

# Image Formation

CSE P576

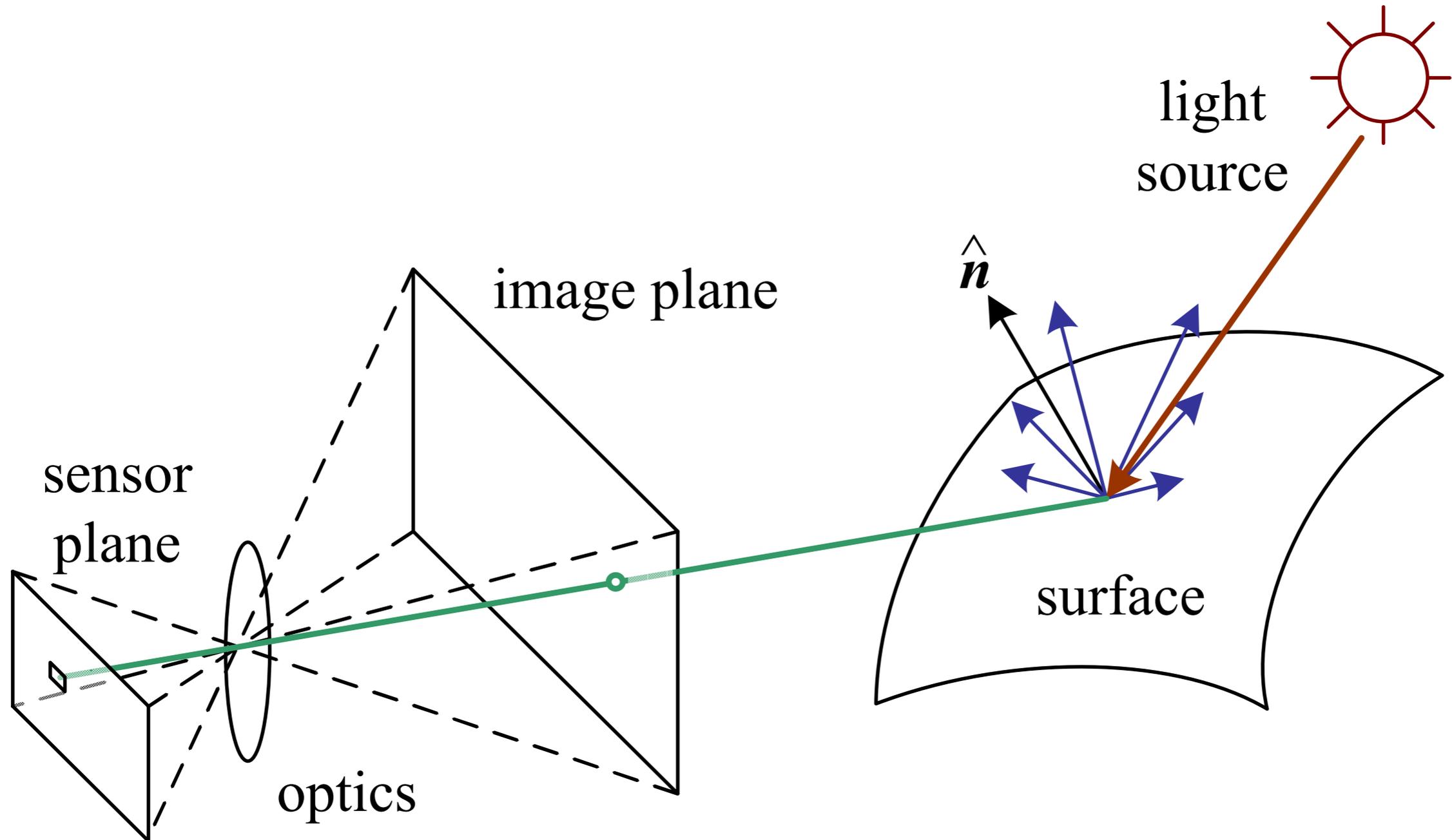
Dr. Matthew Brown

# Image Formation

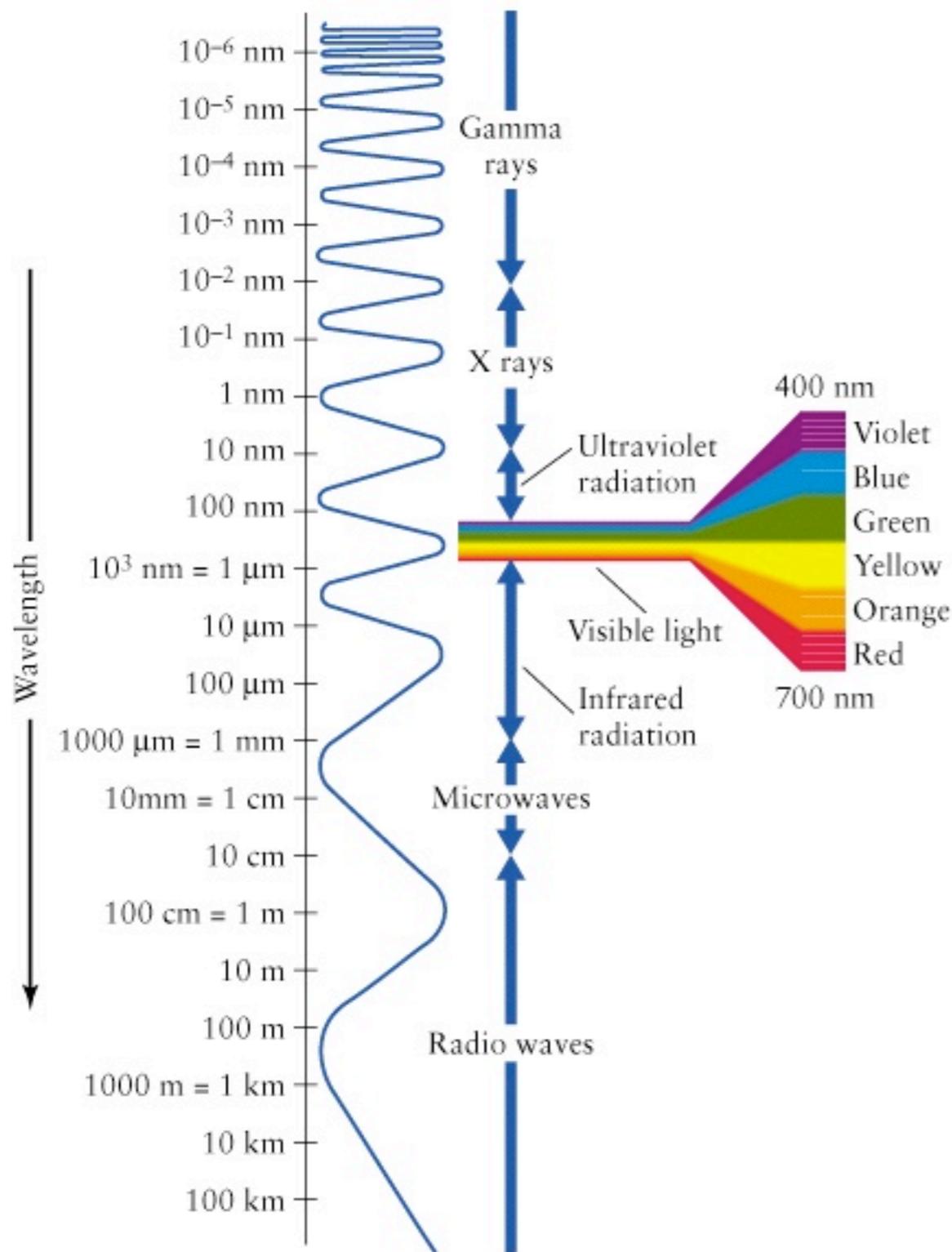
- Light, Optics, Sensing
- The Digital Camera

[ Szeliski Chapter 2 ]

# Image Formation



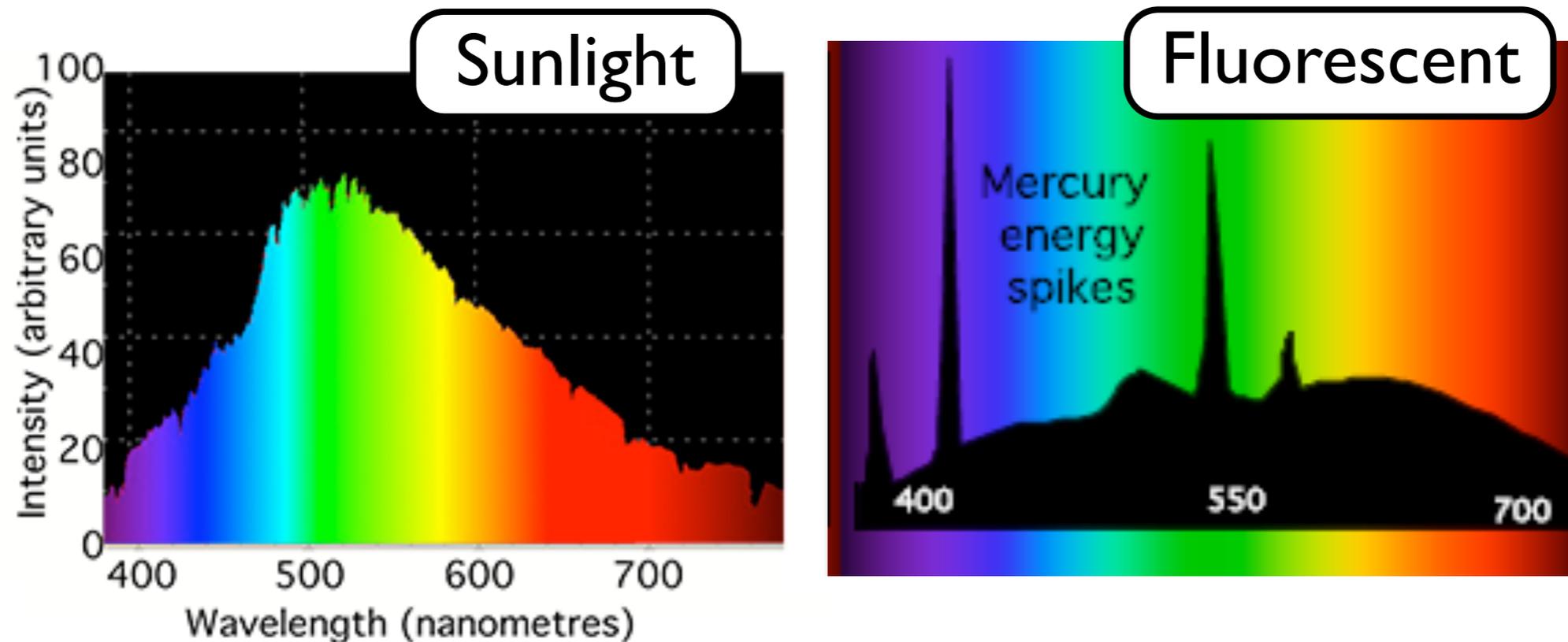
# Light and Colour



- Light is electromagnetic radiation in the 400-700nm band
- This is the peak in the spectrum of sunlight passing through the atmosphere
- Newton's Prism experiment showed that white light is composed of all frequencies
- Black is the absence of light!

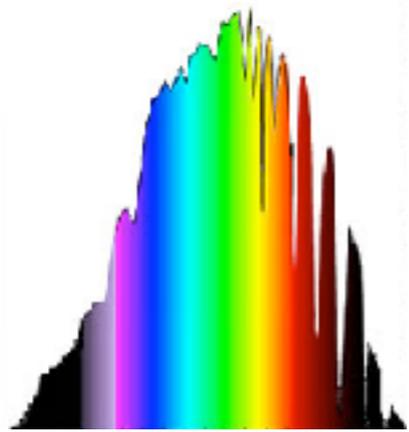
# Spectral Power Distribution

- The spectral distribution of energy in a light ray determines its colour

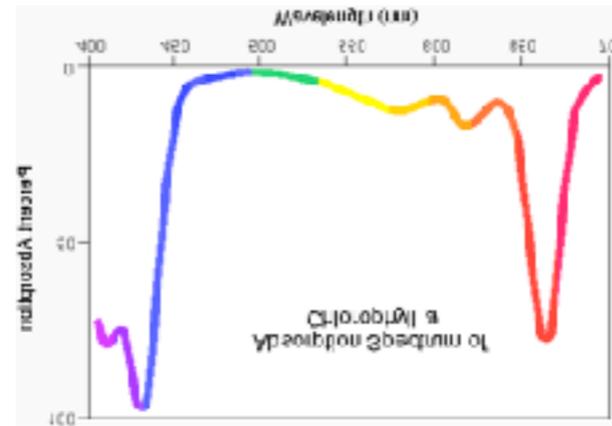
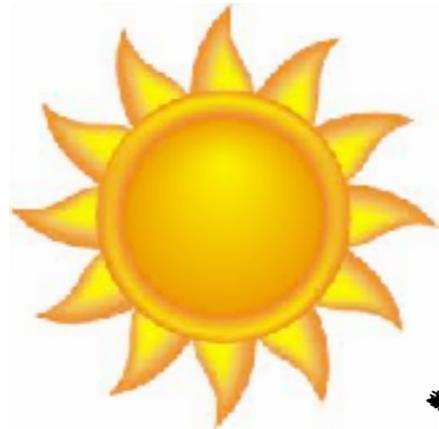


- Surfaces reflect light energy according to a spectral distribution as well
- The combination of incident spectra and reflectance spectra determine the light colour

# Spectral Reflectance Example



$$E(\lambda)$$



$$S(\lambda)$$



$$E(\lambda)S(\lambda)$$



# Surface Reflectance

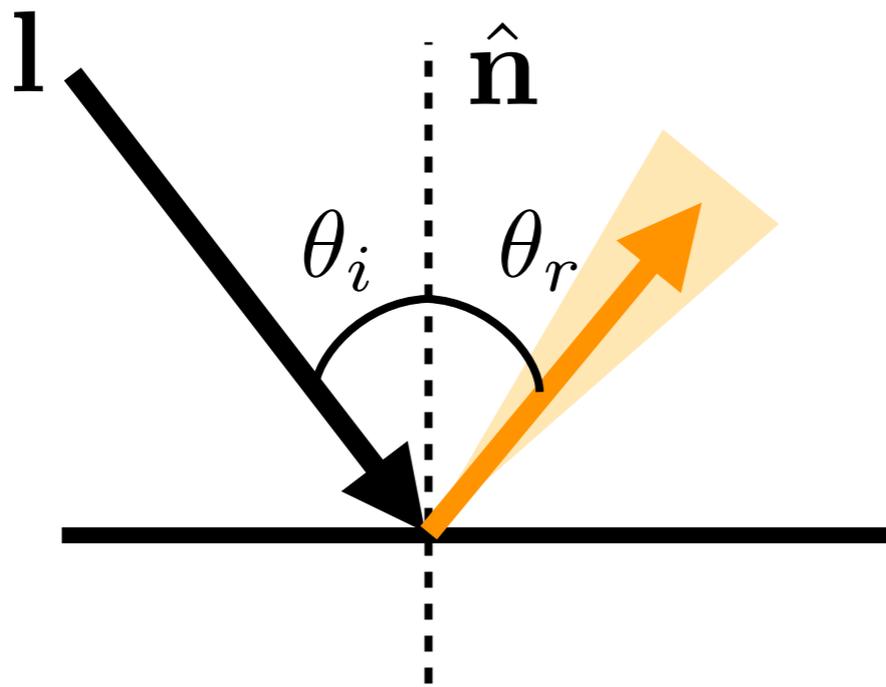
- Reflected intensity also depends on geometry: surface orientation, viewer position, shadows, etc.



It also depends on surface properties, e.g., diffuse or specular

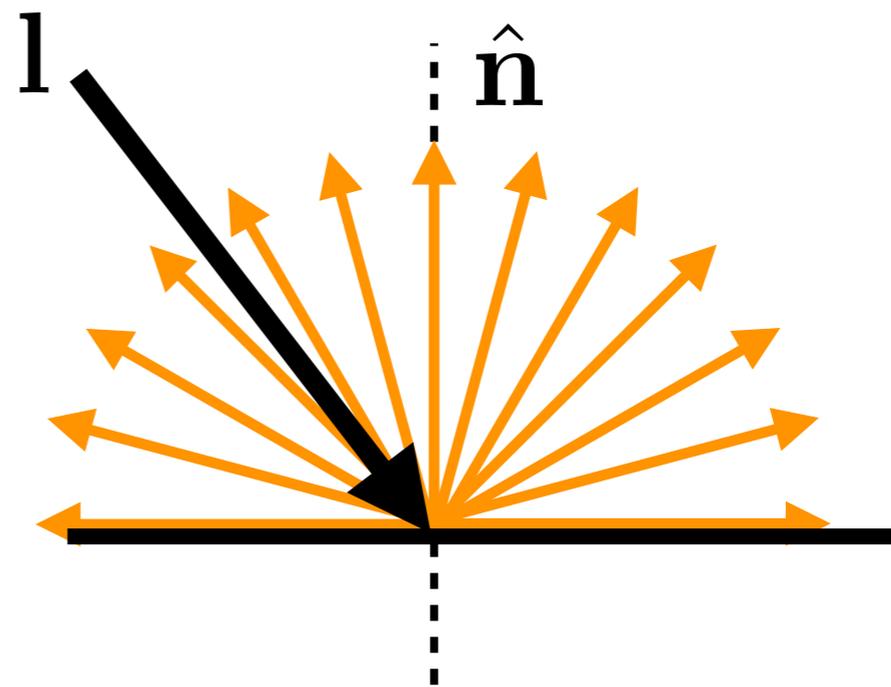
# Diffuse and Specular

- A pure **mirror** reflects light along a line symmetrical about the surface normal
- A pure **diffuse** surface scatters light equally in all directions



Pure Mirror Reflection

$$\theta_i = \theta_r$$



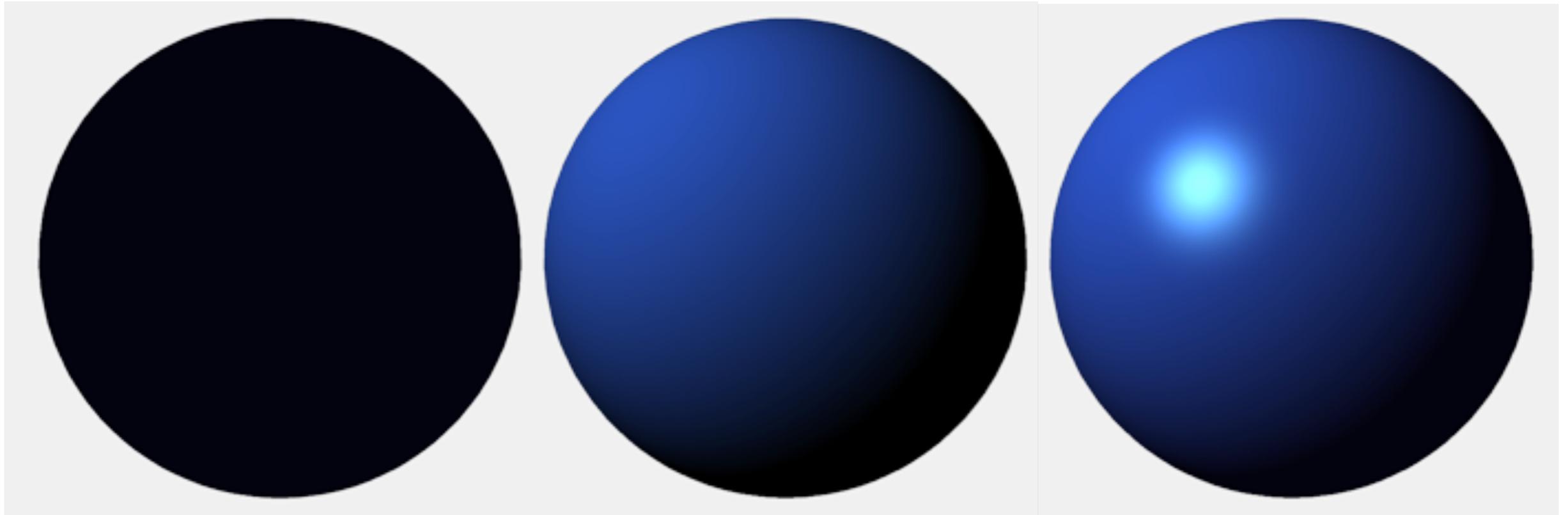
Lambertian Reflection

(Diffuse)

**Specular** surfaces directly reflect over a small angle

# Diffuse and Specular

- A sphere lit with ambient, +diffuse, +specular reflectance



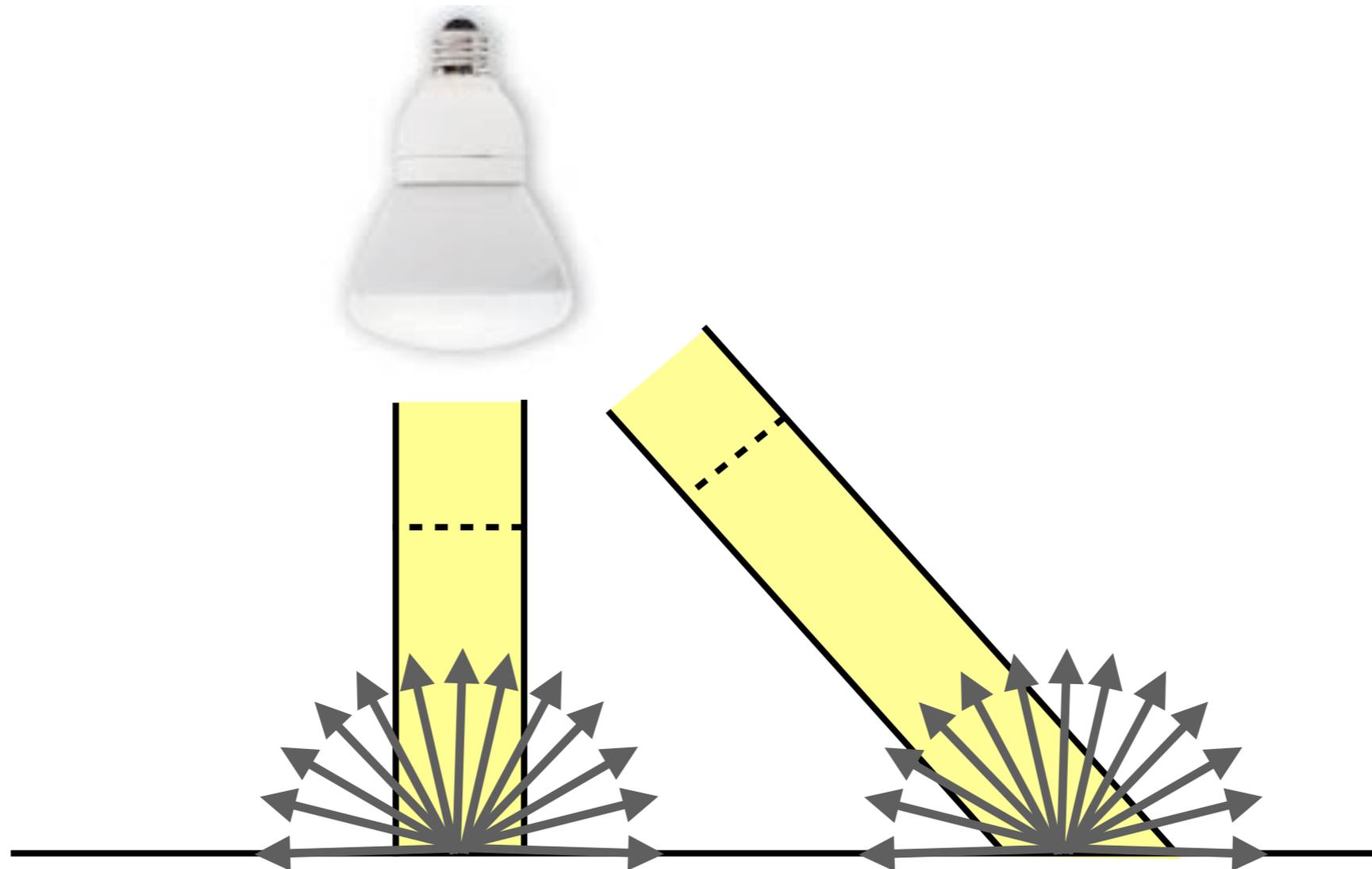
Ambient

+Diffuse

+Specular

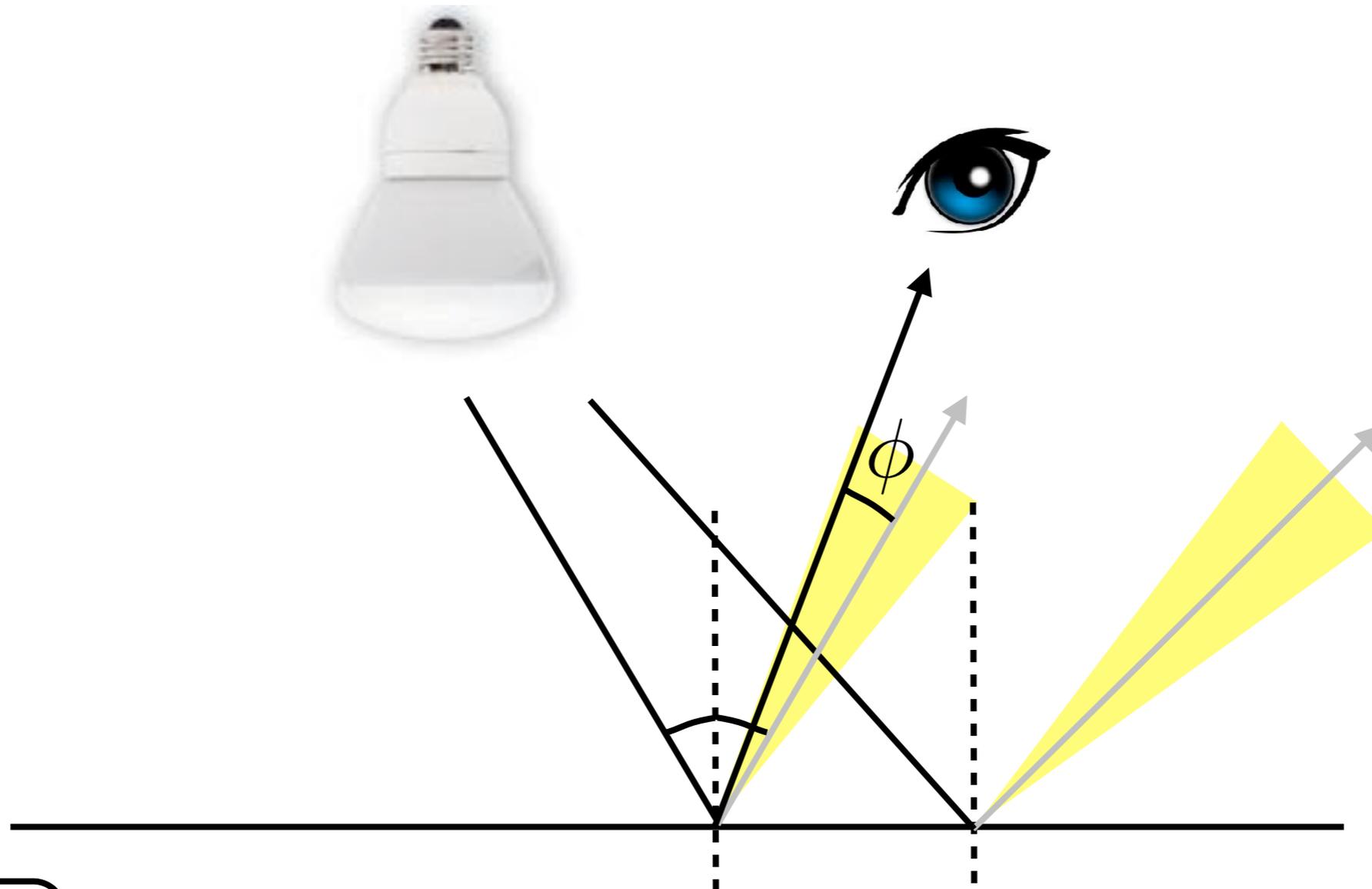
# Diffuse Reflection

- Light is reflected equally in all directions (Lambertian surface)
- But the amount of light reaching unit surface area depends on the angle between the light and the surface...



# Specular Reflection

- Light reflected strongly around the mirror reflection direction
- Intensity depends on viewer position

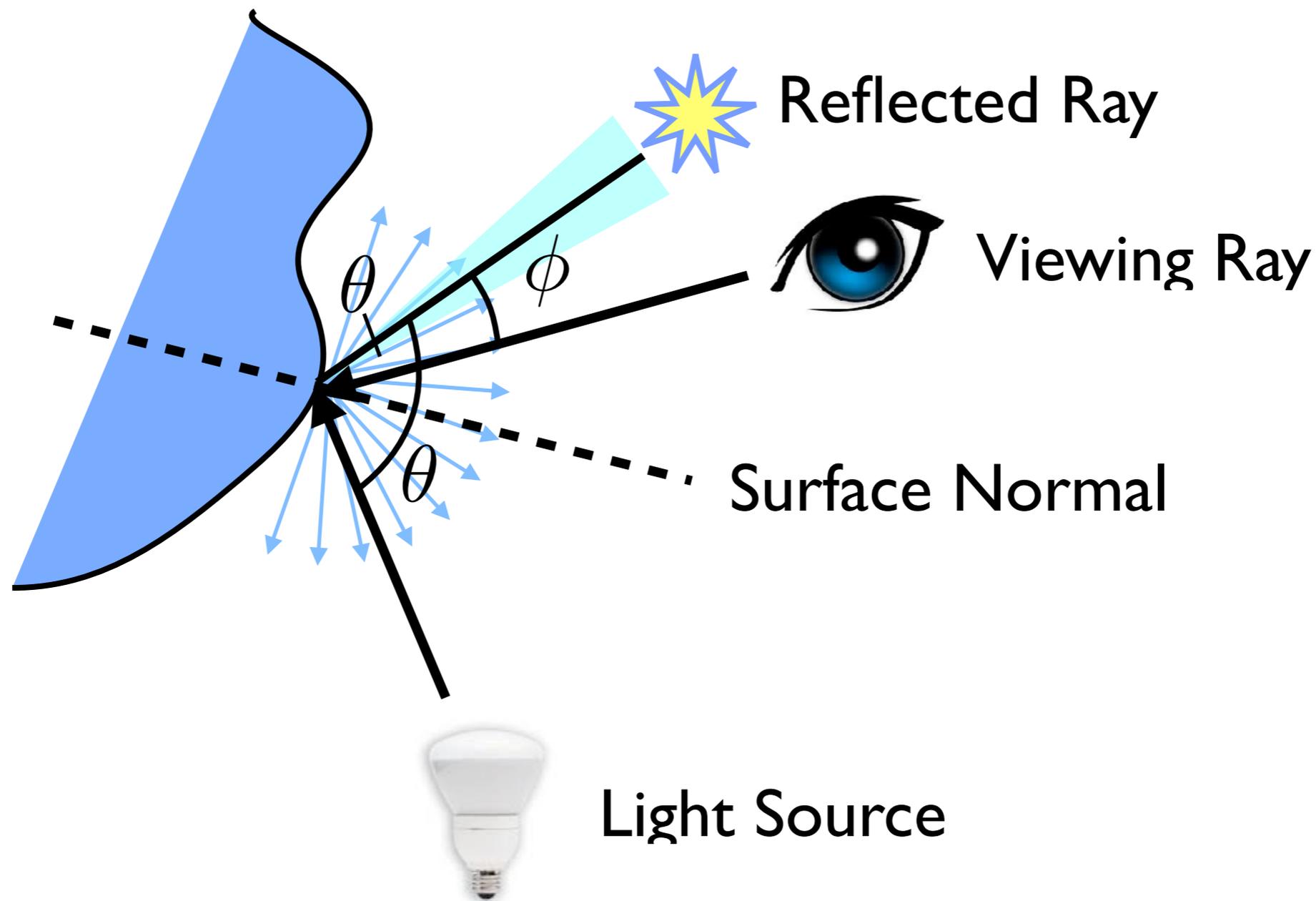


1.2

# Phong Illumination Model

- Includes ambient, diffuse and specular reflection

$$I = k_a i_a + k_d i_d \cos \theta + k_s i_s \cos^\alpha \phi$$



# Reflectance in Vision







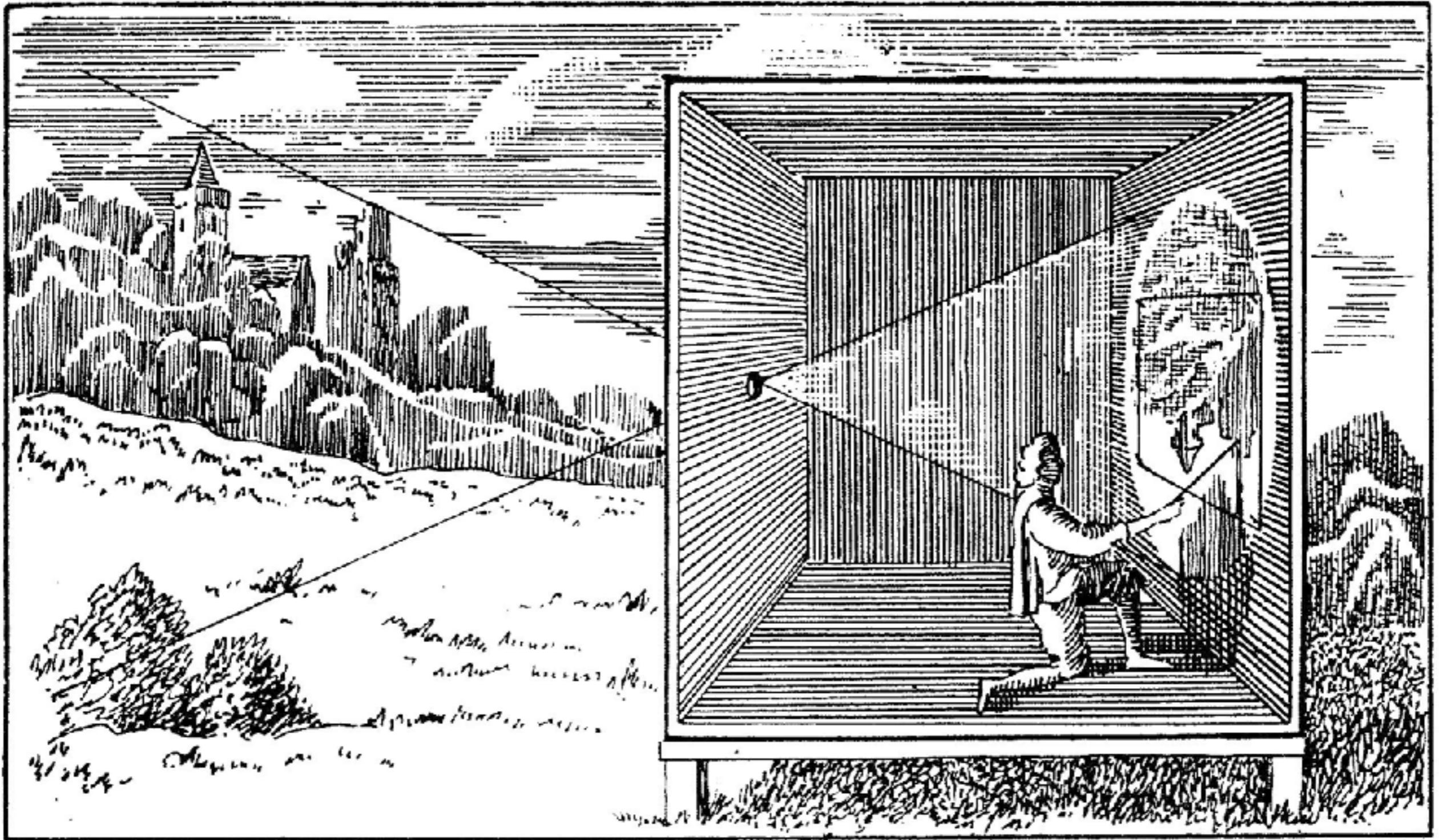
# Reflectance in Vision

- More complex models than Phong are possible with reflected intensity at a given ray an arbitrary function of the surface geometry and lights, see Szeliski 2.2.2 (BRDFs)
- For Computer Vision, understanding reflection can help us to infer shape, e.g., **shape from shading** and **photometric stereo**, we will revisit this later in the course



[ Hertzmann  
Seitz 2003 ]

# Optics



Camera Obscura = “dark room”

# Clifton Observatory

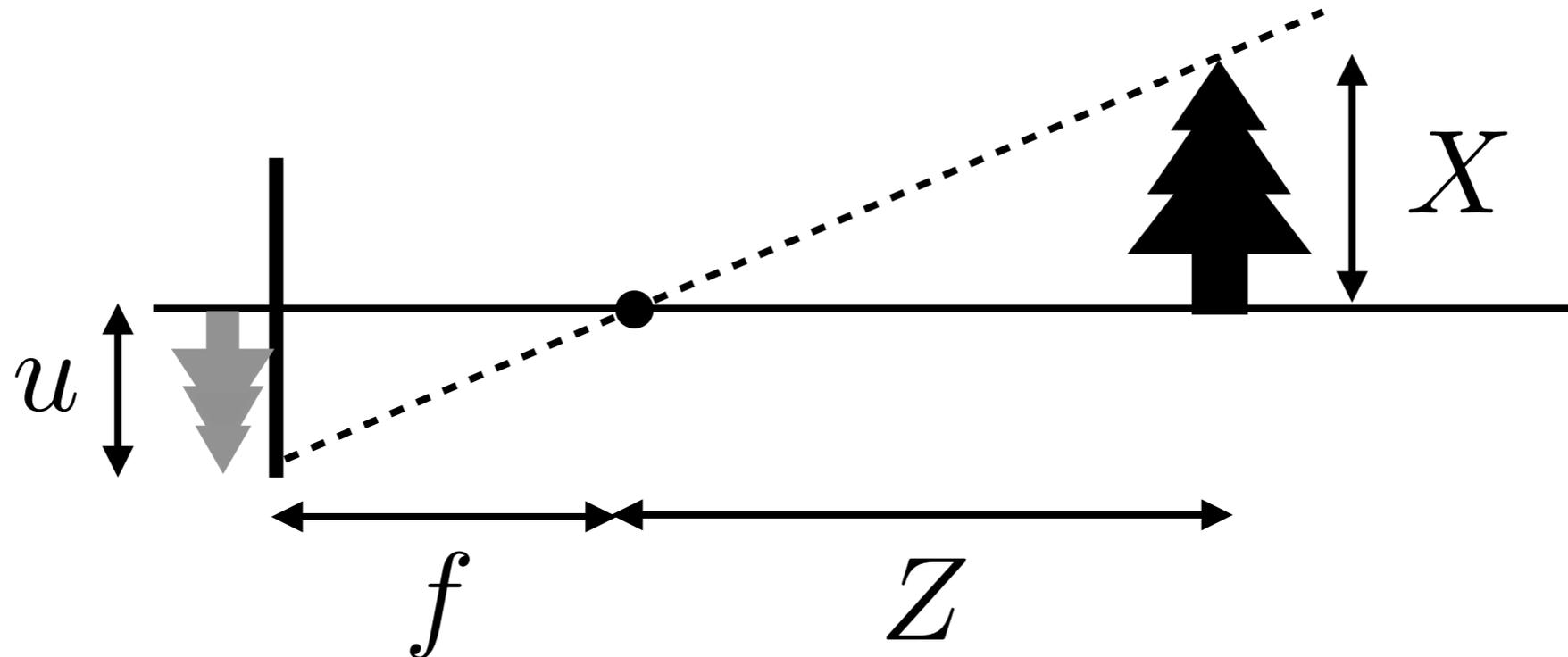


[ P. Chapman ]

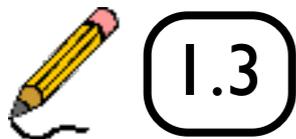
A working camera obscura open to the public

# Pinhole Camera

- All rays pass through a single point (the pinhole)

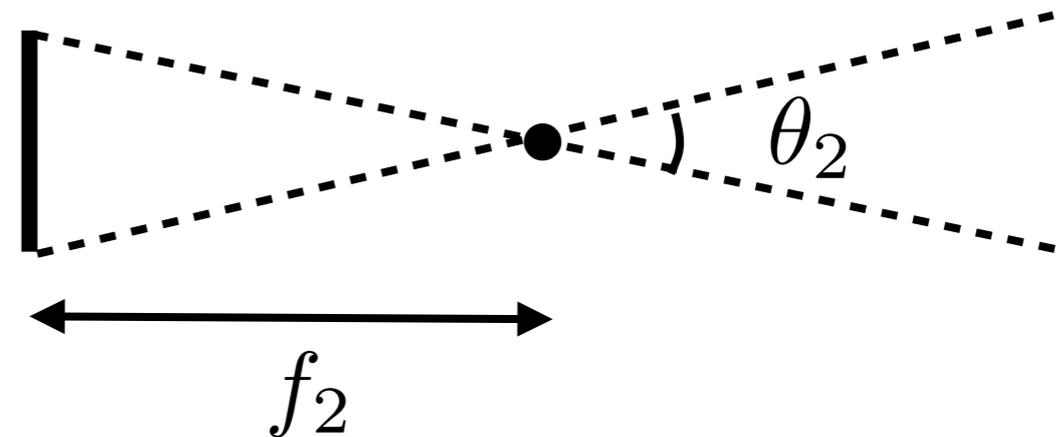
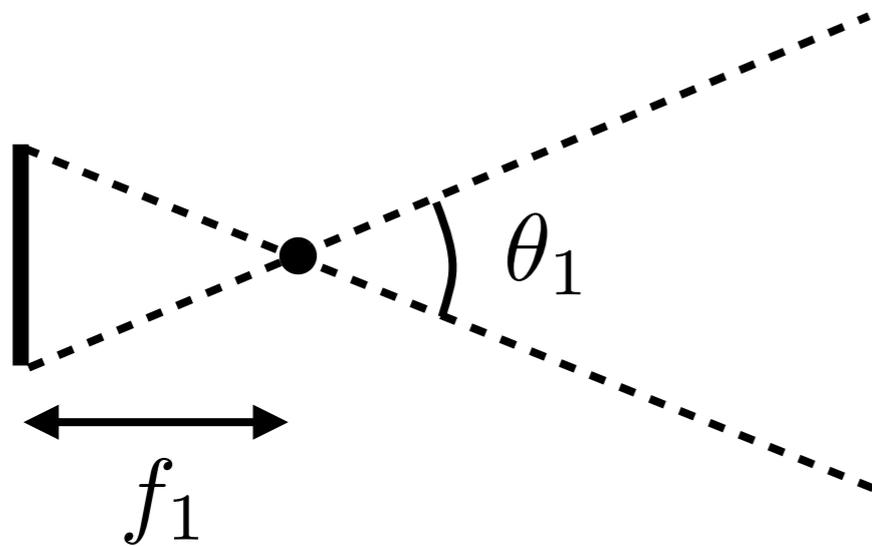


- Similar triangles



# Focal Length

- For a fixed sensor size, focal length determines the field of view (fov)



**Q:** What is the field of view of a full frame (35mm) camera with 50mm lens? 100mm lens?



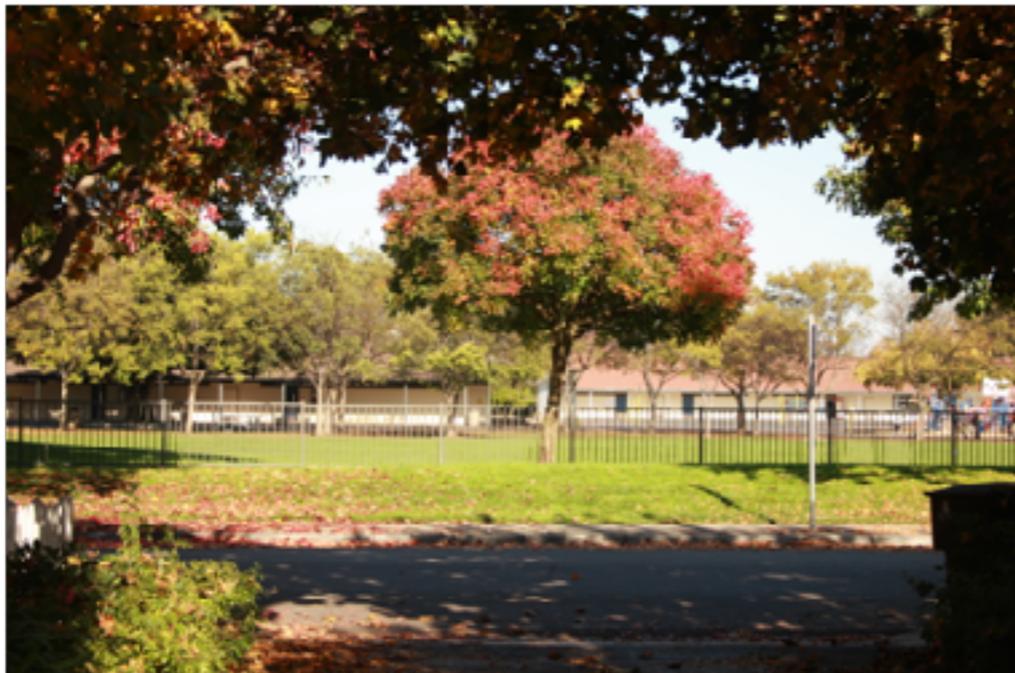
# Focal Length



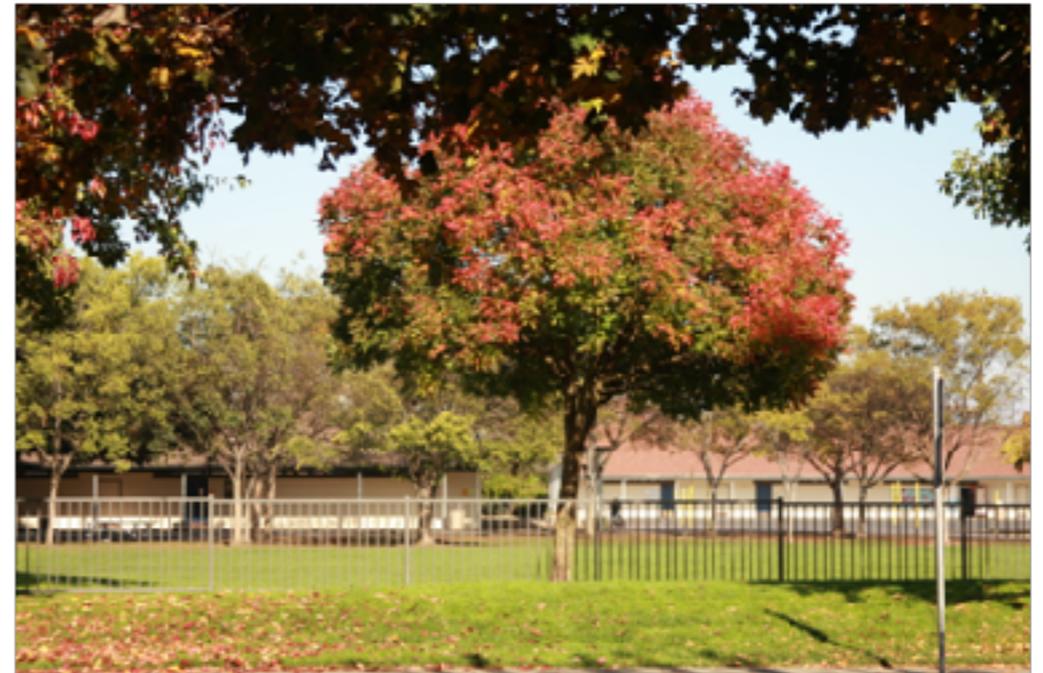
28 mm



35 mm



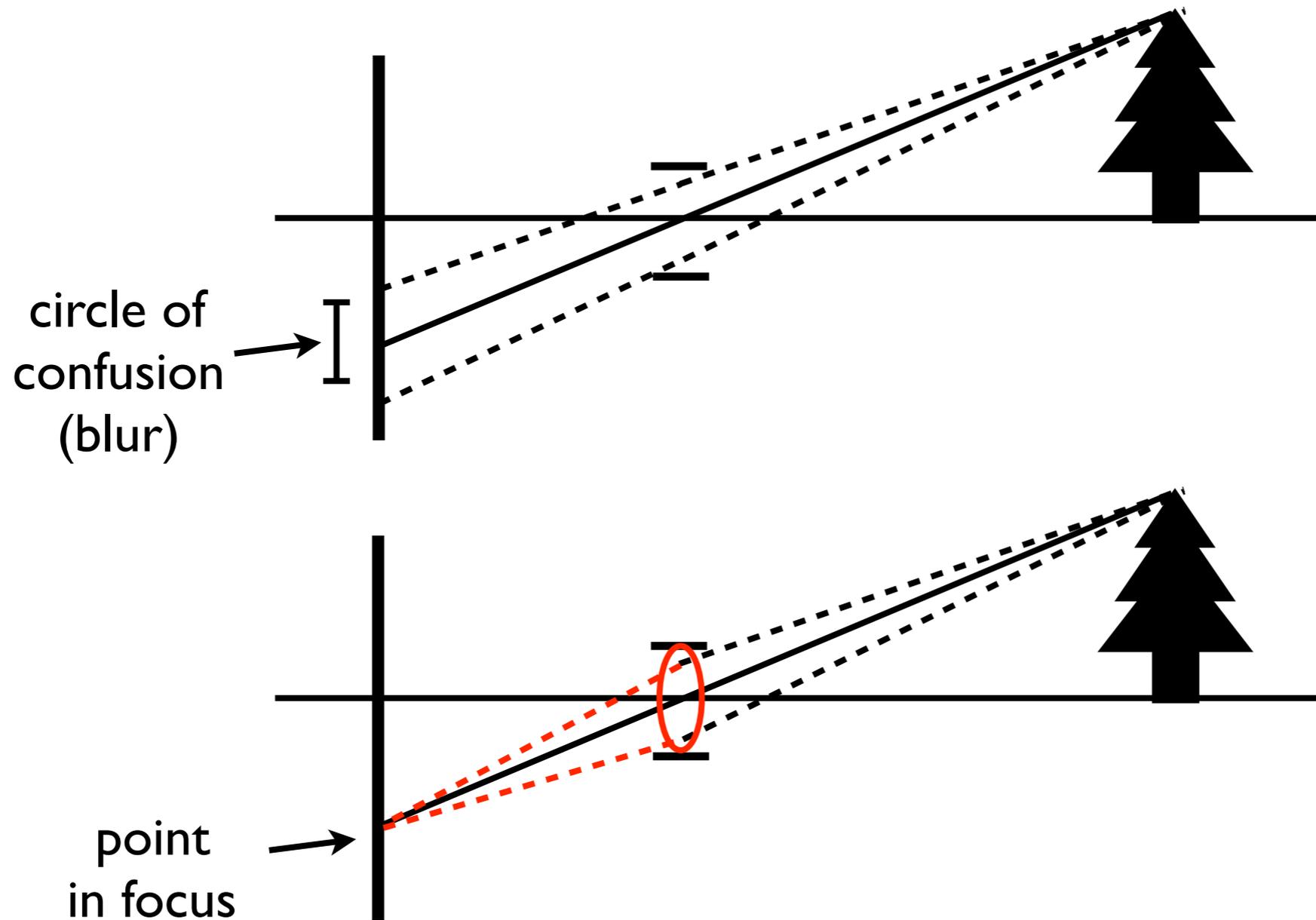
50 mm



70 mm

# Finite Aperture

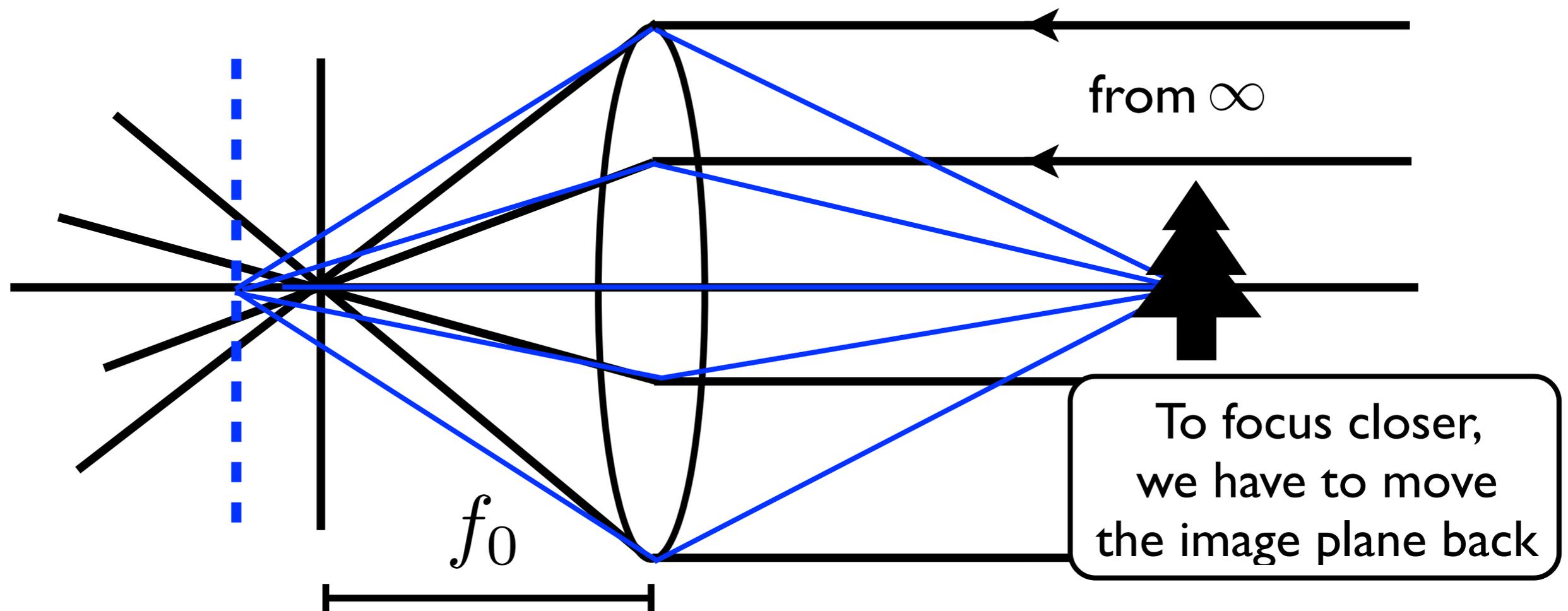
- A real camera must have a finite aperture to get enough light, but this causes **blur** in the image



Solution: use a **lens** to focus light onto the image plane

# Lens Basics

- A lens focuses rays from infinity at the focal length of the lens
- Points passing through the centre of the lens are not bent



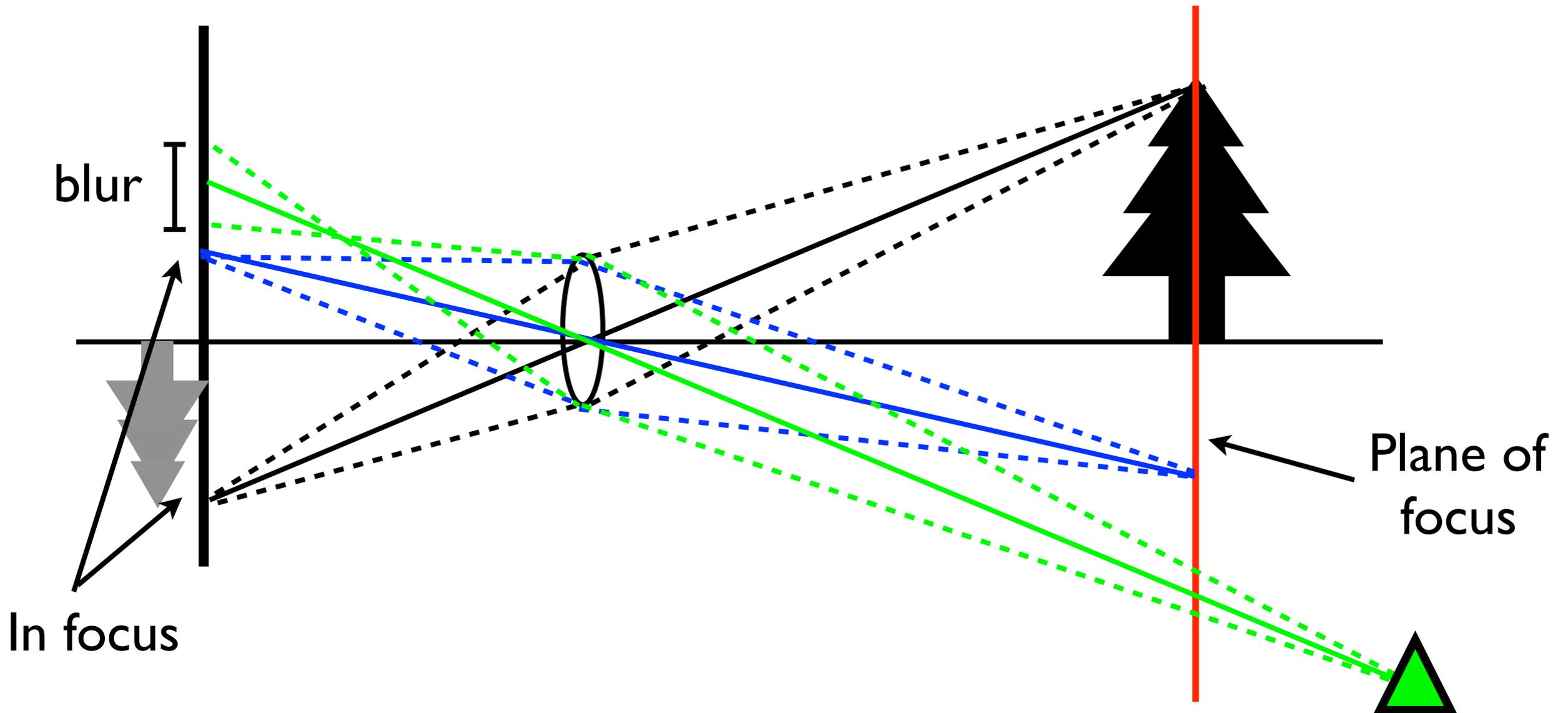
- We can use these 2 properties to find the lens equation



1.5

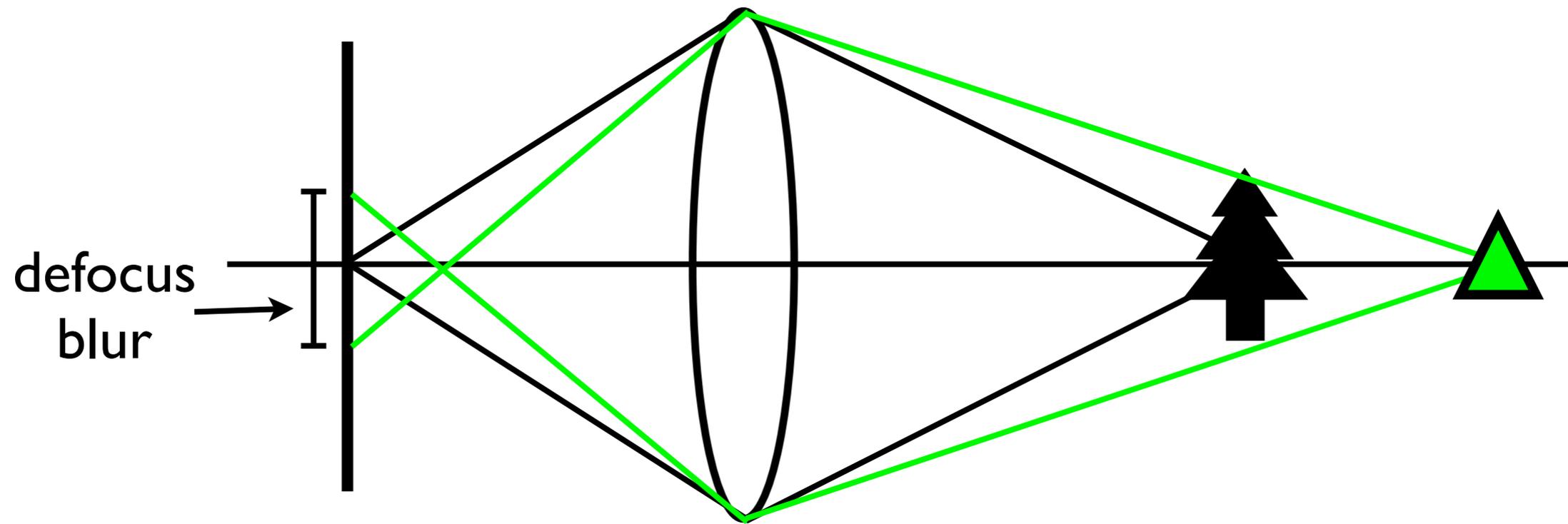
# Lens Basics

- Note that lenses focus all rays from a **plane** in the world

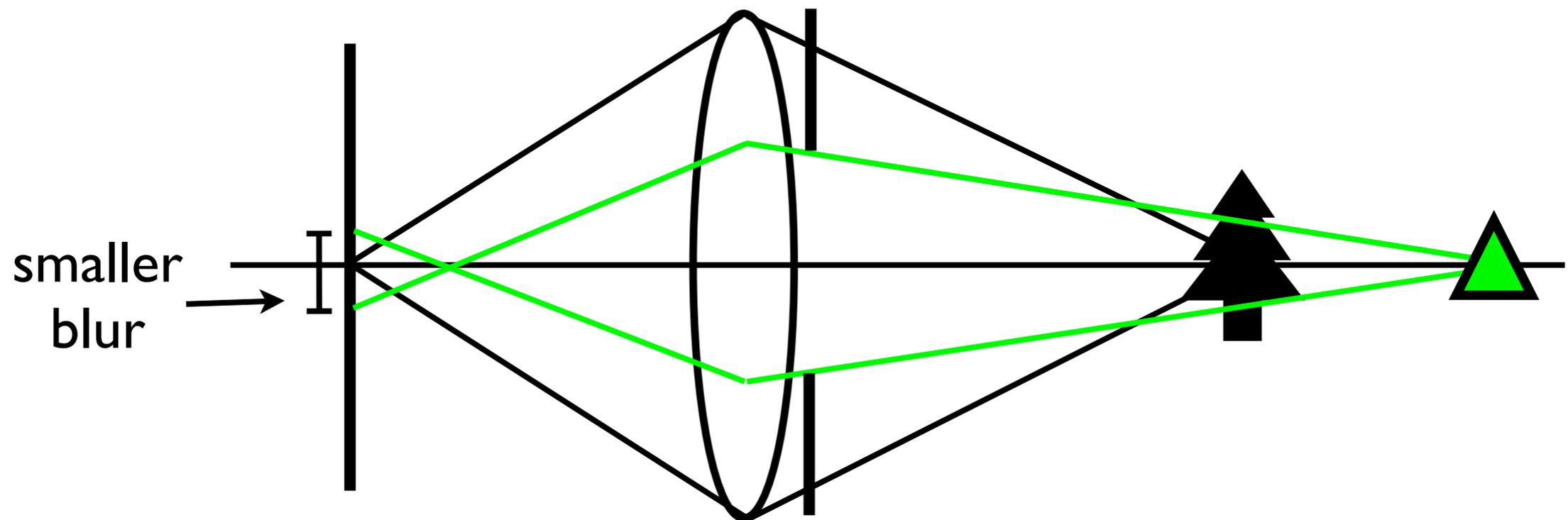


- Objects off the plane are blurred depending on distance

# Effect of Aperture



Smaller aperture  $\Rightarrow$  smaller blur, larger **depth of field**



# Depth of Field

- Photographers use large apertures to give small depth of field



Aperture size =  $f/N$ ,  $\Rightarrow$  large  $N$  = small aperture

# Shutter Speed



# Real Lenses

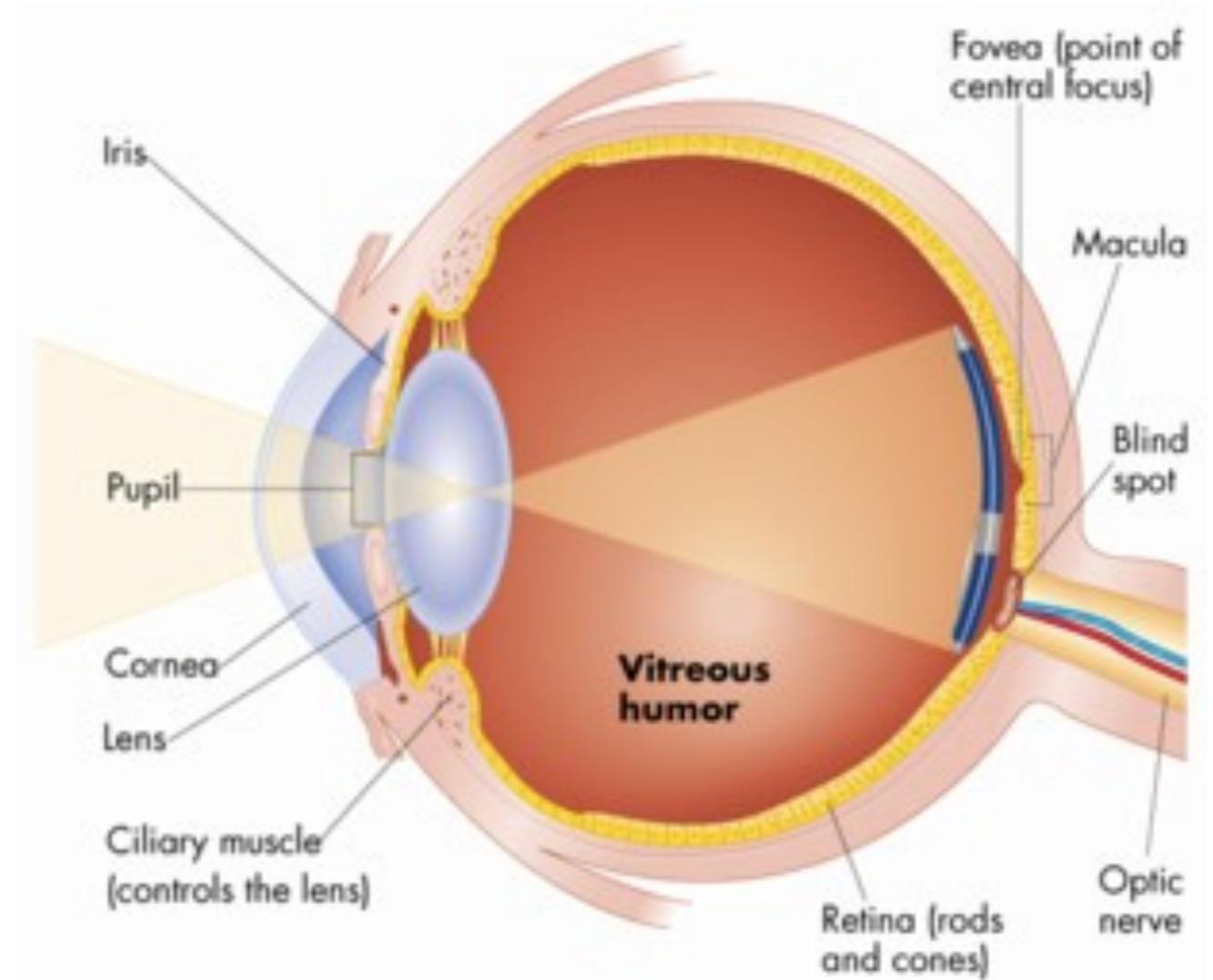


- Multiple stages of positive and negative elements with differing refractive indices
- Deal with issues such as chromatic aberration (different colours bent by different amounts), vignetting (light fall off at image edge) and sharp imaging across the zoom range

# Sensors

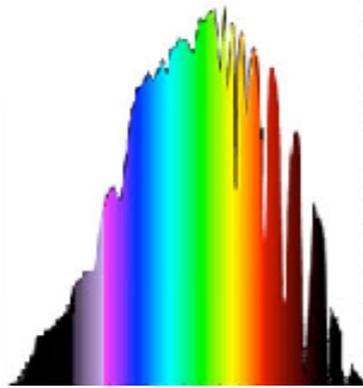


CMOS (or CCD)

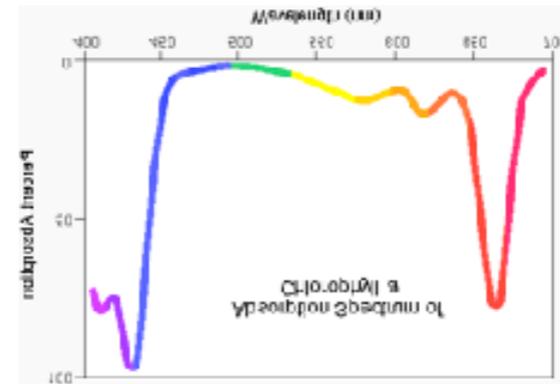
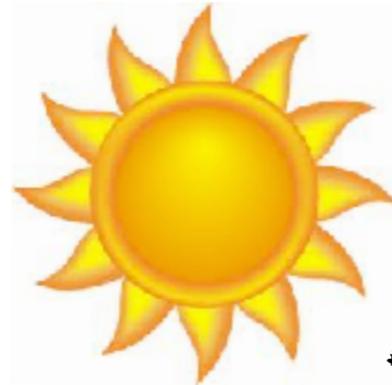


Retina

# Colour Perception



$E(\lambda)$

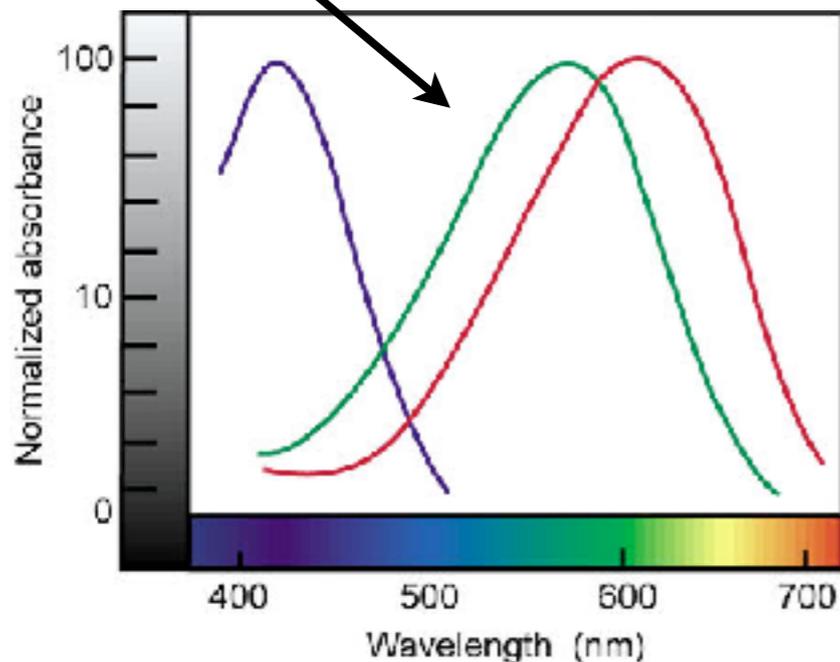


$S(\lambda)$

Cone responses



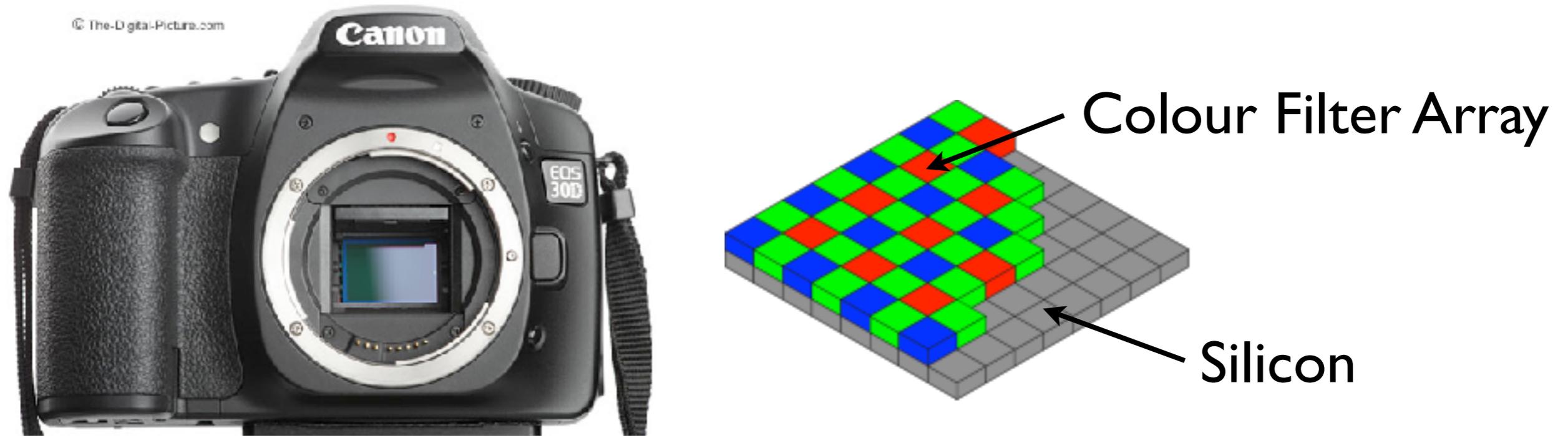
$R_{red}(\lambda)$



Cone excitation (multiply and add):

$$\rho_{red} = \int R_{red}(\lambda) E(\lambda) S(\lambda) d\lambda$$

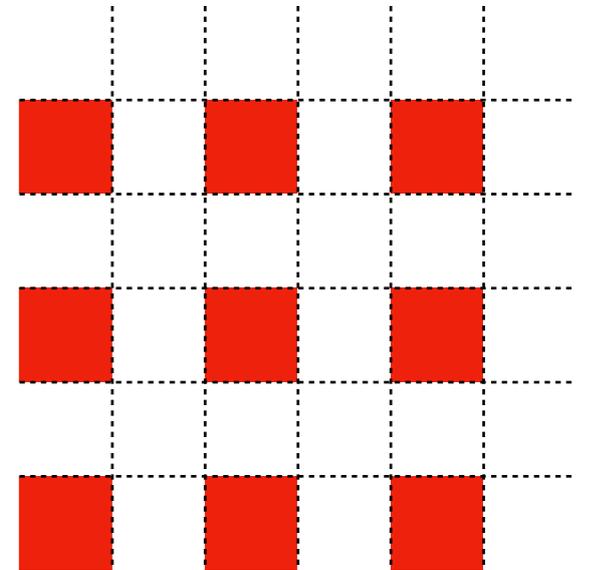
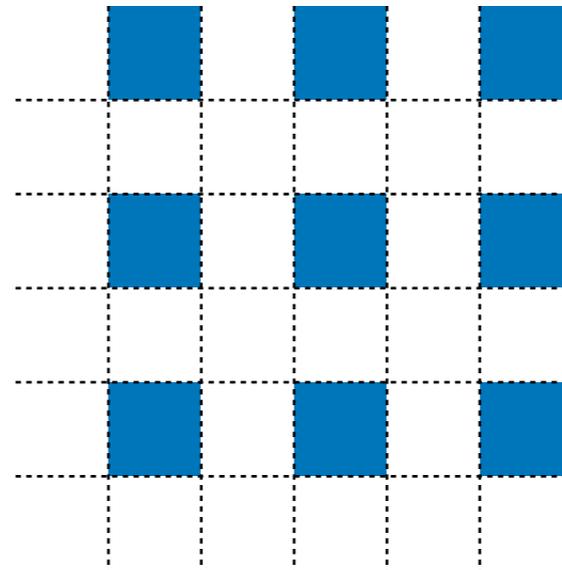
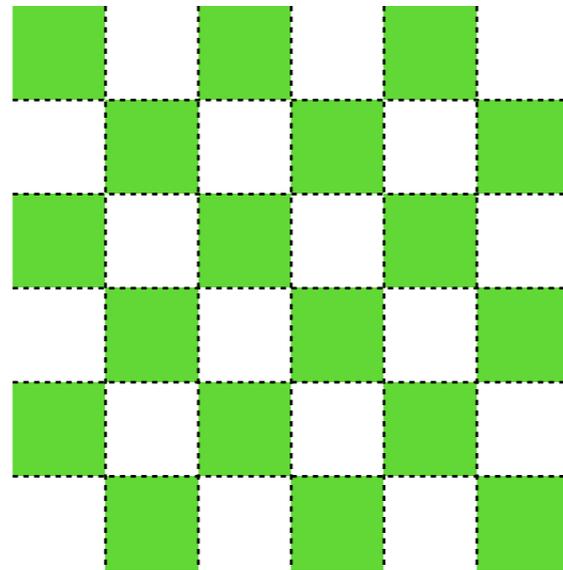
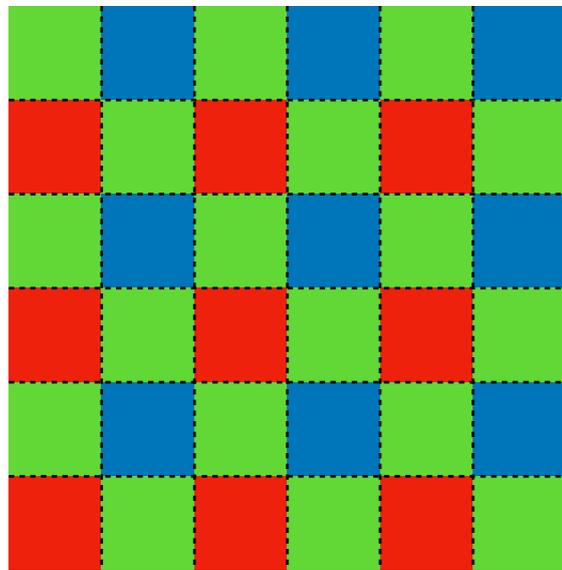
# Digital Sensor



- Analogue image is sampled by a CMOS (or CCD) sensor
- RGB colour filters arranged in a “Bayer” pattern
- Counts from this sensor are camera RAW
- For viewing we need an RGB value per pixel

# Demosaicing

- Each colour channel has different information:



How can we fill in the missing information?

# Demosaicing

- Simple interpolation causes colour errors



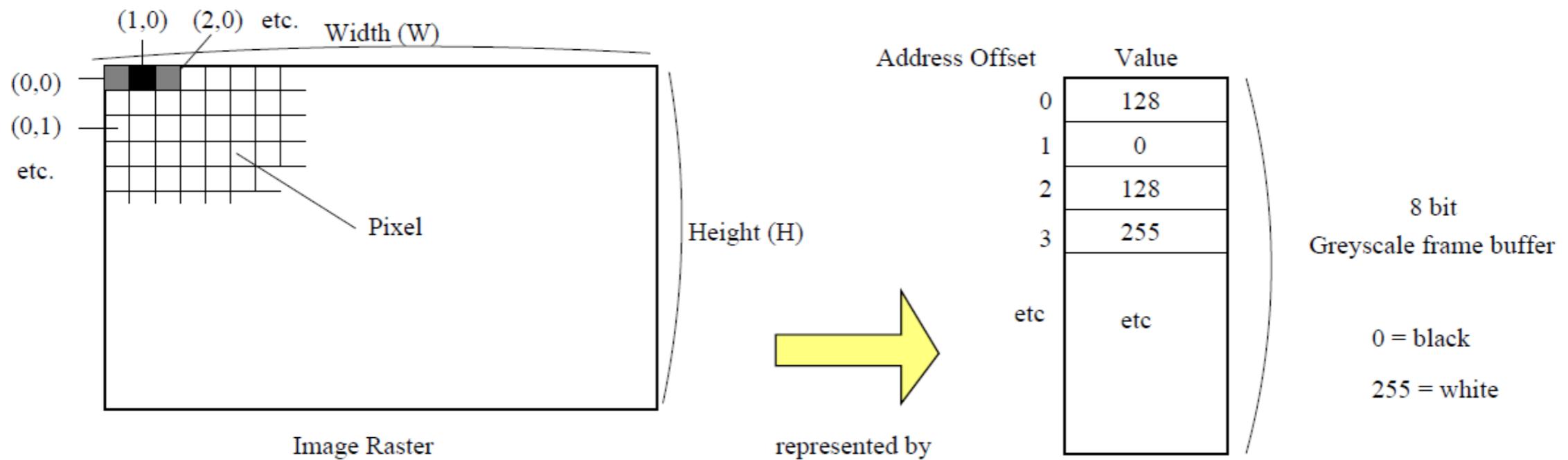
Bilinear interpolation



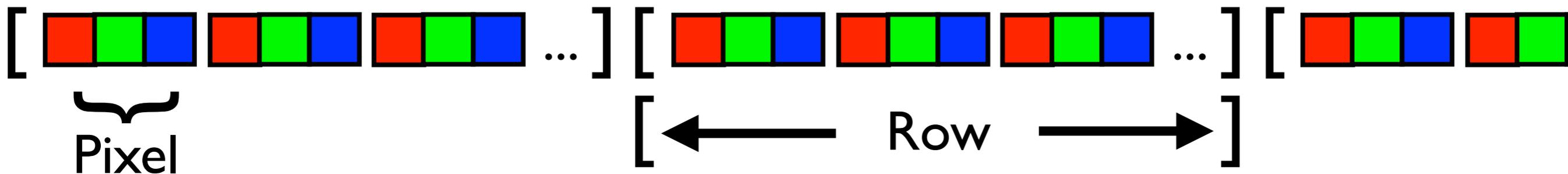
Bennet et al 2006  
(local 2 colour prior)

- Many techniques use edge information from the densely sampled green channel, and some form of image prior
- It can also be tackled via a data-driven approach, e.g., [Gharbi et al. 2016]

# The Digital Image



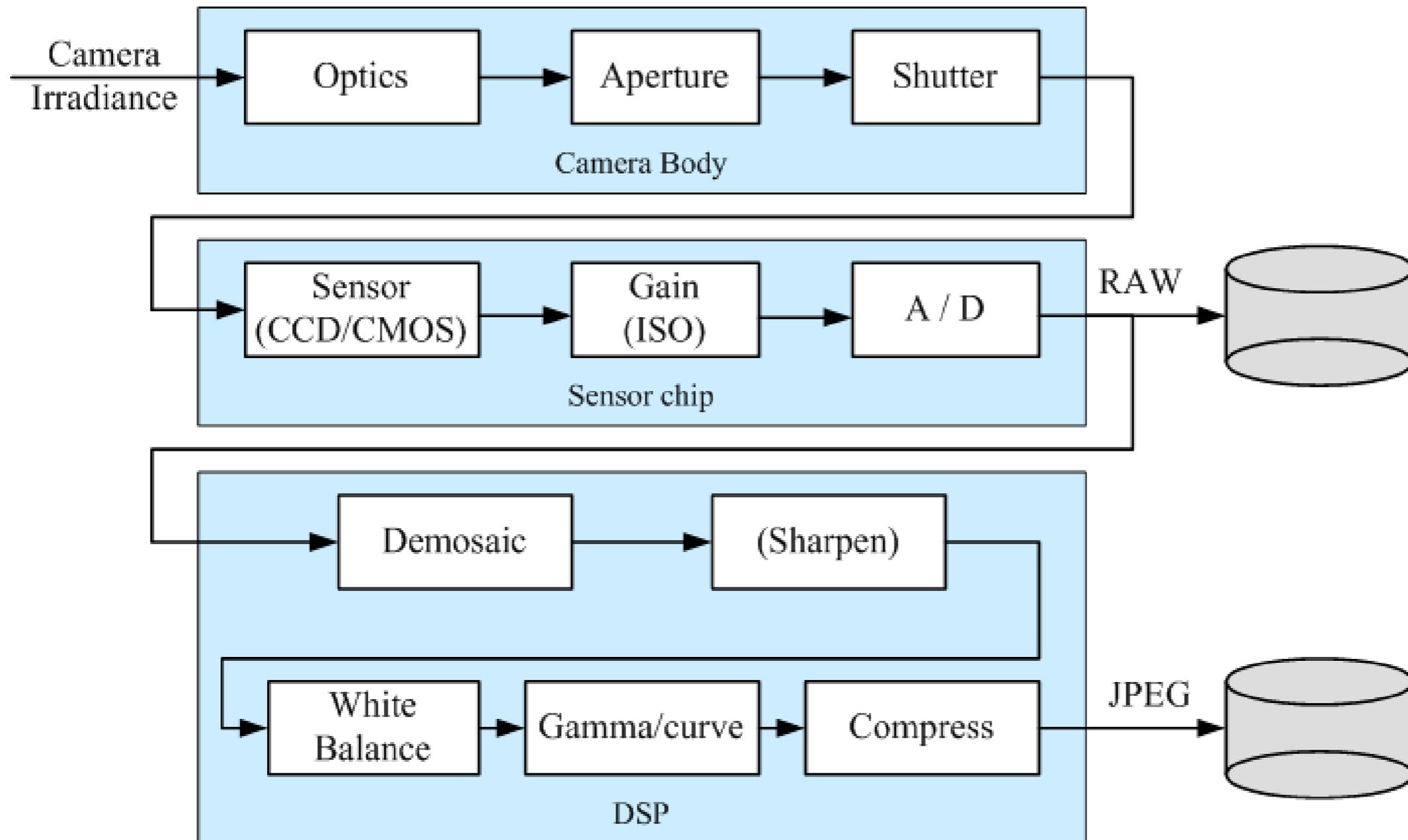
- e.g., arranged in memory with RGB pixels stored in rows:



- Many other possibilities, e.g., BGR, RGBA pixels, row/column major ordering, and rows or columns aligned to power of 2 boundaries

# Digital Camera Processing

- Main stages in a digital camera



# White Balance

- Humans are good at adapting to global illumination conditions: you would still describe a white object as white whether under blue sky or candle light.
- However, when the picture is viewed later, the viewer is no longer correcting for the environment and the illuminant colour typically appears too strong.
- **White balancing** is the process of correcting for the illuminant



- A simple white balance algorithm is to assume the scene is grey on average “greyworld”, state of the art methods use learning, e.g., Barron ICCV 2015

# Gamma Correction

- Equal steps in luminance  $\neq$  equal in perceived brightness

linear luminance (raw) 

|     |     |     |     |     |     |     |     |     |     |     |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 0.0 | 0.1 | 0.2 | 0.3 | 0.4 | 0.5 | 0.6 | 0.7 | 0.8 | 0.9 | 1.0 |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|

equal brightness steps 

|     |     |     |     |     |     |     |     |     |     |     |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 0.0 | 0.1 | 0.2 | 0.3 | 0.4 | 0.5 | 0.6 | 0.7 | 0.8 | 0.9 | 1.0 |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|

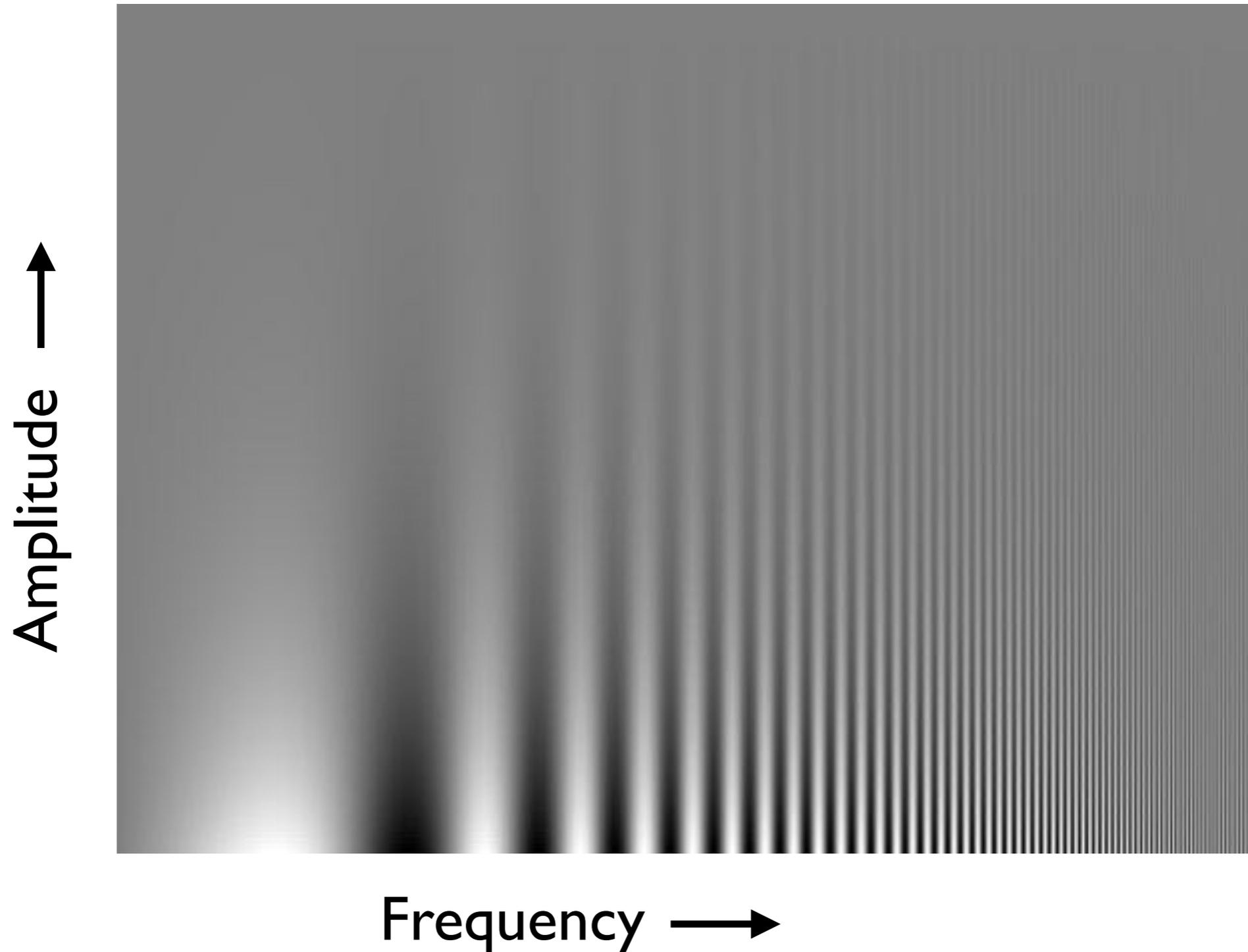
- Equal steps in perceived brightness are achieved by increasingly large steps in luminance (sensor counts)
- Human brightness perception ( $V$ ) follows a power law:

$$L = V^\gamma$$

- Using raw sensor counts wastes bits as we can't differentiate the large values  $\rightarrow$  use **gamma corrected encoding** that allocates more bits to smaller values

# Contrast Sensitivity

- Human visual system is most sensitive to mid-frequencies

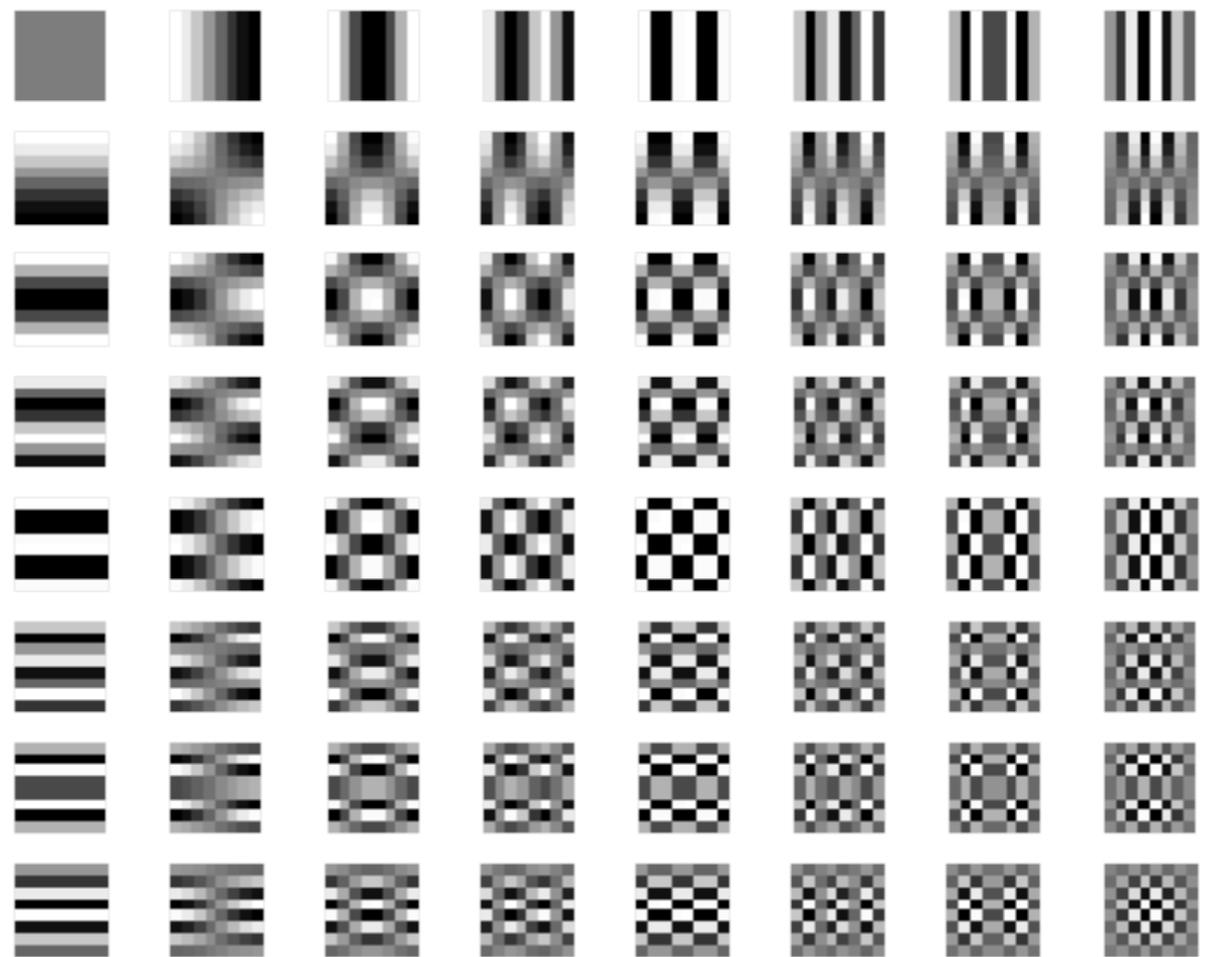


# Discrete Cosine Transform

- Basis functions used in JPEG

$$X(m, n) = \alpha_m \alpha_n \sum_{k=0}^{K-1} \sum_{l=0}^{L-1} x(k, l) \cos \left[ \frac{(2k+1)m\pi}{2K} \right] \cos \left[ \frac{(2l+1)n\pi}{2L} \right]$$

- Energy is concentrated in the low frequency components
- Efficient algorithm to compute (similar to FFT)



8x8 basis functions

# Coefficient Quantisation

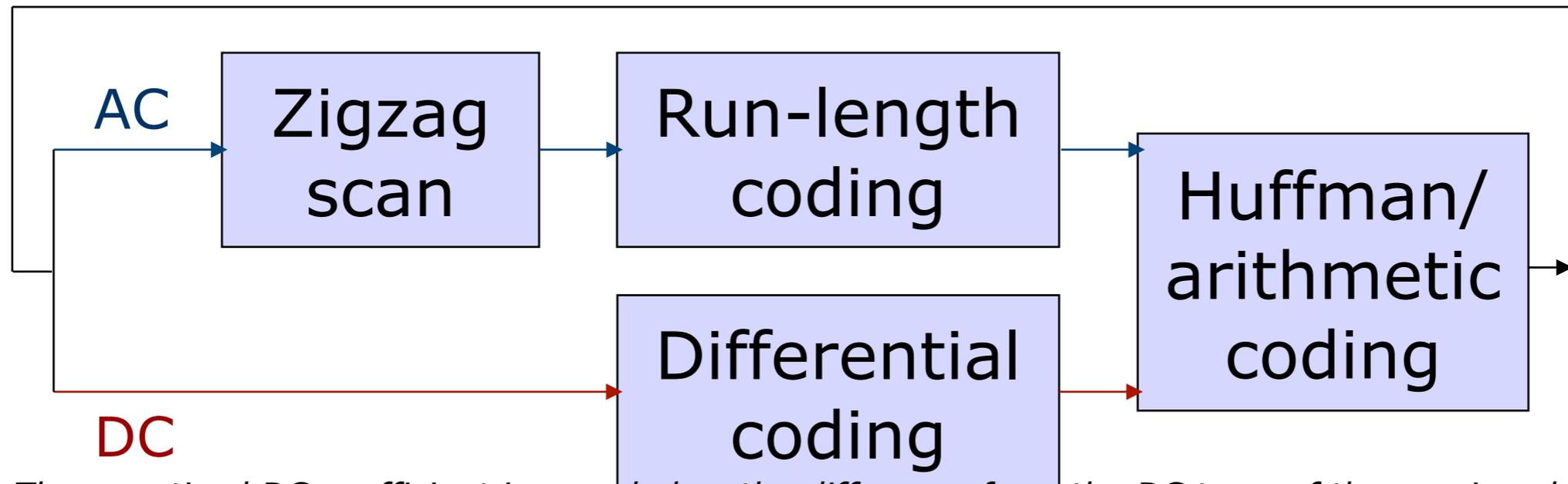
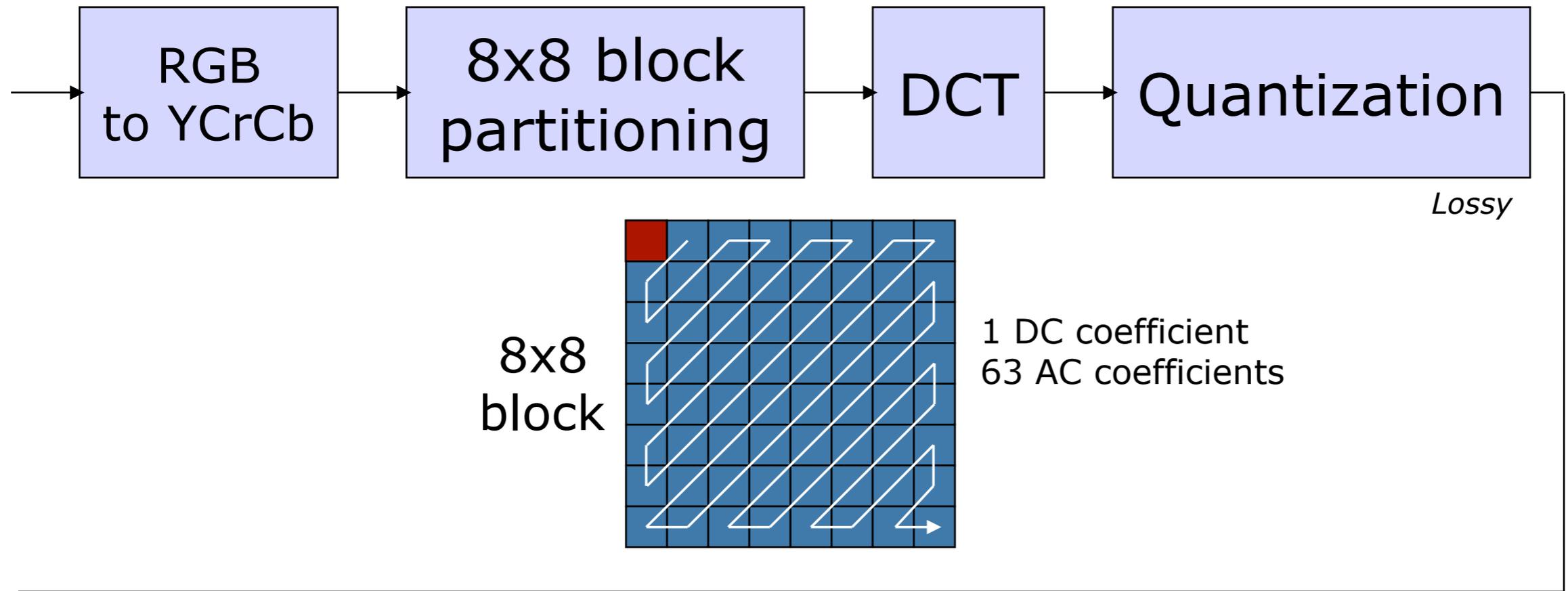
$$F^Q[u, v] = \text{Round} \left( \frac{F[u, v]}{Q[u, v]} \right)$$

- DCT coefficients  $F(u, v)$  are quantised according to a quantisation table
- High frequencies are less important (high factor)
- Quantisation table entries determine the “lossiness” of the compression

$Q[u, v]$

|    |    |    |    |     |     |     |     |
|----|----|----|----|-----|-----|-----|-----|
| 16 | 11 | 10 | 16 | 24  | 40  | 51  | 61  |
| 12 | 12 | 14 | 19 | 26  | 58  | 60  | 55  |
| 14 | 13 | 16 | 24 | 40  | 57  | 69  | 56  |
| 14 | 17 | 22 | 29 | 51  | 87  | 80  | 62  |
| 18 | 22 | 37 | 56 | 68  | 109 | 103 | 77  |
| 24 | 35 | 55 | 64 | 81  | 104 | 113 | 92  |
| 49 | 64 | 78 | 87 | 103 | 121 | 120 | 101 |
| 72 | 92 | 95 | 98 | 112 | 100 | 103 | 99  |

# JPEG Compression



*The quantized DC coefficient is encoded as the difference from the DC term of the previous block to account for the strong correlation between adjacent DC coefficient*

# YCbCr

- Separates luminance (Y) from chrominance (Cb, Cr) = colour



$$Y' = 16 + 65.5R' + 128.6G' + 25.0B'$$
$$Cb = 128 - 37.8R' - 74.2G' + 112B'$$
$$Cr = 128 + 112.0R' - 93.8G' - 18.2B'$$

- Linear transform of RGB
- Primes = gamma correction



Red

Green



Blue



Y

Cr



Cb

# Blurring CbCr



sigma = 1.0

# Blurring CbCr



sigma = 2.0

# Blurring CbCr



sigma = 4.0

# Blurring CbCr



sigma = 8.0

# Blurring CbCr



sigma = 16.0

# Blurring CbCr



sigma = 32.0

# Blurring Y



sigma = 1.0

# Blurring Y



sigma = 2.0

# Blurring Y



sigma = 4.0

# Blurring Y



sigma = 8.0

# Blurring Y



sigma = 16.0

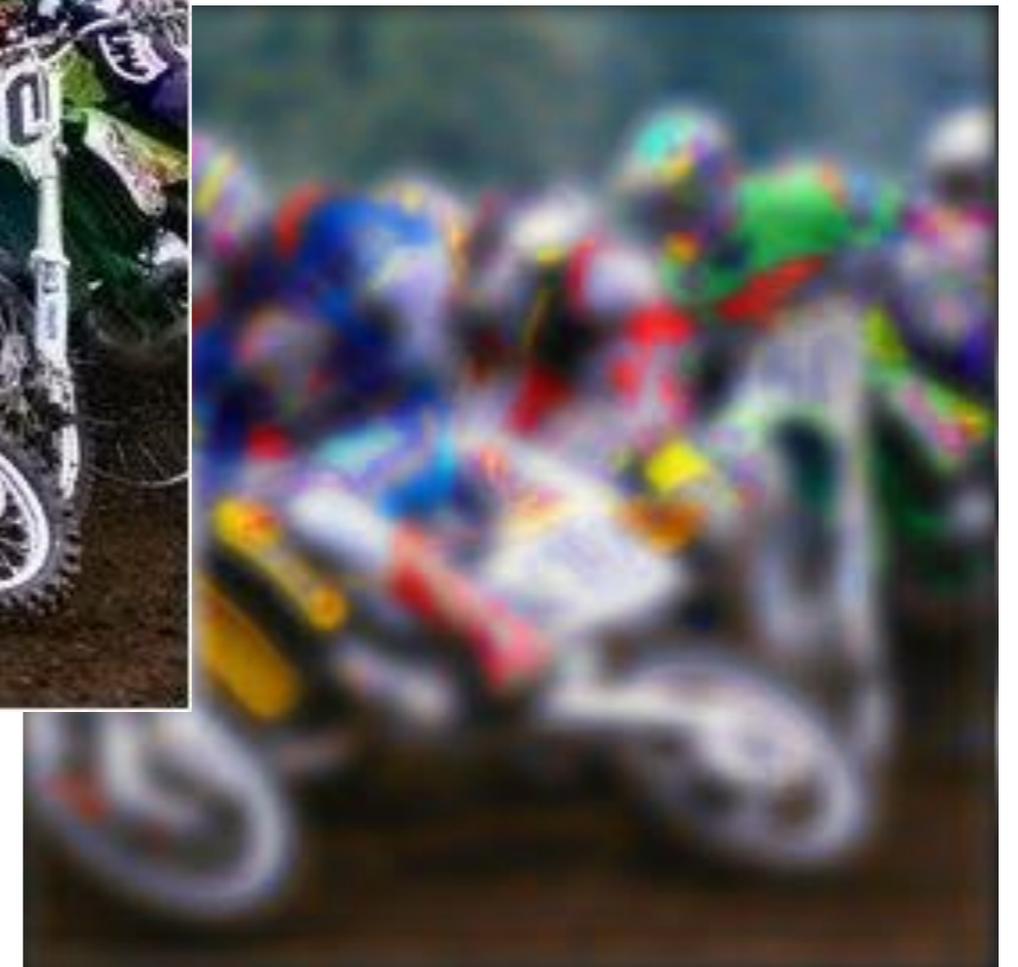
# Subsampling CbCr vs Y



Original



Chrominance  
1/8 scale



Luminance  
1/8 scale

# Compressibility of Chrominance

- Cb+Cr are transmitted at 1/2 size for JPEG



- Note that human vision uses a similar transform to this (opponent colours), also we have fewer cones than rods

# Next Lecture

- Filtering and Pyramids
- Features and Matching