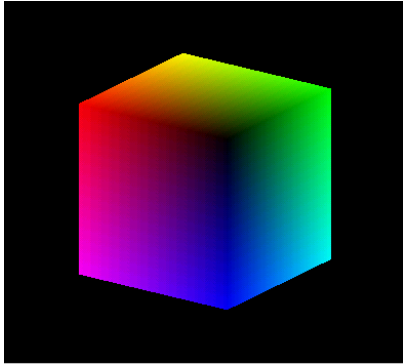
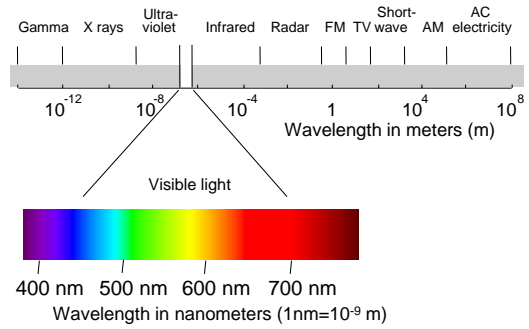


Color Representation

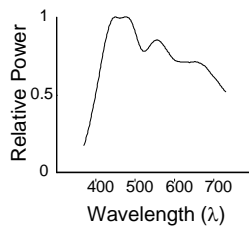


Electromagnetic Radiation - Spectrum

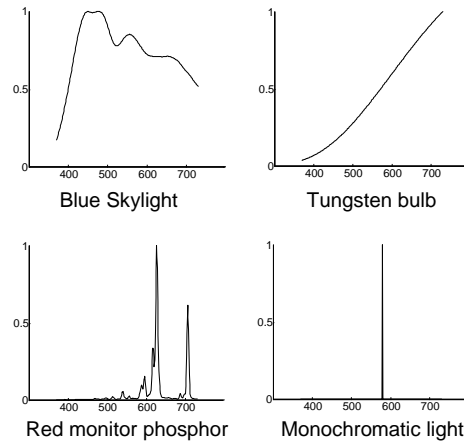


Spectral Power Distribution

The **Spectral Power Distribution (SPD)** of a light is a function $f(\lambda)$ which defines the energy at each wavelength.

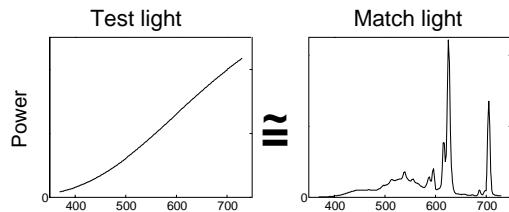
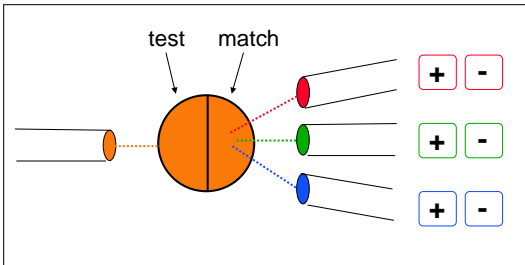


Examples of Spectral power Distributions



Color Matching Experiment

Three primary lights are set to match a test light.



Metamer - two lights that appear the same visually. They might have different SPDs (spectral power distributions).

Trichromatic Color Theory

“tri”=three “chroma”=color
color vision is based on three primaries
(i.e., it is 3 dimensional).

Thomas Young (1773-1829) -

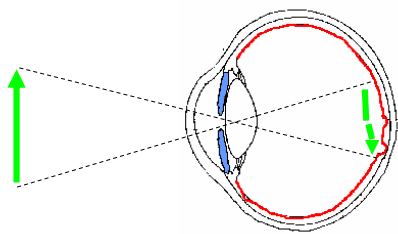
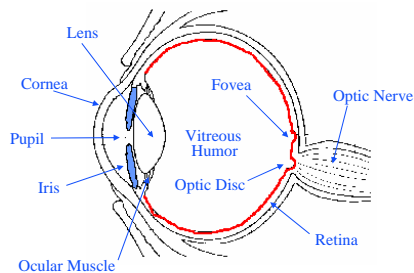
A few different retinal receptors operating with different wavelength sensitivities will allow humans to perceive the number of colors that they do. Suggested 3 receptors.

Helmholtz & Maxwell (1850) -

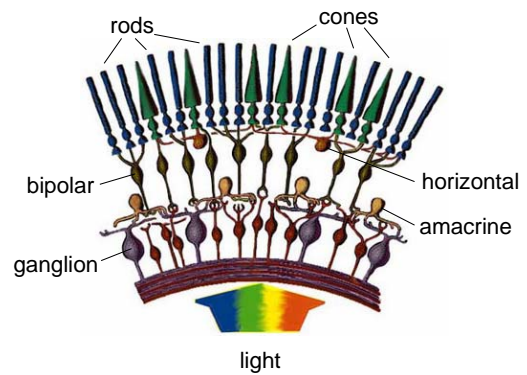
Color matching with 3 primaries.



The Human Eye



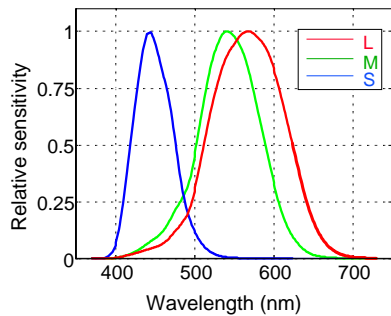
The Human Retina



Retinal Photoreceptors

- Cones -**
- High illumination levels (Photopic vision)
 - Less sensitive than rods.
 - 5 million cones in each eye.
 - Only cones in fovea (aprox. 50,000).
 - Density decreases with distance from fovea.
 - 3 cone types differing in their spectral sensitivity: L, M, and S cones.

Cone Spectral Sensitivity



Linear Color Spaces

Colors in 3D color space can be described as linear combinations of 3 basis colors:
primaries

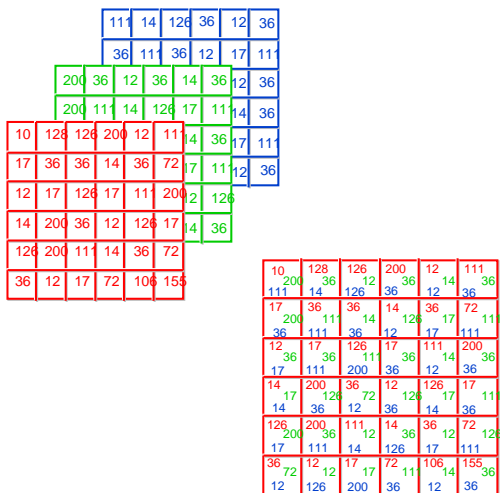
$$\text{Color} = a \cdot \text{Primary 1} + b \cdot \text{Primary 2} + c \cdot \text{Primary 3}$$

The representation of :

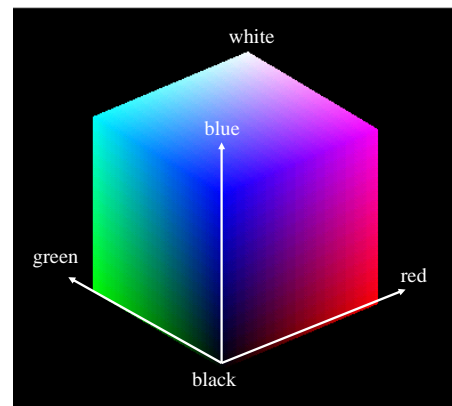


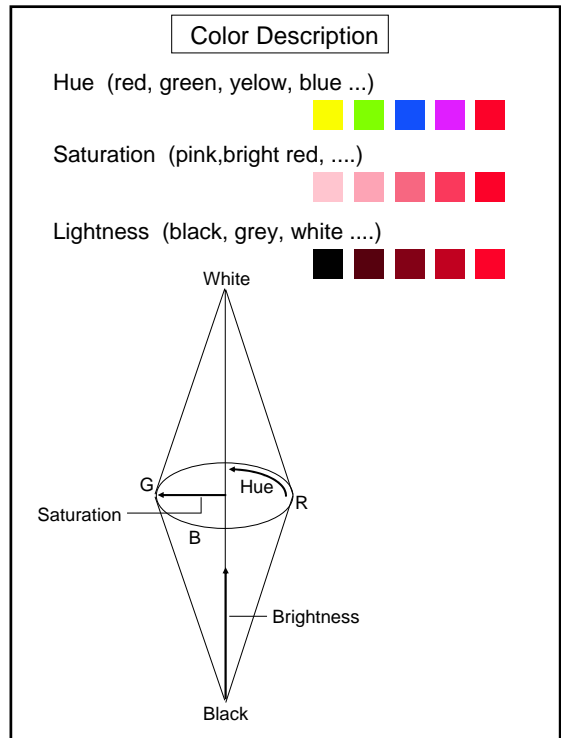
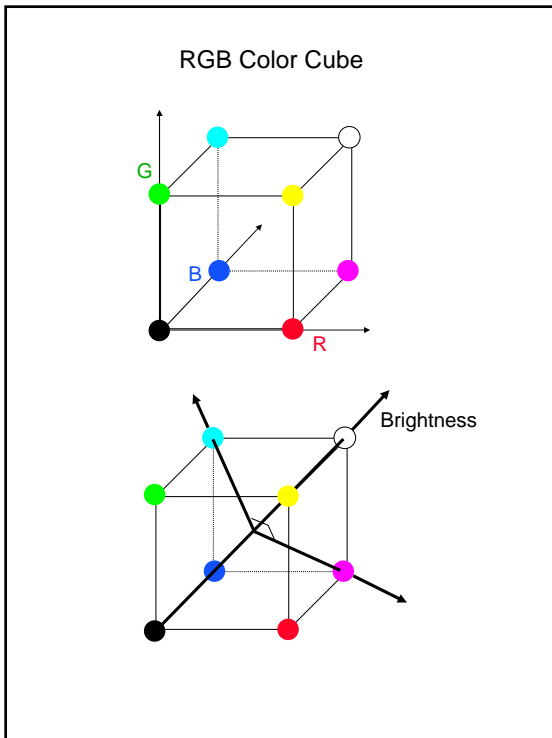
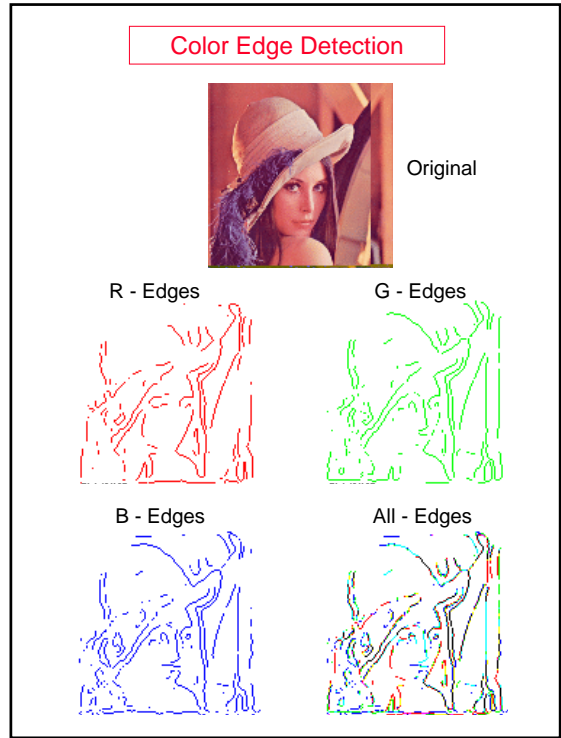
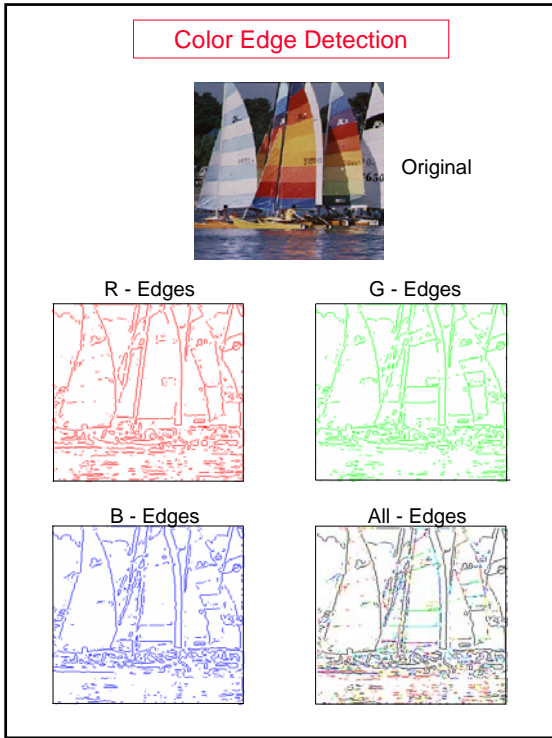
is then given by: (a, b, c)

Rgb Image

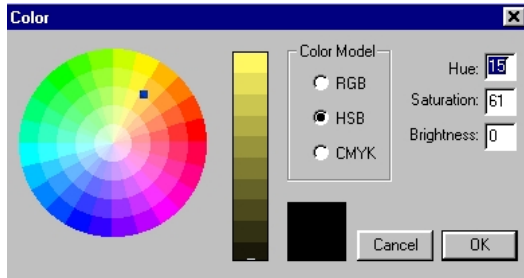


The RGB Cube

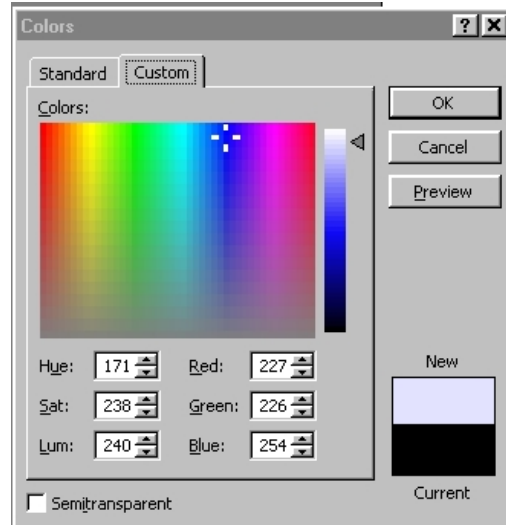




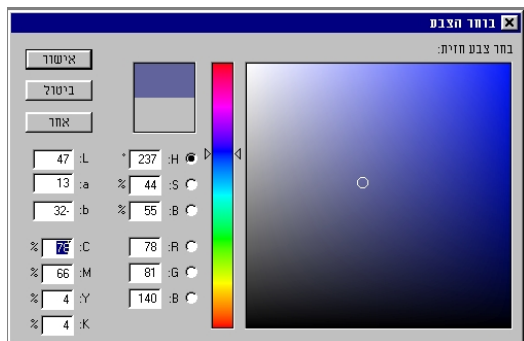
MayuraDraw



PowerPoint



PhotoShop



YIQ - Color Space

NTSC = National Television Systems Committee

Y = luminance

I = red-green

Q = blue-yellow

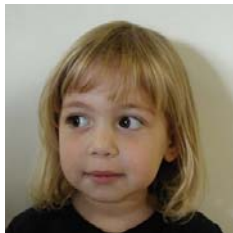
$$\begin{bmatrix} Y \\ I \\ Q \end{bmatrix} = \begin{bmatrix} 0.177 & 0.813 & 0.011 \\ 0.540 & -0.263 & -0.174 \\ 0.246 & -0.675 & 0.404 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

R G B are the CIE-RGB

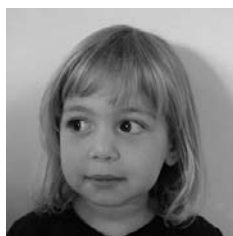


RGB To Monochrome

RGB



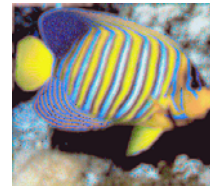
Y



Original



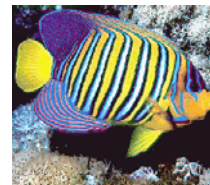
Y - Blur



I - Blur



Q - Blur



Subtractive Color System - CMYK

Printer Dyes:

Cyan = removes red

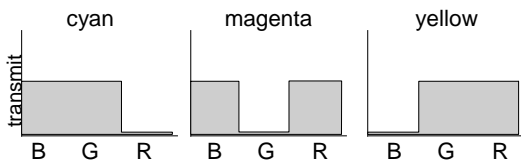
Magenta = removes green

Yellow = removes blue

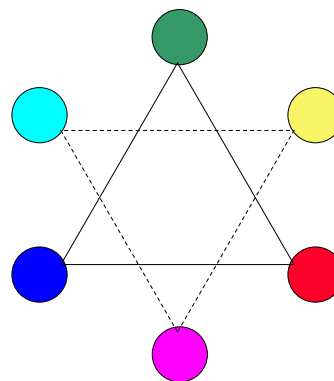
black = removes all



Ideal block dyes:



Opponent Color Wheel

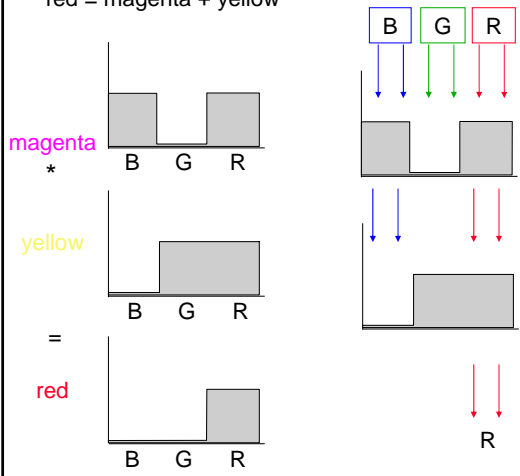


— Additive primaries

- - - Subtractive Primaries

Multiplicative (Subtractive) Color System

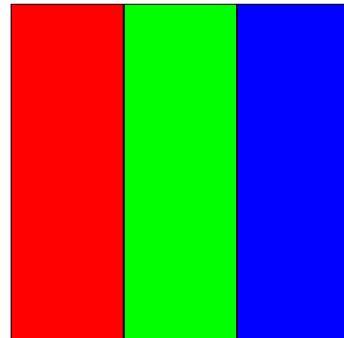
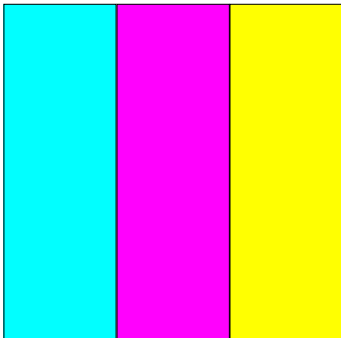
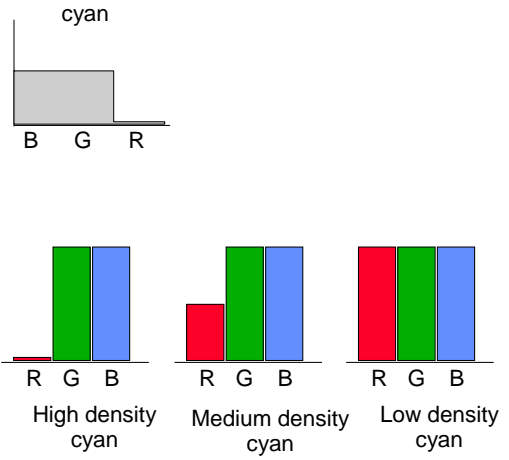
red = magenta + yellow

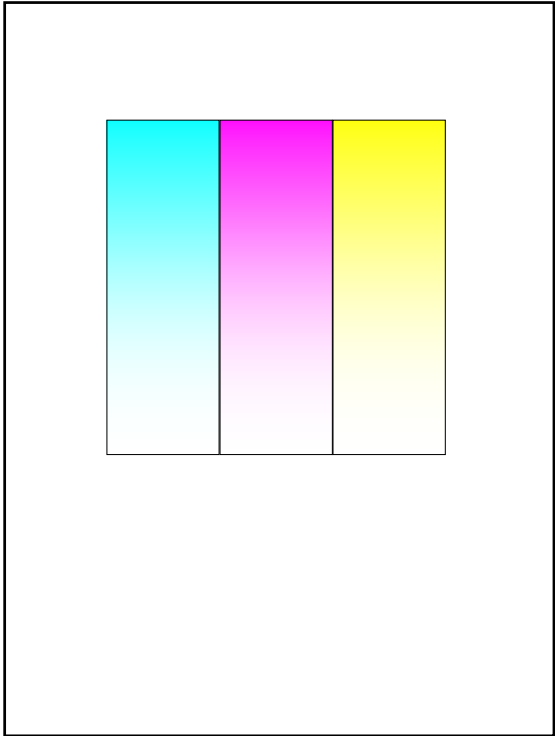


red = magenta + yellow
 green = cyan + yellow
 blue = magenta + cyan

Cyan - controls amount of red in print:

low C = high R (also high G and B)
 high C = low R (high G and B)



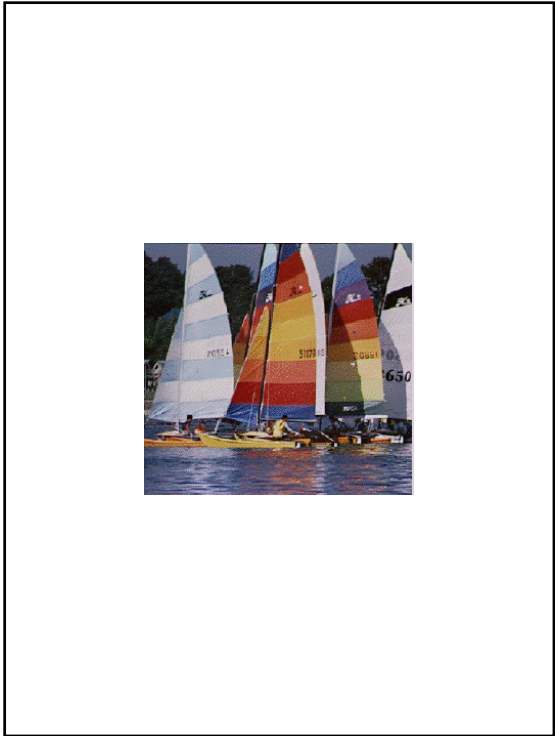
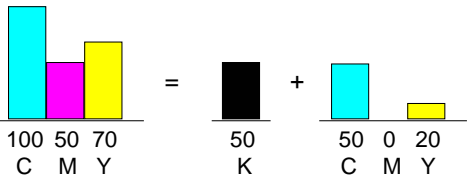


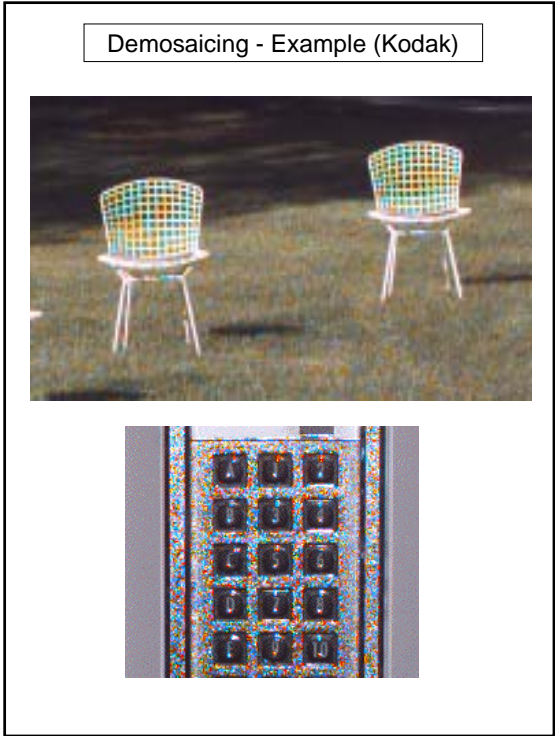
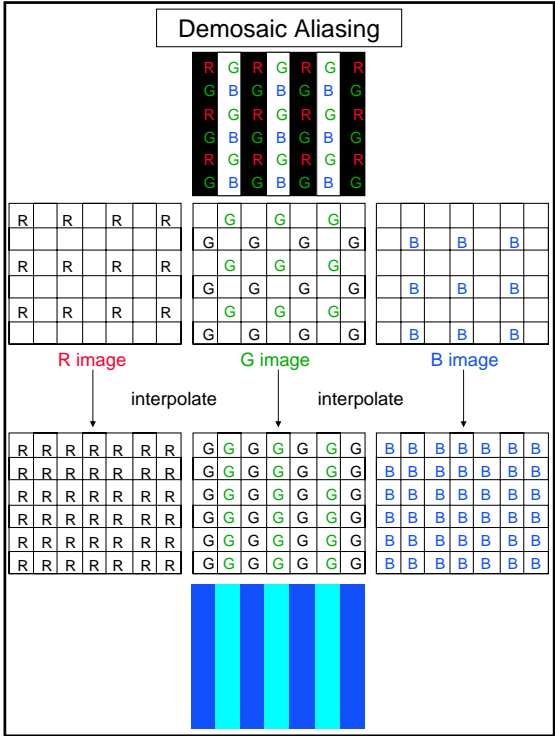
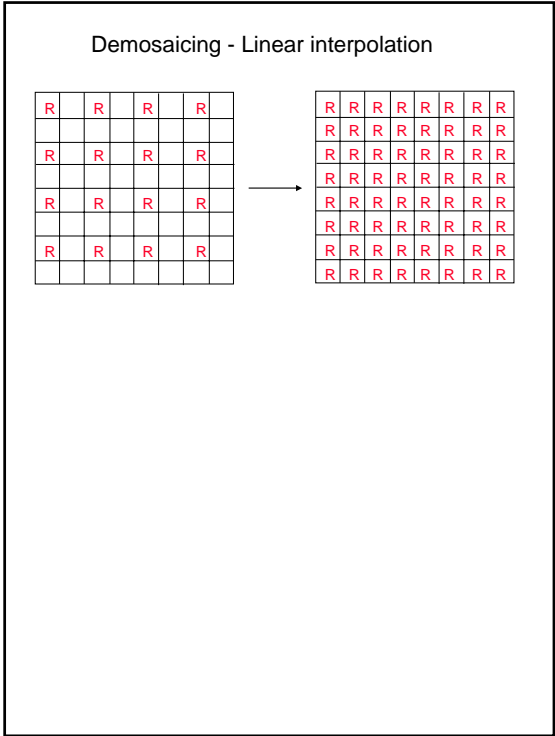
CMY + Black

$C + M + Y = K$ (black)

- Using three inks for black is expensive
- C+M+Y = dark brown not black
- Black instead of C+M+Y is crisper with more contrast.

Undercolor removal -
(gray component replacement)





Demosaicing - Various Approaches

Regularization

Minimize over a functional with a data fit term and an inter-channel color correlation term. (Gamer & Keren)

Minimal Surface

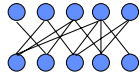
Minimize over a functional with a data fit term and a 5D surface area term. (Beltrami Flow) (Kimmel)

Demosaicing - Various Approaches

Learning Schemes

Learn linear and non linear optimal filters for classes of images (ANN).

(Kapur & Hel-Or)



Demosaicing - Learning Schemes

Channel independent



Perceptron (Linear)

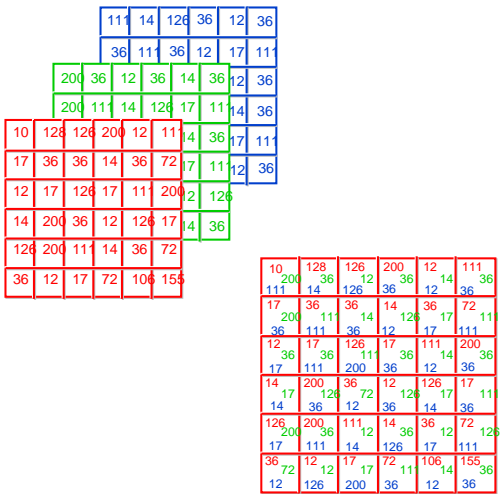


Quadratic - Learned on Image



Quadratic - Learned on Class

Color Quantization



Indexed Image

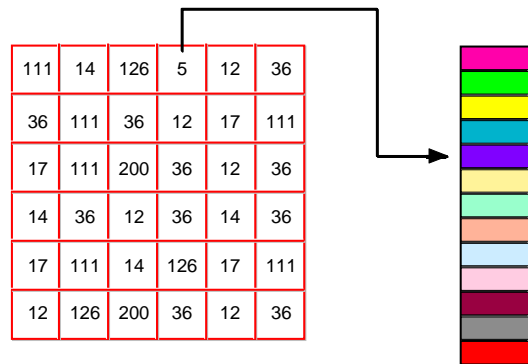


Image Independent Quantization

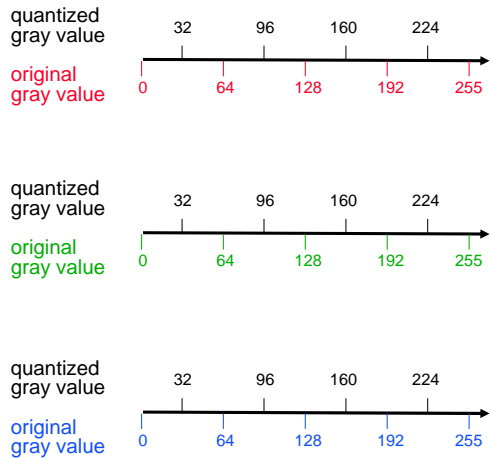


Image Independent Quantization

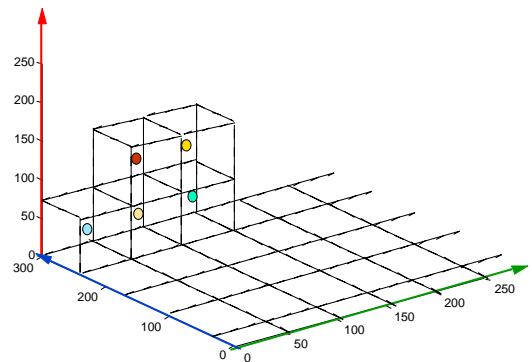
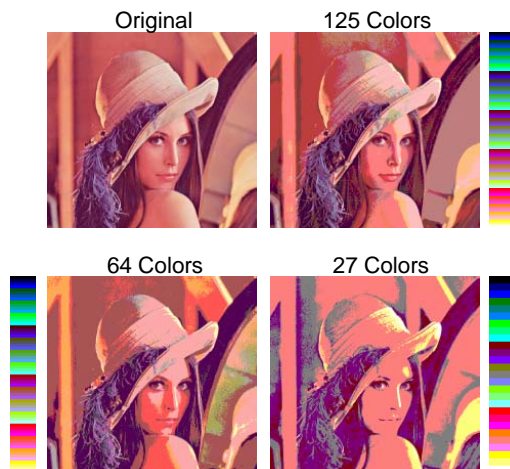
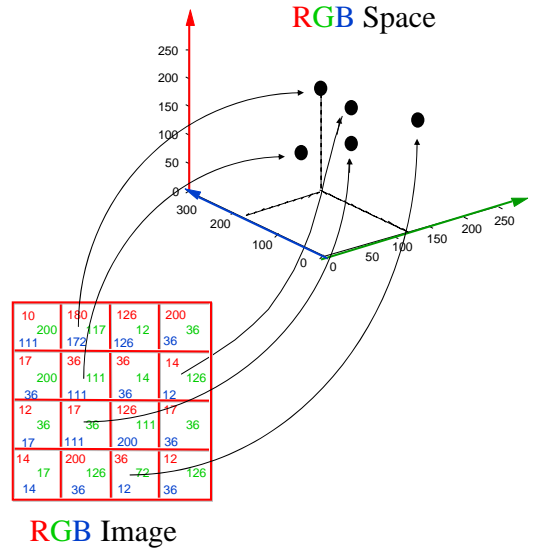
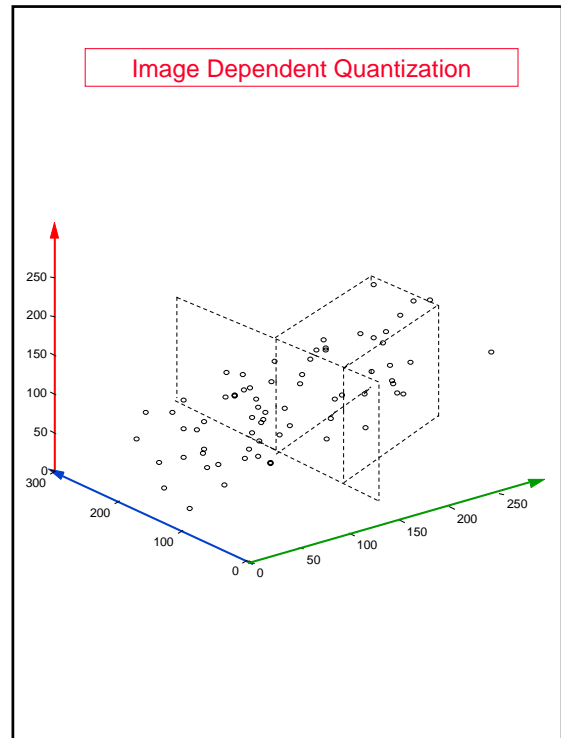
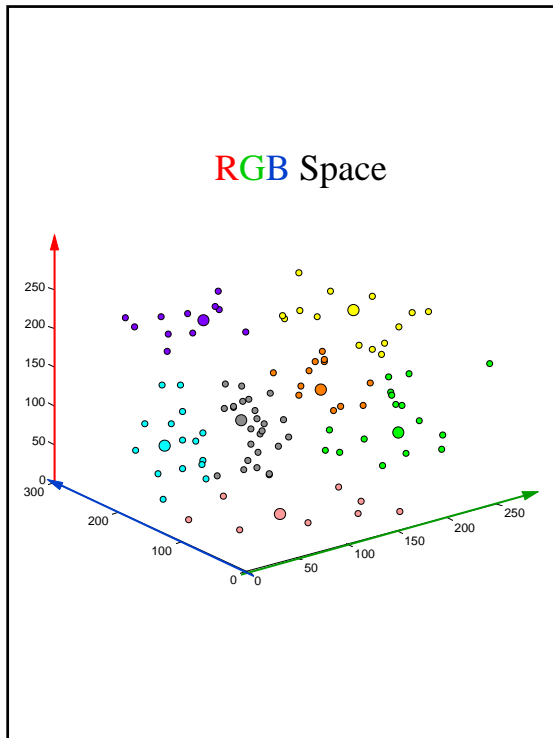


Image Independent Quantization



RGB Space





Clustering using Iso Data

Input: $C = \{c_i\}$, $i=1..n$ - color points.
Output: $S = \{s_j\}$, $j=1..k$ - color indices.

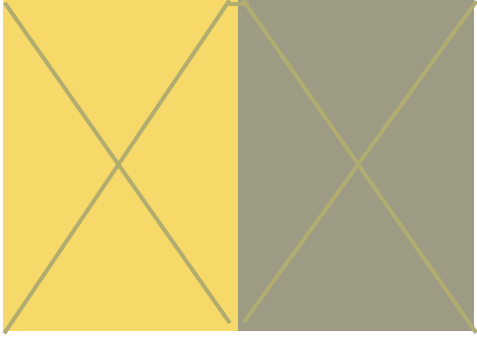
- Distribute s_j , $j=1..k$, uniformly in color space.
- Divide C into k classes based of distances to S .
- For each class j , calculate the mean M_j .
- Set $S_j = M_j$.
- Iterate until convergence.

Image Dependent Quantization

Original

Independent

Dependent



Albers (1975)