### **COMPUTATIONAL ASPECTS OF DIGITAL PHOTOGRAPHY** Light & Color (continued)







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### Administrivia

Assignment 2 available now

- back to programming
- due next Wednesday

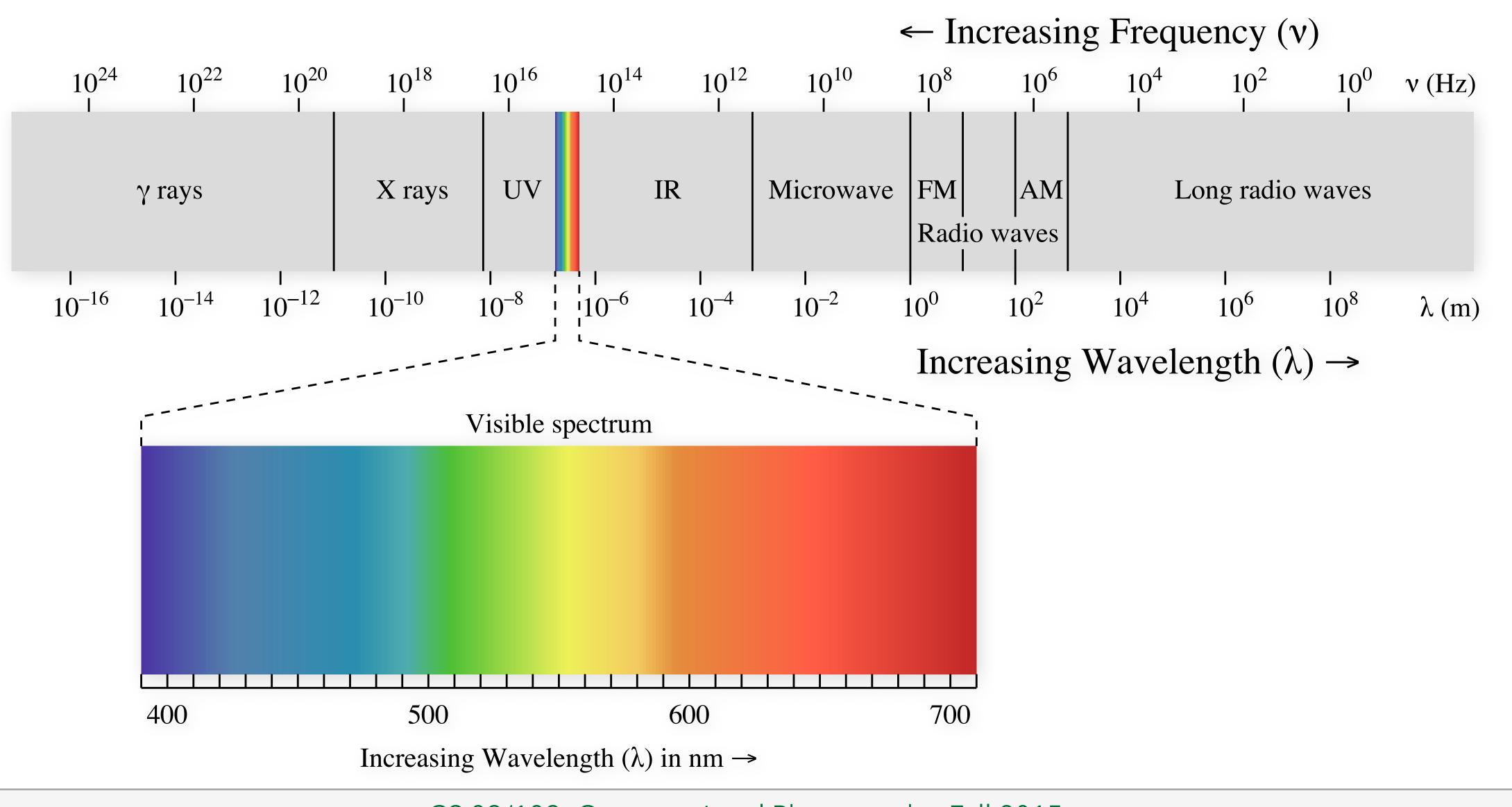


### Last time...

- Light & Color
- Physics background
- Color perception & measurement
- Color reproduction
- Color spaces



## What is light?

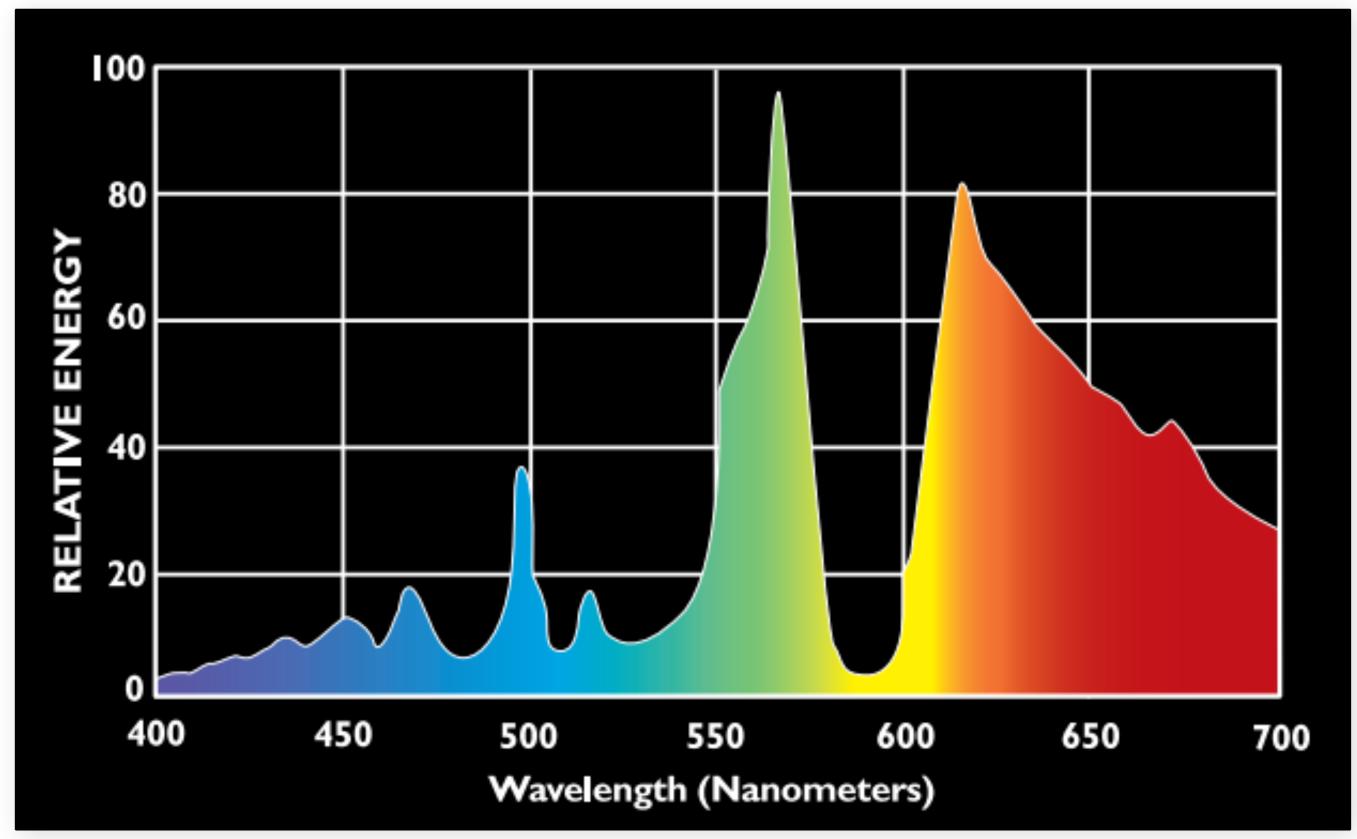




## **Spectral distribution function (SPD)**

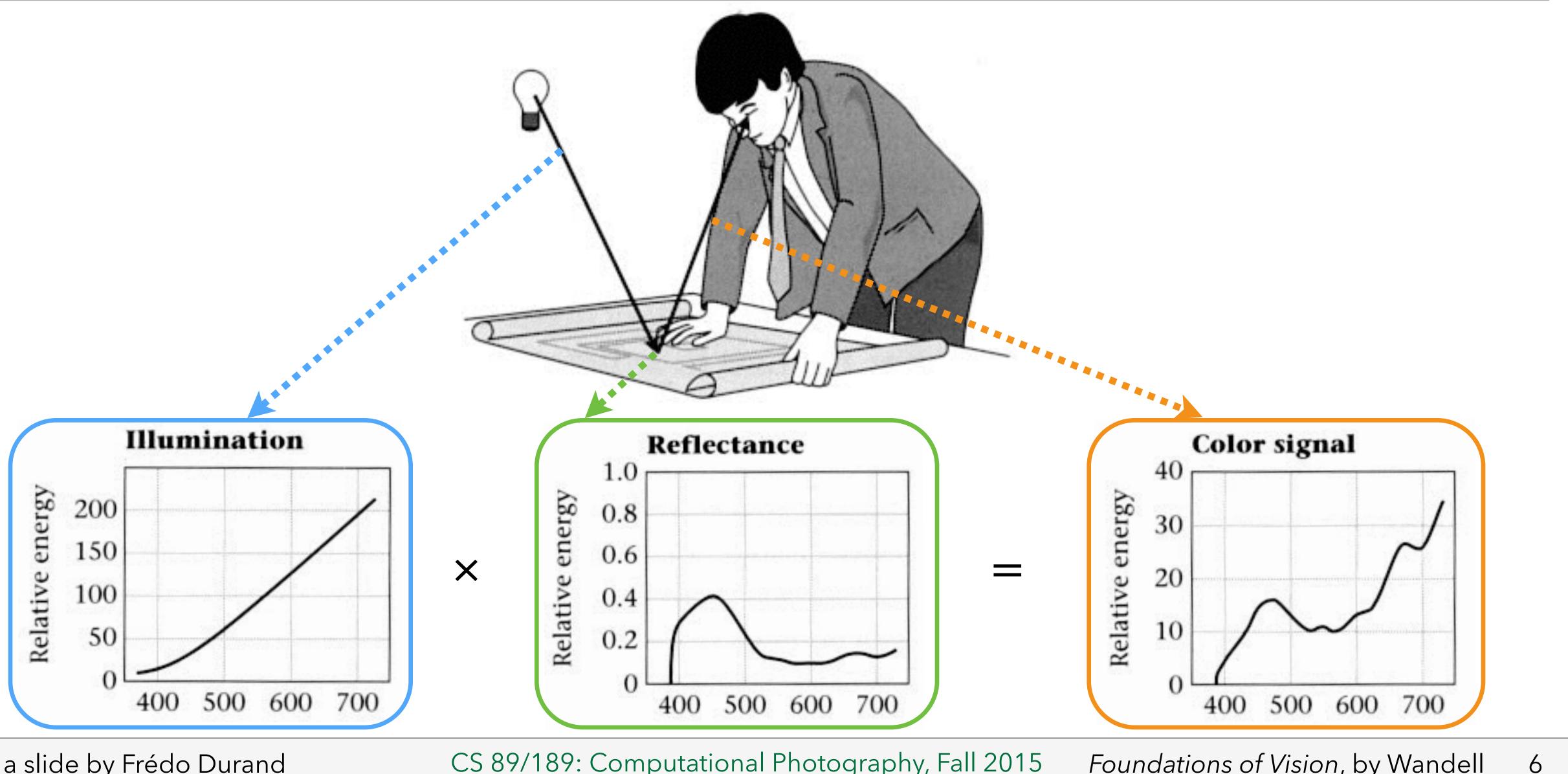
### Light can be a mixture of many wavelengths

- SPD: intensity as a function of wavelength over enter spectrum





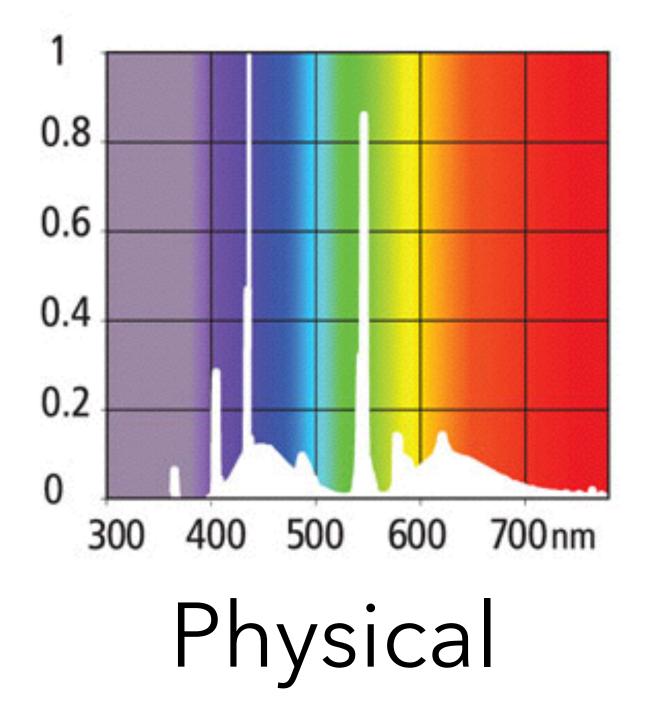
### Light-matter interaction



After a slide by Frédo Durand

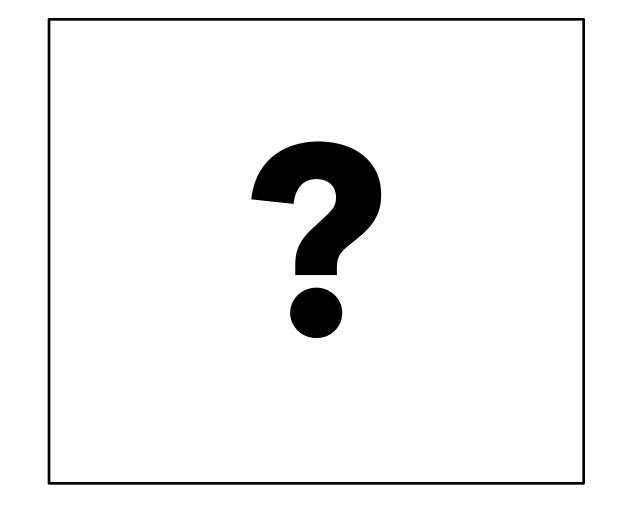
CS 89/189: Computational Photography, Fall 2015 Foundations of Vision, by Wandell

## Physical light to perceptual color



After a slide by Steve Marschner

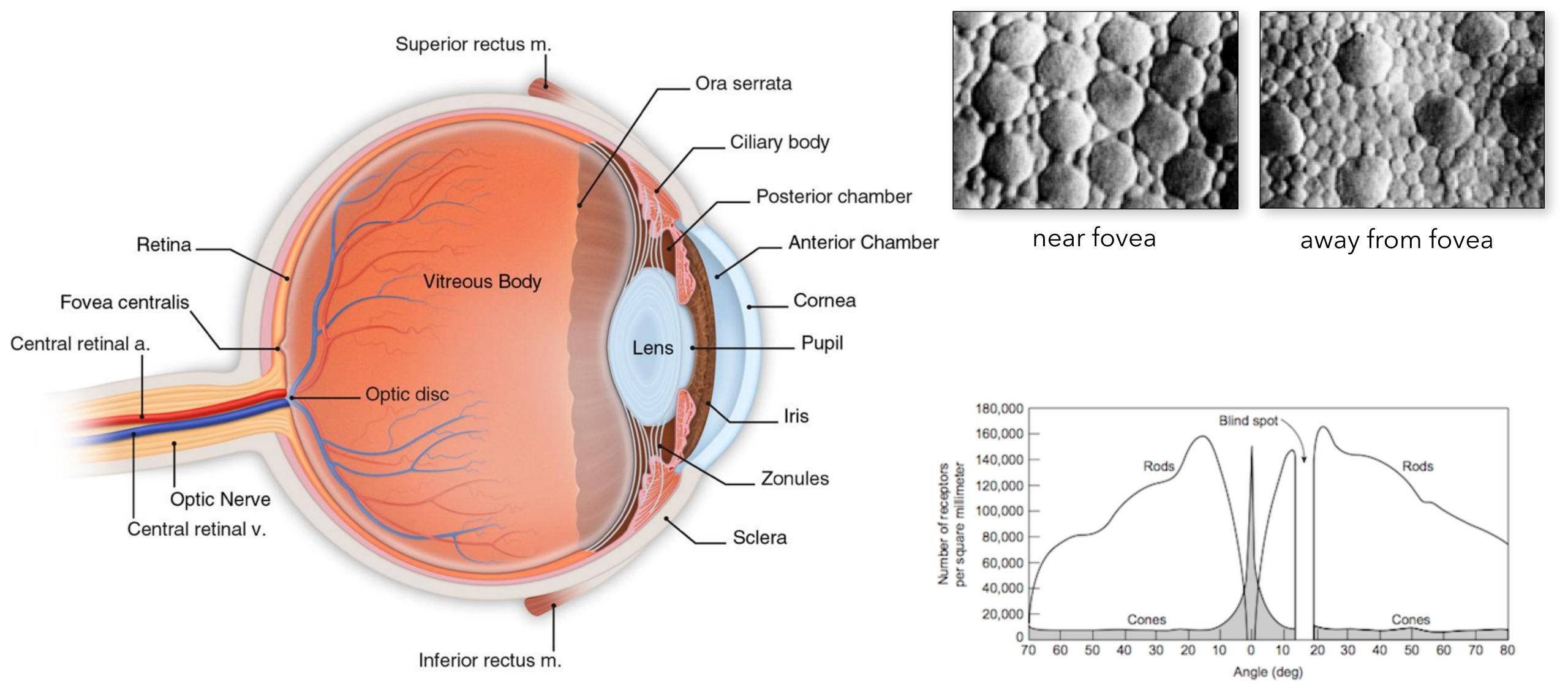
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### Perceptual

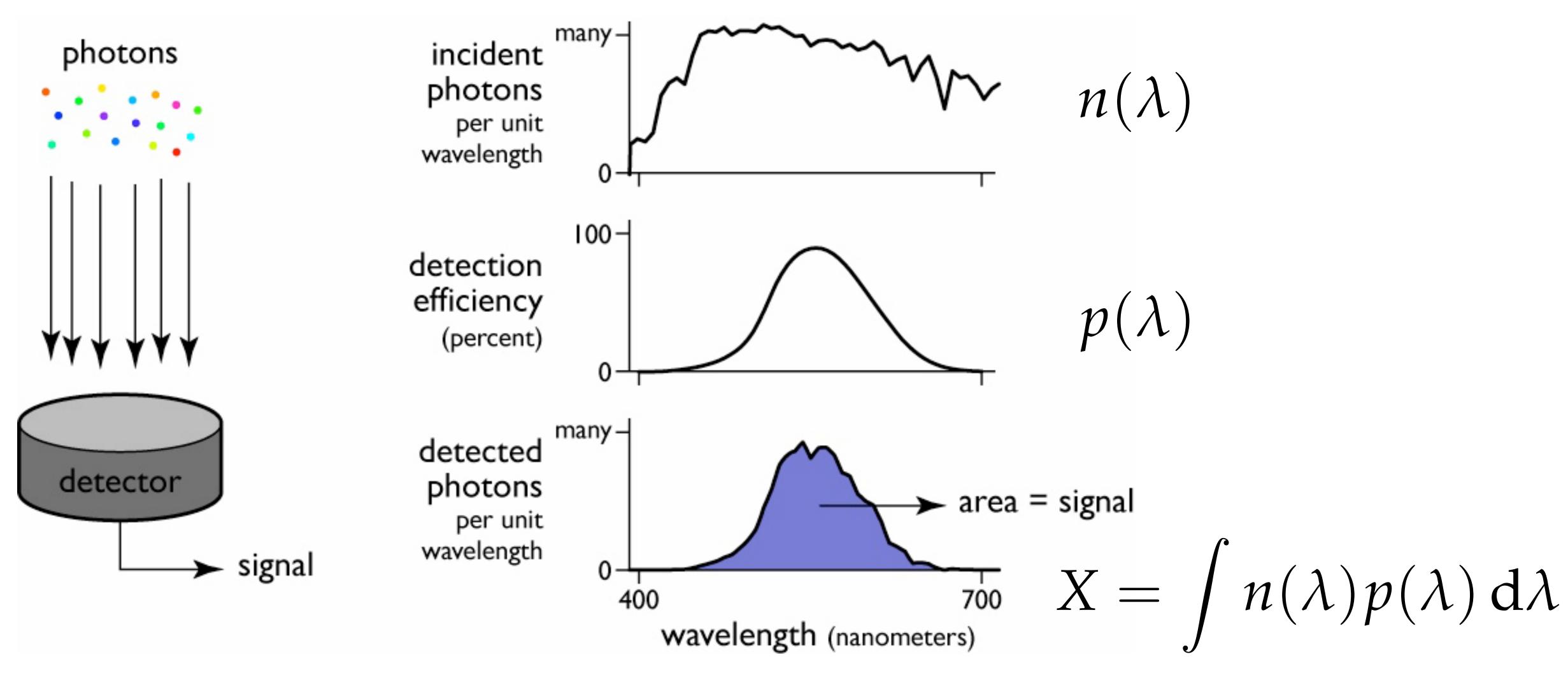


### The eye as a measurement device





## A simple light detector model



After a slide by Steve Marschner

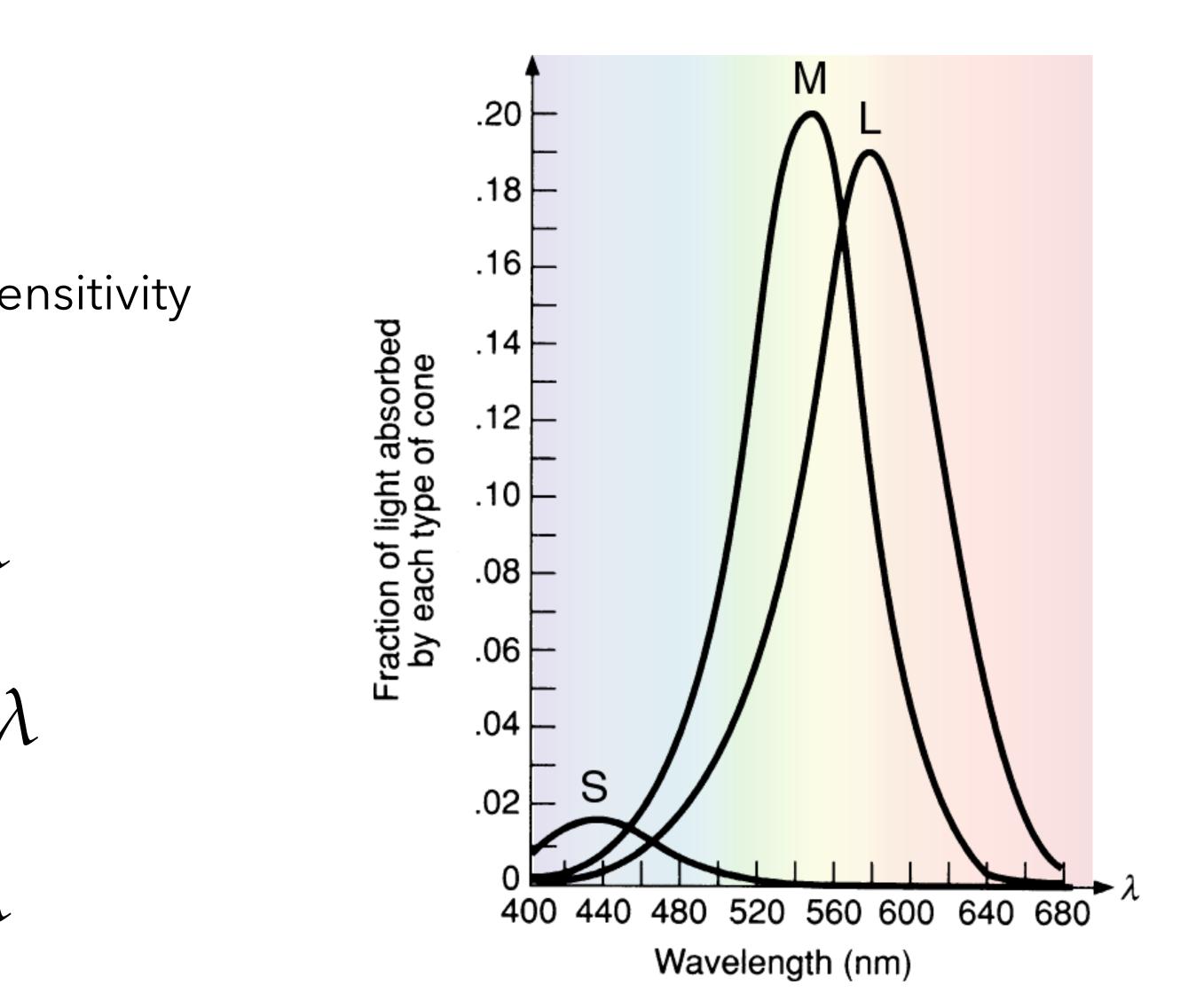




## Light detection math

 $X = \int s(\lambda) r(\lambda) d\lambda$ detector's sensitivity measured signal input spectrum  $S = \int r_S(\lambda) \, s(\lambda) \, \mathrm{d}\lambda$  $M = \int r_M(\lambda) \, s(\lambda) \, d\lambda$  $L = \int r_L(\lambda) s(\lambda) d\lambda$ 

After a slide by Steve Marschner

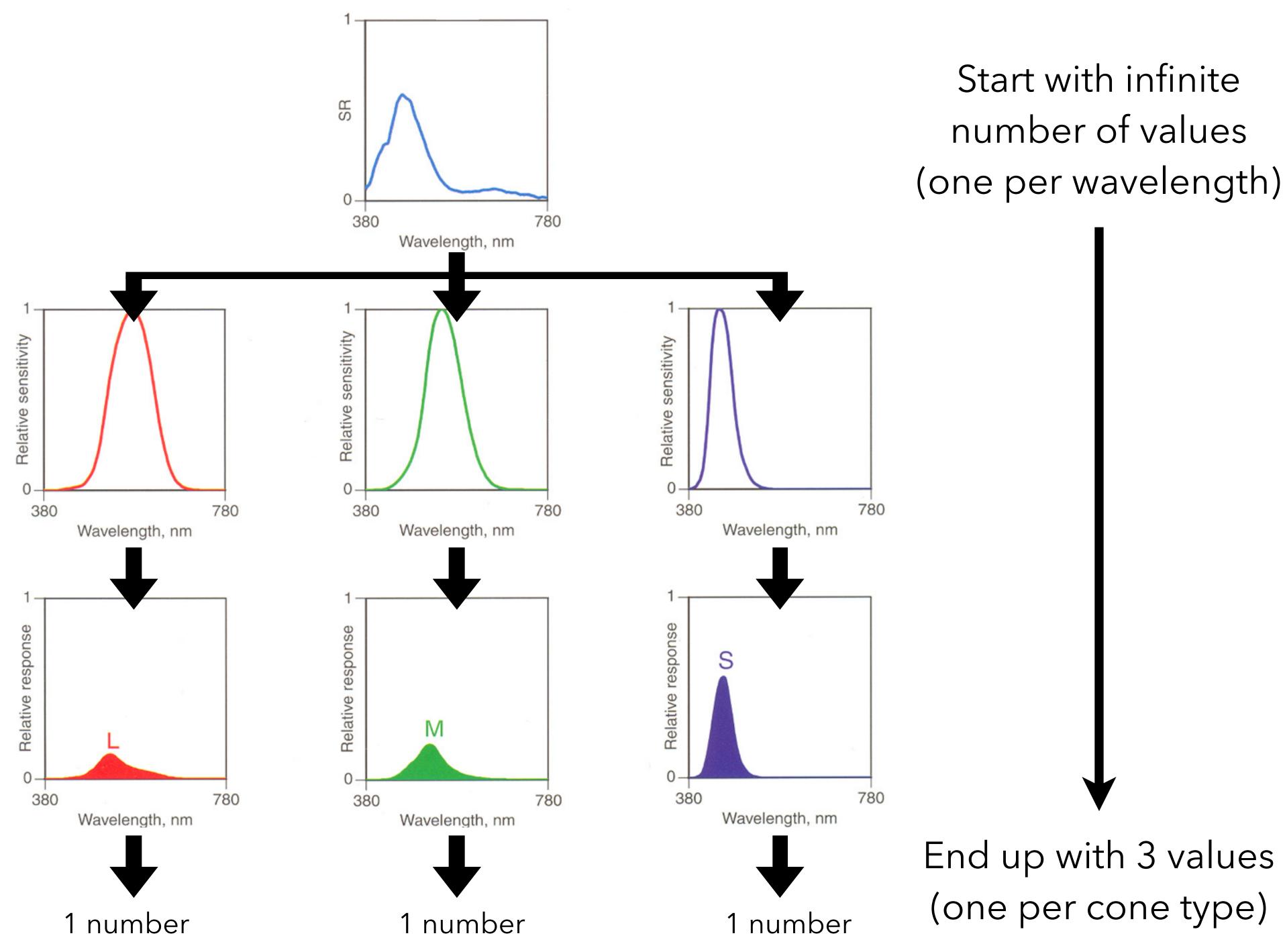




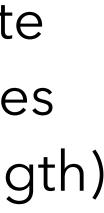
### Stimulus (arbitrary spectrum)



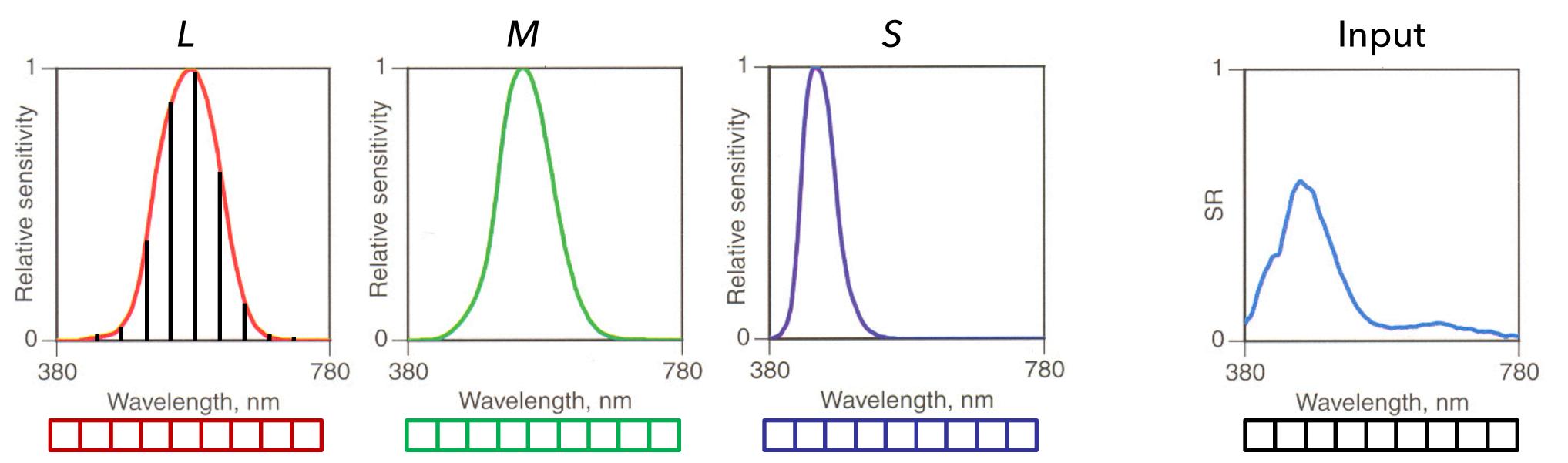
Multiply



Integrate



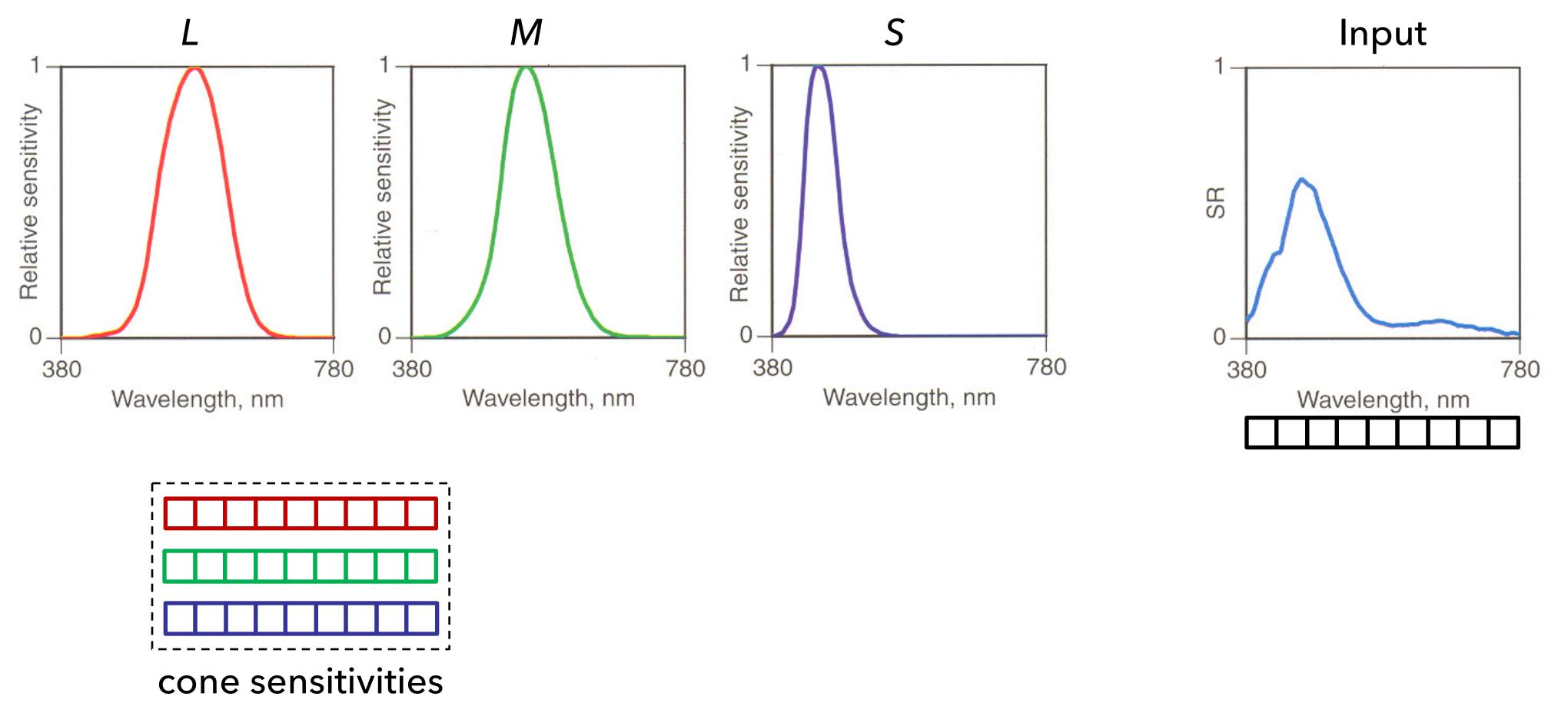




After a slide by Matthias Zwicker



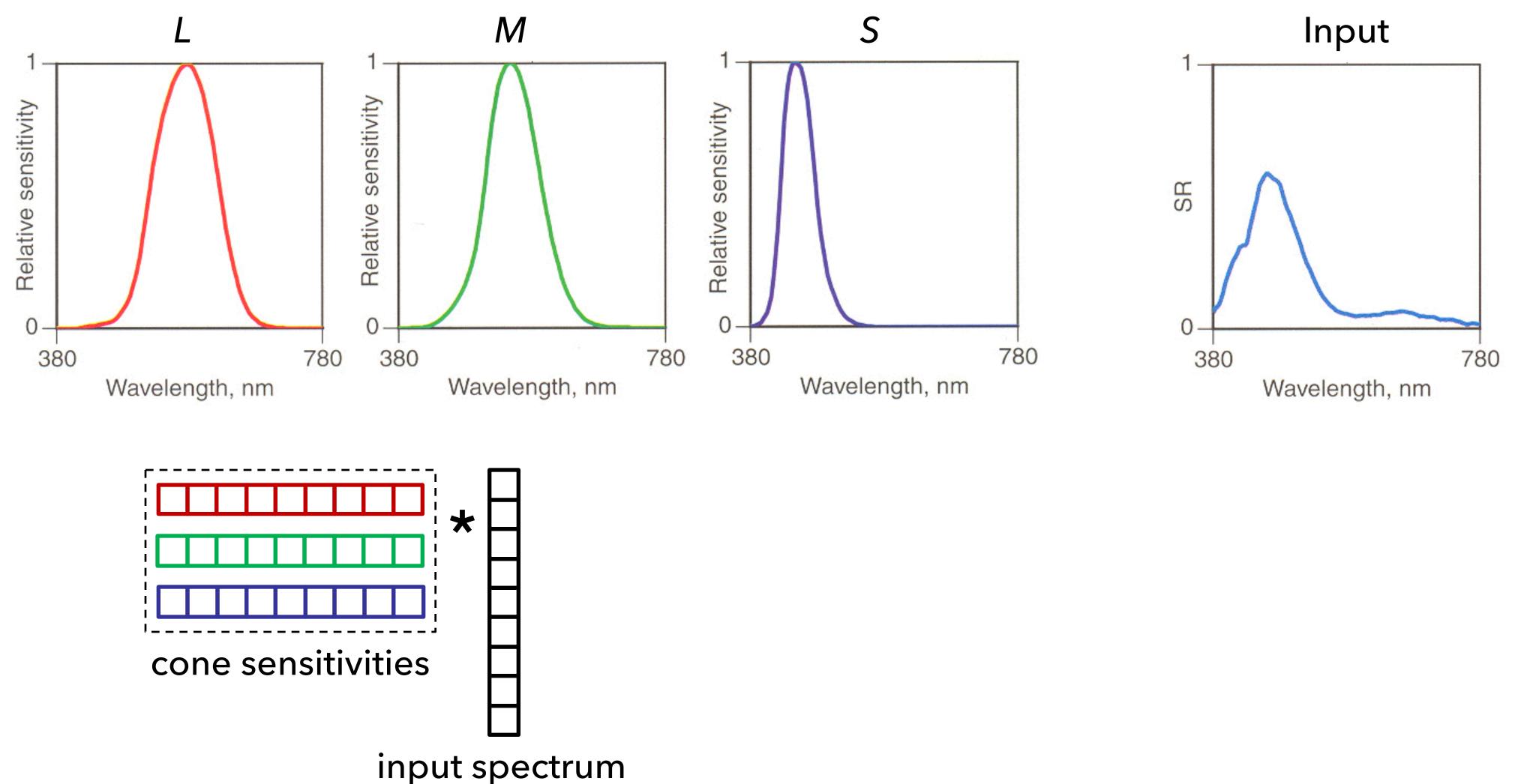




After a slide by Matthias Zwicker

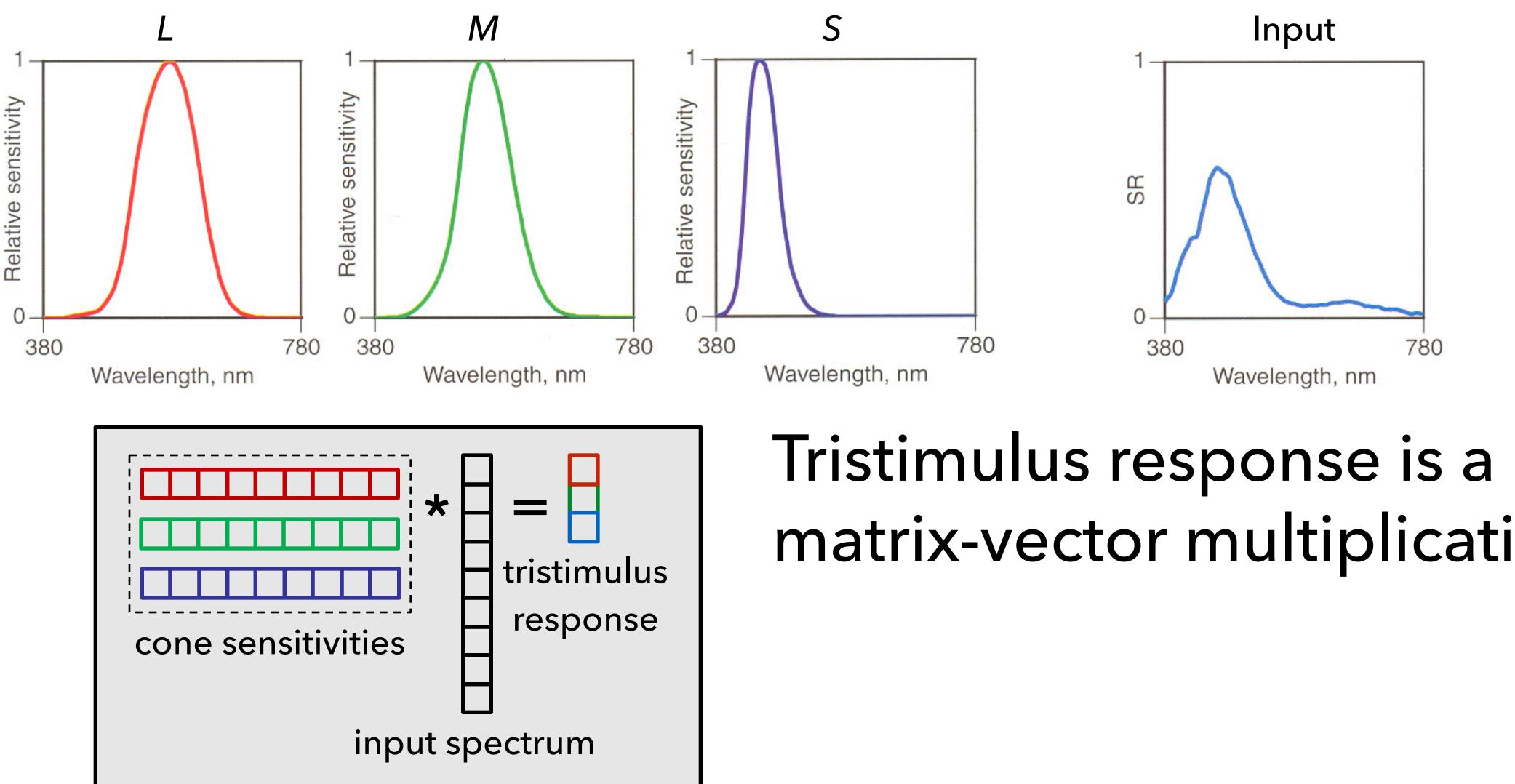






After a slide by Matthias Zwicker





After a slide by Matthias Zwicker

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# matrix-vector multiplication



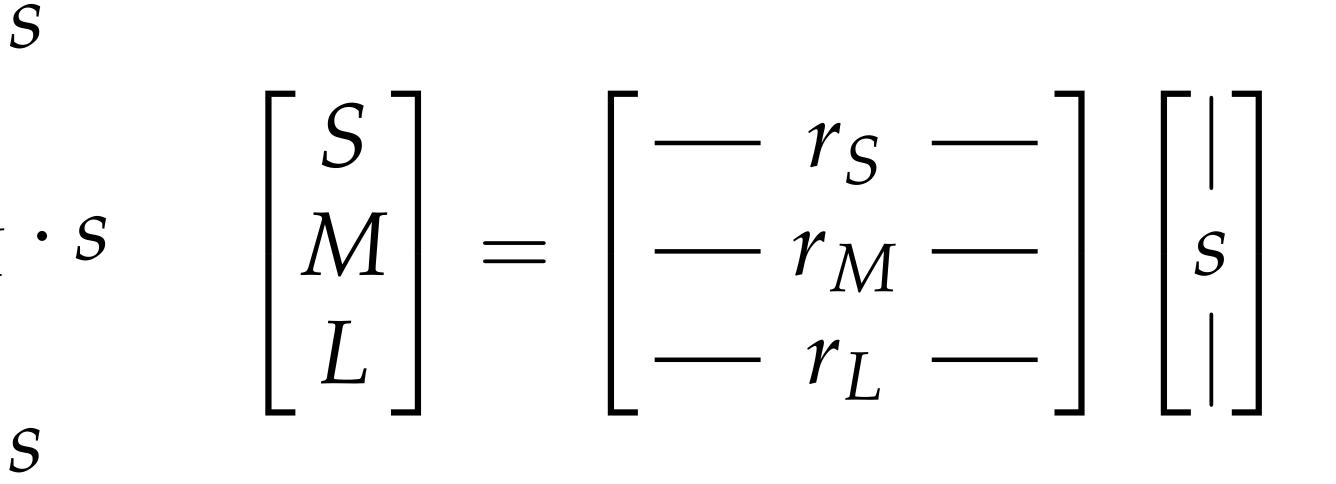


### **Cone responses to a spectrum** *s*

Integral notation:

$$S = \int r_{S}(\lambda) s(\lambda) d\lambda = r_{S} \cdot$$
$$M = \int r_{M}(\lambda) s(\lambda) d\lambda = r_{M}$$
$$L = \int r_{L}(\lambda) s(\lambda) d\lambda = r_{L} \cdot$$

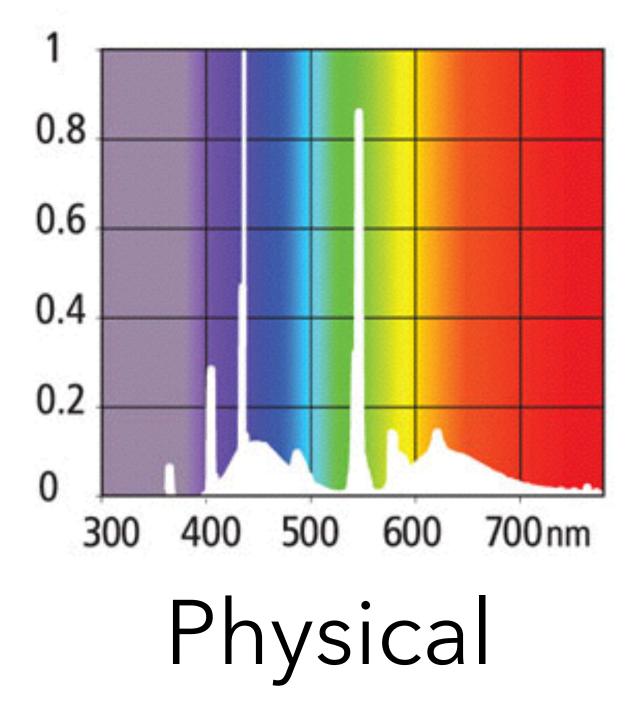
Matrix notation:



### $r_S$ , $r_M$ and $r_L$ are N-dimensional vectors, where $N = \infty$

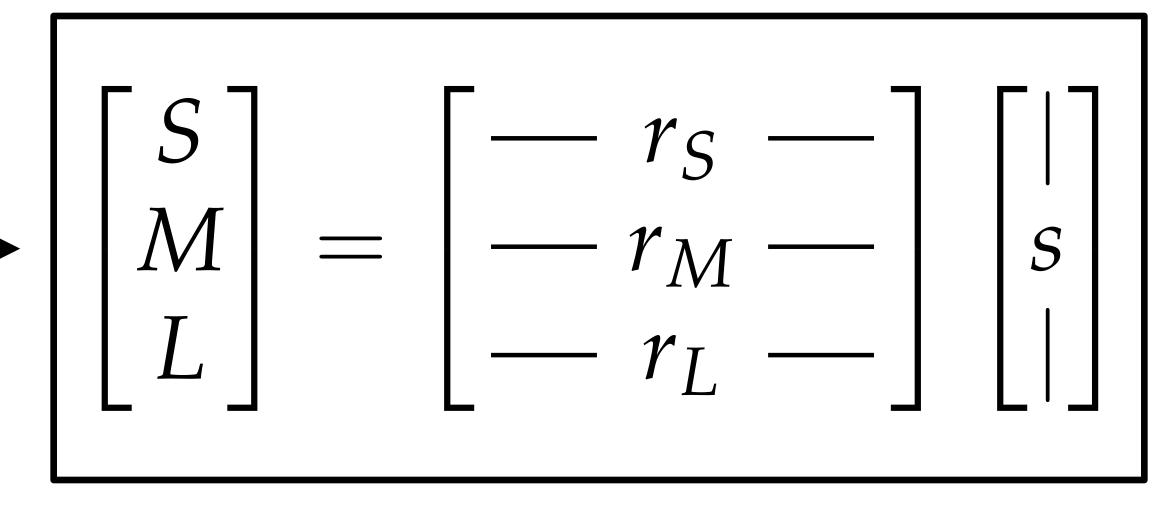


## Physical light to perceptual color



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### Perceptual



## Basic fact of colorimetry

- Take a spectrum (which is an infinity of numbers)
- Eye produces three numbers (a projection to 3D)
- This throws away a lot of information!
- many spectra can produce same S, M, L tristimulus values!
- metamers
- affected by illuminant

After a slide by Steve Marschner



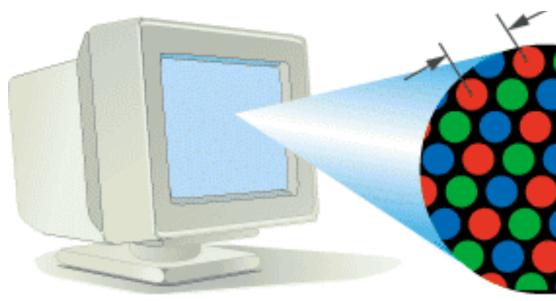




## Warning: tricky thing with color

- Cone responses overlap & are not orthogonal!
- Basis functions for analysis
- eyes, cameras
- are different than for synthesis
- lights, monitors

The RGB in your camera is different than the RGB in your monitor!











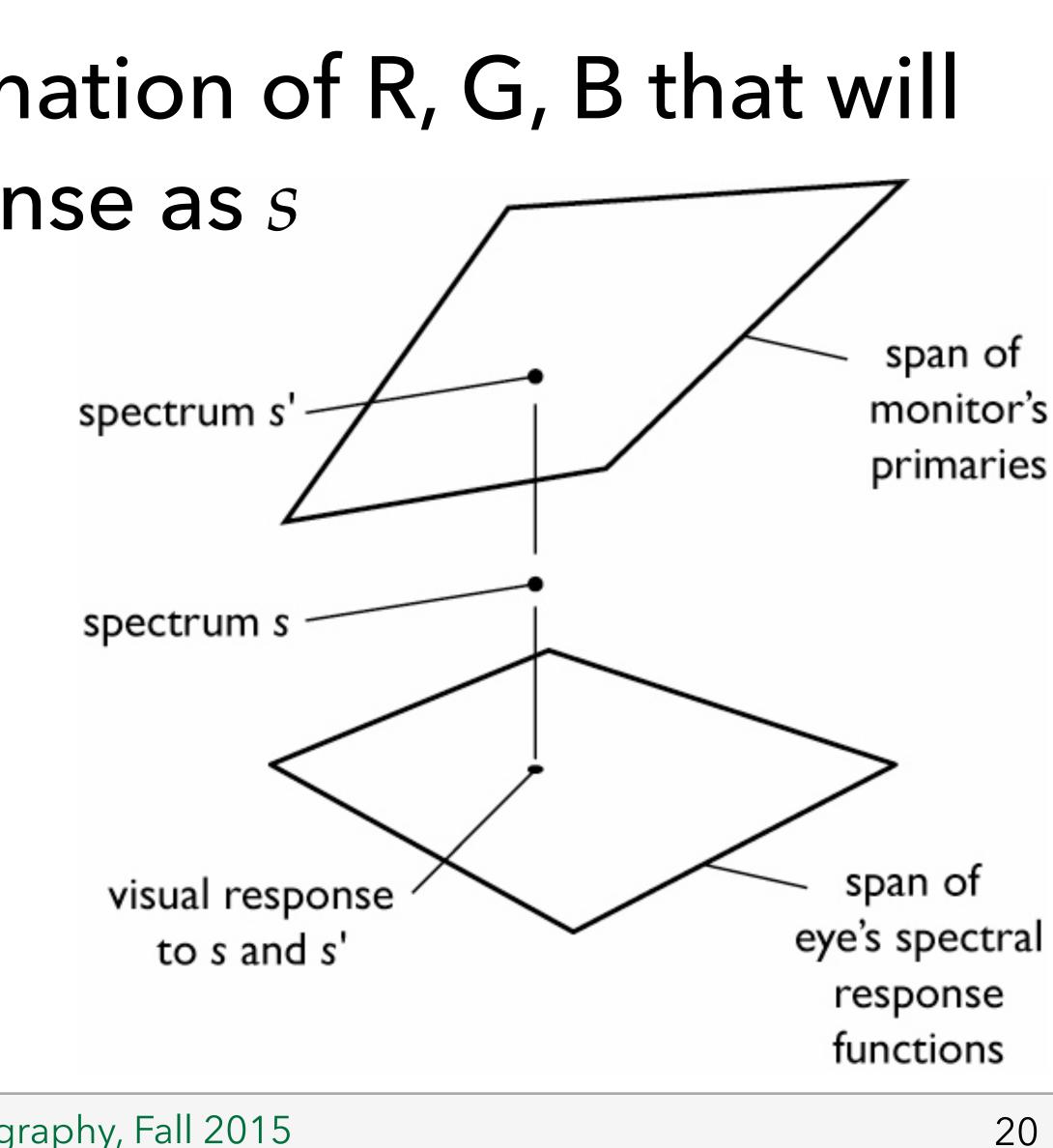
## **Color reproduction (the right way)**

project to the same visual response as s

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### We want to compute the combination of R, G, B that will



### What color do we see when we look at the display?

- Feed C to display

After a slide by Steve Marschner



What color do we see when we look at the display?

- Feed C to display
- Display produces *s*<sub>a</sub>



After a slide by Steve Marschner



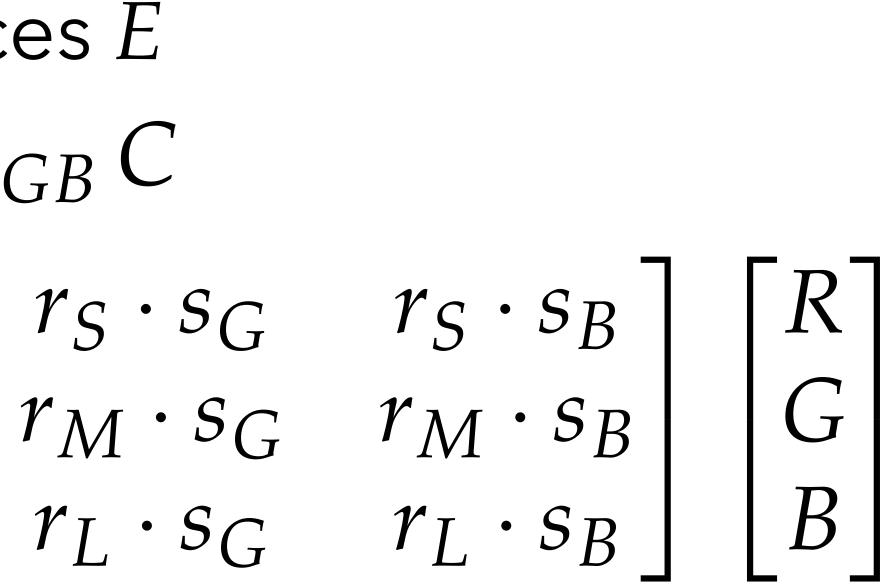
What color do we see when we look at the display?

- Feed C to display
- Display produces s<sub>a</sub>
- Eye looks at s<sub>a</sub> and produces E

$$E = M_{SML} M_{RC}$$

$$I = \begin{array}{c} r_S \cdot s_R \\ r_M \cdot s_R \\ r_L \cdot s_R \end{array}$$

After a slide by Steve Marschner





same:  $M_{SML} s = M_{SML} s_a$ 

### Substitute in expression for $s_a$ ,

$$M_{SML} s = M_{SML} \lambda$$

$$C = (M_{SML} M_{RGB})^{-}$$

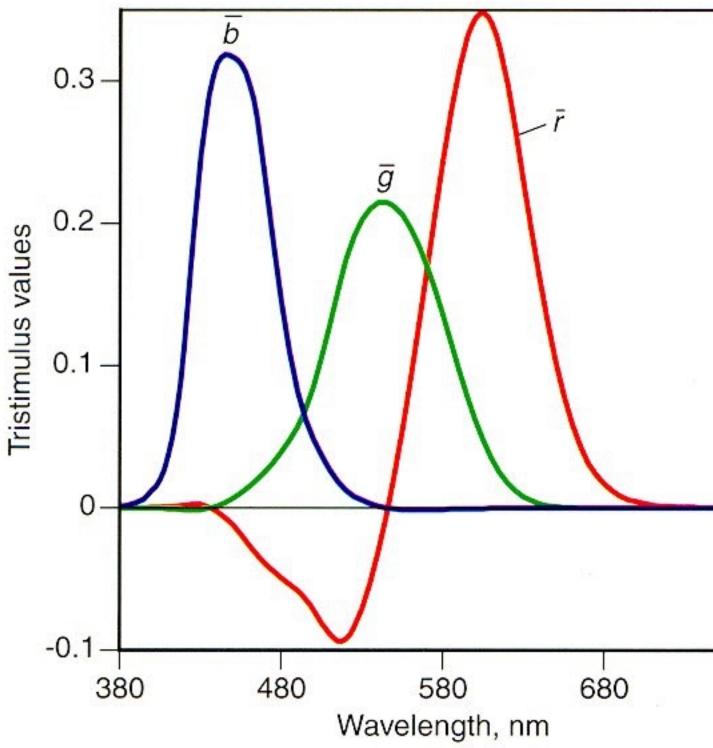
color matching matrix for RGB

After a slide by Steve Marschner

### Goal of reproduction: visual response to s and s<sub>a</sub> is the

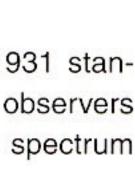
 $M_{RGB}C$ 

 $^{-1}M_{SML}s$ 



These curves are the color-matching functions for the 1931 standard observer, The average results of 17 color-normal observers having matched each wavelength of the equal-energy spectrum with primaries of 435.8 nm, 546.1 nm, and 700 nm.



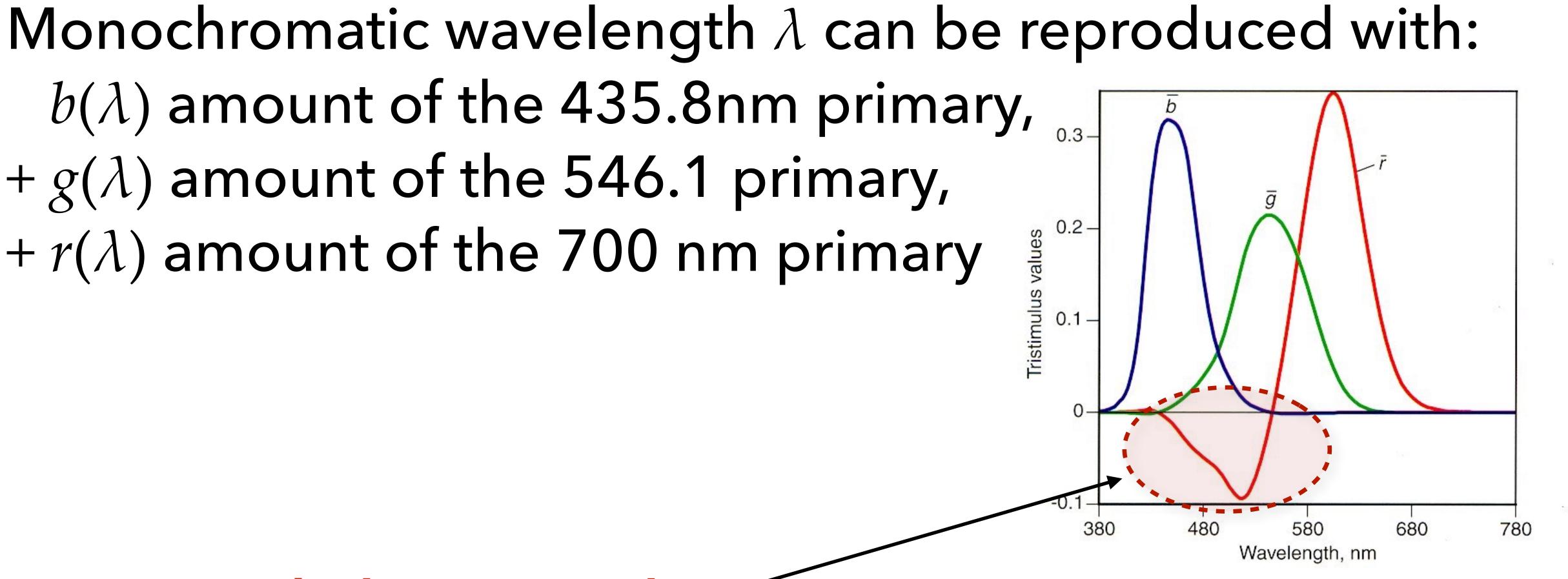




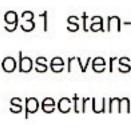
### Meaning of these curves/rows

 $b(\lambda)$  amount of the 435.8nm primary, +  $g(\lambda)$  amount of the 546.1 primary, +  $r(\lambda)$  amount of the 700 nm primary

### **Negative light required?**



These curves are the color-matching functions for the 1931 standard observer, The average results of 17 color-normal observers having matched each wavelength of the equal-energy spectrum with primaries of 435.8 nm, 546.1 nm, and 700 nm.

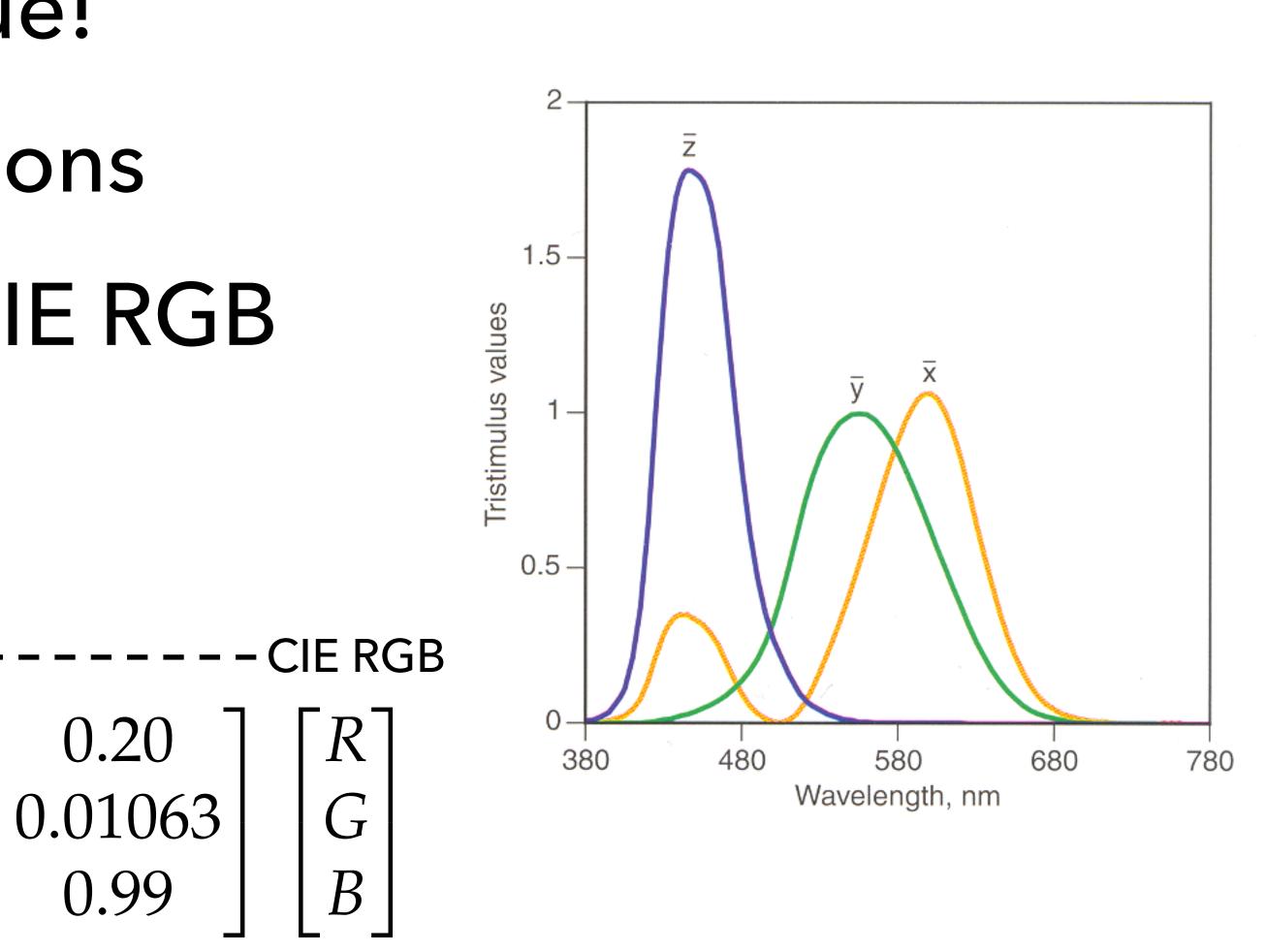




## CIE XYZ color space

Linear algebra to the rescue! Purely positive basis functions Linear transformation of CIE RGB Non-physical primaries

CIE XYZ < 0.490.310.200.176970.812400.01063 0.176970.01 0.00





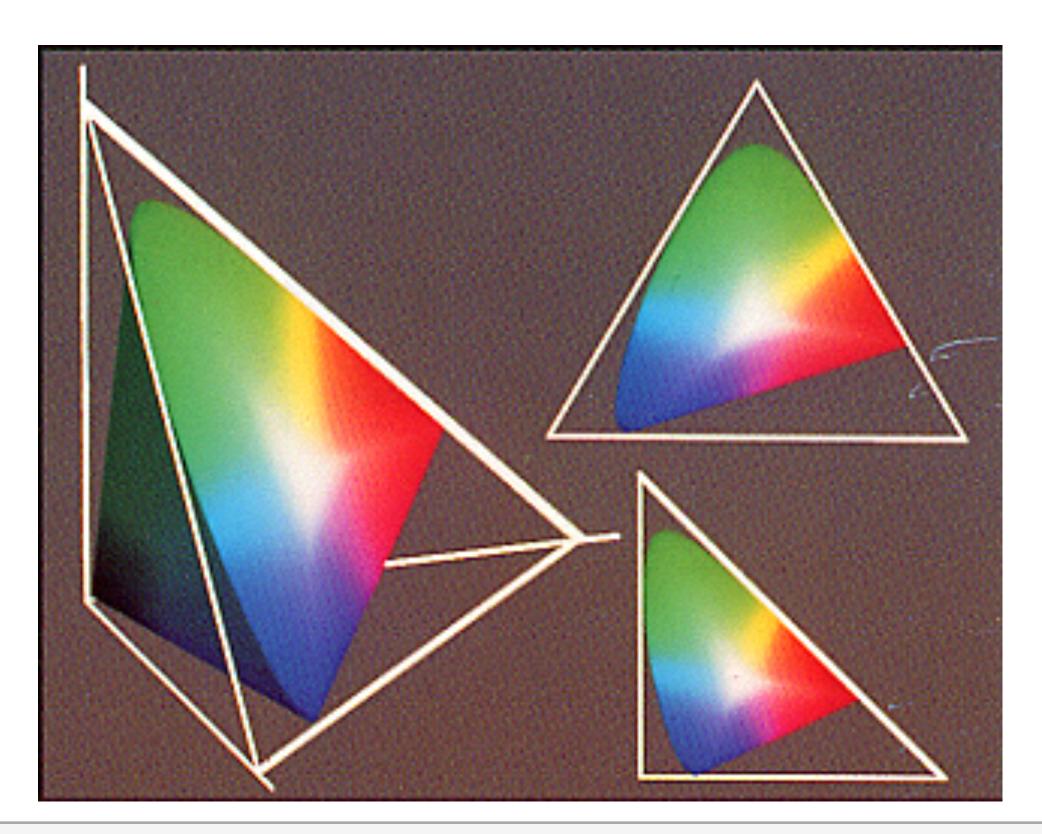
### **CIE XYZ color cone**

3D spaces can be hard to visualize

Chrominance is our notion of color, as opposed to brightness/luminance

Recall that our eyes correct for multiplicative scale factors

- discount light intensity





# Chromaticity Diagram

## The CIE xyY Color Space

 $x = \frac{X}{X + Y + Z}$ 

Chromaticity (x,y) can be derived by normalizing the XYZ color components:

- (x,y) characterize color
- Y characterizes brightness
- Combining xy with Y allows us to represent any color brightness

$$y = \frac{Y}{X + Y + Z}$$

# Plotting on xy plane allows us to see all colors of a single

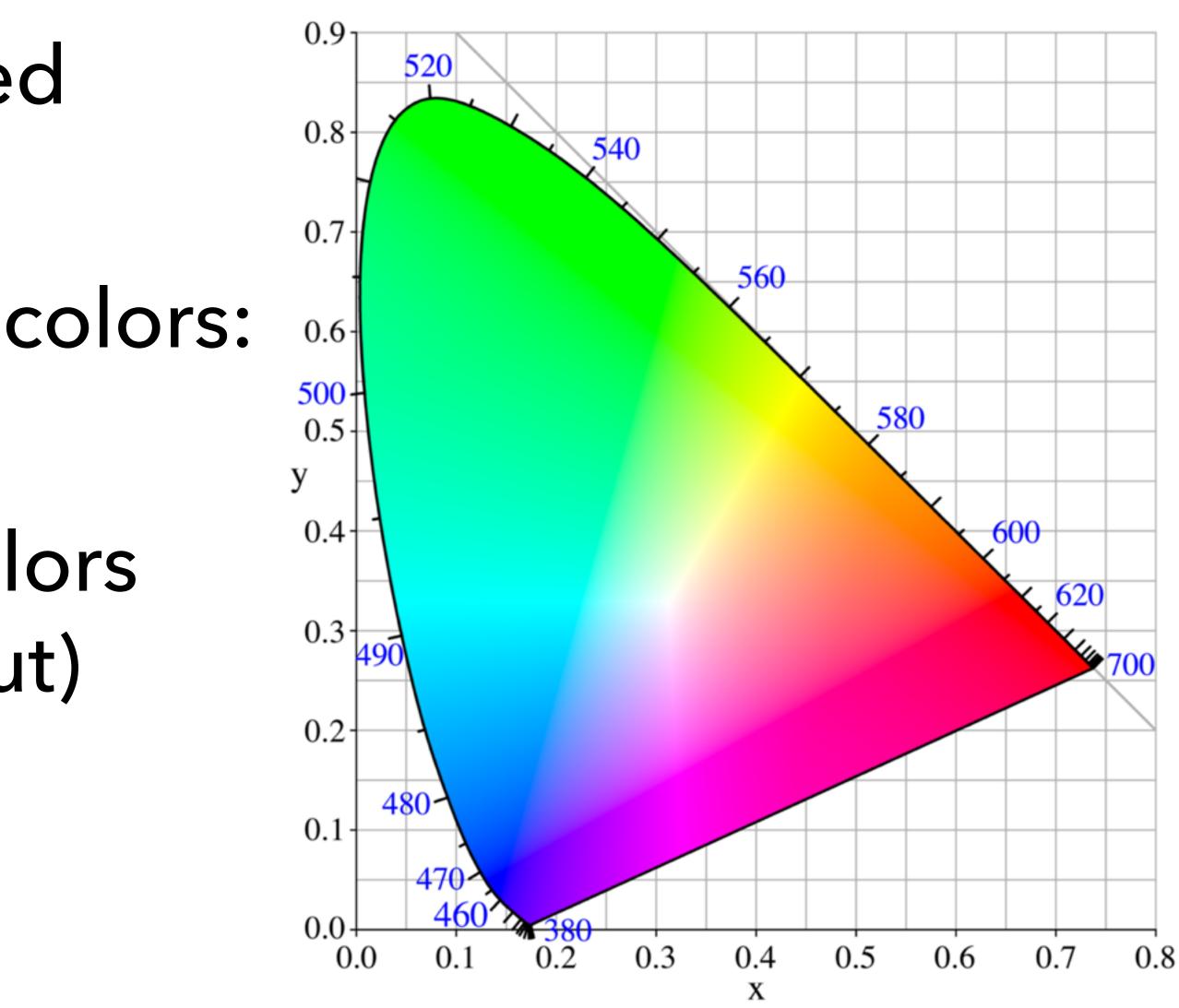


## **CIE Chromaticity Chart**

Spectral colors along curved boundary

Linear combination of two colors: line connecting two points

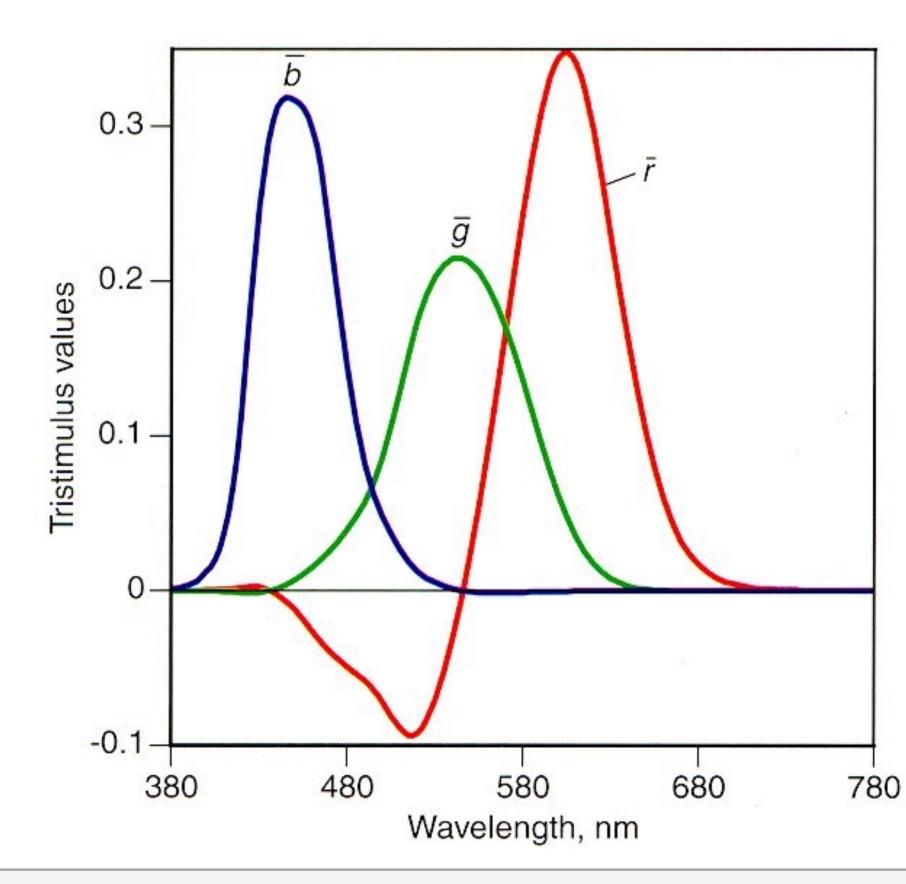
Linear combination of 3 colors span a triangle (color gamut)

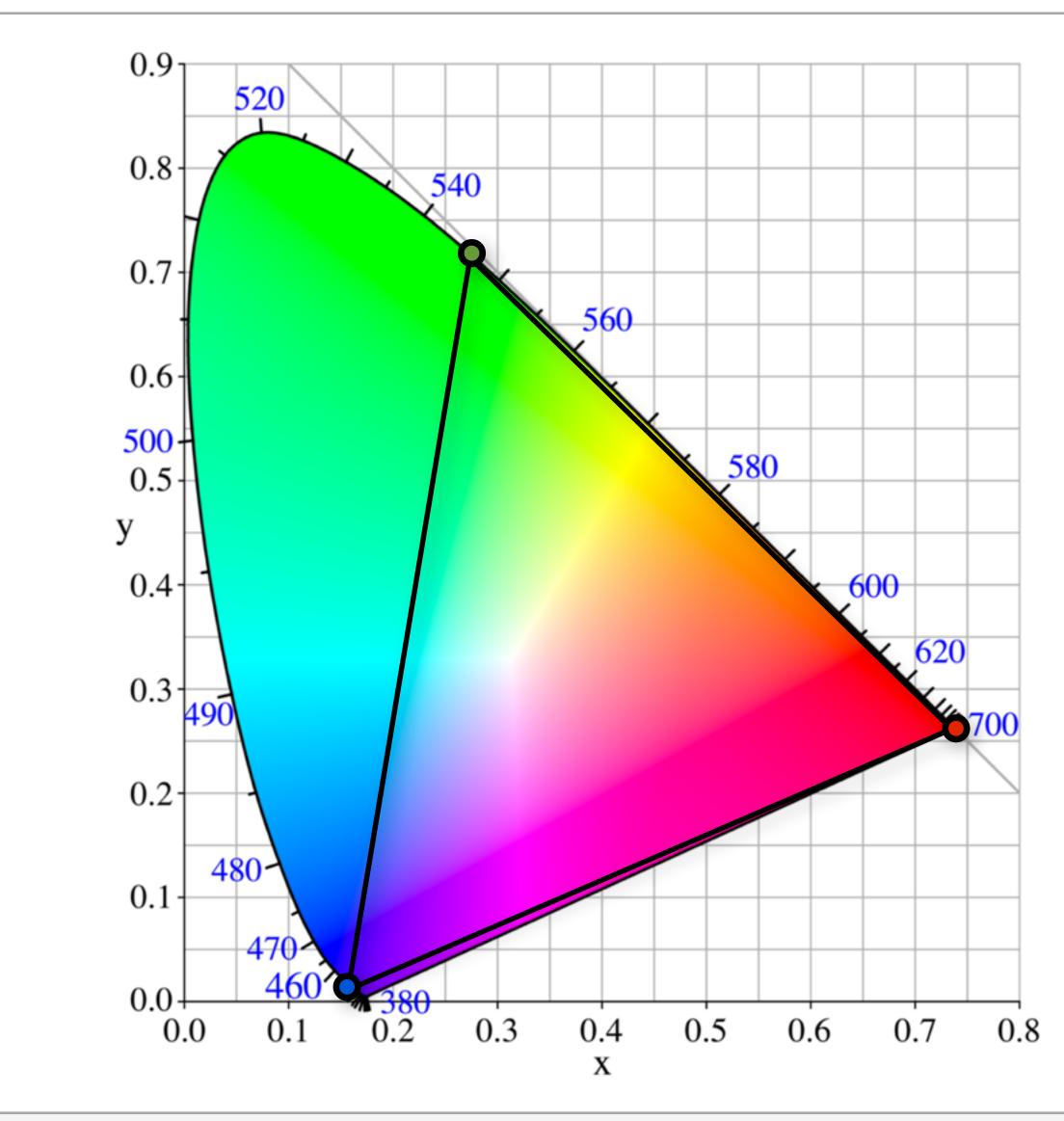




### CIE RGB Color Space

# Color primaries at: 435.8, 546.1, 700.0 nm

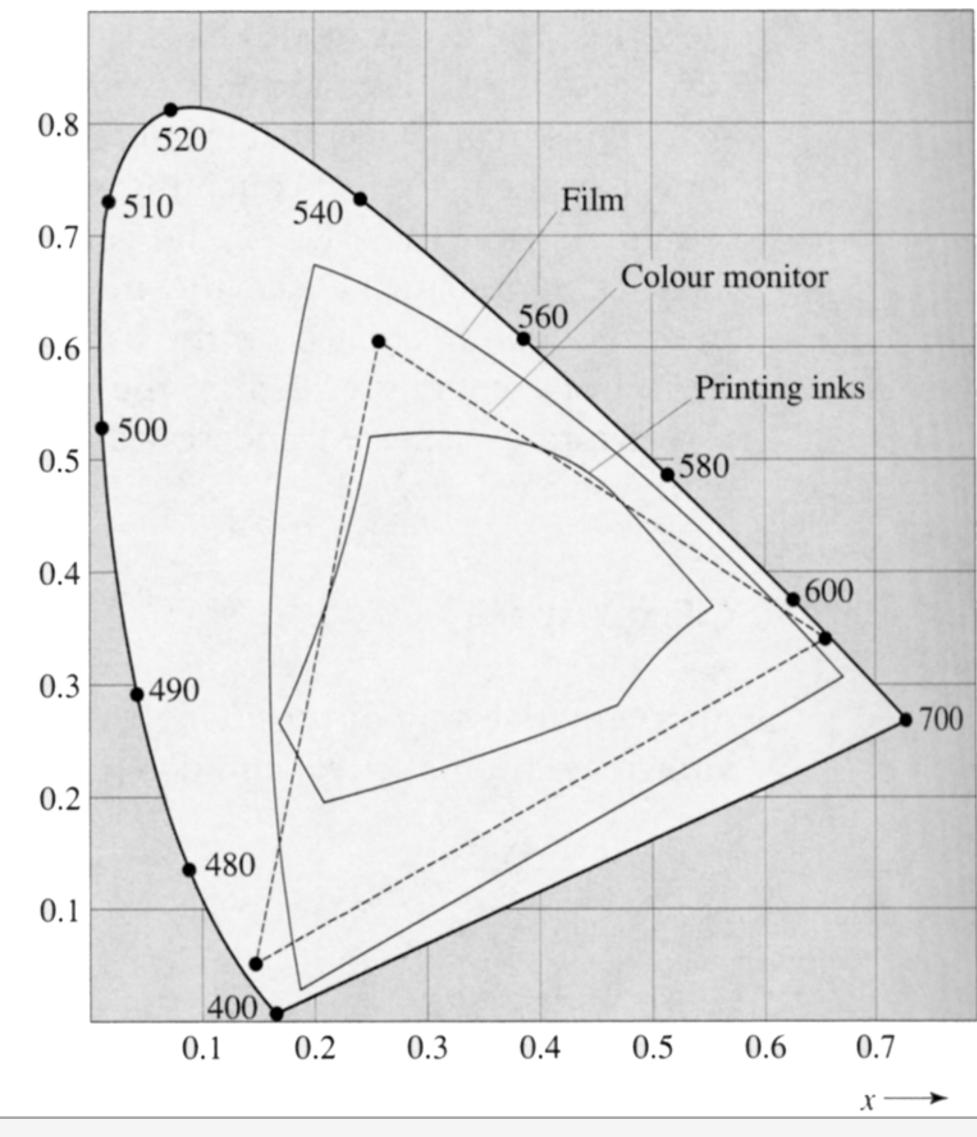






### Color Gamuts

у



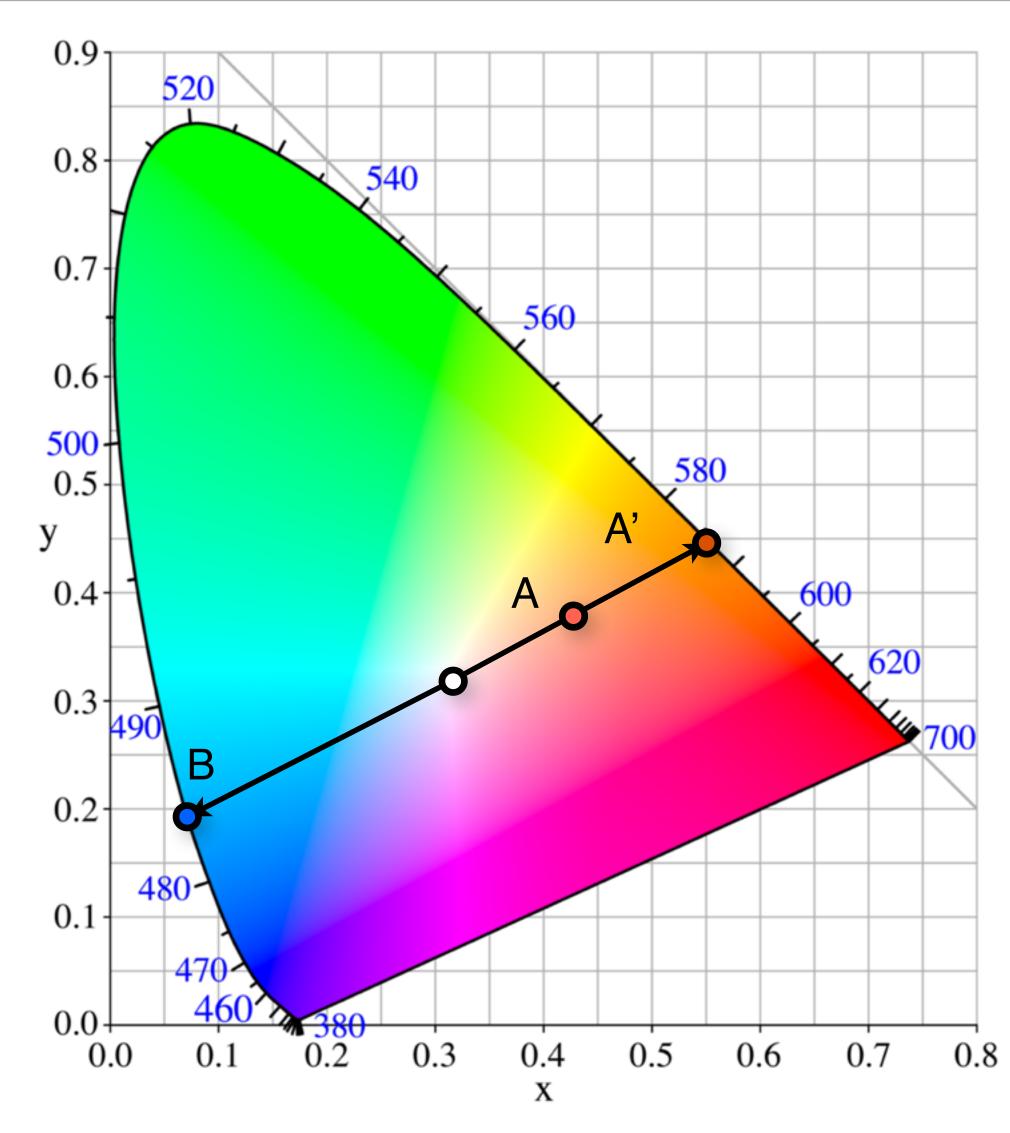
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### **CIE Chromaticity Chart Features**

### White Point

### Dominant wavelength Inverse color





## **Perceptually-Uniform Color Spaces**

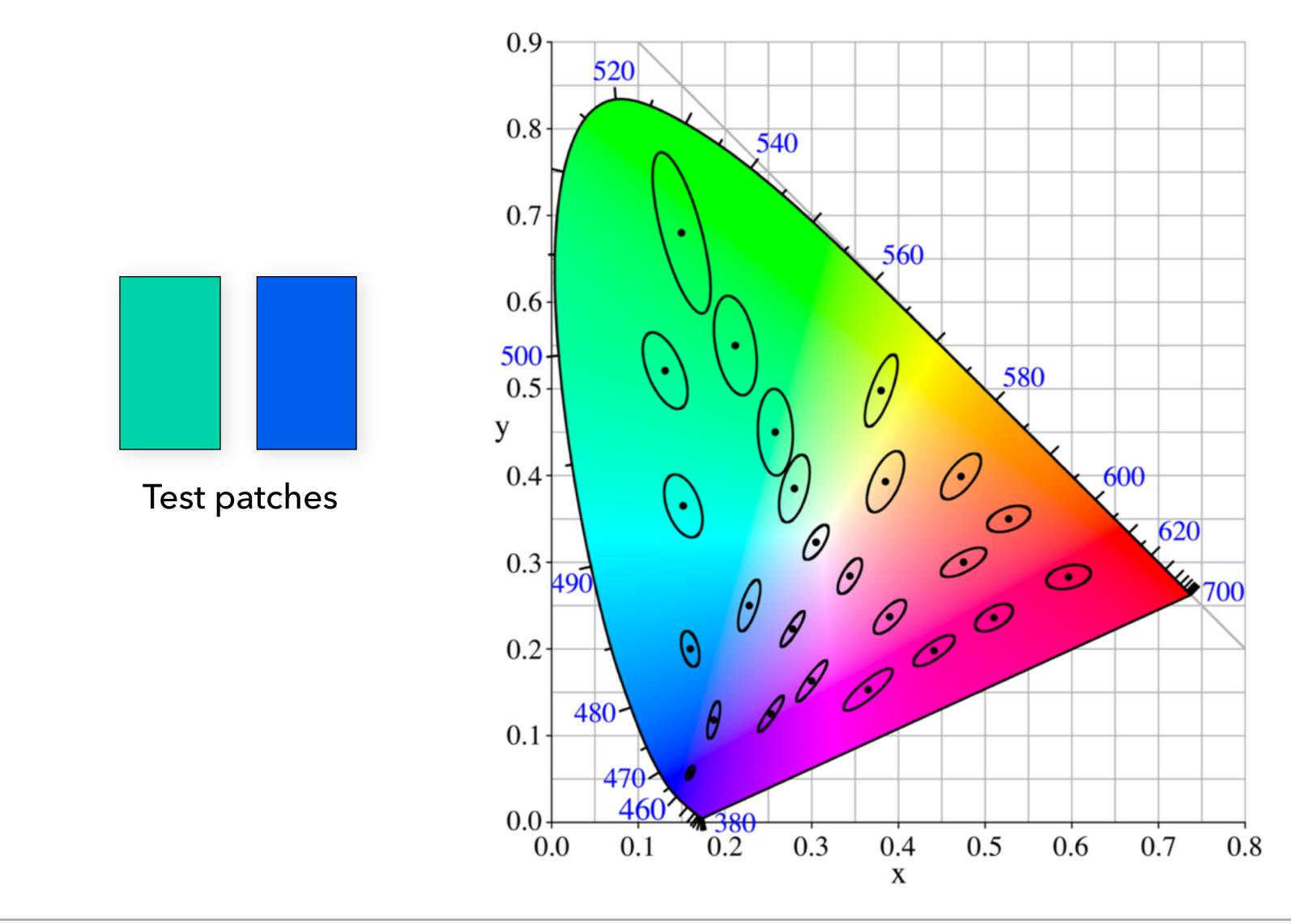
All these color spaces so far are perceptually nonuniform:

- two colors close together in space are not necessarily visually similar
- two colors far apart are not necessarily very different! Measuring "perceptual distance" in color spaces important for many industries

Experiments by MacAdams



### MacAdams Color Ellipses



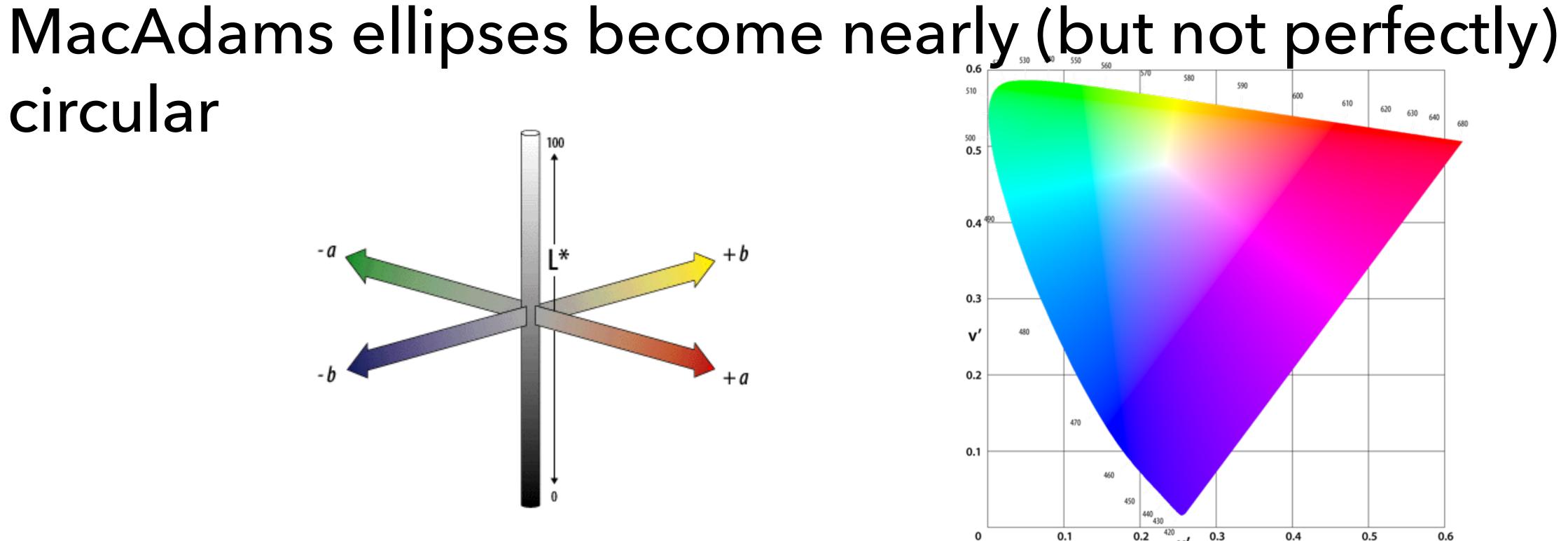
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## **CIELab and CIELuv Color Spaces**

Two attempts to make a perceptually-uniform color space

circular



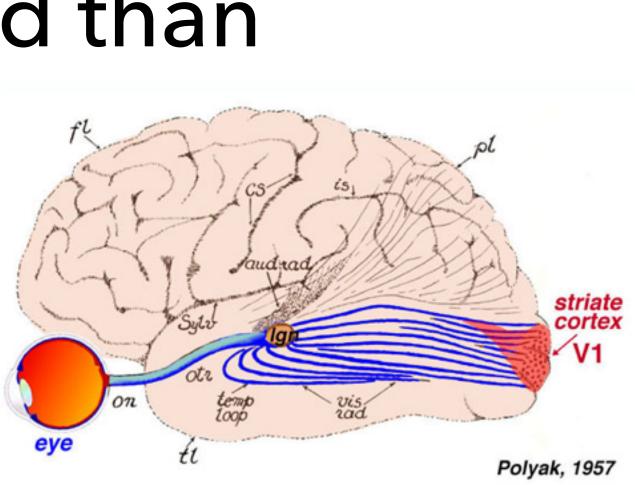


# Higher-level color perception

# Higher-level color perception

Color perception is much more complicated than response of SML cones...

- Visual pathway
- A lot happens after the cones
- But: cone responses are input to further processing





## Color constancy

Also known as chromatic adaptation

Color of object is perceived as the same even under varying illumination

## For example:

- A white sheet of paper under green illumination is still

perceived as white, even though the reflected light is green! The human brain infers the white color from the context, which is "green-ish" too because of the green illumination.



## Color constancy









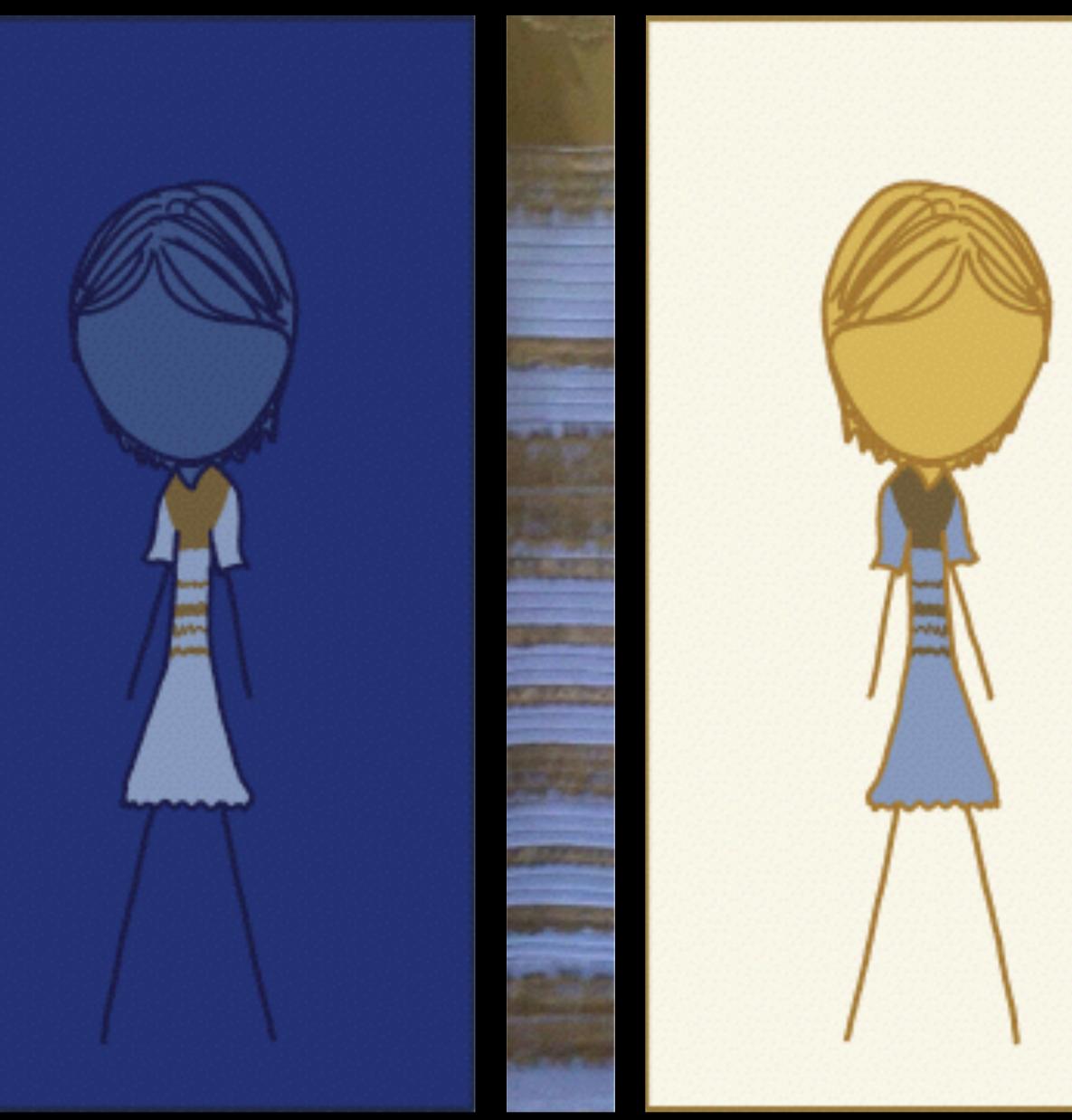
# blue and black?

white and gold?





## Color constancy failure



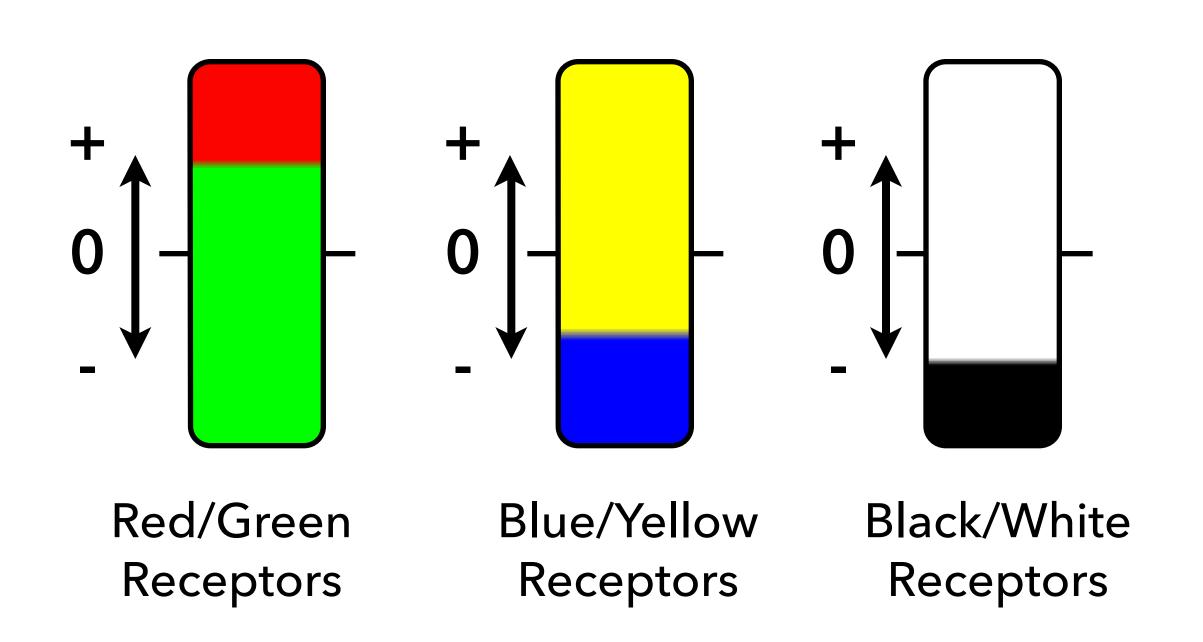
http://xkcd.com/1492/

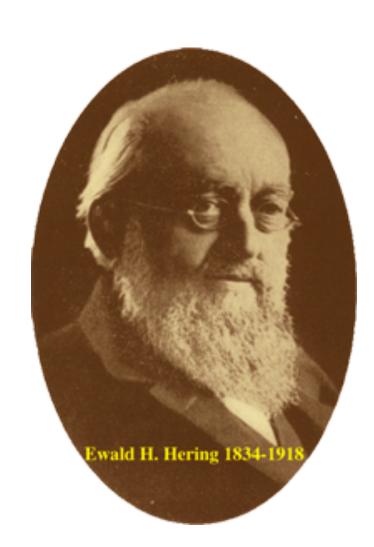


## Hering's opponent process theory (1874)

After sensing by cones, colors are encoded as red versus green, blue versus yellow, and black versus white

Physiological evidence found in the 1950s



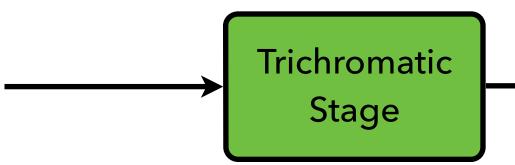




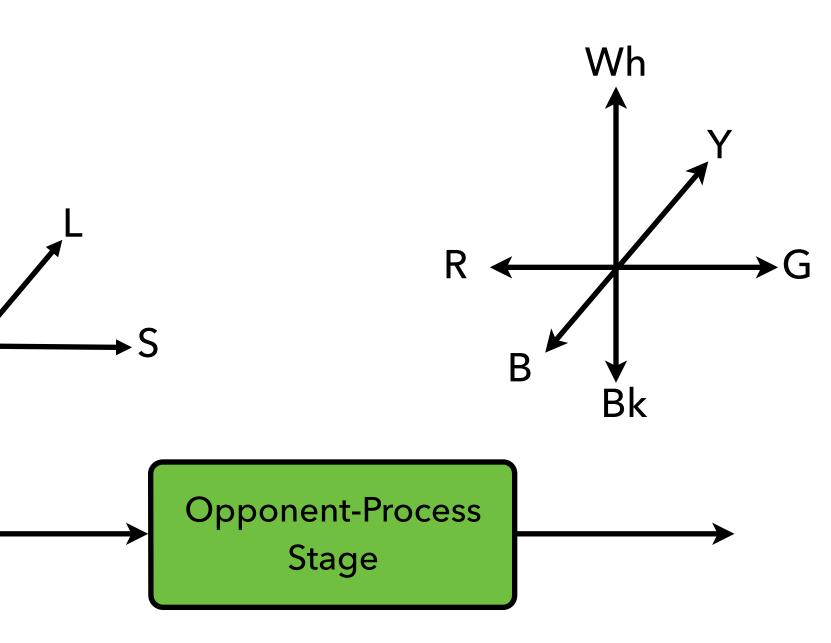


## Dual process theory

- Inputs are LMS cone responses
- Output has a different parameterization:
- Light-dark
- Blue-yellow
- Red-green



Μ





# Color opponents wiring

Sums for brightness

Differences for color opponents

At the end, it's just a 3x3 matrix compared to LMS

*First zone (or stage):* layer of retina with three independent types of cones Second zone (or stage): signals from cones either excite or inhibit second layer of 0 0 neurons, producing blue or yellow red or green opponent signals









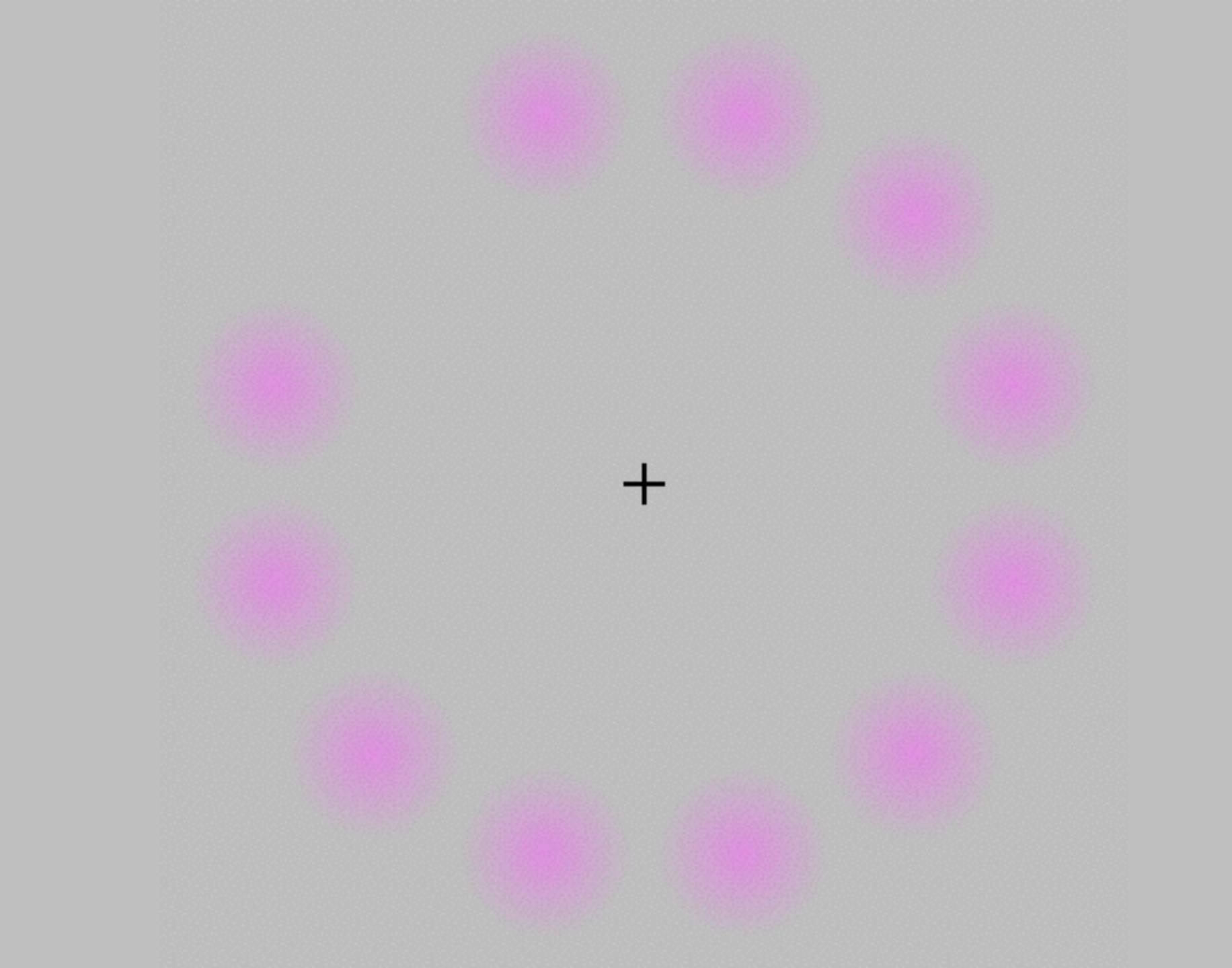
### Image



## **Opponent Colors**

## Afterimage





# Fixate on the + for > 30s. Blinking is OK.

When you eventually move your eyes the image will suddenly reappear or become more vibrant if it didn't disappear.







C Wojciech Jarosz



## **Opponent color spaces**

Luminance, red-green, blue-yellow CIELab

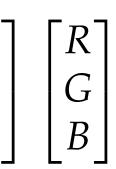
YUV  $\begin{bmatrix} Y' \\ U \\ V \end{bmatrix} = \begin{bmatrix} 0.299 & 0.587 & 0.114 \\ -0.14713 & -0.28886 & 0.436 \\ 0.615 & -0.51499 & -0.10001 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$  $\begin{bmatrix} R \\ G \\ B \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1.13983 \\ 1 & -0.39465 & -0.58060 \\ 1 & 2.03211 & 0 \end{bmatrix} \begin{bmatrix} Y' \\ U \\ V \end{bmatrix}$ YcrCb

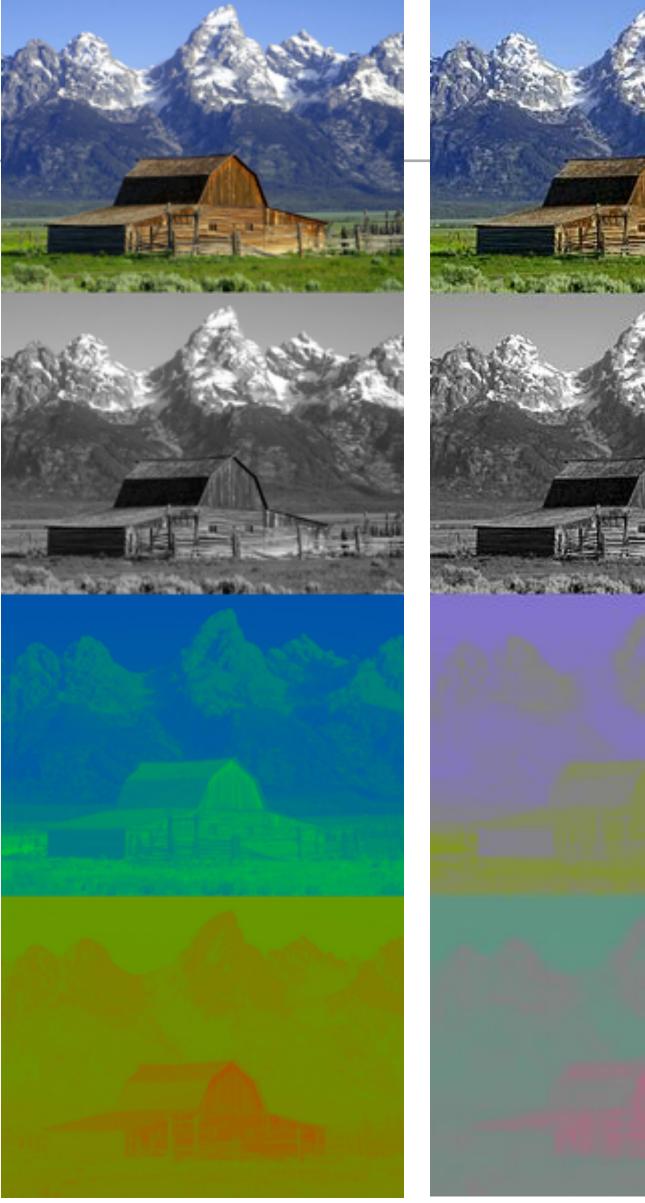
- used a lot in image/video compression

 $Y' = 0 + (0.299 \cdot R'_D) + (0.587 \cdot G'_D) + (0.114 \cdot B'_D)$ 

 $C_B = 128 - (0.168736 \cdot R'_D) - (0.331264 \cdot G'_D) + (0.5)$  $\cdot B'_D$ )

 $C_R = 128 + (0.5 \quad \cdot R'_D) - (0.418688 \cdot G'_D) - (0.081312 \cdot B'_D)$ 





YUV

YCrCb





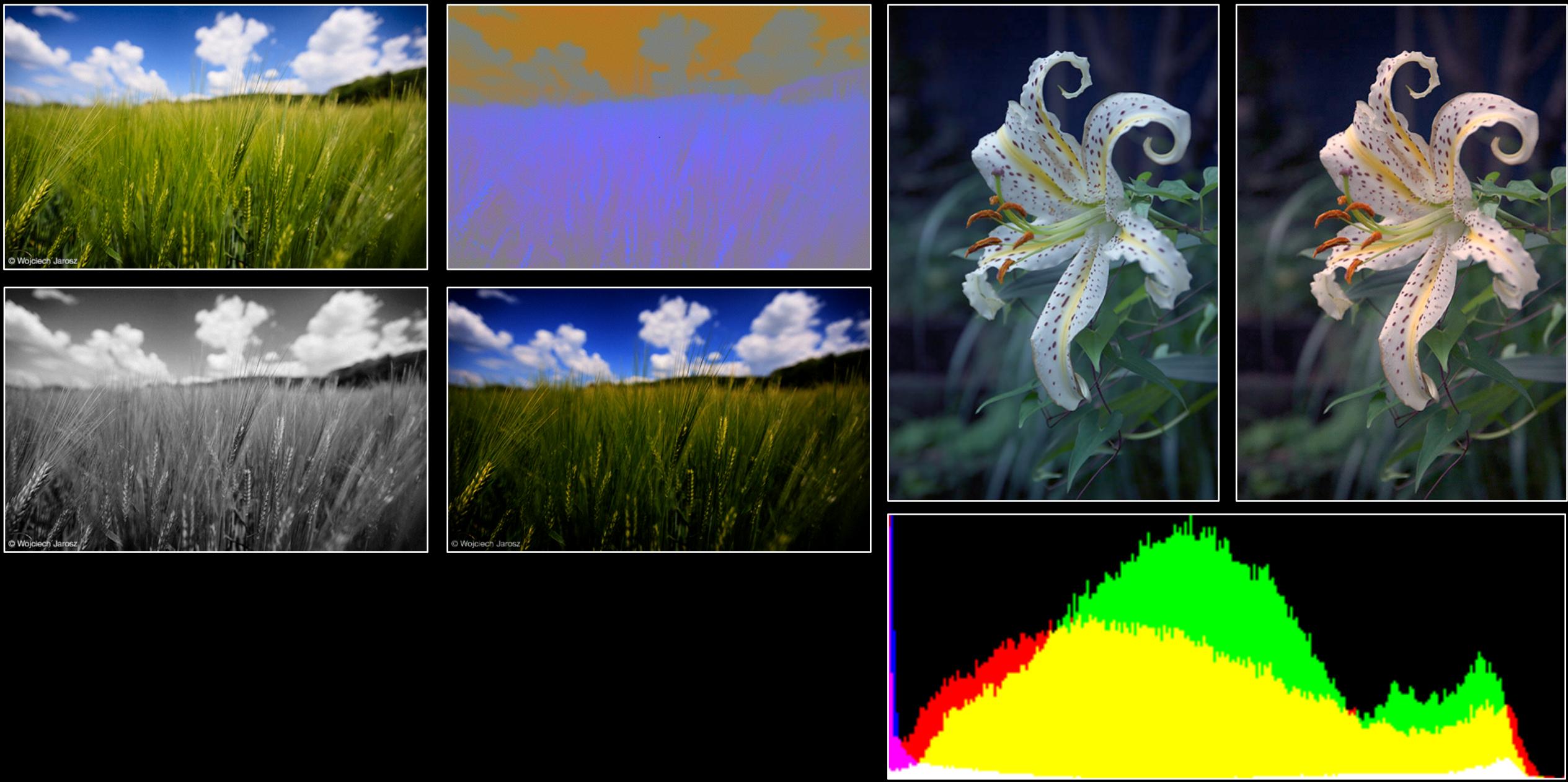


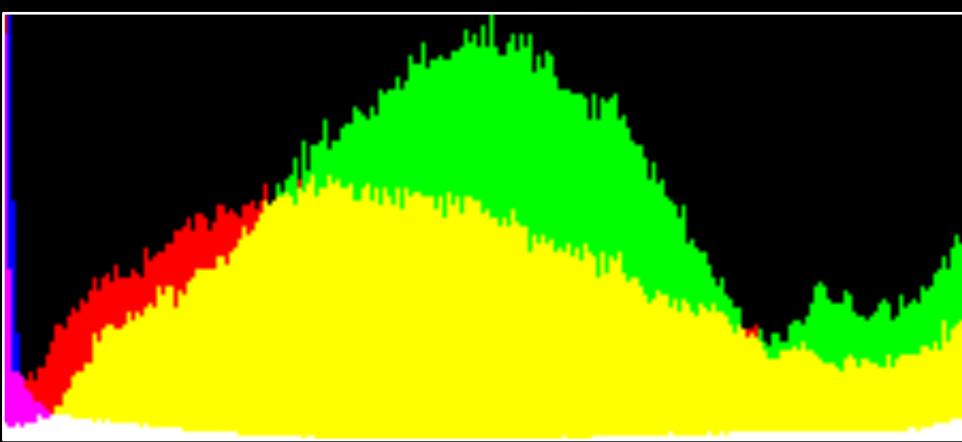
## Programming Assignment 2

- Gamma
- Colorspaces
- color2gray
- luminance-chrominance separation
- luminance-only brightness/ contrast
- RGB⇔YUV

- Spanish castle illusion Grayworld whitebalance Histograms
- autolevels
- visualize RGB histogram
- histogram equalization & histogram matching







## Slide credits

## Frédo Durand Steve Marschner Matthias Zwicker

