

Chapter 6 Color Image Processing

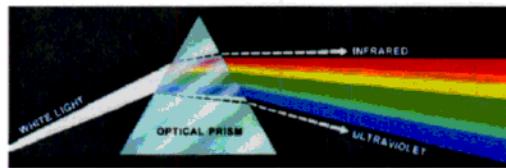


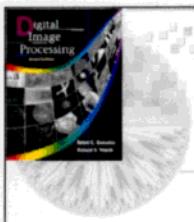
FIGURE 6.1 Color spectrum seen by passing white light through a prism. (Courtesy of the General Electric Co., Lamp Business Division.)

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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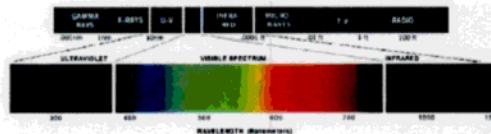


FIGURE 6.2 Wavelengths comprising the visible range of the electromagnetic spectrum.
(Courtesy of the General Electric Co., Lamp Business Division.)

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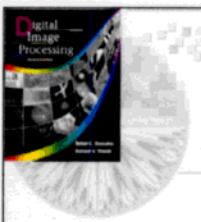
physical units

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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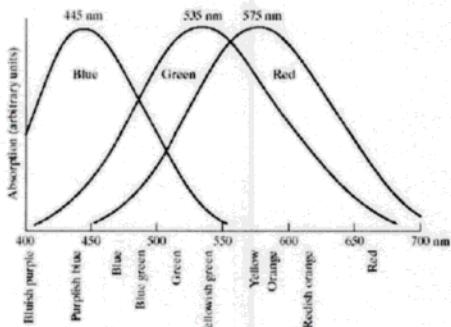


FIGURE 6.3 Absorption of light by the red, green, and blue cones in the human eye as a function of wavelength.

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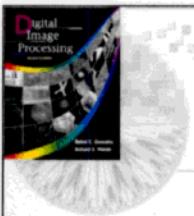
Spectral response of human eye

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

a
b

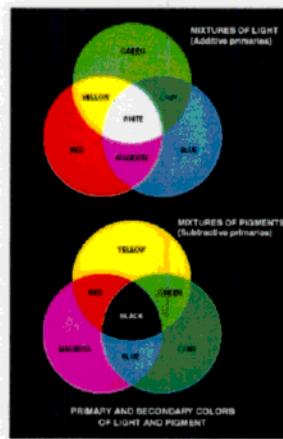


FIGURE 6.4 Primary and secondary colors of light and pigments. (Courtesy of the General Electric Co., Lamp Business Division.)

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} red - green - blue
television monitors

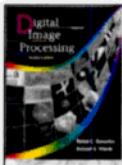
} cyan - magenta - yellow
printing
need black in printing

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

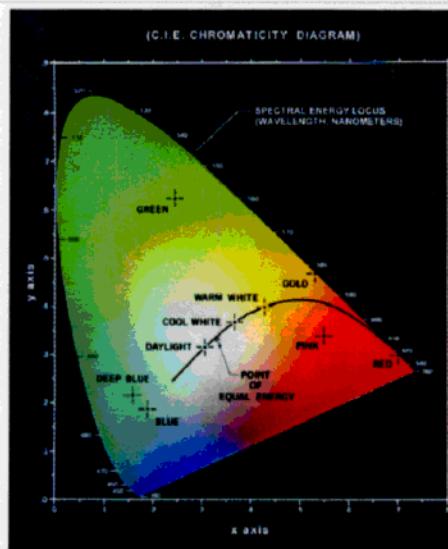


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FIGURE 6.5
Chromaticity diagram.
(Courtesy of the
General Electric
Co., Lamp
Business
Division.)

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

green ↑



+ green ≈ (.25red, .62green)

point of equal energy - white
(.33red, .33green)

Several kinds of white
warm - more red

cool -
natural -
(daylight)

A color can be specified by its tristimulus values

$$\text{red} \quad x = \frac{x}{x+y+z}$$

$$\text{green} \quad y = \frac{y}{x+y+z}$$

$$\text{blue} \quad z = \frac{z}{x+y+z}$$

}

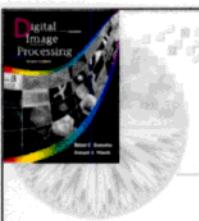
where $x+y+z=1$

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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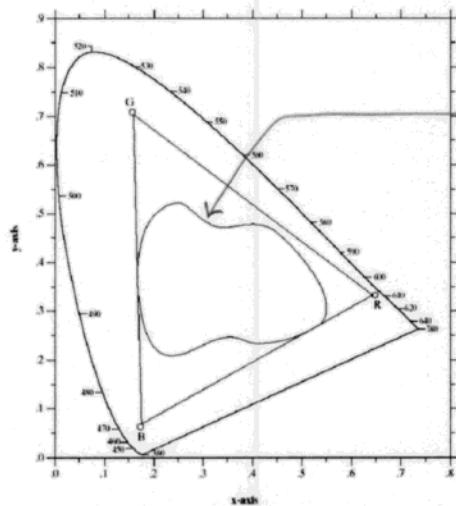
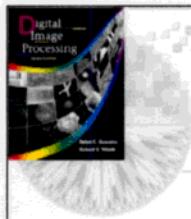


FIGURE 6.6 Typical color gamut of color monitors (triangle) and color printing devices (irregular region).

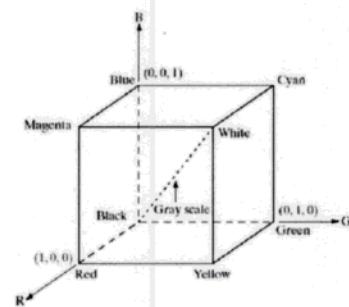
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The RGB points represent the maximum RGB values of an RGB monitor. Since a monitor is an additive process this RGB monitor can create any color within the triangle.



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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).

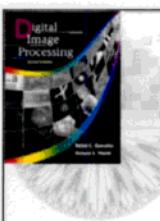


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RGB color model

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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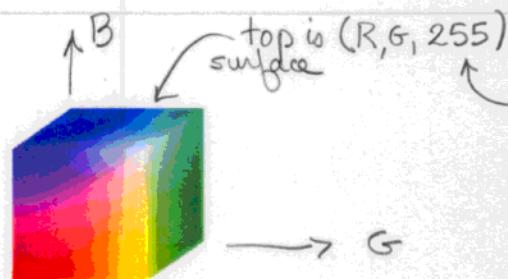
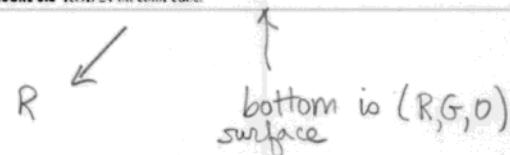
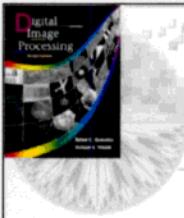


FIGURE 6.8 RGB 24-bit color cube.

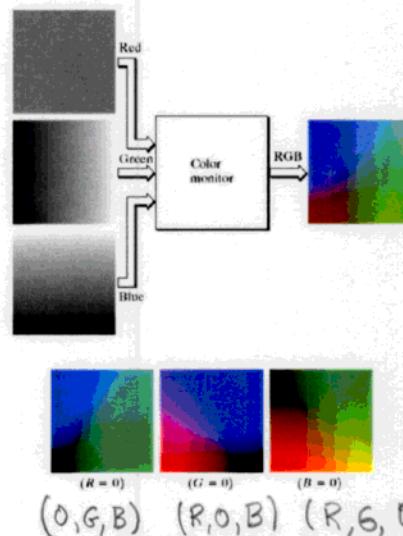


can be represented as
 $[0, 255]$
or
 $[0, 1]$



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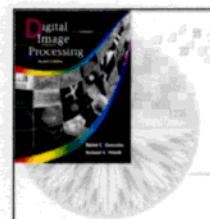
a
b
FIGURE 6.9
(a) Generating the RGB image of the cross-sectional color plane (127, G, B).
(b) The three hidden surface planes in the color cube of Fig. 6.8.



combine color planes
into color image

back planes
looking inside out

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decreasing G

Red:

#FF	\$00	C	\$99
\$66	\$33		\$00

FF FF 99
FF FF CC
FF FF FF →
FF CC FF →
FF 99 FF →

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

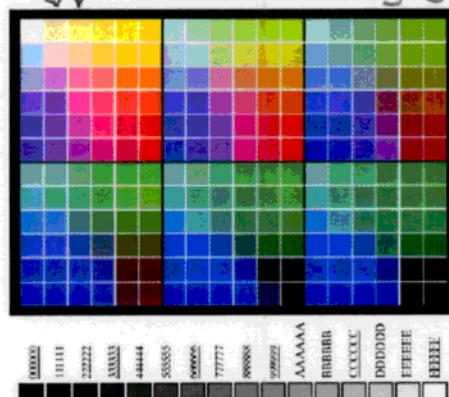


TABLE 6.1
Valid values of each RGB component in a safe color.

a

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).

b

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

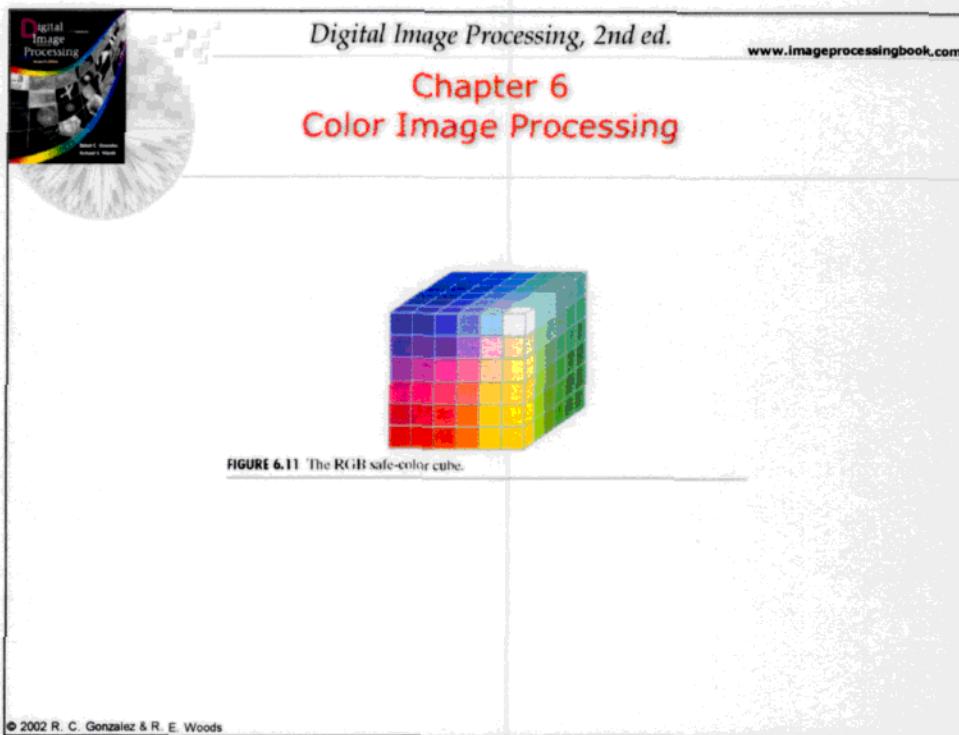


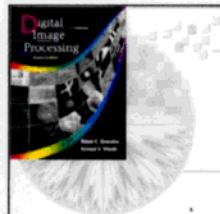
FIGURE 6.11 The RGB safe-color cube.

"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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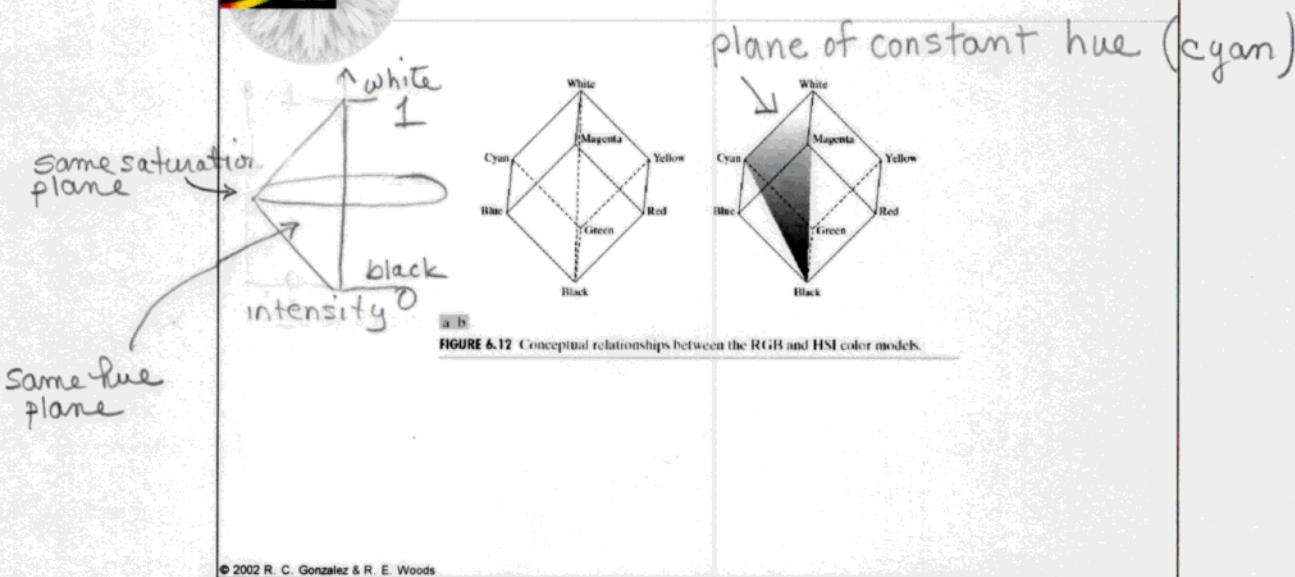


FIGURE 6.12 Conceptual relationships between the RGB and HSI color models.

a b

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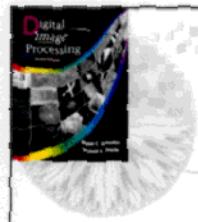
Although RGB is good for generating colors it is not good for describing colors as humans interpret them.

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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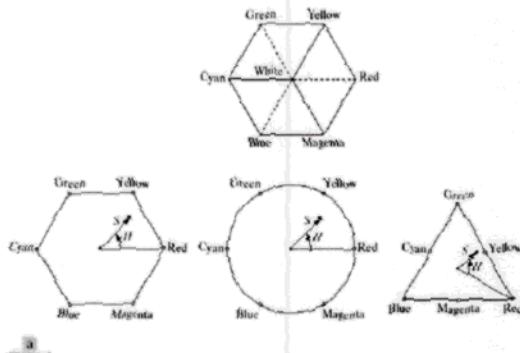
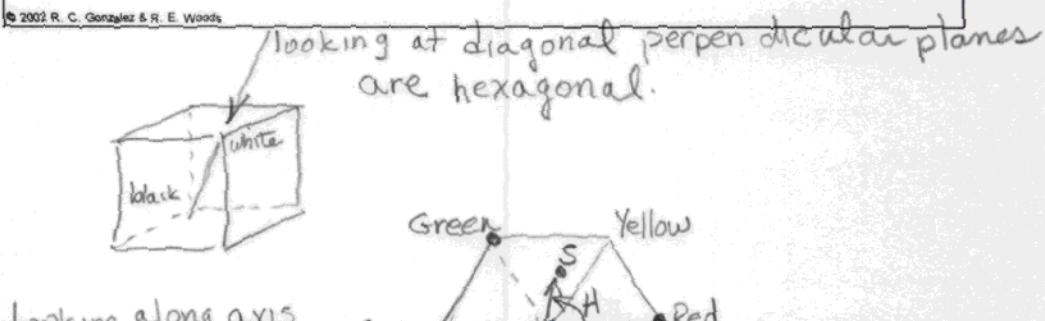
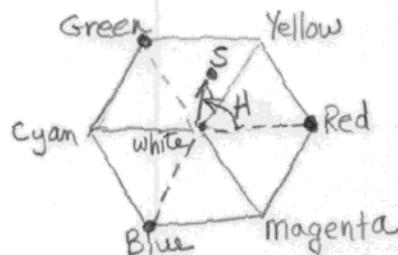


FIGURE 6.13 Hue and saturation in the HSI color model. The dot is an arbitrary color point. The angle from the red axis gives the hue, and the length of the vector is the saturation. The intensity of all colors in any of these planes is given by the position of the plane on the vertical intensity axis.

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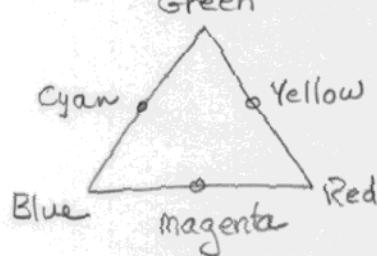
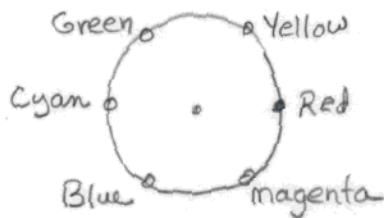
looking along axis



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.





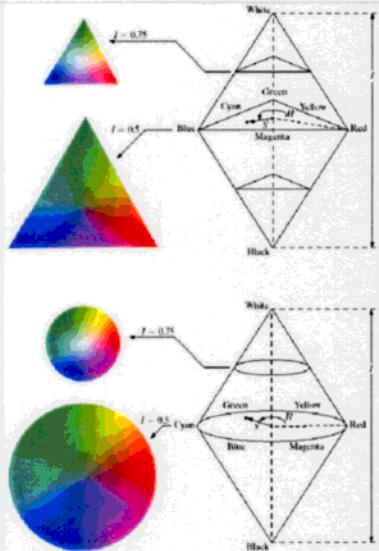
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a

b

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.

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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.

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

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

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

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

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 ($120^\circ \leq H \leq 240^\circ$)

$$\text{compute } H = H - 120^\circ$$

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

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

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

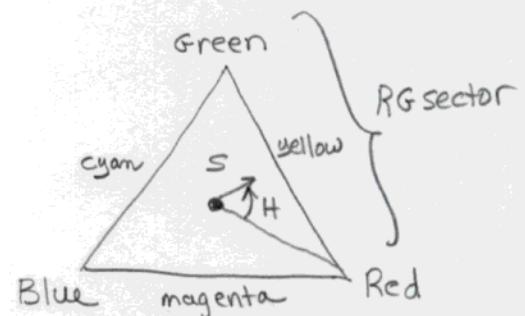
BR sector ($240^\circ \leq H \leq 360^\circ$)

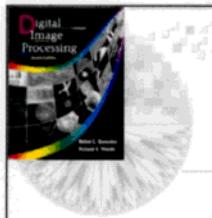
$$\text{compute } H = H - 240^\circ$$

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

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

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





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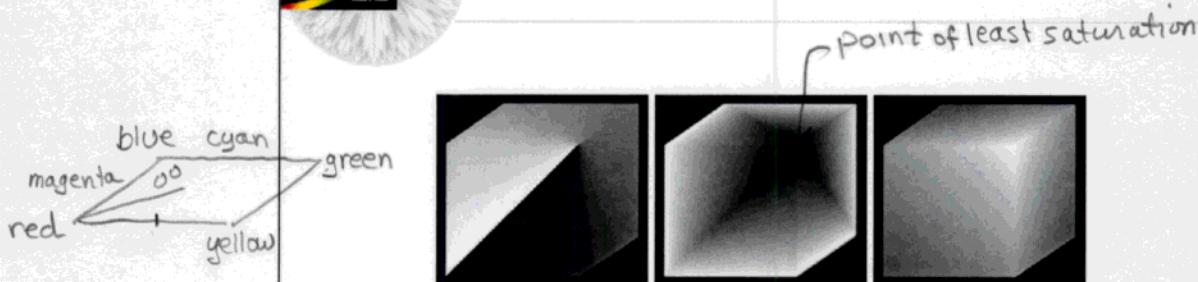
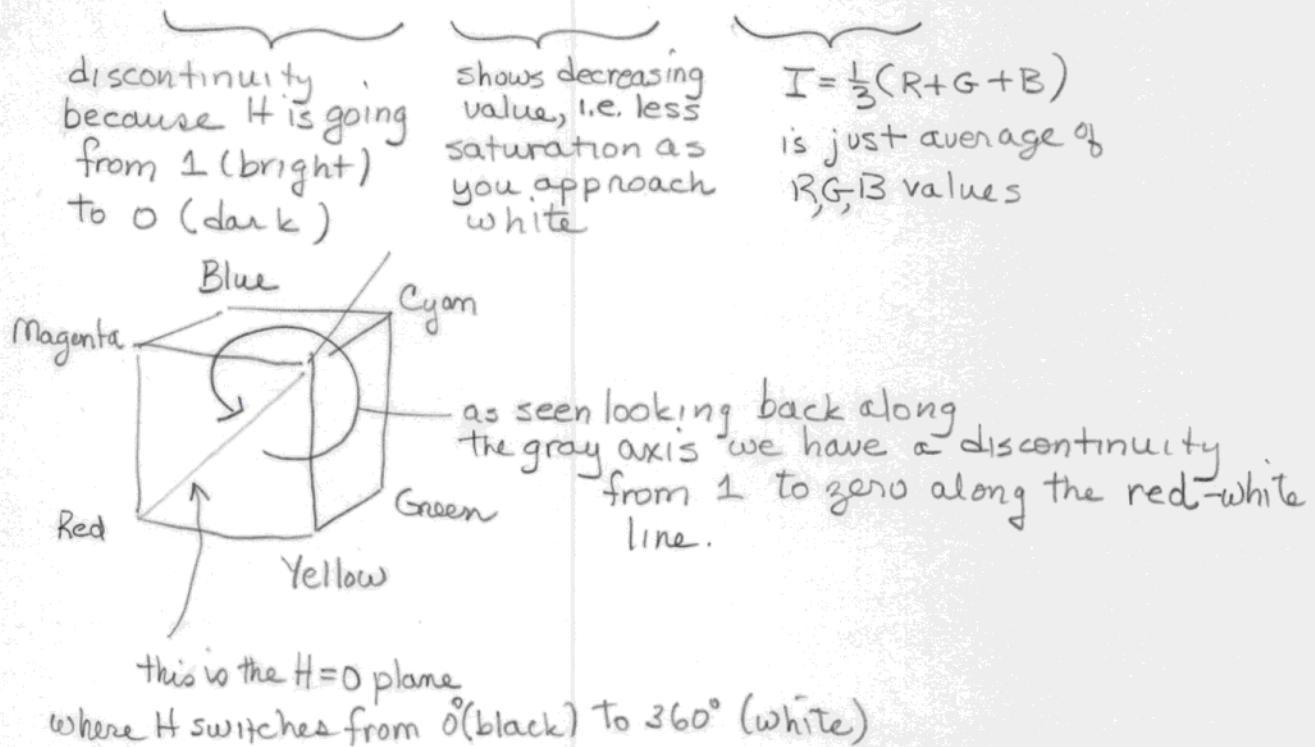


FIGURE 6.15 HSI components of the image in Fig. 6.8. (a) Hue, (b) saturation, and (c) intensity images.

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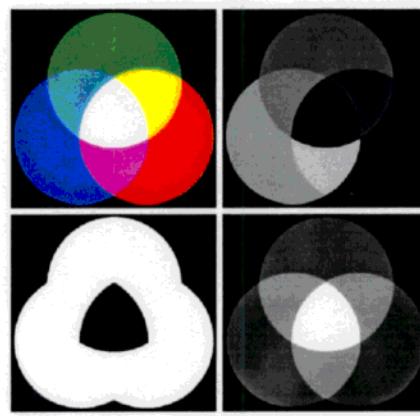




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Primary-
secondary
colors

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



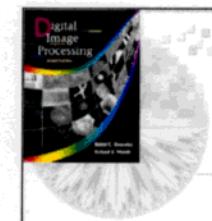
H
component
(angles)

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

I
component
average
intensities

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FIGURE 6.16 (a) RGB image and the components of its corresponding HSI image:
(b) hue, (c) saturation, and (d) intensity.

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H: change blue and green to 0, i.e., red.
blue → green

I: reduce intensity of white by 50%

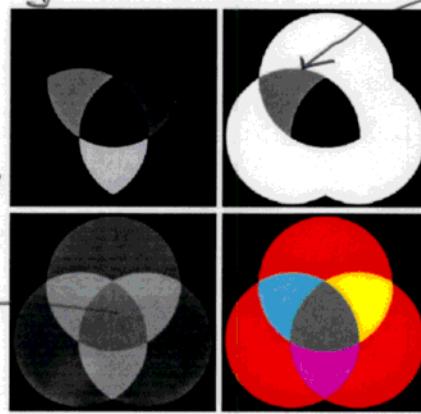


FIGURE 6.17 (a)-(c) Modified HSI component images. (d) Resulting RGB image.
(See Fig. 6.16 for the original HSI images.)

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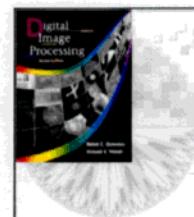
S: reduce cyan saturation by 50%

Transformed color image.

B, G become red
cyan looks washed out
white is now 50% gray

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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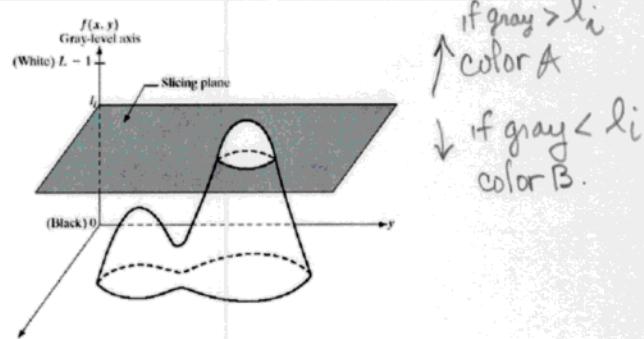
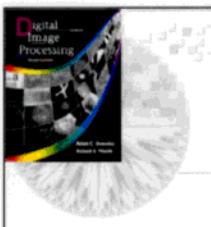


FIGURE 6.18 Geometric interpretation of the intensity-slicing technique.

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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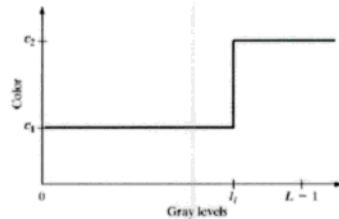


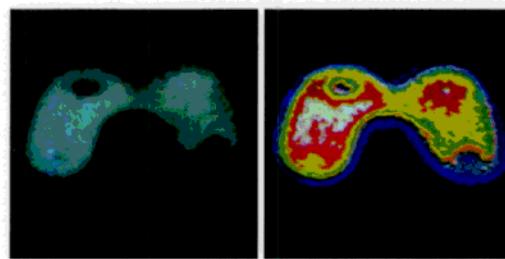
FIGURE 6.19 An alternative representation of the intensity-slicing technique.

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



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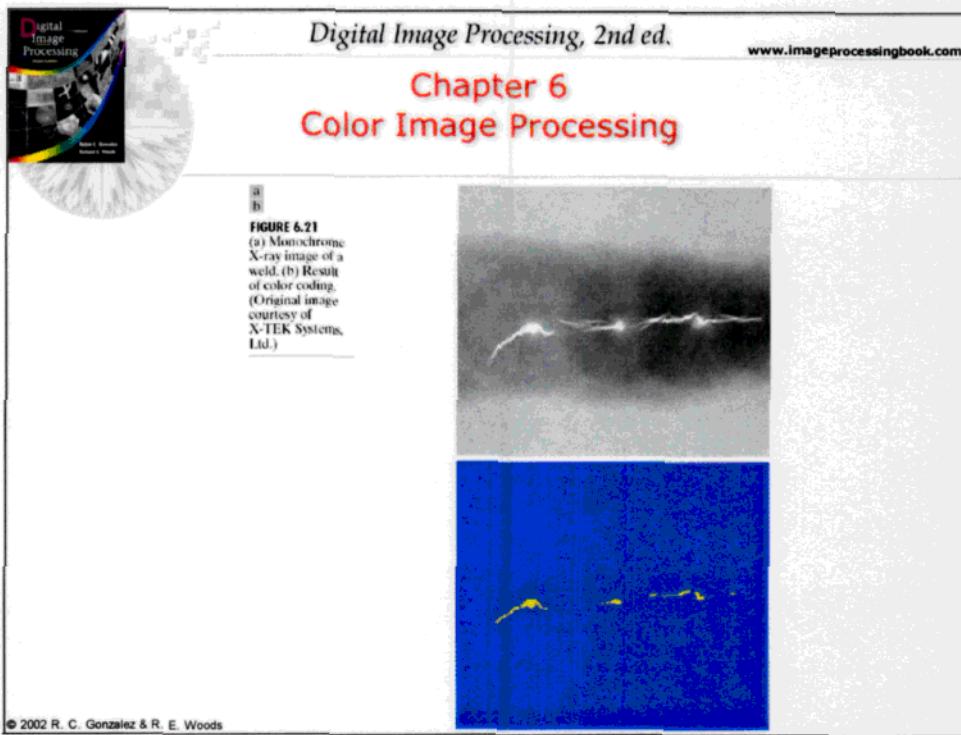


a b

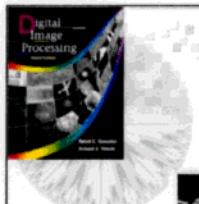
FIGURE 6.20 (a) Monochrome image of the Picker Thyroid Phantom. (b) Result of density slicing into eight colors. (Courtesy of Dr. J. L. Blankenship, Instrumentation and Controls Division, Oak Ridge National Laboratory.)

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intensity slicing with multiple color slices



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

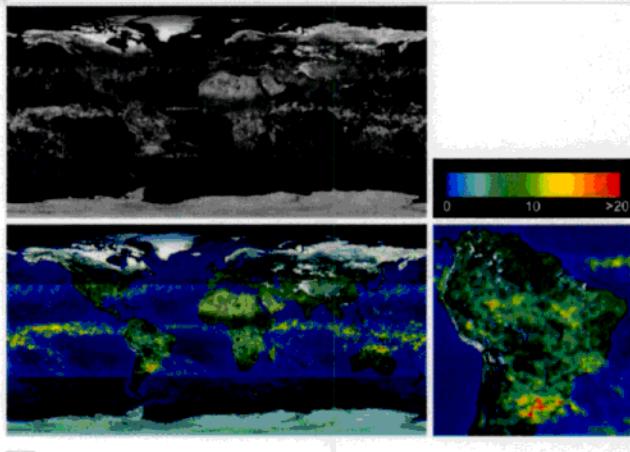


FIGURE 6.22 (a) Gray-scale image in which intensity (in the lighter horizontal band shown) corresponds to average monthly rainfall. (b) Colors assigned to intensity values. (c) Color-coded image. (d) Zoom of the South America region. (Courtesy of NASA.)

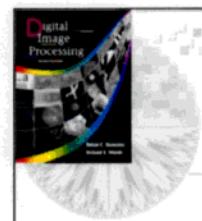
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Combine signals from (Tropical Rainfall Measuring Mission satellite)

- precipitation radar
- microwave imager
- 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

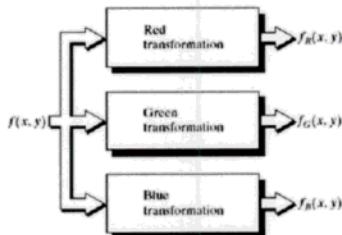
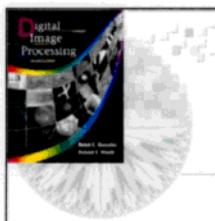


FIGURE 6.23 Functional block diagram for pseudocolor image processing. f_R , f_G , and f_B are fed into the corresponding red, green, and blue inputs of an RGB color monitor.

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



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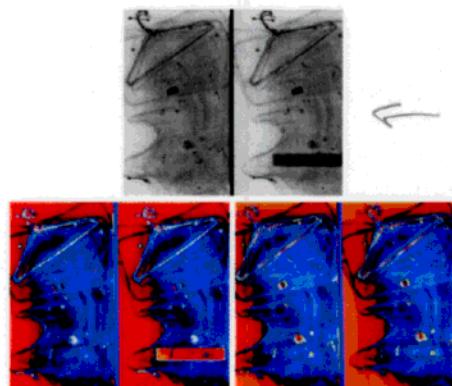


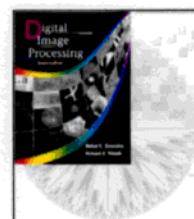
FIGURE 6.24 Pseudocolor enhancement by using the gray-level to color transformations in Fig. 6.25. (Original image courtesy of Dr. Mike Hurwitz, Westinghouse.)

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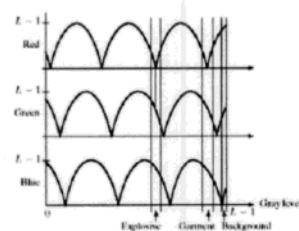
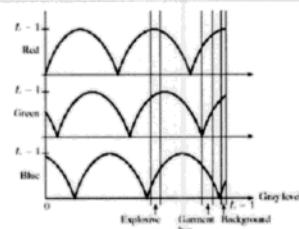
transform
6.25(a)

transform
6.25(b)

2 different transforms.



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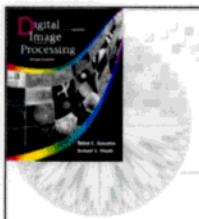


$L = \# \text{ of gray levels}$

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

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transformations used
to make explosives in 6.24 visible.



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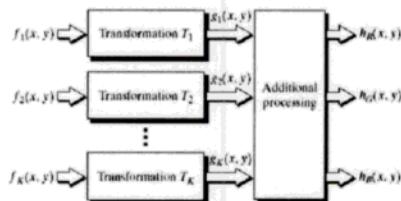


FIGURE 6.26 A pseudocolor coding approach used when several monochrome images are available.

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

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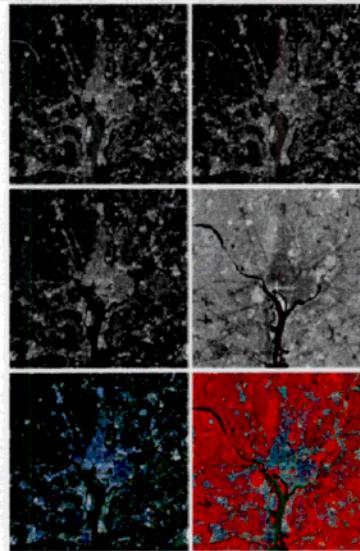


visible red

FIGURE 6.27 (a)-(d) Images in bands 1-4 in Fig. 3.10 (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.)

blue

visible R - G - B
image →



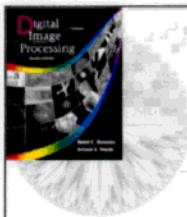
green

near-infrared

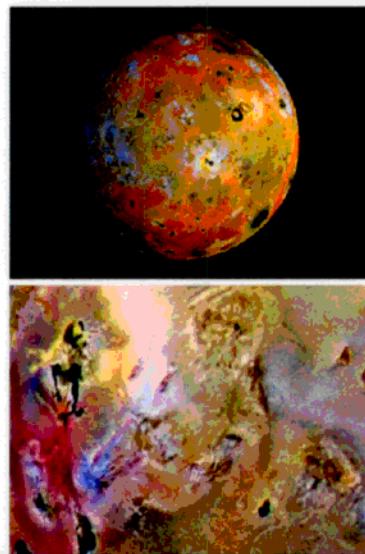
near-infrared - G - B
image

↑
used visible red
for R

↑
used near-infrared
for R.



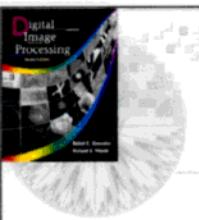
Chapter 6 Color Image Processing



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

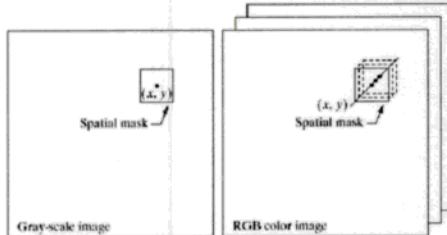
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used a variety of different wavelengths
The newly ejected material is red (different material)
The older material is yellow. (sulfur)



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



for neighborhood averaging operations are equivalent

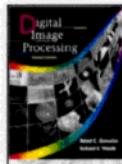
sum and divide
all pixels in
neighborhood

sum and divide all
the vectors in the neighborhood
to get the same result
as averaging each color component
and combining

full-color image processing

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

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



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FIGURE 6.30 A full-color image and its various color-space components. (Original image courtesy of Med-Data Interactive.)

4x5" color negative

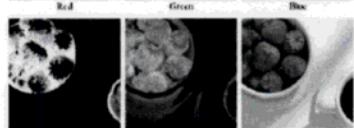
CMYK



RGB



HSI



hue is difficult to interpret since its 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

$$S_3 = kr_3$$

$$S_1 = r_1, \quad S_2 = r_2$$

RGB color space

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

CMY color space

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

$$we \ didn't \ show \ it \ but \ I = \frac{1}{3} [3 - (C + M + Y)] = 1 - \frac{1}{3}(C + M + Y)$$

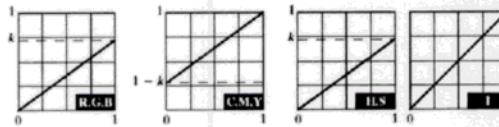
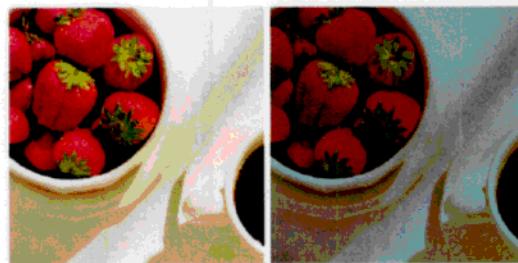
which is why this formula looks a little odd.

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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.)



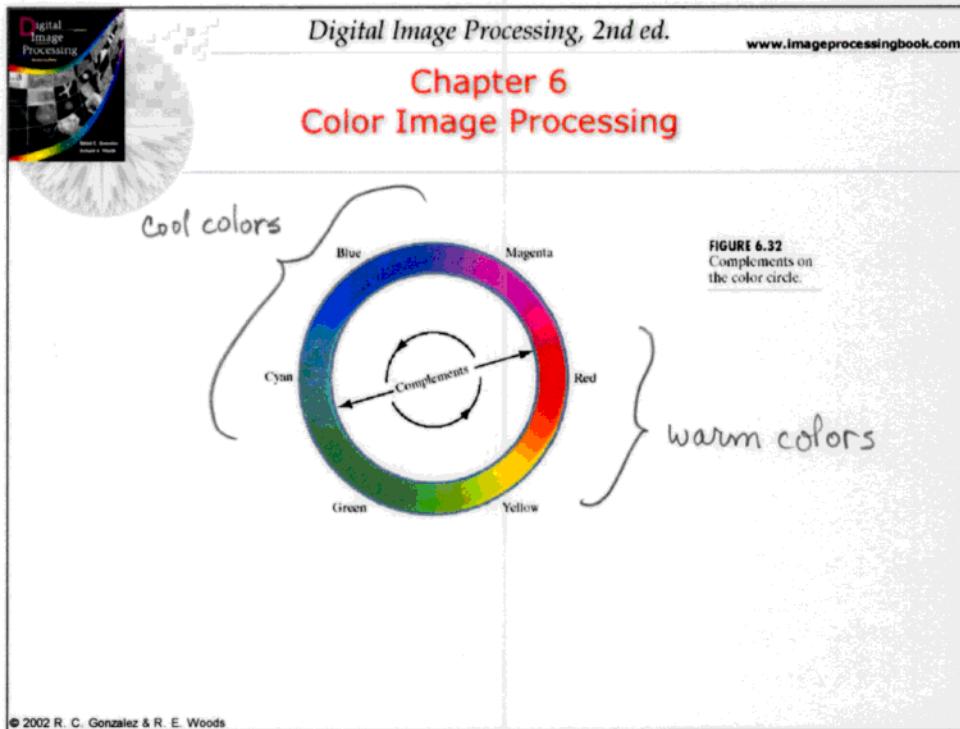
© 2002 R. C. Gonzalez & R. E. Woods.

$$k = 0.7$$

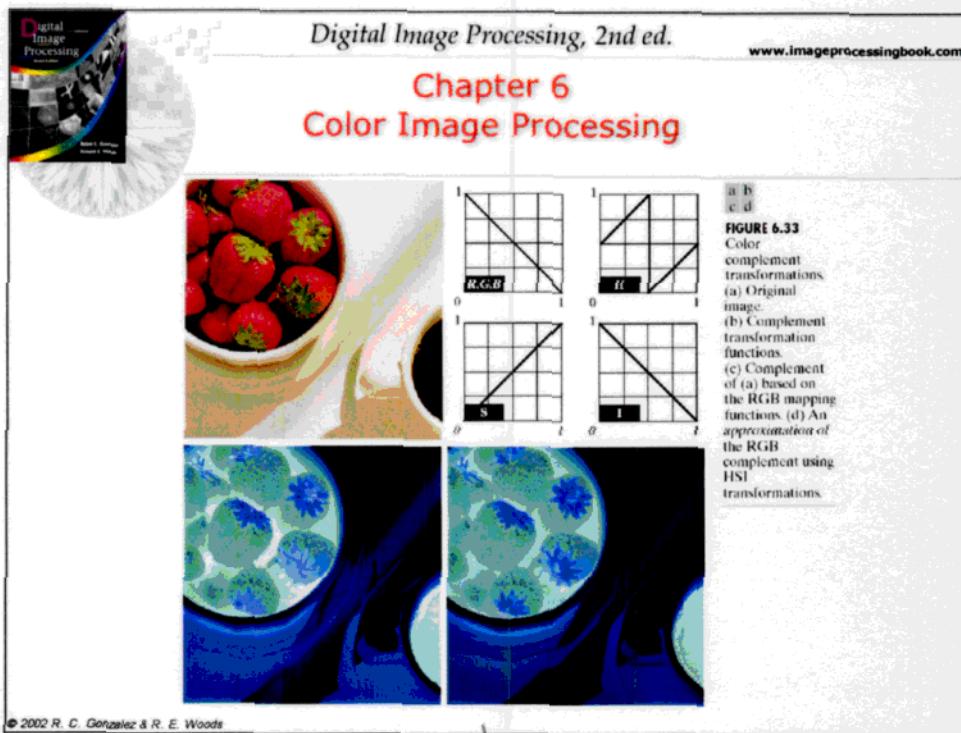
Scale each component in RGB

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

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



Newton's color circle summarizes
the additive properties of colors



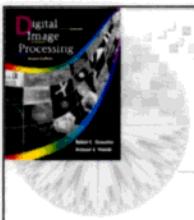
RGB complement

HSI complement

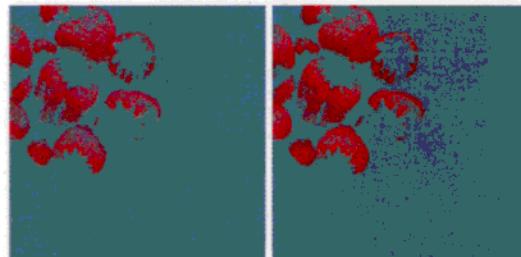
As you decrease
each color (R, G, B)
its complement
becomes evident.

complement not straight forward
See problem 6.18

R \longleftrightarrow cyan
G \longleftrightarrow magenta
B \longleftrightarrow yellow



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a b

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 $(0.5, 0.5, 0.5)$.

sphere might be
slightly better

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

cube, hypercube

$$s_i = \begin{cases} 0.5 & \text{if } |r_j - a_j| > \frac{W}{2} \\ r_i & \text{otherwise} \end{cases} \quad \begin{matrix} \downarrow \\ 1 \leq j \leq n \end{matrix} \quad \begin{matrix} \text{cube of width } W \\ \text{centered at } (a_1, a_2, a_3) \end{matrix}$$

sphere

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

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

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 $L^*a^*b^*$

similar to HSI
by separating color from intensity

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

lightness

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

Red - Green

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

Green - Blue

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

X_w, Y_w, Z_w — reference white tristimulus

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

X, Y, Z — R,G,B tristimulus values

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

colorimetric — colors perceived as identical have identical values

perceptually uniform — color differences are perceived uniformly

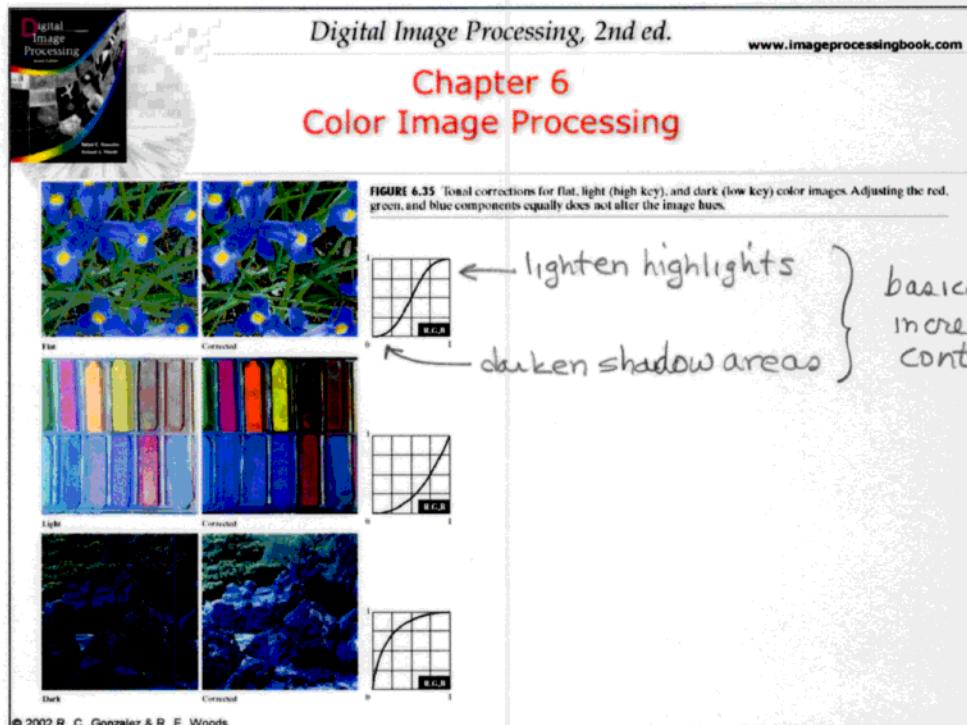
device independent

correction

flat

light

dark

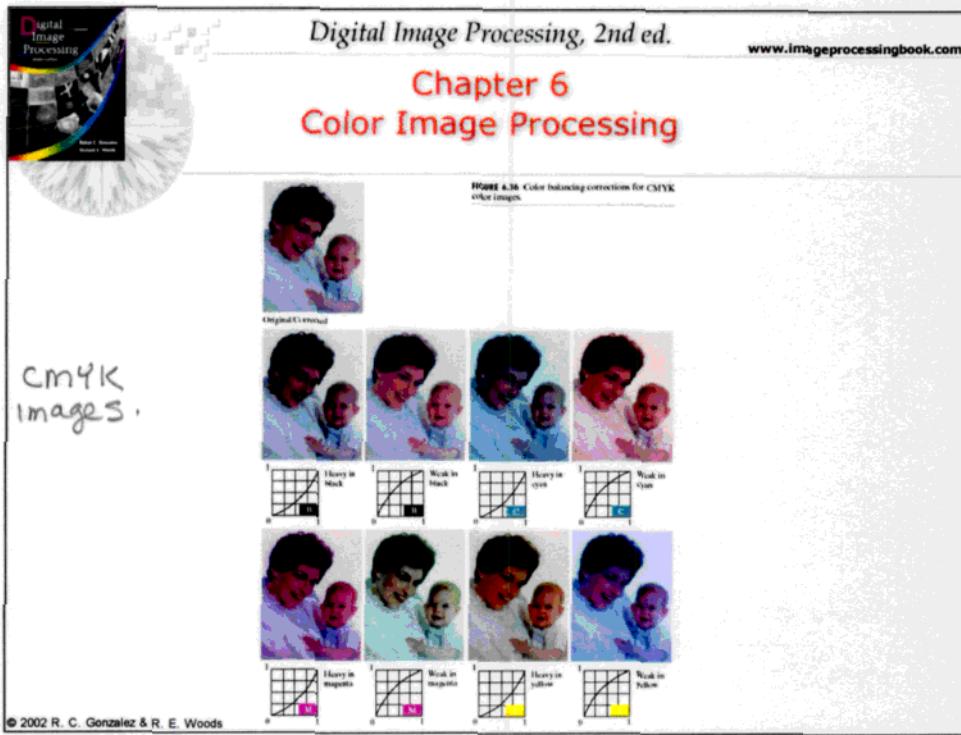


tonal range - key type

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

low-key - " " at low intensities

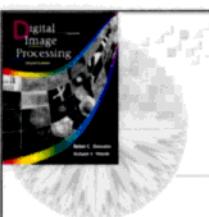
middle-key - " " at intermediate intensities



CMYK
images.

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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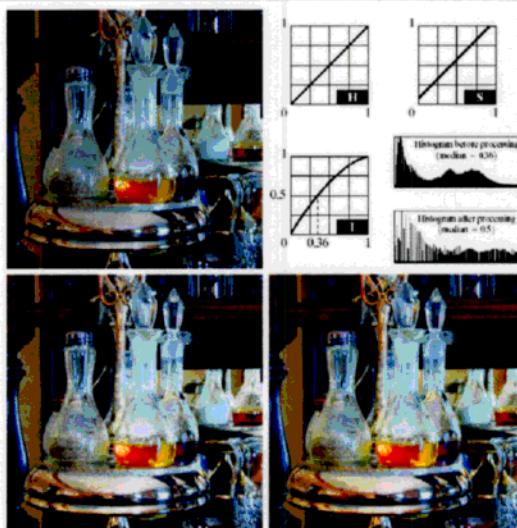


FIGURE 6.37
Histogram equalization (followed by saturation adjustment) in the HSI color space.

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.



We just considered
pixel transforms

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

Consider averaging

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

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



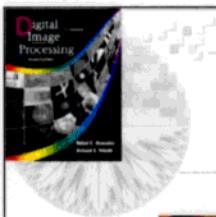
Chapter 6 Color Image Processing



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

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HSI components of previous picture.



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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

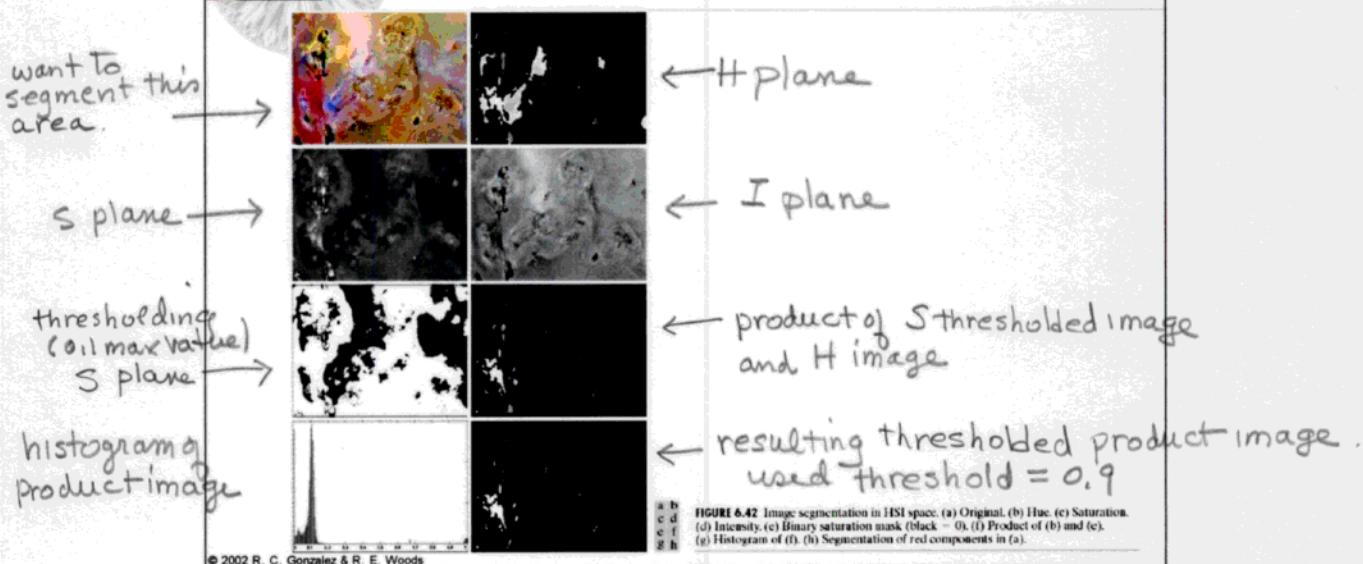
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Color image sharpening using Laplacian

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

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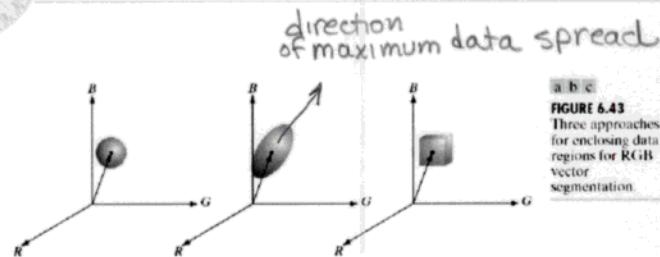


Segmentation (Chap. 10 topic)

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



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

spherical ellipsoidal
happens when you use C^{-1}

box
computationally much more effective
since no squares or square roots
choose sides proportional to standard deviations.

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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

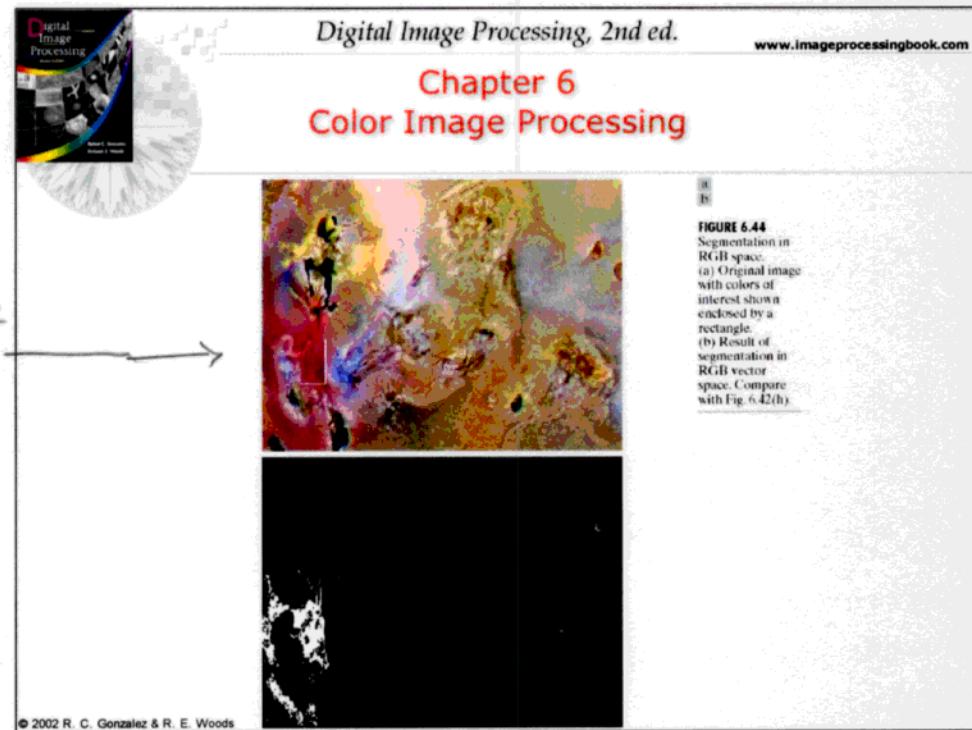
$$D(\underline{z}, \underline{a}) = \|\underline{z} - \underline{a}\| = \sqrt{(\underline{z} - \underline{a})^T (\underline{z} - \underline{a})}$$

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

A generalized distance measure is

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

↑ covariance matrix



rectangle
shows color
we want to
segment

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.

$$\text{mean } \underline{a}$$

$$\text{std deviations } \sigma_R, \sigma_G, \sigma_B$$

2. using box in color space

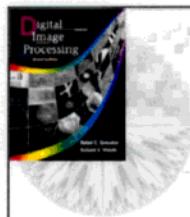
$$a_R \pm 1.25\sigma_R$$

$$a_G \pm 1.25\sigma_G$$

$$a_B \pm 1.25\sigma_B$$

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

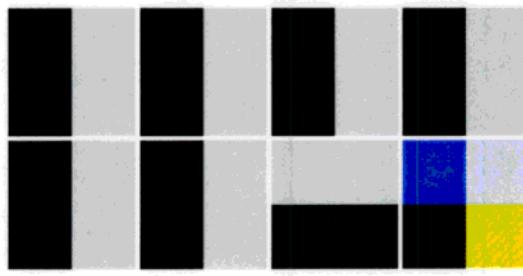


FIGURE 6.45 (a)-(c) R, G, and B component images and (d) resulting RGB color image.
(e)-(g) R, G, and B component images and (h) resulting RGB color image.

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. D. 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

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

further define

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

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

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

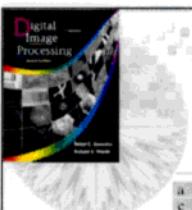
The maximum rate of change of $\underline{c}(x, y)$ at (x, y) is in the direction given by

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

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.



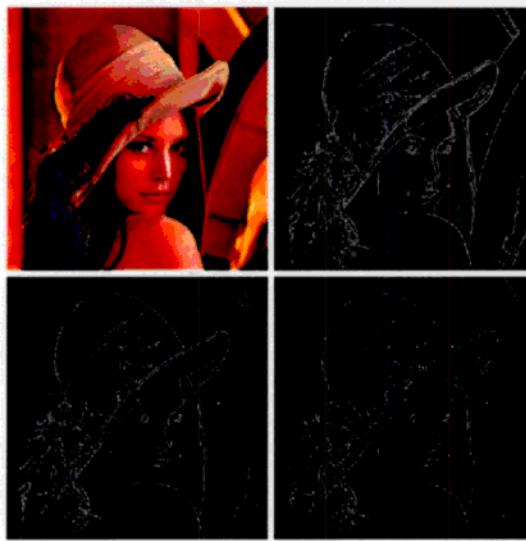
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original
image

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).

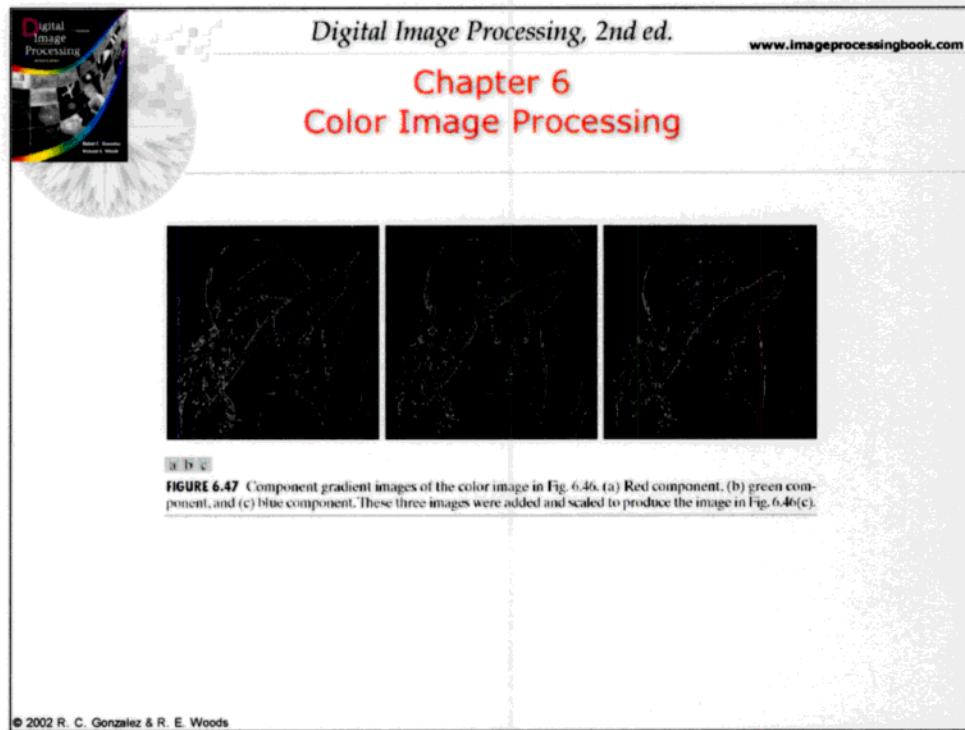
coloredges
computed by
adding
derivates in
each color
plane together

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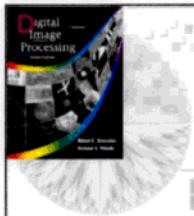


coloredges
computed
used vector
method in
RGB color
space

difference
between
(b) and(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.

noise spreads to all HSI component images

