# Introduction to Computer Vision

#### Taught by

- prof. Luc Van Gool
- Prof. Ender Konukoglu
- Guest starring by prof
- Orcun Goksel

The course comes with a course text that covers most – but not all ! – material. Slide decks for all lectures will be made available on eDoz or similar We got questions about which course to take

Computer Vision (D-INFK), or Image Analysis and Computer vision (this course)

#### IN ANY CASE, DO NOT TAKE BOTH !

If you took the introductory course on CV at D-INFK, then best take Computer Vision

If you did not take that course, then best take Image Analysis and Computer Vision We got questions about which course to take

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# ... it is crucial ...

The central take-home message:

For people vision is their most crucial sense, for good reason



# Vision is important

half our brain is devoted to it

developed many times during evolution

□ it is non-contact

□ it can be implemented with high resolution

works with ambient E-M waves

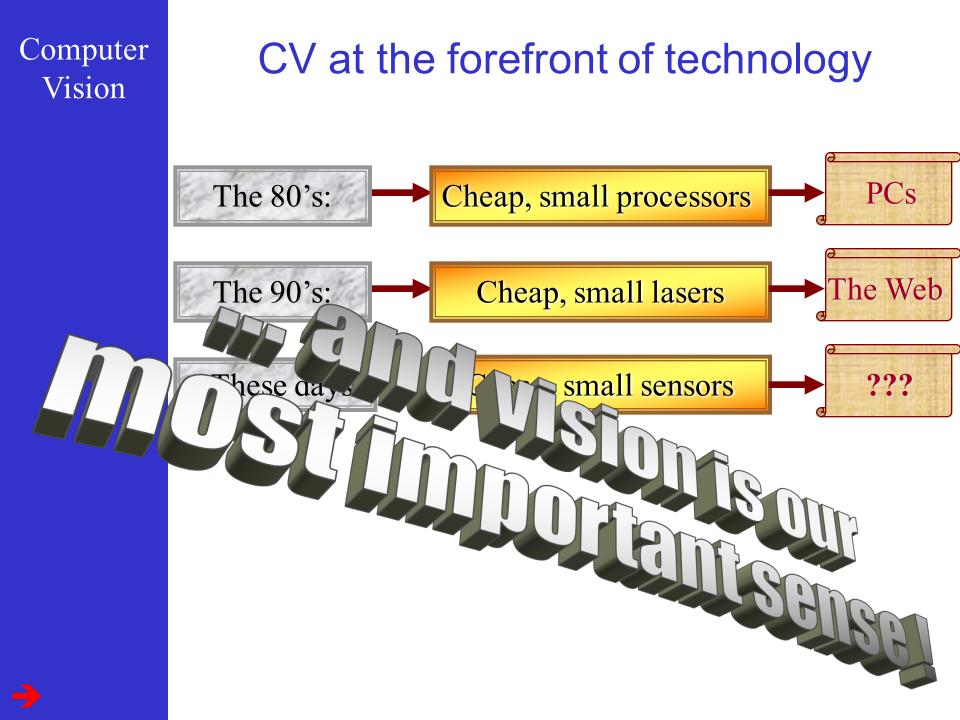
□ yields colour, texture, depth, motion, shape

# ... it is hot ...

The central take-home message:

It is feasible now to let most things see their environment







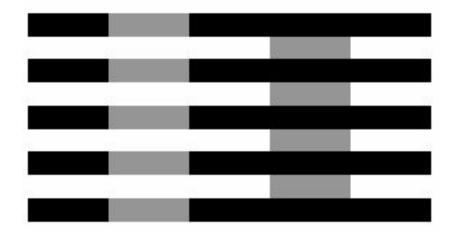
# ... it is intriguing ...

The central take-home message:

# Effective vision needs more than sheer filtering and measuring

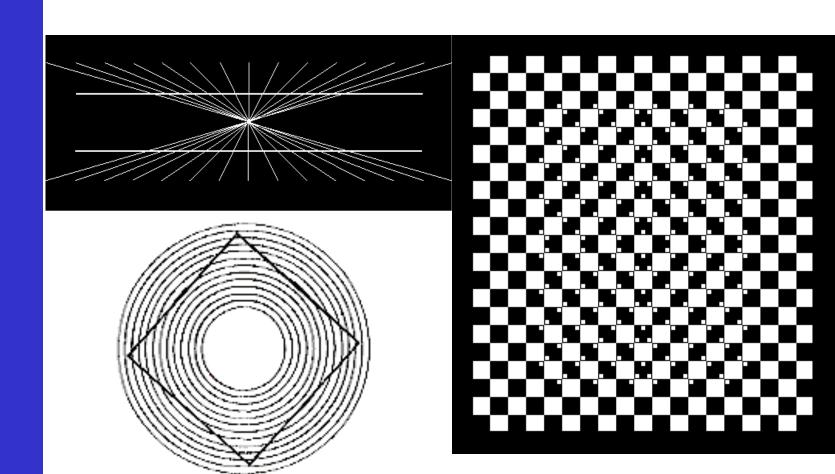


## The perception of intensity

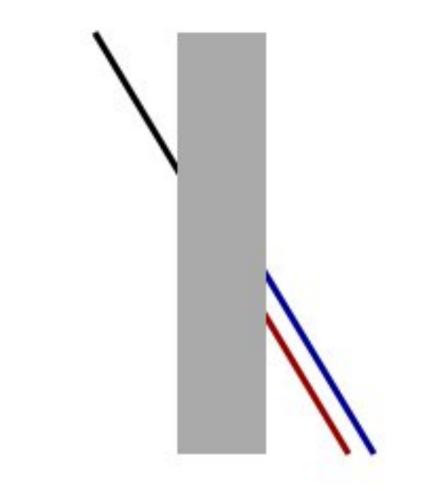




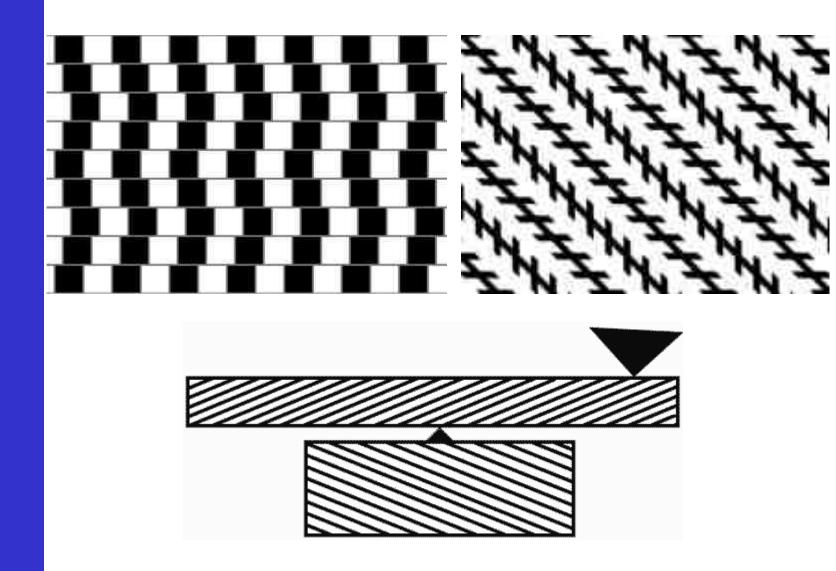
## The perception of lines being straight



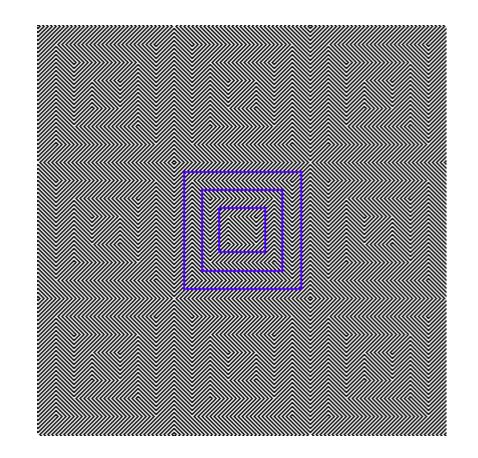
### The perception of colinearity



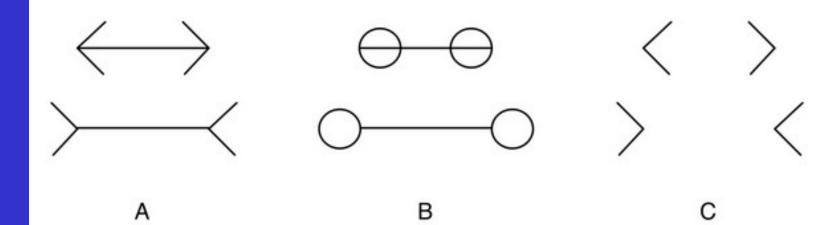
#### The perception of parallelism



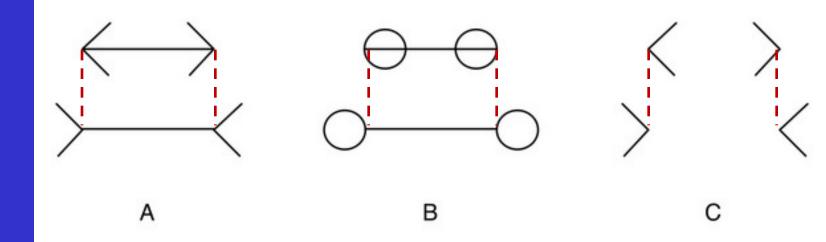
## The perception of squares



The perception of length

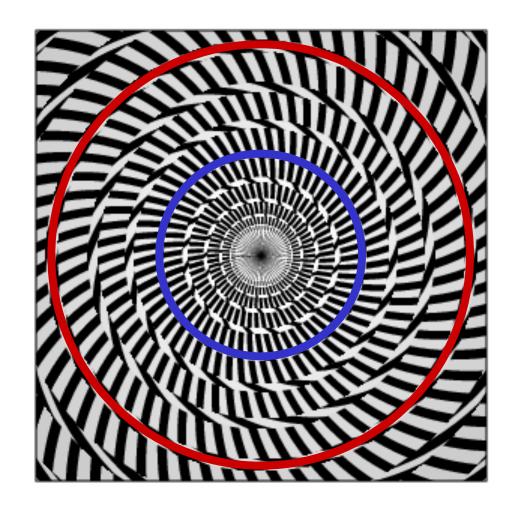


The perception of length



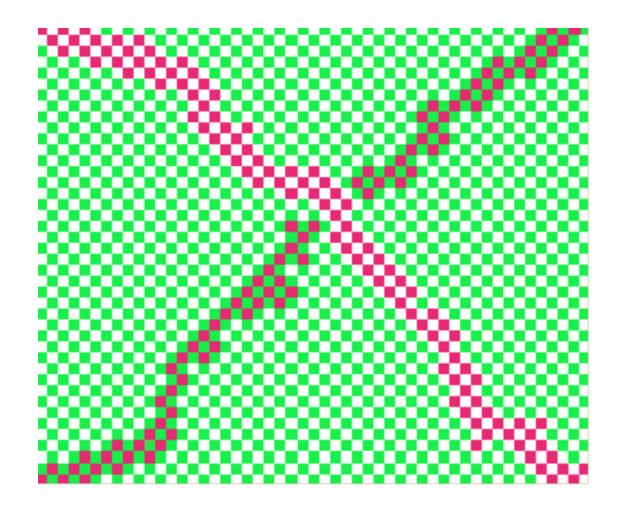
The horizontal lines are equally long...

## The perception of curvatures



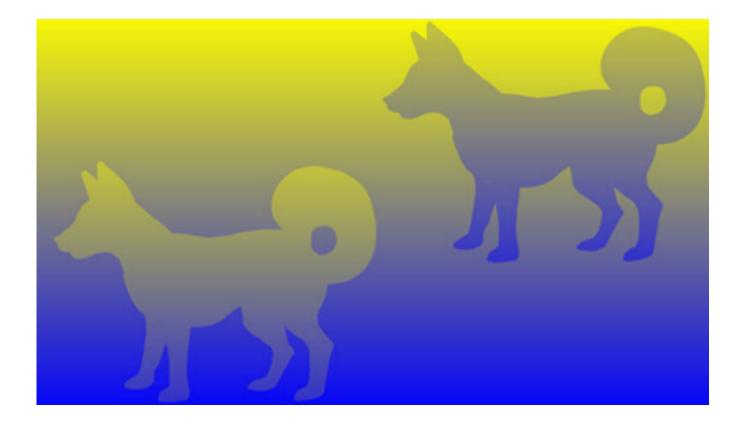
Illusions : interference of differently oriented patterns via adaptation

The perception of color



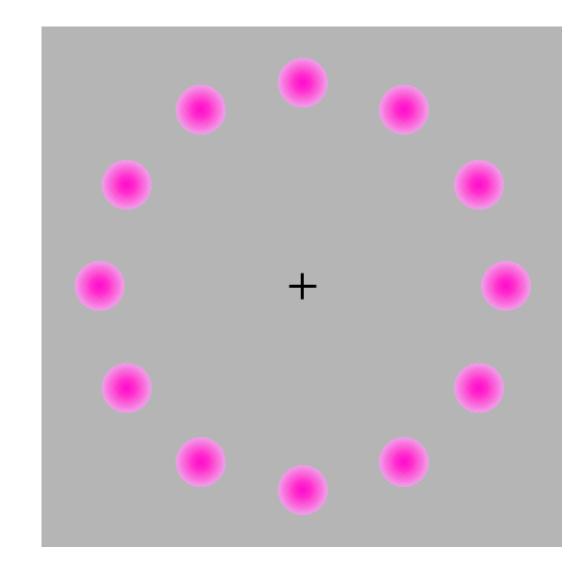
The red squares have equal color...

#### The perception of color



The dogs are identical...

#### The perception of color



Focus on the cross and you see a virtual green dot circling...

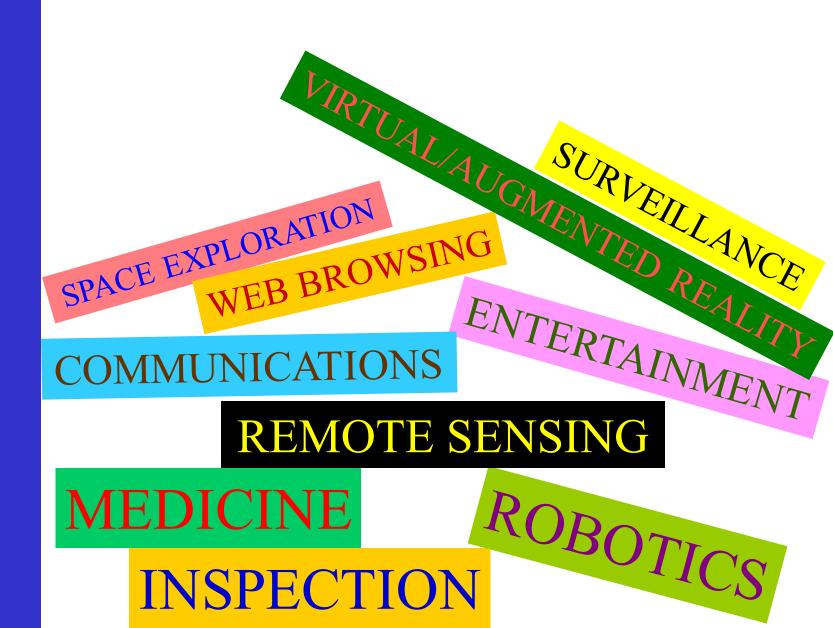
#### The perception of motion



The `barber pole' rotates about the vertical, it does not translate vertically...

# ... it is useful ...

# Applications of computer vision



# The development of computer vision apps

Most early applications where found in production environments, as these allow for *controlled conditions* and *have little uncertainty* 

some areas do not allow for much control: medical IP, remote sensing, surveillance, etc. currently CV is conquering the less controllable areas by storm

# Rationales

- increase productivity
- increase reliability
- increase flexibility
- decrease costs
- assist with quantitative aspects of a job
- guarantee constant vigilance/assistance
- □ realise intelligent man/machine interfaces
- automate complex processes
- □ generate more complete 3D models

```
Computer
Vision
```

# Rationales

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# Rationales

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  increase reliability apps
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Rationales behind computer vision for traditional visual inspection

a well engineered vision system can:

increase productivity

□ increase reliability

□ increase flexibility

decrease costs



# Rationales

- increase productivity
- increase reliability
- □ increase flexibility
- decrease costs

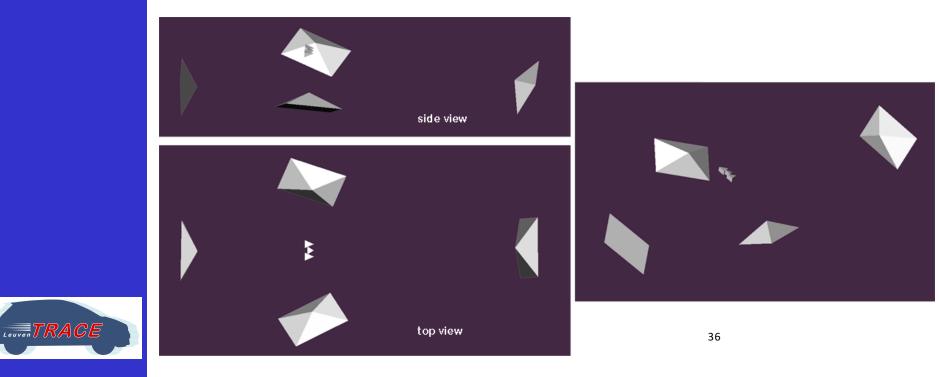
assist with quantitative aspects of a job
 guarantee constant vigilance/assistance
 realise intelligent man/machine interfaces
 automate complete 3D models

## Ex App: autonomous vehicles



# Ex App: autonomous vehicles

- 3 forward normal (1 stereo pair + 1 wide angle) + 4 fish eyes for 360 vision
- Images below show the result of calibration
- We also take into account if cams are behind glass



# Ex App: autonomous vehicles

### car detection:





AUTOSENS 2017

# Ex App: autonomous vehicles

### pedestrian detection:

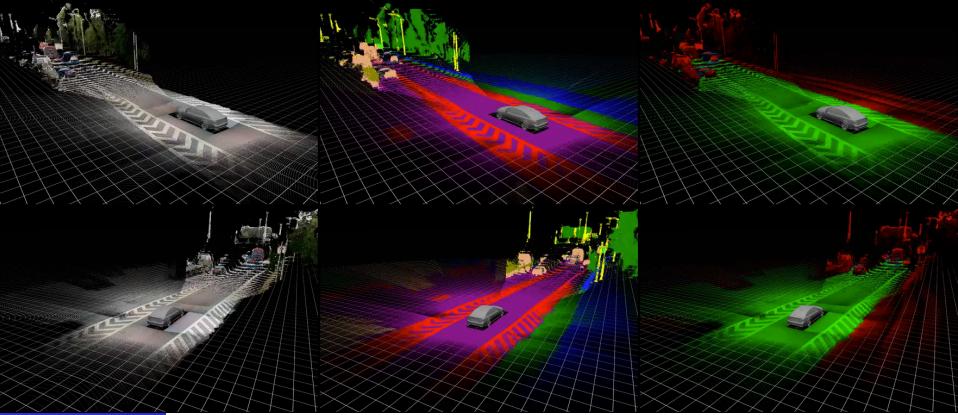




AUTOSENS 2017

# Ex App: autonomous vehicles

### putting vision modalities together:





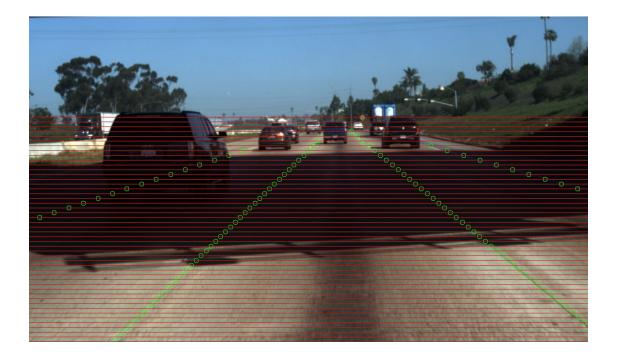
AUTOSENS 2017

# Ex App: autonomous vehicles

### Latest addition: lane marking detection

Annotations based on new benchmark : http://benchmark.tusimple.ai/#/t/1 Markings are annotated as series of connected points on equidistant rows (see example picture below) Note: lanes are also annotated when markings are not present, not visible, or

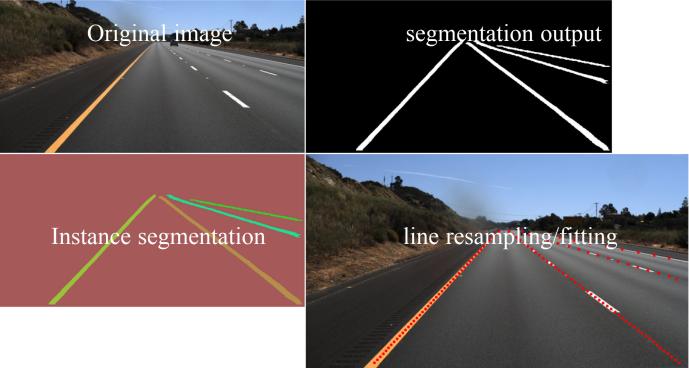
even occluded by cars





Ex App: autonomous vehicles

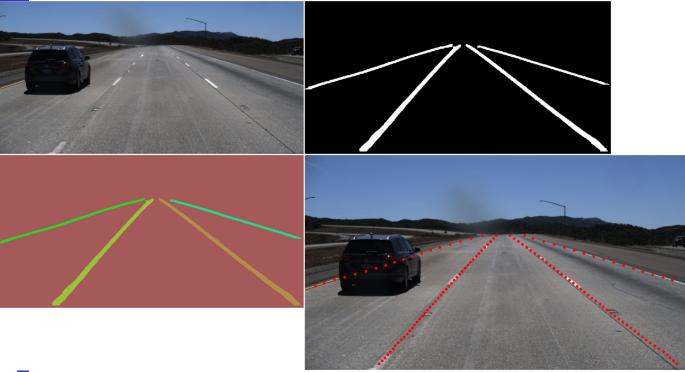
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Ex App: autonomous vehicles

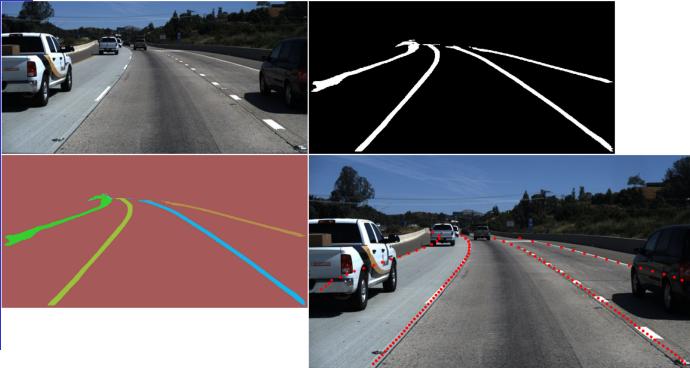
# Latest addition: lane detection





Ex App: autonomous vehicles

# Latest addition: lane detection





# Ex: autonomous mobile platform



# Ex App: image retrieval, captioning, ...

**Describes without errors** 



A person riding a motorcycle on a dirt road.



**Describes with minor errors** 

Two dogs play in the grass.

Somewhat related to the image

A skateboarder does a trick on a ramp.



A little girl in a pink hat is blowing bubbles.



A red motorcycle parked on the side of the road.

#### Unrelated to the image



A dog is jumping to catch a frisbee.



A refrigerator filled with lots of food and drinks.



A yellow school bus parked in a parking lot.



A group of young people playing a game of frisbee.



A herd of elephants walking across a dry grass field.

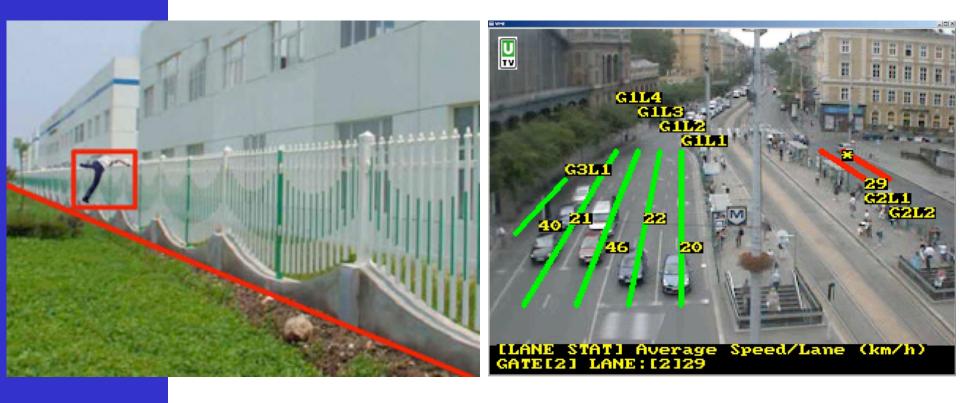


Two hockey players are fighting over the puck.



A close up of a cat laying on a couch.

# Ex App: visual surveillance



# Ex App: Augm. Reality, eg sports



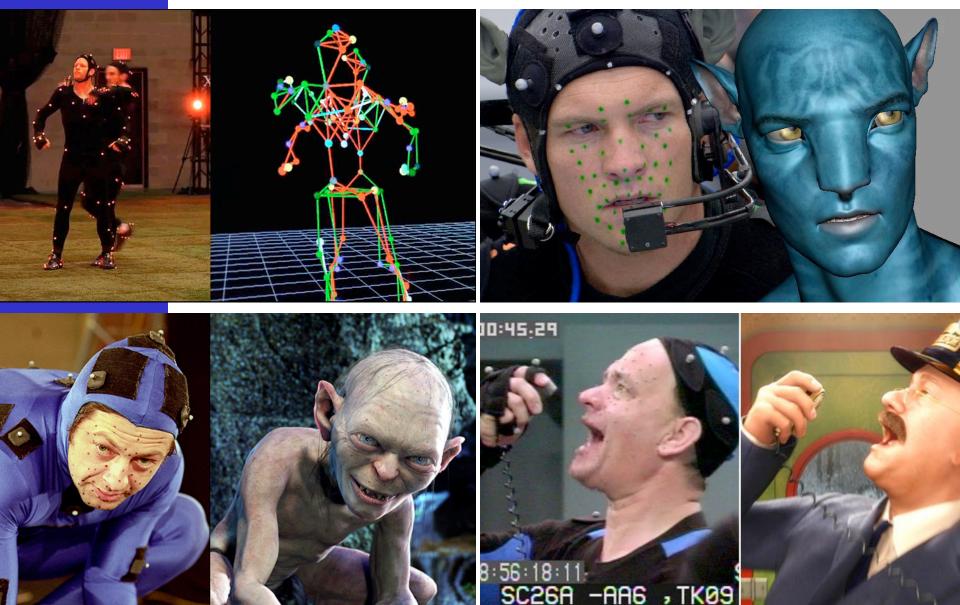


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# Ex App: motion capture for movies/games



# Ex App: facial performance capture

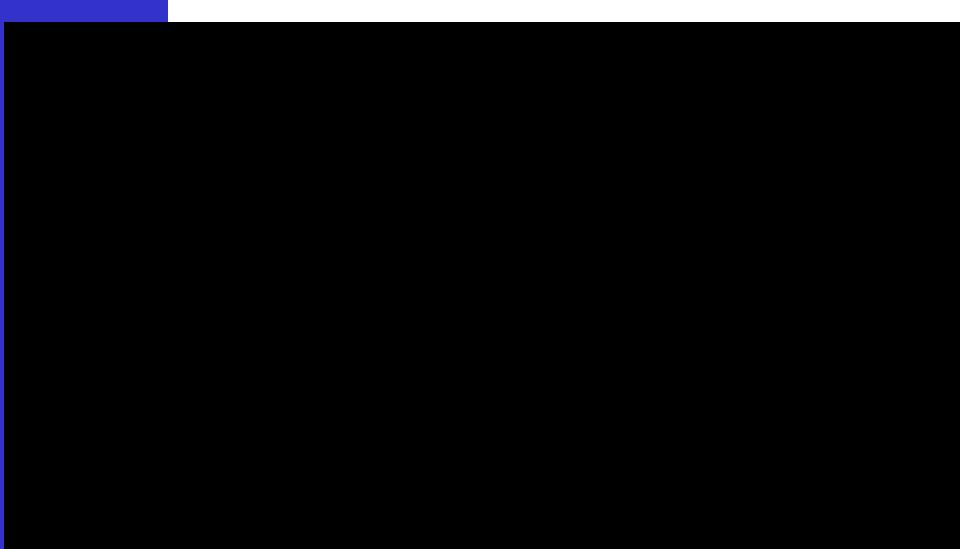
#### COMPUTERCAFE



LC015 Eyetronics Face Capture Test V01 1/291

03 / 11 / 2003

## Ex App: advert replacement



# Ex app: Avatar guidance via 3D facial capture

# Face/Off: Live Facial Puppetry PaperID 102

Developed by our spin-off FaceShift, acquired by Apple. The basis for their user-animated emoji's in new iPhone

# Ex App: computer-assisted surgery



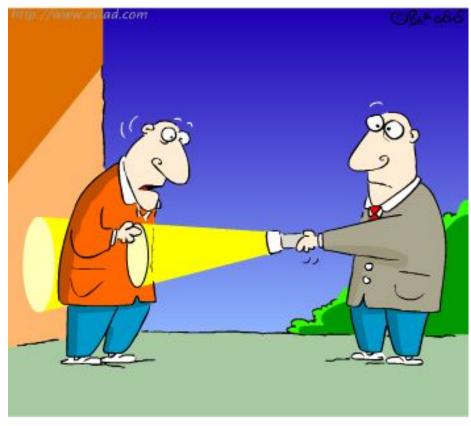
# Mobile mapping



# ... it needs light ...

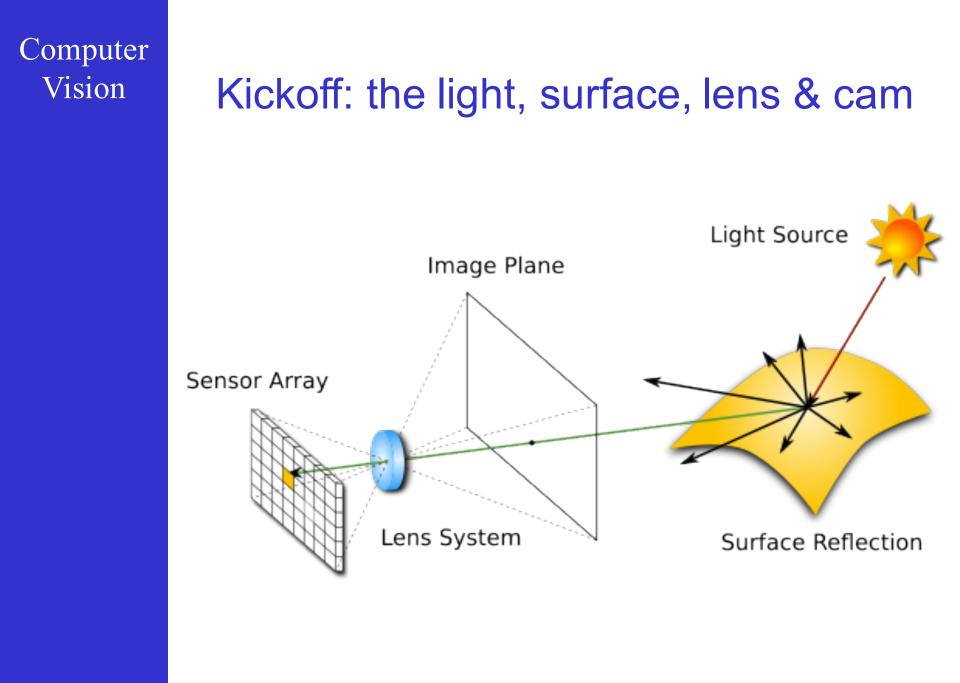
And then there was Light...

no vision without light...... because it is influenced by objects



"What the ...?"





# Computer Vision Kickoff: the light, surface, lens & cam Light Source Image Plane Sensor Array Lens System Surface Reflection





# the nature of light

# interactions with matter





An option on optics

1. Geometrical optics

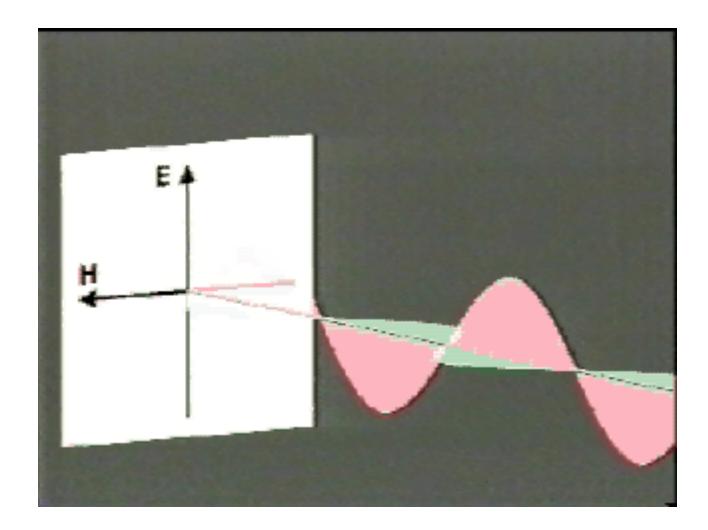
2. Physical optics, or

3. Quantum-mechanical optics

→ wave character

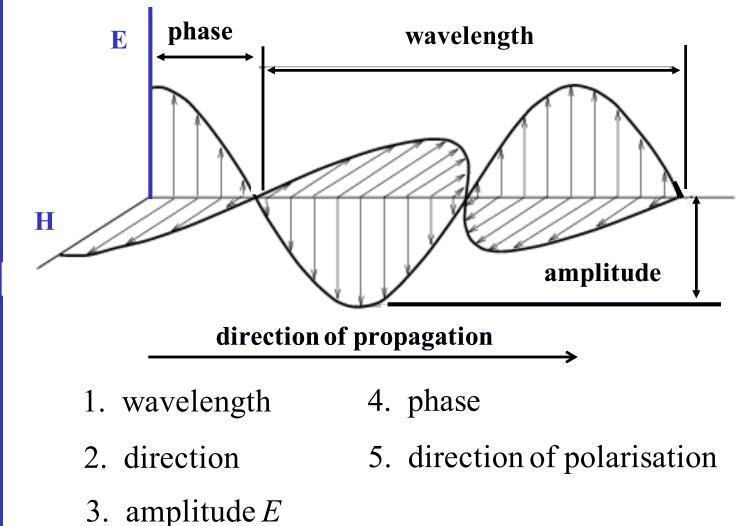


### Light as electromagnetic waves



## Light as electromagnetic waves

Self-sustaining exchange of electric and magnetic fields



### The spectrum

# Normal ambient light is a mixture of wavelengths, polarisation directions, and phases

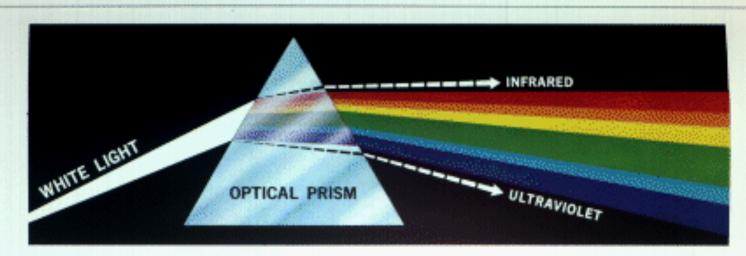
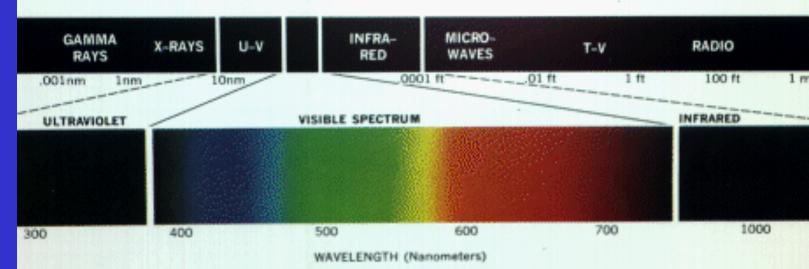
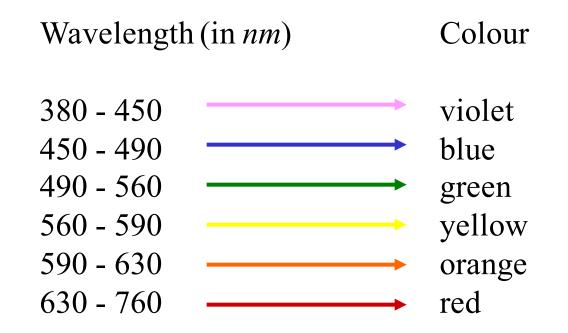


Plate I. Color spectrum seen by passing white light through a prism. (Courtesy of General Electric Co., Lamp Business Division.)





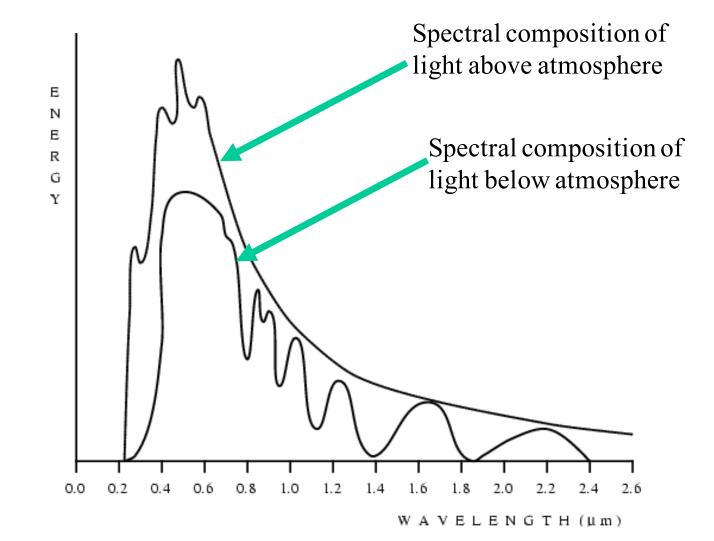
# The visible range



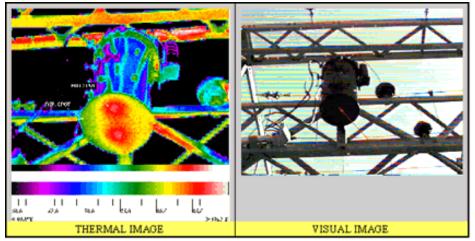
**NOTE** : Cameras may have different spectral sensitivities (i.e. also different from human vision)

### The solar spectrum

Peaks around 500nm, hence human sensitivity for that part of the spectrum



## Also cams for non-visible `light', e.g. infrared



Overheating of transformer coils, with far IR



Near infra-red (NIR) space image

NRG -> RGB for visualization (notice the strong reflection in the NIR for vegetation)

### Interactions with matter

four types :

### phenomenon

absorption scattering reflection refraction

### example

blue water blue sky, red sunset coloured ink dispersion by a prism

### + diffraction

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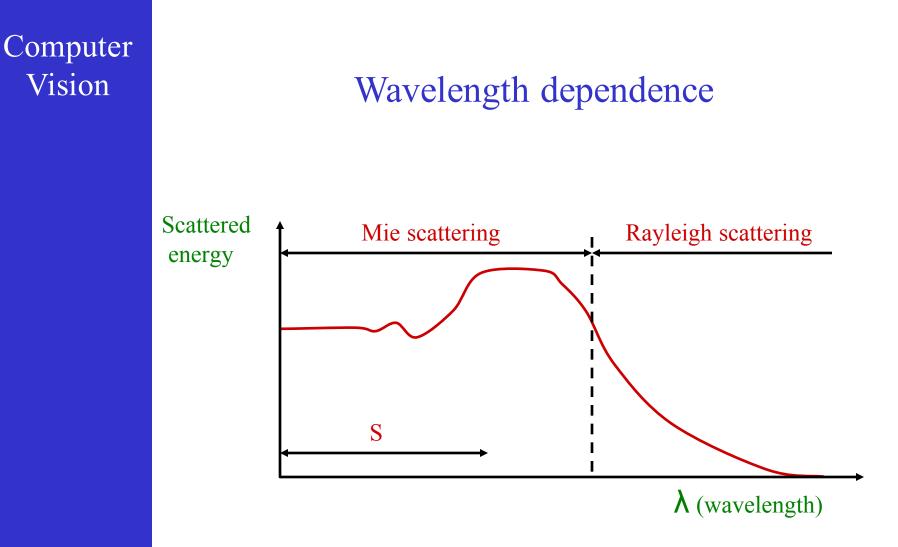
# Scattering

3 types depending on relative sizes of particles and wavelengths:

1. small particles: *Rayleigh* (strongly wavelength dependent)

2. comparable sizes: *Mie* (weakly wavelength dependent)

3. Large particles: *non-selective* (wavelength independent)



Less haze in the infrared (long wavelengths -> little scatter) Looking through clouds by radar (even longer wavelengths) NOTE: without scatter we would wander mainly in the dark

### Atmospheric showcase



### Tyndall effect (blue sky) Red, setting sun Grey clouds



Coloured cloud from volcanic eruption

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### example

blue water blue sky, red sunset coloured ink dispersion by a prism

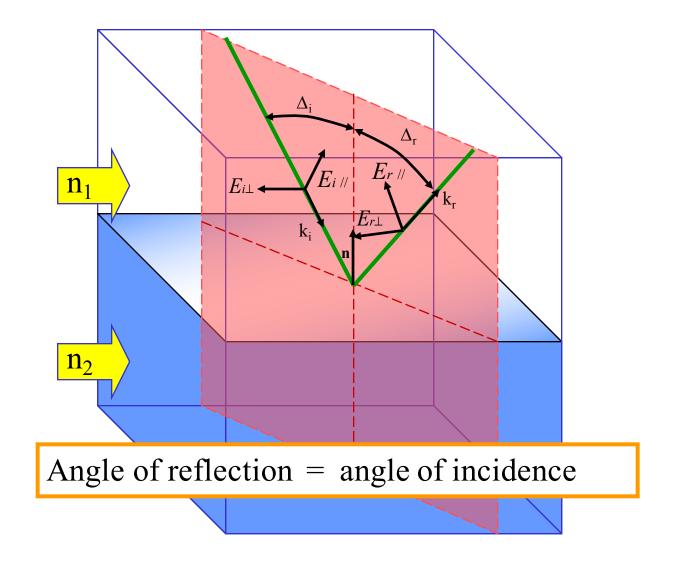
### + diffraction

# (Mirror) reflection

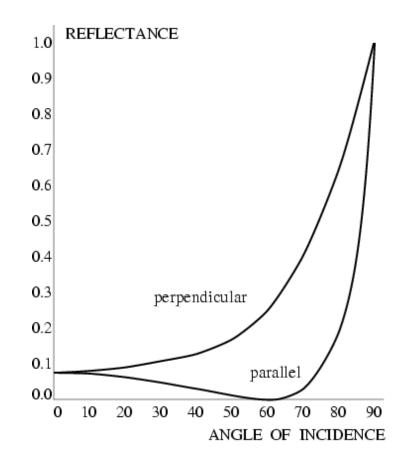




### Reflection



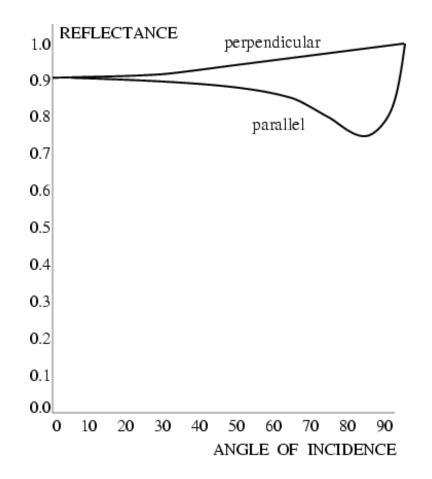
# Reflection : dielectric



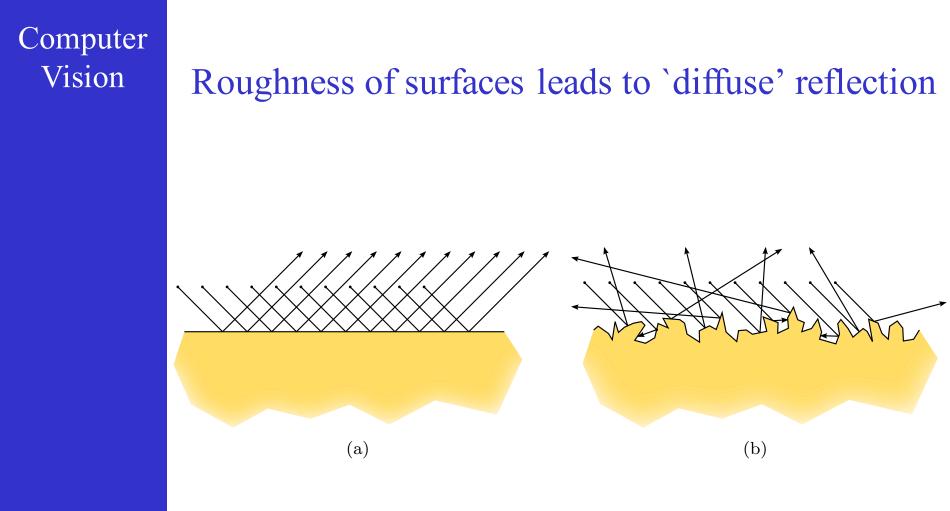
Polarizer at Brewster angle

Full reflection at grazing angles

# Reflection : conductor



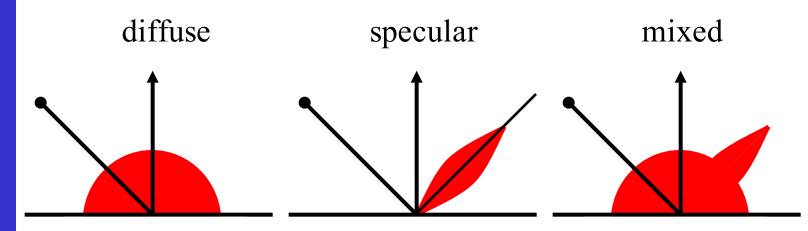
strong reflectors (under all angles) more or less preserve polarization



(a) Mirror or `specular' reflection, (b) diffuse reflection

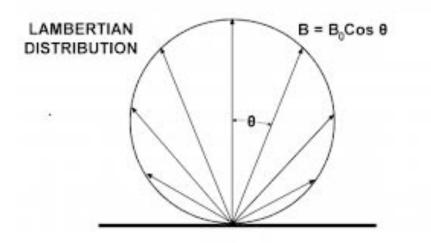
# ... and to mixed reflection for most real surfaces

### three types of reflection :



Note : Lambertian example of diffuse reflection

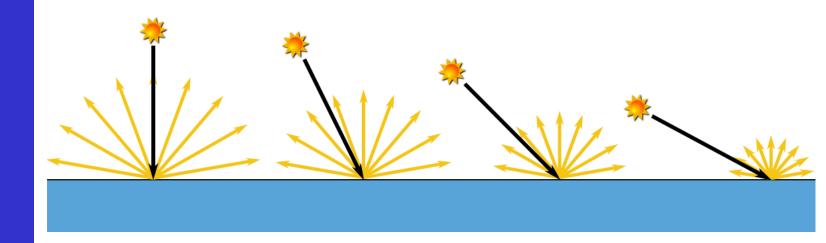
# Lambertian reflection



In case of Lambertian reflection the observed brightness only depends on the direction of the incoming light. If that's fixed the source radiates a flux that goes down with the viewing angle (with  $\cos \theta$ ).

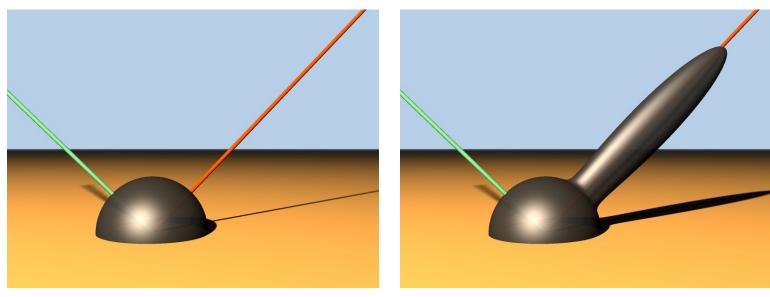
On the other hand, the same area on the camera corresponds to a larger surface area as one looks more obliquely  $(1/\cos\theta)$ And therefore receives light from a larger surface area. *These two factors cancel out, which lets the surface look equally bright from all viewing directions.* 

### Lambertian reflection



A Lambertian surface does *not* look equally bright independent of the lighting direction. Keeping the light intensity fixed but shining more obliquely lowers the amount of light hitting the surface per unit area. Thus there is less light to be reflected and, hence, the surface looks now equally darker from wherever we look.

# BRDFs specify reflection



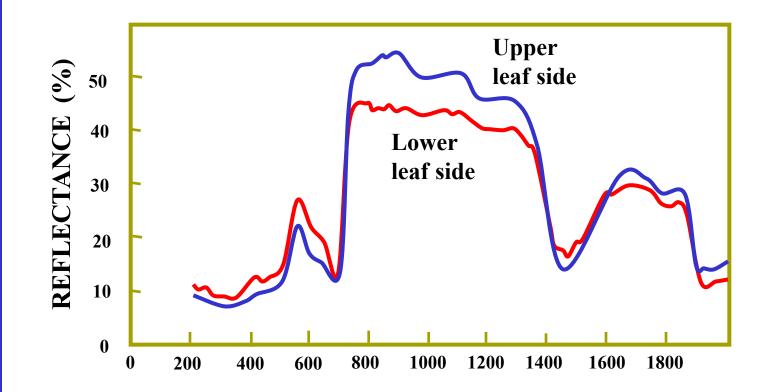
Diffuse example

Mixed example

A BRDF is a 4D function, specifying how much light is being reflected in a direction (red lines) for light coming in from a direction (green lines)

BRDFs are still hard to measure in practice

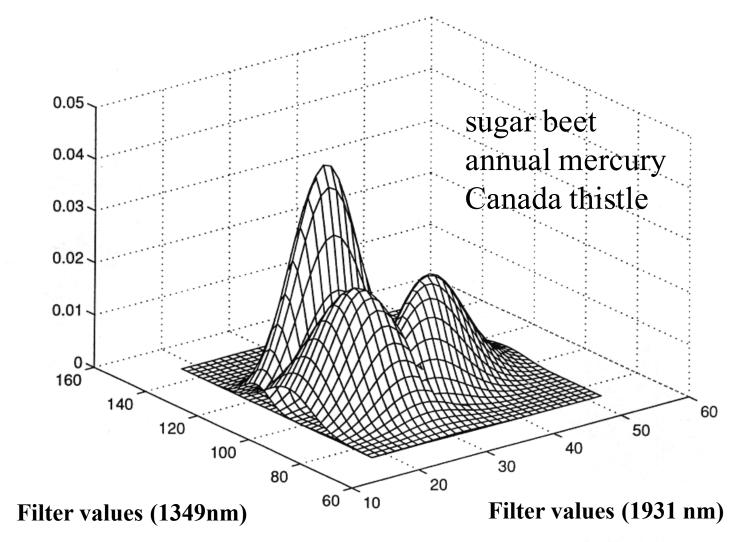
# Spectral reflectance e.g. vegetation



WAVELENGTH (µm)

# Spectral reflectance: ex. app weed detection for Reduced (selective) herbicide spraying

Gaussian probabilities for different classes



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# Ideally: spectral BRDF at all points known



### Interactions with matter

four types :

#### phenomenon

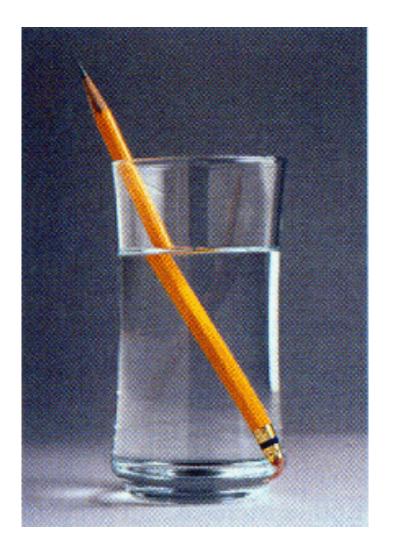
absorption scattering reflection **refraction** 

### example

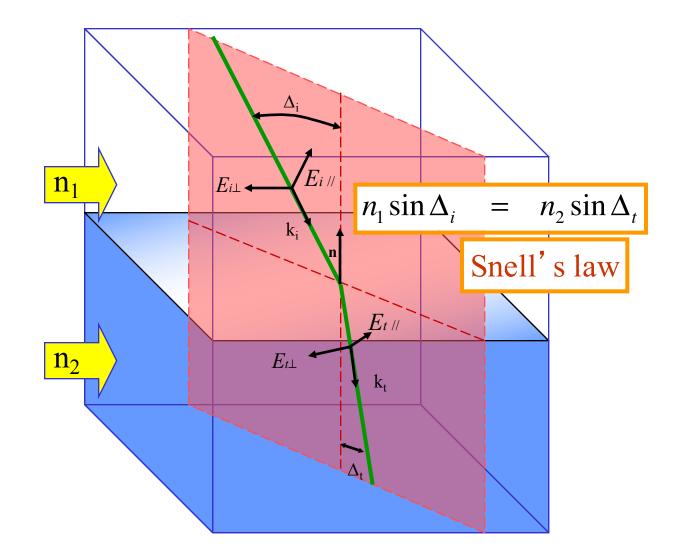
blue water blue sky, red sunset coloured ink dispersion by a prism

### + diffraction

### Refraction

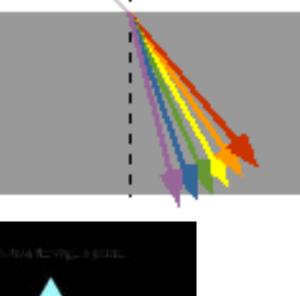


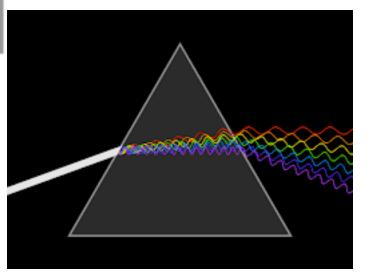
### Refraction



# Dispersion

Refraction is more complicated than mirror reflection: the path orientation of light rays is changed depending on material AND wavelength !!!





### Interactions with matter

four types :

#### phenomenon

absorption

scattering reflection refraction

### example

blue water blue sky, red sunset coloured ink dispersion by a prism

### + diffraction

### Absorption

Dissipation of wavelengths specific for the medium n (index of refraction) Based on resonance frequencies of molecules -> peaks Holes in sky light spectrum observed by Fraunhofer

# ... it needs brains ...

## Perception

Knowledge about HVS is important :

Image quality in interactive systems
 HVS limitations suggest applications
 Might suggest ways to go about

human vision is much more than a bottom-up process of subsequent signal processing steps

# Perception

Knowledge about HVS is important :

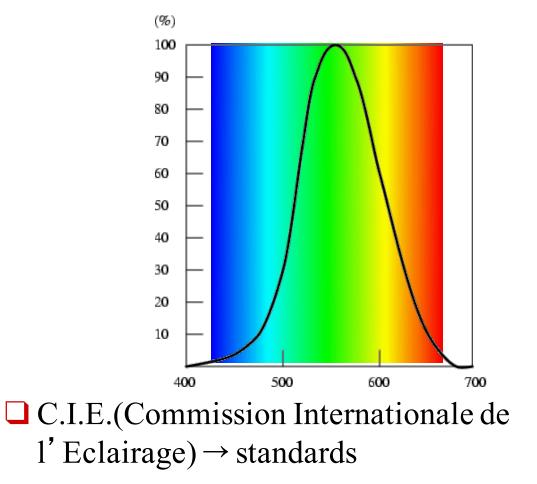
Image quality in interactive systems
 HVS limitations suggest applications
 Might suggest ways to go about

human vision is much more than a bottom-up process of subsequent signal processing steps

current AI or deep learning techniques bring us closer both methodologically and in terms of performance (but still lacks feedback)

# The perception of brightness

Luminous efficiency function : relates radiometry & photometry



# Link radiometry-photometry (Watt to lumen)

*Photometry:* subjective impressions *Radiometry:* objective, physical measurements

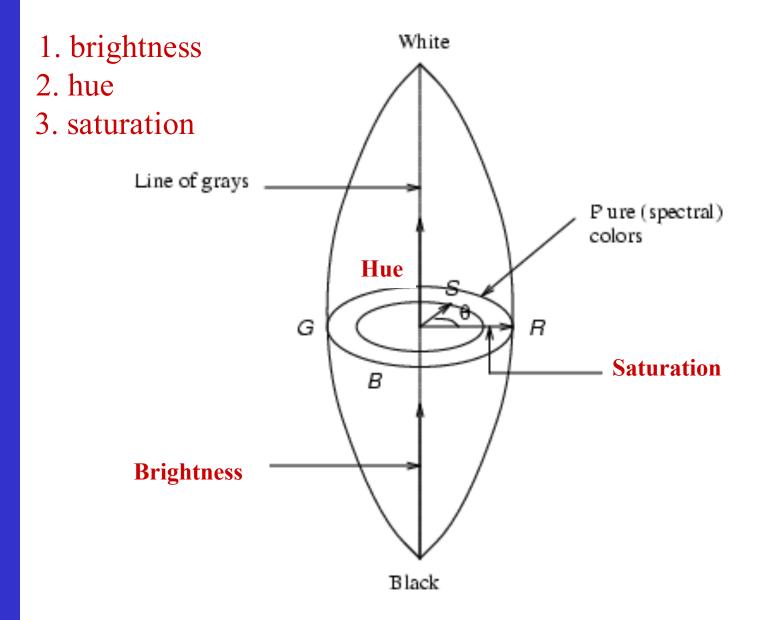
at 555 nm : 11m = 1/683 W = 1.46 mW

for light with spectral radiant flux  $c(\lambda)$ 

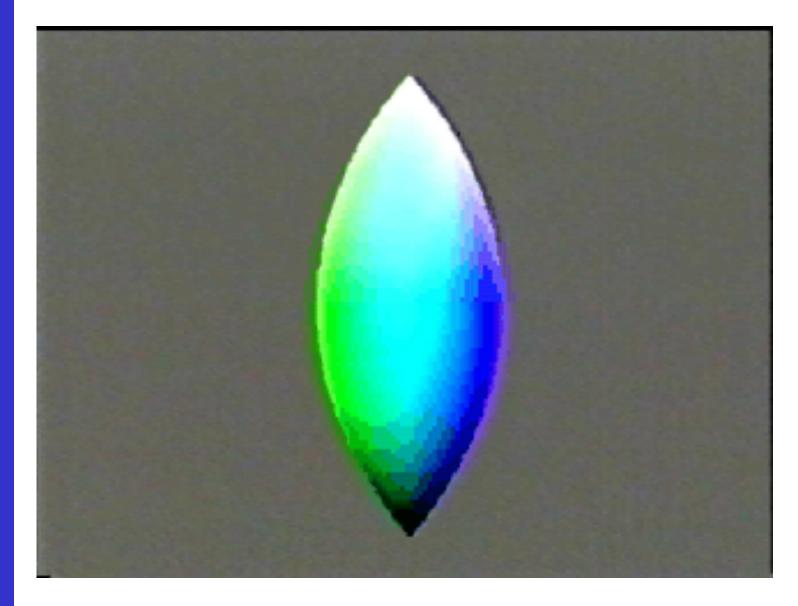
$$l = k \int_{\lambda=0}^{\infty} c(\lambda) v(\lambda) d\lambda$$

with k is 683 lumens/watt

# The perception of colour



# The perception of colour



# The history of colour science

#### Newton $\rightarrow$ spectrum



### Young $\rightarrow$ tristimulus model

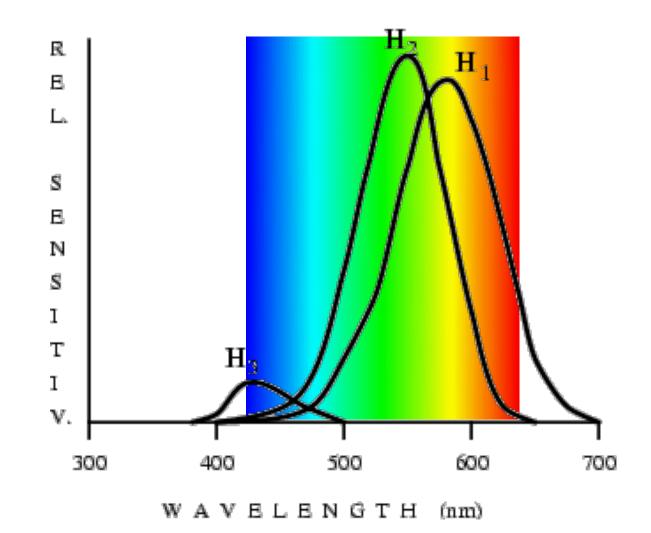


later : physiological underpinning :

3 cone types

### The retinal cones

### 3 types : blue, green, yellow-green



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# The perception of contrast

### *relative luminance* (context) is the key

# The perception of contrast

### *relative luminance* (context) is the key



# The perception of contrast

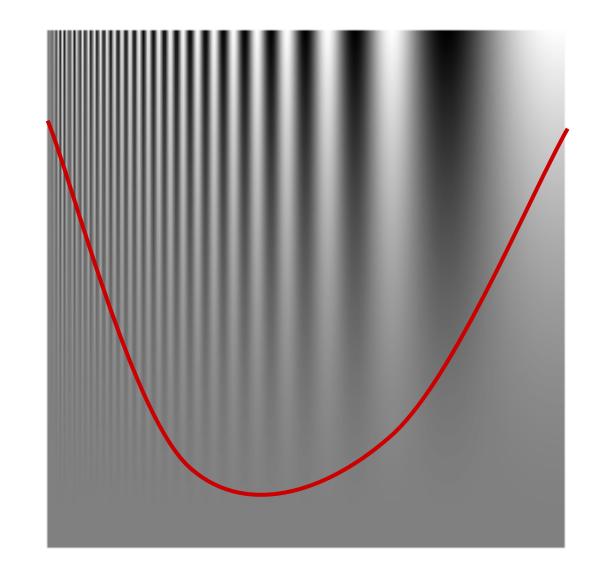
### *relative luminance* (context) is the key



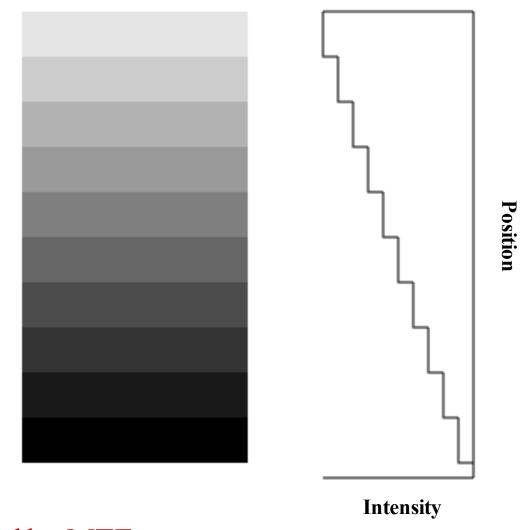
Weber's law: 
$$\frac{\Delta l}{l_{back}} \approx {}^{c}W$$

Relevant for *just noticeable differences (jnds)* and beyond

# The perception of spatial frequency



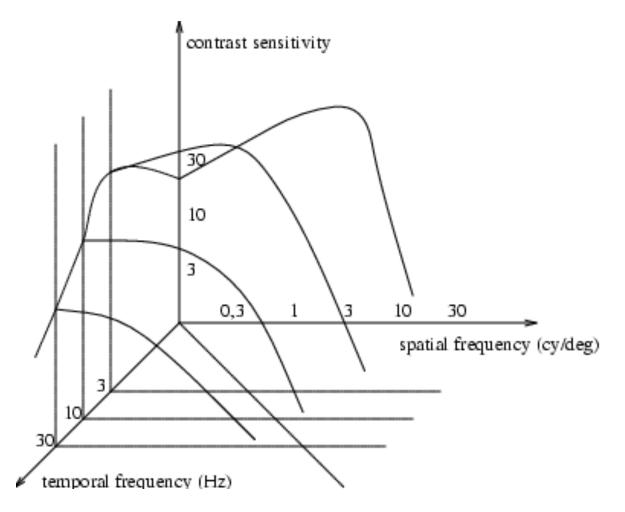
# Mach bands



explained by MTF

→

# The perception of spatio-temporal frequency



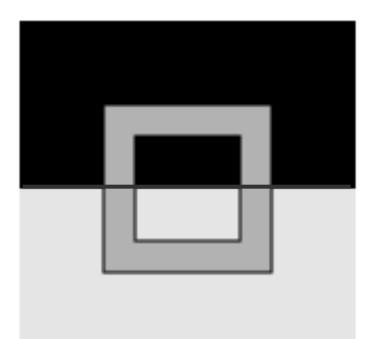
bandpass along both axes

Computer

Vision

# You ain't seen nothing yet

# there is literally more than meets the eye, i.c. a lot of massively parallel processing





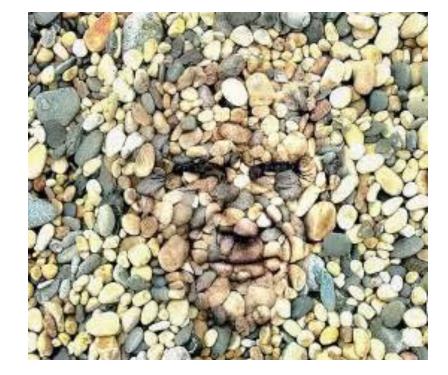
Kanisza (filling in)

# [] [] (T

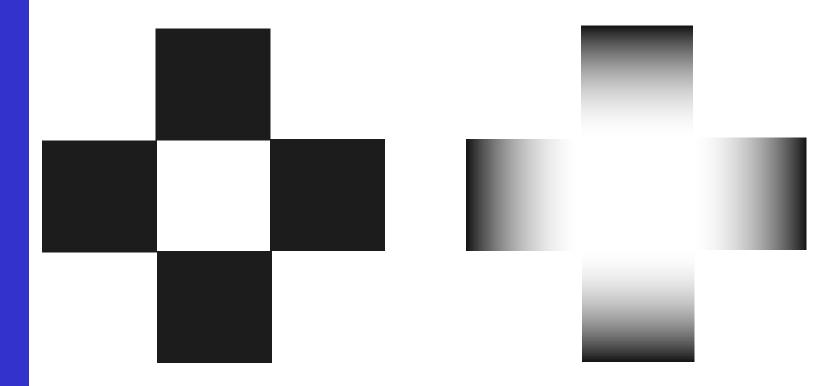
Fill-in : averaging of perceived contrast at edges over regions possibly obtained via extrapolation of the edges



# Our visual system eagerly looks for known patte



## The perception of contrast

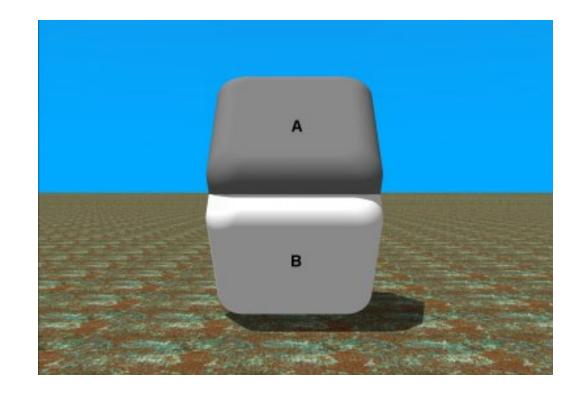


A

В

→

## The perception of intensity



## The role of context

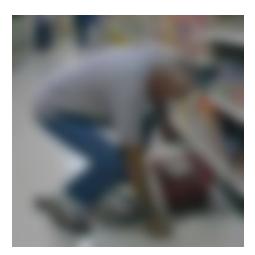




Human vision: Biederman, Bar & Ullman, Palmer,

. . .





## The role of context





## All encircled patterns are identical:





# The role of context



# The role of context



# The role of context

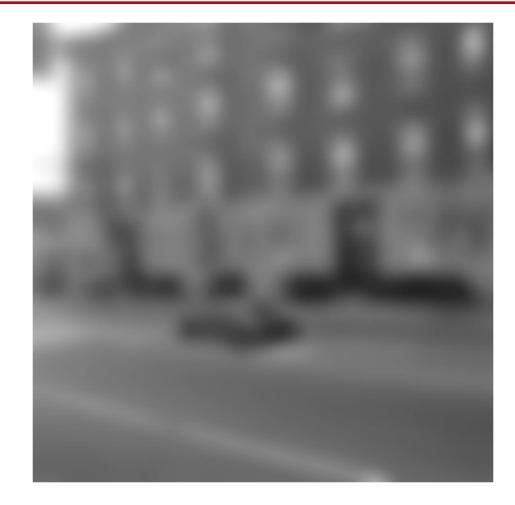


# The role of context



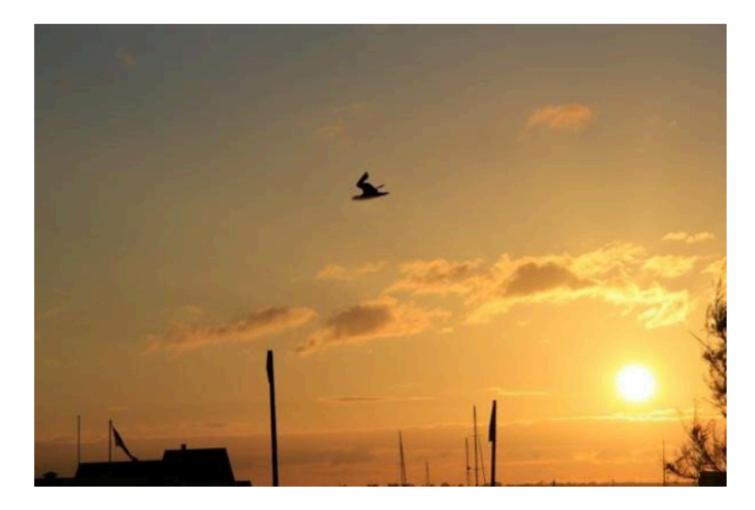
The role of context

human vision is much more than a bottom-up process of subsequent signal processing steps.



Car?

## The role of context



A bird, or a giant rabbit flying on a surf board?

Want to know more?

## Foundations and Trends in

## **Computer Graphics and Vision**

Publishes tutorials / reviews Freely available for ETH students

http://www.nowpublishers.com

