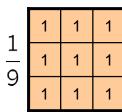
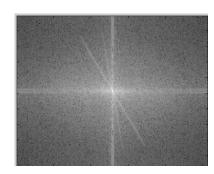
Recap of Filtering

- 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

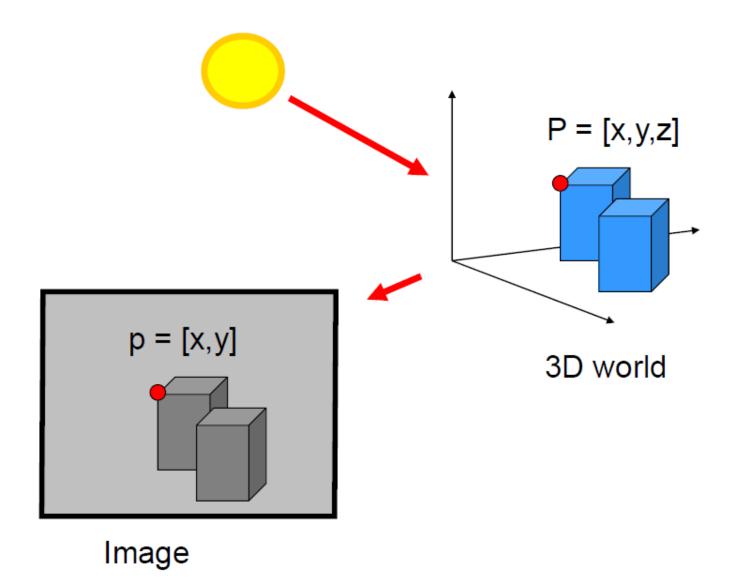
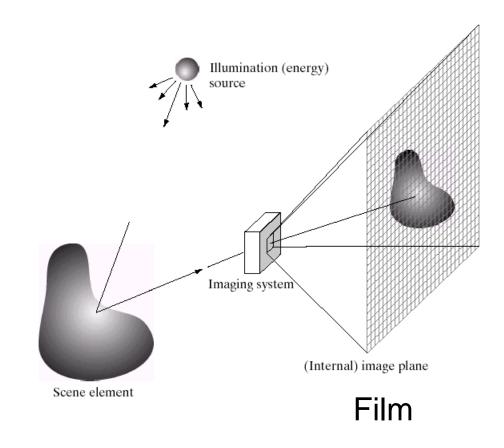
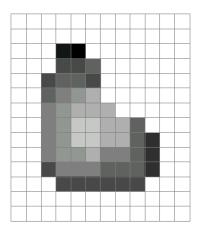
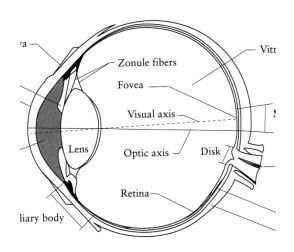


Image Formation



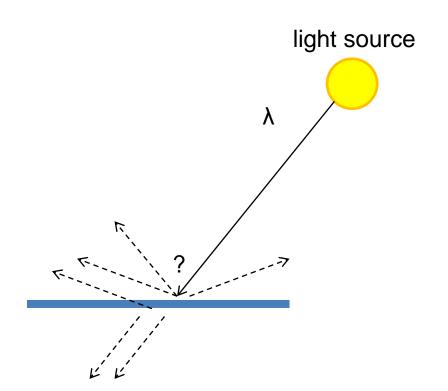


Digital Camera

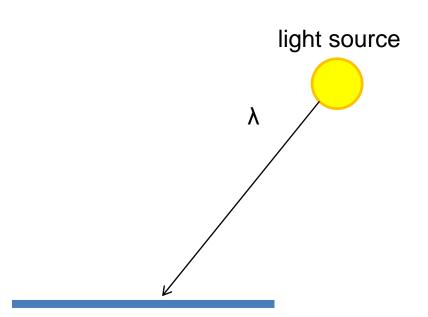


The Eye

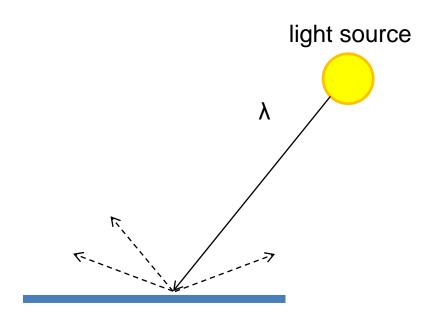
- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



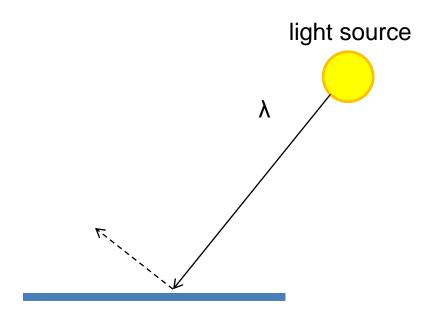
- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



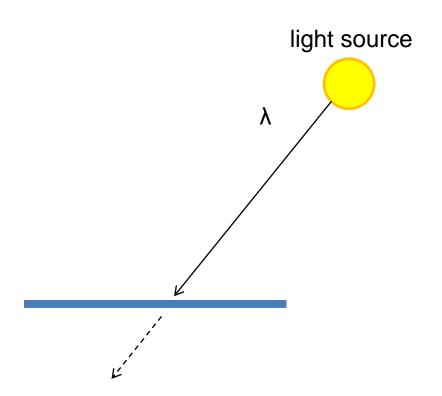
- Absorption
- Diffuse Reflection
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



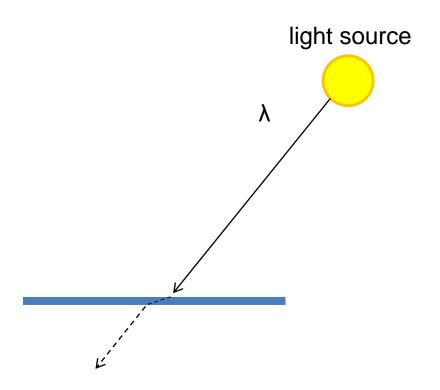
- Absorption
- Diffusion
- Specular Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



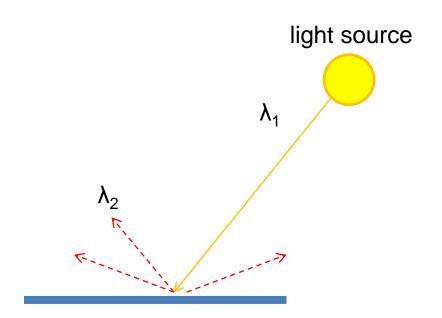
- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



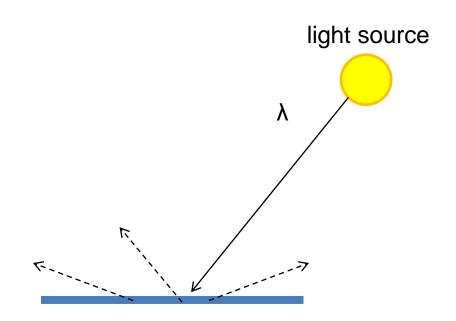
- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



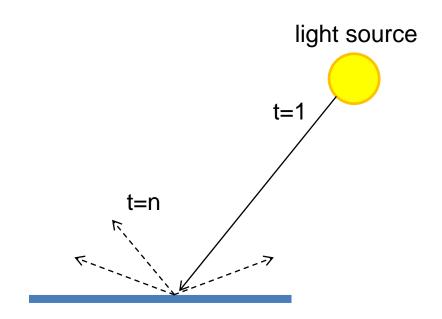
- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



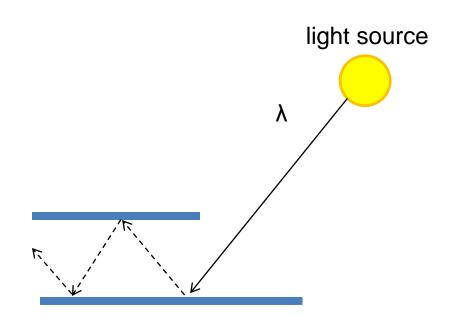
- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



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



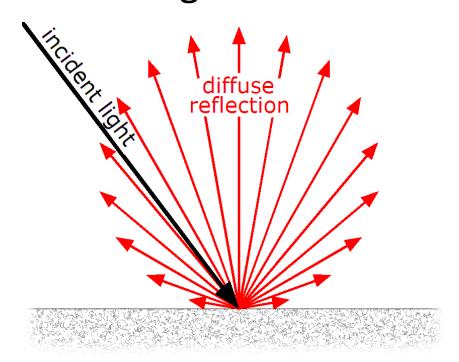
- Absorption
- Diffusion
- Reflection
- Transparency
- Refraction
- Fluorescence
- Subsurface scattering
- Phosphorescence
- Interreflection



(Specular 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.

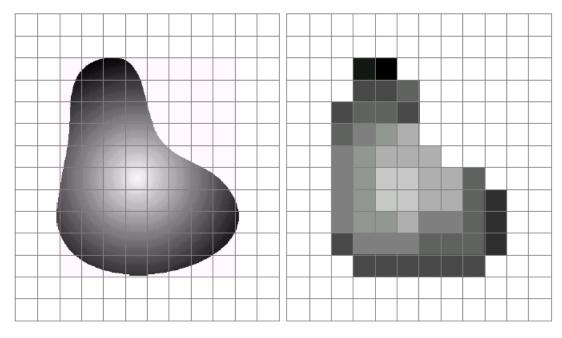


Digital camera



- 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

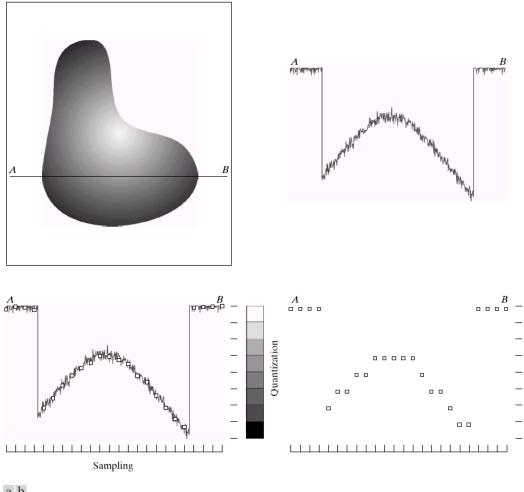


CMOS sensor

a b

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

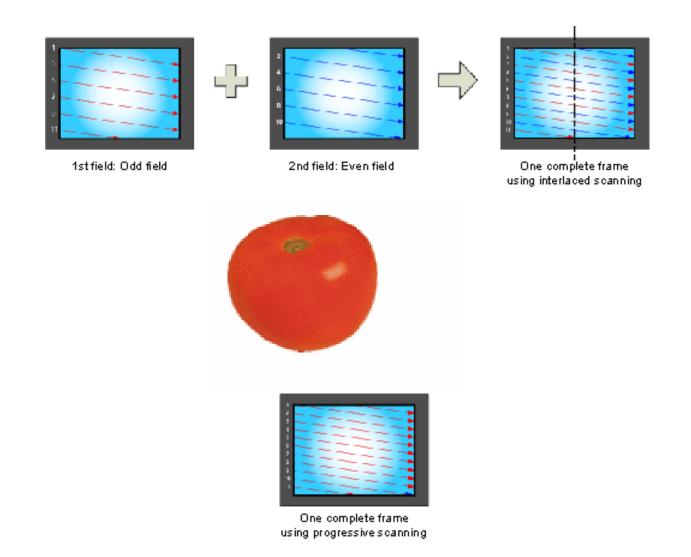
Sampling and Quantization



a b c d

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



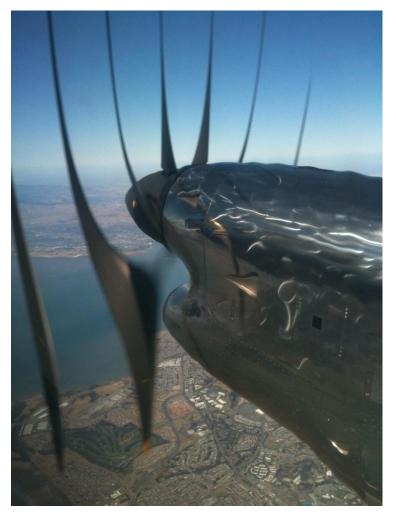
http://www.axis.com/products/video/camera/progressive_scan.htm

Interlaced

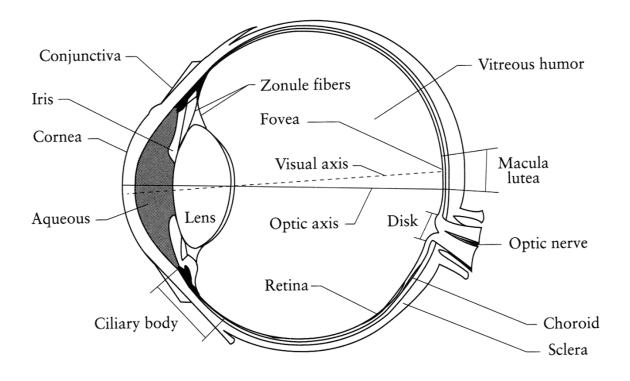


Rolling Shutter





The Eye



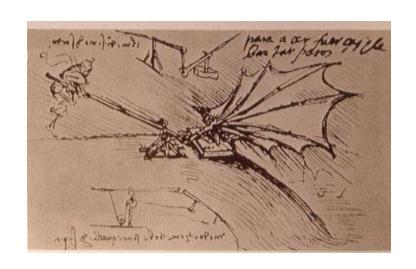
The human eye is a camera!

- 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?

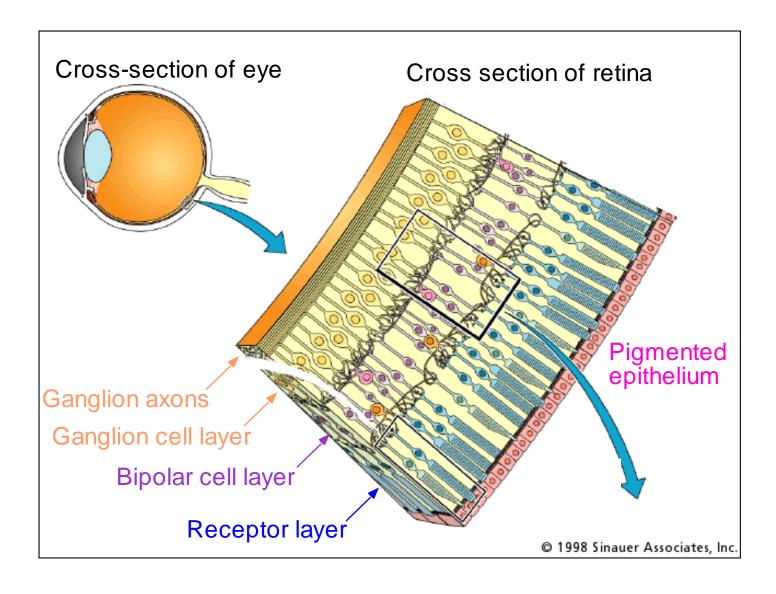
"Can machines fly?" The answer is yes, because airplanes fly.

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

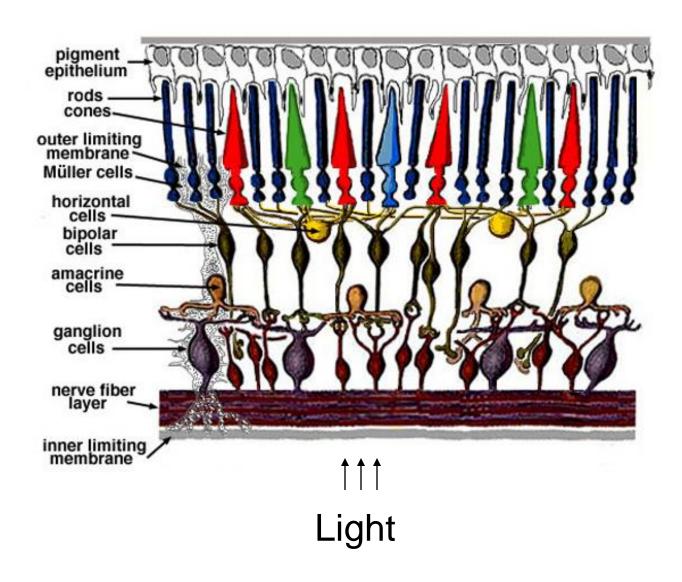
"Can machines think?" Is this question like the first, or like the second?

Source: Norvig

The Retina



Retina up-close



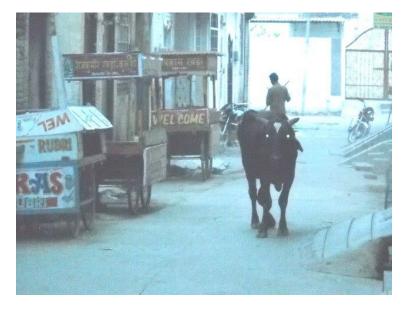
What humans don't have: tapetum lucidum



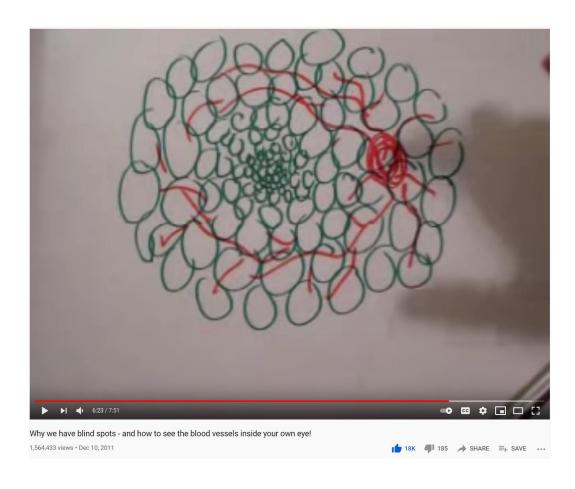




Human eyes can reflect a tiny bit and blood in the retina makes this reflection red.



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



https://www.youtube.com/watch?v=L_W-IXqoxHA

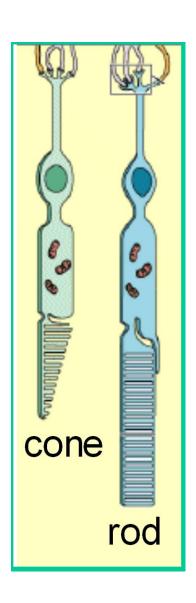
Two types of light-sensitive receptors

Cones

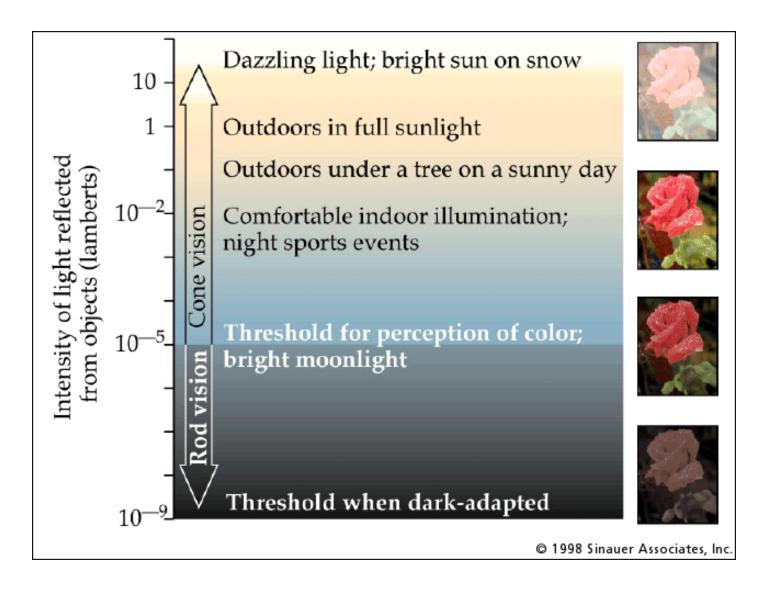
cone-shaped less sensitive operate in high light color vision

Rods

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



Rod / Cone sensitivity



Camera Gamma Curve

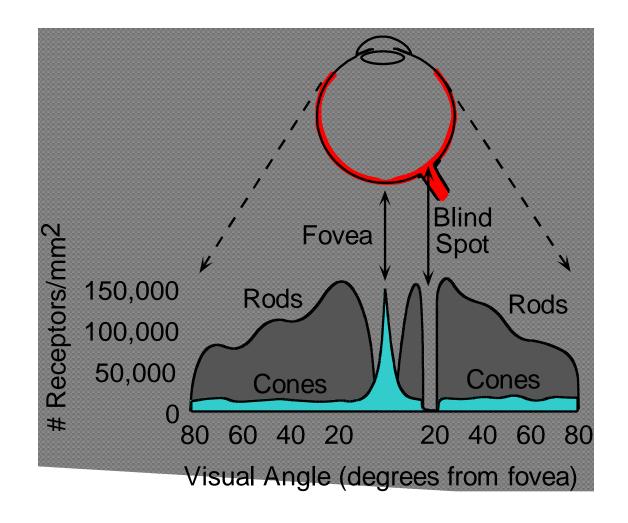
encoded brightness = measured brightness $^{\gamma}$

 γ will typically take values less than 1

In this class (e.g. for project 1), you are manipulating *encoded* brightness values. A pixel with value of "100" is not necessarily twice as bright as a pixel with value "50".



Distribution of Rods and Cones



Night Sky: why are there more stars off-center?
Averted vision: http://en.wikipedia.org/wiki/Averted_vision

Eye Movements

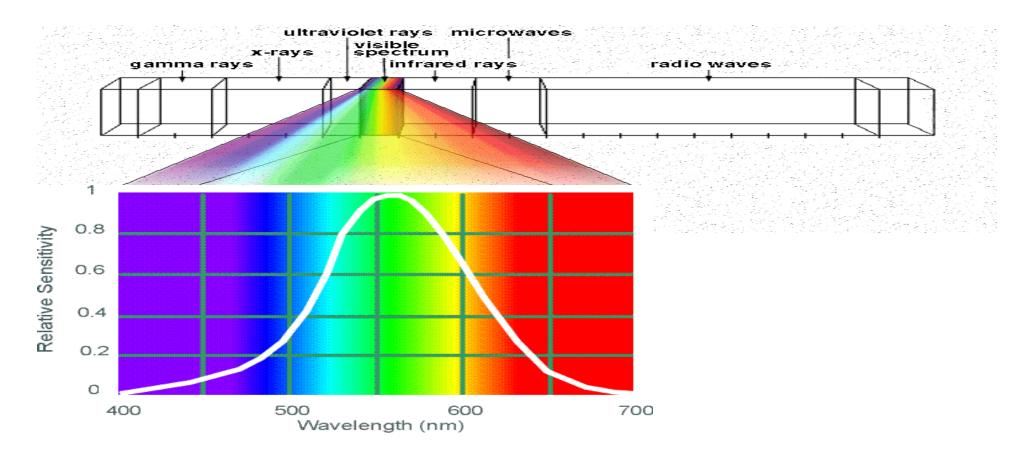
- Saccades
- Can be consciously controlled. Related to perceptual attention.
- 200ms to initiation, 20 to 200ms to carry out. Large amplitude.
- Microsaccades
- Involuntary. Smaller amplitude. Especially evident during prolonged fixation. Function debated.
- Ocular microtremor (OMT)
- Involuntary. high frequency (up to 80Hz), small amplitude.
- Smooth pursuit tracking an object

Slow mo guys – Saccades and CRTs

- https://youtu.be/Fmg9ZOHESgQ?t=21s
- https://youtu.be/3BJU2drrtCM

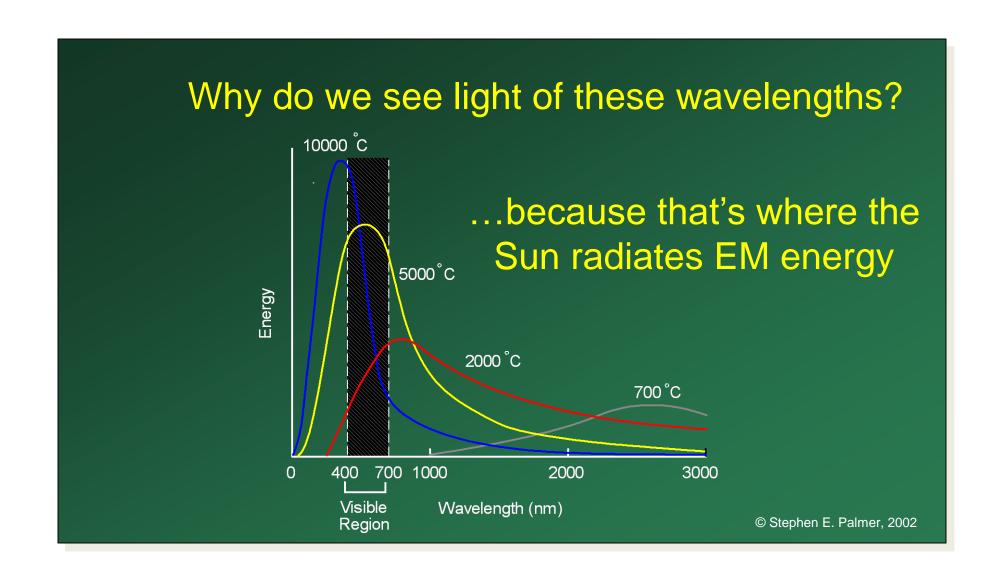


Electromagnetic Spectrum



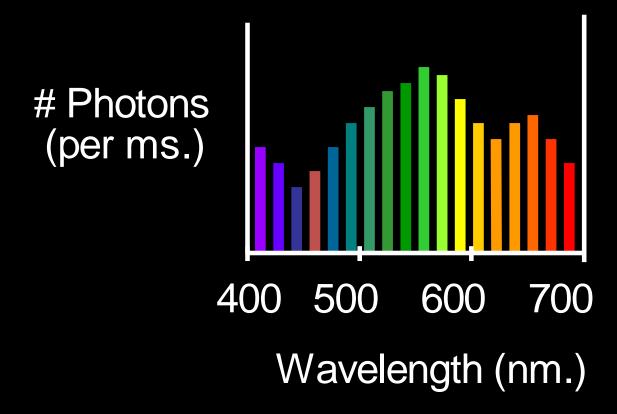
Human Luminance Sensitivity Function

Visible Light



The Physics of Light

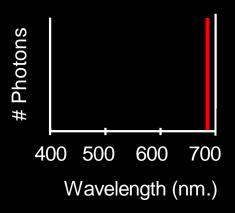
Any patch of light can be completely described physically by its spectrum: the number of photons (per time unit) at each wavelength 400 - 700 nm.



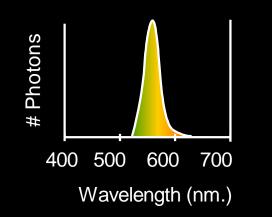
The Physics of Light

Some examples of the spectra of light sources

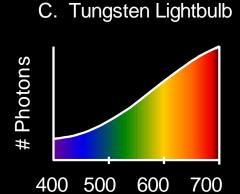


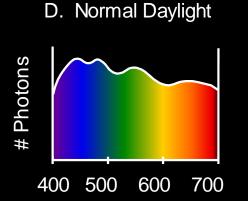


A. Ruby Laser



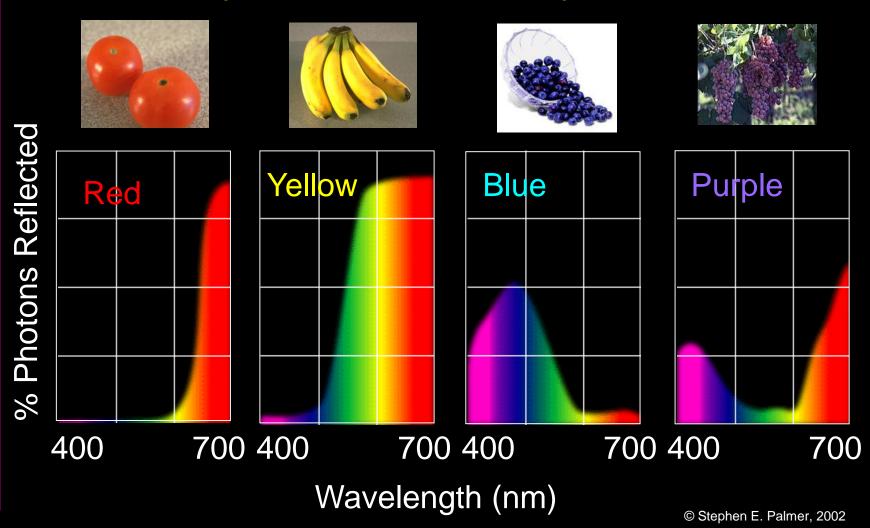
B. Gallium Phosphide Crystal





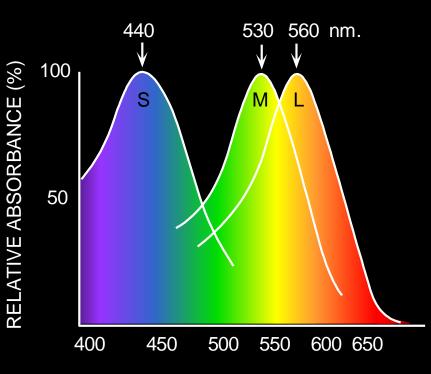
The Physics of Light

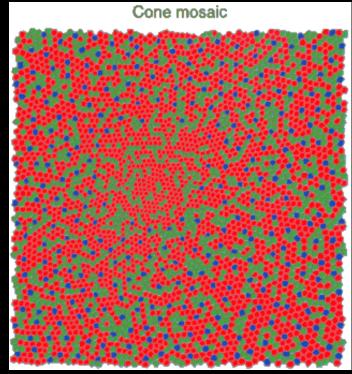
Some examples of the <u>reflectance</u> spectra of <u>surfaces</u>



Physiology of Color Vision

Three kinds of cones:

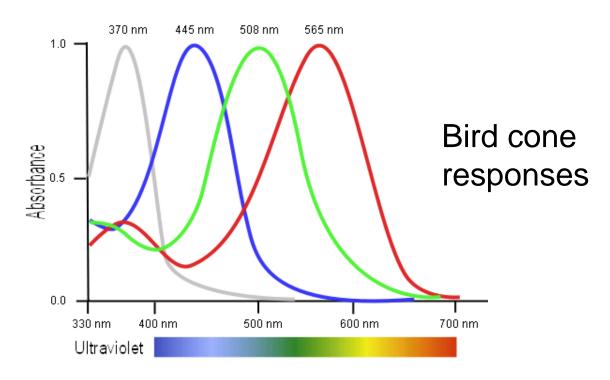




WAVELENGTH (nm.)

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

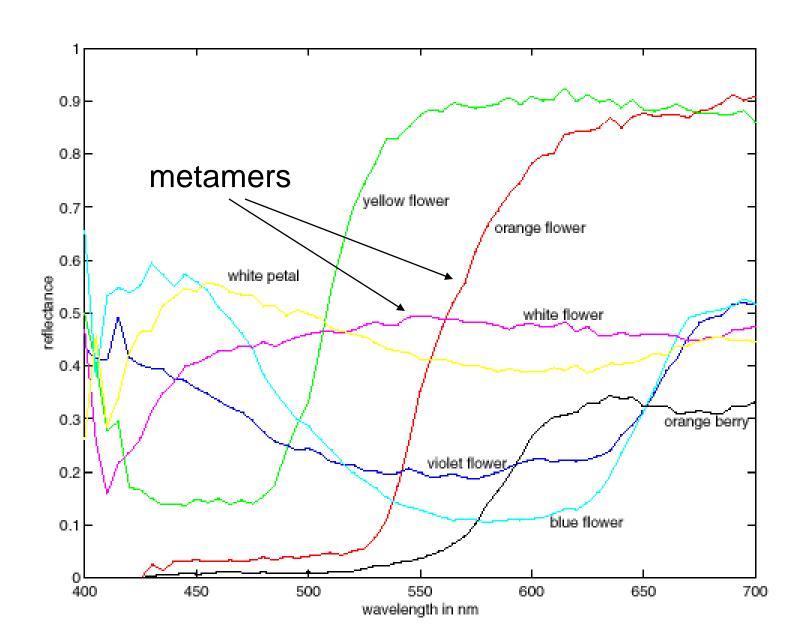
(b) Eutherians

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.

Jacobs GH. Evolution of colour vision in mammals. *Philos Trans R Soc Lond B Biol Sci.* 2009;364(1531):2957-2967. doi:10.1098/rstb.2009.0039

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, [3][4] 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.

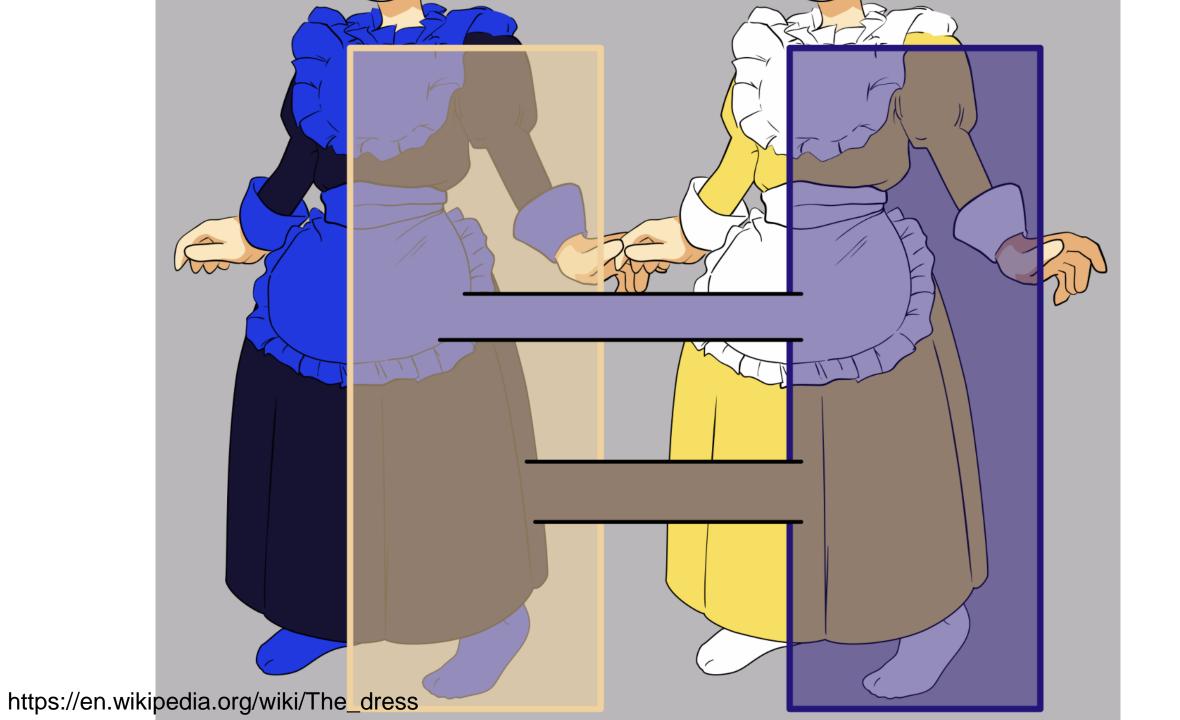
The dress itself, which was identified as a product of the retailer Roman Originals, experienced a major surge in sales as a result of the incident. The retailer also produced a one-off version of the dress in white and gold as a charity campaign.

Contents [hide]

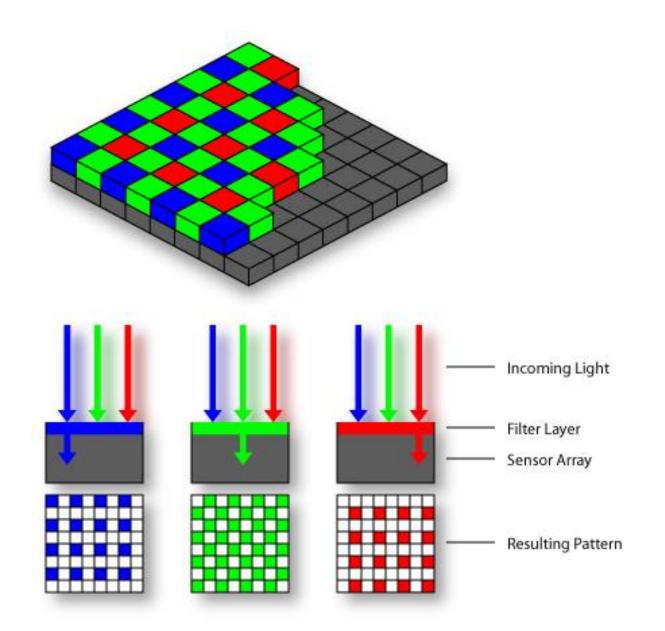
The dress

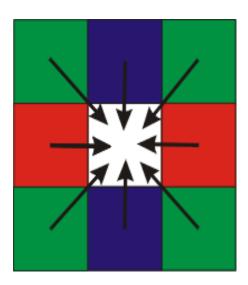


https://en.wikipedia.org/wiki/The_dress



Practical Color Sensing: Bayer Grid





Estimate RGB
 at 'G' cells from
 neighboring
 values

Color Image



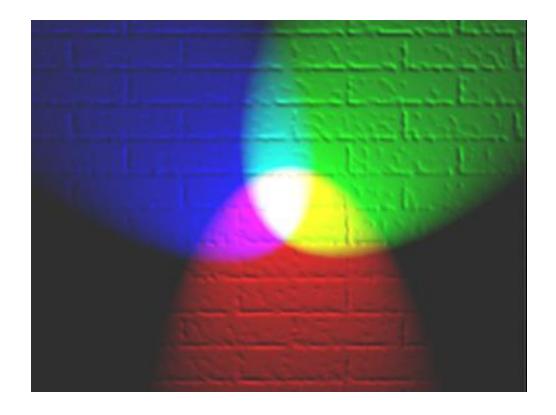
Images in Python

- 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 bth channel
 - im(N-1, M-1, 2) = bottom-right pixel in B-channel

	col	um	n -									\Rightarrow				
row	0.92	0.93	0.94	0.97	0.62	0.37	0.85	0.97	0.93	0.92	0.99	R				
	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	0.92	0.99	ı G		
	0.96	0.95	0.88	0.94	0.56	0.46	0.91	0.87	0.90	0.97	0.95	0.95	0.91			_
	0.71	0.81	0.81	0.87	0.57	0.37	0.80	0.88	0.89	0.79	0.85	0.91	0.92		Ī	B
	0.49	0.62	0.60	0.58	0.50	0.60	0.58	0.50	0.61	0.45	0.33	0.97	0.95	0.92	0.99	
	0.86	0.84	0.74	0.58	0.51	0.39	0.73	0.92	0.91	0.49	0.74	0.79	0.85	0.95	0.91	
	0.96	0.67	0.54	0.85	0.48	0.37	0.88	0.90	0.94	0.82	0.93	0.45	0.33	0.91	0.92	
	0.69	0.49	0.56	0.66	0.43	0.42	0.77	0.73	0.71	0.90	0.99	0.49	0.74	0.97	0.95	
	0.79	0.73	0.90	0.67	0.33	0.61	0.69	0.79	0.73	0.93	0.97	0.82	0.93	0.79	0.85	
W	0.91	0.94	0.89	0.49	0.41	0.78	0.78	0.77	0.89	0.99	0.93	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	0.49	0.74	
			0.91	0.94	0.89	0.49	0.41	0.78	0.78	0.77	0.89	0.99	0.93	0.82	0.93	
					0.05	0.45	0.50	0.00	0.15	U. TZ	0.77	0.75	0.7 1	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	

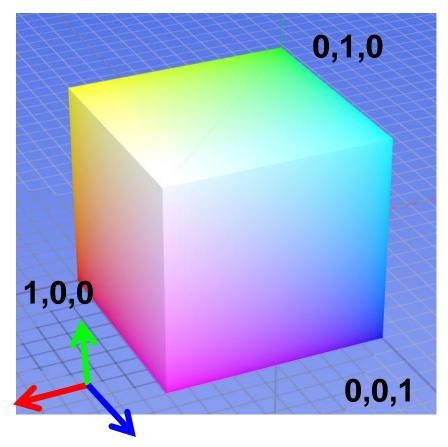
Color spaces

• 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)

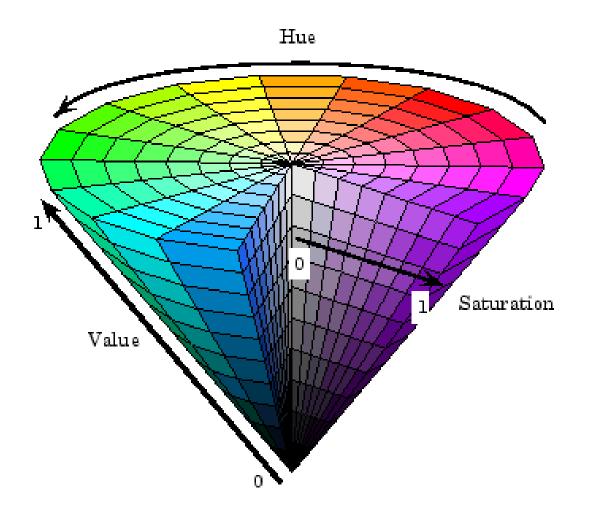


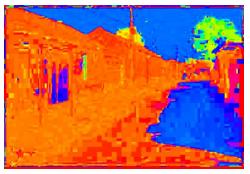
B (R=0,G=0)

Color spaces: HSV



Intuitive color space









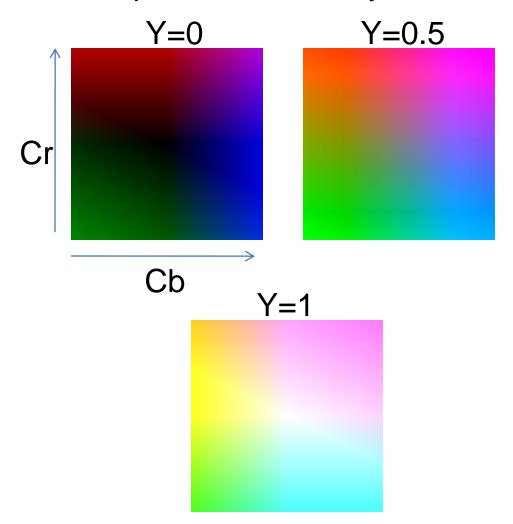
S (H=1,V=1)



V (H=1,S=0)

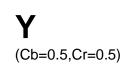
Color spaces: YCbCr

Fast to compute, good for compression, used by TV











Cb (Y=0.5,Cr=0.5)

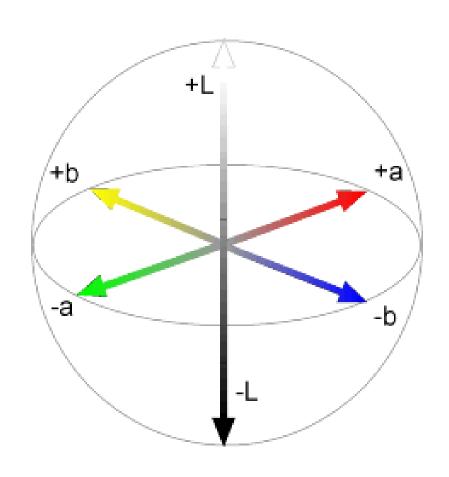


Cr (Y=0.5,Cb=05)

Color spaces: L*a*b*



"Perceptually uniform"* color space





(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



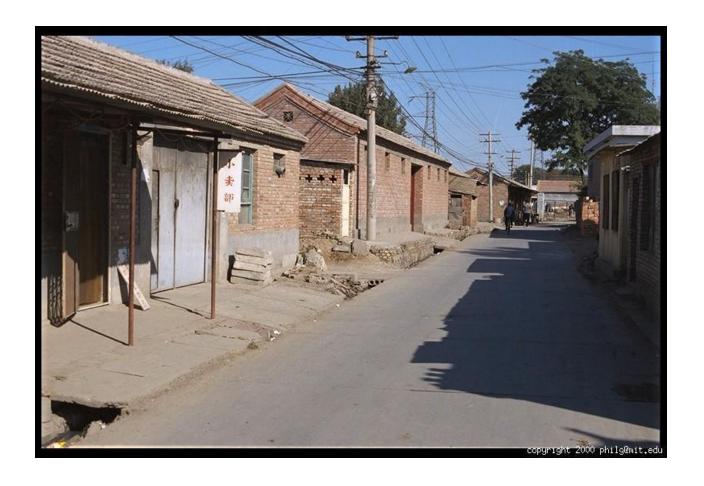
Only color shown – constant intensity

Most information in intensity



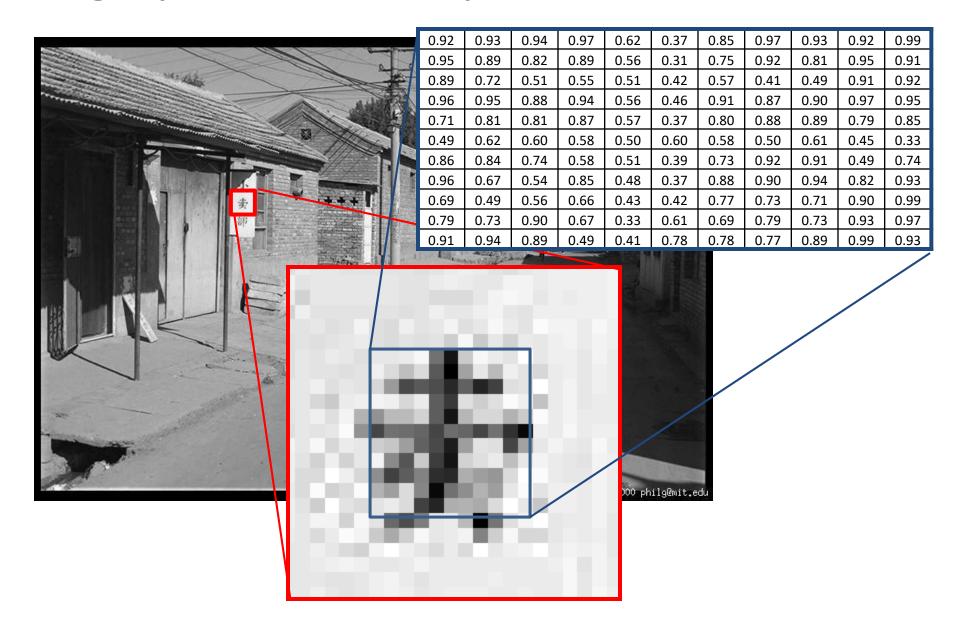
Only intensity shown – constant color

Most information in intensity



Original image

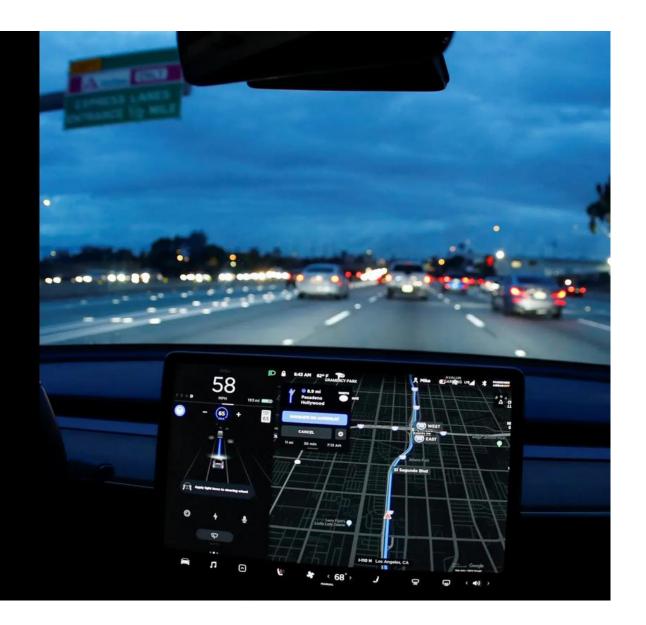
Back to grayscale intensity



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

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

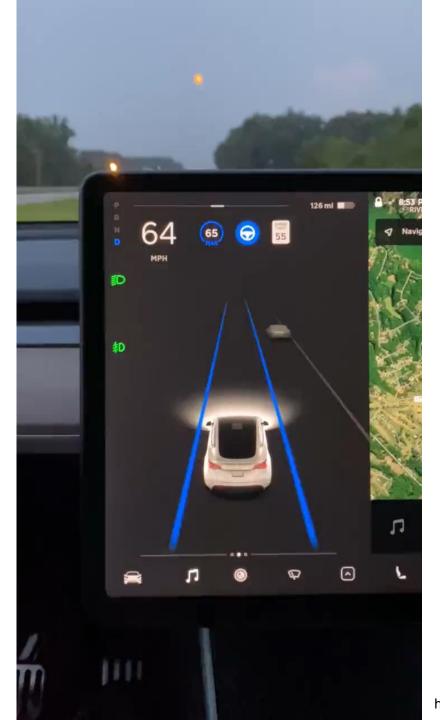
The automaker may have undermined safety in designing its Autopilot driver-assistance system to fit its chief executive's vision, former employees say.

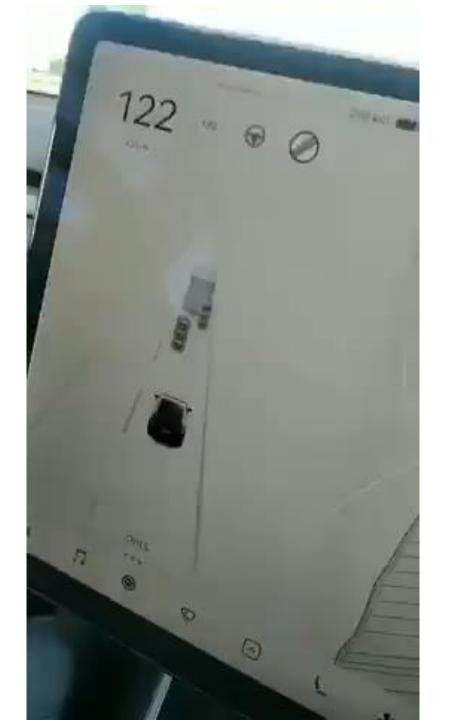


Hardware choices have also raised safety questions. Within Tesla, some argued for pairing cameras with radar and other sensors that worked better in heavy rain and snow, bright sunshine and other difficult conditions. For several years, Autopilot incorporated radar, and for a time Tesla worked on developing 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.

Another driving system with human level eyesight

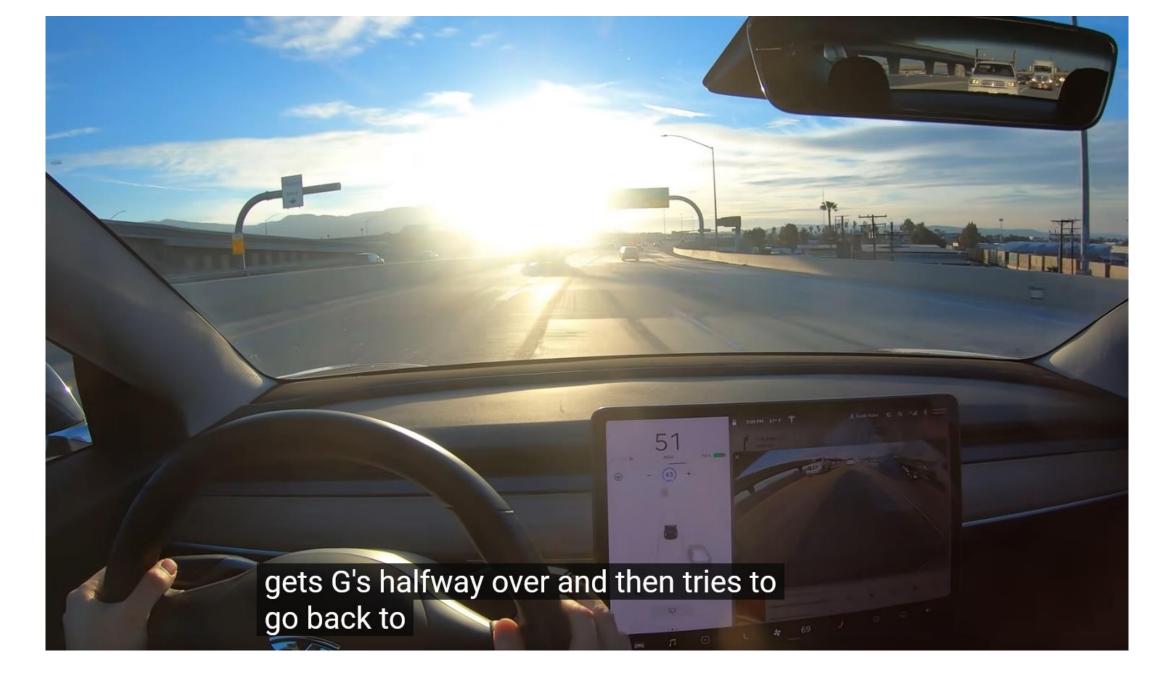






https://www.reddit.com/r/teslamotors/comments/nrs8kf/you_think_ice_cream_truck_stop_signs_are_a_problem/

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



- 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.

Next: Interest points and corners