

MAIA - MedicAl Imaging and Applications  
**Advanced Image Analysis**

Lectures on Advanced Color Image Processing

# B4 - Digital Image Processing: The Camera

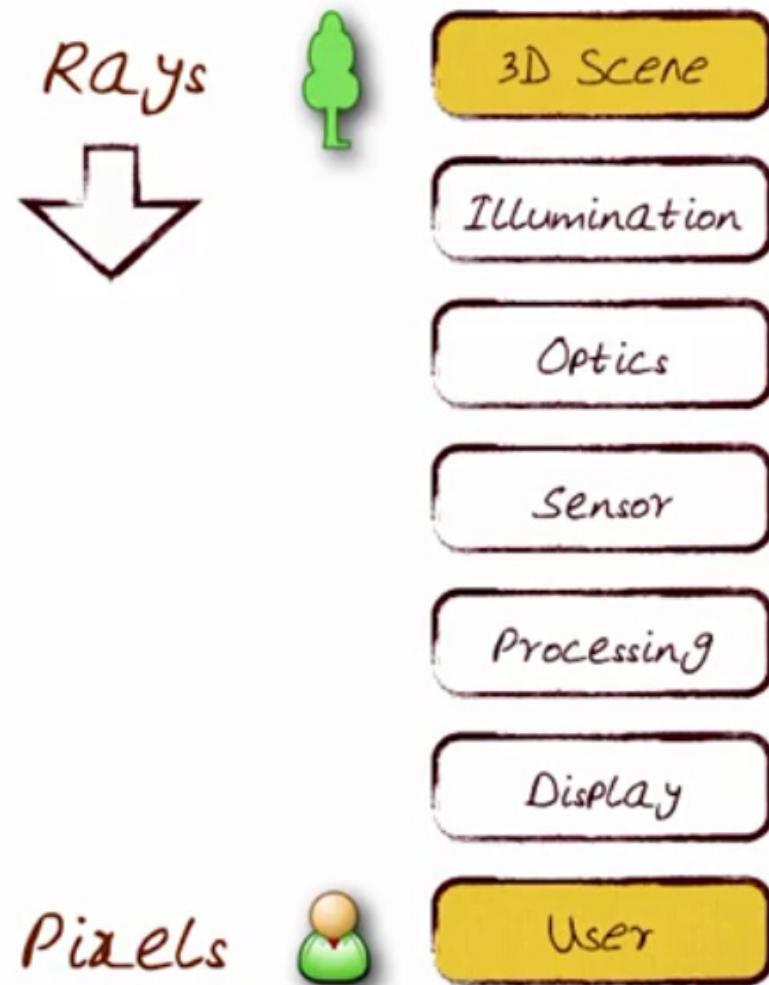
March/2018

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Biomedical Imaging Lab – INESC TEC Porto (Portugal)  
<http://bioimglab.inesctec.pt/>



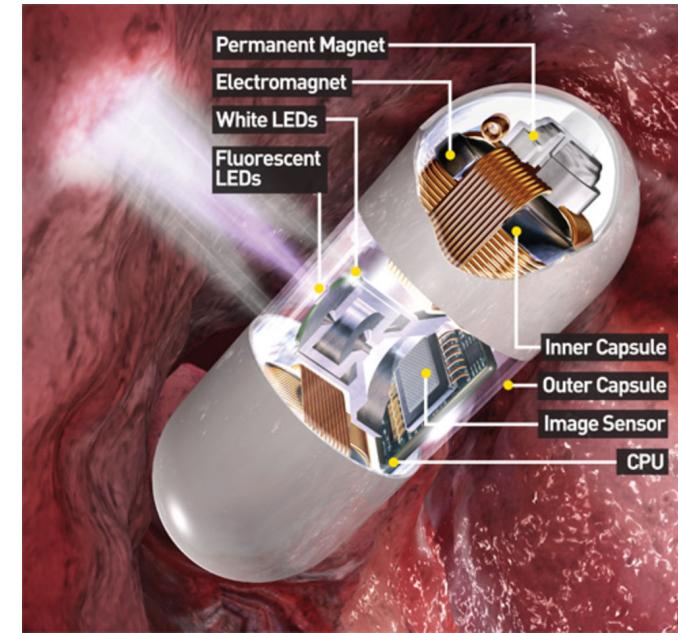
# 1 - The Image Acquisition Pipeline

How do we go from rays to pixels?

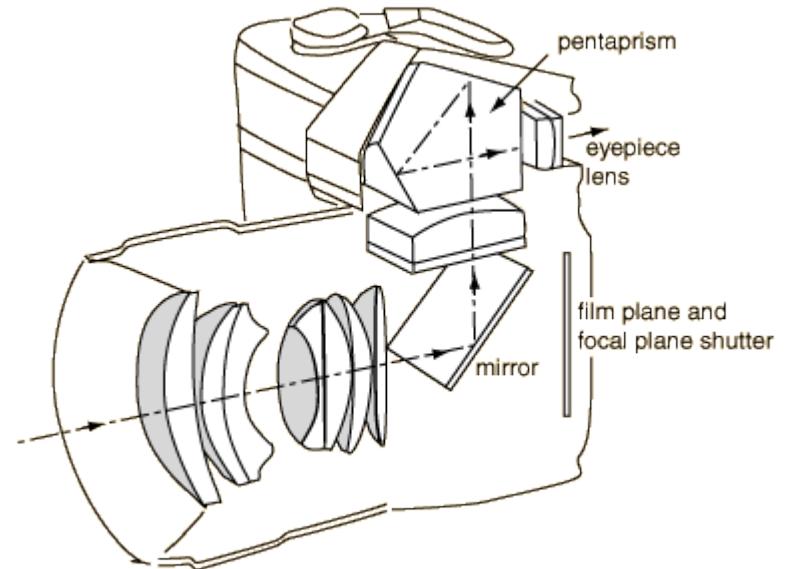


- **B1 - Human Vision System: The Retina and the Brain**
  1. Basics of Human Perception.
  2. Perception and Color. Measuring Color.
  3. Color Spaces. Perceptually Uniform Color Spaces.
  4. Application: Color Image Segmentation. Superpixels.
  5. Introduction to the Retinex model for Human Perception.
  6. Introduction to Computational Image Processing.
  7. Application: A first example - Image Composition.
- **B2 - Digital Image Processing: The Camera**
  1. The Image Acquisition Pipeline.
  2. **Cameras: Lenses, Focus, Exposure, White Balance, Sensors, Output.**
  3. Unconventional Image Acquisition: Panorama.
  4. Application: Panorama Creation.

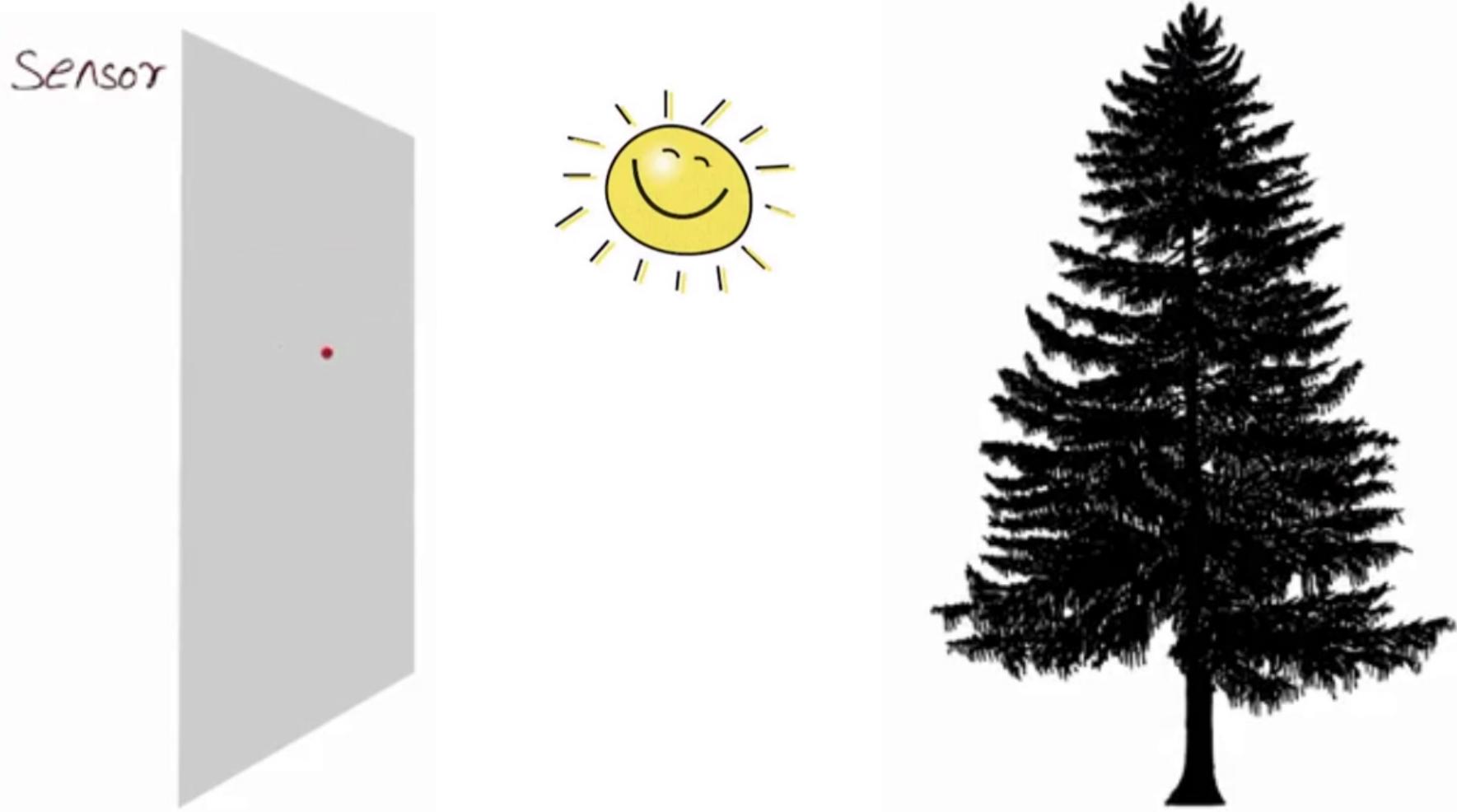
### Cameras and their evolution



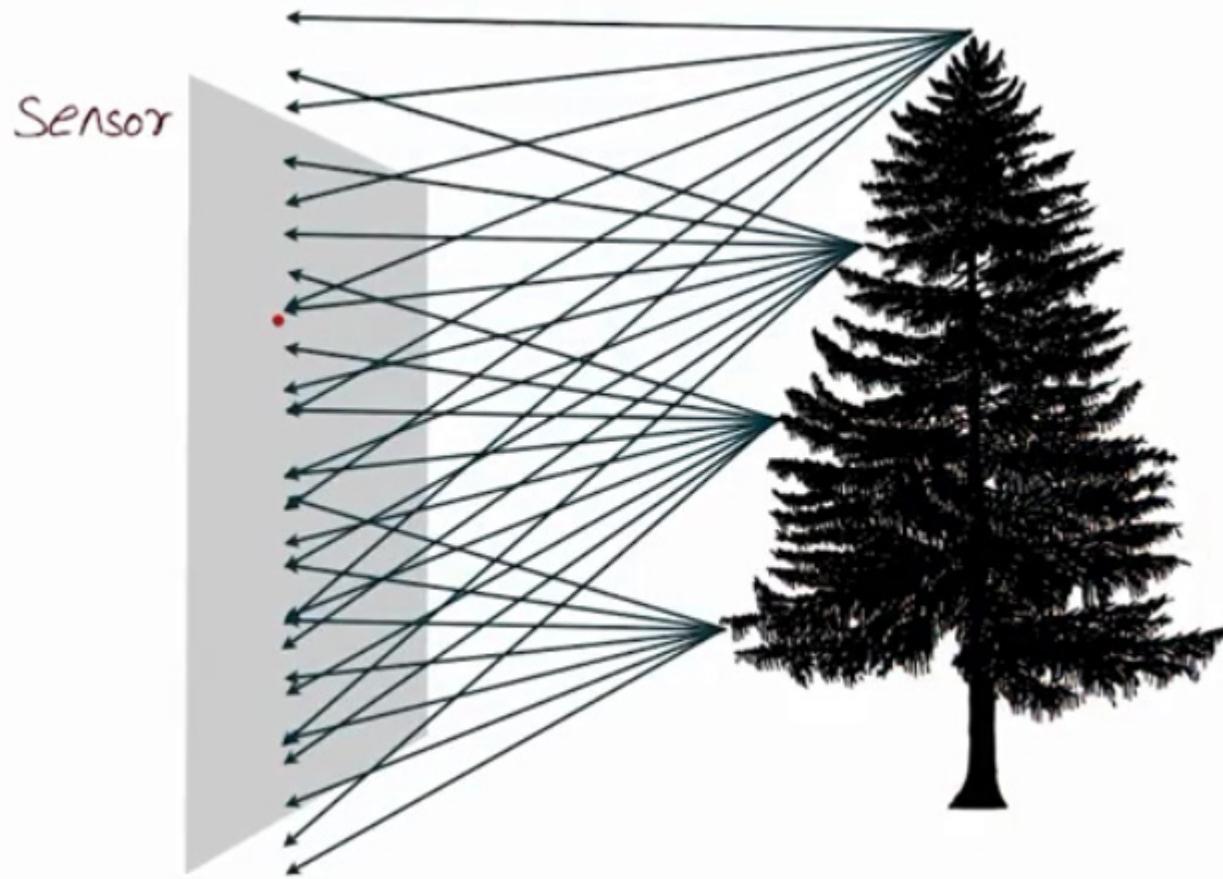
# Single Lens Reflex Cameras



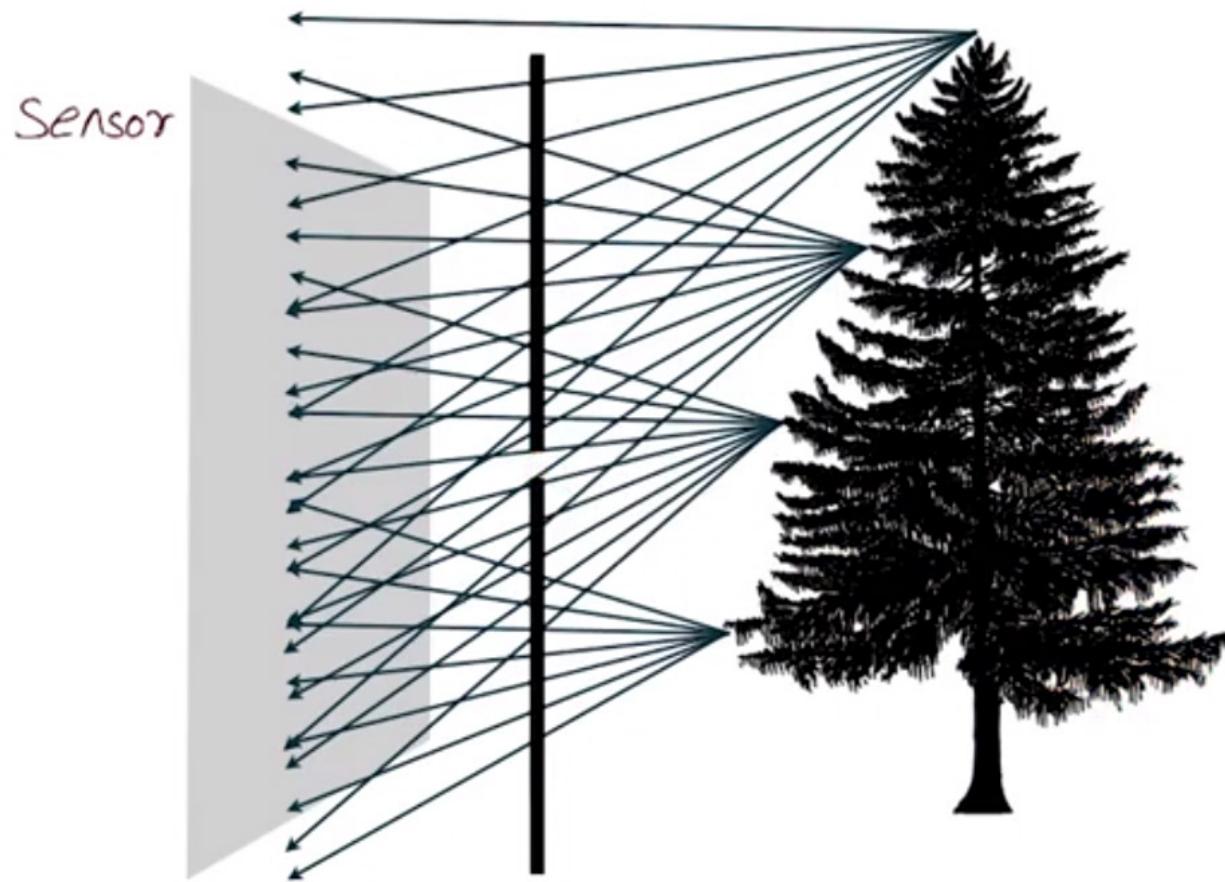
## No Optics Model - Pinhole Camera



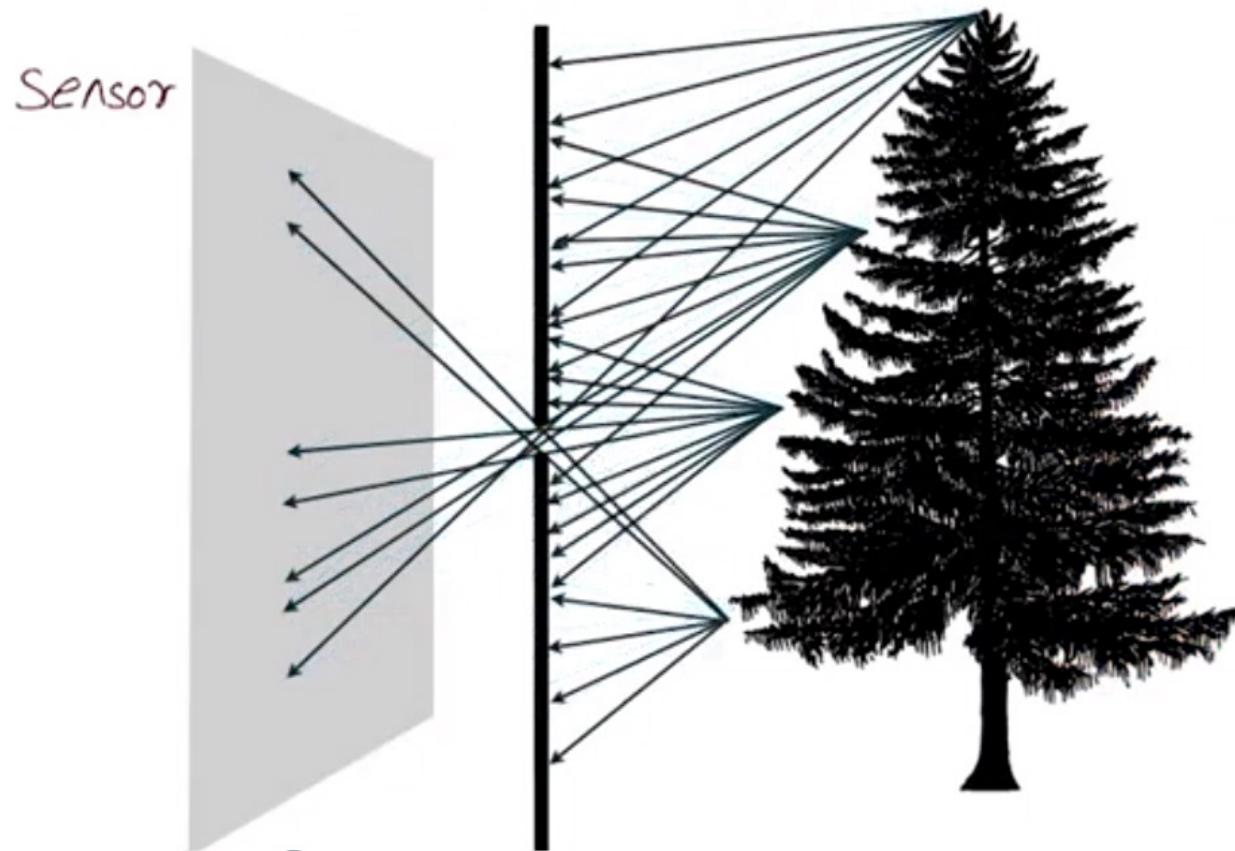
### No Optics Model



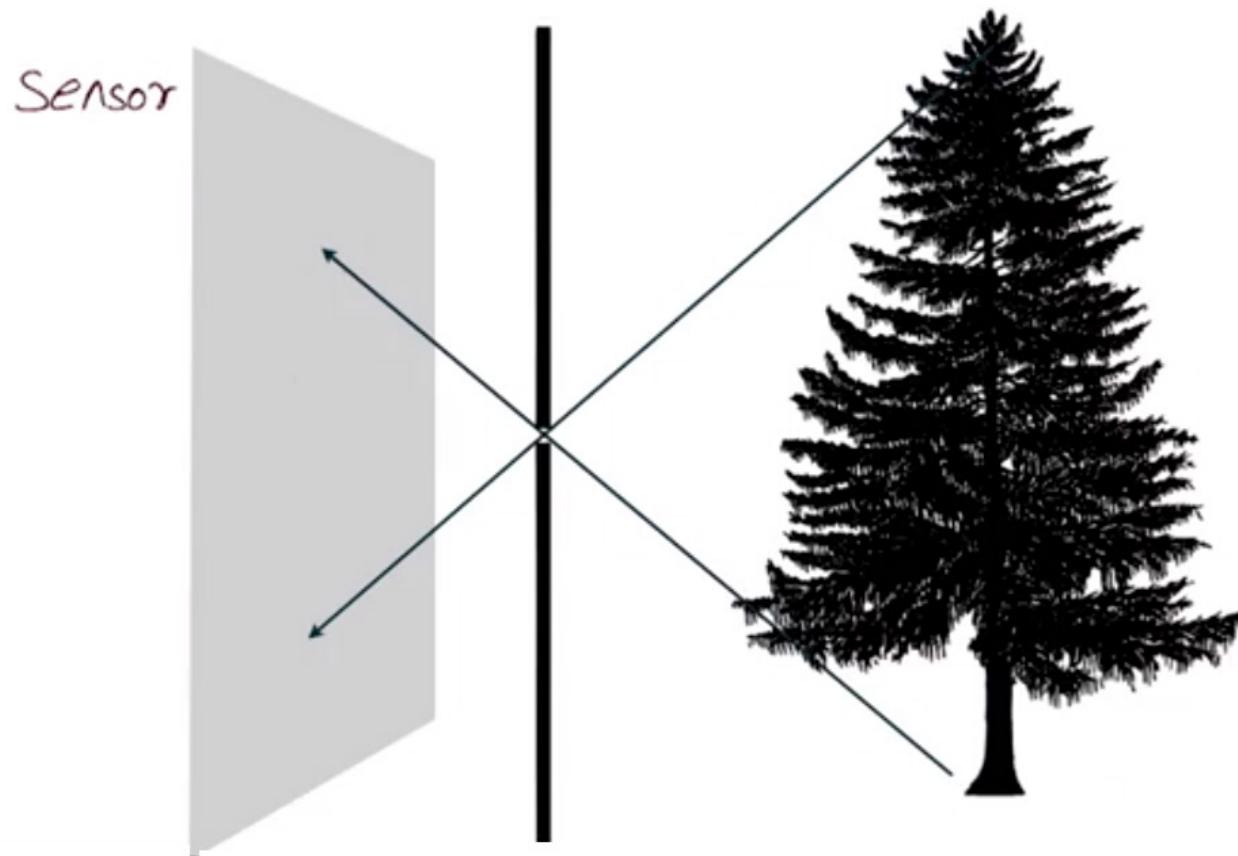
## No Optics Model - Pinhole Camera



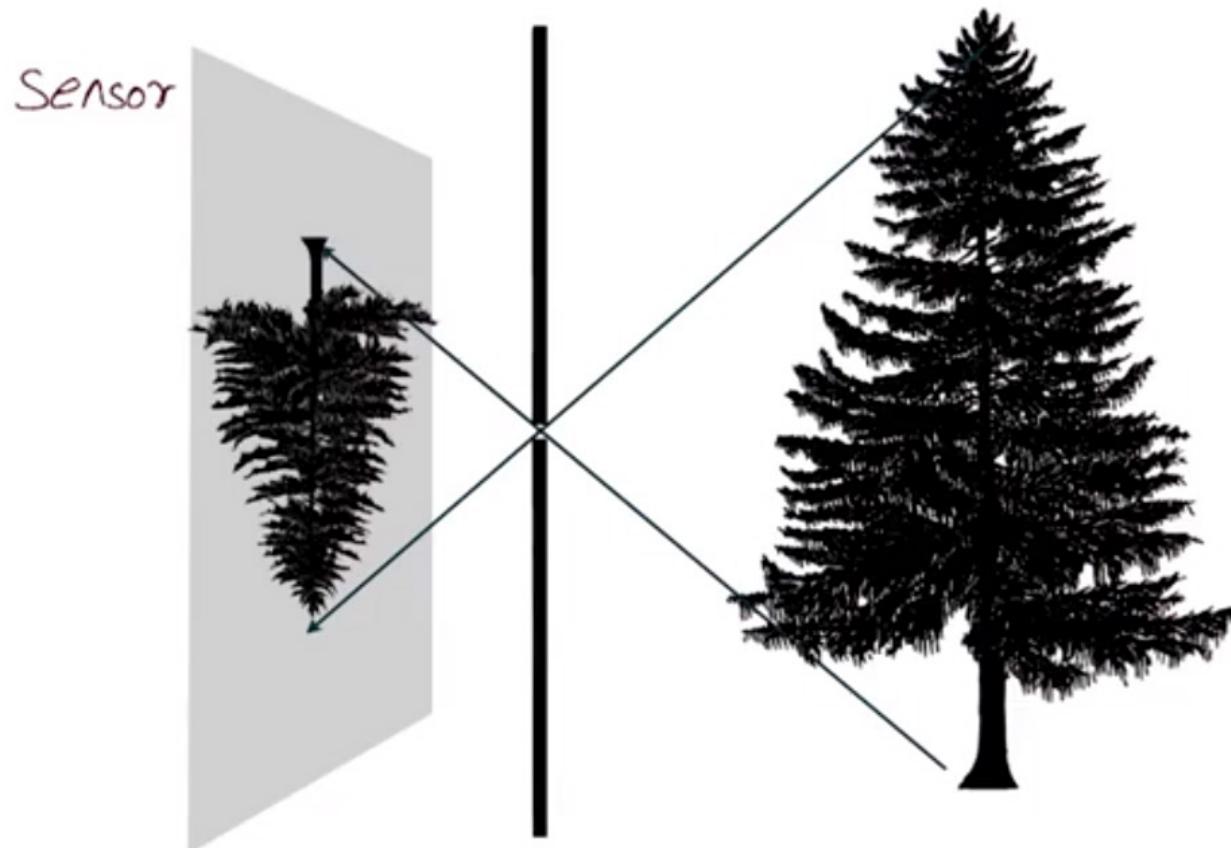
## No Optics Model - Pinhole Camera



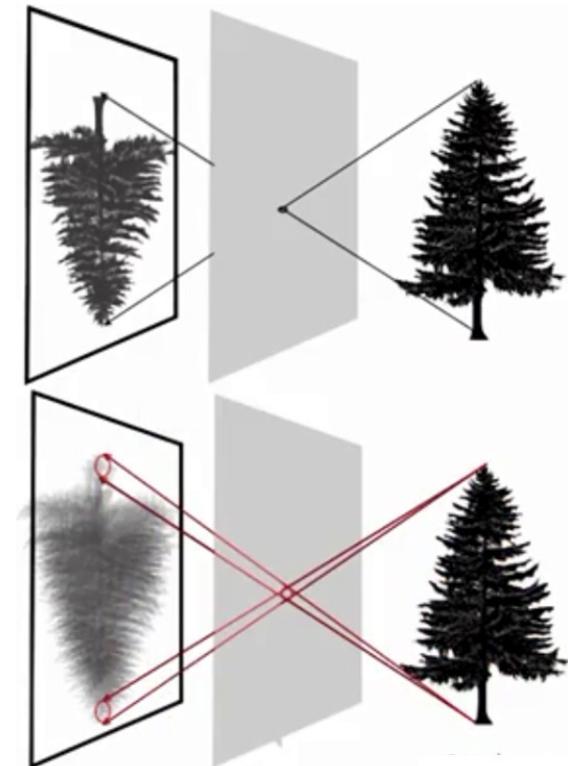
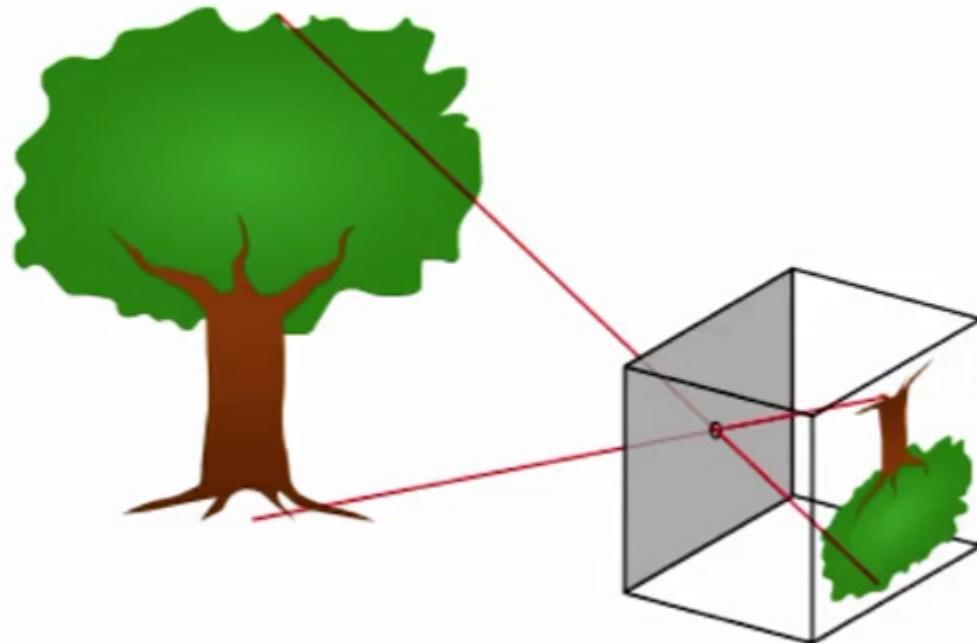
## No Optics Model - Pinhole Camera



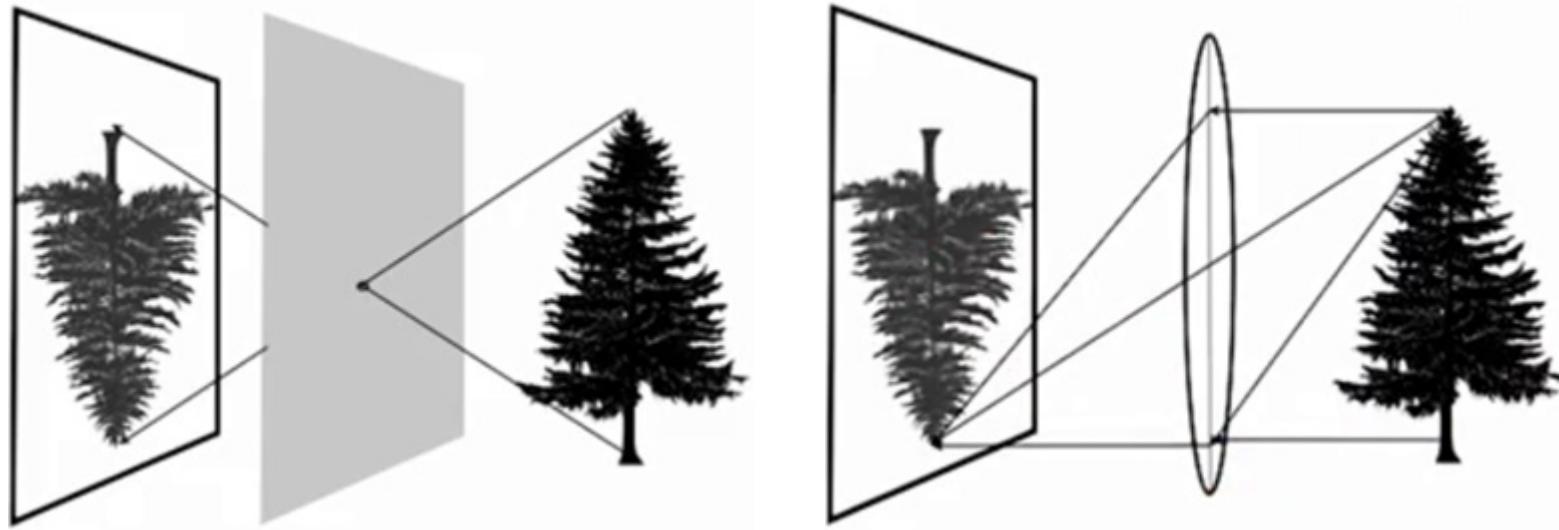
## No Optics Model - Pinhole Camera



## No Optics Model - Pinhole Camera

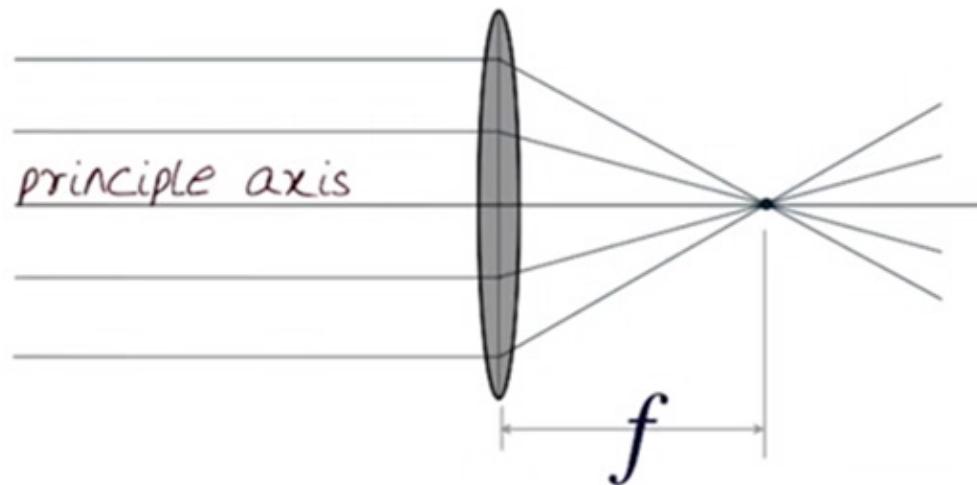


### Optics - From Pinhole to Lenses

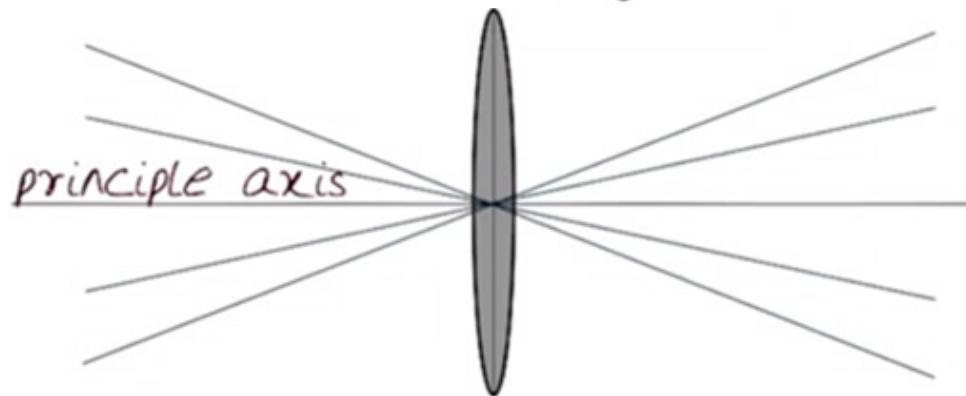


### Optics - From Pinhole to Lenses

**Lens:** specially shaped pieces of glass or other material with refraction index substantially higher than that of air.

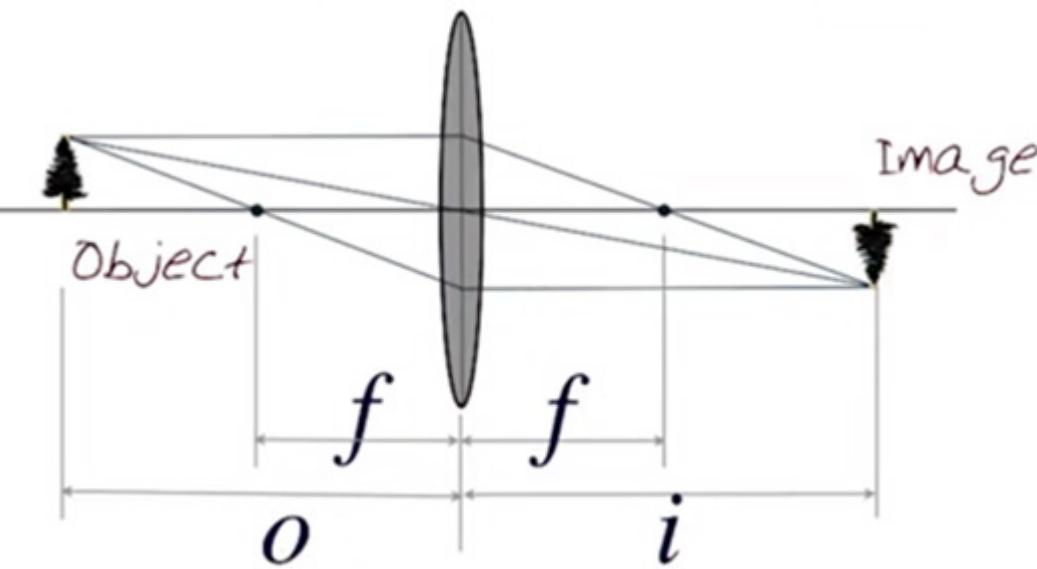


- \* Parallel rays converge to a point located at focal length,  $f$  from lens



- \* Rays going through center of lens do not deviate (functions like a pinhole)

## The Lens Equation

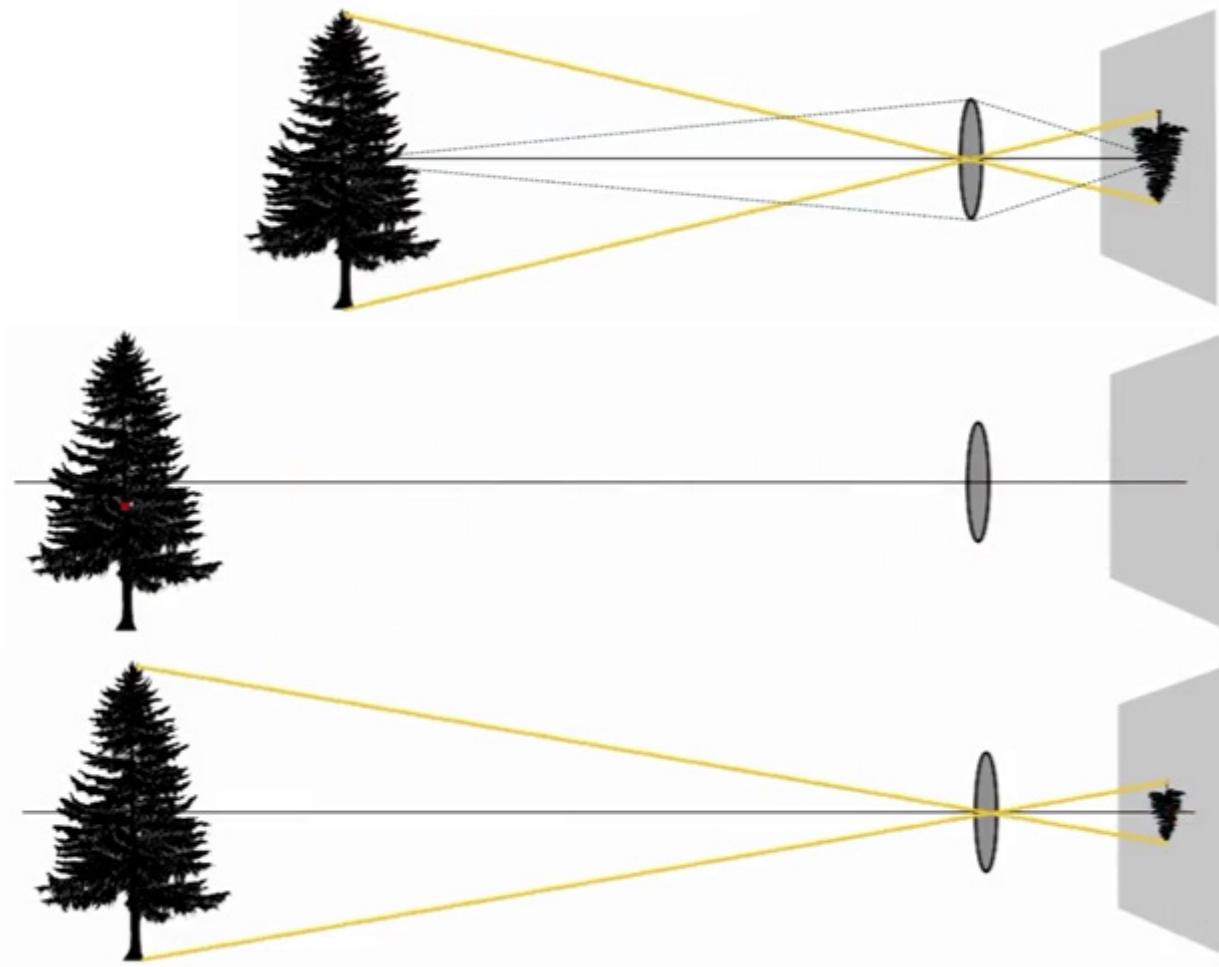


\* Rays from points on a plane parallel to the lens, focus on a plane parallel to the lens on the other side (and upside down)

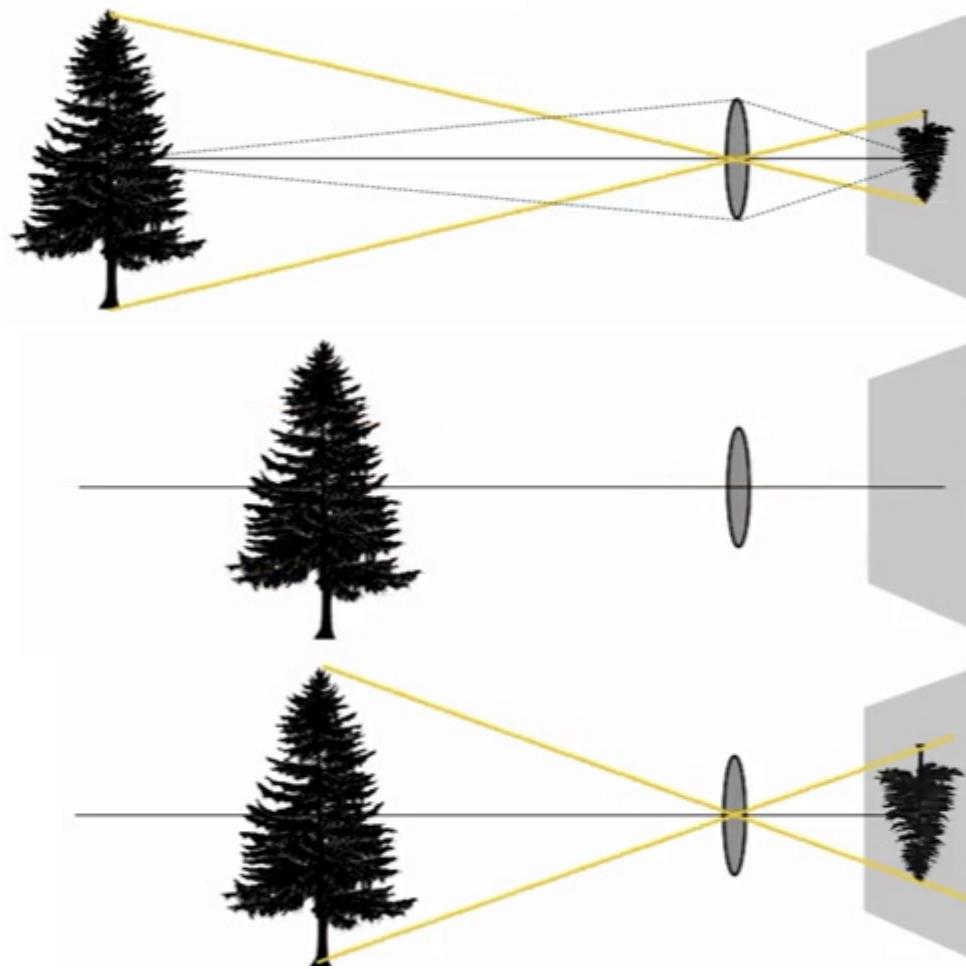
\* Lens Equation

$$\frac{1}{o} + \frac{1}{i} = \frac{1}{f}$$

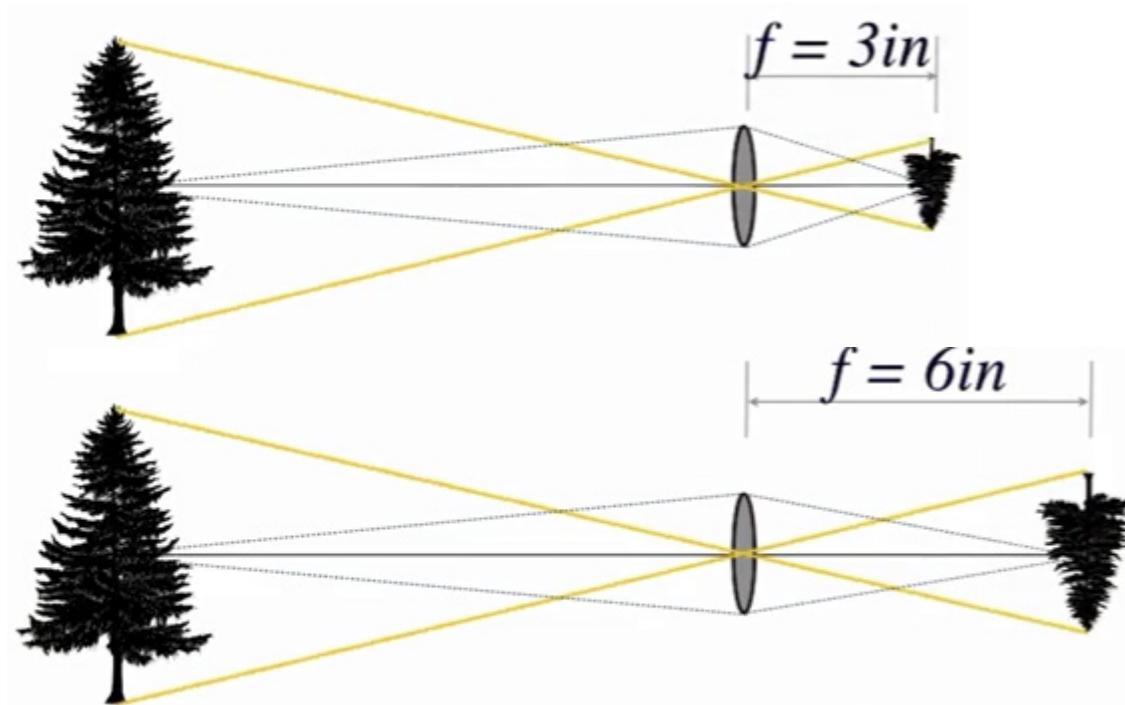
### Playing with distances



### Playing with distances



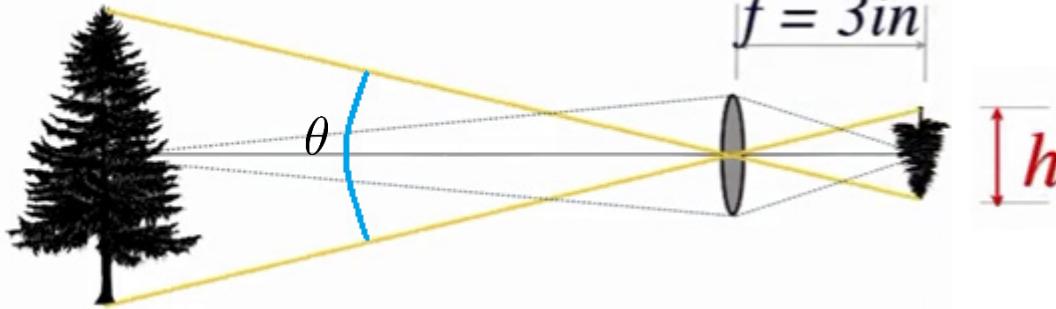
### Optical Zoom



\* longer or  
shorter f

\* focal  
lengths are  
specific to  
lenses

### Field of View



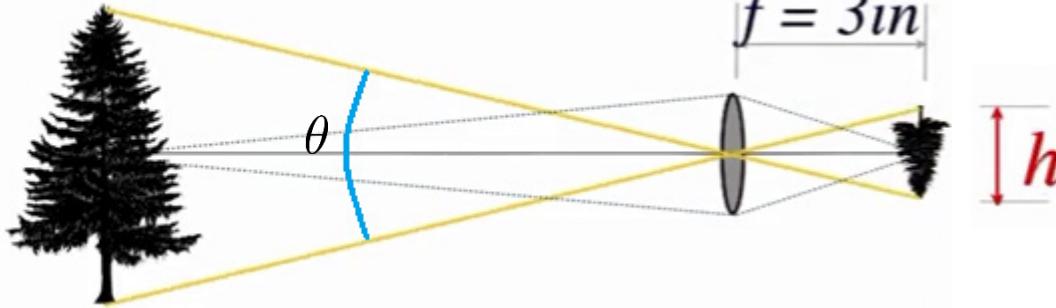
\* Field of View (FOV) depends on sensor size,  $h$

#### QUIZ!

How can you compute the field of view as a function of sensor size and focal length?

Link: <https://goo.gl/forms/RPNXfjZjve4Qd1Gn2>

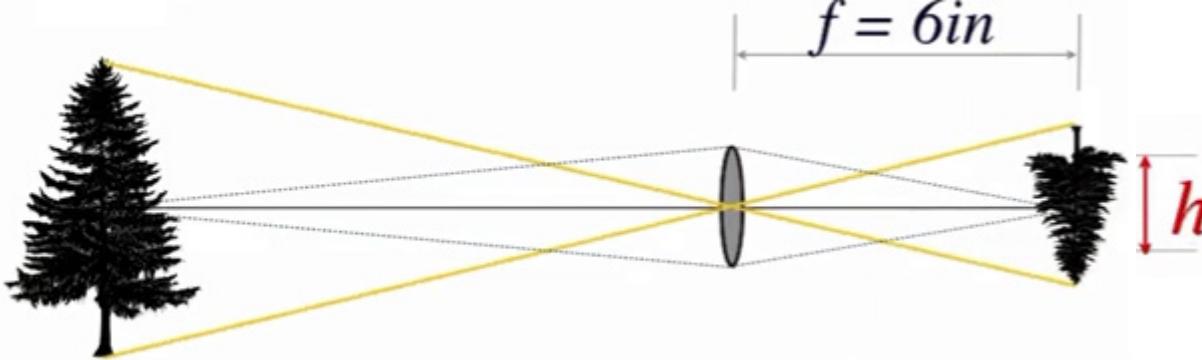
### Field of View



\* Field of View (FOV) depends on sensor size,  $h$

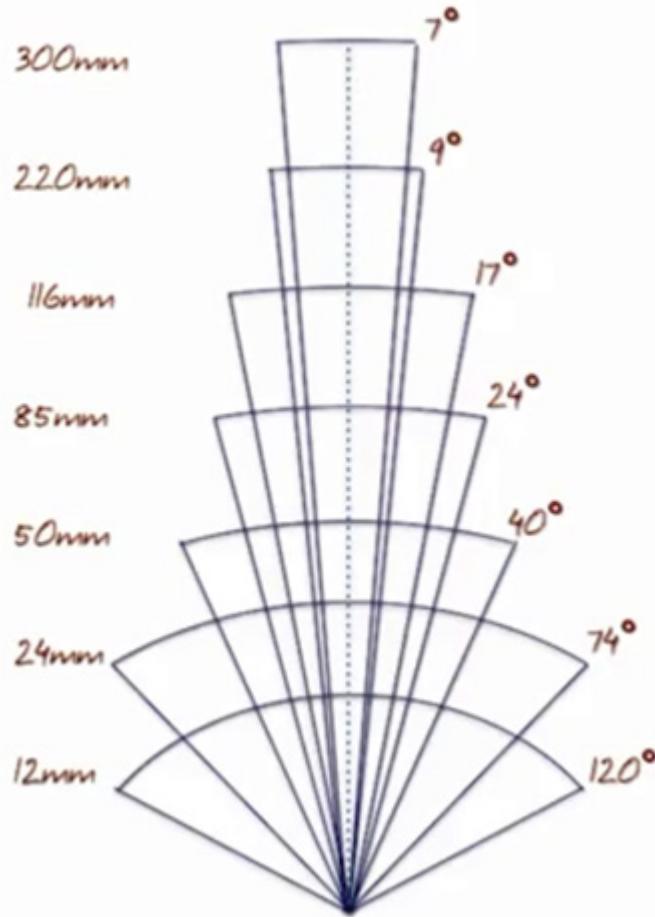
#### QUIZ!

How can you compute the field of view as a function of sensor size and focal length?

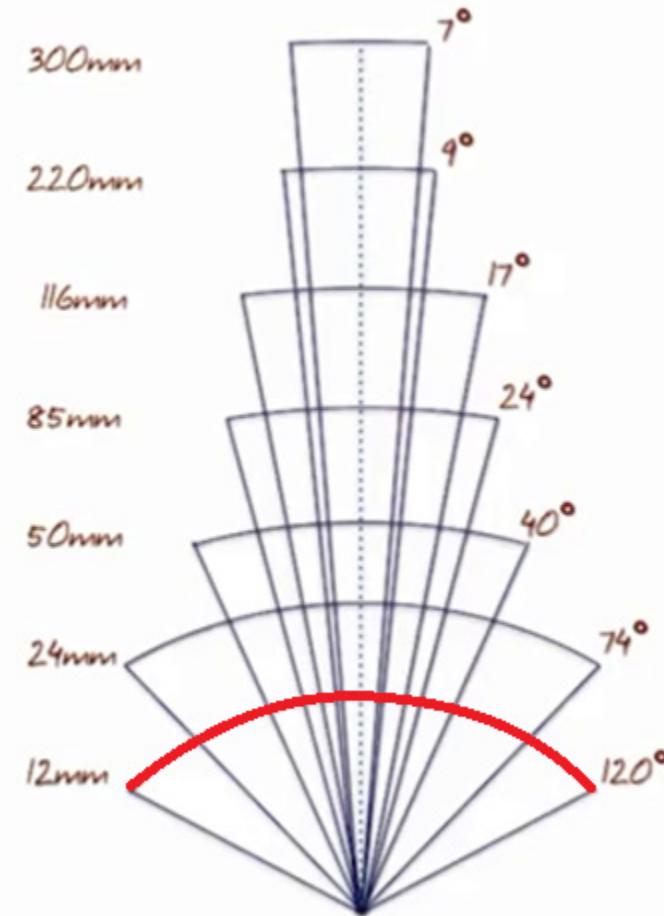


\* Sensor size small, FOV is small too

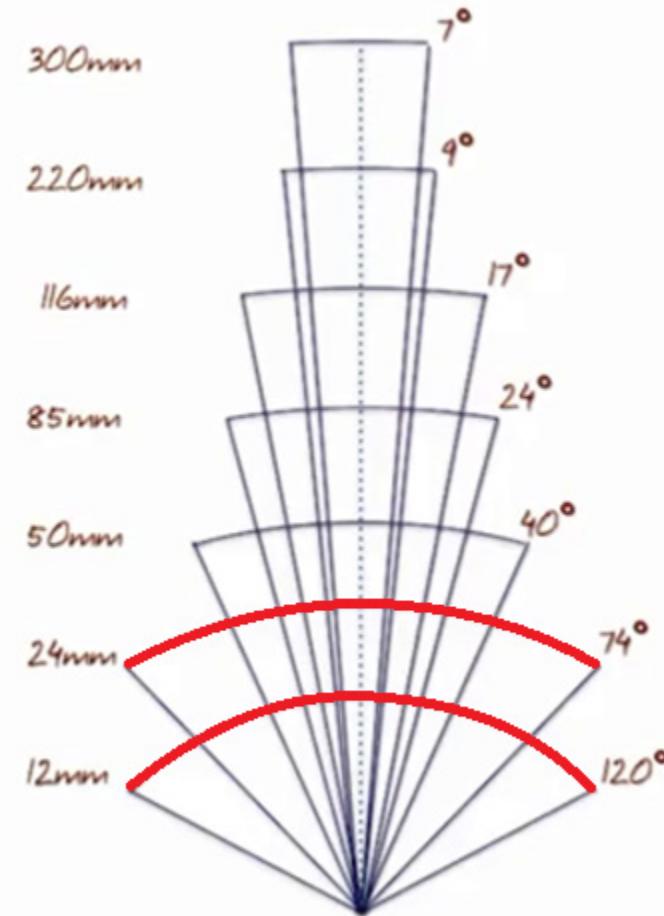
### Optical Zoom 2



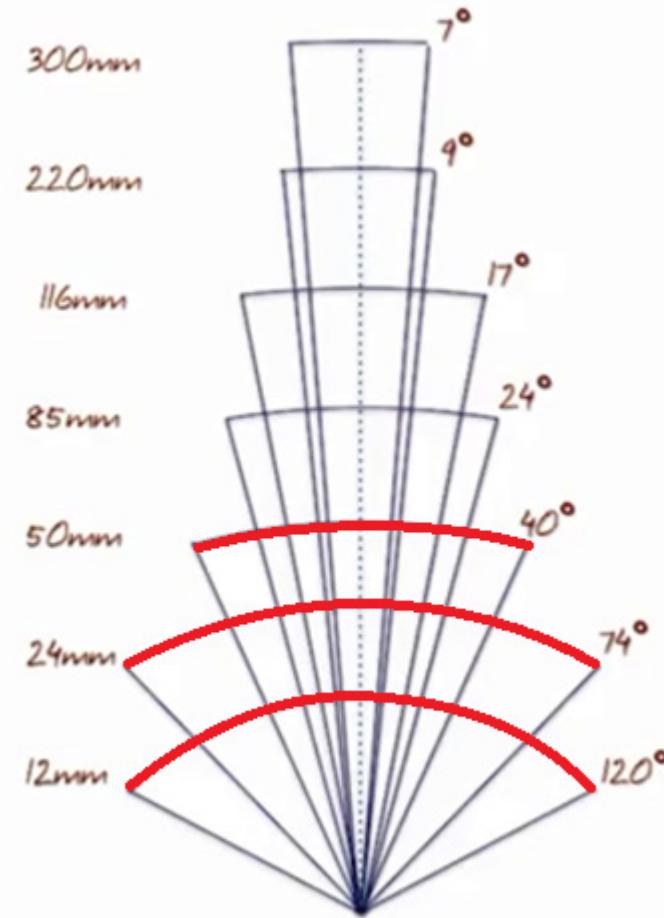
### Optical Zoom 2



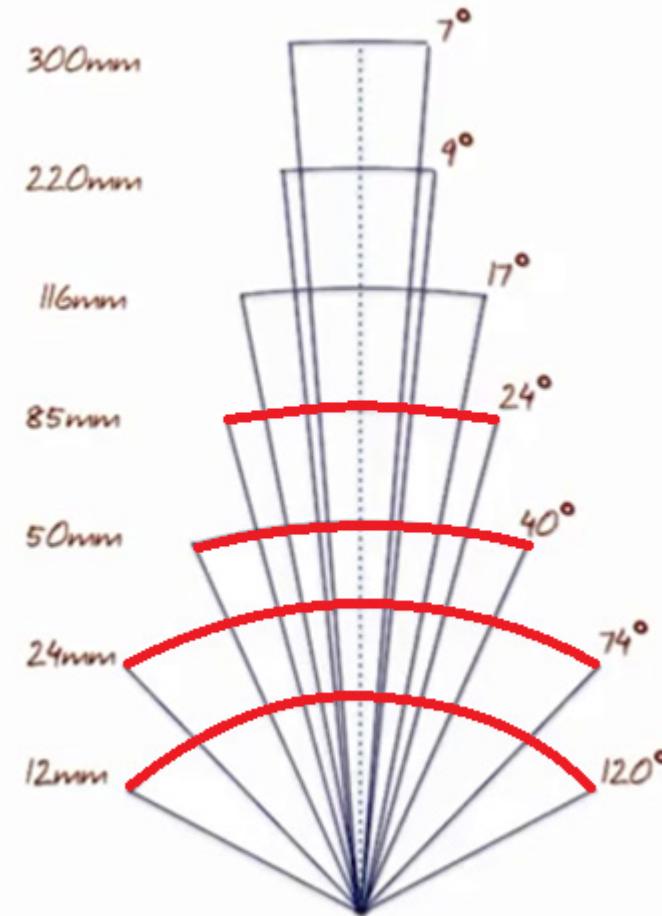
### Optical Zoom 2



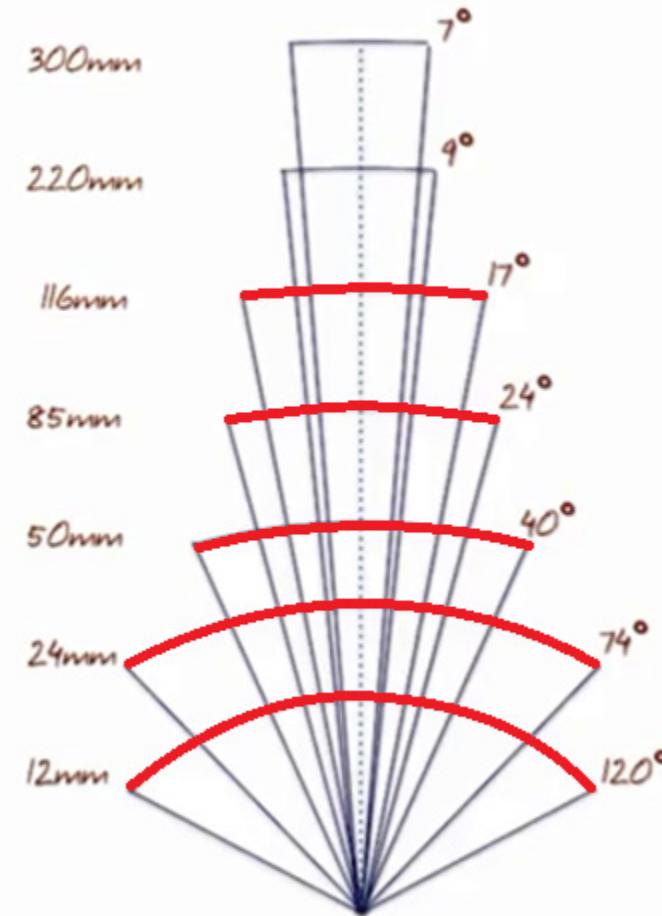
### Optical Zoom 2



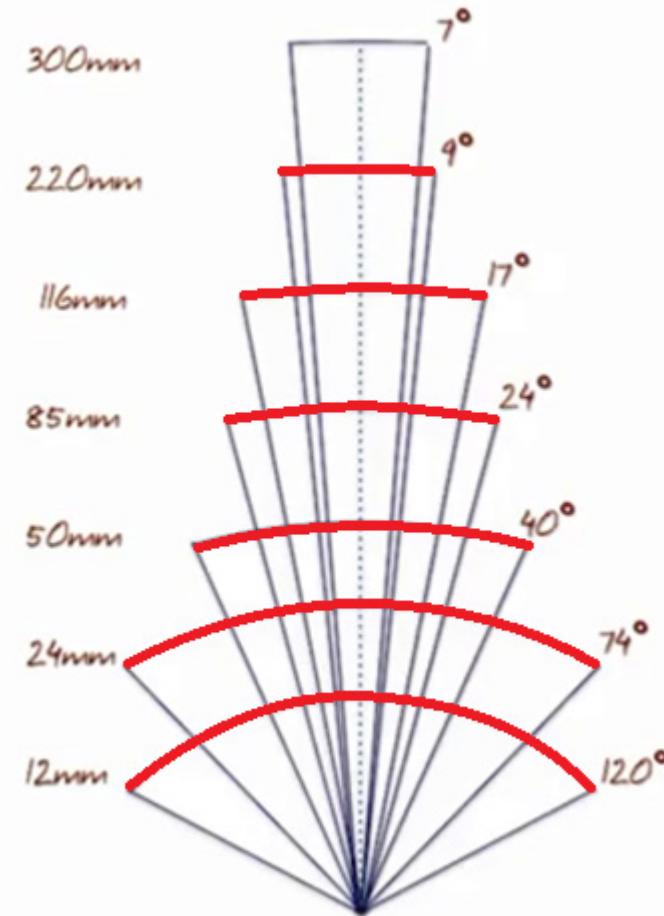
### Optical Zoom 2



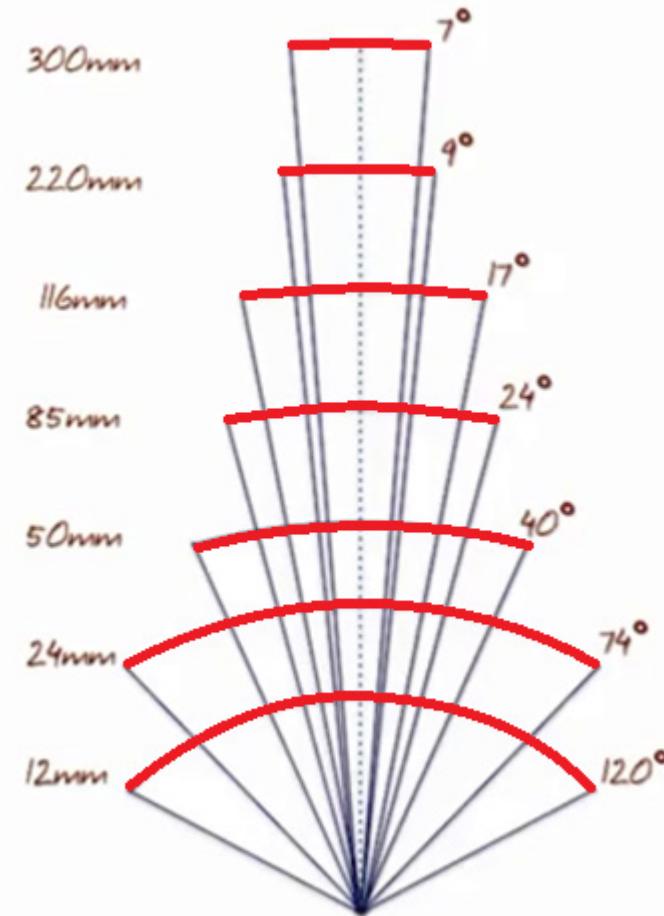
### Optical Zoom 2



### Optical Zoom 2



### Optical Zoom 2



### Common Sensor Sizes

8.8 X 6.6 mm  
(1:2) micro 2/3

4.54 X 3.52mm  
(iPhone 5)

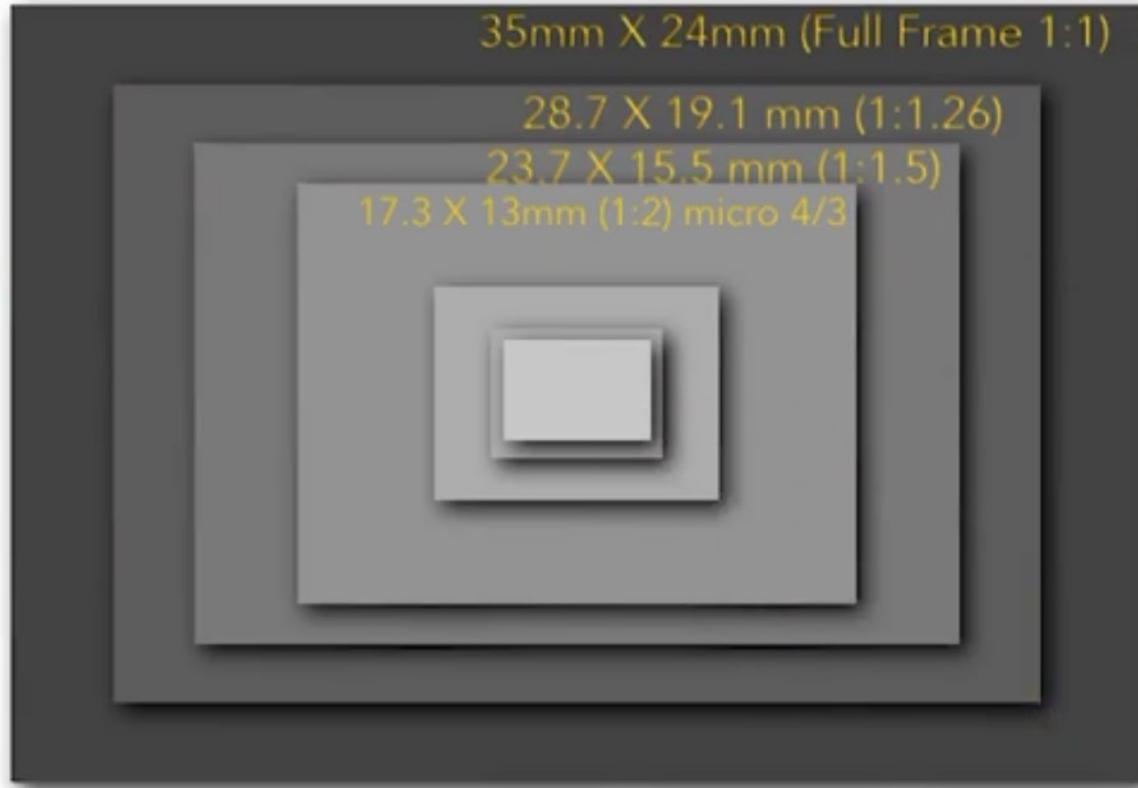
35mm X 24mm (Full Frame 1:1)

28.7 X 19.1 mm (1:1.26)

23.7 X 15.5 mm (1:1.5)

17.3 X 13mm (1:2) micro 4/3

7.25 X 5.33mm



### Focal Length and Perspective



$f = 18\text{mm}$ , 35mm sensor

Distance to 1st Subject = 0.5m   Distance to 1st Subject = 3.0m

Distance to 2nd Subject = 2.0m   Distance to 2nd Subject = 4.5m

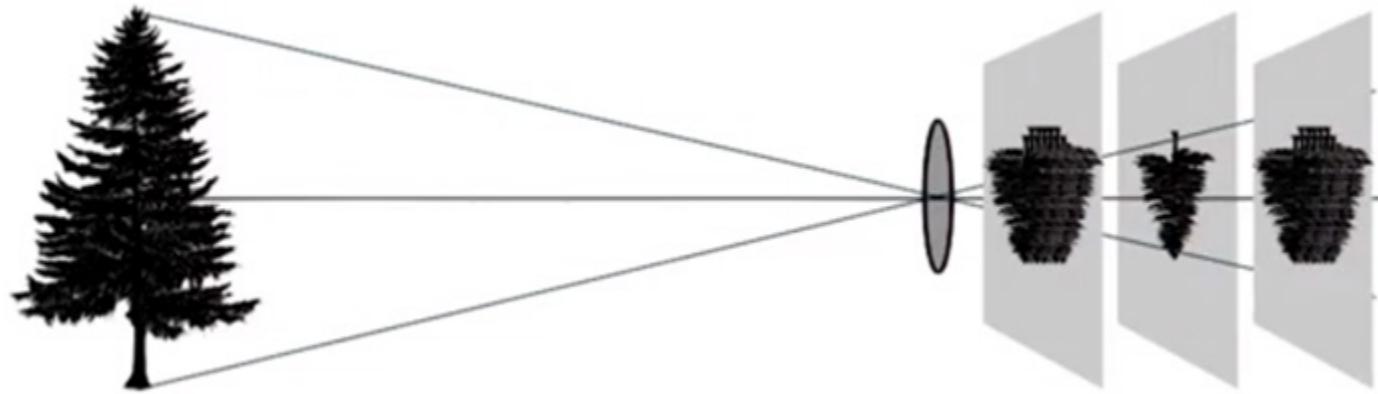
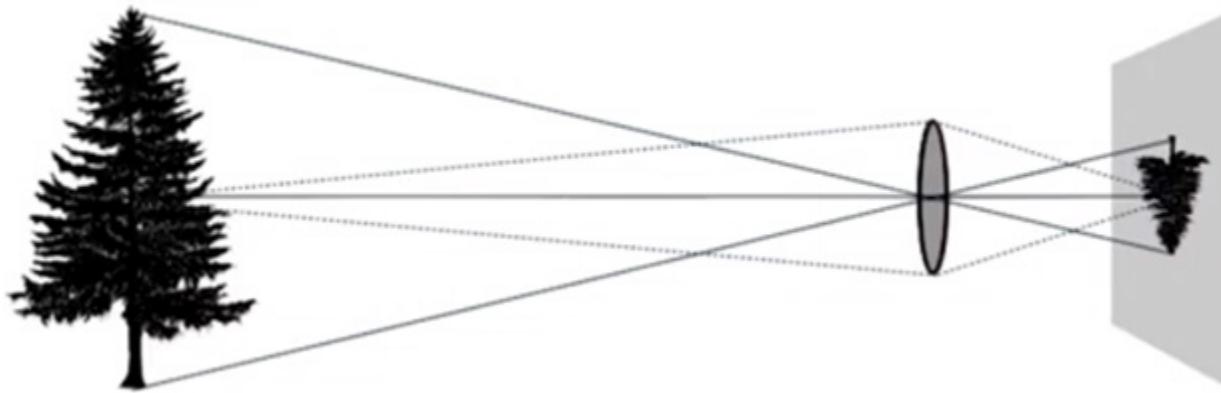


$f = 180\text{mm}$ , 35mm sensor

- \* Changing focal length allows us to move back, and still capture the scene
- \* Changing viewpoint causes perspective changes
- \* See the "vertigo" effect from Hitchcock movies

<https://www.wired.com/2012/12/how-to-make-a-hobbit-with-forced-perspective/>

### How are images focused?





# Exposure



Exposure = Irradiance  $\times$  Time

$$H = E \times T$$

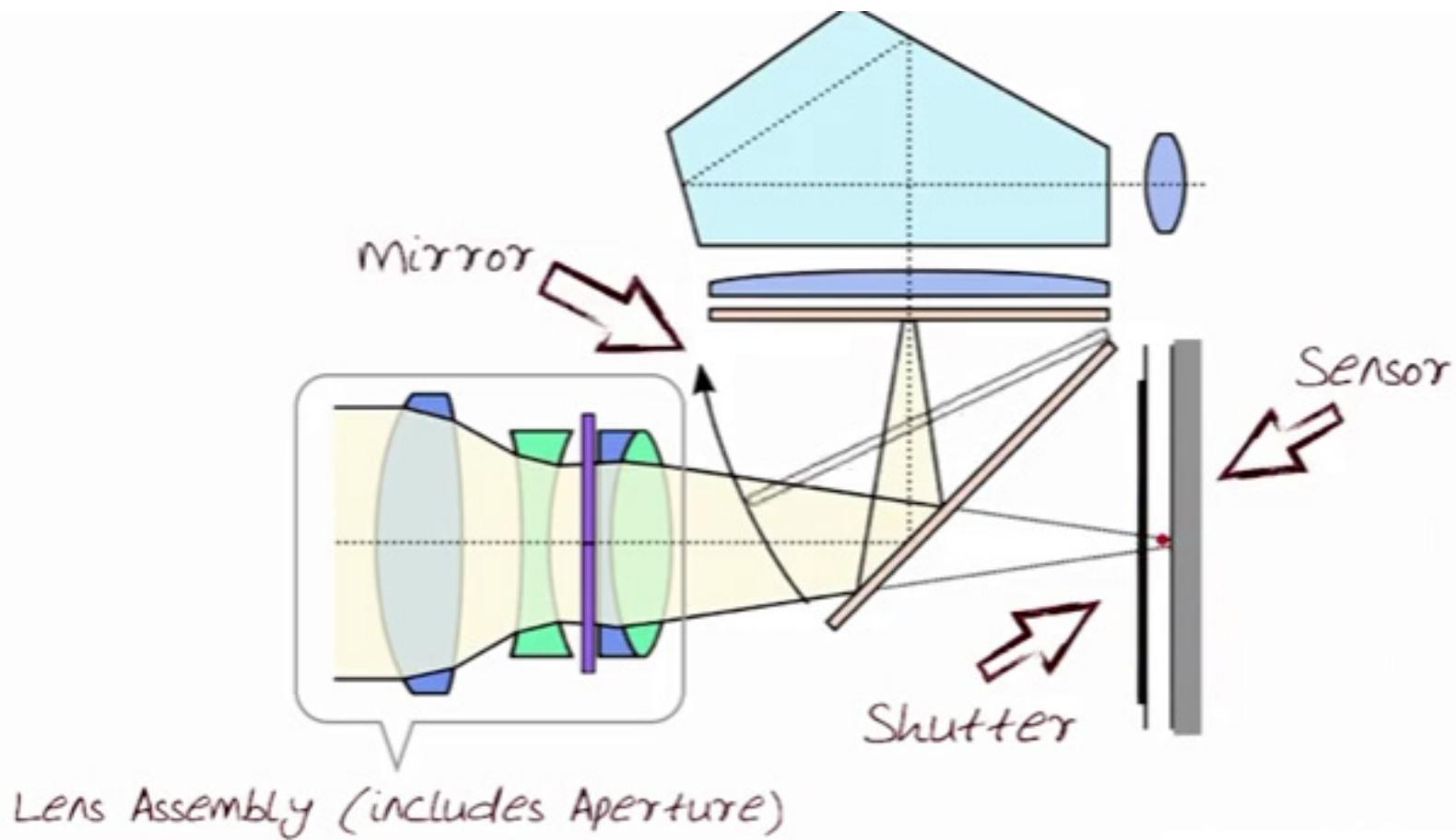
Irradiance ( $E$ )

- \* Amount of light falling on a unit area of sensor per second
- \* Controlled by lens aperture

Exposure Time ( $T$ )

- \* How long the shutter is kept open

### Exposure



<https://www.youtube.com/watch?v=ptfSW4eW25g>

### Shutter Speed

- \* Amount of time the sensor is exposed to light
- \* Usually denoted in fractions of a second (1/2000, 1/1000, ..., 1/250, ..., 1/60, ..., 1/15, ..., 15, 30, Bulb)

## Shutter Speed



### Shutter Speed



## Shutter Speed



## Shutter Speed



## Shutter Speed



### Shutter Speed



<https://www.youtube.com/watch?v=yr3ngmRuGUc>



### Aperture



Irradiance on Sensor → The amount of light captured is proportional to the Area of the Aperture (opening):

$$\text{Area} = \pi \left( \frac{f}{2N} \right)^2$$

- \*  $f$  is the focal length. What is the diameter of the Aperture?
- \* Aperture Number  $N$  usually written as  $f/N$  [F-number]
- \* Aperture Number gives irradiance irrespective of the lens in use



### Aperture



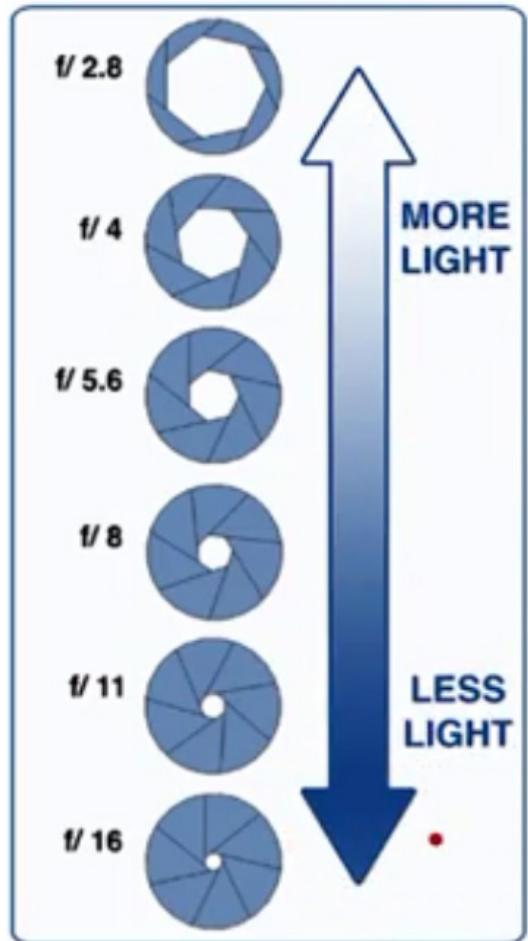
Aperture

$$Area = \pi \left( \frac{f}{2N} \right)^2$$

- \* Doubling  $N$  reduces  $A$  by  $2X$ , and therefore reduces light by  $4X$
- \* from  $f/2.8$  to  $f/5.6$  cuts light by  $4X$
- \* to cut light by  $2X$ , increase  $N$  by  $\sqrt{2}$

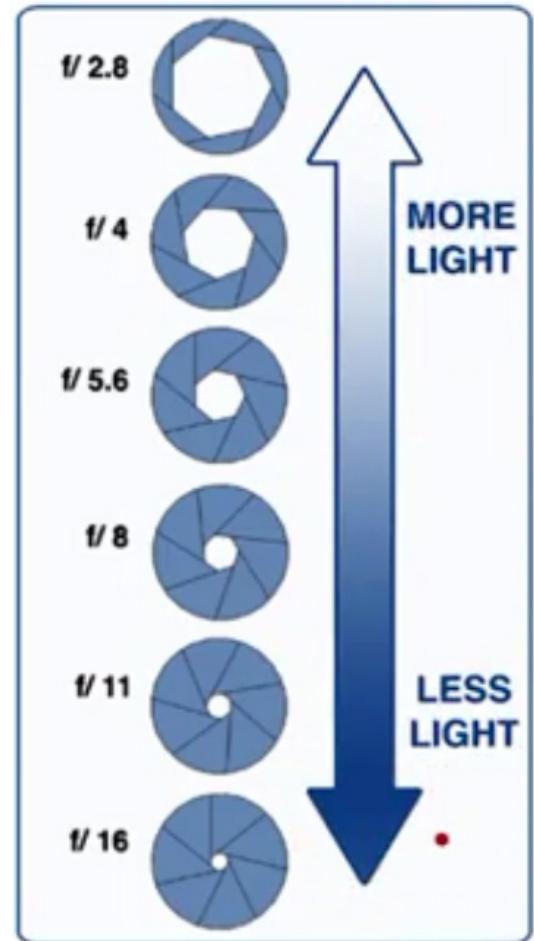


### Aperture



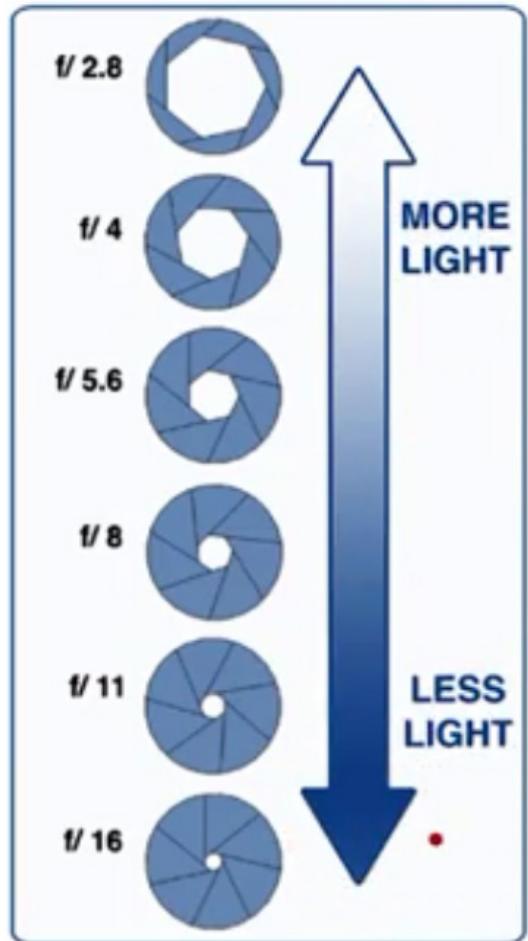


### Aperture



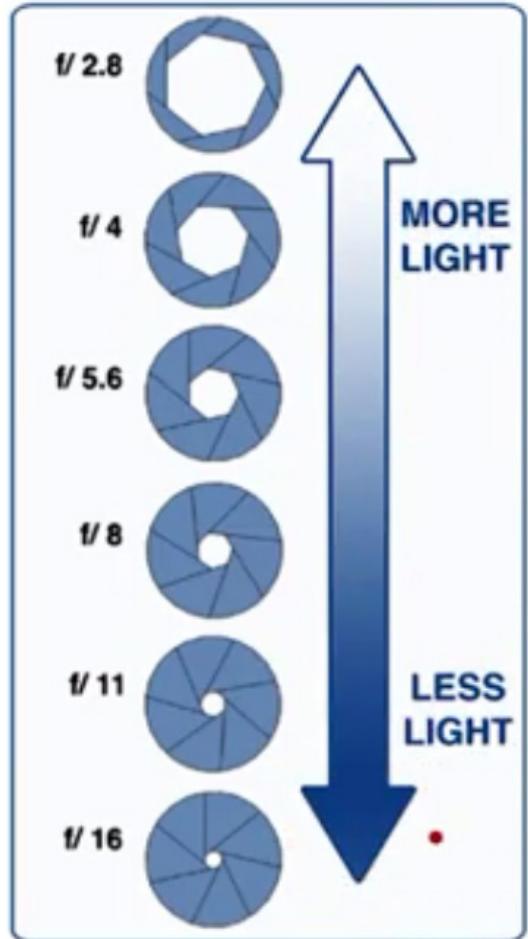


### Aperture



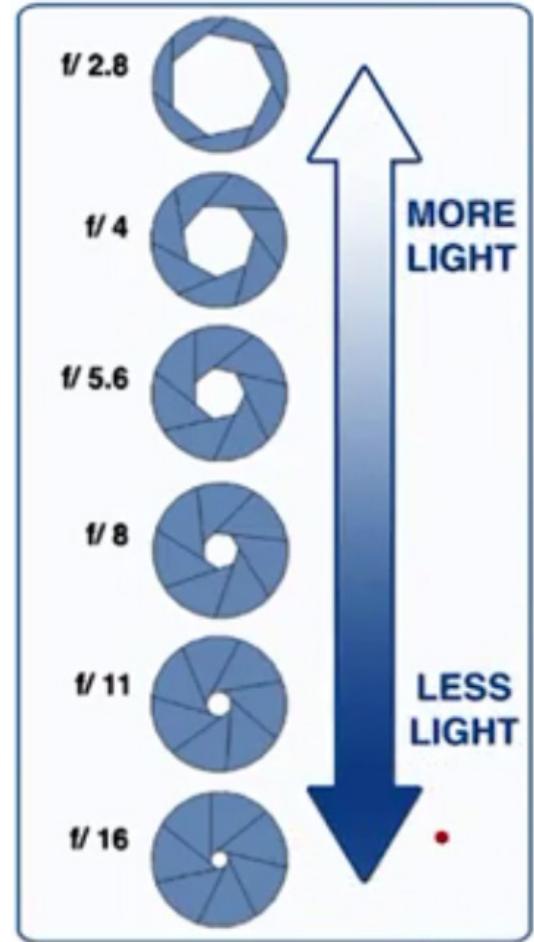


### Aperture



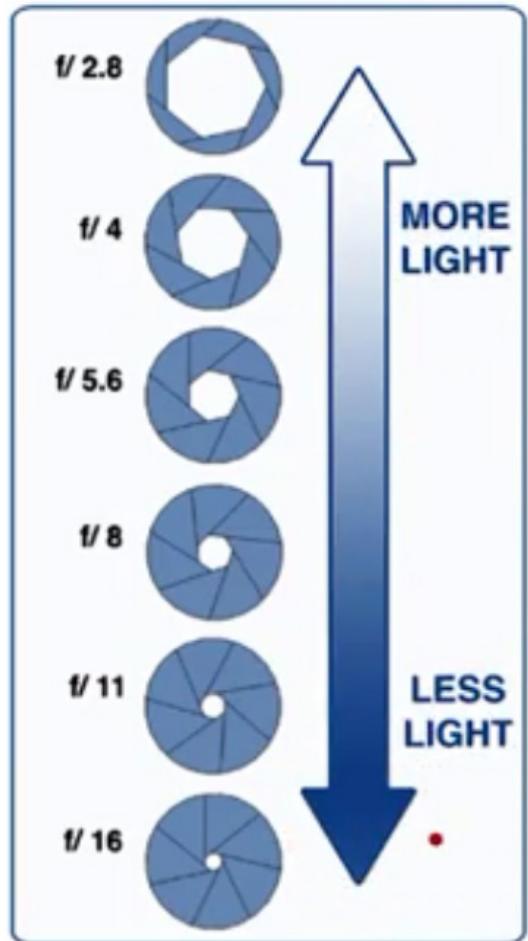


### Aperture



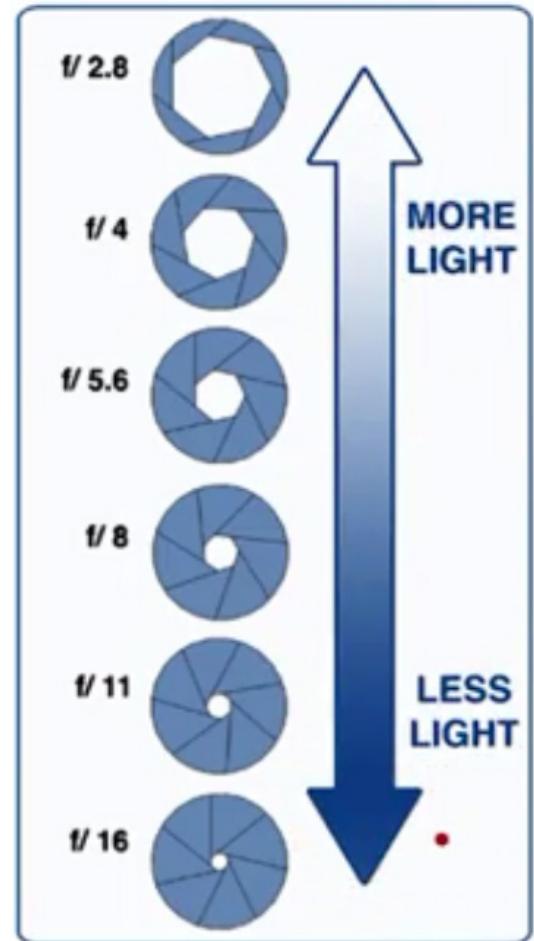


### Aperture





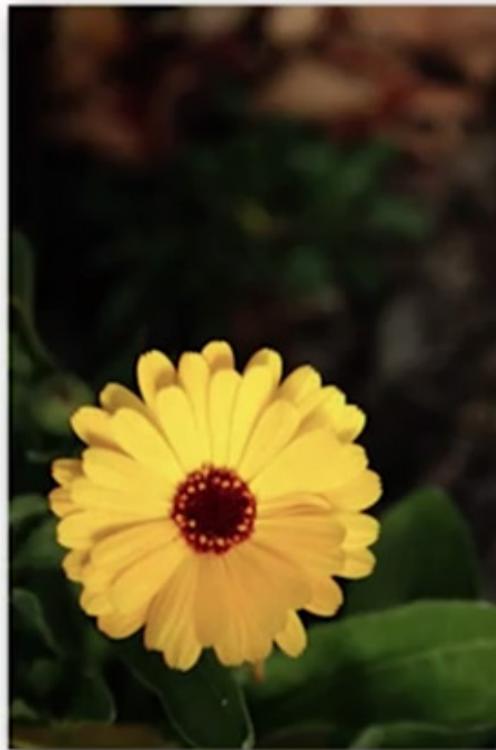
### Aperture



### Sensor Gain - ISO



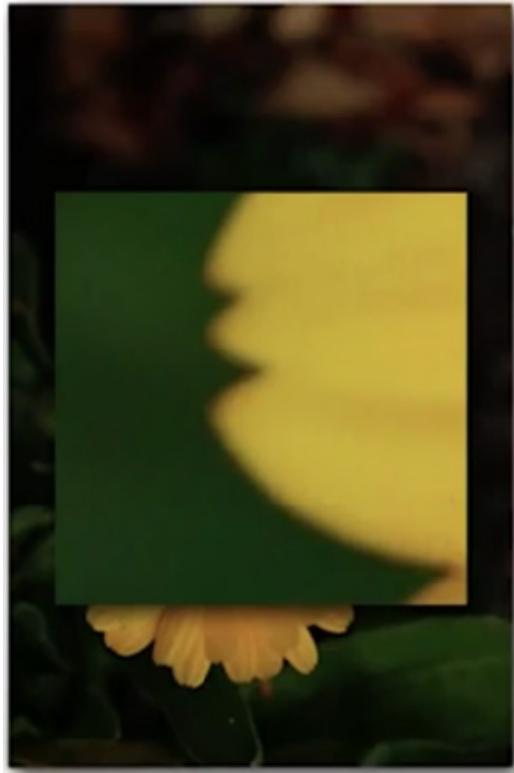
ISO 100



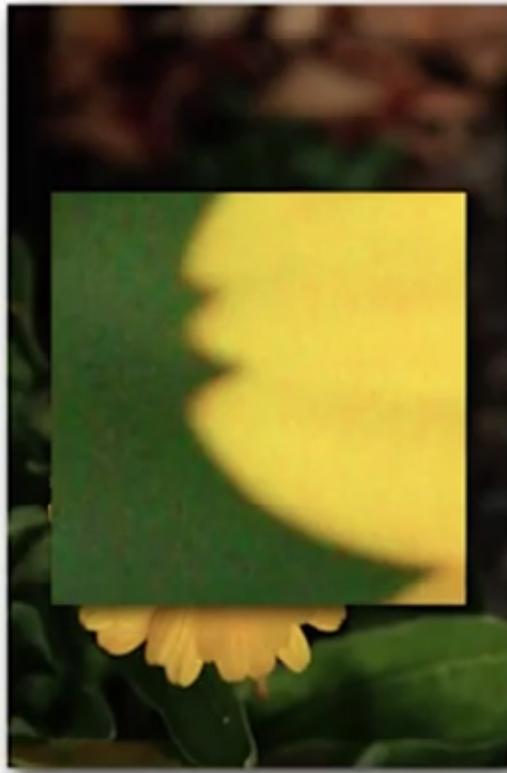
ISO 1600

- \* Third Variable in getting the right Exposure
- \* Film: Sensitivity vs. Grain (of film)
- \* Digital: Sensitivity vs. Noise (of sensor)
- \* Linear: 200 ISO needs half the light of 100 ISO

### Sensor Gain - ISO



ISO 100



ISO 1600

- \* Third Variable in getting the right Exposure
- \* Film: Sensitivity vs. Grain (of film)
- \* Digital: Sensitivity vs. Noise (of sensor)
- \* Linear: 200 ISO needs half the light of 100 ISO

# White Balance

**Human Color Constancy:**

**QUIZ!**

How would you define color constancy?

Reproducing this behavior in our cameras is a problem called  
**Machine Color Constancy**

**But wait!**

**QUIZ!**

If humans have color constancy, why to try  
to reproduce it in digital devices?

**Assumptions:**

- All objects in the scene are **flat**.
- All objects are considered **Lambertian** (no specularities, diffuse reflection)
- Scene uniformly illuminated with a **single global illuminant**.

## White Balance

$$R = \int_{380}^{740} r(\lambda) \times I(\lambda) \times S(\lambda) d\lambda$$

$$G = \int_{380}^{740} g(\lambda) \times I(\lambda) \times S(\lambda) d\lambda$$

$$B = \int_{380}^{740} b(\lambda) \times I(\lambda) \times S(\lambda) d\lambda$$



$$R = I(\lambda_R)S(\lambda_R)$$

$$G = I(\lambda_G)S(\lambda_G)$$

$$B = I(\lambda_B)S(\lambda_B)$$

$$\begin{bmatrix} R \\ G \\ B \end{bmatrix} = \begin{bmatrix} I(\lambda_R) & 0 & 0 \\ 0 & I(\lambda_G) & 0 \\ 0 & 0 & I(\lambda_B) \end{bmatrix} \begin{bmatrix} S(\lambda_R) \\ S(\lambda_G) \\ S(\lambda_B) \end{bmatrix}$$

$$S(\lambda_R) = \frac{R}{I(\lambda_R)}, \quad S(\lambda_G) = \frac{G}{I(\lambda_G)}, \quad S(\lambda_B) = \frac{B}{I(\lambda_B)}$$

## White Balance

$$\begin{bmatrix} R' \\ G' \\ B' \end{bmatrix} = \begin{bmatrix} \frac{1}{I(\lambda_R)} & 0 & 0 \\ 0 & \frac{1}{I(\lambda_G)} & 0 \\ 0 & 0 & \frac{1}{I(\lambda_B)} \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

1. Estimate illuminant components
2. Invert for white-balanced triplet

## Automatic White Balance

### Gray World

Colors in the scene are sufficiently varied, and the average of reflectances is gray. In other words, reflectances are uniformly distributed over  $[0, 1]$ . For each waveband, **impose average 0.5**.

### White Patch

The brightest object in a scene becomes the reference white, it is perceived as white. An estimate of illuminant color is obtained by finding the brightest pixel in the scene. **Impose white presence.**

### Sensors

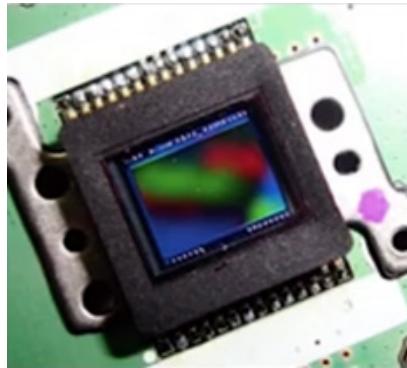


### Photographic Film

- \* Film and Digital Cameras are the same
- \* There have been significant improvements in actuators, and lenses
- \* Difference is how light is trapped and preserved
- \* Chemical process for Film, and Electronic for Digital capture the moment in Time and Space



### External Flash Disk



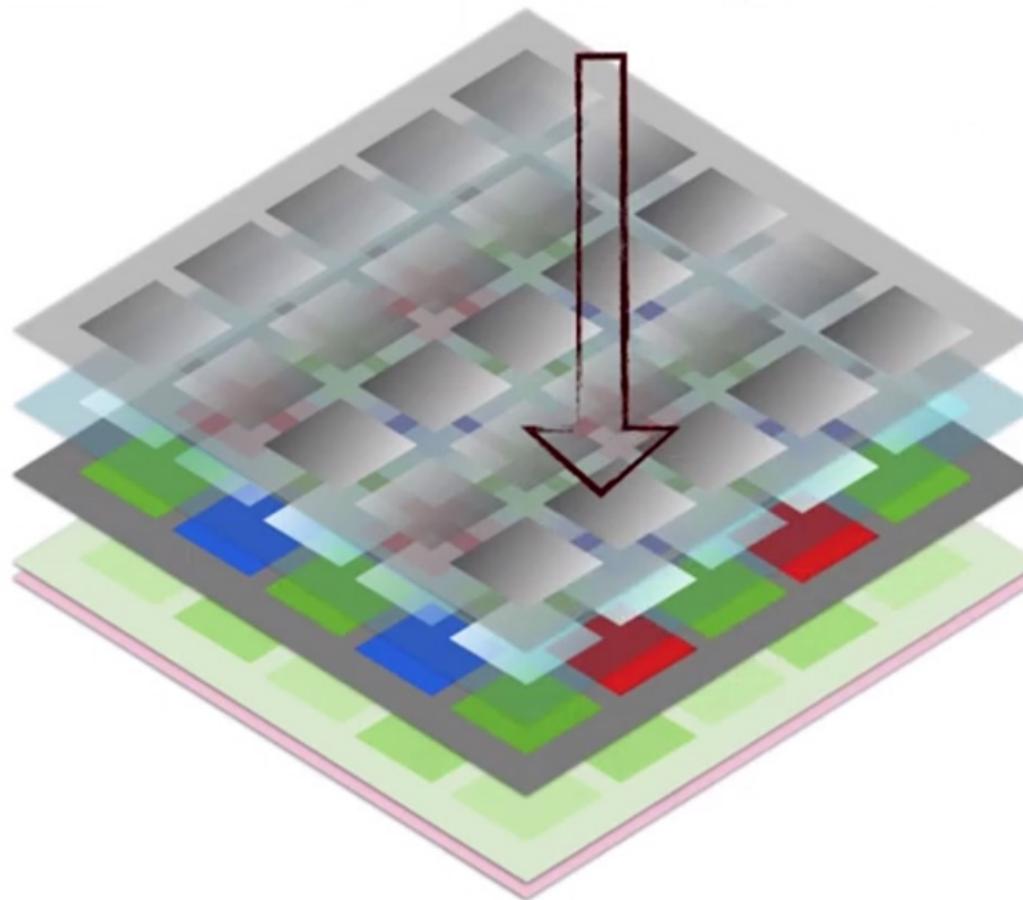
CCD

### Digital Sensors - CCD

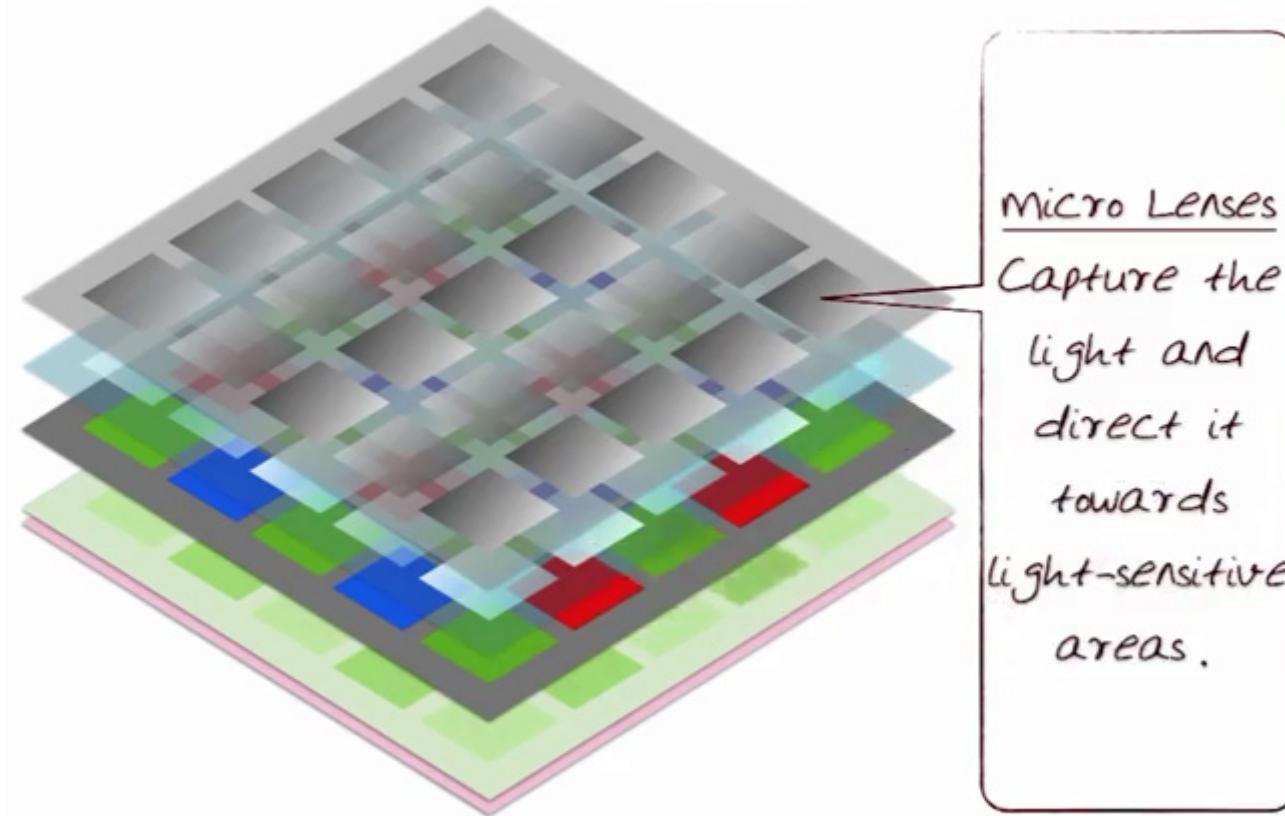
- \* CCD: Charge-Coupled Device, a device for converting electrical charge, into a digital value
- \* Pixels are represented by capacitors, which convert and store incoming photons as electron charges
- \* Willard Boyle and George E. Smith, 1969 (Won a Nobel Prize in Physics in 2009).



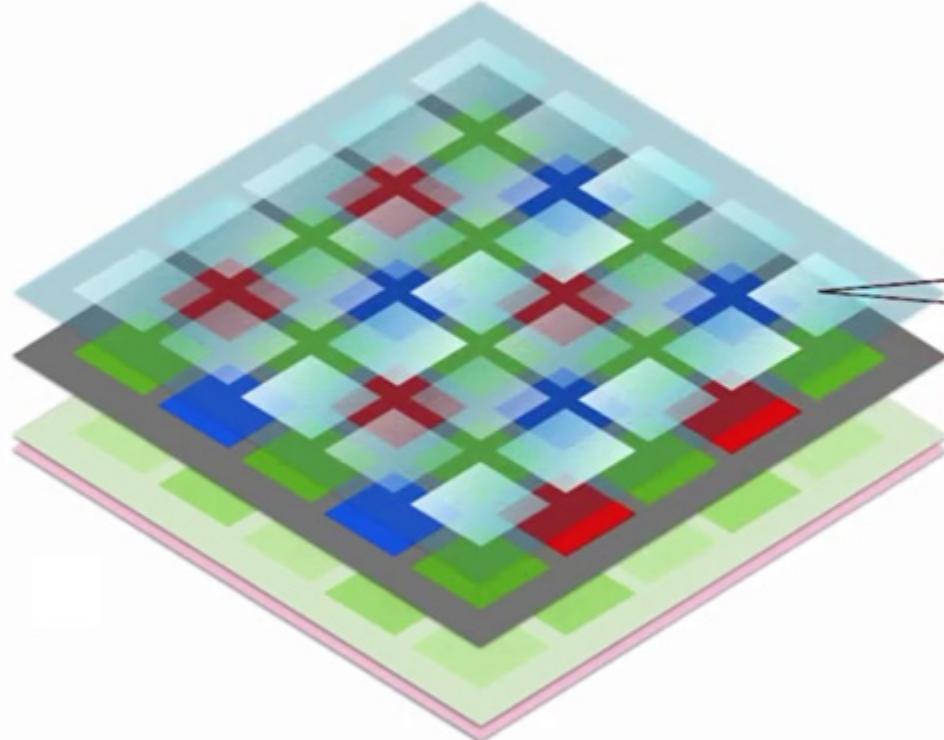
## Digital Sensors - CCD



### Digital Sensors - CCD

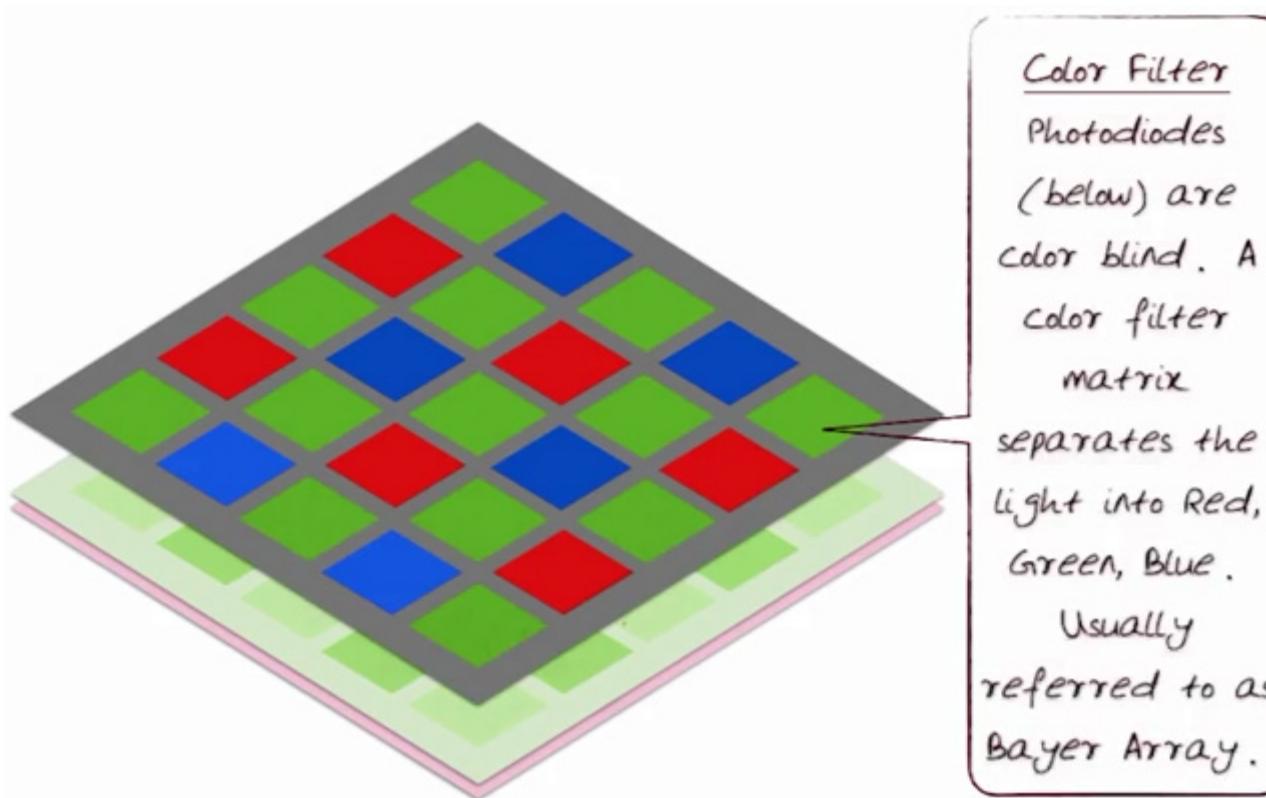


### Digital Sensors - CCD

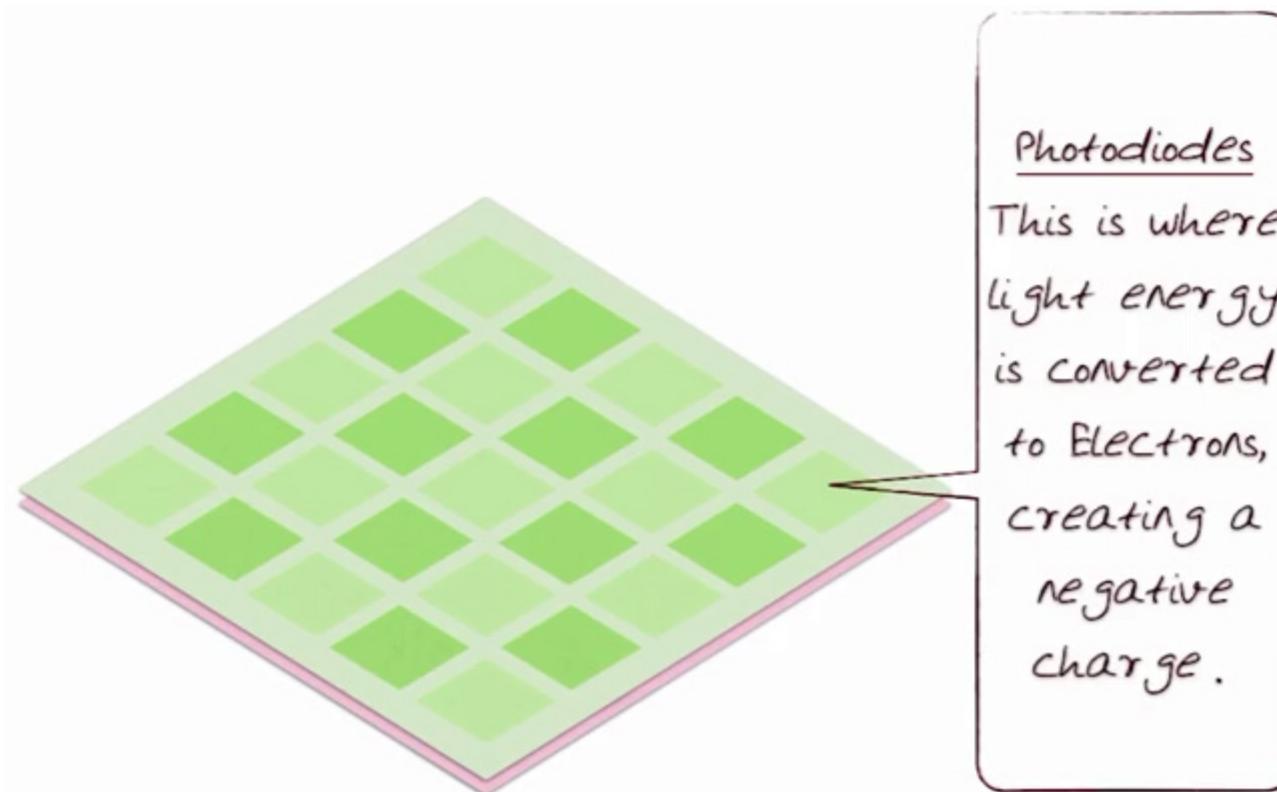


Hot mirror  
Lets visible  
light pass, but  
reflects light  
in the invisible  
part of the  
spectrum  
(depends on  
what kind of  
light to  
capture)

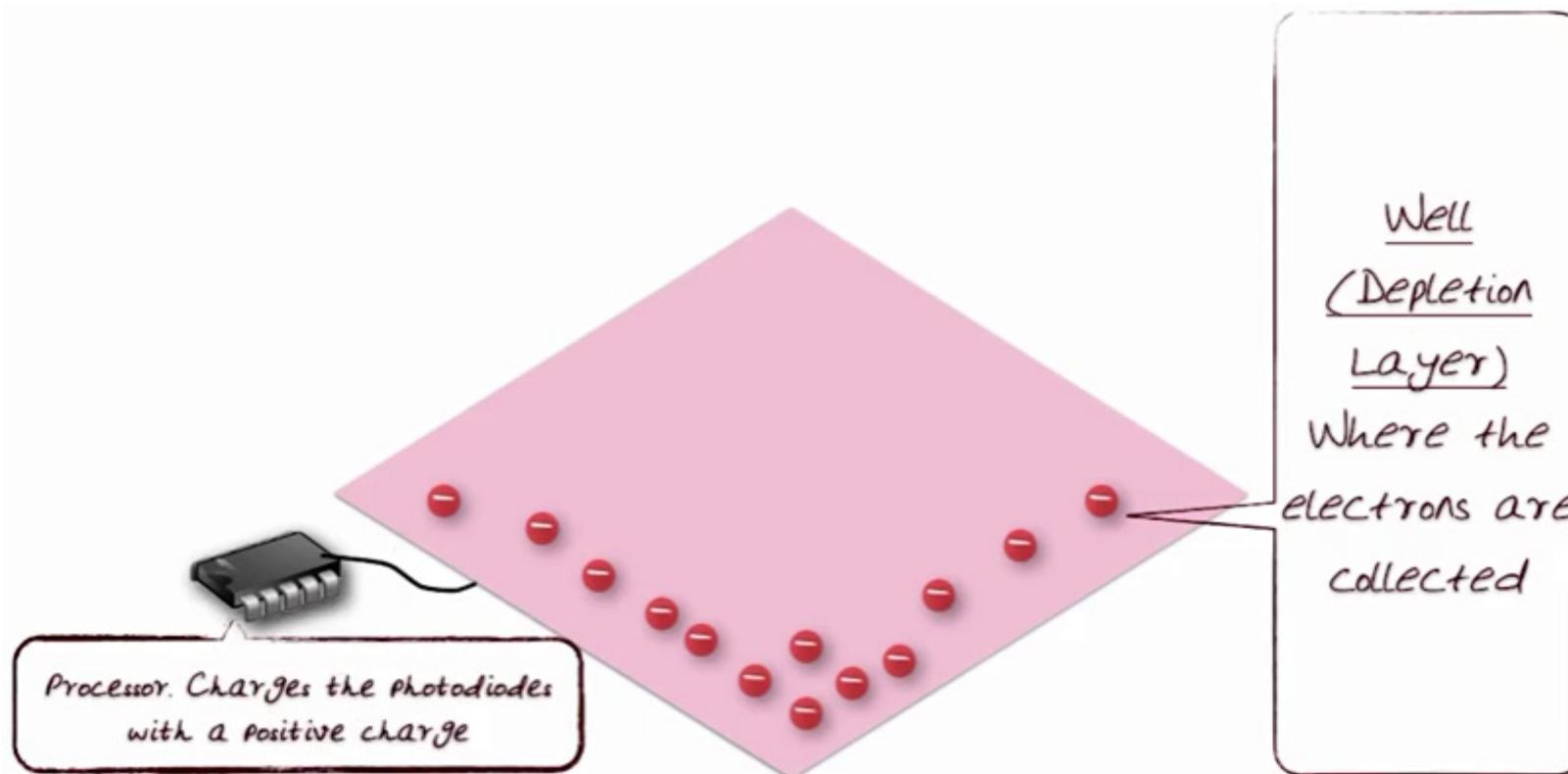
### Digital Sensors - CCD



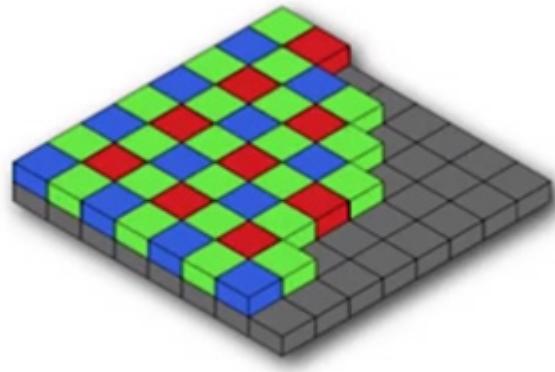
### Digital Sensors - CCD



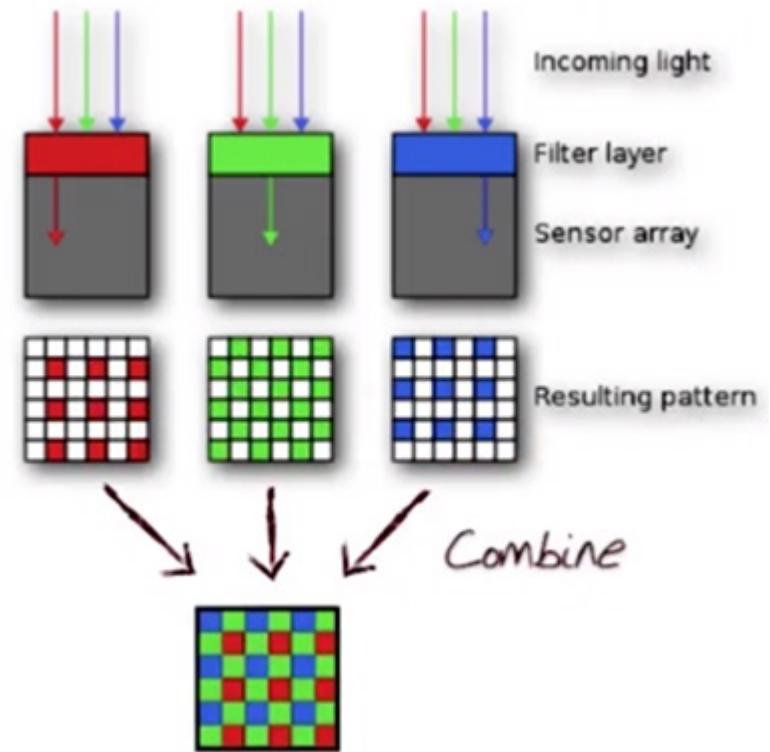
### Digital Sensors - CCD



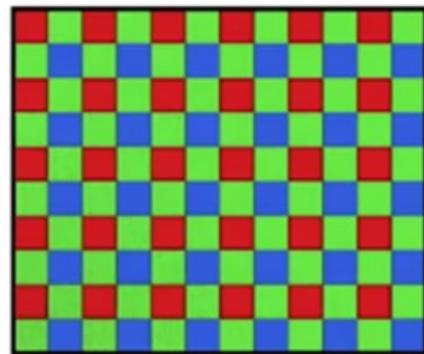
## Bayer Filtering



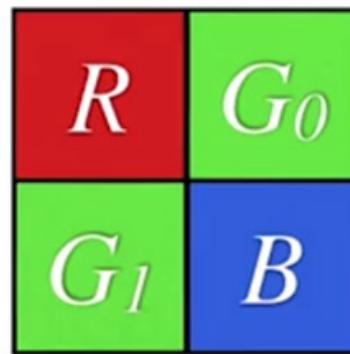
Bayer filter on a sensor



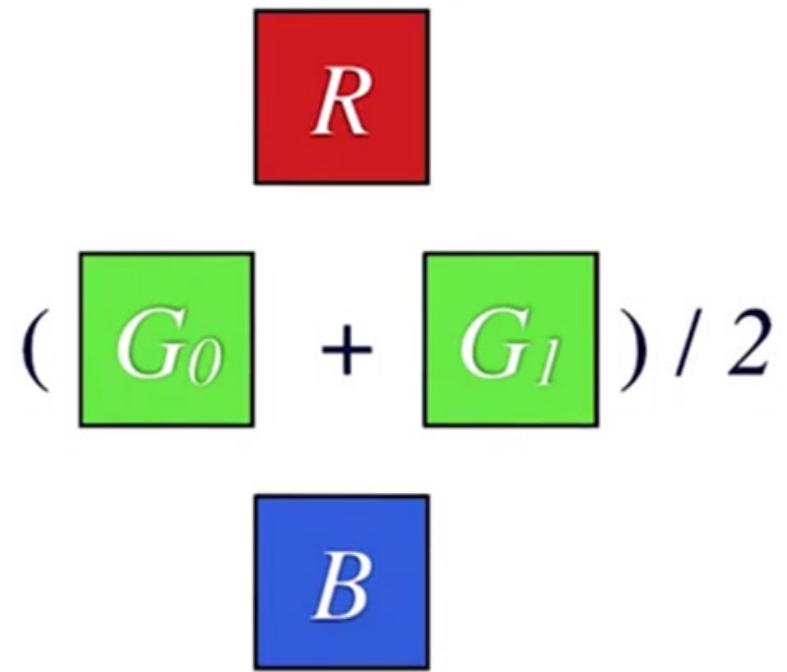
## Bayer to RGB: Demosaicking



Raw input in Bayer  
mosaic format

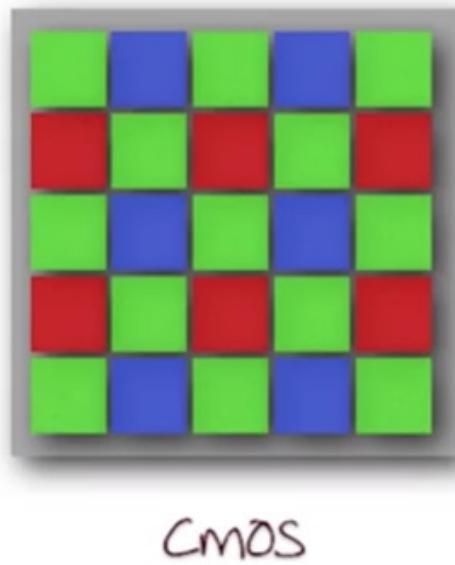
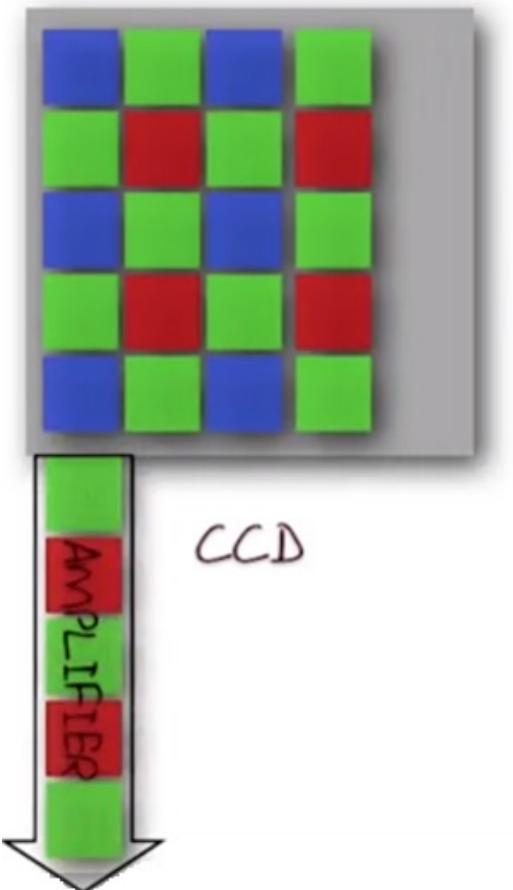


A  $4 \times 4$  subset



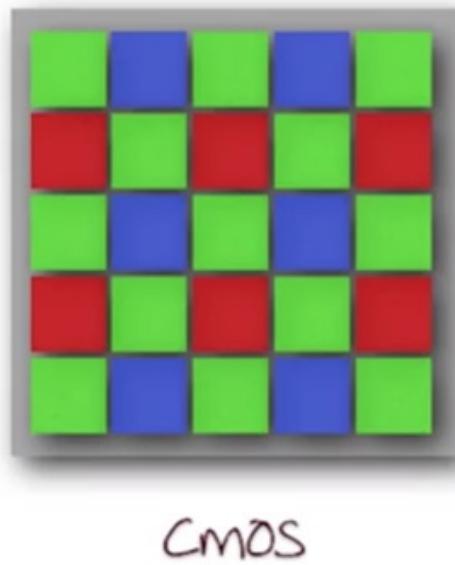
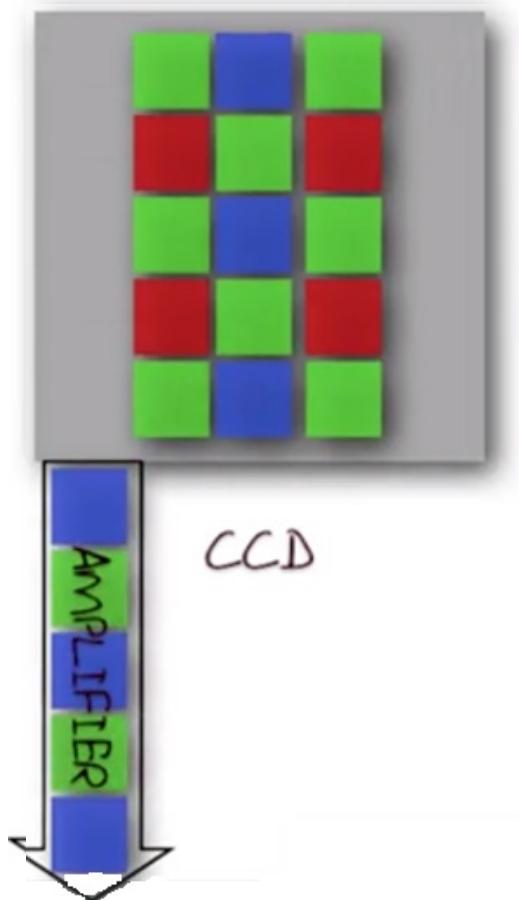
Resulting RGB triple

### Digital Sensors - CMOS



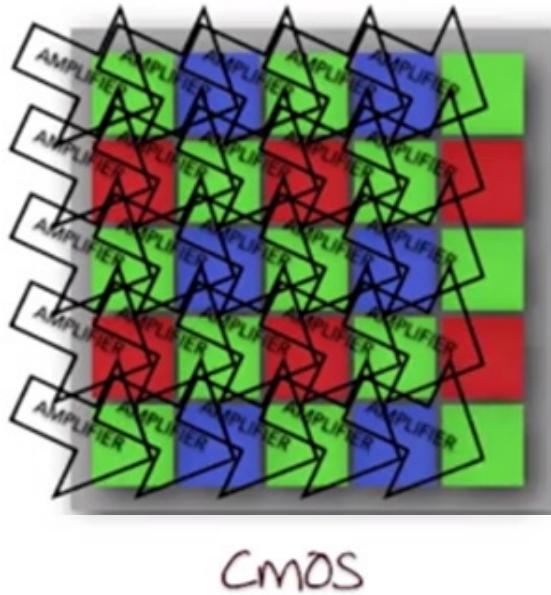
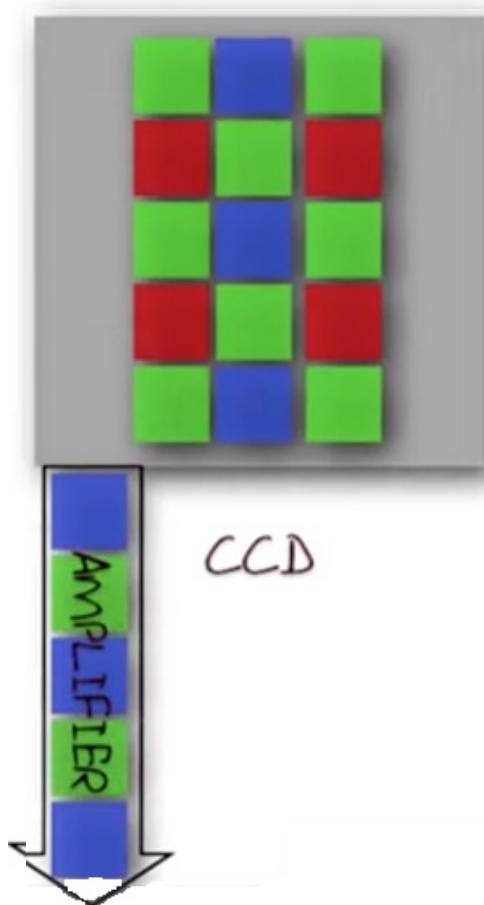
- \* CMOS: Complementary metal Oxide Semiconductor
- \* Photosites in CCD are passive and do no "work"

### Digital Sensors - CMOS



- \* CMOS: Complementary metal Oxide Semiconductor
- \* Photosites in CCD are passive and do no "work"

### Digital Sensors - CMOS



- \* CMOS: Complementary metal Oxide Semiconductor
- \* Photosites in CCD are passive and do no "work"

### Recording in RAW

- \* Contains minimally processed data from the sensor
- \* Image encoded in a device-dependent colorspace
- \* Captures radiometric characteristics of the scene
- \* Viewable image from the camera's sensor data
- \* Like a photographic negative
  - \* has a wider dynamic range or color; preserves most of the information of the captured image

# Color Transformation

For the camera to accurately capture colors matching our perception, the color triplets obtained by the camera sensor should correspond to the cone responses of the human visual system.

However, for a series of practical reasons, emulating cone responses is not practical for image capture.

Still, our goal is that the stimulus the scene would have produced in the human visual system are estimated as accurately as possible.



**QUIZ!**

**Do you remember anything about this transformation?**

## Color Transformation



Transform the (R,G,B) values of the sensor into (X,Y,Z) tristimulus values.

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix} = A \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$



Colorimetric matrix

This process would imply computing a colorimetric matrix for every single triplet (R,G,B), really unpractical.

# Color Transformation

Cameras carry a unique (optimal) colorimetric matrix, computed as follows:

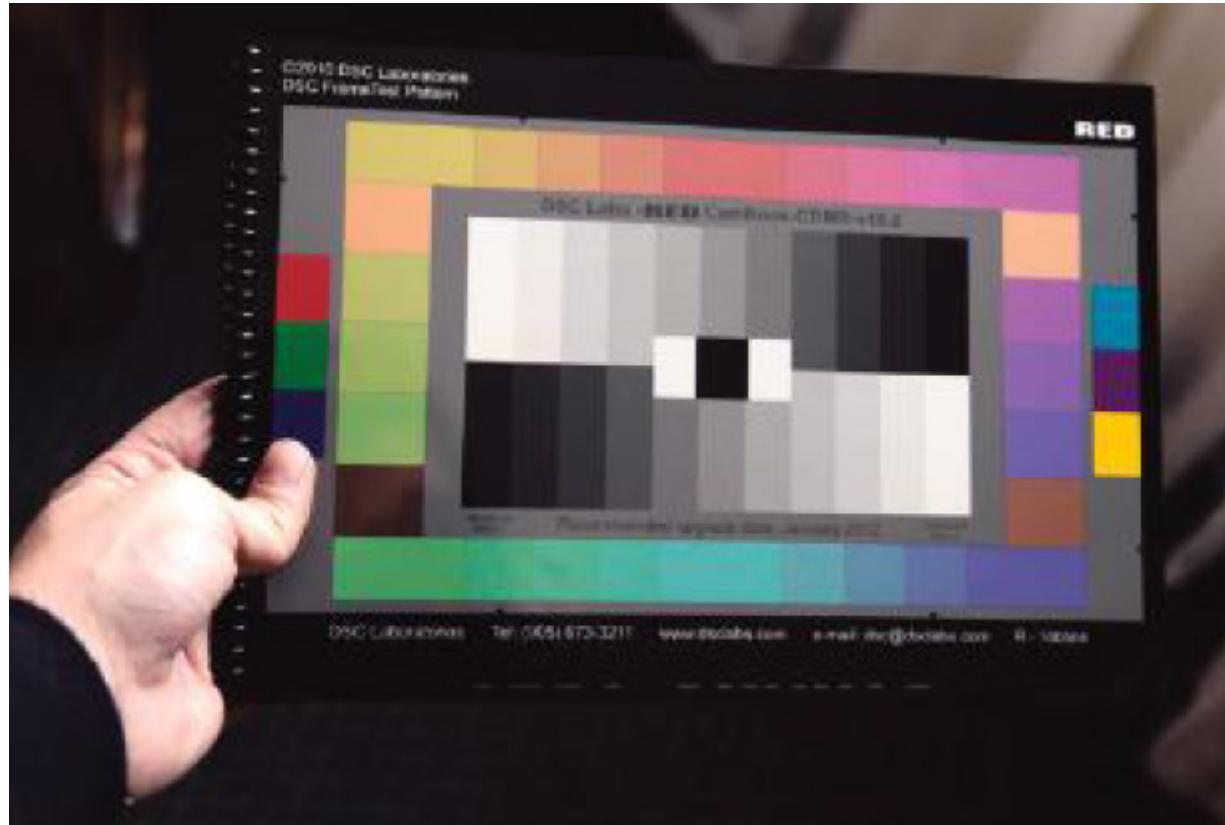
- Build a set of n test patches of representative or important colors.
- Under controlled conditions, with a known illuminant, measure the tristimulus values of the patches with a tristimulus colorimeter obtaining  $(X_i, Y_i, Z_i)$ , for  $1 < i < n$ .
- Under the same conditions, use the camera to measure the  $(R, G, B)$  values of the patches, obtaining  $(R_i, G_i, B_i)$ , for  $1 < i < n$ .
- Compute an optimal matrix A that minimizes the color difference between the target stimulus and corresponding RGB2XYZ transformed points.

Ideally a different matrix should be used for each different scene illuminant.

The above process finds the best colorimetric matrix for the calibration patch set under a given illuminant, and many cameras (most consumer models) use only this one matrix.

Some cameras come with several pre-set matrices computed under different illuminations. e.g., fluorescent light, daylight, halogen, etc.

# Color Transformation



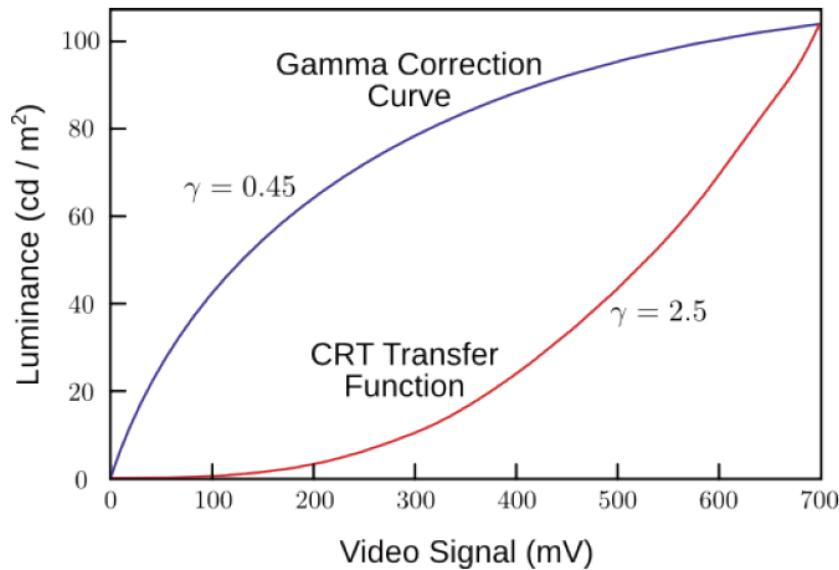
## Color Transformation

A last step that is given before projection into desired display is **Gamma Correction**.

At the onset of broadcast television it was observed that CRTs produced luminance as a non-linear, power function of the device's voltage input.

For correct luminance reproduction on a TV set, the camera's luminance signal needed to be non-linearly scaled with a power function with the inverse exponent of the CRT's power function.

$$L = V^\gamma, \gamma \equiv 2.5$$



But why is gamma correction still used today, when CRTs have become obsolete?

Humans have a perceptual response to luminance that is also non-linear: perceived lightness is roughly the 0.42 power of luminance. Differences in the dark parts of an image are more noticeable than differences of the same amount on bright parts of the same image. We want more bits in dark areas!

# Outline of these lectures - Week 1

- **B1 - Human Vision System: The Retina and the Brain**
  1. Basics of Human Perception.
  2. Perception and Color. Measuring Color.
  3. Color Spaces. Perceptually Uniform Color Spaces.
  4. Application: Color Image Segmentation. Superpixels.
  5. Introduction to the Retinex model for Human Perception.
  6. Introduction to Computational Image Processing.
  7. Application: A first example - Image Composition.
- **B2 - Digital Image Processing: The Camera**
  1. The Image Acquisition Pipeline.
  2. Cameras: Lenses, Exposure, Focus, White Balance, Sensors, Output.
  3. Unconventional Image Acquisition: Panorama.
  4. Application: Panorama Creation.