



Models of early visual perception

Advanced Graphics

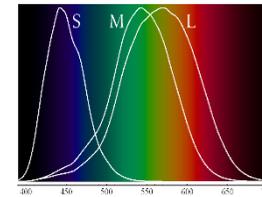
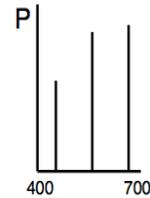
Rafal Mantiuk

Computer Laboratory, University of Cambridge

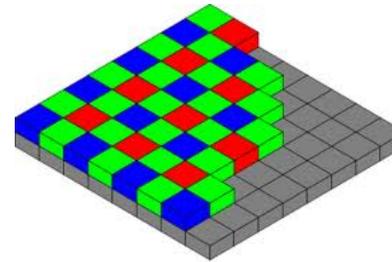
Many technical solutions are motivated by visual perception



Image & video compression



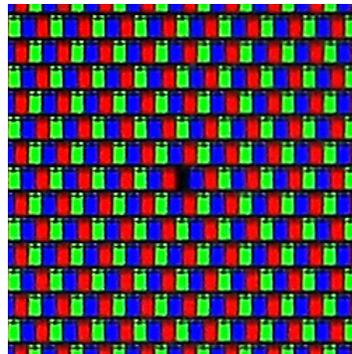
Display spectral emission - metamerism



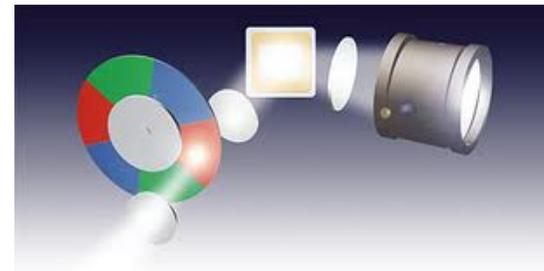
Camera's Bayer pattern



Halftoning



Display's subpixels



Color wheel in DLPs



Perceived brightness of light

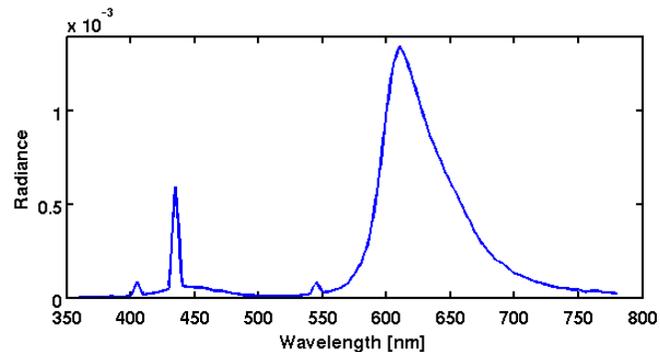
Luminance

- ▶ Luminance – how bright the surface will appear regardless of its colour. Units: cd/m^2

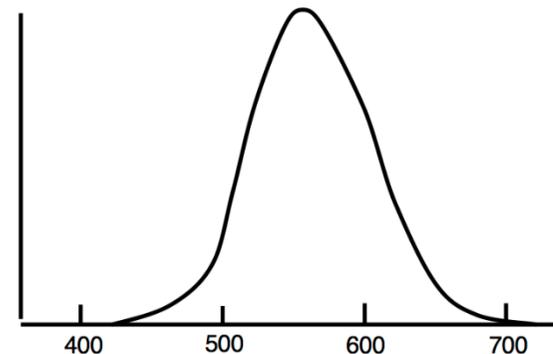
Luminance

$$L_V = \int_0^{\infty} L(\lambda) \cdot V(\lambda) d\lambda$$

Light spectrum (radiance)



Luminous efficiency function (weighting)



Luminance and Luma

▶ Luminance

- ▶ Photometric quantity defined by the spectral luminous efficiency function
- ▶ $L \approx 0.2126 R + 0.7152 G + 0.0722 B$
- ▶ Units: cd/m^2

▶ Luma

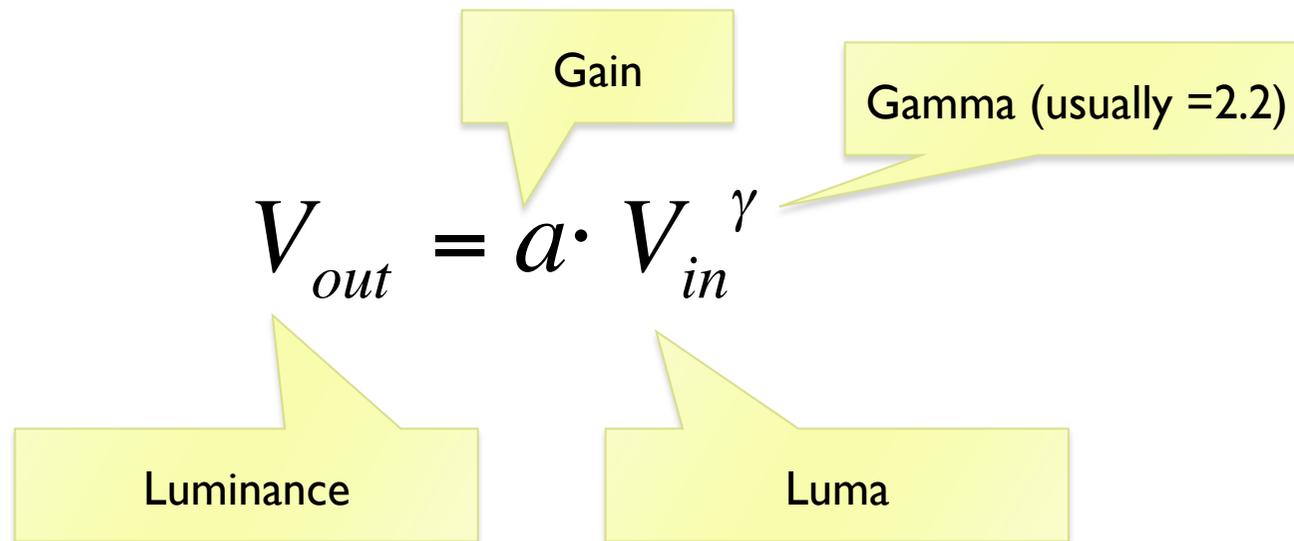
- ▶ Gray-scale value computed from LDR (gamma corrected) image
- ▶ $Y = 0.2126 R' + 0.7152 G' + 0.0722 B'$
- ▶ R' – prime denotes gamma correction

$$R' = R^{1/\gamma}$$

- ▶ Unitless

Gamma correction (reminder – Color lecture)

- ▶ Gamma correction is used to encode luminance or tristimulus color values (RGB) in imaging systems (displays, printers, cameras, etc.)



For color images: $R = a \cdot (R')^{\gamma}$ and the same for green and blue

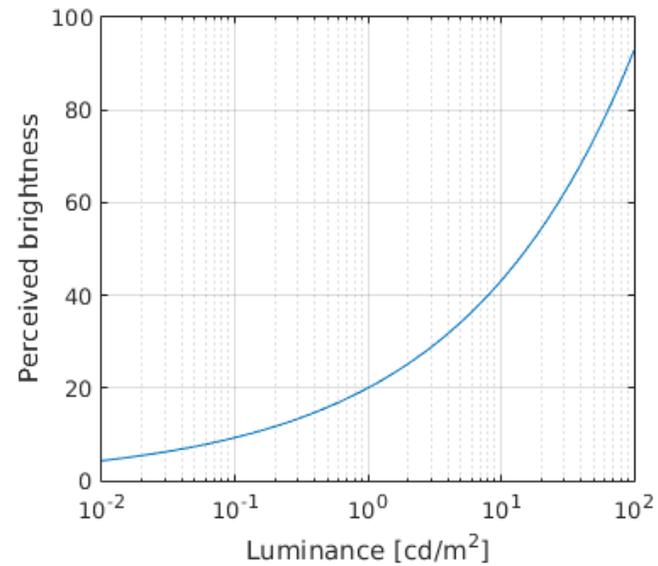
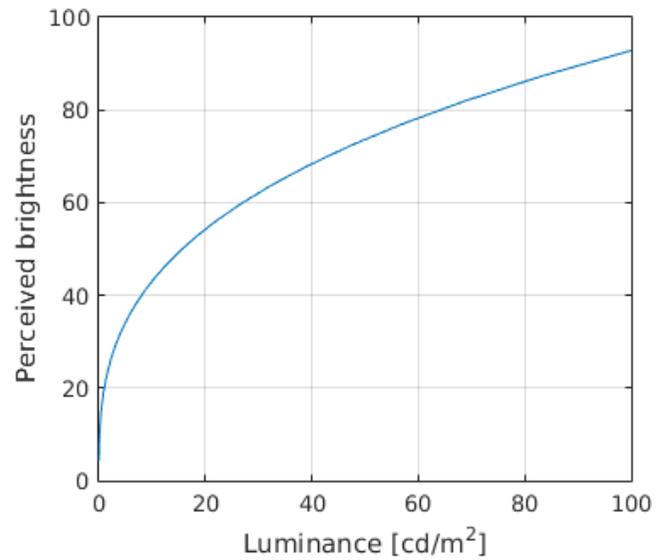
Steven's power law for brightness

- ▶ 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 $\varphi(I) = kI^a$ with four callout boxes pointing to its parts: 'Perceived magnitude' points to $\varphi(I)$, 'Constant' points to k , 'Exponent' points to a , and 'Physical stimulus' points to I .

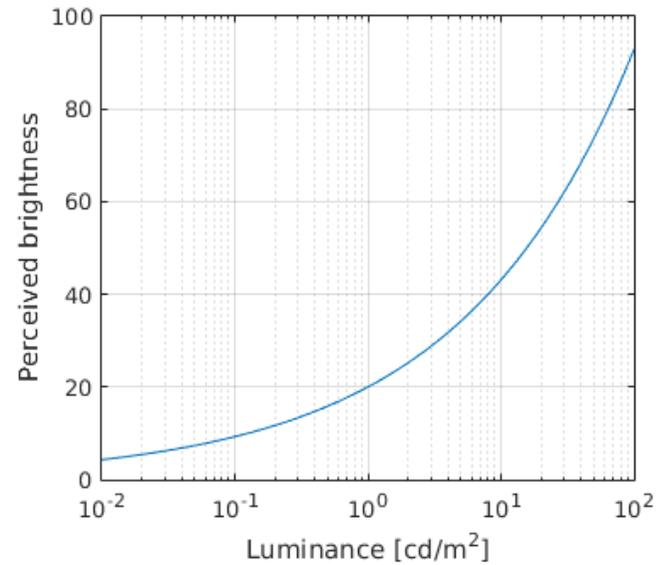
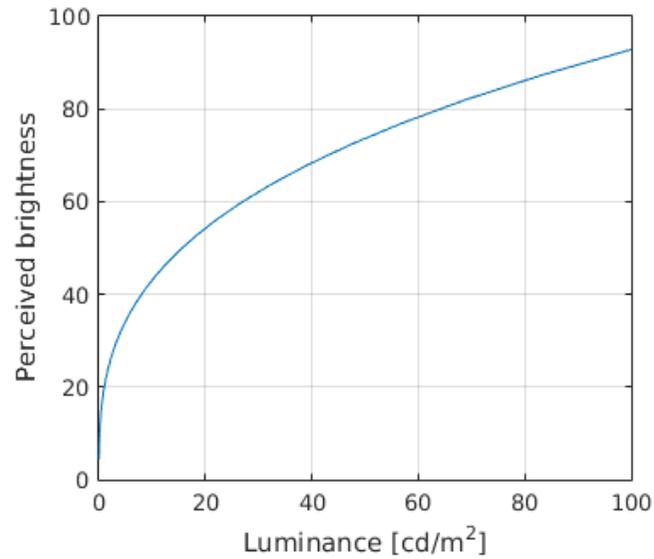
- ▶ For brightness (5 deg target in dark), $a = 0.3$

Steven's law for brightness

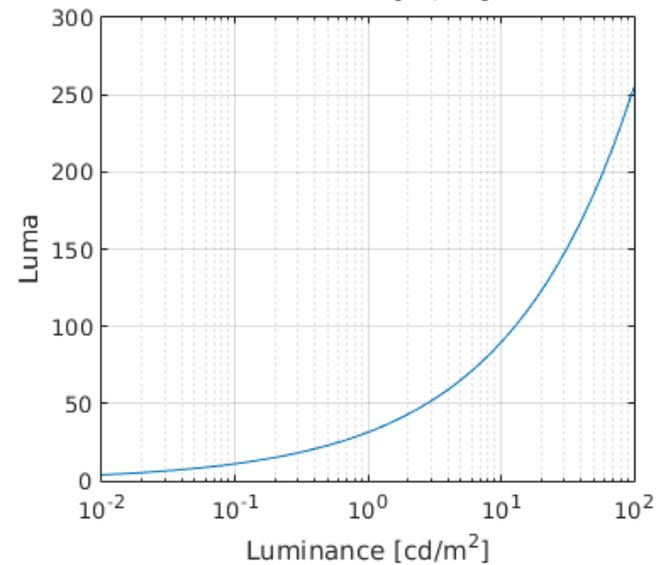
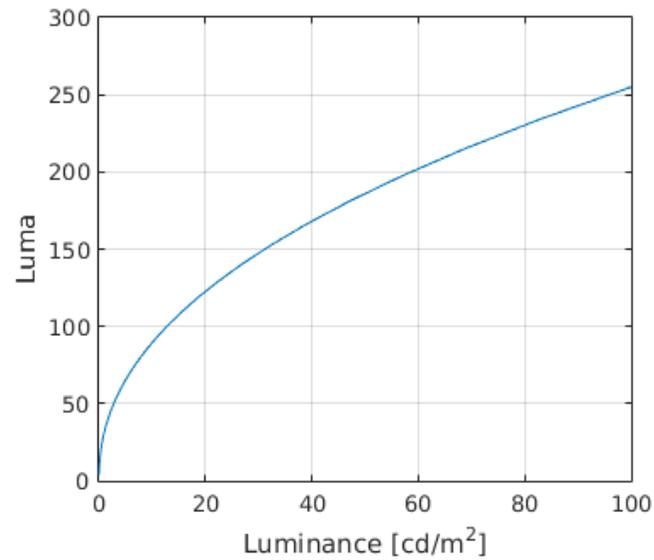


Steven's law vs. Gamma correction

Stevens' law
 $a=0.3$



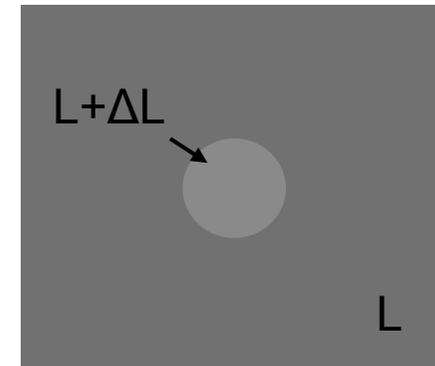
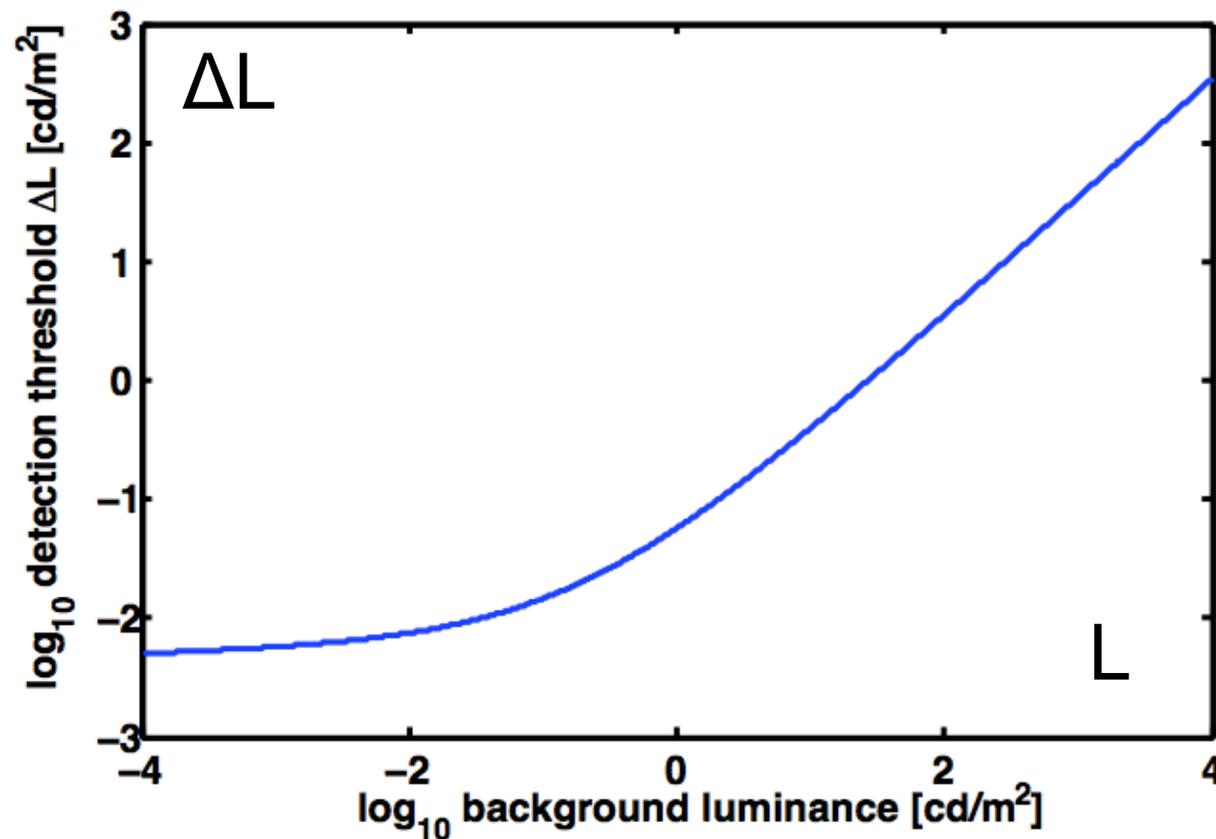
Gamma function
Gamma = 2.2



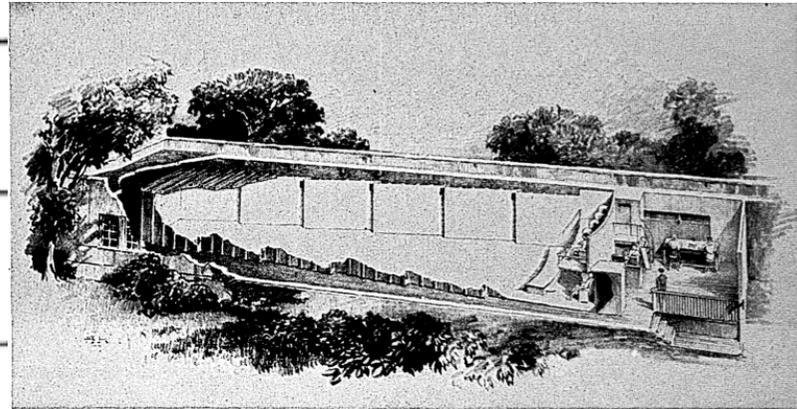
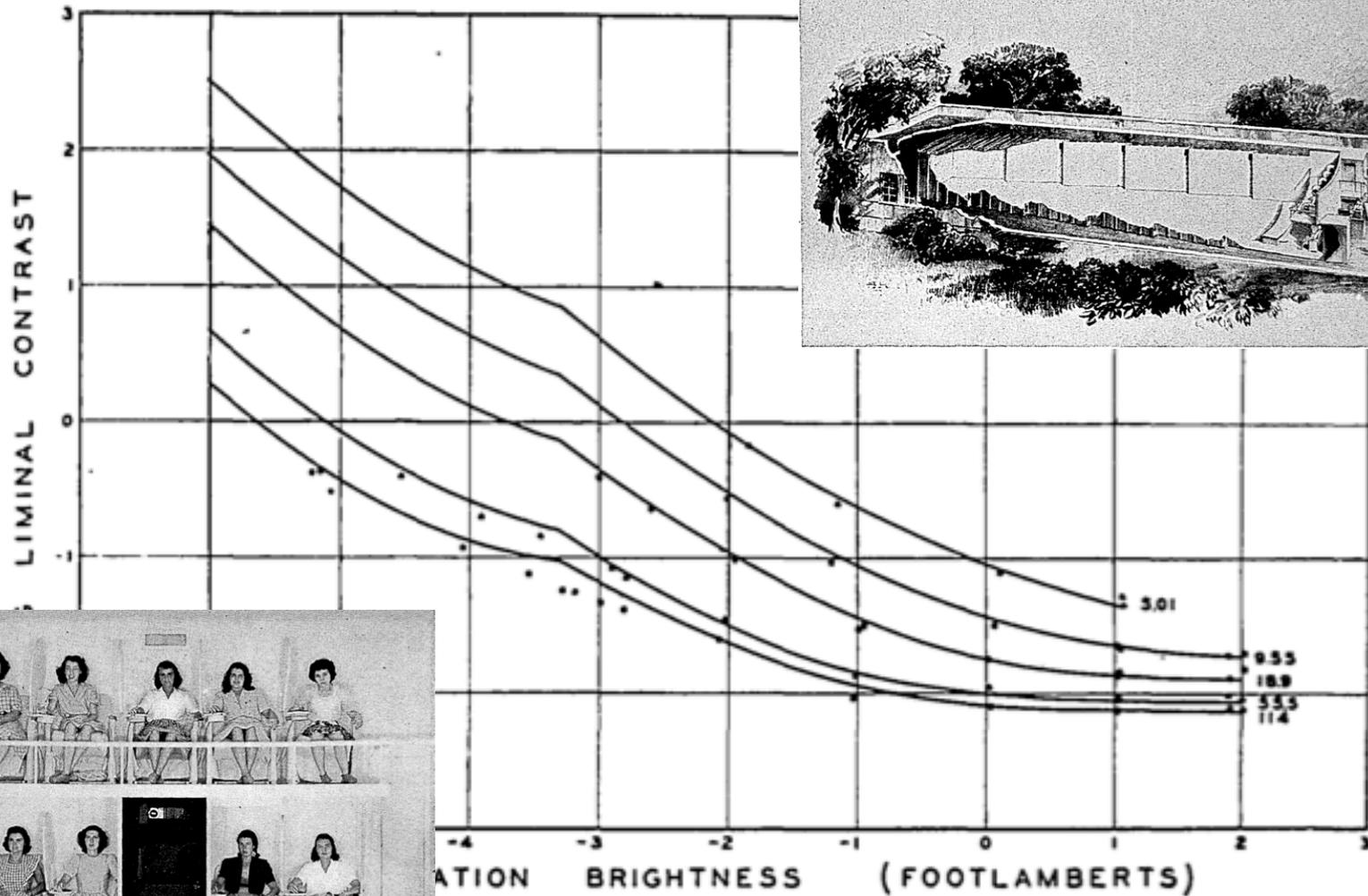
Detection and discrimination

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

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

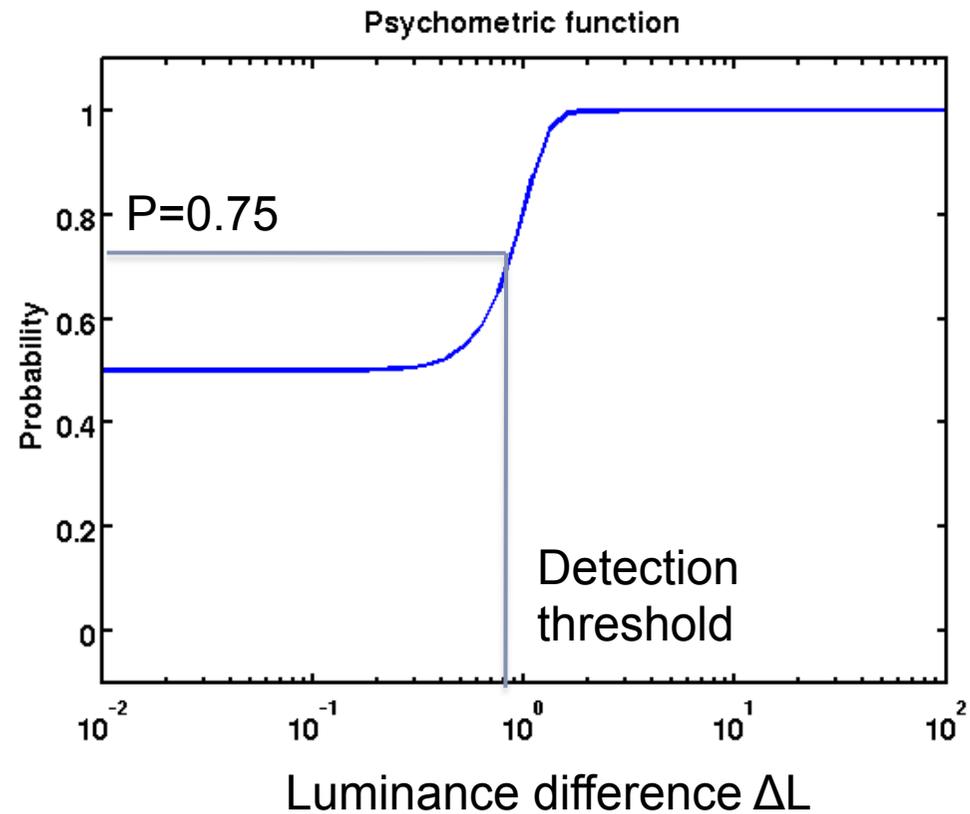
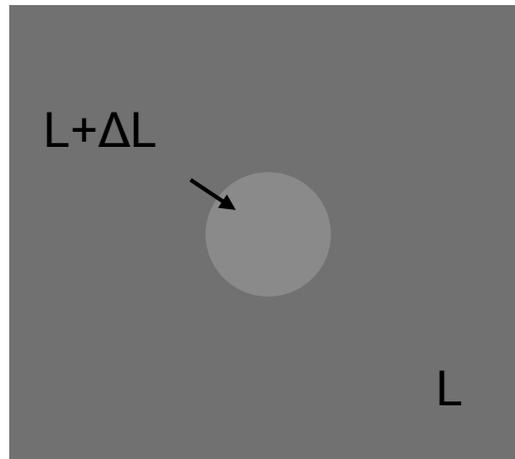


t.v.i. measurements – Blackwell 1946



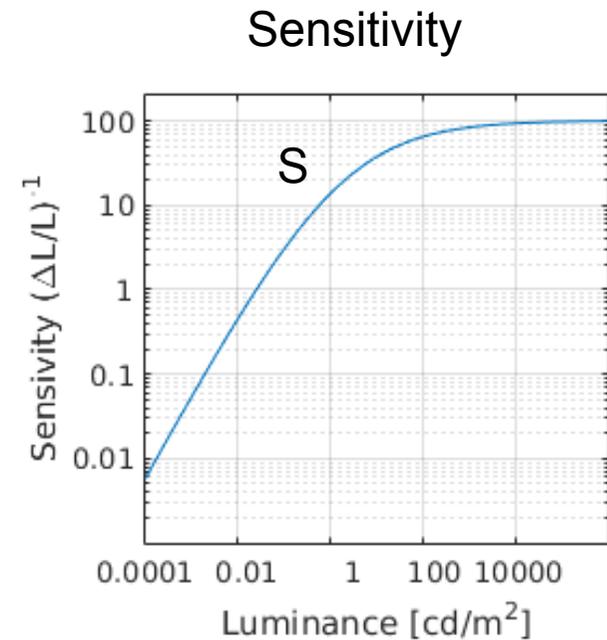
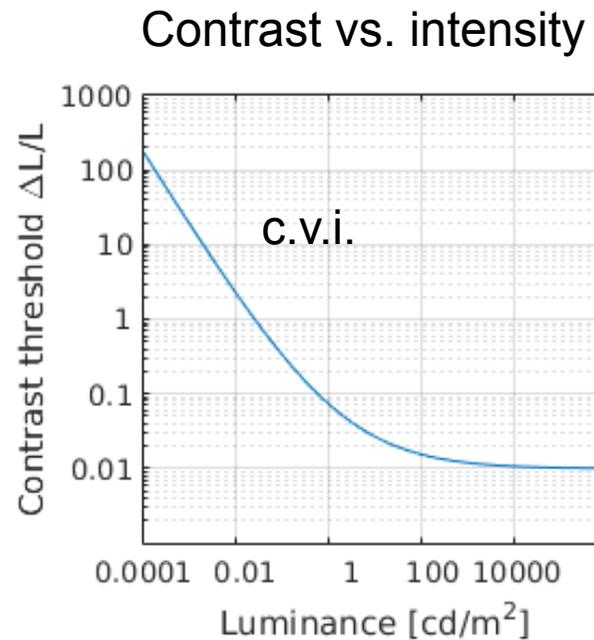
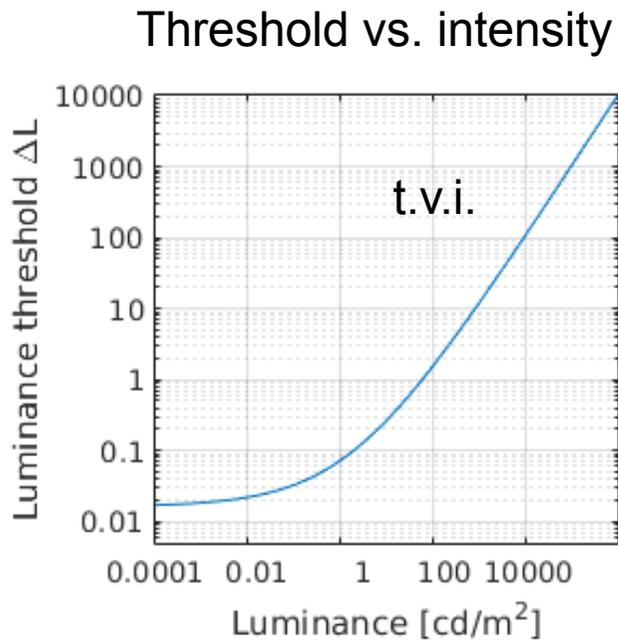
Psychophysics

Threshold experiments



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

- ▶ The same data, different representation



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

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

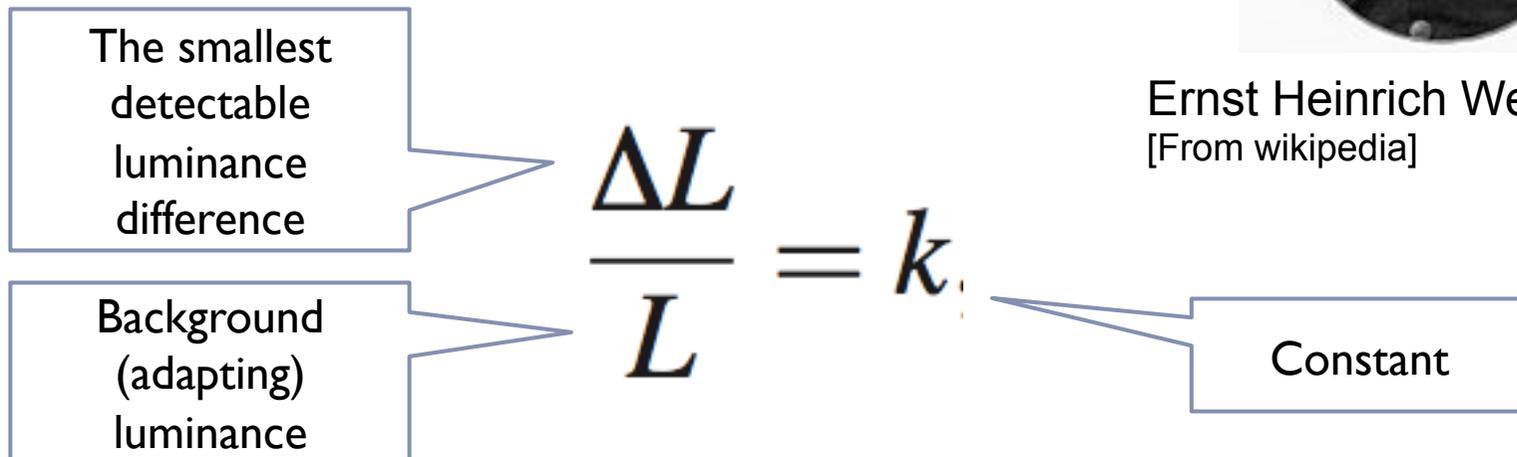
$$S = \frac{1}{T} = \frac{L}{\Delta L}$$

Sensitivity to luminance

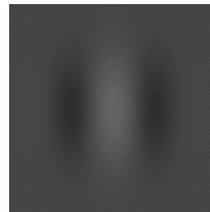
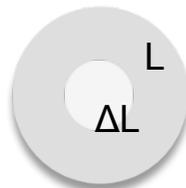
- ▶ Weber-law – the just-noticeable difference is proportional to the magnitude of a stimulus



Ernst Heinrich Weber
[From wikipedia]



Typical stimuli:



Consequence of the Weber-law

- ▶ Smallest detectable difference in luminance

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

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

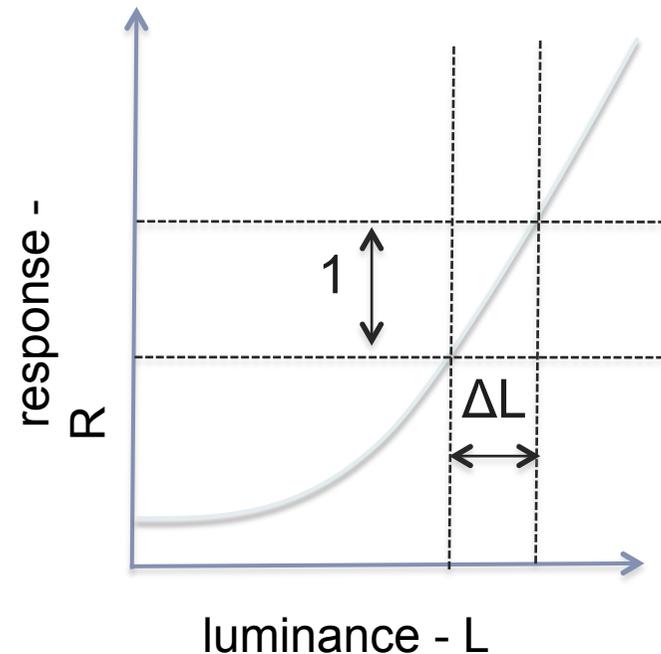
$$\frac{dR}{dl}(L) = \frac{1}{\Delta L(L)}$$

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

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

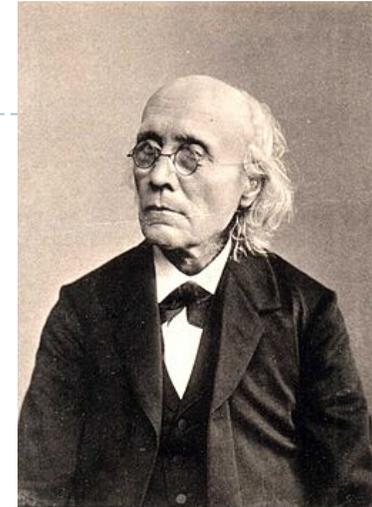
- ▶ the response of the visual system to light is:

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

Fechner law

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

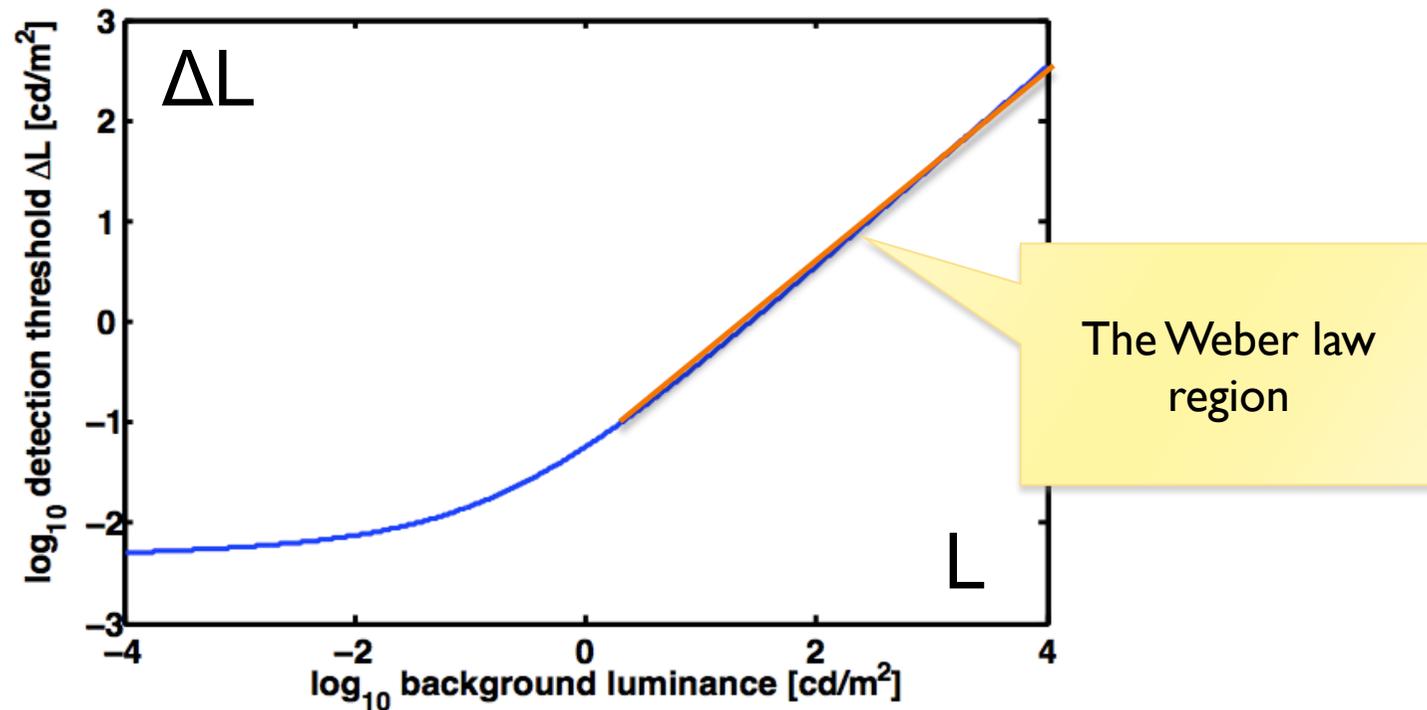
- ▶ Response of the visual system to luminance is **approximately** logarithmic



Gustav Fechner
[From Wikipedia]

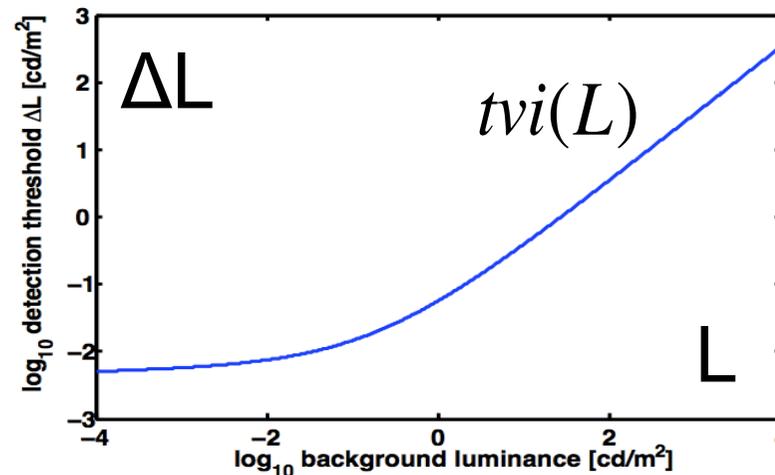
But...the Fechner law does not hold for the full luminance range

- ▶ 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:



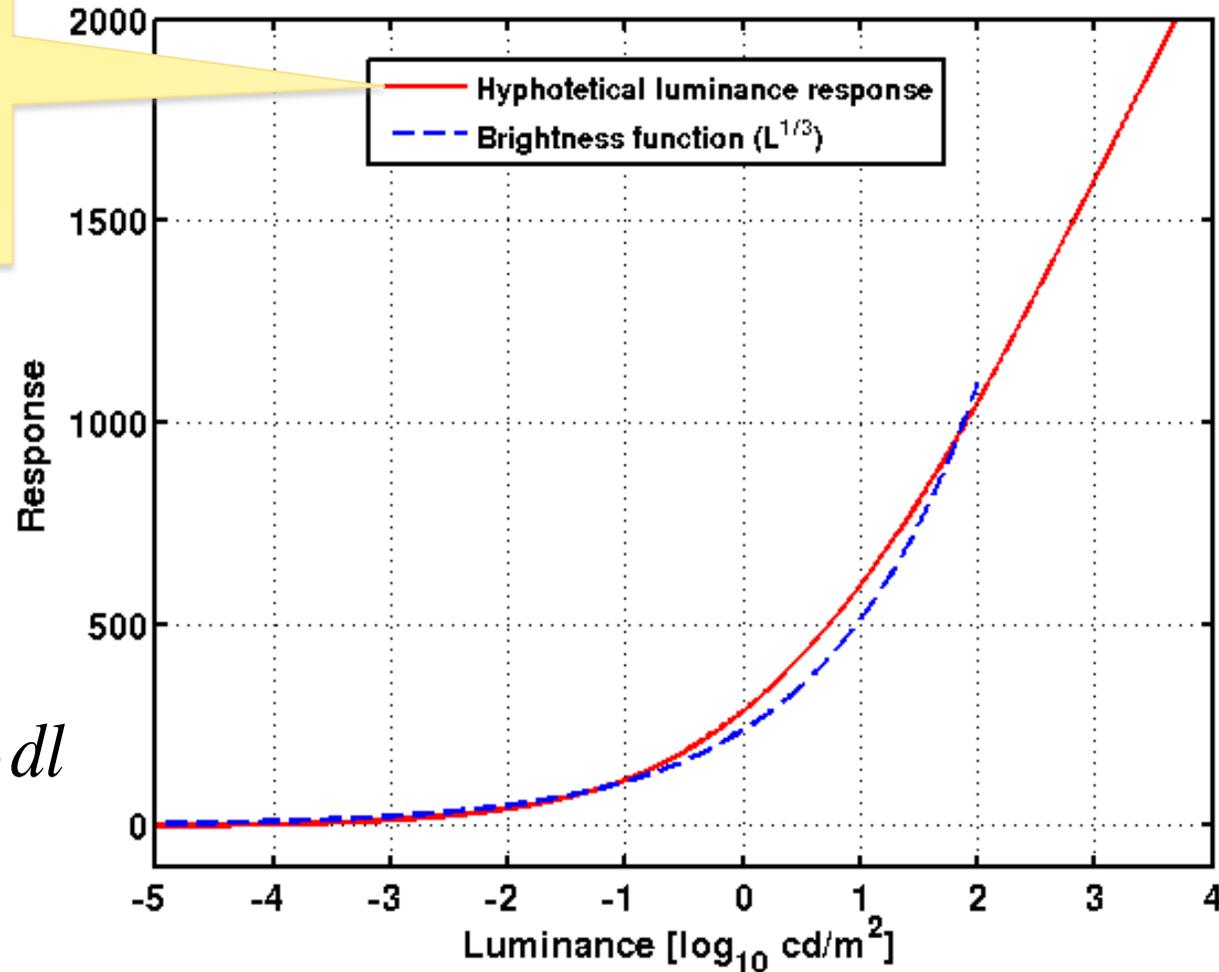
- ▶ we can get more accurate estimate of the “response”:

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

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
- ▶ **Dolby Vision – Perceptual Quantizer**
 - ▶ To encode pixels for high dynamic range images and video

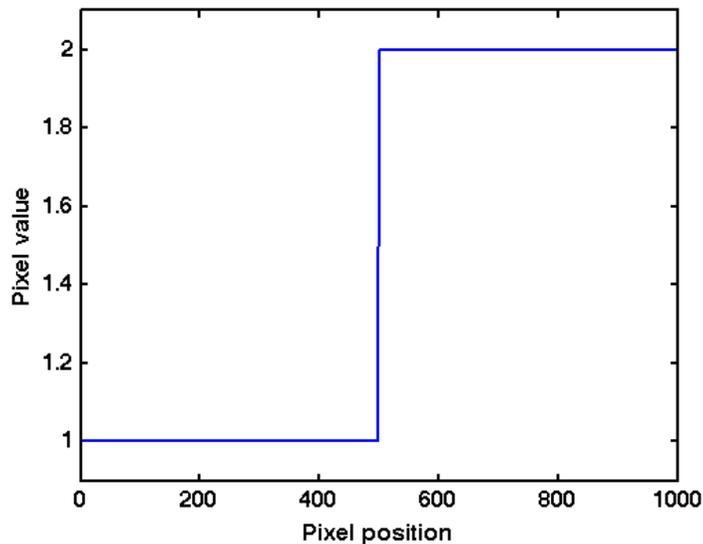


The Future of Vision

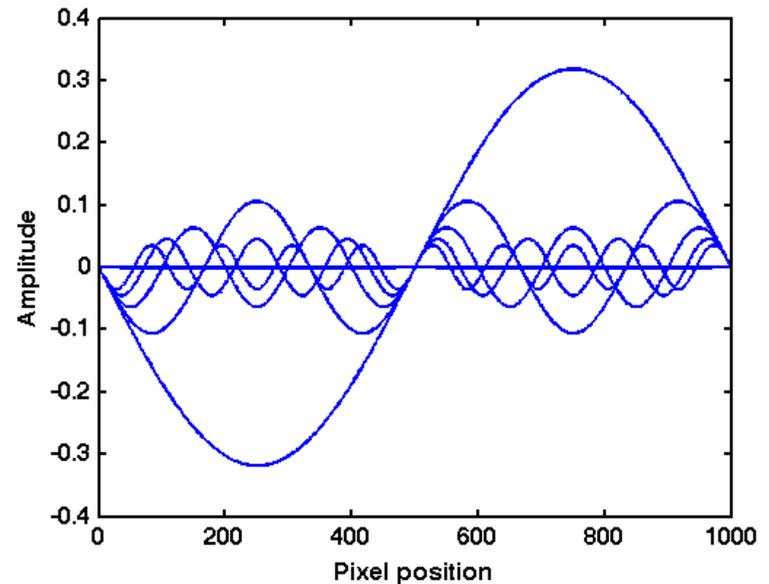
Spatial contrast sensitivity

Fourier analysis

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



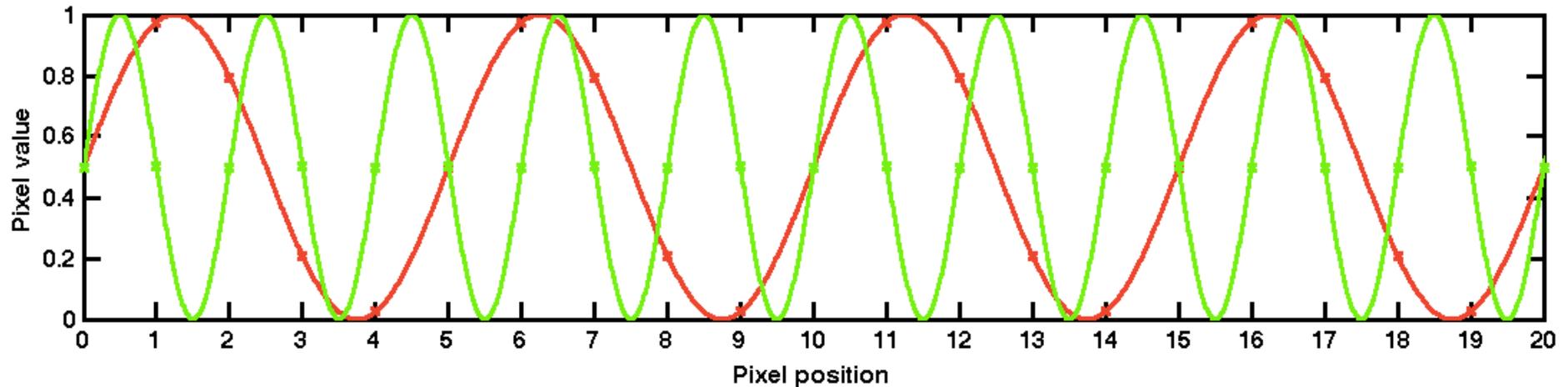
$$= \sum$$



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

Spatial frequency in images

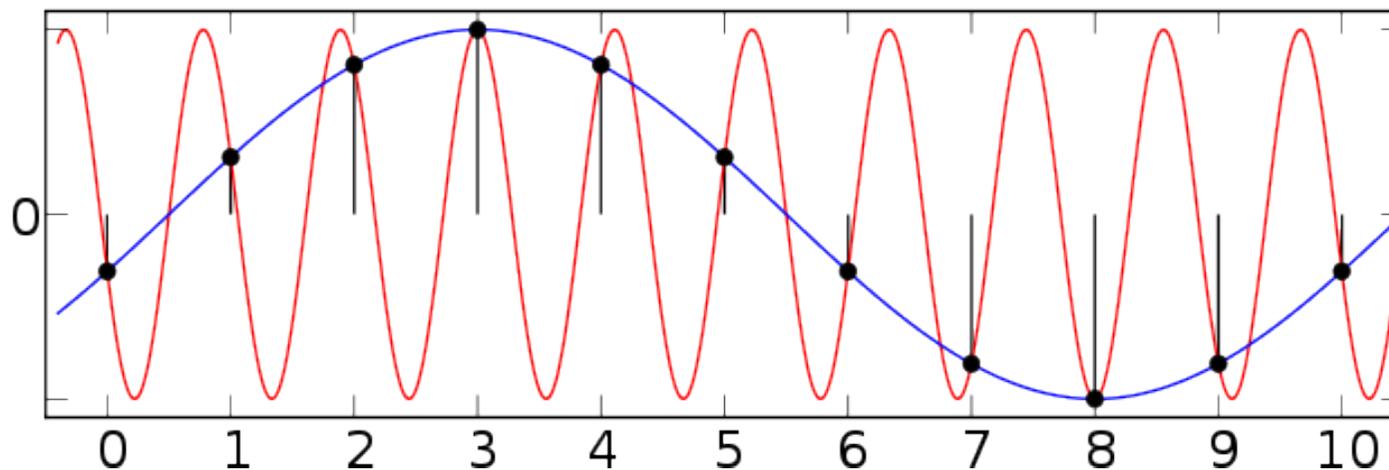
- ▶ Image space units: cycles per sample (or cycles per pixel)



- ▶ 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

- ▶ What is the highest frequency that can be reconstructed for a given sampling density?
 - ▶ Sampling density – how many pixels per image/visual angle/...

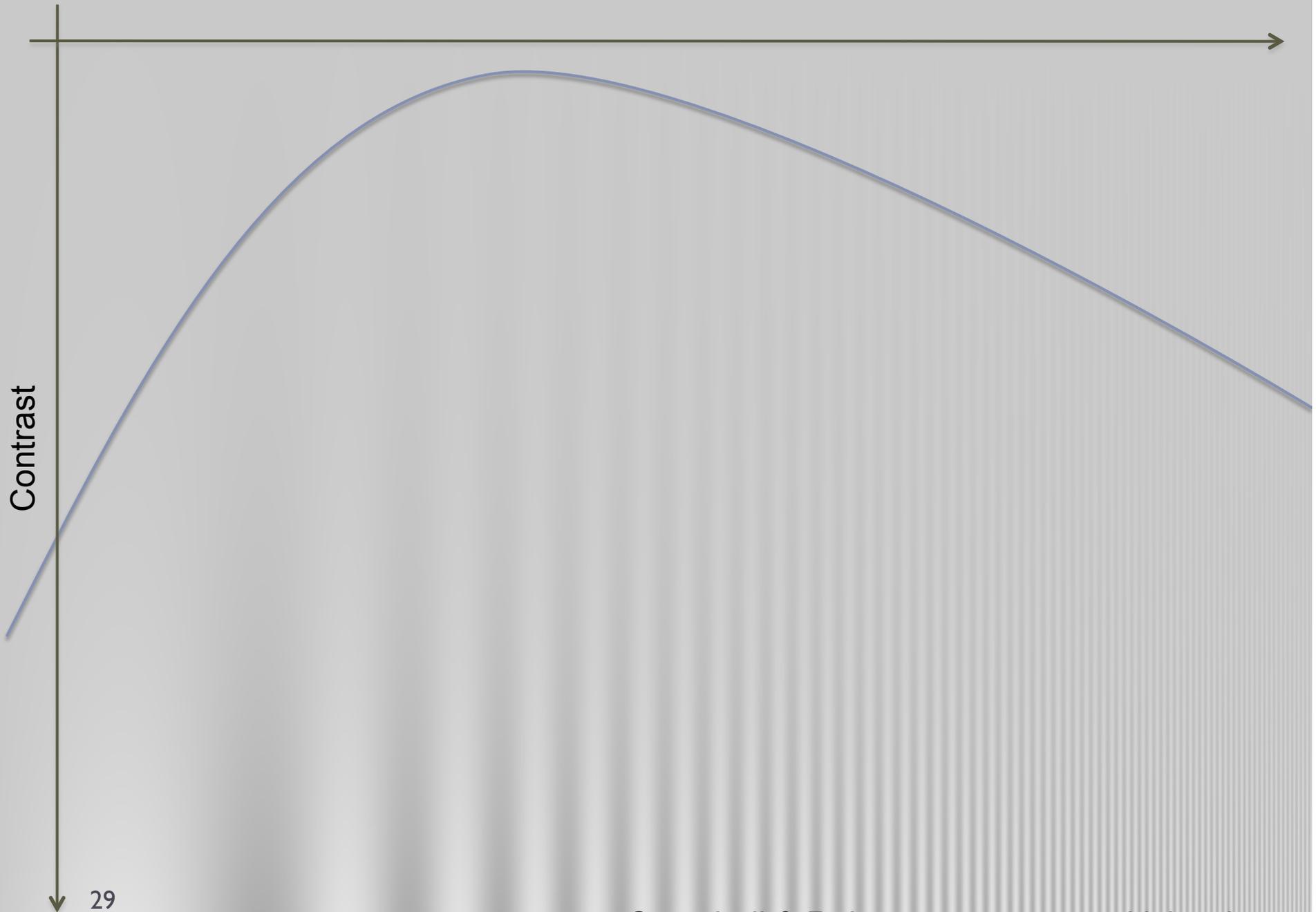


- ▶ Two different sinusoids can be fitted to this set of samples
- ▶ It is possible to fit an infinite number of sinusoids if we allow infinitely high frequency

Nyquist frequency / aliasing

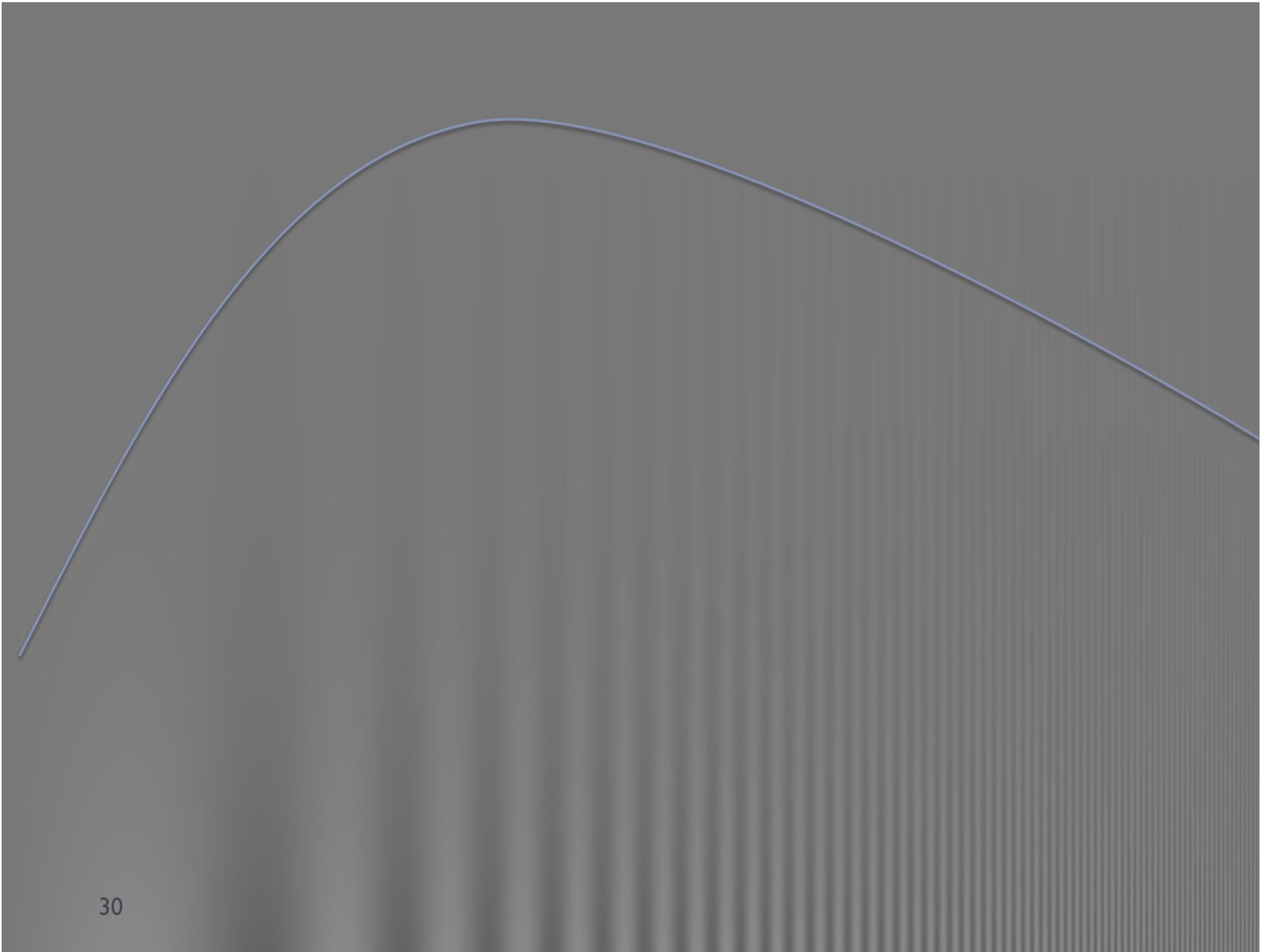
- ▶ Nuquist frequency is the highest frequency that can be represented by a discrete set of uniform samples (pixels)
- ▶ Nuquist frequency = 0.5 sampling rate
 - ▶ For audio
 - ▶ If the sampling rate is 44100 samples per second (audio CD), then the Nyquist frequency is 22050 Hz
 - ▶ For images (visual 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

Spatial frequency [cycles per degree]



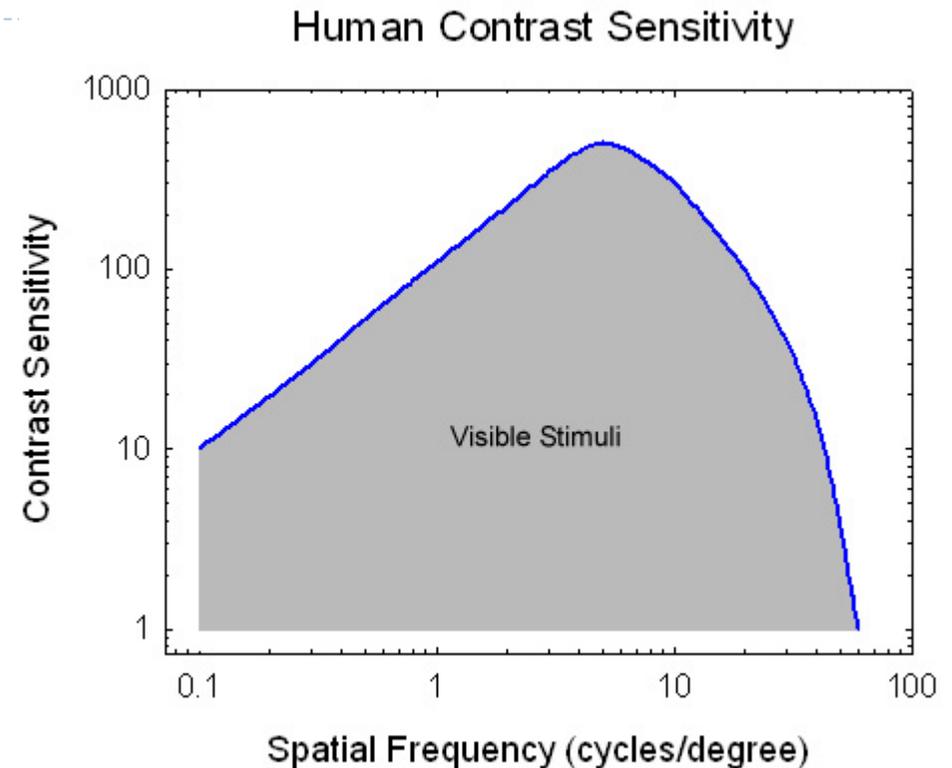
29

Campbell & Robson contrast sensitivity chart



Contrast Sensitivity

- ▶ Sensitivity:
I / threshold contrast
- ▶ Contrast = $\Delta L/L$
- ▶ Maximum acuity 2-5 cycles/degree (0.2 %)
 - ▶ Decrease toward low frequencies: lateral inhibition
 - ▶ Decrease toward high frequencies
 - ▶ Upper limit: 60-70 cycles/degree



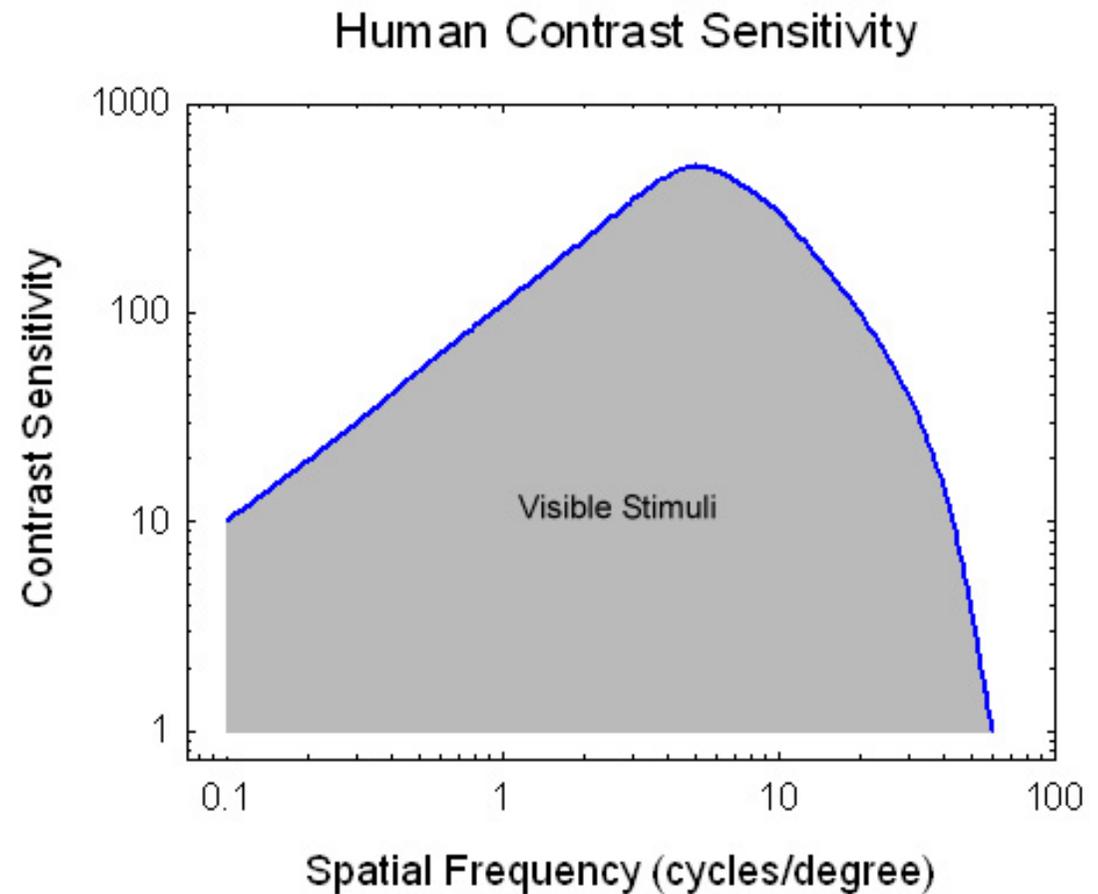
Rationale: if we were sensitive to low frequencies, the vision would be affected by changes of illumination. There are physical limitations to the perception of high frequencies.

Implications of CSF

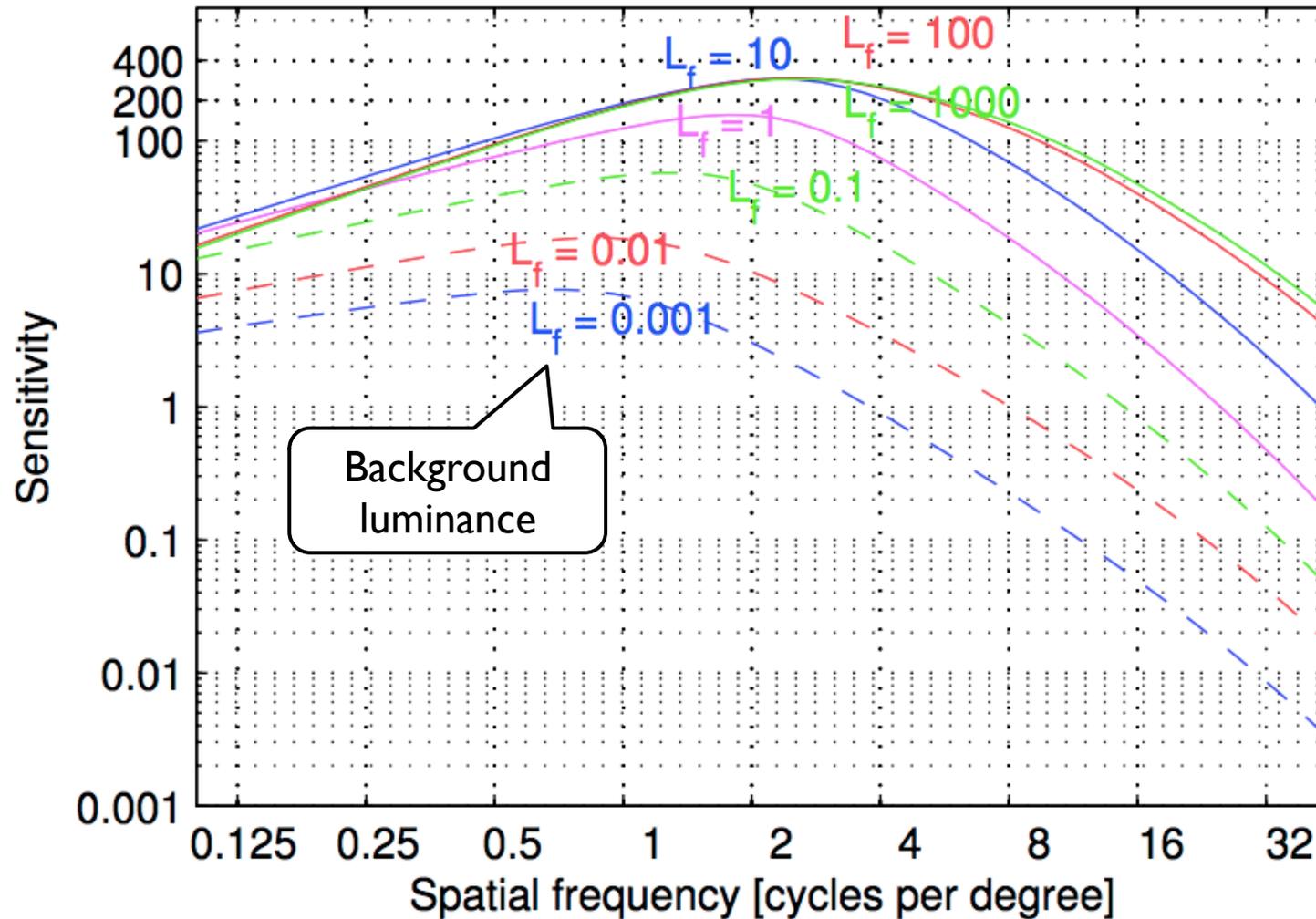
- ▶ As objects get further away, they get smaller; spatial frequencies get higher
 - ▶ At some point we cannot see the details
 - ▶ That is the upper limit of the CSF (60-70 cpd)
- ▶ When we get too close to low frequency patterns, they seem to be constant
 - ▶ The background of this slide contains a smooth gradient
 - ▶ It is well visible when you look at it from a normal viewing distance
 - ▶ Now enlarge the slide to full screen and move your head very close
 - ▶ The gradient should disappear

Explaining the effects with CSF

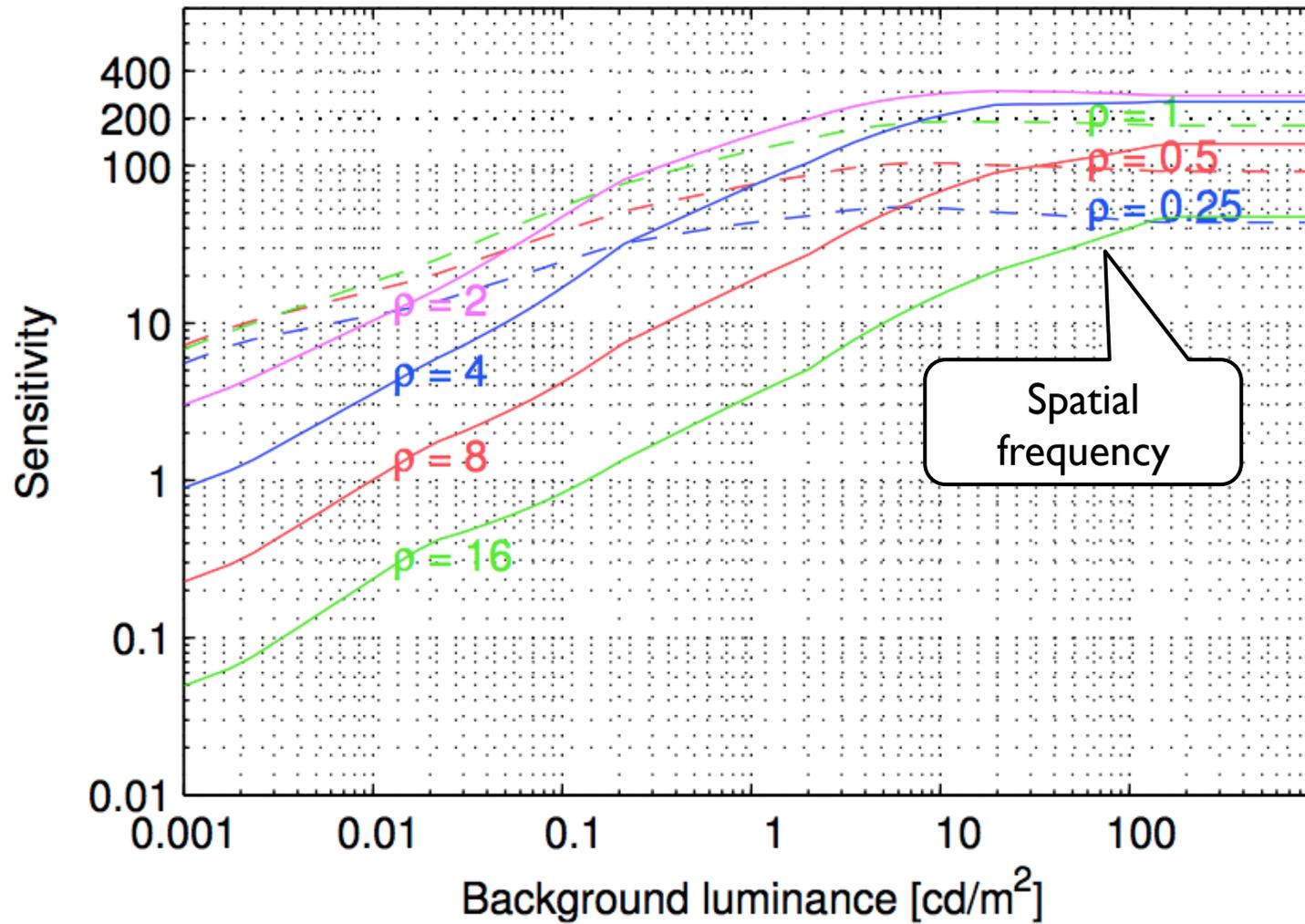
- ▶ Can you explain the effects described on the previous slide using the CSF plot?



CSF as a function of spatial frequency



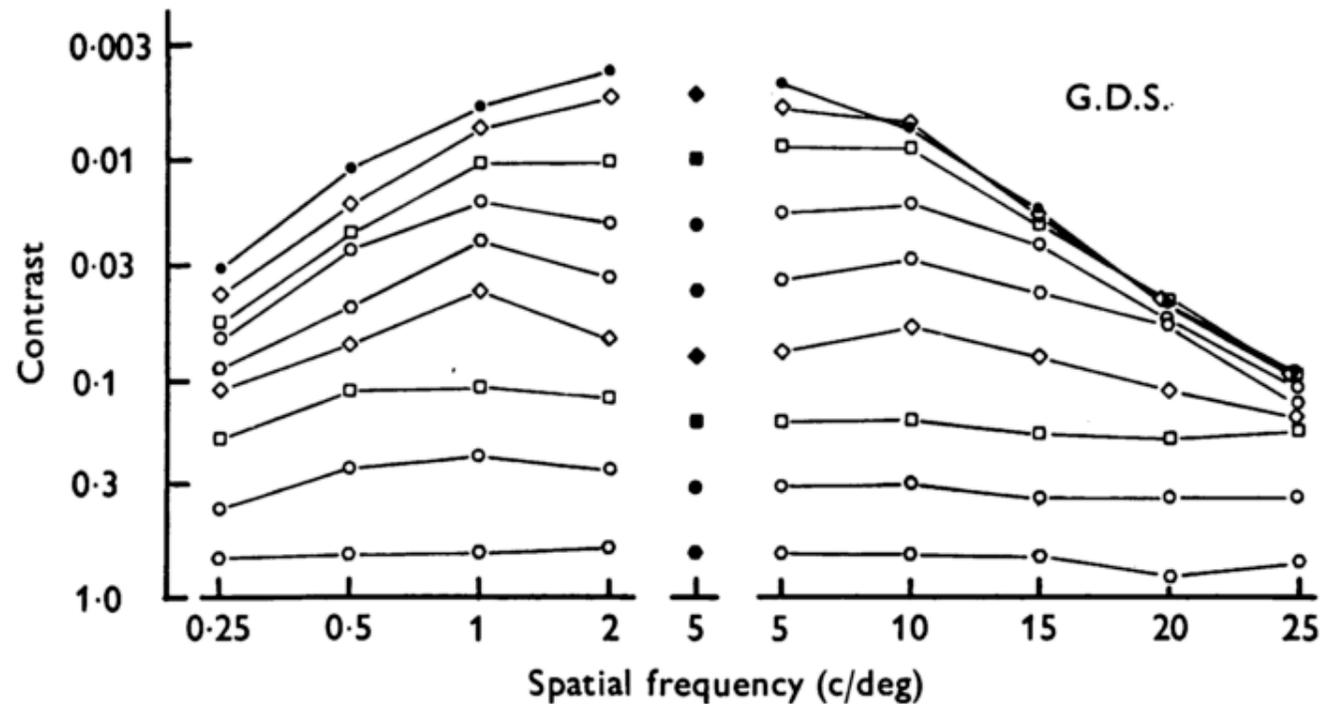
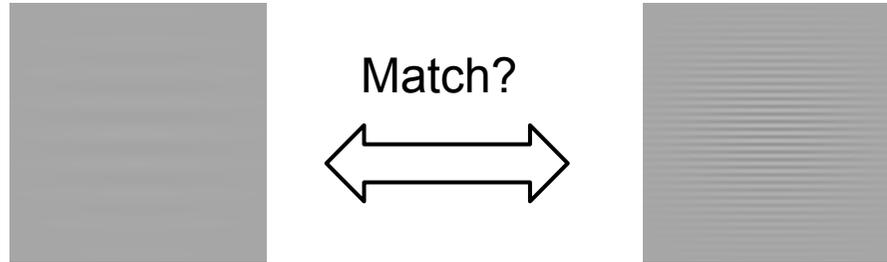
CSF as a function of background luminance



Contrast constancy

Contrast constancy

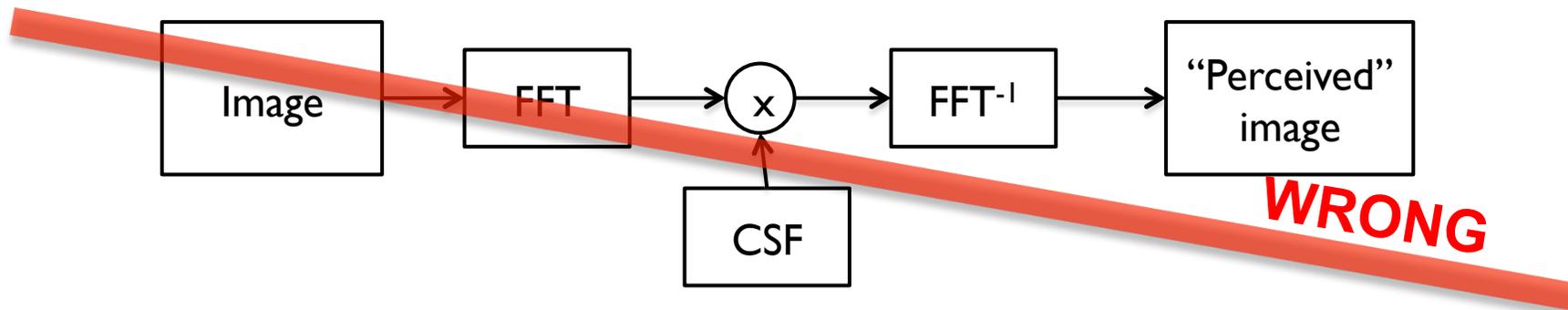
Experiment: Adjust the amplitude of one sinusoidal grating until it matches the perceived magnitude of another sinusoidal grating.



Contrast constancy
No CSF above the detection threshold

Contrast constancy

- ▶ For high (supra-threshold) contrast, the perceived magnitude of contrast does not change with spatial frequency
- ▶ The CSF gets “flat” above the detection threshold
- ▶ A common mistake - use CSF as a linear filter:

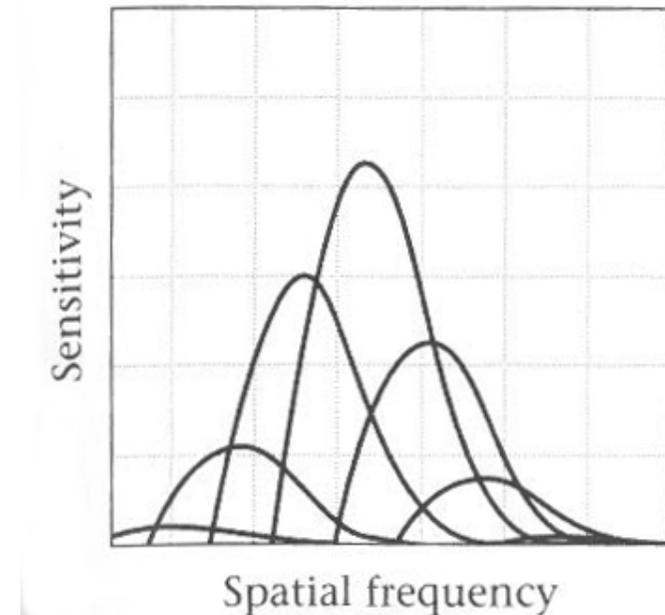


- ▶ why CSF cannot be used as a linear filter?

Multi-resolution models

Spatial-frequency selective channels

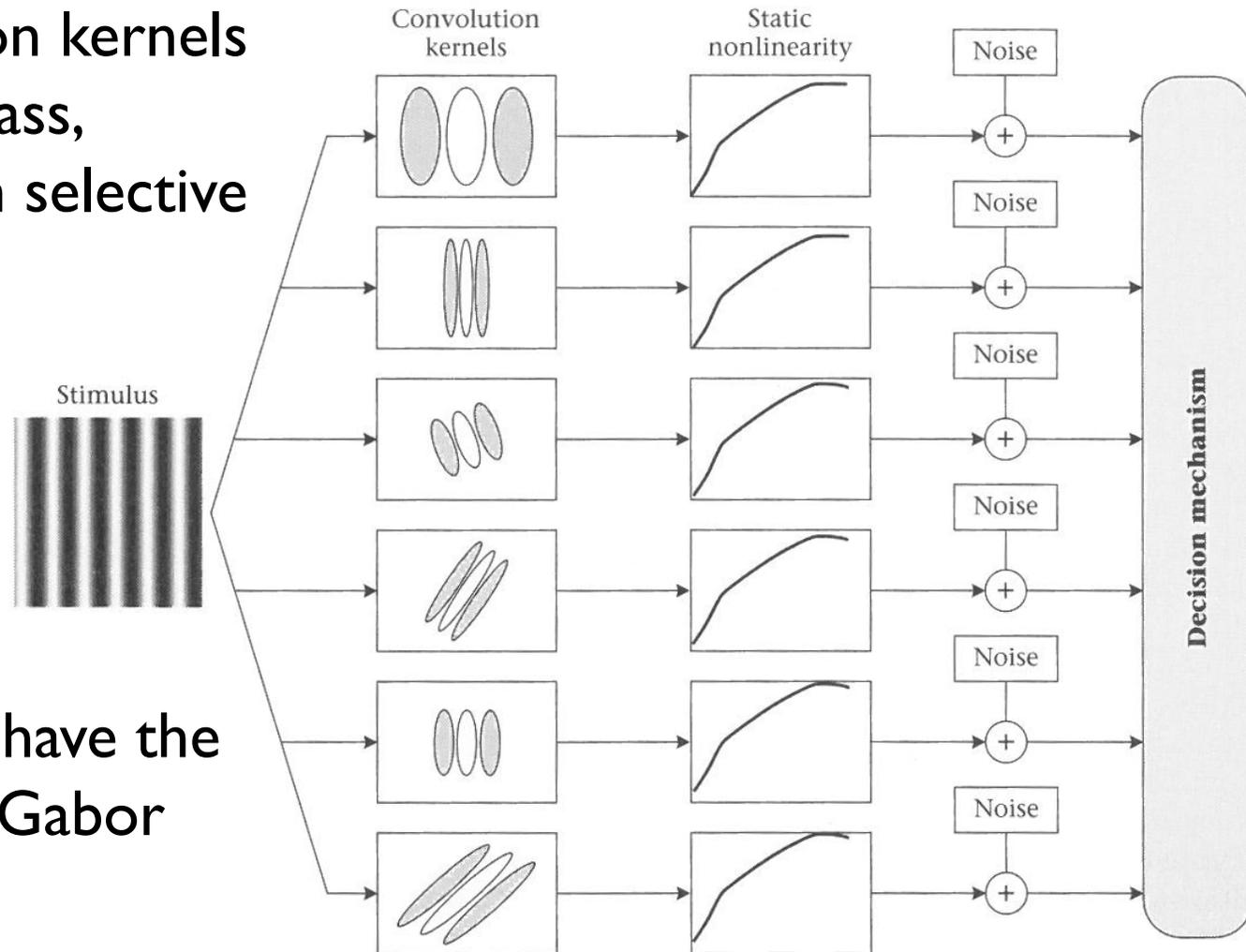
- ▶ The visual information is passed to the visual cortex in multiple channels
 - ▶ The channels are selective to spatial frequency, temporal frequency and orientation
 - ▶ Each channel is affected by different „noise” level
 - ▶ The CSF is the net result of information being passed in noise-affected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



From: Wandell, 1995

Multi-resolution visual model

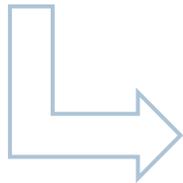
- ▶ Convolution kernels are band-pass, orientation selective filters



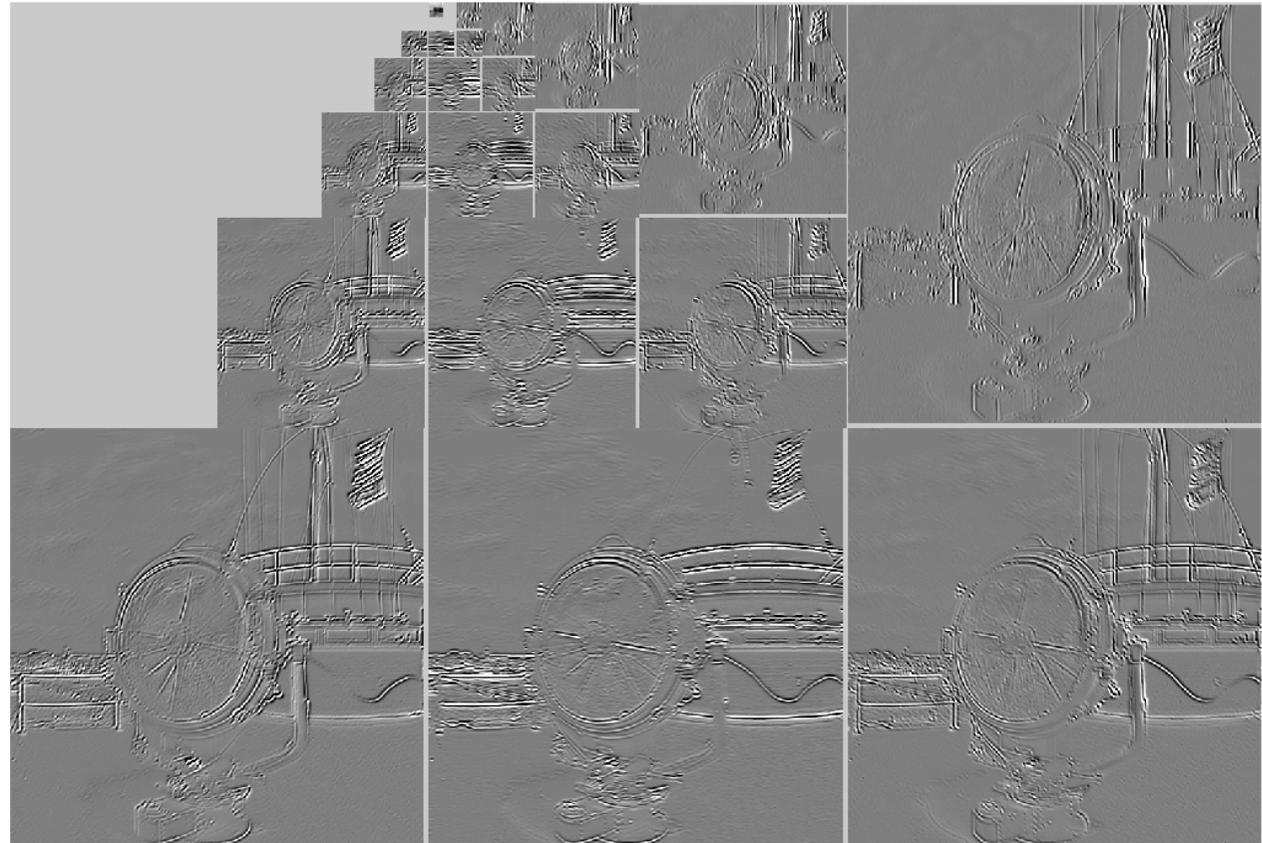
- ▶ The filters have the shape of a Gabor function

From: Wandell, 1995

Multi-scale decomposition



Steerable pyramid
decomposition



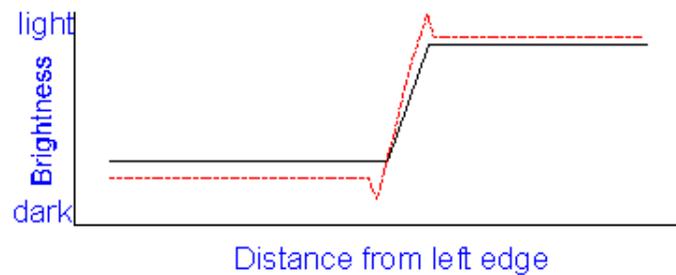
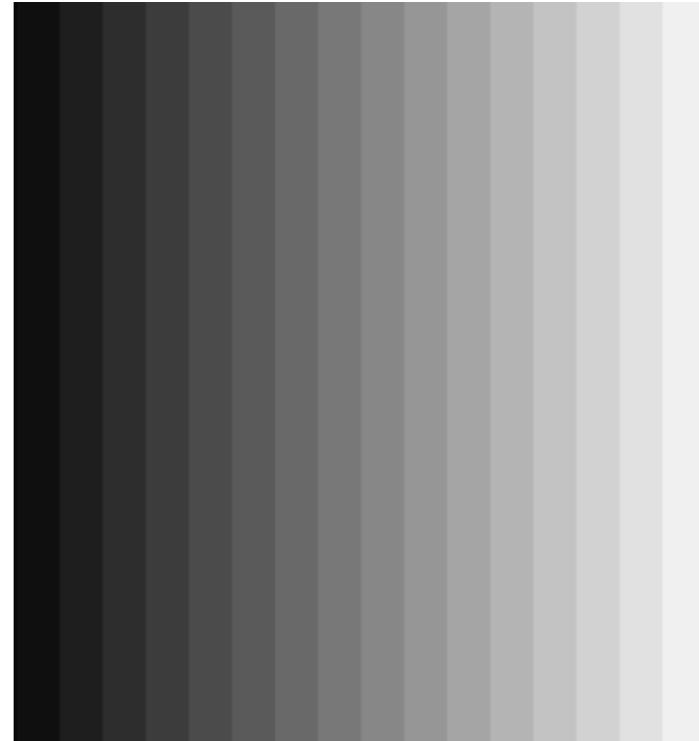
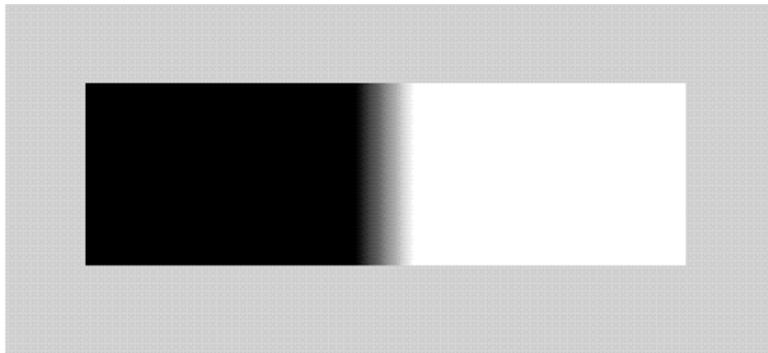
Applications of multi-scale models

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



Mach Bands – evidence for band-pass visual processing

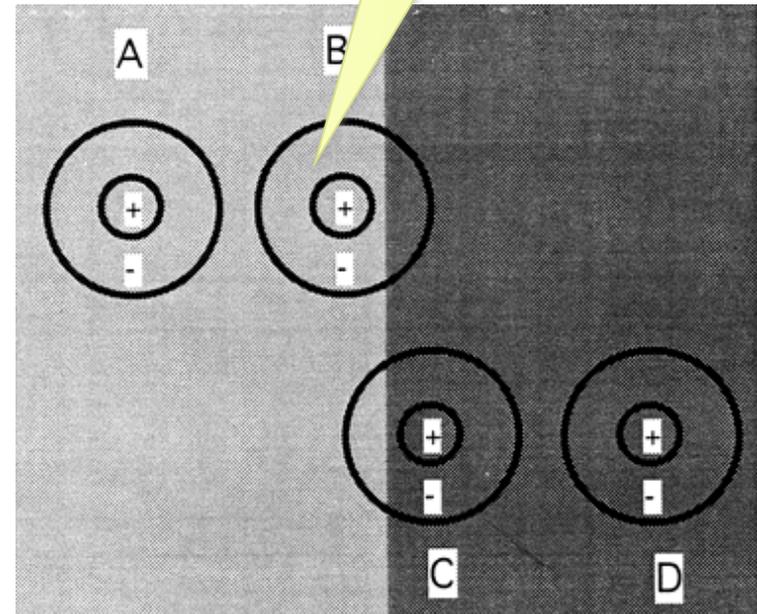
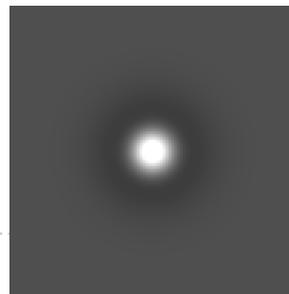
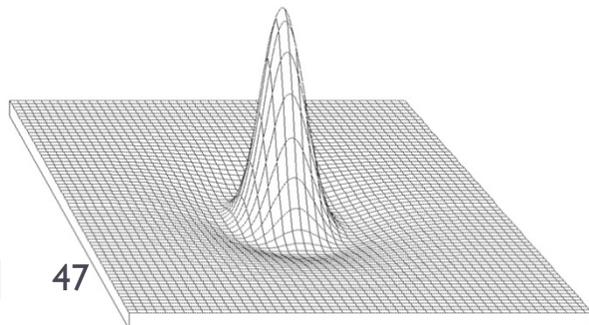
- “Overshooting” along edges
 - Extra-bright rims on bright sides
 - Extra-dark rims on dark sides
- Due to “Lateral Inhibition”



Centre-surround (Lateral Inhibition)

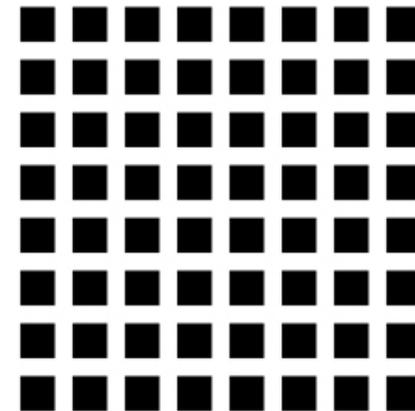
- ▶ “Pre-processing” step within the retina
 - ▶ Surrounding brightness level weighted negatively
 - ▶ A: high stimulus, maximal bright inhibition
 - ▶ B: high stimulus, reduced inhibition & stronger response
 - ▶ D: low stimulus, maximal inhibition
 - ▶ C: low stimulus, increased inhibition & weaker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

Center-surround receptive fields (groups of photoreceptors)



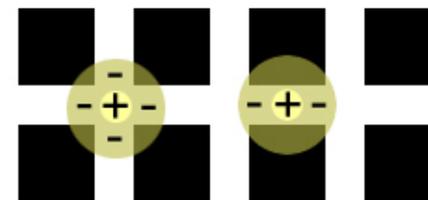
Centre-surround: Hermann Grid

- 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 ?

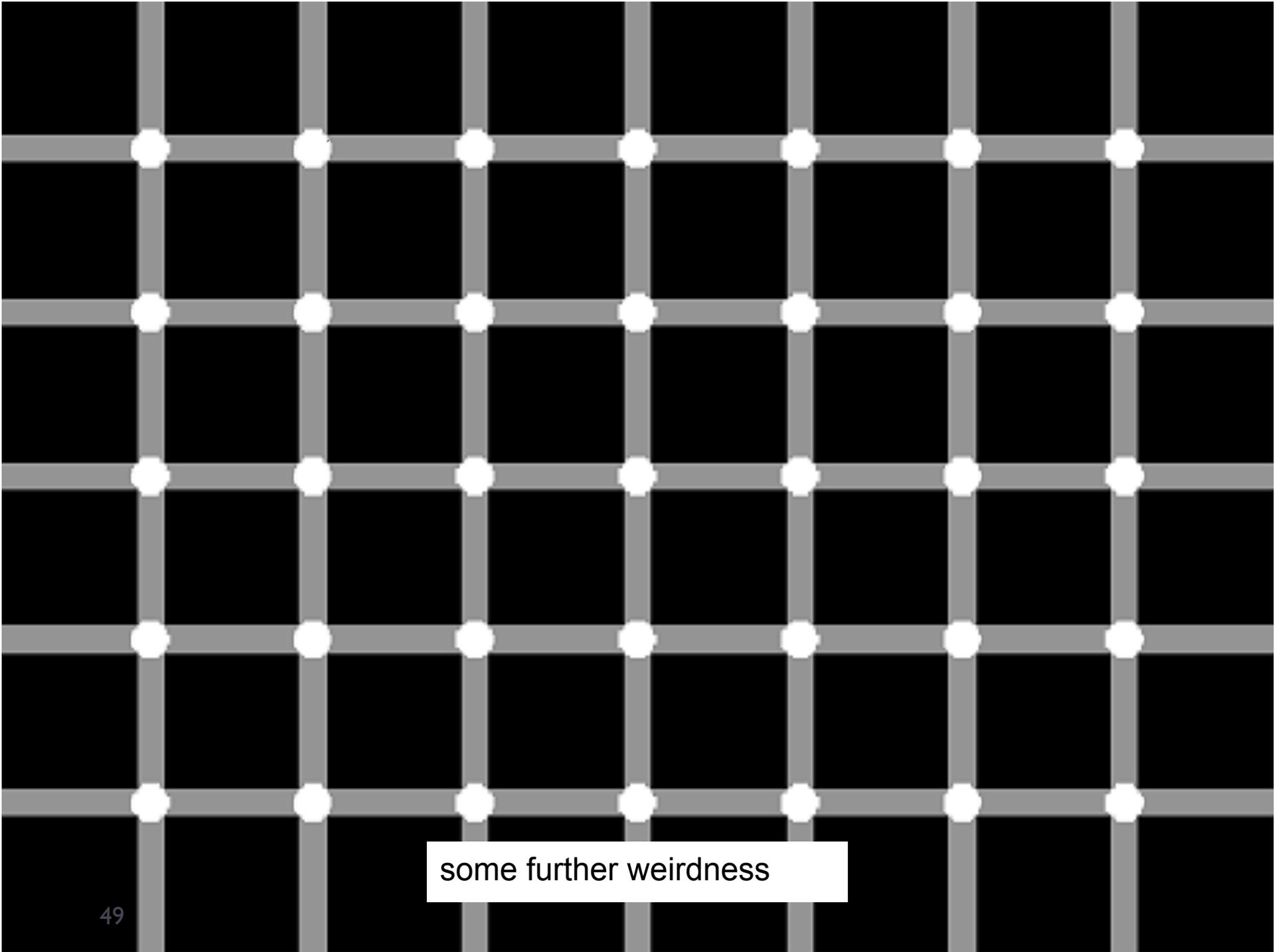


A

B



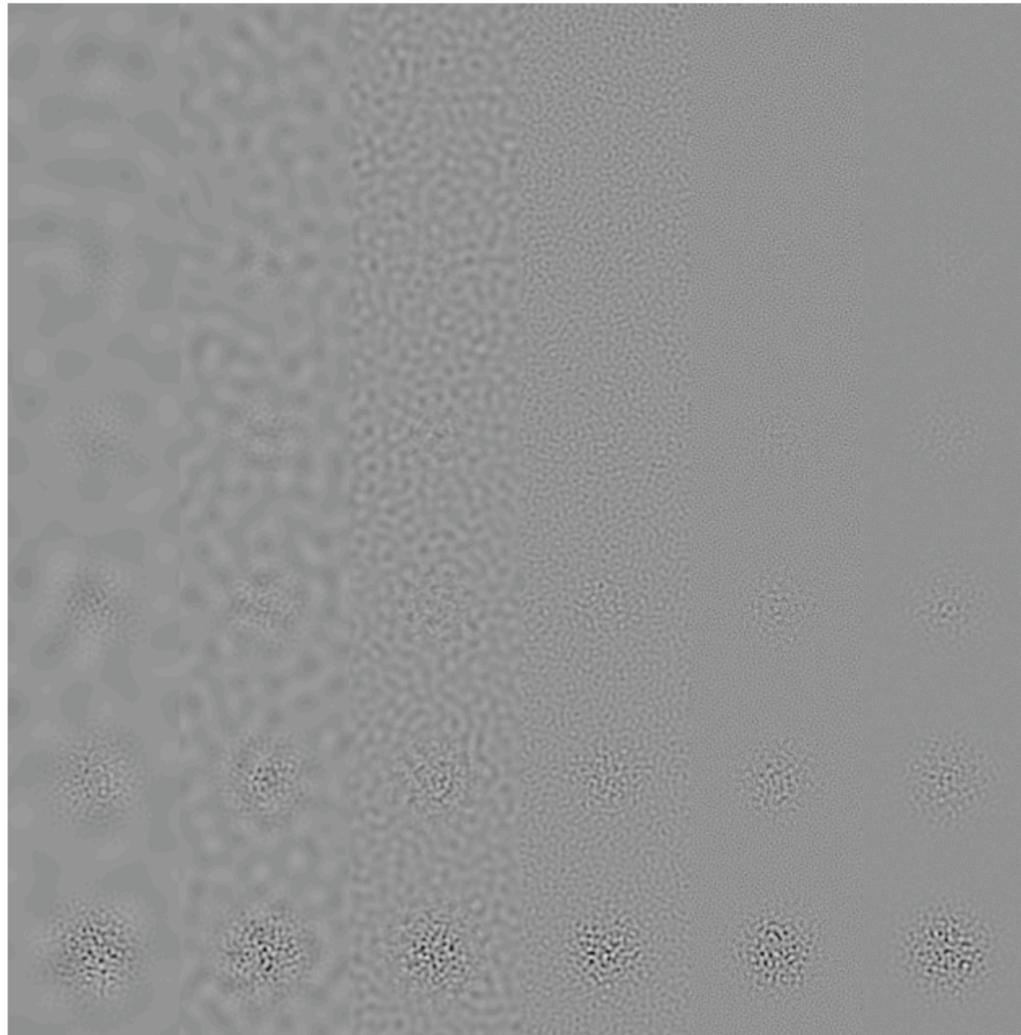
Simulation



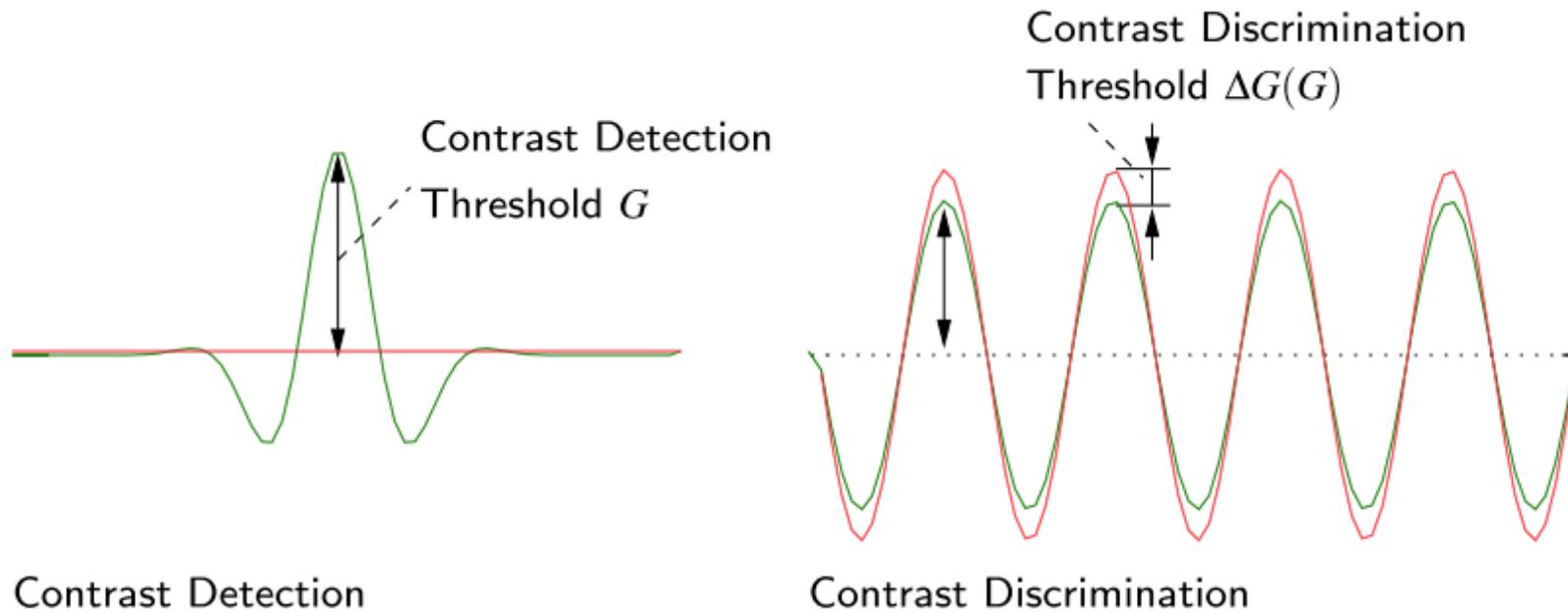
some further weirdness

Contrast masking

In which vertical bar is the pattern the most and the least visible?



Detection vs. discrimination



Definitions of contrast

Simple Contrast

$$C_s = \frac{L_{max}}{L_{min}}$$

Weber Fraction

$$W = \frac{\Delta L}{L_{min}}$$

Logarithmic Ratio

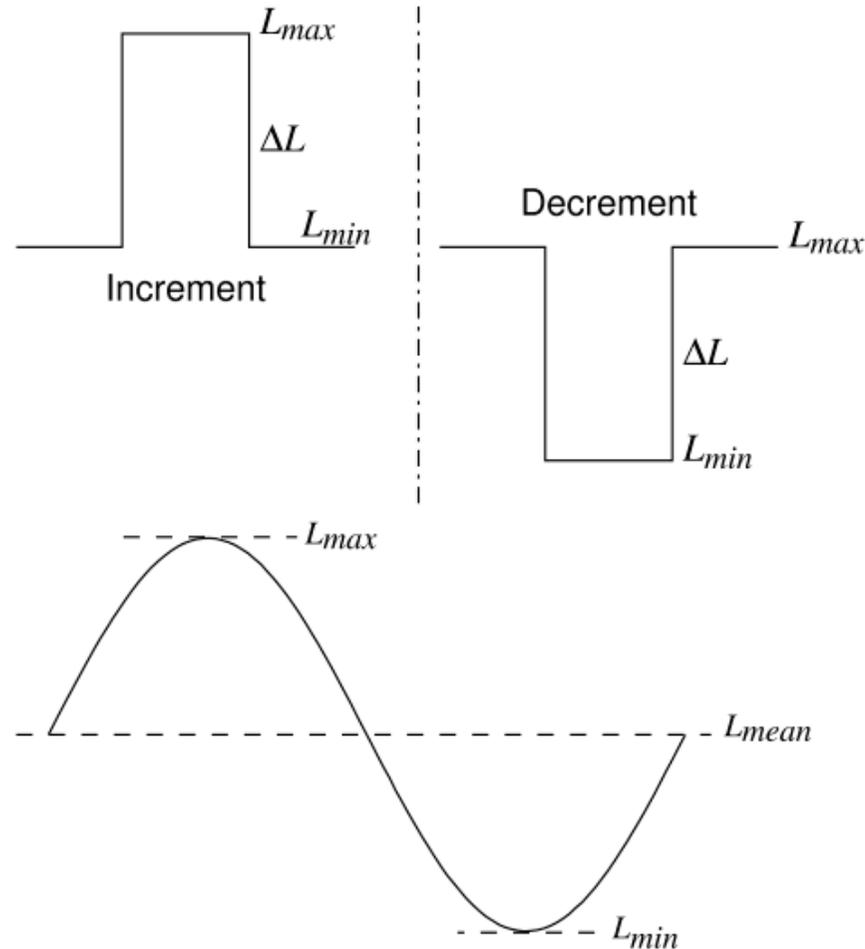
$$G = \log_{10}\left(\frac{L_{max}}{L_{min}}\right)$$

Michelson Contrast

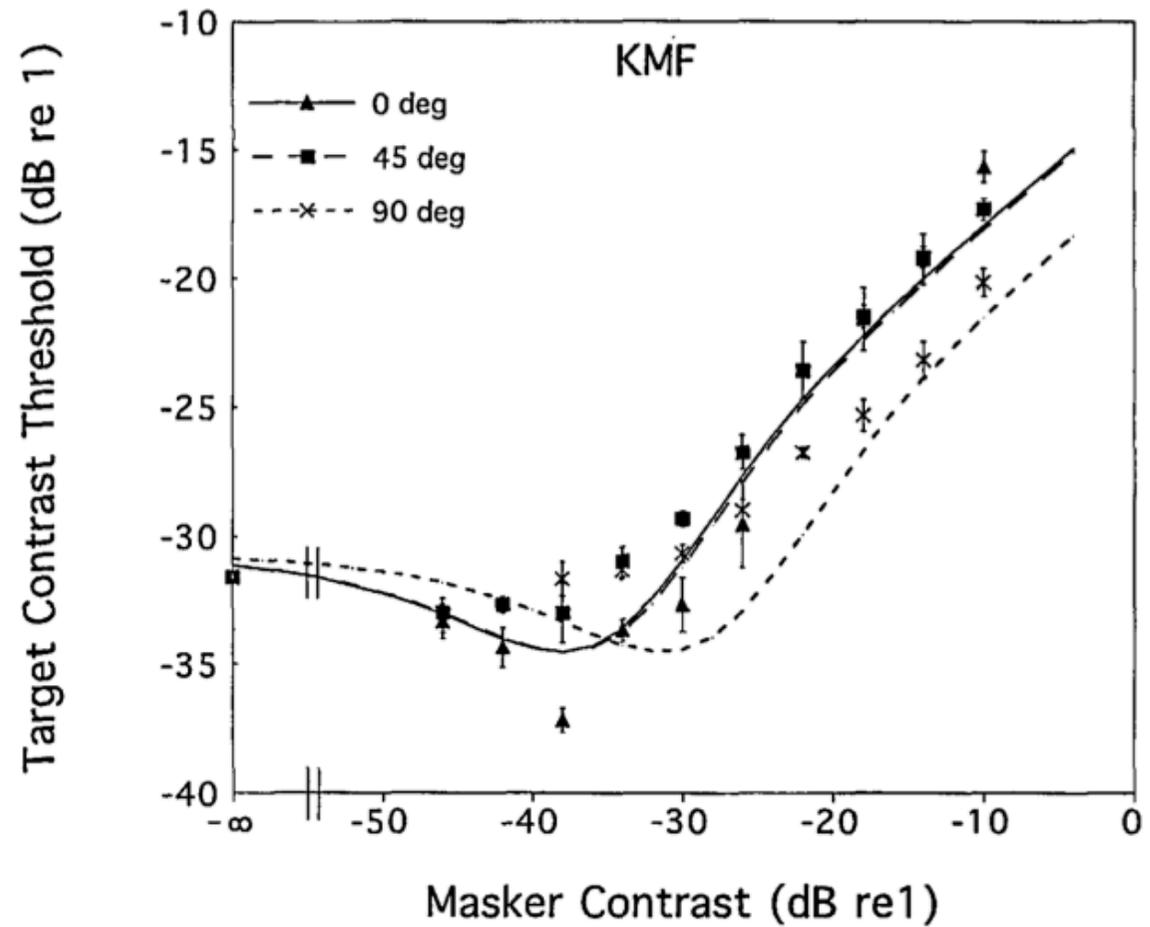
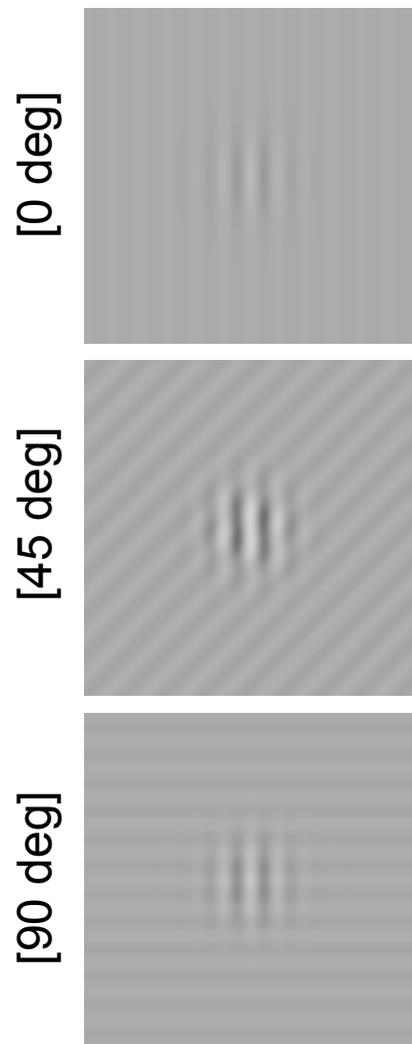
$$M = \frac{|L_{max} - L_{min}|}{L_{max} + L_{min}}$$

Signal to Noise Ratio

$$SNR = 20 \cdot \log_{10}\left(\frac{L_{max}}{L_{min}}\right)$$



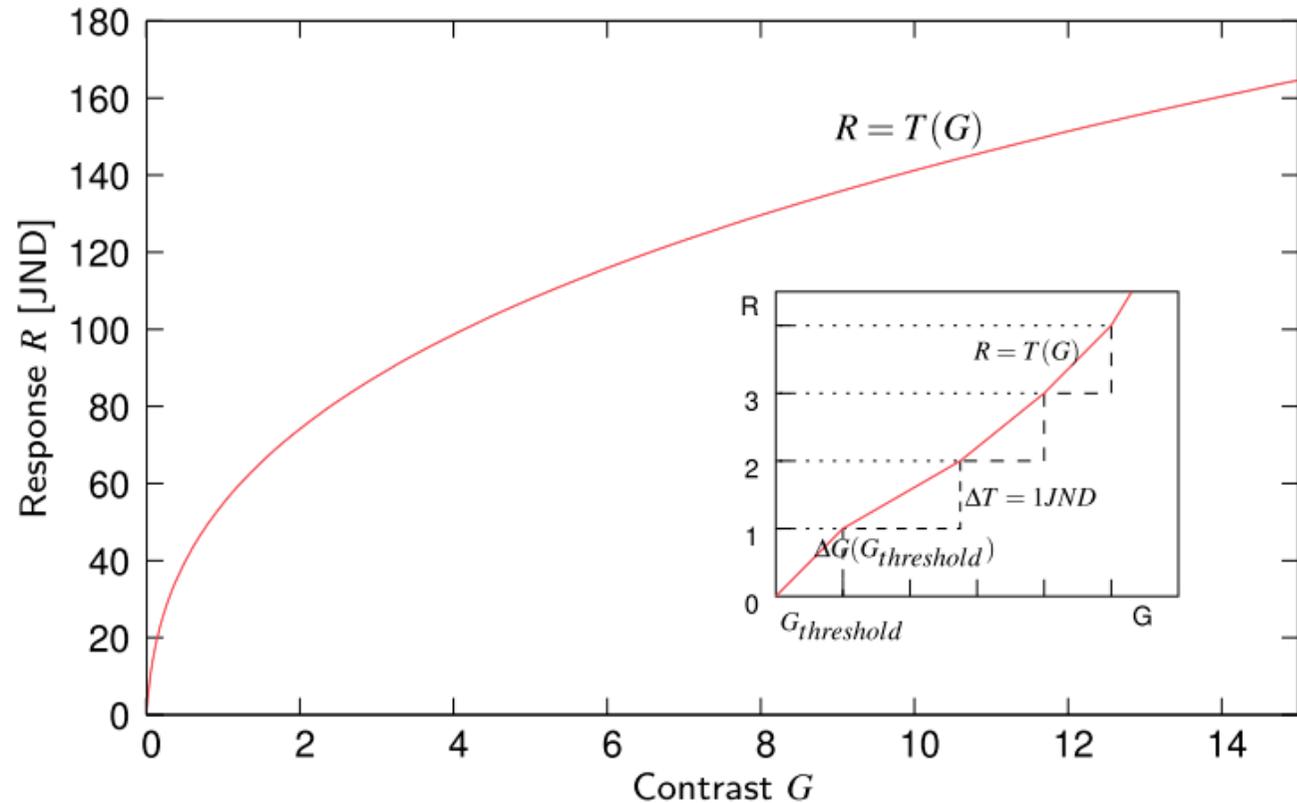
Visual masking experiments - discrimination



[Foley, JOSAA, 1994]

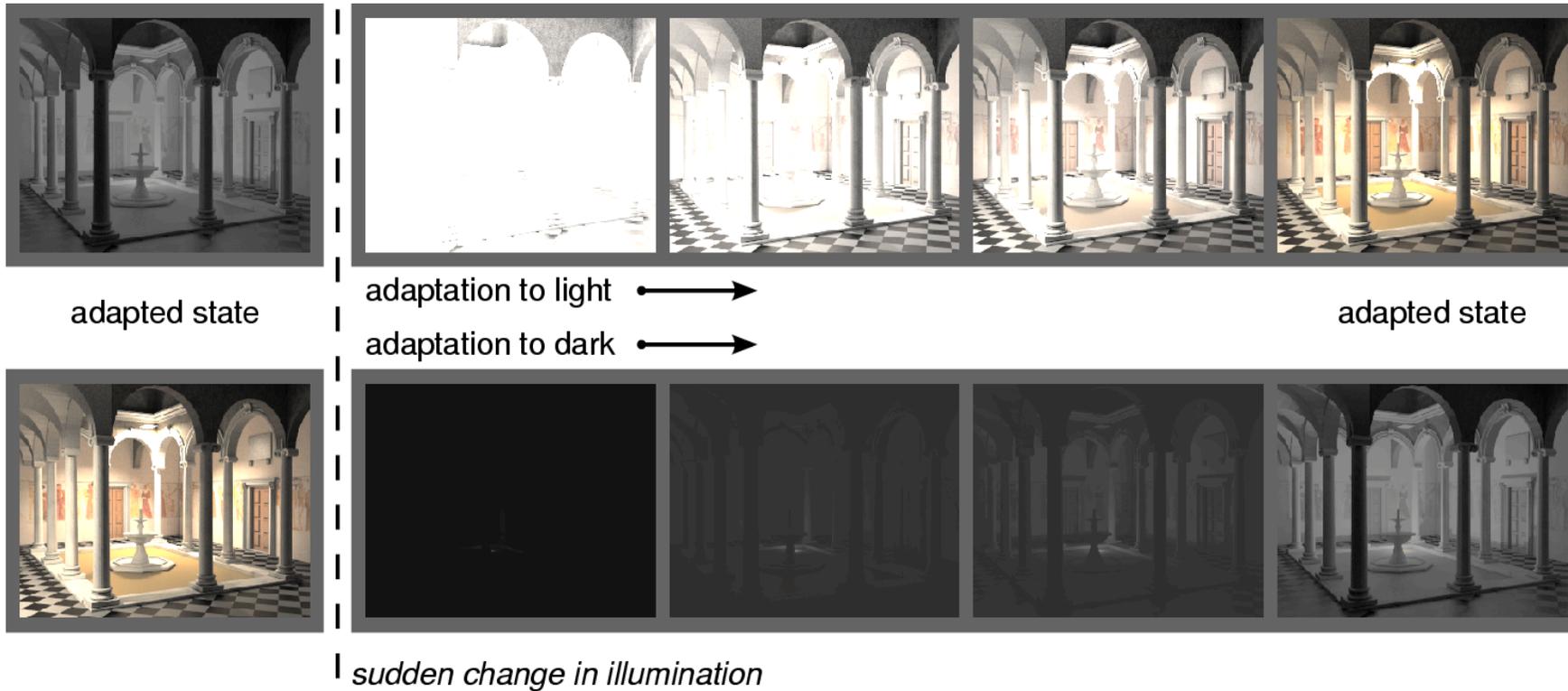
Contrast transducer

- ▶ Similar to JND encoding
- ▶ Trans. from physical contrast into “perceived” contrast



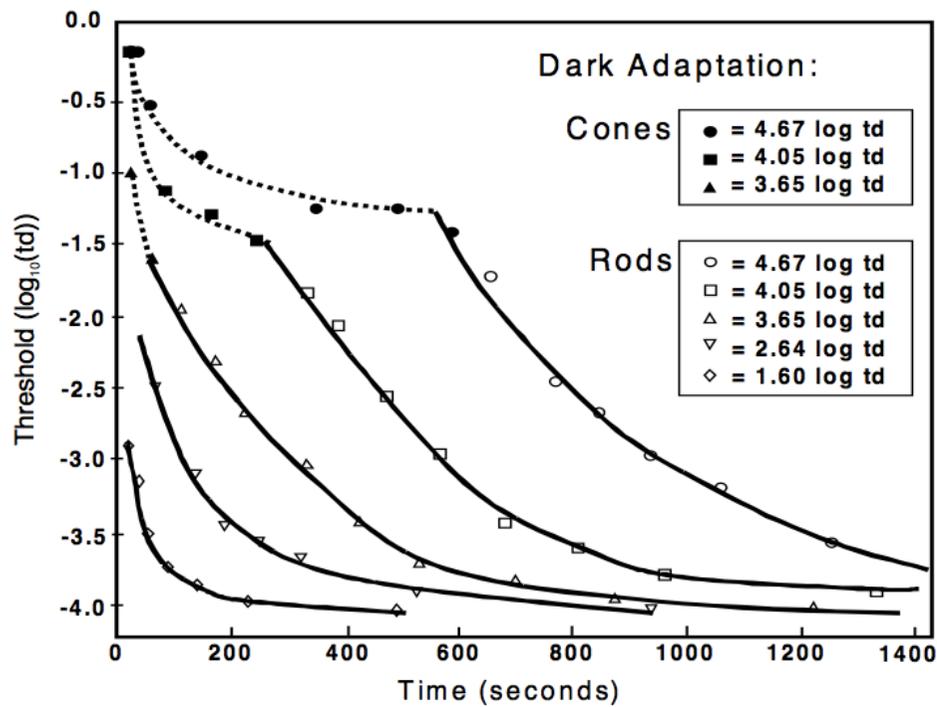
Light and dark adaptation

Light and dark adaptation

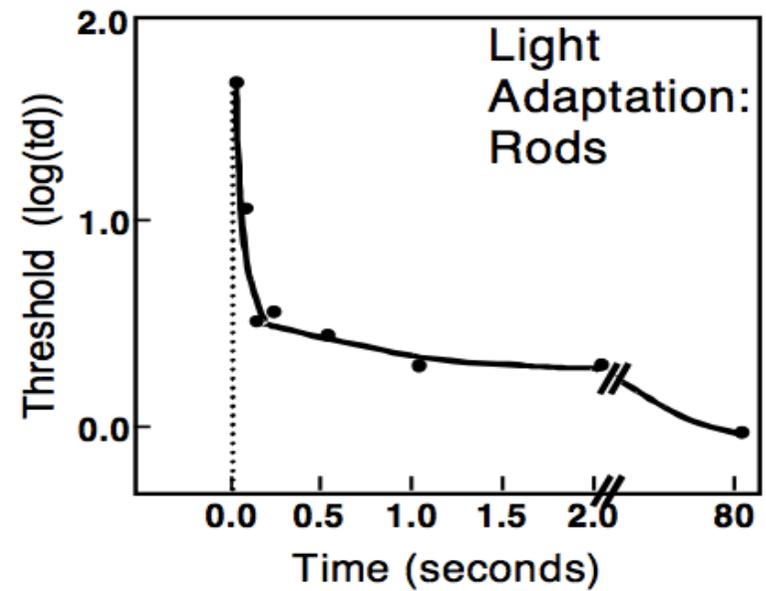
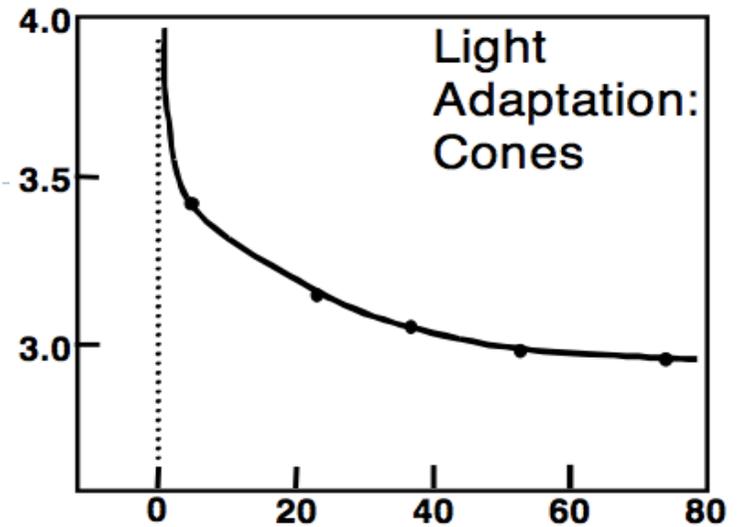


- ▶ Light adaptation: from dark to bright
- ▶ Dark adaptation: from bright to dark (much slower)

Time-course of adaptation



Bright -> Dark



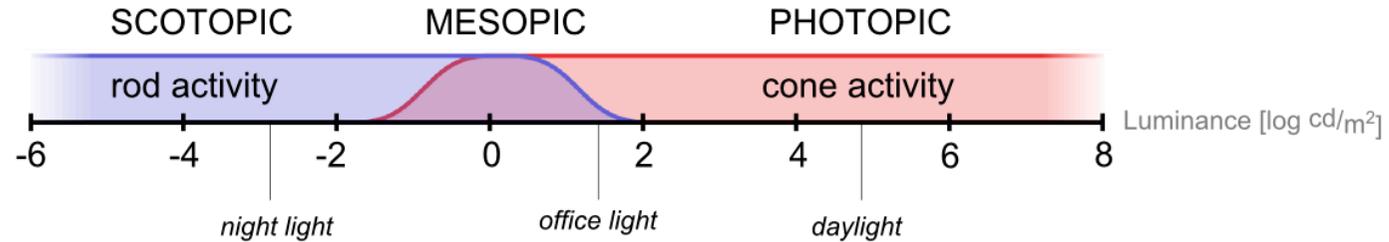
Dark -> Bright

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)

Night and daylight vision

Vision mode:



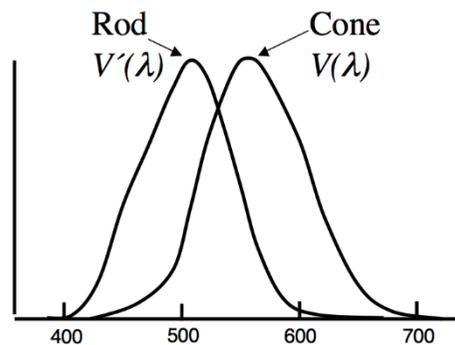
Mode properties:

monochromatic vision
limited visual acuity

good color perception
good visual acuity



Luminous efficiency

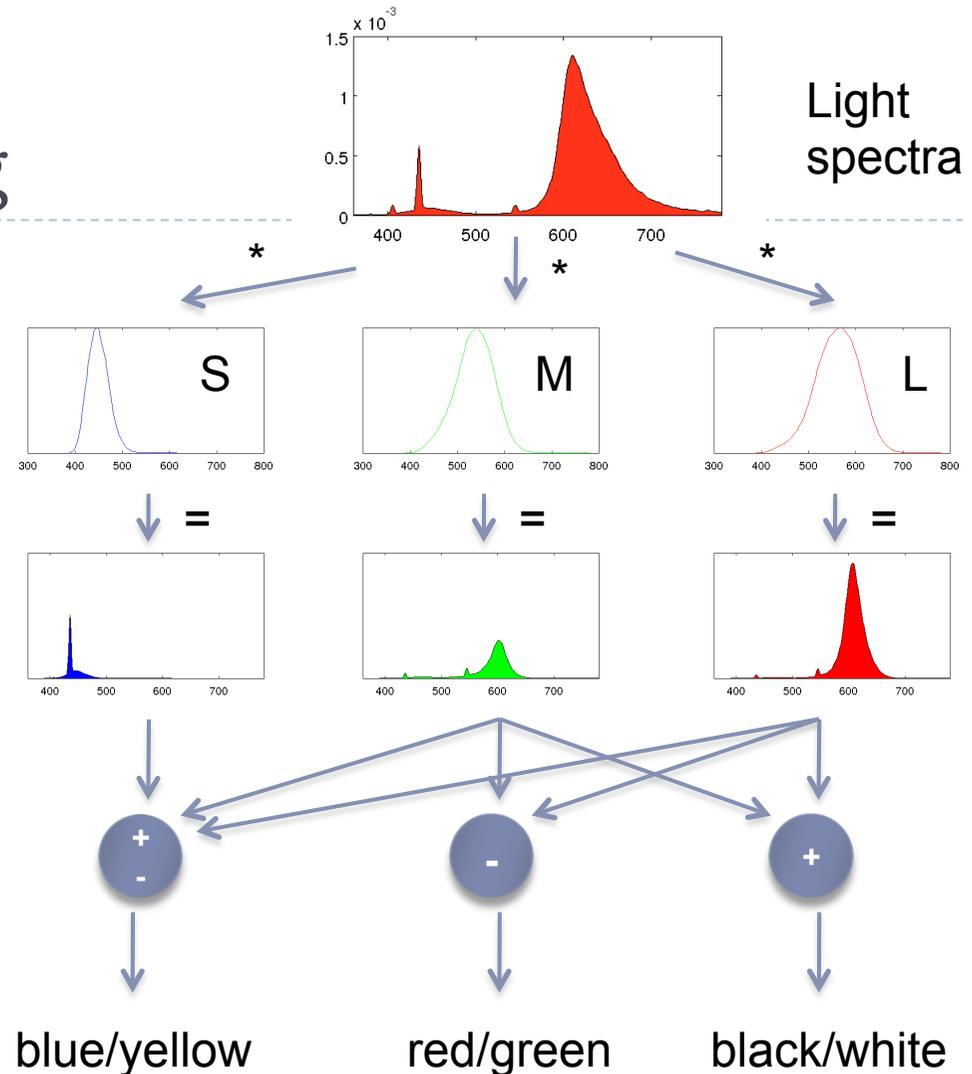




Opponent colours and spatial colour vision

Colour processing

- ▶ Light is sensed by L, M and S, cones
 - ▶ Each cone type is sensitive to different wavelengths
- ▶ Responses from L, M and S cones are combined into three opponent pathways
 - ▶ achromatic (black/white) pathway – luminance
 - ▶ 2 colour opponent pathways
- ▶ Rationale: improve coding efficiency for natural scenes

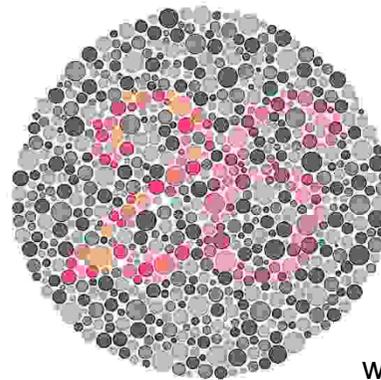
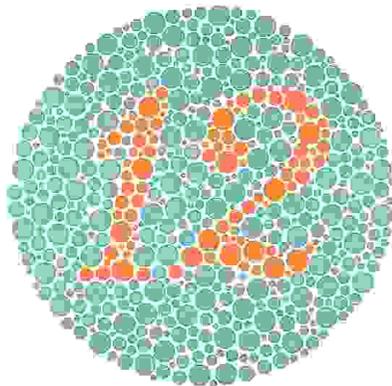


Colour perception

- ▶ **Di-chromaticity (dogs, cats)**
 - ▶ Yellow & blue-violet
 - ▶ Green, orange, red indistinguishable
- ▶ **Tri-chromaticity (humans, monkeys)**
 - ▶ Red, green, blue
 - ▶ Colour-blindness
 - ▶ Most often men, green-red colour-blindness

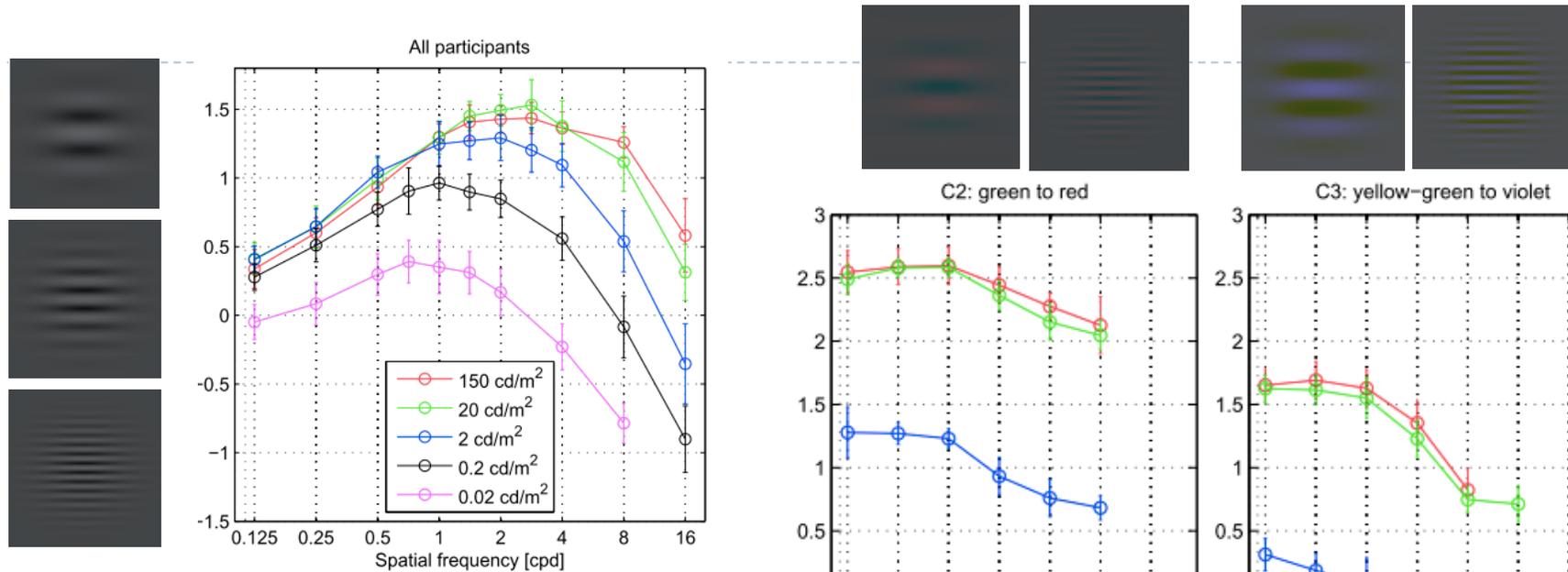


www.lam.mus.ca.us/cats/color/

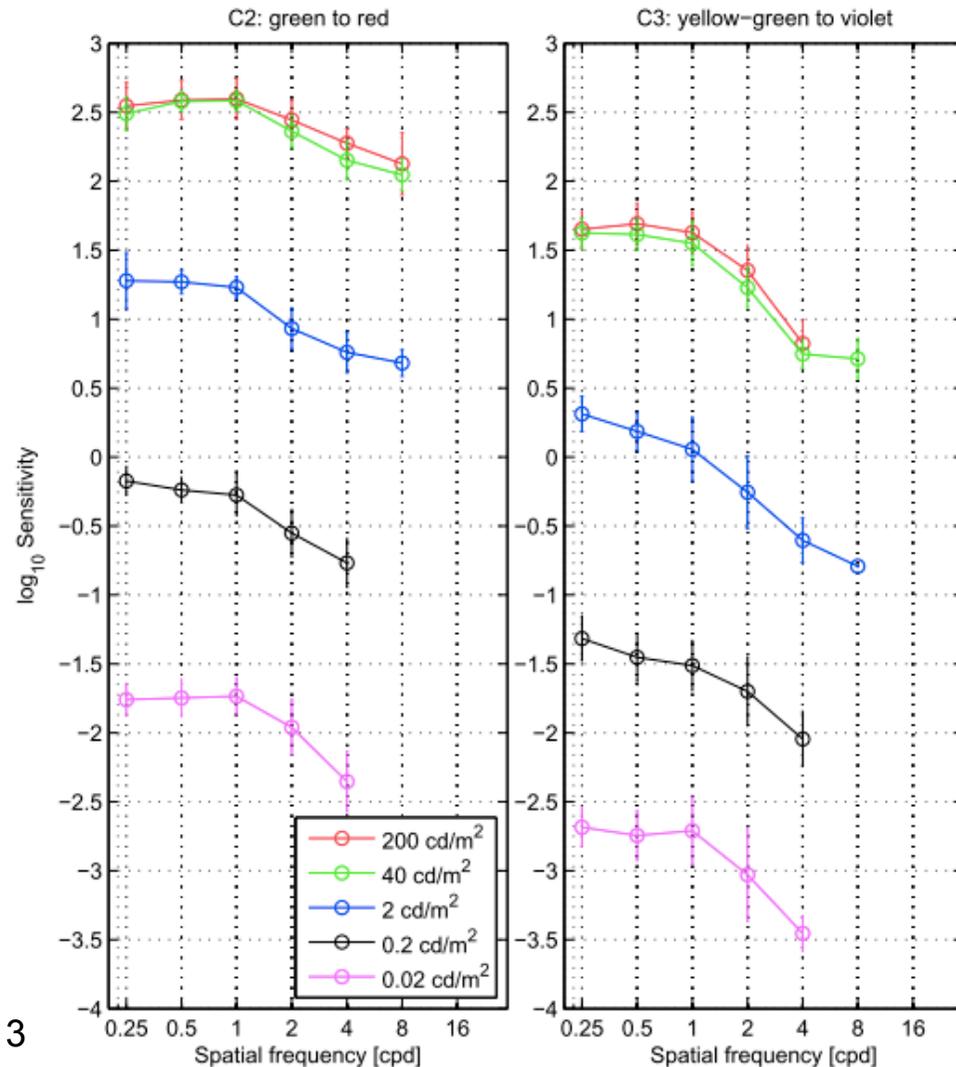


www.colorcube.com/illusions/ClrBlnd.html

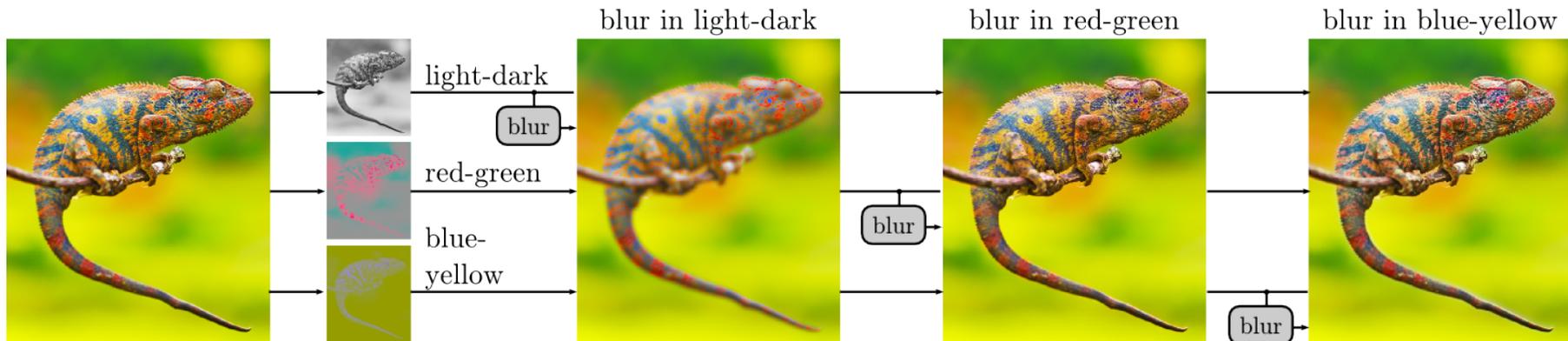
Colour Contrast Sensitivity



- ▶ **Colour vs. luminance vision system**
 - ▶ Higher sensitivity at lower frequencies for colour
 - ▶ High frequencies less visible



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)

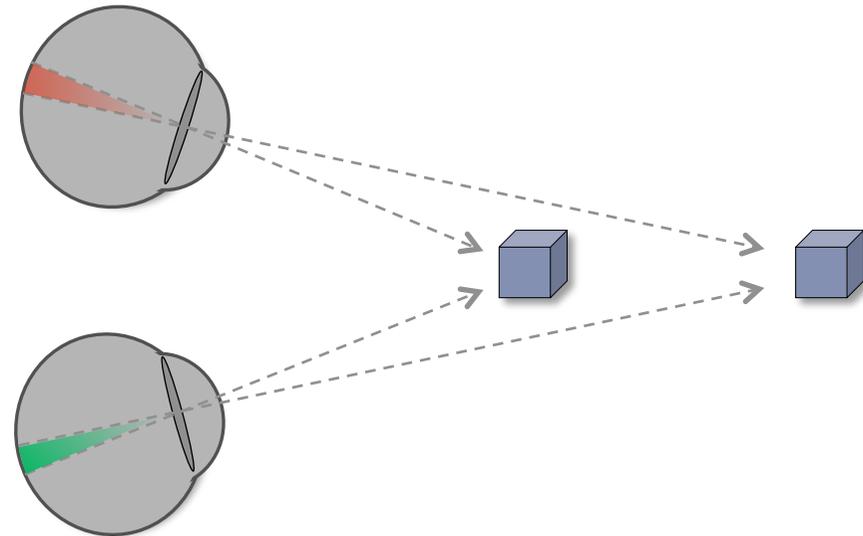
Depth perception

The slides in this section are the courtesy of Piotr Didyk (<http://people.mpi-inf.mpg.de/~pdidyk/>)

Depth perception

We see depth due to depth cues.

Stereoscopic depth cues:
binocular disparity



Depth perception

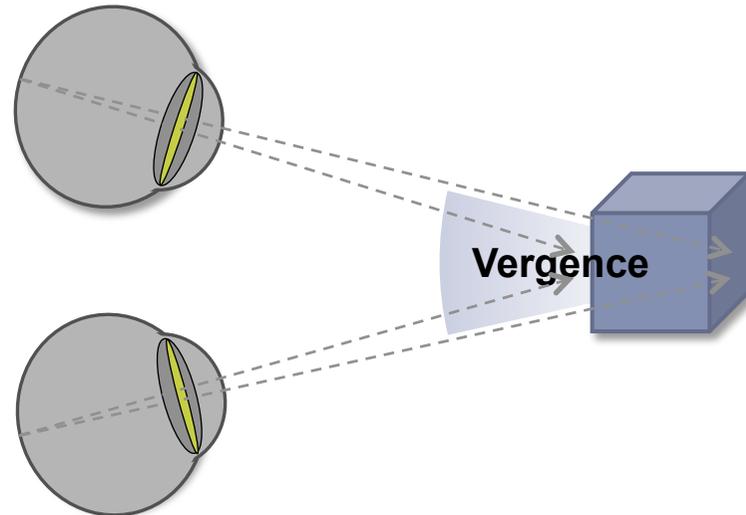
We see depth due to depth cues.

Stereoscopic depth cues:

binocular disparity

Ocular depth cues:

accommodation, vergence



Depth perception

We see depth due to depth cues.

Stereoscopic depth cues:

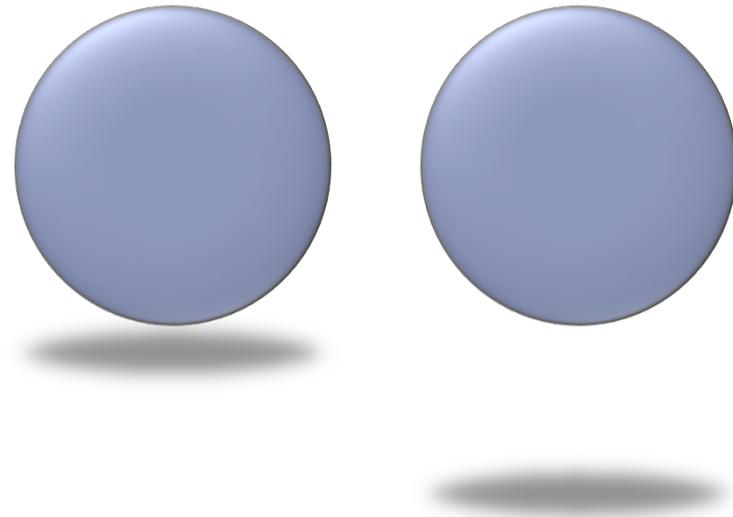
binocular disparity

Ocular depth cues:

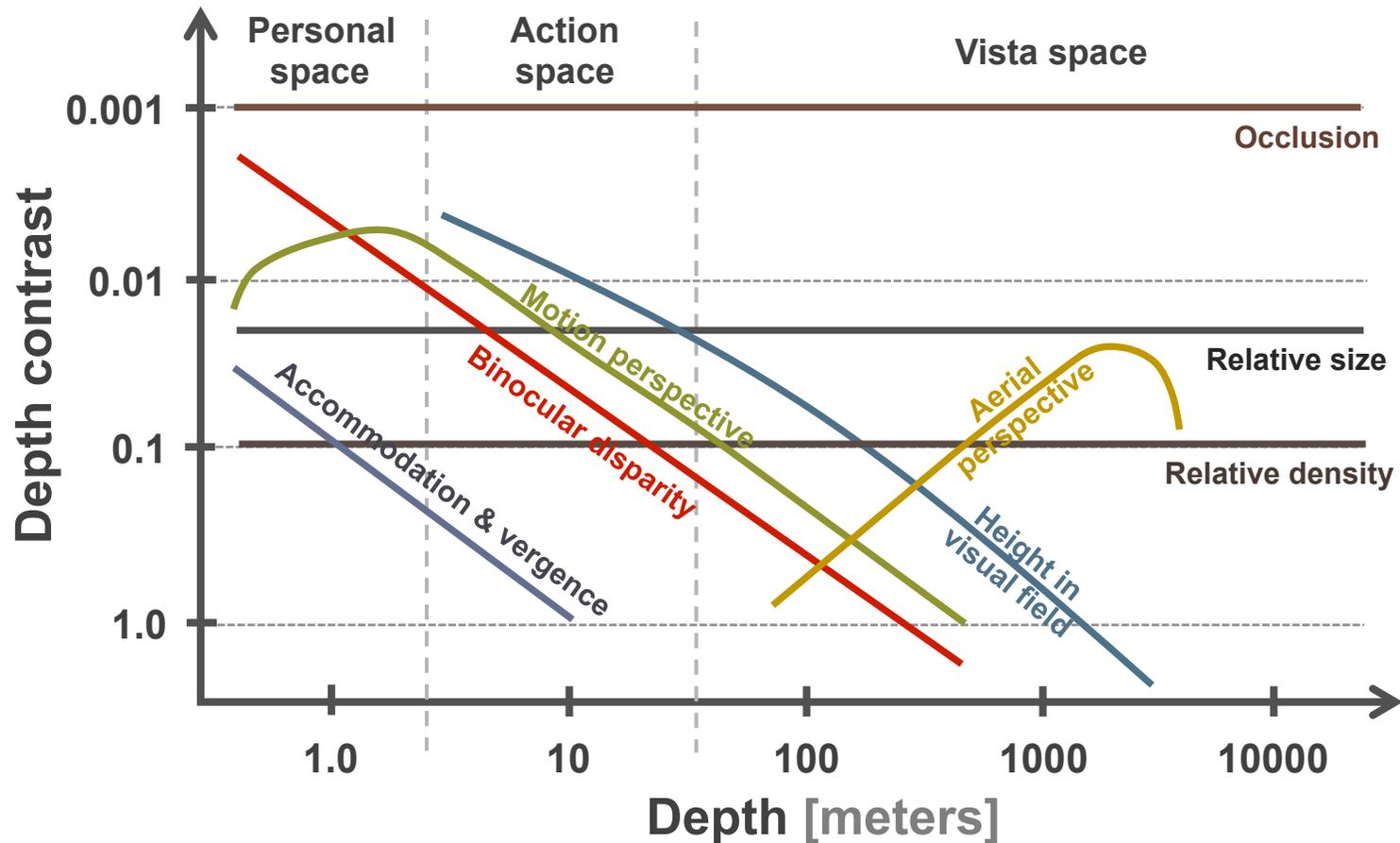
accommodation, vergence

Pictorial depth cues:

occlusion, size, shadows...



Cues sensitivity



"Perceiving layout and knowing distances: The integration, relative potency, and contextual use of different information about depth"
by Cutting and Vishton [1995]

Depth perception

We see depth due to depth cues.

Stereoscopic depth cues:

binocular disparity

Ocular depth cues:

accommodation, vergence

Pictorial depth cues:

occlusion, size, shadows...



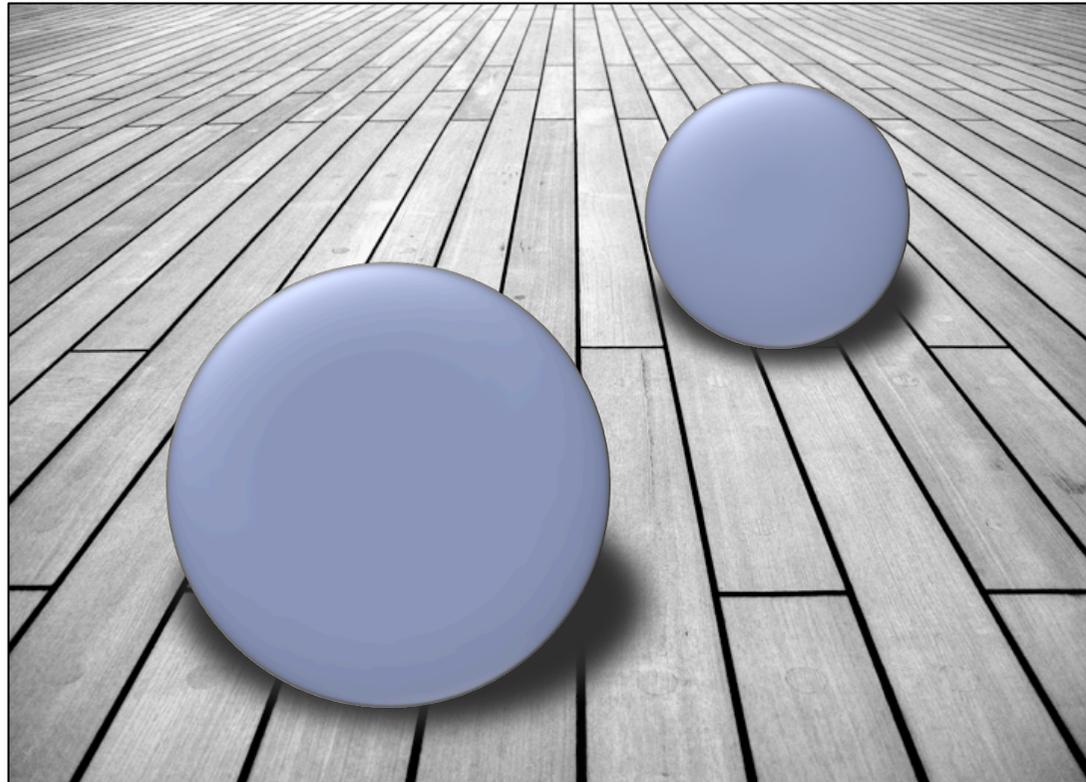
Challenge:
Consistency is
required!



Simple conflict example

Present cues:

- Size
- Shadows
- Perspective
- **Occlusion**



Disparity & occlusion conflict

Objects in front



Disparity & occlusion conflict

**Disparity & occlusion
conflict**



Depth perception

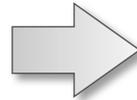
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binocular disparity

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accommodation, vergence

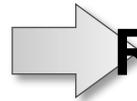


Require 3D space

We cheat our Human Visual System!

Pictorial depth cues:

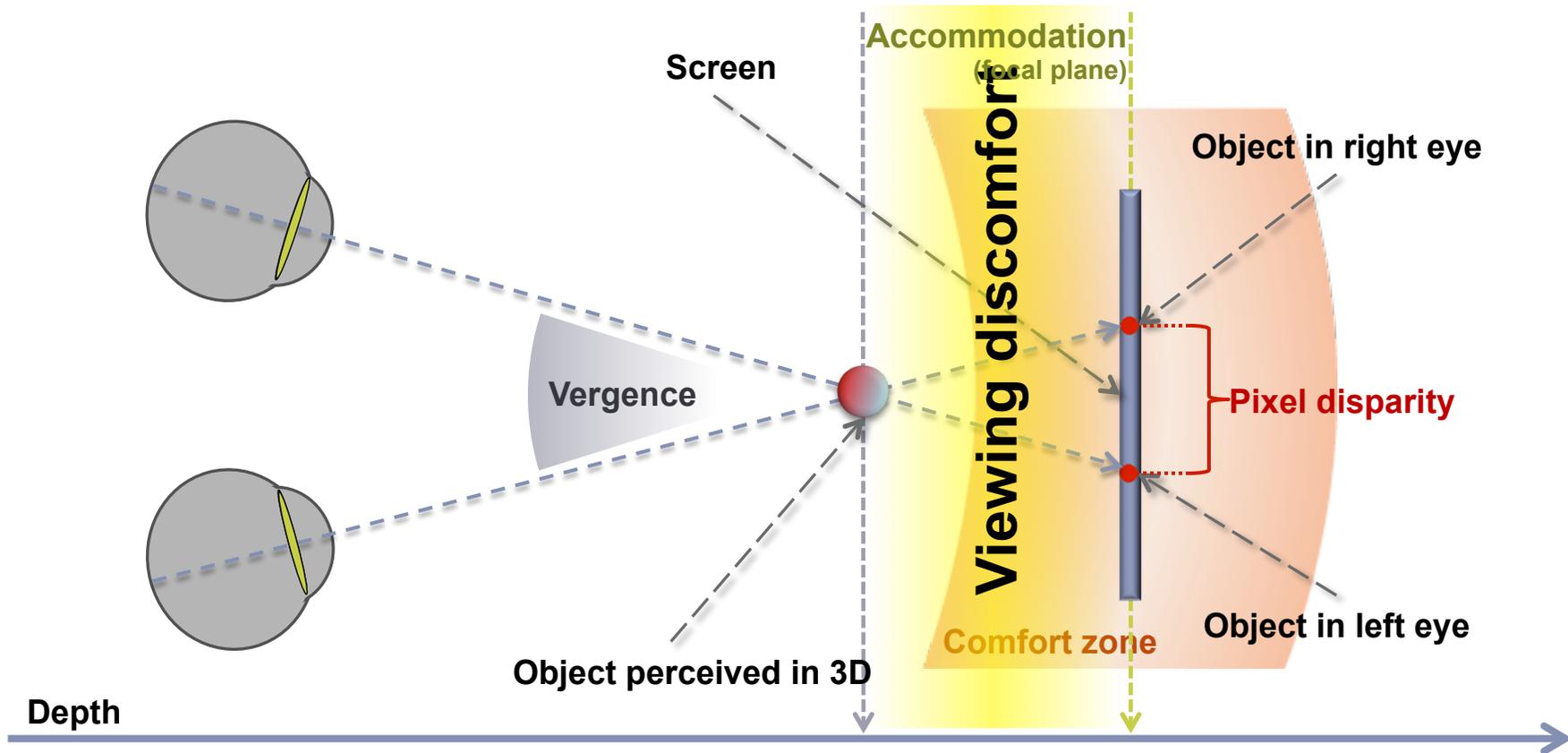
occlusion, size, shadows...



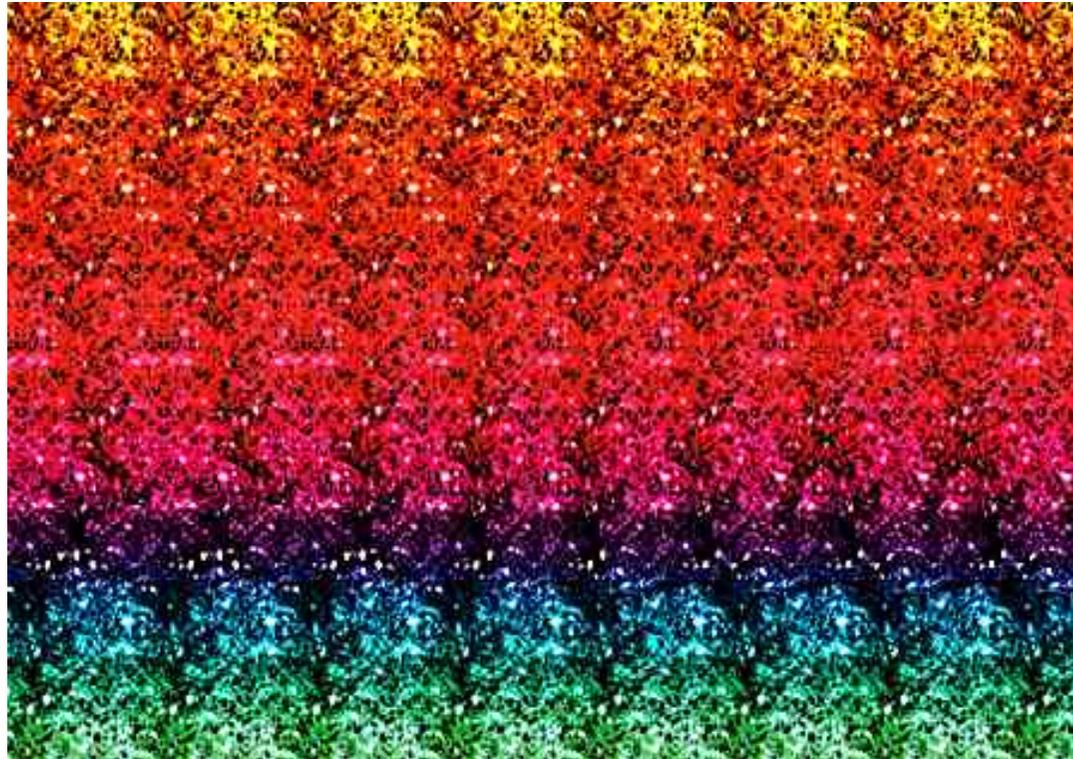
Reproducible on a flat displays



Cheating our HVS



Single Image Random Dot Stereograms



- ▶ Fight the vergence vs. accommodation conflict to see the hidden image

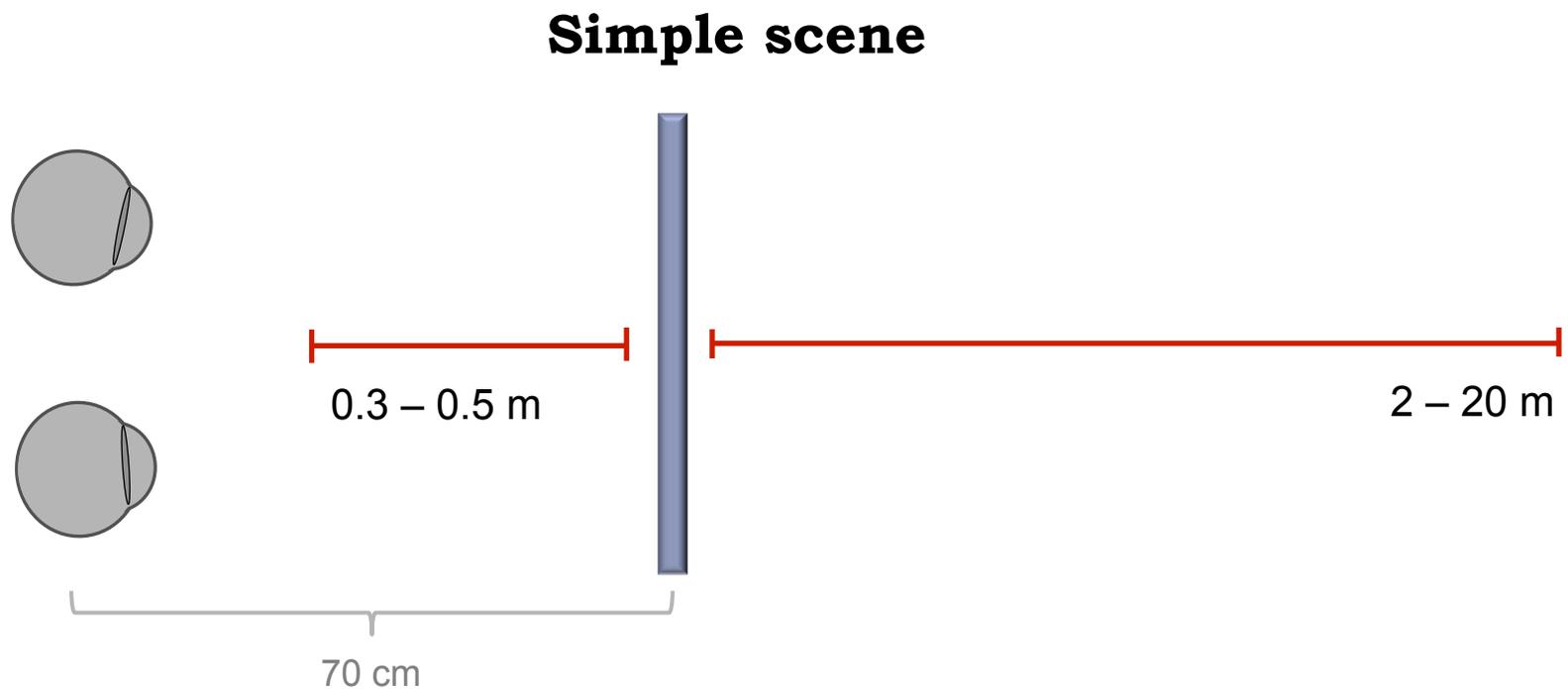
Viewing discomfort



Comfort zones

Comfort zone size depends on:

- Presented content
- Viewing condition



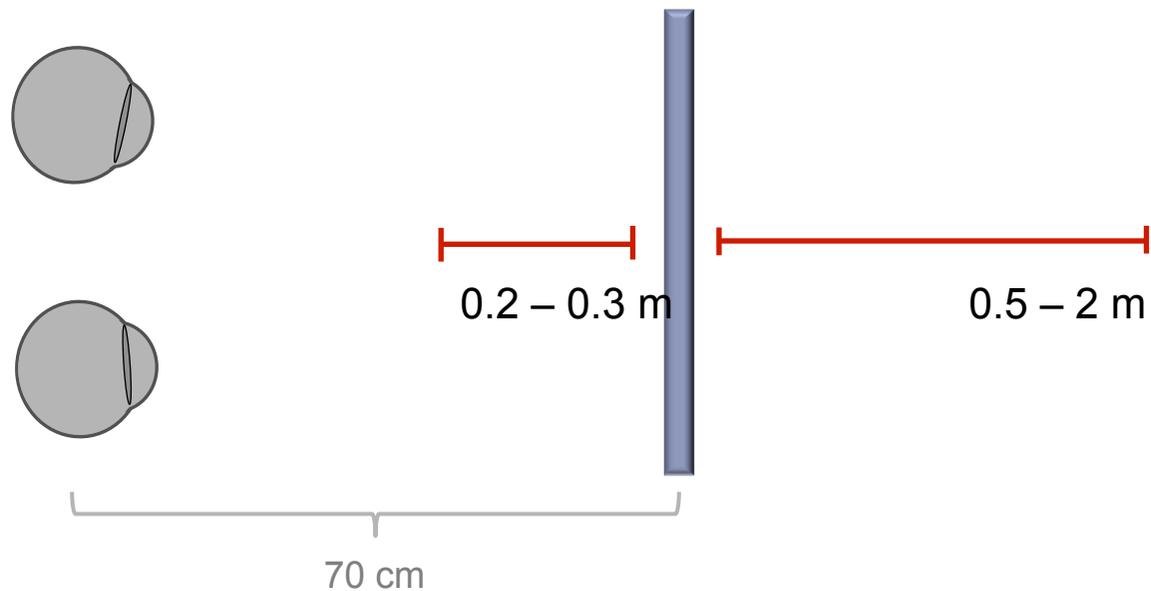
▶ *“Controlling Perceived Depth in Stereoscopic Images” by Jones et al. 2001*

Comfort zones

Comfort zone size depends on:

- Presented content
- Viewing condition

Simple scene, user allowed to look away from screen

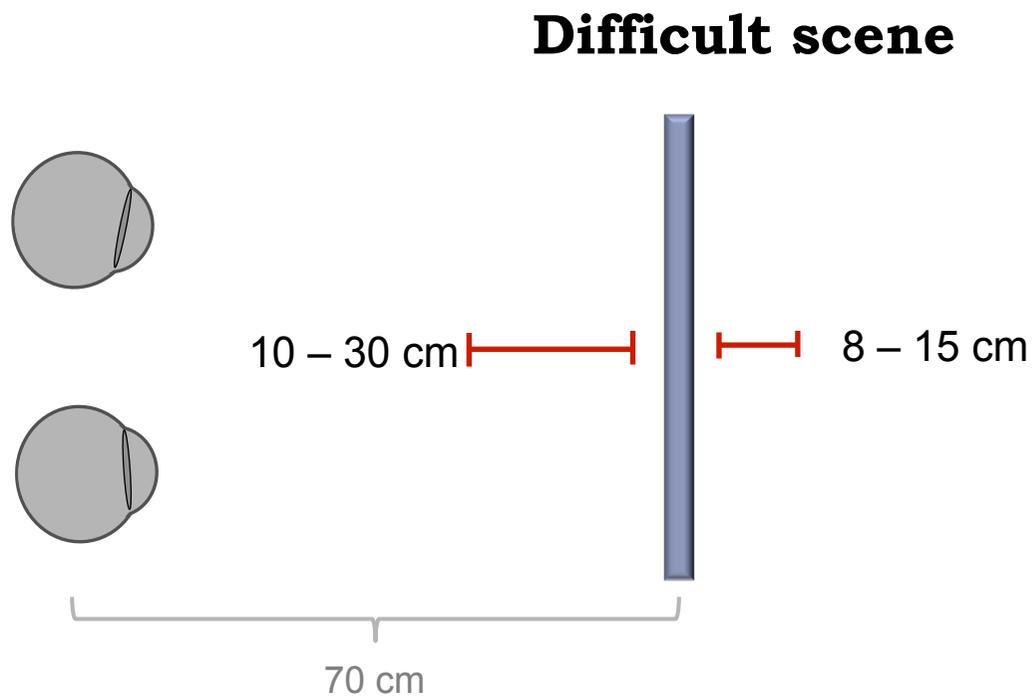


▶ *"Controlling Perceived Depth in Stereoscopic Images" by Jones et al. 2001*

Comfort zones

Comfort zone size depends on:

- Presented content
- Viewing condition



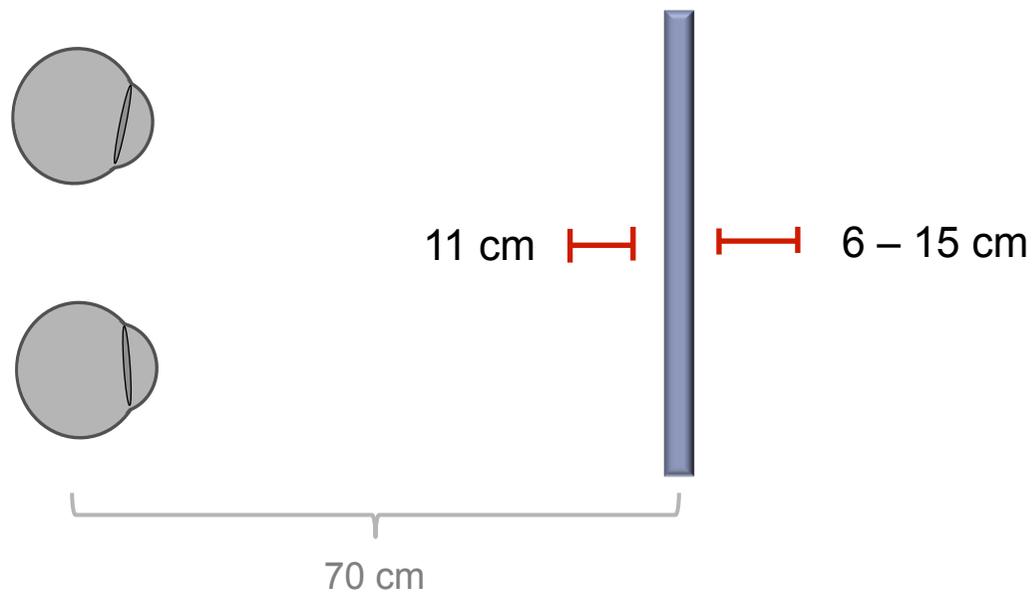
▶ *“Controlling Perceived Depth in Stereoscopic Images” by Jones et al. 2001*

Comfort zones

Comfort zone size depends on:

- Presented content
- Viewing condition

Difficult scene, user allowed to look away from screen



▶ *“Controlling Perceived Depth in Stereoscopic Images” by Jones et al. 2001*

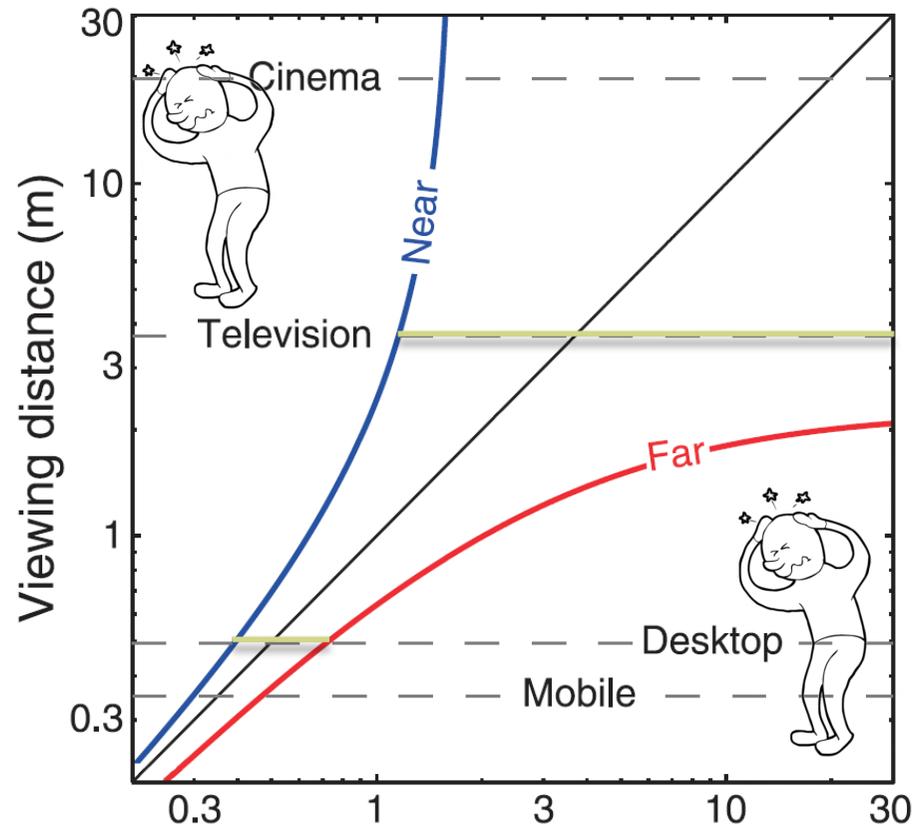
Comfort zones

Comfort zone size depends on:

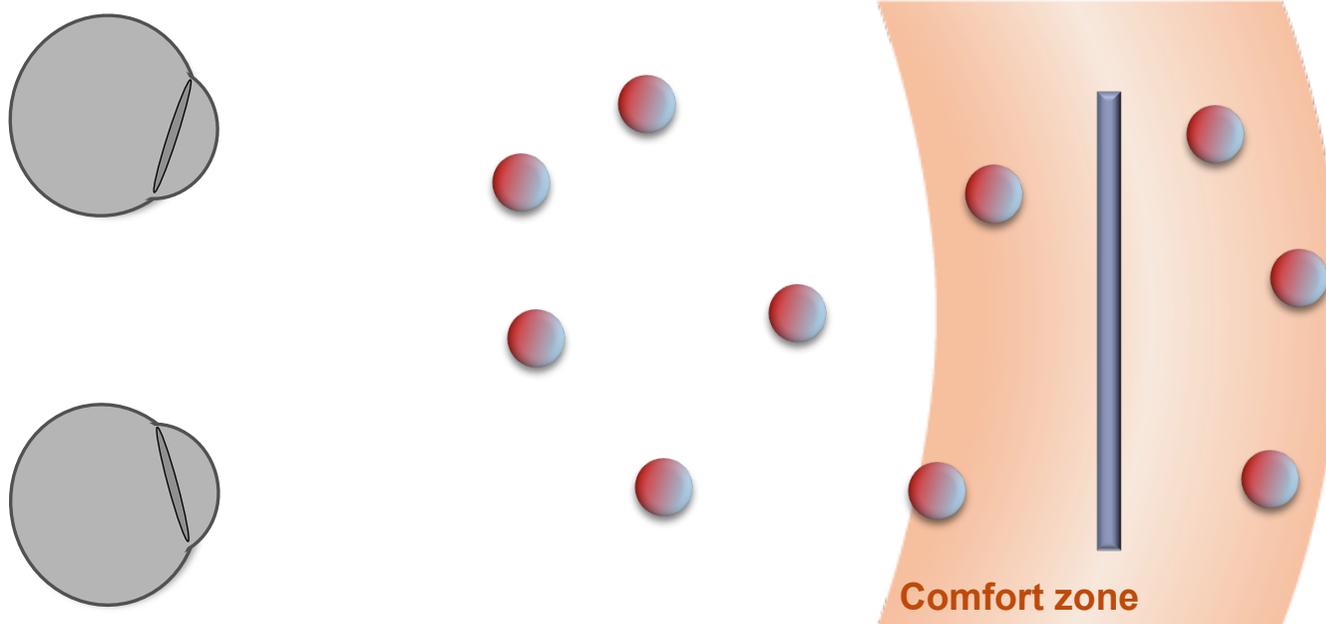
- Presented content
- Viewing condition
- Screen distance

Other factors:

- Distance between eyes
- Depth of field
- Temporal coherence

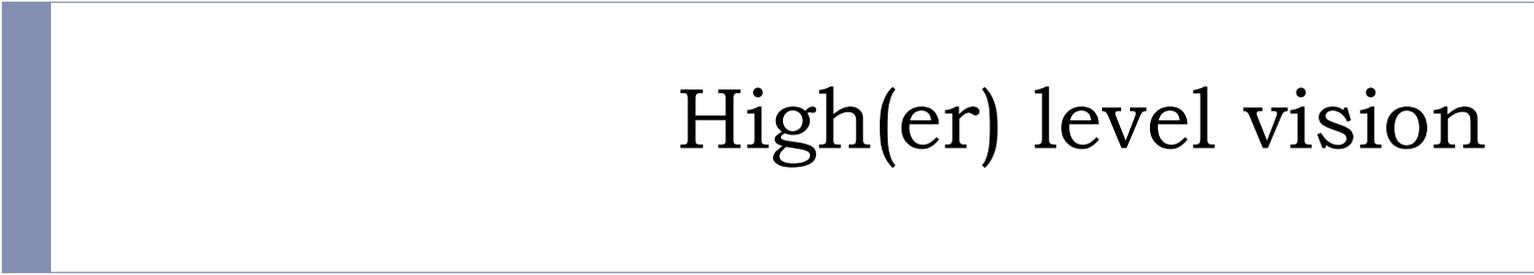


Depth manipulation



Scene manipulation
~~Viewing discomfort~~ Viewing comfort

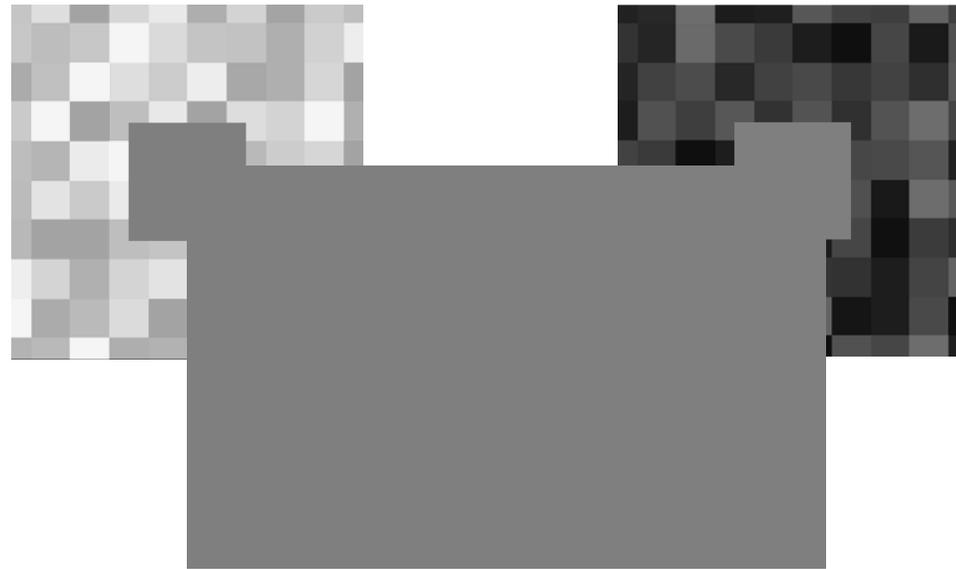




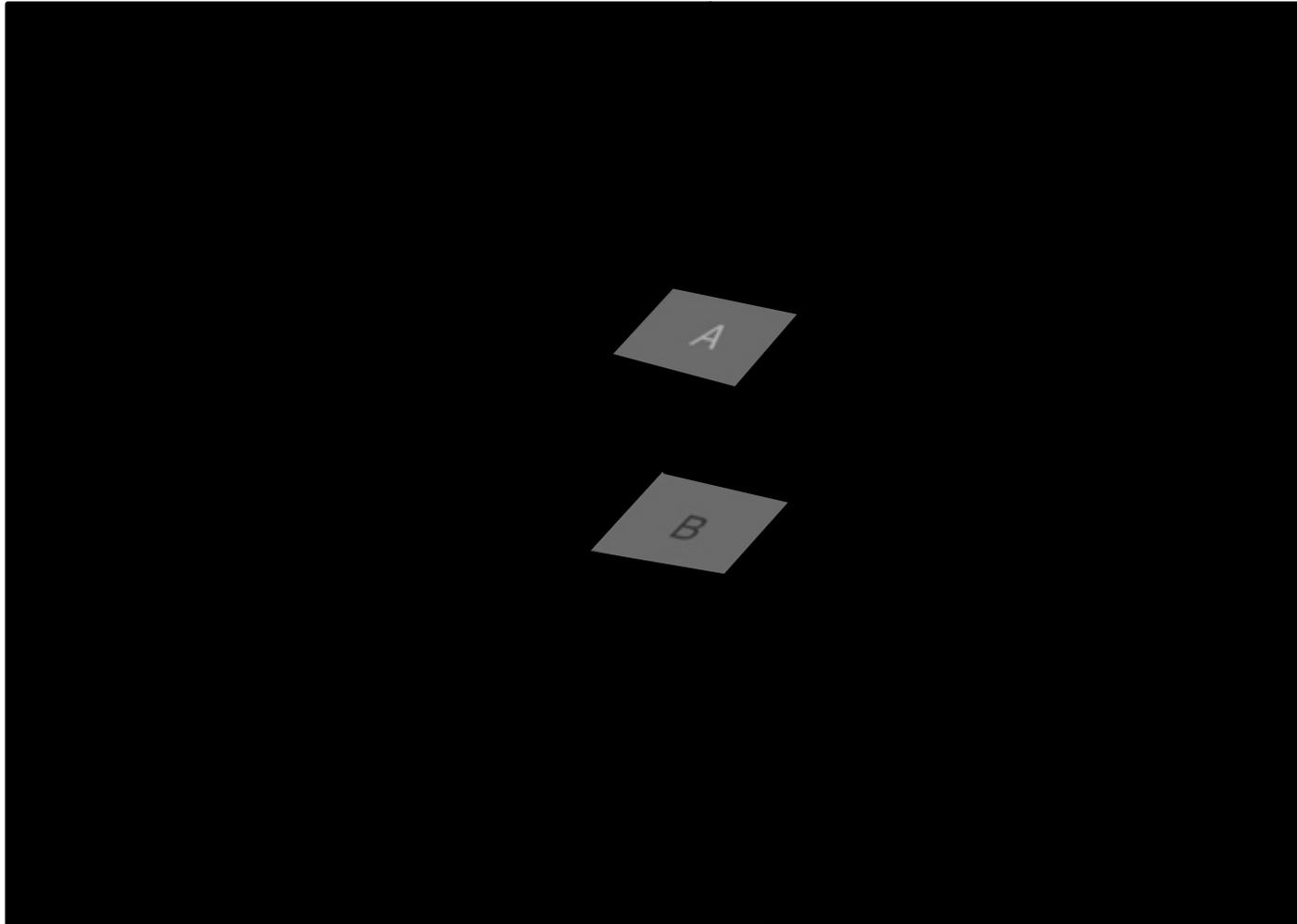
High(er) level vision



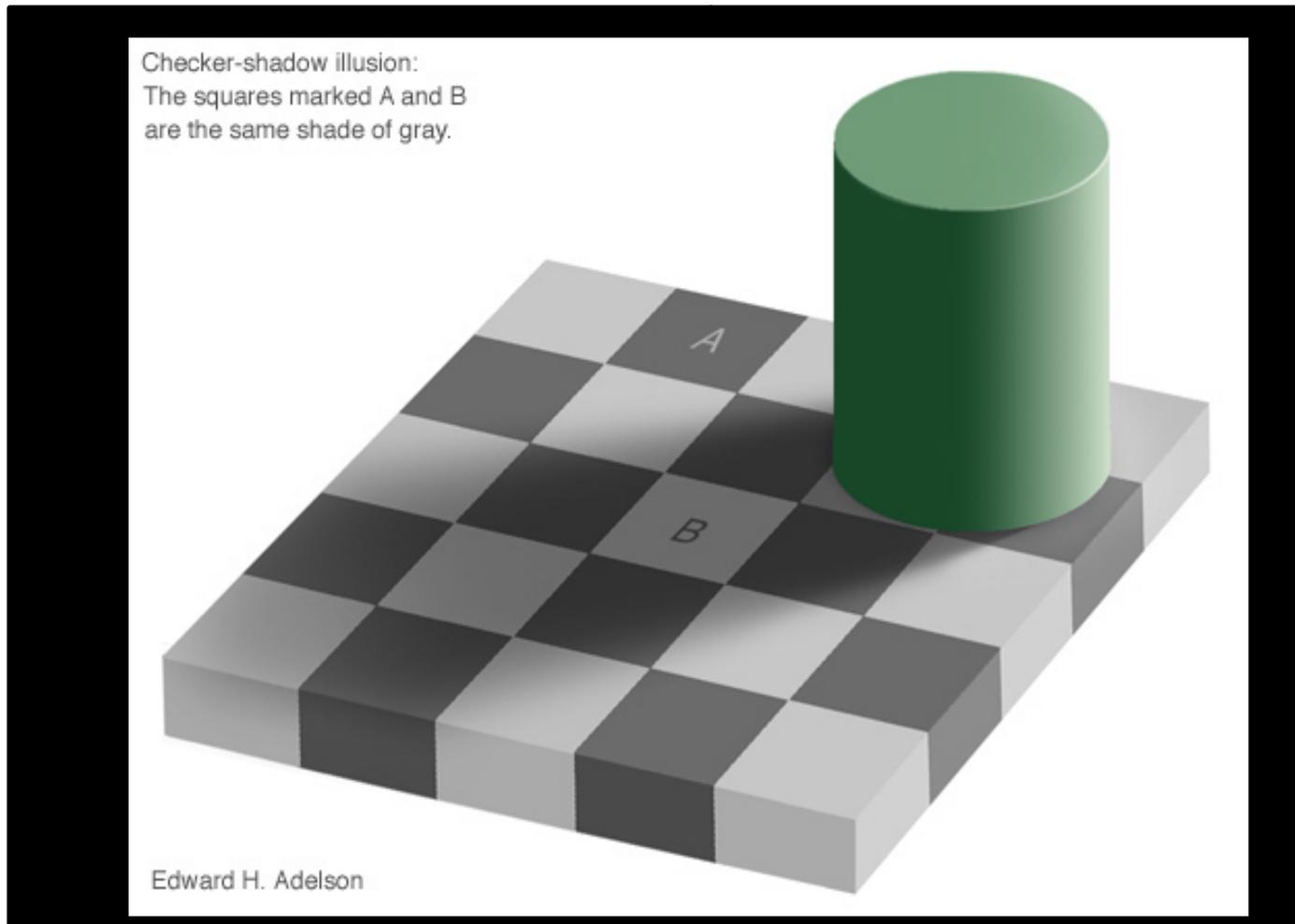
Simultaneous contrast



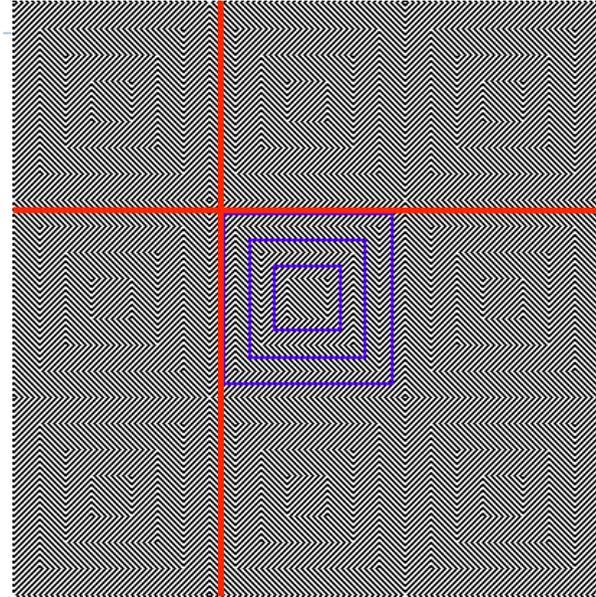
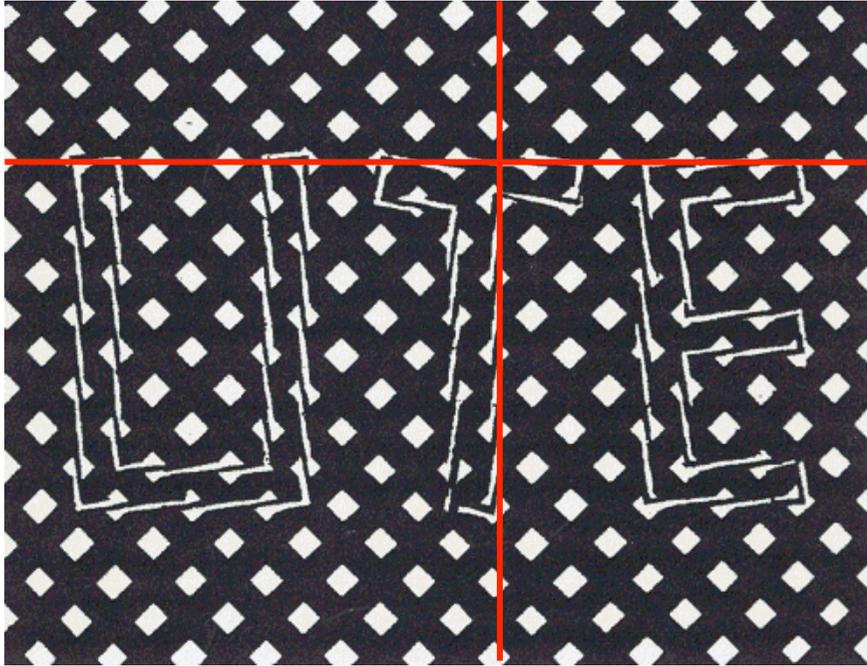
High-Level Contrast Processing



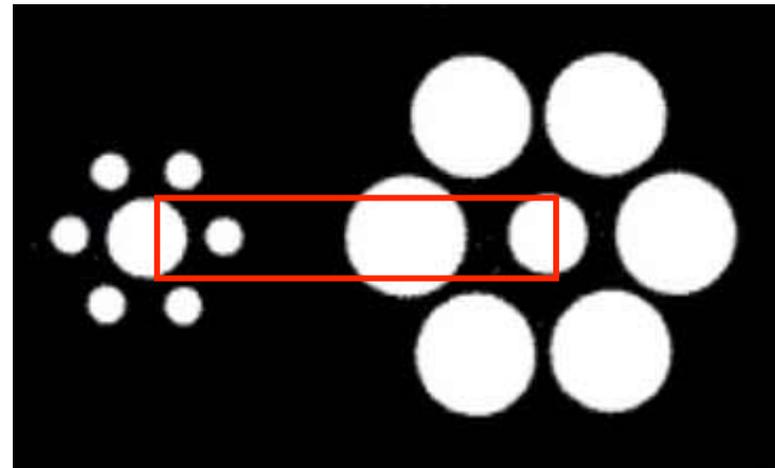
High-Level Contrast Processing



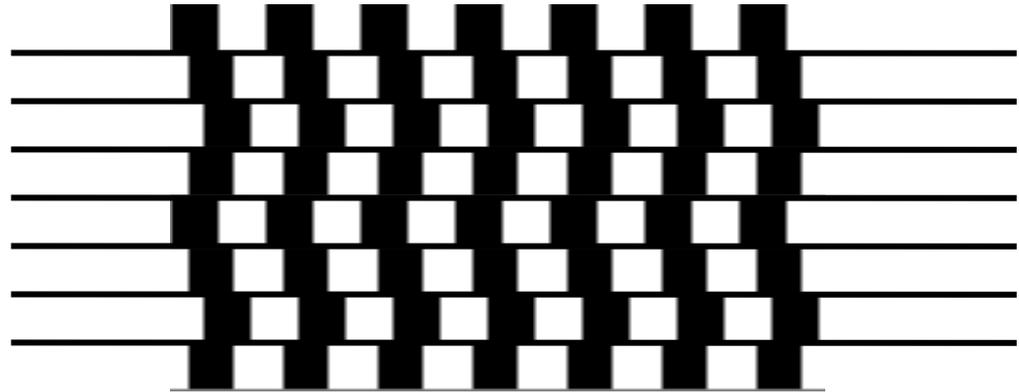
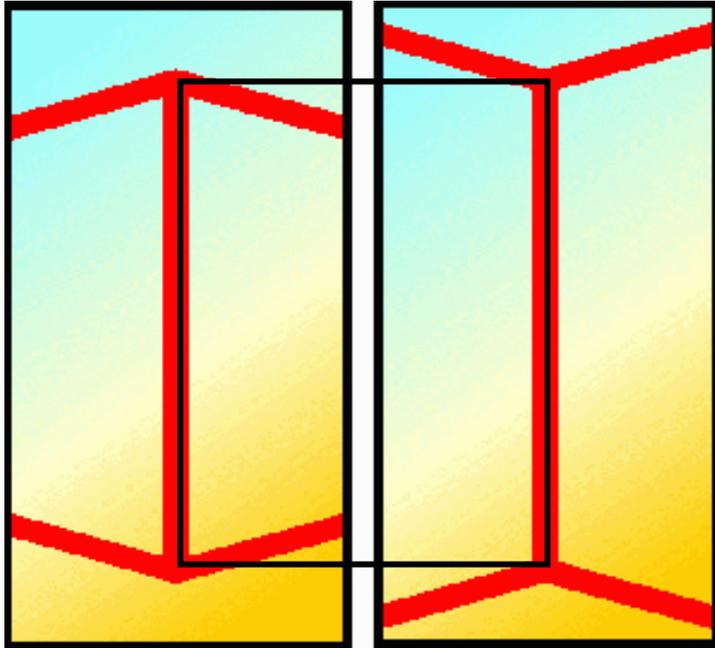
Shape Perception



- Depends on surrounding primitives
 - Directional emphasis
 - Size emphasis



Shape Processing: Geometrical Clues

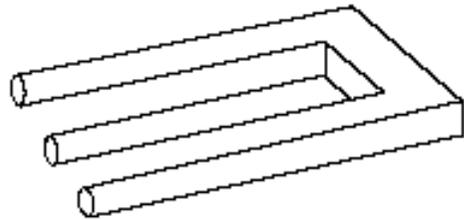
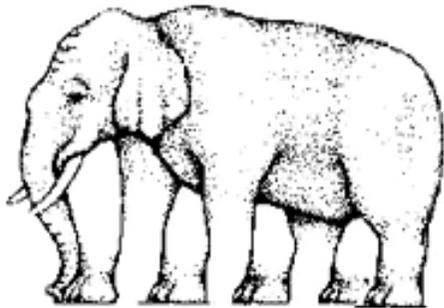


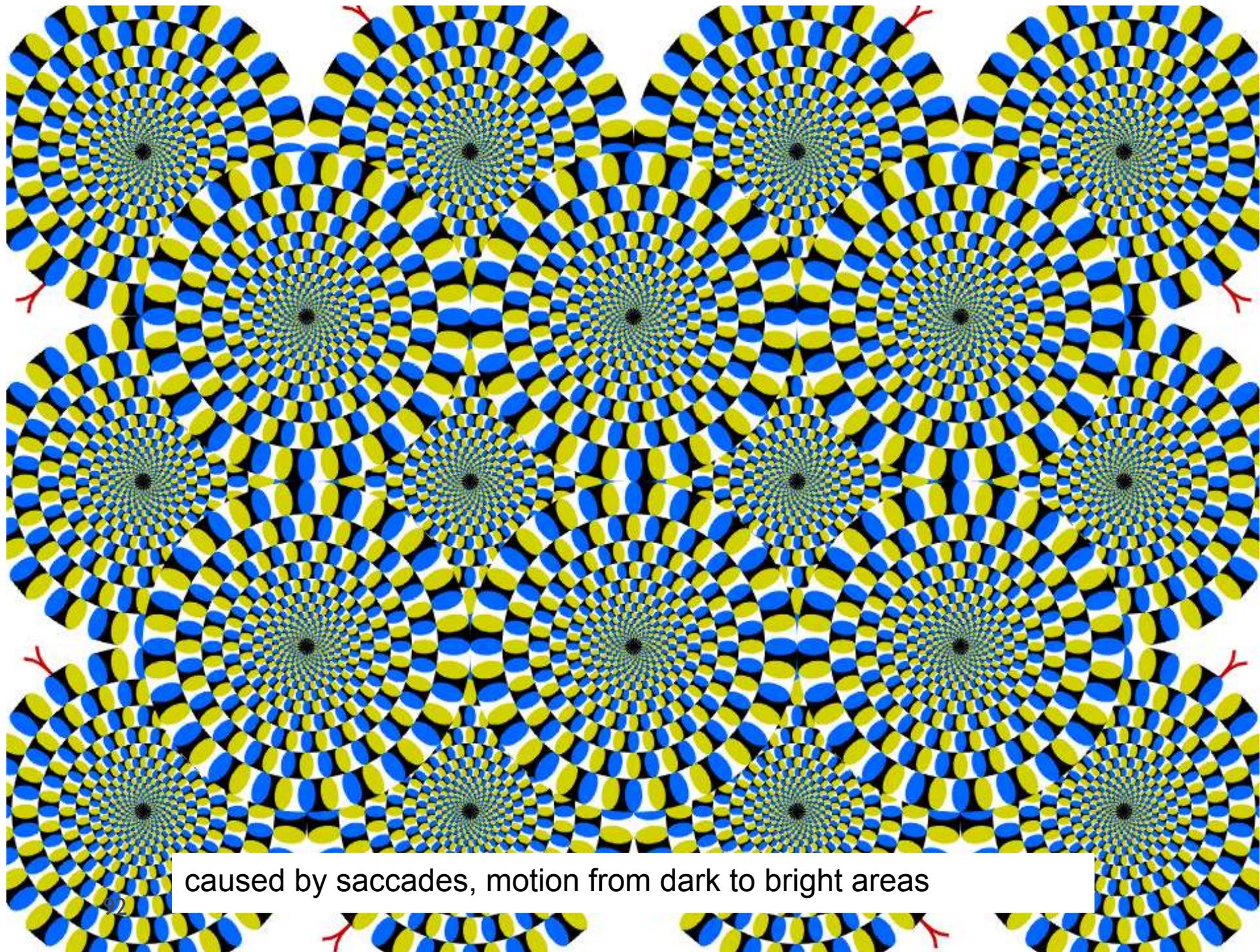
<http://www.panoptikum.net/optischetaeusungen/index.html>

- 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



Another Optical Illusion



- ▶ If you stare for approx. 20 seconds some of you will actually see a giraffe.

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