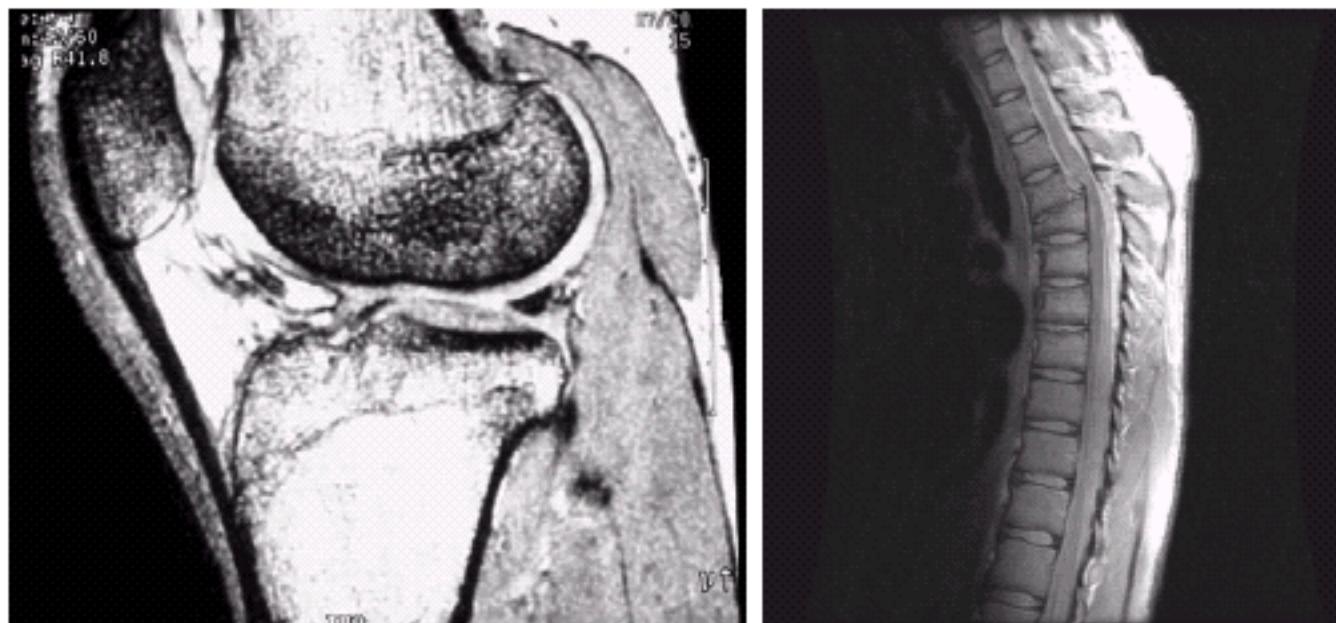
- ◆ Range image for 3D modeling and rendering



Kinect sensor for 3-D map

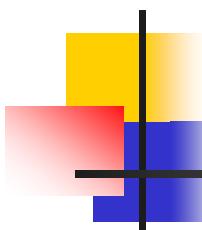


# MRI

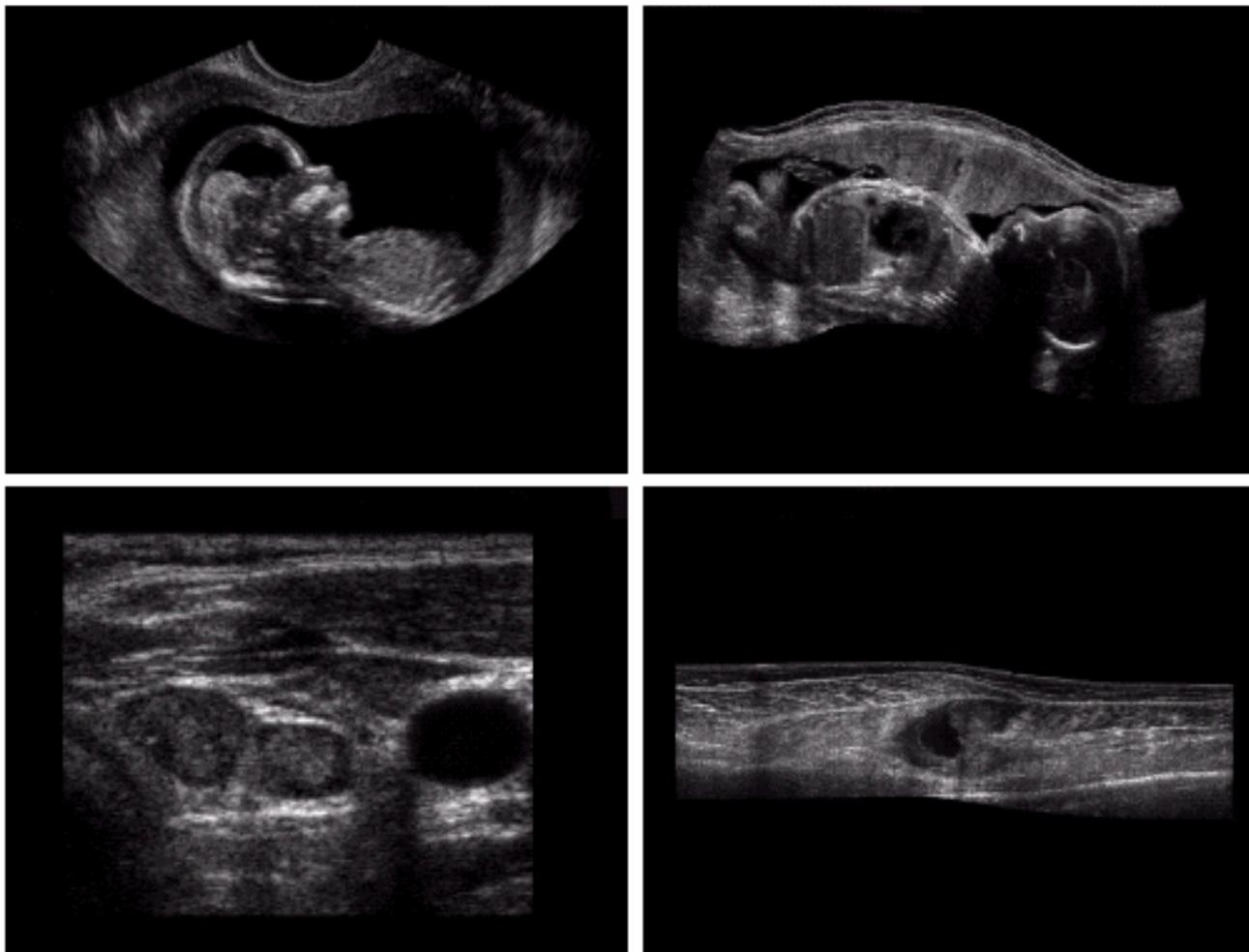


a b

**FIGURE 1.17** MRI images of a human (a) knee, and (b) spine. (Image (a) courtesy of Dr. Thomas R. Gest, Division of Anatomical Sciences, University of Michigan Medical School, and (b) Dr. David R. Pickens, Department of Radiology and Radiological Sciences, Vanderbilt University Medical Center.)



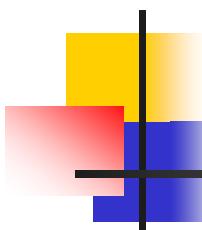
# Ultrasonic Images



a b  
c d

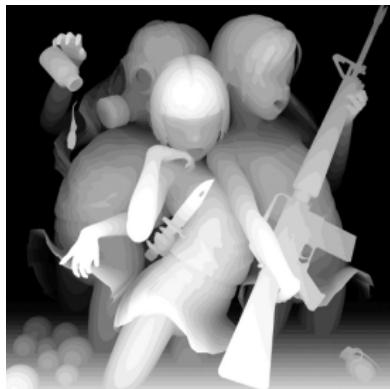
**FIGURE 1.20**  
Examples of ultrasound imaging. (a) Baby.  
(b) Another view of baby.  
(c) Thyroids.  
(d) Muscle layers showing lesion.  
(Courtesy of Siemens Medical Systems, Inc., Ultrasound Group.)

DIP by R. Gonzalez

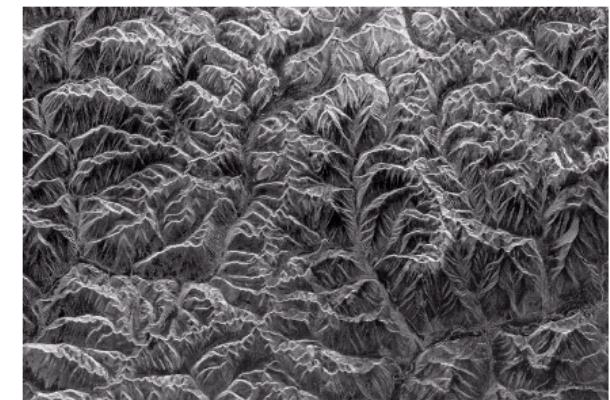


# Laser (range) Image

- Lidar (light detection and ranging)
  - Distance measure, 3D modeling



**FIGURE 1.16**  
Spaceborne radar  
image of  
mountains in  
southeast Tibet.  
(Courtesy of  
NASA.)



Depth map



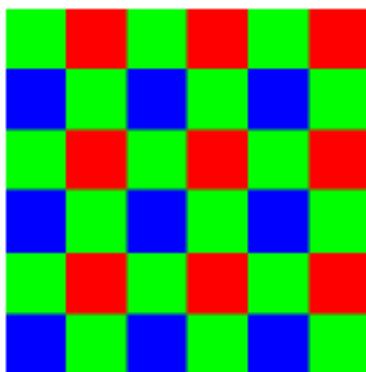
Stereoscopic system



3D model+ image texture mapping

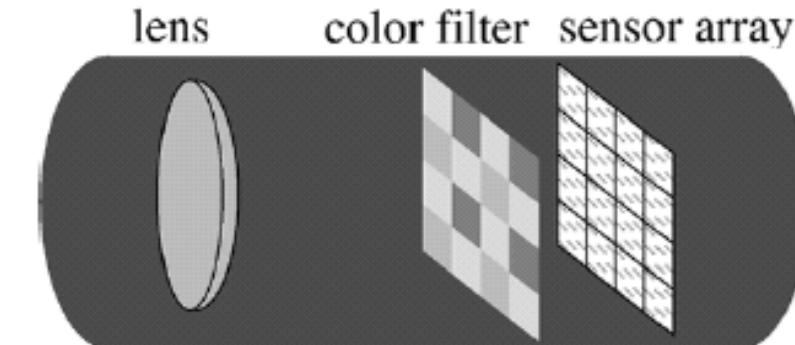
# Digital Color Image Generation (1)

- Bayar CFA (color filter array) pattern
  - Each pixel takes only one color component via mosaicked CFA pattern.
  - Green is sampled at a higher rate than red and blue.
  - Resolution reduction.



Bayar CFA pattern

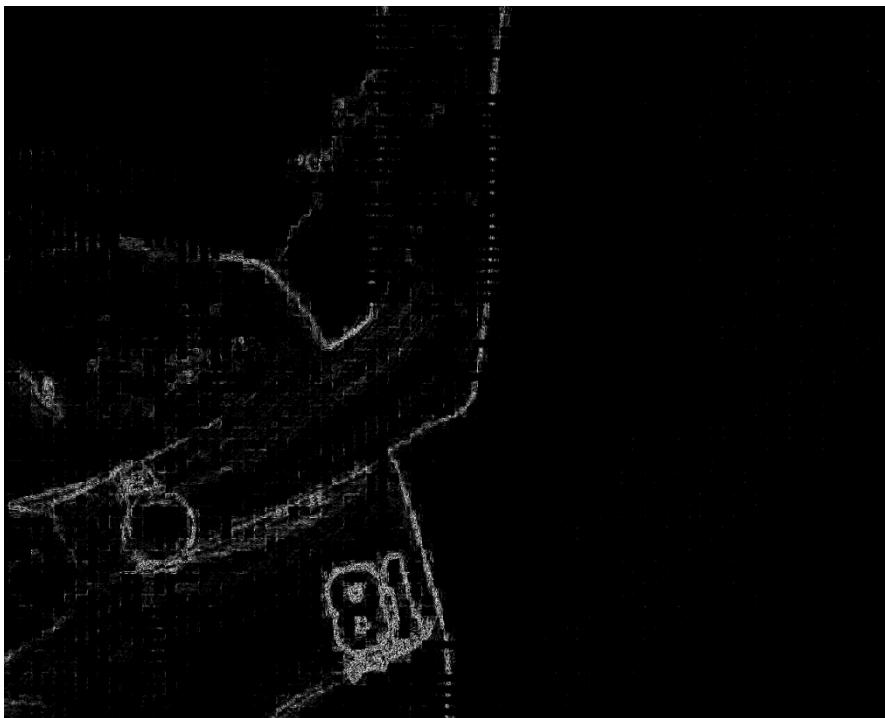
G	R	G	R	G	R
B	G	B	G	B	G
G	R	G	R	G	R
B	G	B	G	B	G
G	R	G	R	G	R
B	G	B	G	B	G



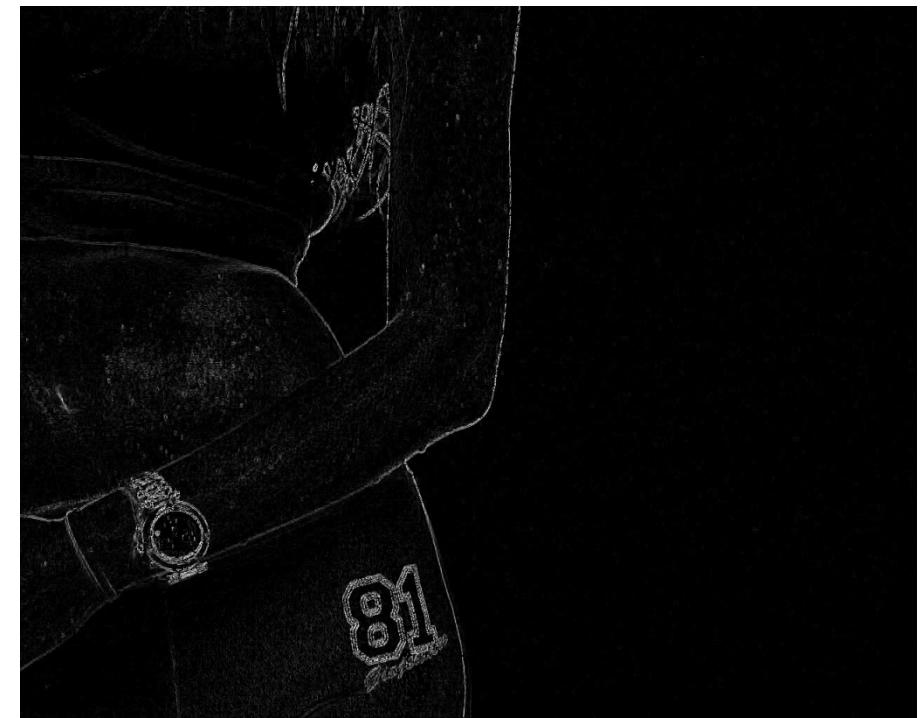
Color image acquisition

# Digital Color Image Generation (2)

- Resolution reduction by CFA
  - Comparison of edge maps



Edge map of Cb-Cr (b-R) components



Edge map of Y (green) components

# Demosaic

- For full resolution color images
  - Interpolation is required for each color channel.
  - Image Demosaic
- Interpolating each color component, (R,G,B)
  - Independent interpolation
  - Using correlations between colors.
- Artifacts generated from
  - Image contents (high frequency)
  - Interpolation algorithms
  - CFA pattern: zippering

# Demosaic Results

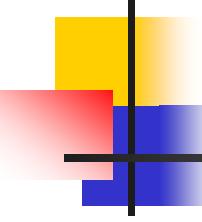
- An exemplary artifact



Horizontal interpolation

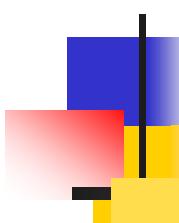
Vertical interpolation

Directional interpolation



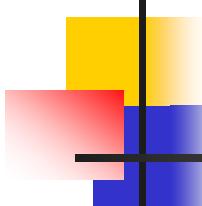
# Image Processing for Acquisition

- Spatial sampling by pixel grid
- Quantization of intensities
- Noise reduction
- Demosaic
- Contrast enhancement
- Color enhancement
- Radiometric compensation
  - White balance, anti-vignette
- Lens distortion compensation
- Compression



# Color Space

Sang Hwa Lee



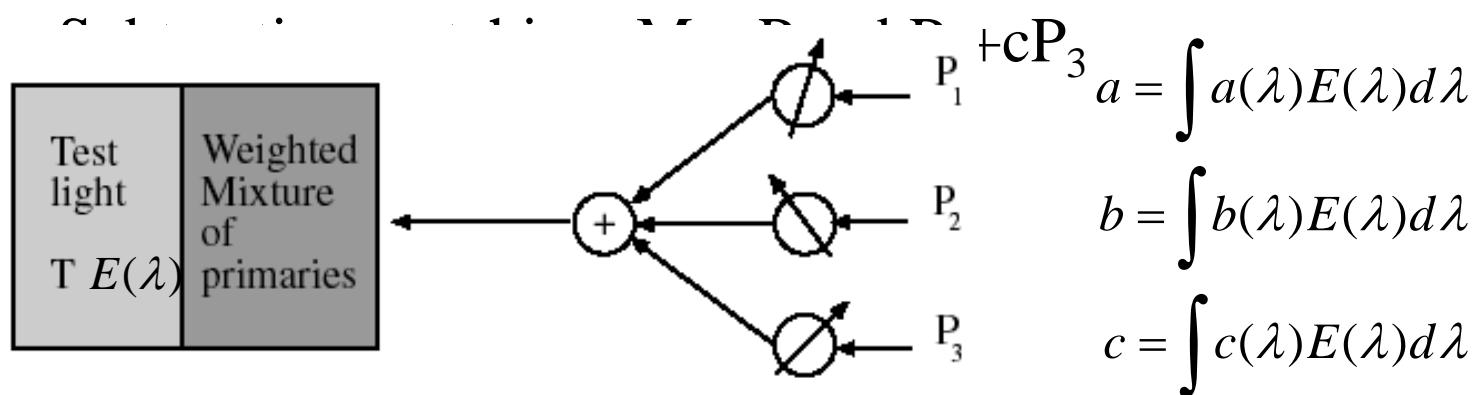
# Color Representation

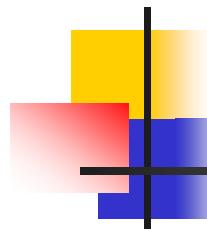
- 3 primaries for 3-D color space

- Reproduce natural colors.
  - Representation in display devices

- Color matching

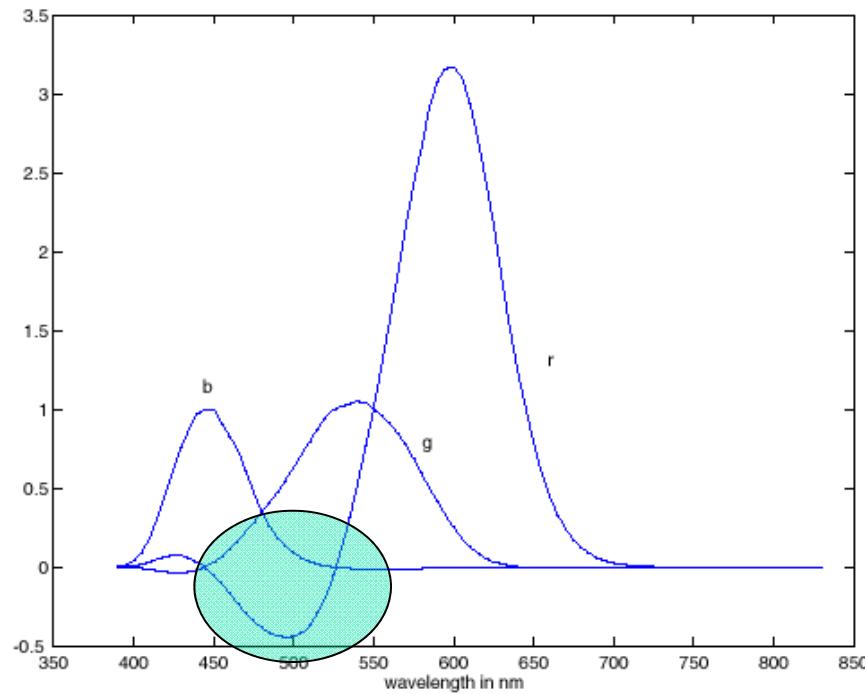
- Linear combination of 3 primaries
  - Additive matching:  $M = aP_1 + bP_2 + cP_3$ 
    - Non-negative bases



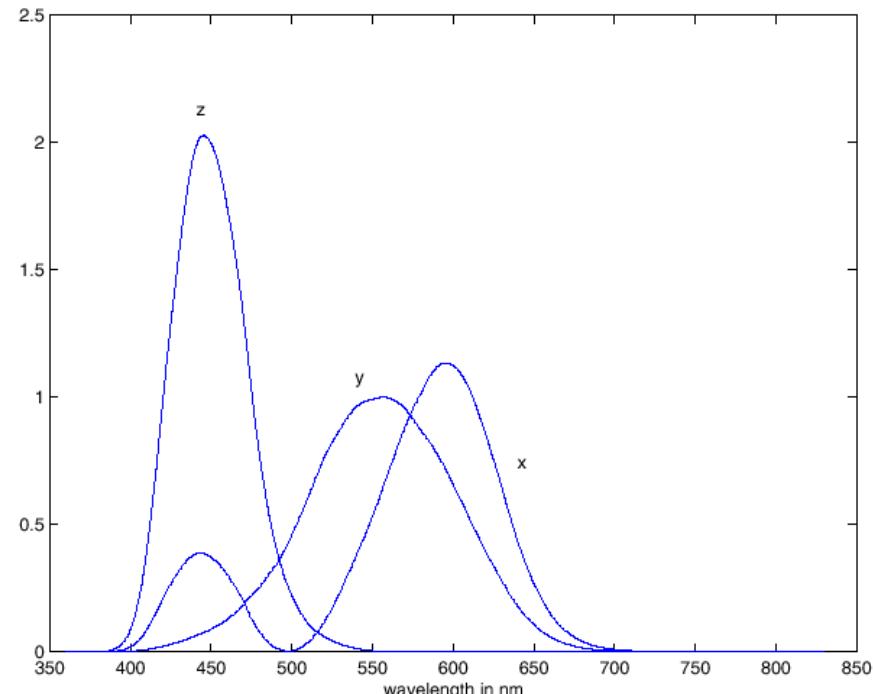


# CIE XYZ Color Space (1)

RGB Color matching function



CIE XYZ matching function



Negative(subtractive)  
- not representable in display

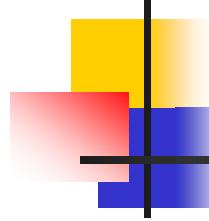
$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} 0.490 & 0.310 & 0.200 \\ 0.177 & 0.813 & 0.011 \\ 0.000 & 0.010 & 0.990 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

x=X/(X+Y+Z)  
y=Y/(X+Y+Z)

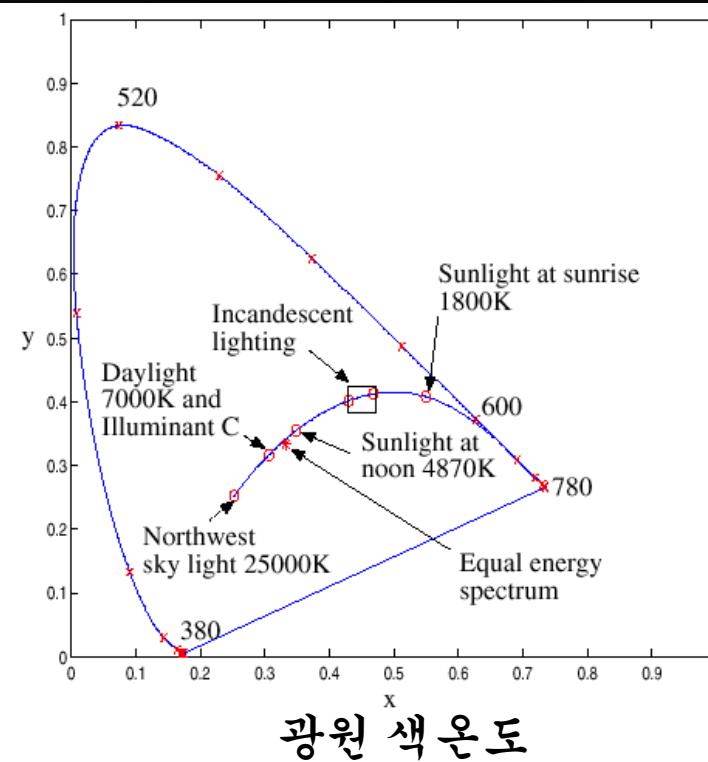
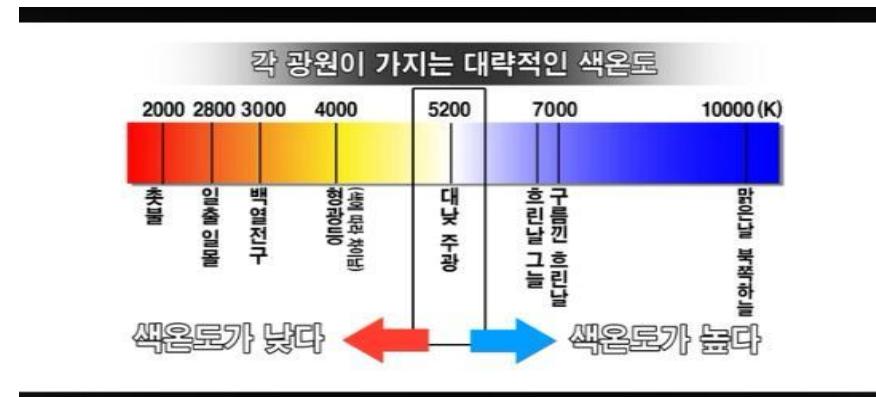
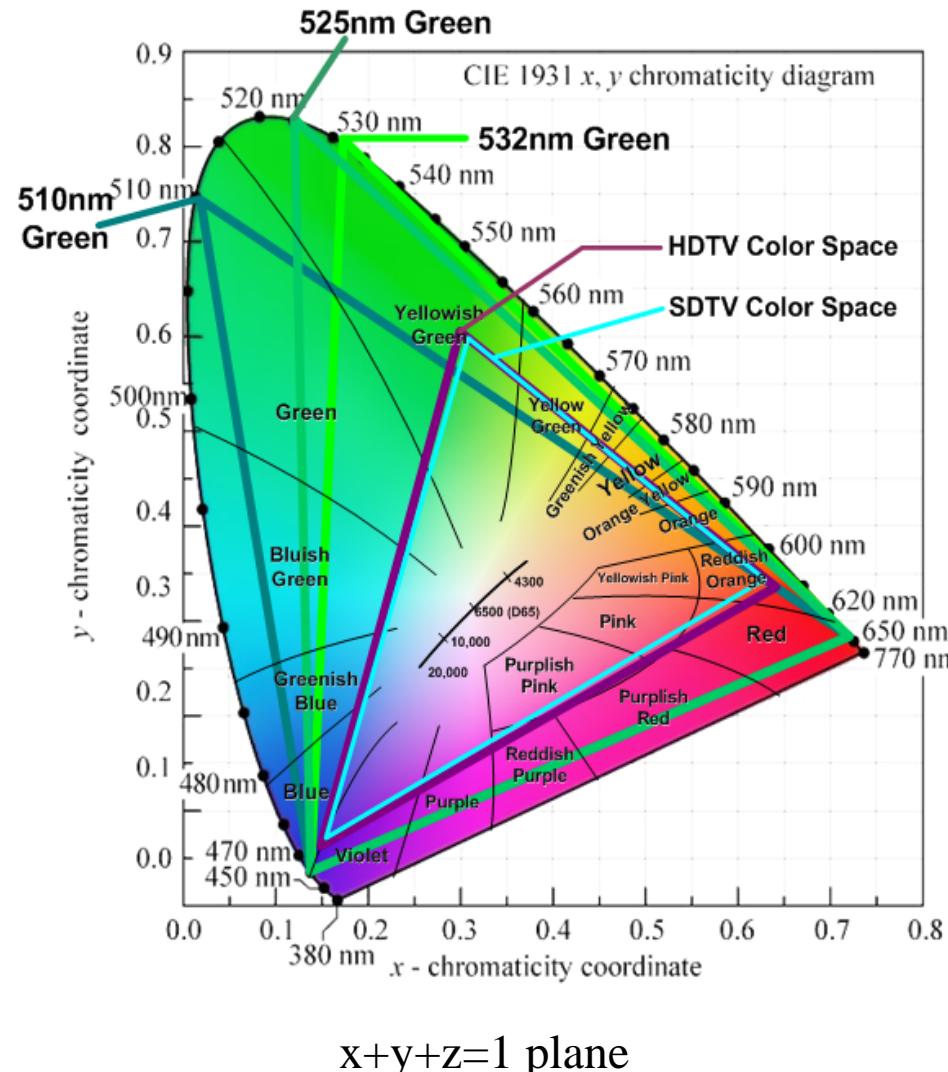
$$X = 0.490R + 0.310G + 0.200B$$

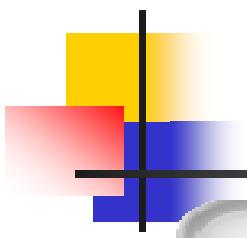
$$Y = 0.177R + 0.813G + 0.011B$$

$$Z = 0.000R + 0.010G + 0.990B$$



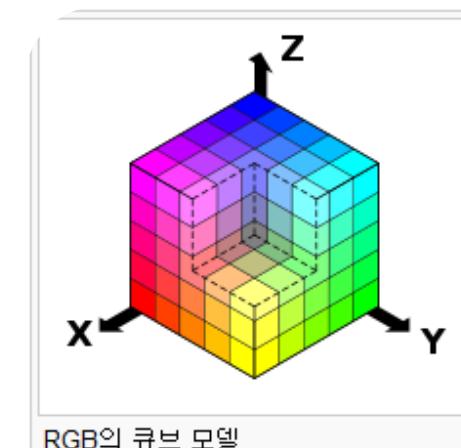
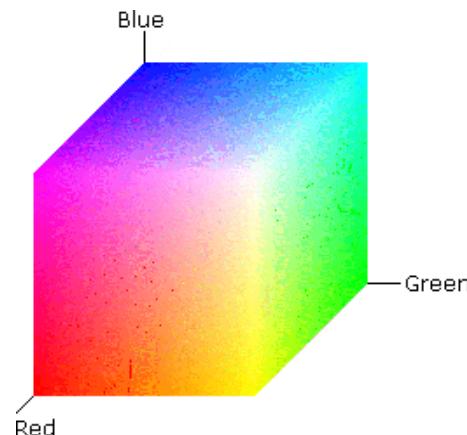
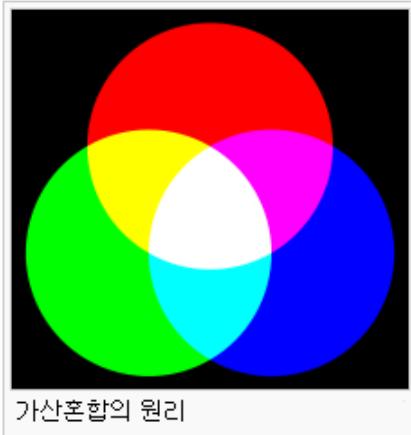
## CIE XYZ Color Space (2)

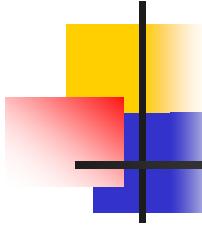




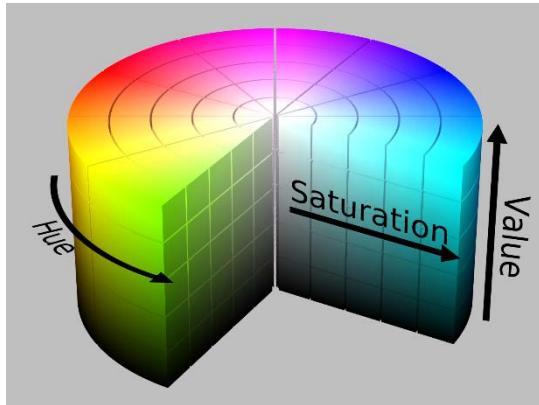
# RGB Color Space

- 3 Primaries of colors
  - Red (645.2nm), Green (526.3nm), Blue (444.4nm)
- 3-D orthogonal rectangular coordinate
  - RGB cube
  - Actually neither independent nor orthogonal to each other
- The more is mixed, the brighter is.
  - Additive mixing
- Variations: sRGB, adobeRGB, ...

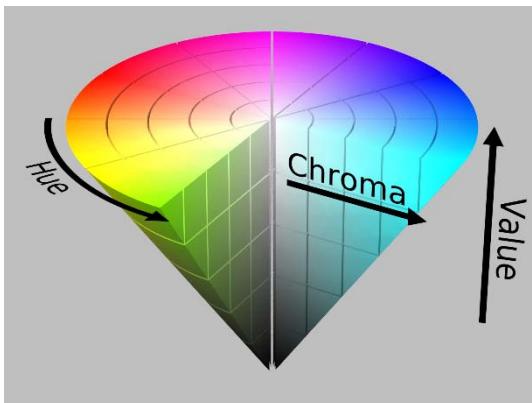




# HSV Color Space (1)

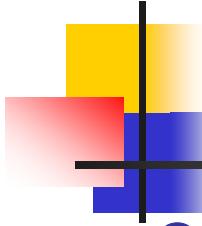


Cylinder HSV



Cone HSV

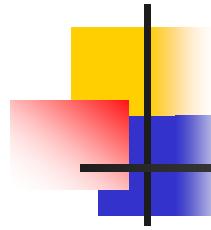
- Cylindrical representation of colors
  - Nonlinear color representation
  - Independent components: (H,S,V)
- Hue
  - Visible light spectrum
  - Red ~ violet mapping to  $0^\circ \sim 360^\circ$
- Saturation
  - Purity of color
  - the radius of cylinder: 0(achromatic)~1(chromatic)
- Value (Intensity)
  - Luminance or brightness
  - the height of cylinder
- Cone Representation
  - Low values -> Hue components are not distinctive.



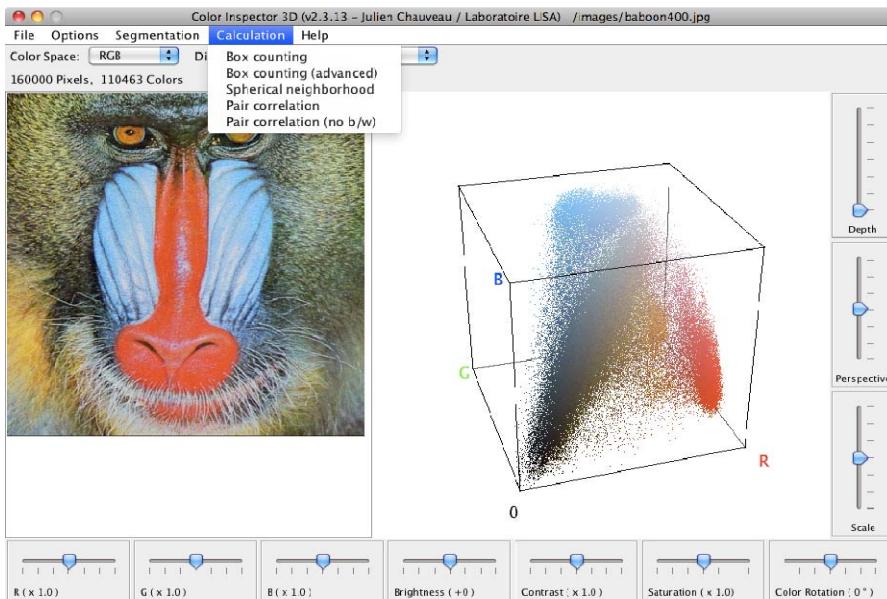
# HSV Color Space (2)

## ● Transform of RGB to HSV

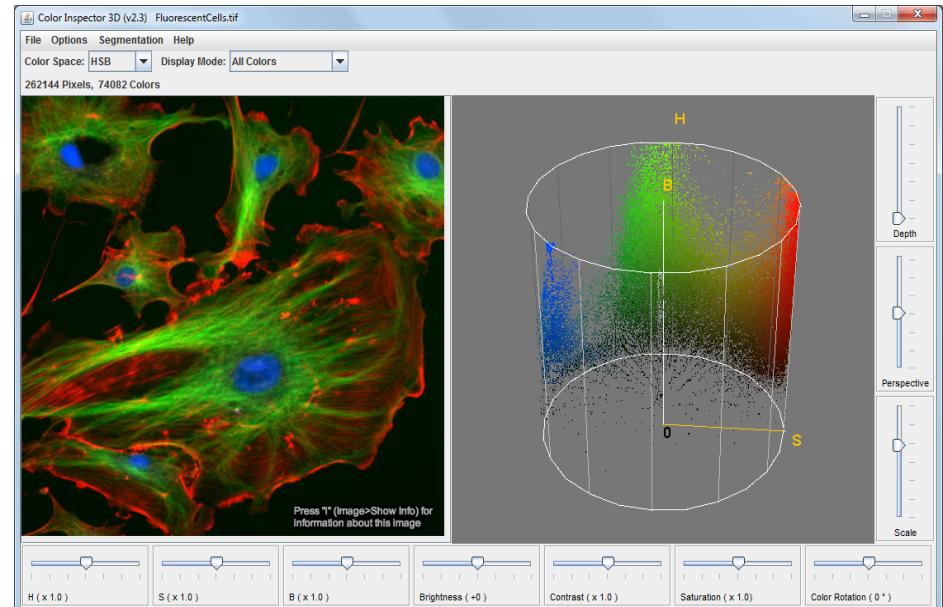
```
RGBtoHSV( float r, float g, float b, float *h, float *s, float *v )
{
    float min, max, delta;
    min = MIN( r, g, b );
    max = MAX( r, g, b );
    *v = max;                                // v
    delta = max - min;
    if( max != 0 )
        *s = delta / max;                  // s
    else {
        // r = g = b = 0                  // s = 0, v is undefined
        *s = 0;
        *h = -1;
        return;
    }
    if( r == max )
        *h = ( g - b ) / delta;          // between yellow & magenta
    else if( g == max )
        *h = 2 + ( b - r ) / delta;      // between cyan & yellow
    else
        *h = 4 + ( r - g ) / delta;      // between magenta & cyan
    *h *= 60;                            // degrees
    if( *h < 0 )
        *h += 360;
}
```



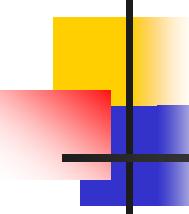
# RGB and HSV



RGB 분포

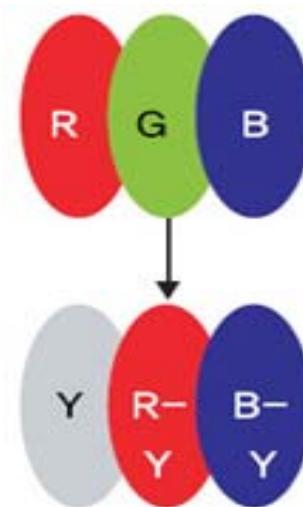


HSV 분포

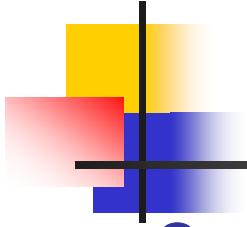


# YUV Color Space (1)

- Rectangular representation of colors
  - Color TV signal transmission (composite color)
- Reducing the redundancy (correlation) among color components
  - Y: Luminance, brightness
  - U, V : Chrominance (color difference)
- Human visual system (HVS) is less sensitive to chrominance than luminance or RGB
  - Spectral reduction : 4:2:0 format encoding
  - Image and video compression: JPEG, MPEG
- Variations: YCbCr, YIQ (NTSC)



Correlation  
reduction



# YUV Color Space (2)

## ● RGB to YUV

$$\begin{bmatrix} Y' \\ U \\ V \end{bmatrix} = \begin{bmatrix} 0.299 & 0.587 & 0.114 \\ -0.14713 & -0.28886 & 0.436 \\ 0.615 & -0.51499 & -0.10001 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

$$\begin{bmatrix} R \\ G \\ B \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1.13983 \\ 1 & -0.39465 & -0.58060 \\ 1 & 2.03211 & 0 \end{bmatrix} \begin{bmatrix} Y' \\ U \\ V \end{bmatrix}$$



## ● RGB to YCbCr

$$Y = 0.299 \times R + 0.587 \times G + 0.114 \times B + 0$$

$$C_b = -0.169 \times R - 0.331 \times G + 0.499 \times B + 128$$

$$C_r = 0.499 \times R - 0.418 \times G - 0.0813 \times B + 128$$

$$R = \text{clamp}(Y + 1.402 \times (Cr - 128))$$

$$G = \text{clamp}(Y - 0.344 \times (Cb - 128) - 0.714 \times (Cr - 128))$$

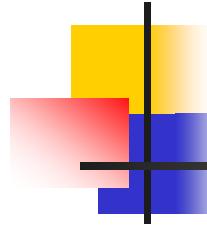
$$B = \text{clamp}(Y + 1.772 \times (Cb - 128))$$



	R	G	B
Black	0	0	0
White	255	255	255
Yellow	255	255	0
Cyan	0	255	255
Green	0	255	0
Magenta	255	0	255
Red	255	0	0
Blue	0	0	255



Y	U	V
16	128	128
235	128	128
210	16	146
170	166	16
145	54	34
107	202	222
82	90	240
41	240	110



# CIE 1976 Lab Color Space

❖ Wider representable, more uniform in color space

- L: luminance (0~100)
- a: green ~ red (-1~+1)
- b: blue ~ yellow (-1~+1)

❖ Non-linear transform

$$L^* = 116f(Y/Y_n) - 16$$

$$a^* = 500[f(X/X_n) - f(Y/Y_n)]$$

$$b^* = 200[f(Y/Y_n) - f(Z/Z_n)]$$

$$f(t) = \begin{cases} t^{1/3} & \text{if } t > (\frac{6}{29})^3 \\ \frac{1}{3} \left(\frac{29}{6}\right)^2 t + \frac{4}{29} & \text{otherwise} \end{cases}$$

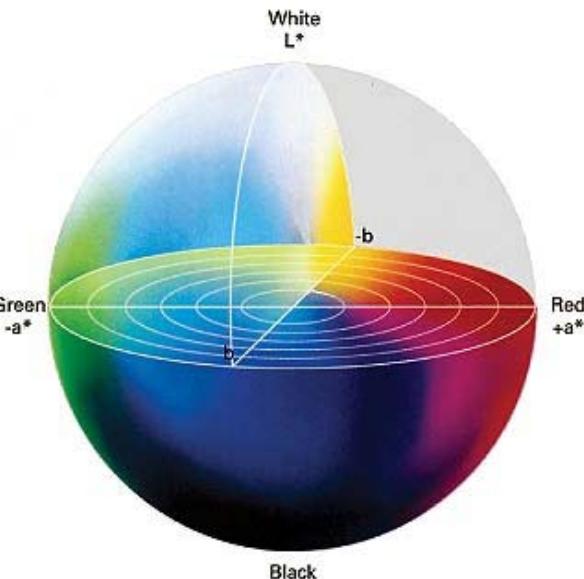
$$Y = Y_n f^{-1} \left( \frac{1}{116} (L^* + 16) \right)$$

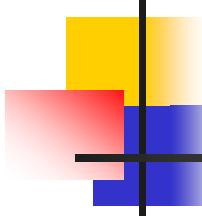
$$X = X_n f^{-1} \left( \frac{1}{116} (L^* + 16) + \frac{1}{500} a^* \right)$$

$$Z = Z_n f^{-1} \left( \frac{1}{116} (L^* + 16) - \frac{1}{200} b^* \right)$$

$$f^{-1}(t) = \begin{cases} t^3 & \text{if } t > \frac{6}{29} \\ 3 \left(\frac{6}{29}\right)^2 \left(t - \frac{4}{29}\right) & \text{otherwise} \end{cases}$$

\* $X_n, Y_n, Z_n$ : normalized XYZ for standard white color



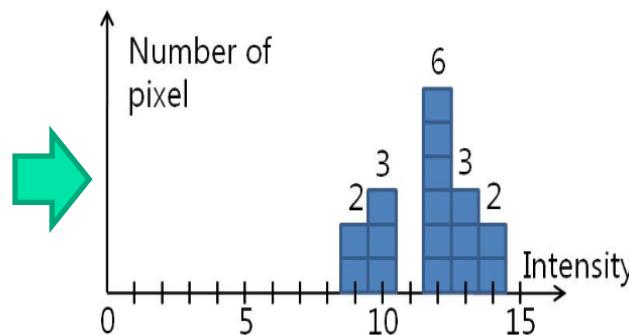


# Histogram (1)

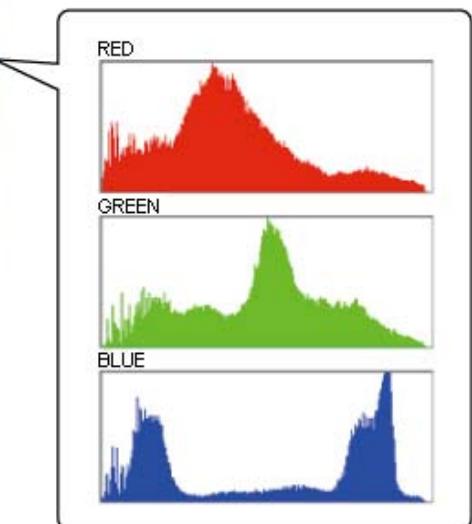
- Histogram: Number of pixels to have an intensity
  - Statistical information: probability function
  - Histogram bin: width of intensities or colors

<Histogram example>

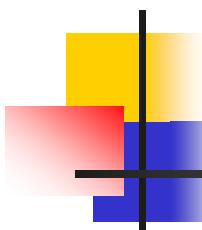
9	13	12	10
12	10	13	12
14	12	12	9
10	13	14	12



Gray-scale

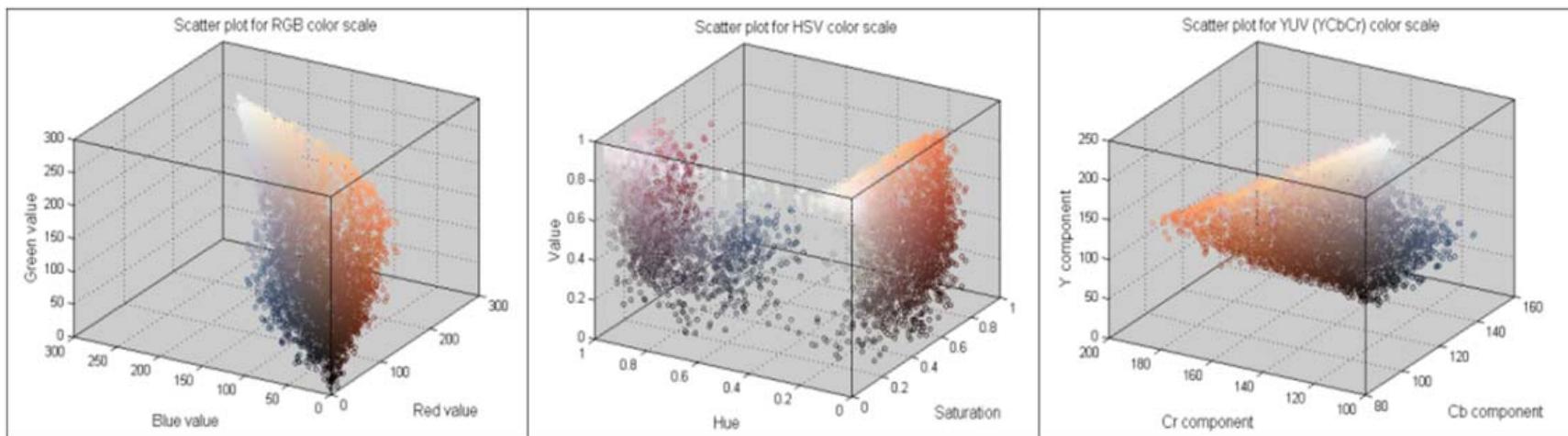


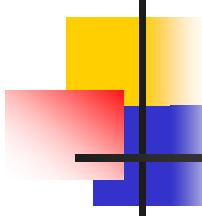
RGB color



# Histogram (2)

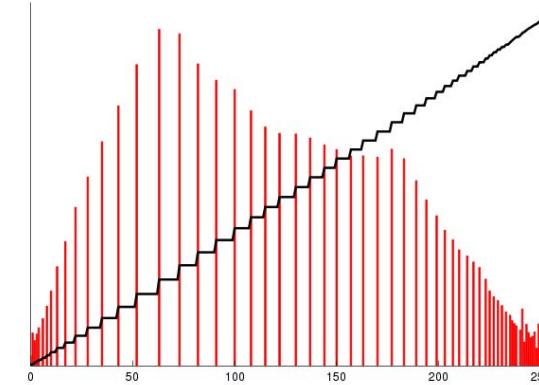
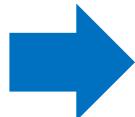
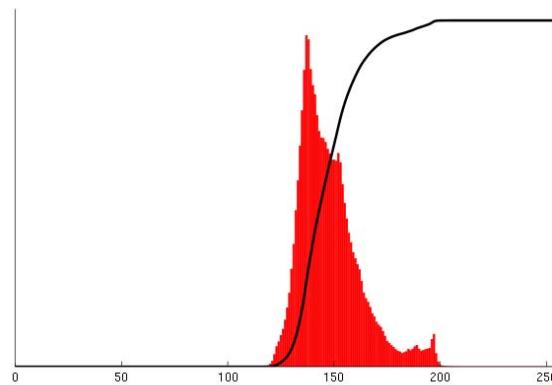
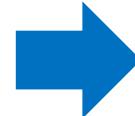
- Histogram of skin colors
  - Face/hand detection
  - RGB  $\Rightarrow$  HSV, YUV





# Histogram Equalization

- Automatic statistical and non-linear modification of intensity distribution



Intensity histogram (red) and  
cumulative histogram(black curve)

Equalized intensity histogram (red) and  
uniform cumulative histogram(black curve)