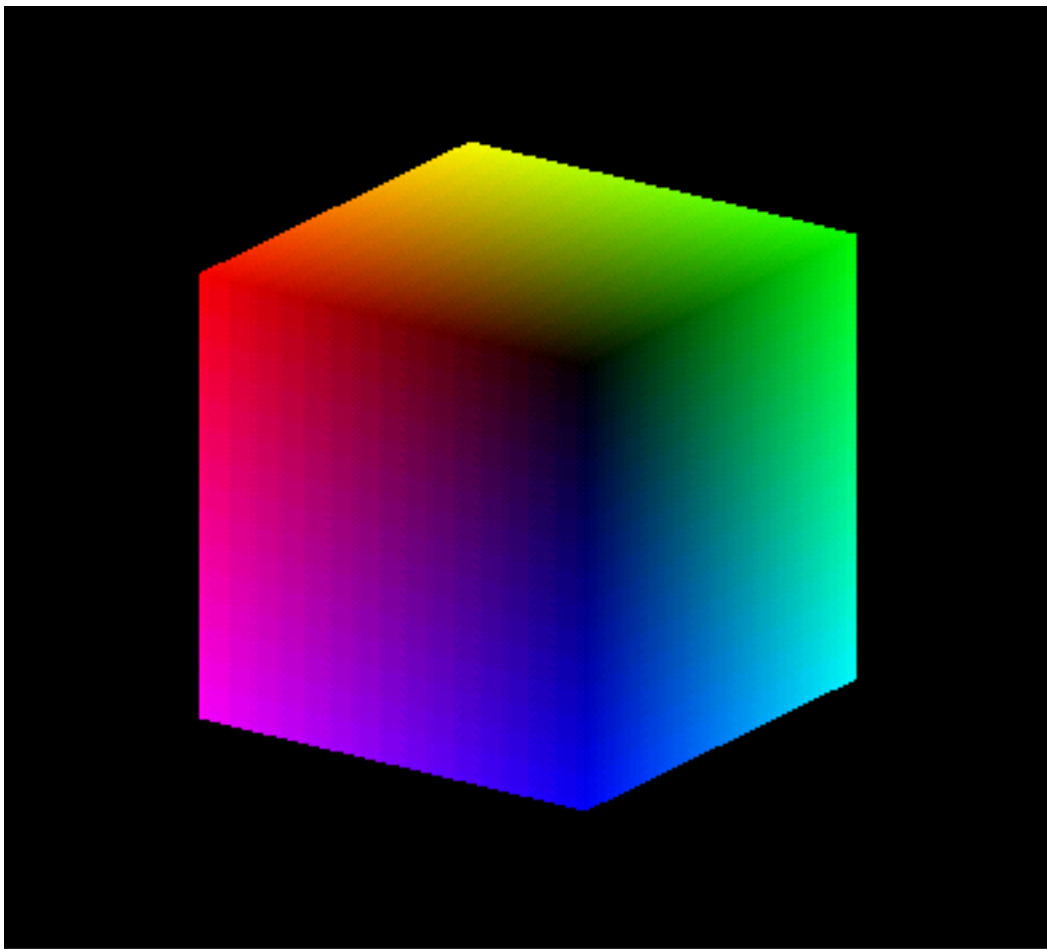
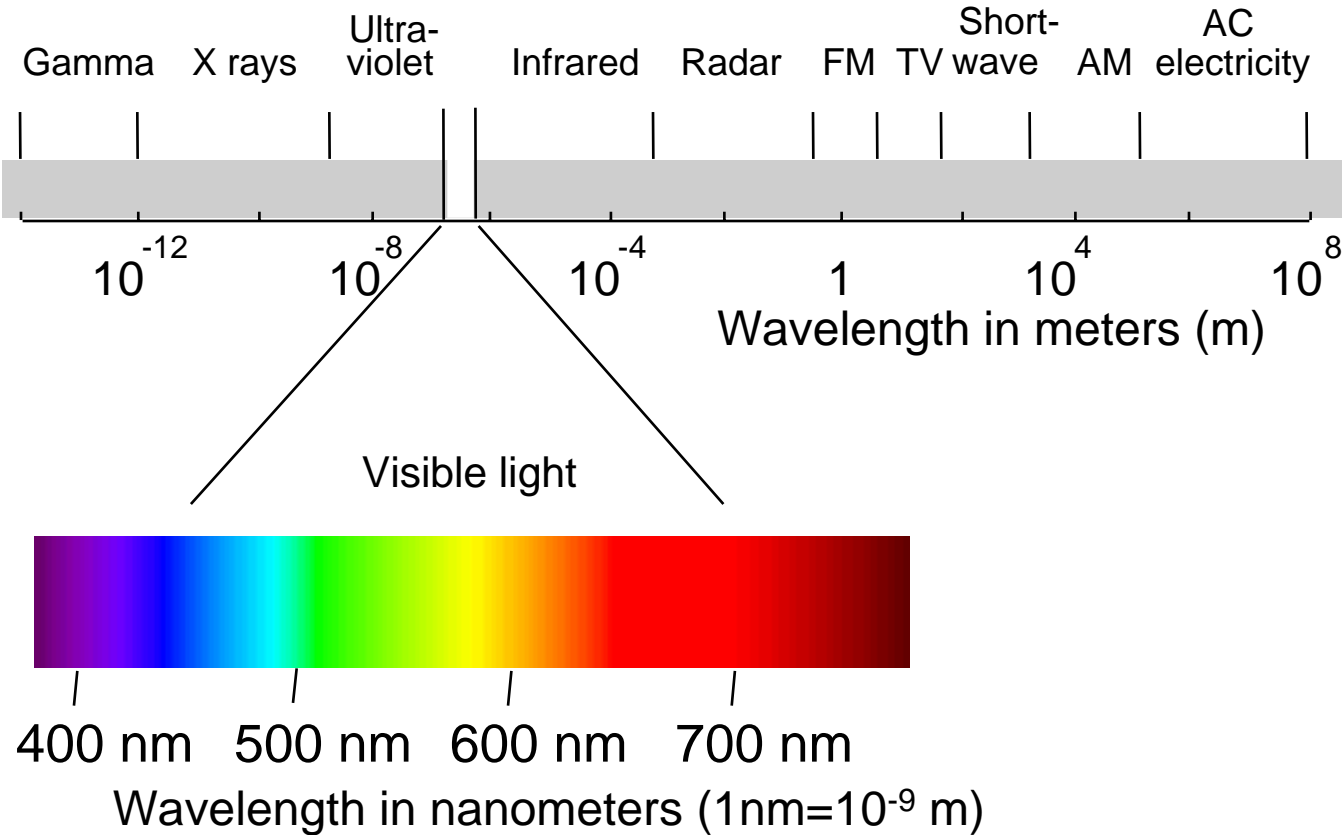


Color Representation

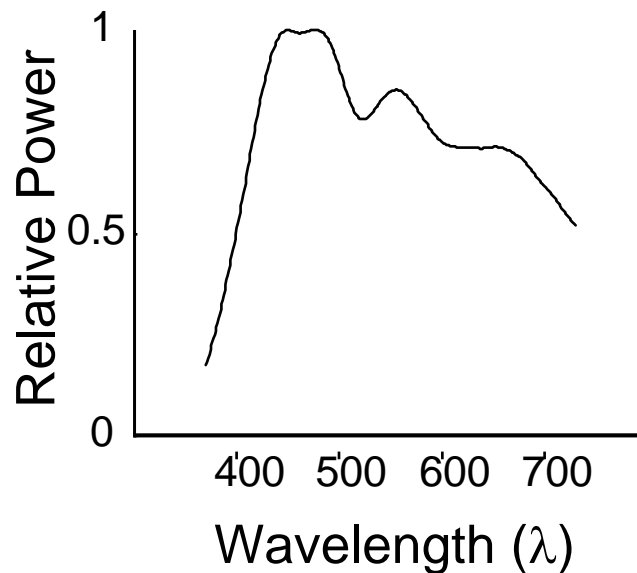


Electromagnetic Radiation - Spectrum

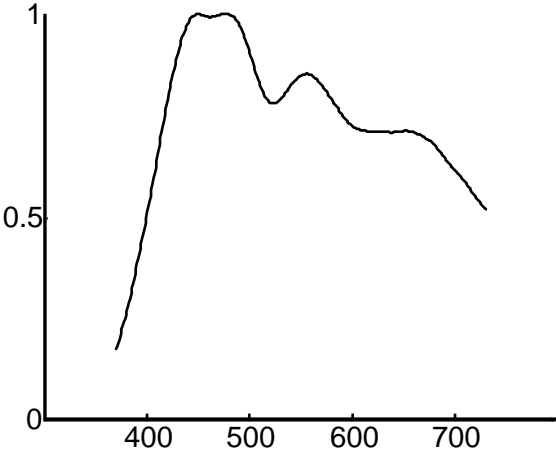


Spectral Power Distribution

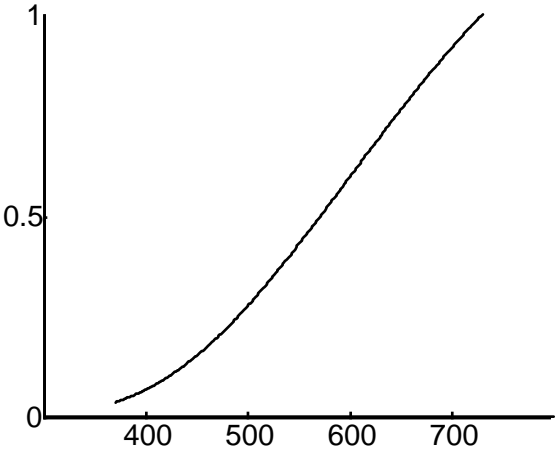
The **Spectral Power Distribution (SPD)** of a light is a function $f(\lambda)$ which defines the energy at each wavelength.



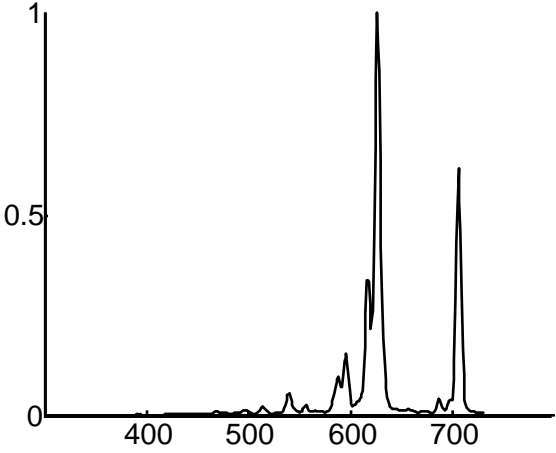
Examples of Spectral power Distributions



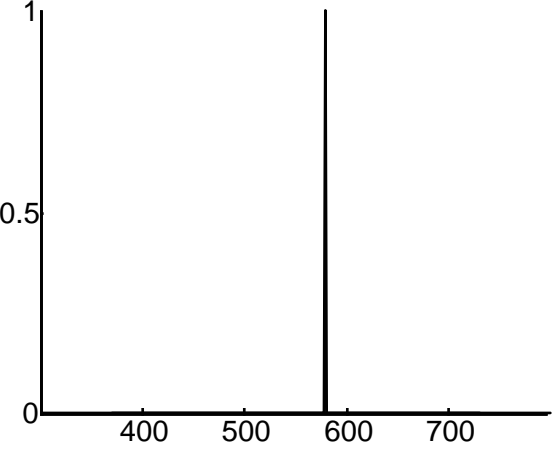
Blue Skylight



Tungsten bulb



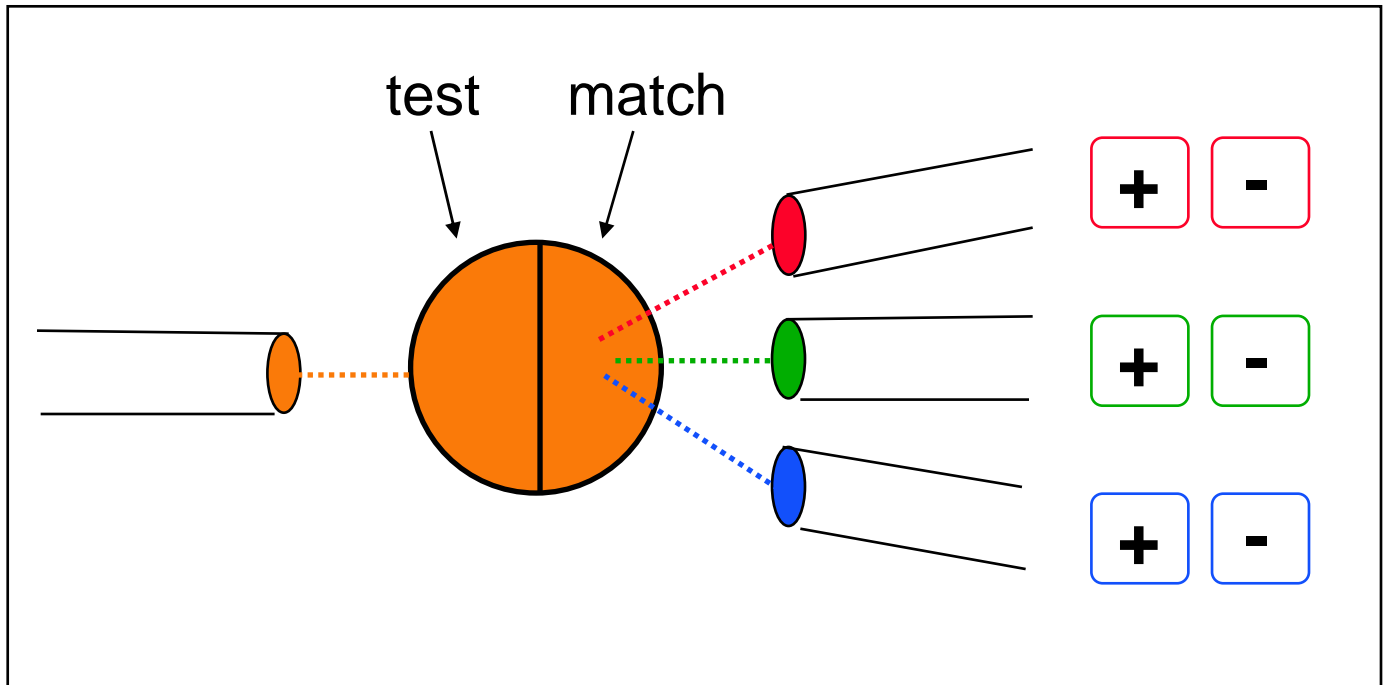
Red monitor phosphor



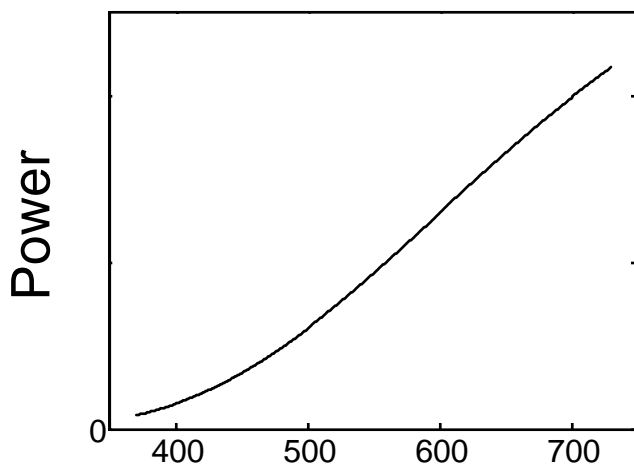
Monochromatic light

Color Matching Experiment

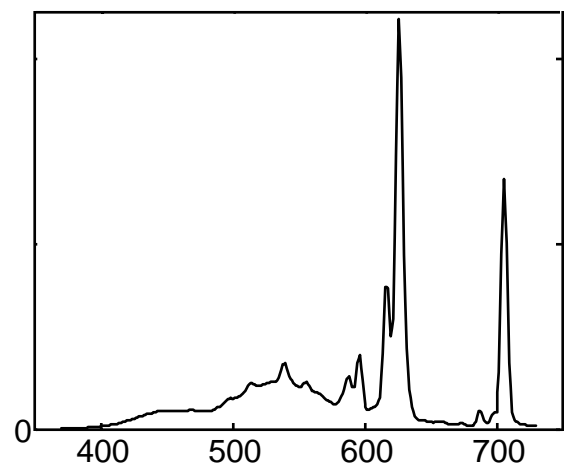
Three primary lights are set to match a test light.



Test light



Match 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).

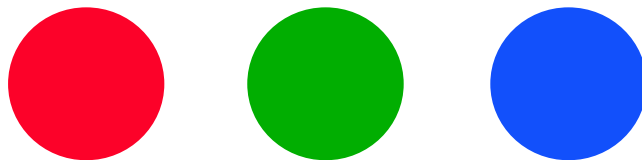
Thomas Young (1773-1829) -

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

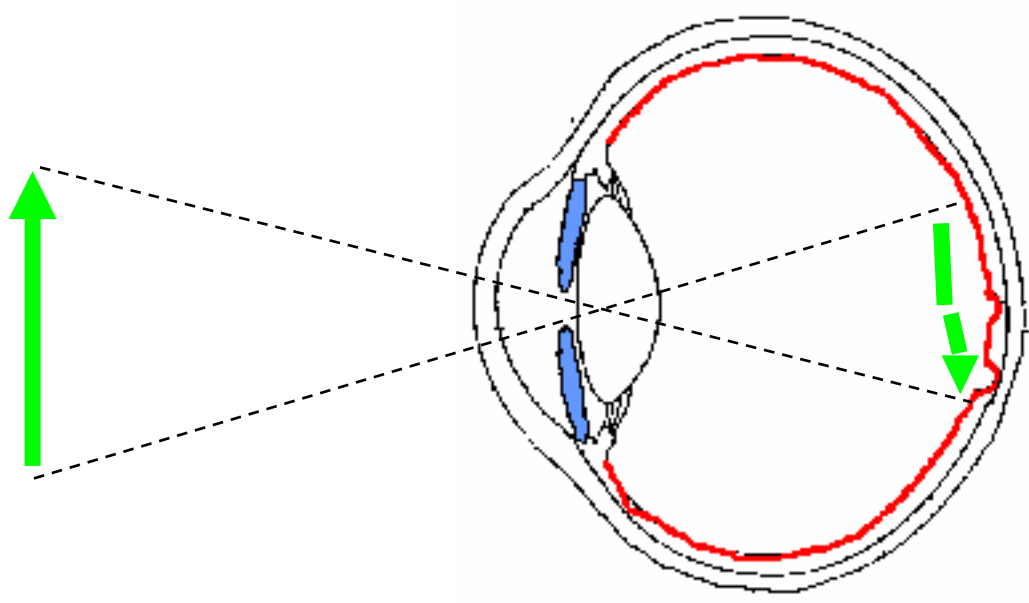
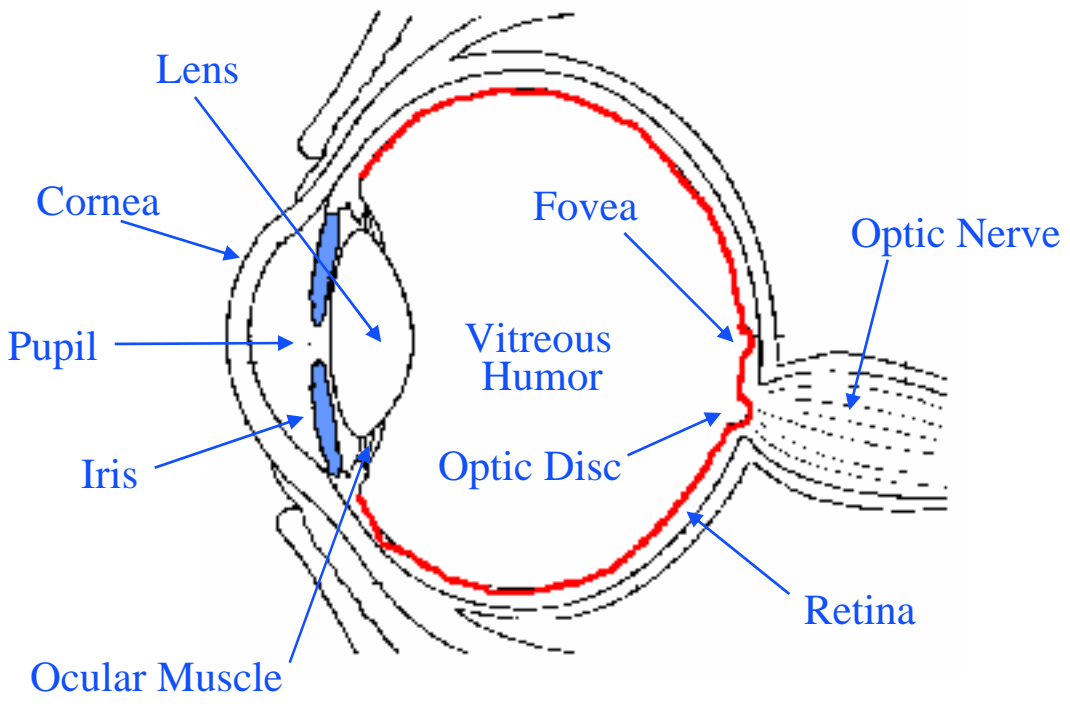
Suggested 3 receptors.

Helmholtz & Maxwell (1850) -

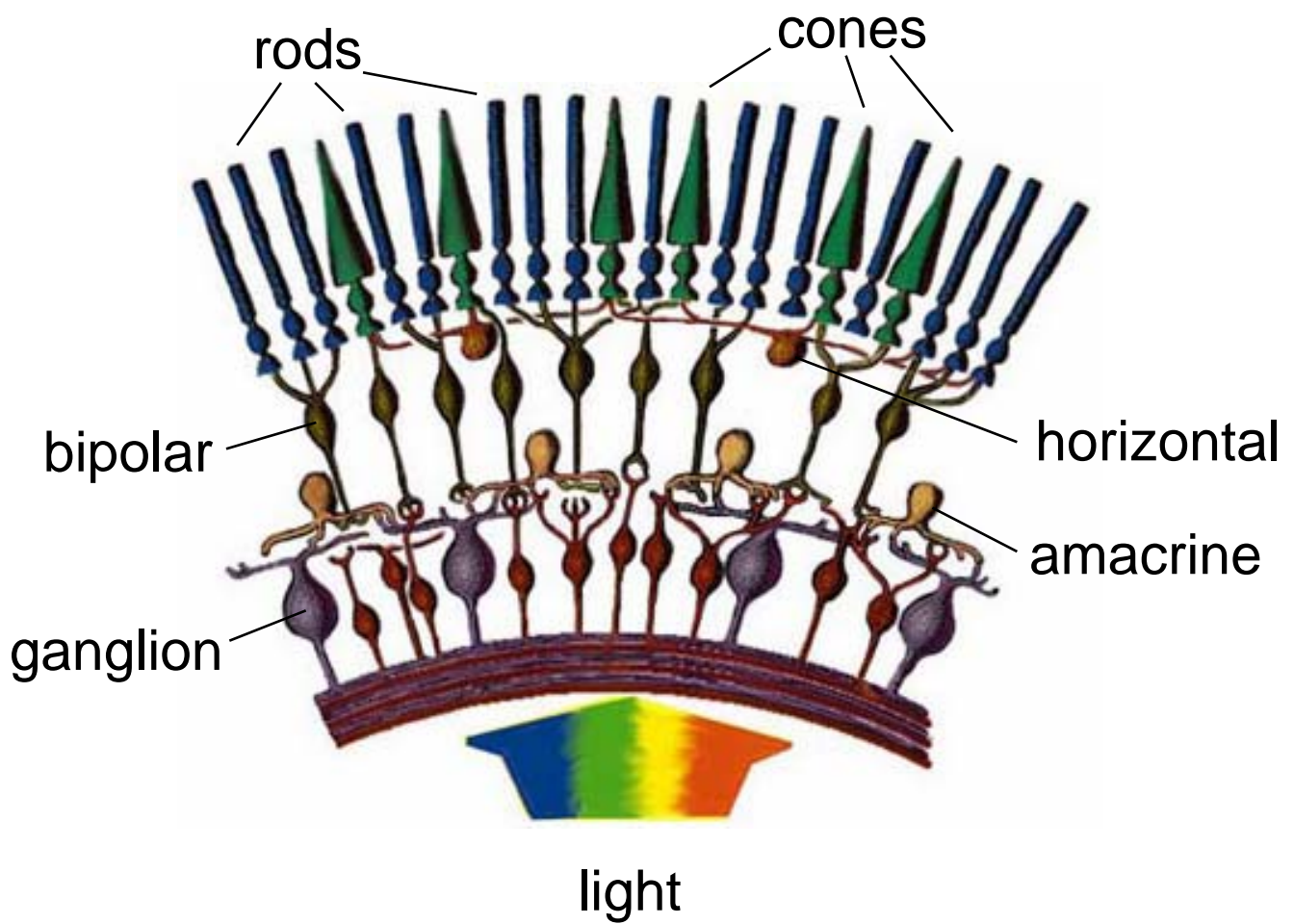
Color matching with 3 primaries.



The Human Eye



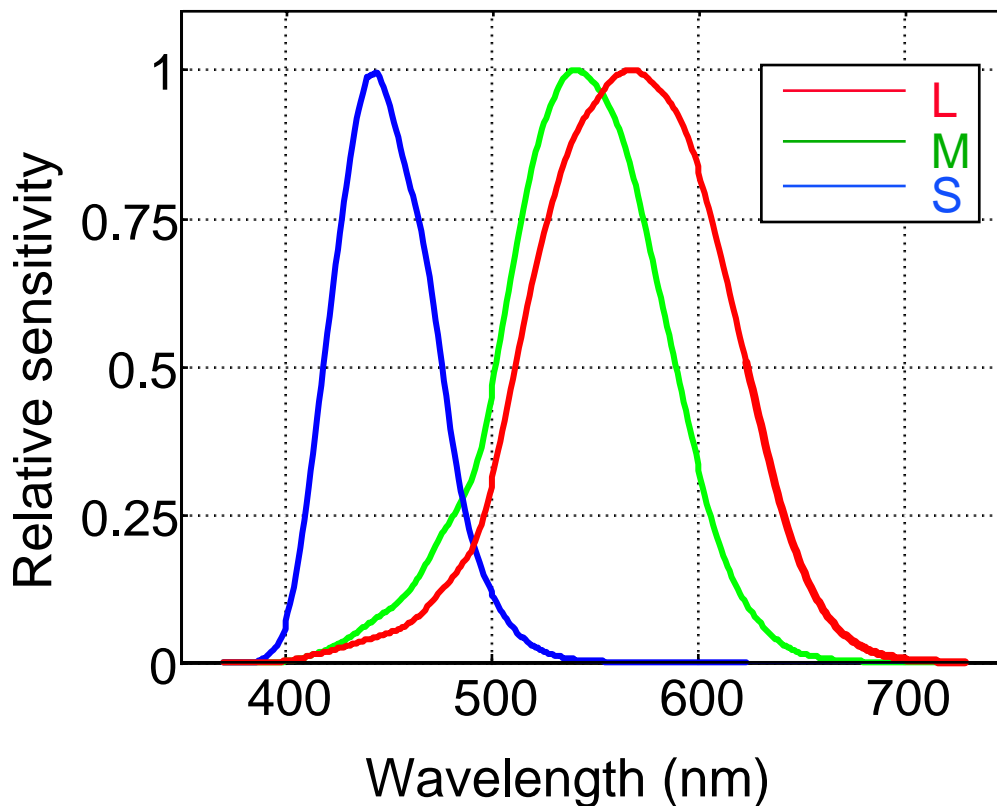
The Human Retina



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.

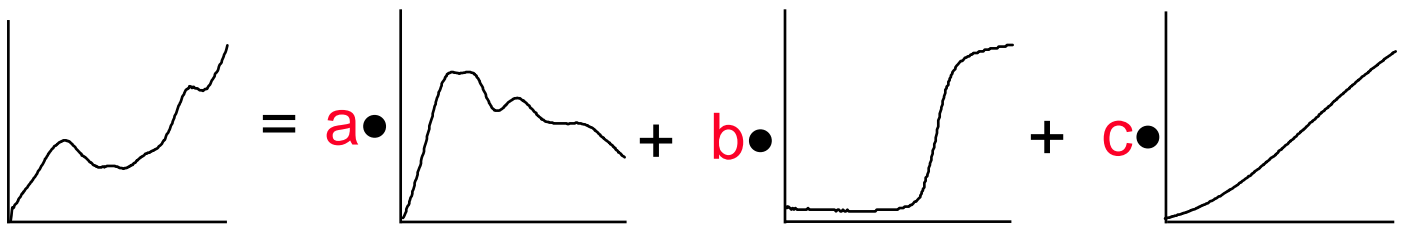
Cone Spectral Sensitivity



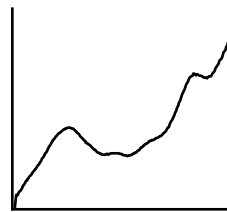
Linear Color Spaces

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

primaries

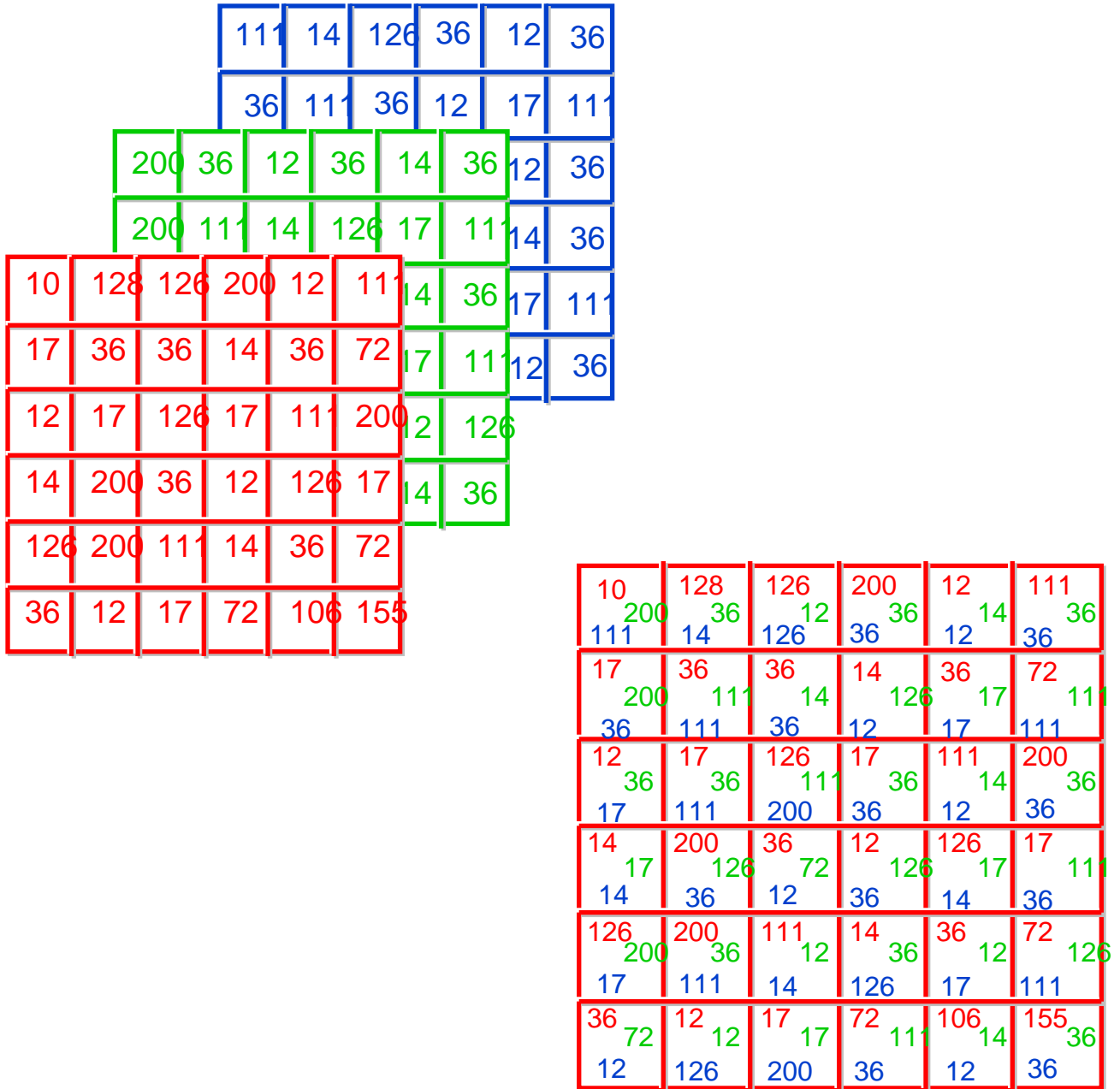


The representation of :

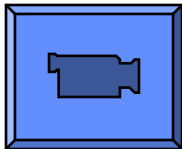
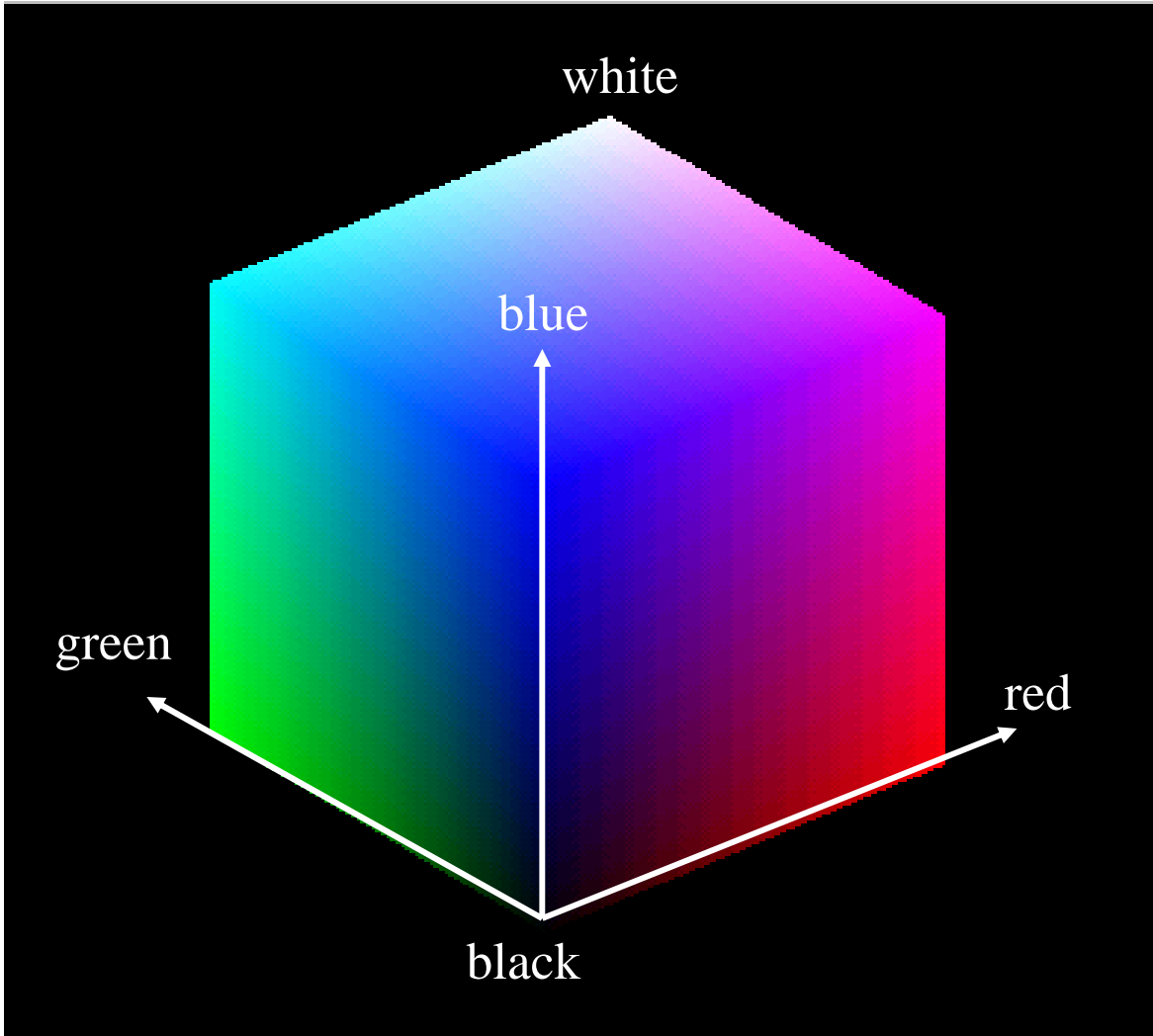


is then given by: **(a, b, c)**

Rgb Image



The RGB Cube

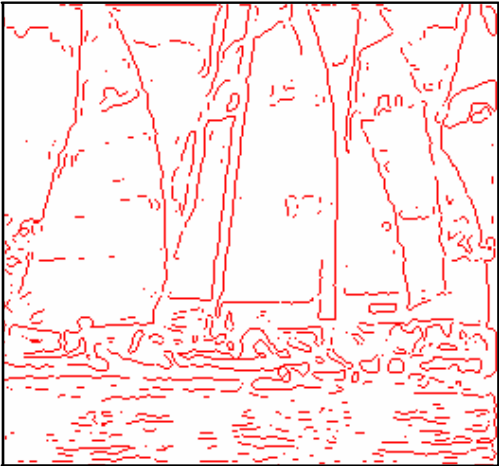


Color Edge Detection

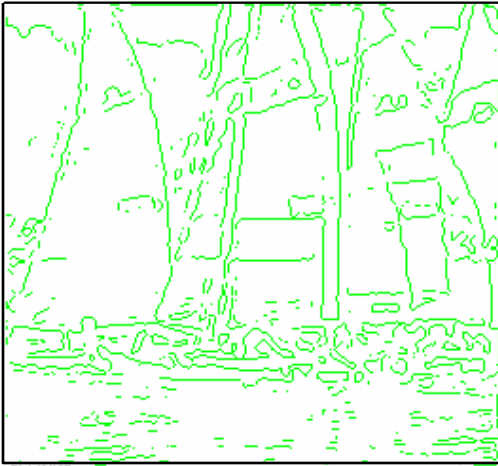


Original

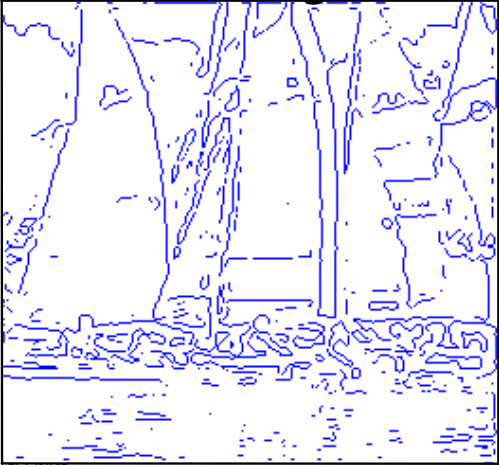
R - Edges



G - Edges



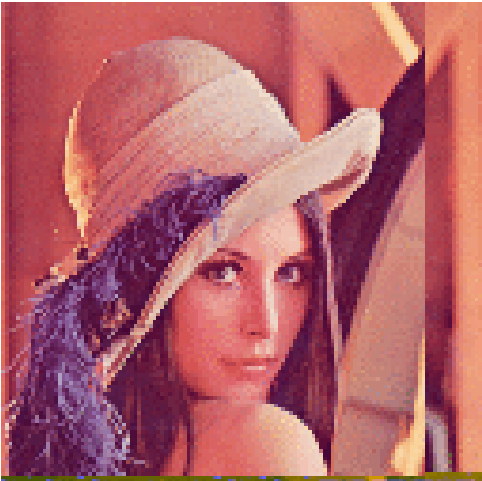
B - Edges



All - Edges



Color Edge Detection



Original

R - Edges



G - Edges



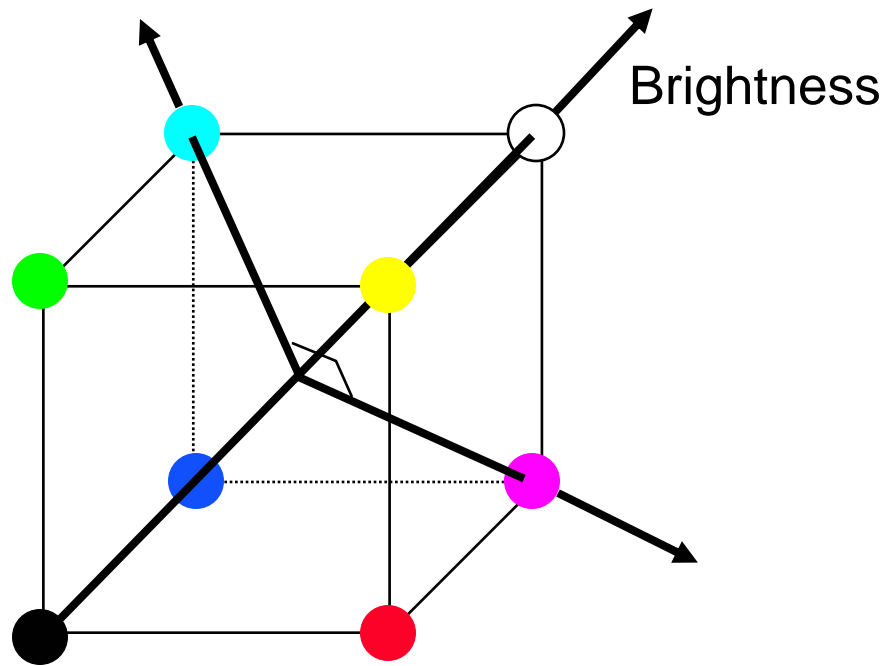
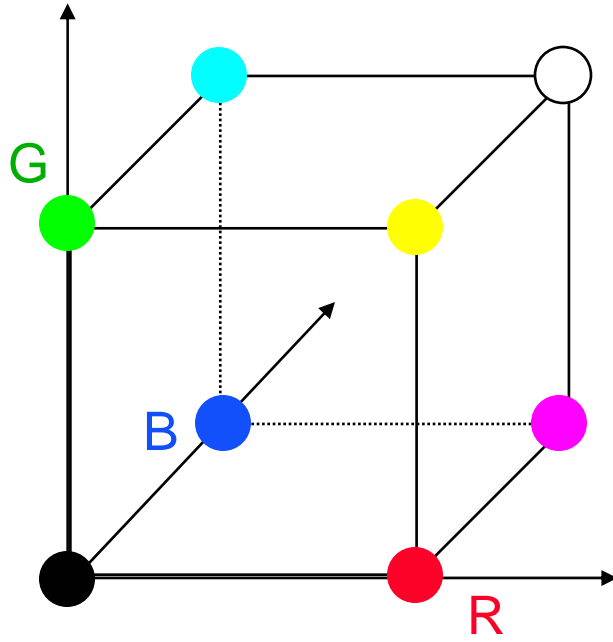
B - Edges



All - Edges



RGB Color Cube



Color Description

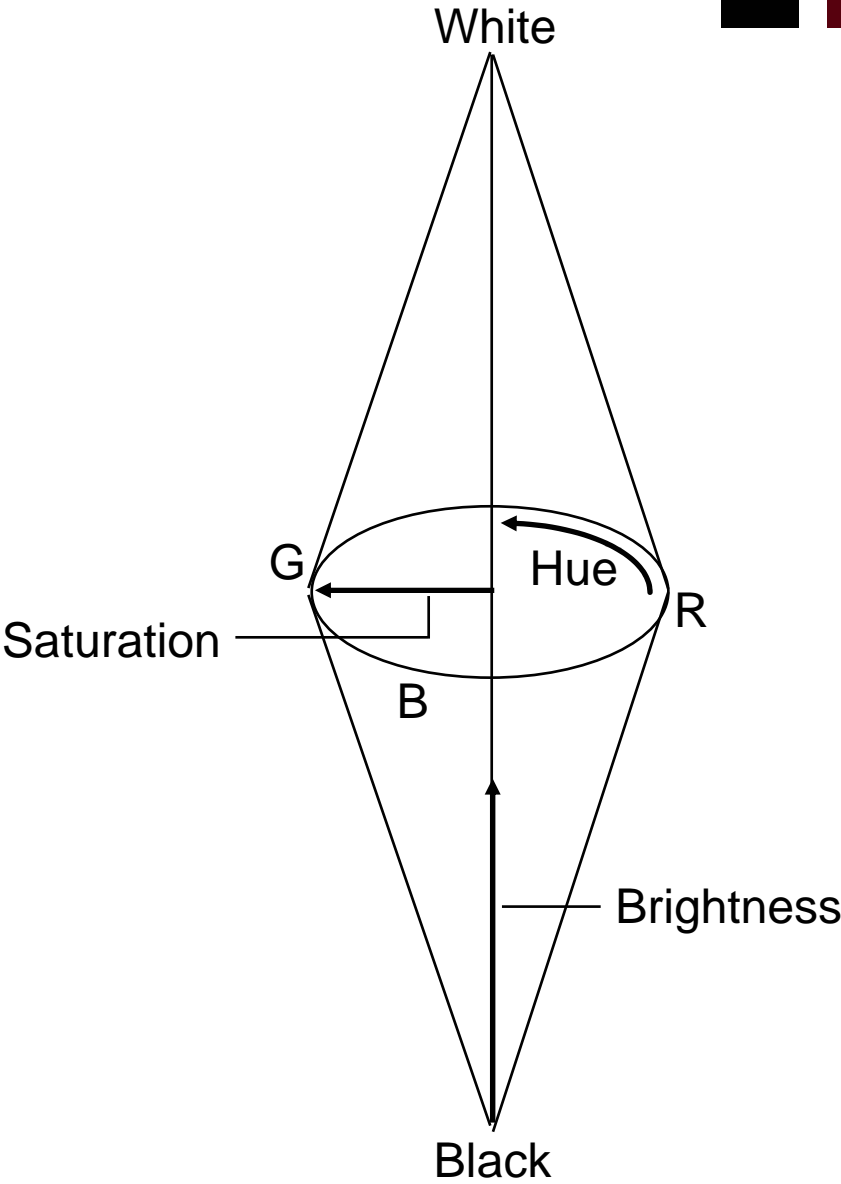
Hue (red, green, yellow, blue ...)



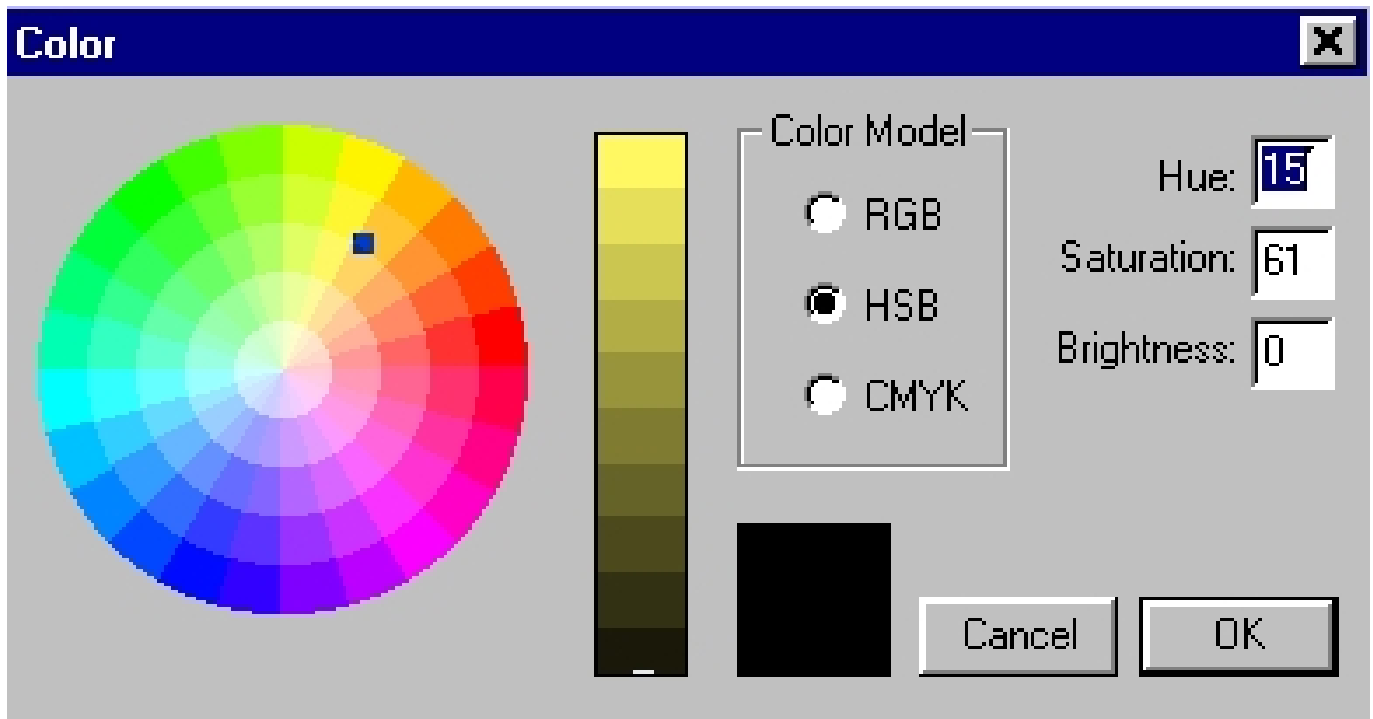
Saturation (pink, bright red,)



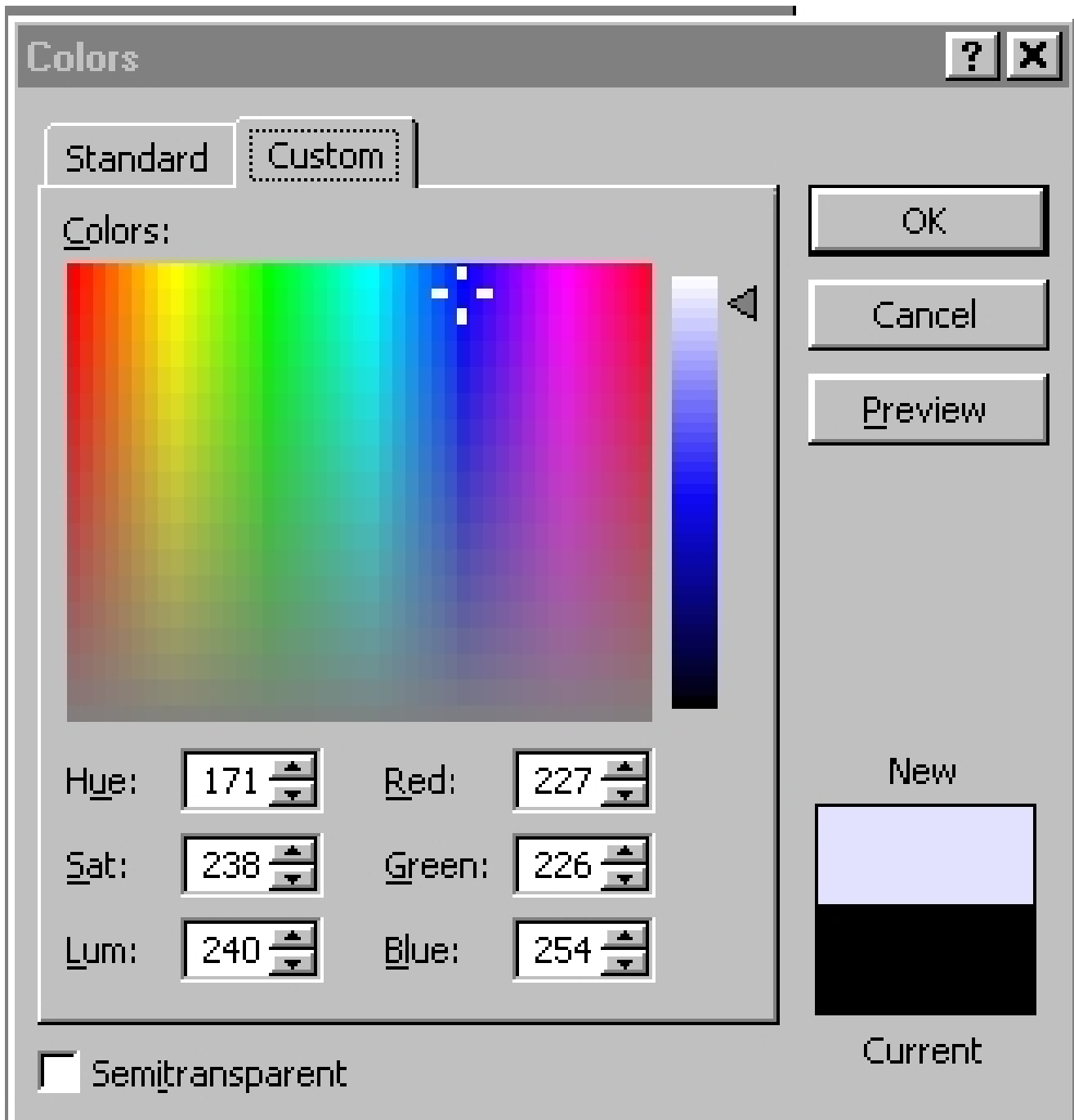
Lightness (black, grey, white)



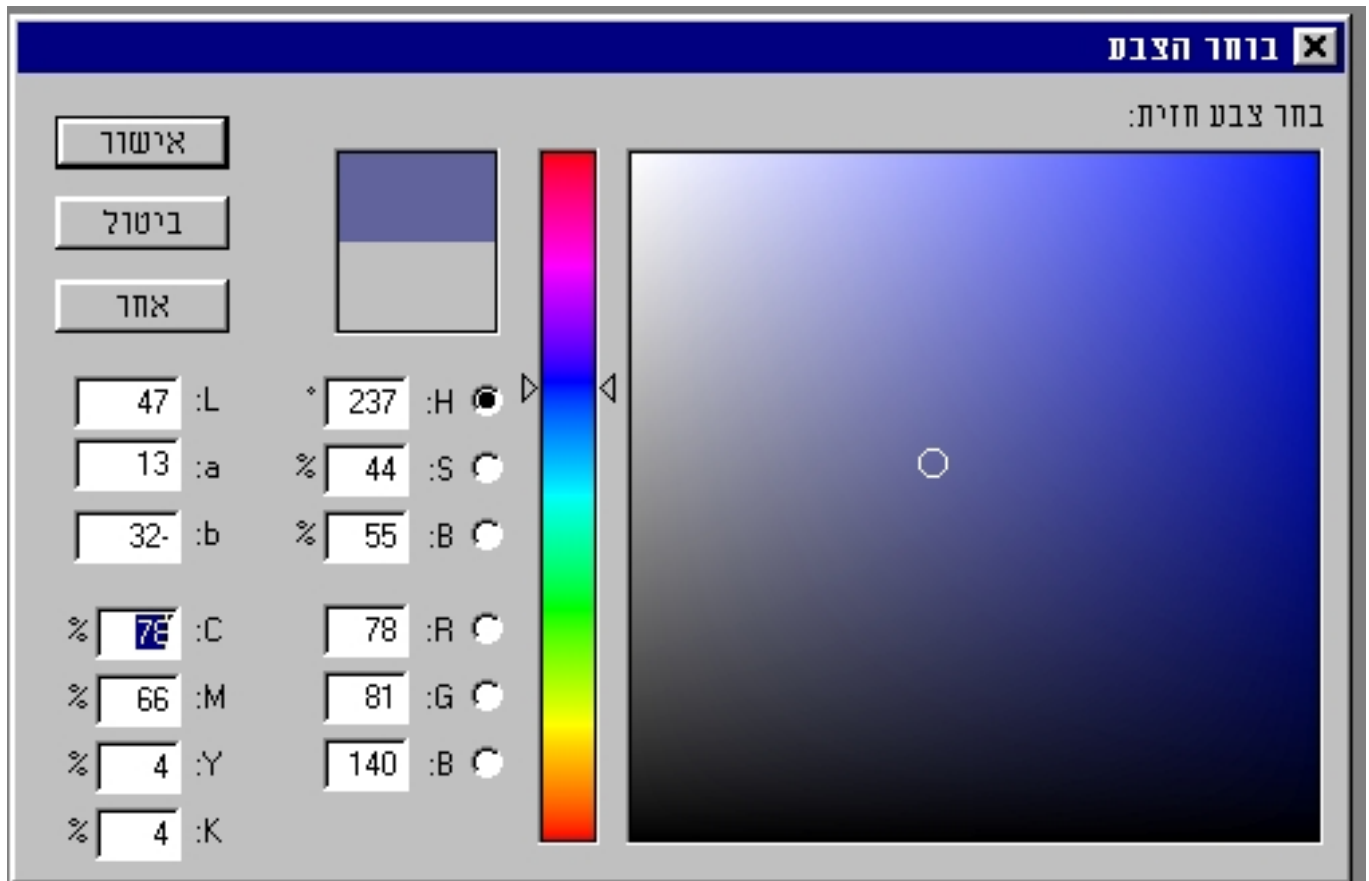
MayuraDraw



PowerPoint



PhotoShop



YIQ - Color Space

NTSC = National Television Systems Committee

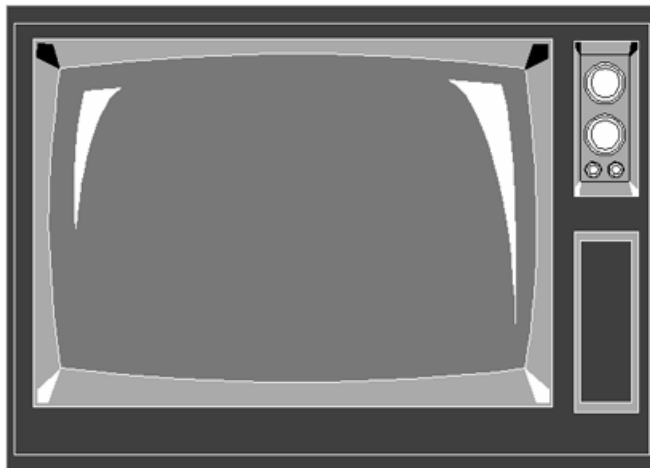
Y = luminance

I = red-green

Q = blue-yellow

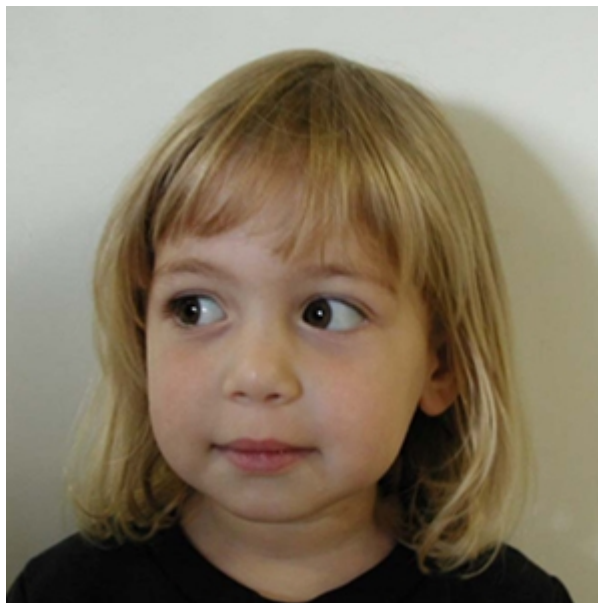
$$\begin{bmatrix} Y \\ I \\ Q \end{bmatrix} = \begin{bmatrix} 0.177 & 0.813 & 0.011 \\ 0.540 & -0.263 & -0.174 \\ 0.246 & -0.675 & 0.404 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

R G B are the CIE-RGB



RGB To Monochrome

RGB



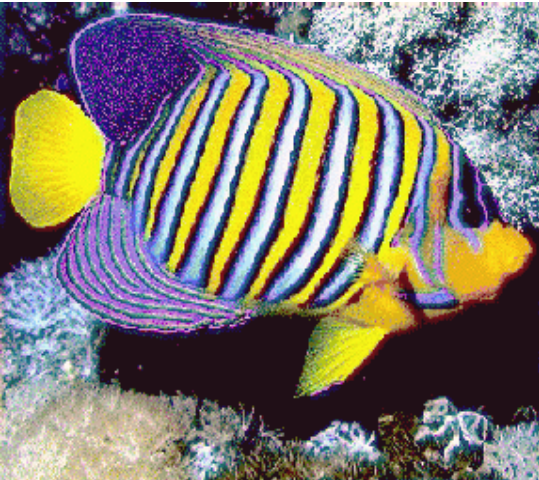
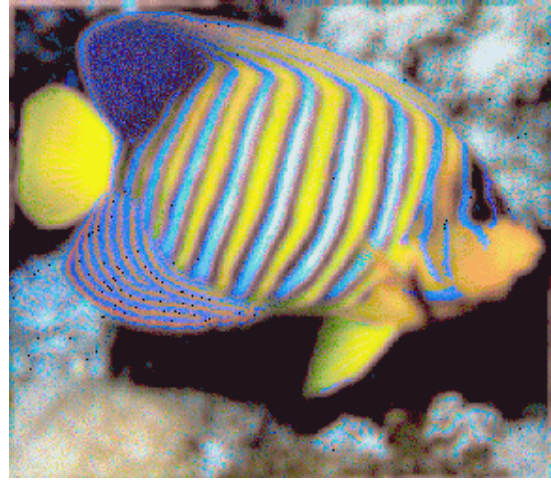
Y



Original



Y - Blur



I - Blur



Q - Blur

Subtractive Color System - CMYK

Printer Dyes:

Cyan = removes red

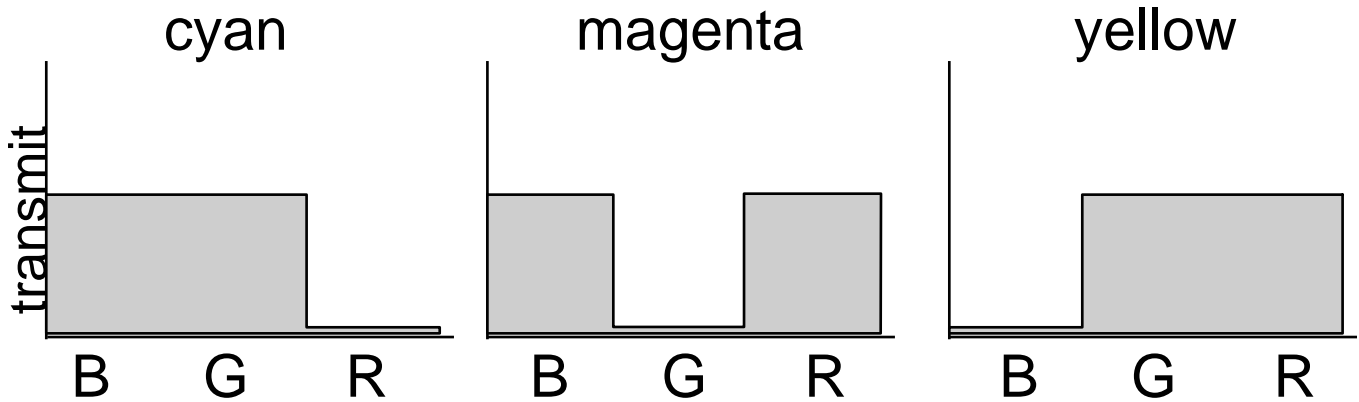
Magenta = removes green

Yellow = removes blue

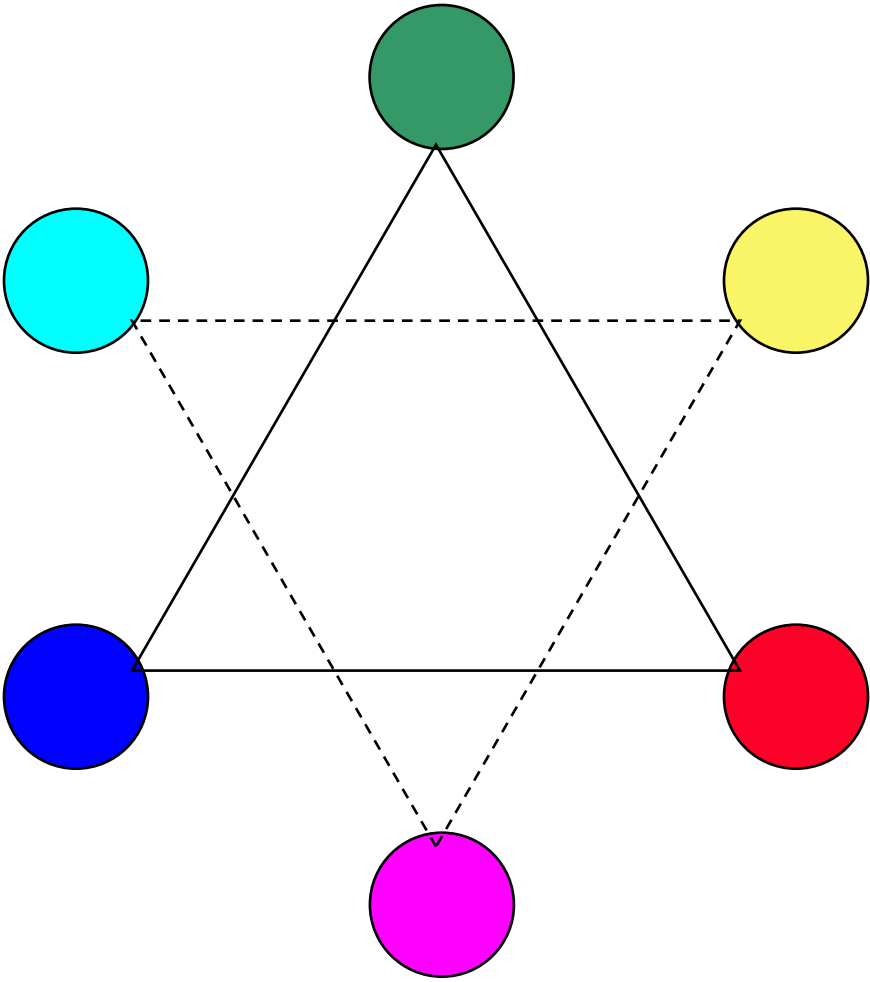
black = removes all



Ideal block dyes:



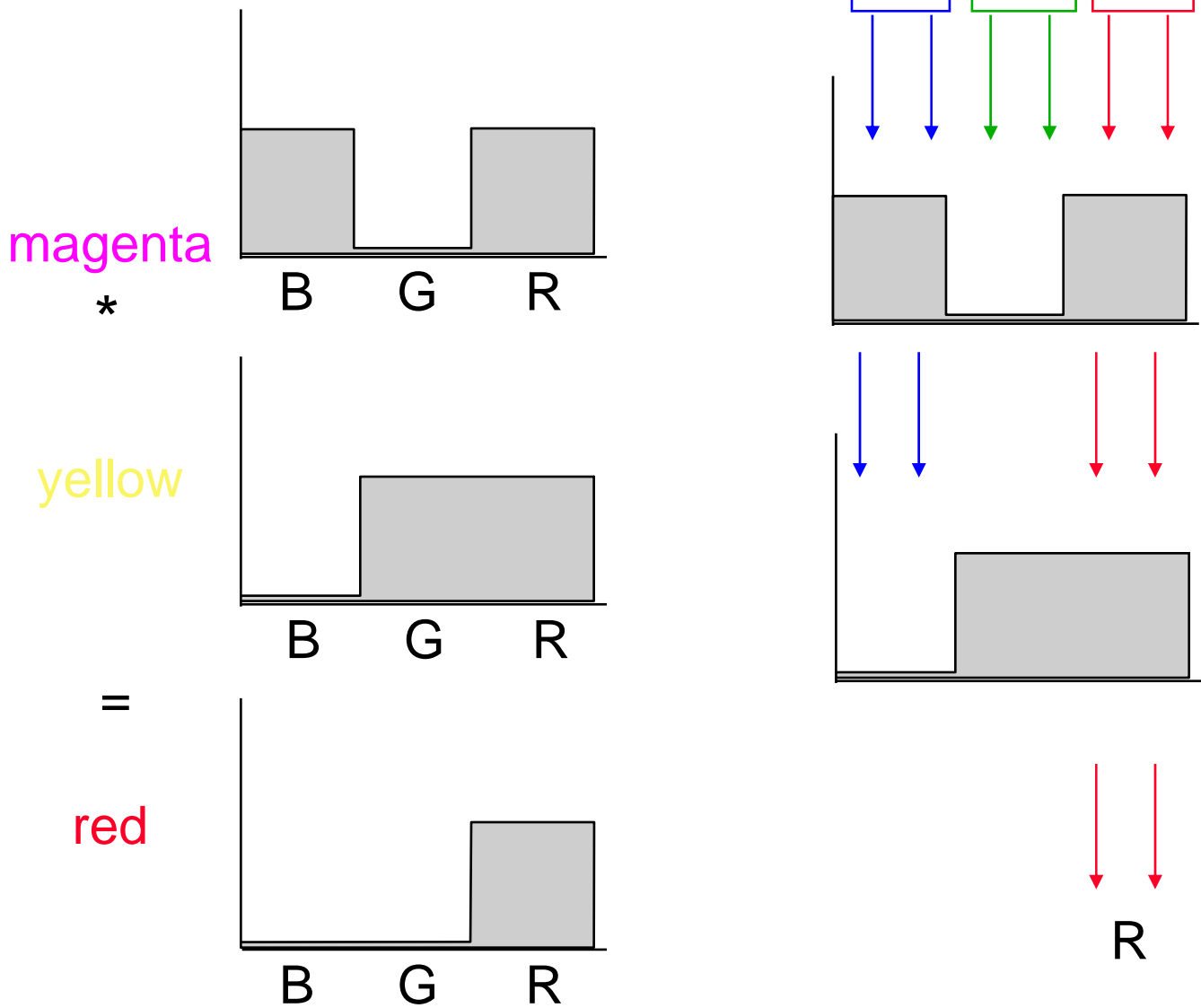
Opponent Color Wheel



—— Additive primaries
----- Subtractive Primaries

Multiplicative (Subtractive) Color System

red = magenta + yellow

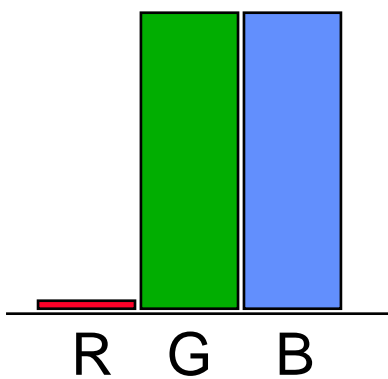
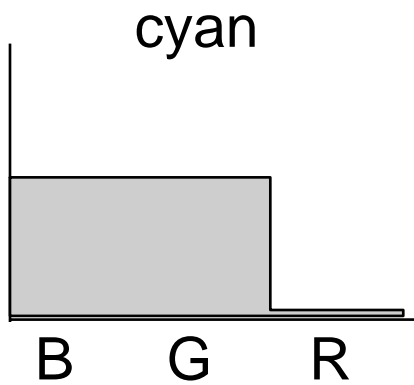


red = magenta + yellow
green = cyan + yellow
blue = magenta + cyan

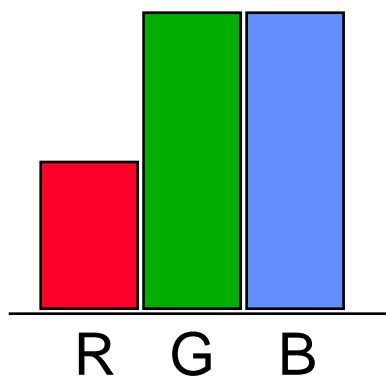
Cyan - controls amount of red in print:

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

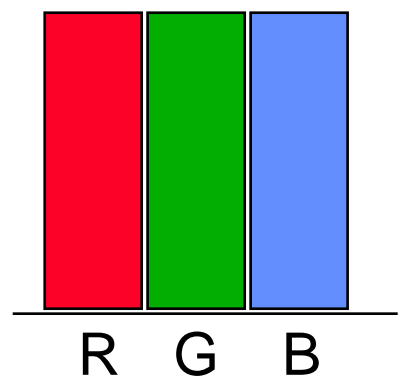
high C = low R (high G and B)



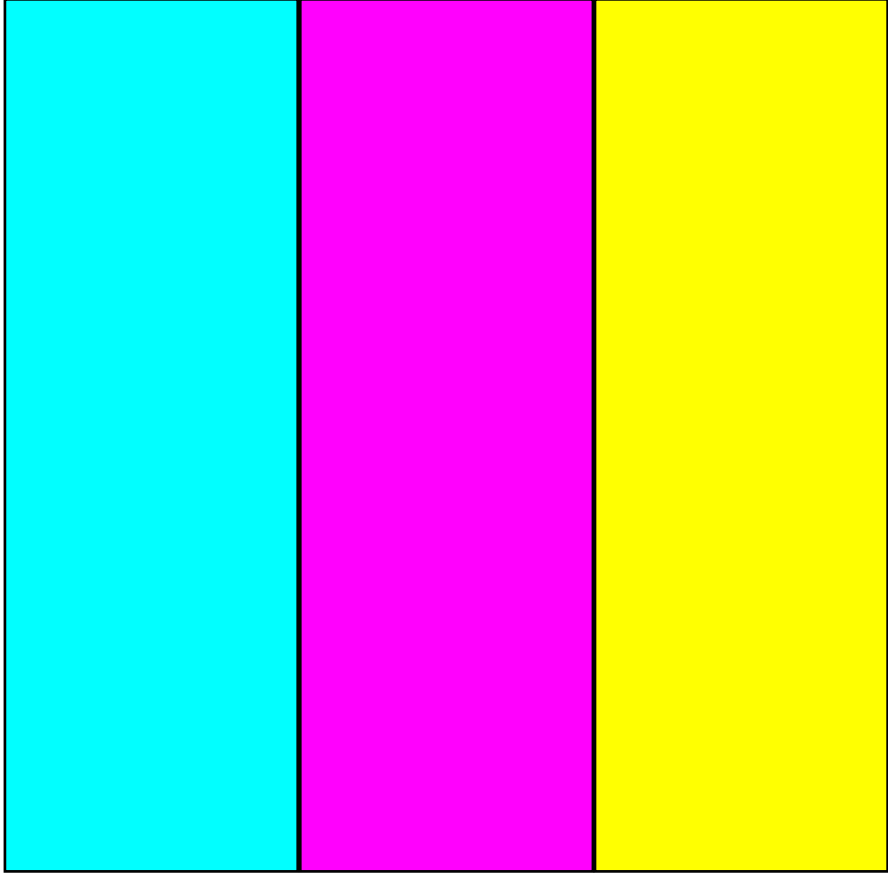
High density
cyan

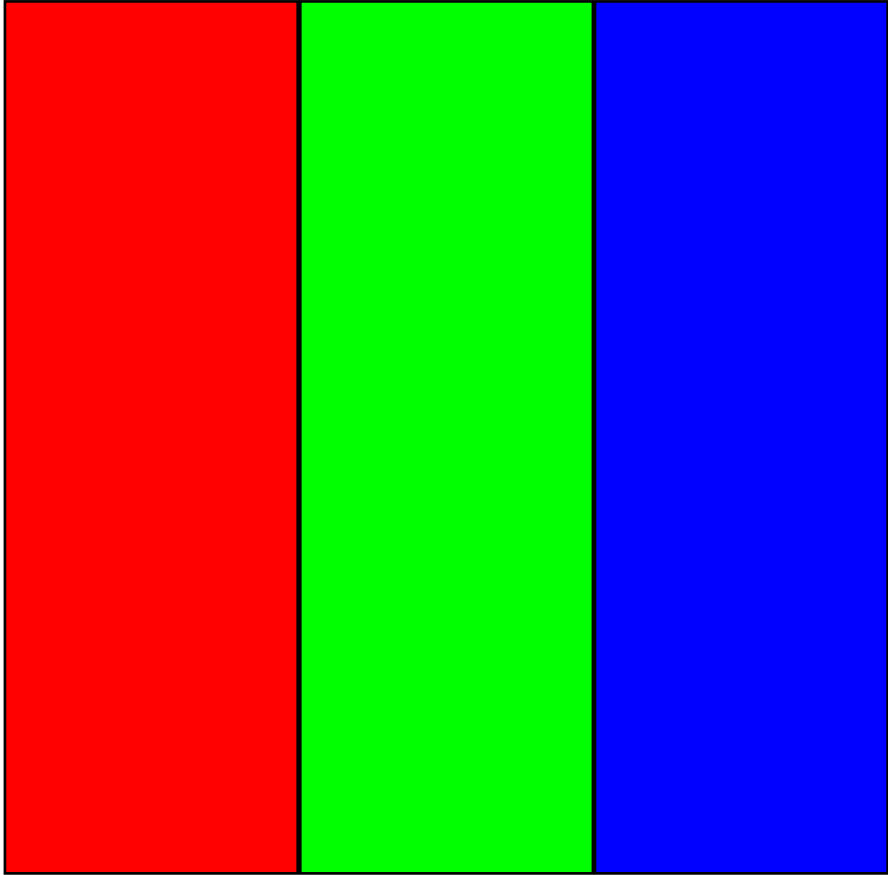


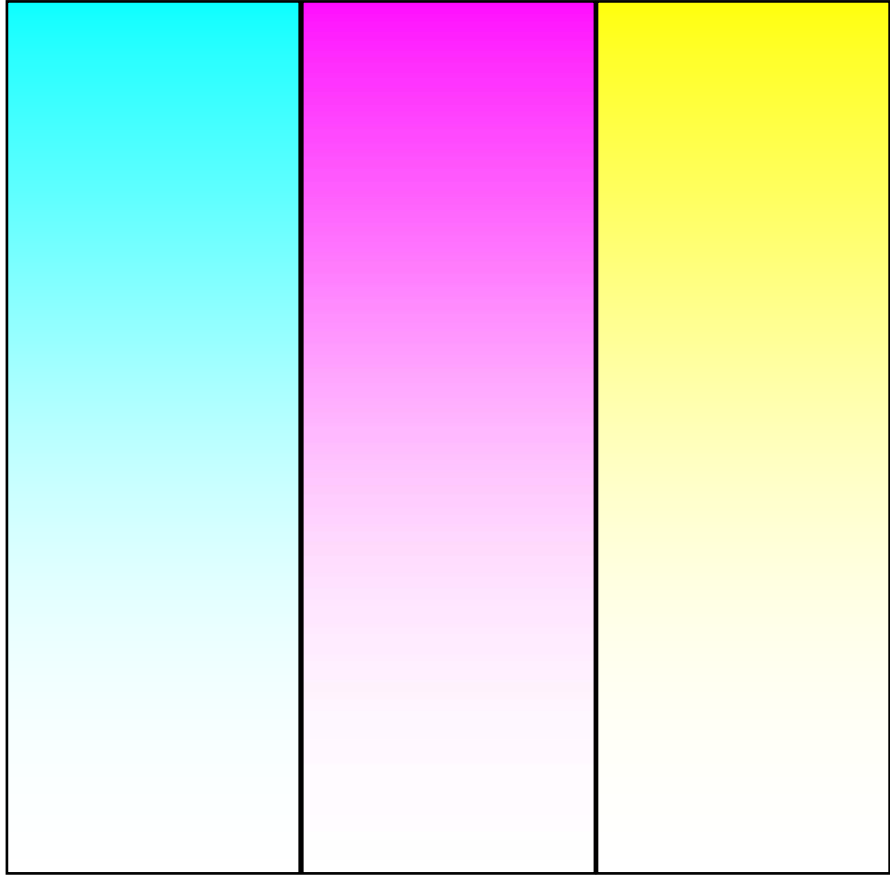
Medium density
cyan



Low density
cyan





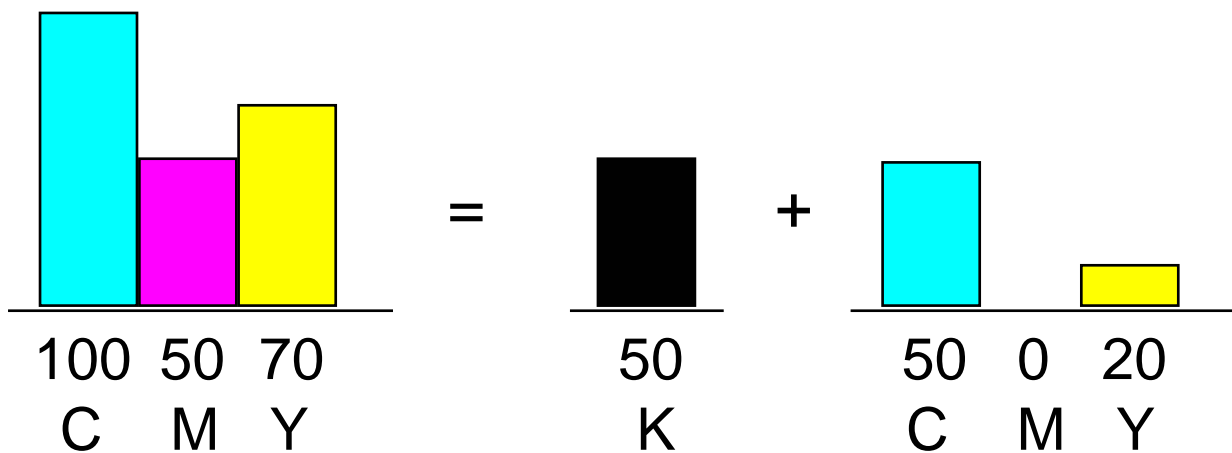


CMY + Black

$$C + M + Y = K \text{ (black)}$$

- 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.

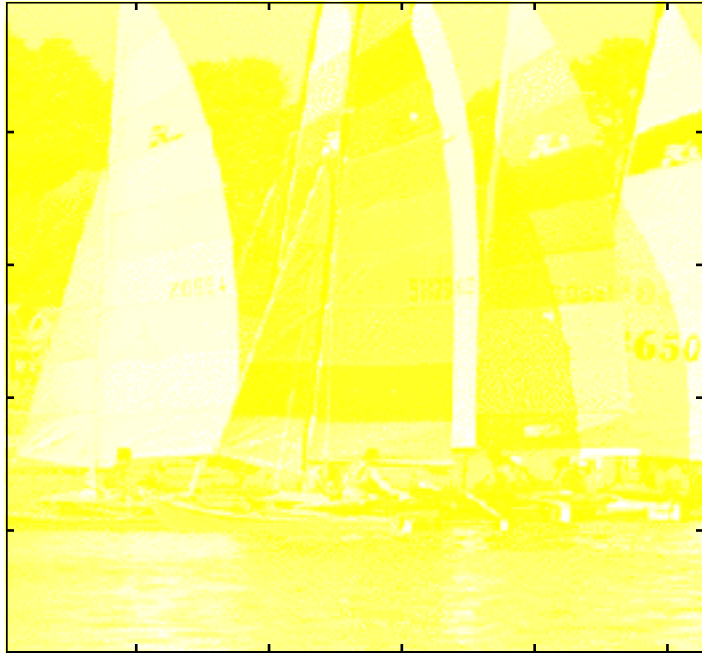
Undercolor removal -
(gray component replacement)





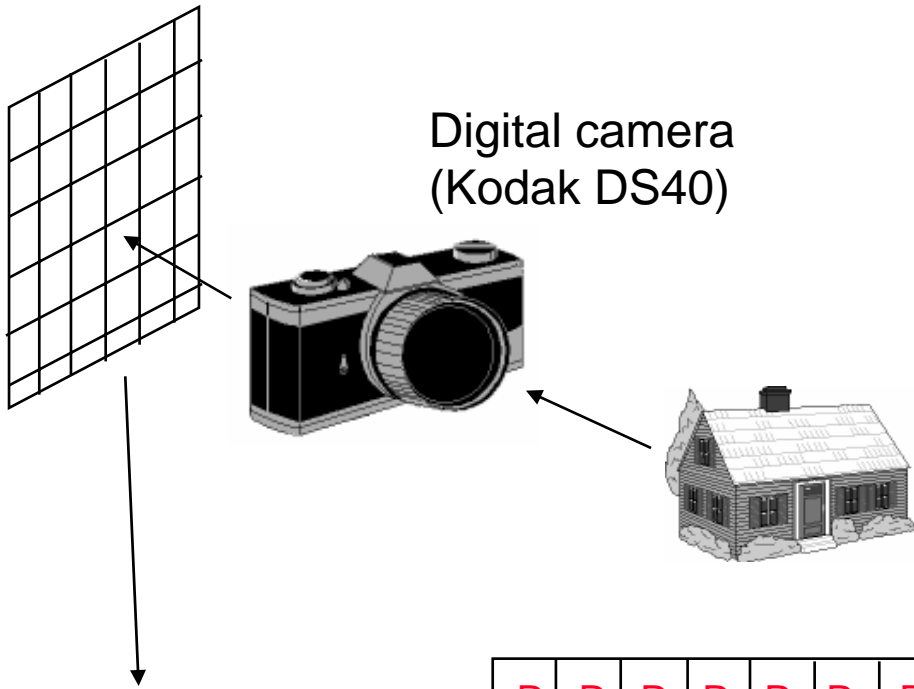








Demosaicing



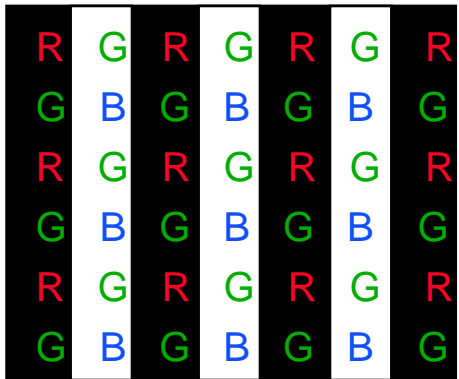
R	G	R	G	R	G	R	G
G	B	G	B	G	B	G	B
R	G	R	G	R	G	R	G
G	B	G	B	G	B	G	B
R	G	R	G	R	G	R	G
G	B	G	B	G	B	G	B
R	G	R	G	R	G	R	G
G	B	G	B	G	B	G	B

R	R	R	R	R	R	R	R				
R	R	R	R	R	R	R	R				
R	R	G	G	G	G	G	G	G			
R	R	G	G	G	G	G	G	G			
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B
R	R	G	G	B	B	B	B	B	B	B	B



demosaic

Demosaic Aliasing



R		R		R		R
R		R		R		R
R		R		R		R

R image



interpolate

R	R	R	R	R	R	R
R	R	R	R	R	R	R
R	R	R	R	R	R	R
R	R	R	R	R	R	R
R	R	R	R	R	R	R
R	R	R	R	R	R	R

	G		G		G	
G		G		G		G
	G		G		G	
G		G		G		G
	G		G		G	
G		G		G		G

G image



interpolate

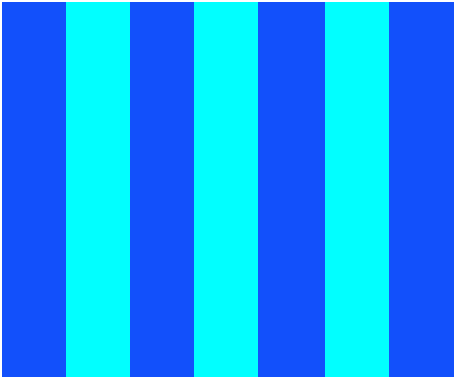
G	G	G	G	G	G	G
G	G	G	G	G	G	G
G	G	G	G	G	G	G
G	G	G	G	G	G	G
G	G	G	G	G	G	G
G	G	G	G	G	G	G

	B		B		B	
	B		B		B	
	B		B		B	

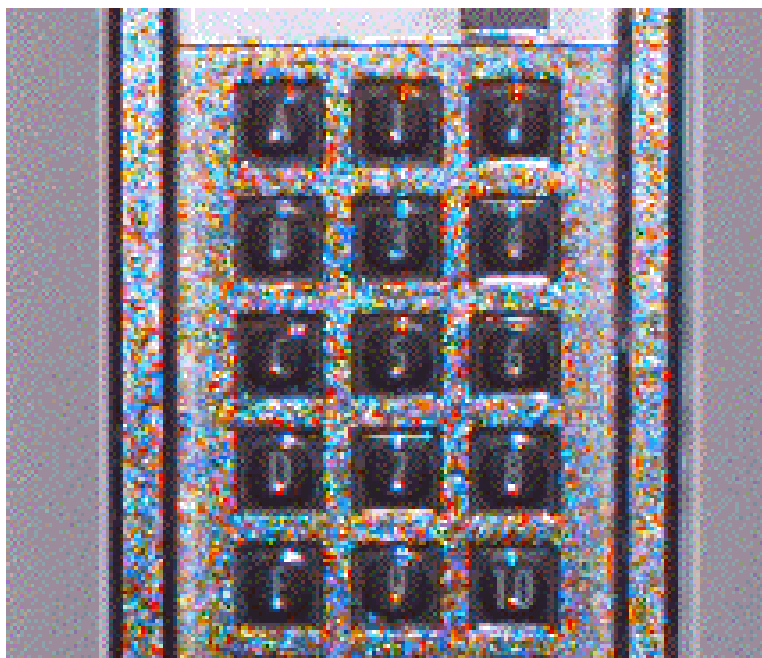
B image



B	B	B	B	B	B	B
B	B	B	B	B	B	B
B	B	B	B	B	B	B
B	B	B	B	B	B	B
B	B	B	B	B	B	B
B	B	B	B	B	B	B



Demosaicing - Example (Kodak)



Demosaicing - Various Approaches

Regularization

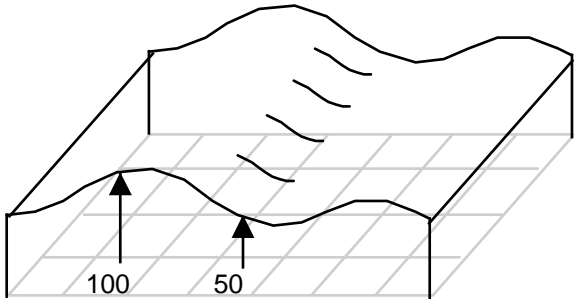
Minimize over a functional with a data fit term and an inter-channel color correlation term.
(Gamer & Keren)

Minimal Surface

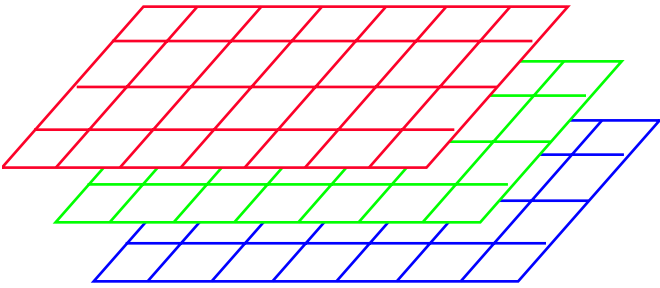
Minimize over a functional with a data fit term and a 5D surface area term. (Beltrami Flow)
(Kimmel)



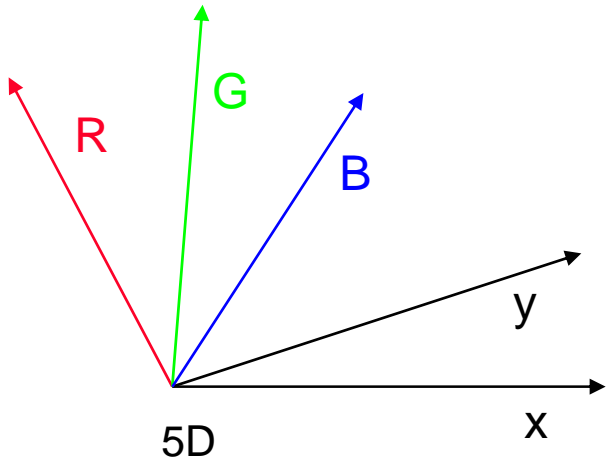
Grayscale Image



3D



RGB Image



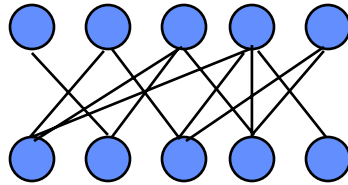
5D

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)

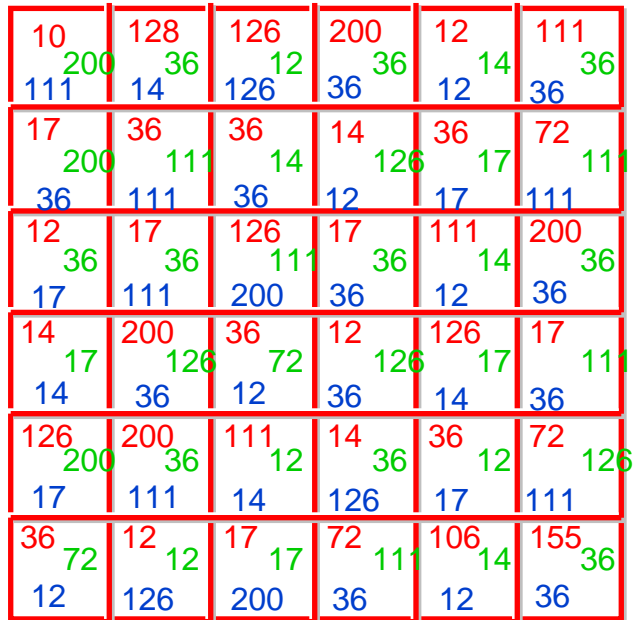
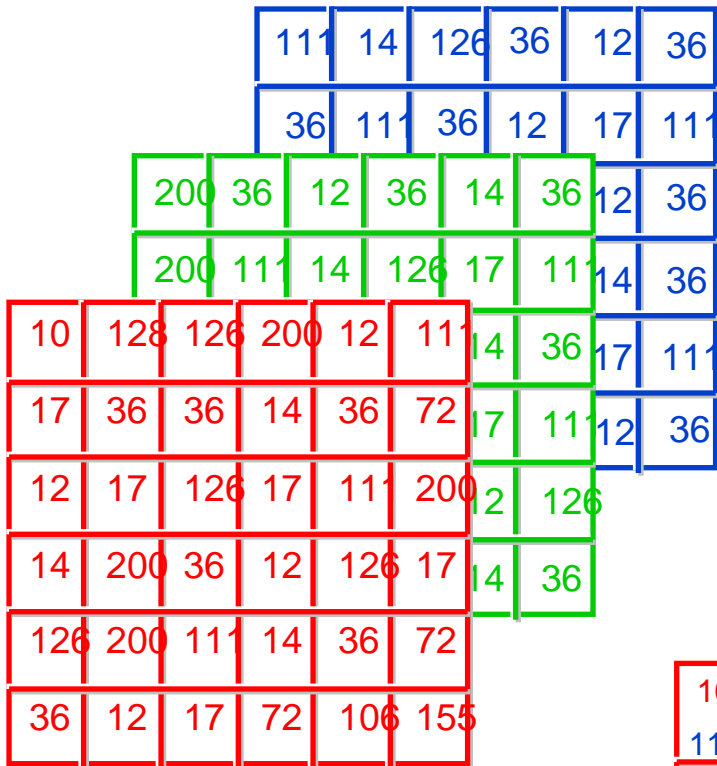


Quadratic - Learned on Image



Quadratic - Learned on Class

Color Quantization



Indexed Image

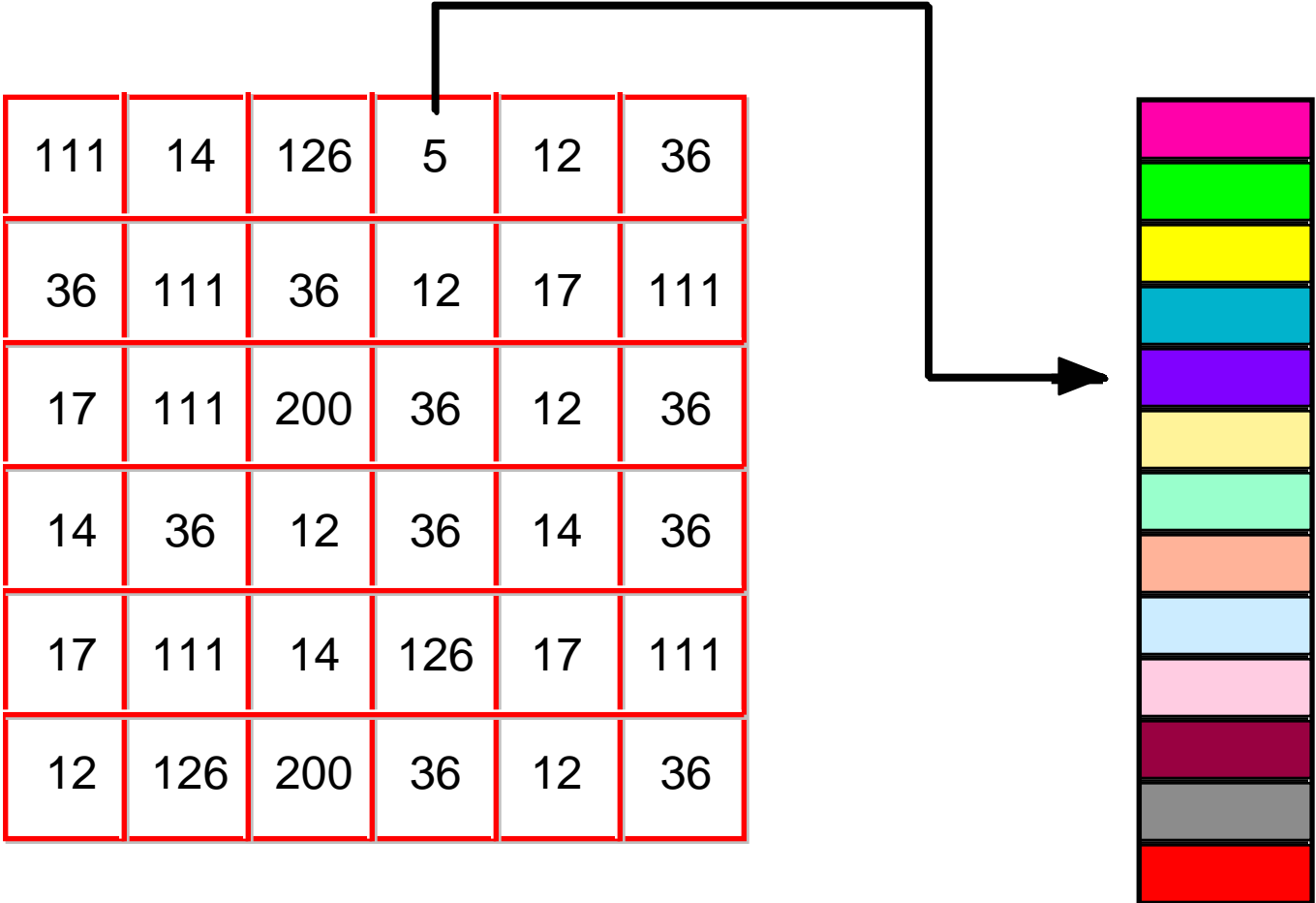
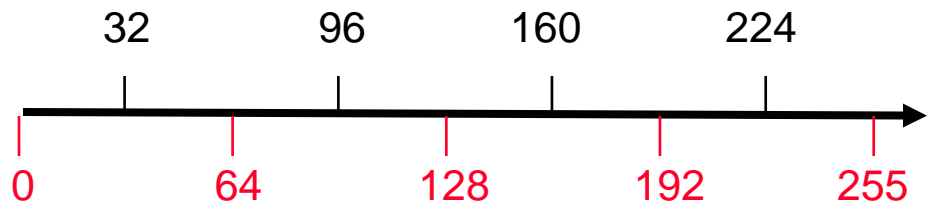


Image Independent Quantization

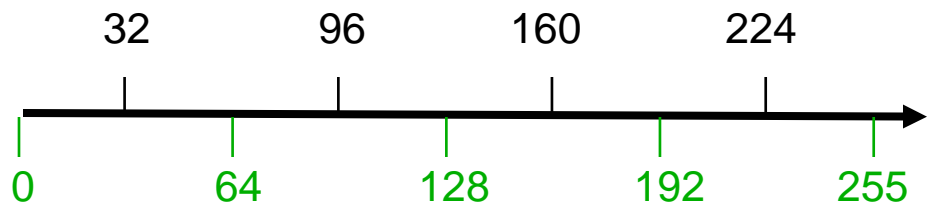
quantized
gray value

original
gray value



quantized
gray value

original
gray value



quantized
gray value

original
gray value

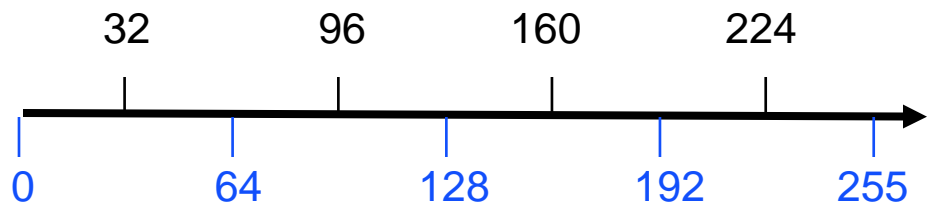


Image Independent Quantization

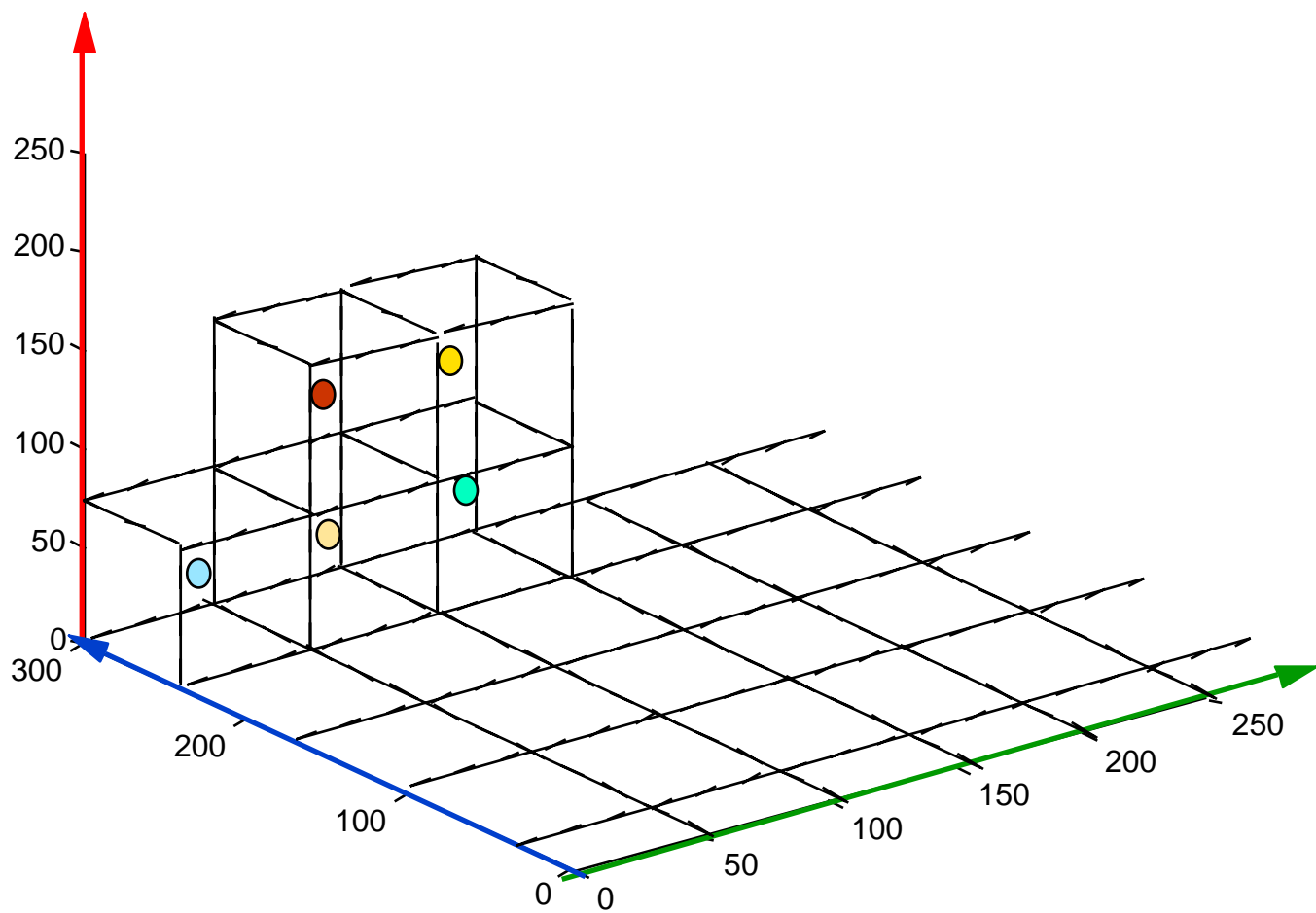


Image Independent Quantization

Original



125 Colors



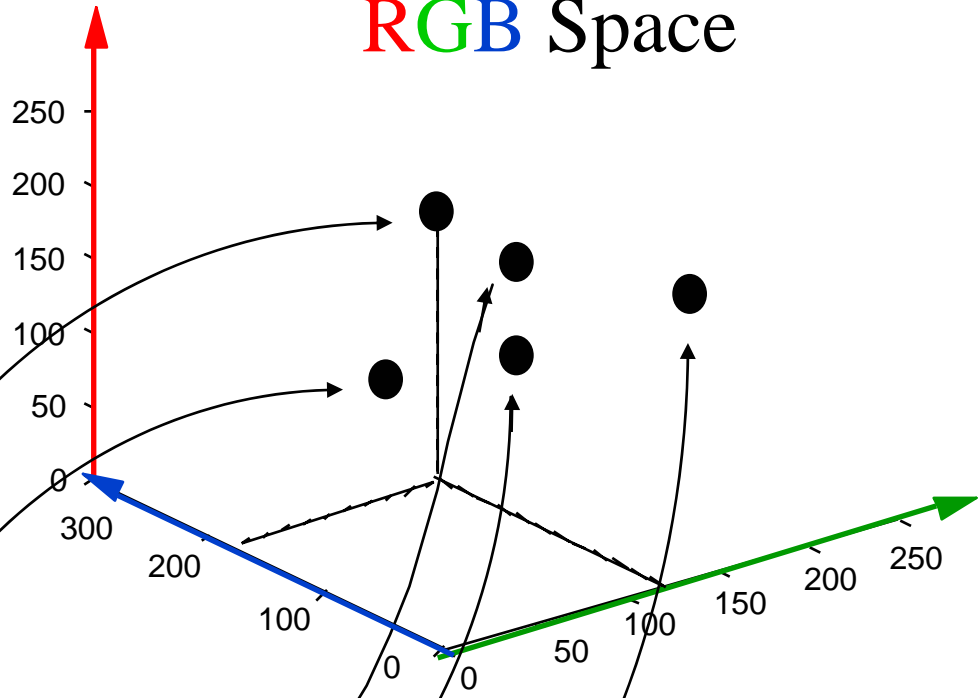
64 Colors



27 Colors



RGB Space



10	180	126	200
200	117	12	36
111	172	126	36
17	36	36	14
200	111	14	126
36	111	36	12
12	17	126	17
36	36	111	36
17	111	200	36
14	200	36	12
17	126	72	126
14	36	12	36

RGB Image

RGB Space

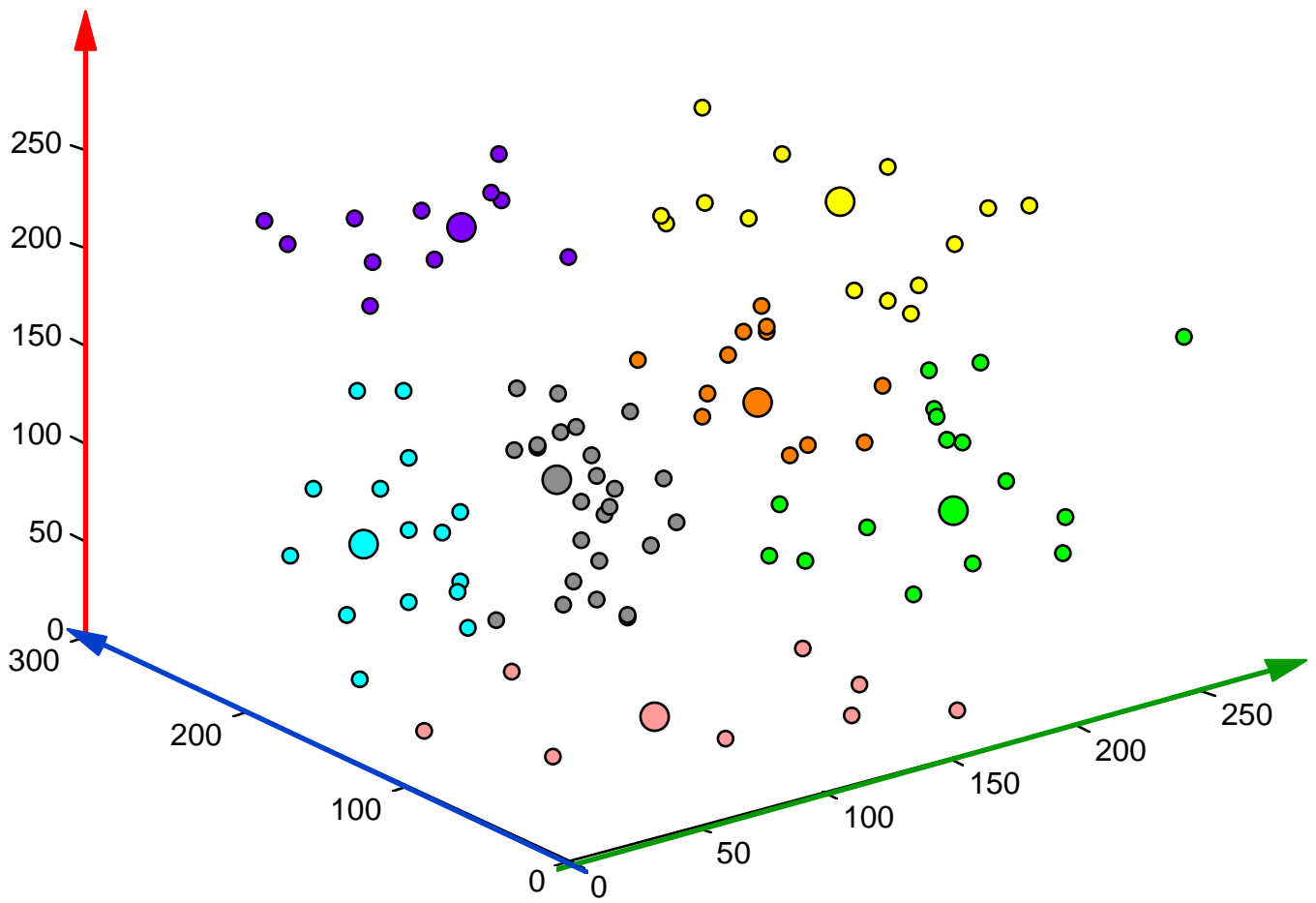
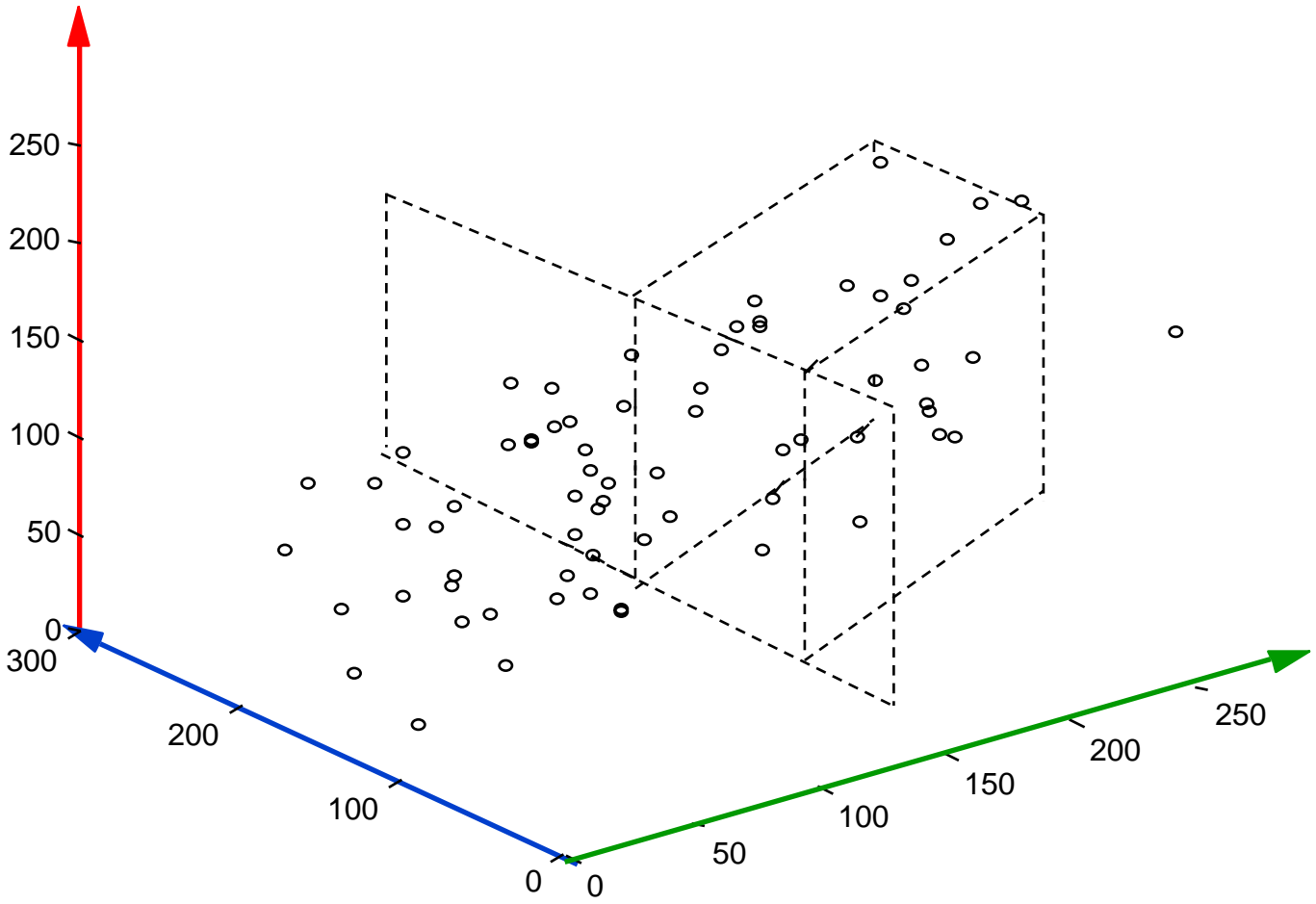


Image Dependent Quantization



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

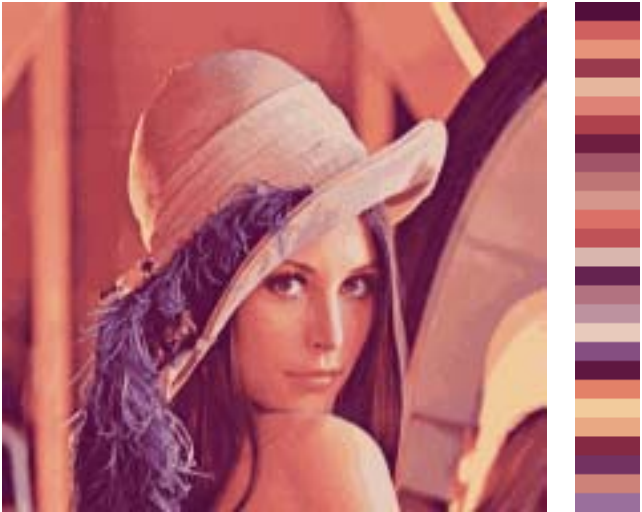
Original

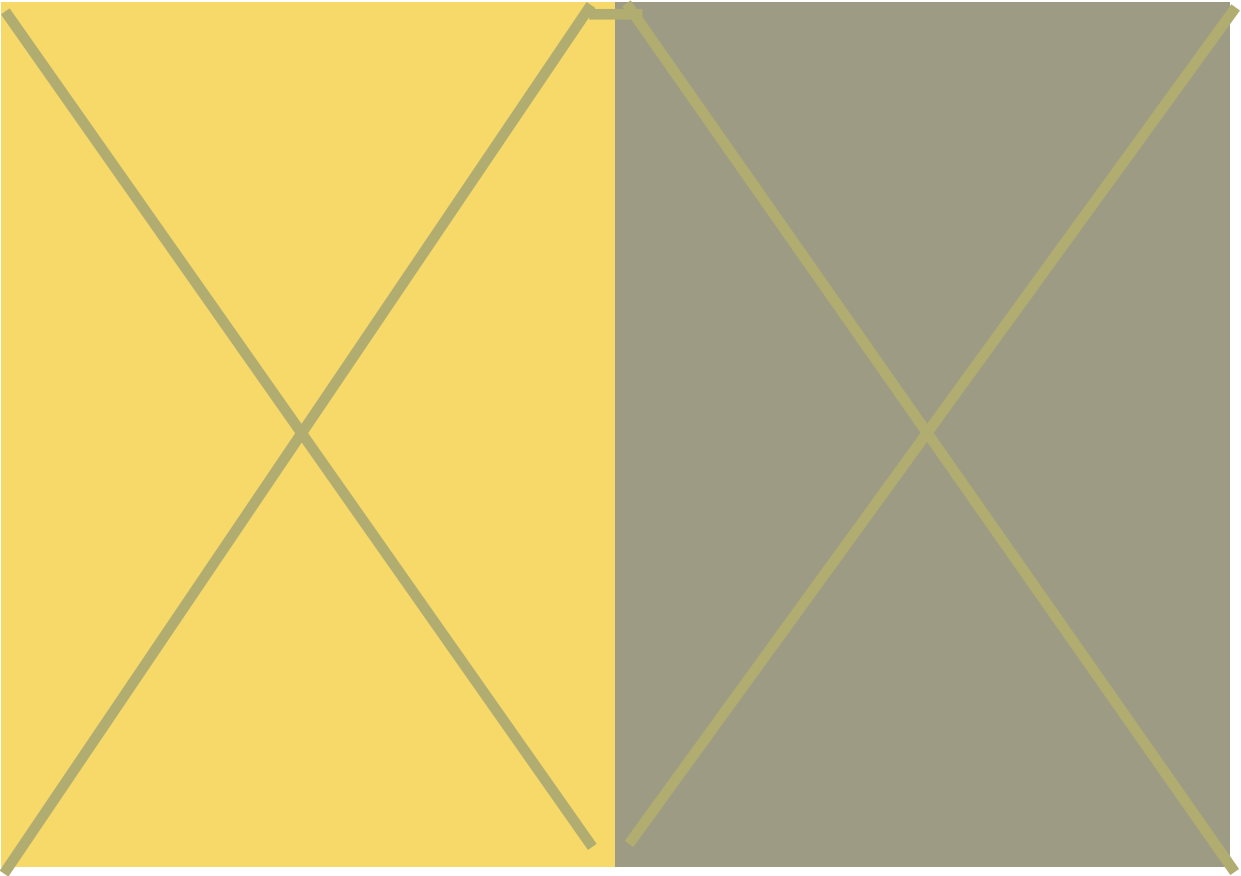


Independent



Dependent





Albers (1975)