

Color Fundamentals

Dr. Eman Gouda



Lecture Material

- Digital Image Processing Fourth Edition Rafael C. Gonzalez and Richard E. Woods
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Digital Image Processing & Human Vision

- Digital Image Processing is built on mathematics but often guided by human intuition and visual judgment.
- Understanding human visual perception is a key first step:
 - How images are formed and perceived by humans.
 - The limitations of human vision (resolution, adaptation to illumination).
 - Comparing human vision vs. electronic imaging devices:
- Helps design techniques aligned with what the human eye can perceive.

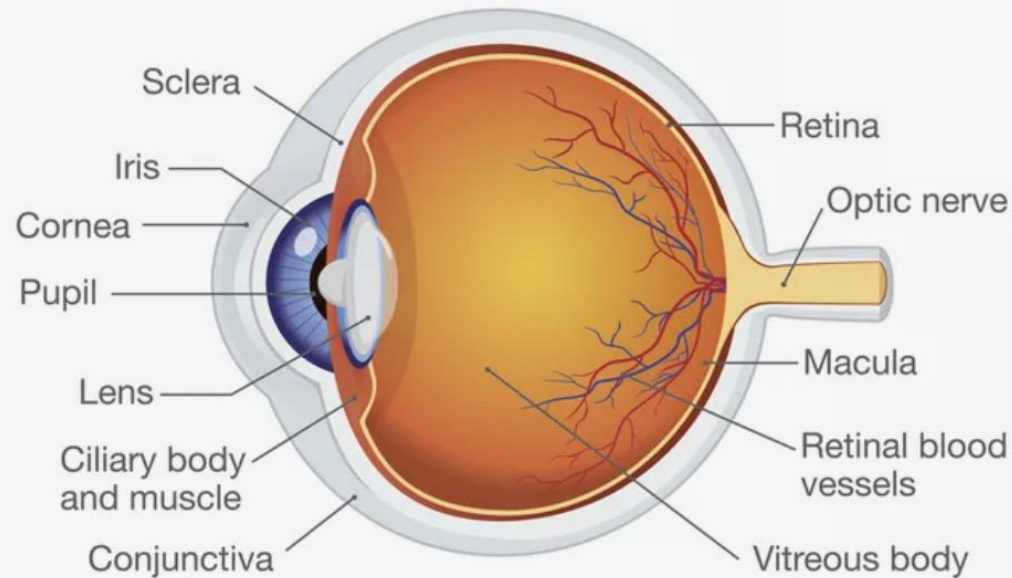
The Human Eye and Vision

- The **eye** provides us with the **sense of sight**.
- It allows us to see and interpret:
 - **Shapes**
 - **Colors**
 - **Dimensions (size & depth)**
- Works by processing the **light reflected or emitted** from objects.
- The eye can detect **bright light** and **dim light**, but it **cannot see in the absence of light**.



The Human Eye and Vision

Human Eye Anatomy



- **Light path:** Cornea → Iris → Lens → Retina
- **Pupil:** Adjustable opening in the iris; regulates light (like a camera aperture).
- **Iris:** Colored part; controls pupil size and amount of incoming light.
- **Lens:** Focuses light rays onto the retina.
- **Retina:** Lining at the back of the eye; contains photoreceptors that send signals to the brain.

Photoreceptors in the Retina

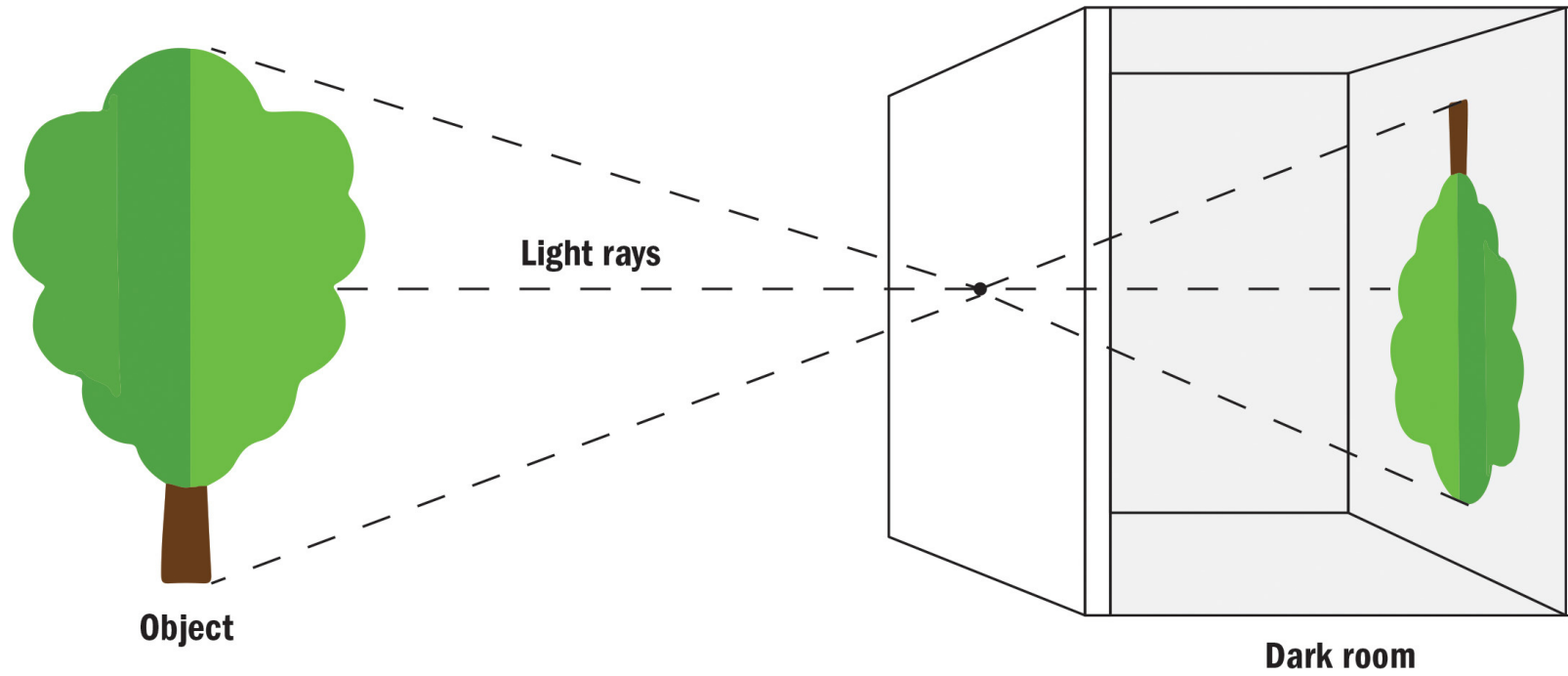
- **Rods**

- Very sensitive to low light (night vision).
- Detect motion and intensity, but not color.
- Example: At night, colors fade and objects look gray.

- **Cones**

- Active in bright light (day vision).
- Responsible for sharp details and color vision.
- 3 types:
 - L-cones → Red
 - M-cones → Green
 - S-cones → Blue

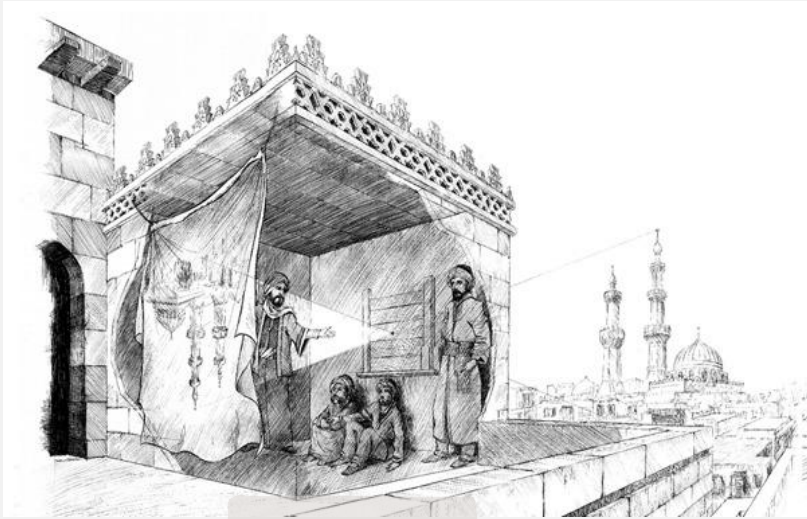
Camera obscuras



Camera obscura (pinhole camera)

- Camera Obscuras date back to **400 BC** (and possibly earlier).
- Light passes through a small hole → projects an **inverted image** inside a dark room.
- **Mozi** (China, 400 BC)
 - Light from an illuminated object passing through a pinhole
 - Creates an **upside-down image** inside a dark chamber.
- **Aristotle** (Greece, 4th Century BC)
 - Discovered a way to view **solar eclipses safely**.
 - Saw crescent-shaped sun projections on the ground through **gaps between tree leaves**
 - Realized that light creates images without direct eye contact with the sun.

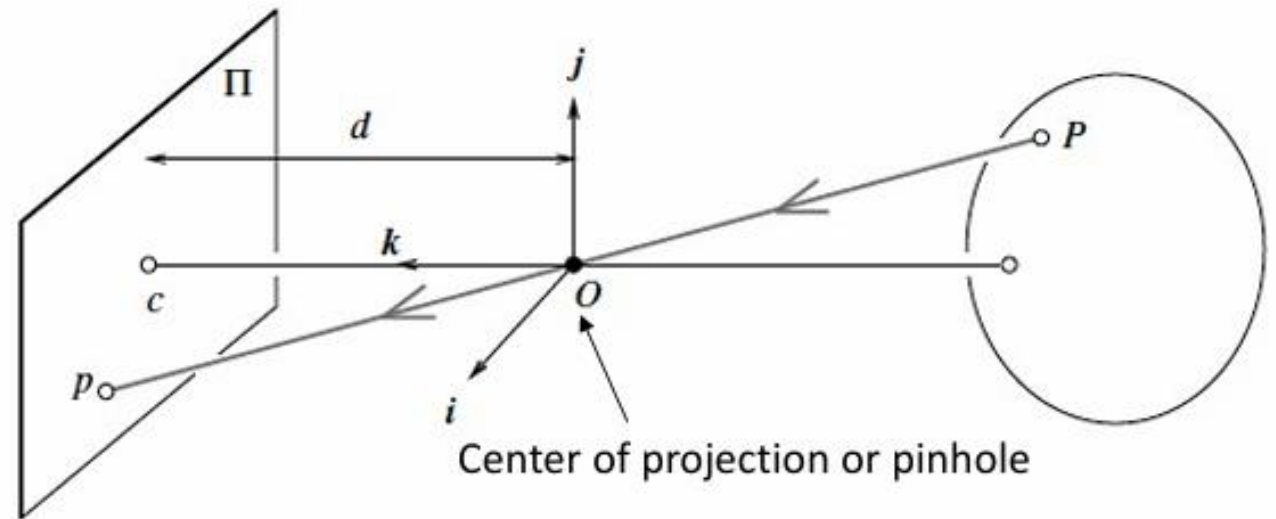
Camera obscura (pinhole camera)



- **Ibn al-Haytham** (11th Century)
 - First to use a **viewing screen** to clearly see the inverted image.
 - The actual **inventor of the Camera Obscura**
 - Conducted **experiments with candles**
- **Leonardo da Vinci** (1502)
 - Suggested the eye works like a camera obscura
- **Giambattista della Porta** (1535–1615)
 - Improved design by adding a concave lens at the pinhole.
 - Produced clearer, sharper images.
- **Johannes Kepler** (1604)
 - First to use the term “Camera Obscura”.

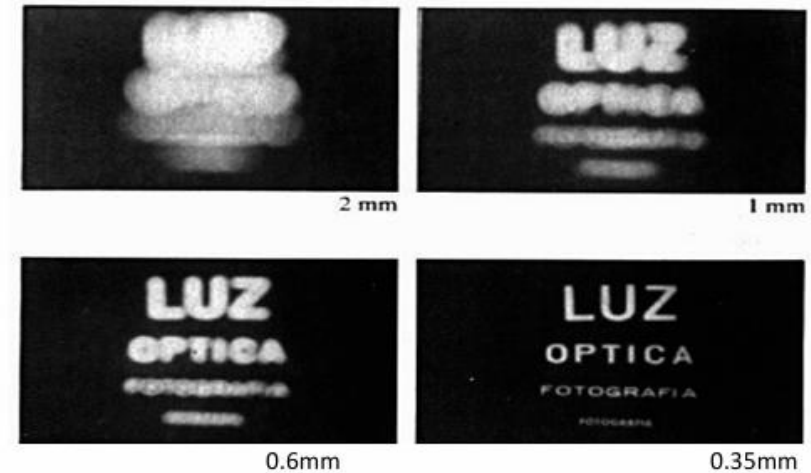
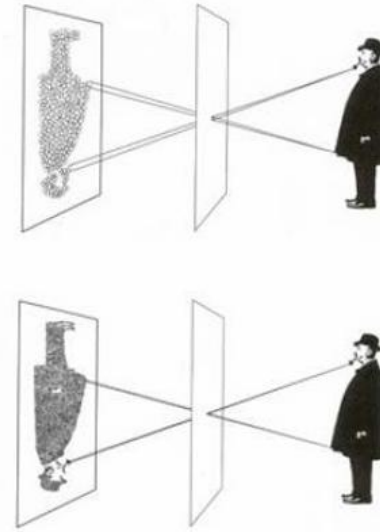
Pinhole Camera Model

- **Light Rays:** Pass through the pinhole and form an inverted image on the image plane.
- **Focal Length (f):** Distance between the pinhole and the image plane.
- **Optical Axis:** A line perpendicular to the image plane (Π) that passes through the pinhole.
- **Image Center (c):** The point where the optical axis intersects the image plane.



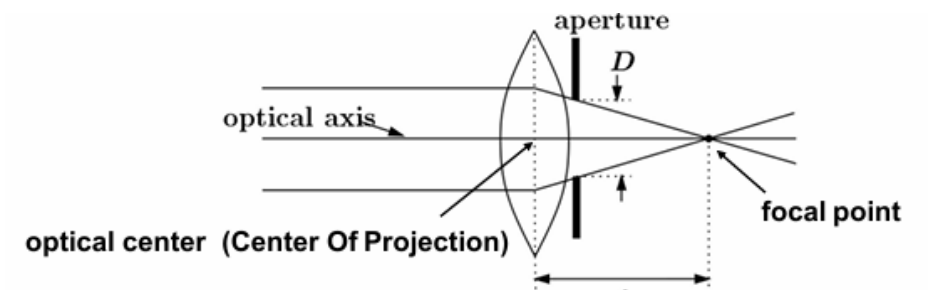
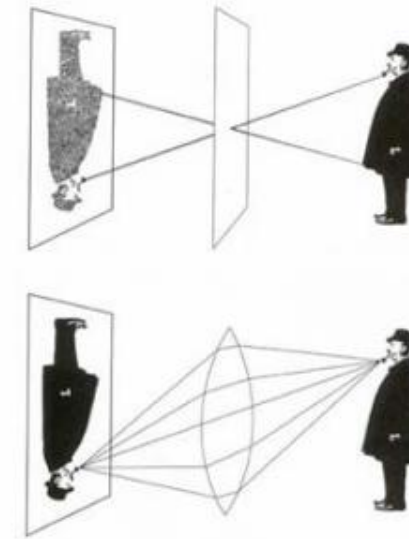
Effect of Pinhole Size

- **Large Aperture:**
 - Allows more light to enter.
 - Light rays spread out → image becomes blurry (not well focused).
- **Small Aperture:**
 - Allows less light to enter.
 - Reduces blurring → image appears sharper, but dimmer.



From Pinhole to Lens

- A **lens** replaces the pinhole.
- Lenses **improve image quality** → sharper and brighter images.
- **Focal Point (F):**
 - Point where parallel rays (to the optical axis) converge.
 - Located at a distance **f** from the lens.



Field of View

- The portion of scene space that actually projects onto the retina of the camera.
- Focal Length (f):
 - Distance from the lens to the image plane/sensor.
 - Small $f \rightarrow$ wide field of view.
 - Large $f \rightarrow$ narrow field of view.

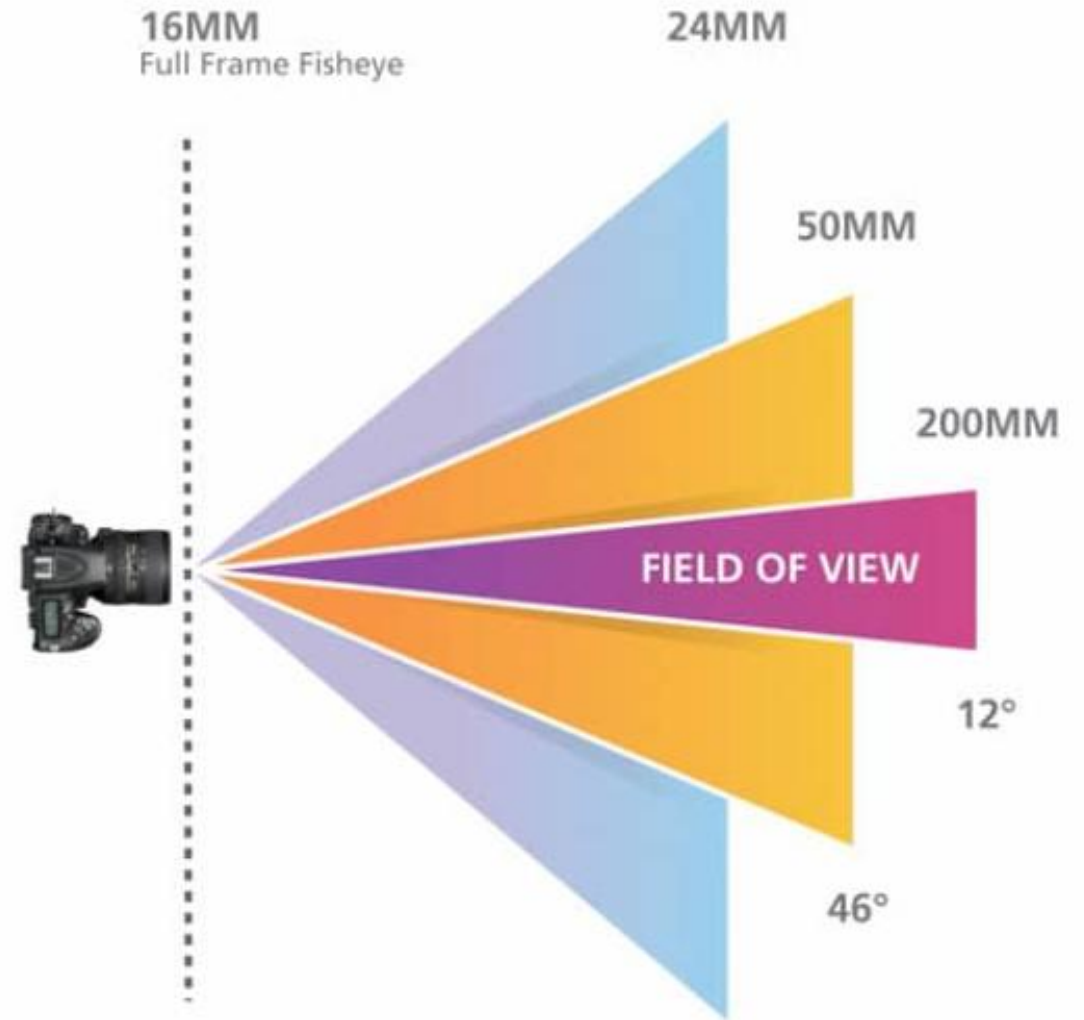
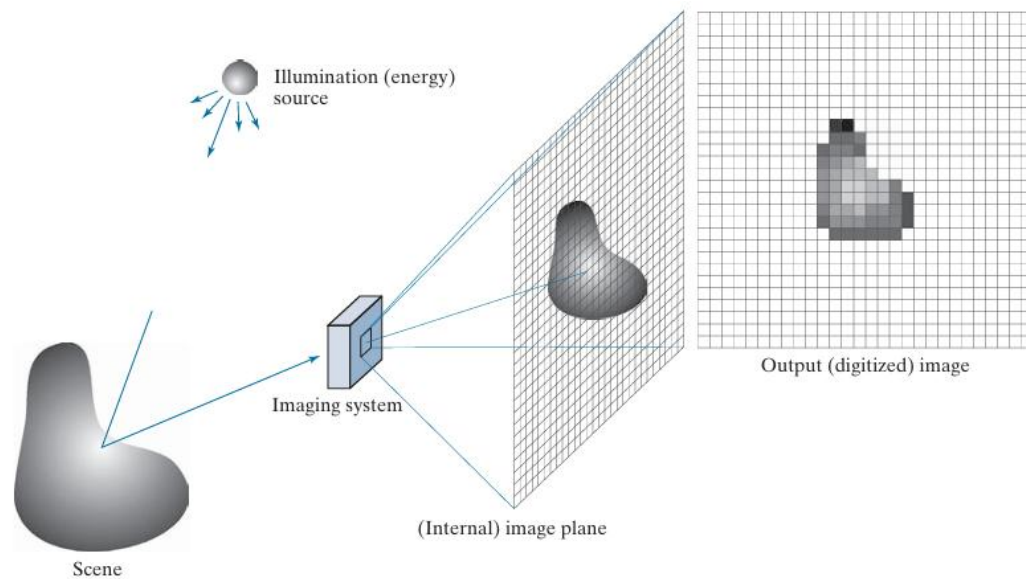


Image Acquisition using Sensor Arrays



- The **sensor** is a **2D grid of pixels**.
- Each **pixel** is a **light-sensitive element**.
- A pixel measures the **intensity of light** falling on it.
- Pixels **convert light into electrical signals**.
- Camera circuitry **translates signals into digital numbers**.
- All pixel values together form the **digital image**.

Simple Image Formation Model

- An image is represented as a function:

$$f(x, y) = i(x, y)r(x, y)$$

- **Illumination $i(x, y)$:**

- Amount of light falling on the scene.
- Depends on the light source (sun, lamp, etc.).

- **Reflectance $r(x, y)$:**

- Proportion of light reflected by the surface.
- Depends on material properties (white paper vs. black cloth).

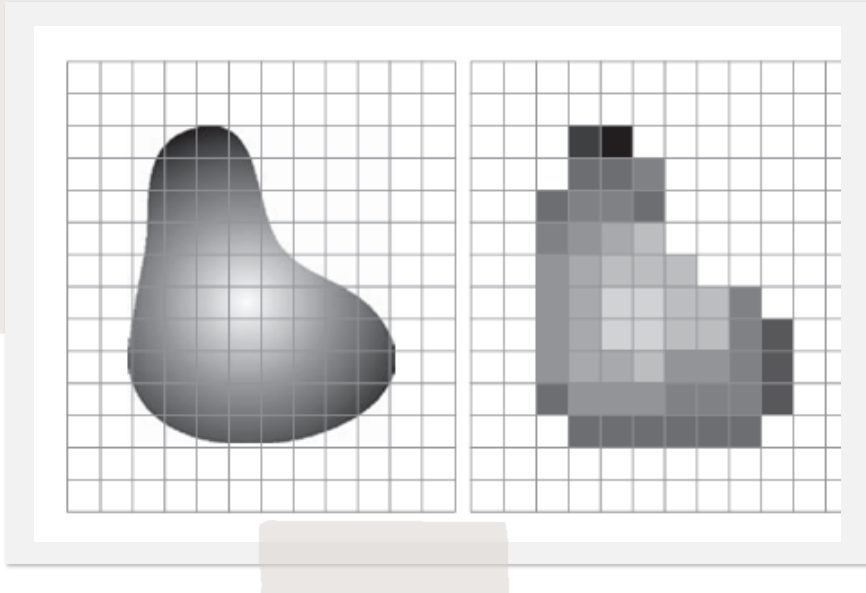
- **Example:**

- Bright light + white surface → high intensity (bright pixel).
- Low light + white surface → medium intensity (gray pixel).
- Bright light + black surface → low intensity (dark pixel).

Image Sampling And Quantization

- Generate digital images from sensed data.
- **Sensor Output:**
 - Continuous voltage waveform.
 - Amplitude & spatial behavior depend on physical phenomenon being sensed.
- **To create a digital image**, Continuous data must be converted to digital format.
- **Two key processes:**
 - **Sampling** → Discretize spatial coordinates (x, y).
 - **Quantization** → Discretize amplitude (intensity values).

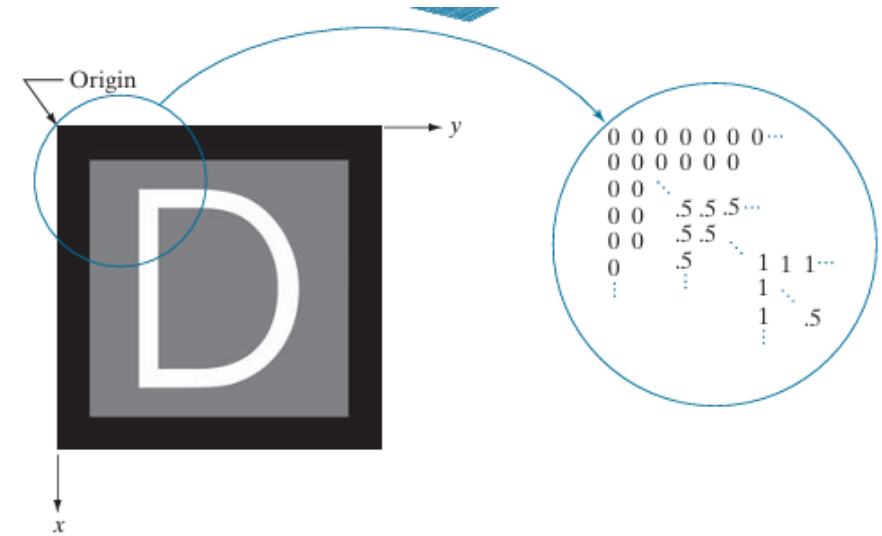
Image Sampling And Quantization



- **Sampling**
 - Number of sensors determines **sampling in both horizontal and vertical directions**.
 - Sampling provides **discrete spatial locations**, but **intensity remains continuous**.
- **Quantization**
 - Converts each sensor's continuous intensity value into a **digital value**.
 - **Intensity scale** is divided into **discrete levels**.
 - Example:
 - **8 levels** from black → white.
 - Each level has a **specific digital value**.
 - Each sample is assigned to the **nearest level**.
- Quality depends on **number of sensors and number of intensity levels**.

Representing Digital Images

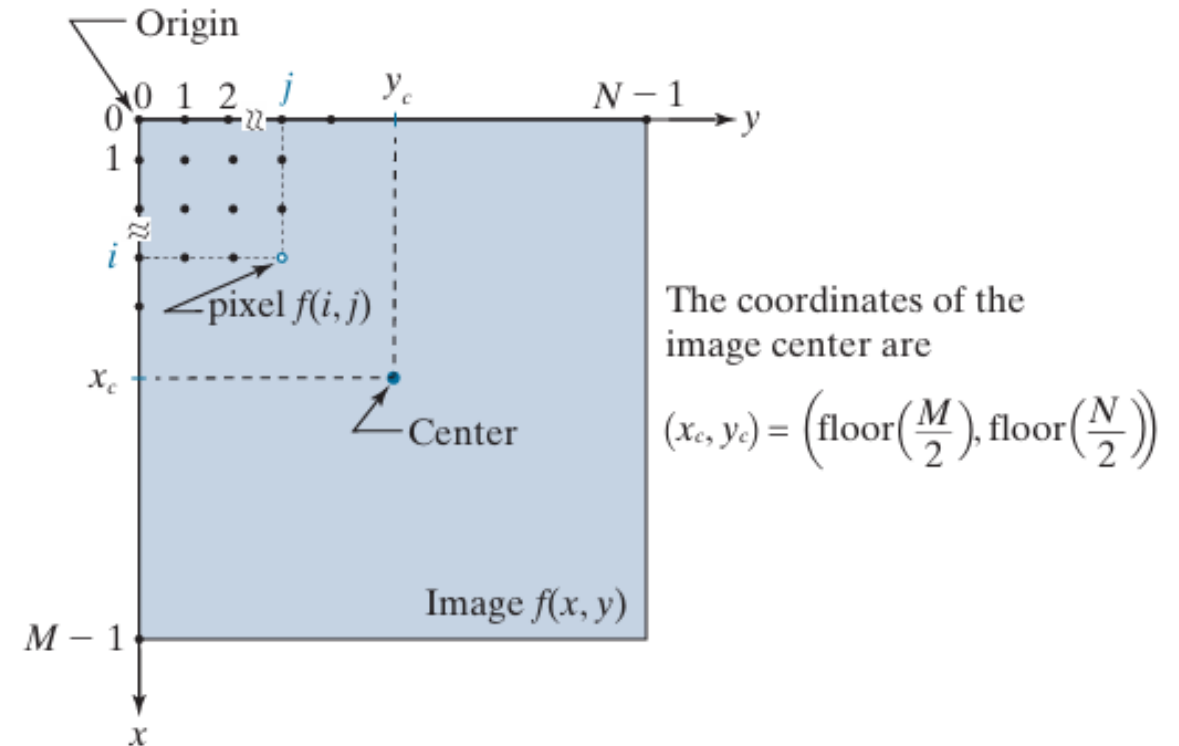
- Converting a continuous image into a digital form
- Digital image obtained via:
 - **Sampling** → discrete spatial locations
 - **Quantization** → convert intensity to discrete values
 - Digital image $f(x,y)$ with: M rows and N columns
 - **Top-left corner:** $f(0,0)$



$$f(x,y) = \begin{bmatrix} f(0,0) & f(0,1) & \dots & f(0,N-1) \\ f(1,0) & f(1,1) & \dots & f(1,N-1) \\ \vdots & \vdots & & \vdots \\ f(M-1,0) & f(M-1,1) & \dots & f(M-1,N-1) \end{bmatrix}$$

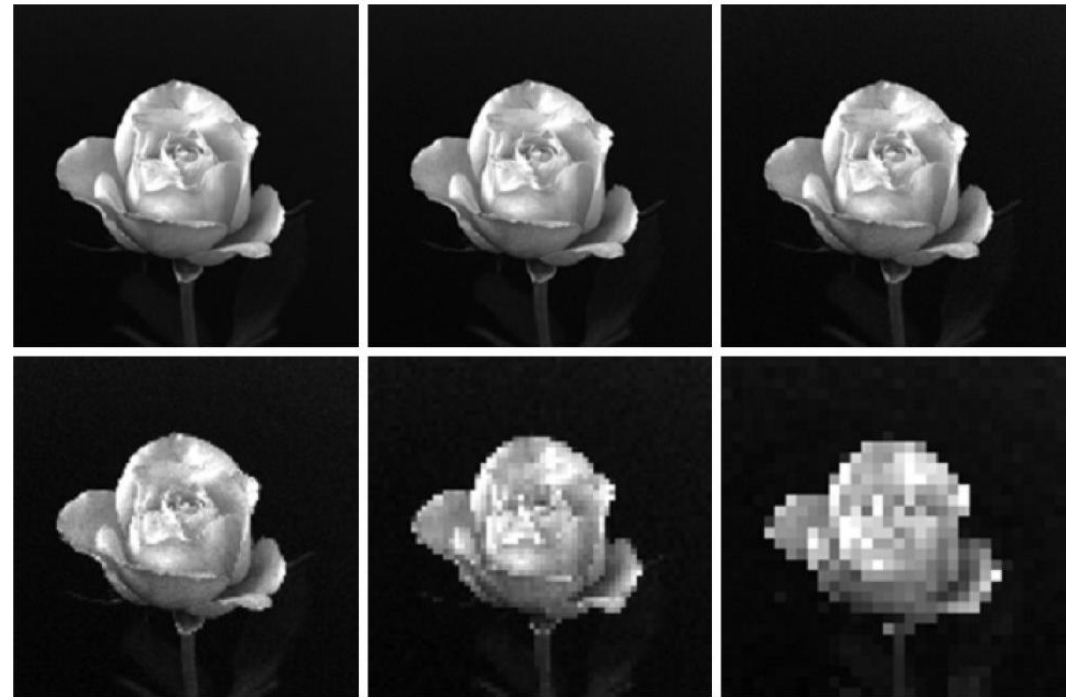
Representing Digital Images

- The origin (0,0) is defined at the top-left corner.
- Digital images are stored as **2D matrices**.
- Center coordinates found by:
 - $\text{center} = (\lfloor M/2 \rfloor, \lfloor N/2 \rfloor)$



Spatial Resolution

- Each pixel records a **small portion of the scene**.
- More pixels → can distinguish **smaller details** → higher spatial resolution.
- Same lens and same distance:
 - 20 Megapixel Camera → more pixels covering the same scene → finer details visible
 - 8 Megapixel Camera → fewer pixels → less detailed image



a	b	c
d	e	f

Intensity Resolution (Depth)

- **Intensity resolution** = smallest discernible change in **intensity/brightness**.
- Determines **how finely we can distinguish shades or brightness levels** in an image.
- Typically, an integer power of 2:
 - 8 bits \rightarrow 256 levels (2^8) \rightarrow most common

Grey-level resolution 2^1



Grey-level resolution 2^2



Grey-level resolution 2^3



Grey-level resolution 2^4



Grey-level resolution 2^5



Grey-level resolution 2^6



Grey-level resolution 2^7



Grey-level resolution 2^8



Why Use Color?

There is a rabbit in this image.

- Powerful descriptor → simplifies object identification and extraction.
- Human visual perception → can distinguish thousands of colors, vs ~24 shades of gray



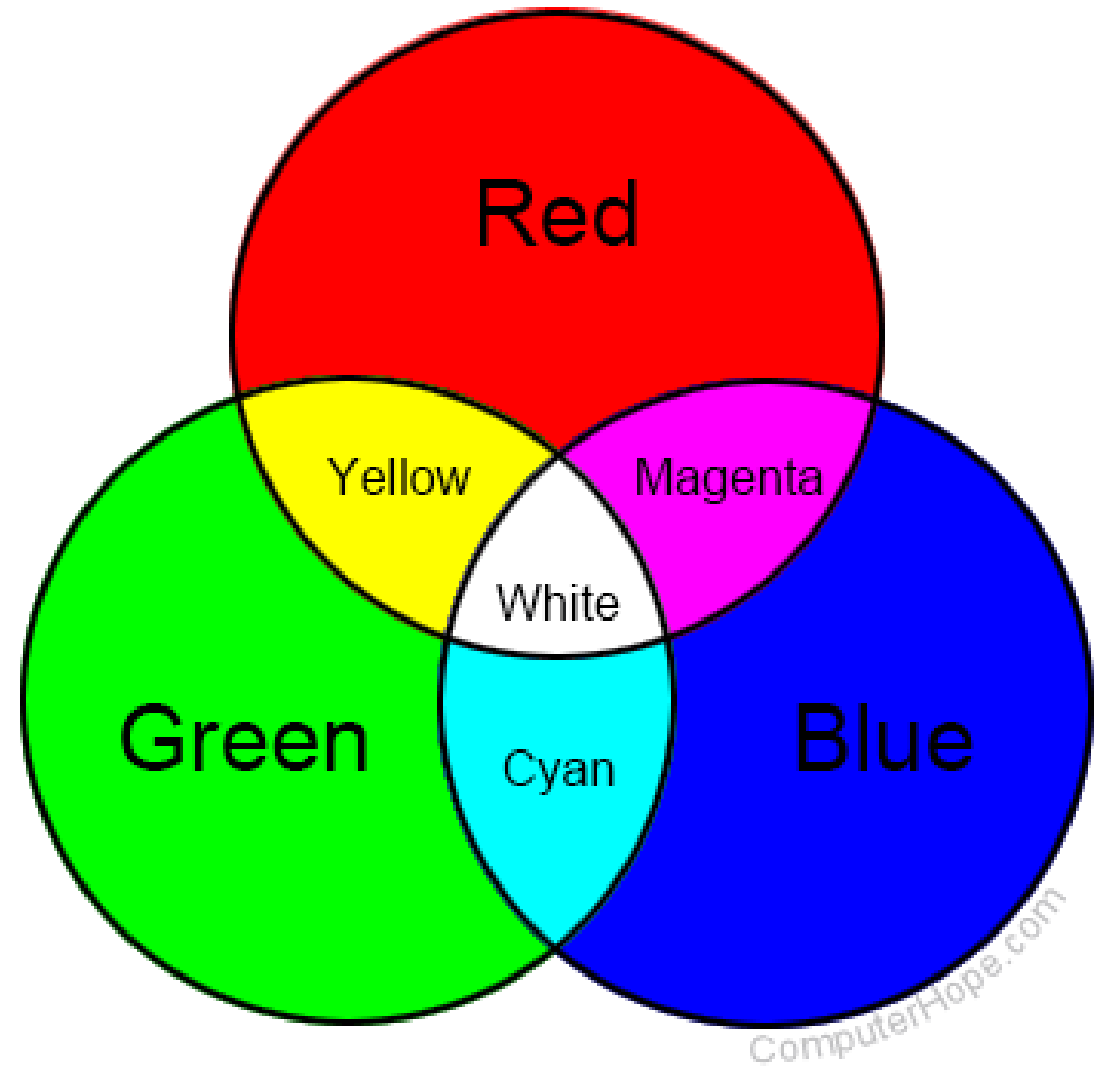


Cones and Color Vision in the Human Eye

- Cones are sensors in the eye responsible for color vision.
 - Cones are divided into three main types:
 - Red-sensitive cones → ~65%
 - Green-sensitive cones → ~33%
 - Blue-sensitive cones → ~2%
 - Human eye perceives **colors as combinations of the primary colors:**
Red (R), Green (G), Blue (B)
-

Primary and Secondary Colors of Light

- **Primary Colors:** Red (R), Green (G), Blue (B)
- Mixing all three primary colors in the correct intensities → **White**
- **Secondary Colors:** Produced by **adding two primary colors:**
 - Magenta (Red + Blue)
 - Cyan (Green + Blue)
 - Yellow (Red + Green)
- Mixing a secondary color with its opposite primary color → **White**
- Pure Green → $R=0, G=255, B=0$

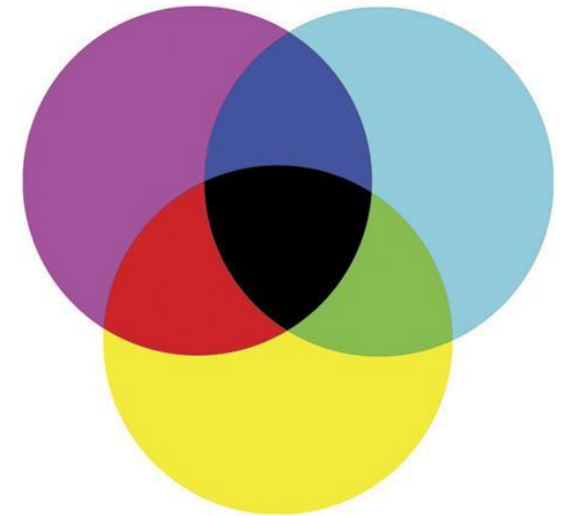


Primary Colors of Light vs Primary Colors of Pigments

- Light colors (RGB) are **additive** → mixing produces **white**
- Pigment colors (CMY) are **subtractive** → mixing absorbs light and can produce **black**



Additive color mixing



Subtractive color mixing

Primary pigment colors

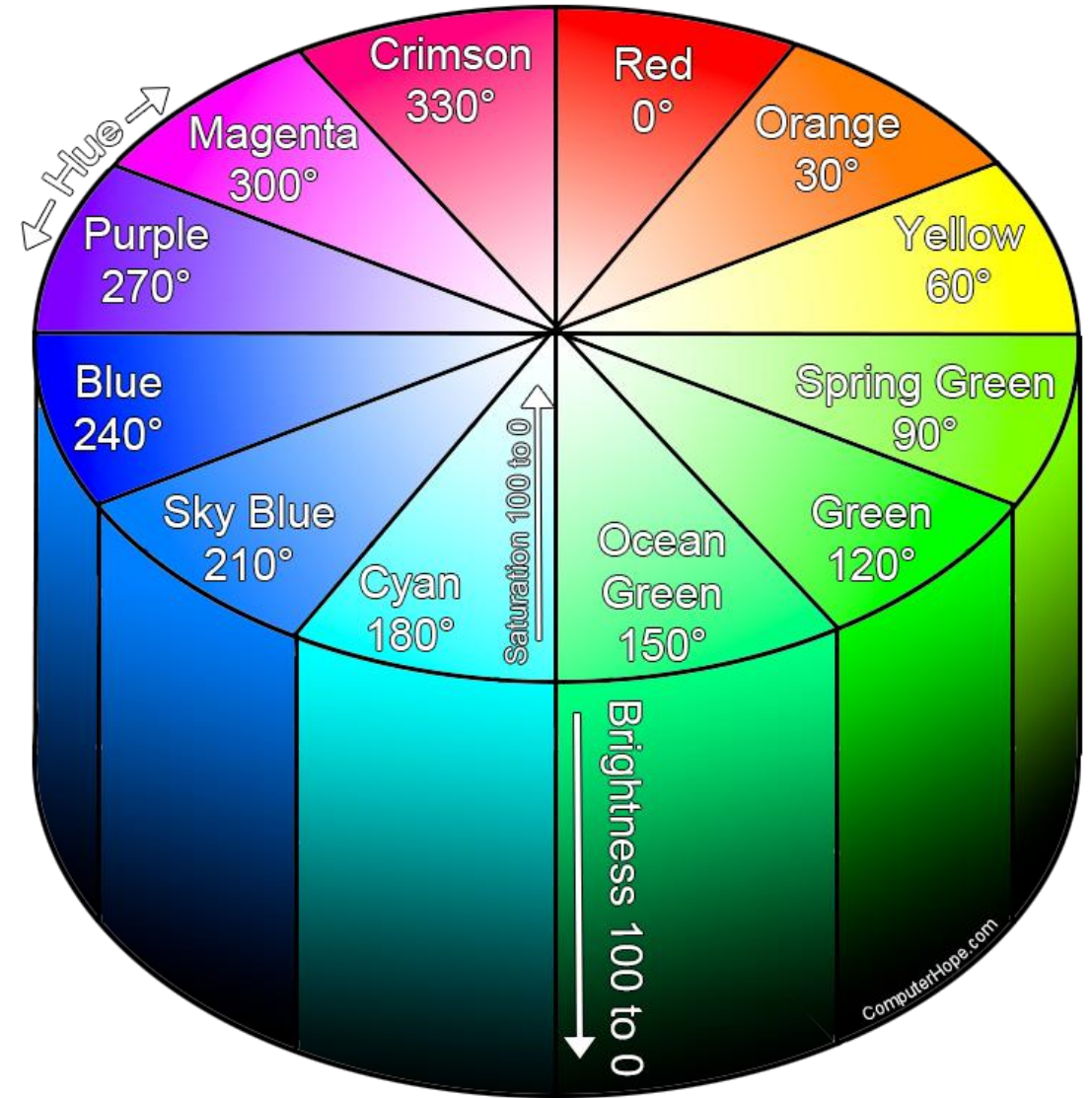
- A primary pigment **absorbs (subtracts) one primary color of light** and reflects or transmits the other two colors
- **Primary pigment colors:**
 - Cyan (absorbs Red, reflects Green + Blue)
 - Magenta (absorbs Green, reflects Red + Blue)
 - Yellow (absorbs Blue, reflects Red + Green)

Secondary Colors of Pigments

- Produced by mixing **two primary pigments**:
 - **Red** ← Magenta + Yellow
 - **Green** ← Cyan + Yellow
 - **Blue** ← Cyan + Magenta
- Mixing all three primary pigments (Cyan, Magenta, Yellow) → Black
- Or mixing **a secondary pigment with its opposite primary** → **Black**

Main Characteristics of Colors

- All colors can be defined using Hue, Saturation, and Brightness (HSB), an alternative to RGB, describing a color's type, intensity, and brightness
 - Brightness
 - Hue
 - Saturation



Brightness

- **Brightness (Lightness / Luminance)**
 - Represents intensity (lightness or darkness)
 - Measures how light or dark a color appears.
 - Adjusting brightness does not change the hue, only how light or dark it looks.



Hue

- Defined by the **dominant wavelength** of light
- Determines the **perceived color** (e.g., red, orange, yellow)
- When we say, “the object is red,” we are describing its hue
- Represented on a 360° color wheel:
 - 0° = Red
 - 120° = Green
 - 240° = Blue
 - 360° = Red (loop)



Saturation

- Refers to the **purity of a color**
- Pure spectrum colors → **fully saturated**
- Adding white light → **reduces saturation**
- Example:
 - Pure Purple → Saturated
 - lavender (Purple + White) → Less Saturated

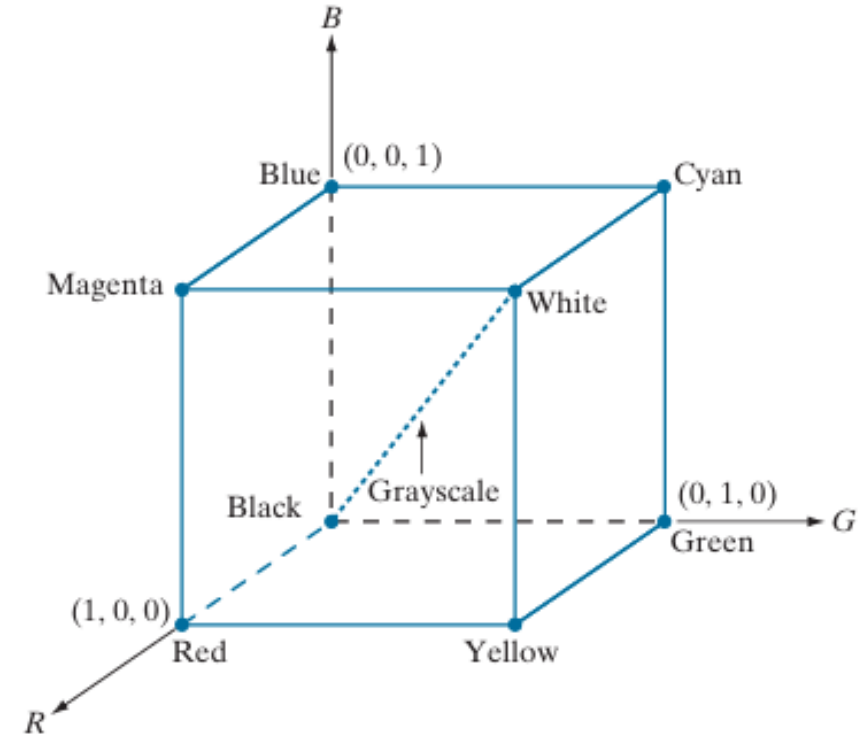


Color Models (Color Spaces / Color Systems)

- Provide a **standard way** to specify colors
- Allow communication of colors using **numbers**, not just names
- Hardware-Oriented Models or Application-Oriented Models
- Examples of Color Models:
 - **RGB Model**
 - **CMY/CMYK model**
 - **HSV Model**

RGB Color Model

- Colors are represented by **3 primary components**:
 - R = Red, G = Green, B = Blue
 - Each color = a combination of R, G, and B
- RGB is represented in a 3D cube
- Each color = a **point** inside or on the cube



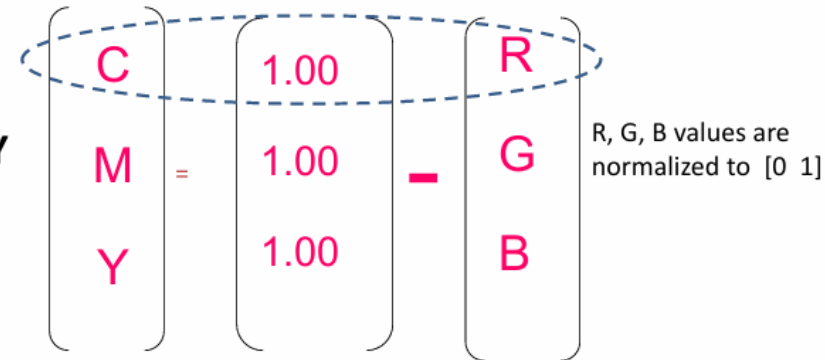
RGB Image Representation

- In digital images, we specify the levels of R, G and B with 8-bit integers
- RGB color pixel is a triplet of values (R, G, B)
- Each RGB color pixel has a depth of 24 bits.

CMY and CMYK Color Models

- **CMY = Cyan, Magenta, Yellow**
 - Secondary colors of **light (RGB)**
 - Primary colors of **pigments/printing**
- Each pigment **removes** one primary color of light
 - **Cyan pigment** absorbs **Red**, reflects (Green + Blue = Cyan)
 - **Magenta pigment** absorbs **Green**, reflects (Red + Blue = Magenta)
 - **Yellow pigment** absorbs **Blue**, reflects (Red + Green = Yellow)

RGB to CMY
conversion



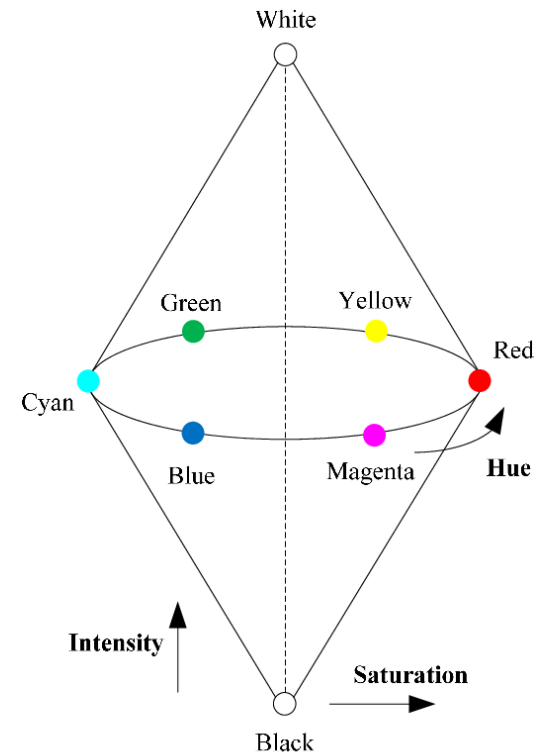
The diagram illustrates the conversion from RGB to CMY. It features three vertical containers. The first container on the left holds the letters 'C', 'M', and 'Y' in pink. The second container in the middle holds the value '1.00' in pink. The third container on the right holds the letters 'R', 'G', and 'B' in pink. A blue dashed oval encircles the 'C' in the first container, the '1.00' in the second, and the 'R' in the third. A pink equals sign is placed between the first and second containers, and a pink minus sign is placed between the second and third containers. To the right of the third container, a text note states: 'R, G, B values are normalized to [0 1]'.

$$\begin{bmatrix} C \\ M \\ Y \end{bmatrix} = \begin{bmatrix} 1.00 \\ 1.00 \\ 1.00 \end{bmatrix} - \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

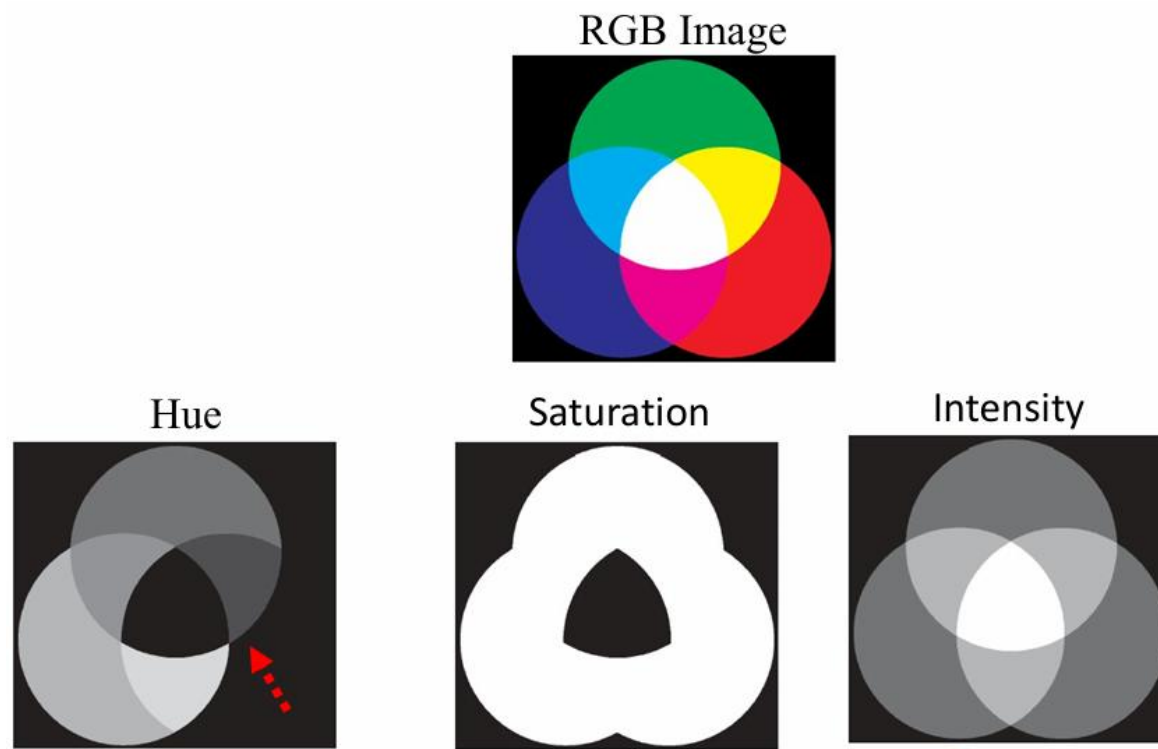
R, G, B values are normalized to [0 1]

HSI Color Model

- RGB and CMYK is not practical for human interpretation.
- **HSI = Hue + Saturation + Intensity**
 - **Hue:** the actual color (red, blue, green...)
 - **Saturation:** purity of the color (dull \leftrightarrow vivid)
 - **Intensity:** brightness/lightness (dark \leftrightarrow bright)



HSI Component Images



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

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

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

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

The background is a solid teal color. Overlaid on this are several stylized, low-angle views of skyscrapers, creating a sense of height and urban architecture. A large, double-lined white diamond shape is centered on the image, framing the text.

Thank You