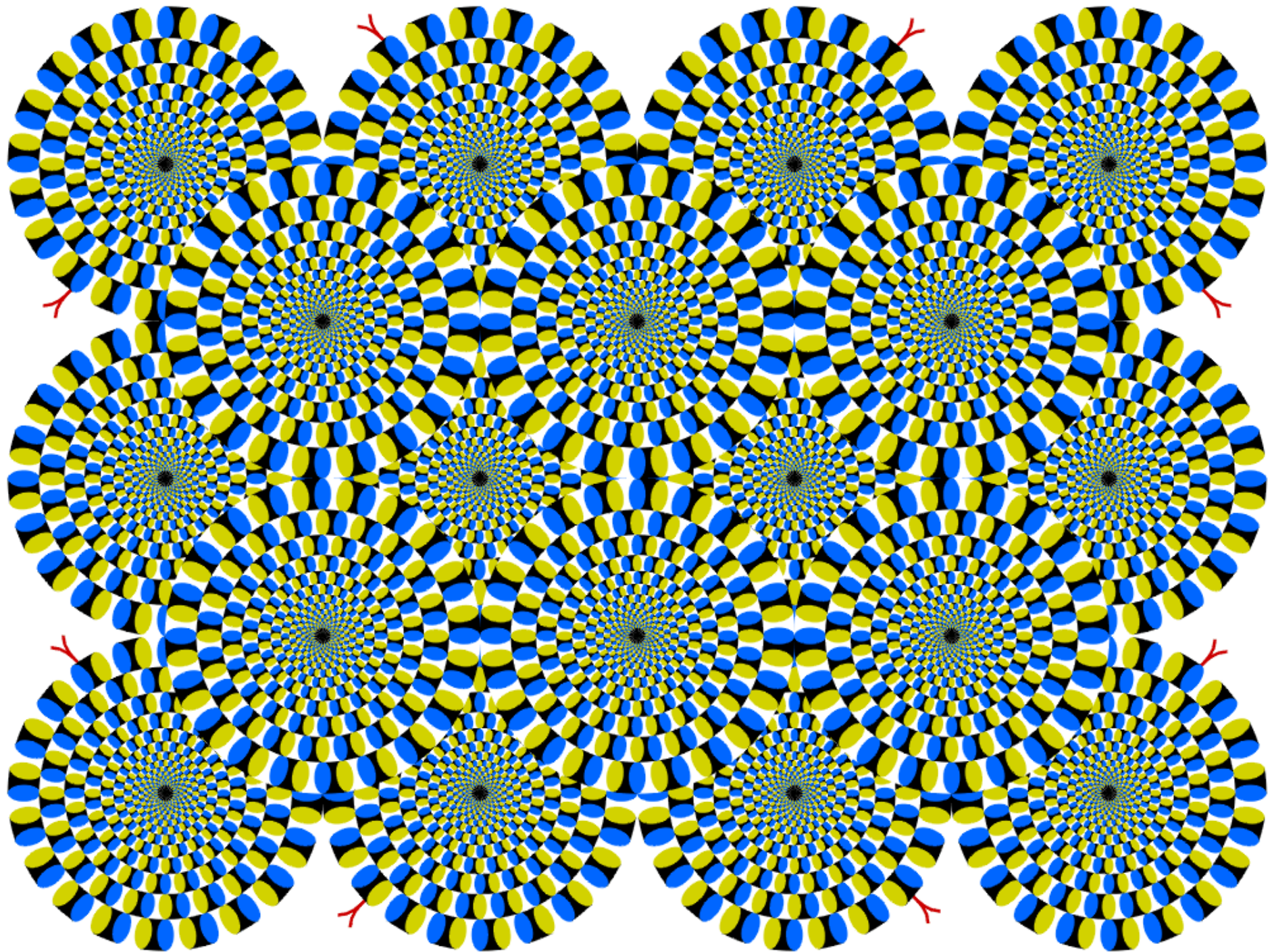
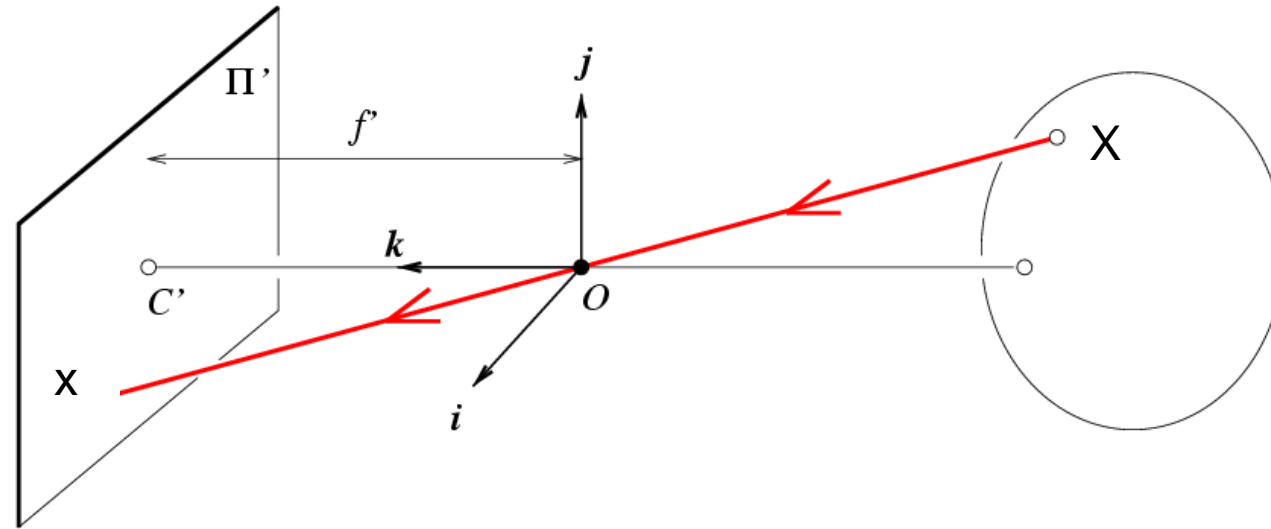


Moti



# Recap: projection



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



$$w \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \alpha & s & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

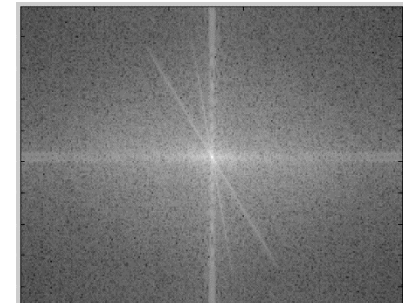
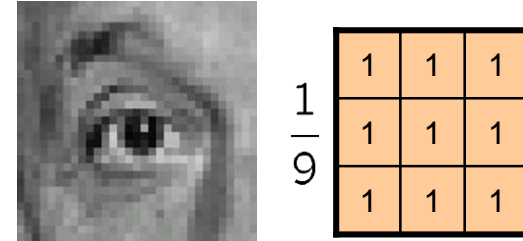
# Relating multiple views



Figure Credit: Bundler: Structure from Motion (SfM) for Unordered Image Collections

# Recap of Filtering

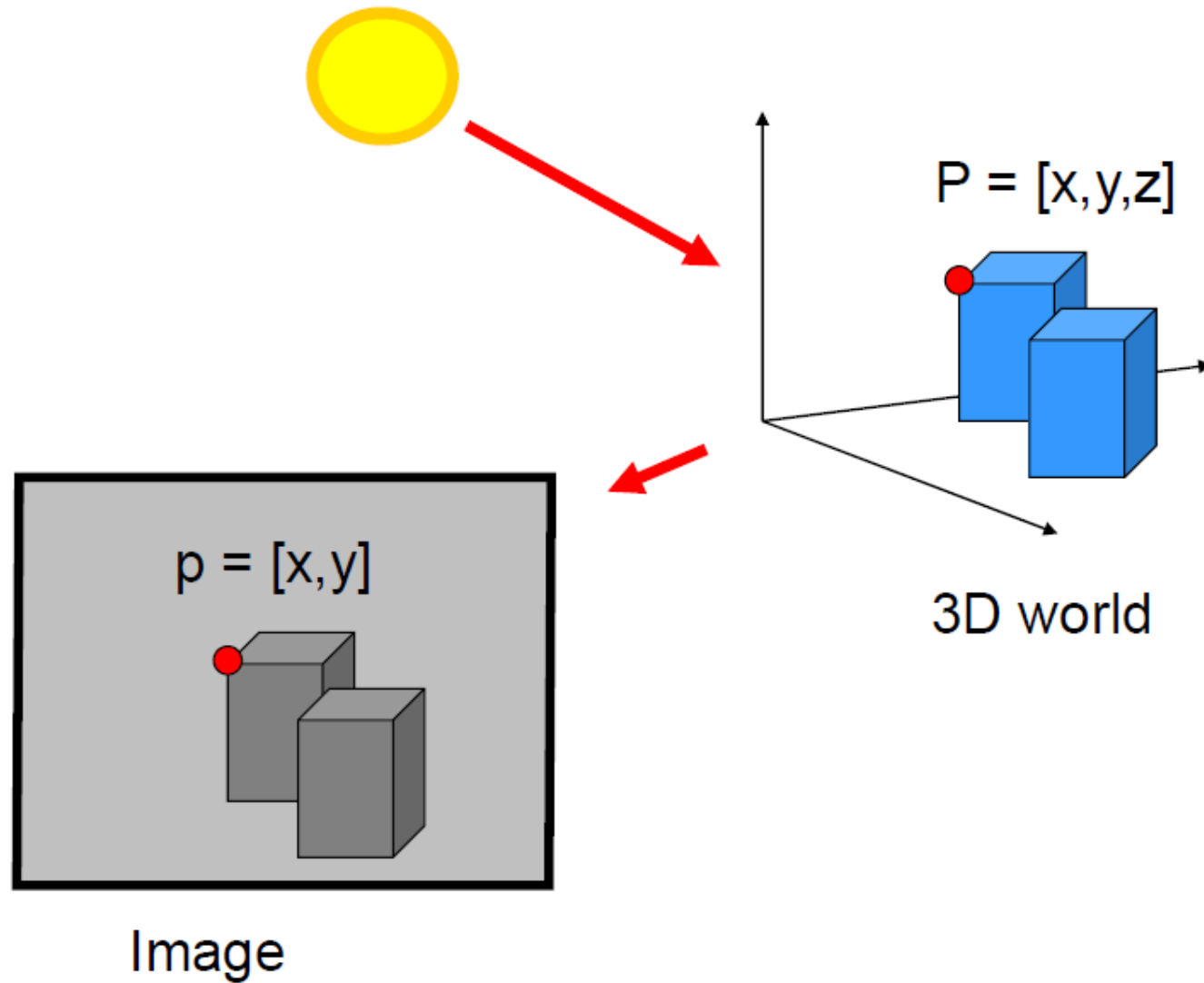
- Linear filtering is dot product at each position
  - Not a matrix multiplication
  - Can smooth, sharpen, translate (among many other uses)
- We can use the Fourier transform to represent images in the frequency domain.
  - Filtering in the spatial domain is multiplication in the frequency domain.



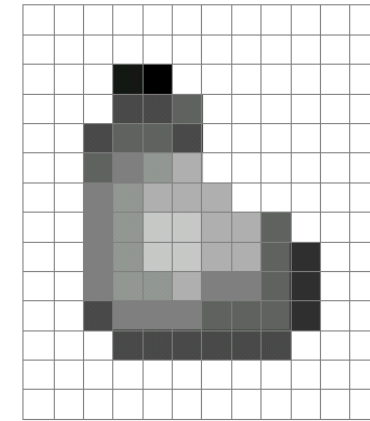
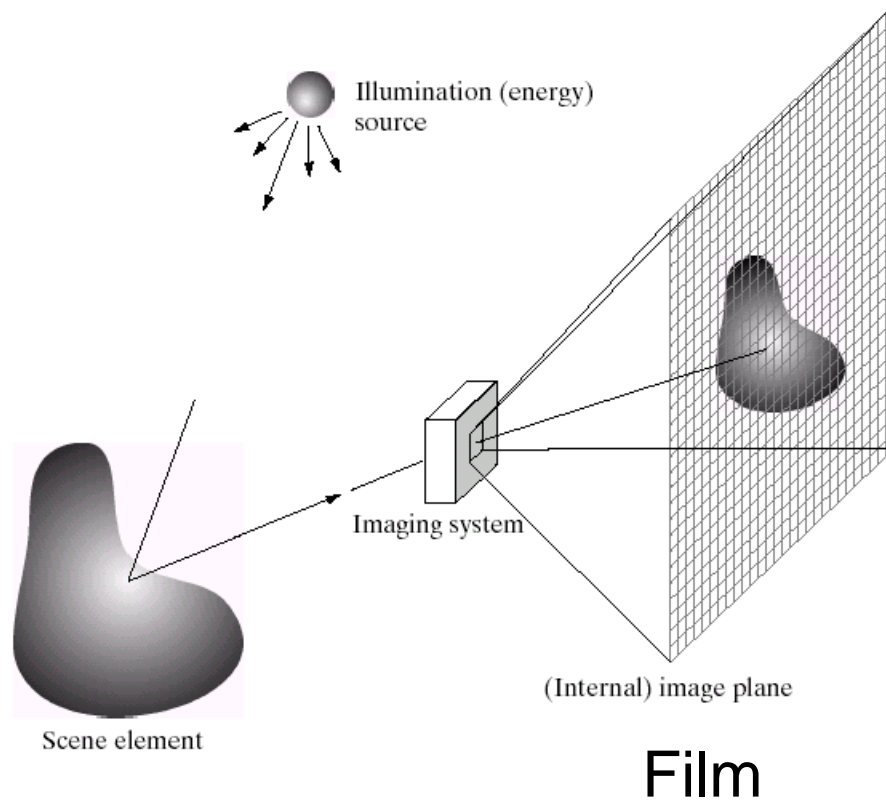
# This lecture

- Image Formation
- Biological Vision
- Light and Color

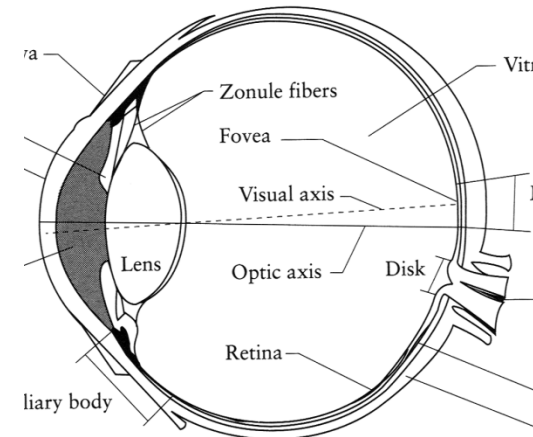
# From the 3D to 2D



# Image Formation



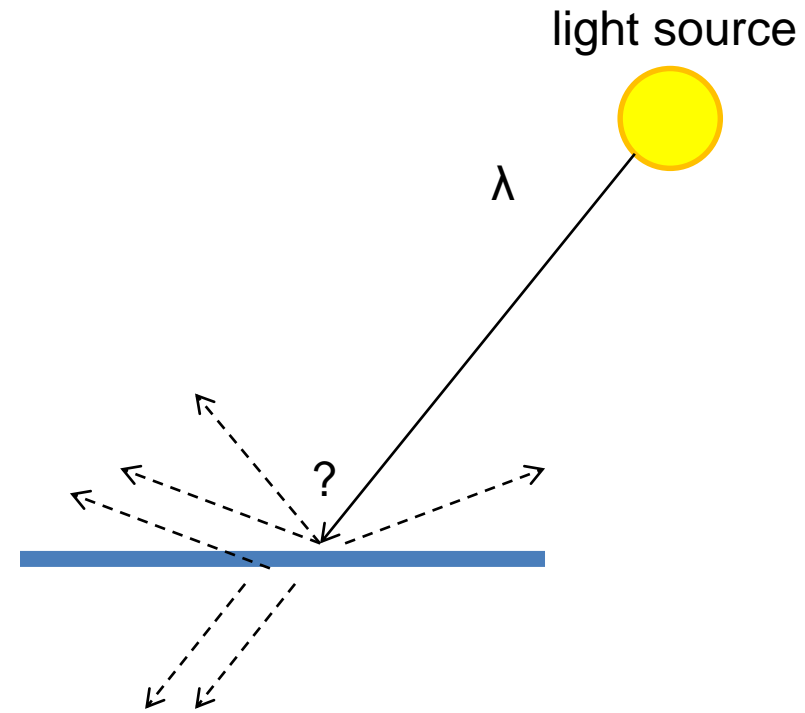
Digital Camera



The Eye

# A photon's life choices

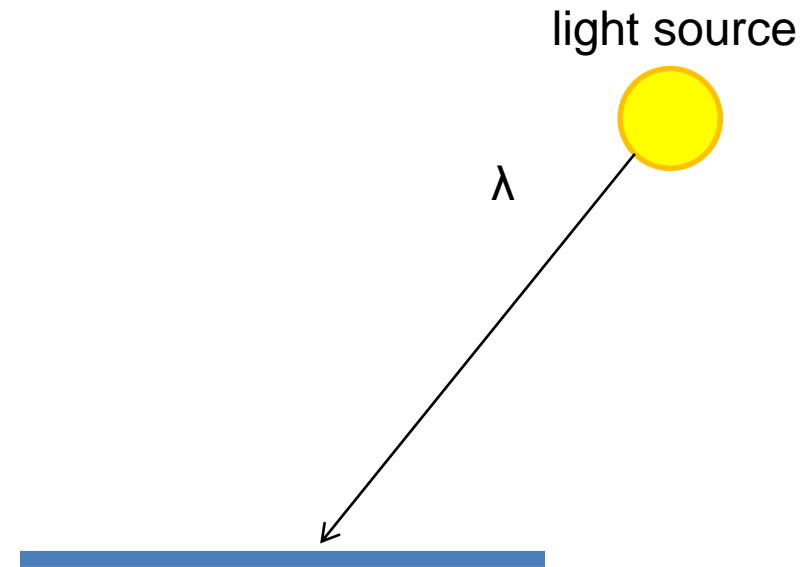
- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection





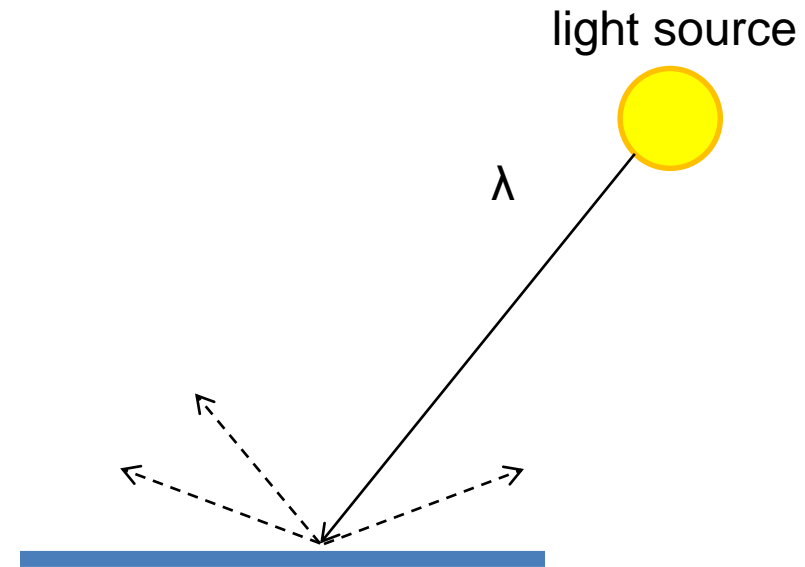
# A photon's life choices

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



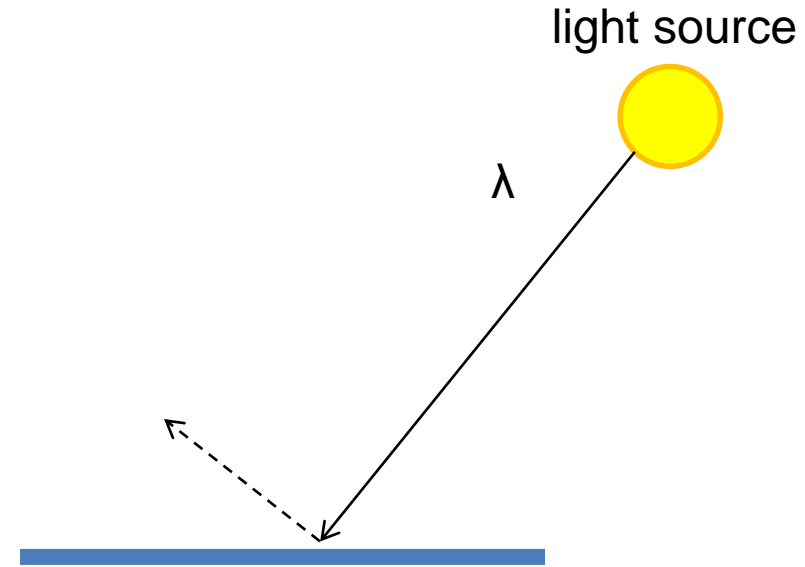
# A photon's life choices

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



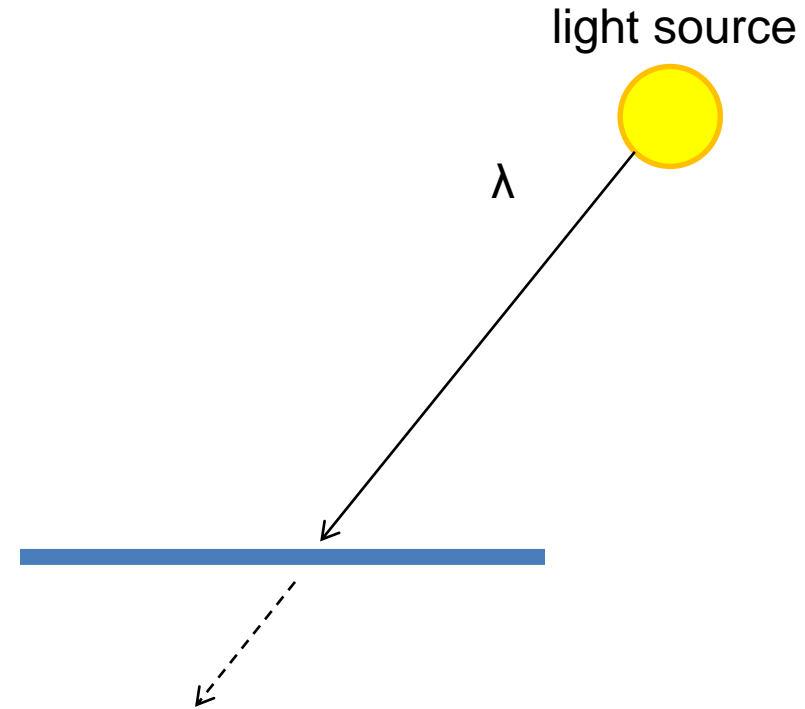
# A photon's life choices

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



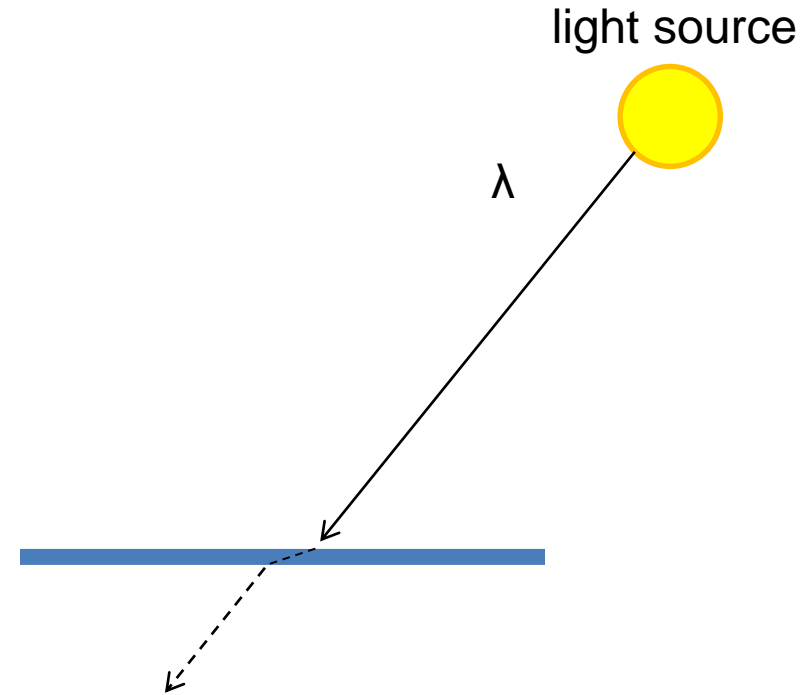
# A photon's life choices

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



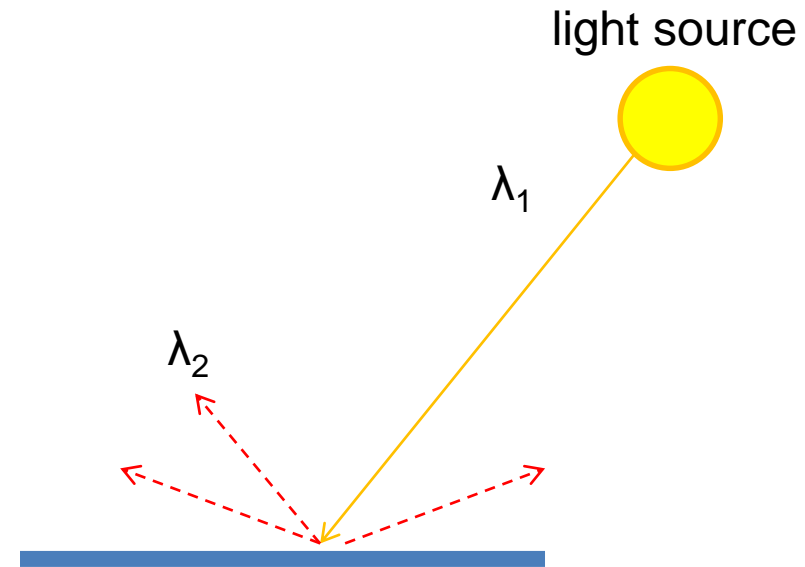
# A photon's life choices

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



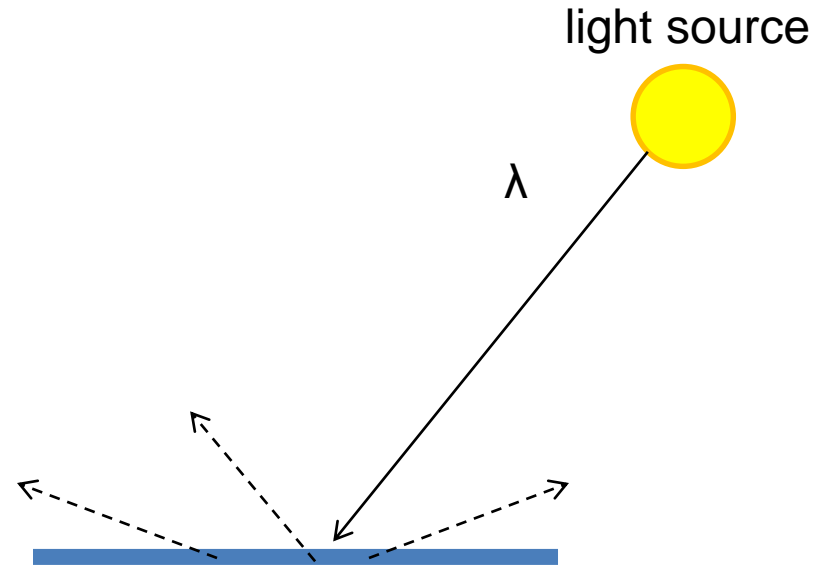
# A photon's life choices

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



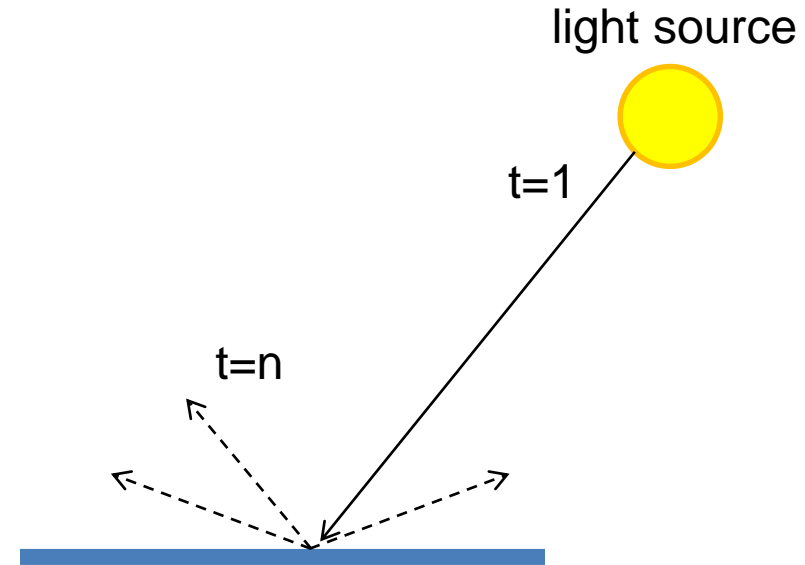
# A photon's life choices

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



# A photon's life choices

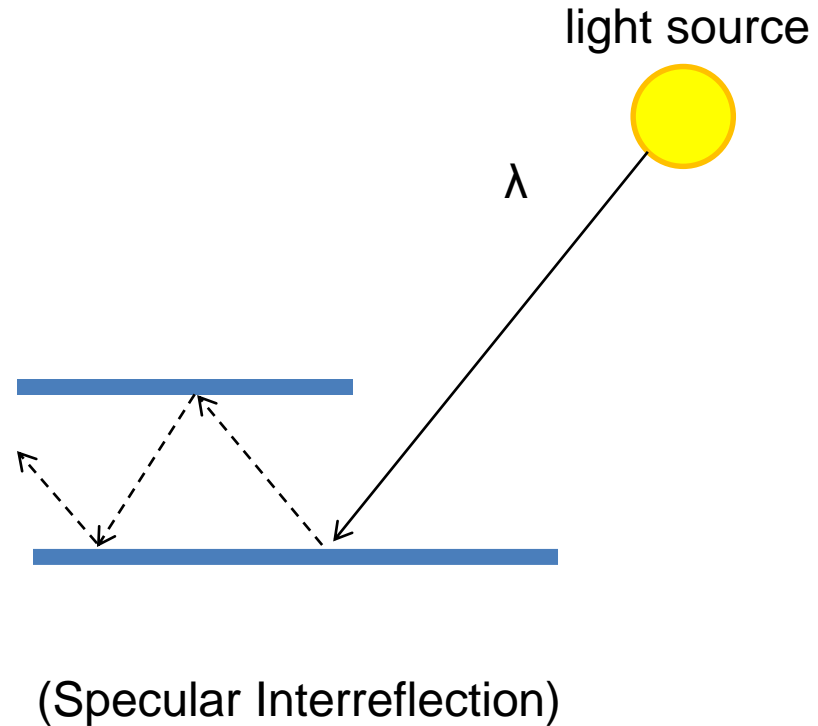
- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- **Phosphorescence**
- Interreflection





# A photon's life choices

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



# Lambertian Reflectance

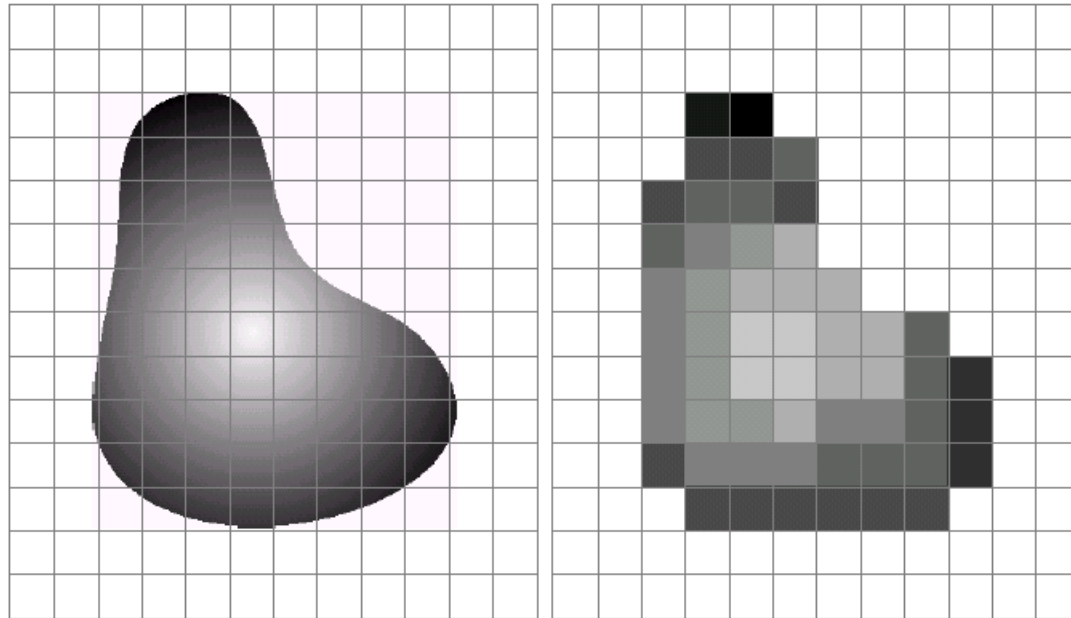
- In computer vision, the complexity of light transport is mostly ignored.
- Surfaces are often assumed to be ideal diffuse reflectors with no dependence on viewing direction.

# 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>

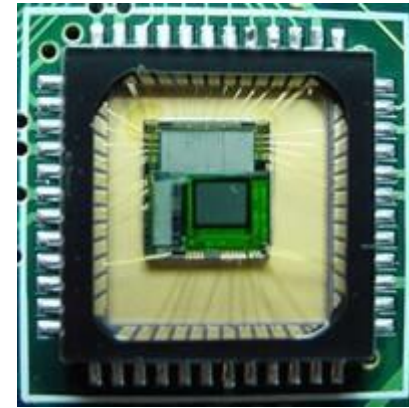
# Sensor Array



a b

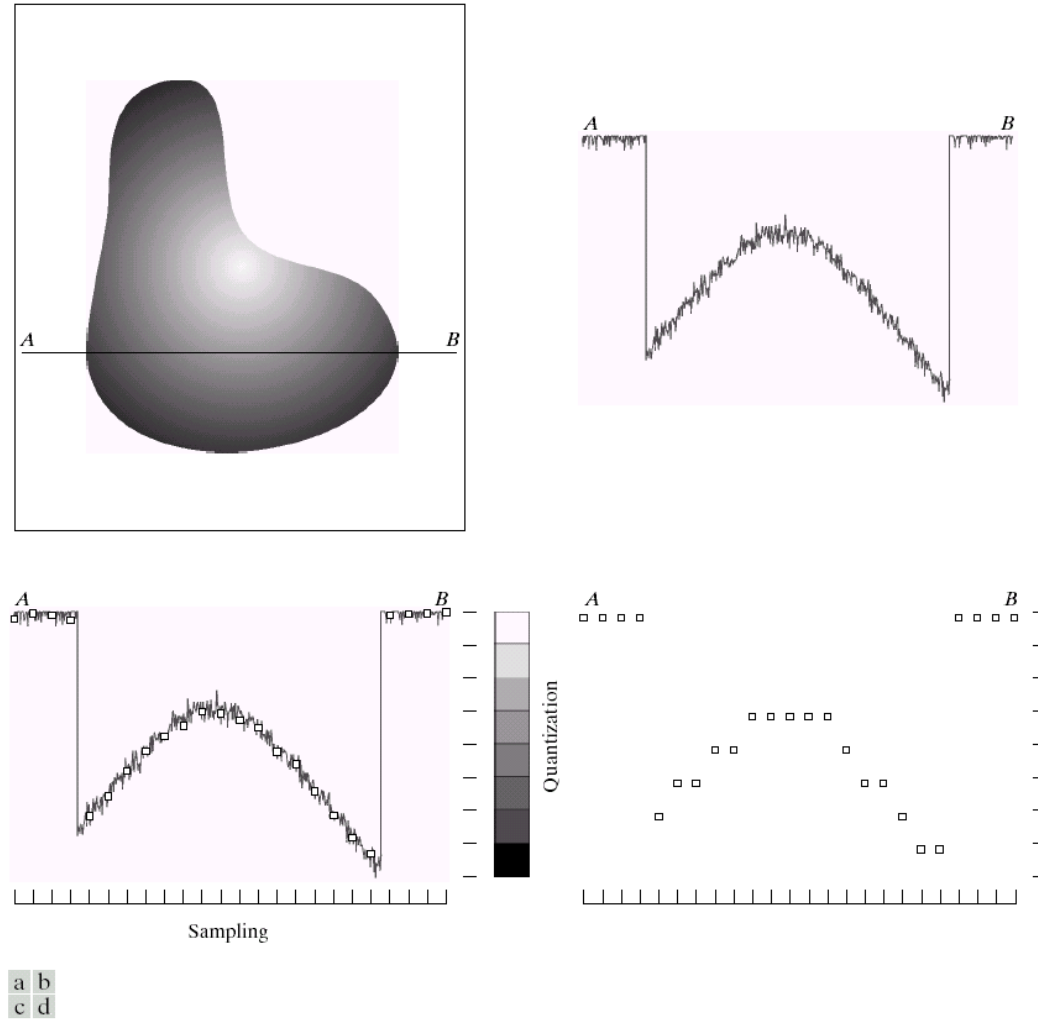
**FIGURE 2.17** (a) Continuous image projected onto a sensor array. (b) Result of image sampling and quantization.

---



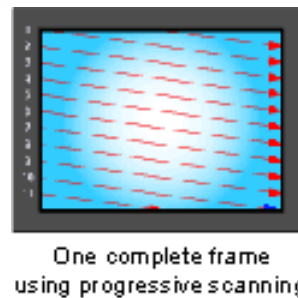
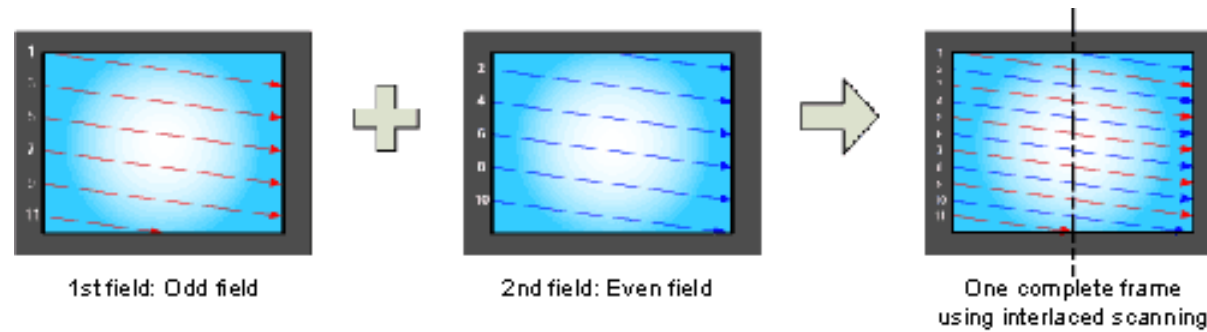
CMOS sensor

# 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



# Progressive scan or Global shutter

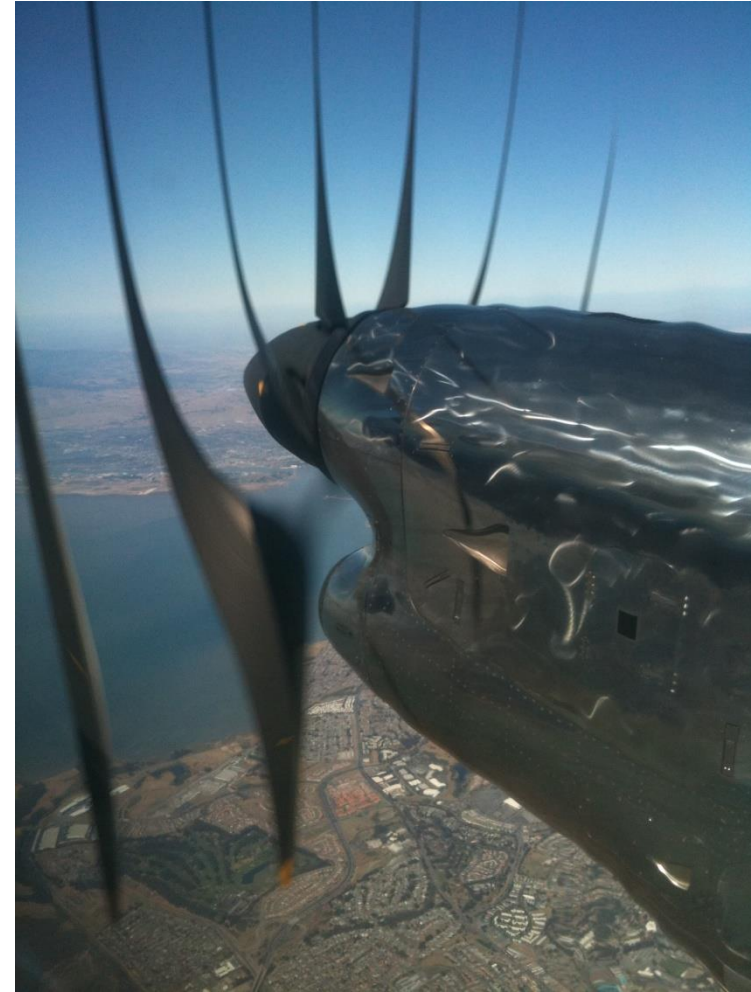


# Interlaced

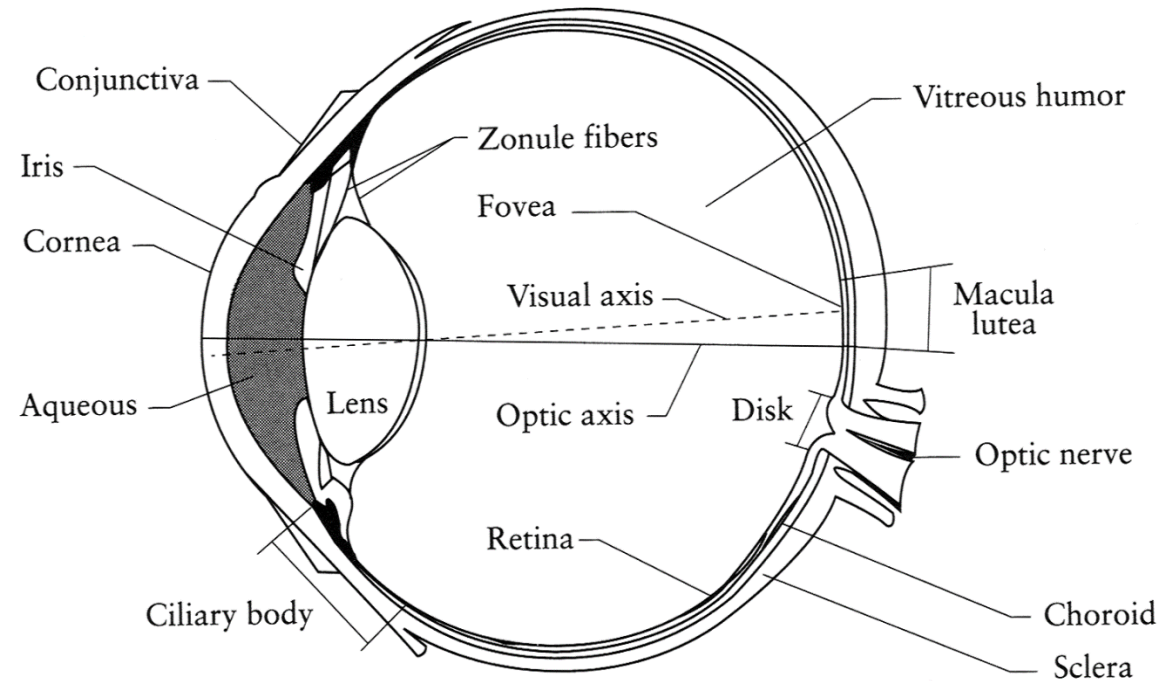




# Rolling Shutter



# The Eye

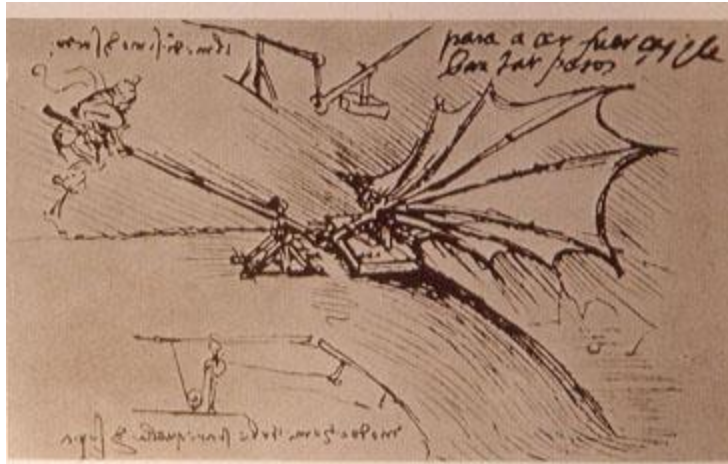


- The human eye is a camera!
  - **Iris** - colored annulus with radial muscles
  - **Pupil** - the hole (aperture) whose size is controlled by the iris
  - What's the “film”?
    - photoreceptor cells (rods and cones) in the **retina**

Aside: why do we care about human vision in this class?

- We don't, necessarily.

# Ornithopters



# Why do we care about human vision?

- We don't, necessarily.
- But cameras necessarily imitate the frequency response of the human eye, so we should know that much.
- Also, computer vision probably wouldn't get as much scrutiny if biological vision (especially human vision) hadn't proved that it was possible to make important judgements from 2d images.

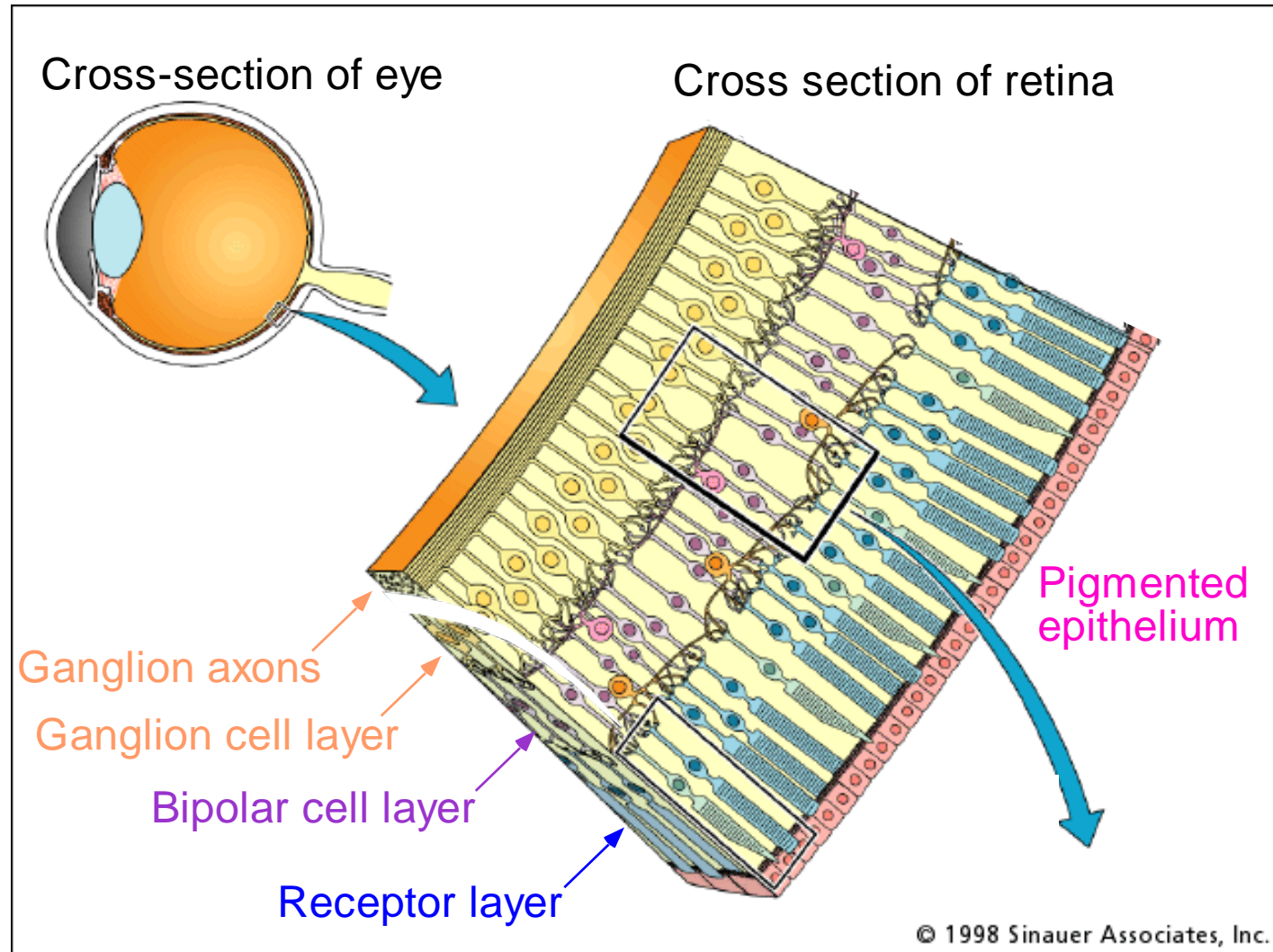
Does computer vision “understand” images?

"Can machines fly?" The answer is yes, because airplanes fly.

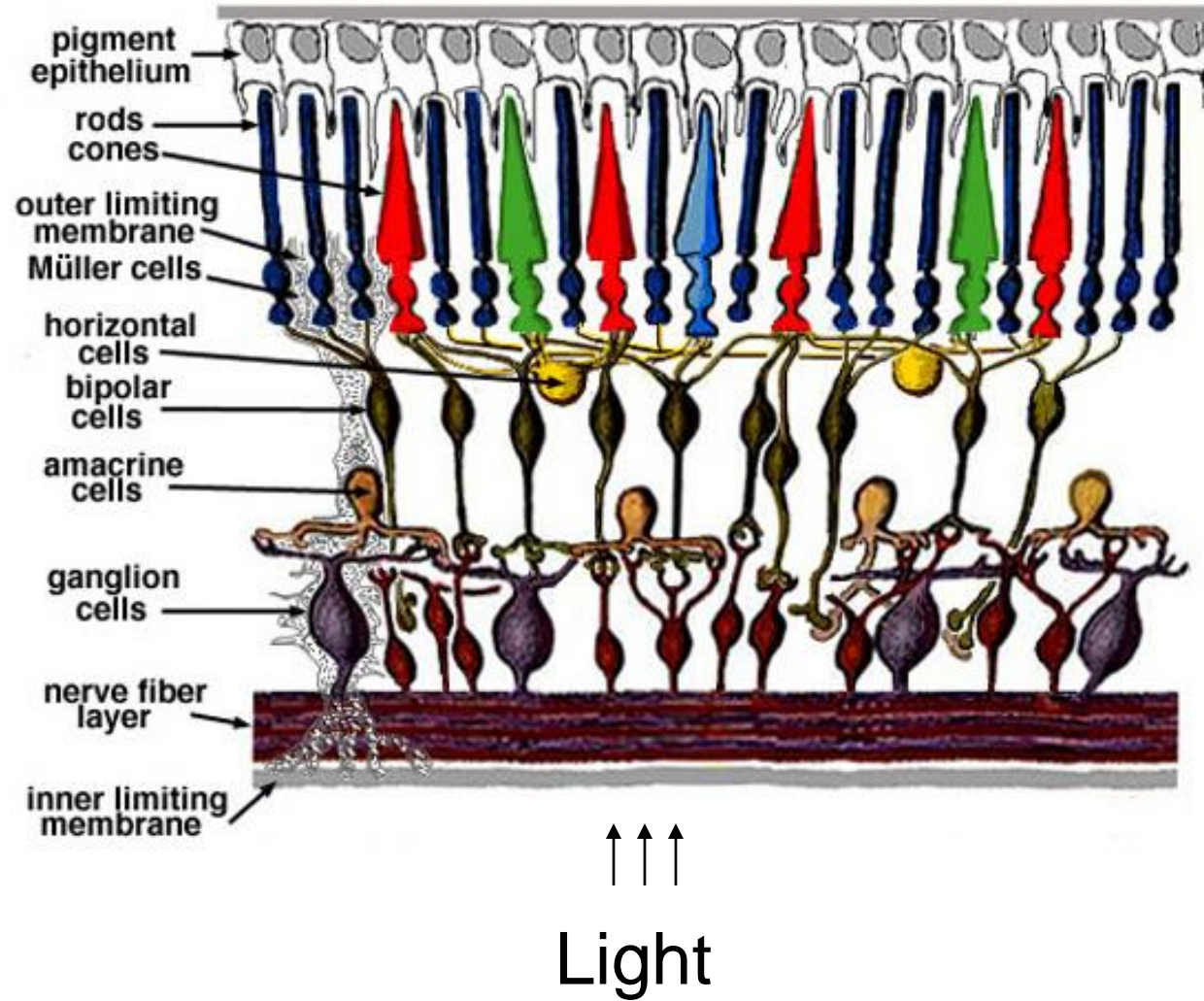
"Can machines swim?" The answer is no, because submarines don't swim.

"Can machines think?" Is this question like the first, or like the second?

# The Retina



# Retina up-close

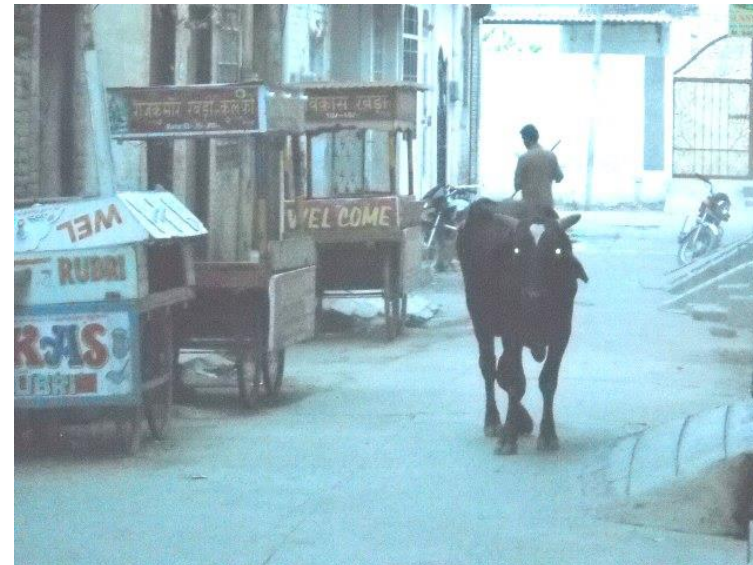




# What humans don't have: tapetum lucidum



Human eyes can reflect a tiny bit and blood in the retina makes this reflection red.



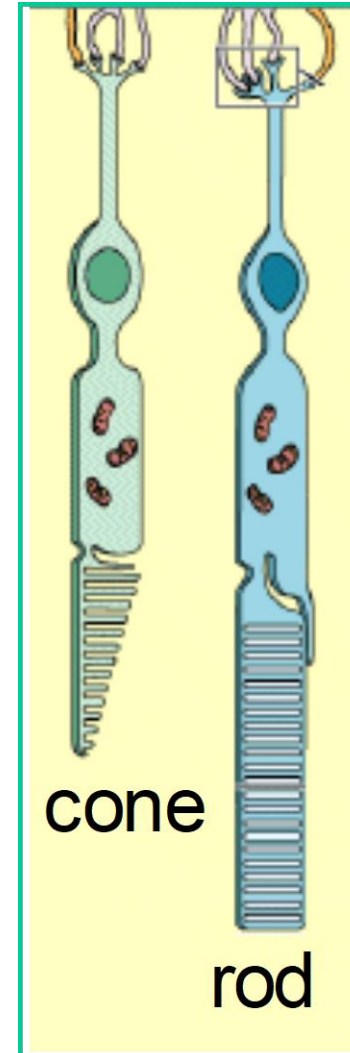
# Two types of light-sensitive receptors

## **C**ones

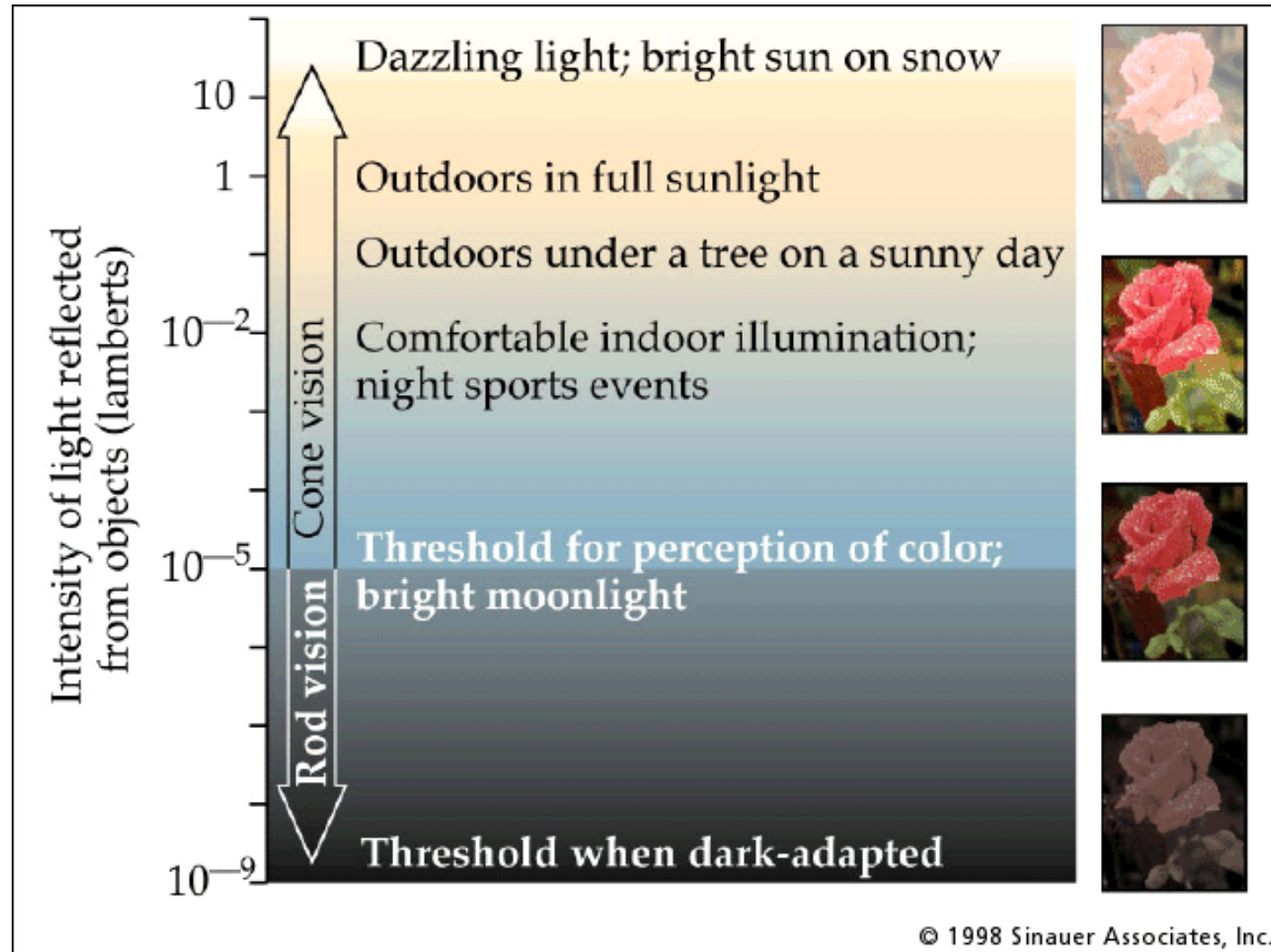
cone-shaped  
less sensitive  
operate in high light  
color vision

## **R**ods

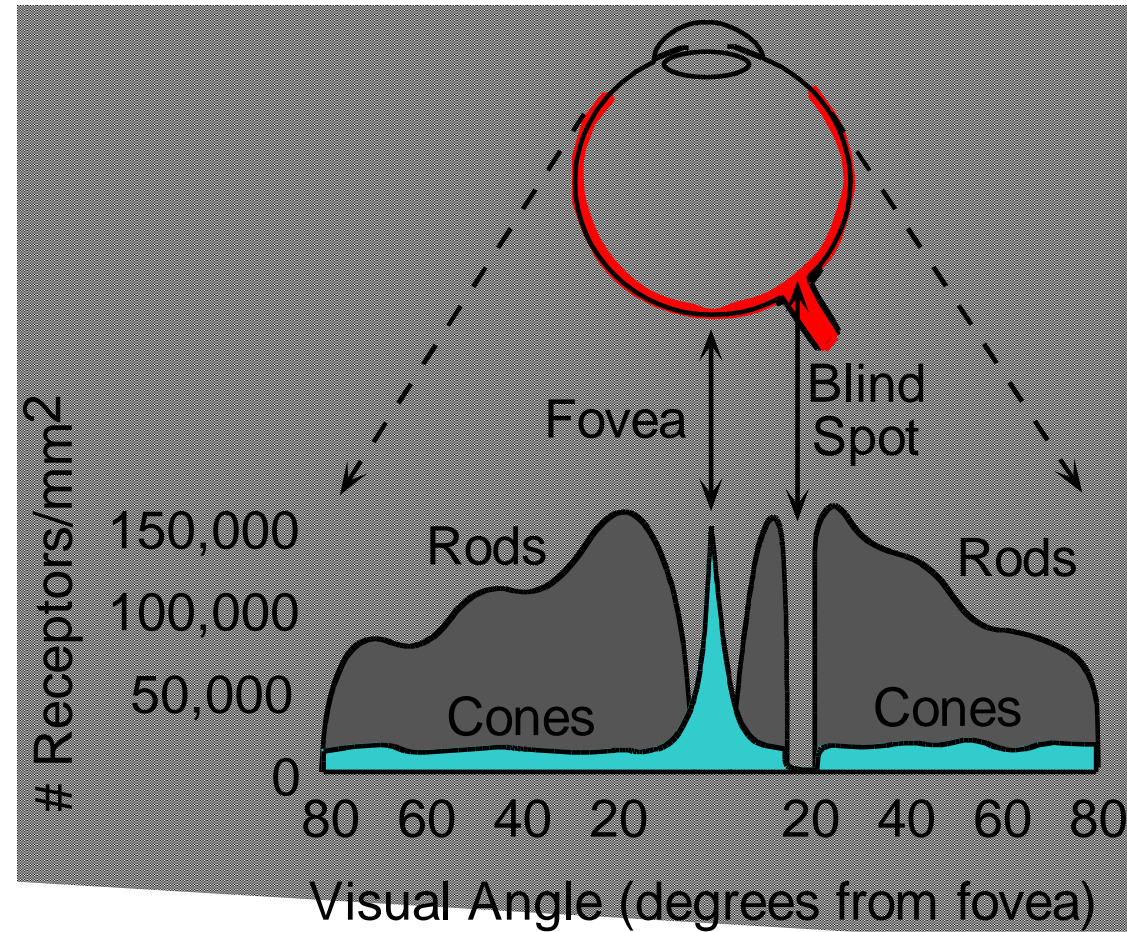
rod-shaped  
highly sensitive  
operate at night  
gray-scale vision



# Rod / Cone sensitivity



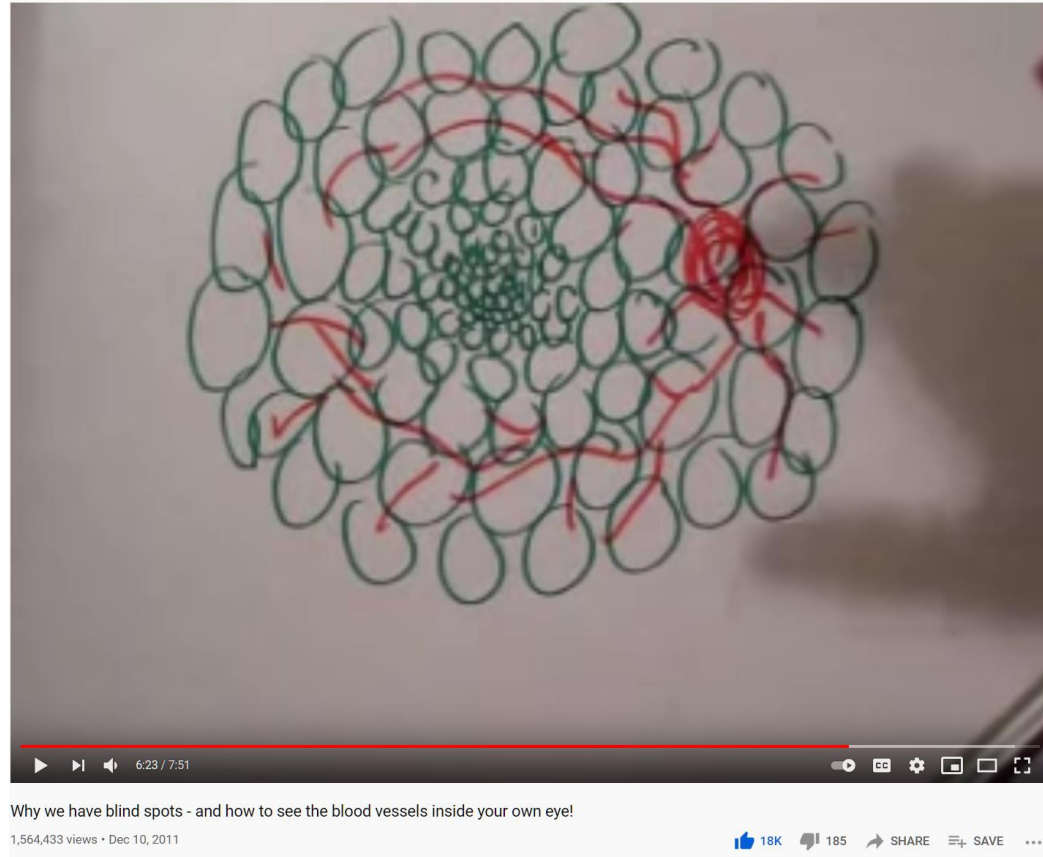
# Distribution of Rods and Cones



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)

Wait, the blood vessels are in front of the photoreceptors??



[https://www.youtube.com/watch?v=L\\_W-IXqoxHA](https://www.youtube.com/watch?v=L_W-IXqoxHA)



# 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.
- Smooth pursuit – tracking an object

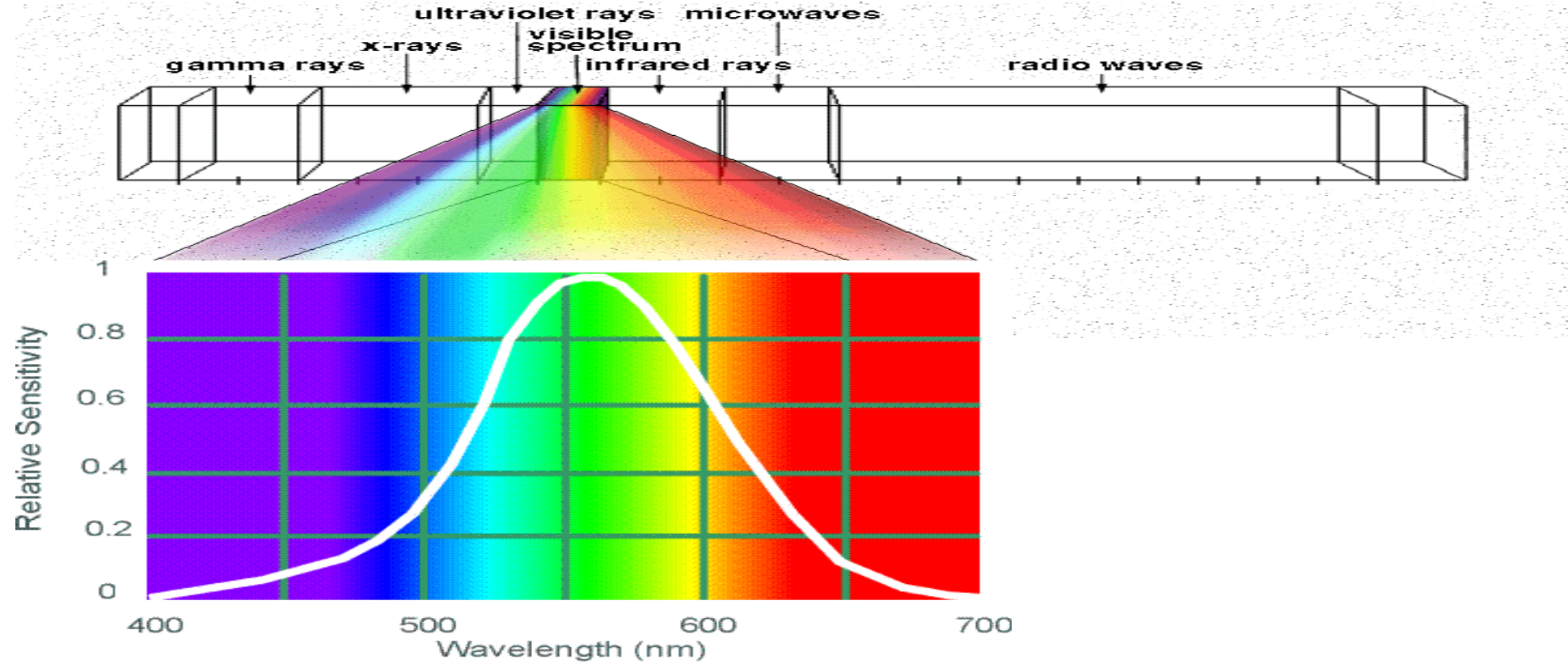
# Intermission

## Slow mo guys – Saccades and CRTs

- <https://youtu.be/Fmg9ZOHESgQ?t=21s>
- <https://youtu.be/3BJU2drirtCM>



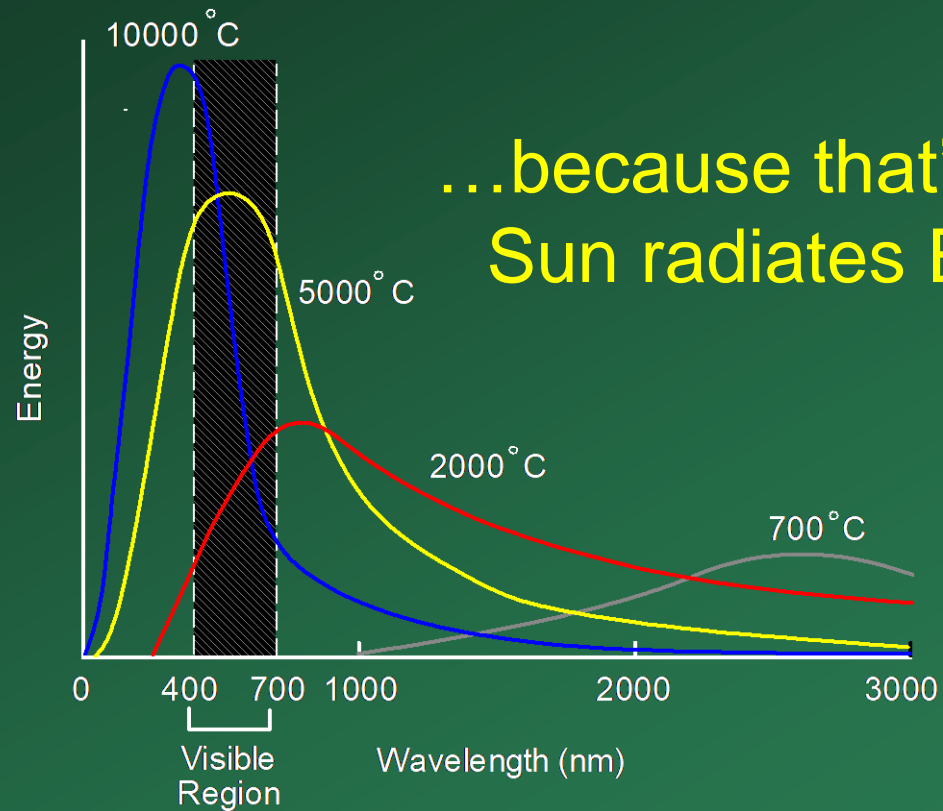
# Electromagnetic Spectrum



Human Luminance Sensitivity Function

# Visible Light

Why do we see light of these wavelengths?

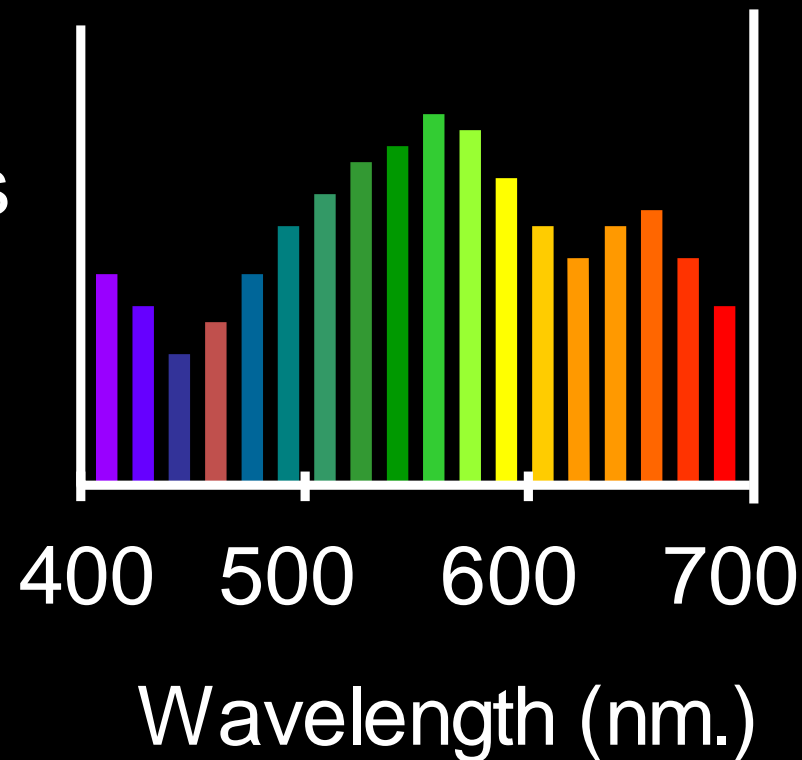


...because that's where the Sun radiates EM energy

# 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.

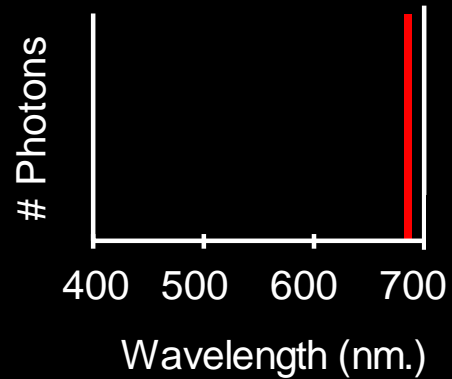
# Photons  
(per ms.)



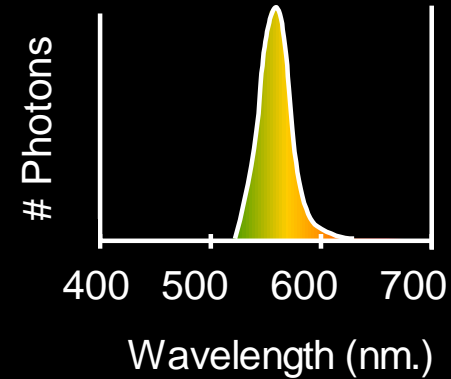
# The Physics of Light

## Some examples of the spectra of light sources

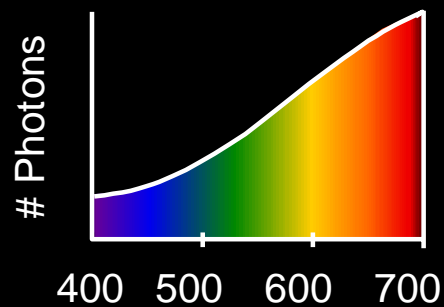
A. Ruby Laser



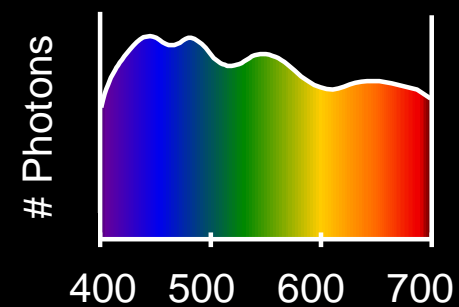
B. Gallium Phosphide Crystal



C. Tungsten Lightbulb

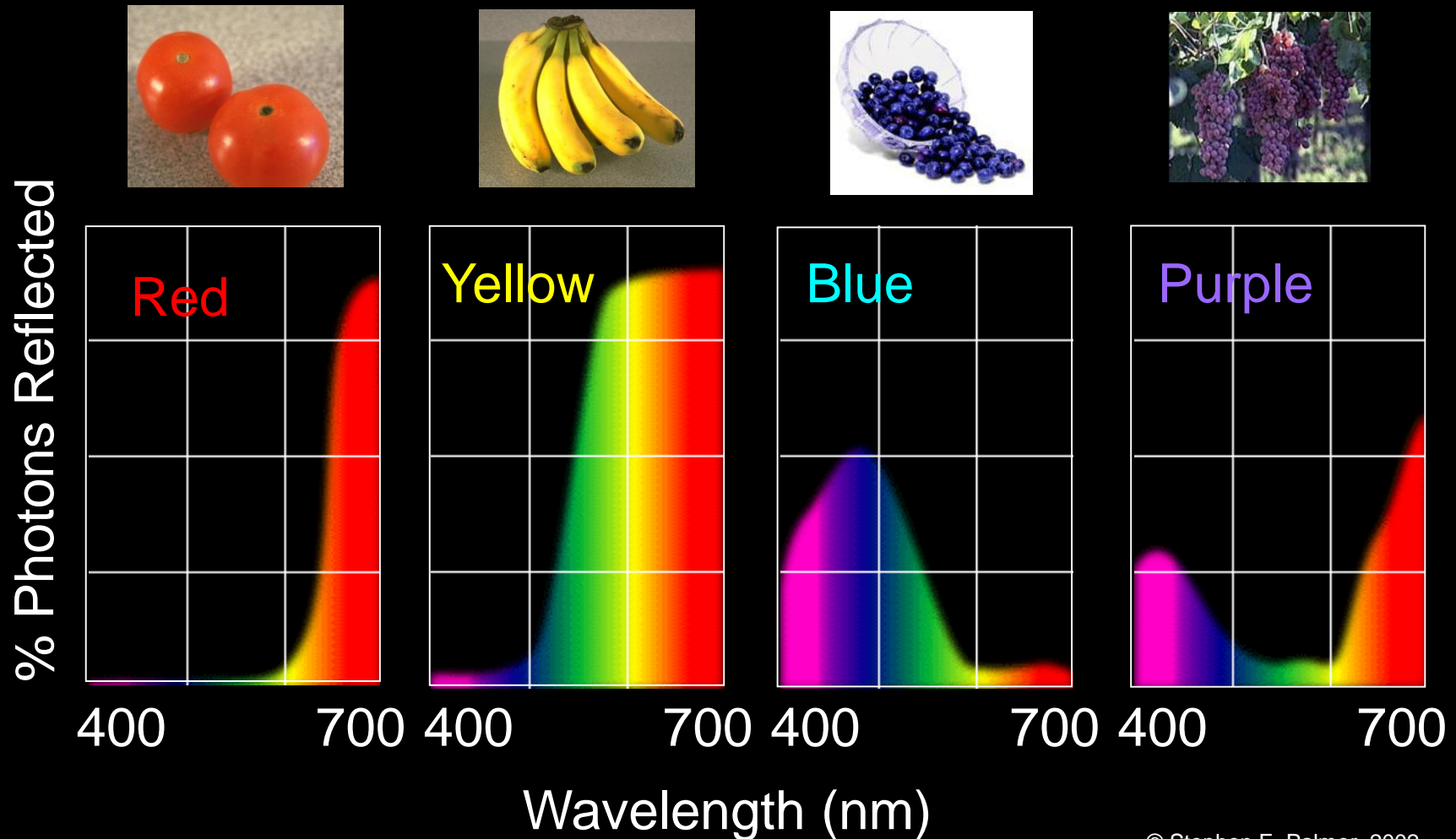


D. Normal Daylight



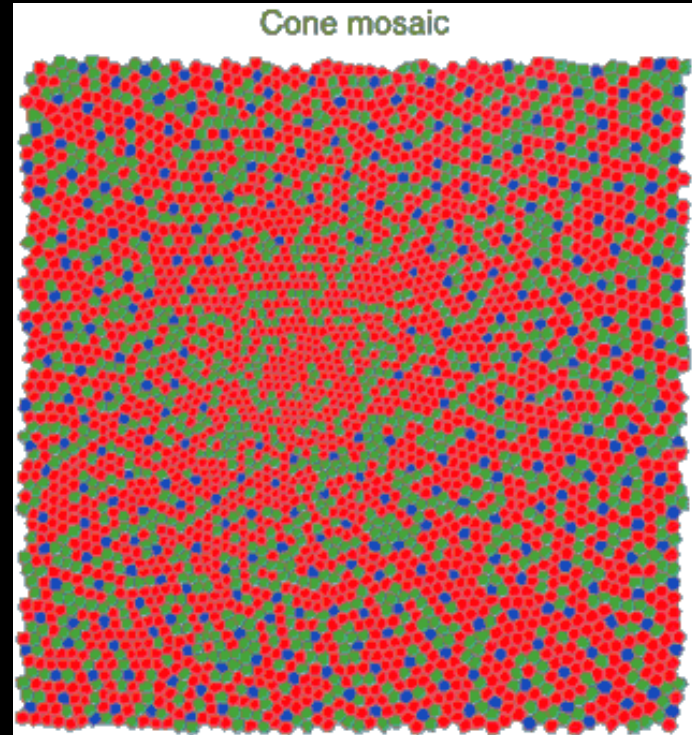
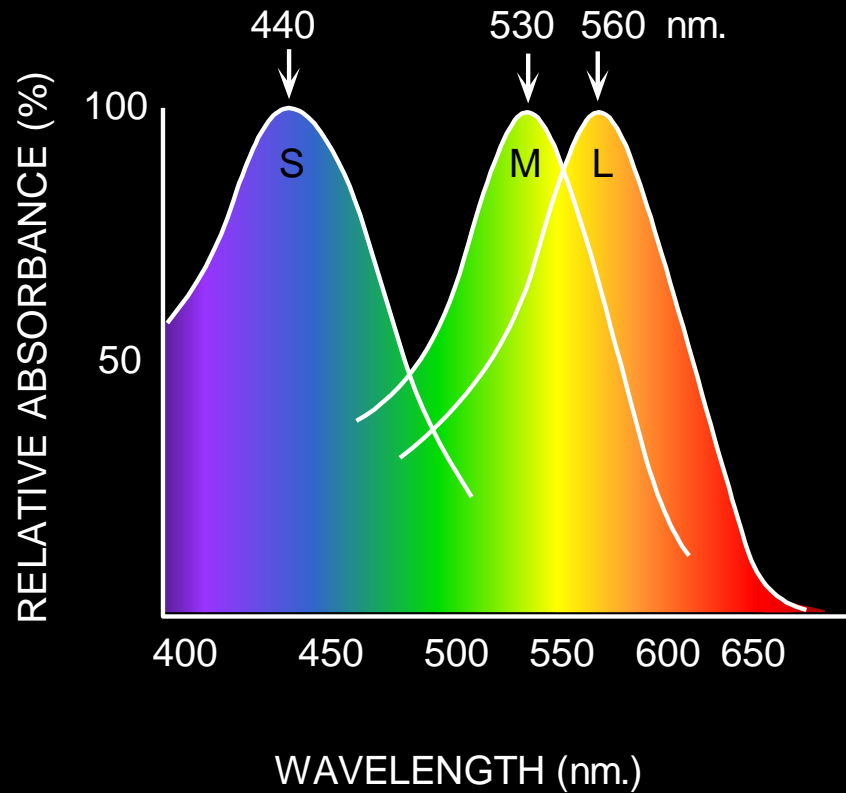
# The Physics of Light

Some examples of the reflectance spectra of surfaces



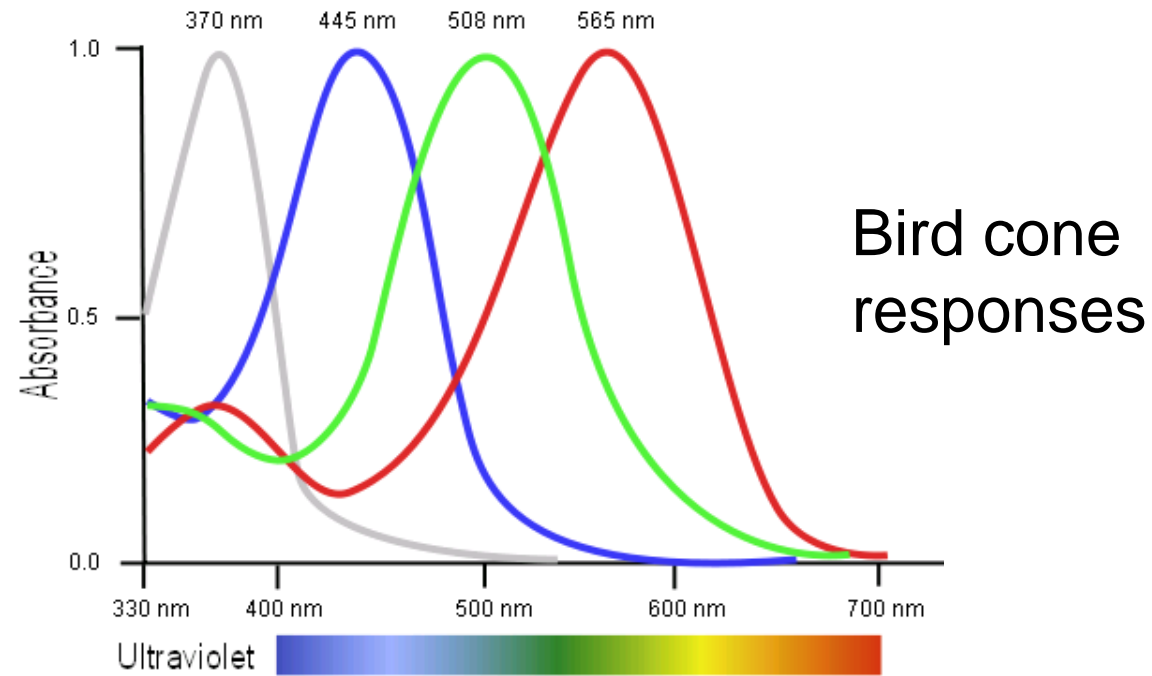
# Physiology of Color Vision

Three kinds of cones:



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

# Tetrachromacy



- Most birds, and many other animals, have cones for ultraviolet light.
- Some humans, mostly female, seem to have slight tetrachromatism.

Table 1.

Cone pigment complements in some eutherian mammals.

order	exemplars	SWS1 pigment <sup>a</sup>	LWS pigment <sup>a</sup>	reference	
Rodentia	<i>Mus</i> (mouse)	UV	M	<a href="#">Jacobs <i>et al.</i> (1991)</a>	
	<i>Rattus</i> (rat)	UV	M	<a href="#">Jacobs <i>et al.</i> (1991)</a>	
	<i>Geomys</i> (gopher)	UV	M	<a href="#">Williams <i>et al.</i> (2005)</a>	
	<i>Cavia</i> (guinea pig)	S	M	<a href="#">Parry &amp; Bowmaker (2002)</a>	
	<i>Spermophilus</i> (squirrel)	S	M	<a href="#">Jacobs <i>et al.</i> (1985)</a>	
	<i>Cricetomys</i> (African rat)	absent	ML <sup>b</sup>	<a href="#">Peichl &amp; Moutairou (1998)</a>	
Lagomorpha	<i>Oryctolagus</i> (rabbit)	S	M	<a href="#">Nuboer <i>et al.</i> (1983)</a>	
primate	<i>Macaca</i> (macaque monkey)	S	M+L	<a href="#">Schnapf <i>et al.</i> (1988)</a>	
	<i>Saimiri</i> (squirrel monkey)	S	poly (3)	<a href="#">Mollon <i>et al.</i> (1984)</a>	
	<i>Aotus</i> (owl monkey)	absent	L	<a href="#">Jacobs <i>et al.</i> (1993b)</a>	
	<i>Alouatta</i> (howler monkey)	S	M+L	<a href="#">Jacobs <i>et al.</i> (1996a)</a>	
	<i>Galago</i> (bushbaby)	absent	L	<a href="#">Deegan II &amp; Jacobs (1996)</a>	
	<i>Lemur</i> (ring-tailed)	S	L	<a href="#">Jacobs &amp; Deegan II (1993)</a>	
	<i>Propithecus</i> (sifaka)	S	poly (2)	<a href="#">Tan &amp; Li (1999)</a>	
	Scandentia	<i>Tupaia</i> (tree shrew)	S	L	<a href="#">Jacobs &amp; Neitz (1986)</a>
	Cetacea	<i>Eschrichtius</i> (whale)	absent	L	<a href="#">Levenson &amp; Dizon (2003)</a>
		<i>Tursiops</i> (dolphin)	absent	L	<a href="#">Fasick <i>et al.</i> (1998)</a>
Artiodactyla	<i>Bos</i> (cow)	S	L	<a href="#">Jacobs <i>et al.</i> (1994)</a>	
	<i>Odocoileus</i> (deer)	S	M	<a href="#">Jacobs <i>et al.</i> (1994)</a>	
	<i>Sus</i> (pig)	S	L	<a href="#">Neitz &amp; Jacobs (1989)</a>	
Perissodactyla	<i>Equus</i> (horse)	S	L	<a href="#">Carroll <i>et al.</i> (2001)</a>	
Carnivora	<i>Felis</i> (cat)	S	L	<a href="#">Loop <i>et al.</i> (1987)</a>	
	<i>Canis</i> (dog)	S	L	<a href="#">Jacobs <i>et al.</i> (1993a)</a>	
	<i>Mustela</i> (ferret)	S	L	<a href="#">Calderone &amp; Jacobs (2003)</a>	
	<i>Ursus</i> (bear)	S	L	<a href="#">Levenson <i>et al.</i> (2006)</a>	
	<i>Crocuta</i> (hyena)	UV/S	ML	<a href="#">Calderone <i>et al.</i> (2003)</a>	
	<i>Phoca</i> (seal)	absent	L	<a href="#">Levenson <i>et al.</i> (2006)</a>	
	<i>Enhydra</i> (otter)	S	L	<a href="#">Levenson <i>et al.</i> (2006)</a>	

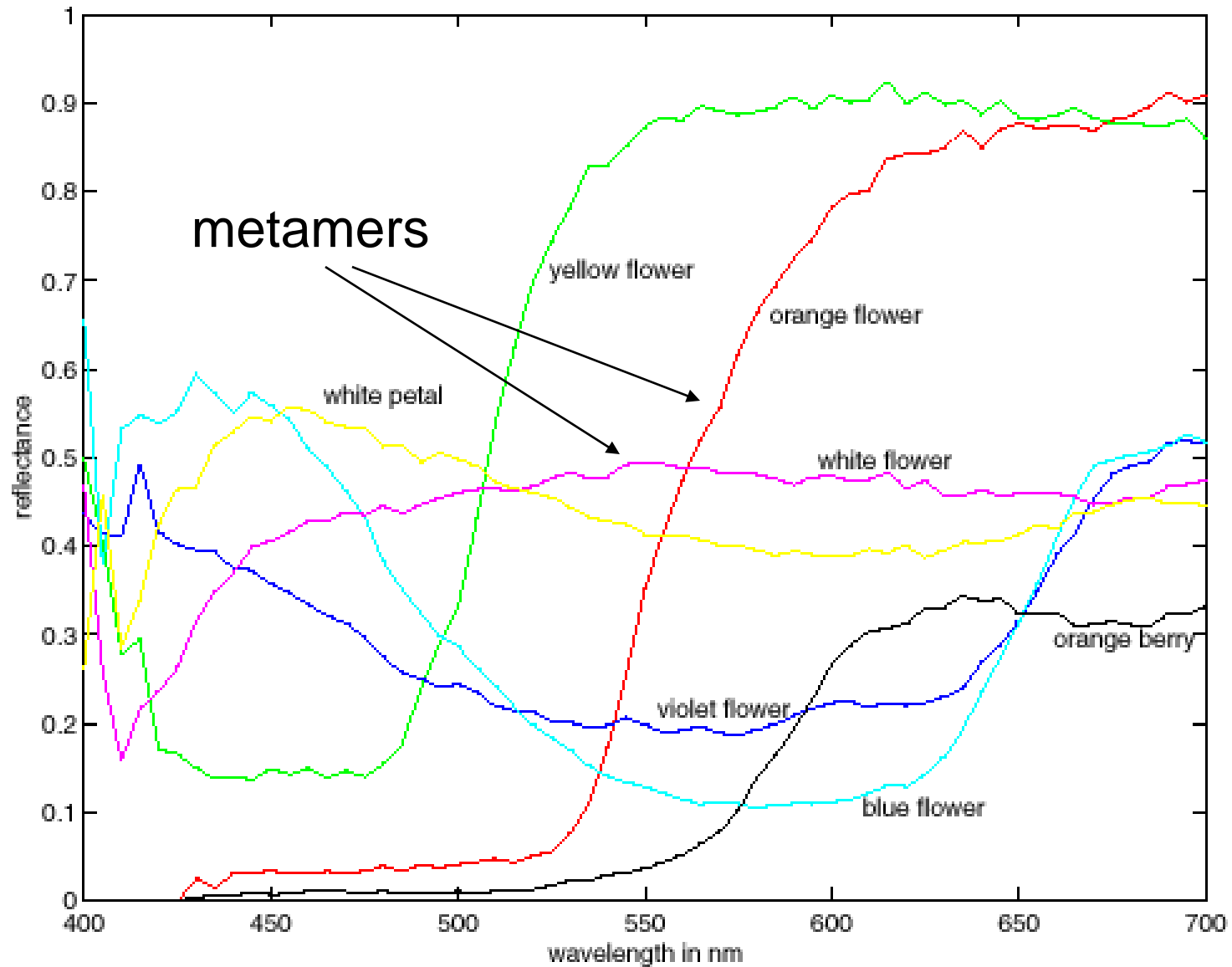
### (b) Eutherians

Representatives from two cone opsin gene families appear in contemporary eutherian mammals and, with the exception of some primates, none of these animals derive more than a single photopigment type from each of their two gene families (*SWS1* and *LWS*; [figure 1](#)). Given that commonality, what cone pigments may have been present in the retinas of the early eutherians? Sequence comparisons of cone opsin genes have suggested that the ancestral mammalian pigments drawn from these two gene families peaked in the UV, at about 360 nm ([Hunt \*et al.\* 2001](#)), and in the long wavelengths at 560 nm ([Yokoyama \*et al.\* 2008](#)). Assuming the visual pigment chromophore of early mammals was 11-*cis*-retinal, the same as that of contemporary mammals, the cone pigments of early eutherian mammals would have had absorption properties similar to those sketched at the bottom of [figure 1](#).

If these deductions are correct, the retinas of the early eutherians were similar to those of the majority of contemporary mammals in containing two types of cone pigment. Such an arrangement could support dichromatic colour vision. Whether it did would additionally depend on there having been at least some degree of selective expression of the two pigment types in separate receptor classes, on these early mammals having a nervous system organized to allow a contrast of signals from the two types of cone, and on them at least occasionally encountering photic environments sufficiently intense to activate neural comparison circuits. The fact that this basic two cone pigment arrangement is largely conserved among contemporary mammals strongly attests to its adaptive utility in our eutherian ancestors and to its probable role in supporting dichromatic colour vision in these early mammals.



# More Spectra



Color can be ambiguous



Color can be ambiguous



# The dress



From Wikipedia, the free encyclopedia

*For other uses, see [The Dress](#).*

**The dress** is a photograph that became a [viral](#) internet sensation on 26 February 2015, when viewers disagreed over whether the dress pictured was coloured black and blue, or white and gold. The phenomenon revealed differences in human colour perception, which have been the subject of ongoing scientific investigations into [neuroscience](#) and [vision science](#), producing a number of papers published in peer-reviewed science journals.

The photo originated from a washed-out colour photograph of a [dress](#) posted on the [social networking service Facebook](#). Within the first week after the surfacing of the image, more than 10 million tweets mentioned the dress, using [hashtags](#) such as [#thedress](#), [#whiteandgold](#), and [#blackandblue](#). Although the colour of the dress was eventually confirmed as black and blue,<sup>[3][4]</sup> the image prompted many discussions, with users discussing their different perceptions of the dress's colour. Members of the scientific community began to investigate the photo for fresh insights into human [colour vision](#).

The dress itself, which was identified as a product of the retailer Roman Originals, experienced a major surge in sales as a result of the incident. The retailer also produced a one-off version of the dress in white and gold as a charity campaign.

**Contents** [\[hide\]](#)

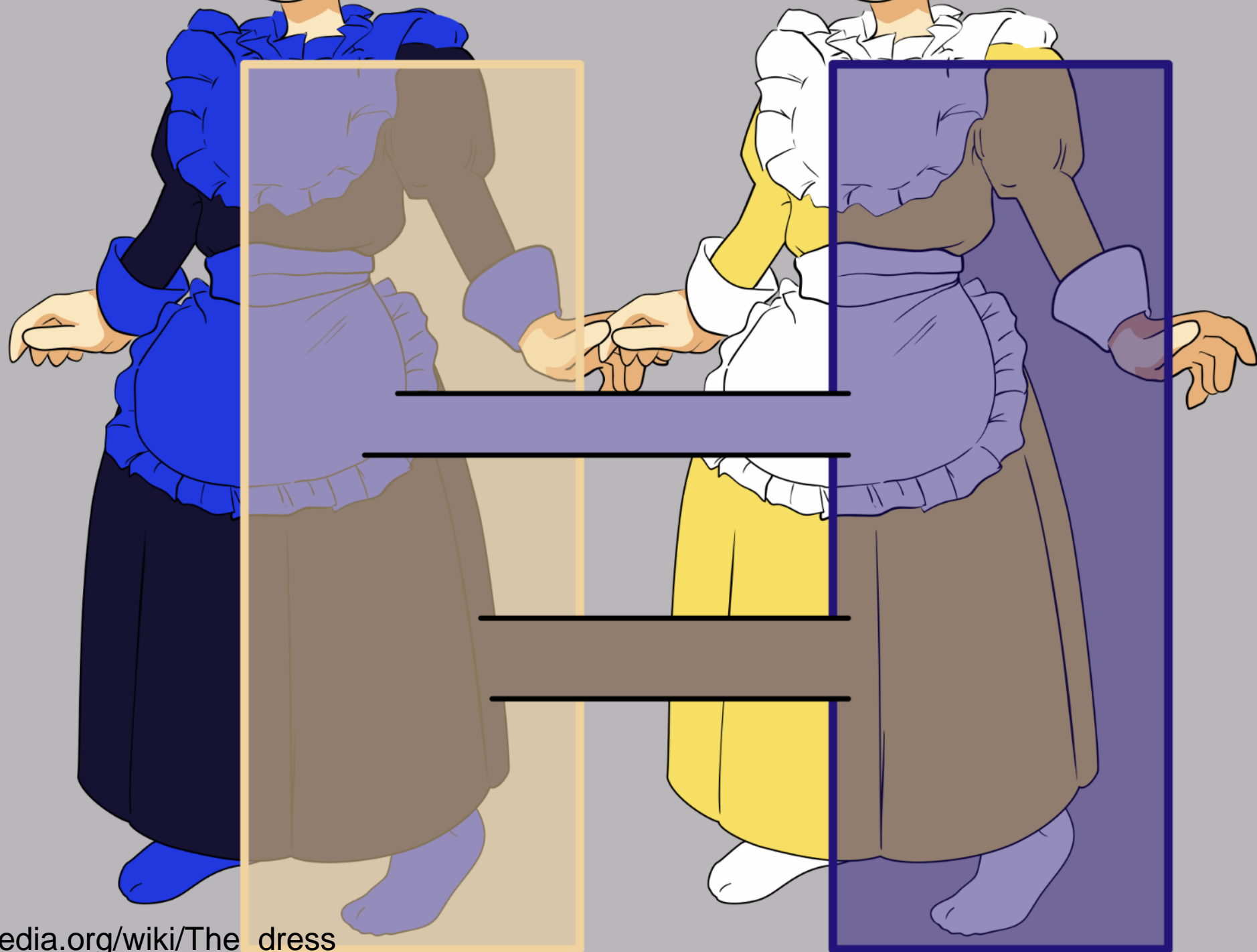
**The dress**



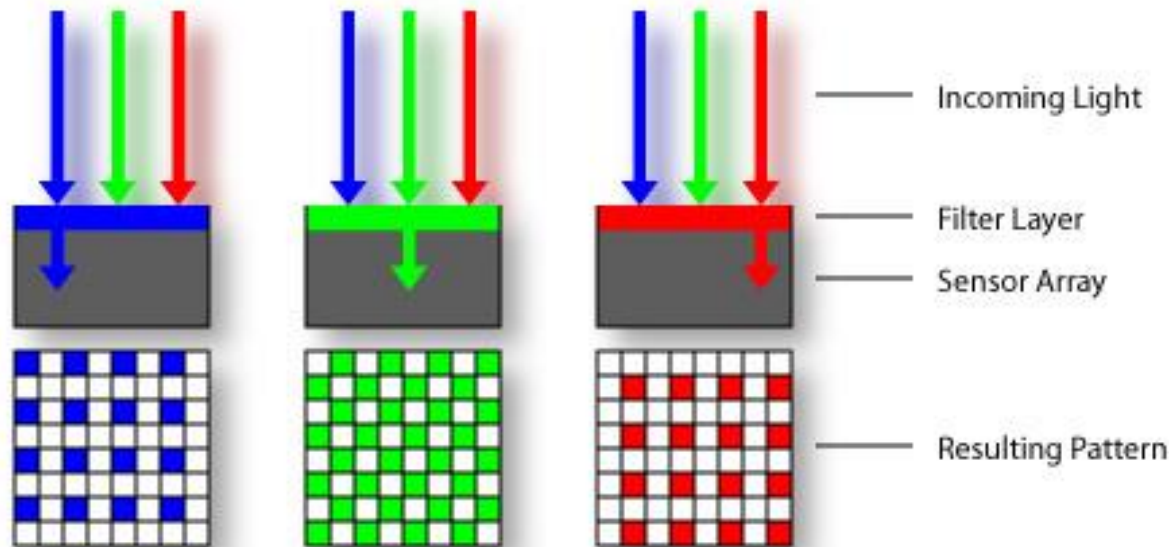
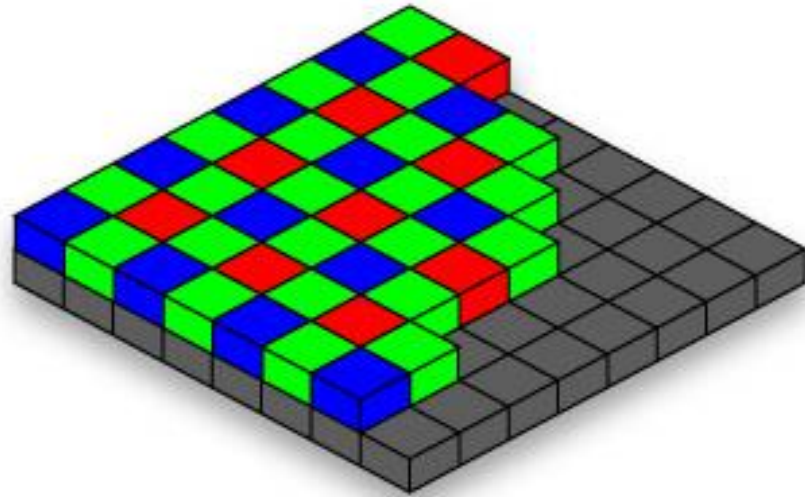
The original *The dress* picture

**Designer**

Roman Originals<sup>[1]</sup>

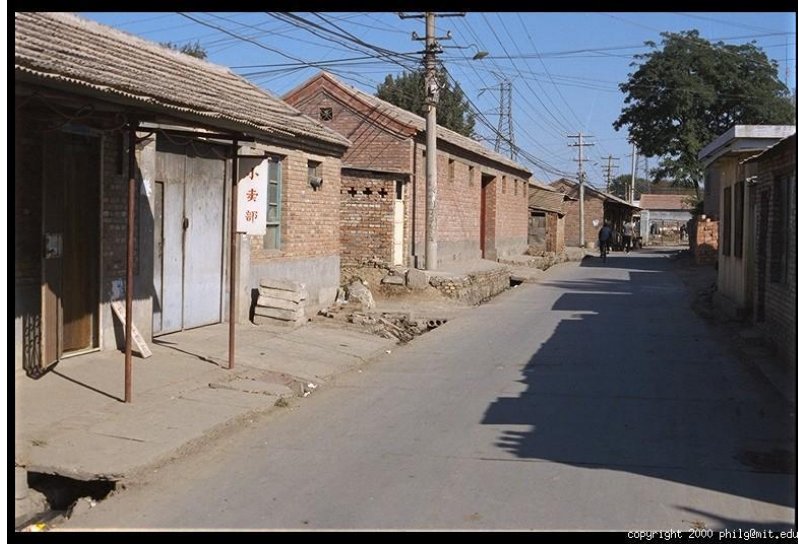


# Practical Color Sensing: Bayer Grid



- Estimate RGB at 'G' cells from neighboring values

# Color Image



# 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

The diagram illustrates an RGB image matrix. A large 10x10 grid of numerical values represents the image. A blue arrow labeled 'row' points downwards on the left side, and another blue arrow labeled 'column' points to the right above the grid. To the right of the grid, three smaller 2x2 grids are shown, labeled 'R', 'G', and 'B', representing the Red, Green, and Blue channels respectively. The 'R' channel is a 10x10 grid, the 'G' channel is a 10x2 grid, and the 'B' channel is a 10x2 grid. The values in the 'R' channel are the first 10 columns of the main grid. The values in the 'G' channel are the 11th and 12th columns of the main grid. The values in the 'B' channel are the 13th and 14th columns of the main grid. The main grid values are as follows:

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



# Images in ~~Matlab~~ Python

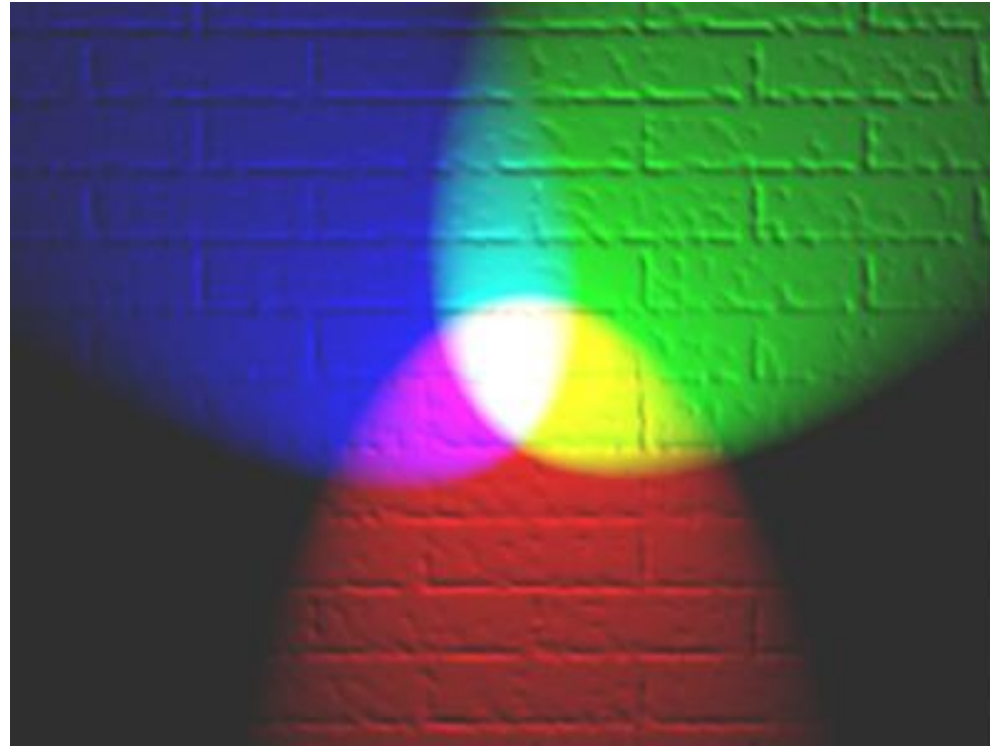
- Images represented as a matrix
- Suppose we have a NxM RGB image called “im”
  - $im(0,0,0)$  = top-left pixel value in R-channel
  - $im(y, x, b)$  = y pixels down, x pixels to right in the  $b^{\text{th}}$  channel
  - $im(N-1, M-1, 2)$  = bottom-right pixel in B-channel

The diagram illustrates an RGB image matrix. A large 10x10 grid of numerical values represents the image. A blue arrow labeled 'row' points downwards on the left side. A blue arrow labeled 'column' points to the right above the grid. To the right of the grid, three smaller 2x2 grids are shown, labeled 'R', 'G', and 'B', representing the individual color channels. The 'R' channel is the top-left 10x10 grid, the 'G' channel is a 2x2 grid to its right, and the 'B' channel is a 2x2 grid to the right of the 'G' channel. The numerical values in the grid are as follows:

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

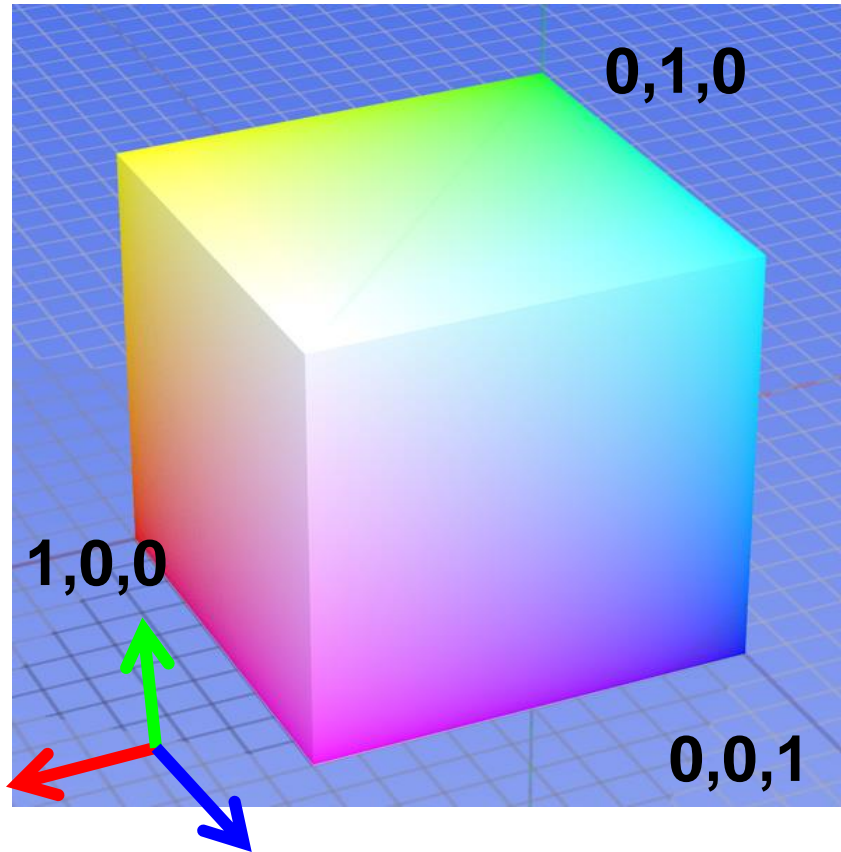
# Color spaces

- How can we represent color?



# Color spaces: RGB

Default color space



Some drawbacks

- Strongly correlated channels
- Non-perceptual



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



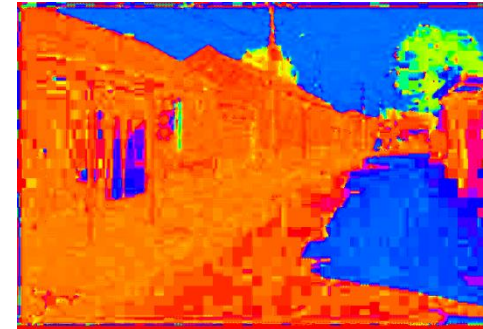
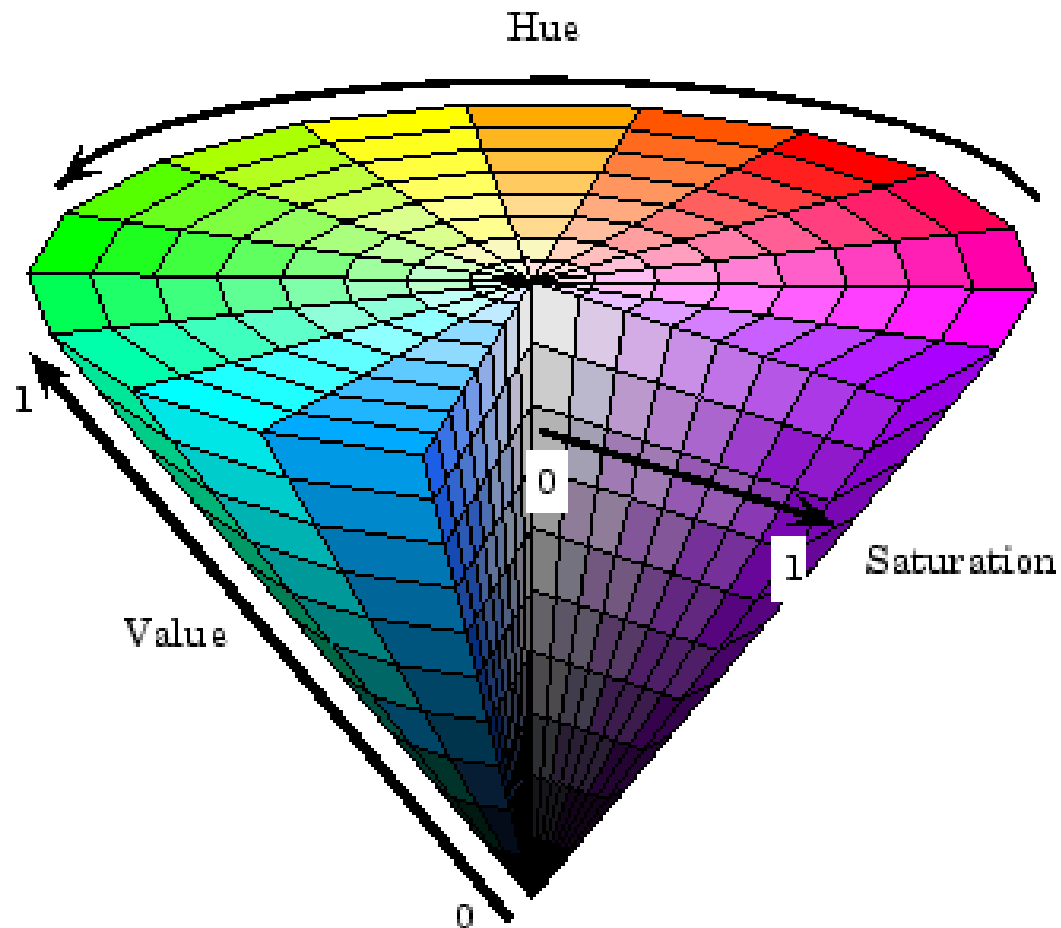
**G**  
(R=0,B=0)



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

# Color spaces: HSV

## Intuitive color space



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



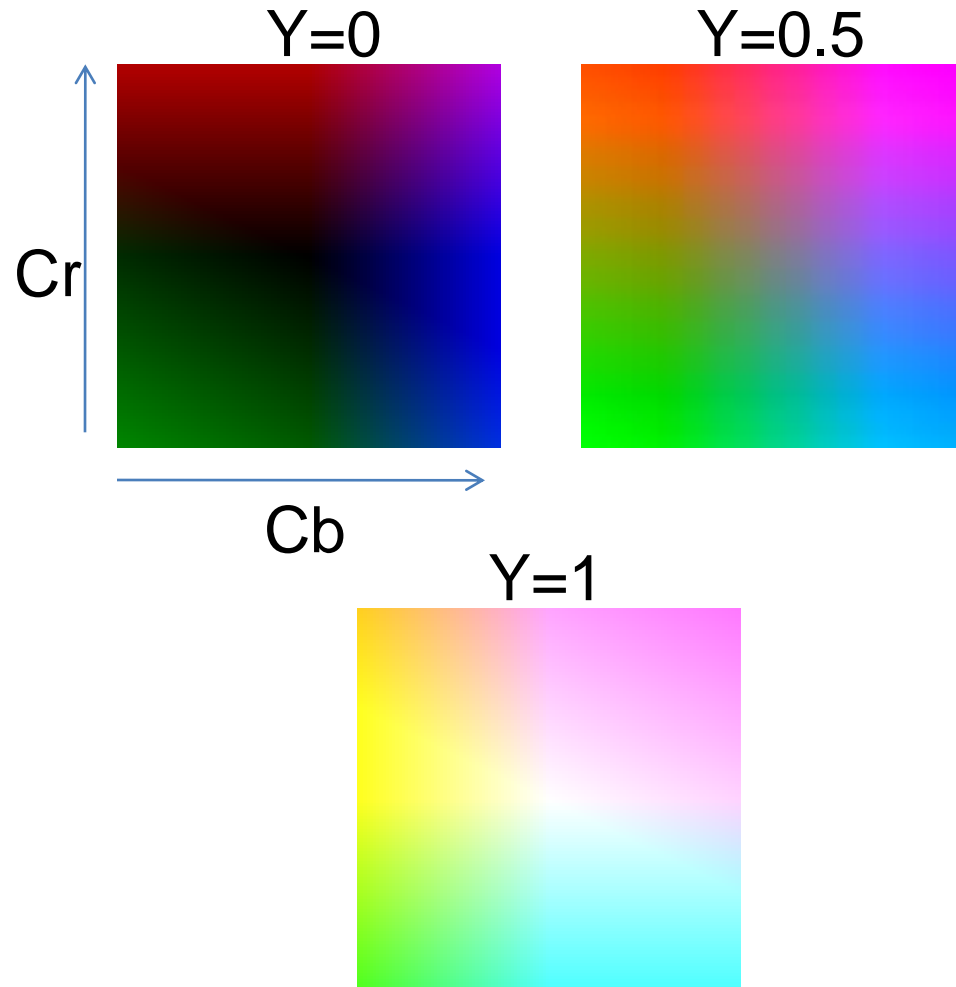
**S**  
(H=1, V=1)



**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)

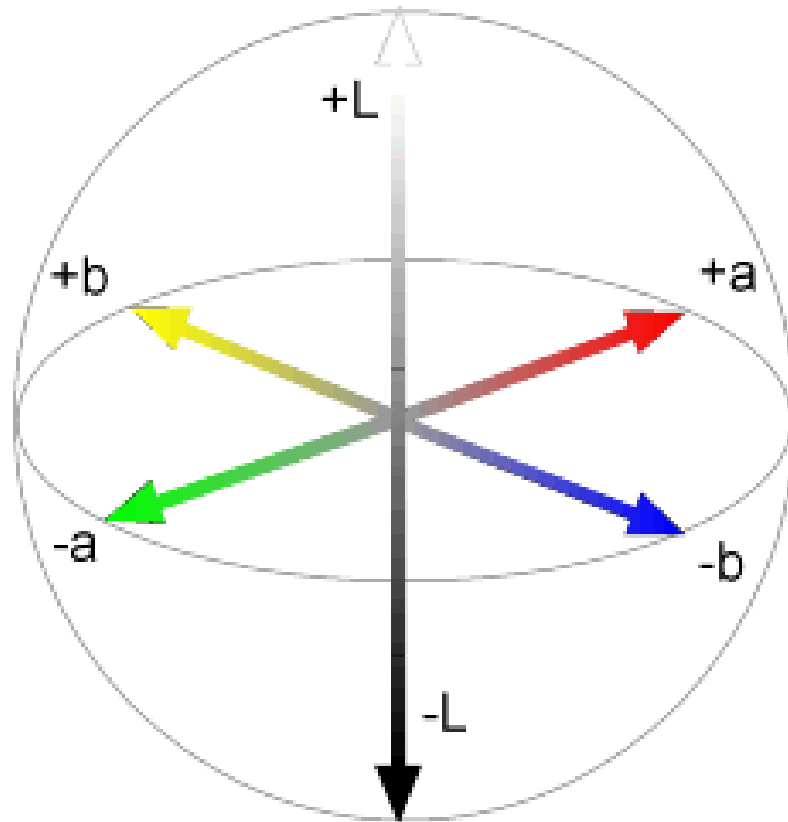


**Cr**  
(Y=0.5,Cb=0.5)

# 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



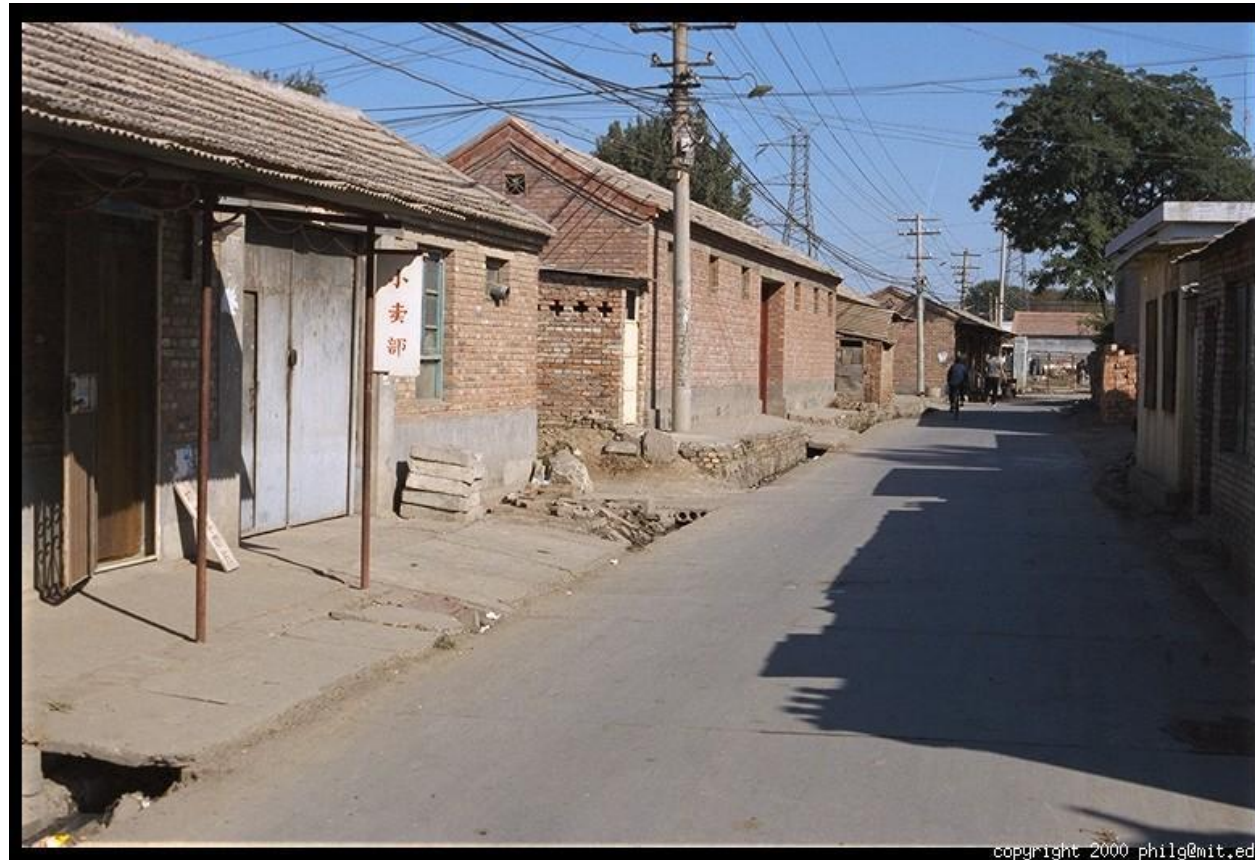
Only color shown – constant intensity

# Most information in intensity



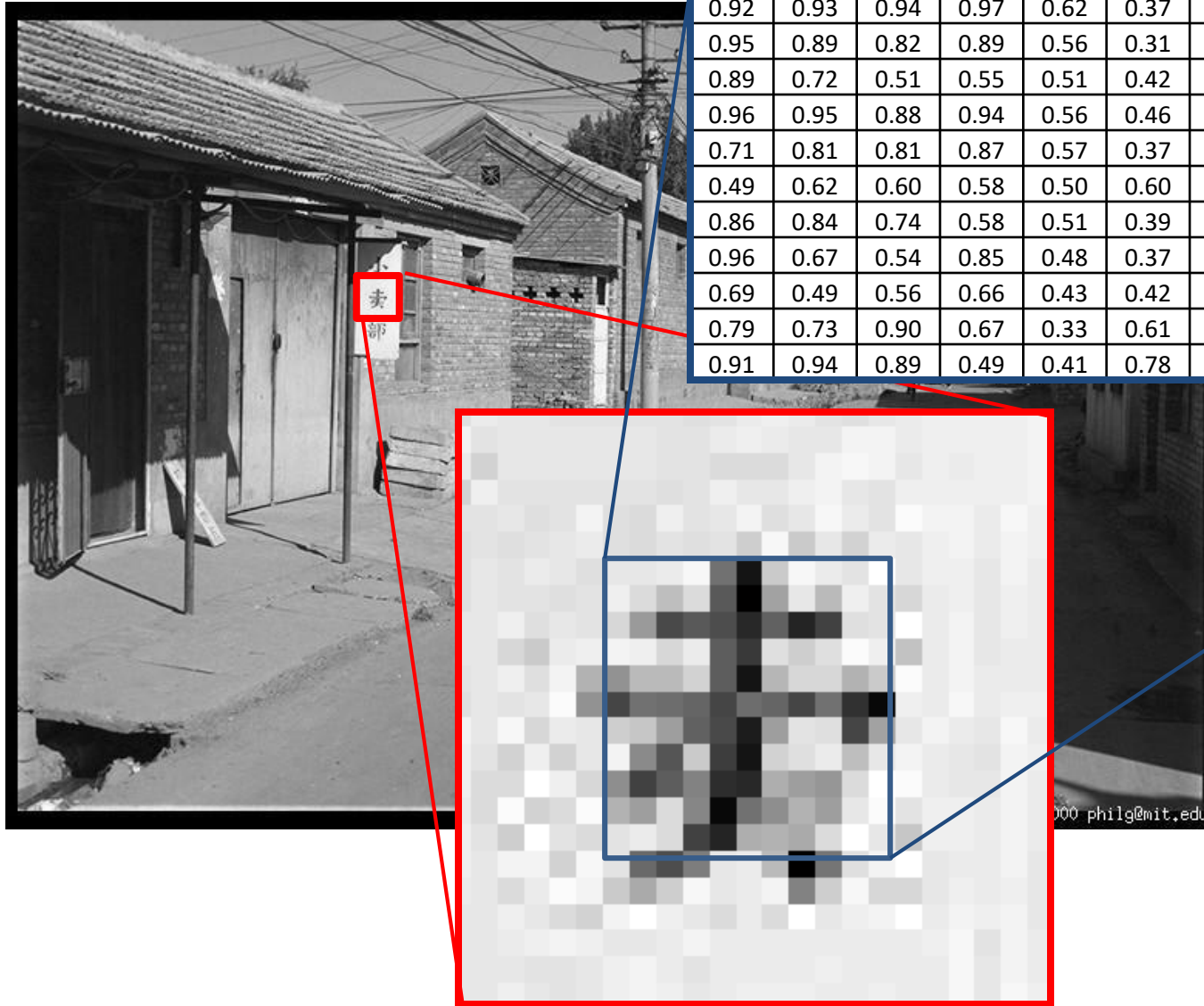
Only intensity shown – constant color

# Most information in intensity



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

Wrap up: Why do we care about cameras and eyes?

## *Inside Tesla as Elon Musk Pushed an Unflinching Vision for Self- Driving Cars*

The automaker may have undermined safety in designing its Autopilot driver-assistance system to fit its chief executive's vision, former employees say.



Hardware choices have also raised safety questions. Within Tesla, some argued for pairing cameras with radar and other sensors that worked better in heavy rain and snow, bright sunshine and other difficult conditions. For several years, Autopilot incorporated radar, and for a time Tesla worked on developing its own radar technology. But three people who worked on the project said Mr. Musk had repeatedly told members of the Autopilot team that humans could drive with only two eyes and that this meant cars should be able to drive with cameras alone.

Schuyler Cullen, who oversaw a team that explored autonomous-driving possibilities at the South Korean tech giant Samsung, said in an interview that Mr. Musk's cameras-only approach was fundamentally flawed. "Cameras are not eyes! Pixels are not retinal ganglia! The F.S.D. computer is nothing like the visual cortex!" said Mr. Cullen, a computer vision specialist who now runs a start-up that is building a new kind of camera-based sensor.

Amnon Shashua, chief executive of Mobileye, a former Tesla supplier that has been testing technology that is similar to the electric-car maker's, said Mr. Musk's idea of using only cameras in a self-driving system could ultimately work, though other sensors may be needed in the short term. He added that Mr. Musk might exaggerate the capabilities of the company's technology, but that those statements shouldn't be taken too seriously.



Another driving system with human level eyesight

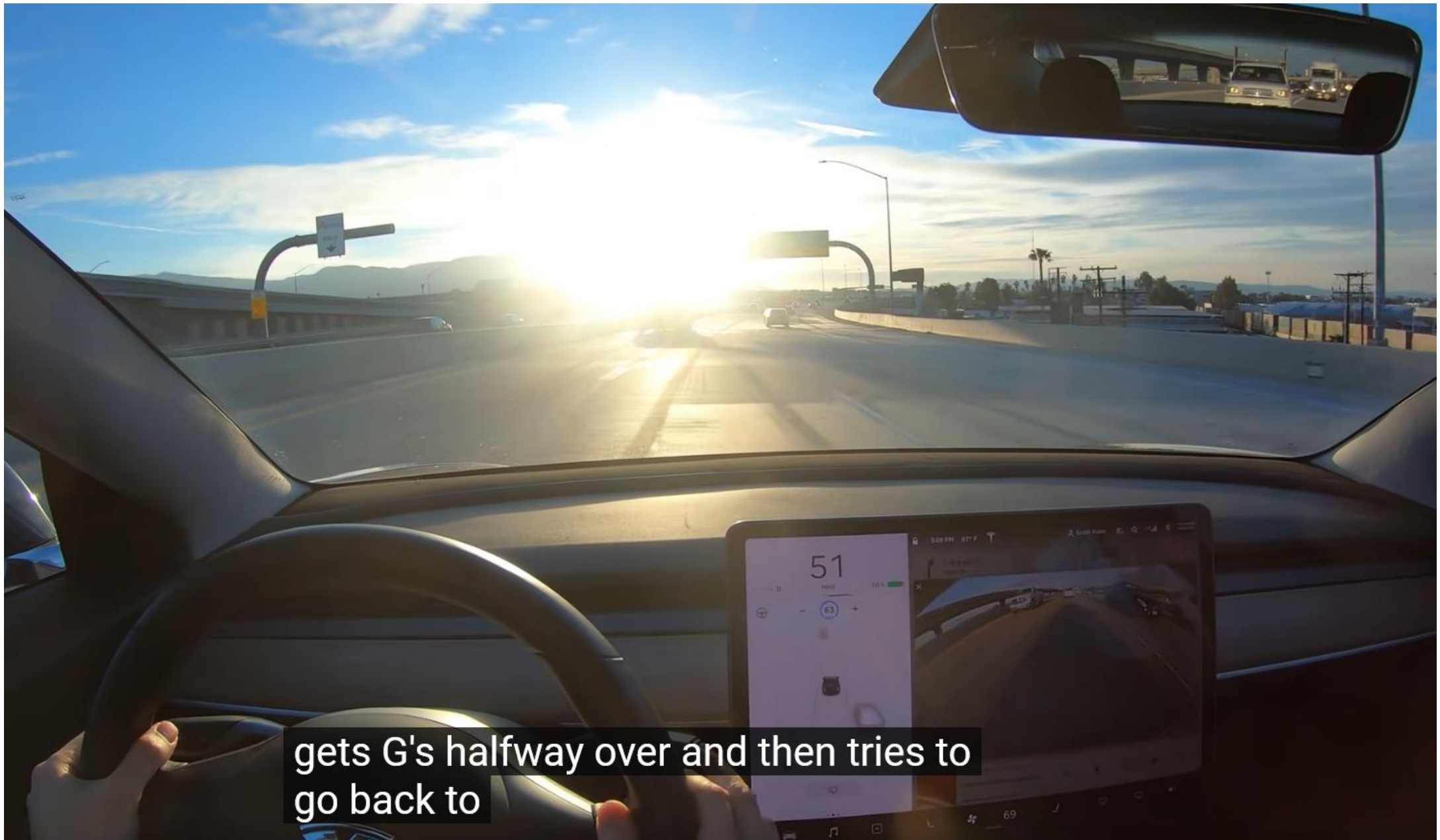






[https://www.reddit.com/r/teslamotors/comments/nrs8kf/you\\_think\\_ice\\_cream\\_truck\\_stop\\_signs\\_are\\_a\\_problem/](https://www.reddit.com/r/teslamotors/comments/nrs8kf/you_think_ice_cream_truck_stop_signs_are_a_problem/)

- On the whole, cameras *are* a reasonable analogy for eyes. They do capture sufficient information for safe driving 99.9% of the time.



gets G's halfway over and then tries to go back to

- On the whole, cameras *are* a reasonable analogy for eyes. They do capture sufficient information for safe driving 99.9% of the time.
  - Imagine remote controlling a vehicle based on a camera feed.
- *But* the computer vision and machine learning methods that interpret the camera images *are not* yet a reasonable analogy for the human brain.