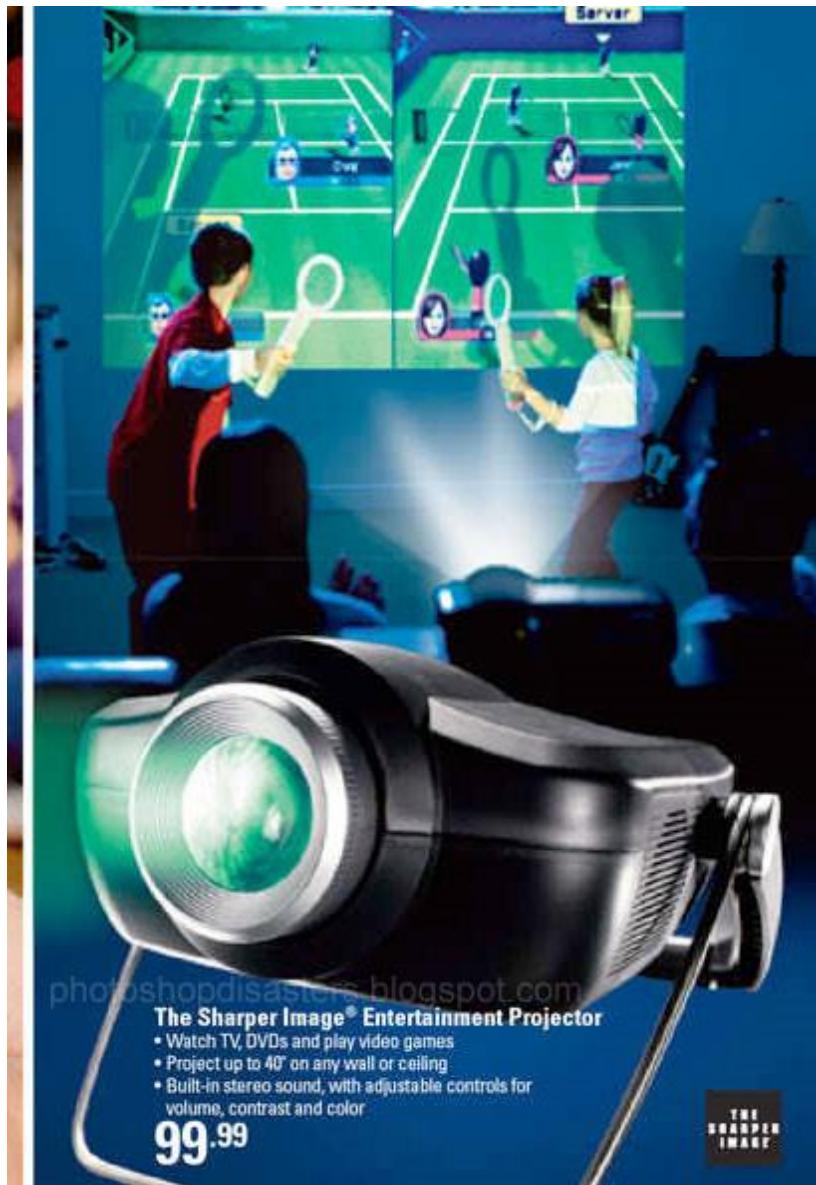






# What is wrong with this picture?

---



# Recap from Lecture 2

---

Pinhole camera model

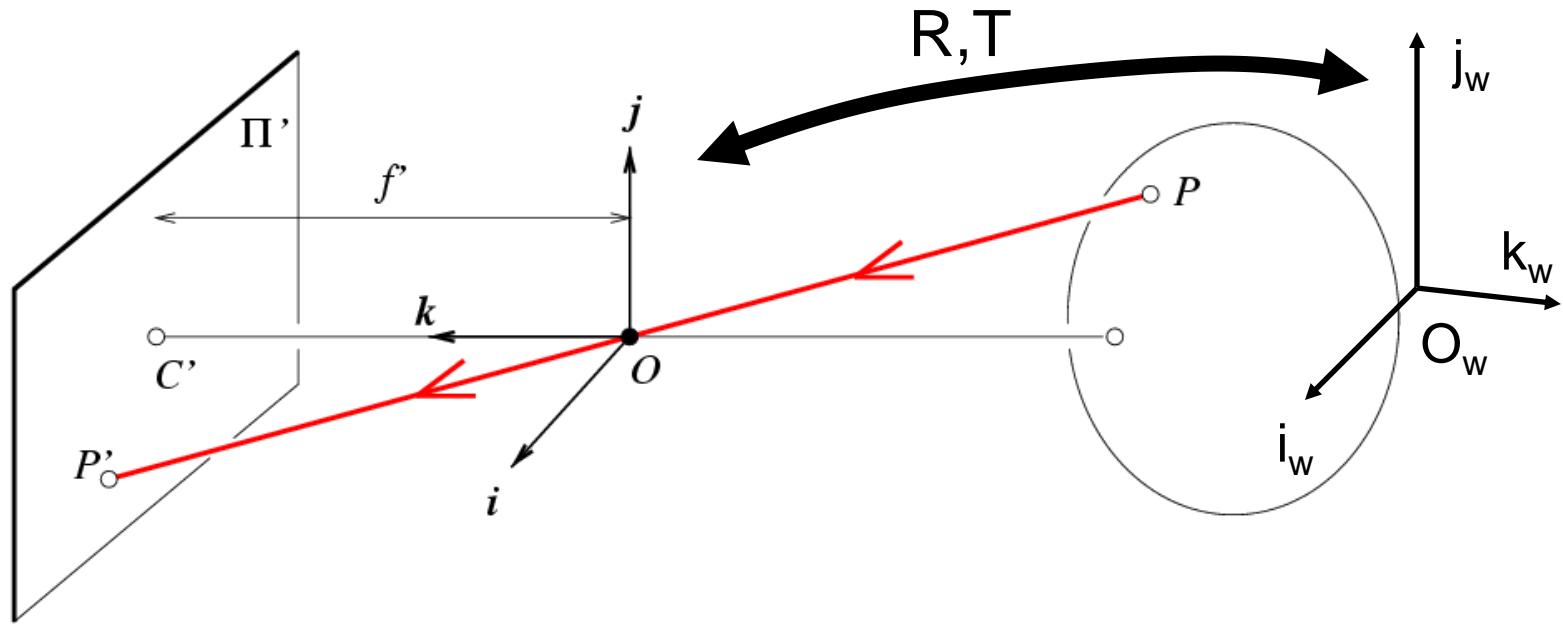
Perspective projections

Focal length and field of view

Remember to use your textbook:

**Chapter 2 of Szeliski**

# Recap - Projection matrix



$$\mathbf{x} = \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}$$

$\mathbf{x}$ : Image Coordinates:  $(u, v, 1)$   
 $\mathbf{K}$ : Intrinsic Matrix (3x3)  
 $\mathbf{R}$ : Rotation (3x3)  
 $\mathbf{t}$ : Translation (3x1)  
 $\mathbf{X}$ : World Coordinates:  $(X, Y, Z, 1)$

# Recap - Projection matrix

---

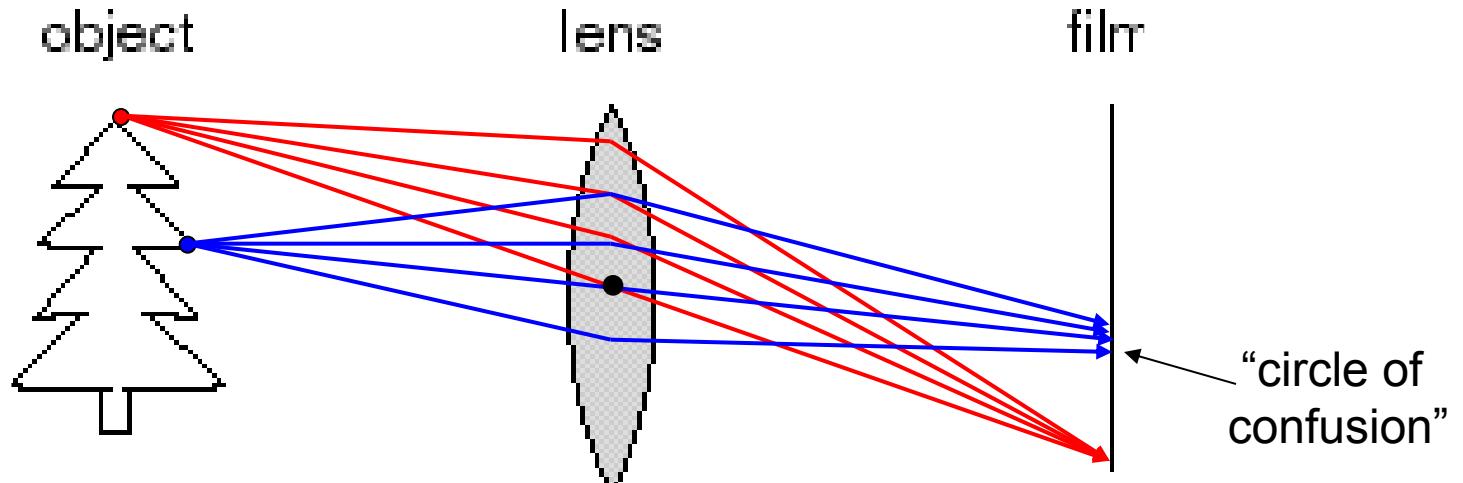


$$\mathbf{x} = \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}$$

$\mathbf{x}$ : Image Coordinates:  $(u, v, 1)$   
 $\mathbf{K}$ : Intrinsic Matrix (3x3)  
 $\mathbf{R}$ : Rotation (3x3)  
 $\mathbf{t}$ : Translation (3x1)  
 $\mathbf{X}$ : World Coordinates:  $(X, Y, Z, 1)$

# Adding a lens

---

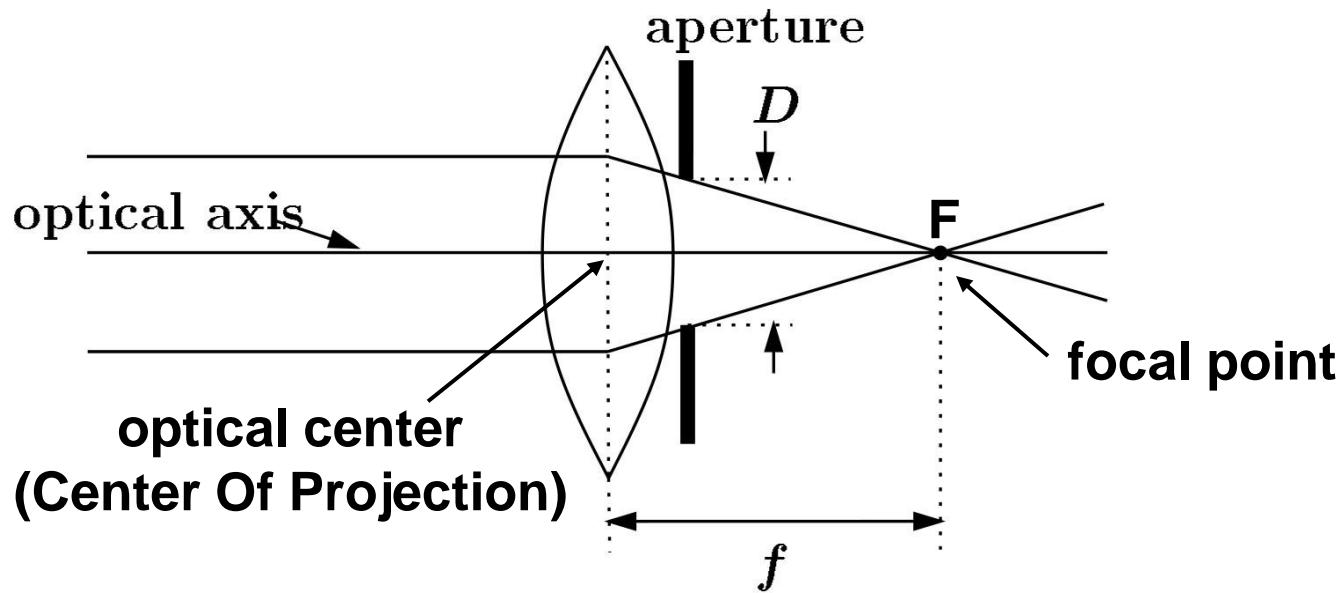


A lens focuses light onto the film

- There is a specific distance at which objects are “in focus”
  - other points project to a “circle of confusion” in the image
- Changing the shape of the lens changes this distance

# Focal length, aperture, depth of field

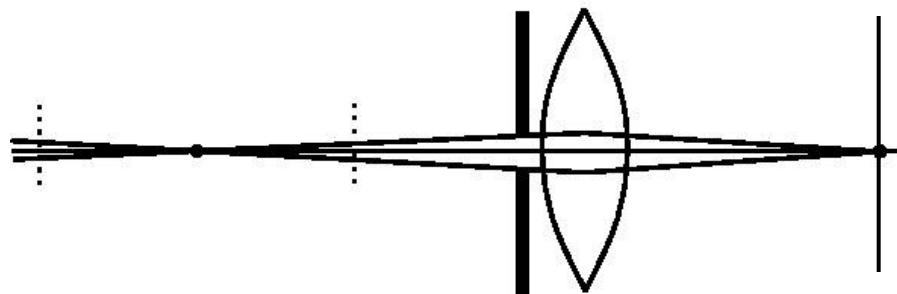
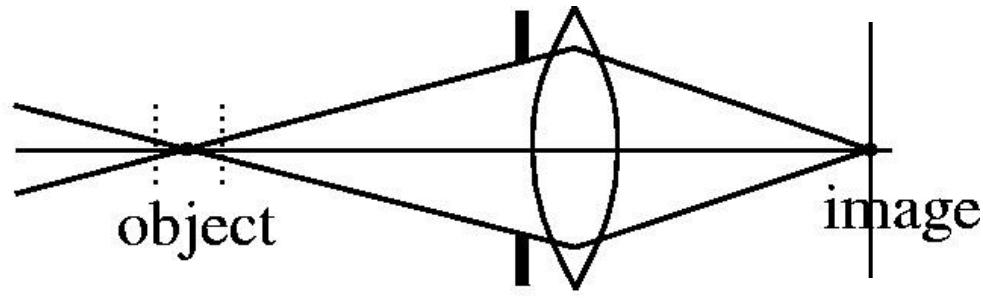
---



A lens focuses parallel rays onto a single focal point

- focal point at a distance  $f$  beyond the plane of the lens
- Aperture of diameter  $D$  restricts the range of rays

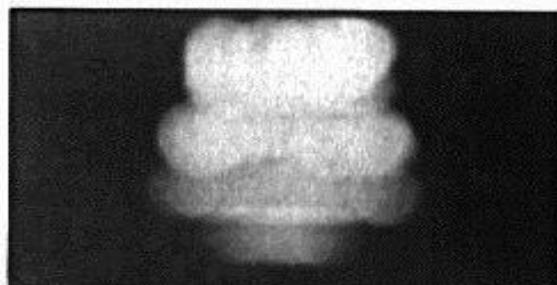
# Depth of field



Changing the aperture size or focal length affects depth of field

# Shrinking the aperture

---



2 mm



1 mm



0.6mm



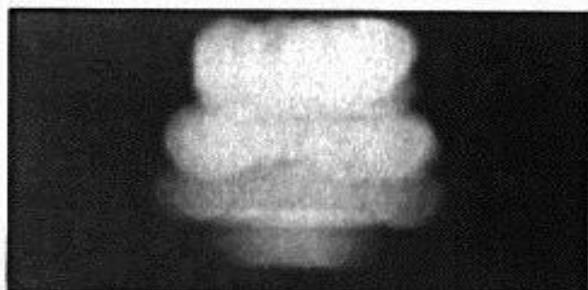
0.35 mm

Why not make the aperture as small as possible?

- Less light gets through
- Diffraction effects

# Shrinking the aperture

---



2 mm



1 mm



0.6mm



0.35 mm



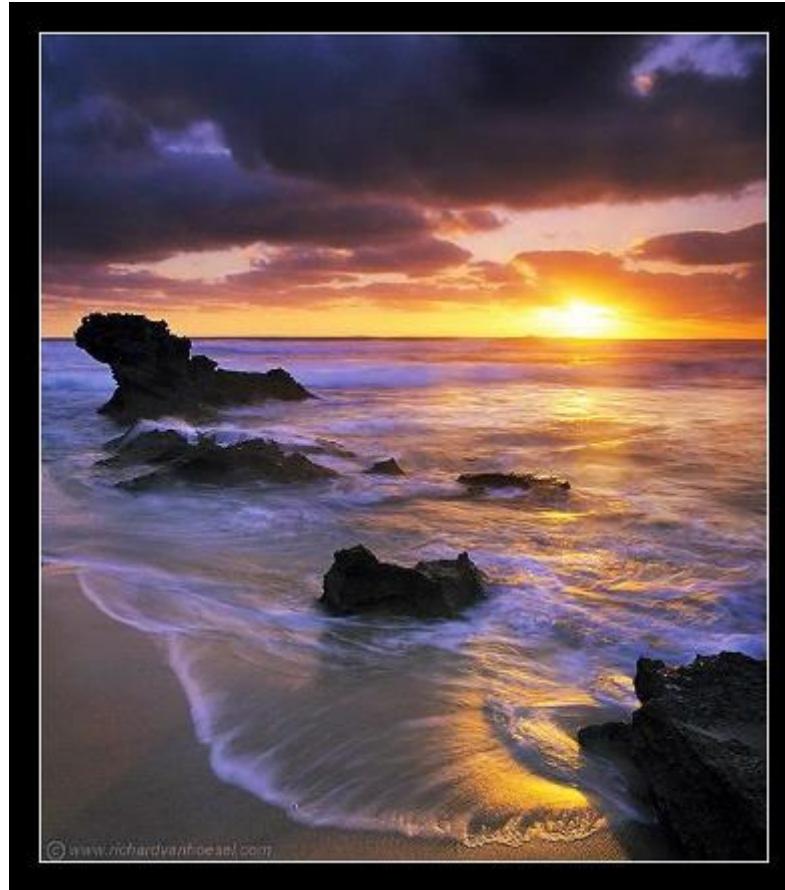
0.15 mm



0.07 mm

# Capturing Light... in man and machine

---

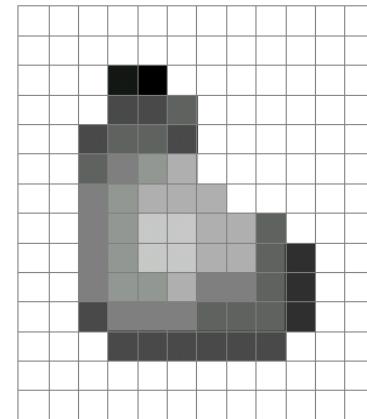
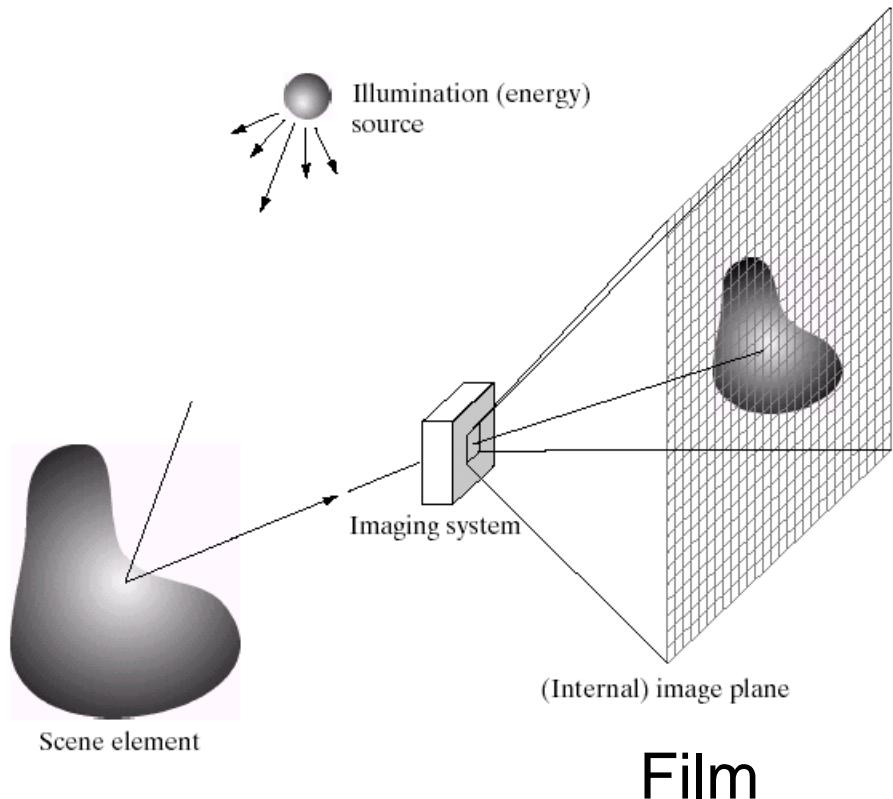


Many slides by  
Alexei A. Efros

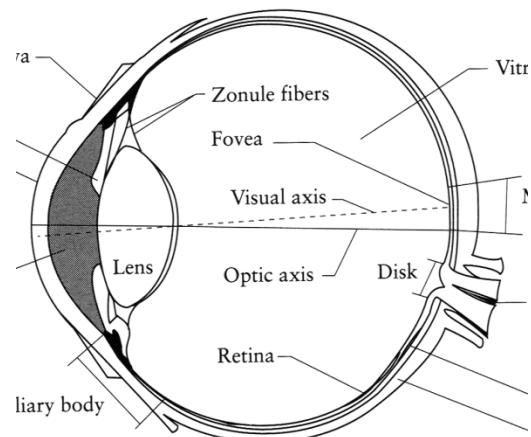
CS 143: Computer Vision  
James Hays, Brown, Fall 2013

# Image Formation

---



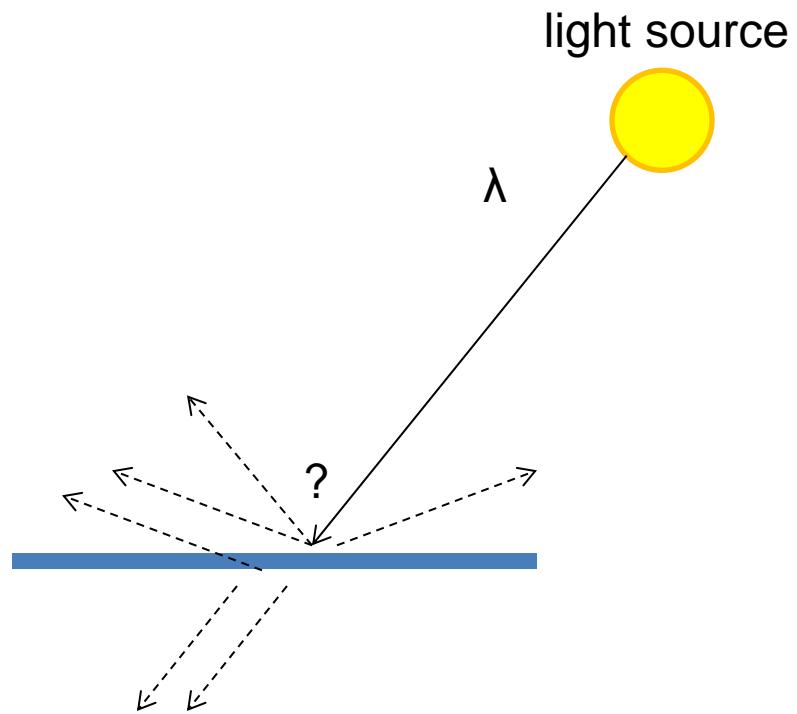
## Digital Camera



## The Eye

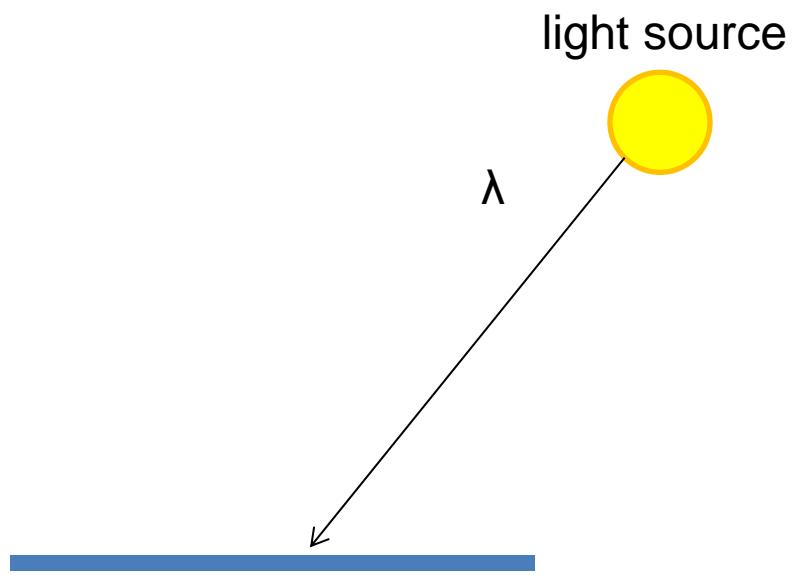
# A photon's life choices

- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



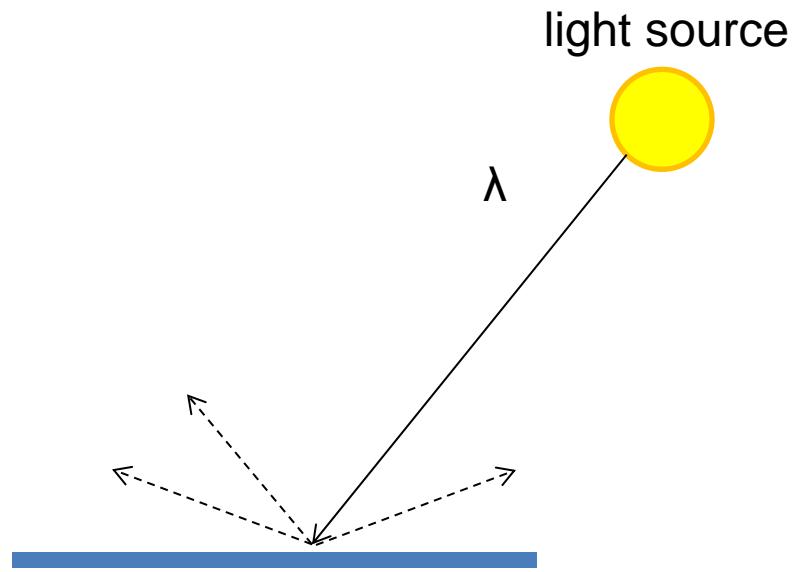
# A photon's life choices

- **Absorption**
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



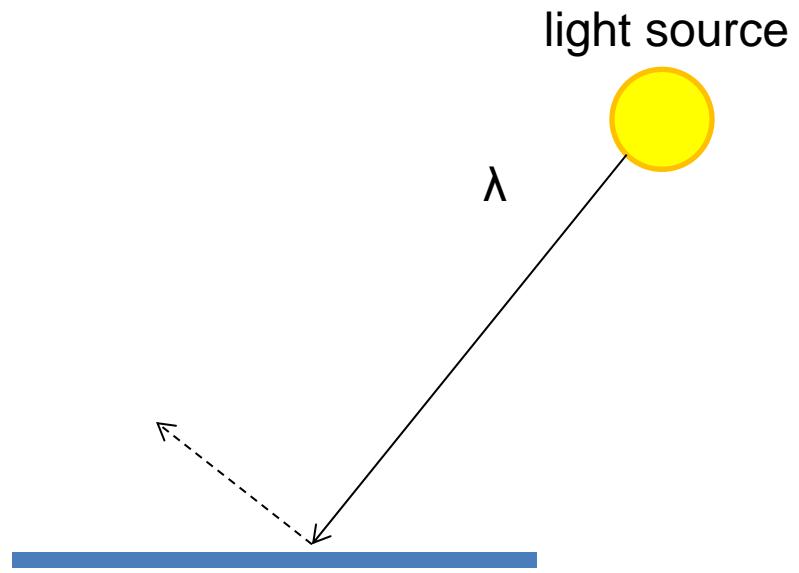
# A photon's life choices

- Absorption
- **Diffuse Reflection**
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



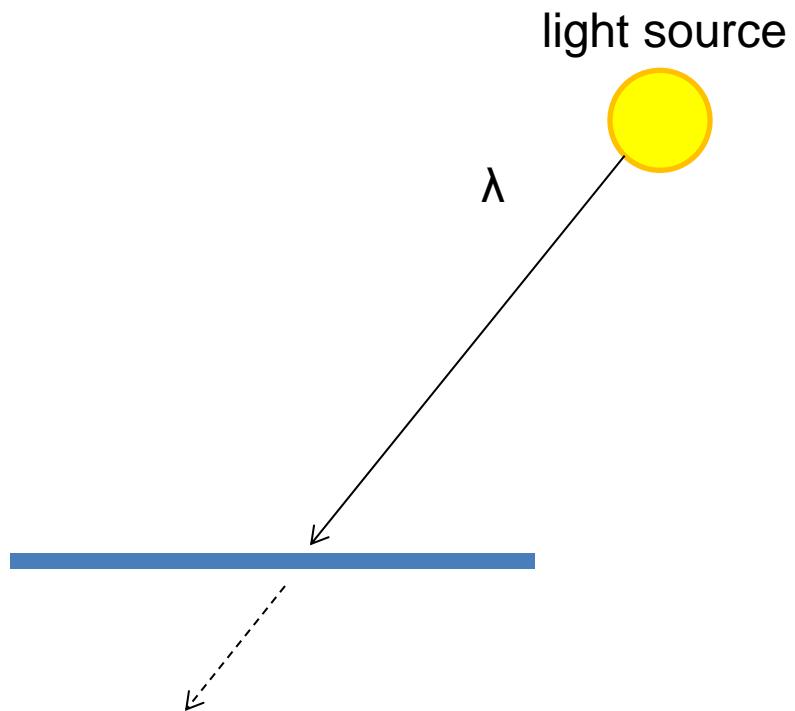
# A photon's life choices

- Absorption
- Diffusion
- **Specular Reflection**
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



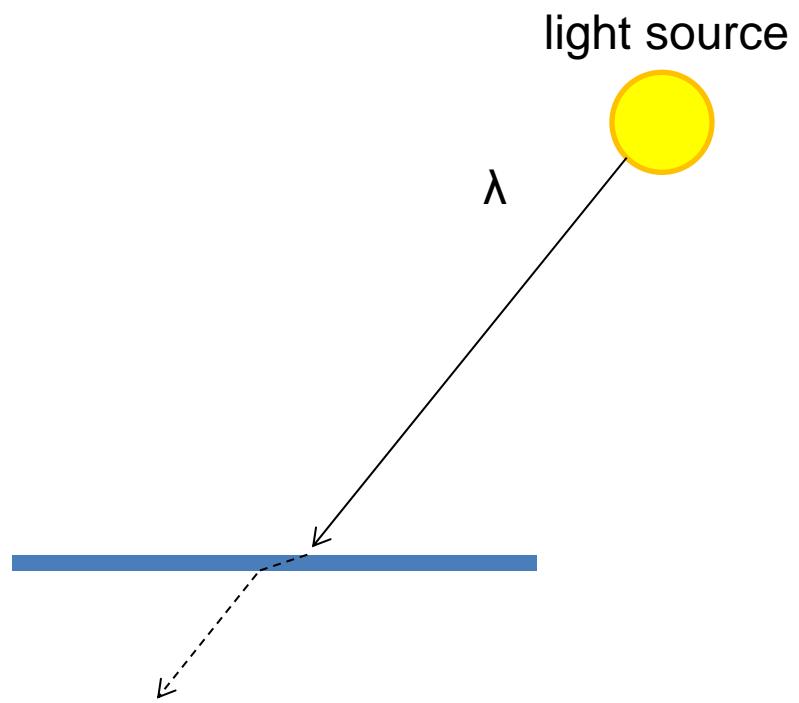
# A photon's life choices

- Absorption
- Diffusion
- Reflection
- **Transparency**
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



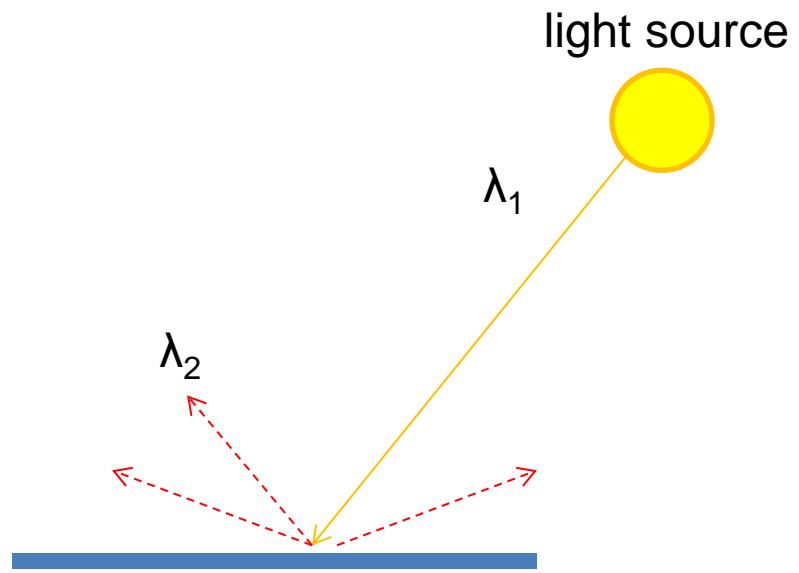
# A photon's life choices

- Absorption
- Diffusion
- Reflection
- Transparency
- **Refraction**
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



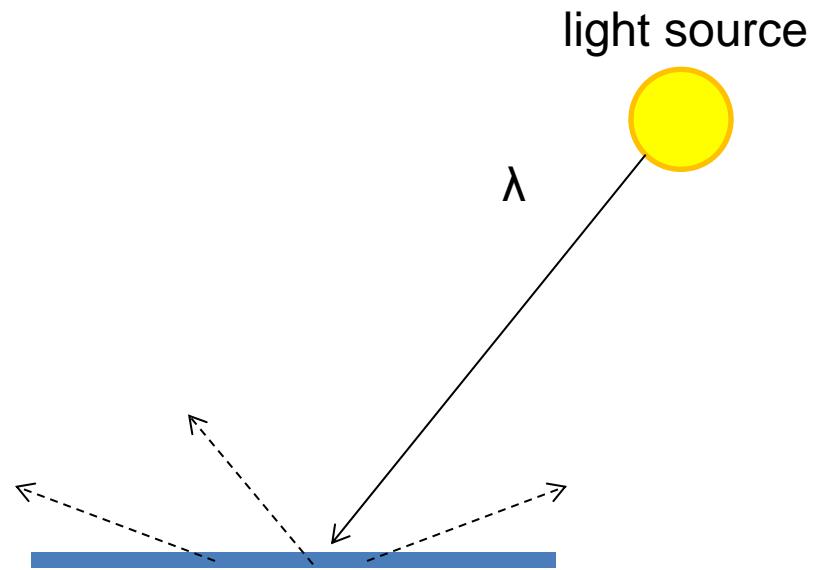
# A photon's life choices

- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- **Fluorescence**
- Subsurface scattering
- Phosphorescence
- Interreflection



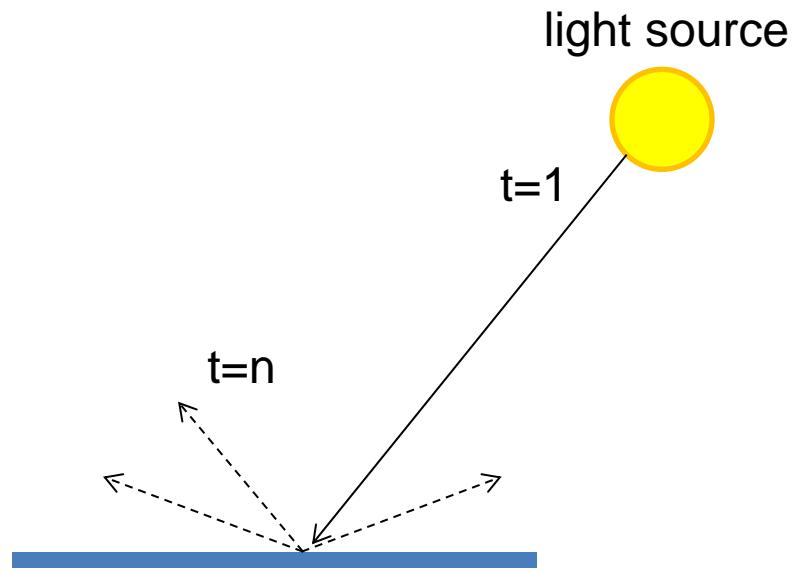
# A photon's life choices

- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- **Subsurface scattering**
- Phosphorescence
- Interreflection



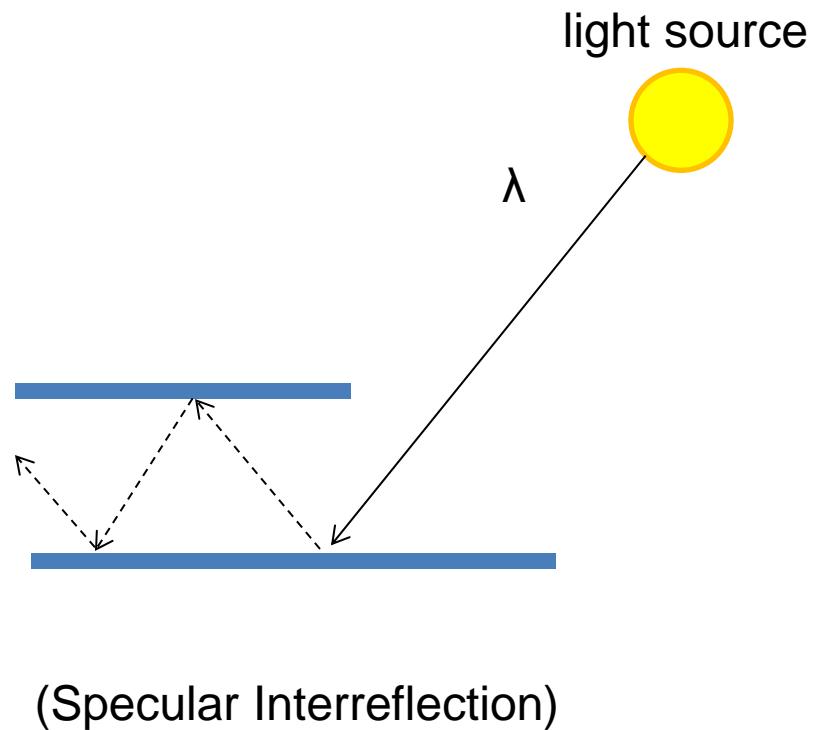
# A photon's life choices

- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- **Phosphorescence**
- Interreflection



# A photon's life choices

- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- **Interreflection**



# Lambertian Reflectance

- In computer vision, surfaces are often assumed to be ideal diffuse reflectors with known dependence on viewing direction.

# Digital camera

---

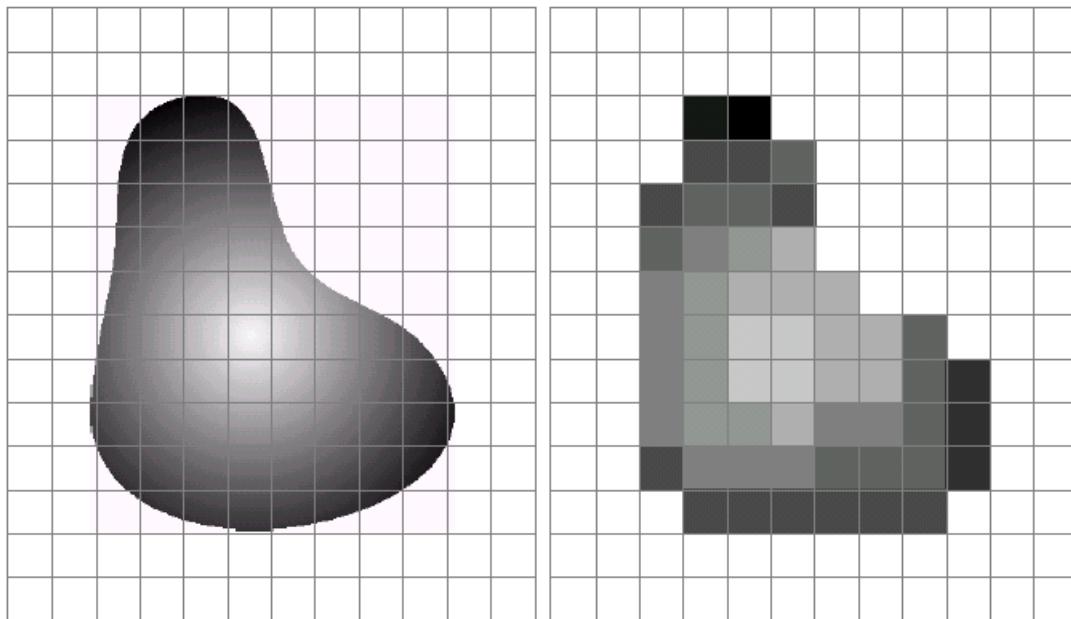


A digital camera replaces film with a sensor array

- Each cell in the array is light-sensitive diode that converts photons to electrons
- Two common types
  - Charge Coupled Device (CCD)
  - CMOS
- <http://electronics.howstuffworks.com/digital-camera.htm>

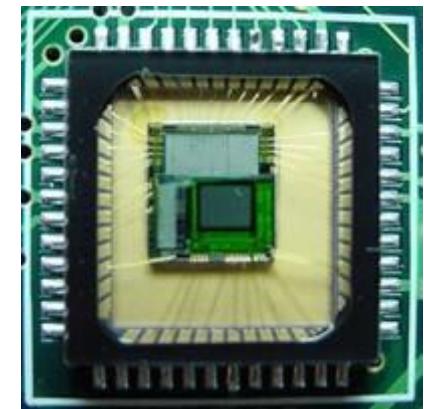
# Sensor Array

---



a b

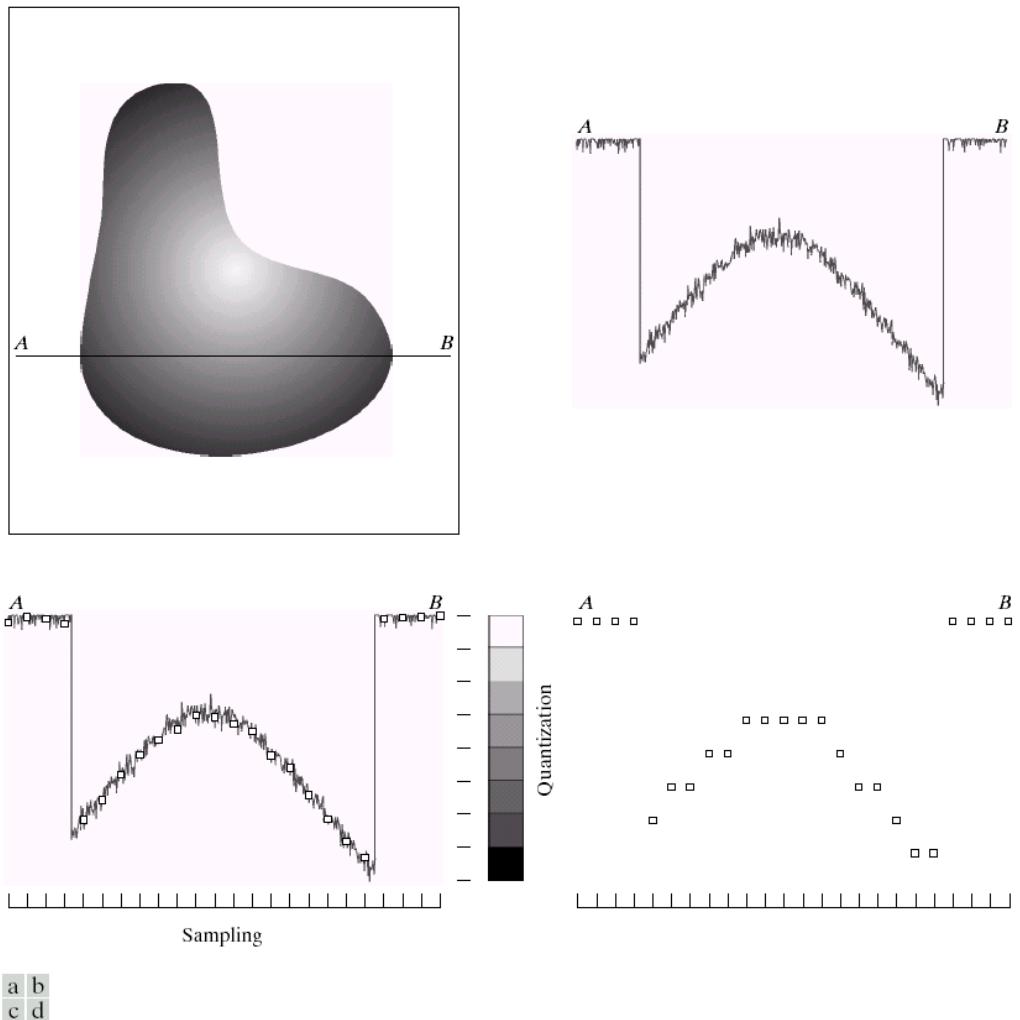
**FIGURE 2.17** (a) Continuous image projected onto a sensor array. (b) Result of image sampling and quantization.



CMOS sensor

# Sampling and Quantization

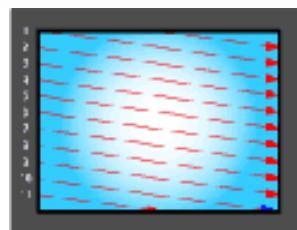
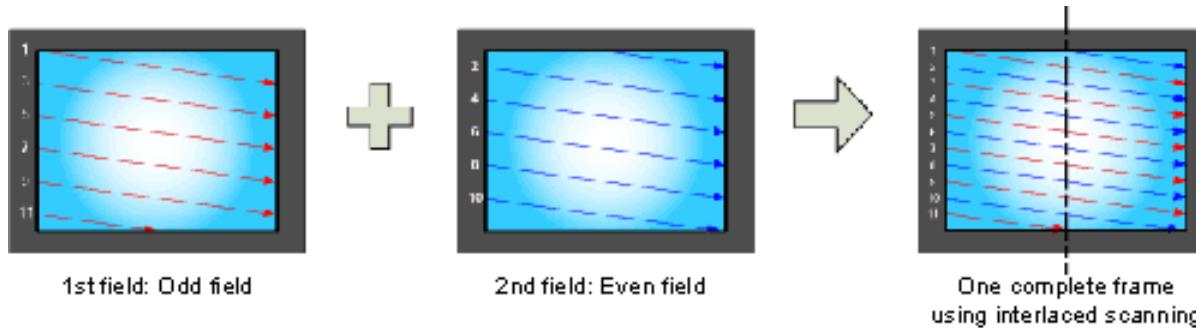
---



**FIGURE 2.16** Generating a digital image. (a) Continuous image. (b) A scan line from *A* to *B* in the continuous image, used to illustrate the concepts of sampling and quantization. (c) Sampling and quantization. (d) Digital scan line.

# Interlace vs. progressive scan

---



One complete frame  
using progressive scanning

# Progressive scan

---



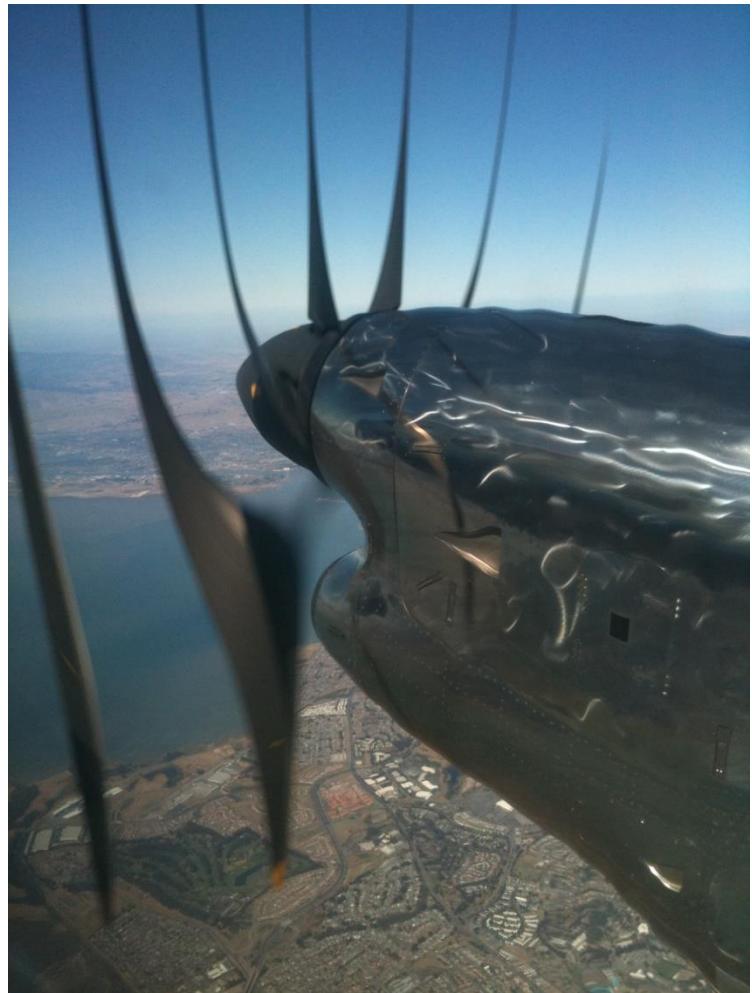
# Interlace

---



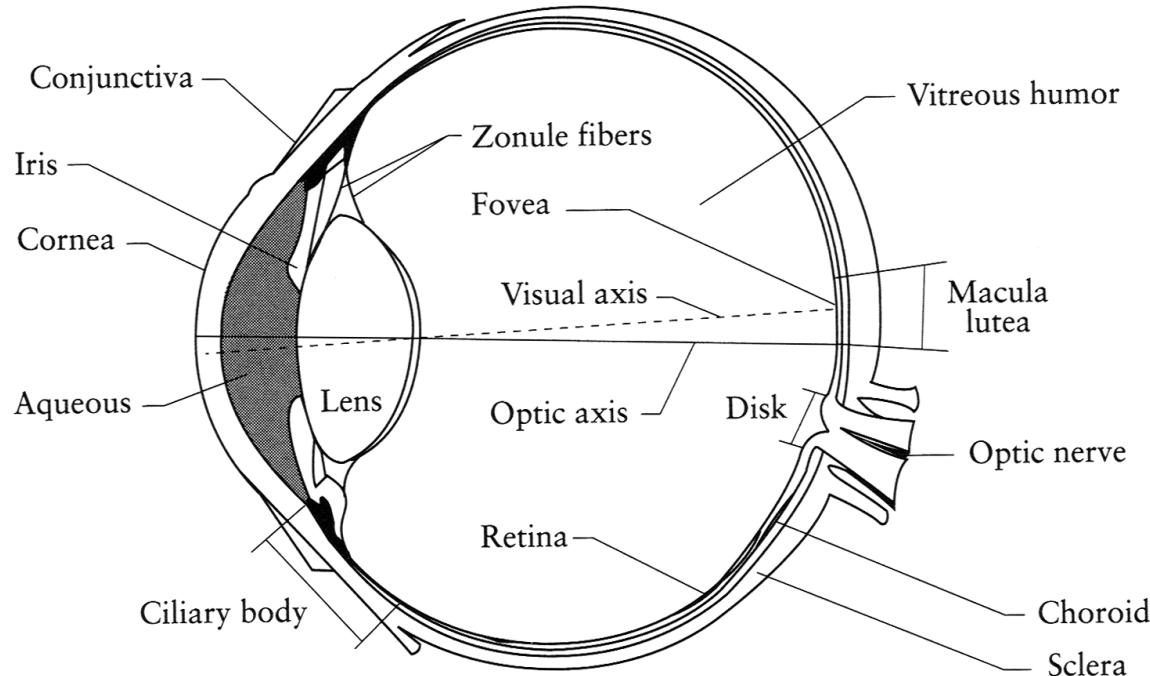
# Rolling Shutter

---



# The Eye

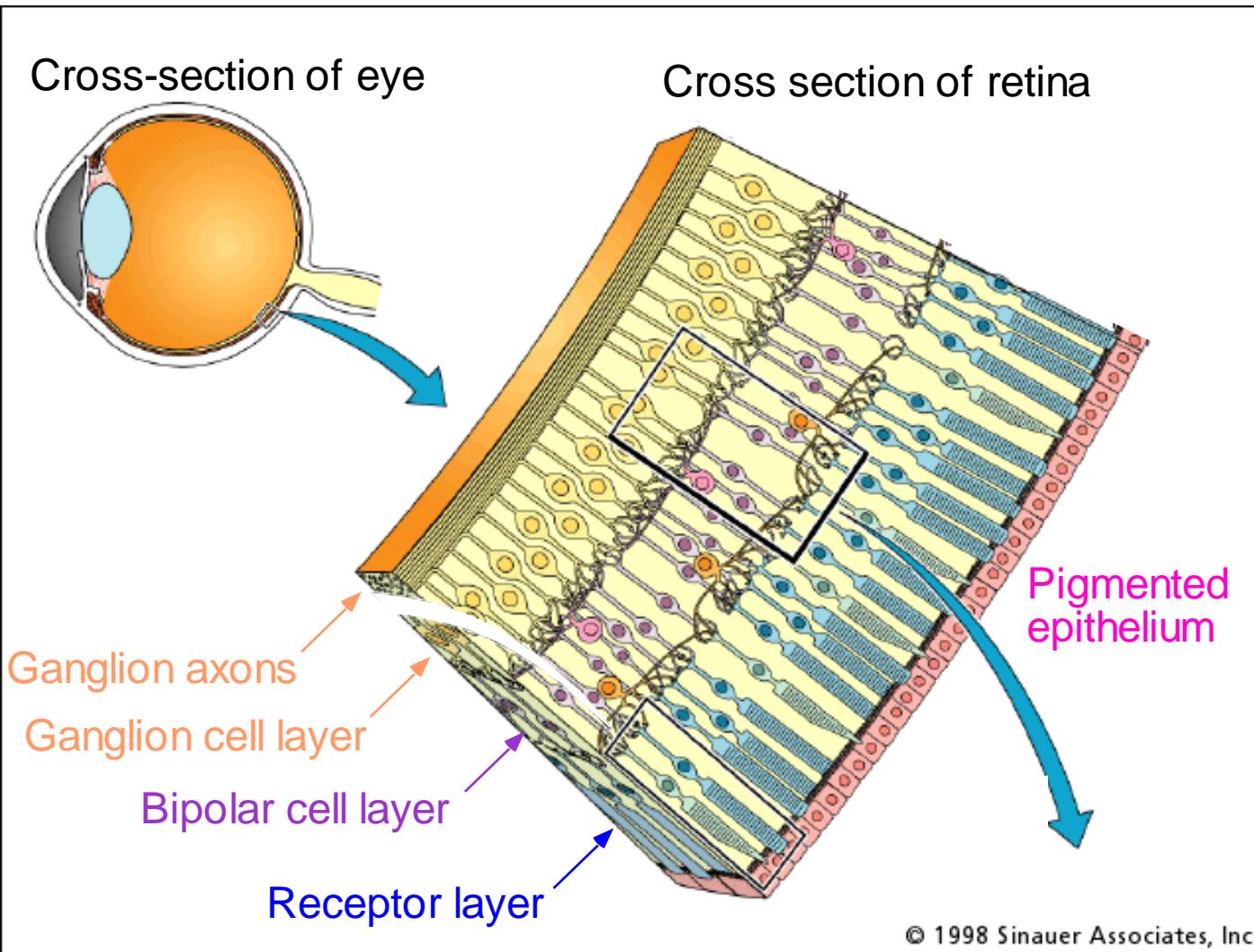
---



The human eye is a camera!

- **Iris** - colored annulus with radial muscles
- **Pupil** - the hole (aperture) whose size is controlled by the iris
- What's the "film"?
  - photoreceptor cells (rods and cones) in the **retina**

# The Retina



# What humans don't have: tapetum lucidum

---



# Two types of light-sensitive receptors

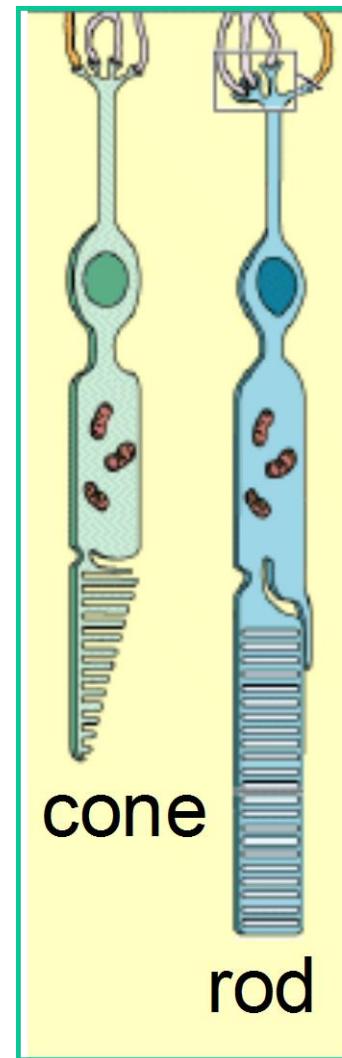
---

## Cones

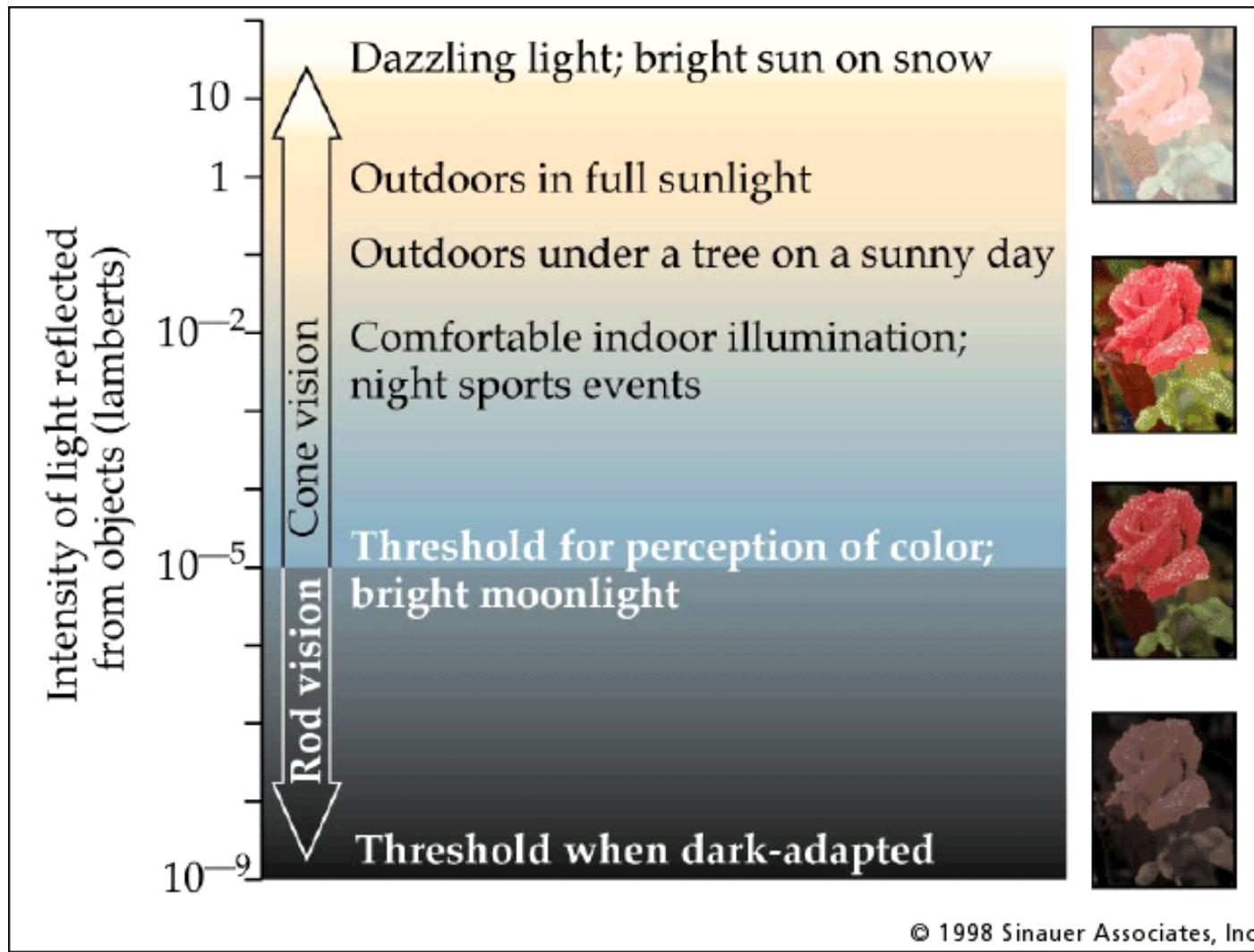
cone-shaped  
less sensitive  
operate in high light  
color vision

## Rods

rod-shaped  
highly sensitive  
operate at night  
gray-scale vision

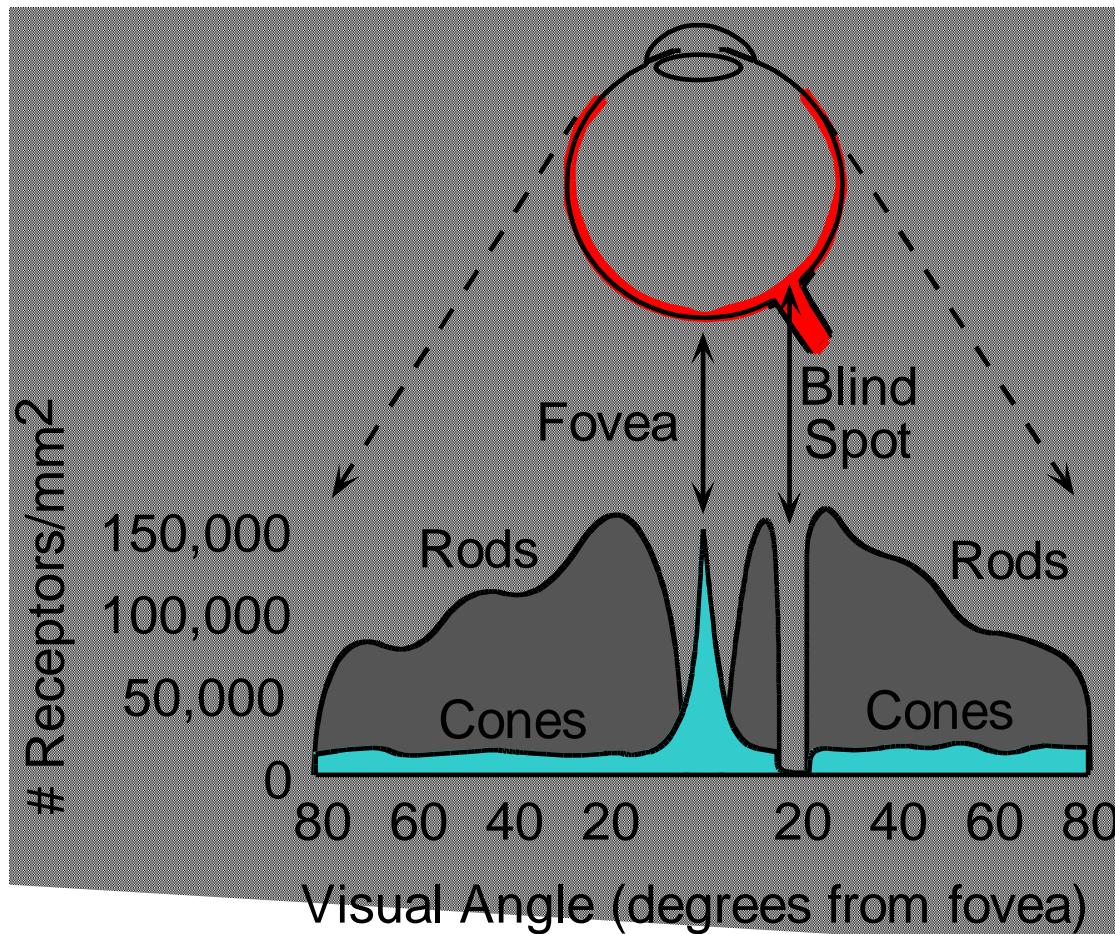


# Rod / Cone sensitivity



# Distribution of Rods and Cones

---



Night Sky: why are there more stars off-center?

Averted vision: [http://en.wikipedia.org/wiki/Averted\\_vision](http://en.wikipedia.org/wiki/Averted_vision)

# Eye Movements

---

## Saccades

Can be consciously controlled. Related to perceptual attention.  
200ms to initiation, 20 to 200ms to carry out. Large amplitude.

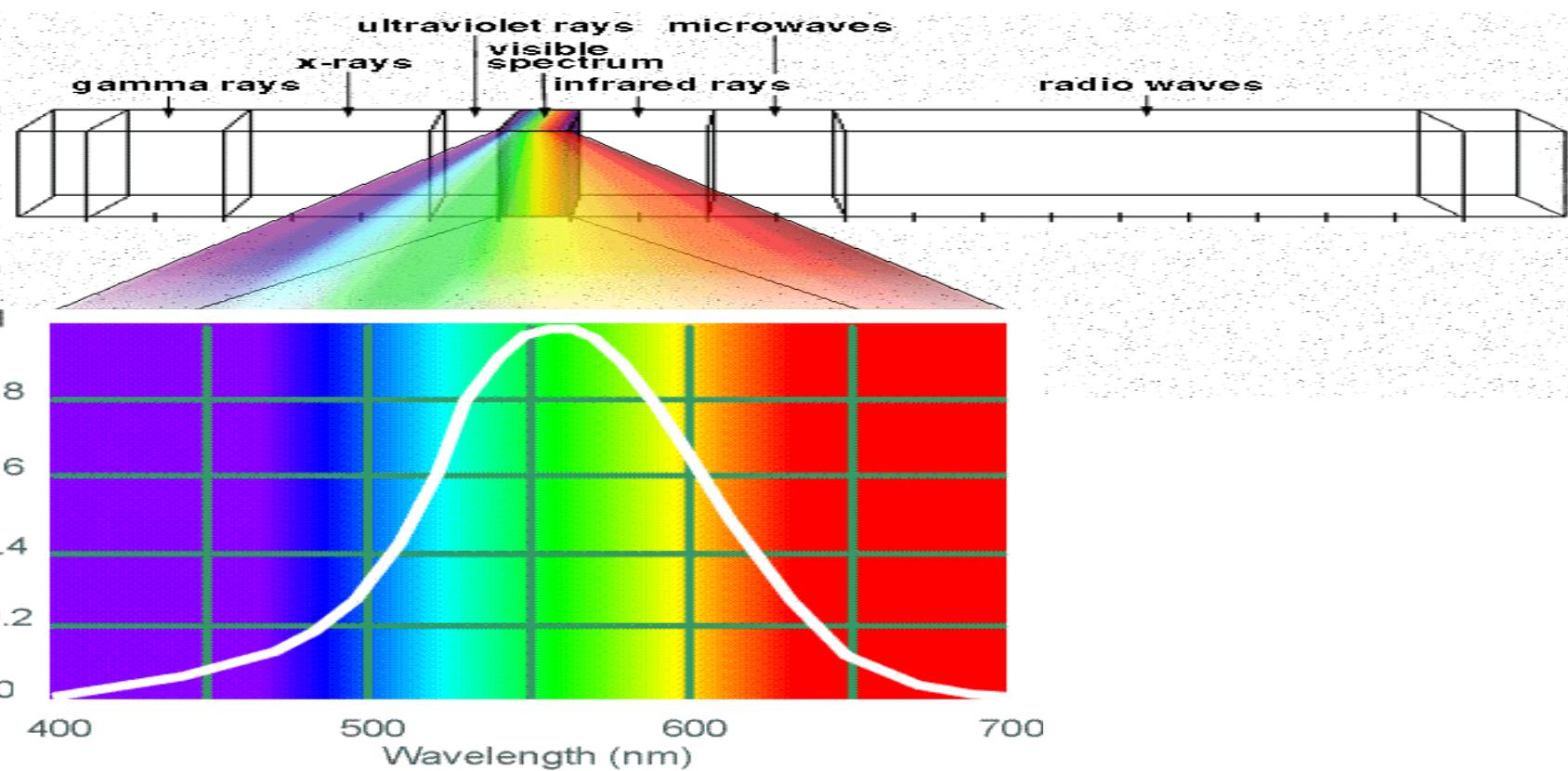
## Microsaccades

Involuntary. Smaller amplitude. Especially evident during prolonged fixation. Function debated.

## Ocular microtremor (OMT)

involuntary. high frequency (up to 80Hz), small amplitude.

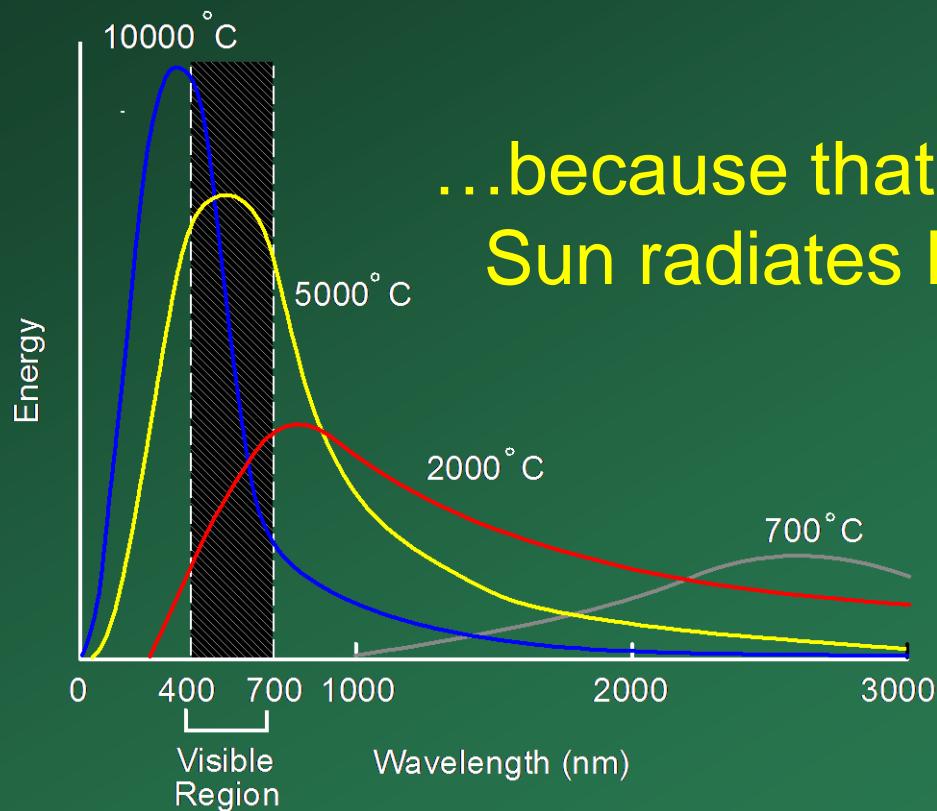
# Electromagnetic Spectrum



Human Luminance Sensitivity Function

# Visible Light

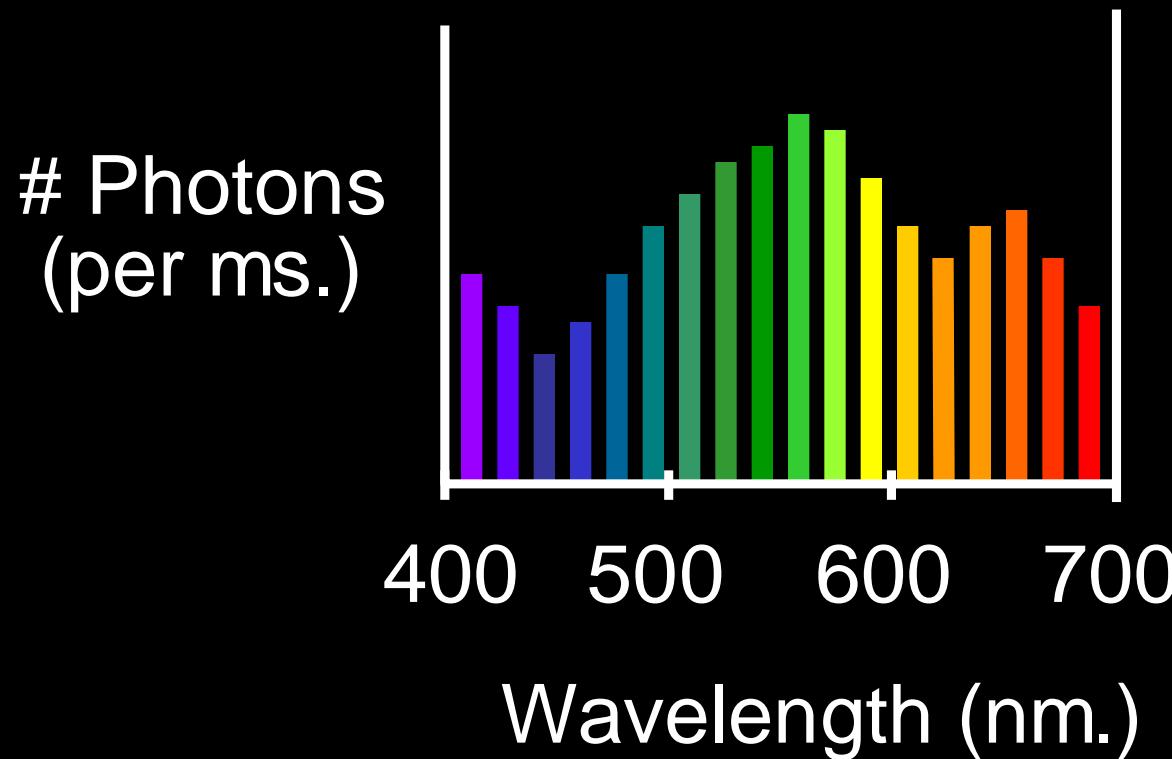
Why do we see light of these wavelengths?



...because that's where the Sun radiates EM energy

# The Physics of Light

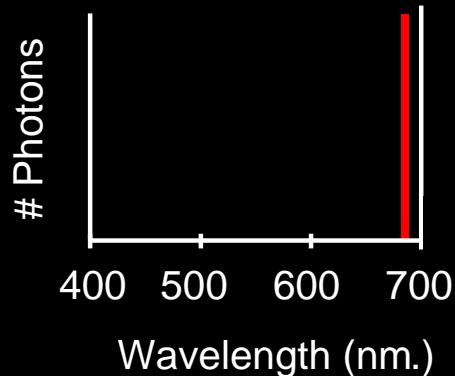
Any patch of light can be completely described physically by its spectrum: the number of photons (per time unit) at each wavelength 400 - 700 nm.



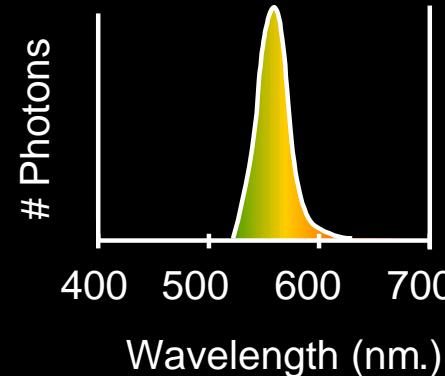
# The Physics of Light

Some examples of the spectra of light sources

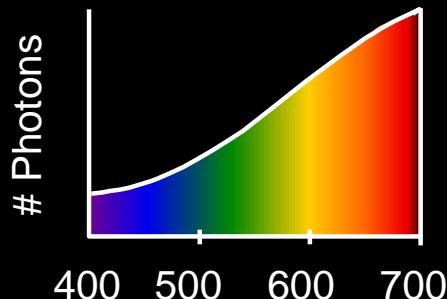
A. Ruby Laser



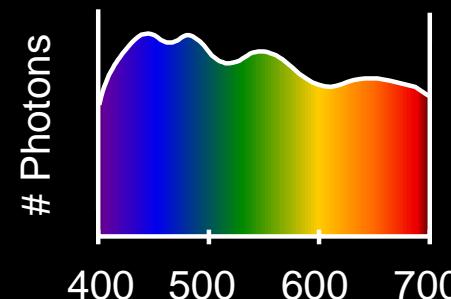
B. Gallium Phosphide Crystal



C. Tungsten Lightbulb

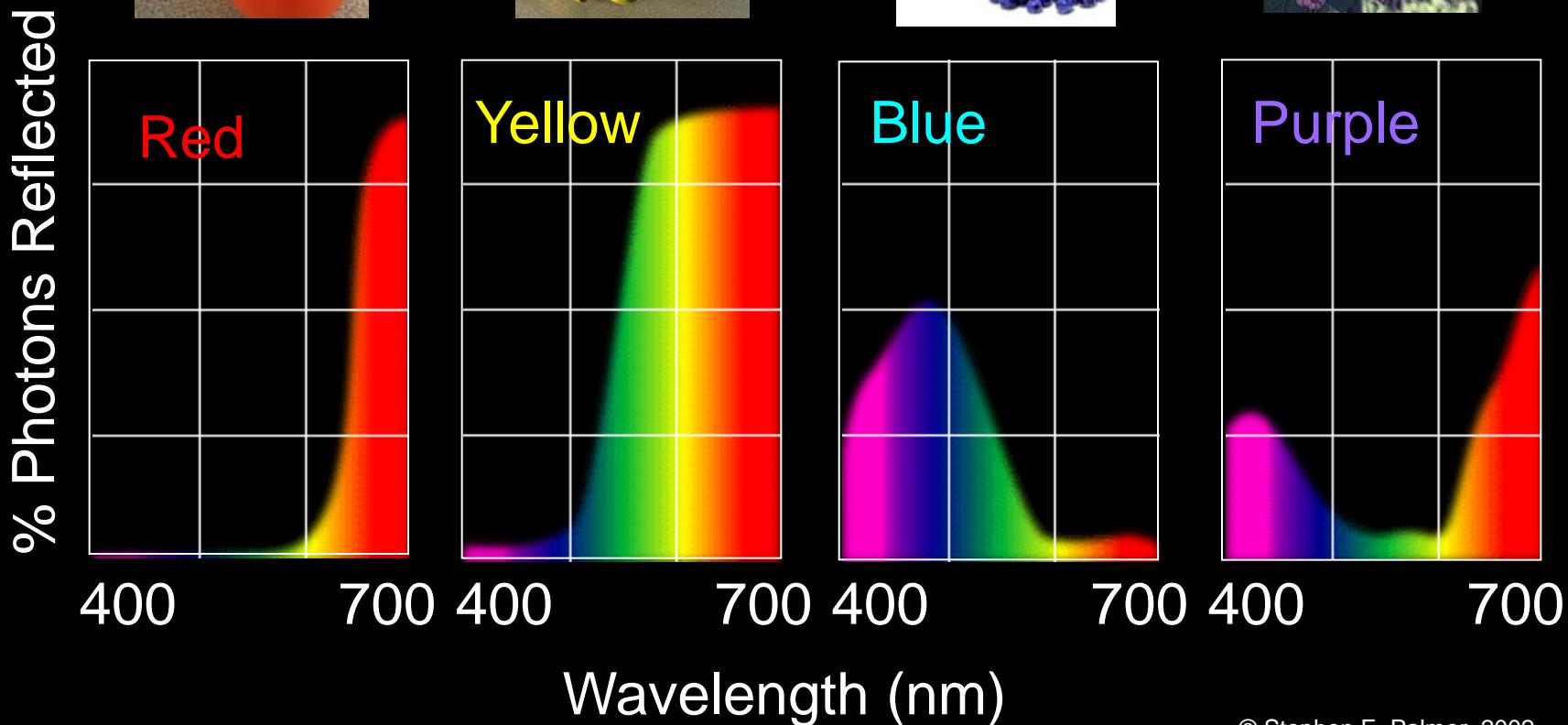


D. Normal Daylight



# The Physics of Light

Some examples of the reflectance spectra of surfaces

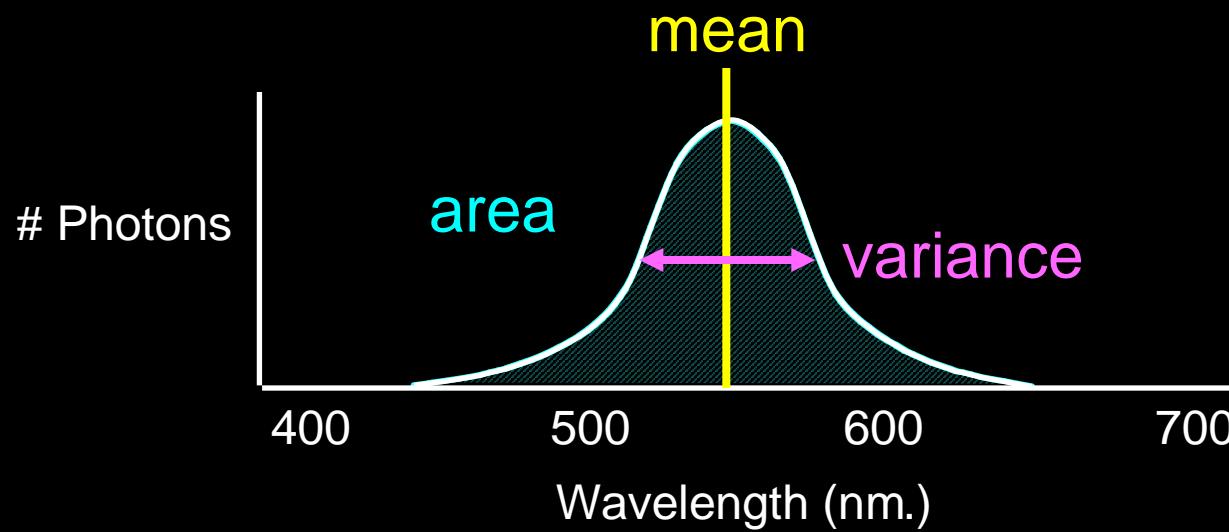


# The Psychophysical Correspondence

There is no simple functional description for the perceived color of all lights under all viewing conditions, but .....

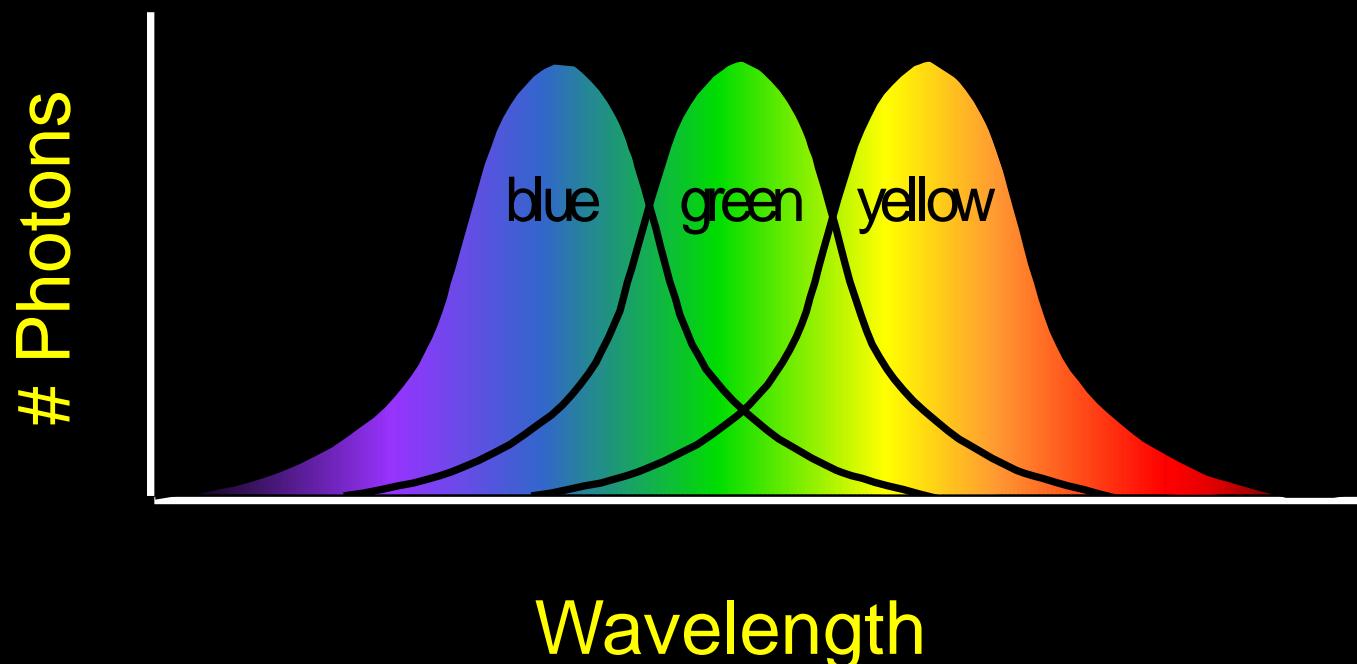
A helpful constraint:

Consider only physical spectra with normal distributions



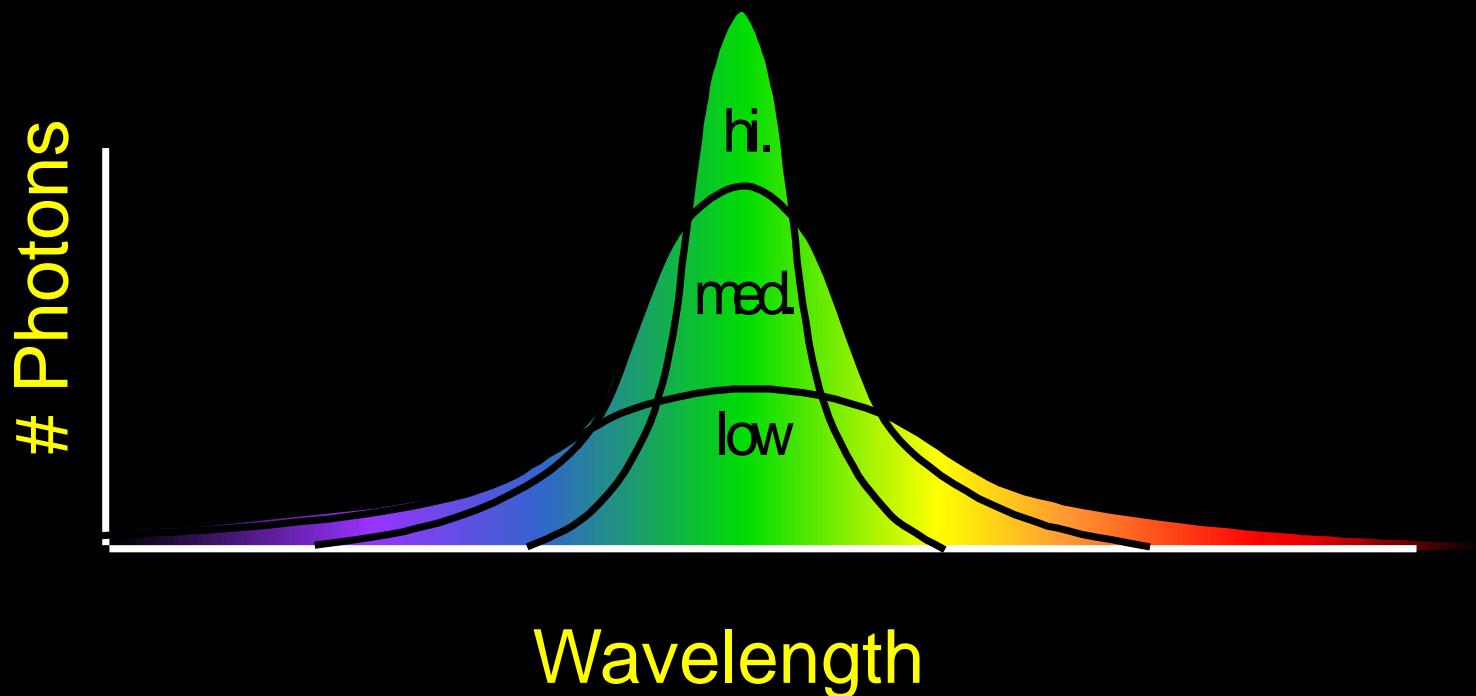
# The Psychophysical Correspondence

Mean  $\longleftrightarrow$  Hue



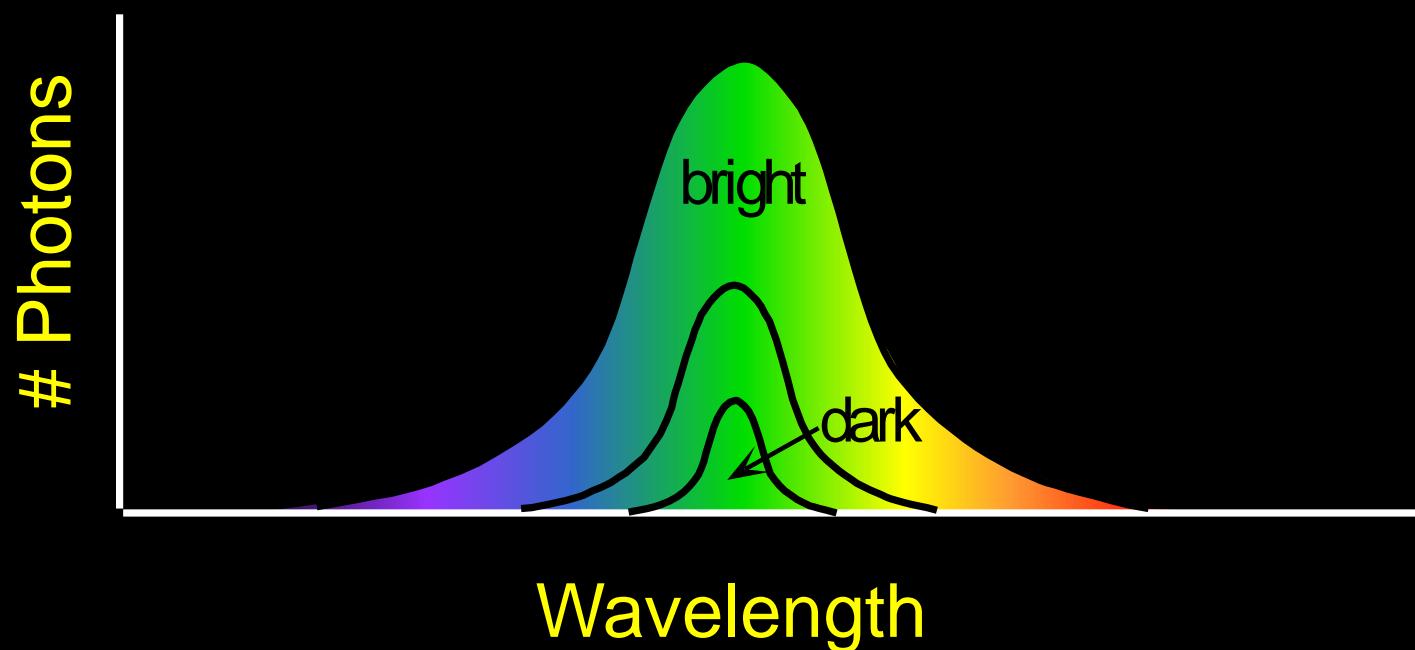
# The Psychophysical Correspondence

Variance  $\longleftrightarrow$  Saturation



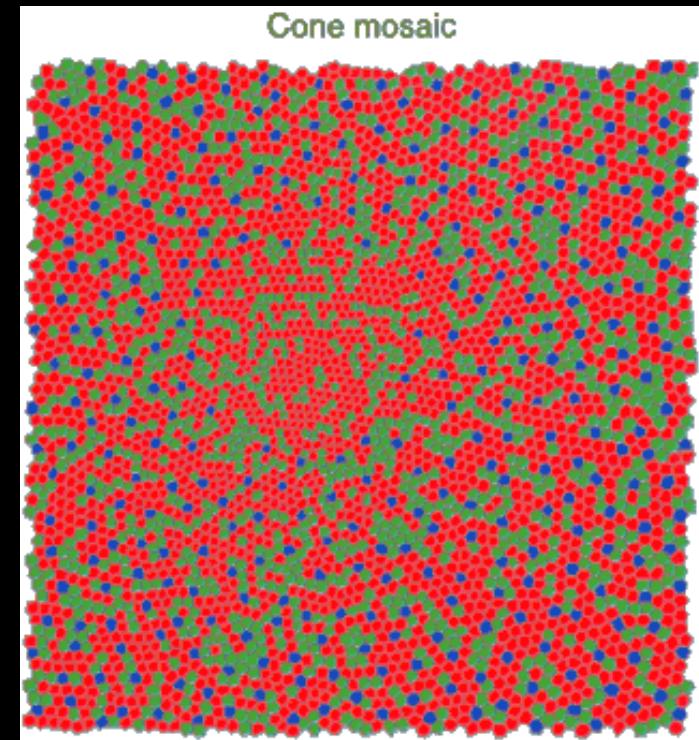
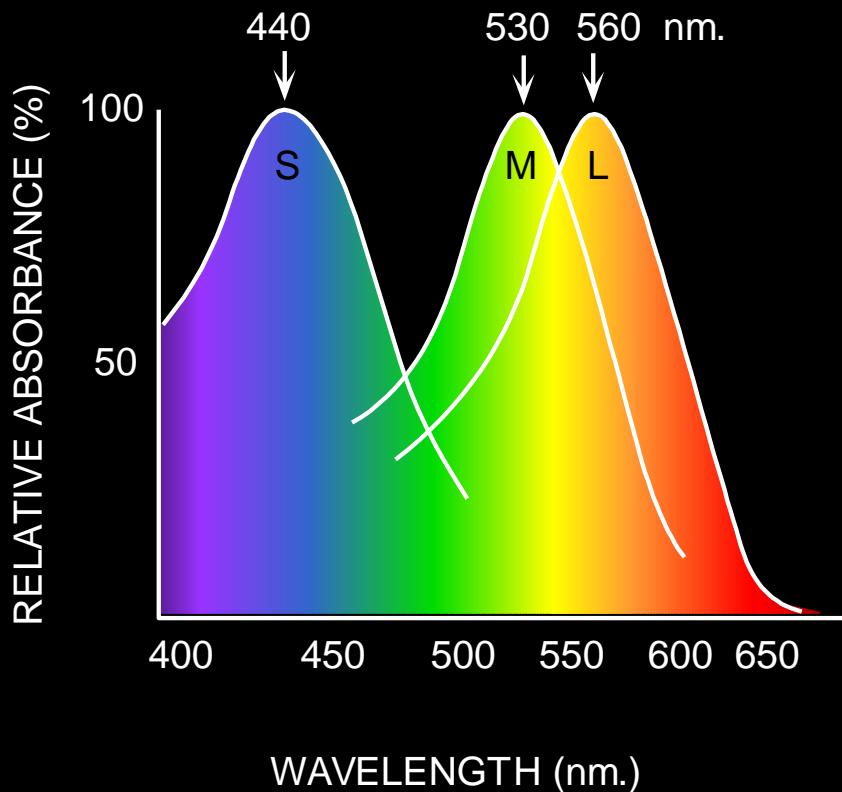
# The Psychophysical Correspondence

Area  $\longleftrightarrow$  Brightness



# Physiology of Color Vision

Three kinds of cones:



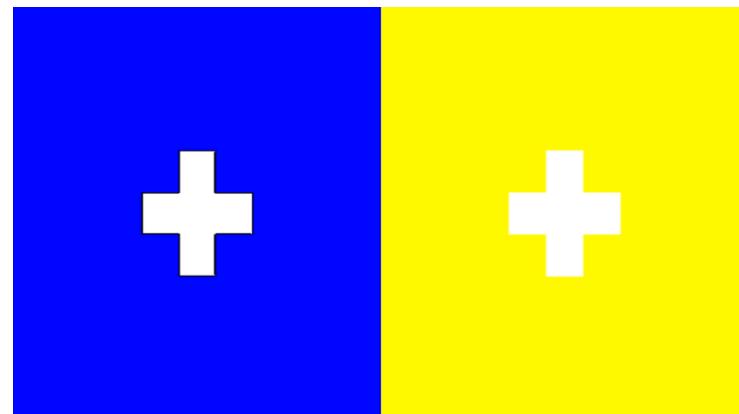
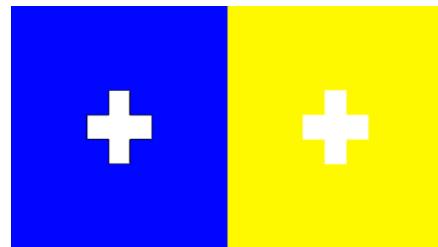
- Why are M and L cones so close?
- Why are there 3?

# Impossible Colors

---

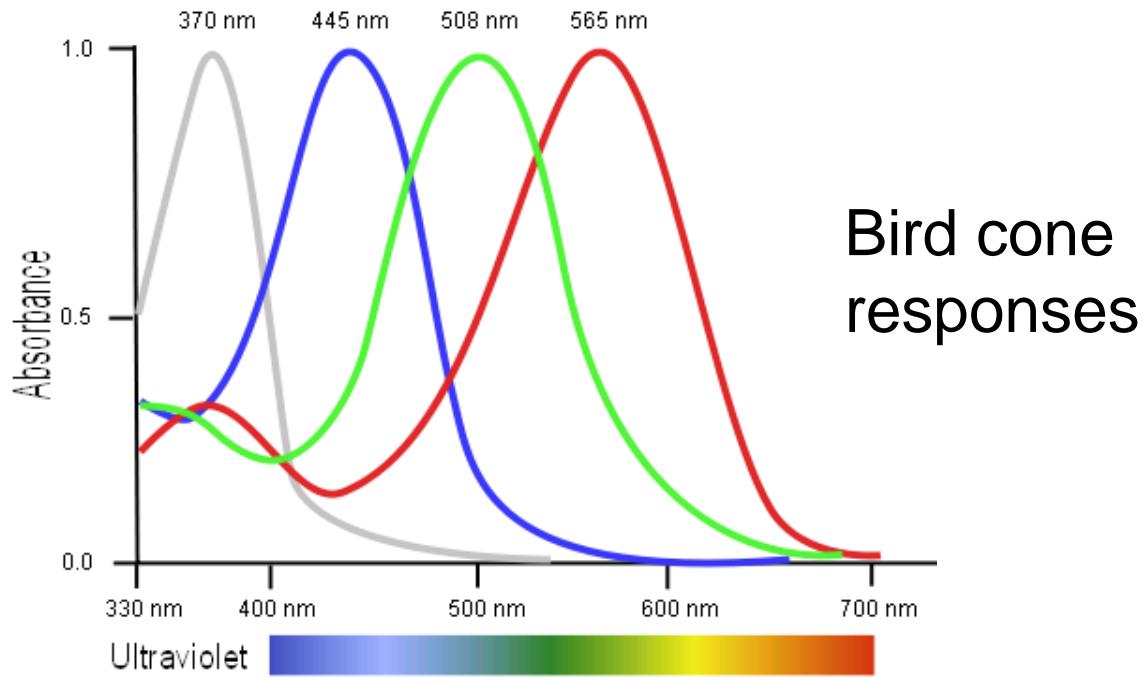
Can you make the cones respond in ways that typical light spectra never would?

[http://en.wikipedia.org/wiki/Impossible\\_colors](http://en.wikipedia.org/wiki/Impossible_colors)



# Tetrachromatism

---

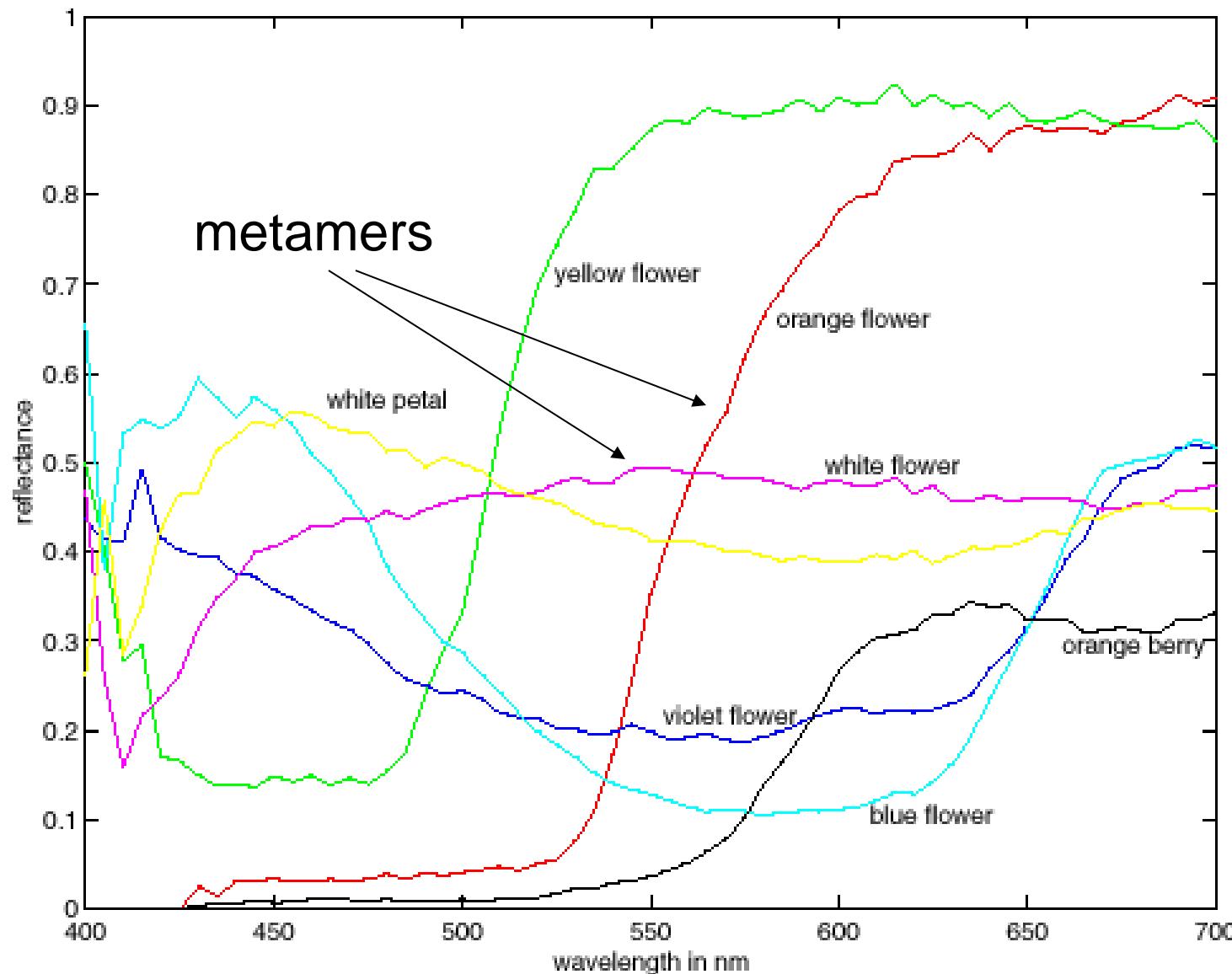


Most birds, and many other animals, have cones for ultraviolet light.

Some humans, mostly female, seem to have slight tetrachromatism.

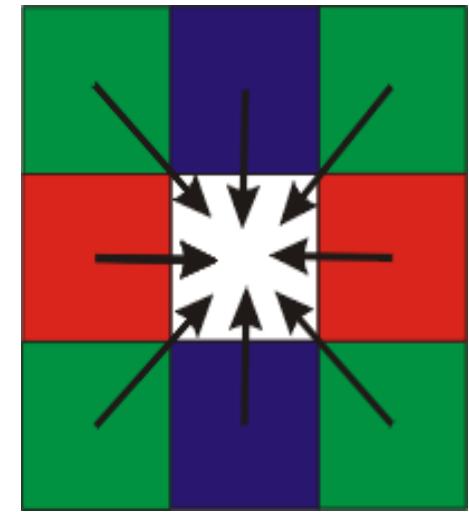
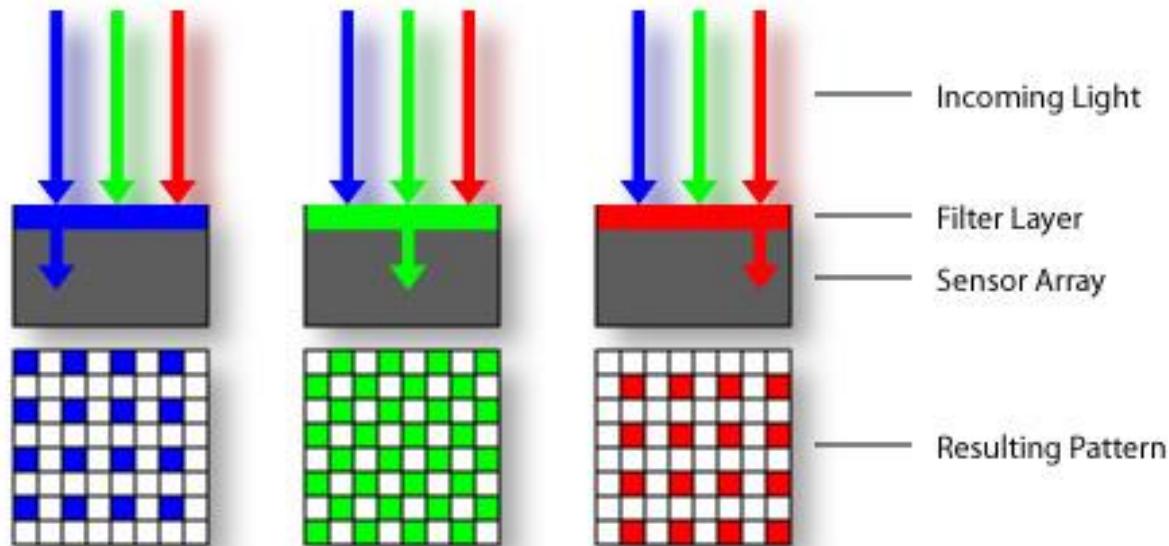
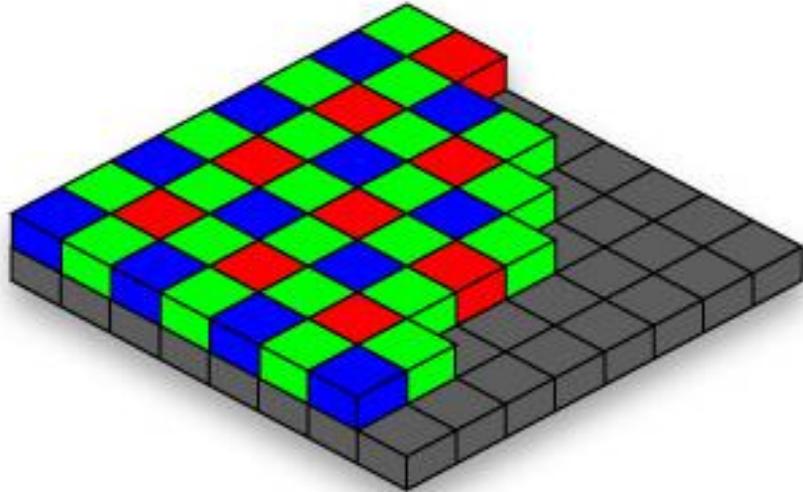
# More Spectra

---



# Practical Color Sensing: Bayer Grid

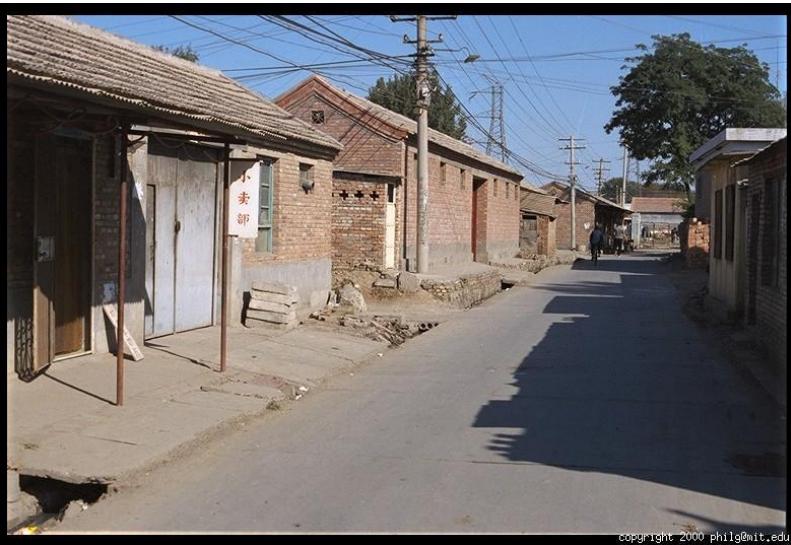
---



Estimate RGB  
at 'G' cells from  
neighboring  
values

# Color Image

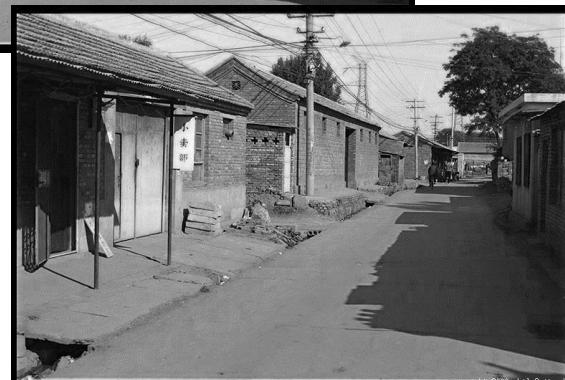
---



R



G



B

# Images in Matlab

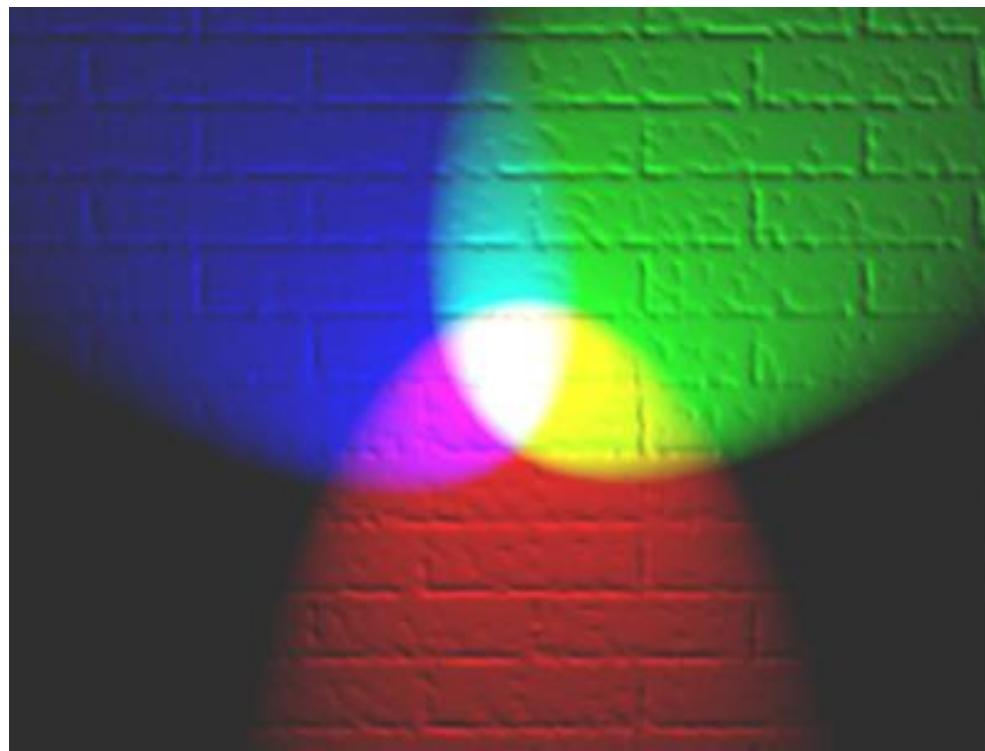
- Images represented as a matrix
- Suppose we have a NxM RGB image called “im”
  - $\text{im}(1,1,1)$  = top-left pixel value in R-channel
  - $\text{im}(y, x, b)$  = y pixels down, x pixels to right in the b<sup>th</sup> channel
  - $\text{im}(N, M, 3)$  = bottom-right pixel in B-channel
- **imread(filename)** returns a uint8 image (values 0 to 255)
  - Convert to double format (values 0 to 1) with **im2double**

column											
row	0.92	0.93	0.94	0.97	0.62	0.37	0.85	0.97	0.93	0.92	0.99
	0.95	0.89	0.82	0.89	0.56	0.31	0.75	0.92	0.81	0.95	0.91
	0.89	0.72	0.51	0.55	0.51	0.42	0.57	0.41	0.49	0.91	0.92
	0.96	0.95	0.88	0.94	0.56	0.46	0.91	0.87	0.90	0.97	0.95
	0.71	0.81	0.81	0.87	0.57	0.37	0.80	0.88	0.89	0.79	0.85
	0.49	0.62	0.60	0.58	0.50	0.60	0.58	0.50	0.61	0.45	0.33
	0.86	0.84	0.74	0.58	0.51	0.39	0.73	0.92	0.91	0.49	0.74
	0.96	0.67	0.54	0.85	0.48	0.37	0.88	0.90	0.94	0.82	0.93
	0.69	0.49	0.56	0.66	0.43	0.42	0.77	0.73	0.71	0.90	0.99
	0.79	0.73	0.90	0.67	0.33	0.61	0.69	0.79	0.73	0.93	0.97
	0.91	0.94	0.89	0.49	0.41	0.78	0.78	0.77	0.89	0.99	0.93
	0.65	0.45	0.56	0.66	0.45	0.42	0.77	0.75	0.71	0.90	0.99
	0.79	0.73	0.90	0.67	0.33	0.61	0.69	0.79	0.73	0.93	0.97
	0.91	0.94	0.89	0.49	0.41	0.78	0.78	0.77	0.89	0.99	0.93
	0.65	0.45	0.56	0.66	0.45	0.42	0.77	0.75	0.71	0.90	0.99
	0.79	0.73	0.90	0.67	0.33	0.61	0.69	0.79	0.73	0.93	0.97
	0.91	0.94	0.89	0.49	0.41	0.78	0.78	0.77	0.89	0.99	0.93
	0.65	0.45	0.56	0.66	0.45	0.42	0.77	0.75	0.71	0.90	0.99
	0.79	0.73	0.90	0.67	0.33	0.61	0.69	0.79	0.73	0.93	0.97
	0.91	0.94	0.89	0.49	0.41	0.78	0.78	0.77	0.89	0.99	0.93

# Color spaces

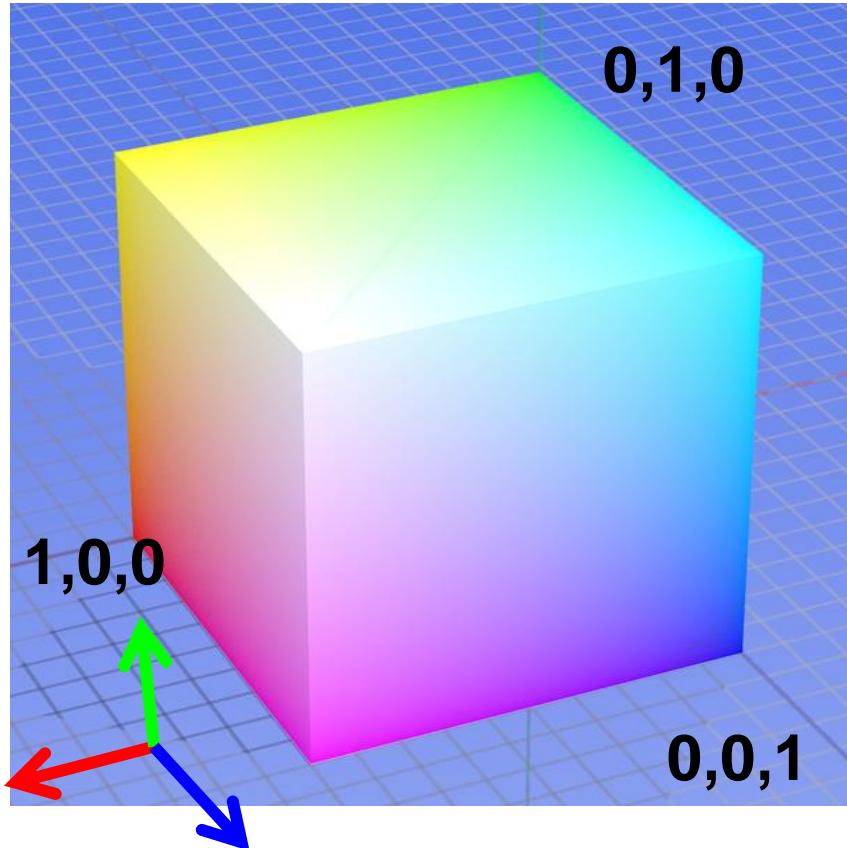
---

How can we represent color?



# Color spaces: RGB

Default color space



**R**  
( $G=0, B=0$ )

**G**  
( $R=0, B=0$ )

**B**  
( $R=0, G=0$ )

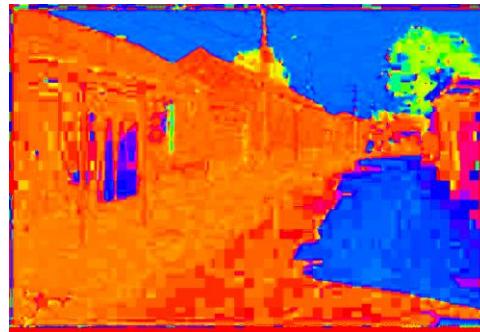
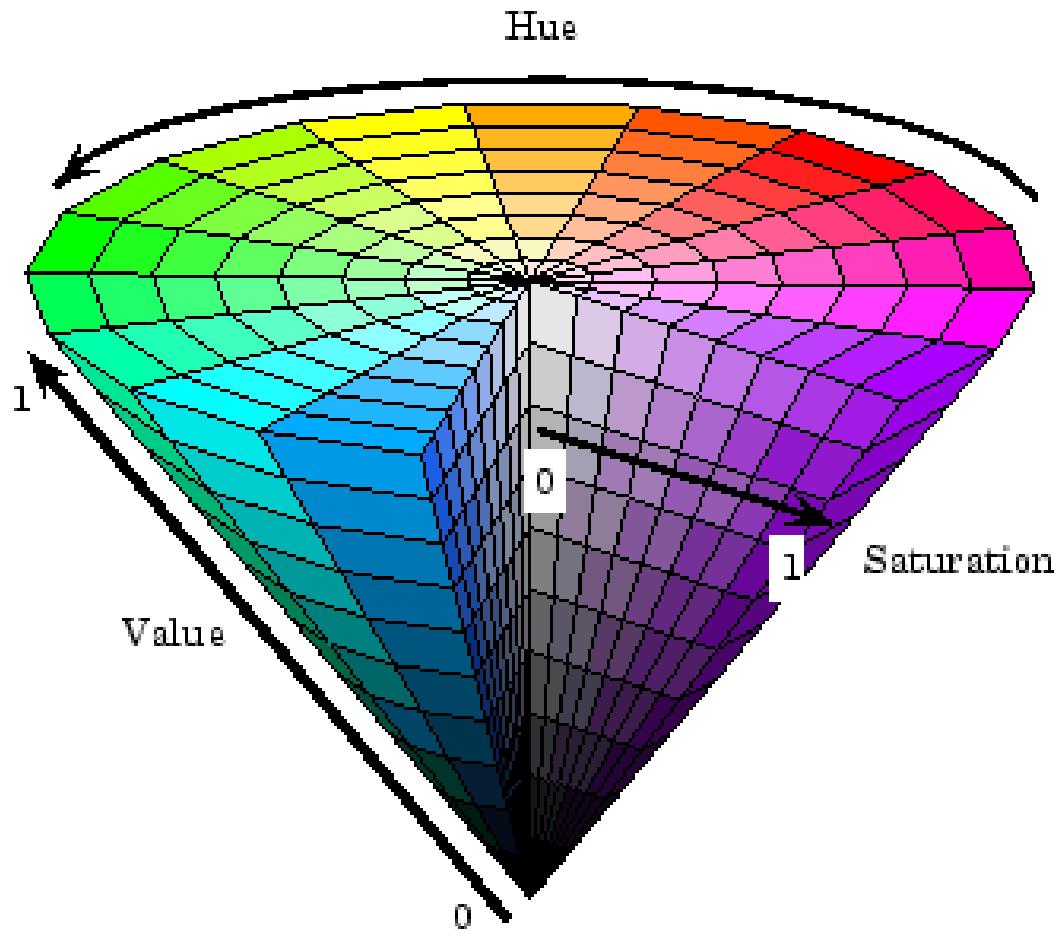
Some drawbacks

- Strongly correlated channels
- Non-perceptual

# Color spaces: HSV



## Intuitive color space



**H**  
( $S=1, V=1$ )



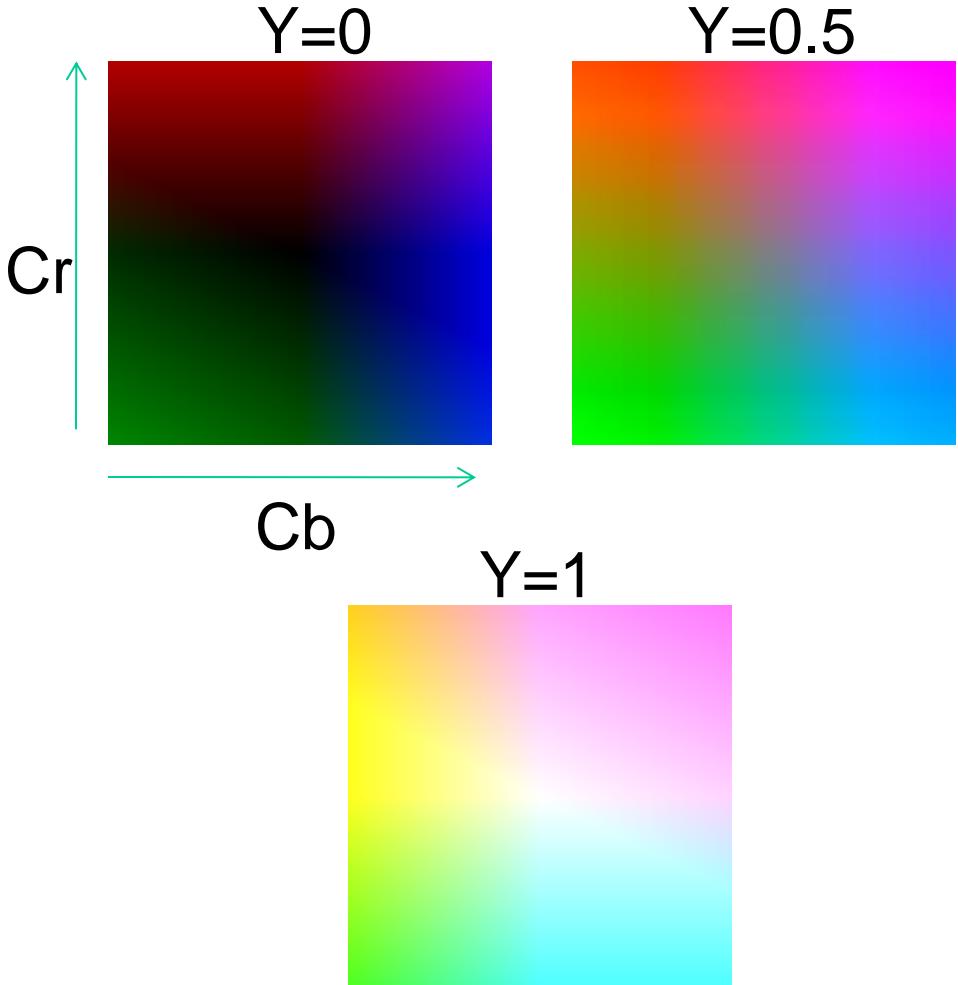
**S**  
( $H=1, V=1$ )



**V**  
( $H=1, S=0$ )

# Color spaces: YCbCr

Fast to compute, good for compression, used by TV



**Y**  
(Cb=0.5,Cr=0.5)



**Cb**  
(Y=0.5,Cr=0.5)

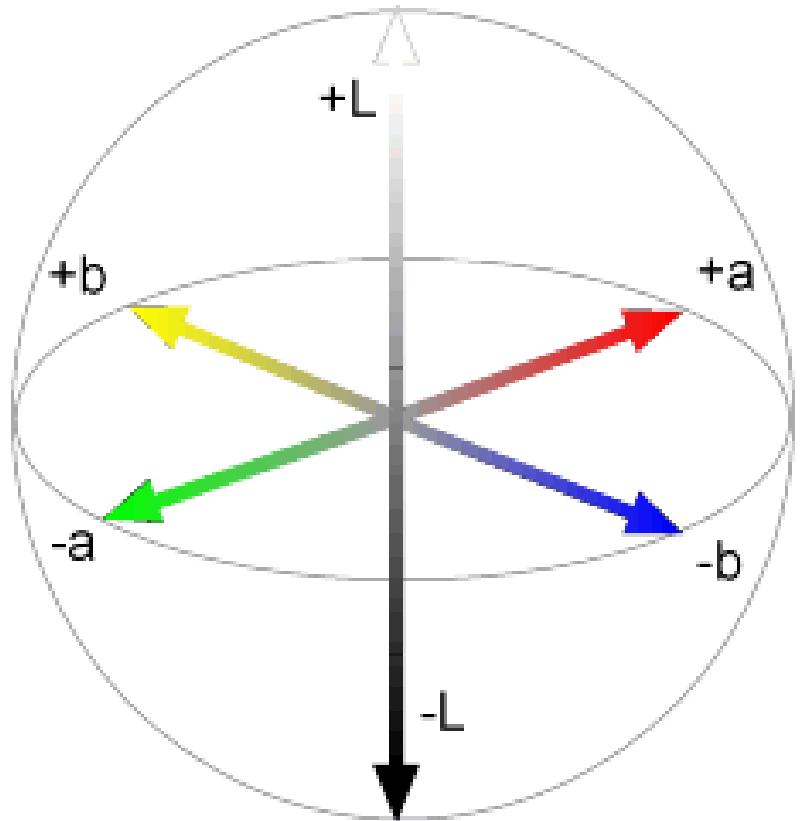


**Cr**  
(Y=0.5,Cb=0.5)

# Color spaces: L\*a\*b\*



“Perceptually uniform”\* color space



**L**  
( $a=0, b=0$ )



**a**  
( $L=65, b=0$ )



**b**  
( $L=65, a=0$ )

---

If you had to choose, would you rather go without luminance or chrominance?

If you had to choose, would you rather go without luminance or chrominance?

# Most information in intensity

---



Only color shown – constant intensity

# Most information in intensity

---

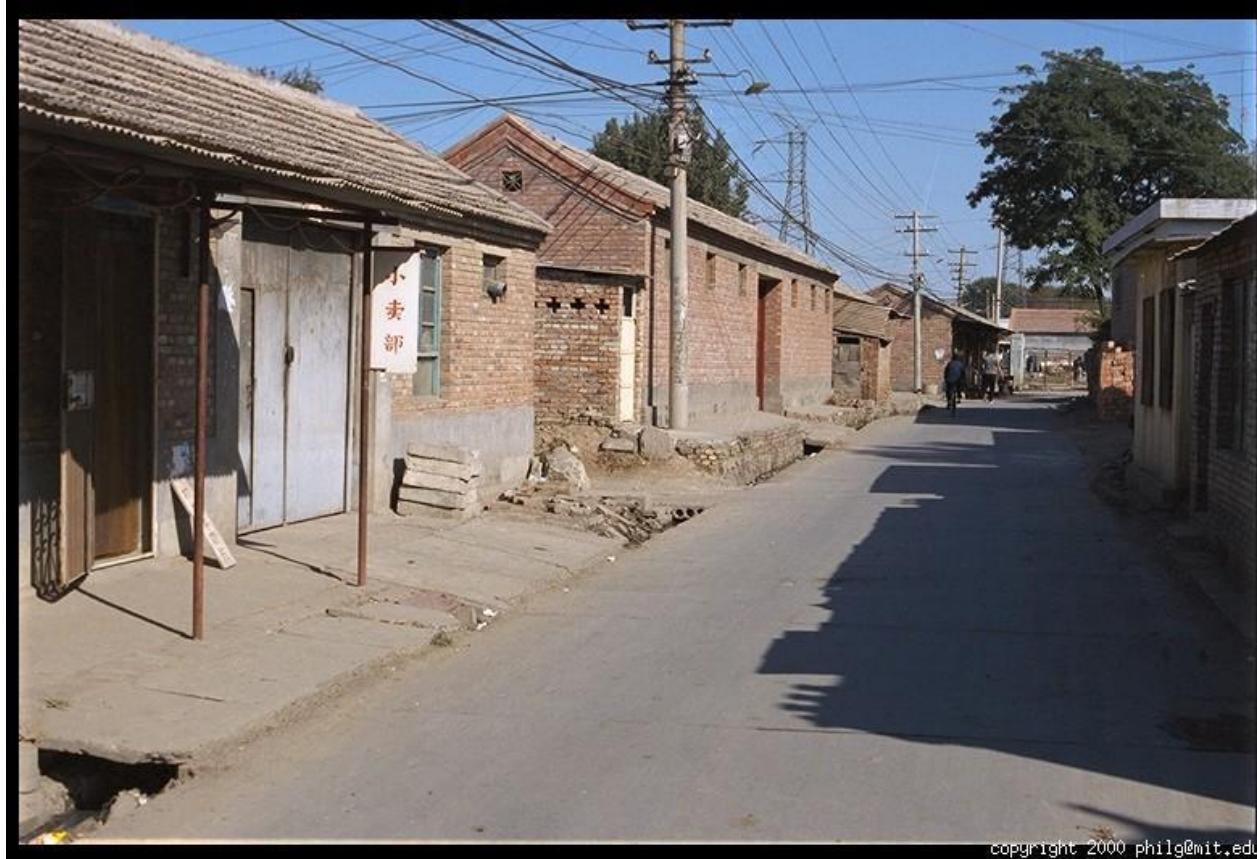


copyright 2000 philg@mit.edu

Only intensity shown – constant color

# Most information in intensity

---



copyright 2000 philg@mit.edu

Original image

# Back to grayscale intensity

---



0.92	0.93	0.94	0.97	0.62	0.37	0.85	0.97	0.93	0.92	0.99
0.95	0.89	0.82	0.89	0.56	0.31	0.75	0.92	0.81	0.95	0.91
0.89	0.72	0.51	0.55	0.51	0.42	0.57	0.41	0.49	0.91	0.92
0.96	0.95	0.88	0.94	0.56	0.46	0.91	0.87	0.90	0.97	0.95
0.71	0.81	0.81	0.87	0.57	0.37	0.80	0.88	0.89	0.79	0.85
0.49	0.62	0.60	0.58	0.50	0.60	0.58	0.50	0.61	0.45	0.33
0.86	0.84	0.74	0.58	0.51	0.39	0.73	0.92	0.91	0.49	0.74
0.96	0.67	0.54	0.85	0.48	0.37	0.88	0.90	0.94	0.82	0.93
0.69	0.49	0.56	0.66	0.43	0.42	0.77	0.73	0.71	0.90	0.99
0.79	0.73	0.90	0.67	0.33	0.61	0.69	0.79	0.73	0.93	0.97
0.91	0.94	0.89	0.49	0.41	0.78	0.78	0.77	0.89	0.99	0.93



# Next Lecture

---

Image Filtering - the core idea for project 1, and all of image processing.

Project 1 is much simpler than the remaining projects.