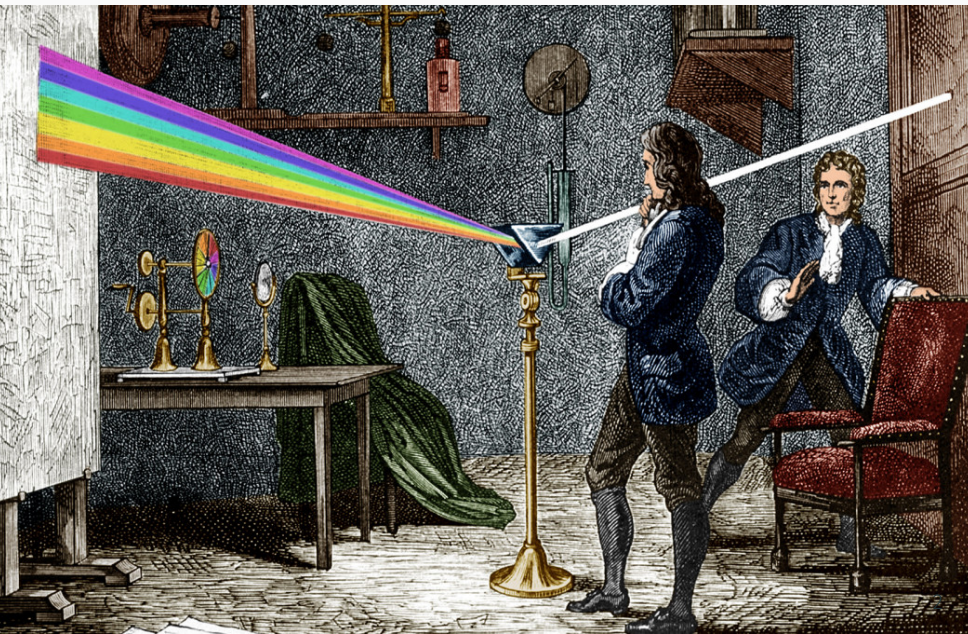


2D COMPUTER GRAPHICS

Diego Nehab
Summer 2020
IMPA

COLOR AND COMPOSITING

THE PRISM EXPERIMENT



MORE THAN VISIBLE LIGHT

Visible light: prism experiment (Newton, 1666)

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Infrared light: thermometers (Herschel, 1800)

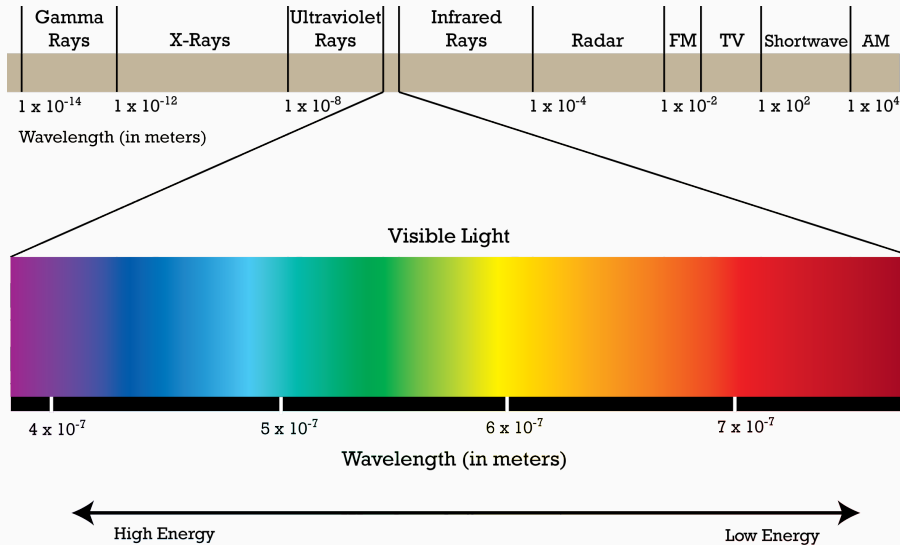
MORE THAN VISIBLE LIGHT

Visible light: prism experiment (Newton, 1666)

Infrared light: thermometers (Herschel, 1800)

Ultraviolet light: silver chloride (Ritter, 1801)

FULL ELECTROMAGNETIC SPECTRUM



Measurement of radiant energy in terms of absolute power

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Wave vs. particle

- Wavelength (λ), frequency ($\nu = \frac{c}{\lambda}$), and amplitude (A)
- Energy ($E = h\nu$, where h is Planck's constant) and flux (Φ)

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Pure spectral light (monochromatic colors)

RADIOMETRY

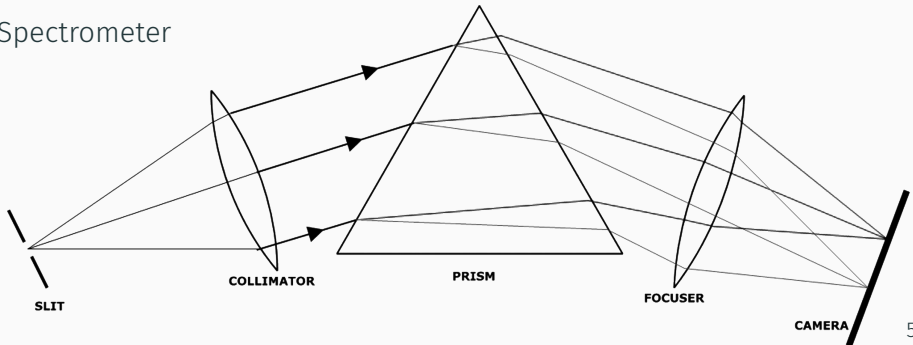
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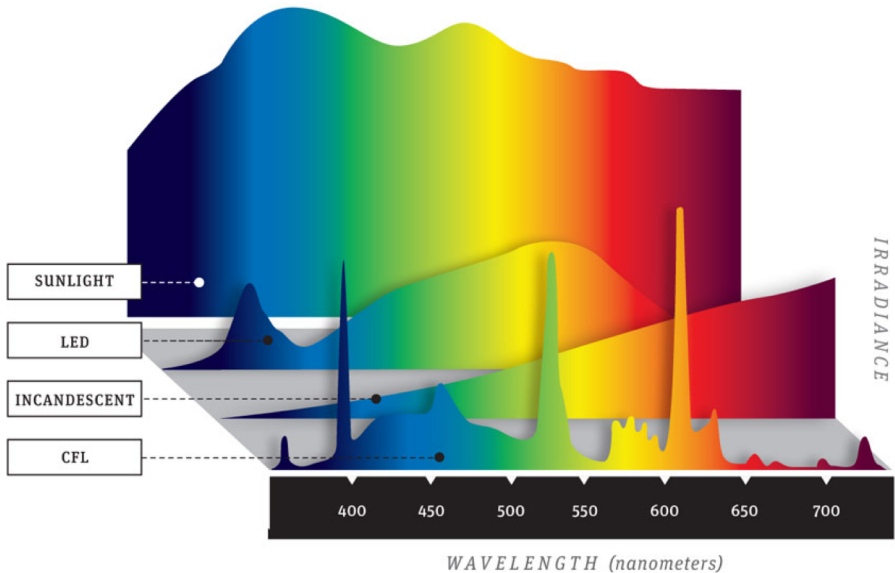
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Spectrometer



COLORS ARE SPECTRAL DISTRIBUTIONS



SPECTRAL REPRESENTATION

As a continuous function of $c(\lambda)$

$$c : \mathbf{R}_{>0} \rightarrow \mathbf{R}_{\geq 0}, \quad \lambda \mapsto A_\lambda$$

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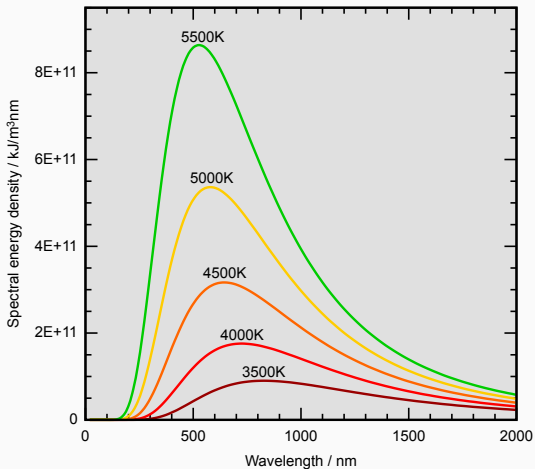
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Light emitter has a spectrum, material properties modulate the reflected spectrum (Fluorescence is something else)

BLACK-BODY RADIATION

$$B(\nu, T) = \frac{2h\nu^3}{c^2} \left(e^{\frac{h\nu}{k_B T}} - 1 \right)^{-1}, \quad \text{where } k_B \text{ is Boltzmann's constant}$$



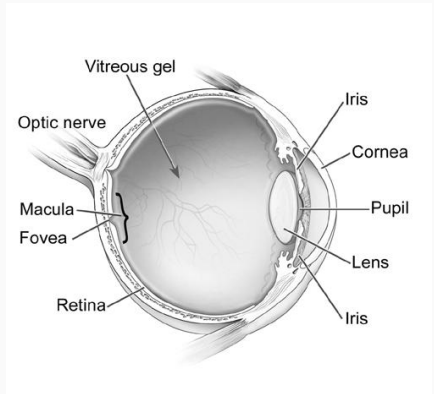
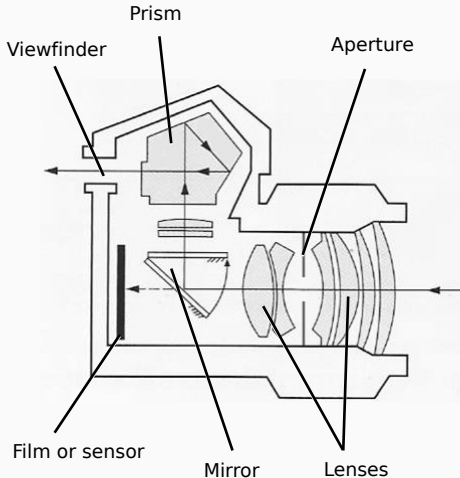
Measurement light in terms of perceived brightness to human eye

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Visible light $\lambda \in [390\text{nm}, 700\text{nm}]$ approximately

PHOTOMETRY

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Well-lit conditions

PHOTOPIC VISION

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Cones: Three types of retinal cells with distinct spectral responses

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Response curves S (short λ), M (medium λ), L (long λ)

- Peaks at $\lambda = 420\text{nm}$, $\lambda = 534\text{nm}$, and $\lambda = 564\text{nm}$
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Are there tetrachromats among us?

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SCOTOPIC VISION

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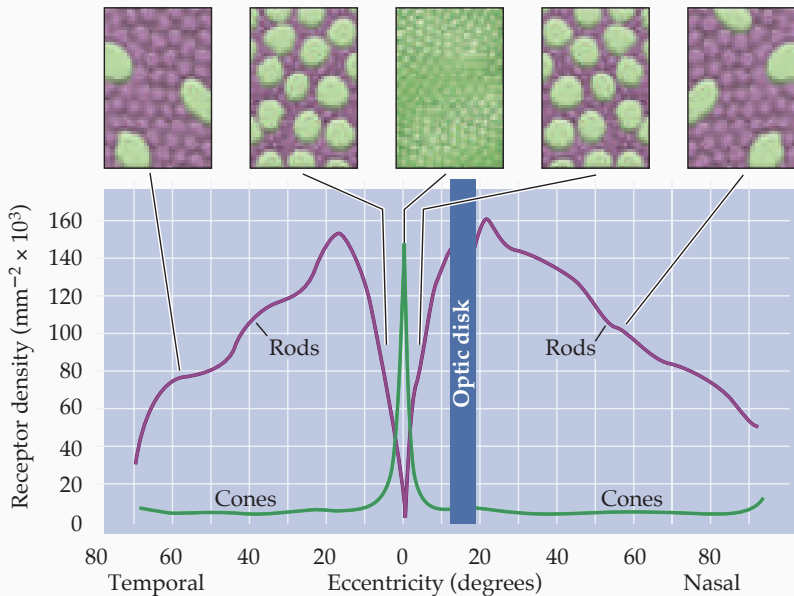
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Response curve

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Things look “gray-bluish” at night

HUMAN PHOTORECEPTOR DISTRIBUTION



LUMINOUS EFFICIENCY FUNCTION

Spectral sensitivity $V(\lambda)$ of human perception of *brightness*

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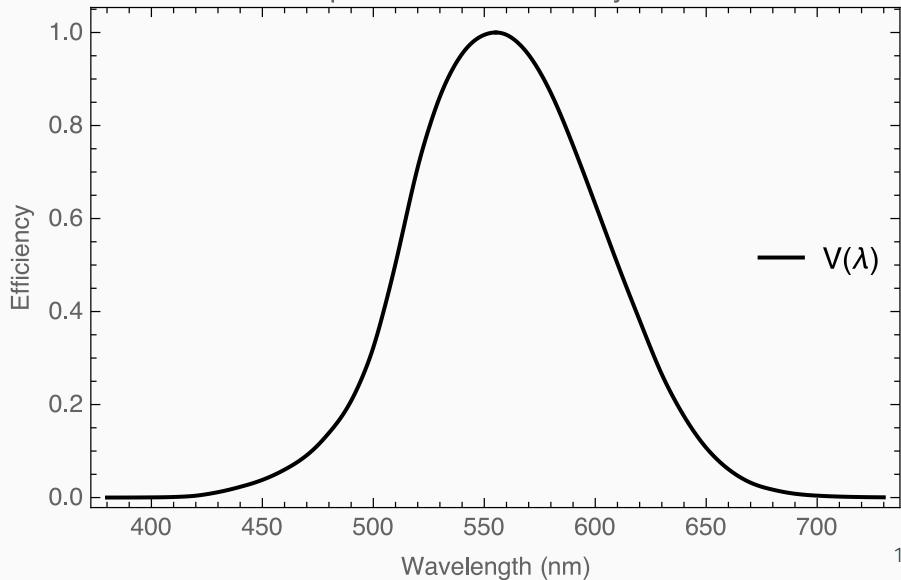
Immense dynamic range 1 : 10^{10} (brightness adaptation)

Convert radiant intensity (W/sr) to luminous intensity (cd)

$$v(c) = \int_{\lambda} c(\lambda)V(\lambda)d\lambda$$

PHOTOPIC LUMINOUS EFFICIENCY FUNCTION

Photopic luminous efficiency function



Nonlinear perceptual response to brightness

Nonlinear perceptual response to brightness

Power law

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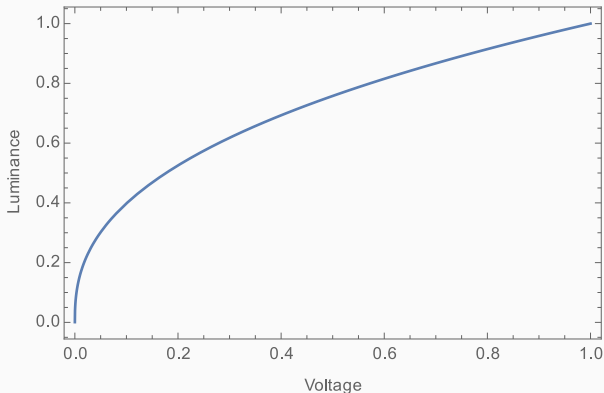
Weber law of just noticeable difference

$$\Delta L^* \approx \frac{Y}{100}$$

GAMMA CORRECTION

Created to compensate for input-output characteristic of CRT displays

$$Y = V^{2.5} = V^\gamma$$

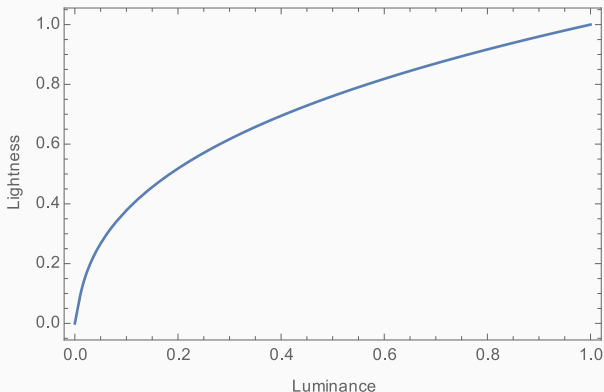


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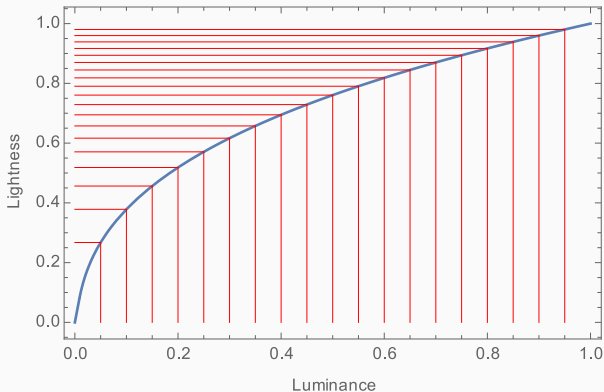


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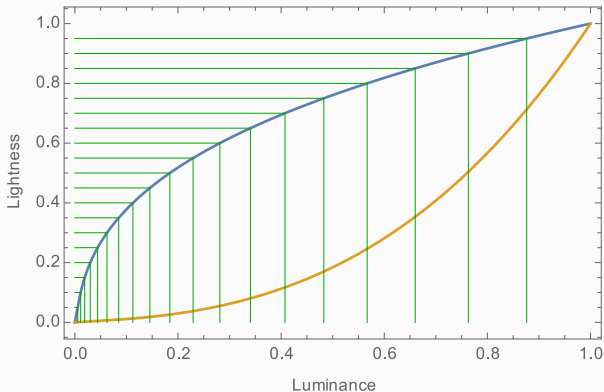


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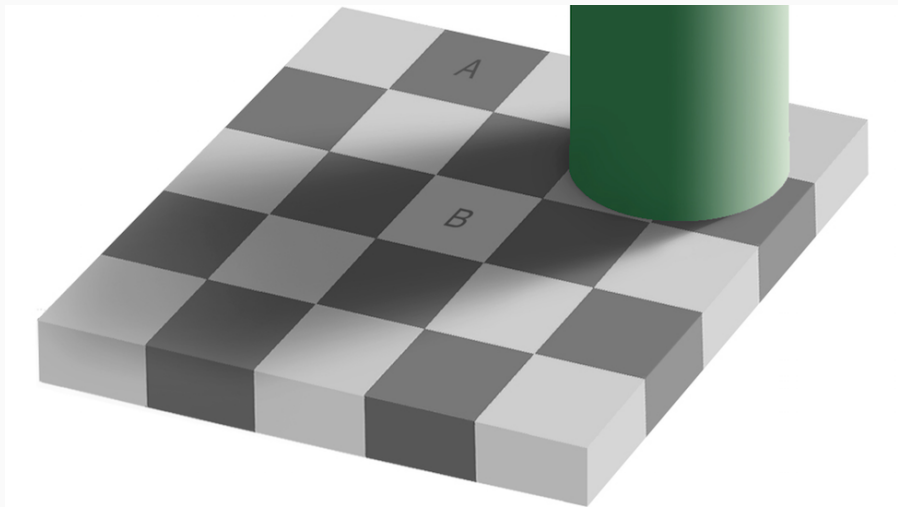
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ILLUSION



1st attempt: Measure spectral distribution of stimulus

- Convex combinations of monochromatic colors
- Could use spectrophotometer to measure $c(\lambda)$.
- But how to would you reproduce it?

MODELING COLOR PERCEPTION

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- Could use spectrophotometer to measure $c(\lambda)$.
- But how to would you reproduce it?

2nd attempt: Measure optical nerve response

- Remove eye, attach wires to cones: The Matrix
- Re-inject signal to reproduce
- Painful, but only 3 values per color

3rd attempt: Linear algebra

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Measuring

- $c(\lambda)$ is the target color's spectral distribution

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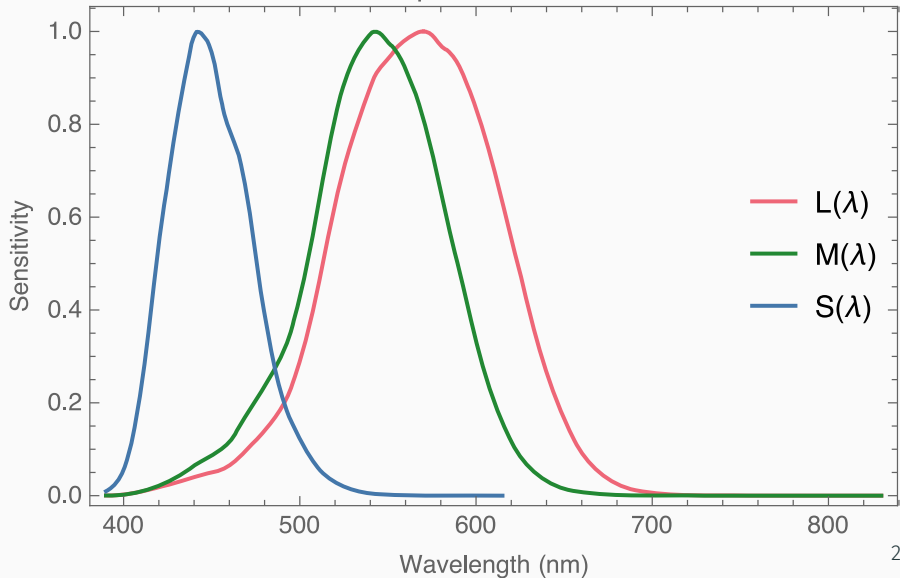
$$\langle f, g \rangle = \int_{-\infty}^{\infty} f(\lambda)g(\lambda)d\lambda$$

- The cone responses to c must be

$$S_c = \langle c, S \rangle, \quad M_c = \langle c, M \rangle, \quad \text{and} \quad L_c = \langle c, L \rangle$$

CONE SPECTRAL SENSITIVITIES (NOT TO SCALE)

Cone spectral sensitivities



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- Assume 3 different stimuli colors $r(\lambda)$, $g(\lambda)$, and $b(\lambda)$

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Stimuli must be linearly independent

Result R_c , G_c , or B_c could be non-convex

- There is no negative light...

SPACE OF VISIBLE COLORS

All convex combinations of visible monochromatic colors

- Could use entire spectrum

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Unnecessary (most of the time) due to *metamerism*

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Obtain R_c , G_c , and B_c directly from c and RGB color matching functions

$$R_c = \langle c, R \rangle$$

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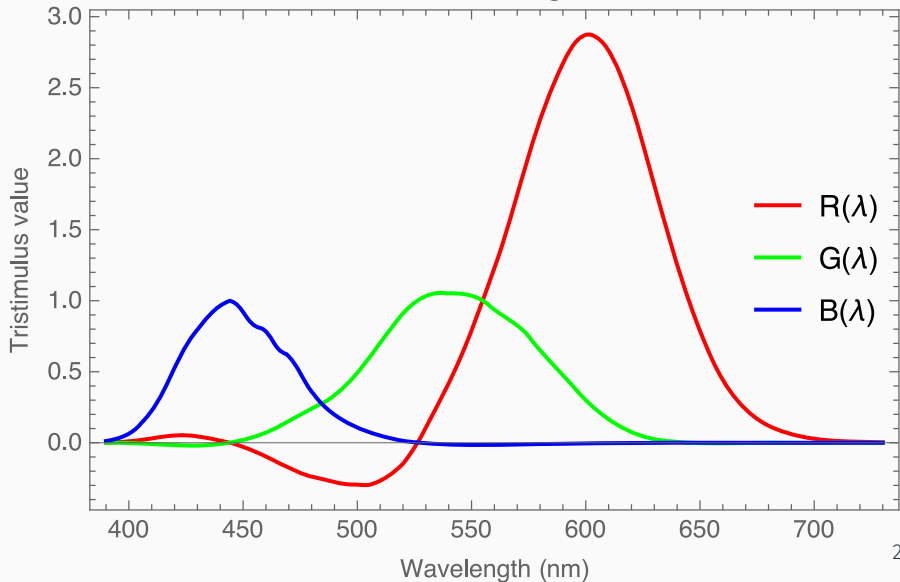
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How to measure color matching functions R , G , and B

CIE 1931 RGB COLOR MATCHING FUNCTIONS

RGB color matching functions



Visible colors always use non-negative coordinates

XYZ COLOR MATCHING FUNCTIONS

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Linear transformation to R, G, B

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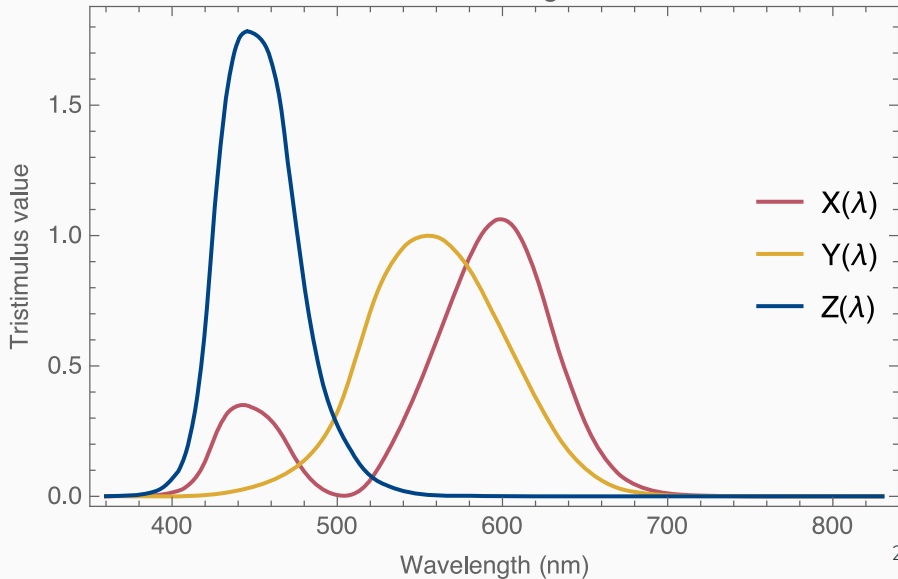
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Z ended up almost equal to S

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XYZ color matching functions



CIE CHROMATICITY DIAGRAM

Similar to \mathbf{RP}^2

- Given $\alpha > 0$, $[\alpha X \quad \alpha Y \quad \alpha Z]$ have same *chromaticity*
- Different *brightness*

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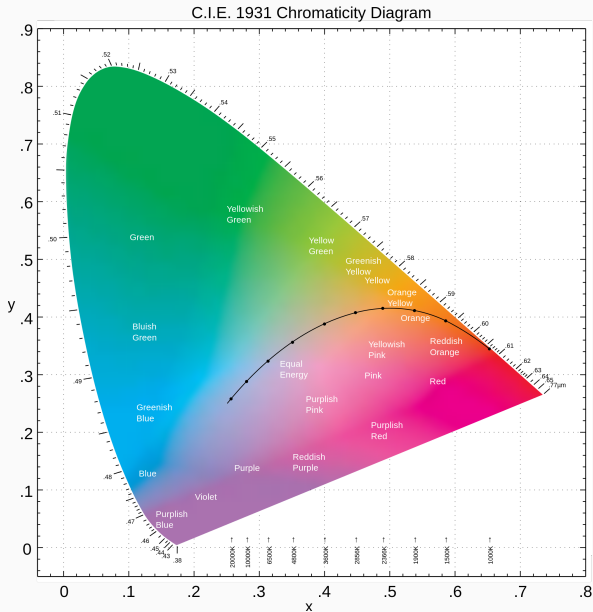
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Separation of chromaticity and brightness

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

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

CIE CHROMATICITY DIAGRAM



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Horseshoe shape

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Color gamut

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Color calibration and matching

OTHER COLOR SPACES

sRGB [IEC Project Team 61966, 1998]

$$\begin{bmatrix} R \\ G \\ B \end{bmatrix} = \begin{bmatrix} \gamma(R_\ell) \\ \gamma(G_\ell) \\ \gamma(B_\ell) \end{bmatrix}, \quad \begin{bmatrix} R_\ell \\ G_\ell \\ B_\ell \end{bmatrix} = \begin{bmatrix} 3.2406 & -1.5372 & -0.4986 \\ -0.9689 & 1.8758 & 0.0415 \\ 0.0557 & -0.2040 & 1.0570 \end{bmatrix} \begin{bmatrix} X_{D65} \\ Y_{D65} \\ Z_{D65} \end{bmatrix}$$

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Perceptual (CIE $L^*a^*b^*$)

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Additive (*CMY* and *CMYK*)

TV (PAL *YUV*, NTSC *YIQ*)

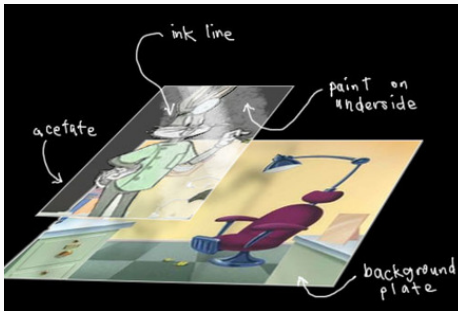
Perceptual (CIE $L^*a^*b^*$)

Opponent color models

TRANSPARENCY

SEMINAL WORK BY PORTER AND DUFF [1984]

Semitransparent color f on top of opaque background color b



- Assume probability of light hitting f is α
- Reflected color (integrated over small area) is

$$f, \alpha \oplus b = \alpha f + (1 - \alpha)b$$

- This is what we call *alpha blending* or the *over operator*

Now imagine f_1, α_1 on top of f_2, α_2 on top of b

- Reflected color is

$$f_1, \alpha_1 \oplus (f_2, \alpha_2 \oplus b) = \alpha_1 f_1 + (1 - \alpha_1)(\alpha_2 f_2 + (1 - \alpha_2)b)$$

COMPOSITING

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$$f_1, \alpha_1 \oplus (f_2, \alpha_2 \oplus b) = \alpha_1 f_1 + (1 - \alpha_1)(\alpha_2 f_2 + (1 - \alpha_2)b)$$

Can we combine f_1, α_1 and f_2, α_2 into a single material f, α ?

$$\begin{aligned}\alpha f + (1 - \alpha)b &= \alpha_1 f_1 + (1 - \alpha_1)(\alpha_2 f_2 + (1 - \alpha_2)b) \\ &= \alpha_1 f_1 + (1 - \alpha_1)\alpha_2 f_2 + (1 - \alpha_1)(1 - \alpha_2)b\end{aligned}$$

COMPOSITING

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So we have

$$\begin{cases} (1 - \alpha)b = (1 - \alpha_1)(1 - \alpha_2)b \\ \alpha f = \alpha_1 f_1 + (1 - \alpha_1)\alpha_2 f_2 \end{cases} \Rightarrow \begin{cases} \alpha = \alpha_1 + (1 - \alpha_1)\alpha_2 \\ \alpha f = \alpha_1 f_1 + (1 - \alpha_1)\alpha_2 f_2 \end{cases}$$

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Setting $\tilde{f} = \alpha f$, $\tilde{f}_1 = \alpha_1 f_1$, and $\tilde{f}_2 = \alpha_2 f_2$, we obtain

$$\begin{cases} \alpha = \alpha_1 + (1 - \alpha_1)\alpha_2 \\ \tilde{f} = \tilde{f}_1 + (1 - \alpha_1)\tilde{f}_2 \end{cases}$$

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This is what we call *pre-multiplied alpha*

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Blending becomes associative

$$\tilde{f}_1, \alpha_1 \oplus (\tilde{f}_2, \alpha_2 \oplus b) = (\tilde{f}_1, \alpha_1 \oplus \tilde{f}_2, \alpha_2) \oplus b$$

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Should we blend front-to-back or back-to-front?

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