

## Chapter 6 Color Image Processing

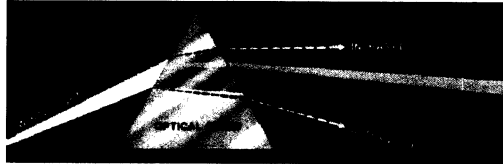


FIGURE 6.1 Color spectrum seen by passing white light through a prism. (Courtesy of the General Electric Co., Lamp Business Division.)

white light split into individual colors using a prism

full-color - image acquired with a full-color sensor

pseudocolor - color assigned to a intensity value or range

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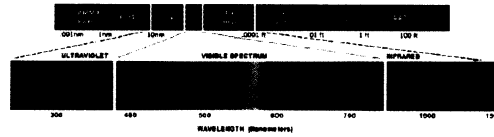


FIGURE 6.2 Wavelengths comprising the visible range of the electromagnetic spectrum. (Courtesy of the General Electric Co., Lamp Business Division.)

blue

red

physical units

radiance (watts) - total energy emitted by a light source

luminance (lumens) - incoming energy as measured by the detector

subjective units

brightness

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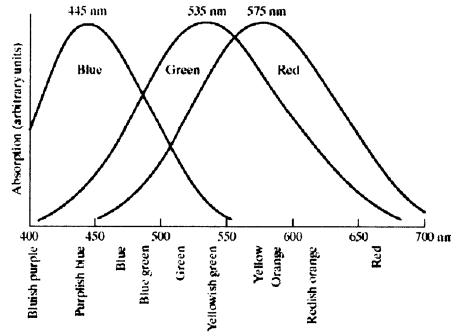


FIGURE 6.3 Absorption of light by the red, green, and blue cones in the human eye as a function of wavelength.

Spectral response of human eye

primary colors do not quite correspond to human eye sensitivity

red (700nm) — 65% of all cones sensitive to red

green (546.1nm) — 33% " green

blue (435.8nm) — 2% " blue

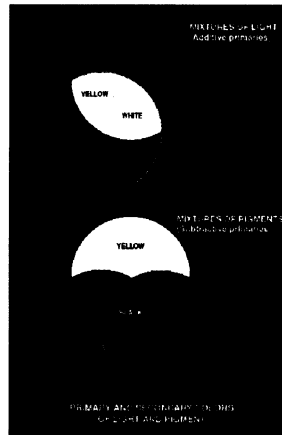
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(primary)  
Additive  
colors  
transmission

Subtractive  
colors  
(secondary)  
reflection

absorbs a primary color and



} red-green-blue  
television monitors

} cyan-magenta-yellow  
printing  
need black in printing

FIGURE 6.4 Primary and secondary colors of light and pigments. (Courtesy of the General Electric Co., Lamp Business Division.)

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yellow - absorbs blue  
and transmits  
red + green = yellow

characteristics of color

brightness

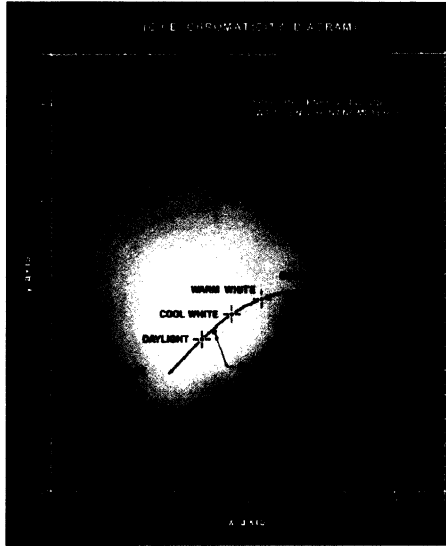
chromaticity {

hue - dominant color seen by an observer

saturation - amount of white light mixed with the color

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FIGURE 6.5  
Chromaticity  
diagram  
(Courtesy of the  
General Electric  
Co., Lamp  
Business  
Division.)



any point on  
boundary is  
fully saturated  
(saturation = 1)

green ↑

+ green  $\approx$  (.25red, .62green)  
point of equal energy - white  
(.33red, .33green)

Several kinds of white  
warm - more red

cool -

→ red natural -  
(daylight)

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A color can be specified by its tristimulus values

red  $x = \frac{X}{x+y+z}$

green  $y = \frac{Y}{x+y+z}$

blue  $z = \frac{Z}{x+y+z}$

where  $x+y+z=1$

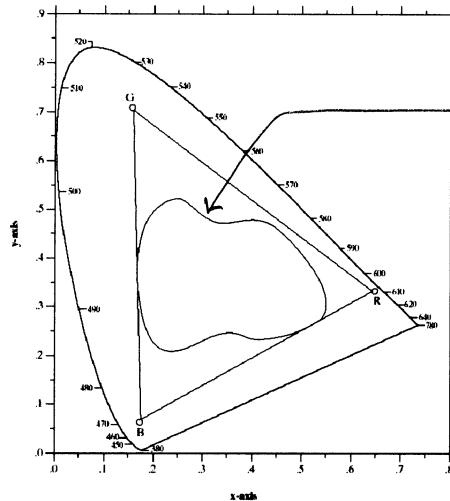
where  $x, y, z$  are the amounts of red, green, and blue needed to form a color

CIE chromaticity diagram

just specify  $x, y$  (red, green) since blue is then determined by  $1-x-y$ .

point of equal energy = CIE standard for white light (saturation = 0 here)

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irregular area  
is typical color  
printing area.

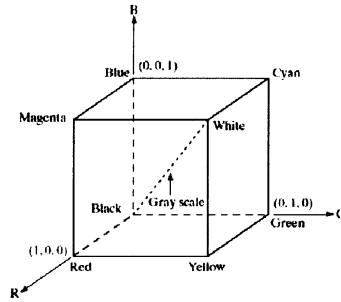
FIGURE 6.6 Typical color gamut of color monitors (triangle) and color printing devices (irregular region).

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The RGB points represent the maximum RGB values of an RGB monitor. Since a monitor is an additive process this RGB monitor can create any color within the triangle.

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**FIGURE 6.7**  
Schematic of the RGB color cube. Points along the main diagonal have gray values, from black at the origin to white at point (1, 1, 1).



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RGB color model

depth is the number of bits used in total

For example, 8-bit RGB is  $8 \times 3 = 24$  bit depth.

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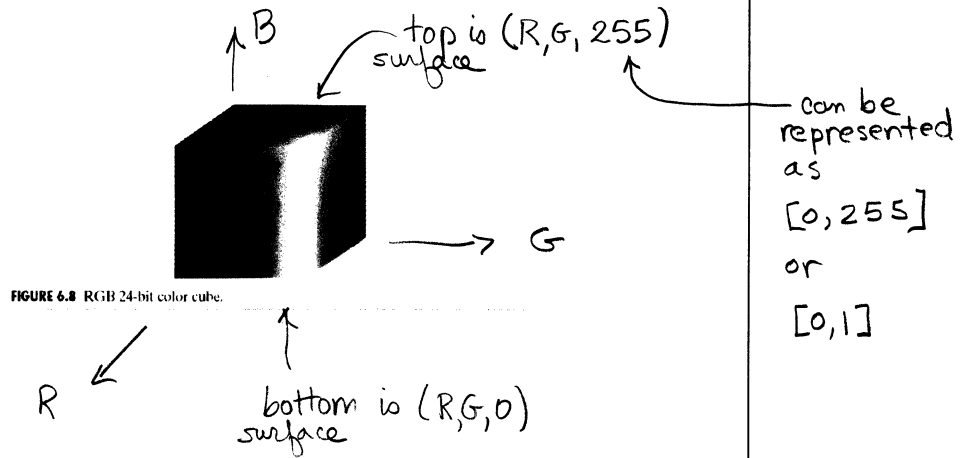
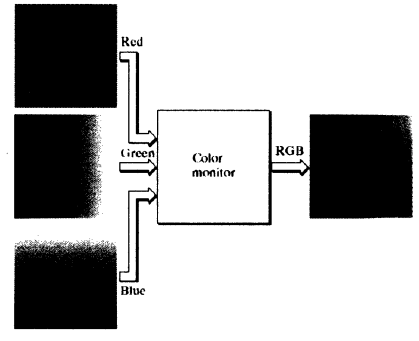


FIGURE 6.8 RGB 24-bit color cube.



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a  
b  
**FIGURE 6.9**  
(a) Generating the RGB image of the cross-sectional color plane (127, G, B).  
(b) The three hidden surface planes in the color cube of Fig. 6.8.



combine color planes into color image



(R,G,B)

(0,G,B)

(R,0,B)

(R,G,0)

back planes looking inside out

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Number System	Color Equivalents					
Hex	00	33	66	99	CC	FF
Decimal	0	51	102	153	204	255

**TABLE 6.1**  
Valid values of each RGB component in a safe color.

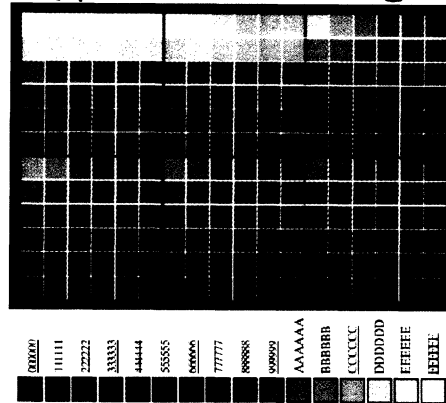
FF FF 99  
FF FF CC  
FF FF FF  
FF CC FF  
FF 99 FF

→ decreasing B

decreasing G

Red:

\$FF	\$CC	\$99
\$66	\$33	\$00



**FIGURE 6.10**  
(a) The 216 safe RGB colors.  
(b) All the grays in the 256-color RGB system (grays that are part of the safe color group are shown underlined).

includes Internet "safe" colors

Many systems in use today restrict themselves to 256 colors for simplicity and speed of generation.

Of the only 216 Internet "safe" colors are reliably reproduced by the operating system.

(R, G, B)

each value can only be 00, 33, 66, 99, CC or FF

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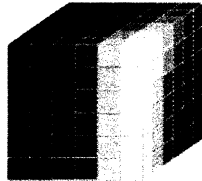


FIGURE 6.11 The RGB safe-color cube.

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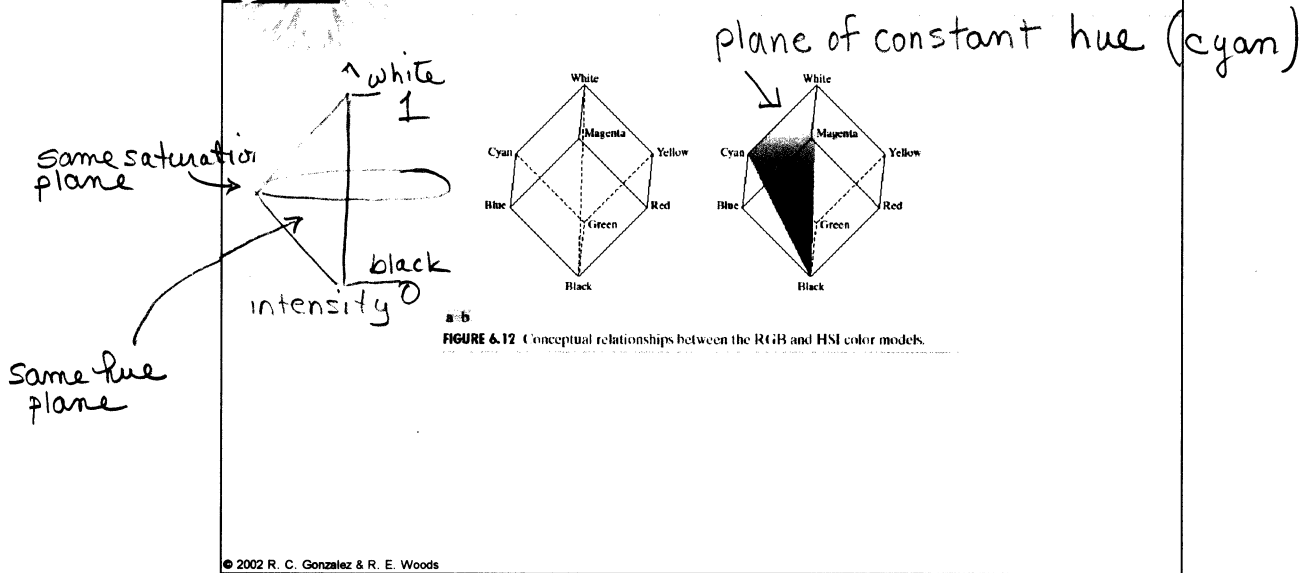
"safe" colors are only on the surfaces (faces).  
No interior colors are "safe".

Color printers & copiers convert RGB to CMY

$$\begin{bmatrix} c \\ m \\ y \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

To get a good black on color printers we add "black"  
as a fourth color.

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Although RGB is good for generating colors it is not good for describing colors as humans interpret them.

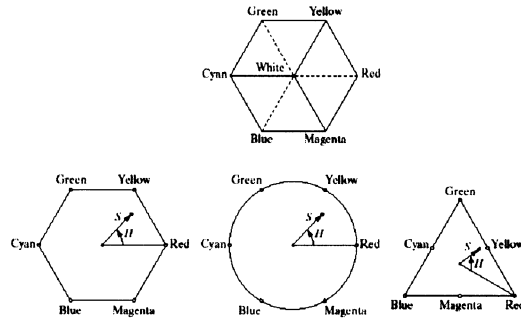
H - hue  
S - saturation } color information decoupled from intensity  
I - intensity } gray scale image

For an RGB color cube the intensity  $I$  is the diagonal from  $(0,0,0)$  black to  $(1,1,1)$  white. The intensity of any RGB color is its projection onto this intensity diagonal.

The plane perpendicular to the gray diagonal in the RGB cube will contain all colors of the same saturation since white does not change.

The plane defined by the gray diagonal and the cube boundaries will contain all colors of the same hue.

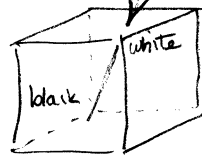
# Chapter 6 Color Image Processing



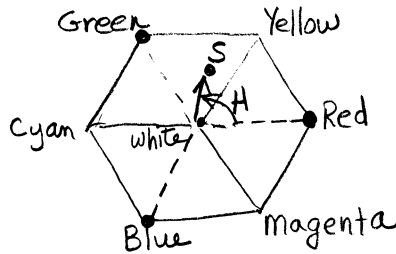
**FIGURE 6.13** Hue and saturation in the HSI color model. The dot is an arbitrary color point. The angle from the red axis gives the hue, and the length of the vector is the saturation. The intensity of all colors in any of these planes is given by the position of the plane on the vertical intensity axis.

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looking at diagonal perpendicular planes are hexagonal.



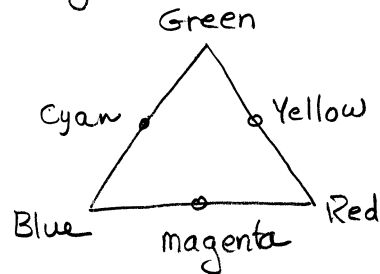
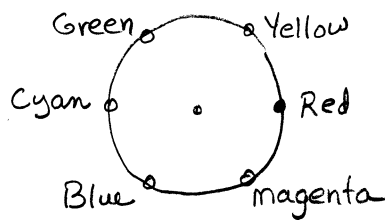
looking along axis



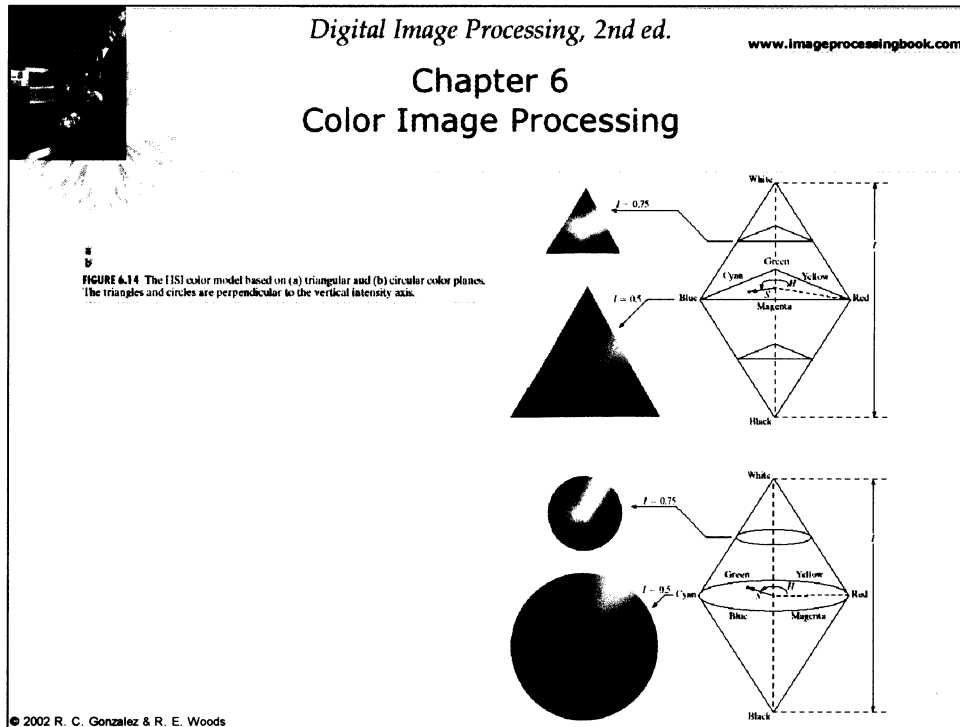
The hue of any point in each plane is its angle from a reference point.

The saturation is the distance from the origin.

Various simplifications of this hexagon are circles and triangles.



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From Figure 6.13 we can use trigonometry to derive the relationships between RGB and an HSI point.

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

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

$$\text{saturation } S = 1 - \frac{3}{R+G+B} \min(R, G, B)$$

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

See book's web site for derivation

## Converting colors from HSI to RGB

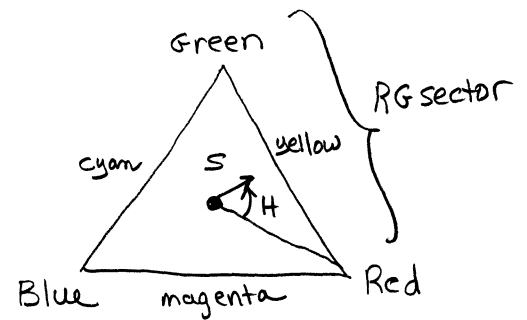
Multiply  $H$  by  $360^\circ$  to convert it back to an angle.  $H$  is usually normalized to  $[0, 1]$ .

### RG Sector ( $0^\circ \leq H \leq 120^\circ$ )

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

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

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



### GB sector ( $120^\circ \leq H \leq 240^\circ$ )

compute  $H = H - 120^\circ$

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

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

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

### BR sector ( $240^\circ \leq H \leq 360^\circ$ )

compute  $H = H - 240^\circ$

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

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

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

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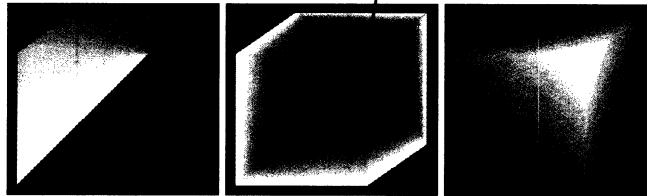
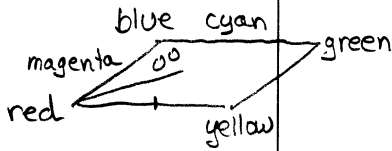


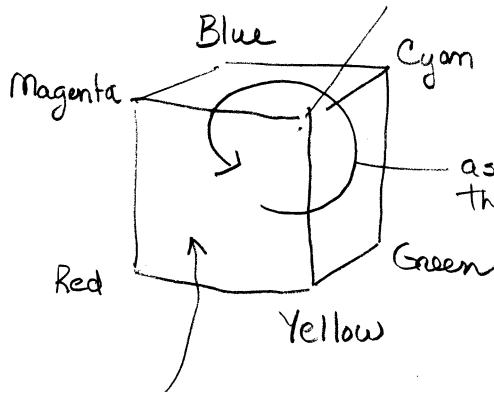
FIGURE 6.15 HSI components of the image in Fig. 6.8. (a) Hue, (b) saturation, and (c) intensity images.

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discontinuity because H is going from 1 (bright) to 0 (dark)

shows decreasing value, i.e. less saturation as you approach white

$I = \frac{1}{3}(R+G+B)$   
is just average of R, G, B values



as seen looking back along the gray axis we have a discontinuity from 1 to zero along the red-white line.

this is the  $H=0$  plane where H switches from  $0^\circ$  (black) to  $360^\circ$  (white)

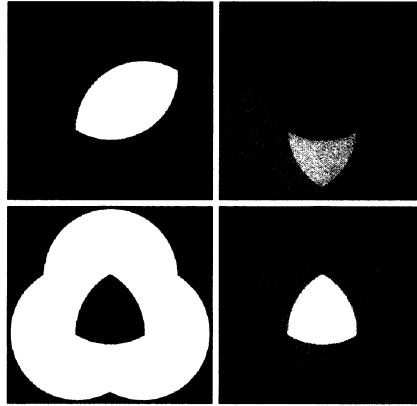


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Primary-  
secondary  
colors

S  
component  
[0,255]  
colors are  
fully saturated  
(on surface of  
RGB cube)

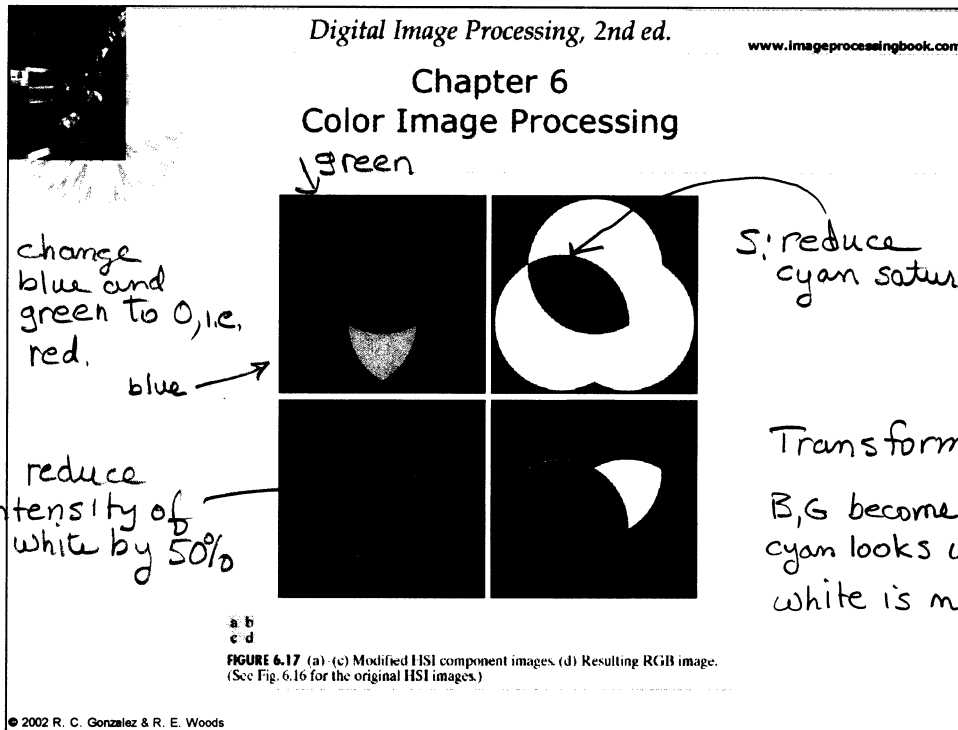


H  
component  
(angles)

black and white  
have zero hue  
red is 0° or black.

I  
component  
average  
intensities

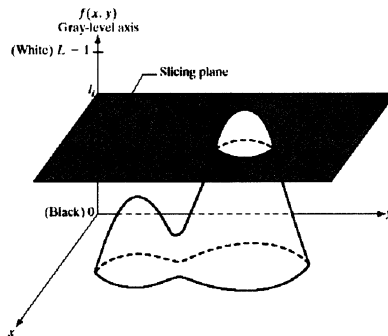
<sup>a b</sup>  
<sub>c d</sub>  
**FIGURE 6.16** (a) RGB image and the components of its corresponding HSI image:  
(b) hue, (c) saturation, and (d) intensity.



To change color in HSI simply change the hue value and convert back to RGB without changing S and I.

To change saturation modify only S.

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↑ if gray > li  
color A  
↓ if gray < li  
color B.

FIGURE 6.18 Geometric interpretation of the intensity-slicing technique.

pseudo color (false color) – assign colors to gray values using some criterion  
– used a lot in data visualization  
↓  
false since these are not real color

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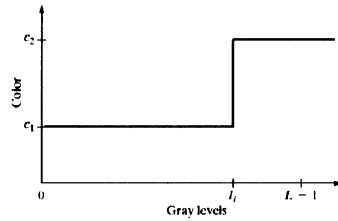
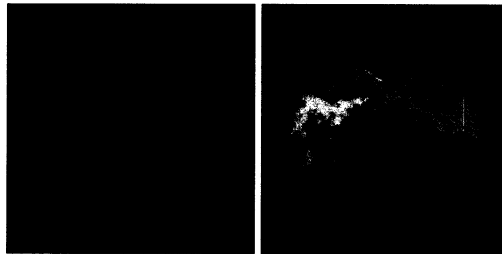


FIGURE 6.19 An alternative representation of the intensity-slicing technique.

Simply another way of describing what is done in Figure 6-18.

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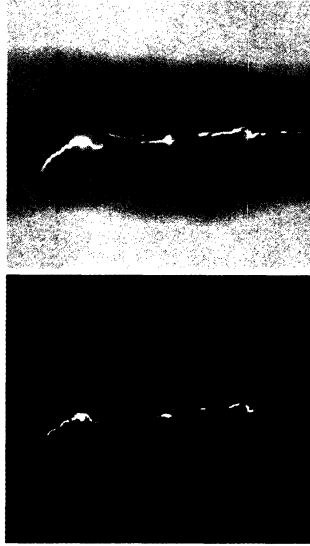
a b

**FIGURE 6.20** (a) Monochrome image of the Picker Thyroid Phantom. (b) Result of density slicing into eight colors. (Courtesy of Dr. J. L. Blankenship, Instrumentation and Controls Division, Oak Ridge National Laboratory.)

intensity slicing with multiple color slices

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**FIGURE 6.21**  
(a) Monochrome X-ray image of a weld. (b) Result of color coding. (Original image courtesy of X-TEK Systems, Ltd.)

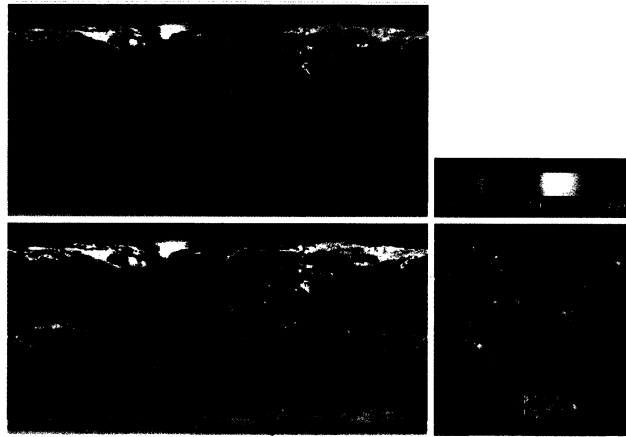


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Simple application in X-ray analysis.  
Cracks allow full X-ray intensity through metal.  
Image simply codes 255 as yellow and all others  
as blue for inspection.

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intensity values  
correspond to  
rainfall



a b  
c d

FIGURE 6.22 (a) Gray-scale image in which intensity (in the lighter horizontal band shown) corresponds to average monthly rainfall. (b) Colors assigned to intensity values. (c) Color-coded image. (d) Zoom of the South America region. (Courtesy of NASA.)

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Combine signals from (Tropical Rainfall Measuring Mission satellite)

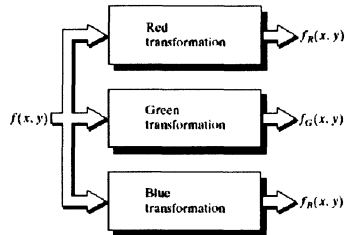
- precipitation radar
- microwave images
- visible/IR scanner.

to estimate average monthly rainfall.

Difficult to see patterns in grayscale. Much easier to see in pseudocolor.

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single gray  
scale input  
image



**FIGURE 6.23** Functional block diagram for pseudocolor image processing.  $f_R$ ,  $f_G$ , and  $f_B$  are fed into the corresponding red, green, and blue inputs of an RGB color monitor.

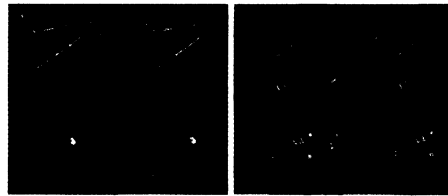
We can use simultaneous  
non-linear transforms to drive  
a color camera.



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← two images  
both gray scale.



Notice how  
explosives show  
up differently.

a  
b c

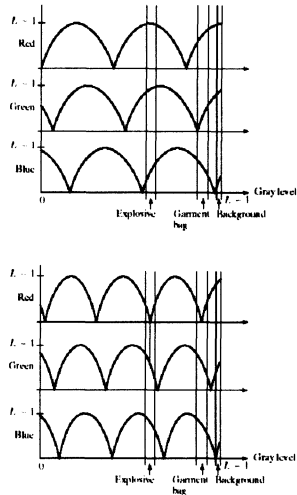
FIGURE 6.24 Pseudocolor enhancement by using the gray-level to color transformations in Fig. 6.25. (Original image courtesy of Dr. Mike Hurwitz, Westinghouse.)

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transform      transform  
6.25(a)      6.25(b)

2 different transforms.

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$L = \# \text{ of gray levels}$

FIGURE 6.25 Transformation functions used to obtain the images in Fig. 6.24.

transformations used.  
to make explosives in 6.24 visible.

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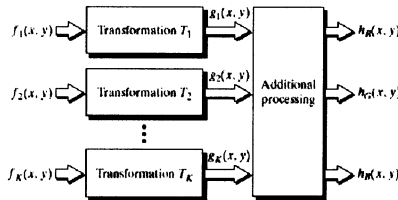


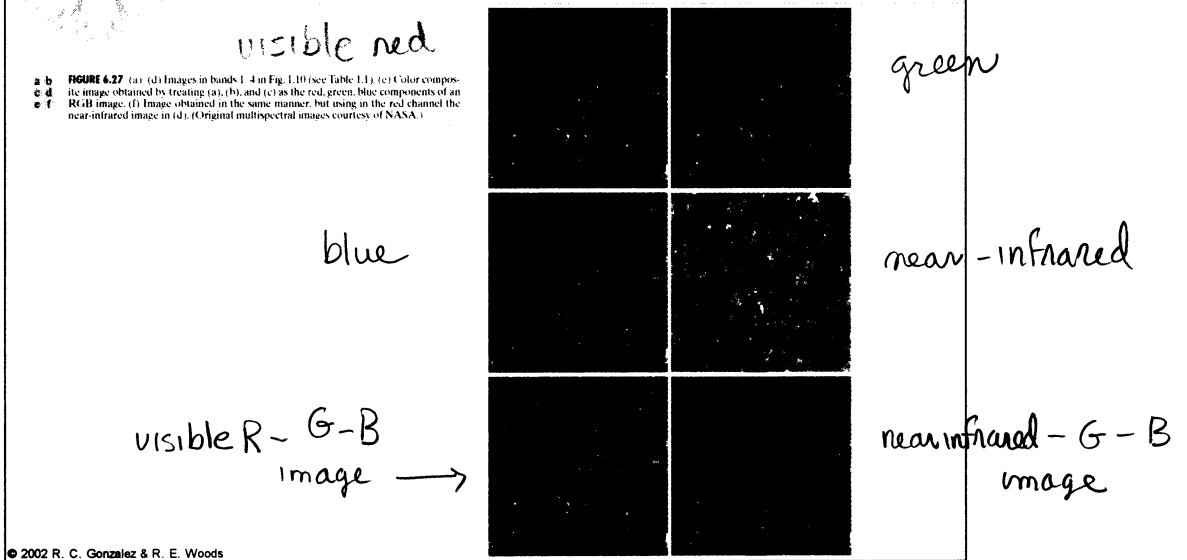
FIGURE 6.26 A pseudocolor coding approach used when several monochrome images are available.

more sophisticated color transformations can be used to combine grayscale images from different sensors as an example.

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**FIGURE 6.27** (a)–(d) Images in bands 1–4 in Fig. 1.10 (see Table 1.1). (e) Color composite image obtained by treating (a), (b), and (c) as the red, green, blue components of an RGB image. (f) Image obtained in the same manner, but using in the red channel the near-infrared image in (d). (Original multispectral images courtesy of NASA.)

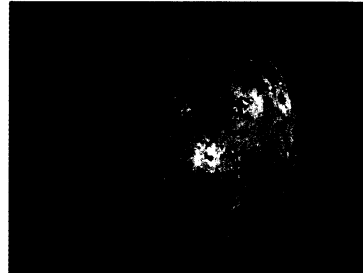


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↑  
used visible red  
for R

↑  
used nearinfrared  
for R.

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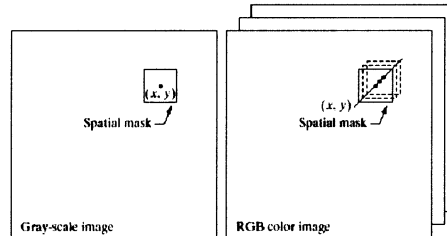


**FIGURE 6.28**  
(a) Pseudocolor  
rendition of  
Jupiter Moon Io.  
(b) A close-up  
(Courtesy of  
NASA.)

used a variety of different wavelengths  
The newly ejected material is red (different material)  
The older material is yellow. (sulfur)

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a b  
**FIGURE 6.29**  
Spatial masks for  
gray-scale and  
RGB color  
images.



for neighborhood averaging operations are equivalent  
 sum and divide all pixels in neighborhood  
 sum and divide all the vectors in the neighborhood to get the same result as averaging each color component and combining

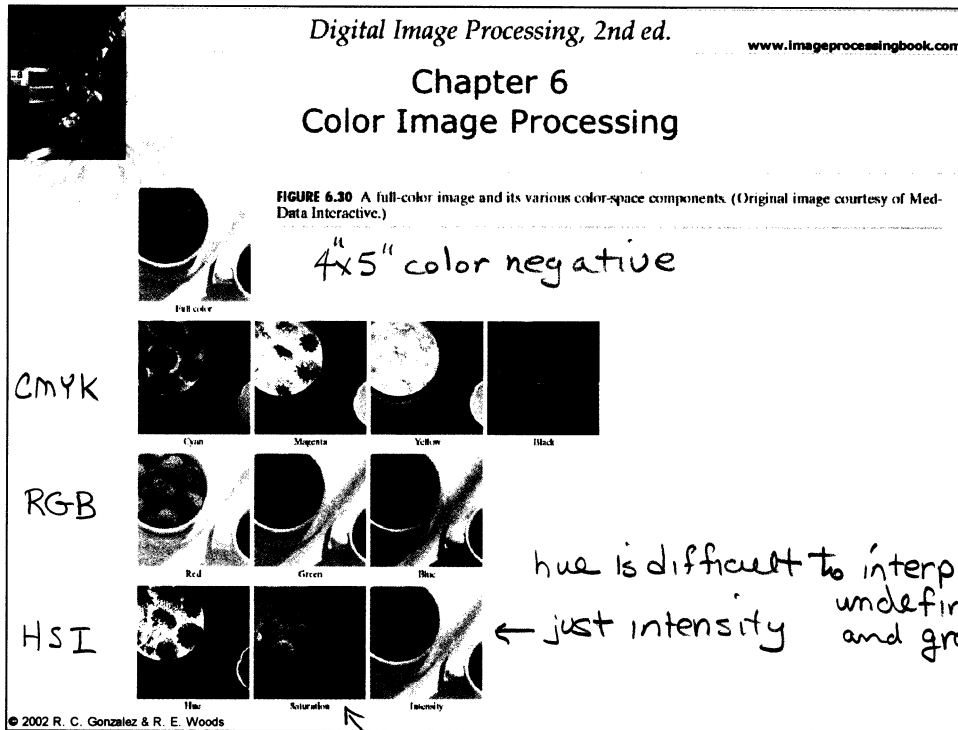
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### full-color image processing

- process each component image separately and combine to form a composite image
- process color vectors (pixels) directly.

$$c(x, y) = \begin{bmatrix} c_R(x, y) \\ c_G(x, y) \\ c_B(x, y) \end{bmatrix} = \begin{bmatrix} R(x, y) \\ G(x, y) \\ B(x, y) \end{bmatrix}$$

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color transformations

$$S_i = T_i(r_1, r_2, \dots, r_n) \quad i=1, \dots, n = \# \text{ of color components}$$

new color components      color components, i.e., R, G, B       $T_i = \text{set of color transformations}$

There are different costs associated with image processing in the different color spaces

to do intensity modification  $g(x, y) = k f(x, y) \quad 0 \leq k \leq 1$

HSI color space

$$s_3 = k r_3$$

$$s_1 = r_1, \quad s_2 = r_2$$

RGB color space

$$s_i = k r_i, \quad i=1, 2, 3$$

CMY color space

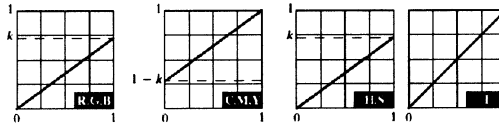
$$s_i = k r_i + (1-k), \quad i=1, 2, 3$$

we didn't show it but  $I = \frac{1}{3} [3 - (c + m + y)] = 1 - \frac{1}{3}(c + m + y)$   
which is why this formula looks a little odd.

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a b  
c d e

**FIGURE 6.31** Adjusting the intensity of an image using color transformations. (a) Original image. (b) Result of decreasing its intensity by 30% (i.e., letting  $k = 0.7$ ). (c)-(e) The required RGB, CMY, and HSI transformation functions. (Original image courtesy of MedData Interactive.)



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Scale each component in RGB

Since  $C = 1 - R$ , etc. This is simply a linear transformation.

These are reversed I is decreased H, S remain the same.



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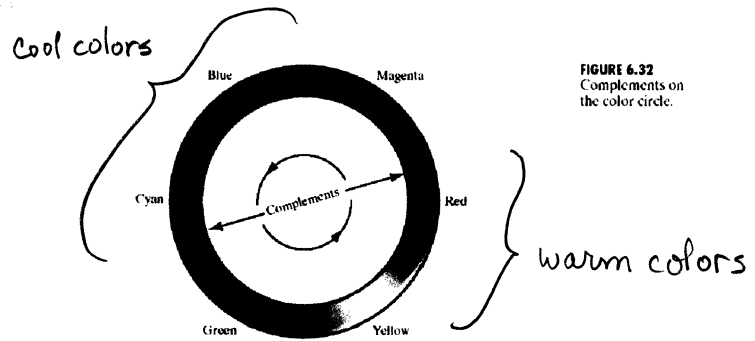
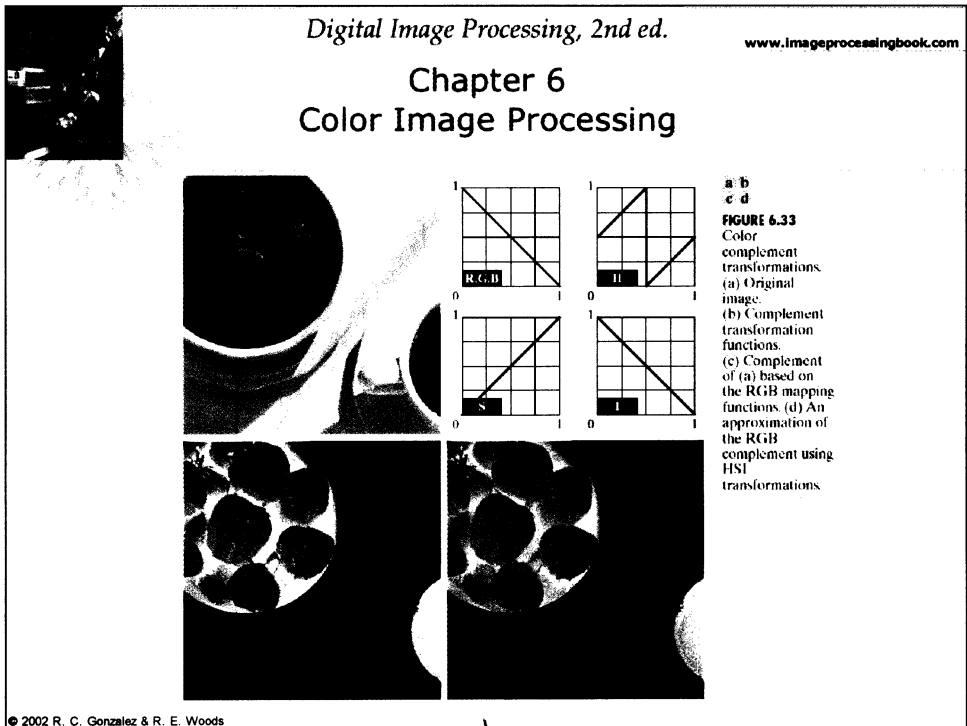


FIGURE 6.32  
Complements on  
the color circle.

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Newton's color circle summarizes  
the additive properties of colors

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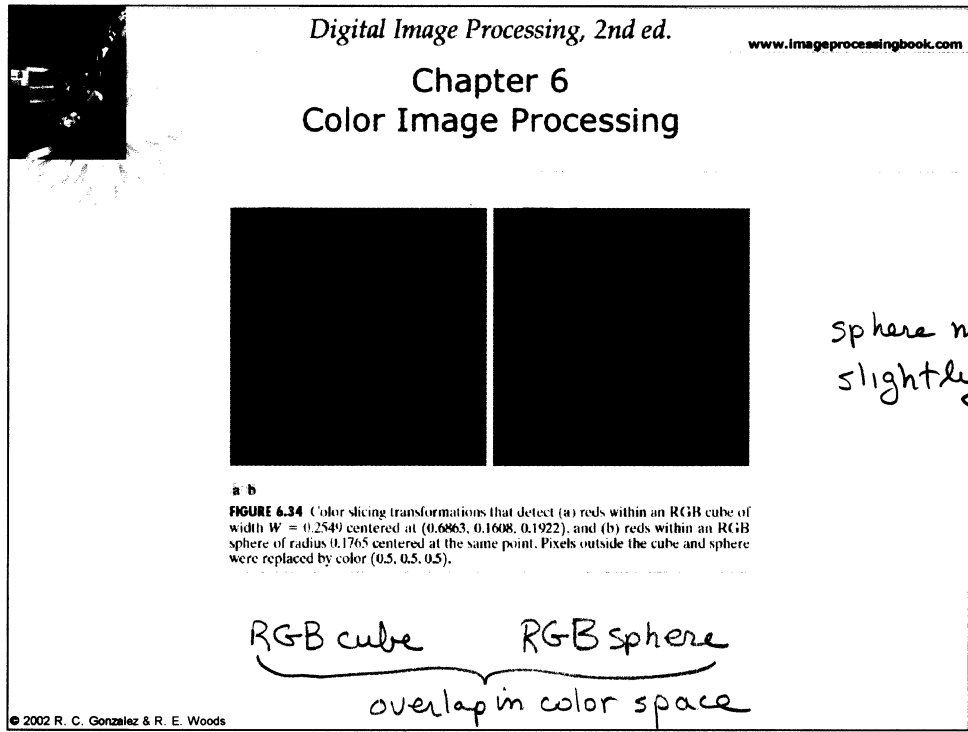
RGB complement

HSI complement

As you decrease each color (R, G, B) its complement becomes evident.

complement not straight forward  
See problem 6.18

- R ↔ cyan
- G ↔ magenta
- B ↔ yellow



color slicing - map colors outside a range of interest to a neutral color

neutral color

cube, hypercube

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

cube of width  $W$  centered at  $(a_1, a_2, a_3)$

sphere

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

$$i = 1, 2, \dots, n$$

#### 6.5.4. Tone/Color correction

need a device-independent color model to get color consistency between monitors & output devices

color management systems

Pantone (used by Adobe)

CIE  $L^*a^*b^*$

$$L^* = 116 \cdot h\left(\frac{Y}{Y_w}\right) - 16$$

$$a^* = 500 \left[ h\left(\frac{X}{X_w}\right) - h\left(\frac{Y}{Y_w}\right) \right]$$

$$b^* = 200 \left[ h\left(\frac{Y}{Y_w}\right) - h\left(\frac{Z}{Z_w}\right) \right]$$

$$\text{where } h(q) = \begin{cases} \sqrt[3]{q} & q > 0.008856 \\ 7.787q + \frac{16}{116} & q \leq 0.008856 \end{cases}$$

$X_w, Y_w, Z_w$  — reference white tristimulus,

perfectly diffused illuminated with CIE D65 light.  
(this is defined to be daylight)

$X, Y, Z$  — R, G, B tristimulus values

similar to HSI  
by separating color from  
intensity

lightness

Red - Green

Green - Blue

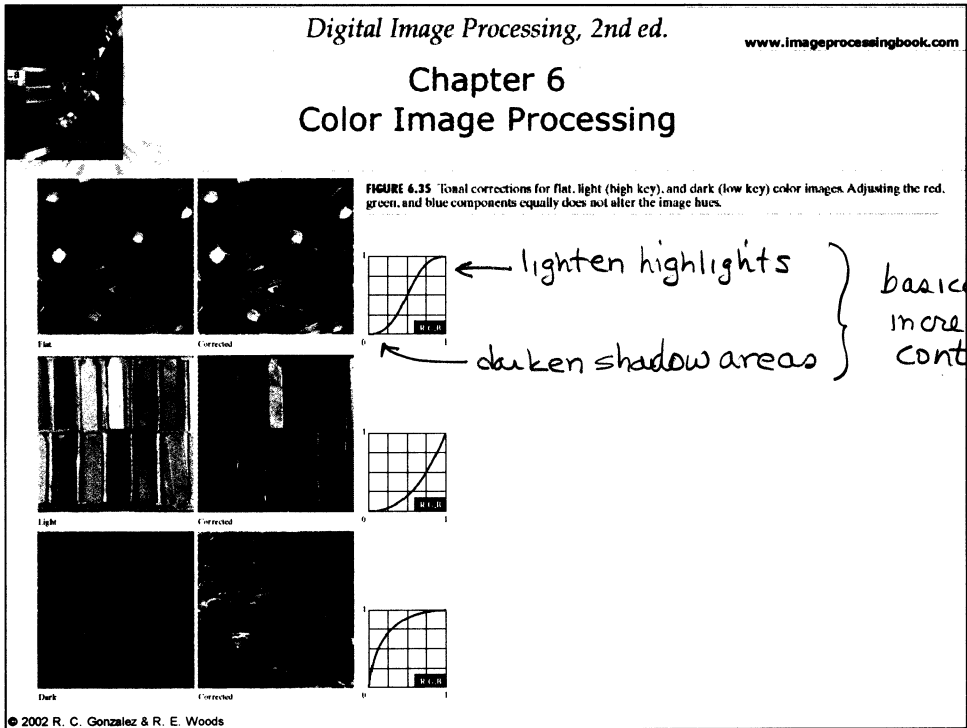
$L^*a^*b^*$  is

colorimetric — colors perceived as identical have identical values

perceptually uniform — color differences are perceived uniformly

device independent

# Chapter 6 Color Image Processing



correction

flat

light

dark

tonal range - key type

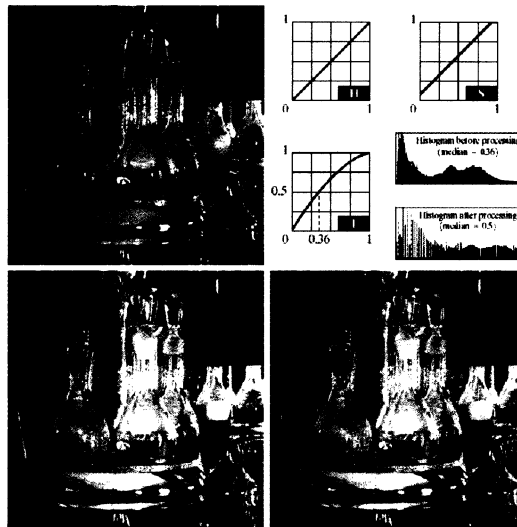
high-key - most information at high (bright) intensities

low-key - " " at low intensities

middle-key - " " at intermediate intensities



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**FIGURE 6.37** Histogram equalization (followed by saturation adjustment) in the HSI color space.

before equalization

histogram equalization of intensity only.

Increase image saturation slightly (after equalization) to make colors look better.

How can you apply histogram equalization to a color image?  
Don't equalize colors independently,  
spread color intensities such as in HSI space.

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FIGURE 6.38  
(a) RGB image.  
(b) Red component image.  
(c) Green component.  
(d) Blue component.

RGB components

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We just considered  
pixel transforms

The next level of processing is  
neighborhood processing  
such as smoothing and sharpening

Consider averaging

$$\bar{c}(x,y) = \frac{1}{K} \sum_{(x,y) \in S_{xy}} c(x,y) \quad K = \# \text{ of pixels}$$

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



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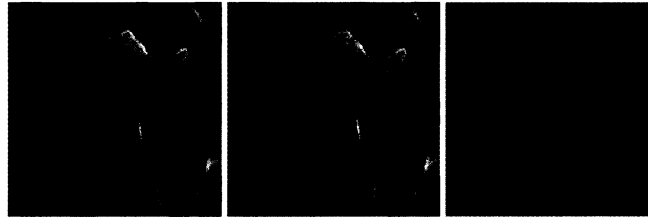


a b c

**FIGURE 6.39** HSI components of the RGB color image in Fig. 6.38(a). (a) Hue. (b) Saturation. (c) Intensity.

HSI components of previous picture.

## Chapter 6 Color Image Processing



a b c

**FIGURE 6.40** Image smoothing with a  $5 \times 5$  averaging mask. (a) Result of processing each RGB component image. (b) Result of processing the intensity component of the HSI image and converting to RGB. (c) Difference between the two results.

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Image smoothing using a  $5 \times 5$  mask.

- (a) smoothing each color plane independently
- (b) smoothing I (intensity) component of HSI image and conversion to RGB. This keeps color accurate.
- (c) No data on how this difference image was computed. Several possibilities

## Chapter 6 Color Image Processing



a b c

FIGURE 6.41 Image sharpening with the Laplacian. (a) Result of processing each RGB channel. (b) Result of processing the intensity component and converting to RGB. (c) Difference between the two results.

processing  
each RGB  
color plane

processing  
only HSI  
intensity  
plane

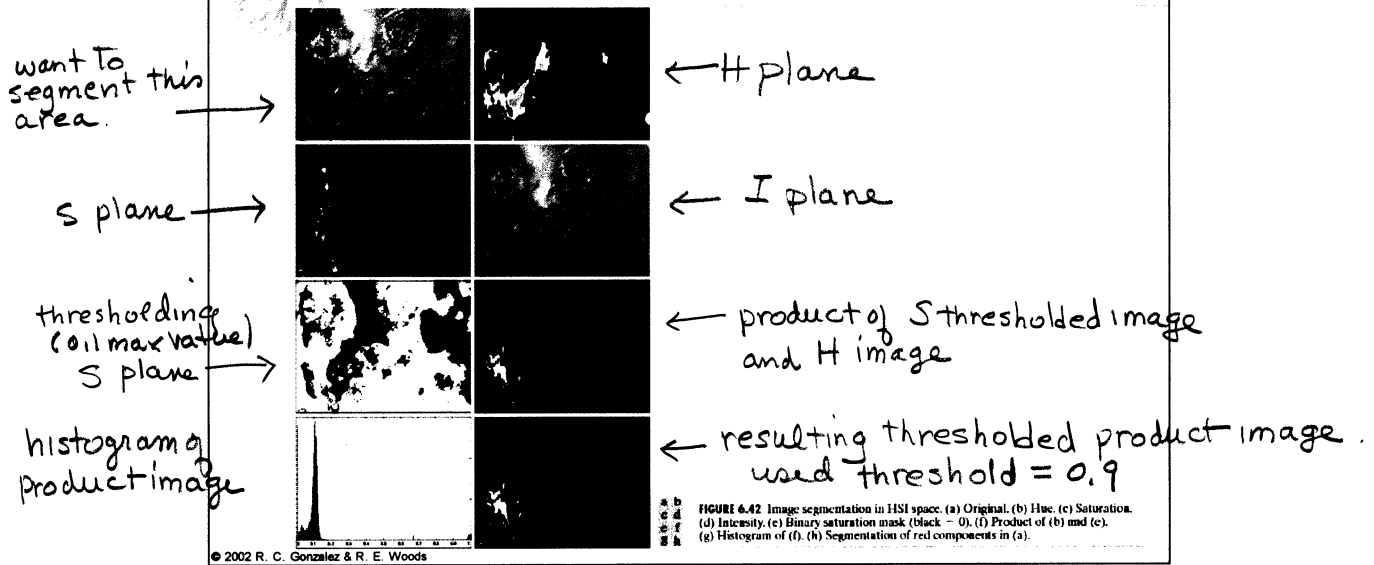
difference (?)  
Image

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Color image sharpening using Laplacian

$$\nabla^2 [c(x,y)] = \begin{bmatrix} \nabla^2 R(x,y) \\ \nabla^2 G(x,y) \\ \nabla^2 B(x,y) \end{bmatrix}$$

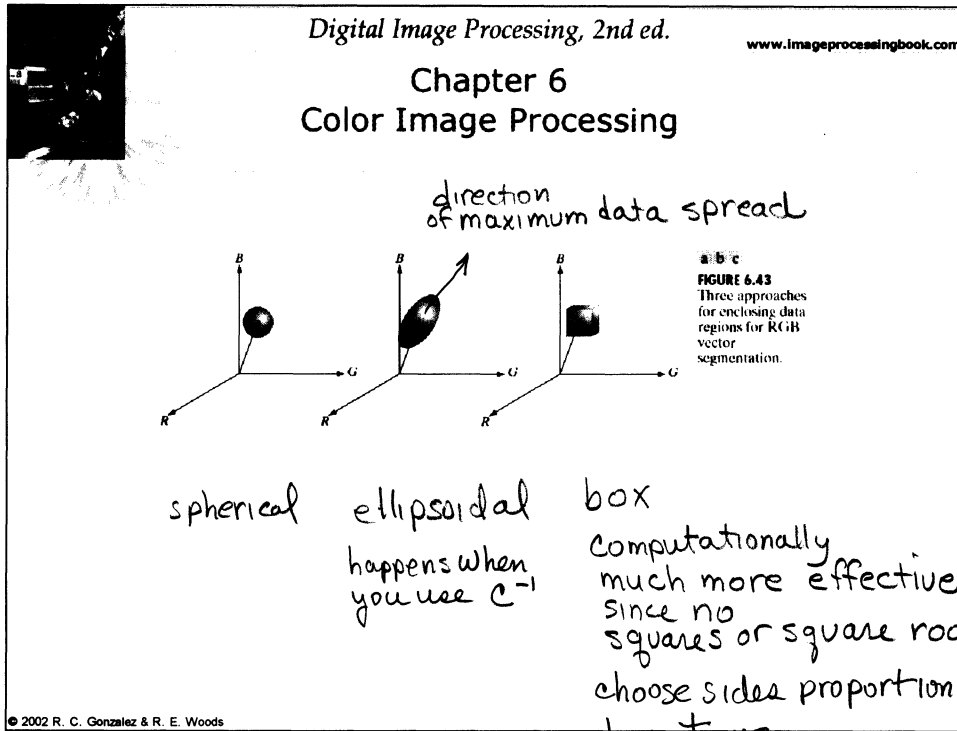
# Chapter 6 Color Image Processing



Segmentation (Chap. 10 topic)

If we want to segment an image based on color the hue (H) image is the most natural to use.

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Segmentation divides an image into constituent regions or objects.

Segmentation usually works better in RGB space than in HSI space.

Objective - classify each color pixel as having a color in the specified range or not

Use Euclidean distance

$$D(\underline{z}, \underline{a}) = \|\underline{z} - \underline{a}\| = \sqrt{(\underline{z} - \underline{a})^T (\underline{z} - \underline{a})}$$

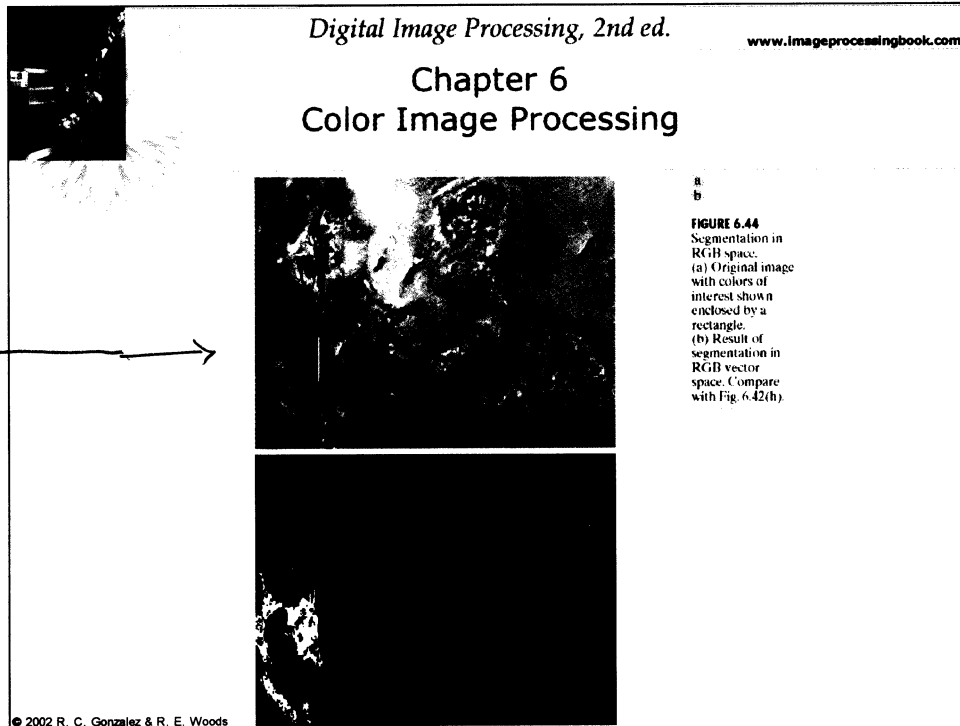
$$D(\underline{z}, \underline{a}) = \sqrt{(z_R - a_R)^2 + (z_G - a_G)^2 + (z_B - a_B)^2}$$

A generalized distance measure is

$$D(\underline{z}, \underline{a}) = \sqrt{(\underline{z} - \underline{a})^T C^{-1} (\underline{z} - \underline{a})}$$

↑  
covariance matrix

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Same image and goal as figure 6.42 but done in RGB space.

Procedure

1. Compute mean and standard deviations of the color contained in the sample rectangle.

mean  $\underline{a}$   
std deviations  $\sigma_R, \sigma_G, \sigma_B$

2. using box in color space

$$a_R \pm 1.25 \sigma_R$$

$$a_G \pm 1.25 \sigma_G$$

$$a_B \pm 1.25 \sigma_B$$

classify pixel as white if inside this box,  
or white outside

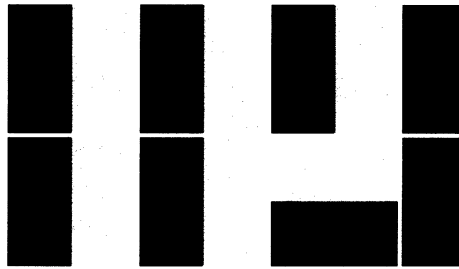
Compare results with 6.42(h)

# Chapter 6 Color Image Processing



RGB edges  
aligned

RGB edges  
not aligned



a b c d  
e f g h

FIGURE 6.45 (a)-(c) R, G, and B component images and (d) resulting RGB color image. (f)-(g) R, G, and B component images and (h) resulting RGB color image.

If we simply add the gradients they would be the same at the center point. Intuitively they cannot be the same.

Computing edges in individual color planes vs. computing directly in color space.

## Alternative vector definition

S. Di Zenzo (1986)

A Note on the Gradient of a Multi-Image  
Computer Vision, Graphics and Image Processing  
Vol. 33, pp. 116-125

We want to define the gradient (magnitude and direction) of the vector  $\underline{c}(x, y)$

For a scalar function  $f(x, y)$  the gradient is a vector pointing in the direction of maximum rate of change of  $f$  at  $(x, y)$

$\hat{r}, \hat{g}, \hat{b}$  be unit vectors along the R, G, B axis of a RGB color space

$$\begin{aligned} \underline{u} &= \frac{\partial R}{\partial x} \hat{r} + \frac{\partial G}{\partial x} \hat{g} + \frac{\partial B}{\partial x} \hat{b} \\ \underline{v} &= \frac{\partial R}{\partial y} \hat{r} + \frac{\partial G}{\partial y} \hat{g} + \frac{\partial B}{\partial y} \hat{b} \end{aligned} \left. \vphantom{\begin{aligned} \underline{u} \\ \underline{v} \end{aligned}} \right\} \begin{array}{l} \text{These can be} \\ \text{computed using a} \\ \text{Sobel operator.} \end{array}$$

further define

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

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

$$g_{xy} = \underline{u} \cdot \underline{v} = \underline{u}^T \underline{v} = \frac{\partial R}{\partial x} \frac{\partial R}{\partial y} + \frac{\partial G}{\partial x} \frac{\partial G}{\partial y} + \frac{\partial B}{\partial x} \frac{\partial B}{\partial y}$$

The maximum rate of change of  $\underline{c}(x, y)$  at  $(x, y)$  is in the direction given by

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

and the rate of change in the direction  $\theta(x, y)$  is given by

$$F(\theta) = \sqrt{\frac{1}{2}(g_{xx} + g_{yy}) + (g_{xx} - g_{yy})\cos 2\theta + 2g_{xy}\sin 2\theta}$$

$\theta$  is given in two orthogonal directions. One is a maximum for  $F$  and the other is a minimum.



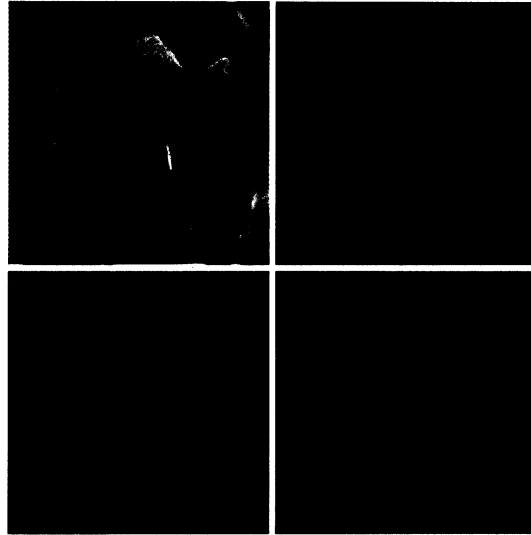
# Chapter 6 Color Image Processing



original image

a b  
c d  
**FIGURE 6.46**  
(a) RGB image.  
(b) Gradient computed in RGB color vector space.  
(c) Gradients computed on a per-image basis and then added.  
(d) Difference between (b) and (c).

colored edges computed by adding derivatives in each color plane together



colored edges computed using vector method in RGB color space

difference between (b) and (c)

## Chapter 6 Color Image Processing



a b c

**FIGURE 6.47** Component gradient images of the color image in Fig. 6.46. (a) Red component, (b) green component, and (c) blue component. These three images were added and scaled to produce the image in Fig. 6.46(c).

Individual RGB gradient images  
Added and scaled to produce RGB gradient image.

## Chapter 6 Color Image Processing

a b  
c d  
**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).]



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Added Gaussian noise to the R, G, and B color planes.  
RGB image with noise shown in (d).

## Chapter 6 Color Image Processing



a b c

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

significantly degraded  
due to non-linearity of  
cosine and min in  
transformations

intensity is average  
which tends to  
reduce noise.

HSI components of the noisy RGB image.

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RGB image  
Salt & Pepper noise  
in Green channel



**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.

noise spreads to all HSI component images

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original  
image



**FIGURE 6.51**  
Color image  
compression.  
(a) Original RGB  
image. (b) Result  
of compressing  
and  
decompressing  
the image in (a).

Compressed and  
decompressed  
using JPEG 2000  
slight blurring  
due to lossy technique