



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

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

pseudocolor - color assigned to a intensity value or range



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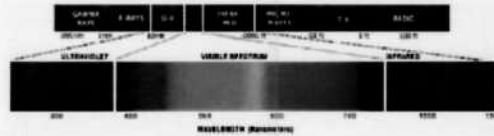


FIGURE 6.2 Wavelengths comprising the visible range of the electromagnetic spectrum. (Courtesy of the General Electric Co., Lamp Business Division.)

blue

red

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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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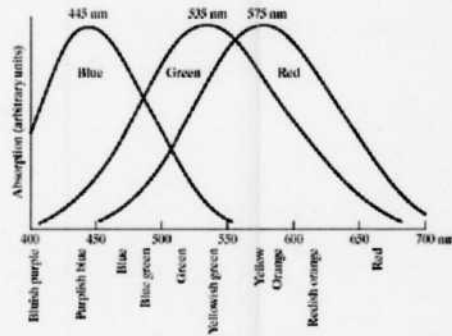


FIGURE 6.3 Absorption of light by the red, green, and blue cones in the human eye as a function of wavelength.

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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

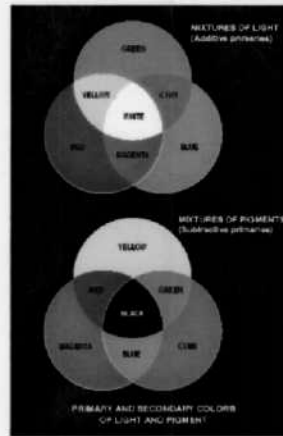


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(primary)
Additive
colors
transmission

Subtractive
colors
(secondary)
reflection

absorbs a primary color and



} red-green-blue
television monitors

} cyan-magenta-yellow
printing
need black in printing

11
12

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

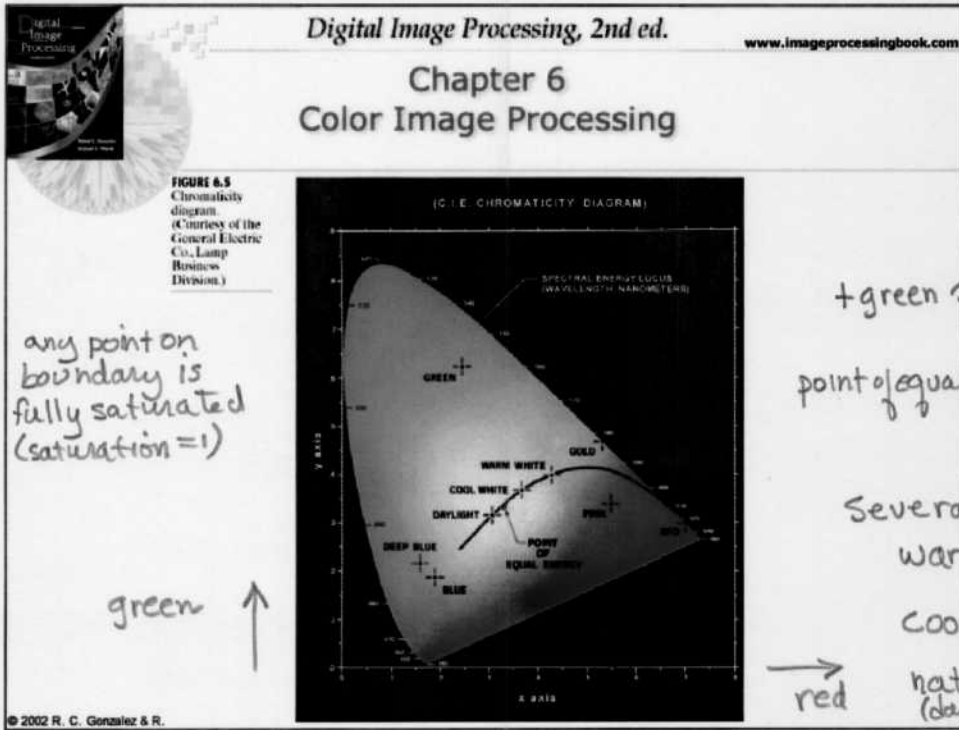
characteristics of color

brightness

chromaticity {

hue - dominant color seen by an observer

saturation - amount of white light mixed with
the color



A color can be specified by its tristimulus values

$$\left. \begin{aligned} \text{red} \quad x &= \frac{X}{x+Y+Z} \\ \text{green} \quad y &= \frac{Y}{x+Y+Z} \\ \text{blue} \quad z &= \frac{Z}{x+Y+Z} \end{aligned} \right\} \text{where } x+y+z=1$$

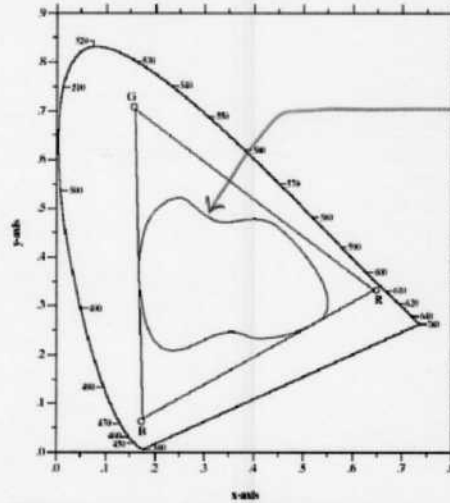
where x, y, z are the amounts of red, green, and blue needed to form a color

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

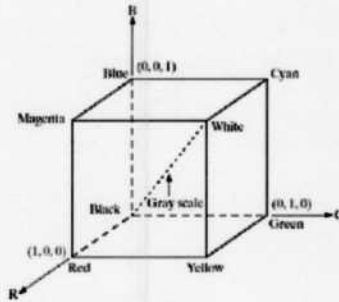
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The RGB points represent the maximum RGB values of an RGB monitor. Since a monitor is an additive process this RGB monitor can create any color within the triangle.



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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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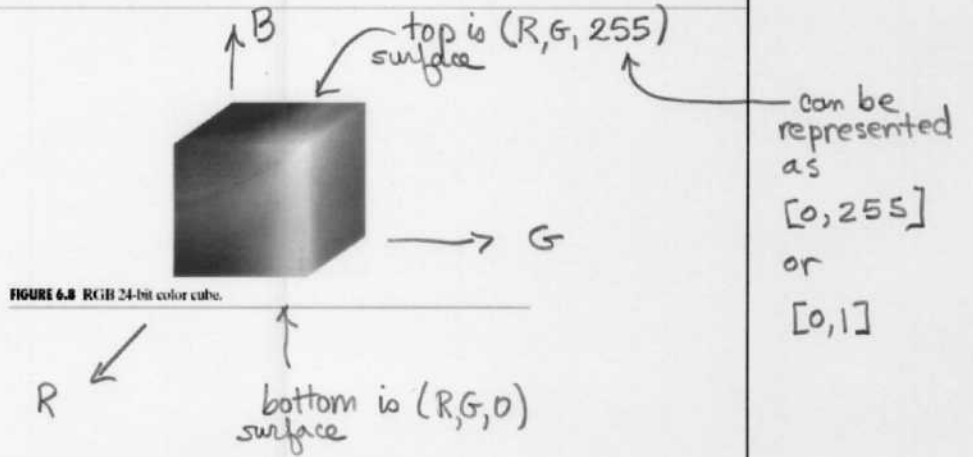
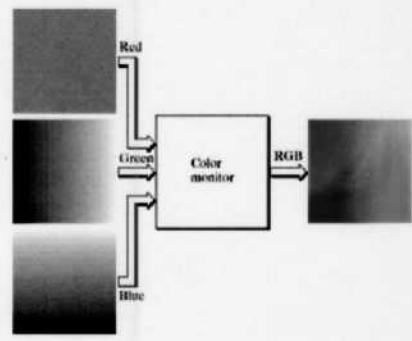


FIGURE 6.8 RGB 24-bit color cube.

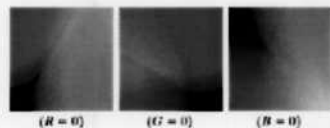


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FIGURE 6.9
(a) Generating the RGB image of the cross-sectional color plane (127, G, B).
(b) The three hidden surface planes in the color cube of Fig. 6.8.



combine color planes into color image



(R, G, B)

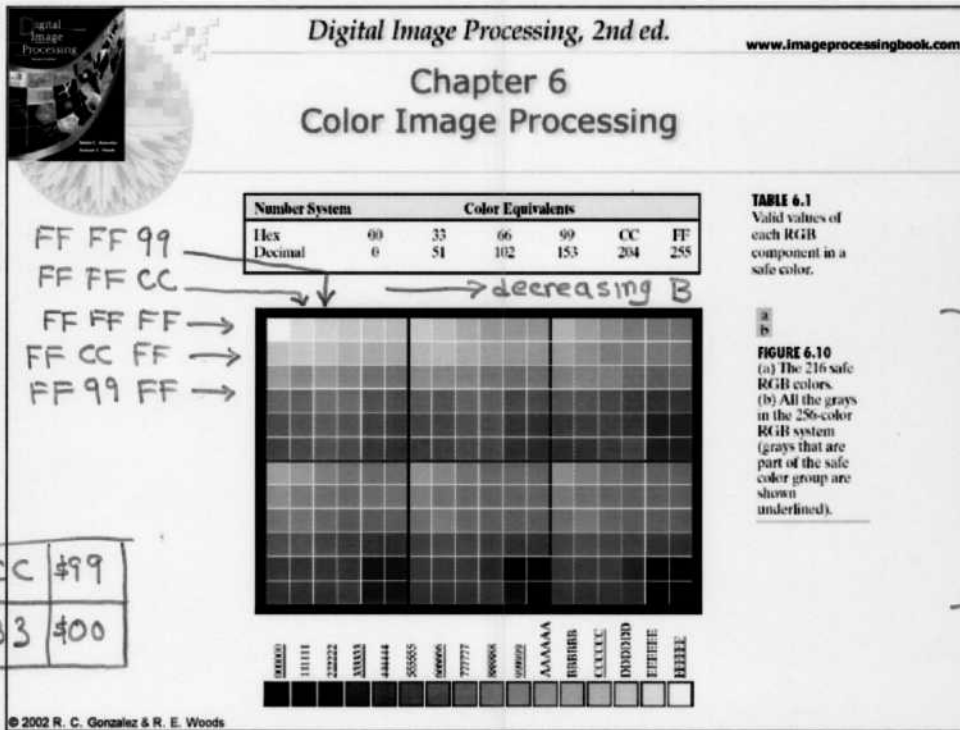
(0, G, B)

(R, 0, B)

(R, G, 0)

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back planes looking inside out



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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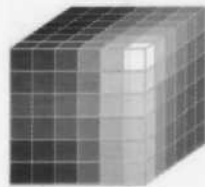


FIGURE 6.11 The RGB safe-color cube.

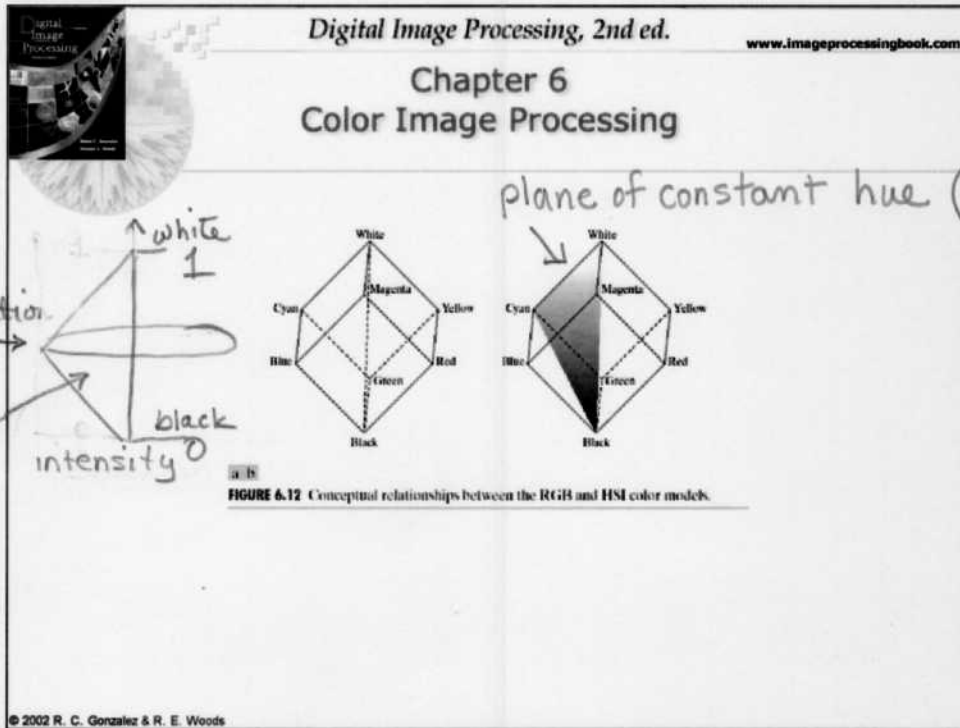
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"safe" colors are only on the surfaces (faces).
No interior colors are "safe".

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.



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

H - hue	}	color information decoupled from intensity
S - saturation		
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.



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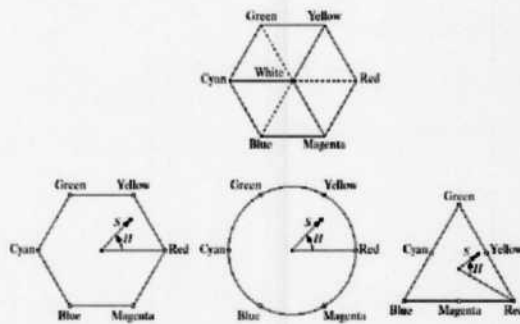
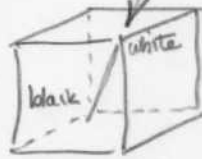


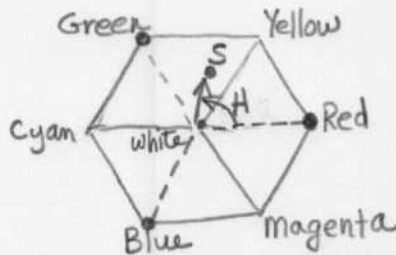
FIGURE 6.13 Hue and saturation in the HSI color model. The dot is an arbitrary color point. The angle from the red axis gives the hue, and the length of the vector is the saturation. The intensity of all colors in any of these planes is given by the position of the plane on the vertical intensity axis.

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looking at diagonal perpendicular planes are hexagonal.

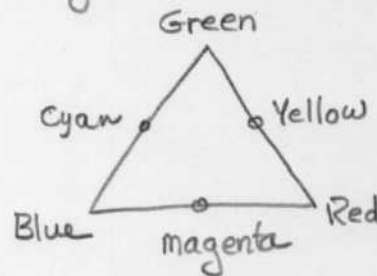
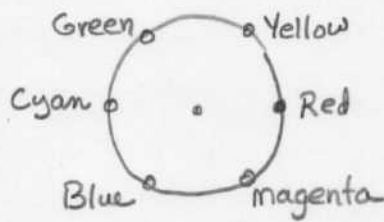


looking along axis



The hue of any point in each plane is its angle from a reference point.
The saturation is the distance from the origin.

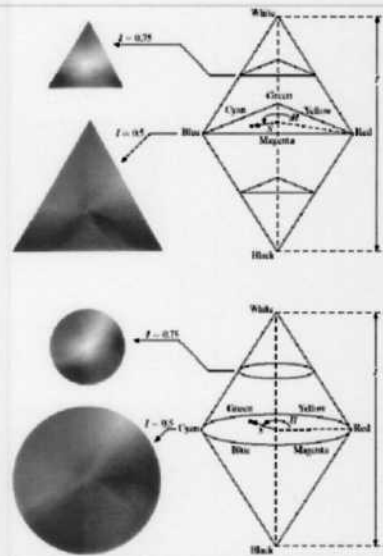
Various simplifications of this hexagon are circles and triangles.





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FIGURE 6.14 The HSI color model based on (a) triangular and (b) circular color planes. The triangles and circles are perpendicular to the vertical intensity axis.



↑ This is the gray diagonal, i.e. the intensity axis.

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

$$\text{Define } \theta = \cos^{-1} \left\{ \frac{\frac{1}{2} [(R-G) + (R-B)]}{\left[\frac{1}{4} [(R-G)^2 + (R-B)(G-B)] \right]^{1/2}} \right\}$$

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

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

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

See book's web site for derivation

Converting colors from HSI to RGB

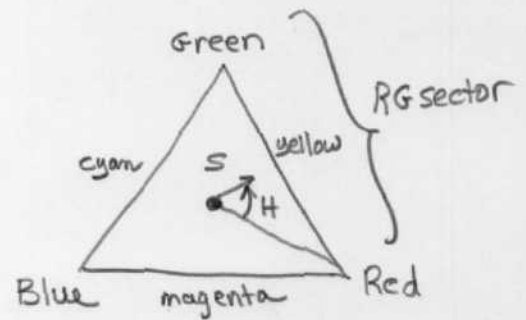
Multiply H by 360° : to convert it back to an angle. H is usually normalized to $[0, 1]$.

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 $H = H - 240^\circ$

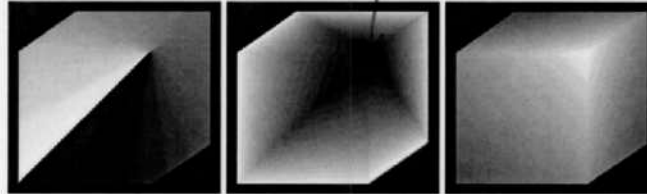
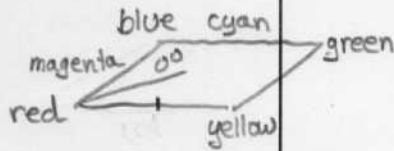
$$G = I(1-s)$$

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

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



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point of least saturation

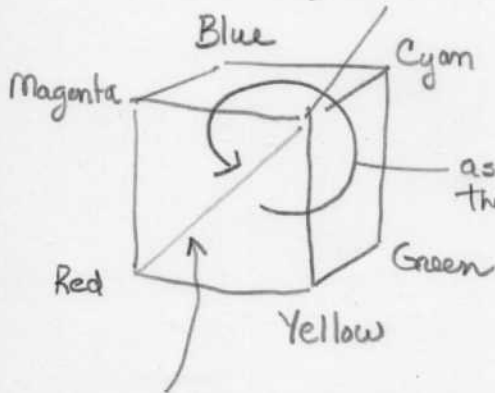
FIGURE 6.15 HSI components of the image in Fig. 6.8. (a) Hue, (b) saturation, and (c) intensity images

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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

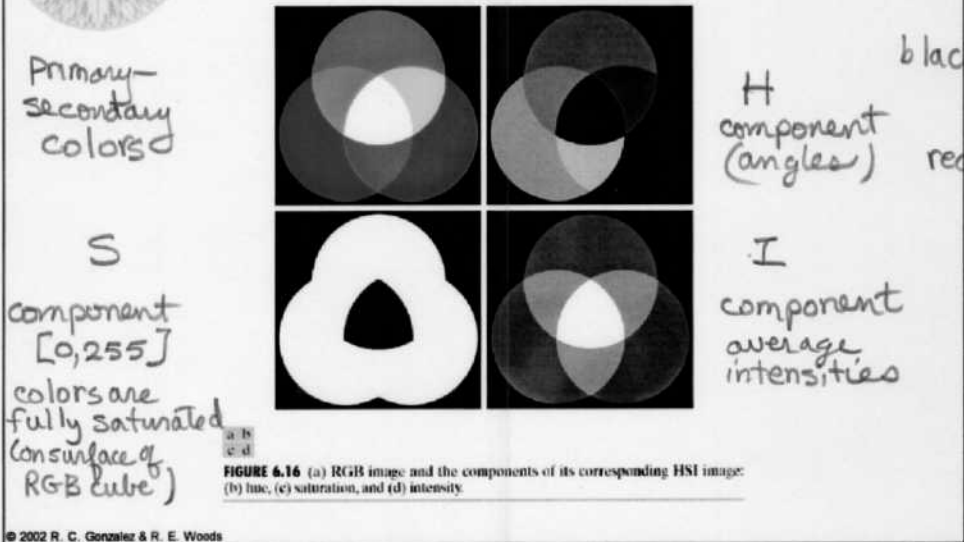


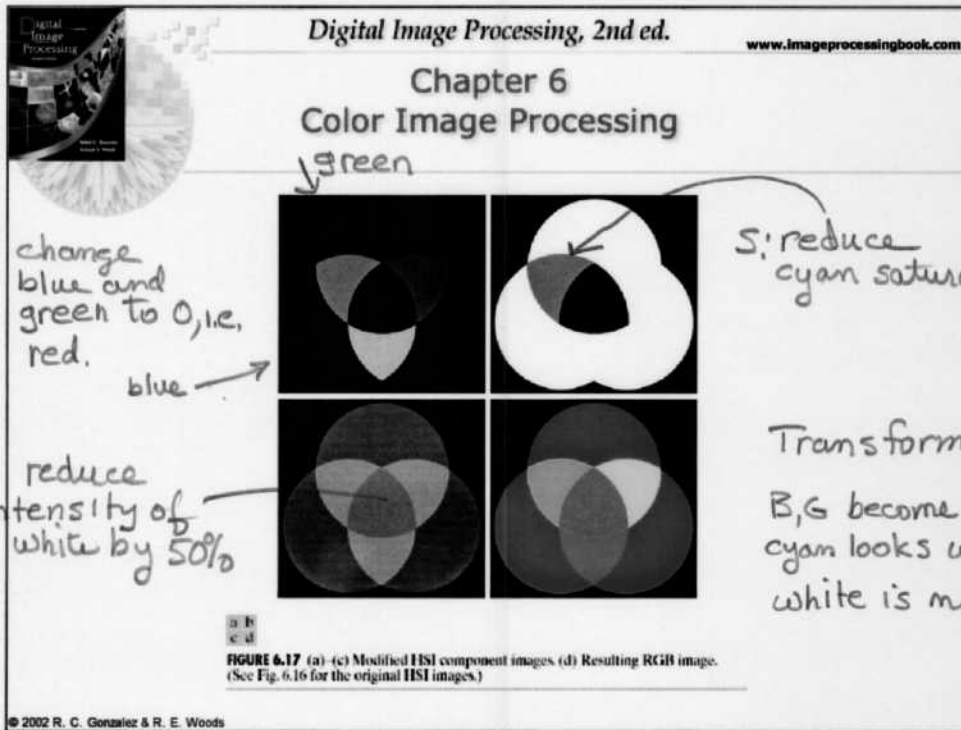
as seen looking back along the gray axis we have a discontinuity from 1 to zero along the red-white line.

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



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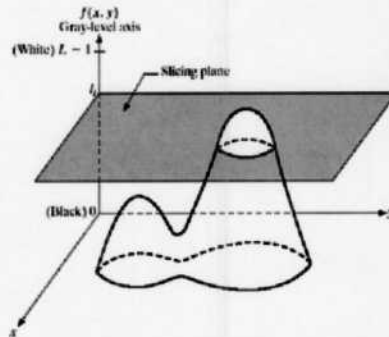


To change color in HSI simply change the hue value and convert back to RGB without changing S and I.

To change saturation modify only S,



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

FIGURE 6.18 Geometric interpretation of the intensity-slicing technique.

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



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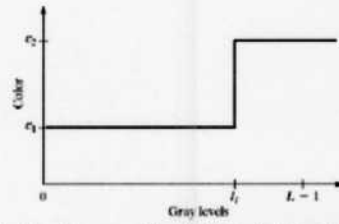
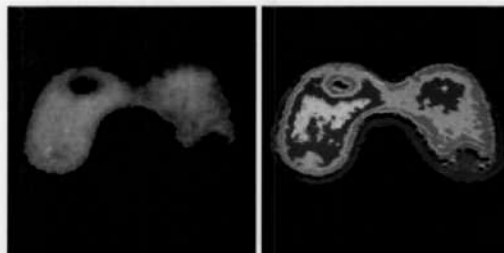


FIGURE 6.19 An alternative representation of the intensity-slicing technique.

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



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(a) (b)

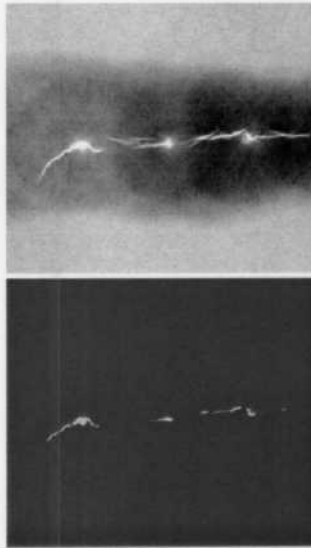
FIGURE 6.20 (a) Monochrome image of the Picker Thyroid Phantom. (b) Result of density slicing into eight colors. (Courtesy of Dr. J. L. Bankenship, Instrumentation and Controls Division, Oak Ridge National Laboratory.)

intensity slicing with multiple color slices



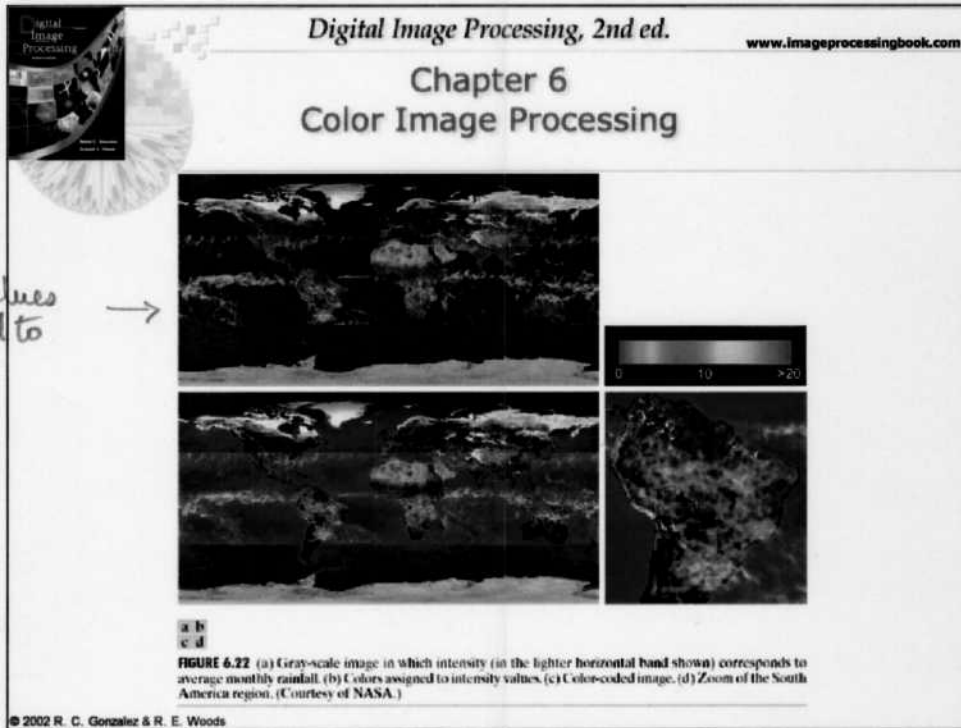
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FIGURE 6.21
(a) Monochrome X-ray image of a weld. (b) Result of color coding. (Original image courtesy of X-TEK Systems, Ltd.)



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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.



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

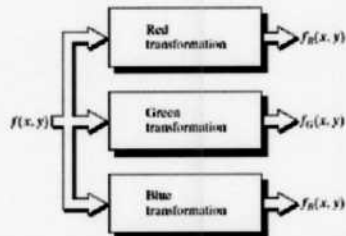
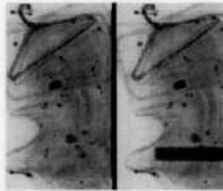


FIGURE 6.23 Functional block diagram for pseudocolor image processing. f_R , f_G , and f_B are fed into the corresponding red, green, and blue inputs of an RGB color monitor.

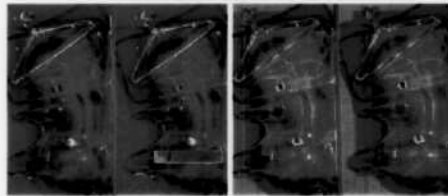
We can use simultaneous
non-linear transforms to drive
a color camera.



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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.)

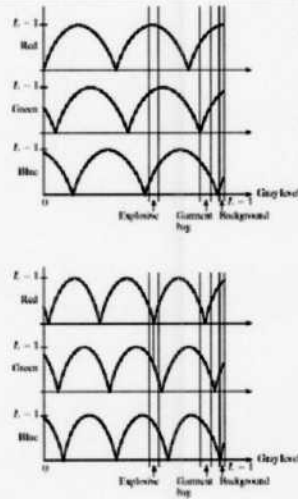
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transform transform
6.25(a) 6.25(b)

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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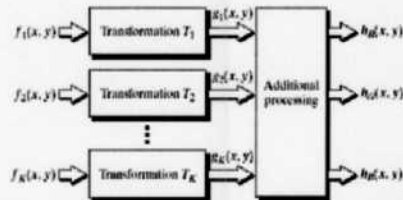


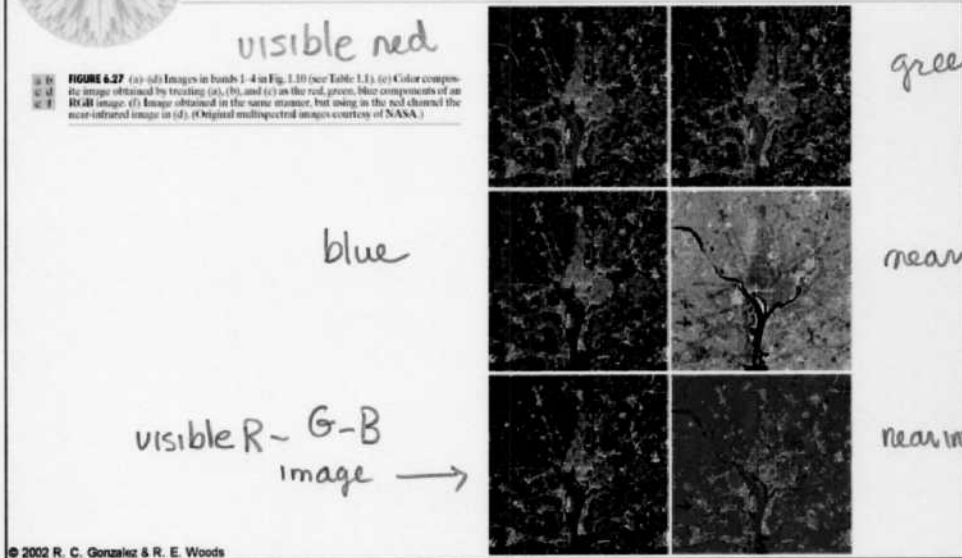
FIGURE 6.26 A pseudocolor coding approach used when several monochrome images are available.

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). (Digital multispectral images courtesy of NASA.)



↑
used visible red
for R

↑
used near-infrared
for R.



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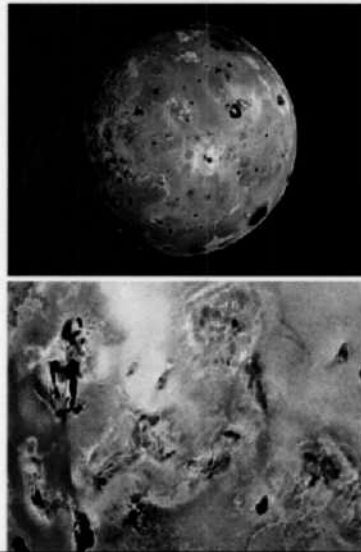


FIGURE 6.28
(a) Pseudocolor
rendition of
Jupiter moon Io.
(b) A close-up
(Courtesy of
NASA.)

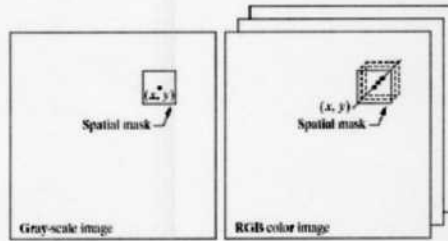
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used a variety of different wavelengths
The newly ejected material is red (different material)
The older material is yellow. (sulfur)



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



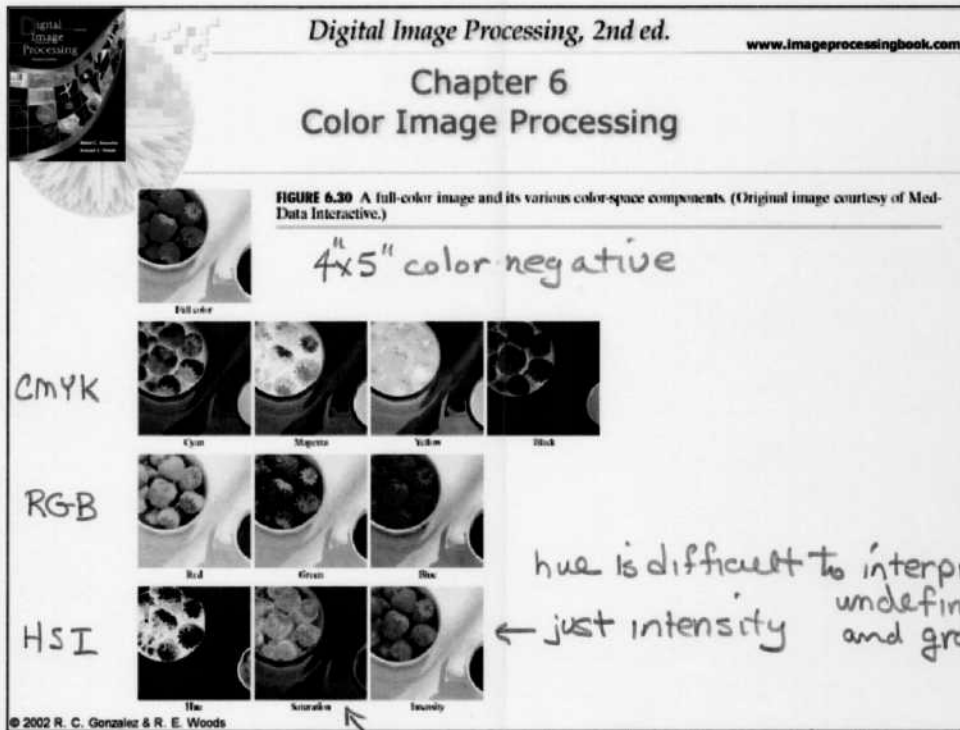
for neighborhood averaging operations are equivalent
 sum and divide all pixels in neighborhood
 sum and divide all the vectors in the neighborhood to get the same result as averaging each color component and combining

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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.

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



hue is difficult to interpret since it's undefined for white, black and gray
 ← just intensity

strawberries are highly saturated

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
 $T_i =$ 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

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

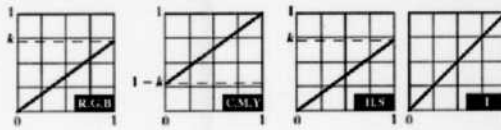
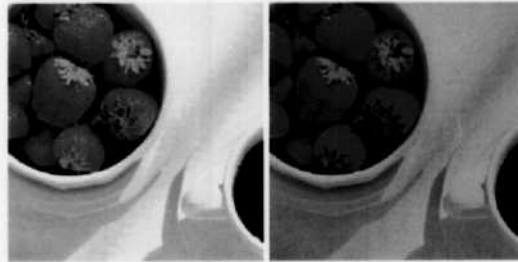
We didn't show it but $I = \frac{1}{3} [3 - (c+m+y)] = 1 - \frac{1}{3}(c+m+y)$
 which is why this formula looks a little odd.



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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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$k=0.7$

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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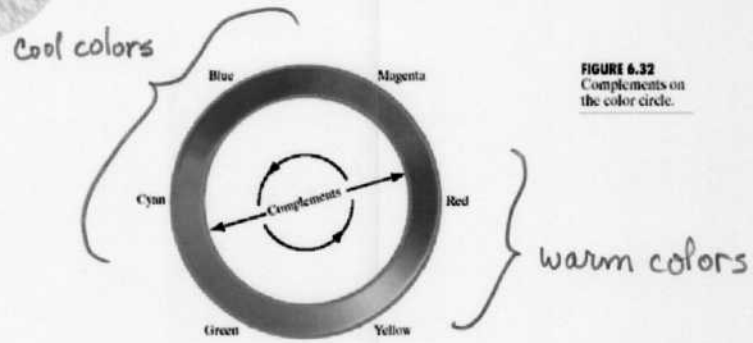


FIGURE 6.32
Complements on
the color circle.

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Newton's color circle summarizes
the additive properties of colors



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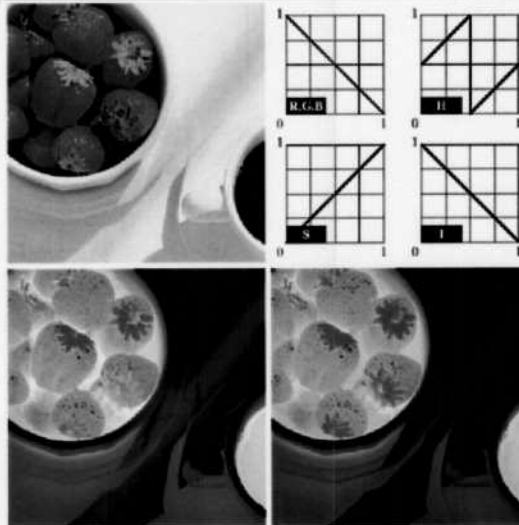


FIGURE 6.33
Color complement transformations. (a) Original image. (b) Complement transformation functions. (c) Complement of (a) based on the RGB mapping functions. (d) An approximation of the RGB complement using HSI transformations.

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RGB complement

HSI complement.

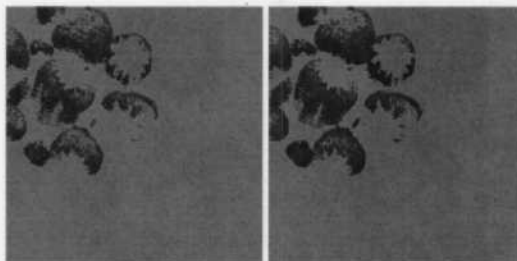
As you decrease each color (R, G, B) its complement becomes evident.

- R ↔ cyan
- G ↔ magenta
- B ↔ yellow

complement not straight forward
See problem 6.18



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sphere might be slightly better

FIGURE 6.34 Color slicing transformations that detect (a) reds within an RGB cube of width $W = 0.2549$ centered at $(0.6963, 0.1608, 0.1922)$, 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 $(0.5, 0.5, 0.5)$.

RGB cube RGB sphere
└──────────┬──────────┘
 overlap in color space

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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

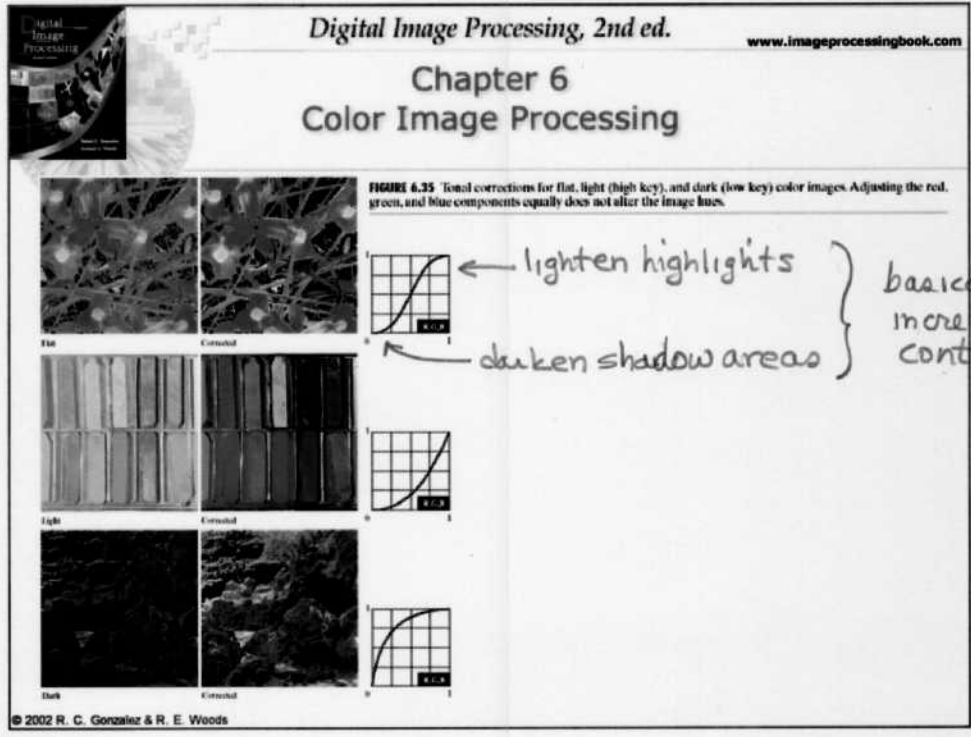
Green - Blue

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

colorimetric - colors perceived as identical have identical values

perceptually uniform - color differences are perceived uniformly

device independent



tonal range - key type

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

low-key - " " at low intensities

middle-key - " " at intermediate intensities

Digital Image Processing, 2nd ed. www.imageprocessingbook.com

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FIGURE 6.36 Color balancing correction for CMYK color images.

CMYK images.

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Easiest way to evaluate color imbalance in an image is to analyze a known color, such as whites or skin.

Simple transformations to either boost or lighten a CMYK image.



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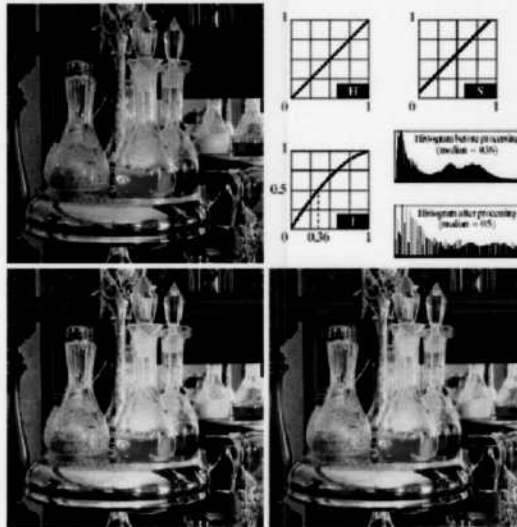


FIGURE 6.37
Histogram equalization (followed by saturation adjustment) in the HSI color space.

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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a b
c d
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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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.



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a b c

FIGURE 6.40 Image smoothing with a 5×5 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 5×5 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



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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

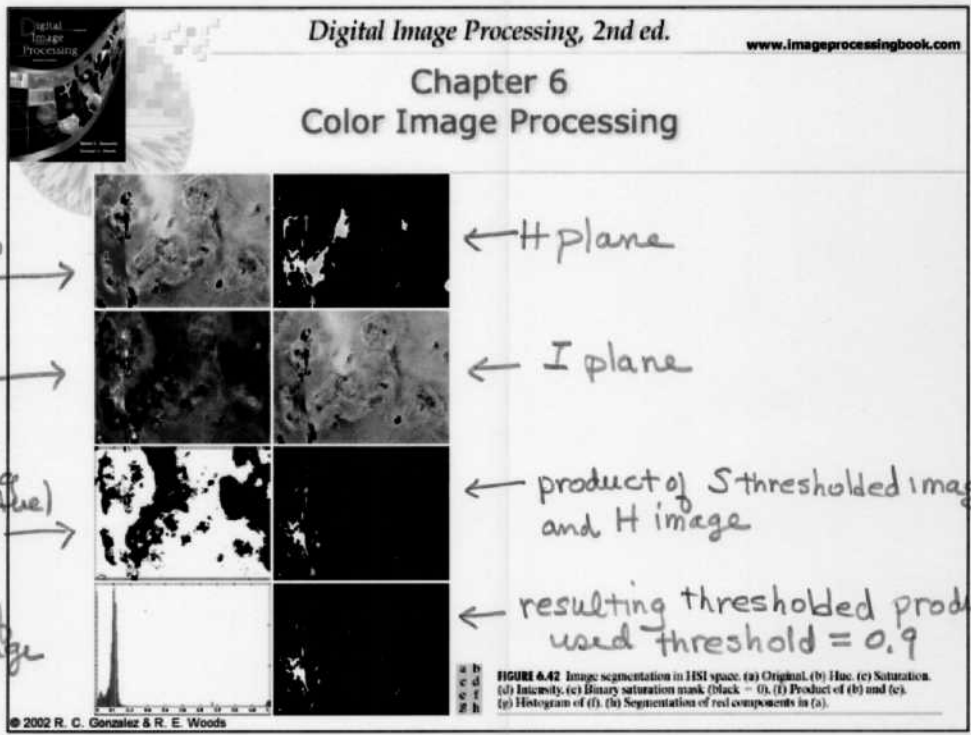
{
 processing
 only HSI
 intensity
 plane

{
 difference (?)
 image

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Color image sharpening using Laplacian

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



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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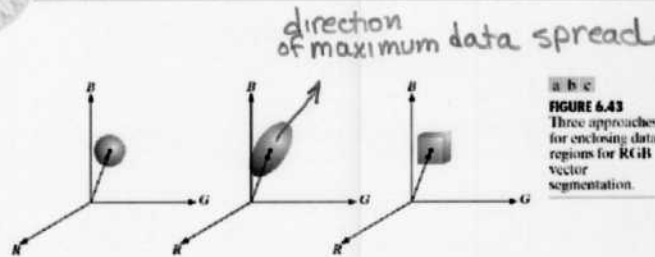


FIGURE 6.43
Three approaches for enclosing data regions for RGB vector segmentation.

spherical ellipsoidal box
 happens when you use C^{-1} computationally much more effective since no squares or square roots choose sides proportional to standard deviations.

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

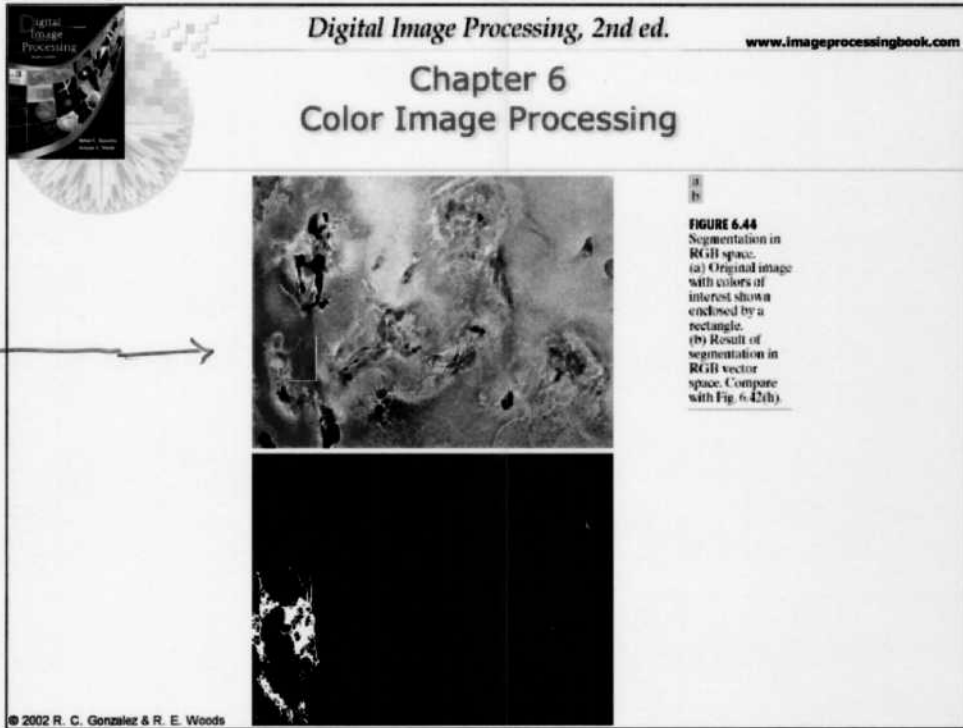
$$D(\underline{z}, \underline{a}) = \|\underline{z} - \underline{a}\| = \sqrt{(\underline{z} - \underline{a})^T (\underline{z} - \underline{a})}$$

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

A generalized distance measure is

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

↑ covariance matrix



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

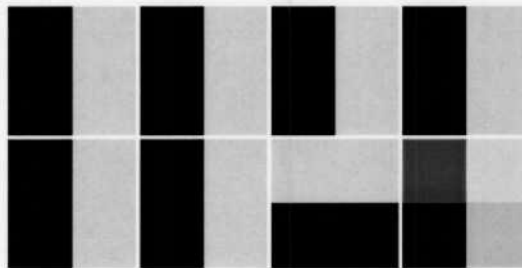
Compare results with 6.42(h)



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RGB edges
aligned

RGB edges
not aligned



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

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.

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

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

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

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

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



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original image

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).



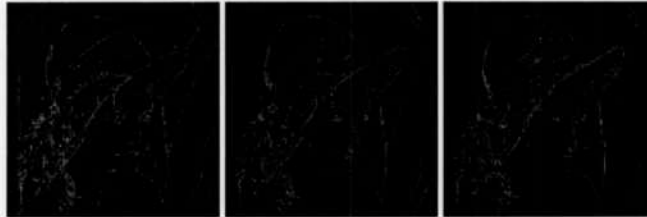
colored edges computed by adding derivatives in each color plane together

colored edges computed using vector method in RGB color space

difference between (b) and (c)



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(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.



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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).



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(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.

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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.

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noise spreads to all HSI component images



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original
image



a b
e d
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