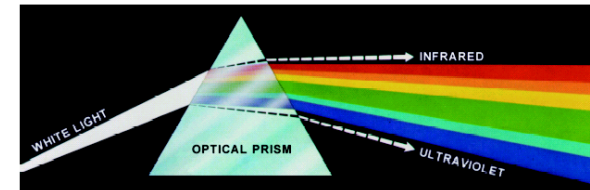


# Digital Image Processing

## Chapter 6:

### Color Image Processing

#### Spectrum of White Light

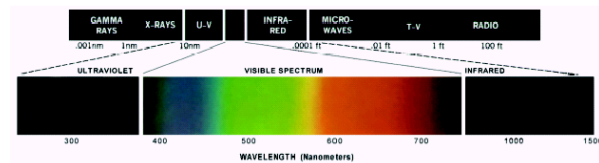


**FIGURE 6.1** Color spectrum seen by passing white light through a prism. (Courtesy of the General Electric Co., Lamp Business Division.)

1666 Sir Isaac Newton, 24 year old, discovered white light spectrum.

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

#### Electromagnetic Spectrum

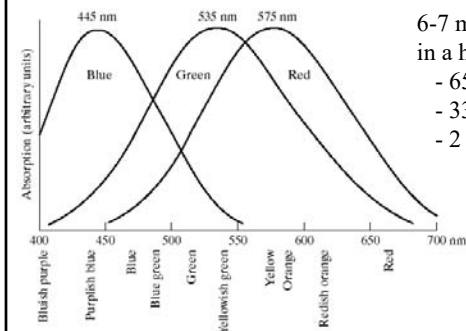


Visible light wavelength: from around 400 to 700 nm

- For an achromatic (monochrome) light source, there is only 1 attribute to describe the quality: **intensity**
- For a chromatic light source, there are 3 attributes to describe the quality:
  - Radiance** = total amount of energy flow from a light source (Watts)
  - Luminance** = amount of energy received by an observer (lumens)
  - Brightness** = intensity

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

#### Sensitivity of Cones in the Human Eye



6-7 millions cones  
in a human eye

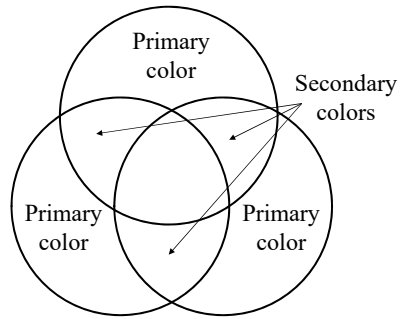
- 65% sensitive to **Red light**
- 33% sensitive to **Green light**
- 2 % sensitive to **Blue light**

**Primary colors:**  
Defined CIE in 1931  
**Red** = 700 nm  
**Green** = 546.1 nm  
**Blue** = 435.8 nm

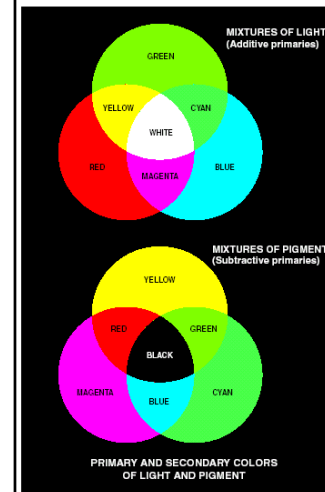
CIE = Commission Internationale de l'Eclairage  
(The International Commission on Illumination)

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Primary and Secondary Colors



### Primary and Secondary Colors (cont.)



Additive primary colors: RGB  
use in the case of light sources  
such as color monitors

RGB add together to get white

Subtractive primary colors: CMY  
use in the case of pigments in  
printing devices

White subtracted by CMY to get  
Black

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Color Matching Theory

- Young's observation
  - Any uniform color can be matched by projecting three different light sources onto a screen
- In the 1920's, an international effort on the part of a number of physicists led to the CIE standard color theory
- The effort attempted to simplify and standardize
- There have been many refinements, but the basic theory still stands
- Theory useful for matching pigments (all paint stores have spectrophotometers) and design of color media systems

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### Color Characterization

Hue: dominant color corresponding to a dominant wavelength of mixture light wave

Saturation: Relative purity or amount of white light mixed with a hue (inversely proportional to amount of white light added)

Brightness: Intensity

Hue } Chromaticity  
Saturation }

amount of red (X), green (Y) and blue (Z) to form any particular color is called *tristimulus*.

### Tristimulus Values

$I(\lambda)$  — spectral energy distribution of a source

$X, Y, Z$  — tristimulus values of the source

$$X = \int_{400}^{700} \bar{x}(\lambda) I(\lambda) d\lambda$$

$$Y = \int_{400}^{700} \bar{y}(\lambda) I(\lambda) d\lambda$$

$$Z = \int_{400}^{700} \bar{z}(\lambda) I(\lambda) d\lambda$$

Chromaticities - trichromatic coefficients

$$x = \frac{X}{X + Y + Z}$$

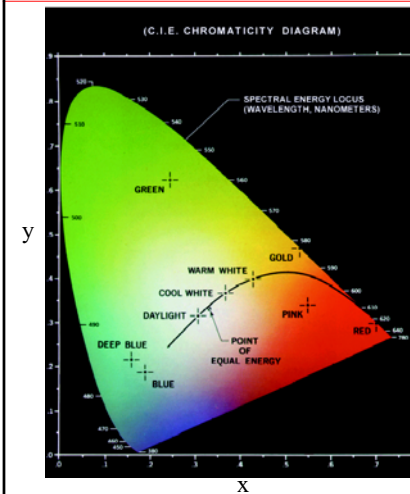
$$y = \frac{Y}{X + Y + Z}$$

$$z = \frac{Z}{X + Y + Z}$$

$$x + y + z = 1$$

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### CIE Chromaticity Diagram



Trichromatic coefficients:

$$x = \frac{X}{X + Y + Z}$$

$$y = \frac{Y}{X + Y + Z}$$

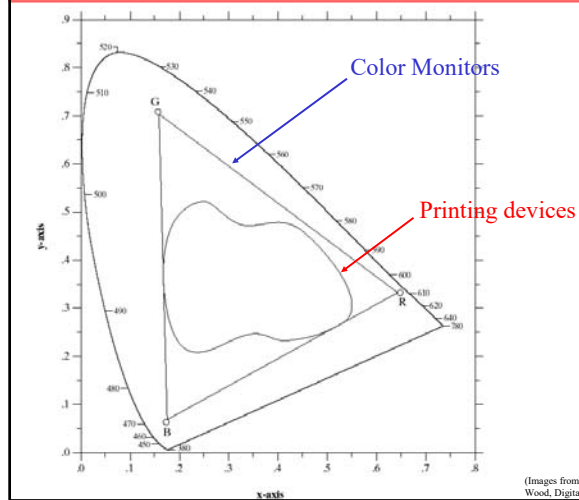
$$z = \frac{Z}{X + Y + Z}$$

$$x + y + z = 1$$

Points on the boundary are fully saturated colors

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Color Gamut of Color Monitors and Printing Devices

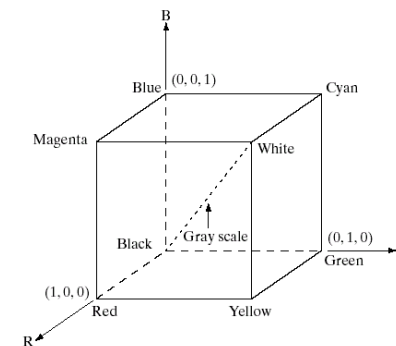


(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### RGB Color Model

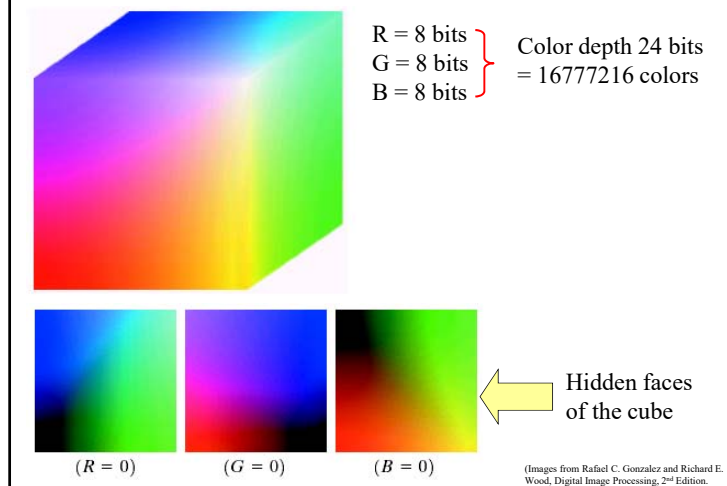
Purpose of color models: to facilitate the specification of colors in some standard

RGB color models:  
- based on cartesian coordinate system

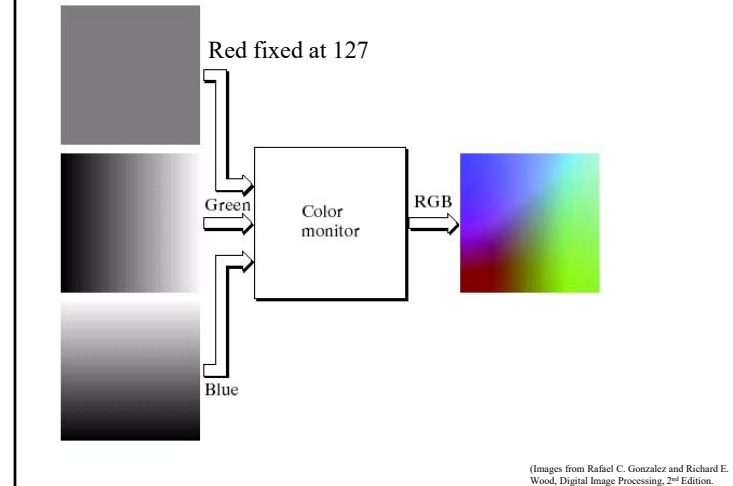


(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### RGB Color Cube

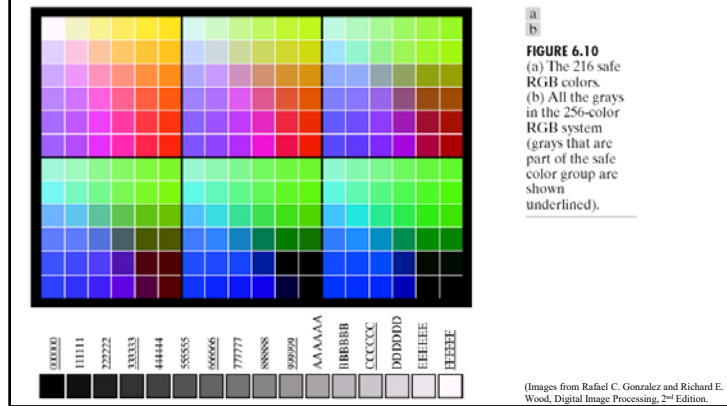


### RGB Color Model (cont.)

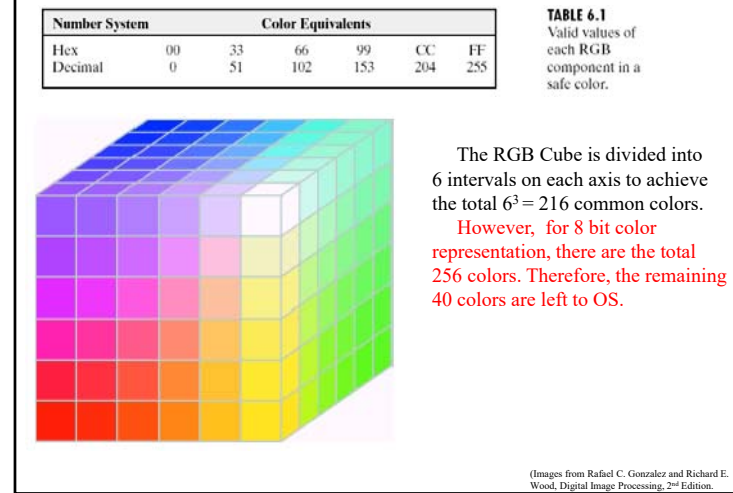


### Safe RGB Colors

Safe RGB colors: a subset of RGB colors.  
 There are 216 colors common in most operating systems.

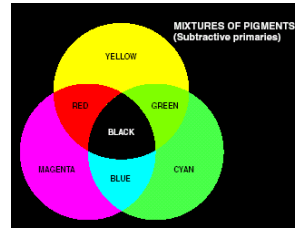


### RGB Safe-color Cube



### CMY and CMYK Color Models

C = Cyan  
M = Magenta  
Y = Yellow  
K = Black



$$\begin{bmatrix} C \\ M \\ Y \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

### HSI Color Model

RGB, CMY models are not good for human interpreting

HSI Color model:

Hue: Dominant color

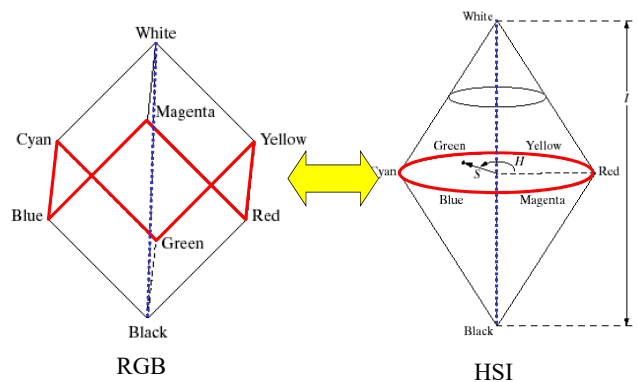
Saturation: Relative purity (inversely proportional to amount of white light added)

Color carrying information

Intensity: Brightness

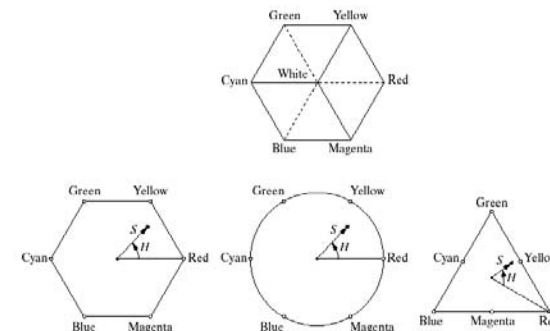
(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Relationship Between RGB and HSI Color Models



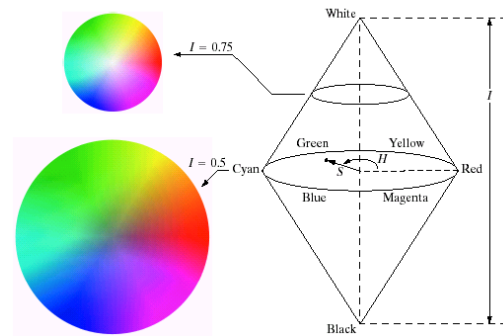
(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Hue and Saturation on Color Planes



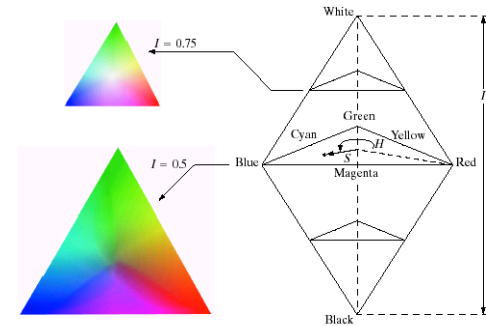
1. A dot in the plane is an arbitrary color.
2. Hue is an angle from a red axis.
3. Saturation is a distance to the point.

### HSI Color Model (cont.)



Intensity is given by a position on the vertical axis.

### HSI Color Model



Intensity is given by a position on the vertical axis.

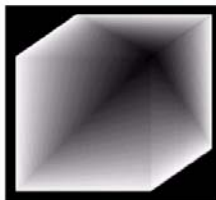
### Example: HSI Components of RGB Cube



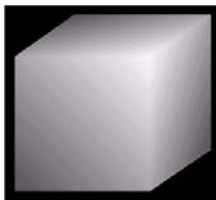
RGB Cube



Hue



Saturation



Intensity

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Converting Colors from RGB to HSI

$$H = \begin{cases} \theta & \text{if } B \leq G \\ 360 - \theta & \text{if } B > G \end{cases}$$

$$\theta = \cos^{-1} \left\{ \frac{\frac{1}{2}[(R-G) + (R-B)]}{[(R-G)^2 + (R-B)(G-B)]^{1/2}} \right\}$$

$$S = 1 - \frac{3}{R+G+B}$$

$$I = \frac{1}{3}(R+G+B)$$

### Converting Colors from HSI to RGB

RG sector:  $0 \leq H < 120$

$$R = I \left[ 1 + \frac{S \cos H}{\cos(60^\circ - H)} \right]$$

$$B = I(1 - S)$$

$$G = 1 - (R + B)$$

BR sector:  $240 \leq H \leq 360$

$$H = H - 240$$

$$B = I \left[ 1 + \frac{S \cos H}{\cos(60^\circ - H)} \right]$$

$$G = I(1 - S)$$

$$R = 1 - (G + B)$$

GB sector:  $120 \leq H < 240$

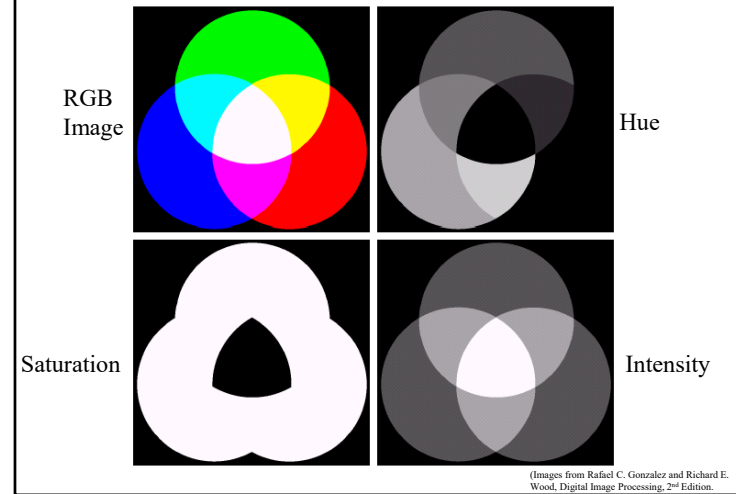
$$H = H - 120$$

$$R = I(1 - S)$$

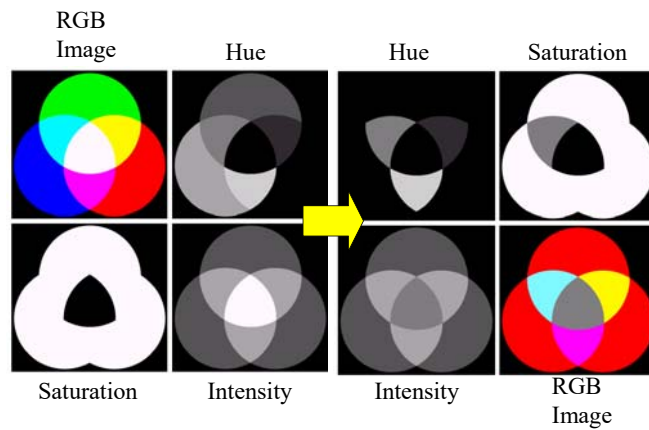
$$G = I \left[ 1 + \frac{S \cos H}{\cos(60^\circ - H)} \right]$$

$$B = 1 - (R + G)$$

### Example: HSI Components of RGB Colors



### Example: Manipulating HSI Components



### Color Image Processing

There are 2 types of color image processes

1. Pseudocolor image process: Assigning colors to gray values based on a specific criterion. Gray scale images to be processed may be a single image or multiple images such as multispectral images

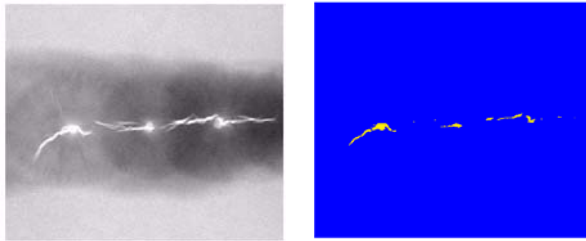
2. Full color image process: The process to manipulate real color images such as color photographs.

### Pseudocolor Image Processing

Pseudo color = false color : In some case there is no “color” concept for a gray scale image but we can assign “false” colors to an image.

Why we need to assign colors to gray scale image?

Answer: Human can distinguish different colors better than different shades of gray.

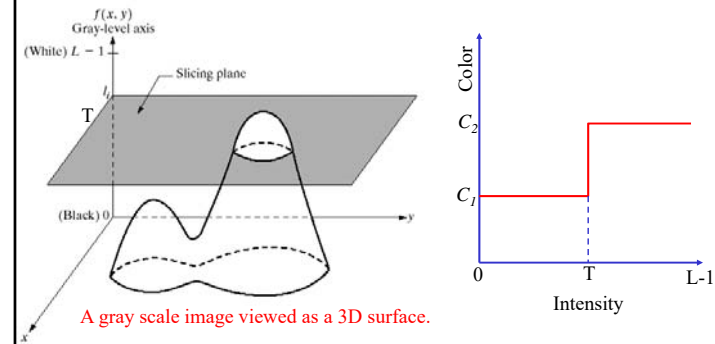


(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

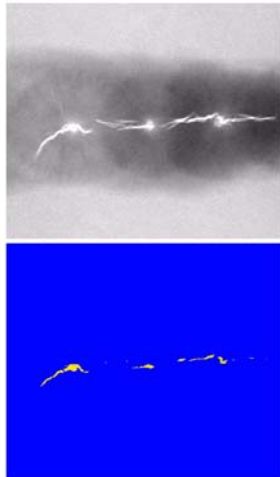
### Intensity Slicing or Density Slicing

Formula:

$$g(x, y) = \begin{cases} C_1 & \text{if } f(x, y) \leq T \\ C_2 & \text{if } f(x, y) > T \end{cases} \quad \begin{matrix} C_1 = \text{Color No. 1} \\ C_2 = \text{Color No. 2} \end{matrix}$$



### Intensity Slicing Example



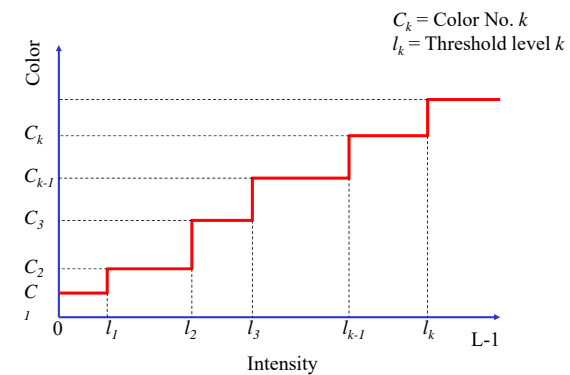
An X-ray image of a weld with cracks

After assigning a yellow color to pixels with value 255 and a blue color to all other pixels.

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Multi Level Intensity Slicing

$$g(x, y) = C_k \quad \text{for } l_{k-1} < f(x, y) \leq l_k$$

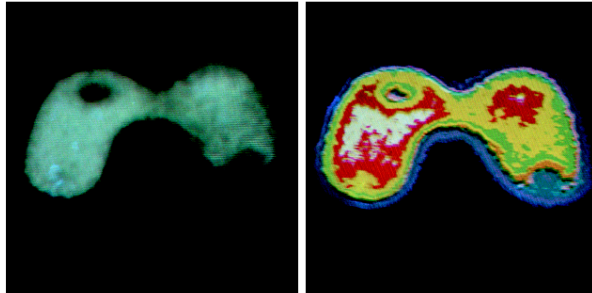




### Multi Level Intensity Slicing Example

$$g(x, y) = C_k \quad \text{for } l_{k-1} < f(x, y) \leq l_k$$

$C_k$  = Color No.  $k$   
 $l_k$  = Threshold level  $k$

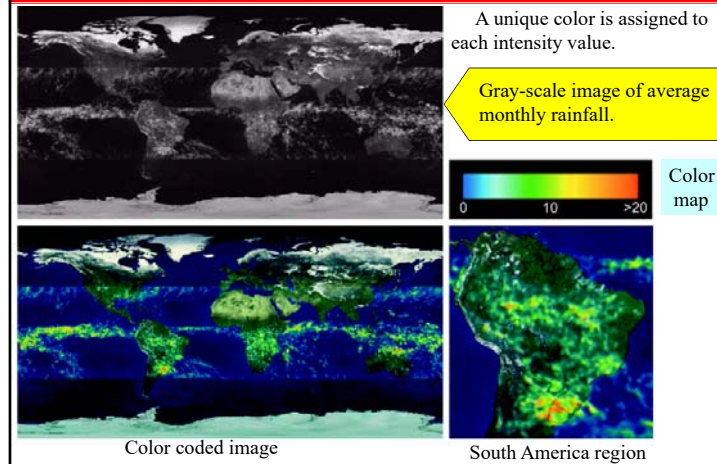


An X-ray image of the Picker Thyroid Phantom.

After density slicing into 8 colors

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

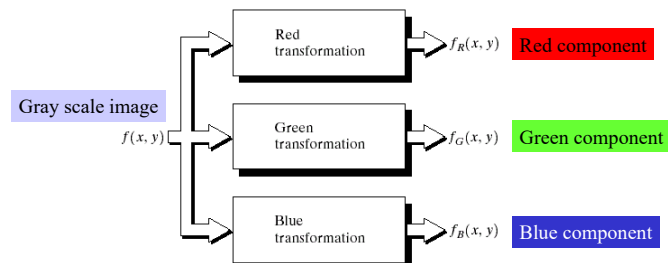
### Color Coding Example



(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

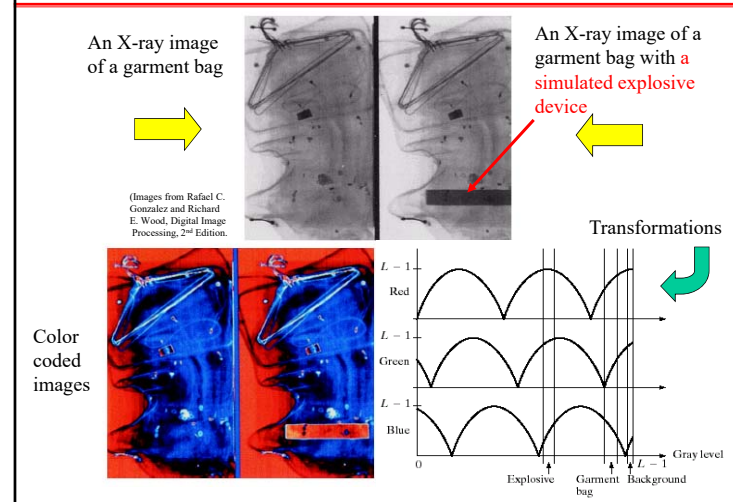
### Gray Level to Color Transformation

Assigning colors to gray levels based on specific mapping functions

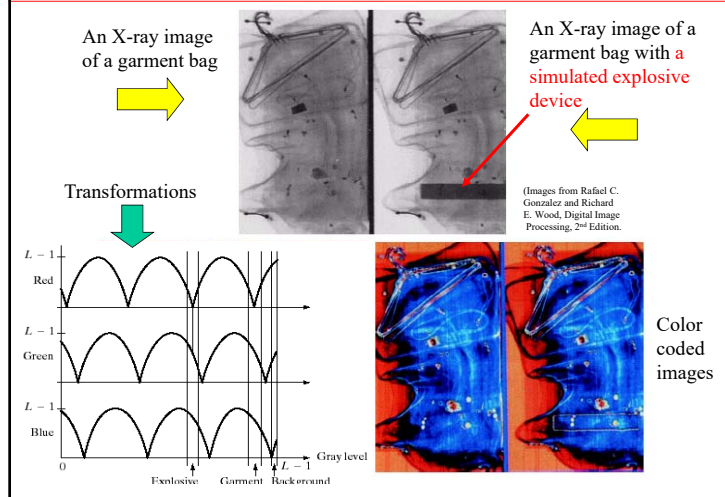


(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Gray Level to Color Transformation Example

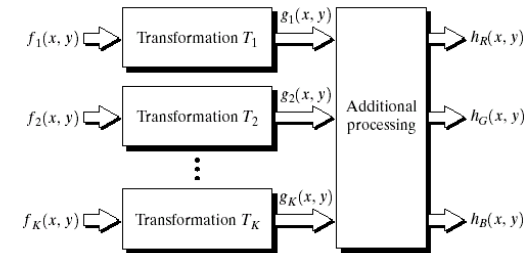


### Gray Level to Color Transformation Example



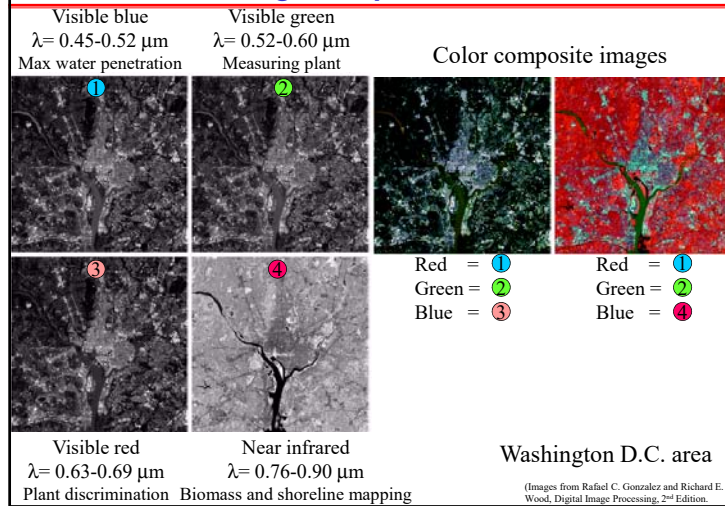
### Pseudocolor Coding

Used in the case where there are many monochrome images such as multispectral satellite images.

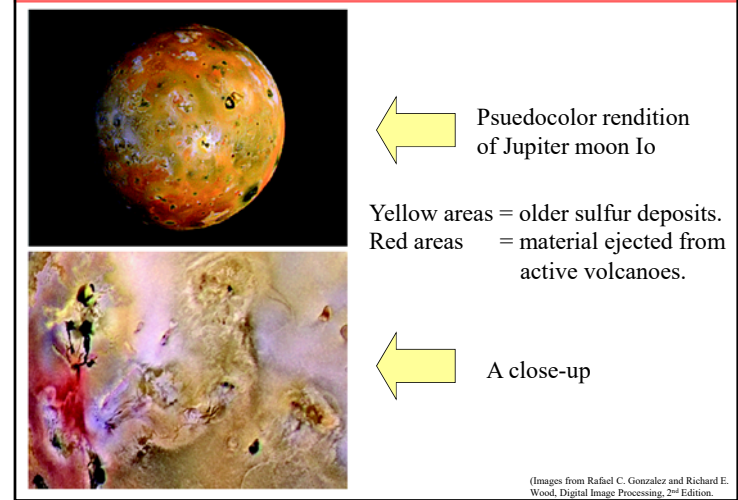


(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Pseudocolor Coding Example




### Pseudocolor Coding Example

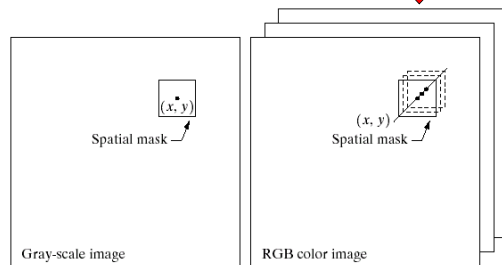


## Basics of Full-Color Image Processing

2 Methods:

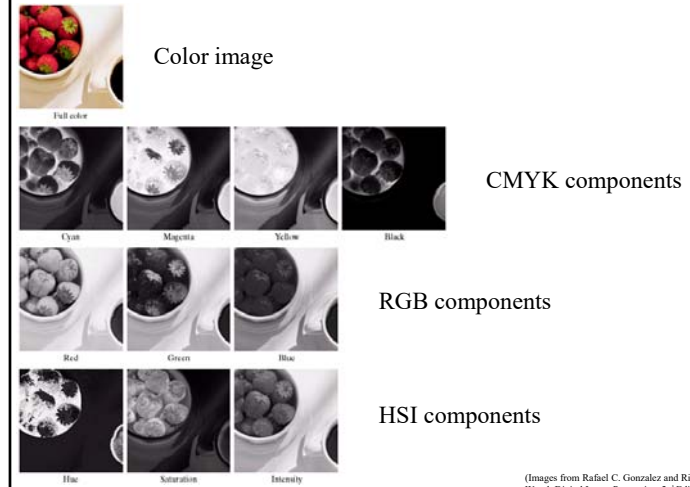
1. Per-color-component processing: process each component separately.
2. Vector processing: treat each pixel as a vector to be processed.

Example of per-color-component processing: smoothing an image  
By smoothing each RGB component separately. 



(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

## Example: Full-Color Image and Various Color Space Components



(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

## Color Transformation

Use to **transform colors to colors**.

Formulation:

$$g(x, y) = T[f(x, y)]$$

$f(x, y)$  = input color image,  $g(x, y)$  = output color image

$T$  = operation on  $f$  over a spatial neighborhood of  $(x, y)$

When **only data at one pixel is used** in the transformation, we can express the transformation as:

$$s_i = T_i(r_1, r_2, \dots, r_n) \quad i = 1, 2, \dots, n$$

Where  $r_i$  = color component of  $f(x, y)$  For RGB images,  $n = 3$   
 $s_i$  = color component of  $g(x, y)$

## Example: Color Transformation

Formula for RGB:

$$s_R(x, y) = kr_R(x, y)$$

$$s_G(x, y) = kr_G(x, y)$$

$$s_B(x, y) = kr_B(x, y)$$

Formula for HSI:

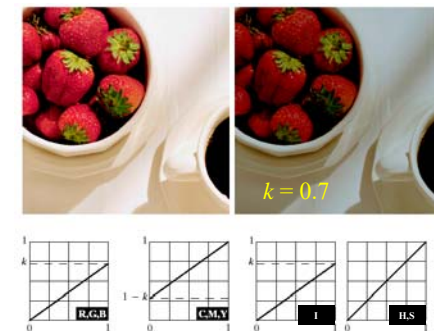
$$s_I(x, y) = kr_I(x, y)$$

Formula for CMY:

$$s_C(x, y) = kr_C(x, y) + (1 - k)$$

$$s_M(x, y) = kr_M(x, y) + (1 - k)$$

$$s_Y(x, y) = kr_Y(x, y) + (1 - k)$$

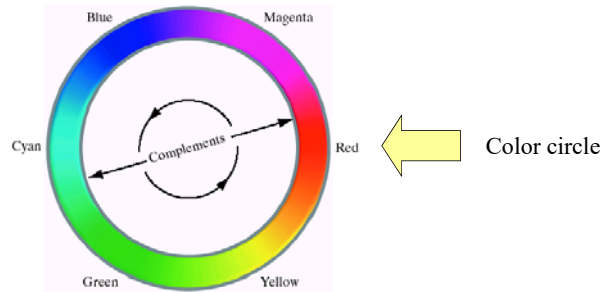


These 3 transformations give the same results.

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

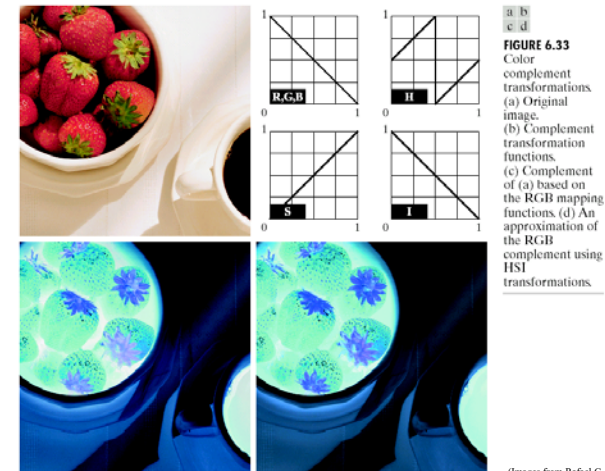
## Color Complements

Color complement replaces each color with its opposite color in the color circle of the Hue component. This operation is analogous to image negative in a gray scale image.



(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

## Color Complement Transformation Example



(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

## Color Slicing Transformation

We can perform “slicing” in color space: if the color of each pixel is far from a desired color more than threshold distance, we set that color to some specific color such as gray, otherwise we keep the original color unchanged.

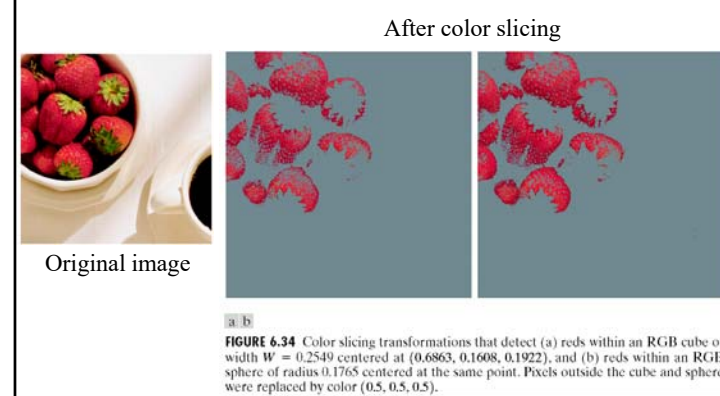
$$s_i = \begin{cases} 0.5 & \text{if } \left[ |r_j - a_j| > \frac{W}{2} \right]_{\text{any } 1 \leq j \leq n} \\ r_i & \text{otherwise} \end{cases} \quad i = 1, 2, \dots, n$$

or

$$s_i = \begin{cases} 0.5 & \text{if } \sum_{j=1}^n (r_j - a_j)^2 > R_0^2 \\ r_i & \text{otherwise} \end{cases} \quad i = 1, 2, \dots, n$$

→ Set to gray  
→ Keep the original color

## Color Slicing Transformation Example

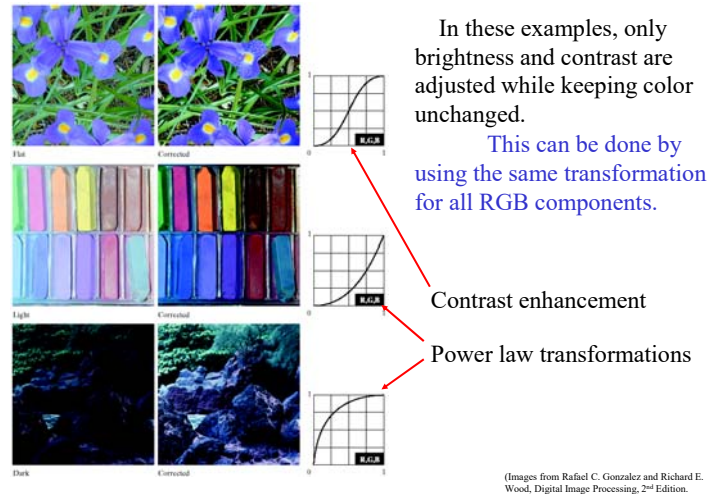


**FIGURE 6.34** Color slicing transformations that detect (a) reds within an RGB cube of width  $W = 0.2549$  centered at  $(0.6863, 0.1608, 0.1922)$ , and (b) reds within an RGB sphere of radius  $0.1765$  centered at the same point. Pixels outside the cube and sphere were replaced by color  $(0.5, 0.5, 0.5)$ .

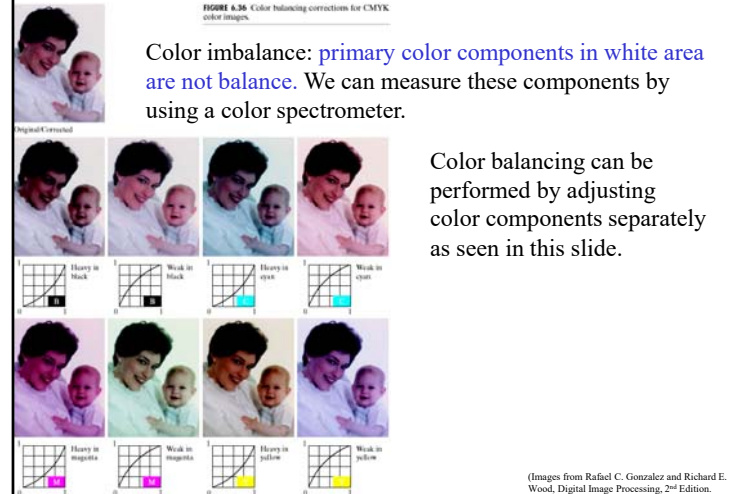
(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)



### Tonal Correction Examples



### Color Balancing Correction Examples



### Histogram Equalization of a Full-Color Image

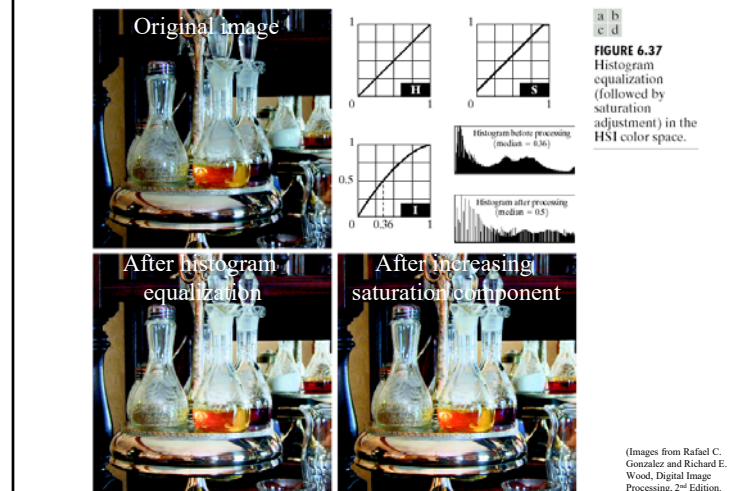
- ❖ Histogram equalization of a color image can be performed by adjusting color intensity uniformly while leaving color unchanged.
- ❖ The HSI model is suitable for histogram equalization where **only Intensity (I) component is equalized**.

$$s_k = T(r_k) = \sum_{j=0}^k p_r(r_j)$$

$$= \sum_{j=0}^k \frac{n_j}{N}$$

where  $r$  and  $s$  are intensity components of input and output color image.

### Histogram Equalization of a Full-Color Image



## Color Image Smoothing

### 2 Methods:

1. **Per-color-plane method:** for RGB, CMY color models  
Smooth each color plane using moving averaging and  
the combine back to RGB

$$\bar{\mathbf{c}}(x, y) = \frac{1}{K} \sum_{(x, y) \in S_{xy}} \mathbf{c}(x, y) = \begin{bmatrix} \frac{1}{K} \sum_{(x, y) \in S_{xy}} R(x, y) \\ \frac{1}{K} \sum_{(x, y) \in S_{xy}} G(x, y) \\ \frac{1}{K} \sum_{(x, y) \in S_{xy}} B(x, y) \end{bmatrix}$$

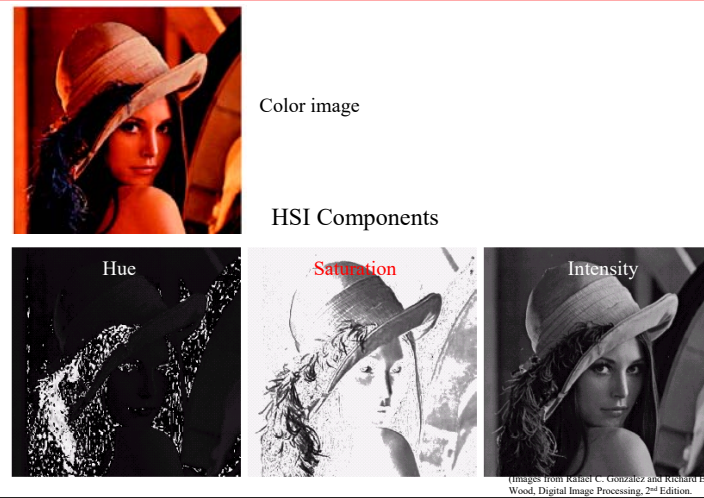
2. Smooth only Intensity component of a HSI image while leaving  
H and S unmodified.

Note: 2 methods are not equivalent.

## Color Image Smoothing Example (cont.)



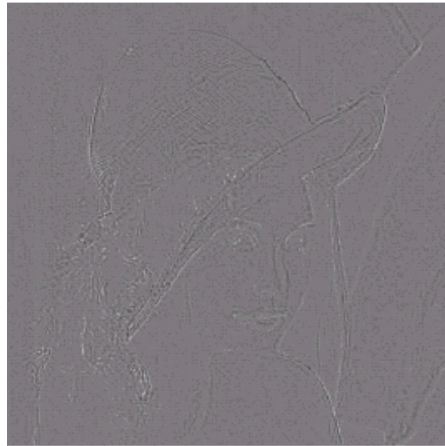
## Color Image Smoothing Example (cont.)



## Color Image Smoothing Example (cont.)



### Color Image Smoothing Example (cont.)



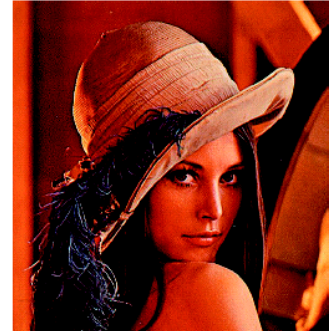
Difference between smoothed results from 2 methods in the previous slide.

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.

### Color Image Sharpening

We can do in the same manner as color image smoothing:

1. Per-color-plane method for RGB, CMY images
2. Sharpening only I component of a HSI image



Sharpening all RGB components



Sharpening only I component of HSI

### Color Image Sharpening Example (cont.)



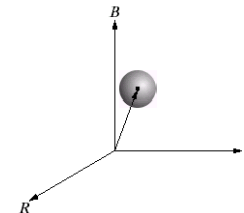
Difference between sharpened results from 2 methods in the previous slide.

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.

### Color Segmentation

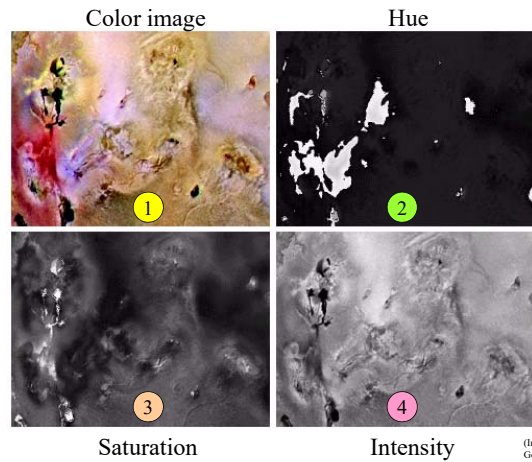
#### 2 Methods:

1. Segmented in HSI color space:  
A thresholding function based on color information in H and S Components. We rarely use I component for color image segmentation.
2. Segmentation in RGB vector space:  
A thresholding function based on distance in a color vector space.



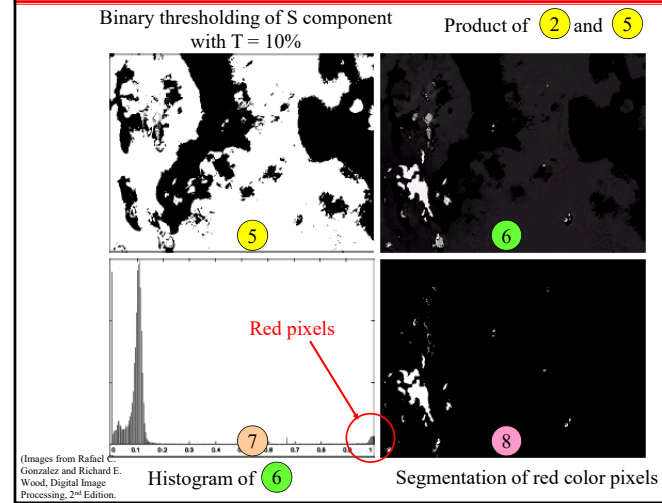
(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.

### Color Segmentation in HSI Color Space



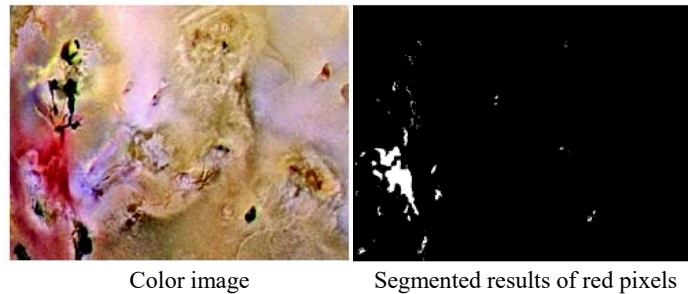
(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Color Segmentation in HSI Color Space (cont.)



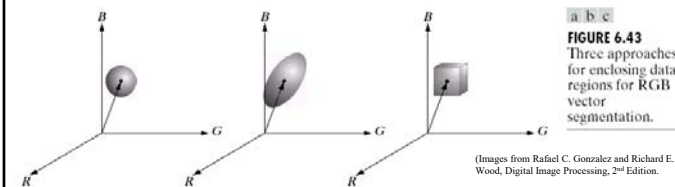
(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Color Segmentation in HSI Color Space (cont.)



(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Color Segmentation in RGB Vector Space



(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

1. Each point with (R,G,B) coordinate in the vector space represents one color.
2. Segmentation is based on distance thresholding in a vector space

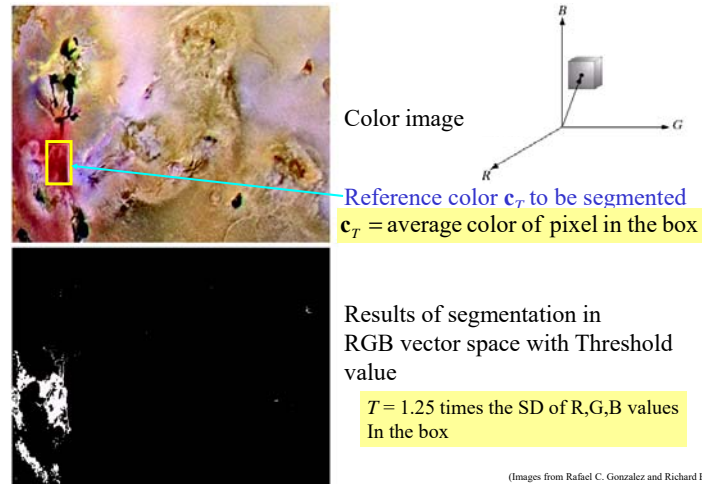
$$g(x, y) = \begin{cases} 1 & \text{if } D(\mathbf{c}(x, y), \mathbf{c}_T) \leq T \\ 0 & \text{if } D(\mathbf{c}(x, y), \mathbf{c}_T) > T \end{cases}$$

$D(\mathbf{u}, \mathbf{v})$  = distance function

$\mathbf{c}_T$  = color to be segmented.  
 $\mathbf{c}(x, y)$  = RGB vector at pixel (x,y).

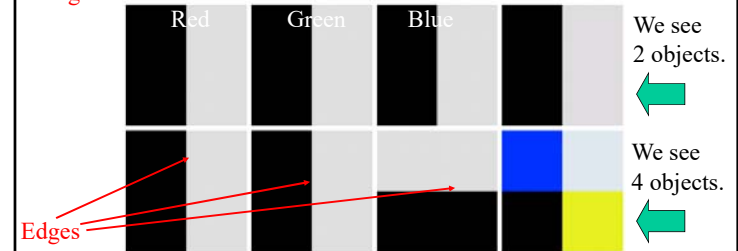


### Example: Segmentation in RGB Vector Space



### Gradient of a Color Image

Since gradient is defined only for a scalar image, there is no concept of gradient for a color image. **Compute gradient of each color component and combine the results to get the gradient of a color image?**



### Gradient of a Color Image (cont.)

One way to compute the maximum rate of change of a color image which is close to the meaning of gradient is to use the following formula: **Gradient computed in RGB color space:**

$$F(\theta) = \left\{ \frac{1}{2} [(g_{xx} + g_{yy}) + (g_{xx} - g_{yy}) \cos 2\theta + 2g_{xy} \sin 2\theta] \right\}^{\frac{1}{2}}$$

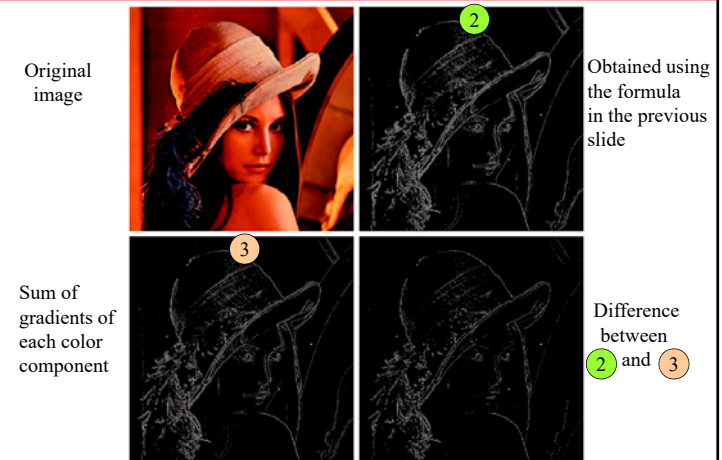
$$\theta = \frac{1}{2} \tan^{-1} \left[ \frac{2g_{xy}}{(g_{xx} - g_{yy})} \right]$$

$$g_{xx} = \left| \frac{\partial R}{\partial x} \right|^2 + \left| \frac{\partial G}{\partial x} \right|^2 + \left| \frac{\partial B}{\partial x} \right|^2$$

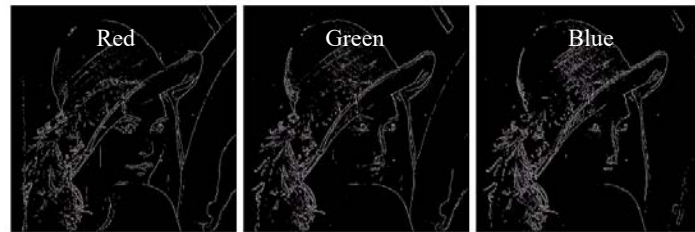
$$g_{yy} = \left| \frac{\partial R}{\partial y} \right|^2 + \left| \frac{\partial G}{\partial y} \right|^2 + \left| \frac{\partial B}{\partial y} \right|^2$$

$$g_{xy} = \frac{\partial R}{\partial x} \frac{\partial R}{\partial y} + \frac{\partial G}{\partial x} \frac{\partial G}{\partial y} + \frac{\partial B}{\partial x} \frac{\partial B}{\partial y}$$

### Gradient of a Color Image Example



### Gradient of a Color Image Example



Gradients of each color component

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2<sup>nd</sup> Edition.)

### Noise in Color Images

Noise can corrupt each color component independently.

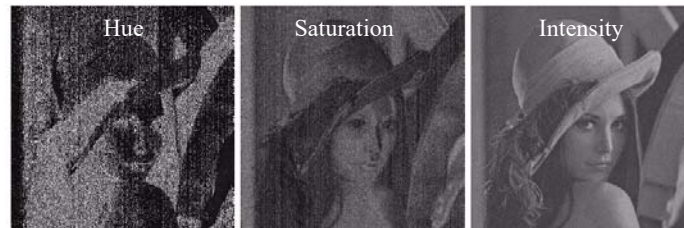


**FIGURE 6.48**  
(a)–(c) Red, green, and blue component images corrupted by additive Gaussian noise of mean 0 and variance 800. (d) Resulting RGB image. [Compare (d) with Fig. 6.46(a).]

Noise is less noticeable in a color image

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2<sup>nd</sup> Edition.)

### Noise in Color Images



a b c

**FIGURE 6.49** HSI components of the noisy color image in Fig. 6.48(d). (a) Hue. (b) Saturation. (c) Intensity.

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2<sup>nd</sup> Edition.)

### Noise in Color Images



**FIGURE 6.50**  
(a) RGB image with green plane corrupted by salt-and-pepper noise. (b) Hue component of HSI image. (c) Saturation component. (d) Intensity component.

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2<sup>nd</sup> Edition.)

### Color Image Compression

Original image



→ JPEG2000 File

After lossy compression with ratio 230:1

(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.)

### Transformation of random variables

$$Y = g(X)$$

$$F_Y(y_0) = P(Y \leq y_0) = P[X \leq x_0] = F_X(x_0)$$

$$\int_0^{y_0} f_Y(y) dy = \int_0^{x_0} f_X(x) dx$$

$$f_Y(y) = f_X(g^{-1}(y)) \frac{dg^{-1}(y)}{dy}$$

$$f_Y(y) = f_X(x) \frac{dx}{dy}$$

### Homework – due next class

1. [Clustering] What is k-means clustering? Make up the matlab code for k-means clustering and run the code for your own sample data. Explain the results.
2. [Color Image Segmentation] For a given set of color images, apply k-means clustering for segmentation. Try *different color systems* and discuss on the results. Is there any way to improve it? Discuss on improvements and present some experimental results if any.
3. [Color Edge Detection] Find methods for color edge finding in the literature. Categorize and summarize different methods with references. Mention representative methods and compare the performances. Code more than one methods and compare the results.

### Sample images for clustering

