



Models of early visual perception

Advanced Graphics and Image Processing

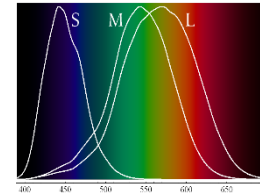
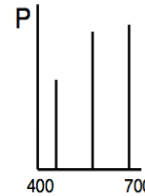
Rafal Mantiuk

Computer Laboratory, University of Cambridge

Many graphics/display solutions are motivated by visual perception



Image & video compression

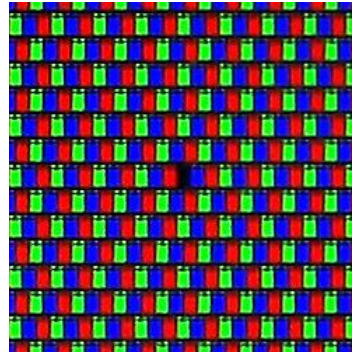


*

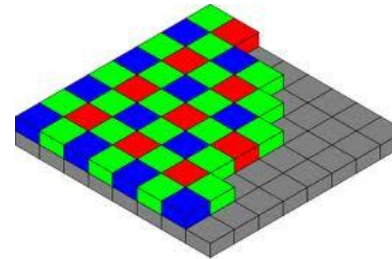
Display spectral emission - metamerism



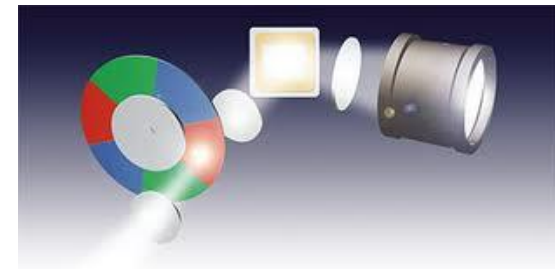
Halftoning



Display's subpixels



Camera's Bayer pattern



Color wheel in DLPs

Perceived brightness of light

Luminance (again)

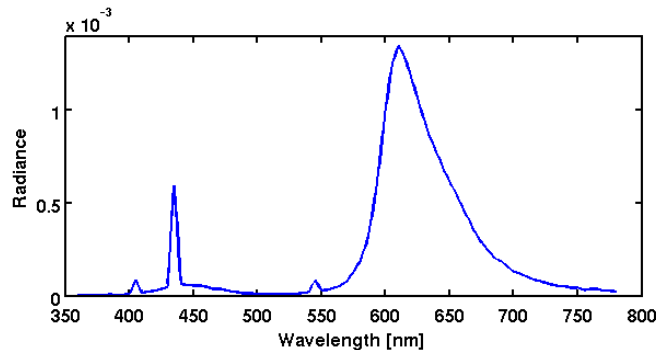
- ▶ Luminance – measure of light weighted by the response of the achromatic mechanism. Units: cd/m^2

Luminance

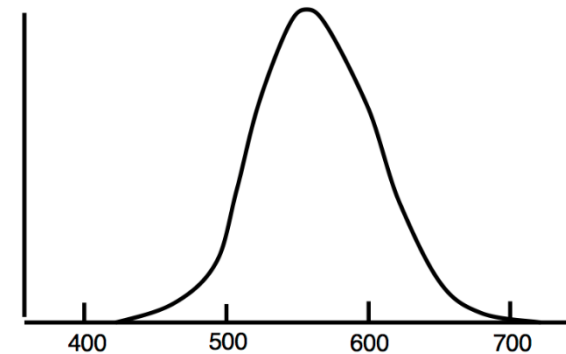
$$L_V = \int_{350}^{700} kL(\lambda)V(\lambda)d\lambda$$

$$k = \frac{1}{683.002}$$

Light spectrum (radiance)



Luminous efficiency function (weighting)



Steven's power law for brightness

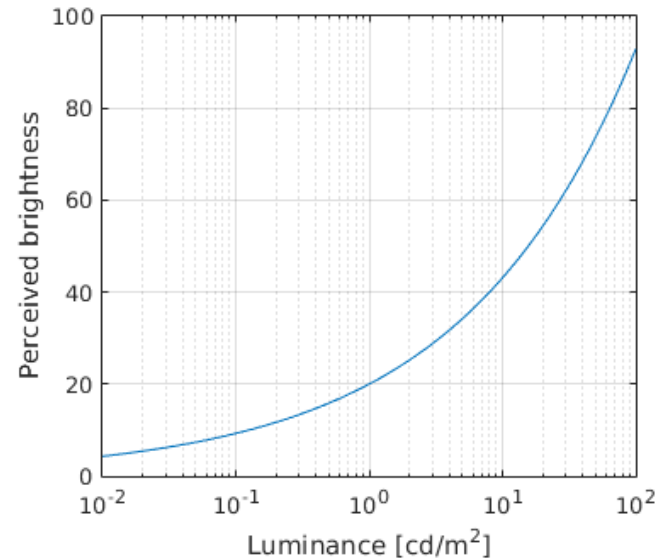
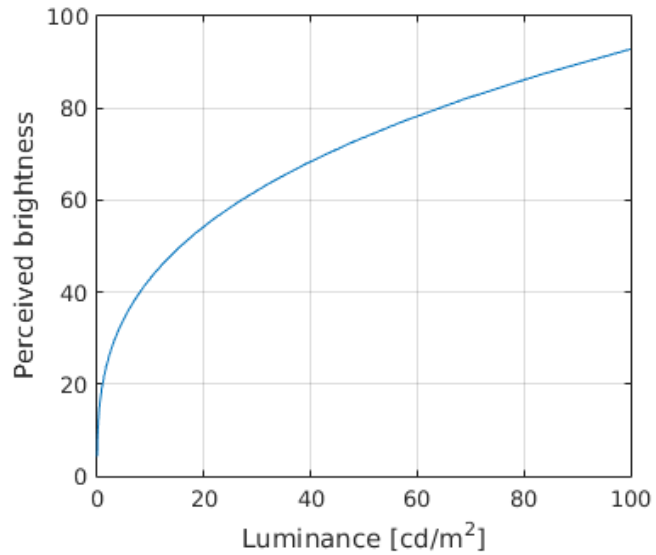
- ▶ Stevens (1906-1973) measured the perceived magnitude of physical stimuli
 - ▶ Loudness of sound, tastes, smell, warmth, electric shock and brightness
 - ▶ Using the magnitude estimation methods
 - ▶ Ask to rate loudness on a scale with a known reference
- ▶ All measured stimuli followed the power law:

The diagram shows the equation $j(I) = kI^a$ with four callout boxes pointing to its parts: 'Perceived magnitude' points to $j(I)$, 'Constant' points to k , 'Physical stimulus' points to I , and 'Exponent' points to a .

$$j(I) = kI^a$$

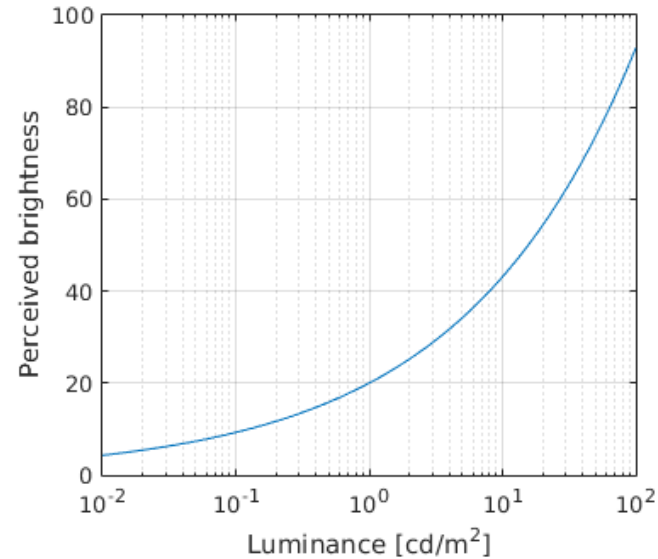
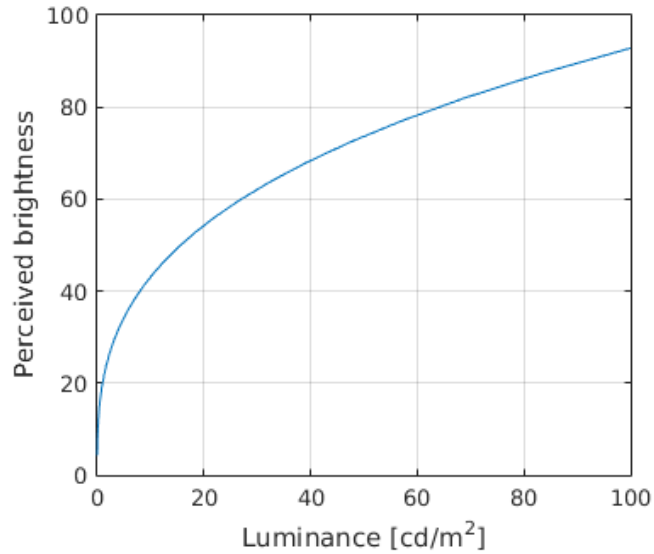
- ▶ For brightness (5 deg target in dark), $a = 0.3$

Steven's law for brightness

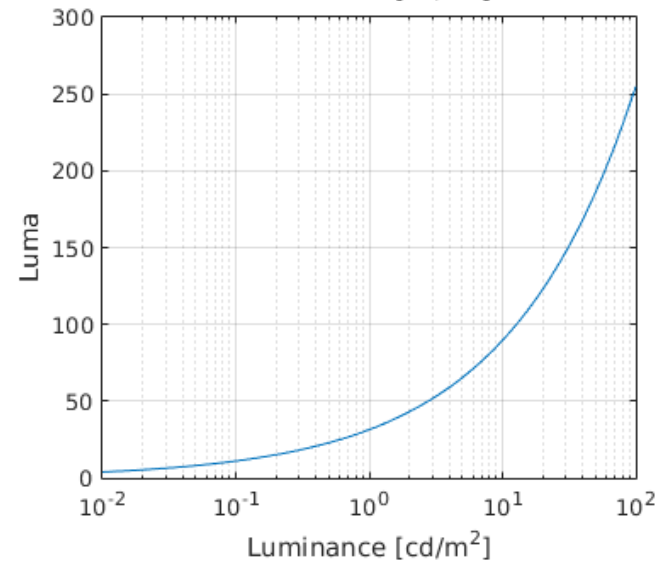
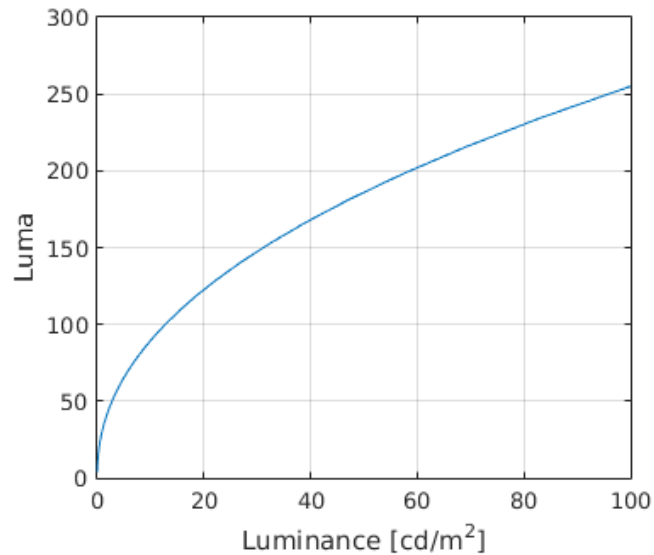


Steven's law vs. Gamma correction

Stevens' law
 $a=0.3$

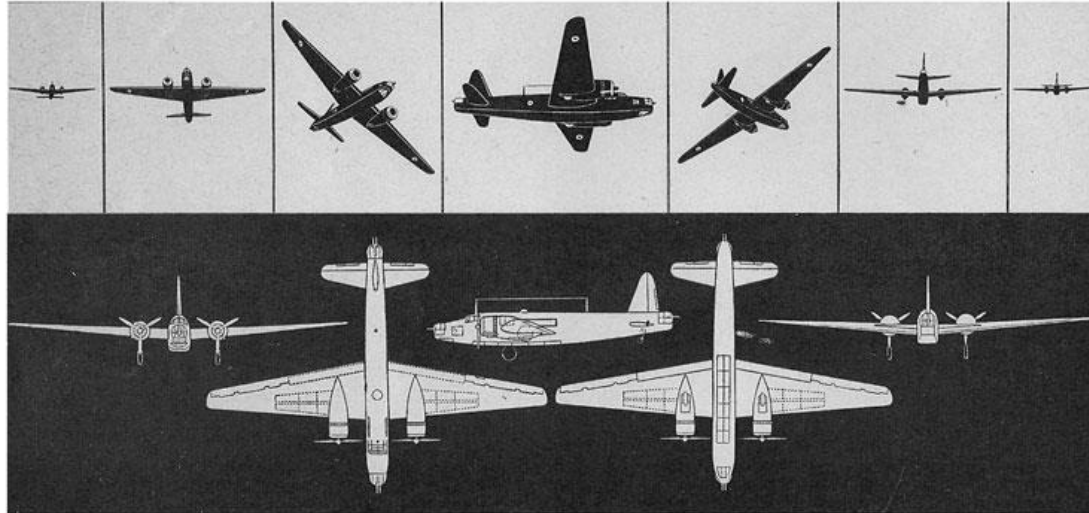


Gamma function
Gamma = 2.2



Detection and discrimination

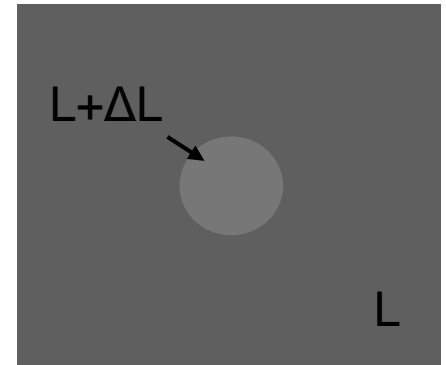
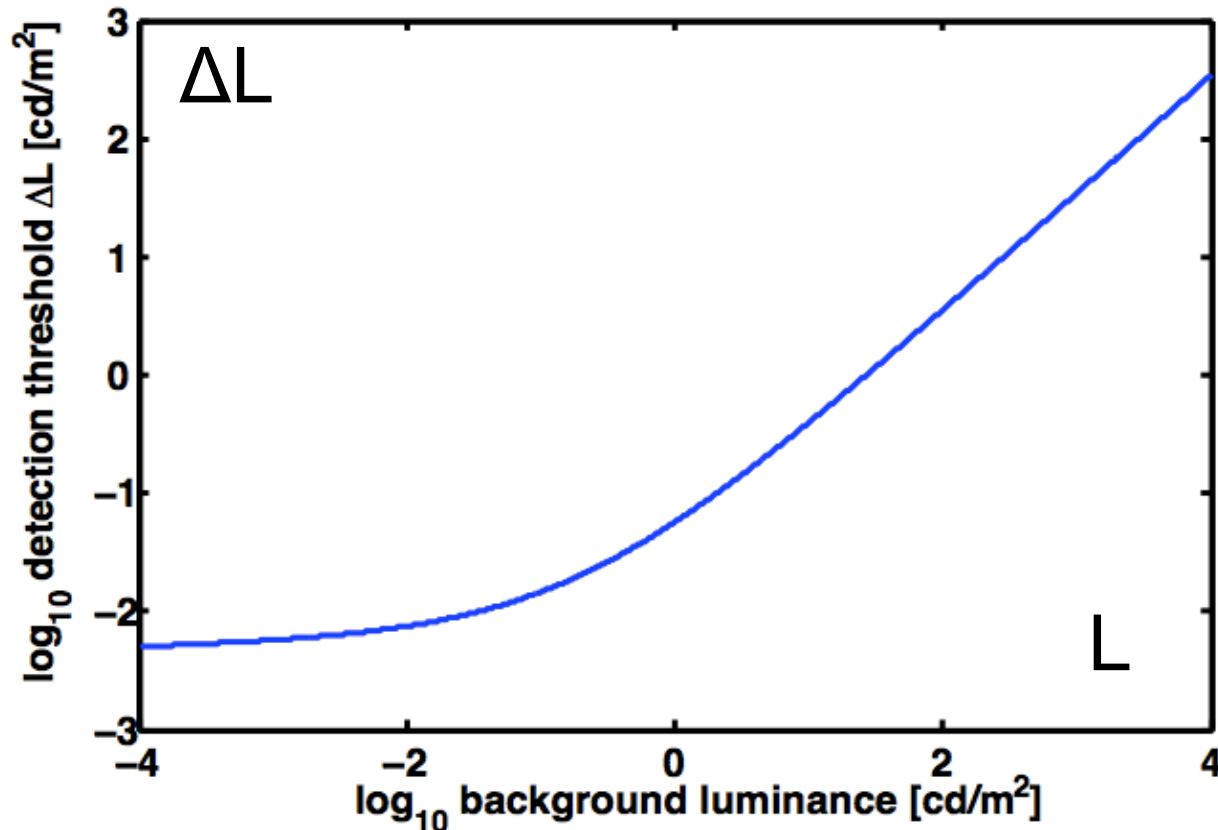
Detection thresholds



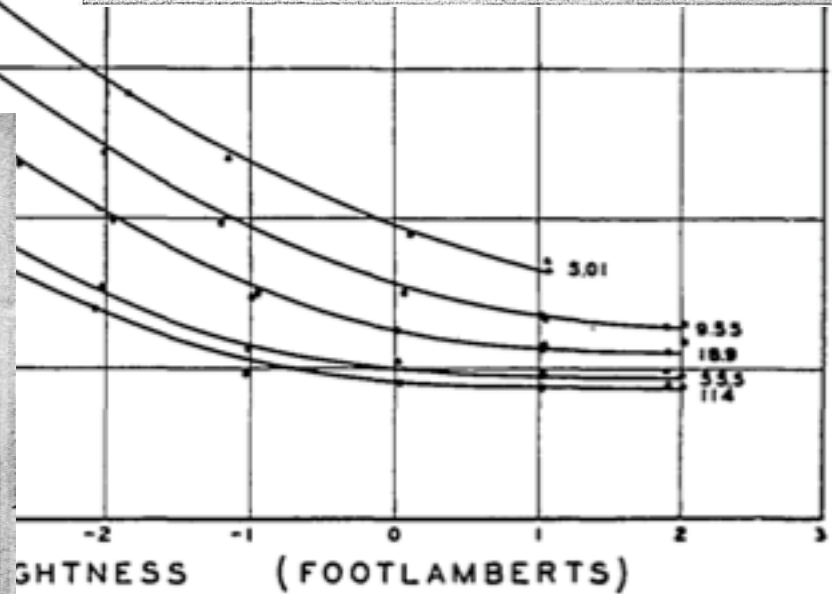
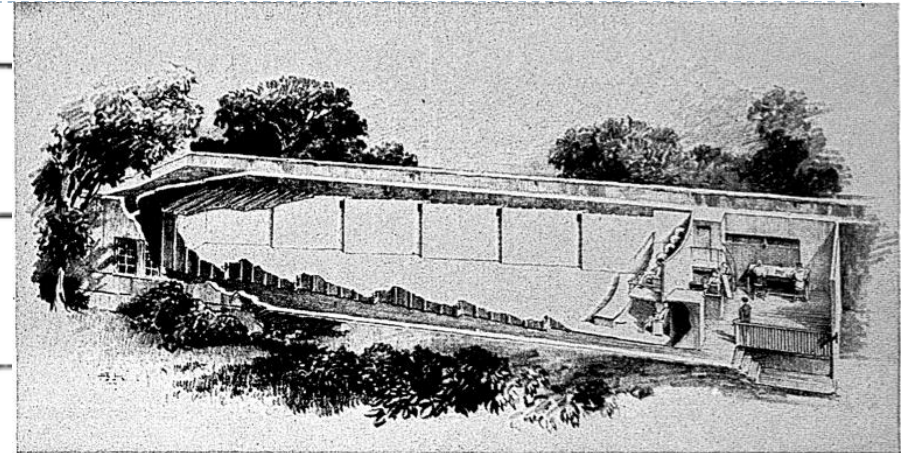
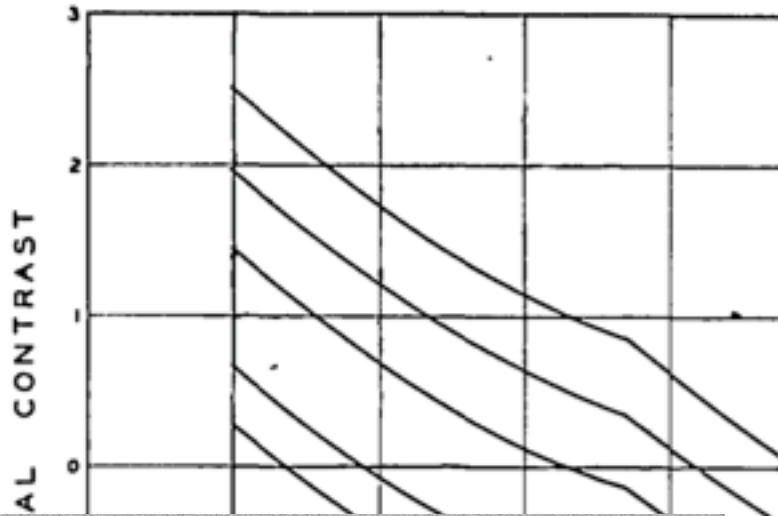
- ▶ **The smallest detectable difference between**
 - ▶ the luminance of the object and
 - ▶ the luminance of the background

Threshold versus intensity (t.v.i.) function

- ▶ The smallest detectable difference in luminance for a given background luminance

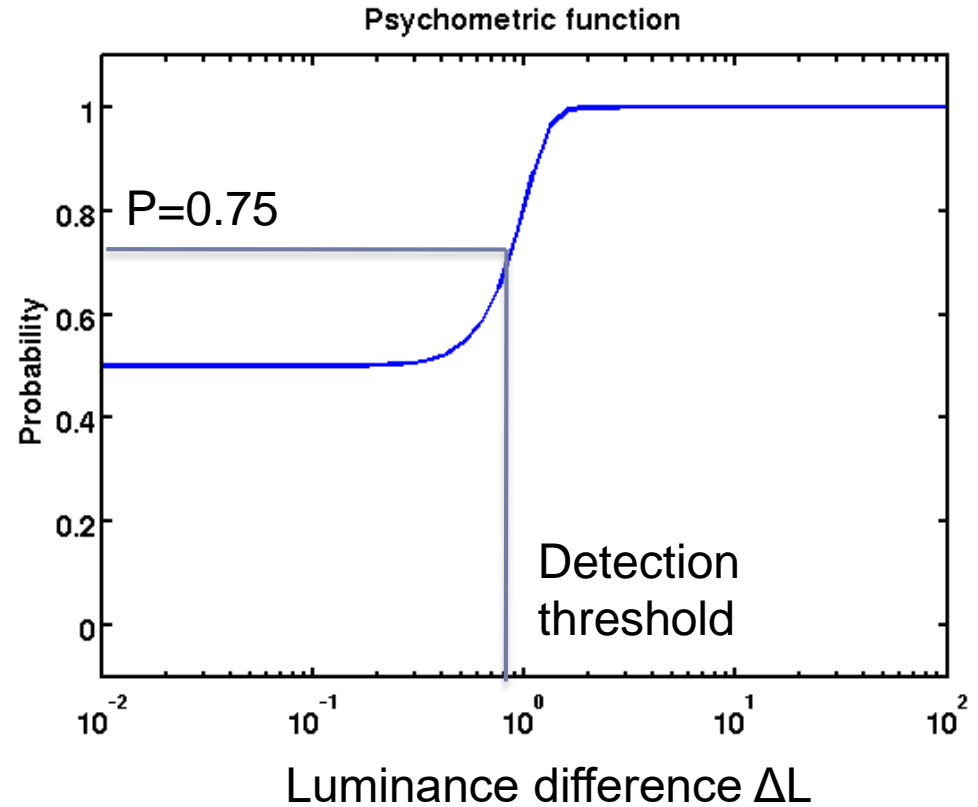
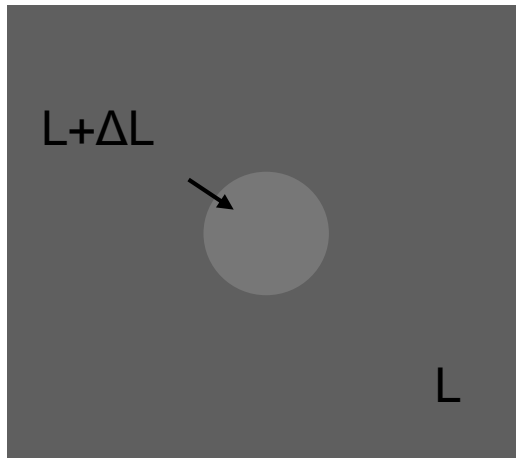


t.v.i. measurements – Blackwell 1946



Psychophysics

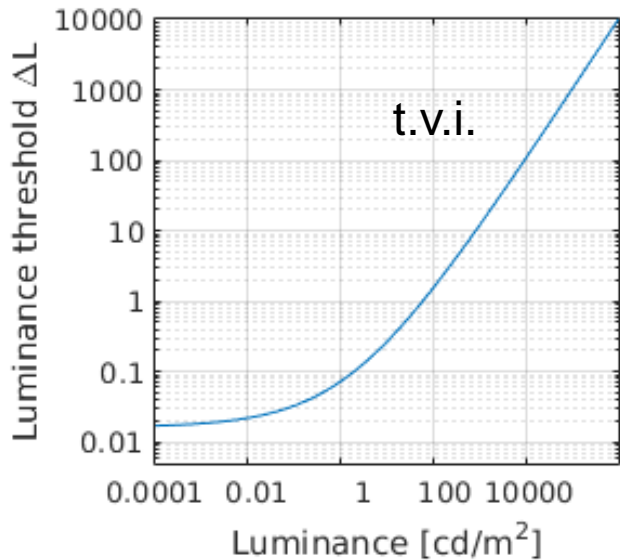
Threshold experiments



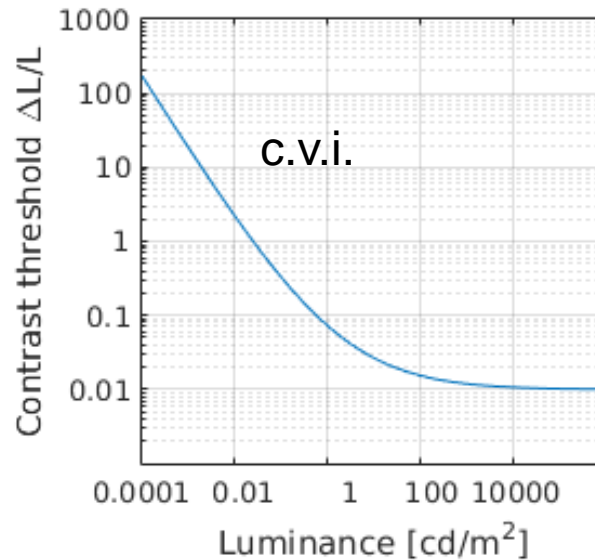
t.v.i function / c.v.i. function / Sensitivity

- ▶ The same data, different representation

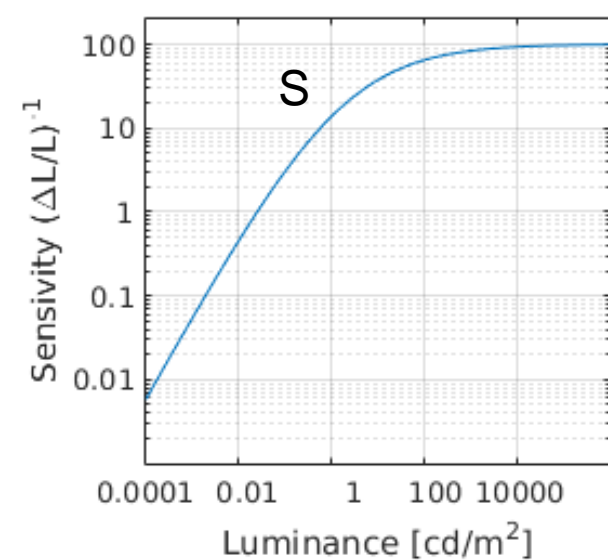
Threshold vs. intensity



Contrast vs. intensity



Sensitivity



$$\Delta L = L_{disk} - L_{background}$$

$$T = \frac{\Delta L}{L}$$

$$S = \frac{1}{T} = \frac{L}{\Delta L}$$

Sensitivity to luminance

- ▶ Weber-law – the just-noticeable difference is proportional to the magnitude of a stimulus



Ernst Heinrich Weber
[From wikipedia]

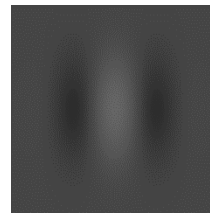
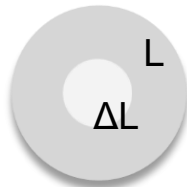
The smallest detectable luminance difference

Background (adapting) luminance

$$\frac{\Delta L}{L} = k$$

Constant

Typical stimuli:



Consequence of the Weber-law

- ▶ Smallest detectable difference in luminance

$$\frac{\Delta L}{L} = k$$

For k=1%

L	ΔL
100 cd/m ²	1 cd/m ²
1 cd/m ²	0.01 cd/m ²

- ▶ Adding or subtracting luminance will have different visual impact depending on the background luminance
- ▶ Unlike LDR luma values, luminance values are **not** perceptually uniform!

How to make luminance (more) perceptually uniform?

- ▶ Using “Fechnerian” integration

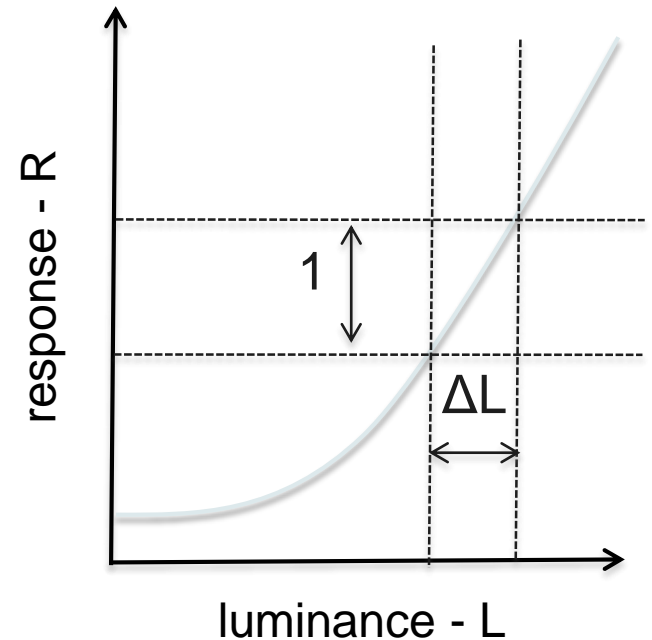
$$\frac{dR}{dl}(L) = \frac{1}{\Delta L(L)}$$

Derivative of response

Detection threshold

Luminance transducer:

$$R(L) = \int_0^L \frac{1}{\Delta L(l)} dl$$



Assuming the Weber law

$$\frac{\Delta L}{L} = k$$

- ▶ and given the luminance transducer

$$R(L) = \int_0^L \frac{1}{\Delta L(l)} dl$$

- ▶ the response of the visual system to light is:

$$R(L) = \int \frac{1}{kL} dL = \frac{1}{k} \ln(L) + k_1$$

Fechner law

$$R(L) = a \ln(L)$$

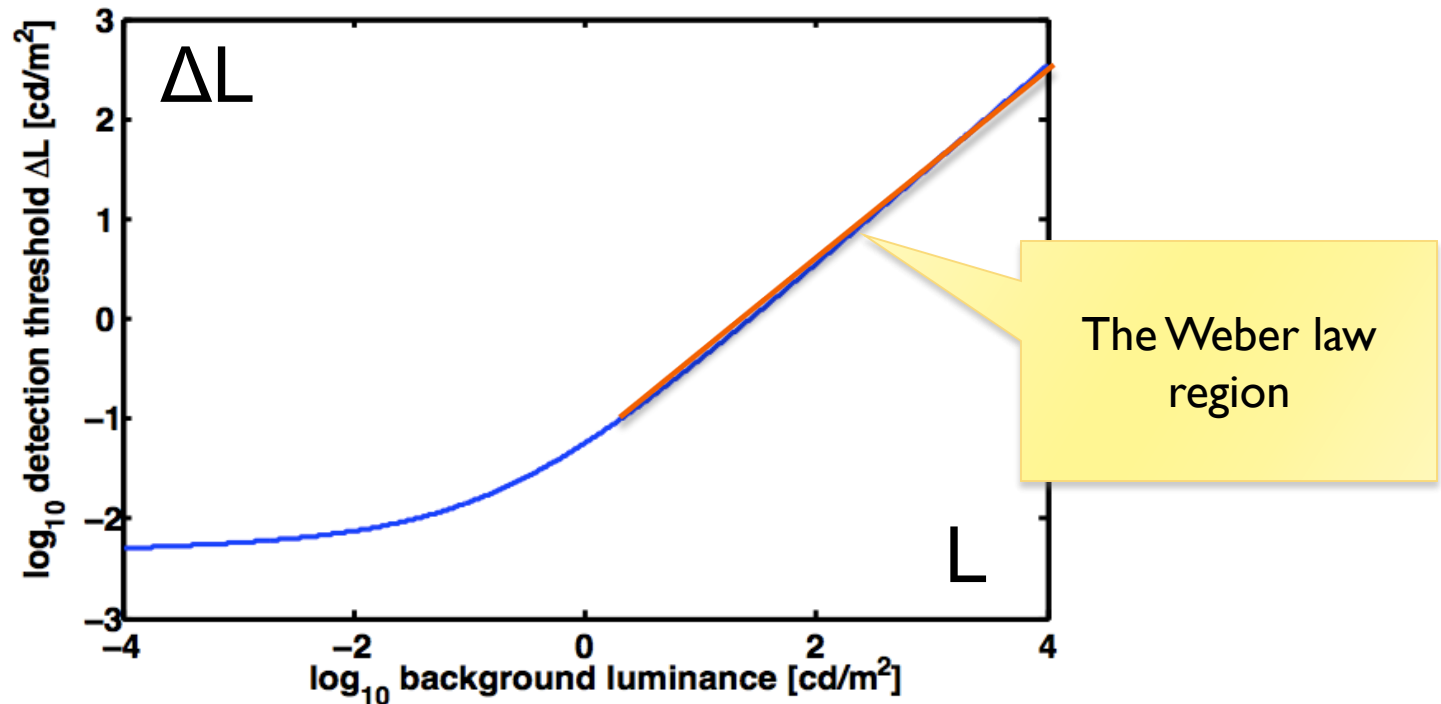
- ▶ Response of the visual system to luminance is **approximately** logarithmic



Gustav Fechner
[From Wikipedia]

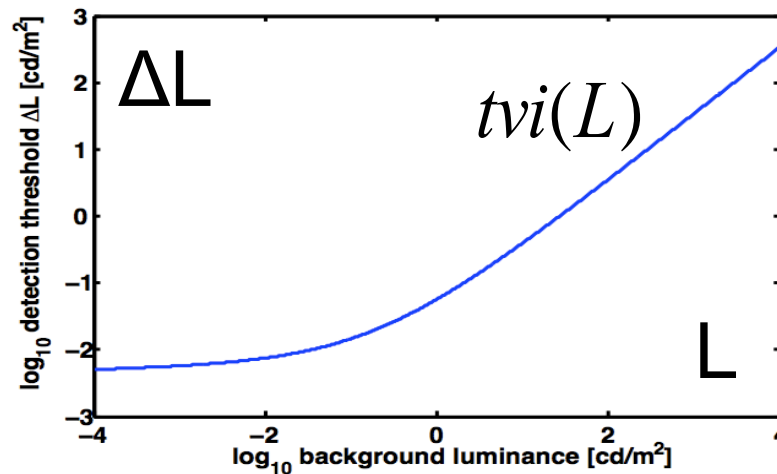
But...the Fechner law does not hold for the full luminance range

- ▶ Because the Weber law does not hold either
- ▶ Threshold vs. intensity function:



Weber-law revisited

- ▶ If we allow detection threshold to vary with luminance according to the t.v.i. function:



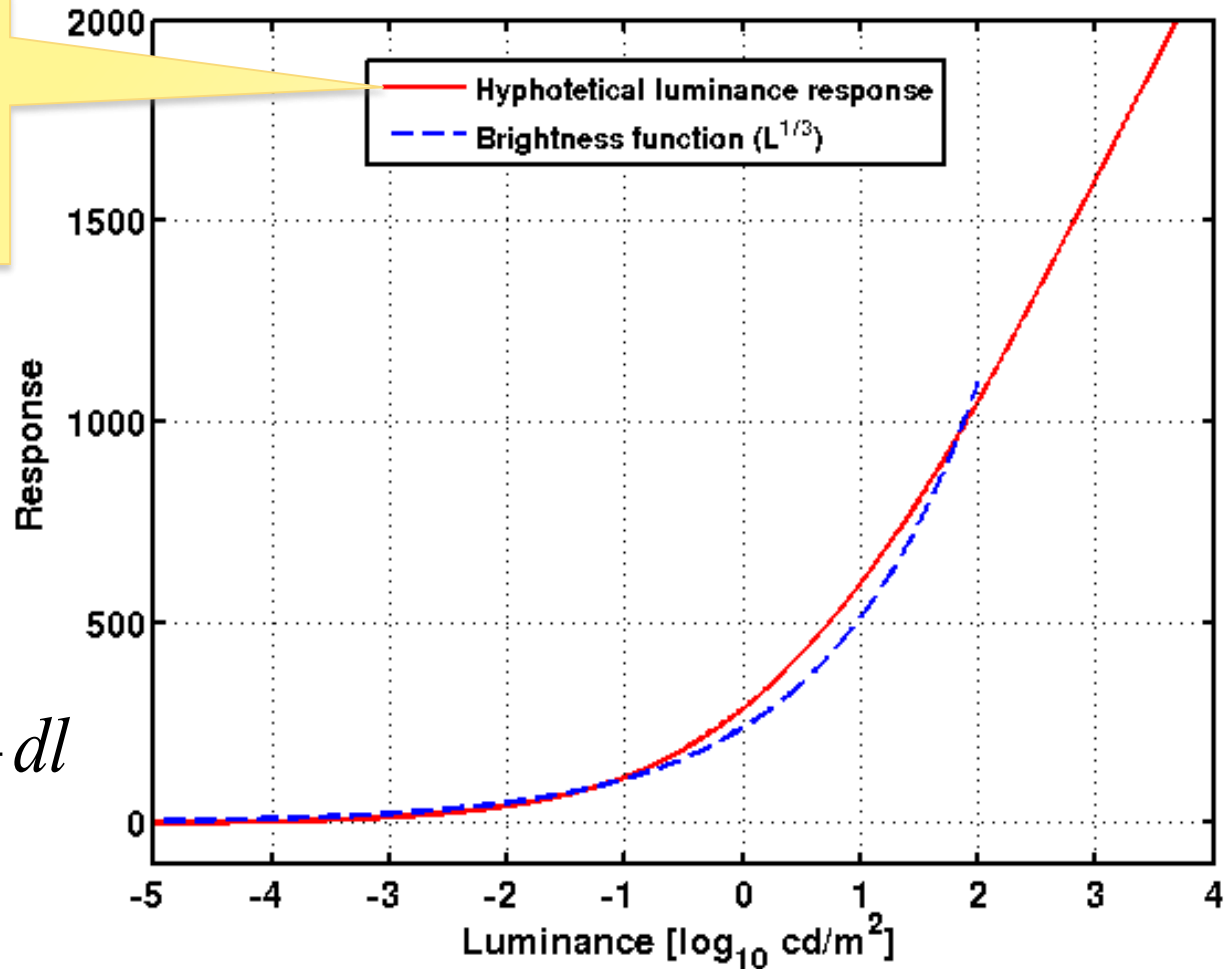
- ▶ we can get a more accurate estimate of the “response”:

$$R(L) = \int_0^L \frac{1}{tvi(l)} dl$$

Fechnerian integration and Stevens' law

R(L) - function derived from the t.v.i. function

$$R(L) = \int_0^L \frac{1}{tvi(l)} dl$$

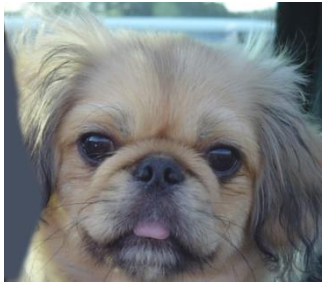


Applications of JND encoding – R(L)

- ▶ **DICOM grayscale function**
 - ▶ Function used to encode signal for medical monitors
 - ▶ 10-bit JND-scaled (just noticeable difference)
 - ▶ Equal visibility of gray levels
- ▶ **HDMI 2.0a (HDR10)**
 - ▶ PQ (Perceptual Quantizer) encoding
 - ▶ Dolby Vision
 - ▶ To encode pixels for high dynamic range images and video



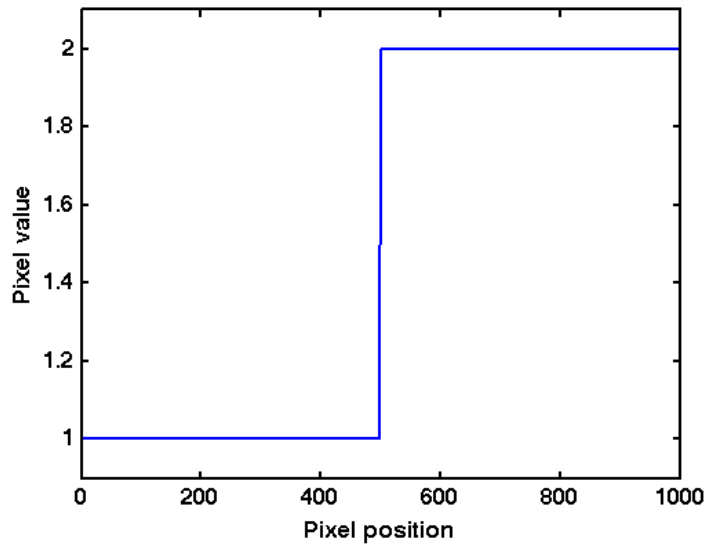
The Future of Vision



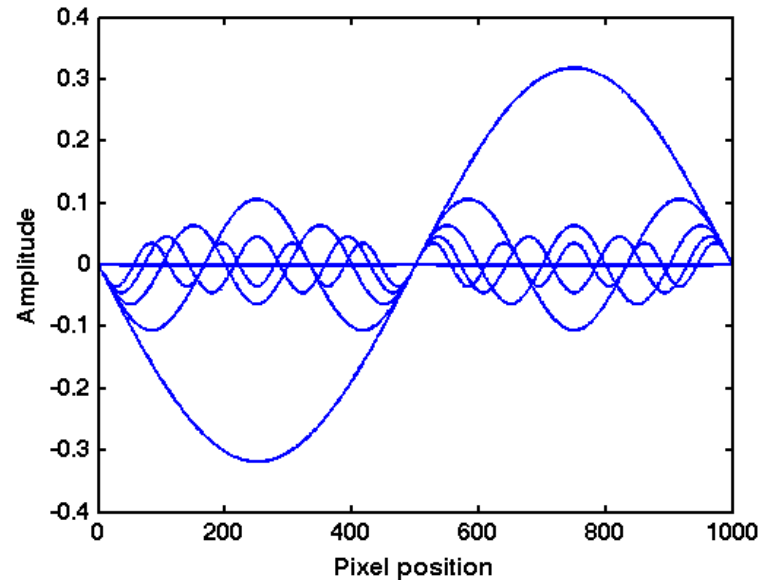
Spatial contrast sensitivity

Fourier analysis

- ▶ Every N-dimensional function (including images) can be represented as a sum of sinusoidal waves of different frequency and phase



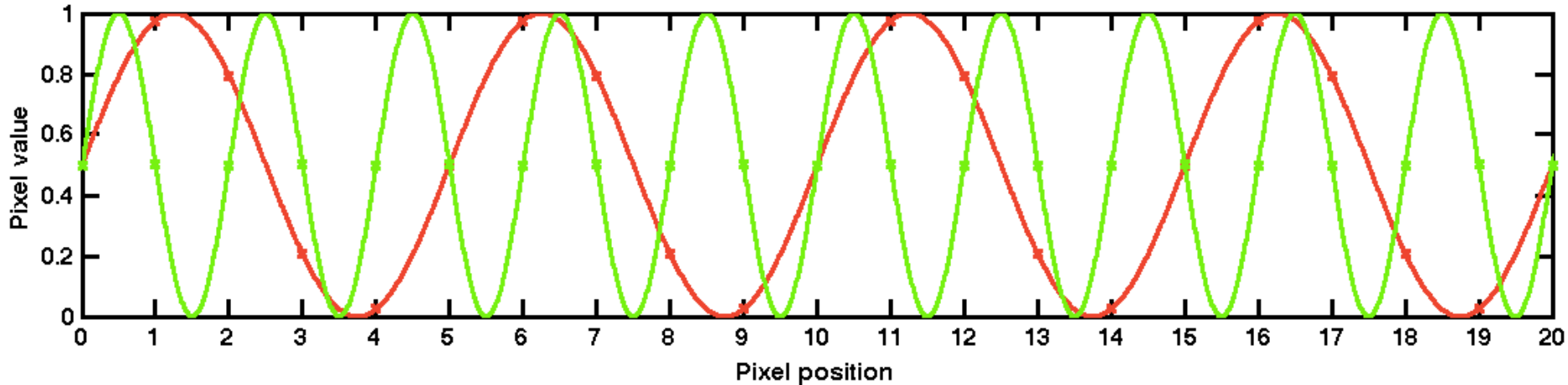
$$= \hat{a}$$



- ▶ Think of “equalizer” in audio software, which manipulates each frequency

Spatial frequency in images

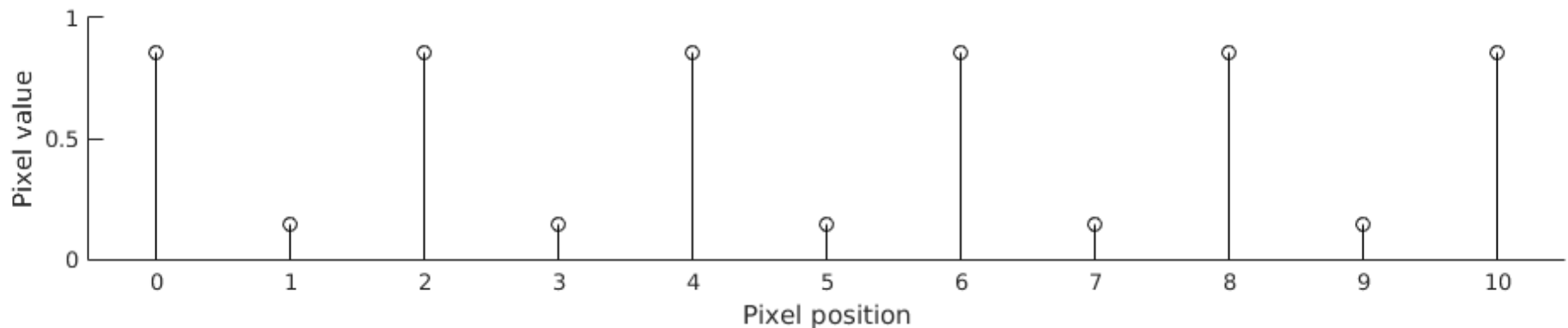
- ▶ Image space units: cycles per sample (or cycles per pixel)



- ▶ What are the screen-space frequencies of the red and green sinusoid?
- ▶ The visual system units: cycles per degree
 - ▶ If the angular resolution of the viewed image is 55 pixels per degree, what is the frequency of the sinusoids in cycles per degree?

Nyquist frequency

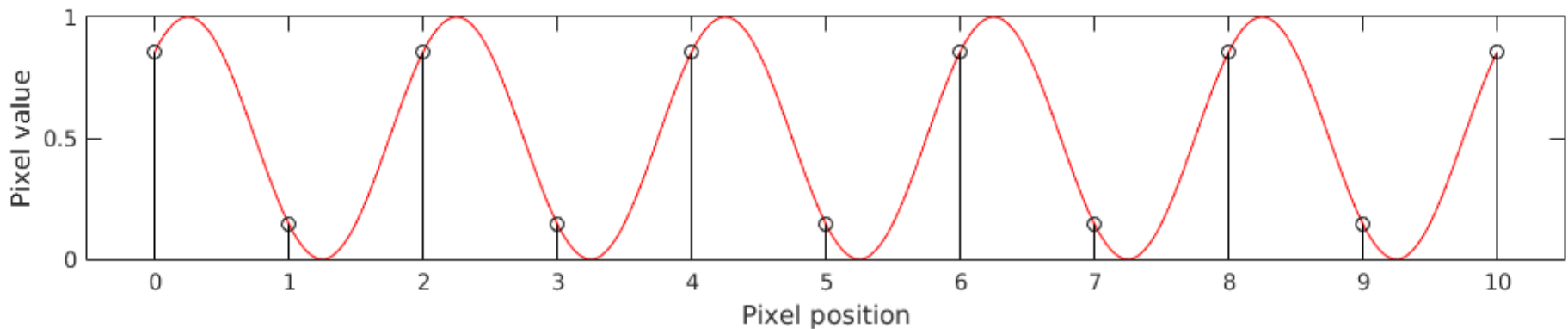
- ▶ Sampling density restricts the highest spatial frequency signal that can be (uniquely) reconstructed
 - ▶ Sampling density – how many pixels per image/visual angle/...



- ▶ Any number of sinusoids can be fitted to this set of samples
- ▶ It is possible to fit an infinite number of sinusoids if we allow infinitely high frequency

Nyquist frequency

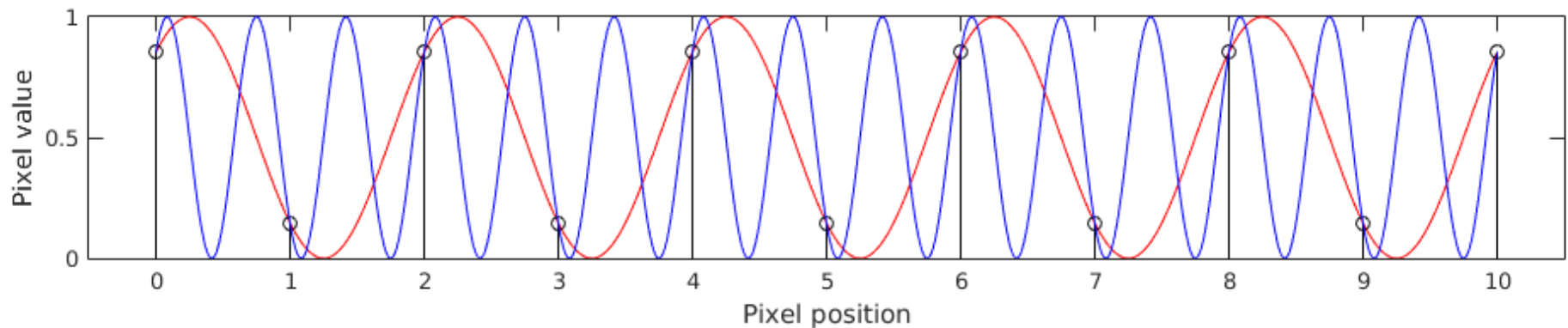
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Nyquist frequency

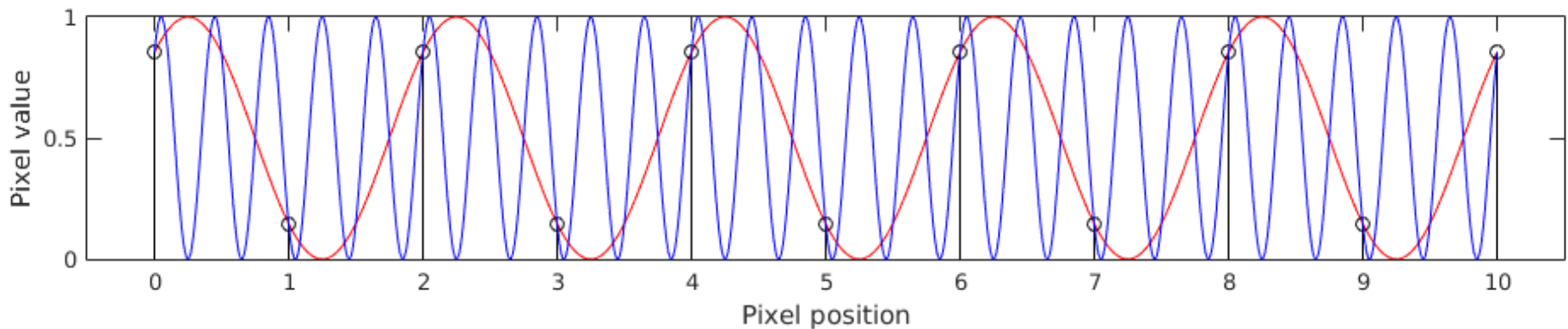
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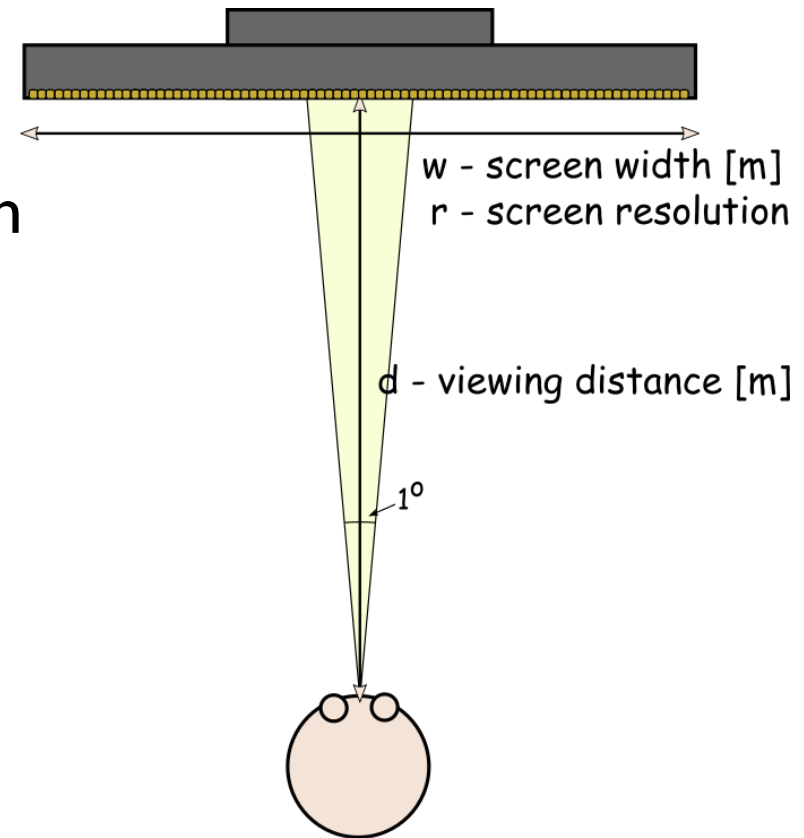
- ▶ Any number of sinusoids can be fitted to this set of samples
- ▶ It is possible to fit an infinite number of sinusoids if we allow infinitely high frequency

Nyquist frequency / aliasing

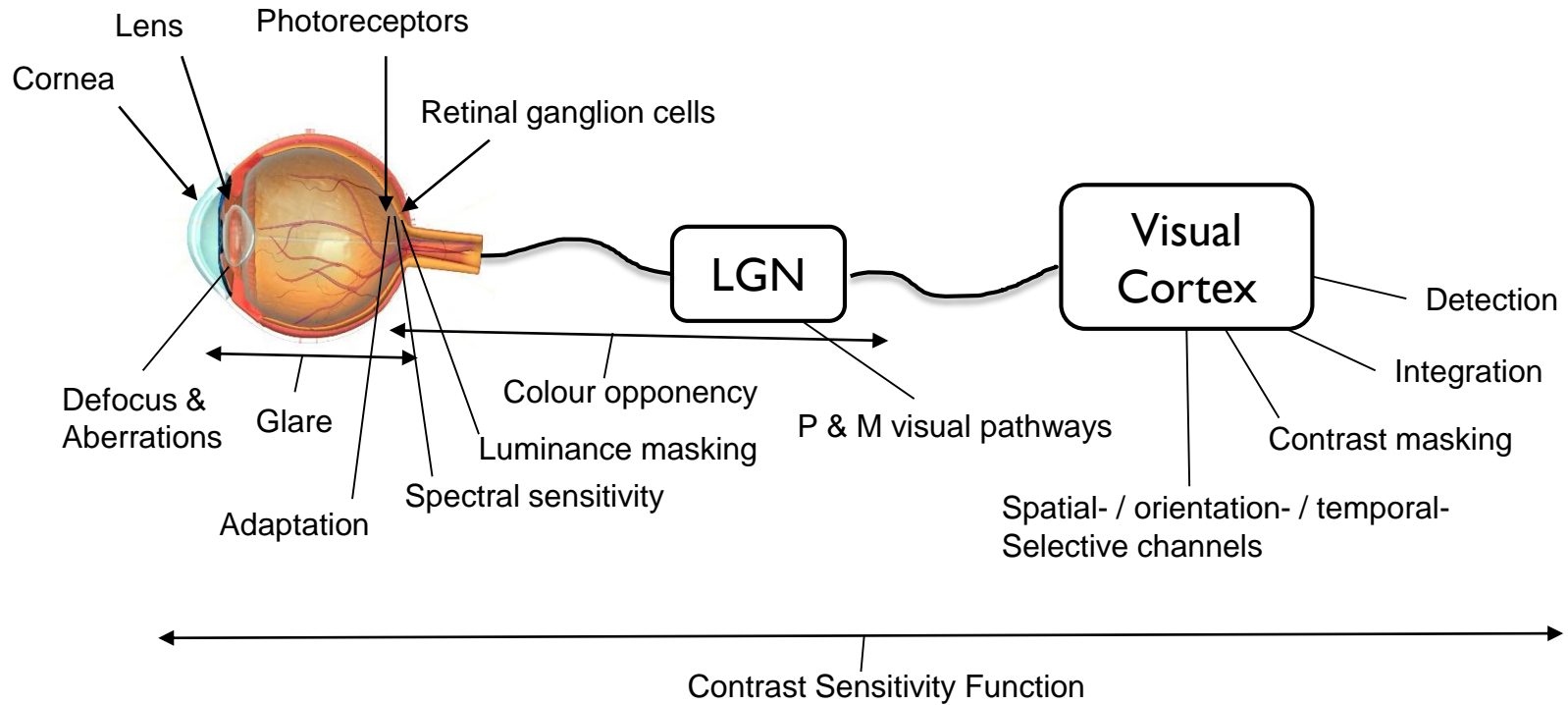
- ▶ Nuquist frequency is the highest frequency that can be represented by a discrete set of uniform samples (pixels)
- ▶ Nuquist frequency = 0.5 sampling rate
 - ▶ For audio
 - ▶ If the sampling rate is 44100 samples per second (audio CD), then the Nyquist frequency is 22050 Hz
 - ▶ For images (visual degrees)
 - ▶ If the sampling rate is 60 pixels per degree, then the Nyquist frequency is 30 cycles per degree
- ▶ When resampling an image to lower resolution, the frequency content above the Nyquist frequency needs to be removed (reduced in practice)
 - ▶ Otherwise **aliasing** is visible

Resolution and sampling rate

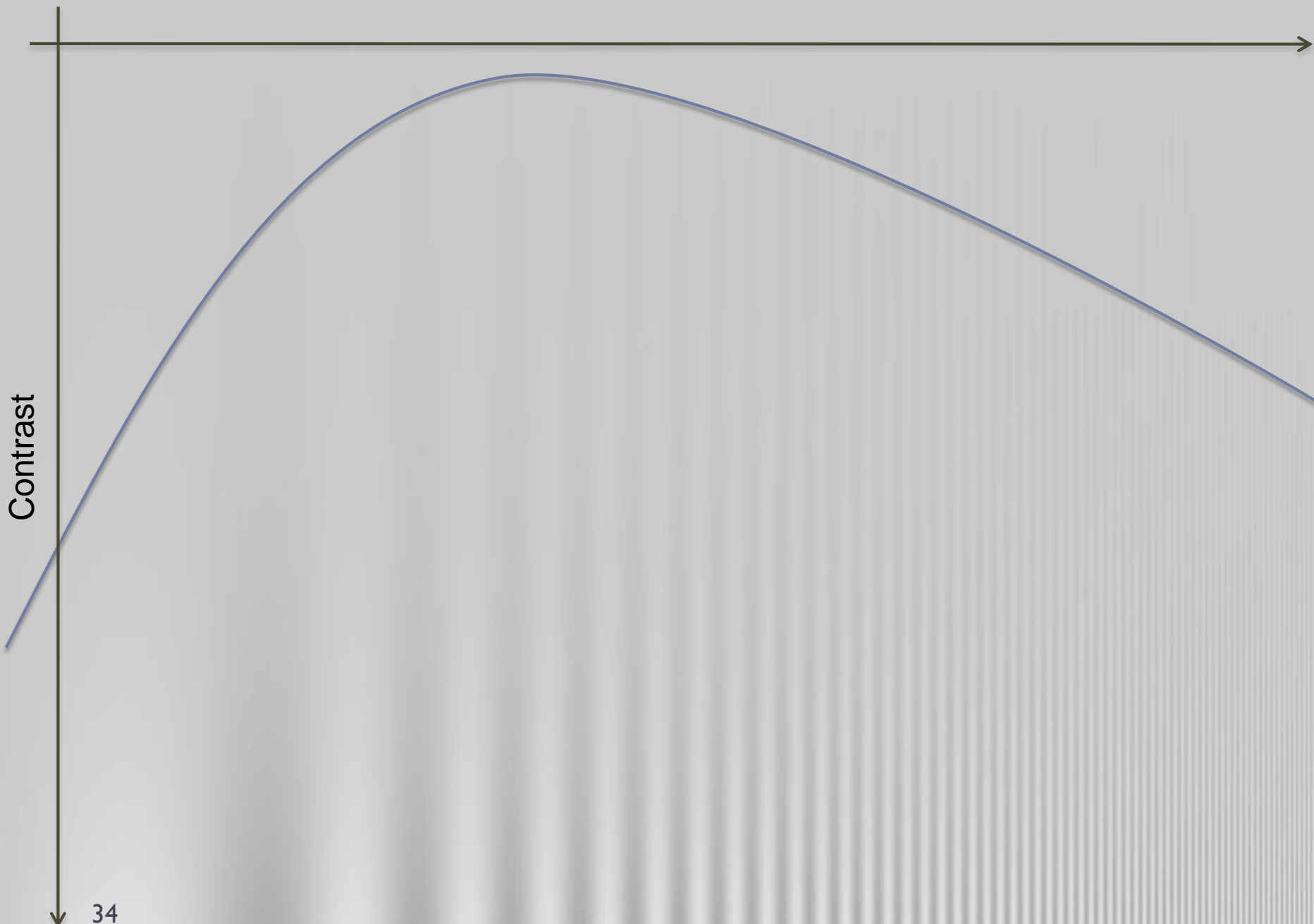
- ▶ **Pixels per inch [ppi]**
 - ▶ Does not account for vision
- ▶ **The visual resolution depends on**
 - ▶ screen size
 - ▶ screen resolution
 - ▶ viewing distance
- ▶ **The right measure**
 - ▶ Pixels per visual degree [ppd]
 - ▶ In frequency space
 - ▶ Cycles per visual degree [cpd]
 - ▶ Nyquist frequency [cpd] = $\frac{1}{2}$ of [ppd]



Modeling visual system



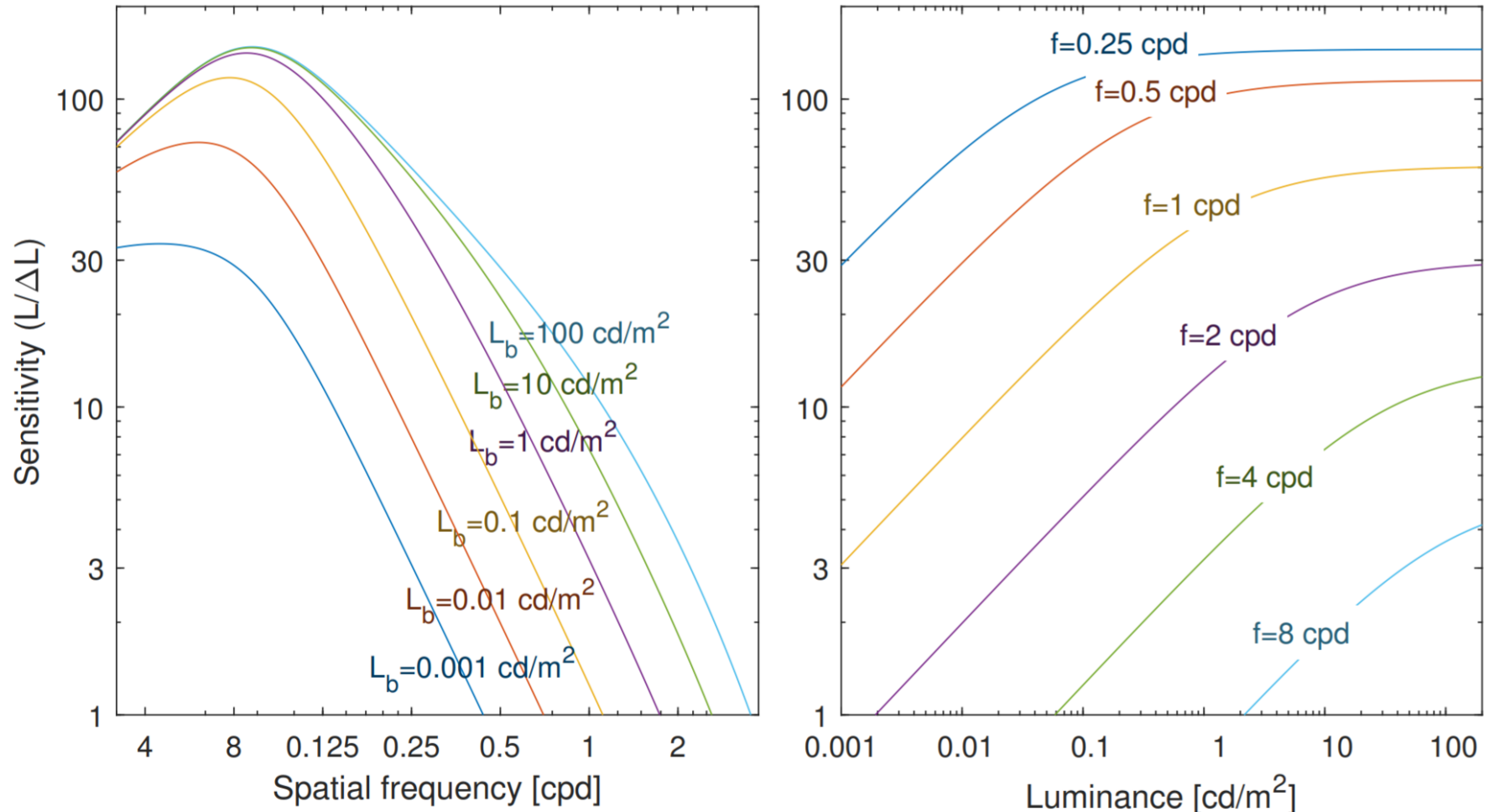
Spatial frequency [cycles per degree]



34

Campbell & Robson contrast sensitivity chart

CSF as a function of spatial frequency and background luminance



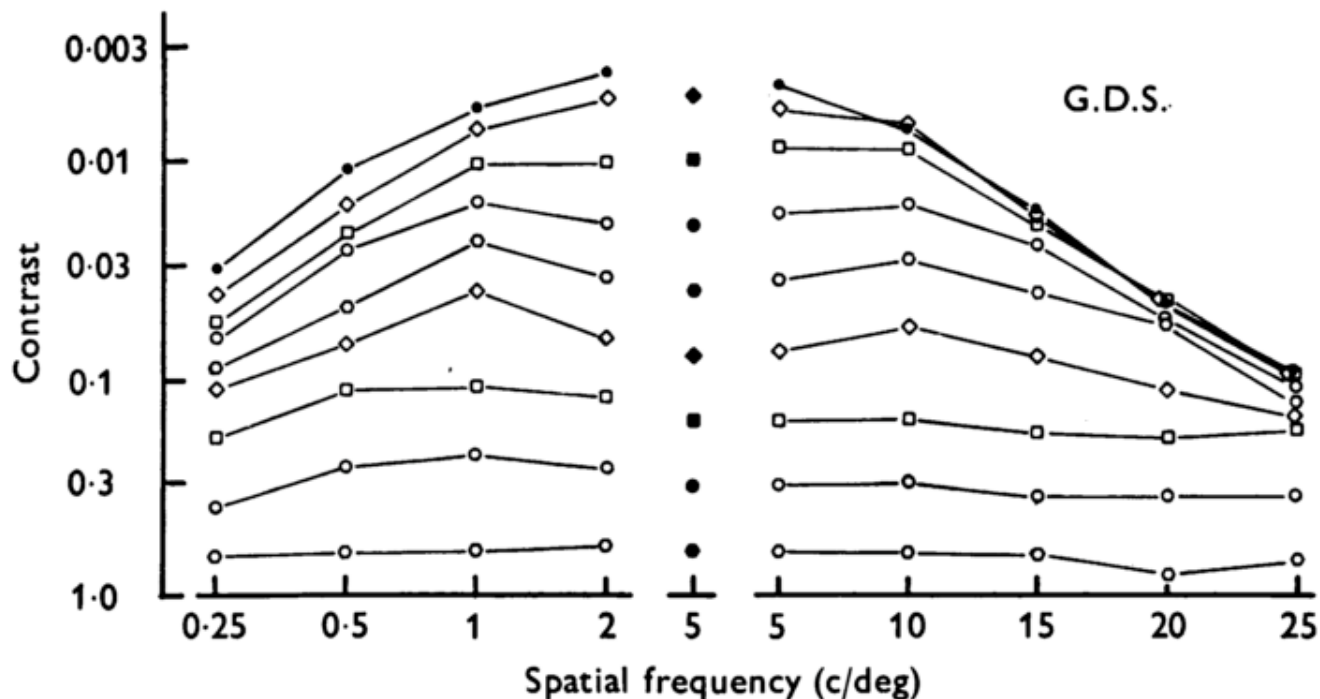
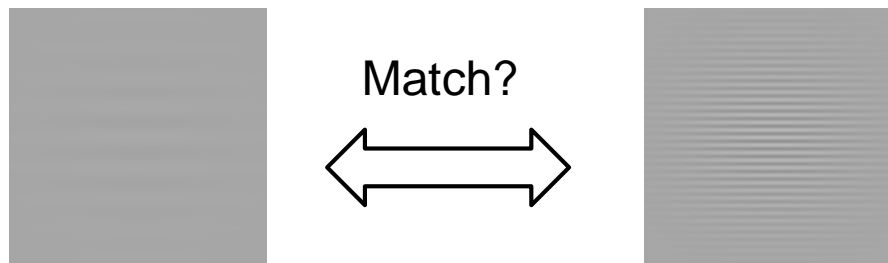


Contrast constancy



Contrast constancy

Experiment: Adjust the amplitude of one sinusoidal grating until it matches the perceived magnitude of another sinusoidal grating.



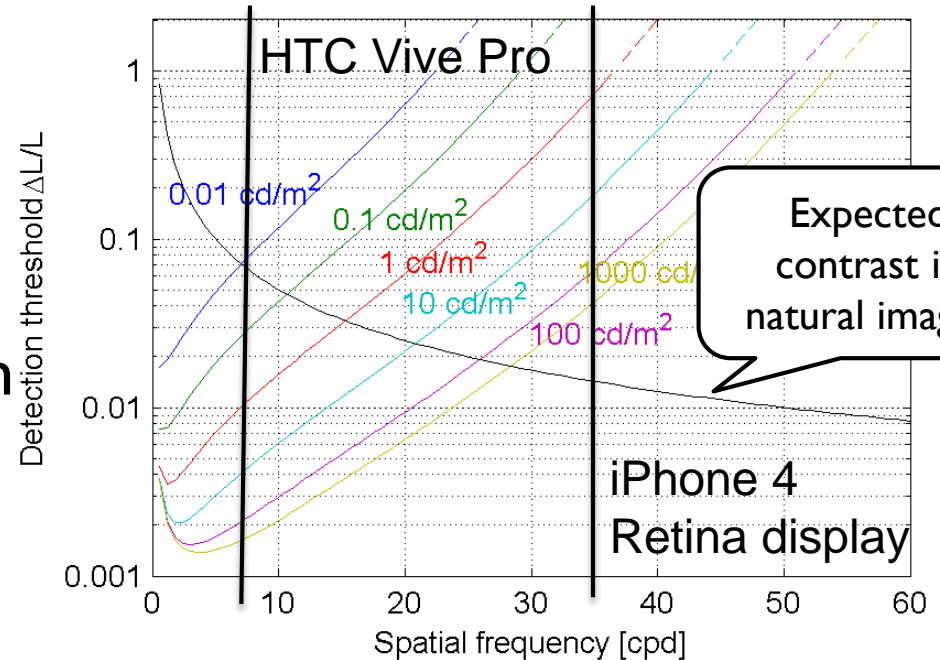
Contrast constancy
No CSF above the detection threshold

CSF and the resolution

- ▶ CSF plotted as the detection contrast

$$\frac{\Delta L}{L_b} = S^{-1}$$

- ▶ The contrast below each line is invisible
- ▶ Maximum perceivable resolution depends on luminance



CSF models:

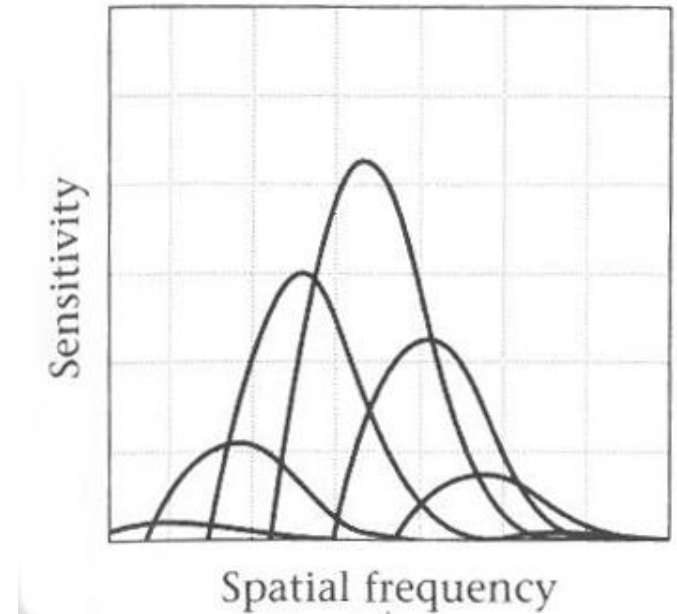
Barten, P. G. J. (2004).

<https://doi.org/10.1117/12.537476>

Multi-resolution models

Spatial-frequency selective channels

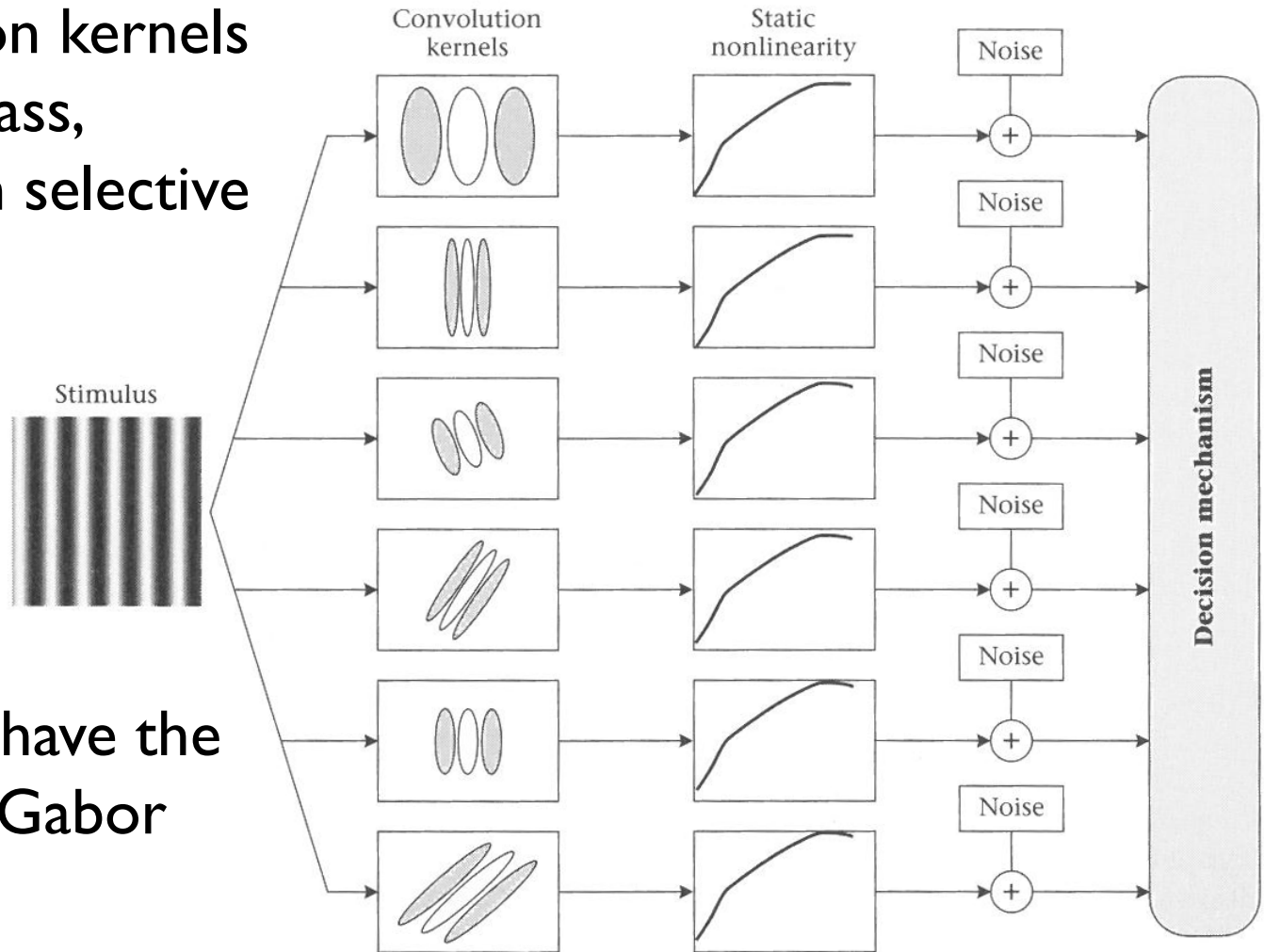
- ▶ The visual information is decomposed in the visual cortex into multiple channels
 - ▶ The channels are selective to spatial frequency, temporal frequency and orientation
 - ▶ Each channel is affected by different „noise” level
 - ▶ The CSF is the net result of information being passed in noise-affected visual channels



From: Wandell, 1995

Multi-resolution visual model

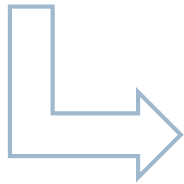
- ▶ Convolution kernels are band-pass, orientation selective filters



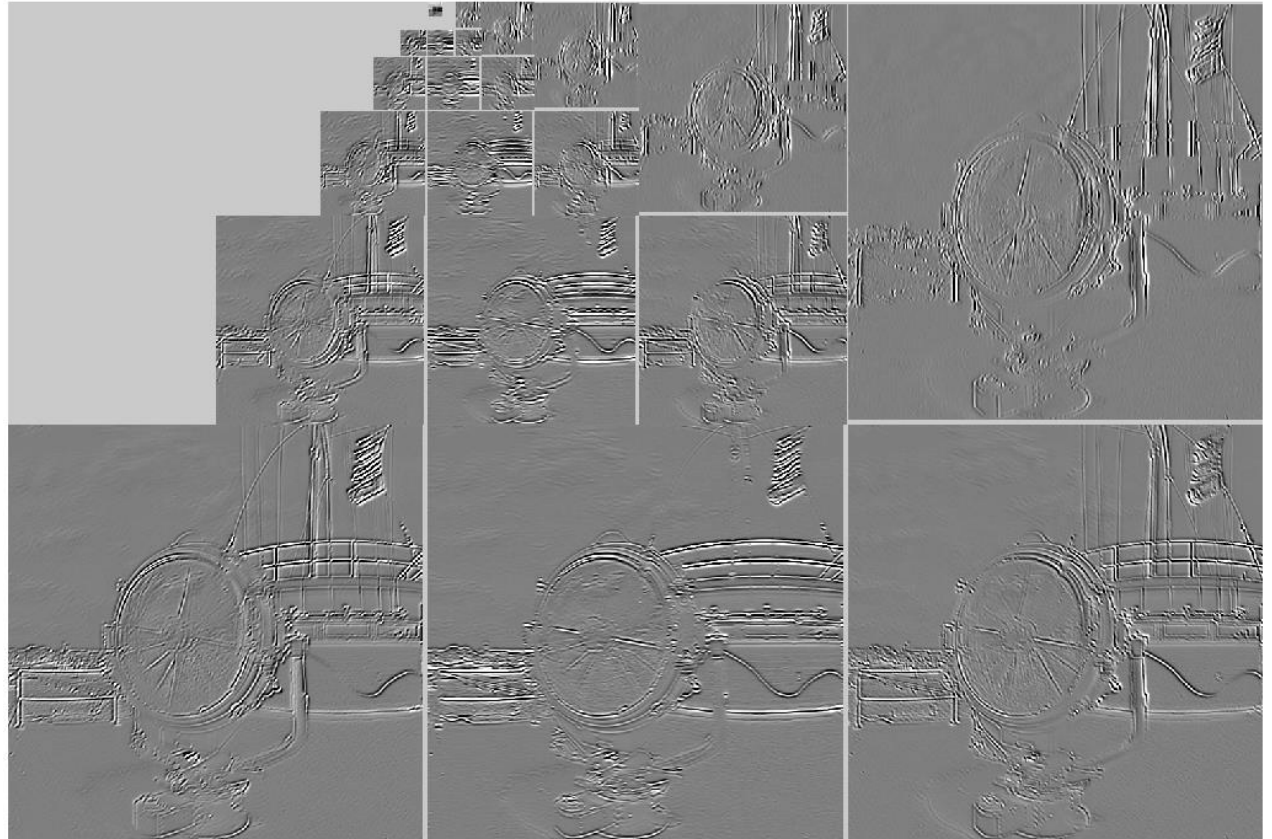
- ▶ The filters have the shape of a Gabor function

From: Wandell, 1995

Multi-scale decomposition

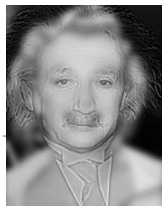
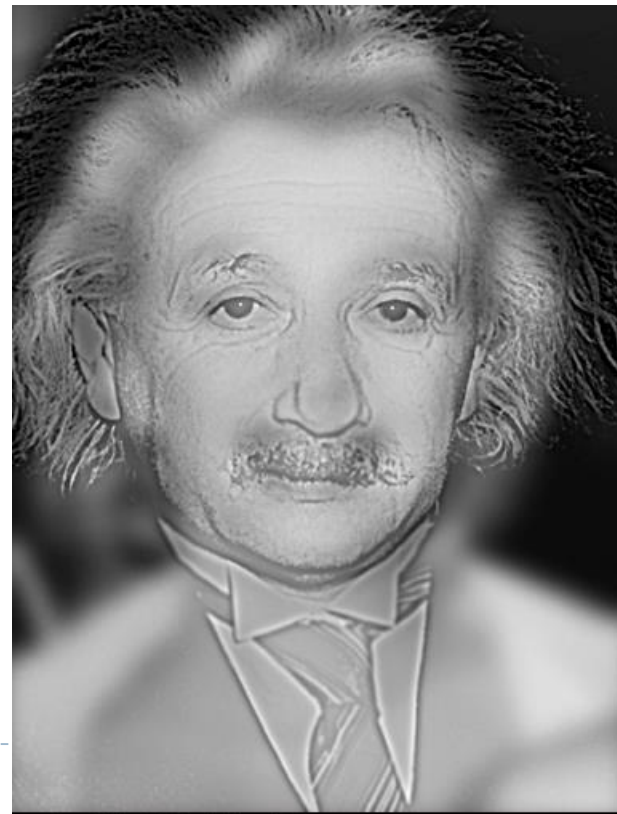


Steerable pyramid
decomposition



Applications of multi-scale models

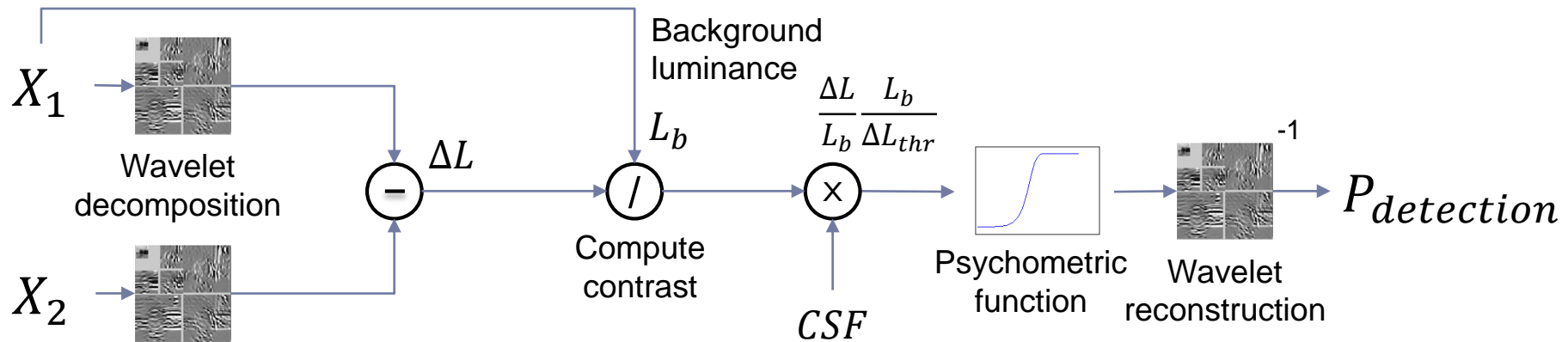
- ▶ **JPEG2000**
 - ▶ Wavelet decomposition
- ▶ **JPEG / MPEG**
 - ▶ Frequency transforms
- ▶ **Image pyramids**
 - ▶ Blending & stitching
 - ▶ Hybrid images



Predicting visible differences with CSF

- ▶ But we can use CSF to find the probability of spotting a difference between a pair of images X_1 and X_2 :

$$p(f[X_1] = f[X_2] | X_1, X_2, CSF)$$

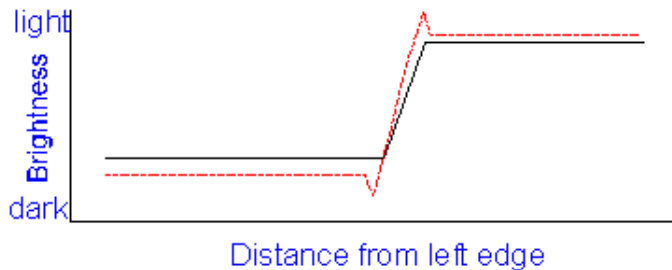
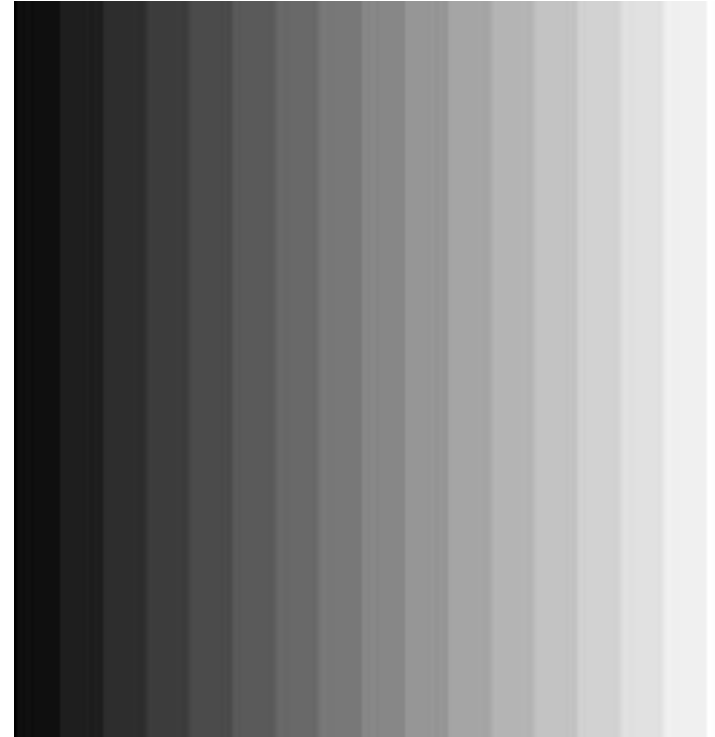
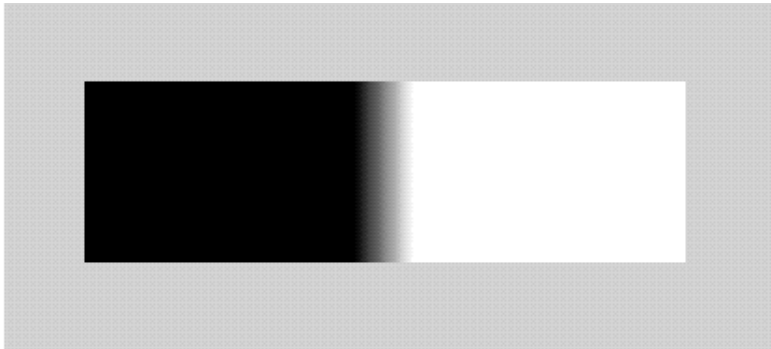


(simplified) Visual Difference Predictor

Daly, S. (1993).

Mach Bands – evidence for band-pass visual processing

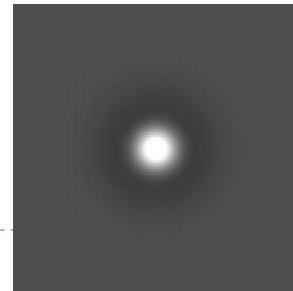
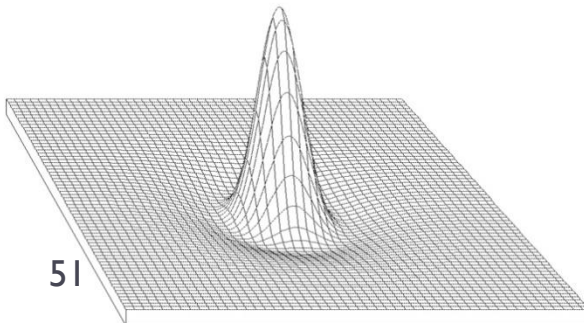
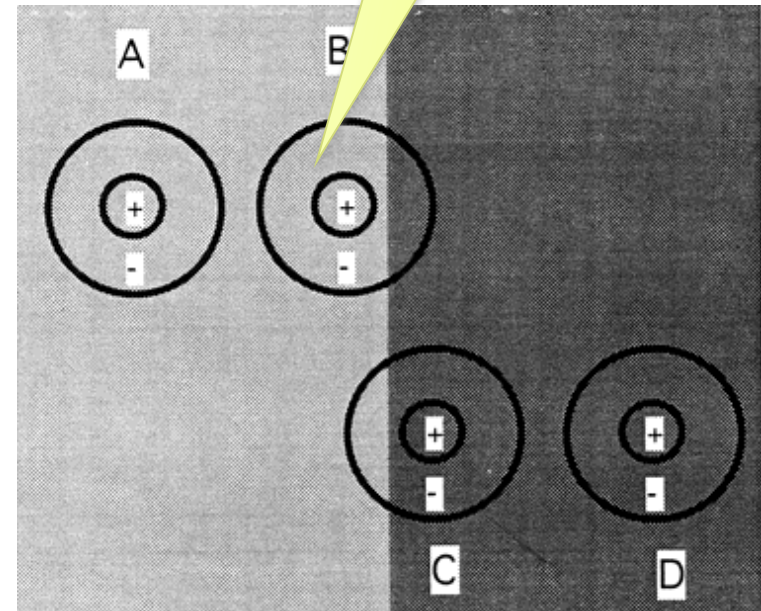
- “Overshooting” along edges
 - Extra-bright rims on bright sides
 - Extra-dark rims on dark sides
- Due to “Lateral Inhibition”



Centre-surround (Lateral Inhibition)

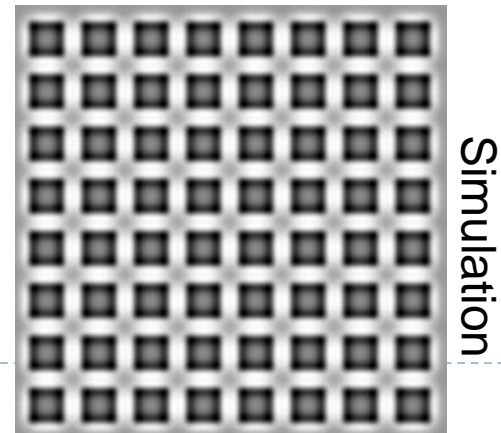
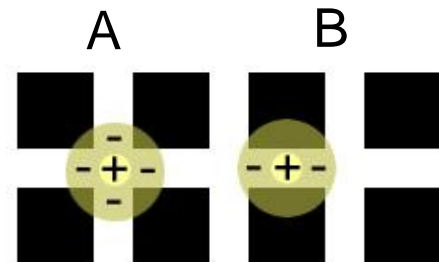
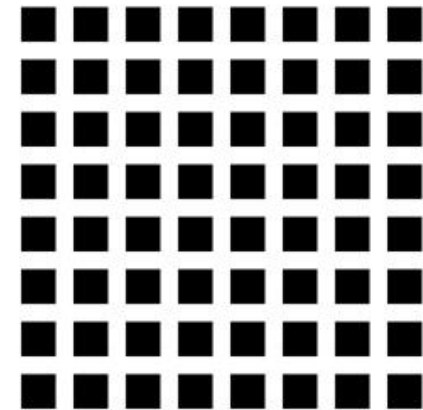
- ▶ “Pre-processing” step within the retina
 - ▶ Surrounding brightness level weighted negatively
 - ▶ A: high stimulus, maximal bright inhibition
 - ▶ B: high stimulus, reduced inhibition & stronger response
 - ▶ D: low stimulus, maximal inhibition
 - ▶ C: low stimulus, increased inhibition & weaker response

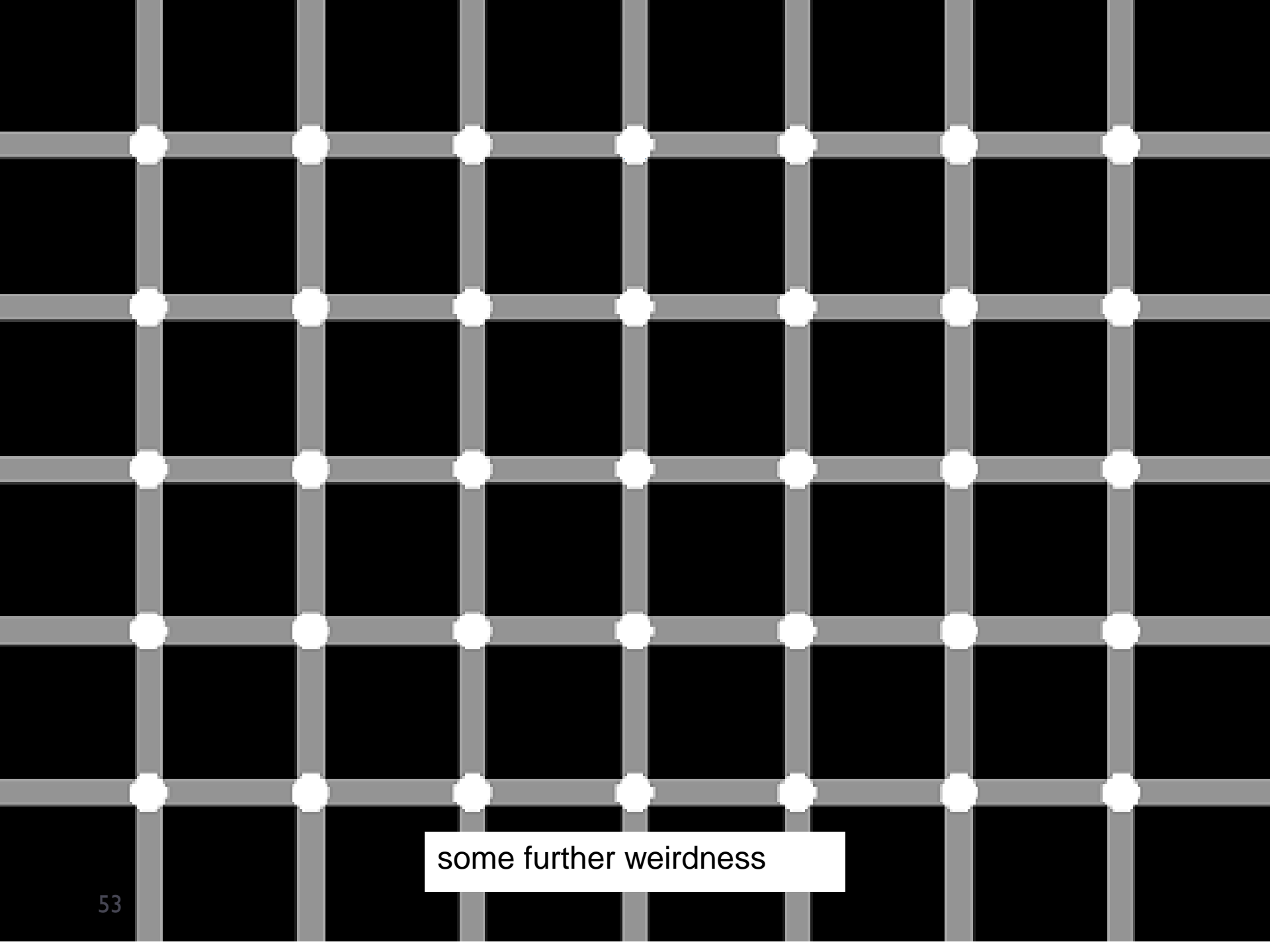
Center-surround
receptive fields
(groups of
photoreceptors)



Centre-surround: Hermann Grid

- Dark dots at crossings
- Explanation
 - Crossings (A)
 - More surround stimulation (more bright area)
 - ⇒ Less inhibition
 - ⇒ Weaker response
 - Streets (B)
 - Less surround stimulation
 - ⇒ More inhibition
 - ⇒ Greater response
- Simulation
 - Darker at crossings, brighter in streets
 - Appears more steady
 - What if reversed ?

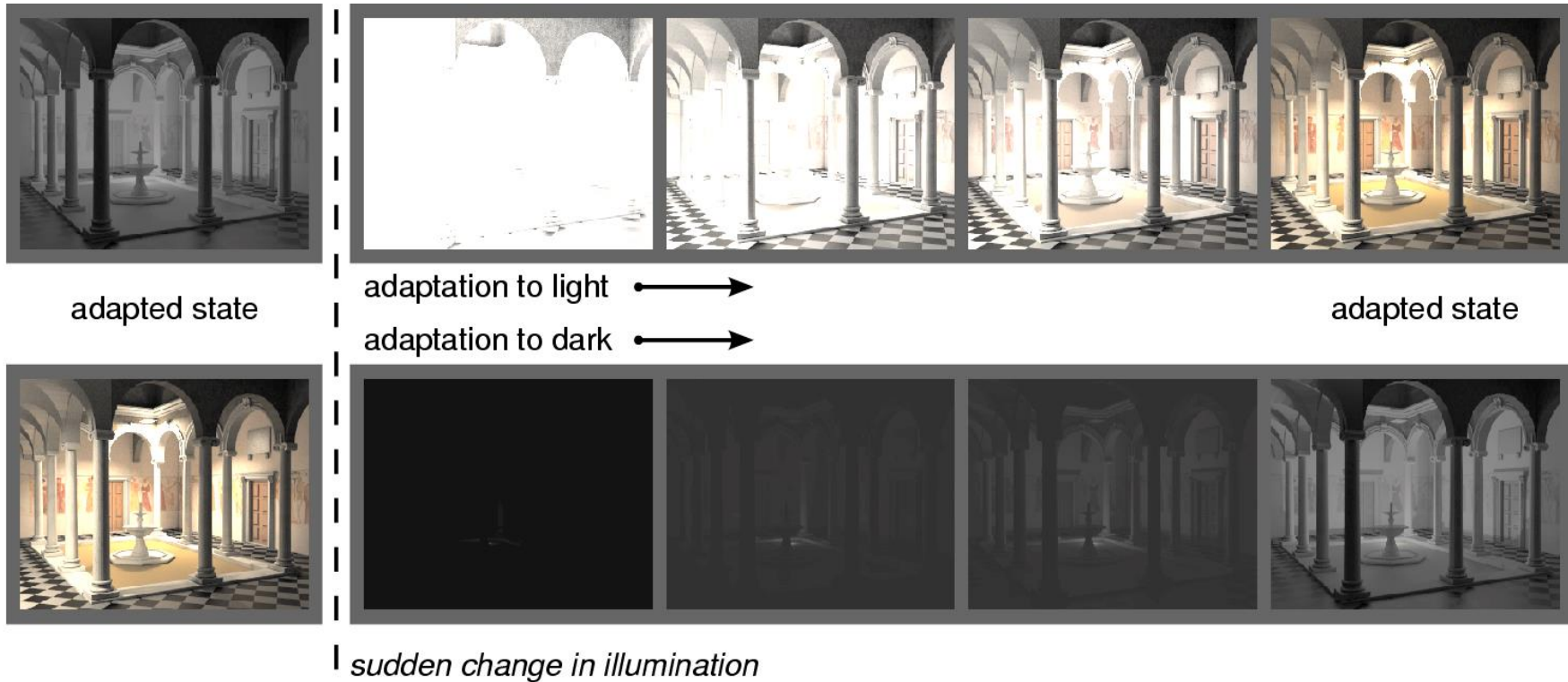




some further weirdness

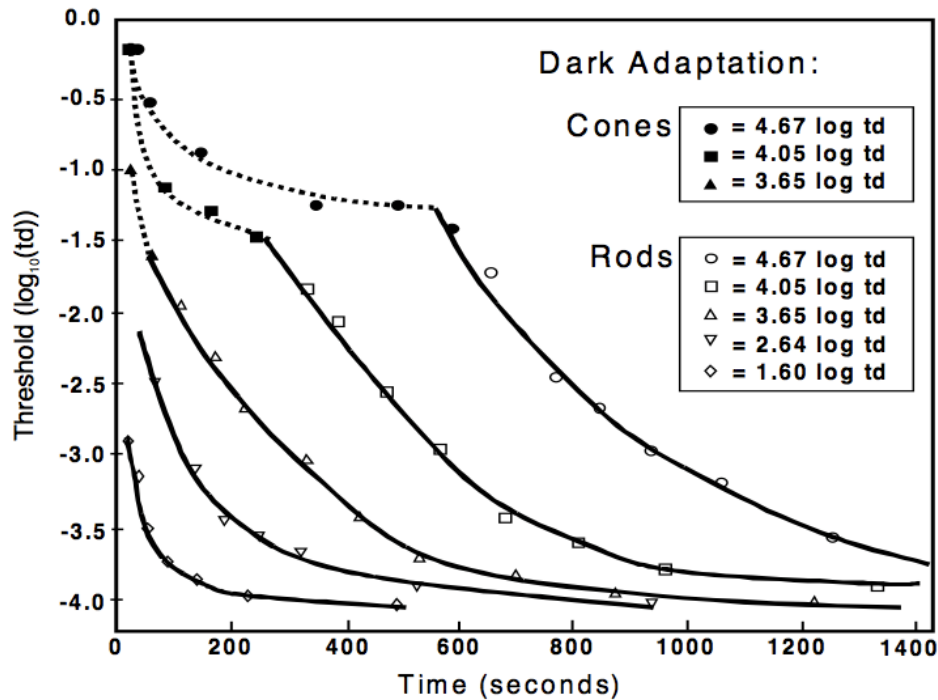
Light and dark adaptation

Light and dark adaptation

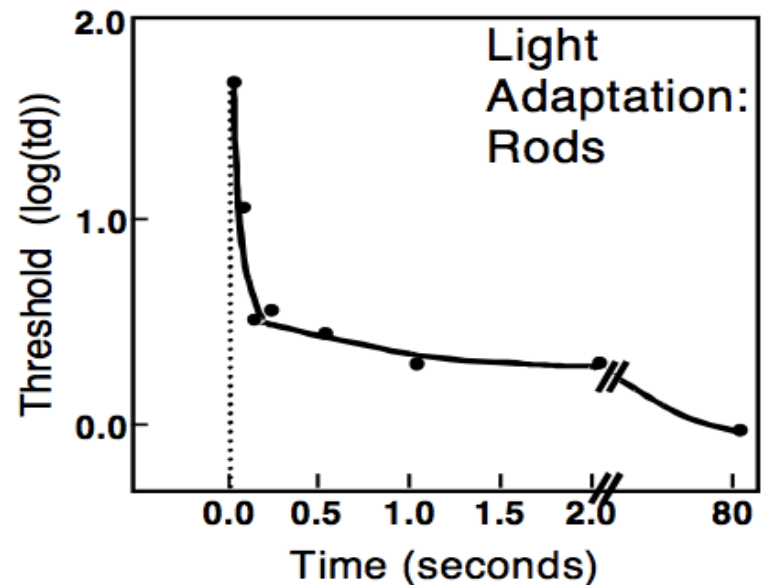
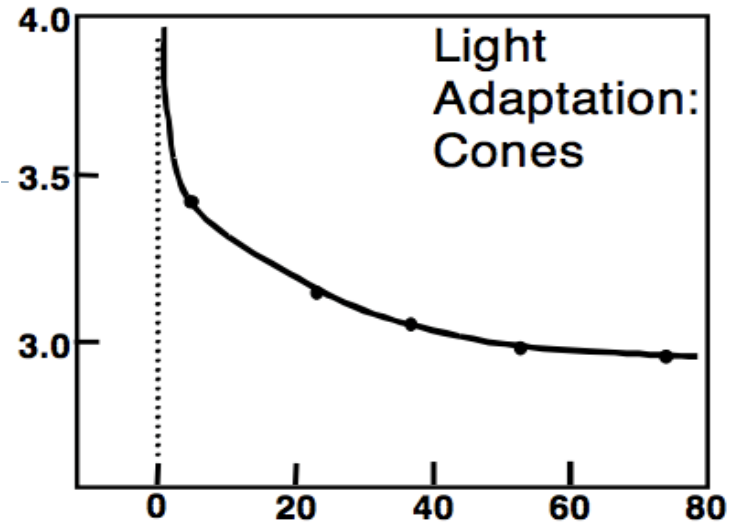


- ▶ Light adaptation: from dark to bright
- ▶ Dark adaptation: from bright to dark (much slower)

Time-course of adaptation



Bright -> Dark



Dark -> Bright

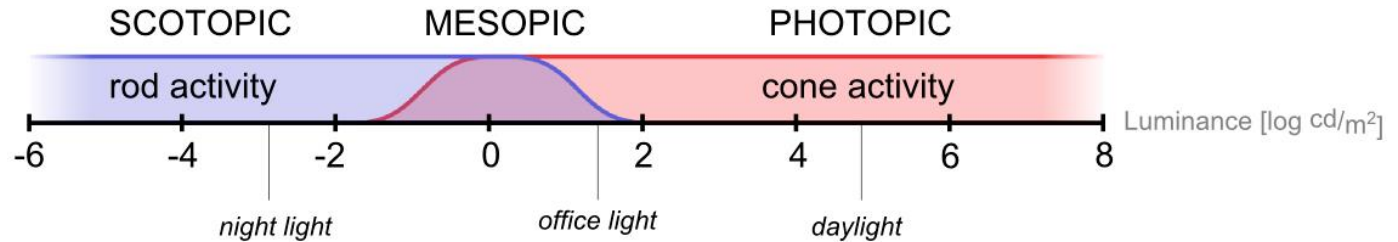


Temporal adaptation mechanisms

- ▶ **Bleaching & recovery of photopigment**
 - ▶ Slow asymmetric (light → dark, dark → light)
 - ▶ Reaction times (1-1000 sec)
 - ▶ Separate time-course for rods and cones
- ▶ **Neural adaptation**
 - ▶ Fast
 - ▶ Approx. symmetric reaction times (10-3000 ms)
- ▶ **Pupil**
 - ▶ Diameter varies between 3 and 8 mm
 - ▶ About 1:7 variation in retinal illumination

Night and daylight vision

Vision mode:



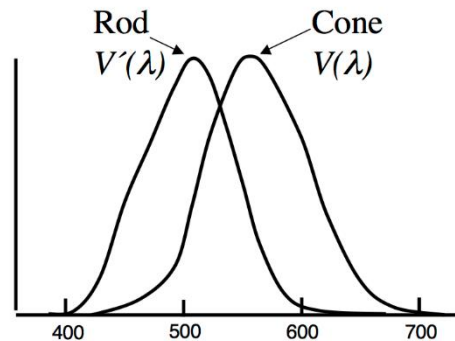
Mode properties:

monochromatic vision
limited visual acuity

good color perception
good visual acuity



Luminous efficiency



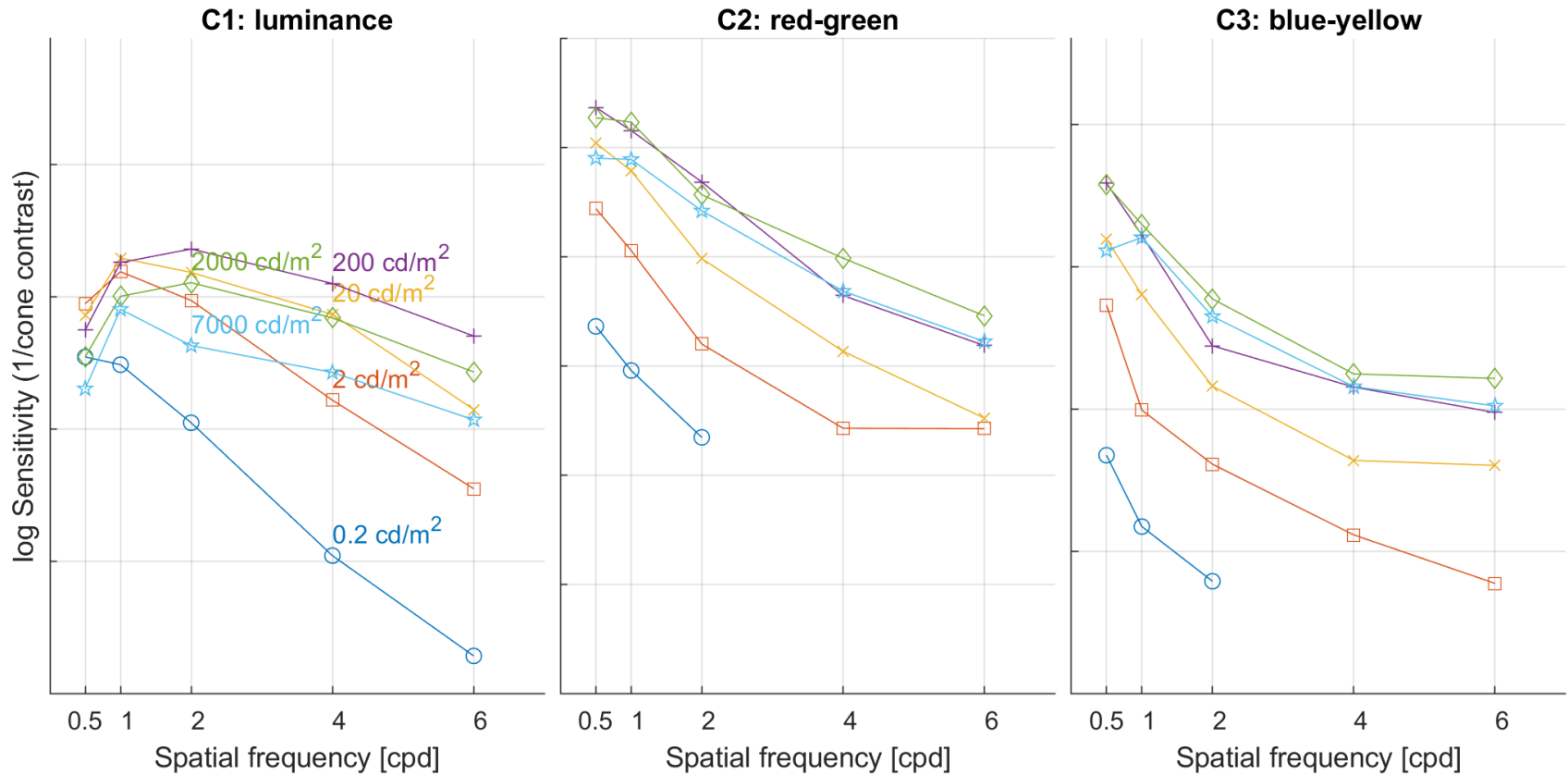


Spatial colour vision

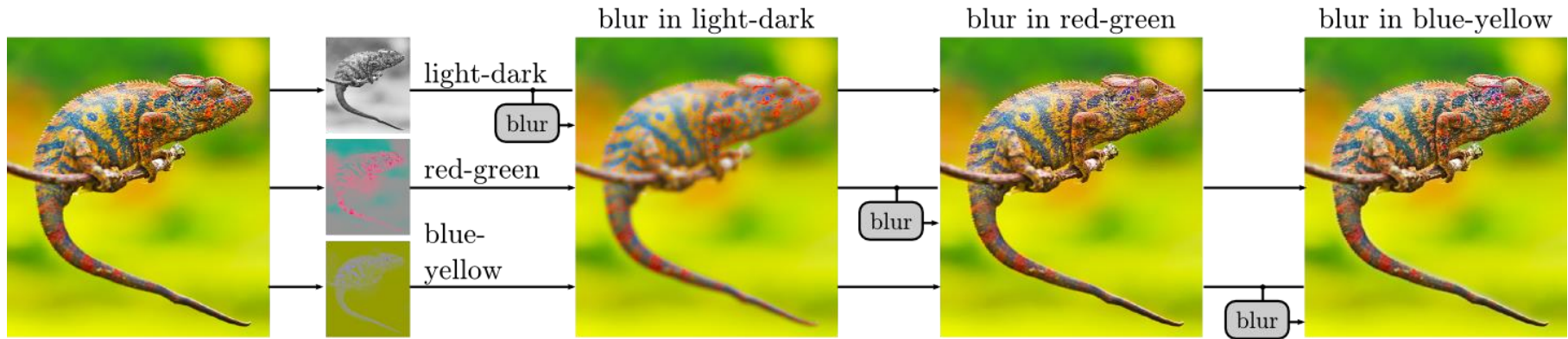
Spatio-chromatic CSF



Color CSF across the luminance range



Visibility of blur



- ▶ The same amount of blur was introduced into light-dark, red-green and blue-yellow colour opponent channels
- ▶ The blur is only visible in light-dark channel
- ▶ This property is used in image and video compression
 - ▶ Sub-sampling of colour channels (4:2:1)



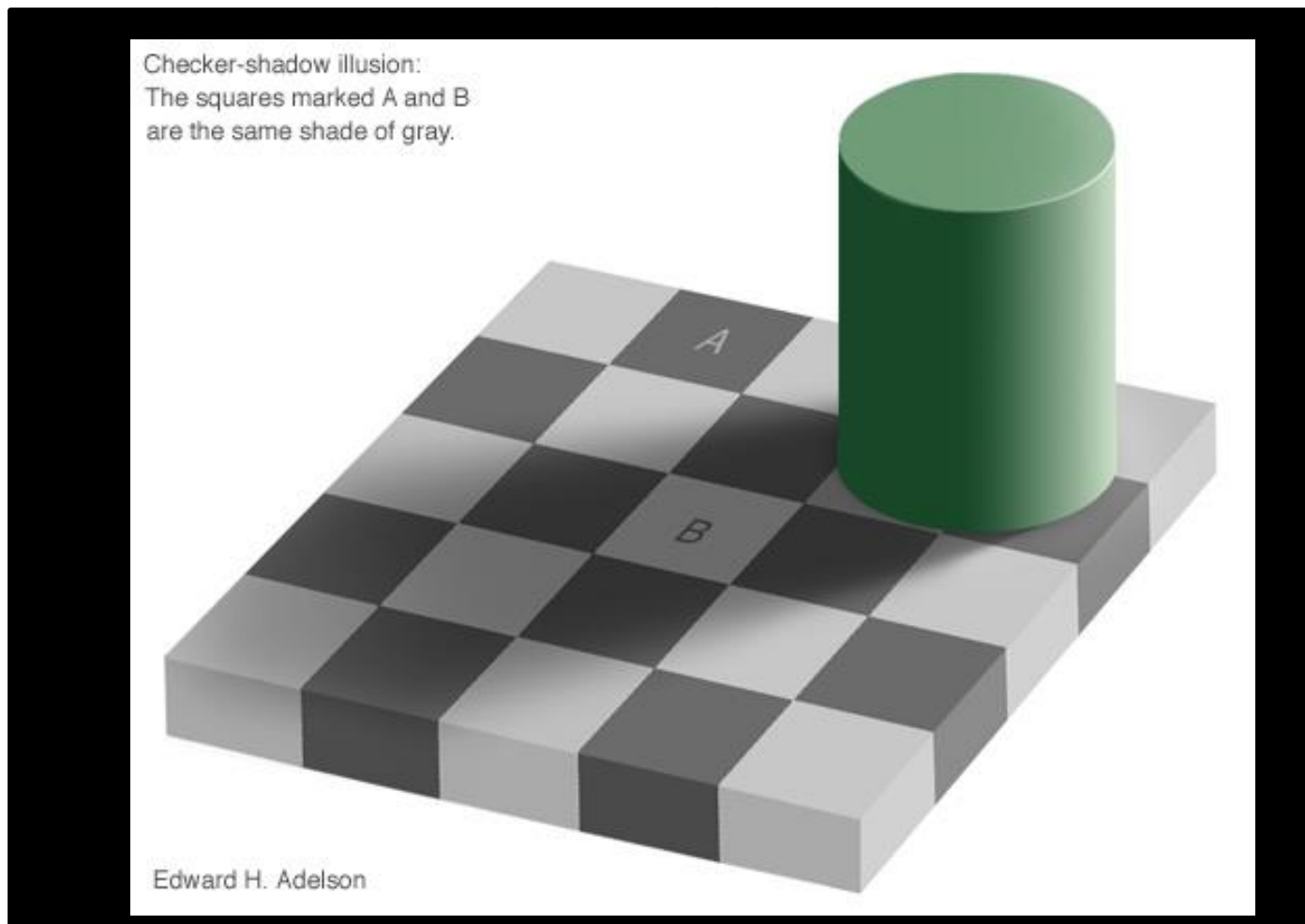
High(er) level vision



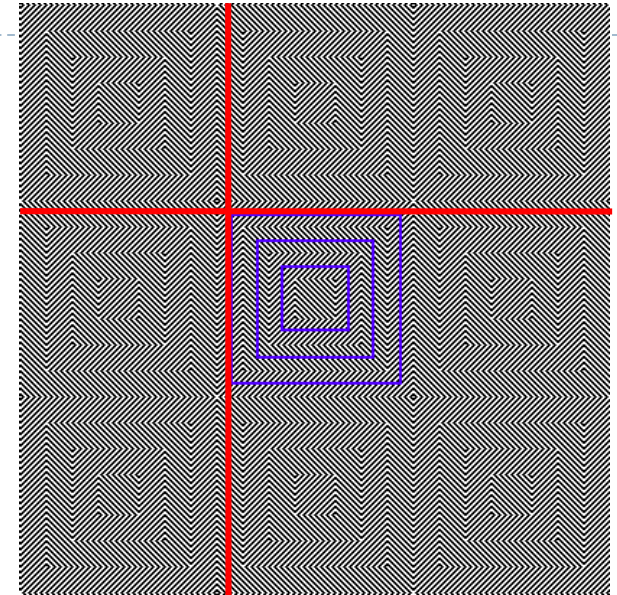
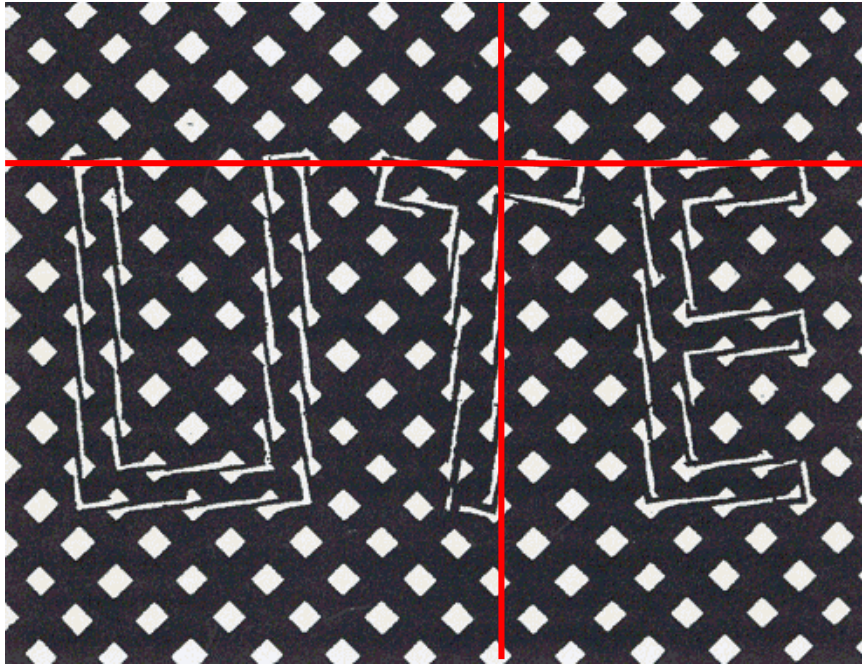
Simultaneous contrast



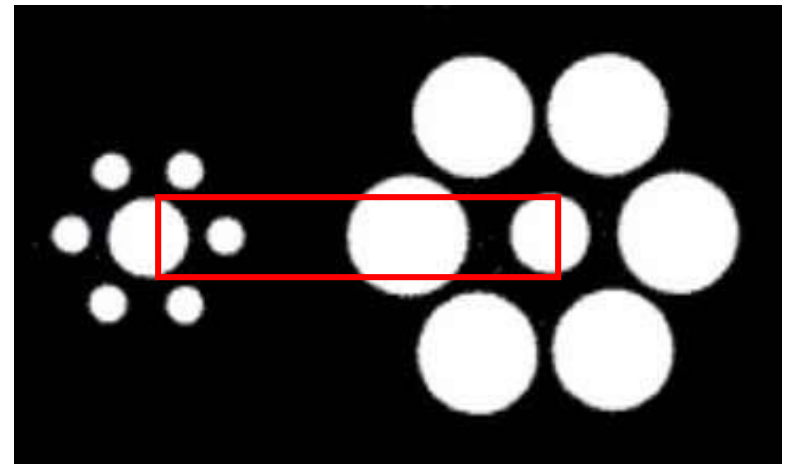
High-Level Contrast Processing



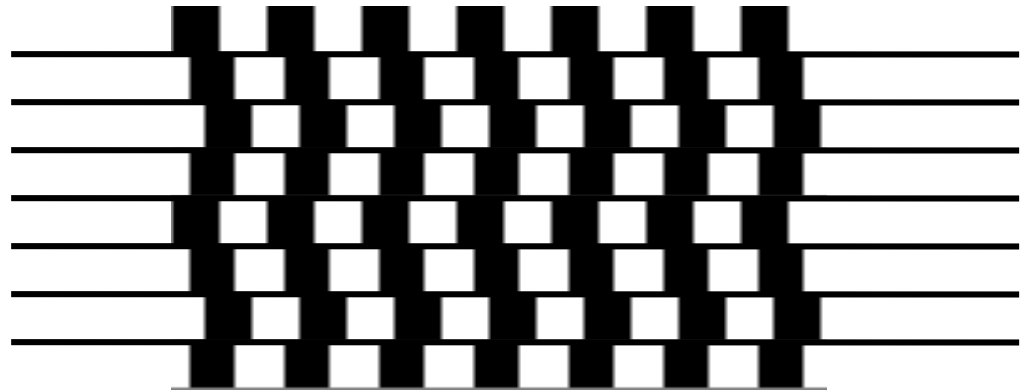
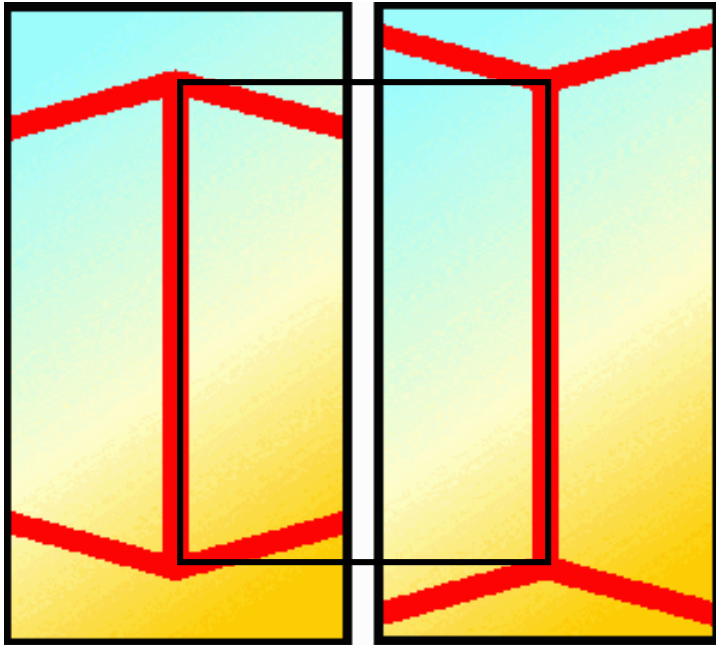
Shape Perception



- Depends on surrounding primitives
 - Directional emphasis
 - Size emphasis



Shape Processing: Geometrical Clues

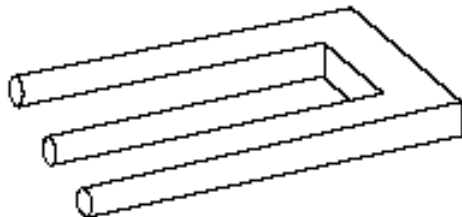
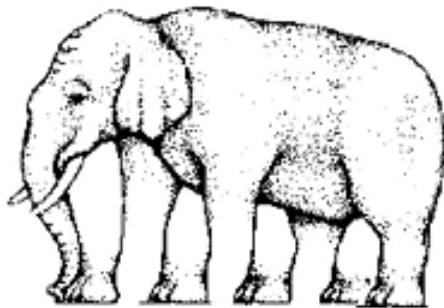


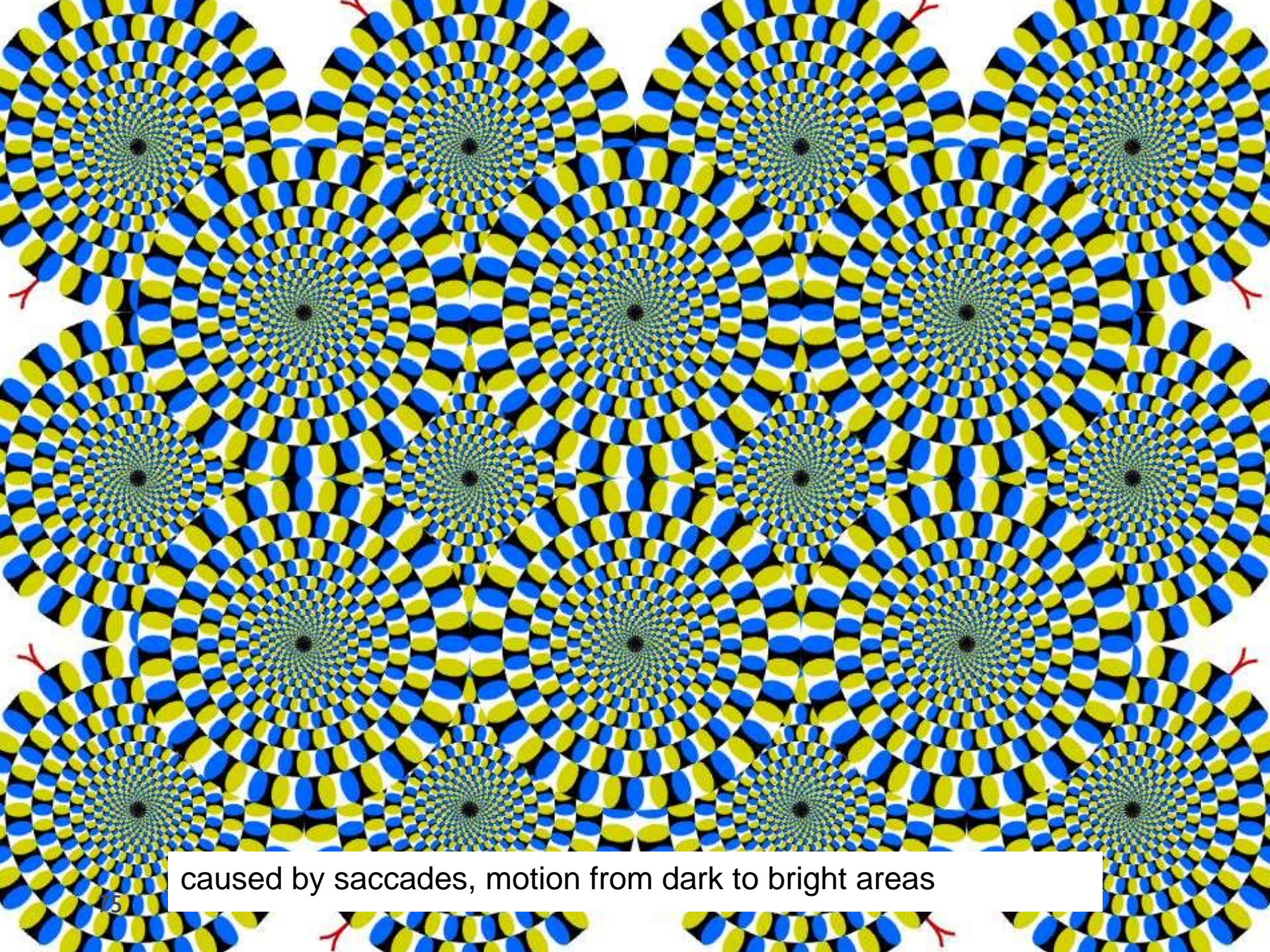
<http://www.panoptikum.net/optischetaeusungen/index.html>

- Automatic geometrical interpretation
 - 3D perspective
 - Implicit scene depth

Impossible Scenes

- Escher et.al.
 - Confuse HVS by presenting contradicting visual clues
 - Local vs. global processing





caused by saccades, motion from dark to bright areas

Law of closure



References

- ▶ Wandell, B.A. (1995). *Foundations of vision*. Sinauer Associates.
- ▶ Mantiuk, R. K., Myszkowski, K., & Seidel, H. (2015). High Dynamic Range Imaging. In *Wiley Encyclopedia of Electrical and Electronics Engineering*. Wiley.
 - ▶ Section 2.4
 - ▶ Available online:
http://www.cl.cam.ac.uk/~rkm38/hdri_book.html