



Chapter-7

Color Image Processing

Color Image Processing

"It is only after years of preparation that the young artist should touch color - not color used descriptively, that is, but as a means of personal expression".

Henri Matisse

"For a long time I limited myself to one color - as a form of discipline".

Pablo Picasso

Color Image Processing Topics

- Color fundamentals
- Color models
- Pseudocolor image processing
- Color image smoothing and sharpening
- Color edge detection
- Noise in color images
- Color perception models

7.1 Color Fundamentals

- In 1666 Sir Isaac Newton discovered that when a beam of sunlight passes through a glass prism, the emerging beam is split into a spectrum of colors

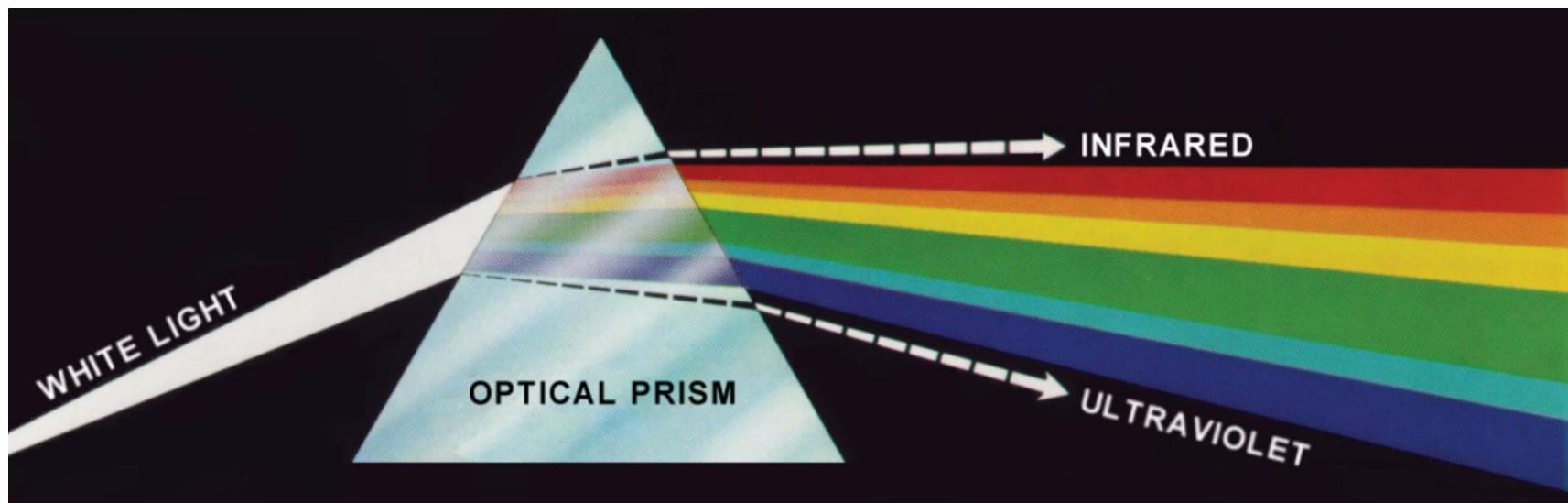
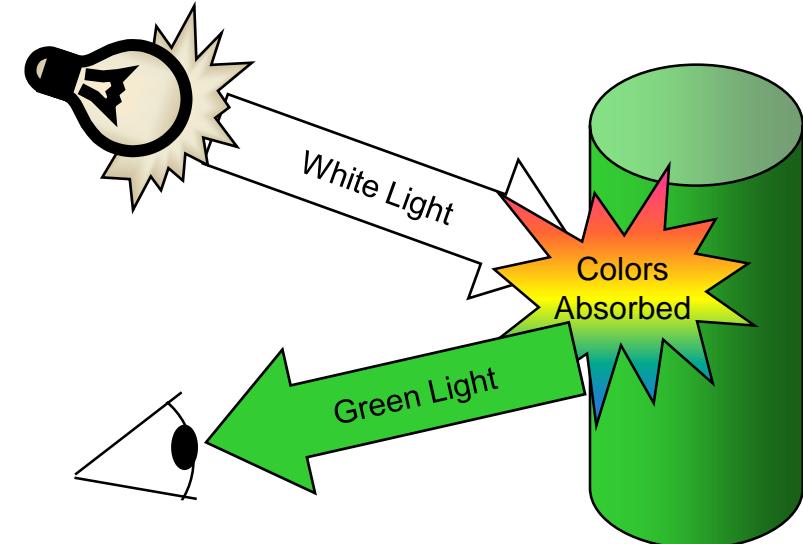


FIGURE 7.1
Color spectrum
seen by passing
white light through
a prism.
(Courtesy of the
General Electric
Co., Lighting
Division.)

Color Fundamentals (cont...)

- The colors that humans and most animals perceive in an object are determined by the nature of the light reflected from the object
- For example, green objects reflect light with wavelengths primarily in the range of 500 – 570 nm while absorbing most of the energy at other wavelengths.



Electromagnetic Spectrum

- Chromatic light (Visible by human eyes) spans the electromagnetic spectrum from around 400 to 700 nm.
- Human color vision is achieved through 6 to 7 million cones in each eye.
- For an achromatic (monochrome) light source, there is only one attribute to describe the quality: **Intensity**

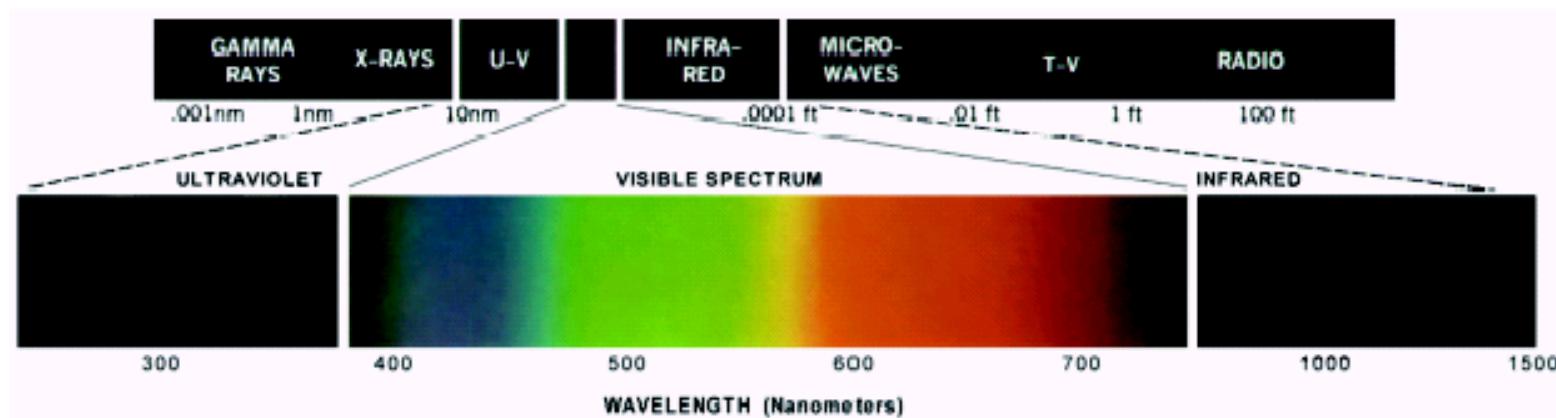


FIGURE 7.2

Wavelengths comprising the visible range of the electromagnetic spectrum. (Courtesy of the General Electric Co., Lighting Division.)

Color Fundamentals (cont...)

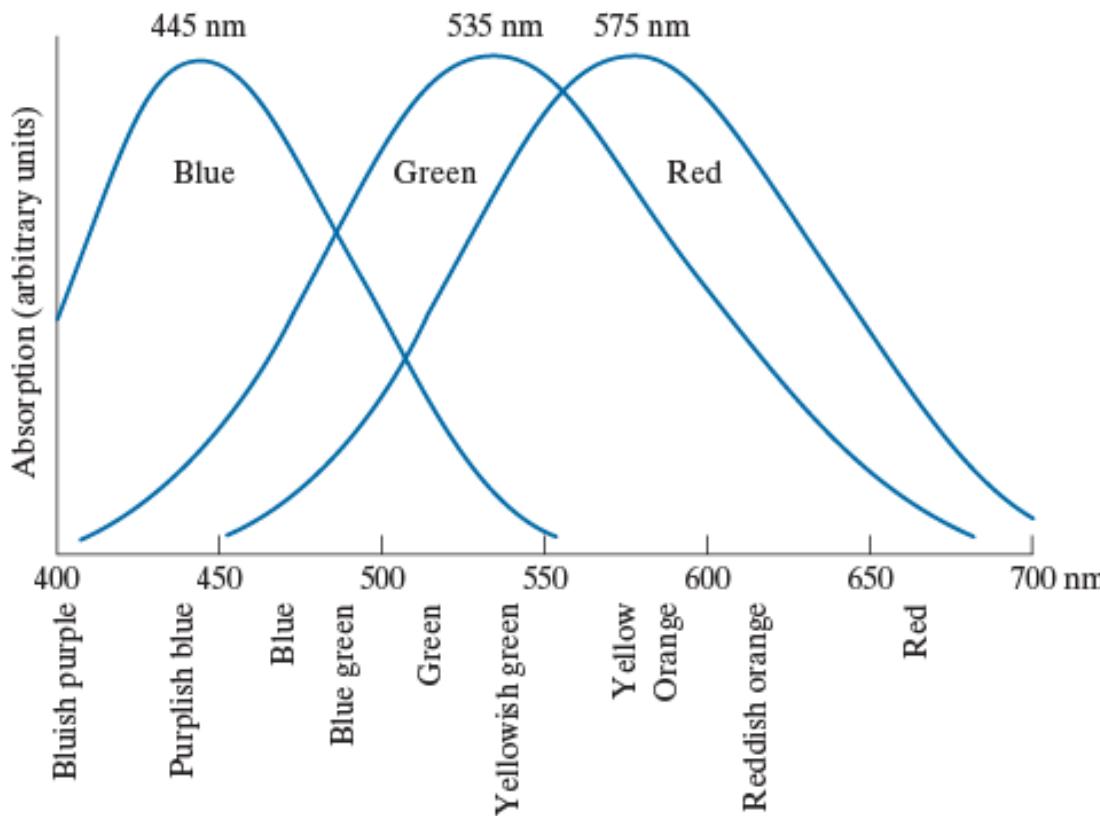
- For a chromatic light source, three attributes describe the quality :
 - **Radiance**: Total amount of energy that flows from the light source (measured in **watts**)
 - **Luminance**: Amount of energy an observer perceives from the light source (measured in **lumens**)
 - Note: We can have high radiance, but low luminance
 - **Brightness (= Intensity)**: A subjective (practically **unmeasurable**) notion that embodies the achromatic notion of intensity of light

Color Fundamentals (cont...)

- 6-7 million **cones** in the human eye can be divided into three principal **color sensing** groups:
 - 66% of these cones are sensitive to **red** light
 - 33% to **green** light
 - 2% to **blue** light
- **Absorption curves** for the different cones have been determined **experimentally**
- Strangely, these do not match the CIE standards for **red (700nm)**, **green (546.1 nm)** and **blue (435.8 nm)** light because the standards were developed well before the experiments could be done!

Color Fundamentals (cont...)

Note: The curves centered at Green and Red are very close



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

FIGURE 7.3
Absorption of light by the red, green, and blue cones in the human eye as a function of wavelength.

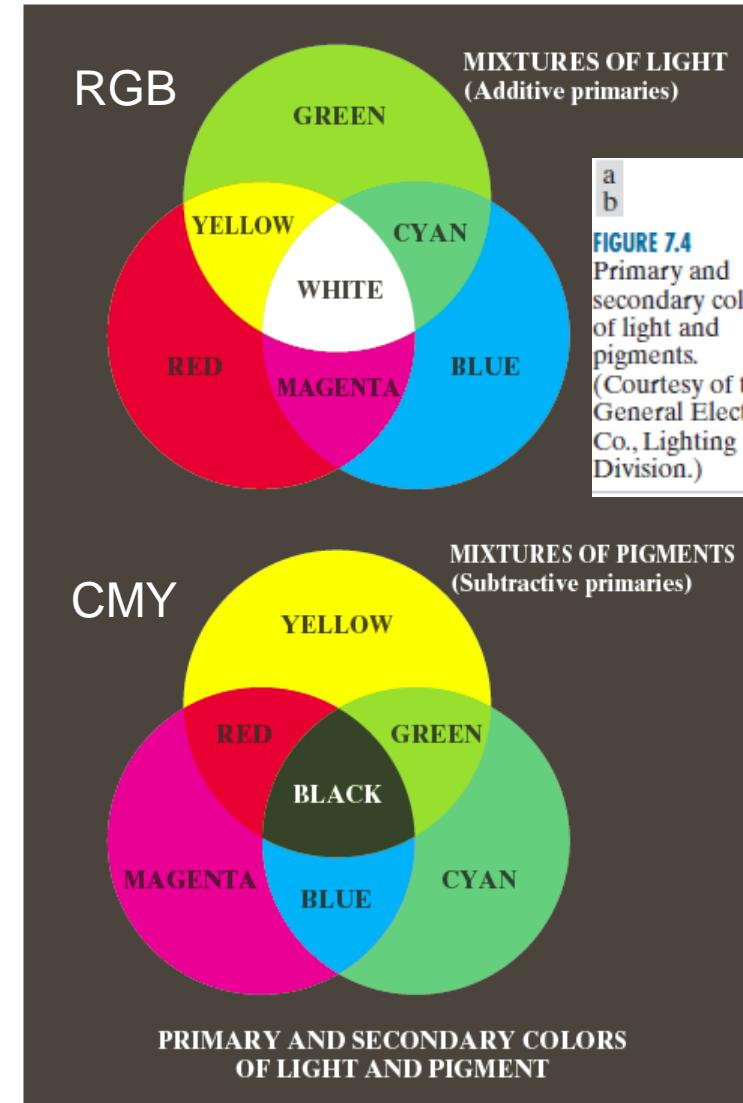
CIE = Commission Internationale de l'Eclairage
 (The International Commission on Illumination)

Color Fundamentals (cont...)

- Wrong : Linear combination of the three primaries (R, G, B) may produce all visible colors
- This would be true only if the centers of the three curves are shifted
 - This means that the wavelengths must change but then we no longer have the same primaries!
- The three curves are not a basis

Additive, Subtractive, Primary and Secondary Colors

- **Additive primary colors:**
 - **RGB** used in case of light sources such as **color monitors**
 - The primary colors can be added to produce the secondary colors.
 - Mixing the three primaries (RGB) produces white.
 - Mixing a secondary with its opposite primary produces white (e.g. red+cyan).
- **Subtractive primary colors:**
CMY used in the case of pigments in printing devices
 - e.g., Subtracting Green from White paper produces Magenta & so on



Color Fundamentals (cont...)

Important difference:

- Additive Primary colors (red, green, blue)
- Subtractive Primary colors of pigments (colorants)
 - A color that subtracts or absorbs a primary color of light and reflects the other two.
 - e.g., Magenta = Red + Blue (Green is subtracted from White)
 - Pigment Colors: Cyan, Magenta and Yellow (CMY)
 - Proper combination of pigment primaries produces black.
 - **Save Money: No pigment needed to produce white**

Color Fundamentals (cont...)

The characteristics generally used to distinguish one color from another are :

- **Brightness:** The achromatic notion of **intensity**
- **Hue:** Dominant wavelength in a mixture of light waves
 - Dominant color perceived by an observer,
 - When we say, "red" or "orange" object → That's Hue
- **Saturation:** Relative purity or the amount of white light mixed with a hue. (inversely proportional to amount of white light added)
 - Pure colors → Fully saturated, e.g., red or blue ... but
 - Pink (red + white) → Less saturated
 - Lavender (violet + white) → Less saturated

Color Characterization

- Hue and Saturation taken together: *Chromaticity*

Hue +
Saturation } Chromaticity

- Therefore, *any color* is characterized by its
 - Brightness and
 - Chromaticity.
- The amounts of *red, green and blue* needed to form a particular color are called *Tristimulus values*, denoted by *X, Y and Z*.

Color Fundamentals (cont...)

- A color is then specified by its *Trichromatic coefficients*:

RED	GREEN	BLUE
$x = \frac{X}{X + Y + Z}$	$y = \frac{Y}{X + Y + Z}$	$z = \frac{Z}{X + Y + Z}$

$$x + y + z = 1$$

- For any visible wavelength, the tristimulus values needed to produce specific color corresponding to particular wavelength are obtained by curves compiled by extensive experimentation.

CIE Chromaticity Diagram

- Specifying colors systematically can be achieved using the CIE chromaticity diagram.
- On this diagram the **x-axis** represents the proportion of **Red** and the **y-axis** represents the proportion of **Green** used to produce a specific color.
- The proportion of **blue** used in a color is calculated as:

$$z = 1 - (x + y)$$

CIE Chromacity Diagram (cont...)

- Shows color composition as a function of x (red) and y (green)
- Point marked "Green"
 - 62% green, 25% red and 13% blue. (sum=1)
- Point marked "Red"
 - 32% green, 67% red and 1% blue. (sum=1)
- The diagram is useful for color mixing.

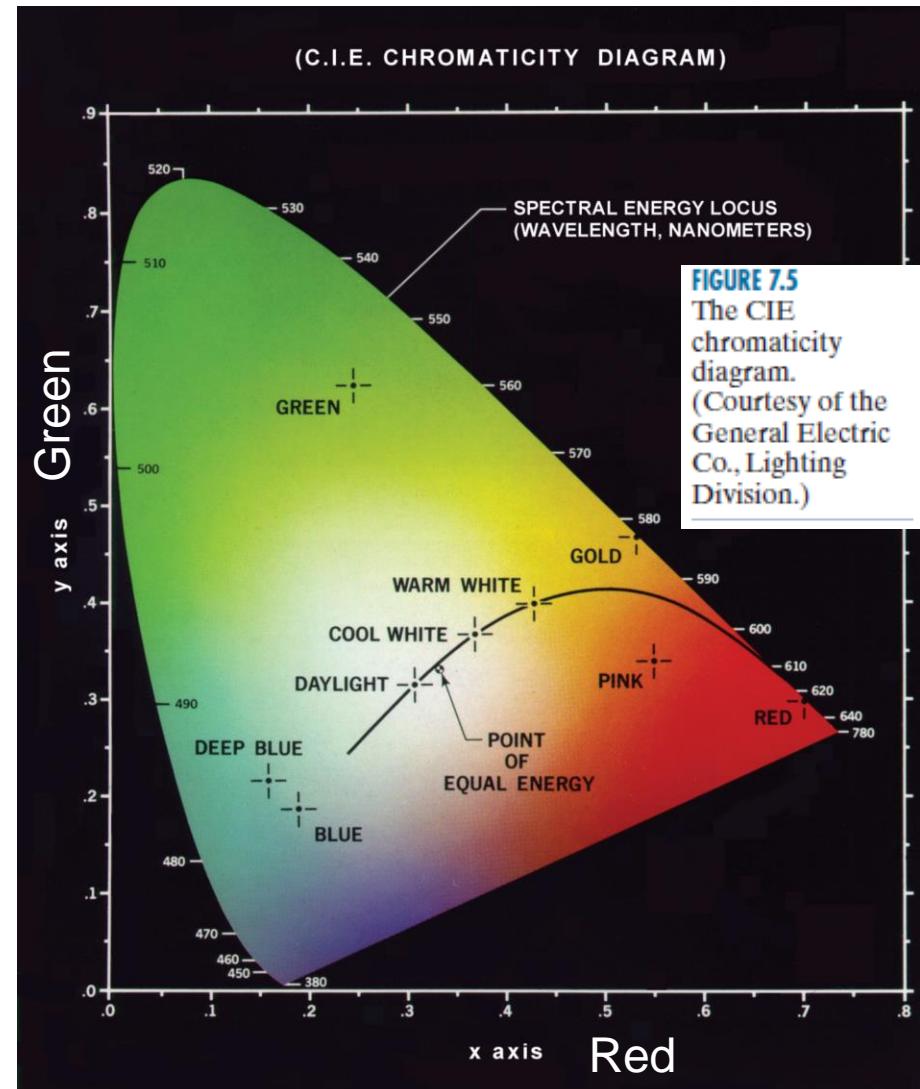


FIGURE 7.5
The CIE chromaticity diagram.
(Courtesy of the General Electric Co., Lighting Division.)



Boundary:
Different pure
spectrum
colors →
saturated

(C.I.E. CHROMATICITY DIAGRAM)

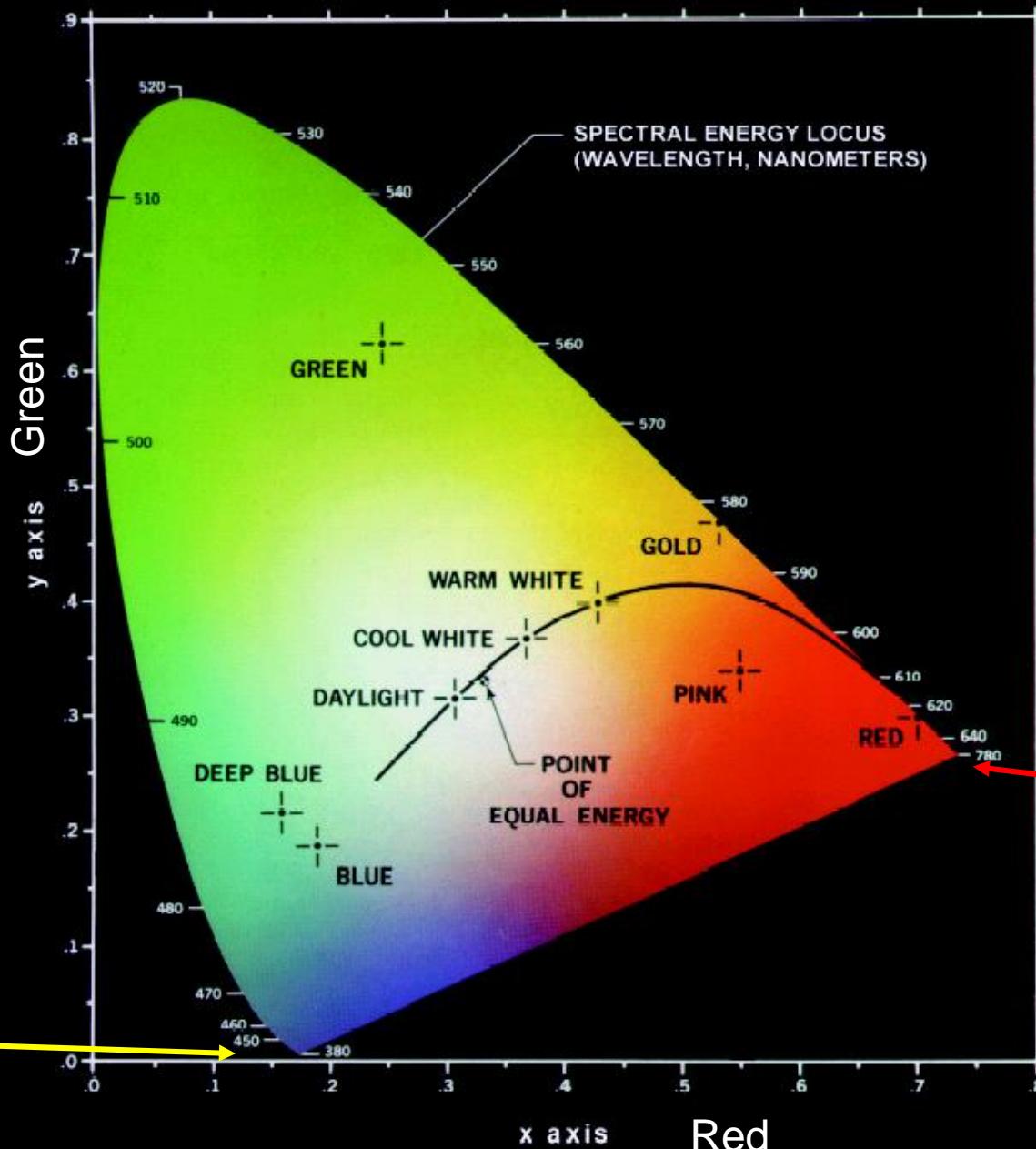


FIGURE 7.5
The CIE
chromaticity
diagram.
(Courtesy of the
General Electric
Co., Lighting
Division.)

CIE Chromacity Diagram (cont...)

- Any color located on the **boundary** of the chromaticity chart is **fully saturated (Pure colors)**.
- The point of equal energy (**PEE**) has **equal amounts of red, green and blue.** (= pure white)
 - It is the CIE standard for pure white.

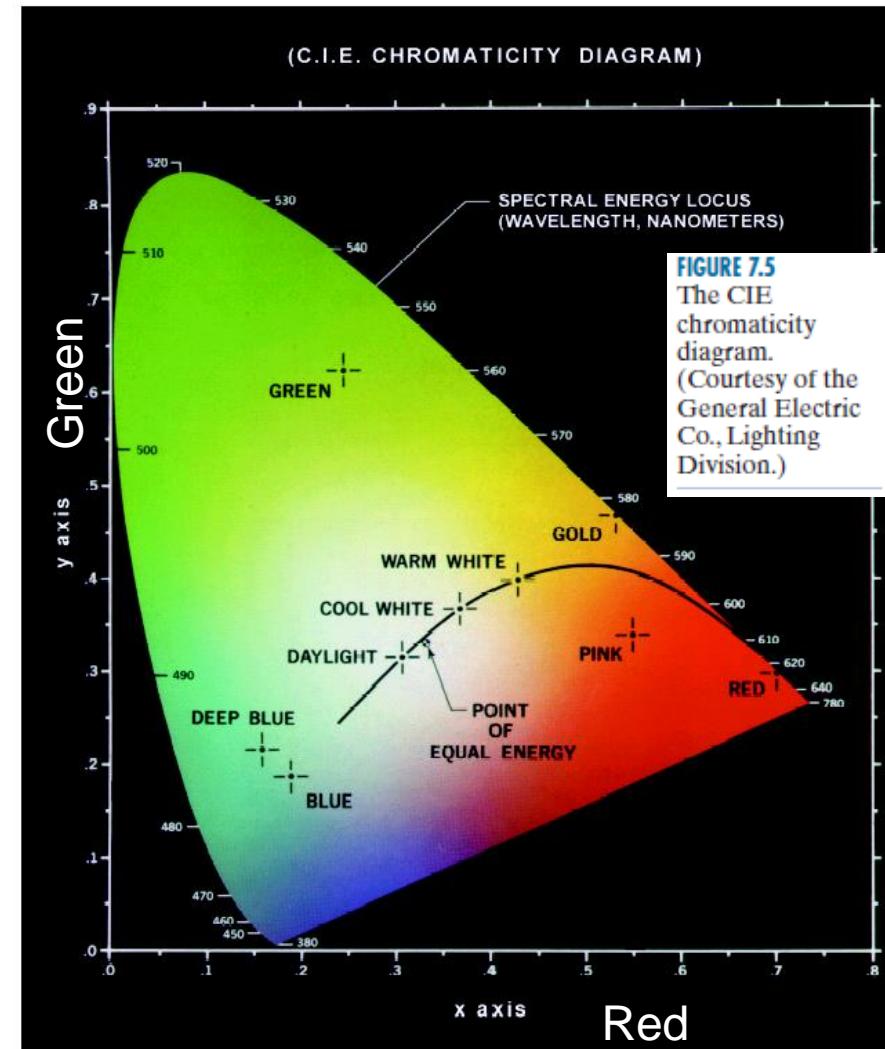
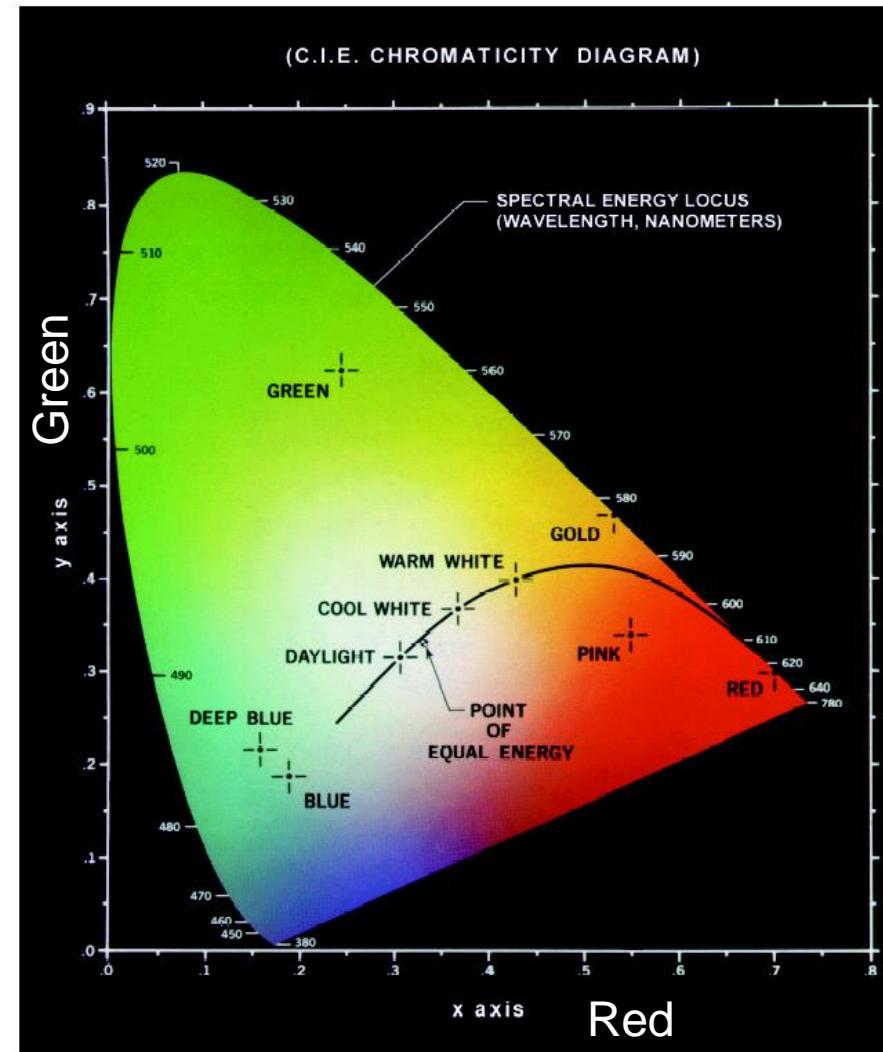


FIGURE 7.5
The CIE chromaticity diagram.
(Courtesy of the General Electric Co., Lighting Division.)

CIE Chromacity Diagram (cont...)

- Any straight line joining two points in the diagram defines all the different colors that can be obtained by combining these two colors additively.
- A line drawn from the PEE to any point on the boundary defines all the shades of that particular color.

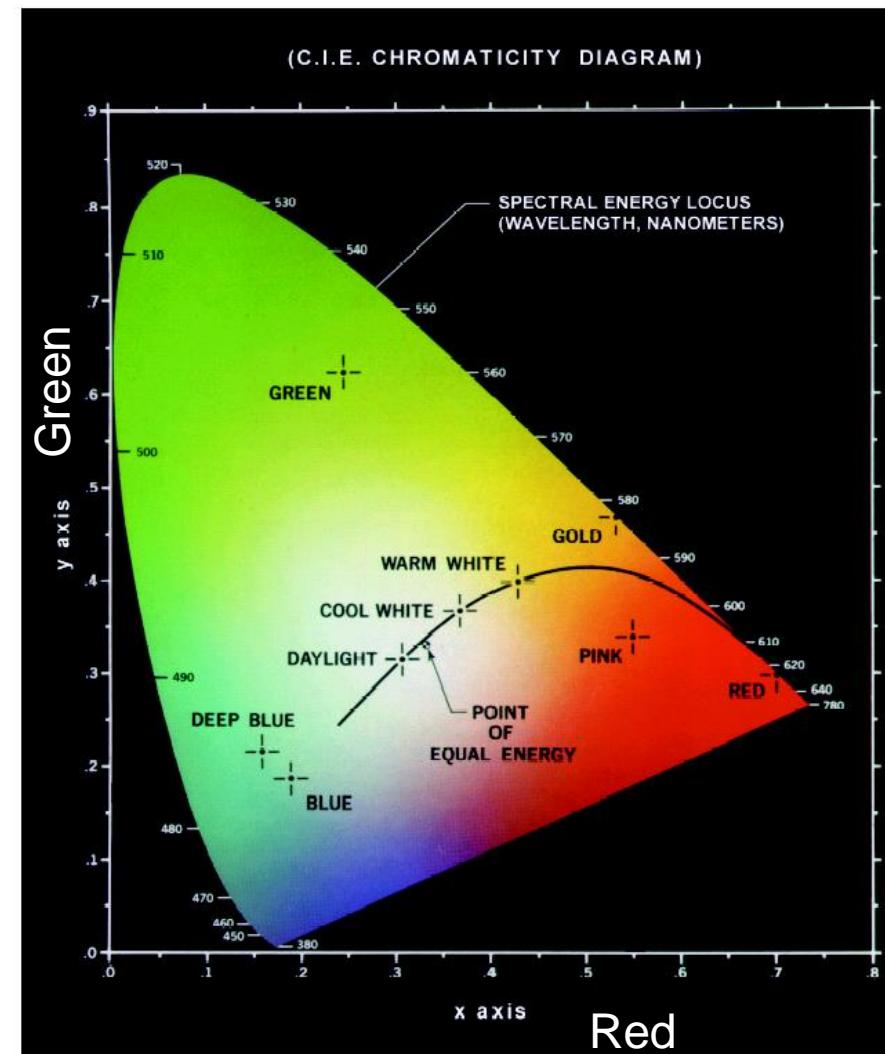
FIGURE 7.5
The CIE chromaticity diagram.
(Courtesy of the General Electric Co., Lighting Division.)



CIE Chromacity Diagram (cont...)

- By combining any three given colors we may obtain the colors enclosed in the triangle defined by the three initial colors.

FIGURE 7.5
 The CIE chromaticity diagram.
 (Courtesy of the General Electric Co., Lighting Division.)



CIE Chromacity Diagram (cont...)

- A triangle with vertices at any three fixed pure colors **can not** enclose the entire color region.

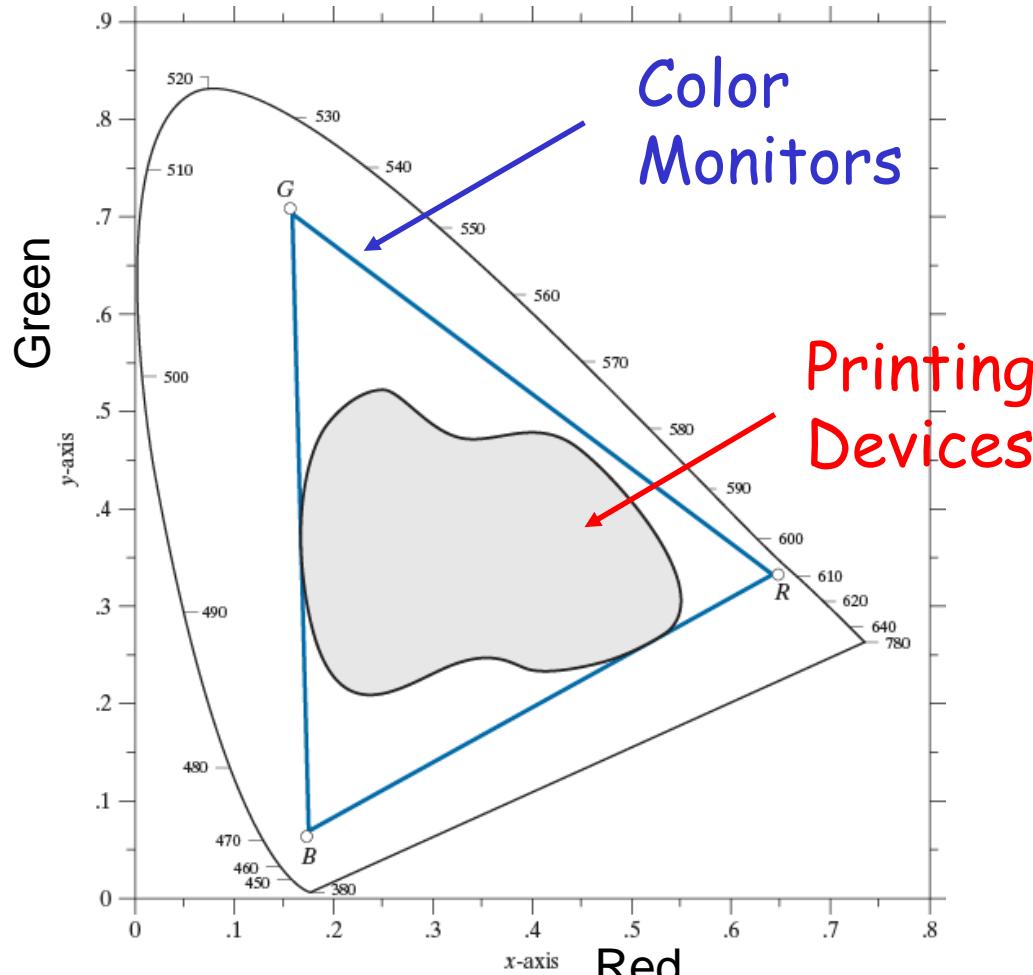


FIGURE 7.6
Illustrative color gamut of color monitors (triangle) and color printing devices (shaded region).

Color gamut of Color Monitors and Printing Devices

- The triangle shows the typical color gamut produced by RGB monitors.
- The entire color range cannot be displayed based on any three colors.
- The irregular shape is the gamut achieved by high quality color printers.

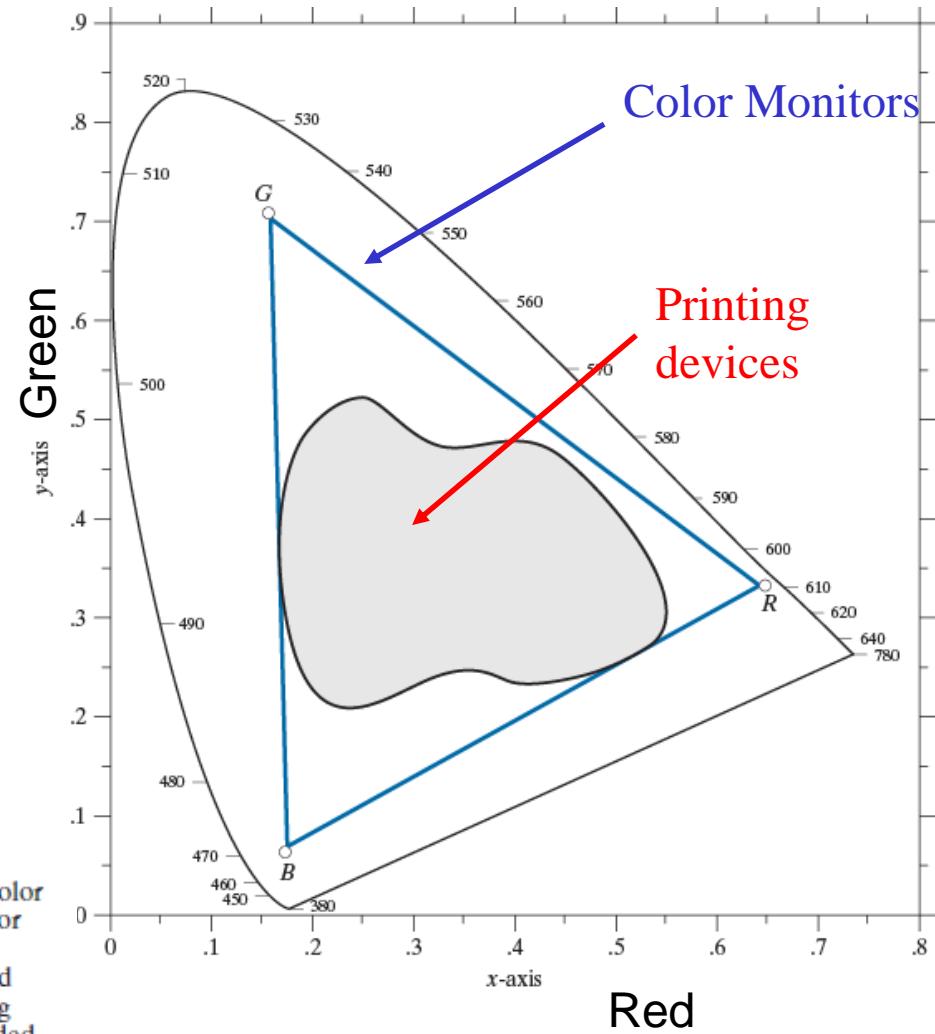
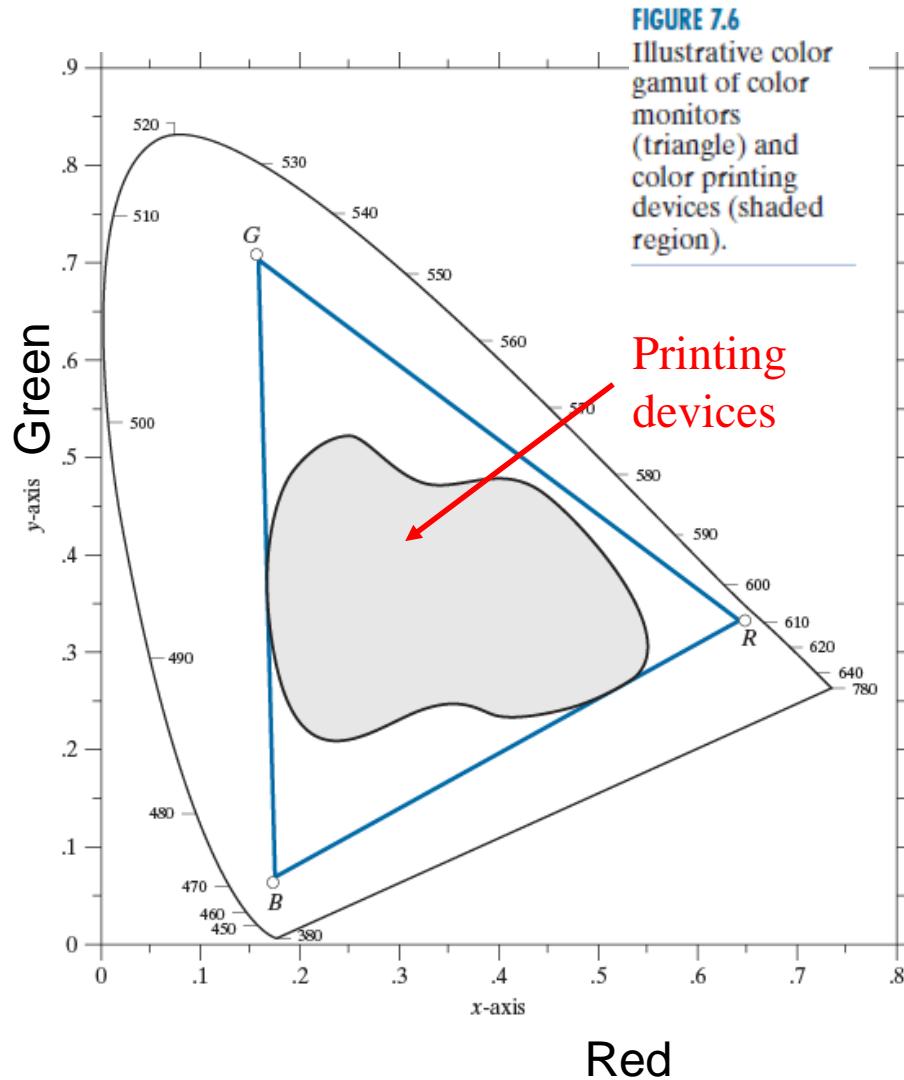


FIGURE 7.6
Illustrative color gamut of color monitors (triangle) and color printing devices (shaded region).



- The boundary of the printing gamut is irregular because printing is a combination of additive and subtractive color mixing.
- This is a more difficult process to control than that of displaying colors on a monitor using additive RGB primaries.

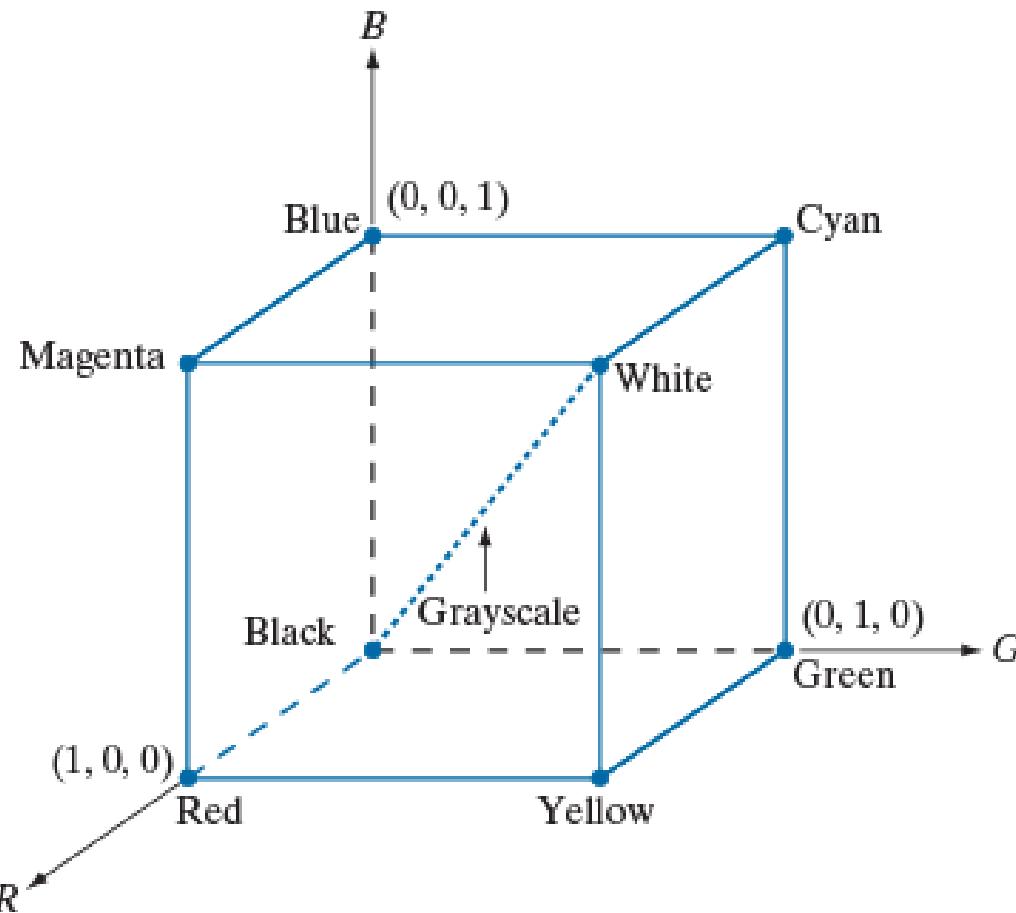


7.2 Color Models

- Purpose of color models: To facilitate specification of colors in some standard
- From previous discussion it is obvious that there are different ways to model color.
- We will consider two most popular models used in color image processing:
 - RGB (Red Green Blue)
 - HSI (Hue Saturation Intensity)

7.2 RGB - For Monitors

- In the RGB model each color appears in its primary spectral components of red, green and blue.



RGB color models:

- Based on Cartesian coordinate system

FIGURE 7.7
Schematic of the RGB color cube. Points along the main diagonal have gray values, from black at the origin to white at point $(1, 1, 1)$.



- RGB values are at 3 corners on primary axis.
- Cyan, Magenta and Yellow are at three other corners of the cube.
- Black is at the origin.
- White is the corner furthest from the origin.
- Different colors are points on or inside the cube represented by RGB vectors.

FIGURE 7.7
Schematic of the RGB color cube. Points along the main diagonal have gray values, from black at the origin to white at point $(1, 1, 1)$.

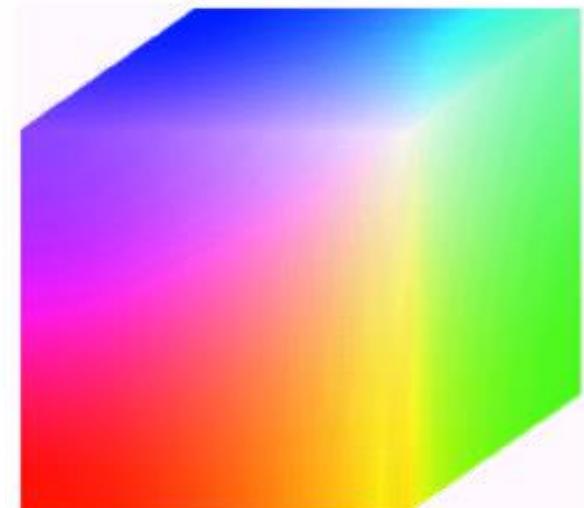
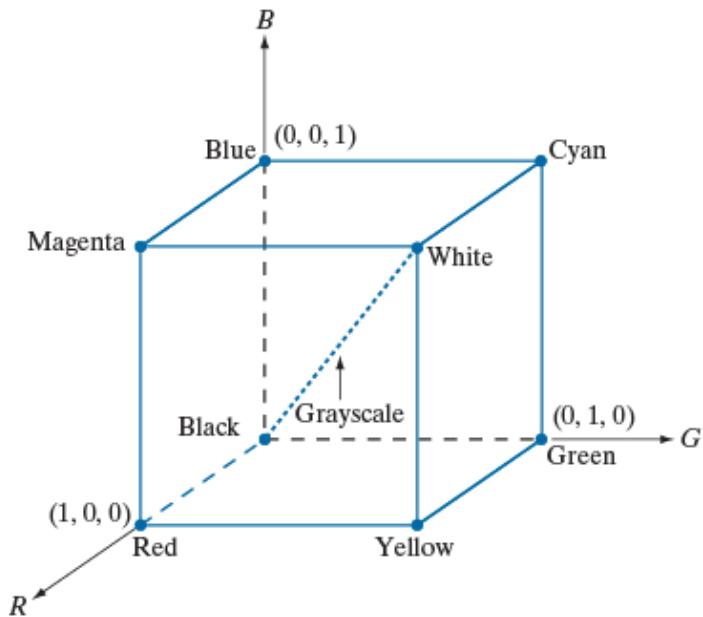


FIGURE 7.8
A 24-bit RGB color cube.

RGB (cont...)

- Images represented in the RGB color model consist of three component images - one for each primary color.
- When fed into a monitor these images are combined to create a composite color image.
- The number of bits used to represent each pixel is referred to as the color depth.
- A 24-bit image is often referred to as a full-color image as it allows $(2^8)^3 = 16,777,216$ colors.

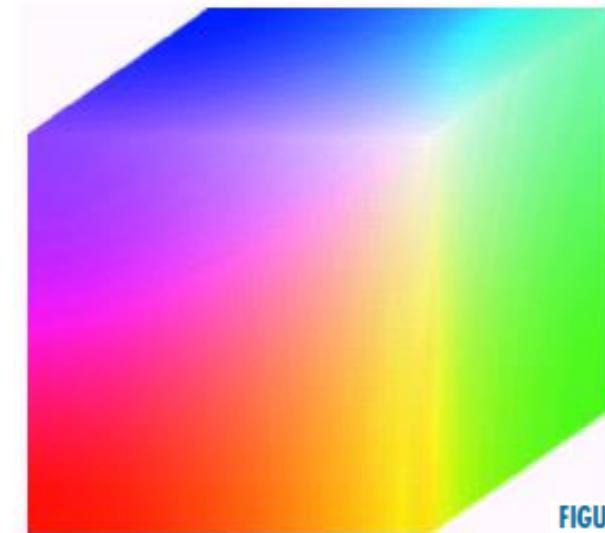


FIGURE 7.8
A 24-bit RGB color cube.

$R = 8 \text{ bits}$
 $G = 8 \text{ bits}$
 $B = 8 \text{ bits}$
}
 Color depth 24 bits
 $= 16777216 \text{ colors}$



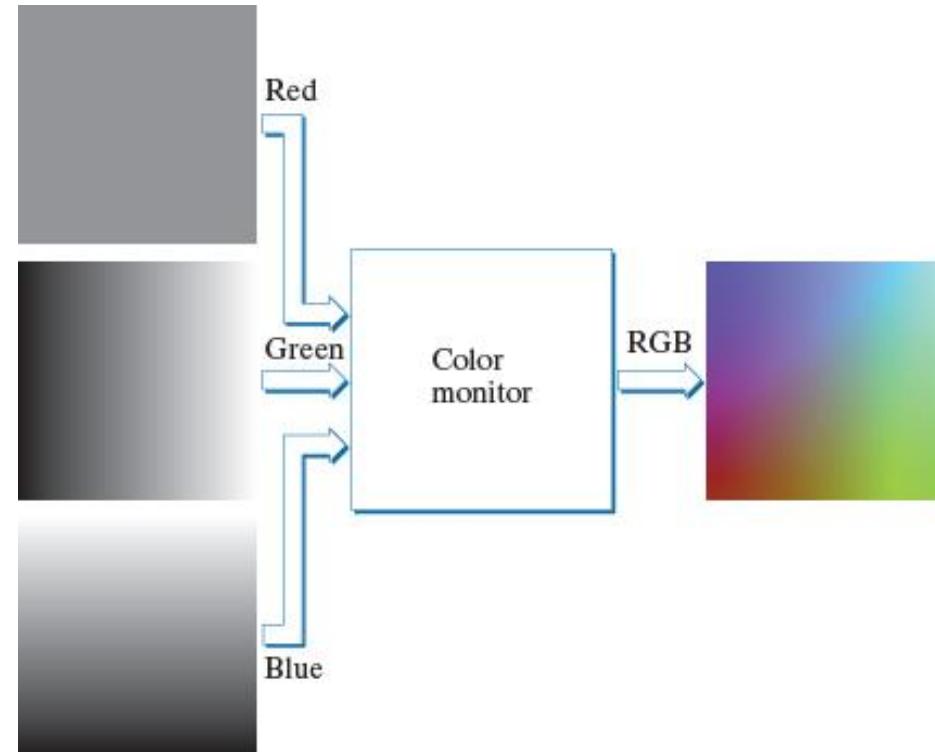
RGB (cont...)

a
b

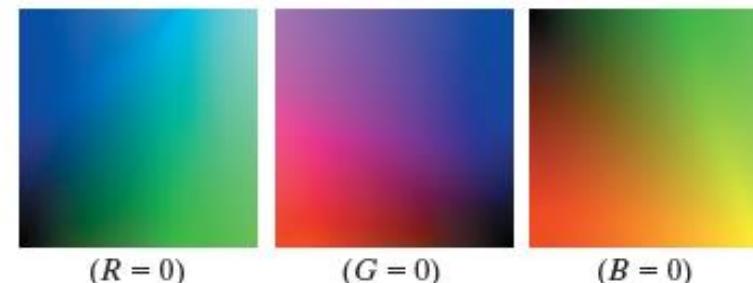
FIGURE 7.9

(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. 7.8.

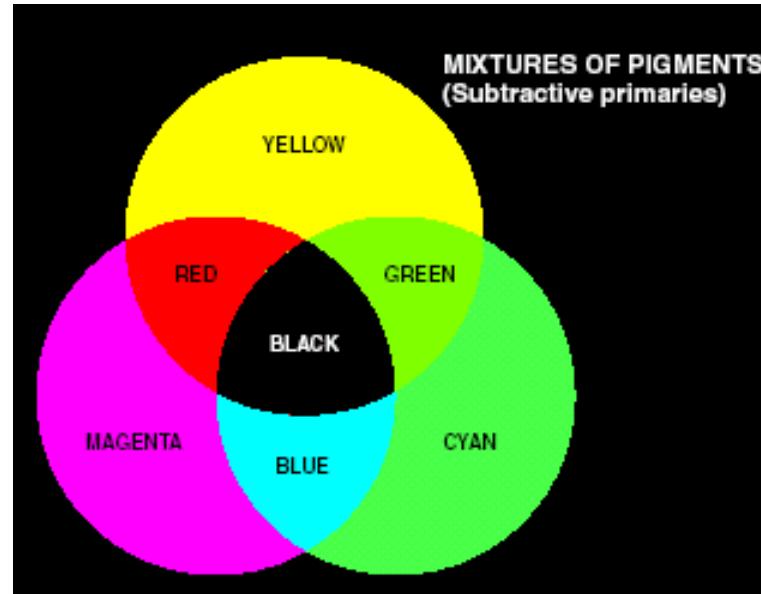


These are scalar valued pixels



CMY and CMYK Color Models

Used in
Color Printers



C = Cyan (Subtracts Red from white=1)

M = Magenta (Subtracts Green from white=1)

Y = Yellow (Subtracts Blue from white=1)

K = Black = $C + M + Y$ (instead, use Black pigment to save money)

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

Conversion of CMY to CMYK

- Begin by, $K = \min(C, M, Y)$
- For $K=1$, we have pure black and no color, i.e.,

$$C = 0, \quad M = 0, \quad Y = 0$$

- Otherwise,

$$C = (C - K) / (1 - K)$$

$$M = (M - K) / (1 - K)$$

$$Y = (Y - K) / (1 - K)$$

All values are in
the range [0,1]

- The conversion of CMYK back to CMY are,

$$C = C * (1 - K) + K$$

$$M = M * (1 - K) + K$$

$$Y = Y * (1 - K) + K$$

The HSI Color Model

- RGB is useful for hardware implementations and is serendipitously related to the way in which the human visual system works.
- However, RGB is not a particularly intuitive way in which to describe colors.
- Rather when people describe colors they tend to use hue, saturation and brightness.
- RGB is great for color generation, but HSI is great for color description.

The HSI Color Model (cont...)

Recall:

- **Hue:** A color attribute that describes a pure color (pure yellow, orange or red).
- **Saturation:** Measure of how much a pure color is diluted with white light.
- **Intensity:** Brightness is nearly impossible to measure because it is so subjective. Instead we use intensity. Intensity is the same achromatic notion that is seen in grey level images.

HSI Color Model

- RGB and CMY models are not good for human interpretation

HSI Color model:

Hue: Dominant color (i.e., Wavelength)

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

Intensity: Brightness

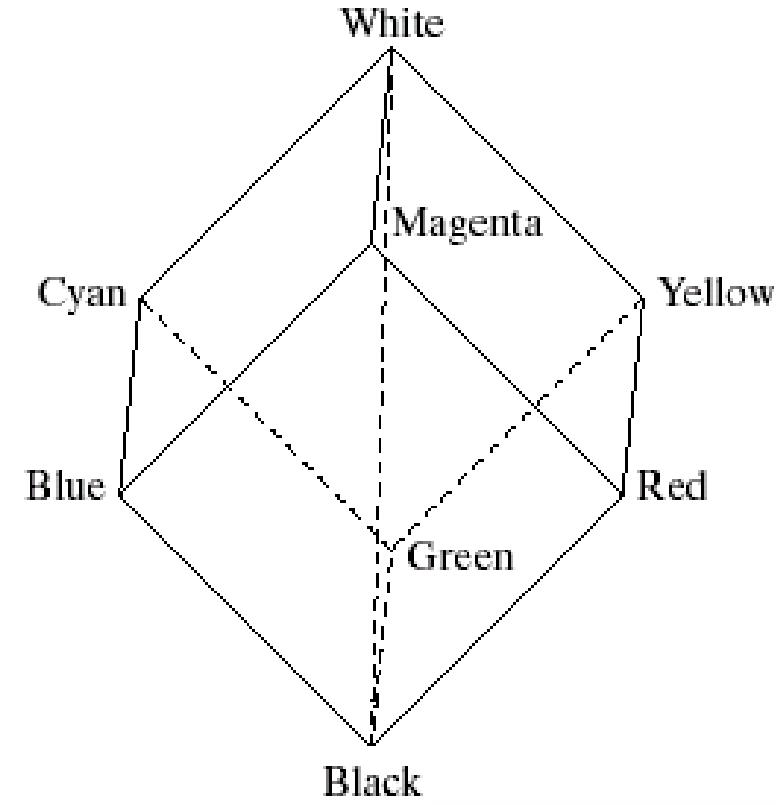
Color carrying information

HSI, Intensity & RGB

- Intensity can be extracted from RGB images.
- However, human perception of color does not refer to percentages of RGB.
- Recall: The diagonal on the RGB color cube runs from black to white.
- Now consider if we stand this cube on the black vertex and position the white vertex directly above it.

Intensity Axis

- The intensity component of any color can be determined by passing a plane perpendicular to the intensity axis and containing the color point.
- The intersection of the plane with the intensity axis gives the intensity component of the color.



a b

FIGURE 7.10
Conceptual relationships between the RGB and HSI color models.

Saturation of a Color

- The saturation of a color (percentage of white missing from the color) increases as a function of distance from the intensity axis.
- We say, Pure or Saturated color at the edges

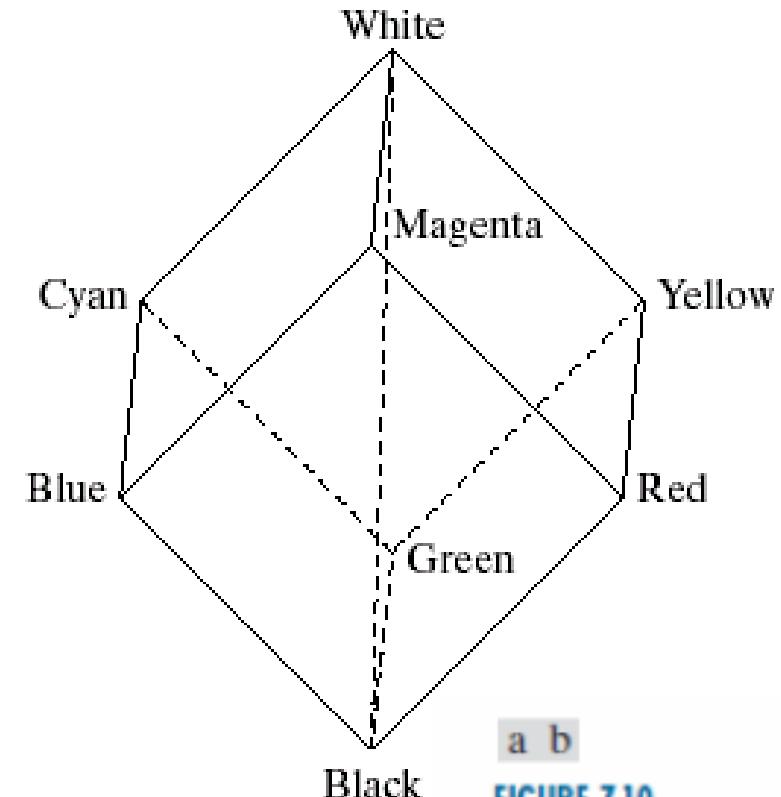


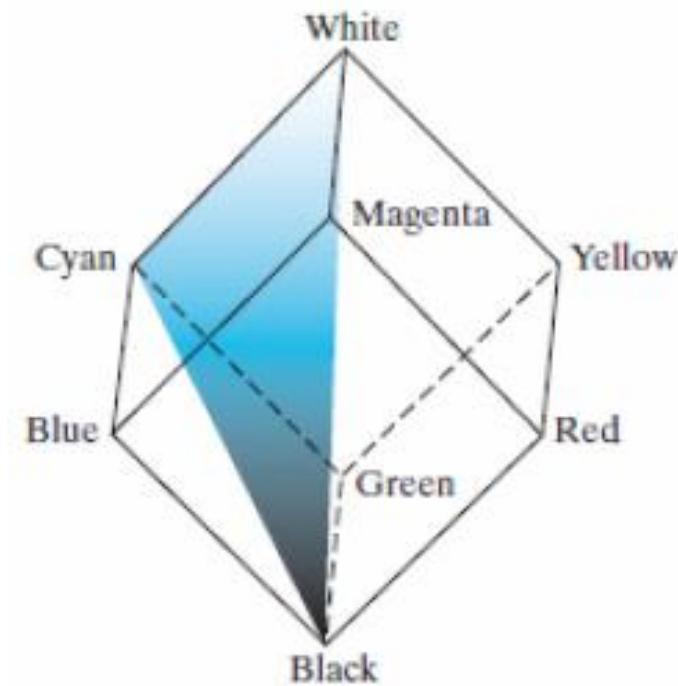
FIGURE 7.10

Conceptual relationships between the RGB and HSI color models.



HSI, Hue & RGB

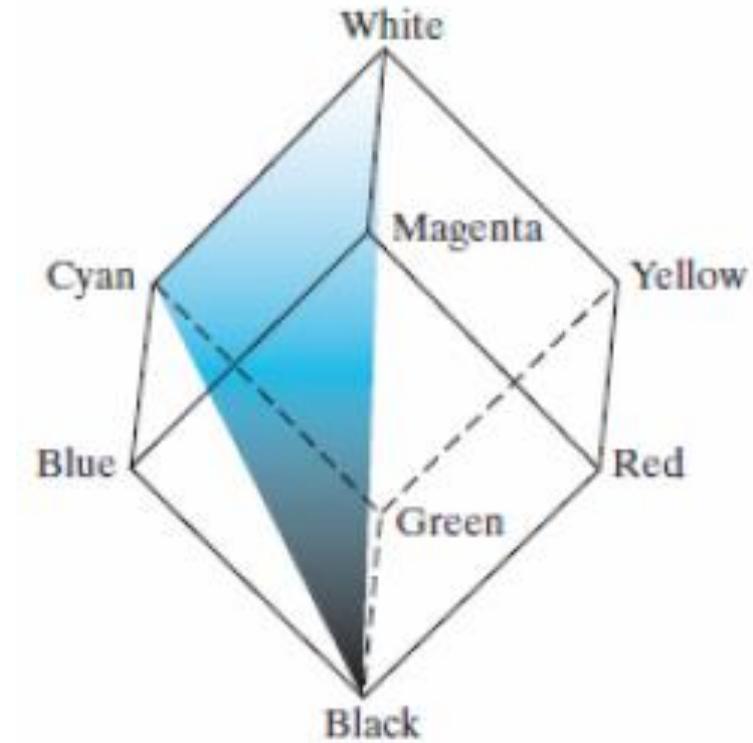
- In a similar way we can extract the hue (Color) from the RGB color cube.
- Consider a plane defined by the three points cyan, black and white.
- All points contained in this plane must have the same hue (cyan) as black and white cannot contribute hue information to a color.



a b

FIGURE 7.10
Conceptual relationships between the RGB and HSI color models.

- Rotating the shaded plane around the intensity axis we obtain different hues.
- Conclusion:
 - The HSI values can be obtained from the RGB values.
 - We need to work the geometric formulas.

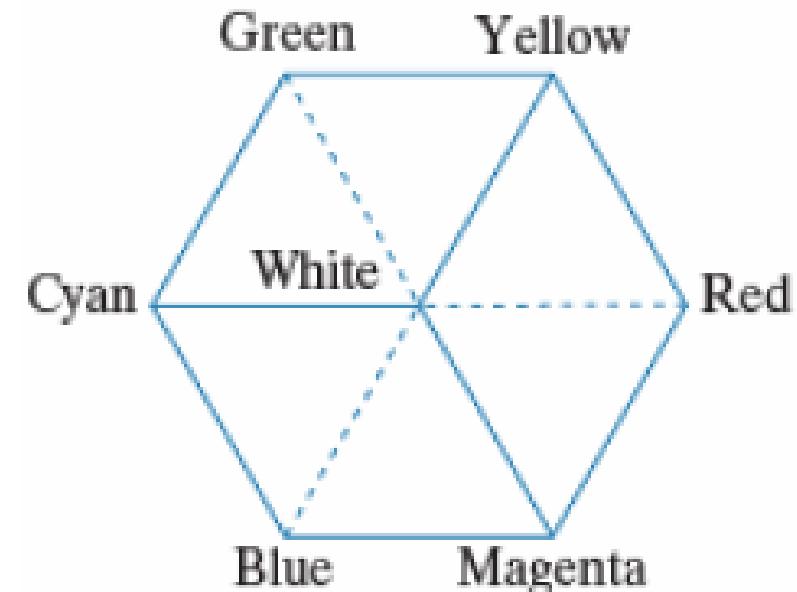


a b

FIGURE 7.10
Conceptual
relationships
between the RGB
and HSI color
models.

The HSI Color Model

- If looked straight down at the RGB cube as it was arranged previously a hexagonal shape can be seen with each primary color separated by 120° and secondary colors at 60° from the primaries.
- **HSI Model:** Composed of a vertical intensity axis and the locus of color points that lie on planes perpendicular to that axis.



The HSI Color Model (cont...)

- Hexagonal shape at an arbitrary color point
 - **Hue (H):** Determined by the angle from a reference point, usually red ($H=0$).
 - **Saturation (S):** Distance from the origin to the point.
 - **Intensity (I):** Determined by how far up the vertical intensity axis this hexagonal plane sits
 - Not apparent from this diagram

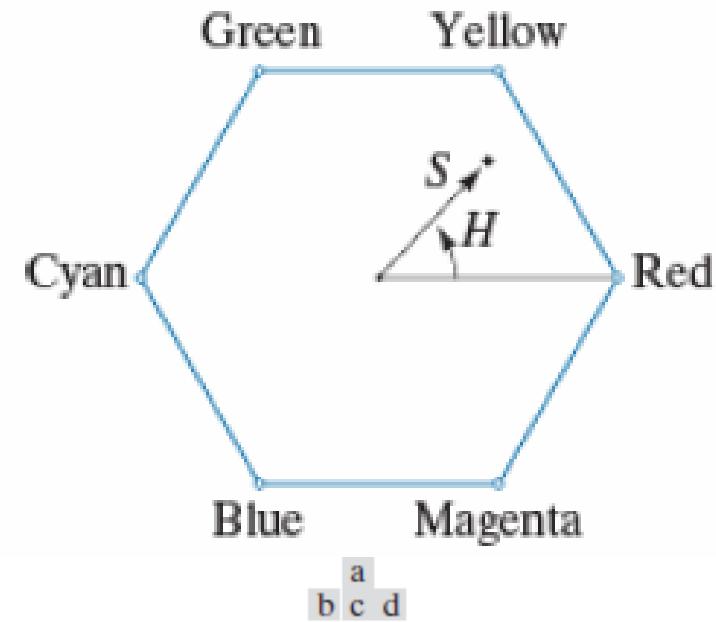


FIGURE 7.11
 Hue and saturation in the HSI color model. The dot is any color point. The angle from the red axis gives the hue. The length of the vector is the saturation. The intensity of all colors in any of these planes is given by the position of the plane on the vertical intensity axis.

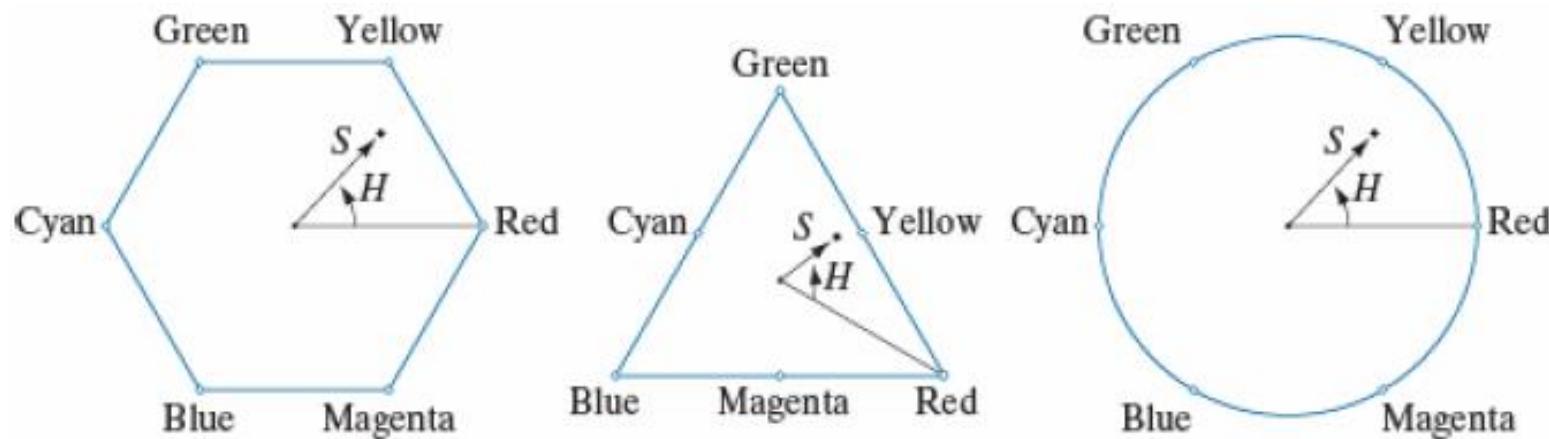
The HSI Color Model (cont...)

- Since the only important parameters are the angle and the length of the saturation vector this plane is also often represented as a circle or a triangle.

a
b c d

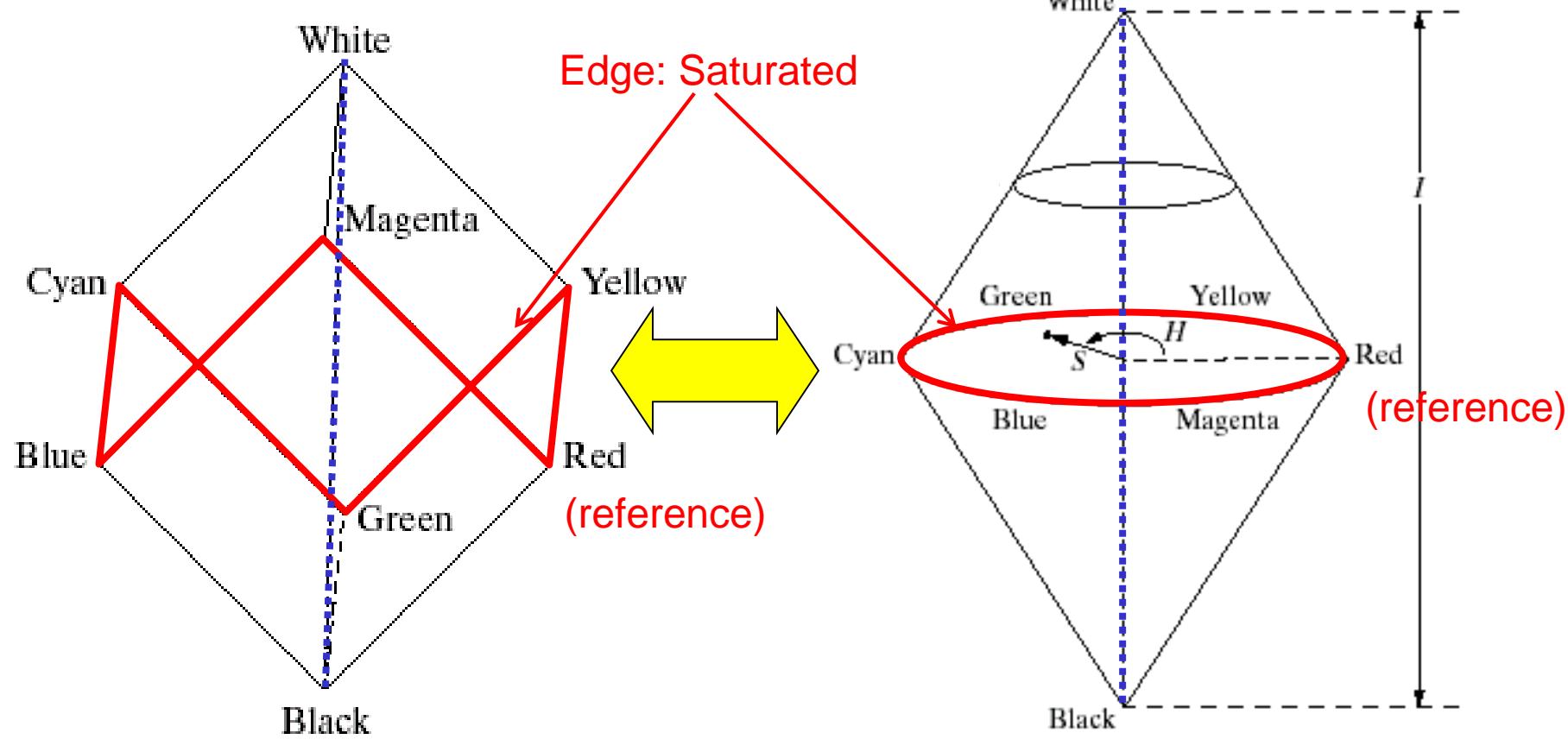
FIGURE 7.11

Hue and saturation in the HSI color model. The dot is any color point. The angle from the red axis gives the hue. 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.





Relationship Between RGB and HSI Color Models



RGB

HSI

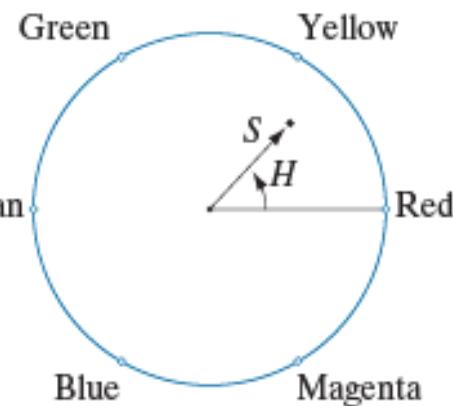
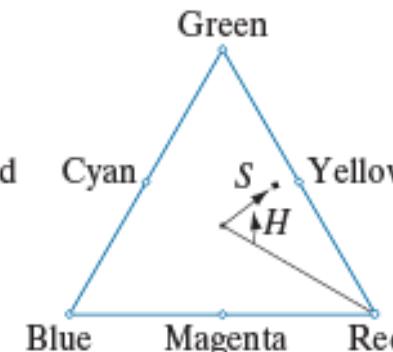
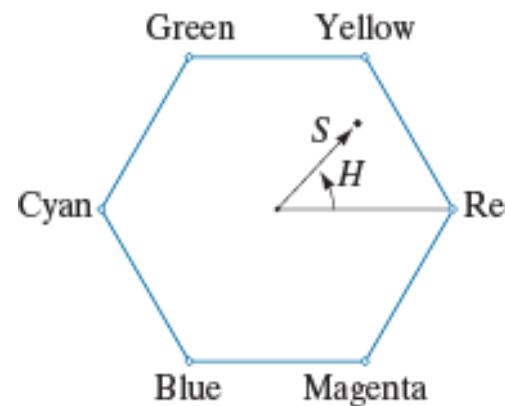
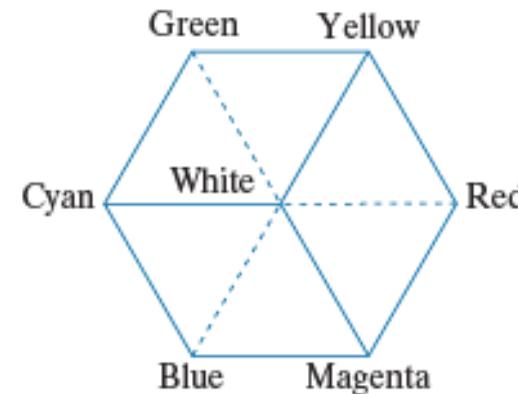


Hue and Saturation on Color Planes

a
b c d

FIGURE 7.11

Hue and saturation in the HSI color model. The dot is any color point. The angle from the red axis gives the hue. 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.



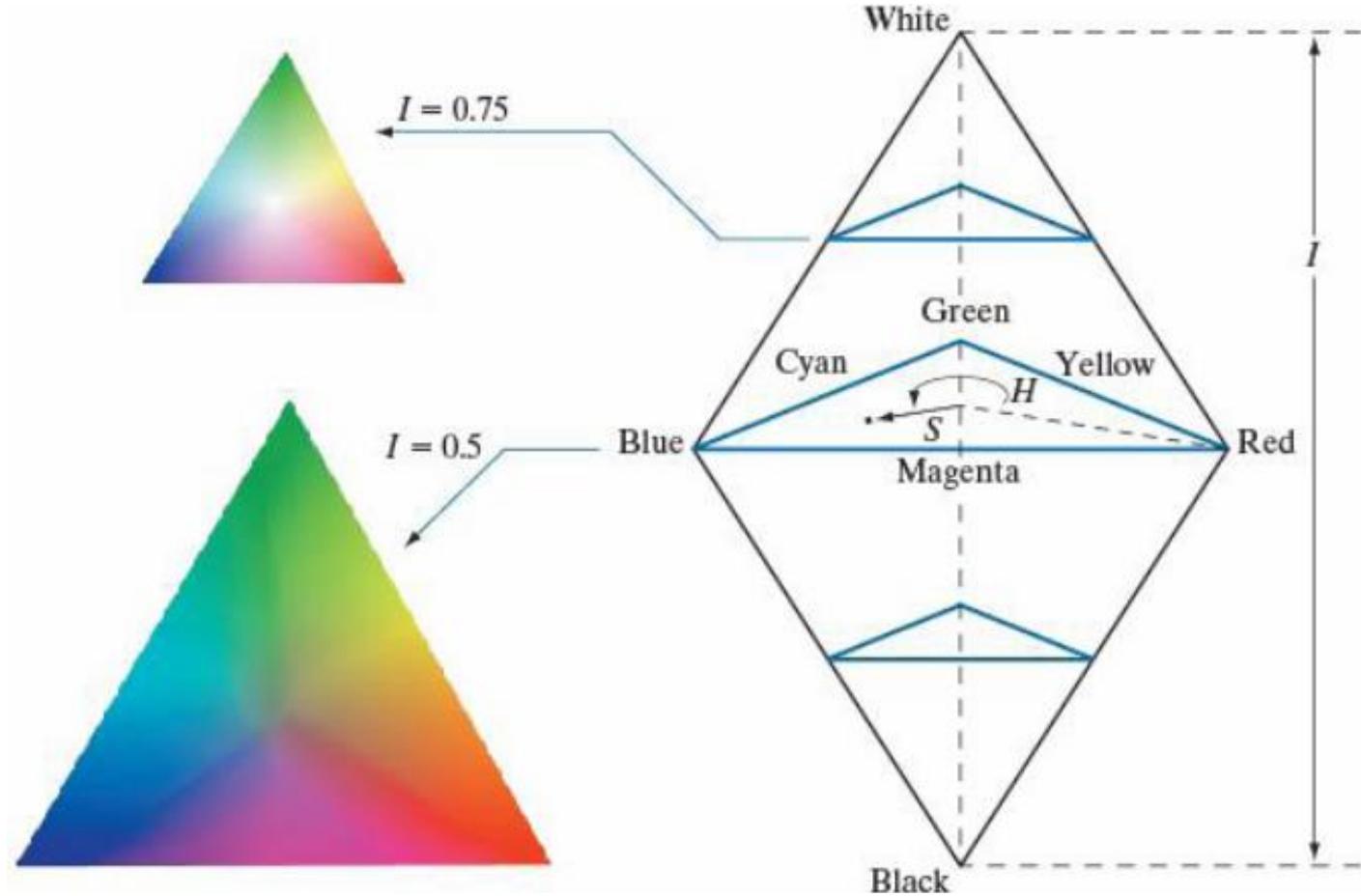
1. A Dot in the plane: An arbitrary color
2. Hue (Dominant color): An angle from the RED (reference)
3. Saturation: Distance to the Dot from Center (Max at edges)

HSI Color Model (contd..)

a
b

FIGURE 7.12

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



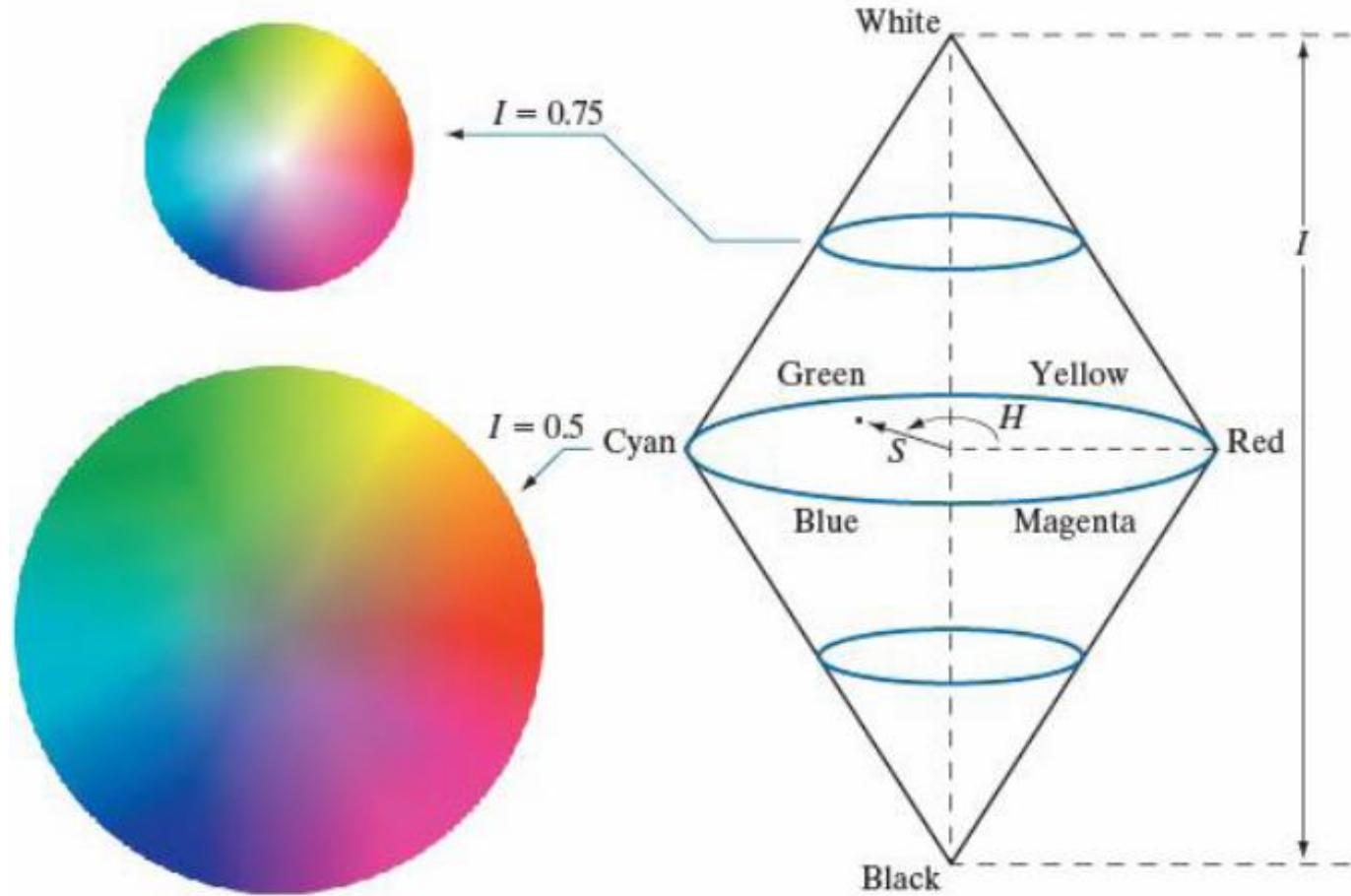
Intensity is given by a position on the vertical axis.

HSI Color Model (contd..)

a
b

FIGURE 7.12

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



Intensity is given by a position on the vertical axis.

Converting From RGB To HSI

- Given a color as R , G , and B , the H , S , and I values are calculated as follows:

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

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

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

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

Converting From HSI To RGB

RG sector: $0 \leq H < 120$

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

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

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

Given a color as H, S, and I the R, G, and B values are found using these equations

BR sector:

GB sector: $120 \leq H < 240$

$$H = H - 120$$

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

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

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

$240 \leq H \leq 360$

$$H = H - 240$$

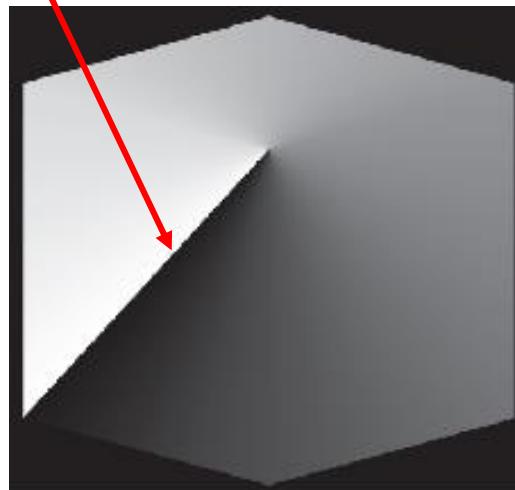
$$B = I \left[1 + \frac{S \cos H}{\cos(60^\circ - H)} \right]$$

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

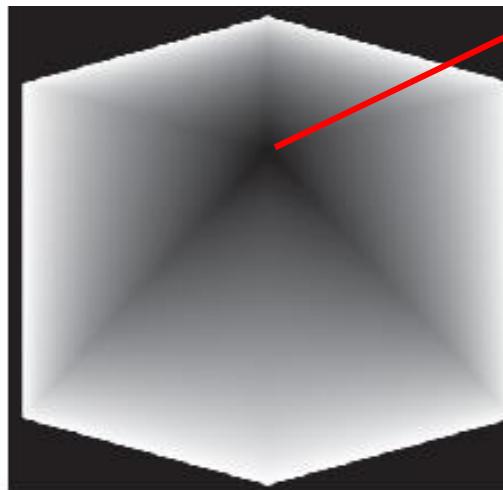
$$R = 1 - (G + B)$$

H, S, and I Components of RGB Color Cube

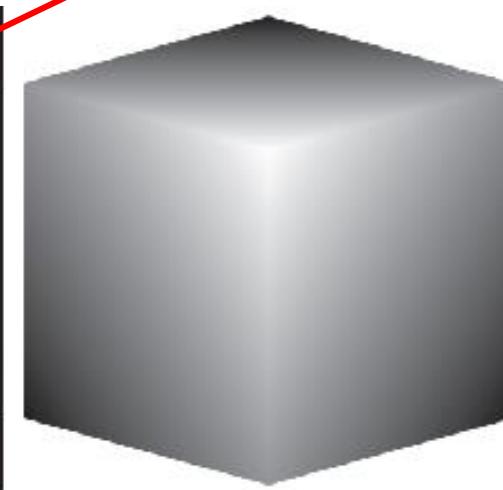
Cross-over from 0°
(white) to 360° (black)



Hue



Saturation



Intensity

a b c



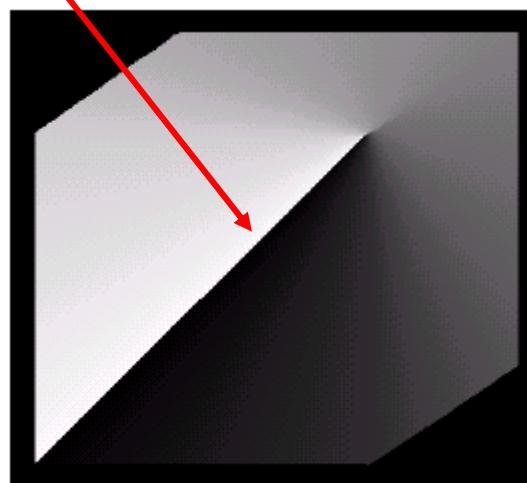
FIGURE 7.8
A 24-bit RGB
color cube.

RGB Color Cube

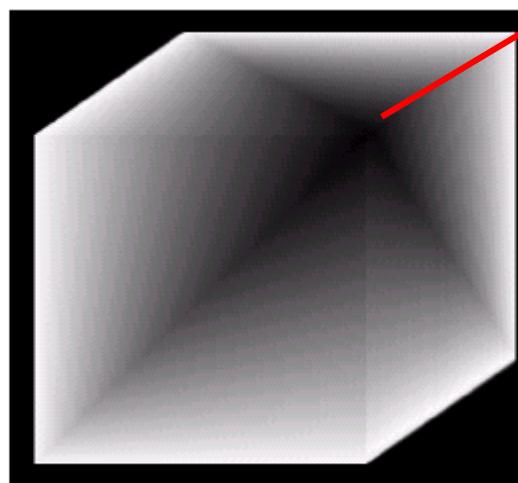
Low → Less
saturated

H, S, and I Components of RGB Color Cube

Cross-over from 0°
(white) to 360° (black)



Hue



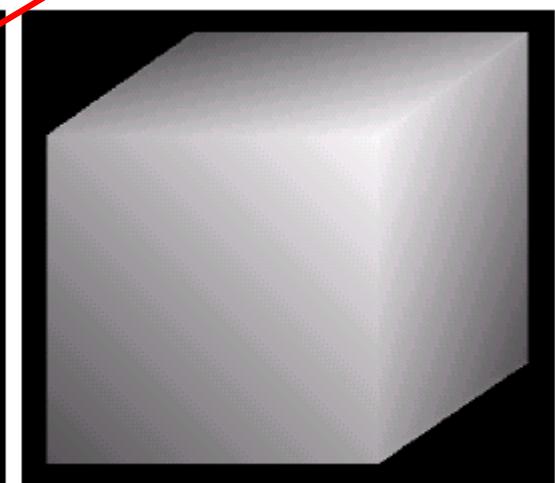
Saturation

RGB Color Cube



FIGURE 6.8 RGB
24-bit color cube.

Low → Less
saturated



Intensity

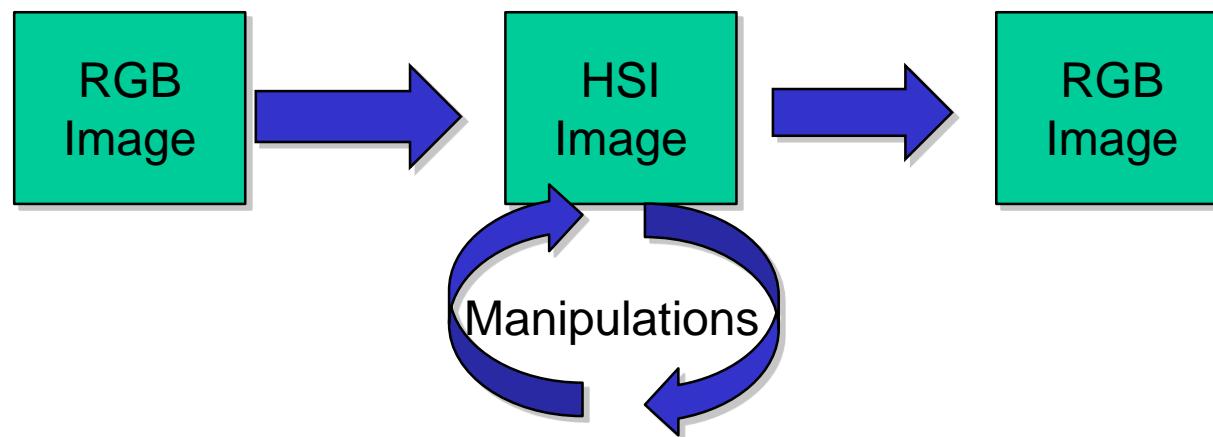
a b c

3rd Edition

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

Manipulating Images in the HSI Model

- Steps for manipulating an image under the HSI model:
 - First convert image from RGB to HSI
 - Perform manipulations under HSI as desired
 - Finally convert the image back from HSI to RGB

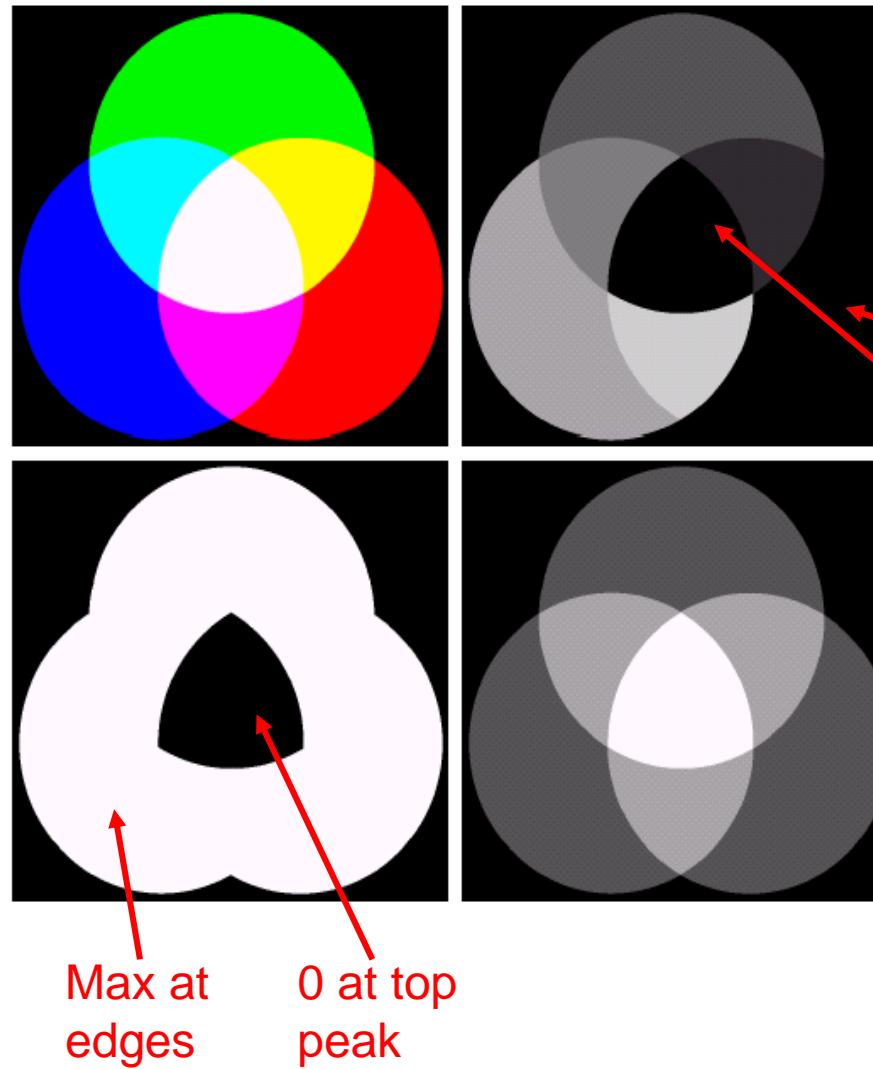


RGB Image:
Primary and
Secondary
Colors

Saturation
(distance
from origin)

a b
c d

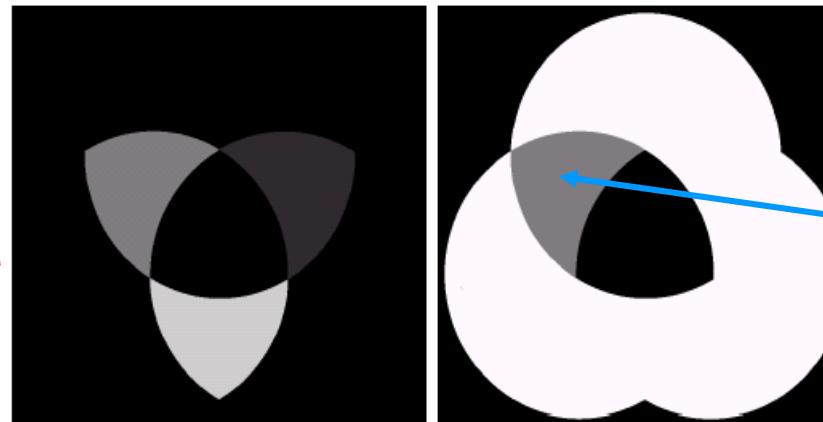
FIGURE 7.14
(a) RGB image
and the
components of
its corresponding
HSI image:
(b) hue,
(c) saturation, and
(d) intensity.



Manipulating HSI Components of Images

Hue component:

Change blue and green pixels to 0 in Fig 7.14(b) → converts to Red

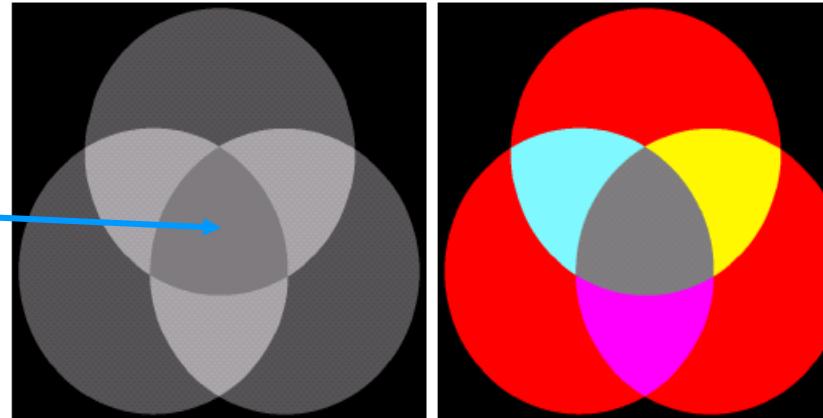


Saturation Component:

Reduce by half the saturation of the cyan $[1 \rightarrow 0.5]$ region in Fig 7.14(c)

Intensity component:

Reduce by half the intensity of the central white region $[1 \rightarrow 0.5]$ in Fig 7.14(d)



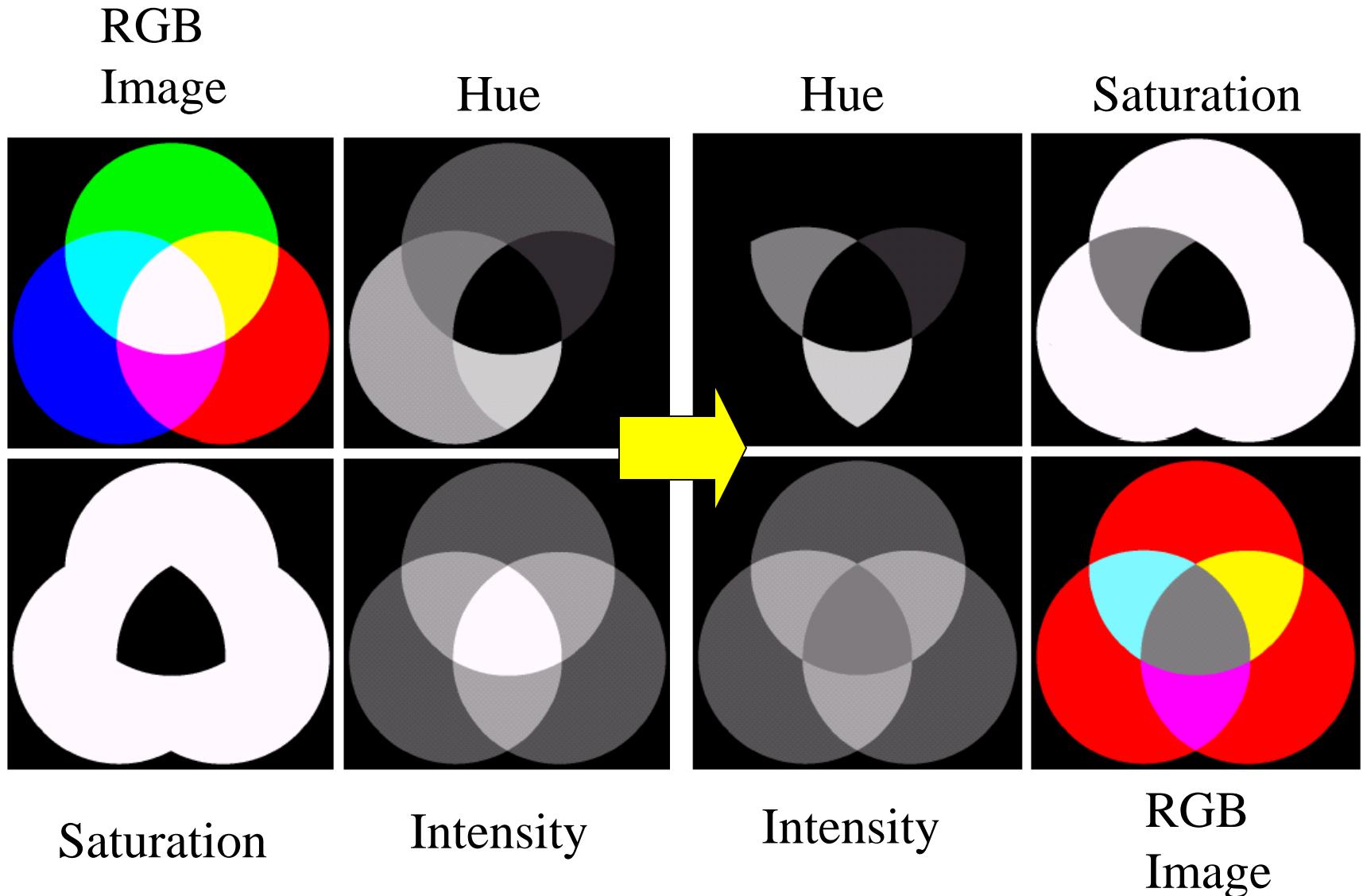
The modified result image

a	b
c	d

FIGURE 7.15

(a)-(c) Modified HSI component images.
(d) Resulting RGB image. (See Fig. 7.14 for the original HSI images.)

Example: Manipulating HSI Components



Two types of color image processing

- **Pseudocolor image processing:** Assign colors to gray values based on a specific criterion
 - Gray scale images to be processed may be a single image or multiple bands in images such as multispectral or hyperspectral images.
- **Full color image processing:** Manipulate real color images such as color photographs

7.3 Pseudocolor Image Processing

Pseudo color = false color : In some cases there is no “color” concept for a gray scale image. But “false” colors can be assigned to an image.

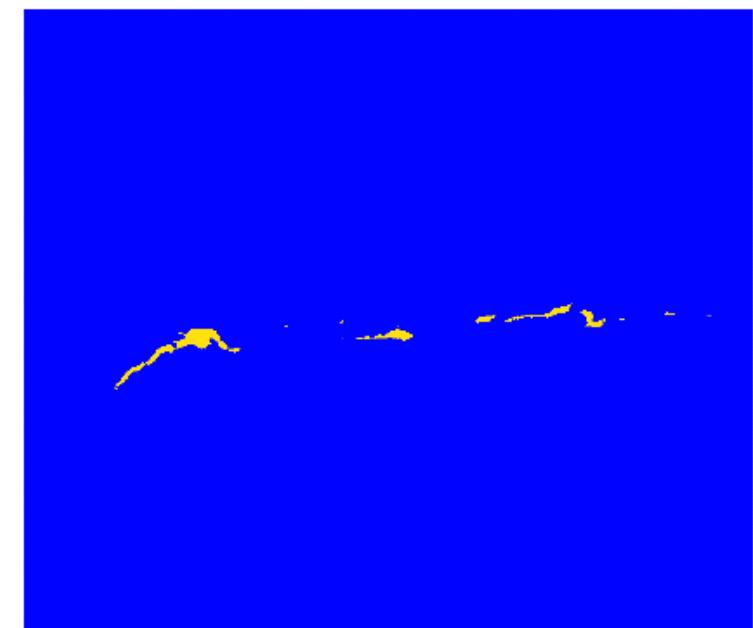
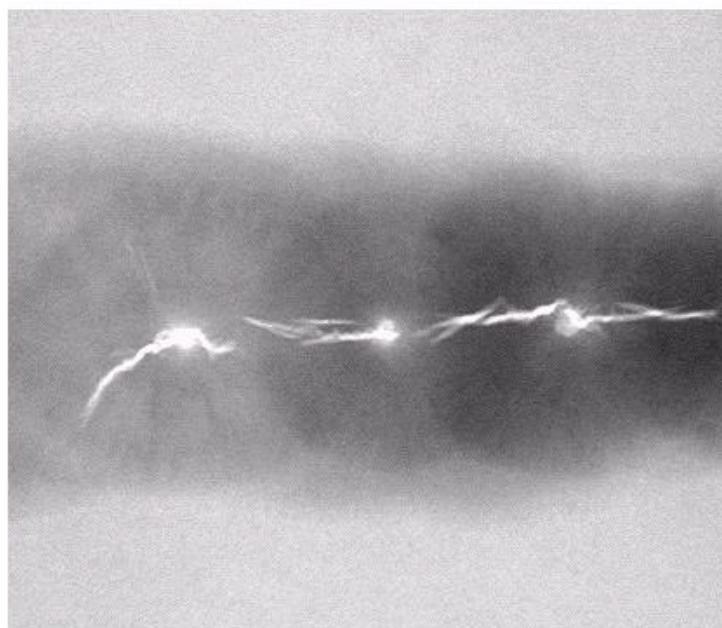
Why we need to assign colors to gray scale image?

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

a b

FIGURE 7.19

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



7.3 Pseudocolor Image Processing

- Pseudocolor (also called **false color**) image processing consists of assigning colors to gray values based on a specific criterion.
- The principle use of Pseudocolor image processing is **for human visualization**.
 - Humans can discern between thousands of color shades and intensities, compared to only about two dozen or so shades of gray

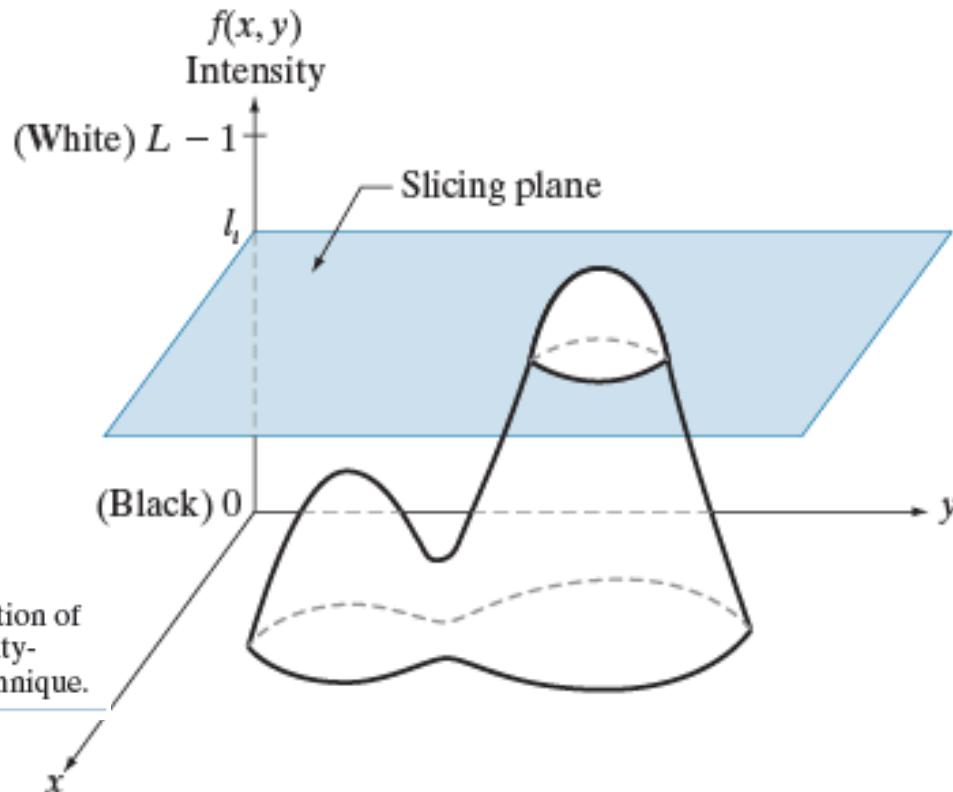


- **Intensity Slicing and Color Coding:** Simplest kind of pseudocolor image processing.
- First consider an image as a 3D function mapping spatial coordinates to intensities
 - Can be considered as heights:
 - ✓ Black → Minimum intensity
 - ✓ White → Maximum intensity
- Consider placing planes at certain levels parallel to the coordinate plane.
- If a value is on one side of such a plane it is rendered in one color, but a different color if on the other side.

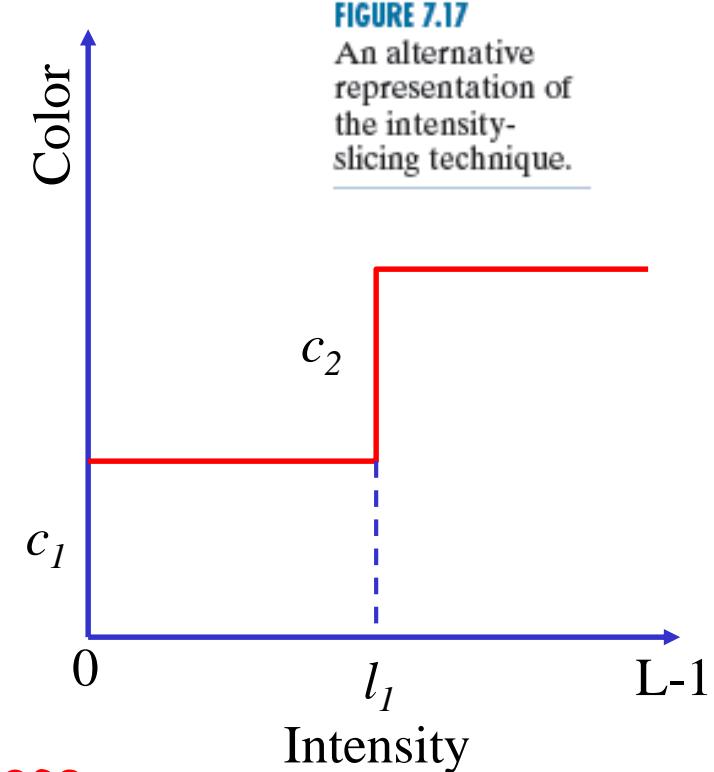
Intensity Slicing or Density Slicing

Formula:

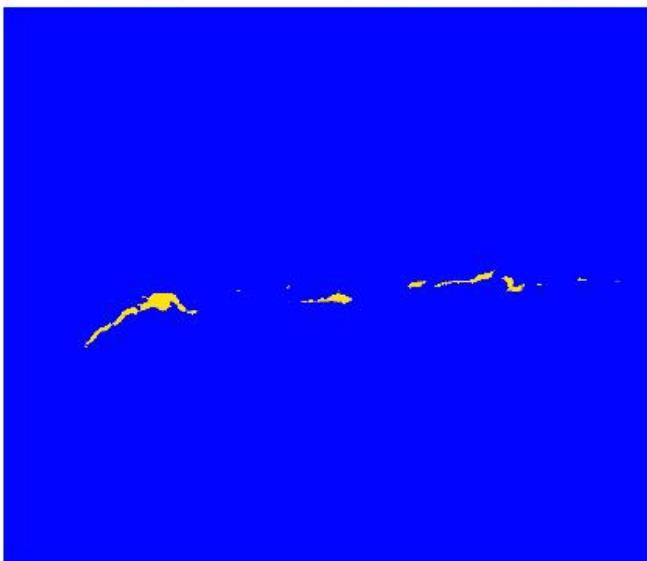
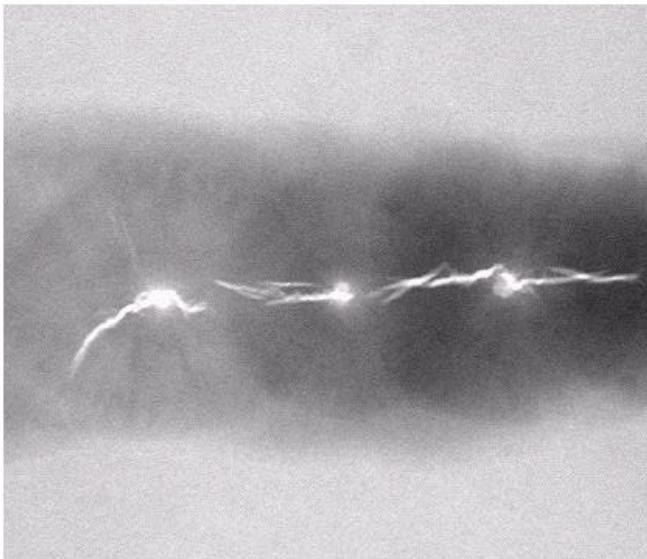
$$f(x, y) = \begin{cases} c_1 & \text{if } f(x, y) \leq l_1 \\ c_2 & \text{if } f(x, y) > l_1 \end{cases} \rightarrow \begin{aligned} c_1 &= \text{Color No. 1} \\ c_2 &= \text{Color No. 2} \end{aligned}$$



A gray scale image viewed as a 3D surface.



Intensity Slicing Example



An X-ray image of a weld with cracks

a b

FIGURE 7.19

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

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

Summary of more general Intensity Slicing :

- Let $[0, L-1]$ represent the grey scale
- Let l_0 represent Black $[f(x, y) = 0]$ and let l_{L-1} represent White $[f(x, y) = L-1]$
- Suppose P planes perpendicular to the intensity axis are defined at levels l_1, l_2, \dots, l_P
- Assume $0 < P < L-1$: P -planes partition the grey scale into $P + 1$ intervals V_1, V_2, \dots, V_{P+1}

- Grey level color assignments can then be made according to the relation:

$$f(x, y) = c_k \quad \text{if } f(x, y) \in V_k$$

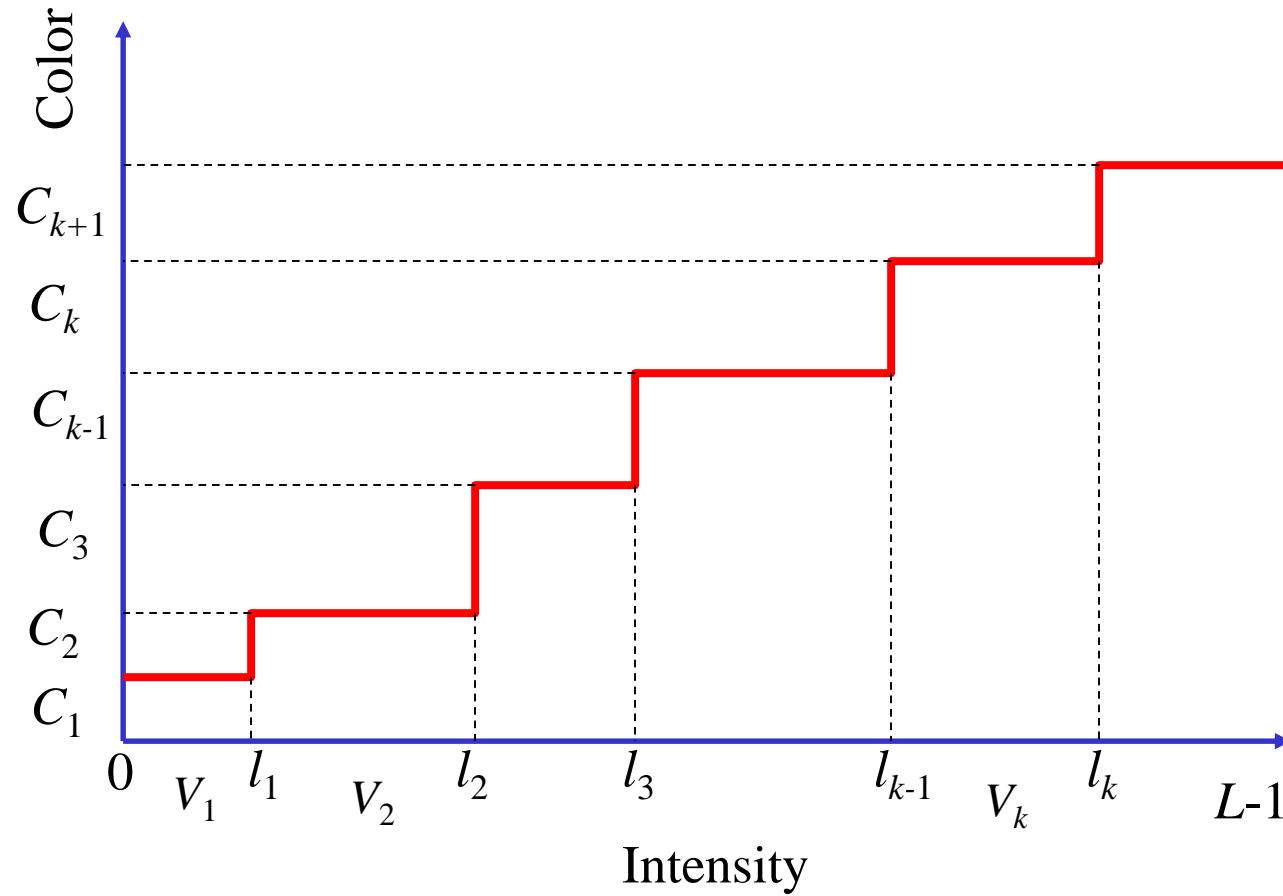
where,

- c_k : Color associated with the k^{th} intensity level V_k
- V_k : Defined by the partitioning planes between $l = k - 1$ and $l = k$

Multi-Level Intensity Slicing

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

C_k = Color No. k
 l_k = Threshold level k





Multi Level Intensity Slicing Example

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

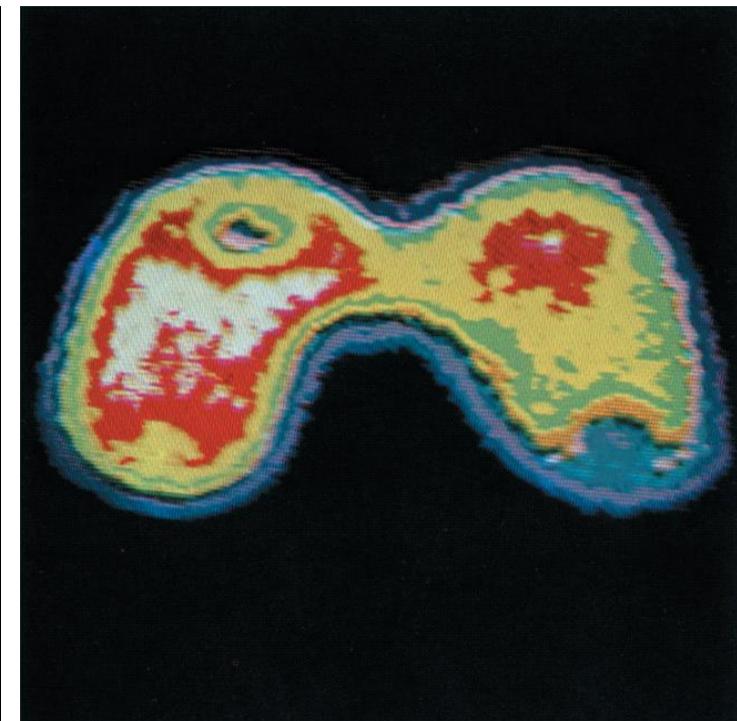
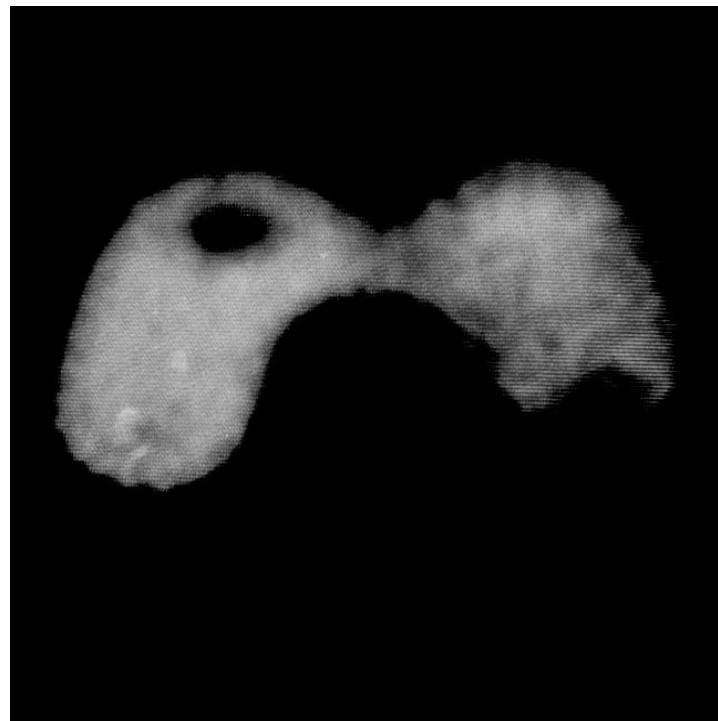
C_k = Color No. k

l_k = Threshold level k

a b

FIGURE 7.18

(a) Grayscale image of the Picker Thyroid Phantom.
(b) Result of intensity slicing using eight colors.
(Courtesy of Dr. J. L. Blankenship,
Oak Ridge National Laboratory.)

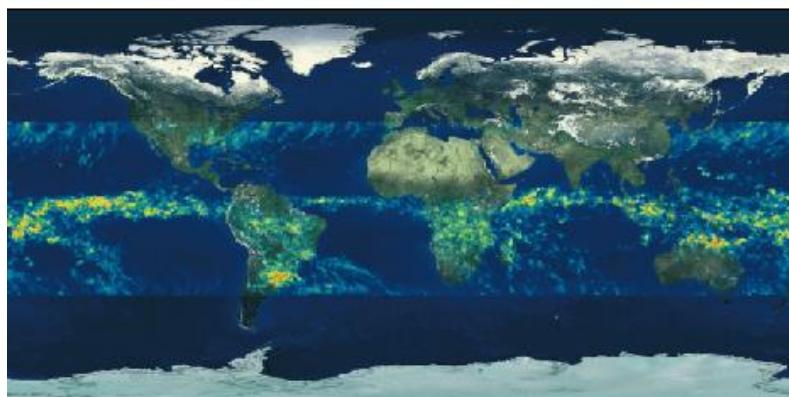
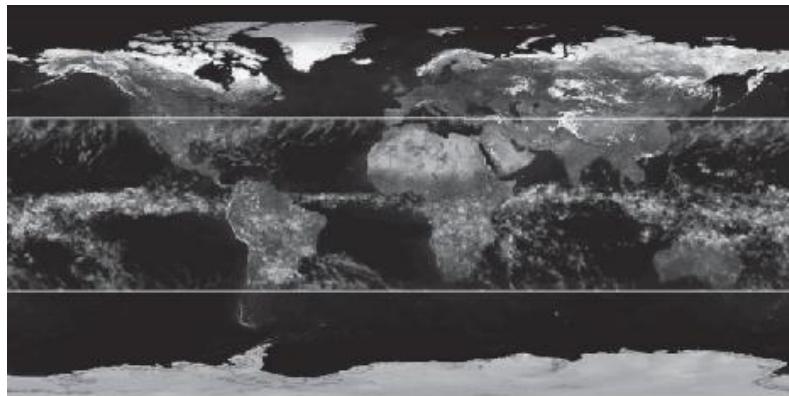


An X-ray image of the Picker Thyroid Phantom

After density slicing into 8 colors



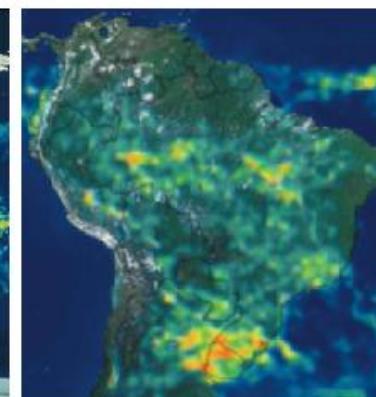
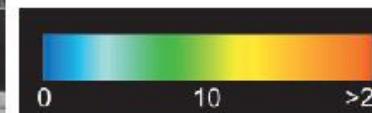
Pseudocolor Image Processing for Rainfall Detection



Color coded image

a b
c d

Each pixel in these images represents a physical land area whose size depends on the resolution of the sensors.

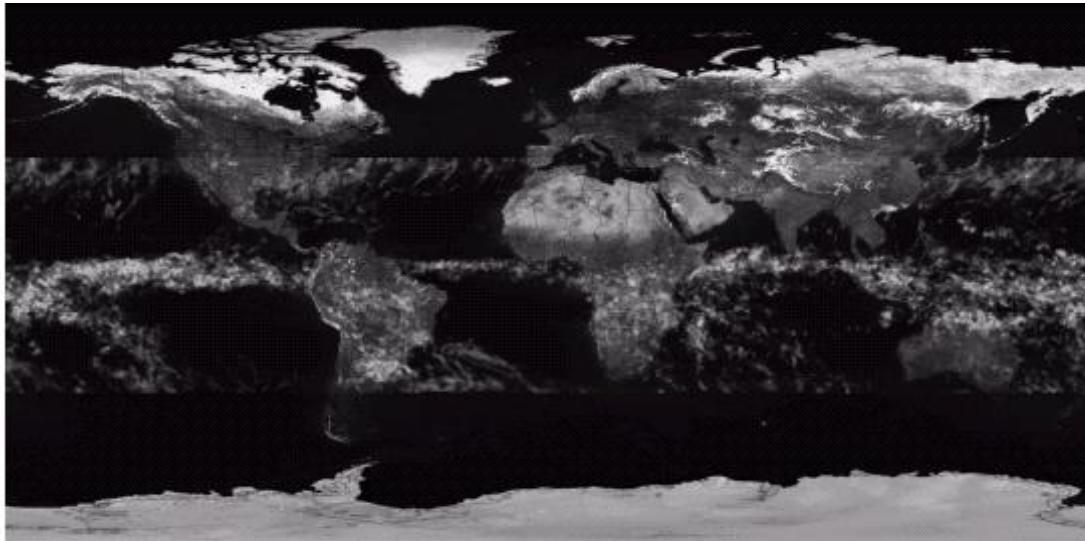


Gray-scale image of average monthly rainfall.

South America region

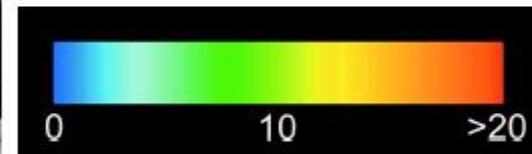
FIGURE 7.20 (a) Grayscale image in which intensity (in the horizontal band shown) corresponds to average monthly rainfall. (b) Colors assigned to intensity values. (c) Color-coded image. (d) Zoom of the South American region. (Courtesy of NASA.)

Color Coding Example

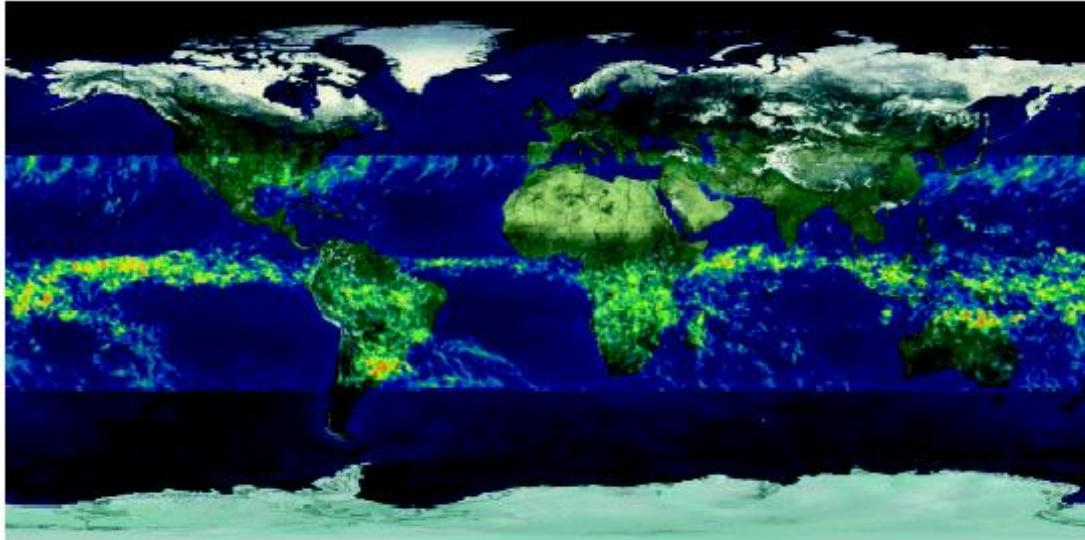


A unique color is assigned to each intensity value.

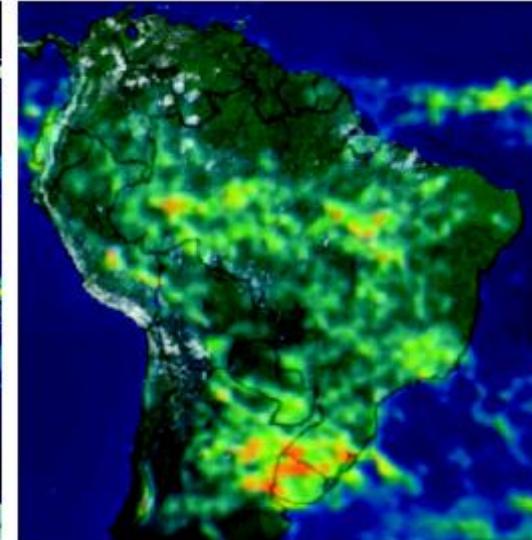
Gray-scale image of average monthly rainfall.



Color map



Color coded image



South America region

Electrical Engineering

7.3 Gray Level to Color Transformation

Assigning colors to gray levels based on specific mapping functions

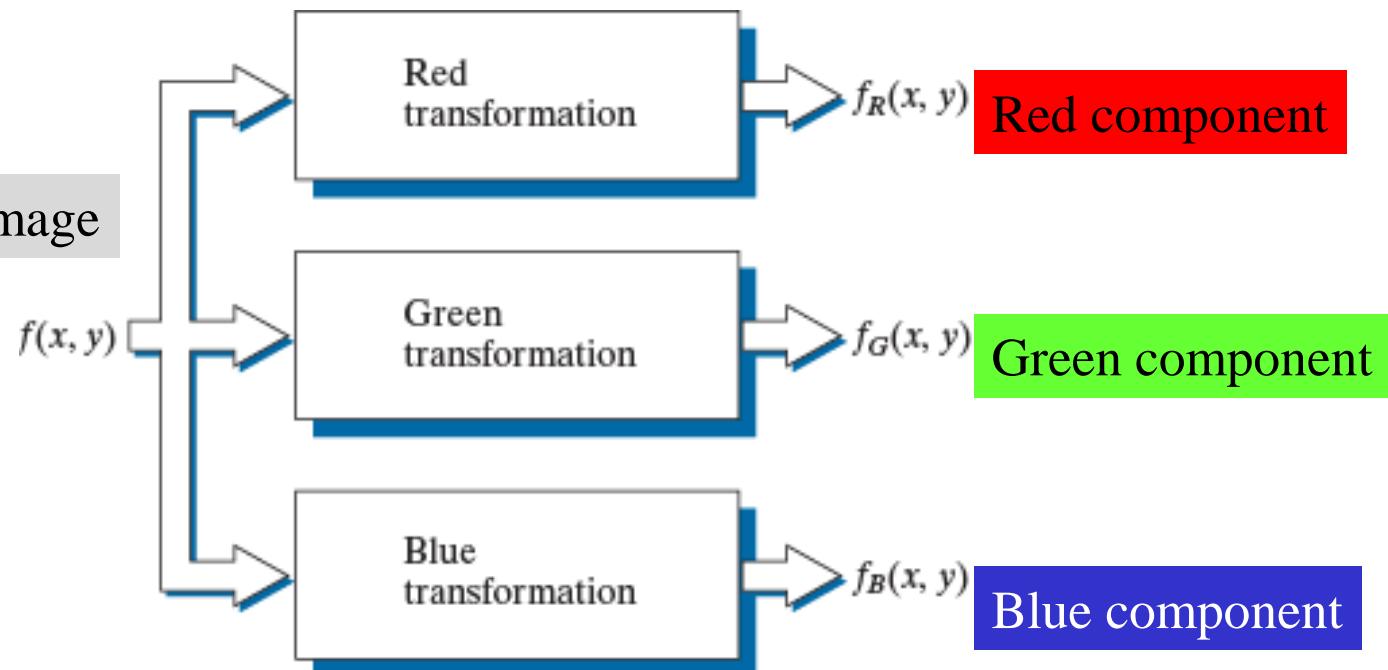


FIGURE 7.21

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

- **Gray level to color transformation for achieving a wider range of pseudocolor enhancement results.**

7.3 Pseudocolor Image Processing Intensity to Color Transformation

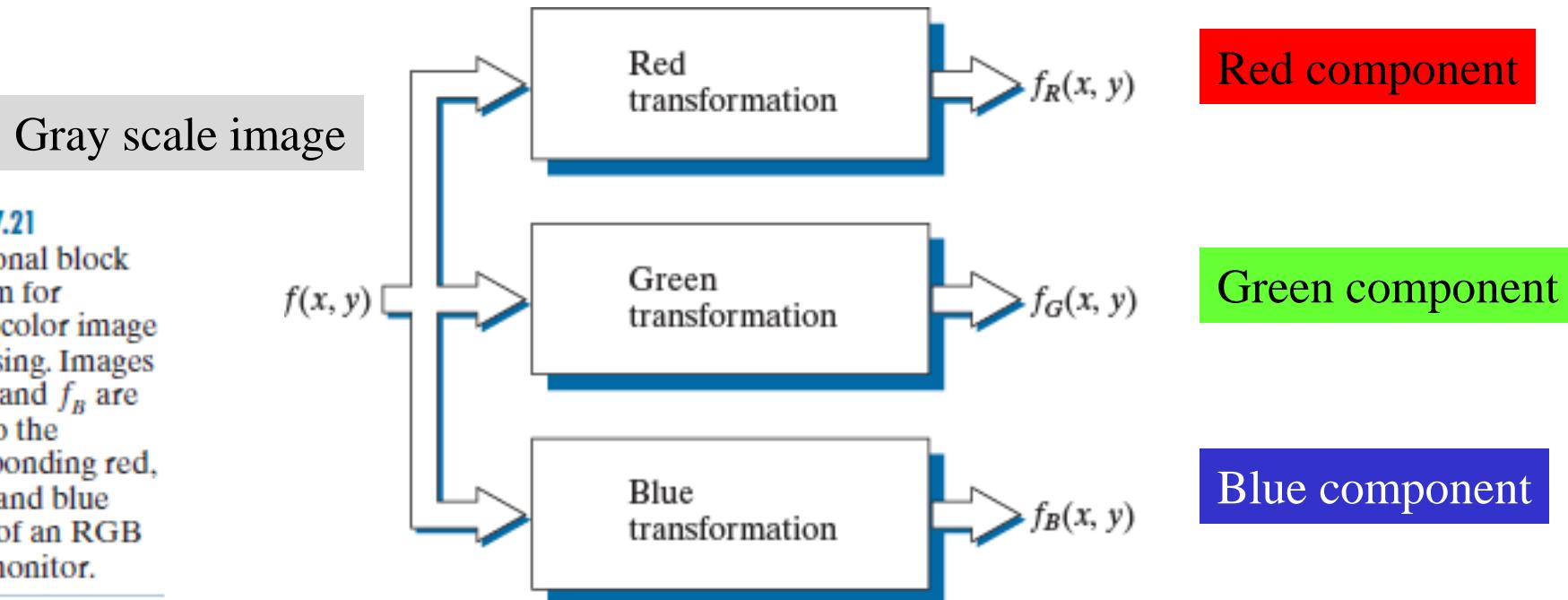


FIGURE 7.21
Functional block diagram for pseudocolor image processing. Images f_R , f_G , and f_B are fed into the corresponding red, green, and blue inputs of an RGB color monitor.

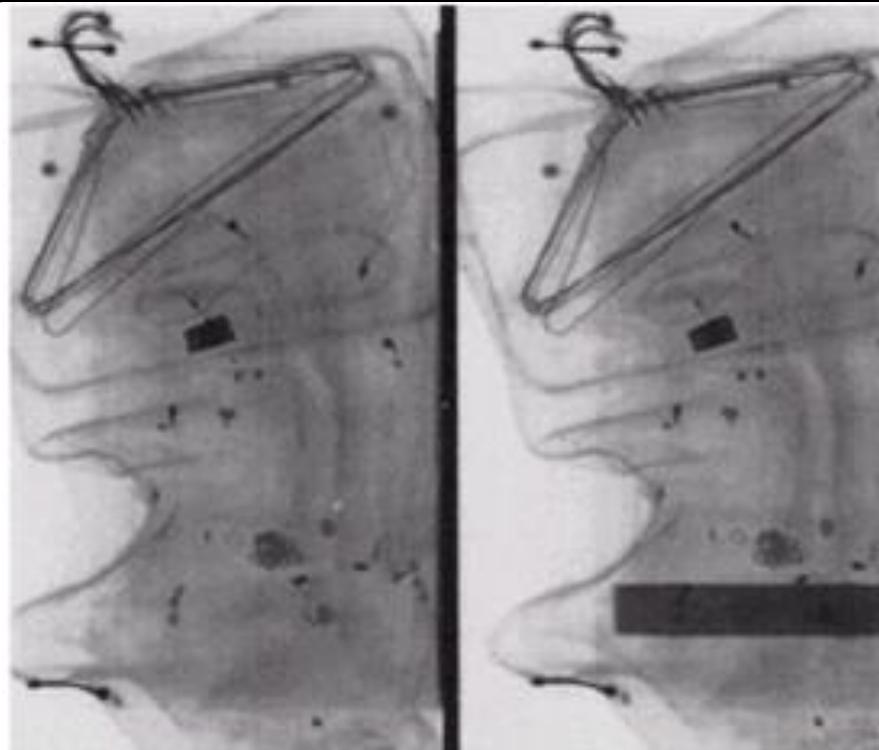
- Three *independent transformations* of intensity
- Results fed into R, G, B channels *separately*
- Resulting composite image highlights certain image parts



Pseudocolor Image Processing Intensity to Color Transformation (cont.)

a
b c

FIGURE 7.22
Pseudocolor enhancement by using the gray level to color transformations in Fig. 7.23. (Original image courtesy of Dr. Mike Hurwitz, Westinghouse.)



- X-ray images from airport scanning system
- The image on the right contains *simulated plastic explosives*

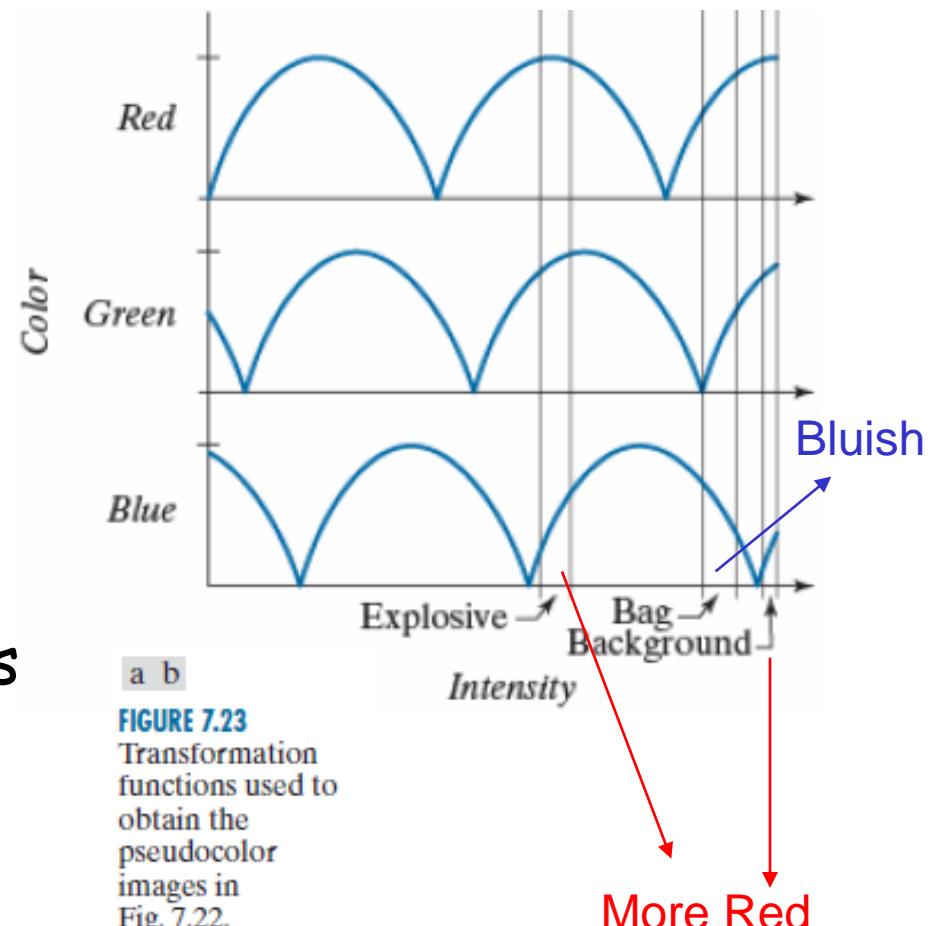


Pseudocolor Image Processing

Intensity to Color Transformation (cont.)

- Sinusoidal transformation functions
- Changing the phase or the frequency of the transformation functions can emphasize ranges in the gray scale:
 - A small change in the phase between the transformations assigns a strong color to the pixels with intensities in the valleys.

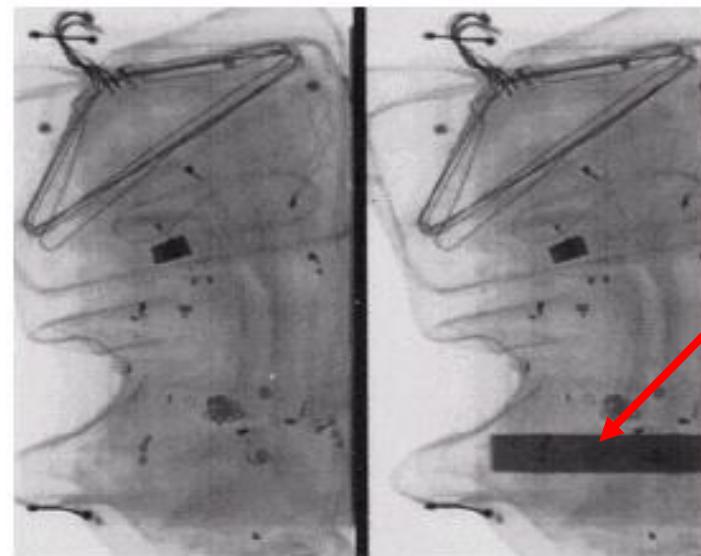
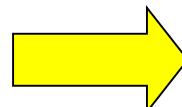
3 Independent Transformations



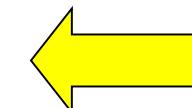


Gray Level to Color Transformation-1

An X-ray image
of a garment bag



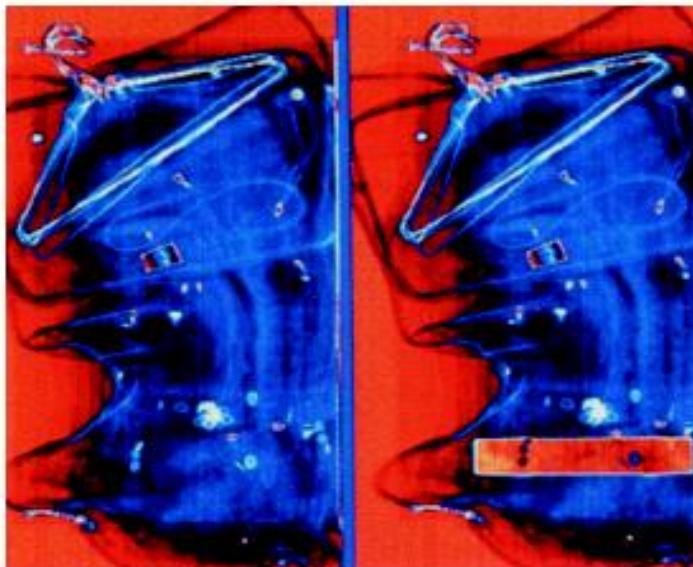
An X-ray image of a
garment bag with a
**a simulated explosive
device**



a
b
c

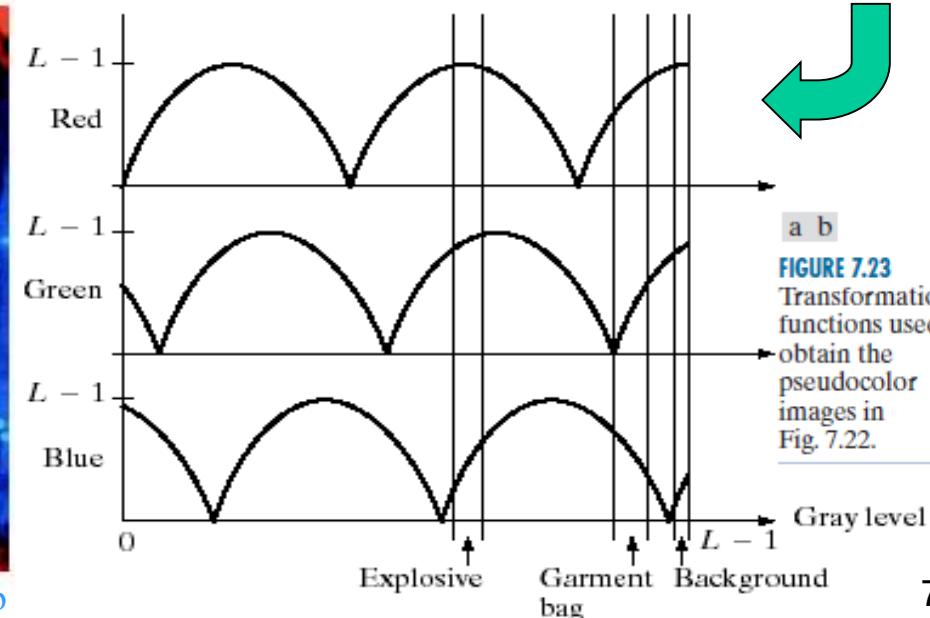
FIGURE 7.22
Pseudocolor
enhancement by
using the gray
level to color
transformations in
Fig. 7.23. (Original
image courtesy of
Dr. Mike Hurwitz,
Westinghouse.)

Color
coded
images



Arbab

3 Independent
Transformations



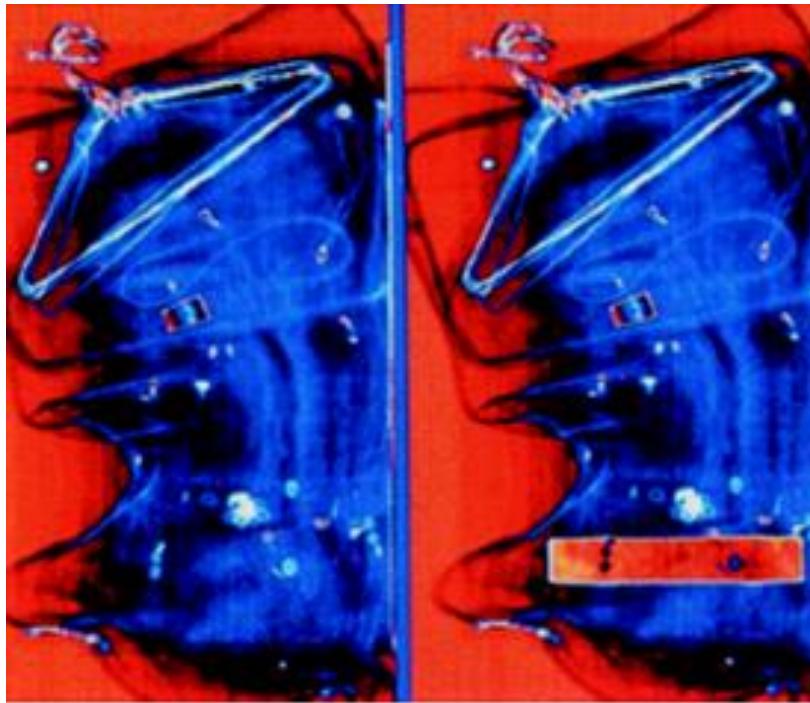
a b

FIGURE 7.23
Transformation
functions used to
obtain the
pseudocolor
images in
Fig. 7.22.

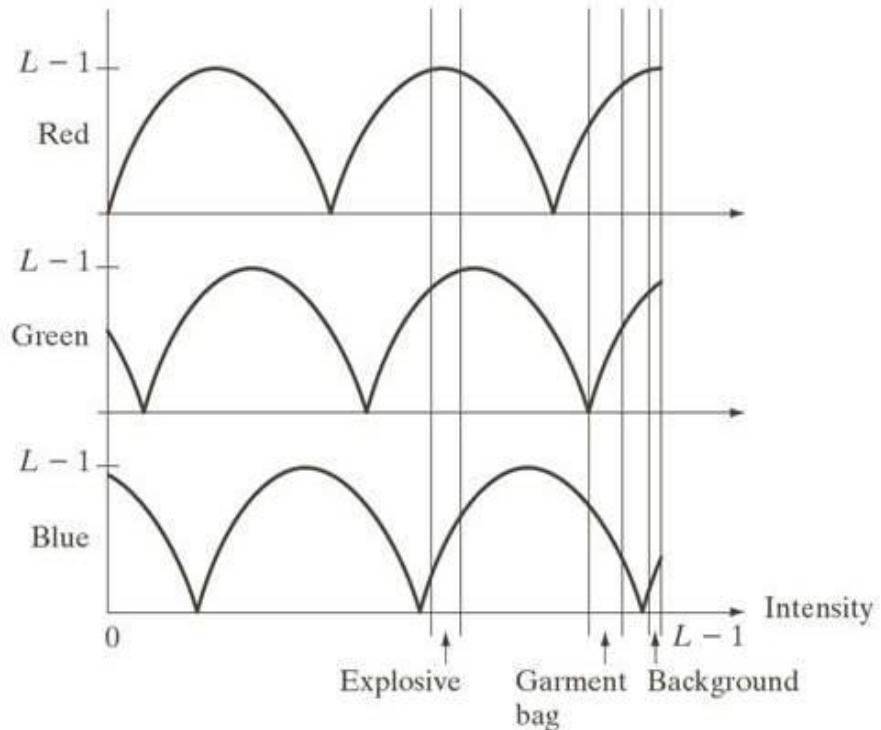


Pseudocolor Image Processing

Intensity to Color Transformation (cont.)



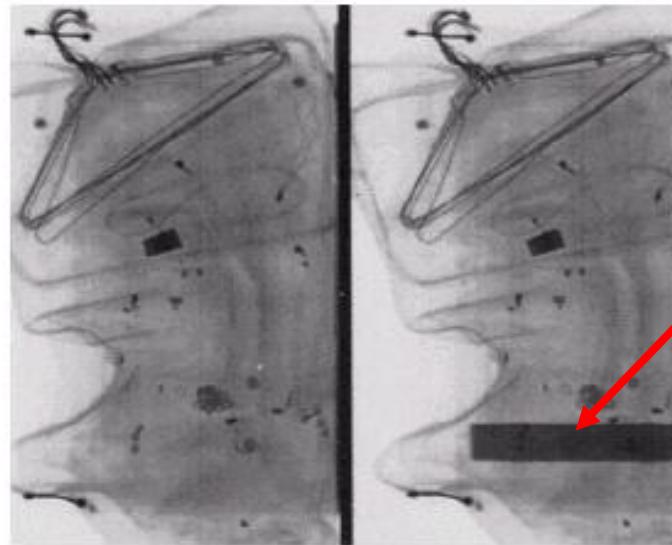
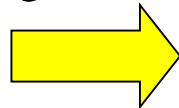
3 Independent Transformations



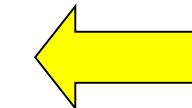
- Background and explosives are coded with approximately the same color although they differ
- This is due to the periodicity of the sine waves
- Done on a case-by-case basis

Gray Level to Color Transformation-2

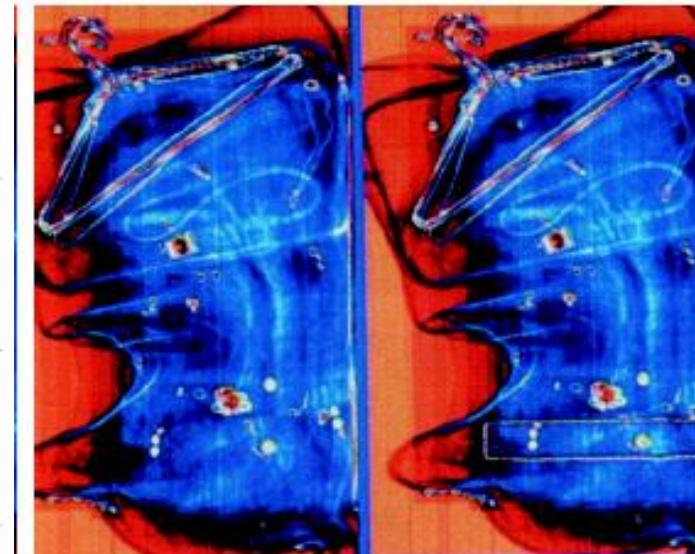
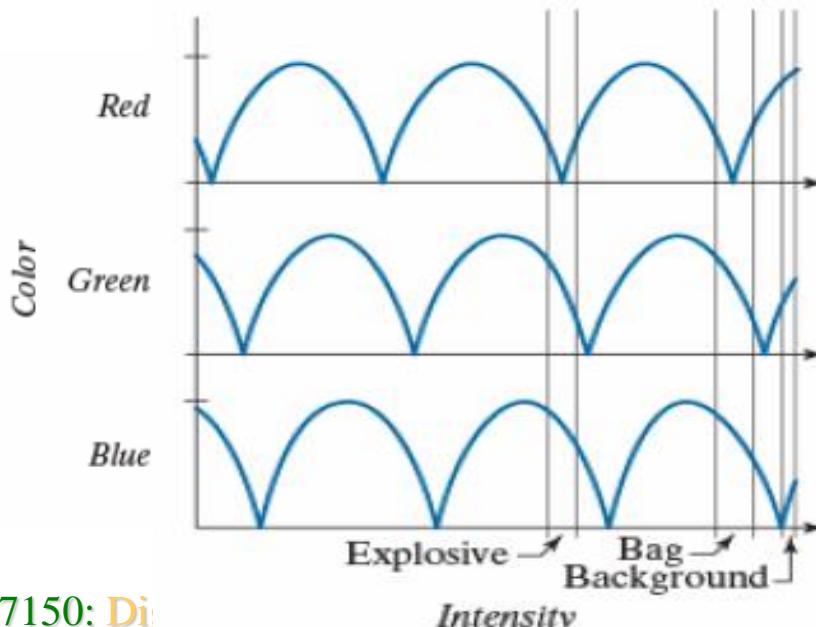
An X-ray image
of a garment bag



An X-ray image of a
garment bag with a
**simulated explosive
device**



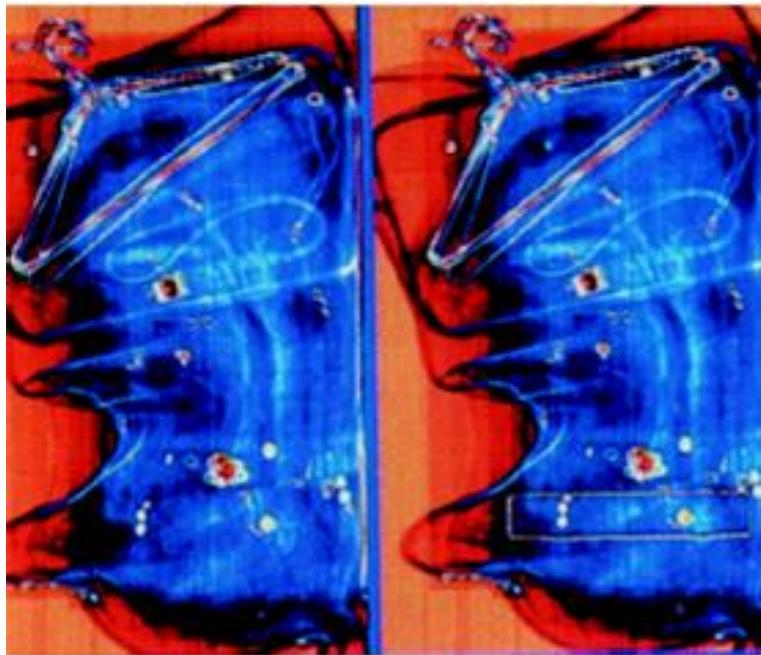
Transformations



Color
coded
images



Pseudocolor Image Processing Intensity to Color Transformation (cont.)



3 Independent Transformations

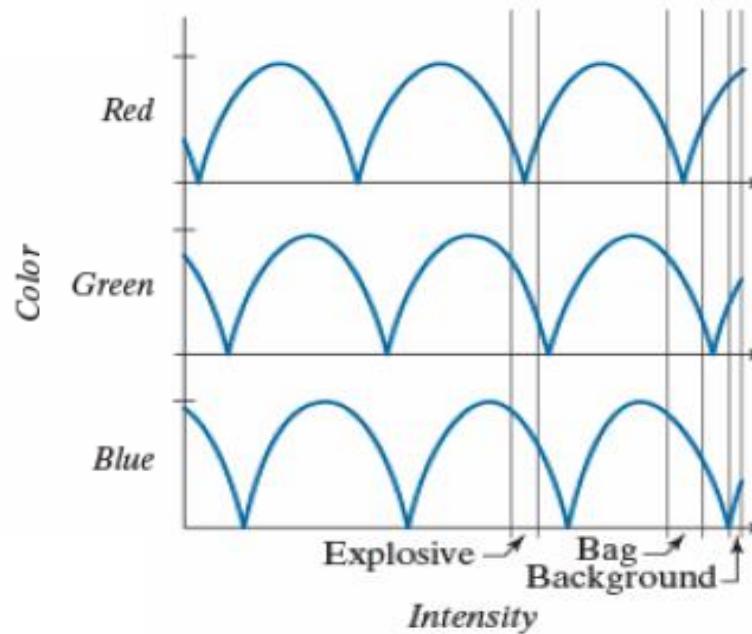


FIGURE 7.23
Transformation functions used to obtain the pseudocolor images in Fig. 7.22.

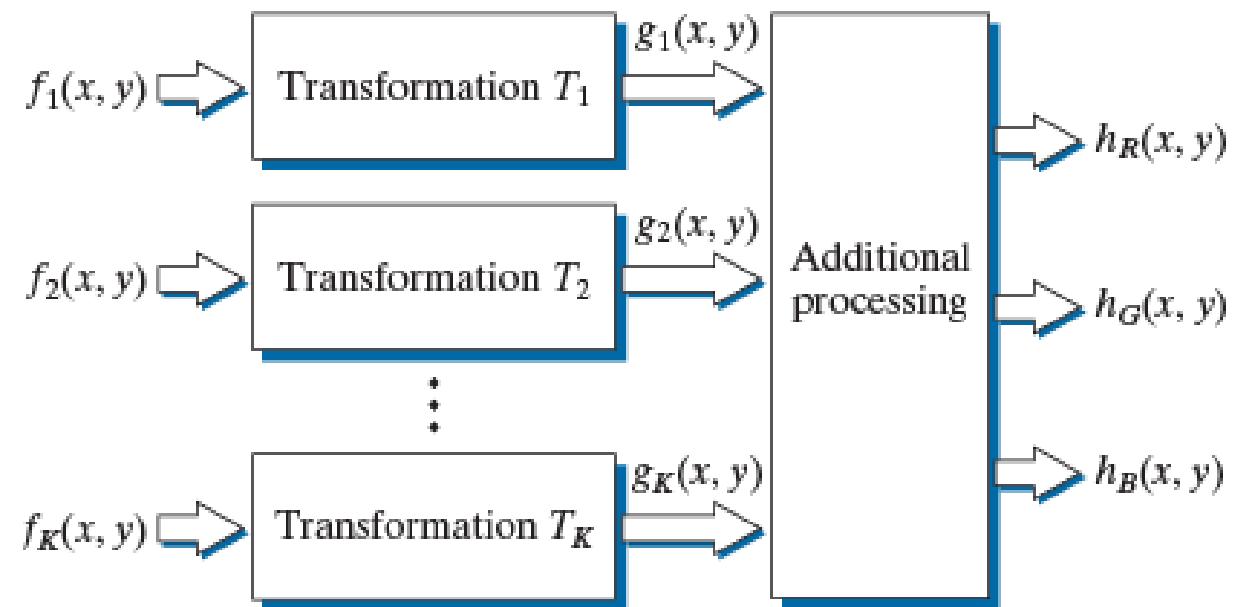
- Explosives and bag content are mapped by similar transformations and were assigned to the same color.
- The observer may “see through” the explosives and not mistake them for the background.

Pseudocolor Coding

- Used if there are **multiple monochrome images** such as multispectral satellite images
 - Sensors collecting at different spectral bands

FIGURE 7.24

A pseudocolor coding approach using multiple grayscale images. The inputs are grayscale images. The outputs are the three components of an RGB composite image.



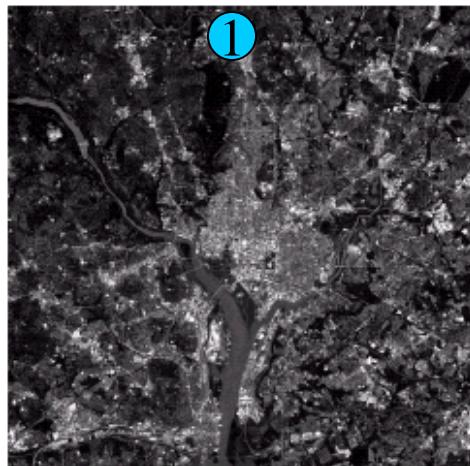
Pseudocolor Coding Example

Visible blue

$\lambda = 0.45\text{-}0.52 \mu\text{m}$

Max water penetration

1

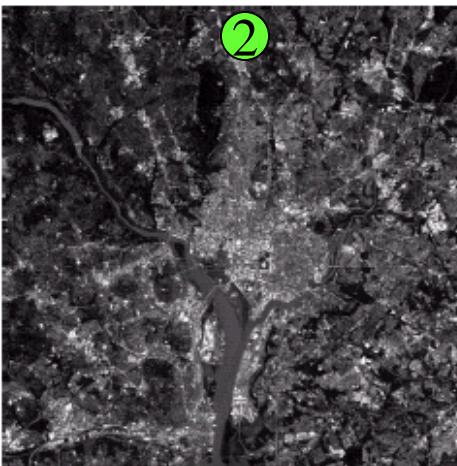


Visible green

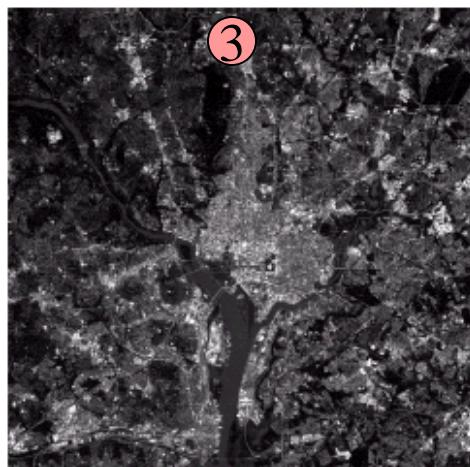
$\lambda = 0.52\text{-}0.60 \mu\text{m}$

Measuring plant

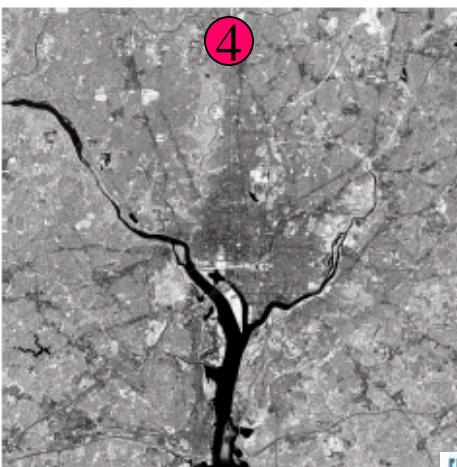
2



3



4



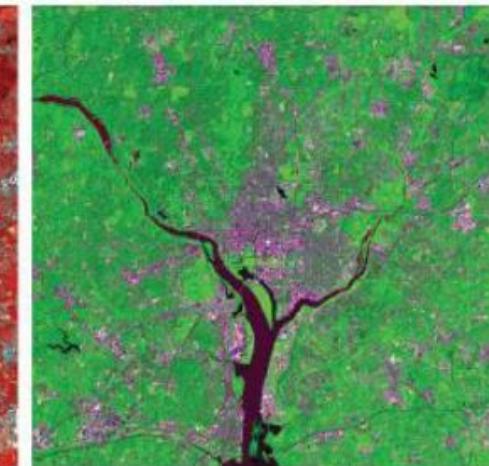
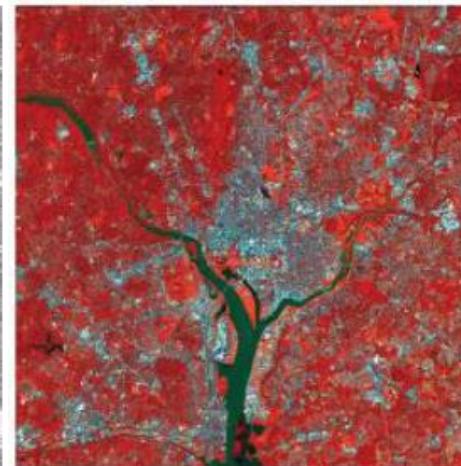
Visible red

$\lambda = 0.63\text{-}0.69 \mu\text{m}$

Plant discrimination Biomass and shoreline mapping

Washington D.C. area

Color composite images



Red = 4

Green = 2

Blue = 3

Red = 3

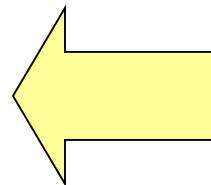
Green = 4

Blue = 1

Shows difference
between Biomass
and man-made
objects

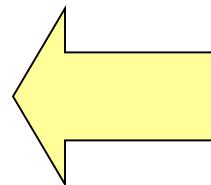
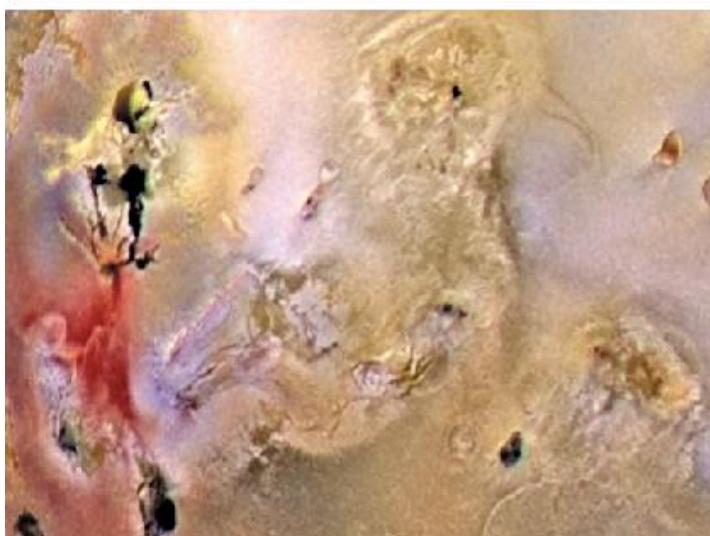
FIGURE 7.25 (a)-(d) Red (R), green (G), blue (B), and near-infrared (IR) components of a LANDSAT multispectral image of the Washington, D.C. area. (e) RGB color composite image obtained using the IR, G, and B component images. (f) RGB color composite image obtained using the R, IR, and B component images. (Original multispectral images courtesy of NASA.)

Pseudocolor Coding Example



Pseudocolor rendition
of Jupiter moon Io

Yellow areas = Older sulfur deposits
Red areas = Material ejected
from active volcanoes



A close-up

a

b

FIGURE 7.26
(a) Pseudocolor
rendition of
Jupiter Moon Io.
(b) A close-up.
(Courtesy of
NASA.)

7.4 Basics of Full-Color Image Processing

Full color image processing: Manipulate real color images such as color photographs

Two Processing Methods:

- 1. Per-color-component processing:** Process each component (R or G or B) image separately and then form composite
- 2. Vector processing:** Treat each pixel as a vector (with RGB bands) to be processed

- Two processing methods:
 - (1) Process each channel (or color component) separately, as if the color image were three gray scale images
 - (2) Work with color pixels directly represented as a vector and process all 3 channels for each pixel
- In order for per-color-component and vector-based processing to be equivalent, two conditions have to be satisfied
 - The process has to be applicable to both vectors and scalars.
 - The operation on each component of a vector must be independent of the other components.

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



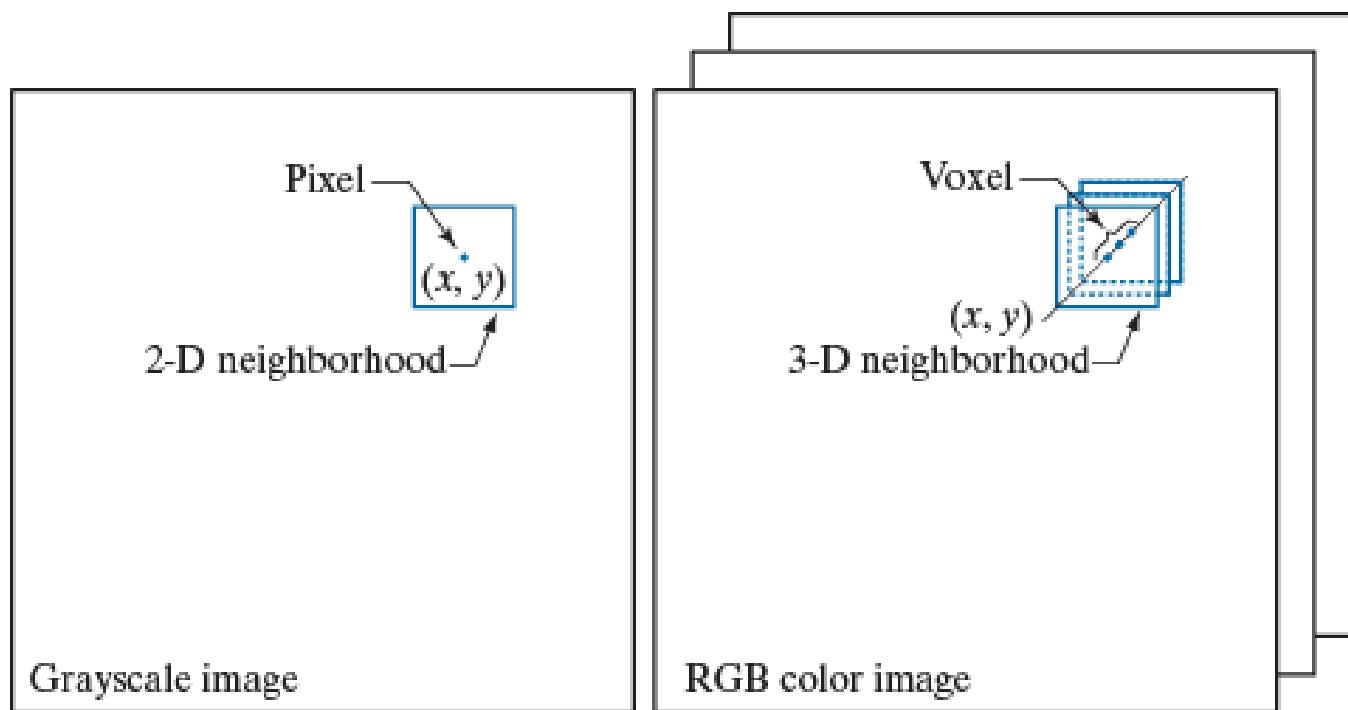
Basics of Full-Color Image Processing

- Example of per-color-component processing: Smoothing an image by smoothing each RGB component separately.
- Both approaches used in later sections

a b

FIGURE 7.27

Spatial neighborhoods for grayscale and RGB color images. Observe in (b) that a *single* pair of spatial coordinates, (x, y) , addresses the same spatial location in all three images.



Let c represent an arbitrary vector in RGB color space:

$$c = \begin{bmatrix} c_R \\ c_G \\ c_B \end{bmatrix} = \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

At coordinates (x, y) ,

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



7.5 Color Transformations

- Process color images within a single color model
- No RGB-> HSI or HSI-> RGB conversion
- Gray scale image transformations may also be applied to each color separately.

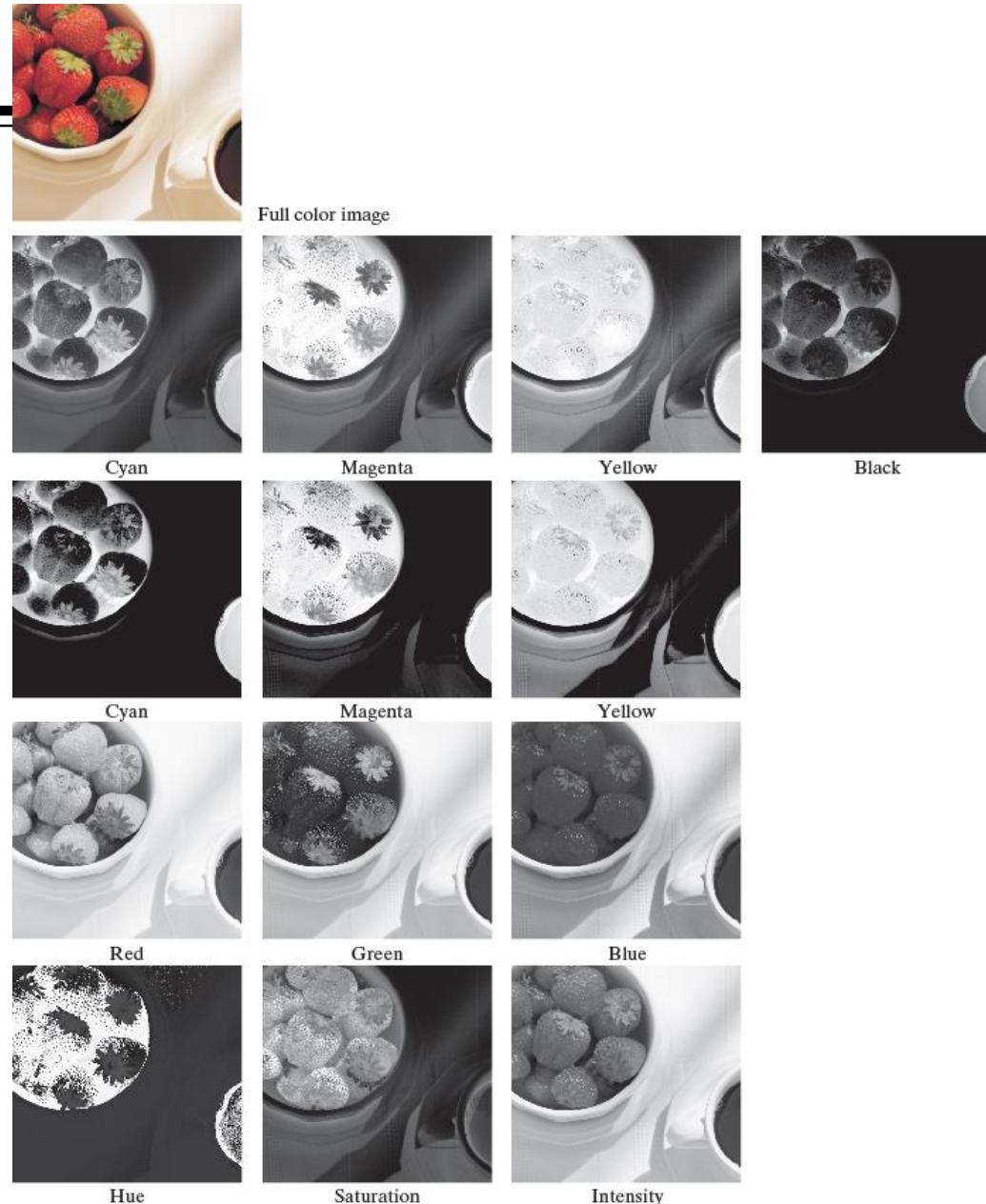


FIGURE 7.28 A full-color image and its various color-space components. (Original image courtesy of MedData Interactive)



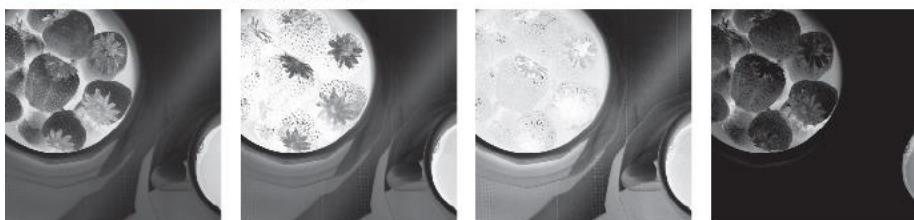
Example: Full-Color Image and Interpretation of Color Space Components



FIGURE 7.28 A full-color image and its various color-space components. (Original image courtesy of MedData Interactive.)

Color image

Full color image



Cyan

Magenta

Yellow

Black

- Black-> 0 & white-> 1

CMYK components

- More magenta & yellow mixing up to form red

CMY components

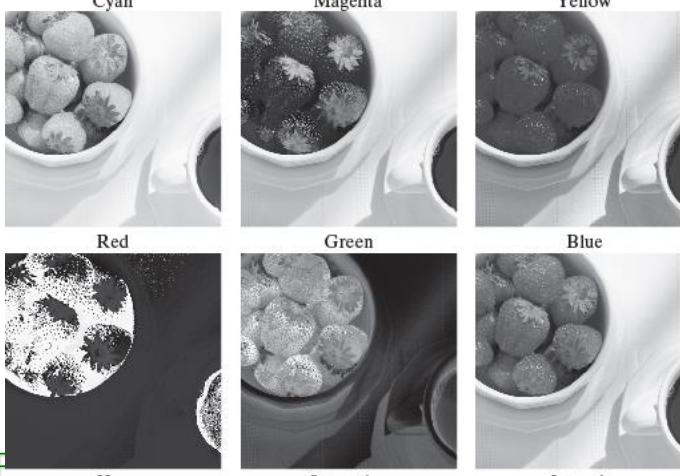
- More magenta & yellow mixing up to form red

RGB components

- More red, very little blue or green

HSI components

- Intensity: Monochrome of color image
- Strawberries are pure → i.e., saturated
- Hue is most difficult to interpret
(discontinuity where 0 and 360 meet)



7.5 Color Transformations

Used to transform colors to colors.

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

$f(x, y)$ = Input color image

$g(x, y)$ = Output color image

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

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

$$s_i = T_i(r_i) \quad i = 1, 2, \dots, n \quad \text{For RGB images, } n = 3$$

where, r_i = Color component of $f(x, y)$

s_i = Color component of $g(x, y)$



Example: Color Transformation - Intensity Manipulation

Formula for RGB:

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

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

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

Formula for HSI:

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

Formula for CMY:

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

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

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

Formula for CMYK:

$$s_C(x, y) = r_C(x, y)$$

$$s_M(x, y) = r_M(x, y)$$

$$s_Y(x, y) = r_Y(x, y)$$

$$s_K(x, y) = kr_K(x, y) + (1 - k)$$

3rd Edition

Intensity Reduction by 30% ($k=0.7$)

- The output is the same regardless of the color space

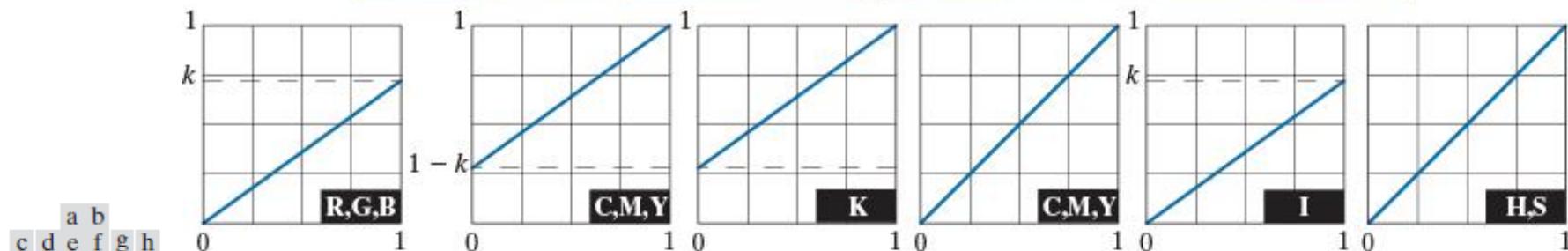


FIGURE 7.29 Adjusting the intensity of an image using color transformations. (a) Original image. (b) Result of decreasing its intensity by 30% (i.e., letting $k = 0.7$). (c) The required RGB mapping function. (d)–(e) The required CMYK mapping functions. (f) The required CMY mapping function. (g)–(h) The required HSI mapping functions. (Original image courtesy of MedData Interactive.)

Color Complements

- For enhancement of dark regions.
- Opposite hues in the color circle (e.g., reds replaced by cyan)
- Straightforward in RGB space
- No equivalent transformation in HSI space.
 - The saturation (S) component cannot be computed from the S component of the original image (because of the equations for S in slide-47)

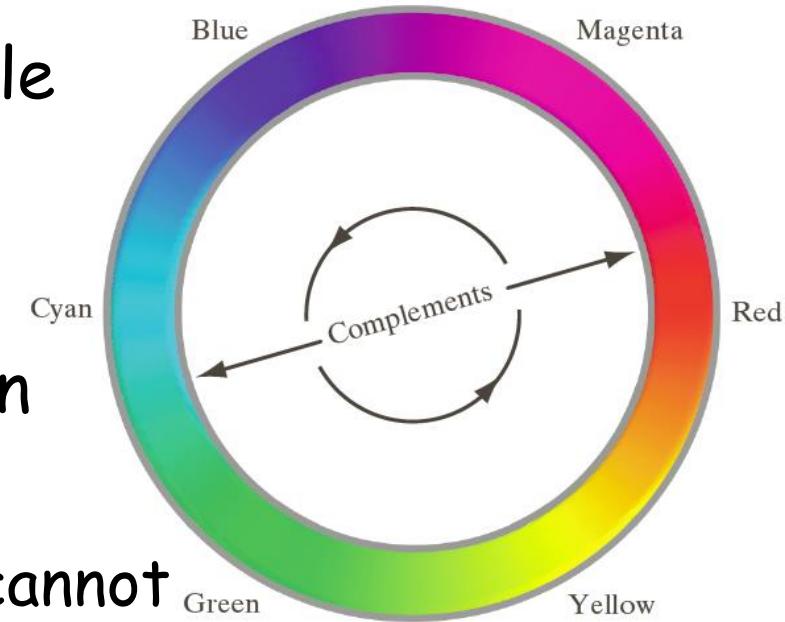


FIGURE 7.30
Color
complements on
the color circle.

Color Complements

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

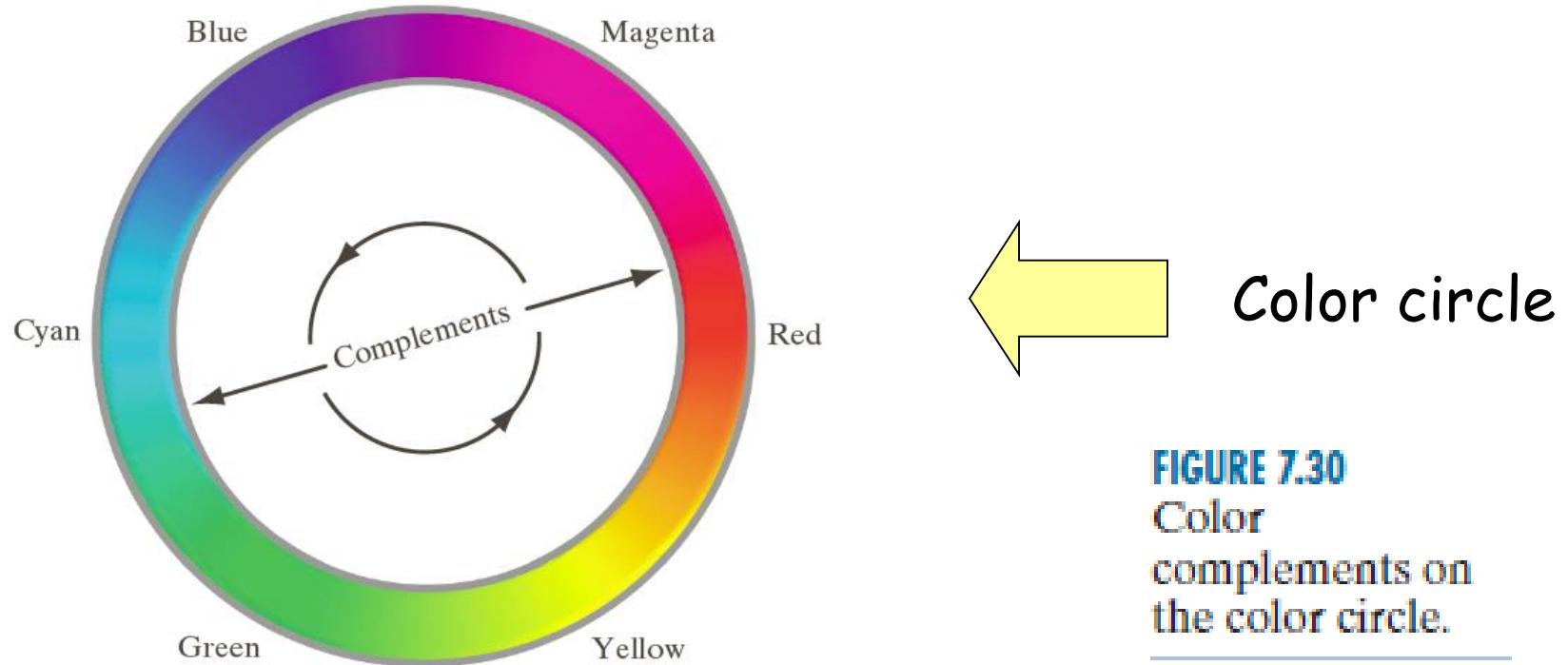
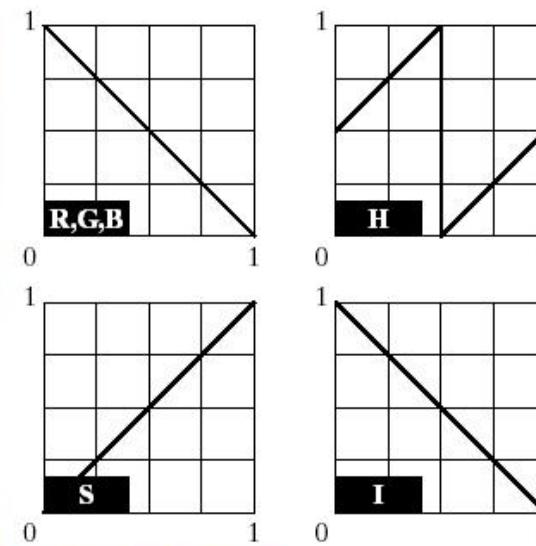


FIGURE 7.30
Color
complements on
the color circle.

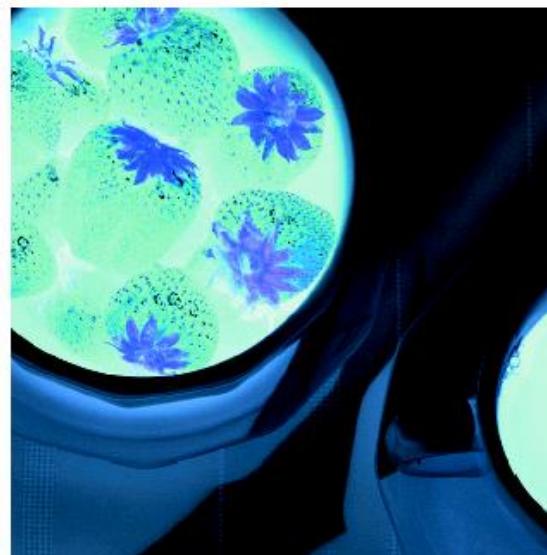
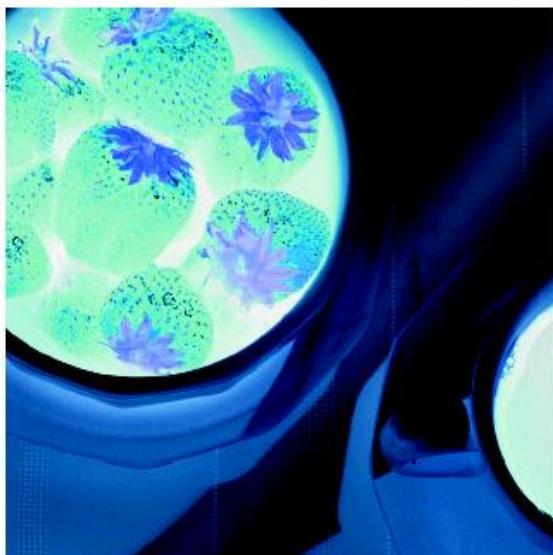


Color Complement Transformation Example



a
b
c
d

FIGURE 7.31
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.



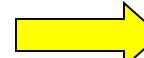
Color Slicing Transformation

- Useful for **separating objects from surroundings**
- If the color of a pixel **is far from a desired color** than threshold distance, then set that color to some specific color such as gray. Otherwise keep the original color unchanged. $(a_1, a_2, \dots, a_n) \rightarrow$ Average color components

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

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

Set to gray 

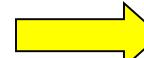
Keep the original color 

Or, use a sphere

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

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

Set to gray 

Keep the original color 

Color Slicing - Edible part



Original image

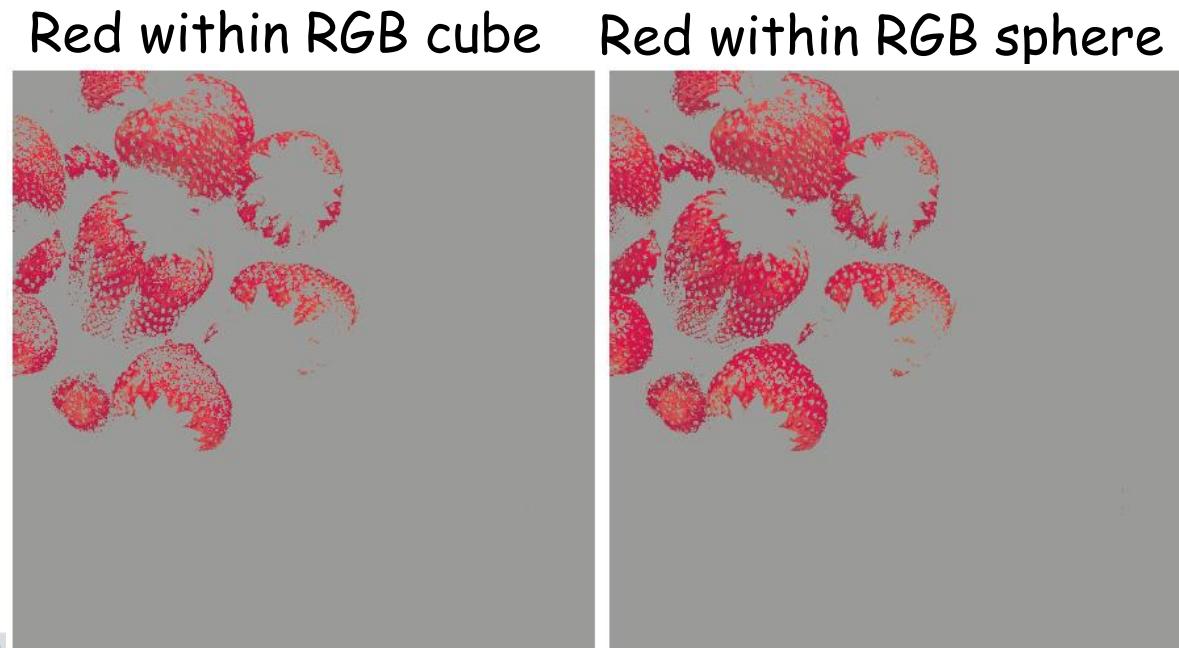


FIGURE 7.32 Color-slicing transformations that detect (a) reds within an RGB cube of width $W = 0.2549$ centered at $(0.6863, 0.1608, 0.1922)$, and (b) reds within an RGB sphere of radius 0.1765 centered at the same point. Pixels outside the cube and sphere were replaced by color $(0.5, 0.5, 0.5)$.

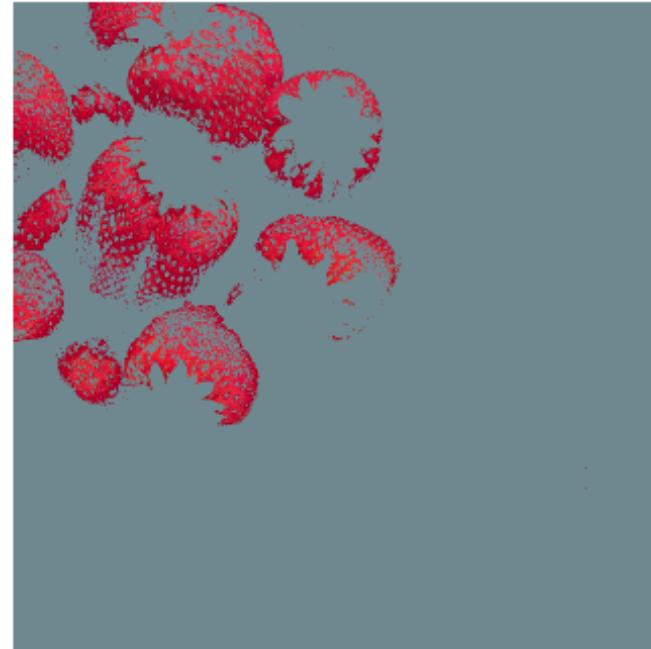
- **Color slicing** for separating objects from their surroundings.
- A first step towards image segmentation.

Color Slicing Transformation Example

After color slicing



Original image



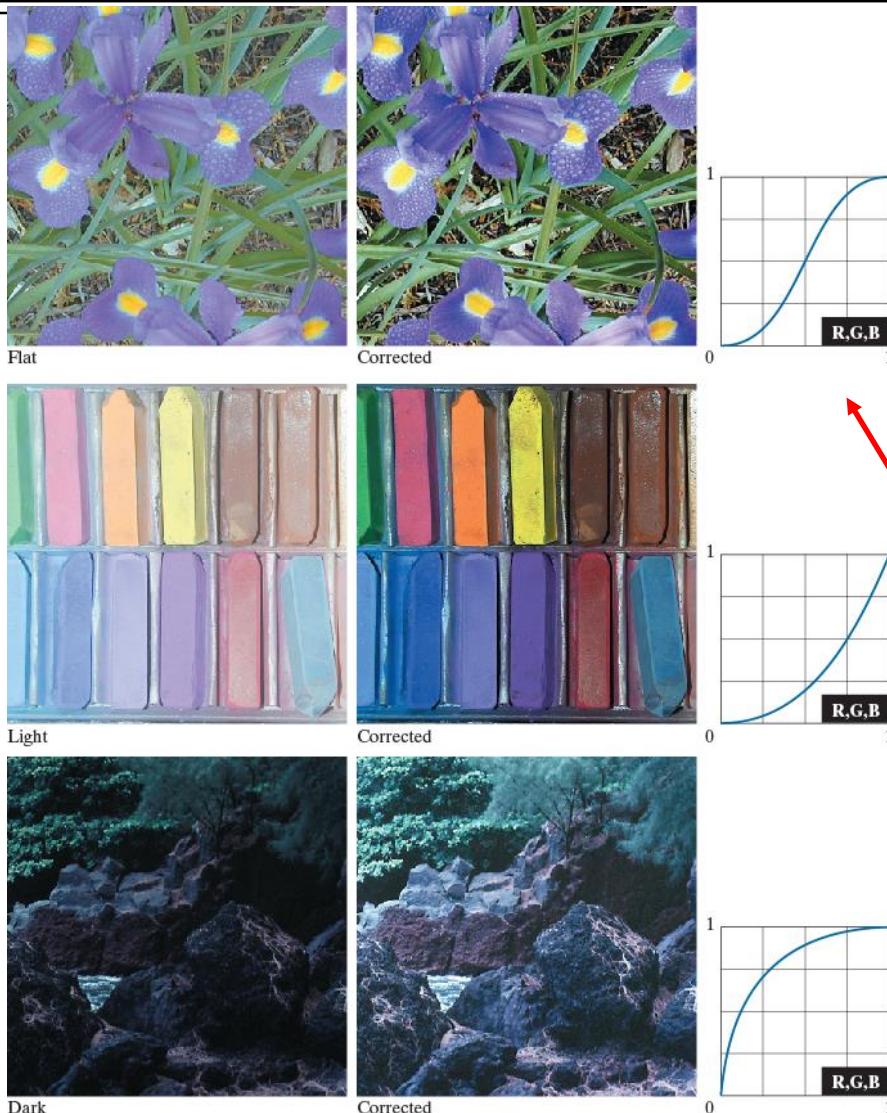
a



b

FIGURE 7.32 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 $(0.5, 0.5, 0.5)$.

Tonal Correction Examples



- Only brightness and contrast are adjusted while keeping color unchanged.
 - This can be done by using the same transformation for all RGB or CMYK components.
 - For HSI: Only Intensity is changed
- Contrast enhancement
- Power law transformations

FIGURE 7.33 Tonal corrections for flat, light (high key), and dark (low key) color images. Adjusting the red, green, and blue components equally does not always alter the image hues significantly.

- The approaches used so far are interactive.
- Histogram equalization of a color image can be performed automatically by **adjusting color intensity uniformly while leaving the colors (i.e., hues) unchanged**
- The **HSI model is suitable** for histogram equalization where **only Intensity (I) component is equalized**

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

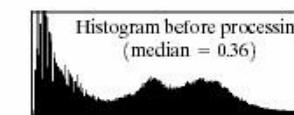
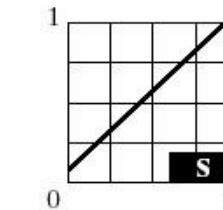
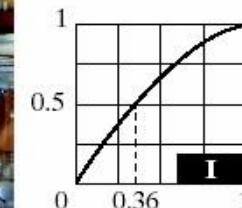
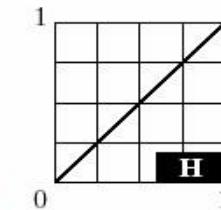
where, r and s are intensity components of input and output color images



Histogram Equalization



Histogram Equalization
of Intensity without
altering hue/saturation



a
b
c
d

FIGURE 7.35
Histogram
equalization
(followed by
saturation
adjustment) in the
HSI color space.

Increased
Saturation
after
Histogram
Equalization
→ More
vibrant oil
and vinegar
colors



Two Methods:

1. **Per-color-plane method:** For RGB and CMY color models
Smooth each color plane using moving average and then combine back to RGB

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

2. **Smooth only Intensity (I) component (faster)** of an HSI image while leaving H and S unmodified

Note: These two methods are not equivalent

Color Image Smoothing Example (cont.)

Color image



Green



Red



Blue



a b
c d

FIGURE 7.36
(a) RGB image.
(b) Red
component image.
(c) Green
component.
(d) Blue
component.

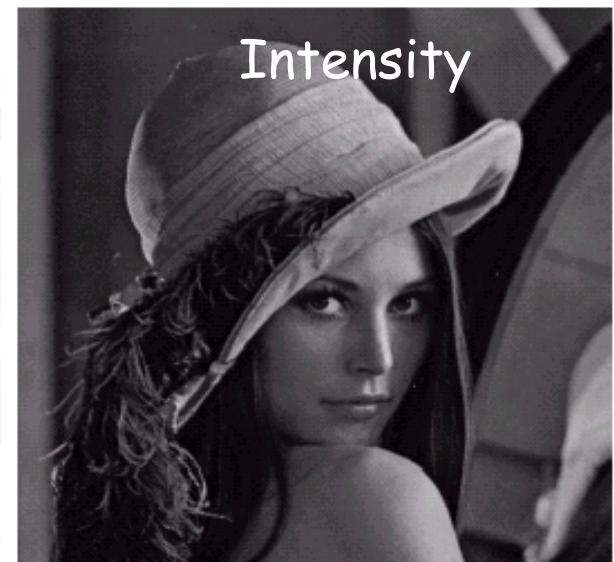
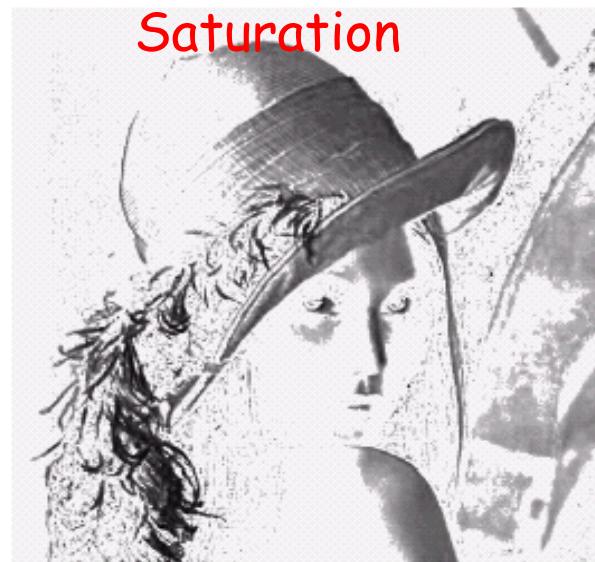


Color image

a b c

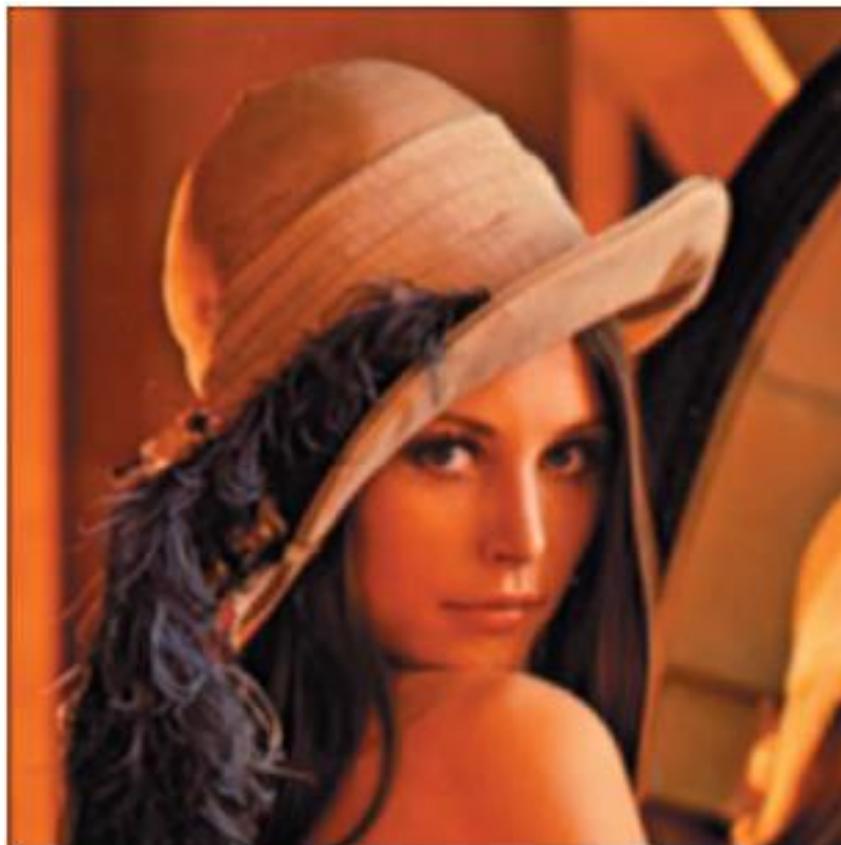
FIGURE 7.37 HSI components of the RGB color image in Fig. 7.36(a). (a) Hue. (b) Saturation. (c) Intensity.

HSI Components

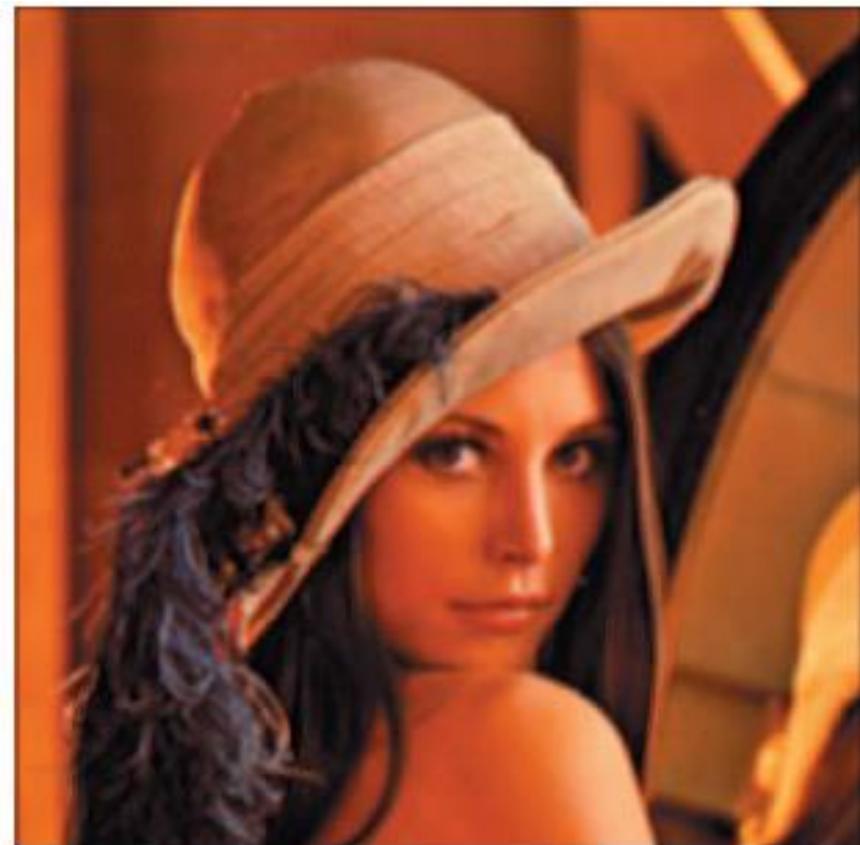




Color Image Smoothing Example (cont.)



Smooth all RGB components



Smooth only I component of HSI
(faster)

a b c

FIGURE 7.38 Image smoothing with a 5×5 averaging kernel. (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.



Color Image Smoothing Example (cont.)



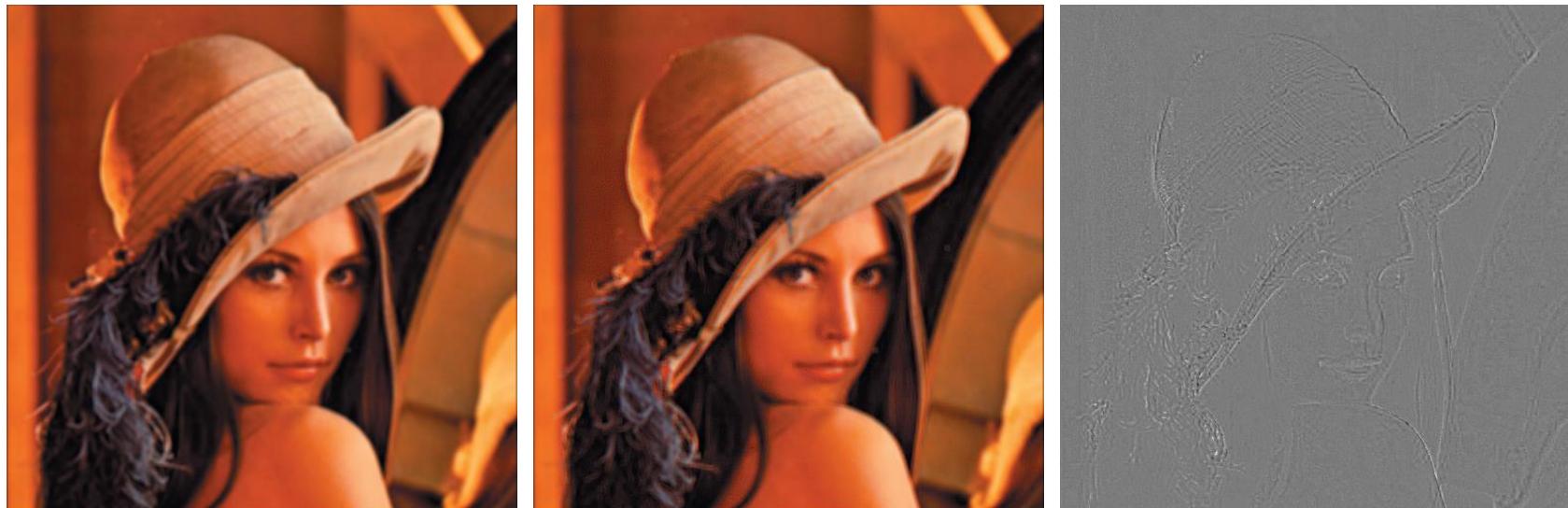
a b c

FIGURE 7.38 Image smoothing with a 5×5 averaging kernel. (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.

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

Color Image Smoothing (cont...)

- HSI decouples intensity from color. Suitable for processing only the intensity component of an image.



a b c

FIGURE 7.38 Image smoothing with a 5×5 averaging kernel. (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.



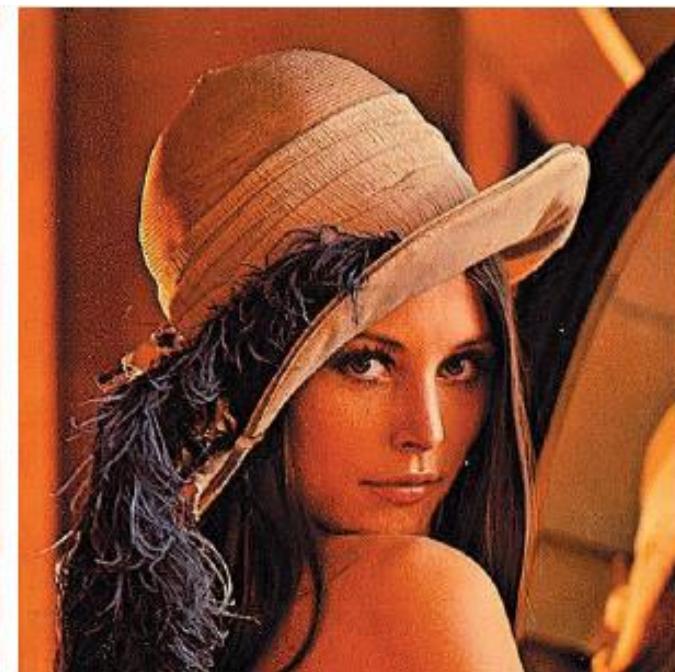
Color Image Sharpening

Can be done in the same manner as color image smoothing:

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



Sharpening all RGB components



Sharpening only I component of HSI

a b c

FIGURE 7.39 Image sharpening using the Laplacian. (a) Result of processing each RGB channel. (b) Result of processing the HSI intensity component and converting to RGB. (c) Difference between the two results.



Color Image Sharpening Example (cont.)



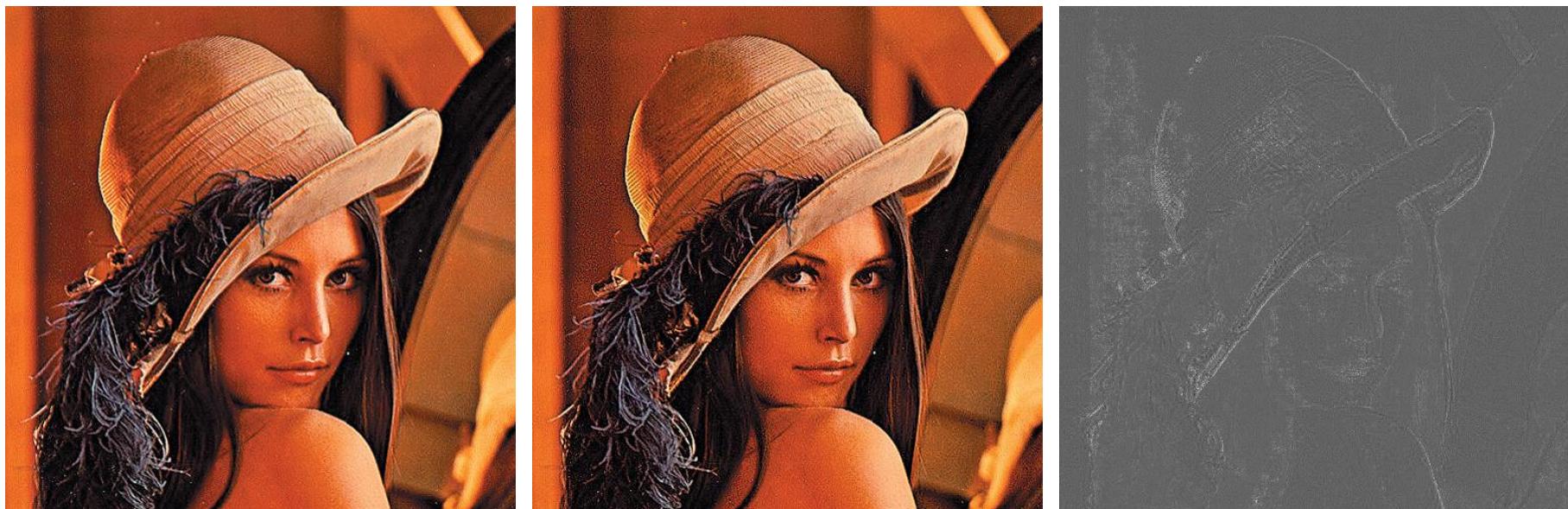
a b c

FIGURE 7.39 Image sharpening using the Laplacian. (a) Result of processing each RGB channel. (b) Result of processing the HSI intensity component and converting to RGB. (c) Difference between the two results.

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

Color Image Sharpening

- The difference becomes more pronounced by increasing the filter size.



a b c

FIGURE 7.39 Image sharpening using the Laplacian. (a) Result of processing each RGB channel. (b) Result of processing the HSI intensity component and converting to RGB. (c) Difference between the two results.

7.8 Noise in Color Images

- The noise models discussed for grayscale images are also applicable to color images.
- However, in many applications, a color channel may be more (or less) affected than the other channels.
- For instance, using a red color filter in a CCD camera may affect the red component of the image (CCD sensors are noisier at low levels of illumination).
- We will take a brief look of how noise carries over when converting from one color model to another.



Noise in Color Images

Noise can corrupt each color component independently.

a
b
c
d

FIGURE 7.46
(a)–(c) Red, green, and blue 8-bit component images corrupted by additive Gaussian noise of mean 0 and standard deviation of 28 intensity levels.
(d) Resulting RGB image.
[Compare (d) with Fig. 7.44(a).]



FIGURE 7.44
(a) RGB image.

Noise is less noticeable in a color image

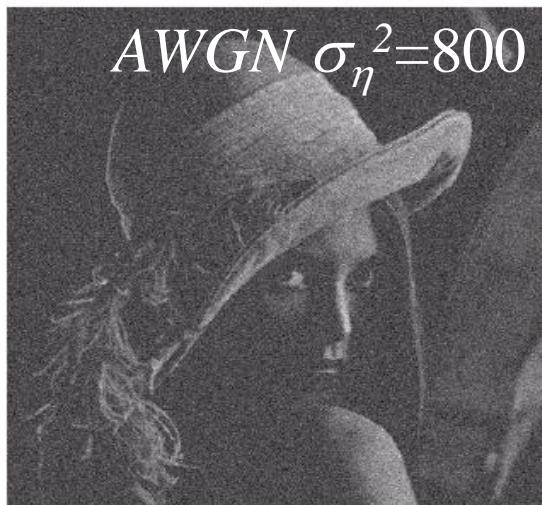
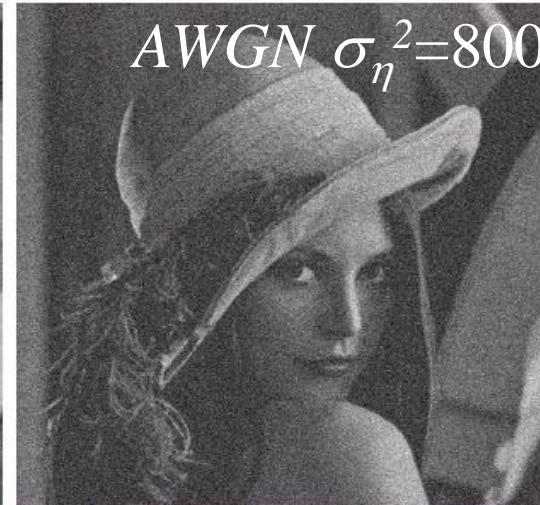
Noise in Color Images

Noise can corrupt each color component independently.

a	b
c	d

FIGURE 6.48

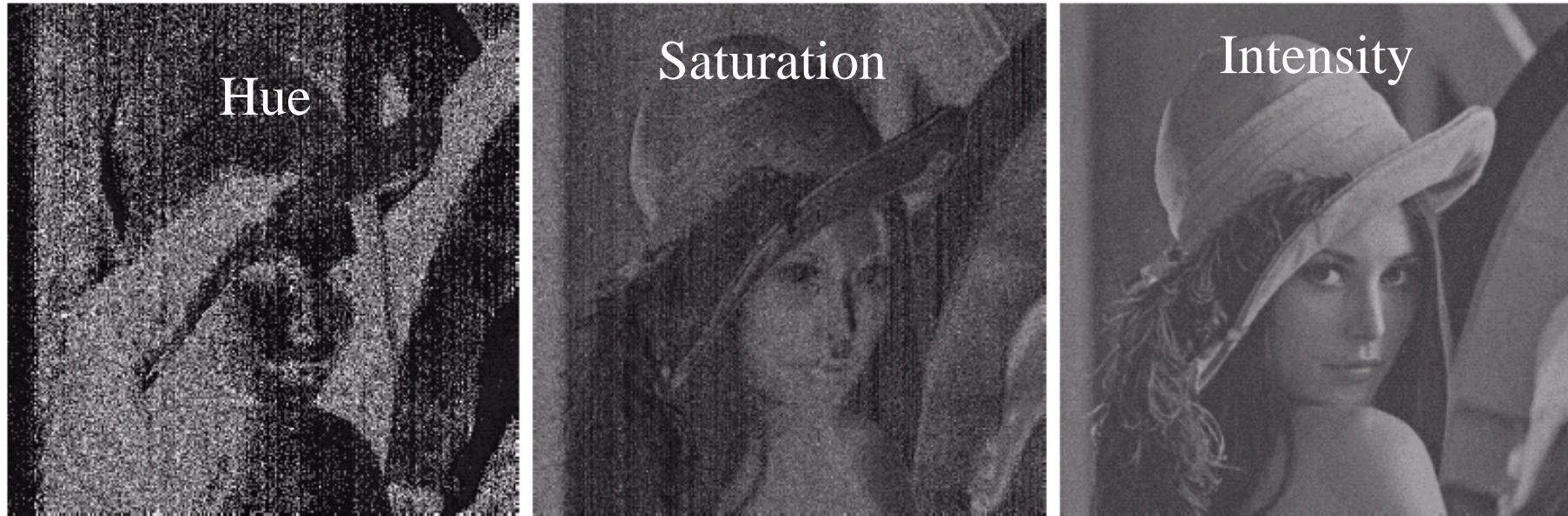
(a)–(c) Red, green, and blue component images corrupted by additive Gaussian noise of mean 0 and variance 800. (d) Resulting RGB image. [Compare (d) with Fig. 6.46(a).]



Noise is less
noticeable
in a color
image

3rd Edition

Noise in Color Images

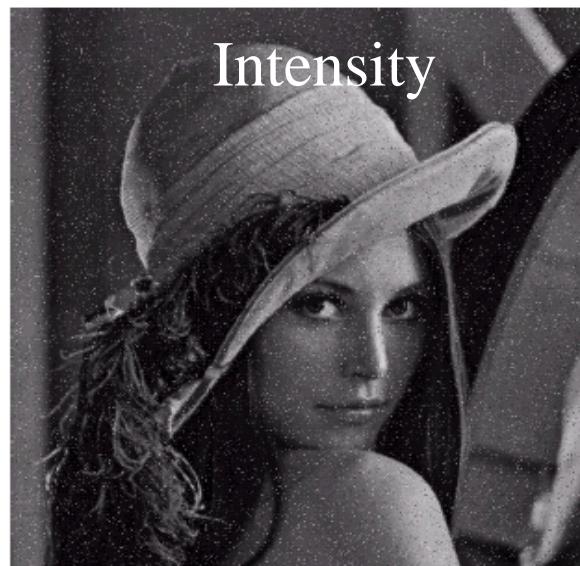
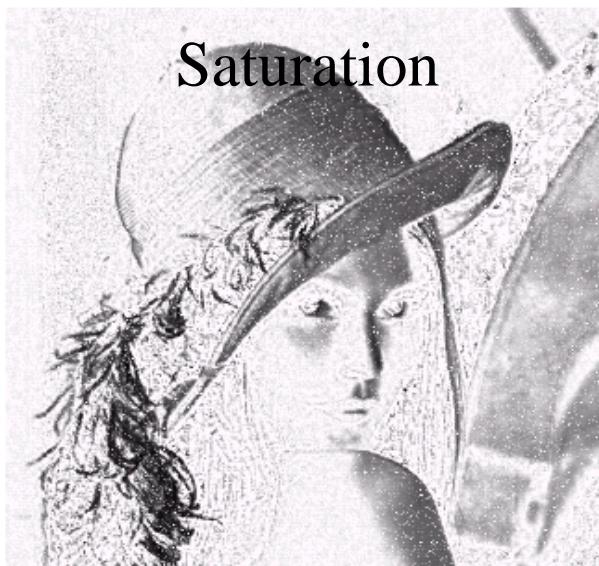
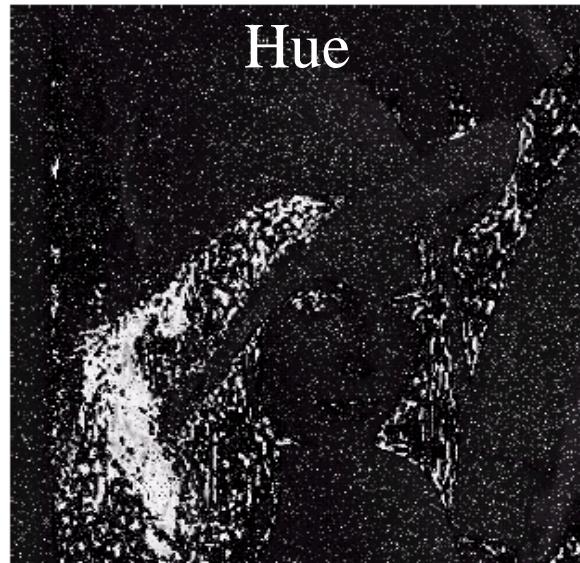
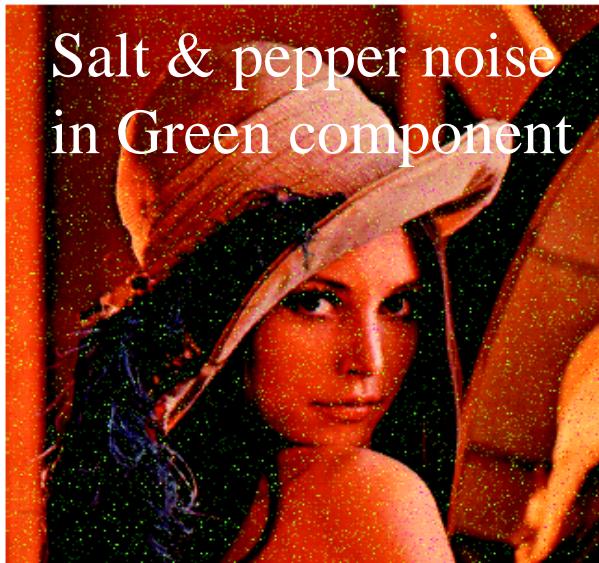


a b c

FIGURE 7.47 HSI components of the noisy color image in Fig. 7.46(d). (a) Hue. (b) Saturation. (c) Intensity.

- The hue and saturation components are significantly degraded. This is due to the nonlinearity of the \cos and \min operations used in the transformation.
- The intensity component is smoother due to averaging of the three noisy RGB components.

Noise in Color Images



a
b
c
d

FIGURE 7.48

- (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 in Color Images (cont...)



- When only one channel is affected by noise, conversion to HSI spreads the noise to all HSI components images
- This is due to the transformation that makes use of all RGB components to compute each HSI components

a
b
c
d

FIGURE 7.48

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

Greyscale from RGB

- Sometimes a single value at each pixel makes processing easier
- This value is usually the intensity or 'grey value'
- We can convert an RGB image to greyscale using
 - Simple average of red, green and blue, i.e.

$$i = (r+g+b)/3 \text{ or}$$

- Weighted average, i.e.

$$i = a_1r + a_2g + a_3b \text{ where } a_1 + a_2 + a_3 = 1$$

where, i is the grey value and r , g , and b are the red green and blue values



Average

Original

Weighted

- Since our eyes are more sensitive to green light one commonly used formula:

$$i = 0.30r + 0.59g + 0.11b$$

Acknowledgements

The slides are primarily based on the figures and images in the Digital Image Processing textbook by Gonzalez and Woods:

- http://www.imageprocessingplace.com/DIP-3E/dip3e_book_images_downloads.htm

In addition, slides have been adopted and modified from the following sources:

- http://www.cs.uoi.gr/~cnikou/Courses/Digital_Image_Processing
- <http://www.comp.dit.ie/bmacnamee/gaip.htm>
- <http://baggins.nottingham.edu.my/~hssooihock/G52IIP/>
- <http://gear.kku.ac.th/~nawapak/178353.html>
- <https://cs.nmt.edu/~ip/index.html>