



# Vision Algorithms for Mobile Robotics

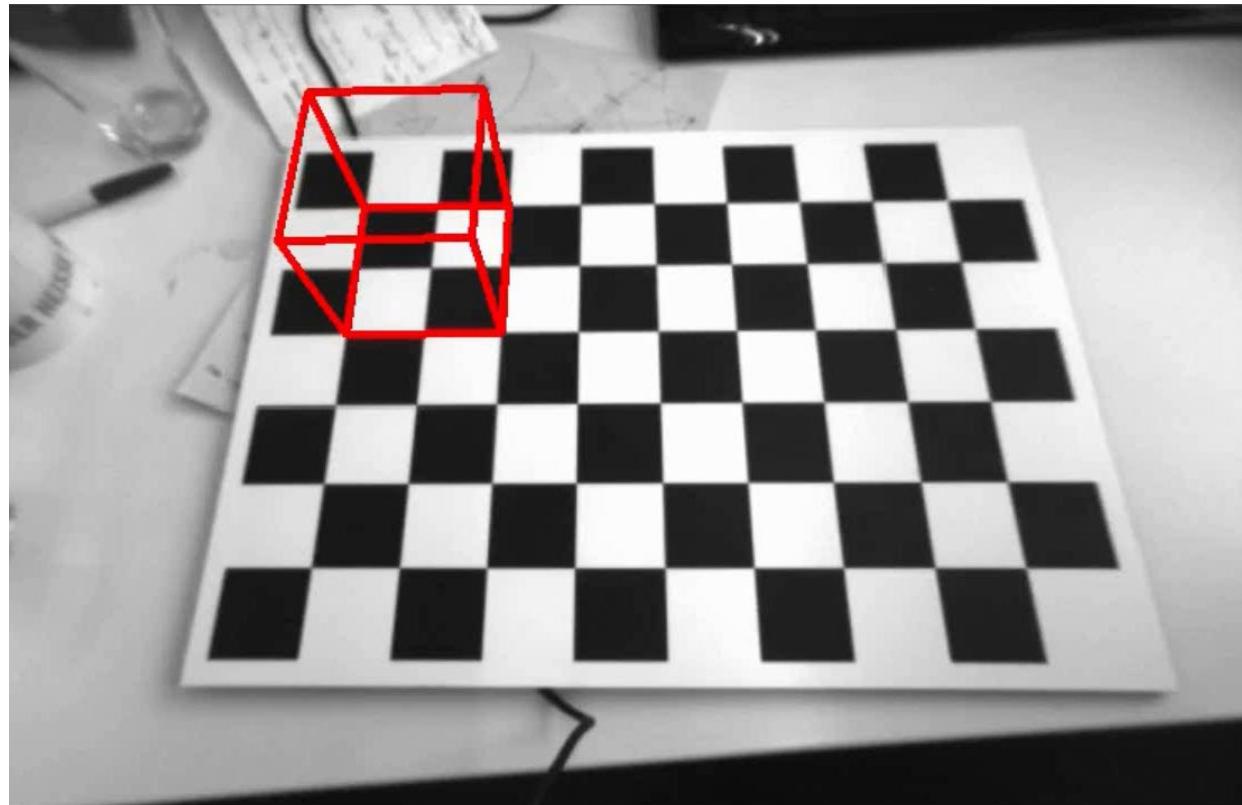
## Lecture 02 Image Formation

Davide Scaramuzza

<http://rpg.ifi.uzh.ch>

# Lab Exercise 1 - Today

Implement an augmented reality wireframe cube and practice the perspective projection

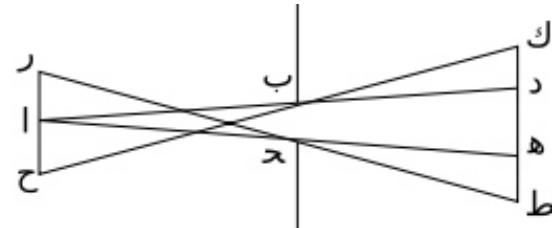


# Today's Outline

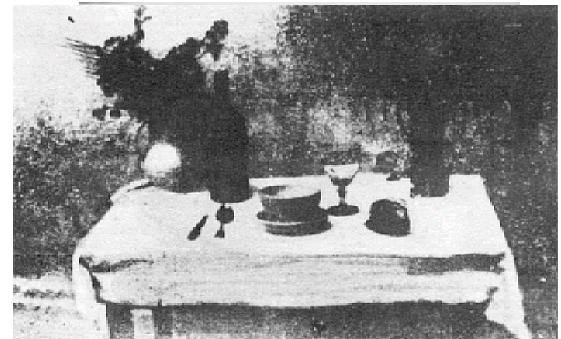
- Image Formation
- Other camera parameters
- Digital camera
- Perspective camera model
- Lens distortion

# Historical context

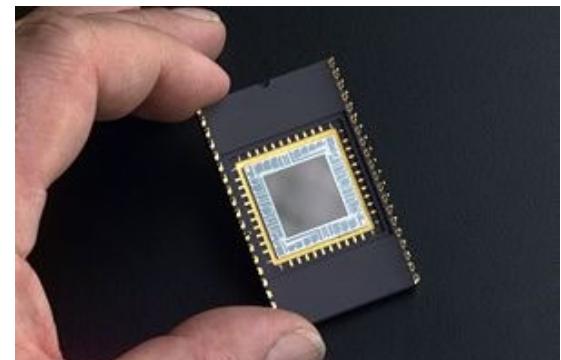
- **Pinhole model:** Mozi (470-390 BCE), Aristotle (384-322 BCE)
- Principles of optics (including lenses):  
Alhacen (965-1039)
- **Camera obscura:** Leonardo da Vinci (1452-1519), Johann Zahn (1631-1707)
- **First photo (heliography):** Joseph Nicéphore Niépce (1827)
- **Daguerreotypes** (1839)
- Photographic film (Eastman, 1888, founder of Kodak)
- Cinema (Lumière Brothers, 1895)
- Color Photography (Lumière Brothers, 1908)
- Television (Baird, Farnsworth, Zworykin, 1920s)
- **First consumer camera with CCD:**  
Sony Mavica (1981)
- **First fully digital camera:** Kodak DCS100 (1990)



Alhacen's notes



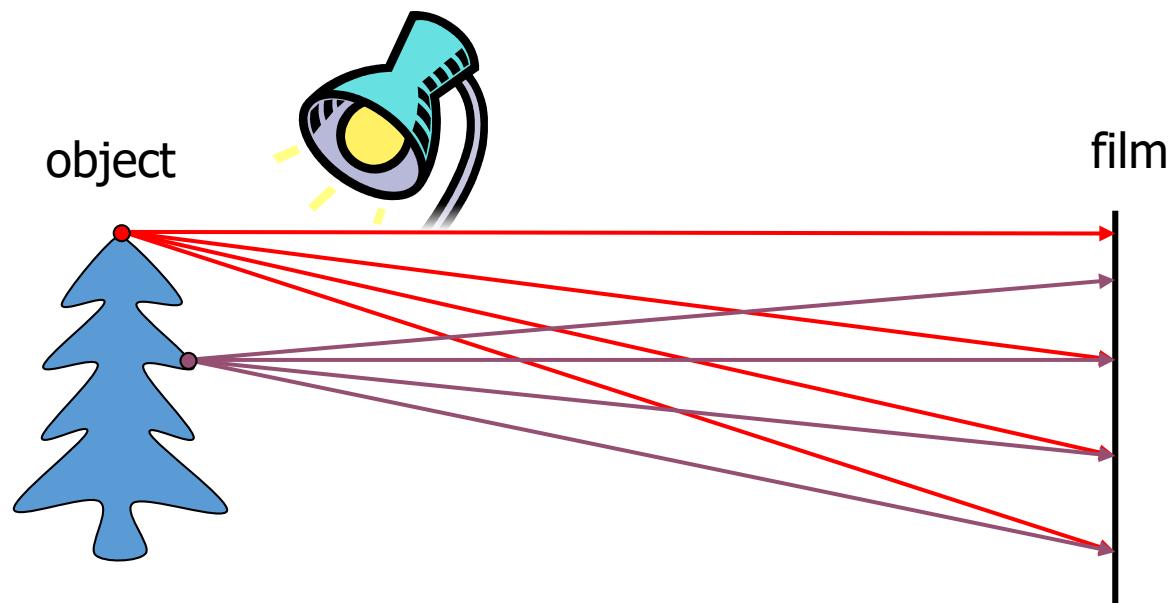
First surviving photo ever: "View from the Window at Le Gras", by Niépce, 1827



CCD chip

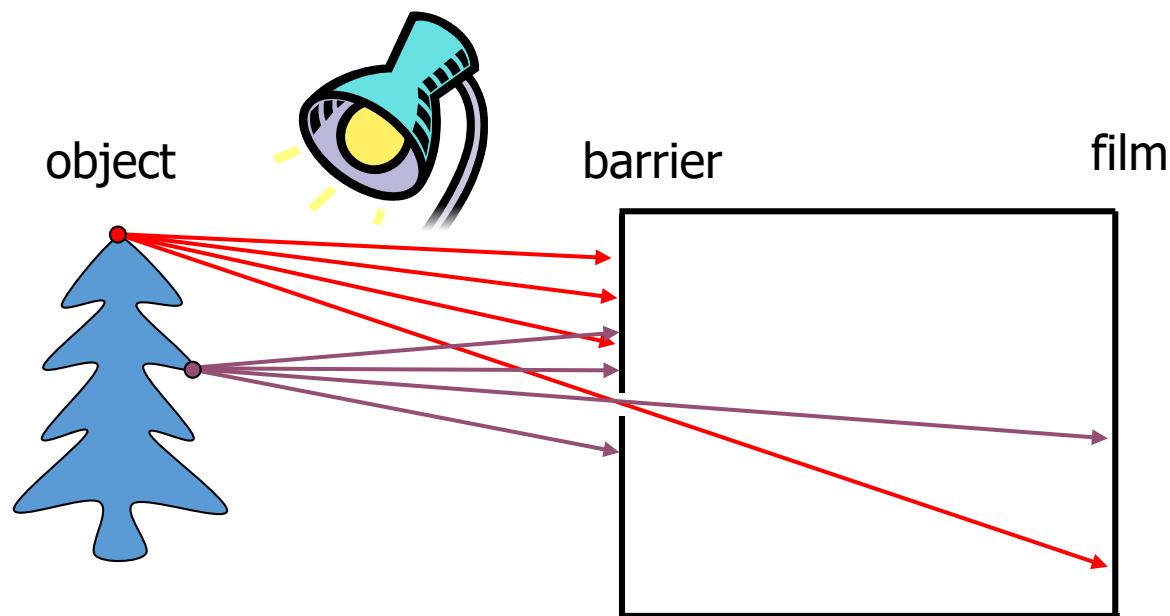
# How to form an image

- Would this work?



# Camera obscura

- Add a barrier to block off most of the rays
  - This reduces blurring
  - The opening is known as the **aperture**



# Home-made camera obscura

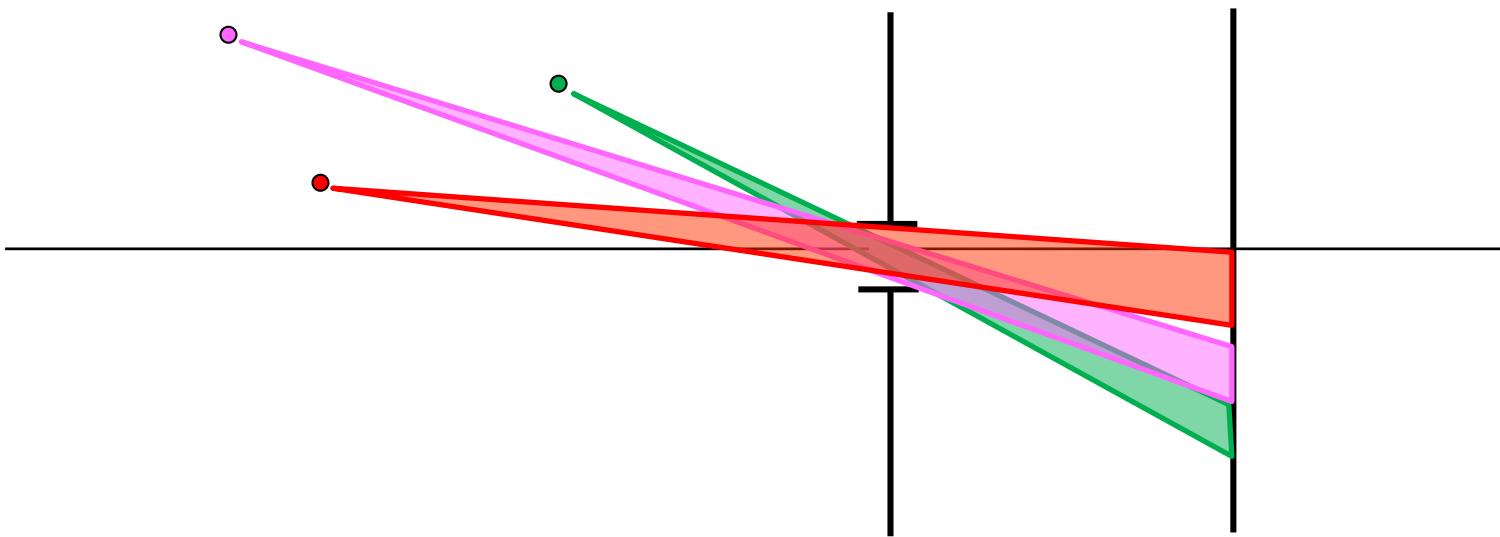
- Image is **inverted**
- **Depth of the room** is the **focal length**
- How can we **reduce the blur?**



How to build a camera obscura at home ([link](#))

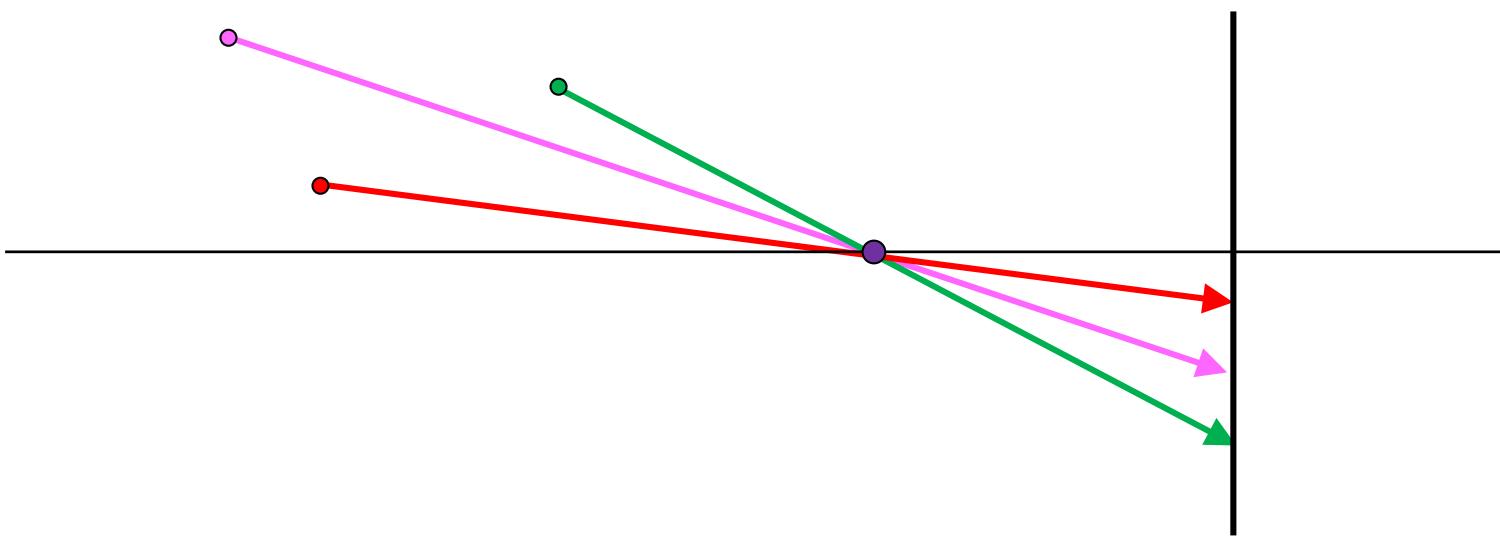
# Effects of the Aperture Size

- A large **aperture** makes the **image blurry** because a cone of light is let through from each world point



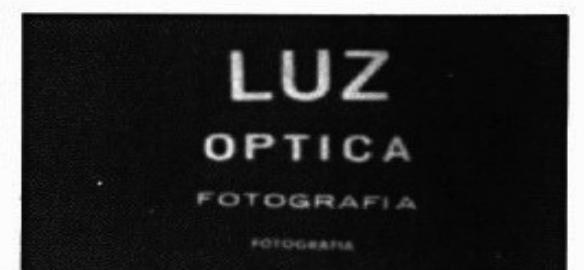
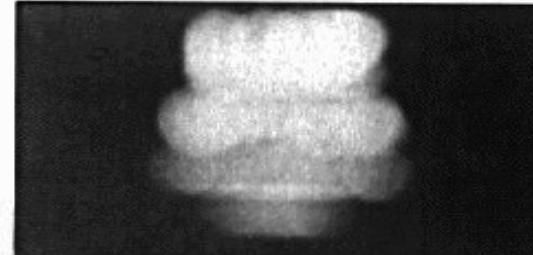
# Effects of the Aperture Size

- **Shrinking the aperture makes the image sharper**
- The ideal aperture is a **pinhole** that only lets through one ray of light from each world point



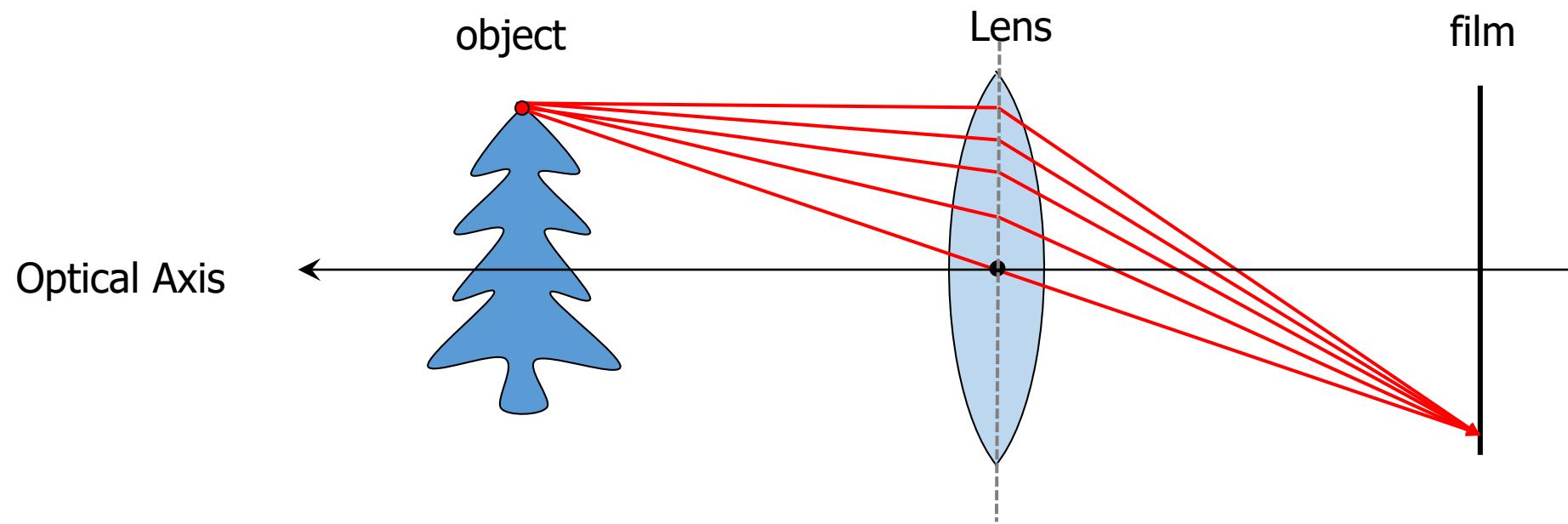
# Why not making the aperture as small as possible?

- With **small apertures**, less light gets through → must **increase exposure time**
- If aperture gets **too small**, **diffraction effects** start to appear



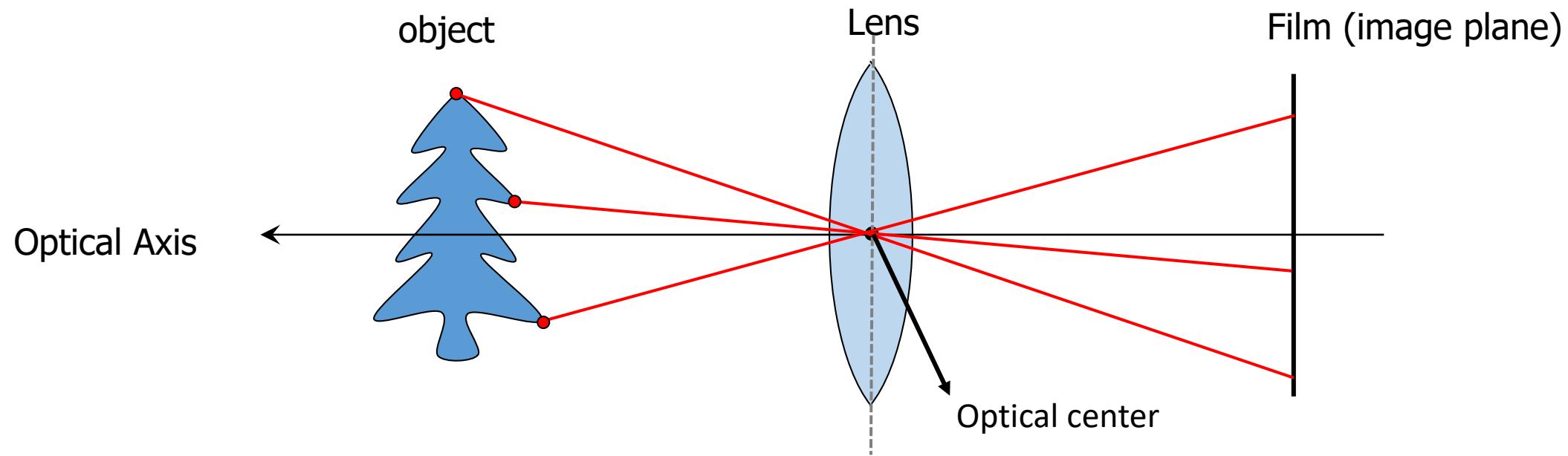
# Image formation using a converging lens

- A **thin converging lens** focuses light onto the film satisfying two properties:



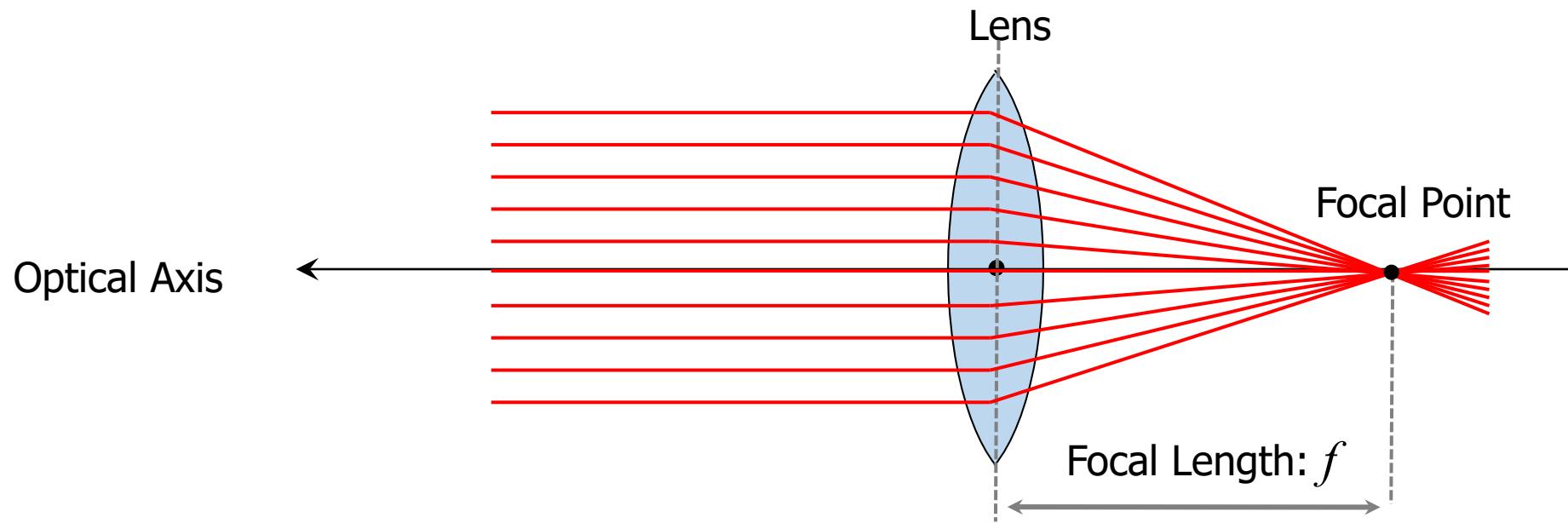
# Image formation using a converging lens

- A **thin converging lens** focuses light onto the film satisfying two properties:
  1. Rays passing through the **Optical Center** are not deviated



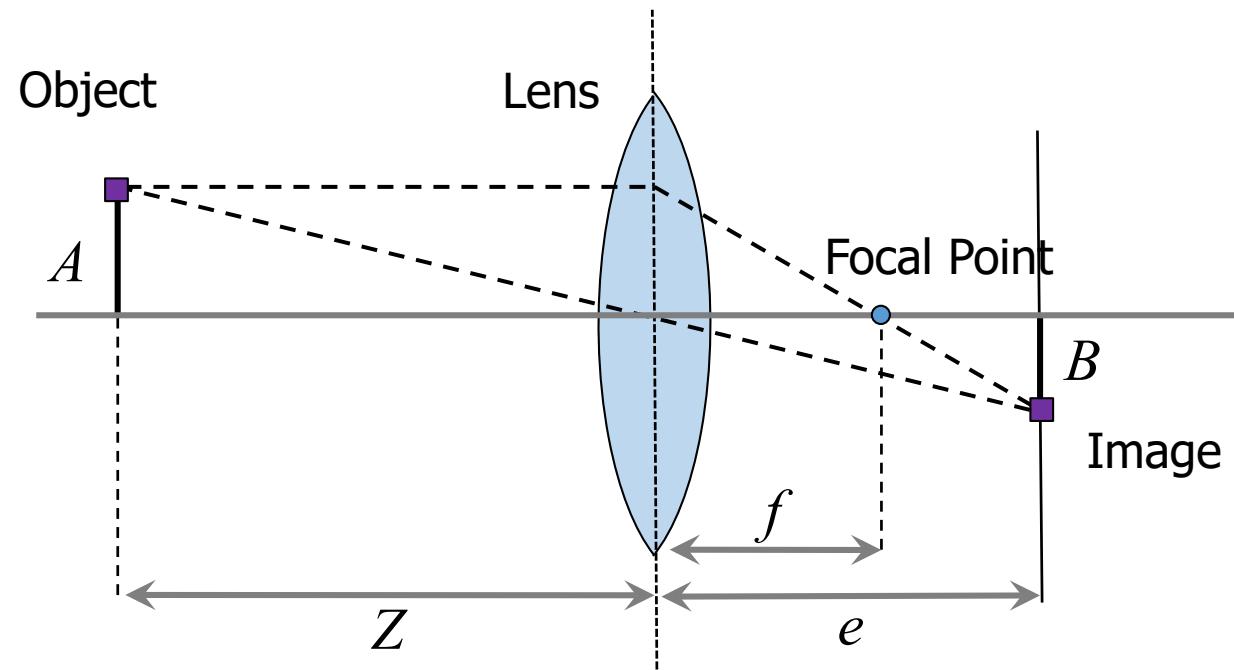
# Image formation using a converging lens

- A **thin converging lens** focuses light onto the film satisfying two properties:
  1. Rays passing through the **Optical Center** are not deviated
  2. All rays parallel to the **Optical Axis** converge at the **Focal Point**



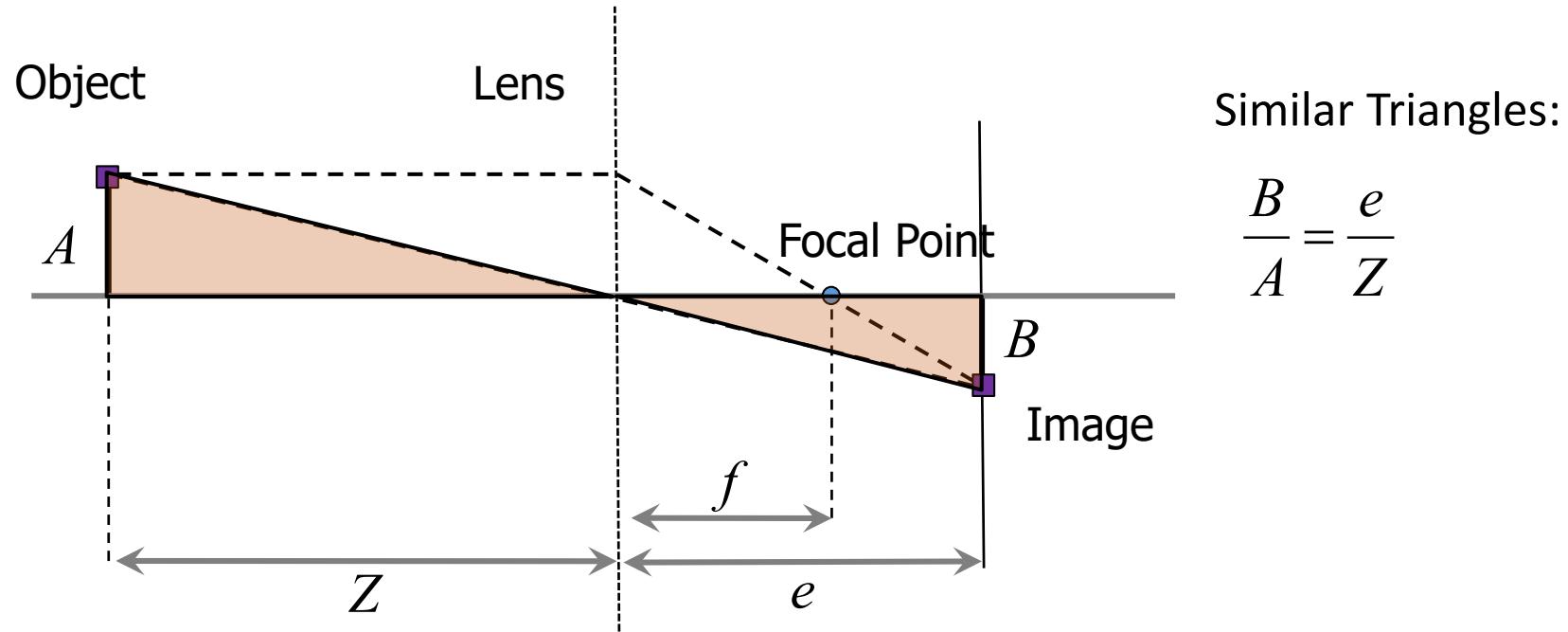
# Thin lens equation

- Find a relationship between  $f$ ,  $Z$ , and  $e$



# Thin lens equation

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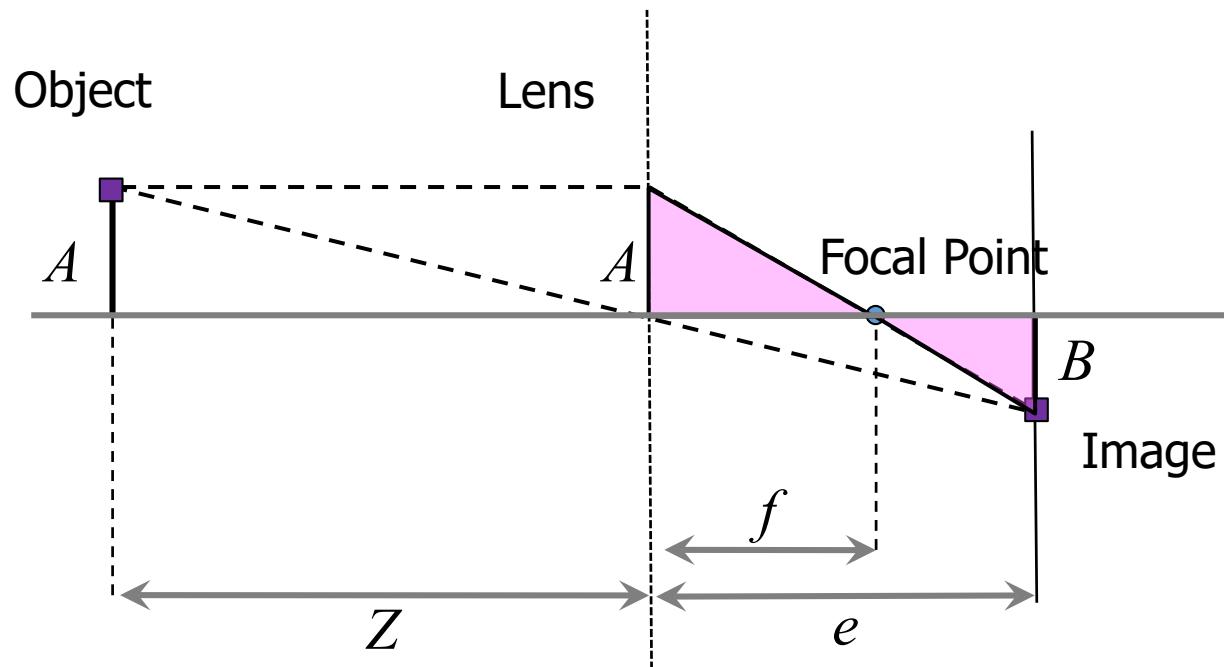


Similar Triangles:

$$\frac{B}{A} = \frac{e}{Z}$$

# Thin lens equation

- Find a relationship between  $f$ ,  $Z$ , and  $e$



Similar Triangles:

$$\frac{B}{A} = \frac{e}{Z}$$

$$\frac{B}{A} = \frac{e-f}{f} = \frac{e}{f} - 1$$

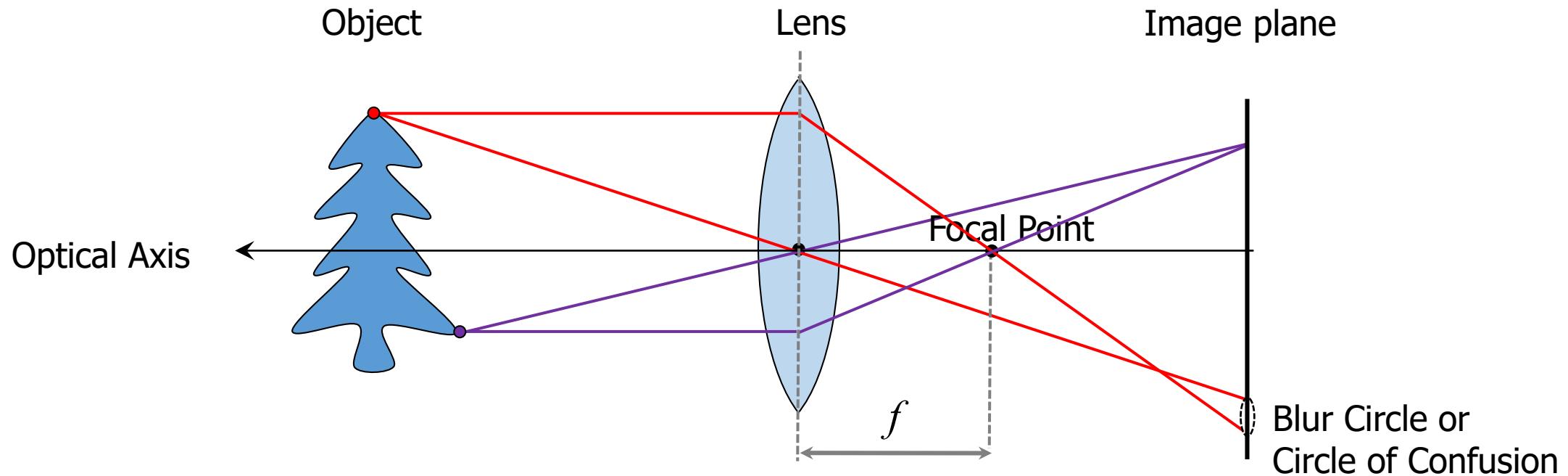
$$\left. \begin{aligned} \frac{e}{f} - 1 &= \frac{e}{Z} \\ \frac{1}{f} &= \frac{1}{Z} + \frac{1}{e} \end{aligned} \right\} \frac{1}{f} = \frac{1}{Z} + \frac{1}{e}$$

"Thin lens equation"

- Any object point satisfying this equation is in focus
- Can I use this to measure distances?

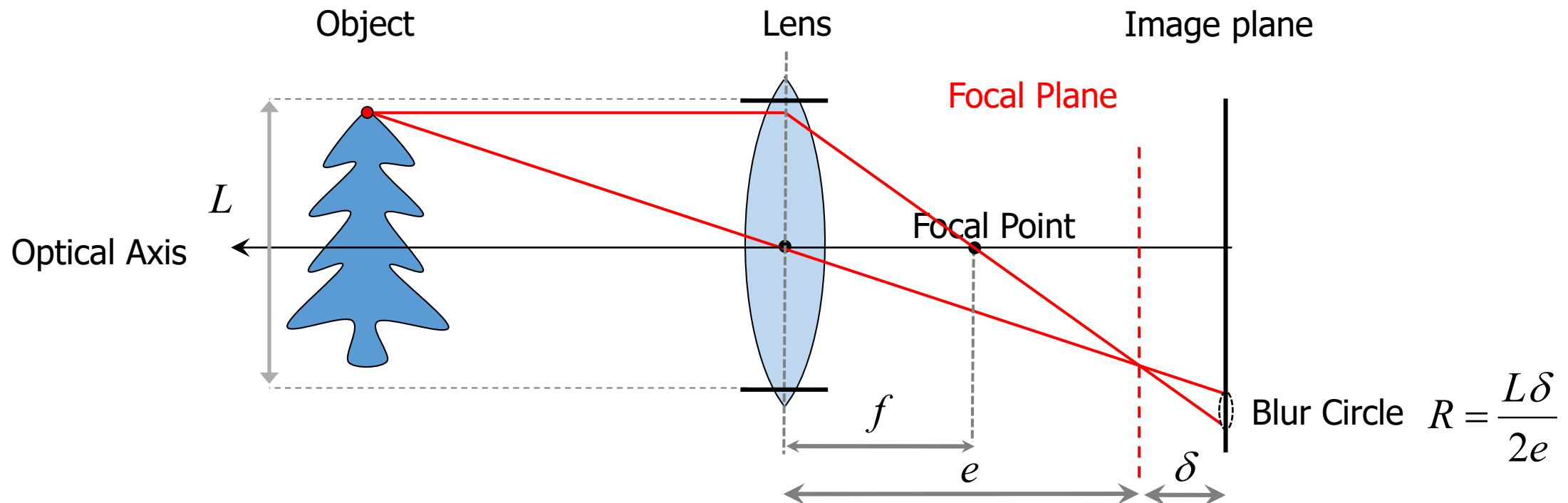
# “In focus”

- For a given point on the object, there is a specific distance between the lens and the film, at which the object appears **in focus** in the image
- Other points project to a **blur circle** in the image



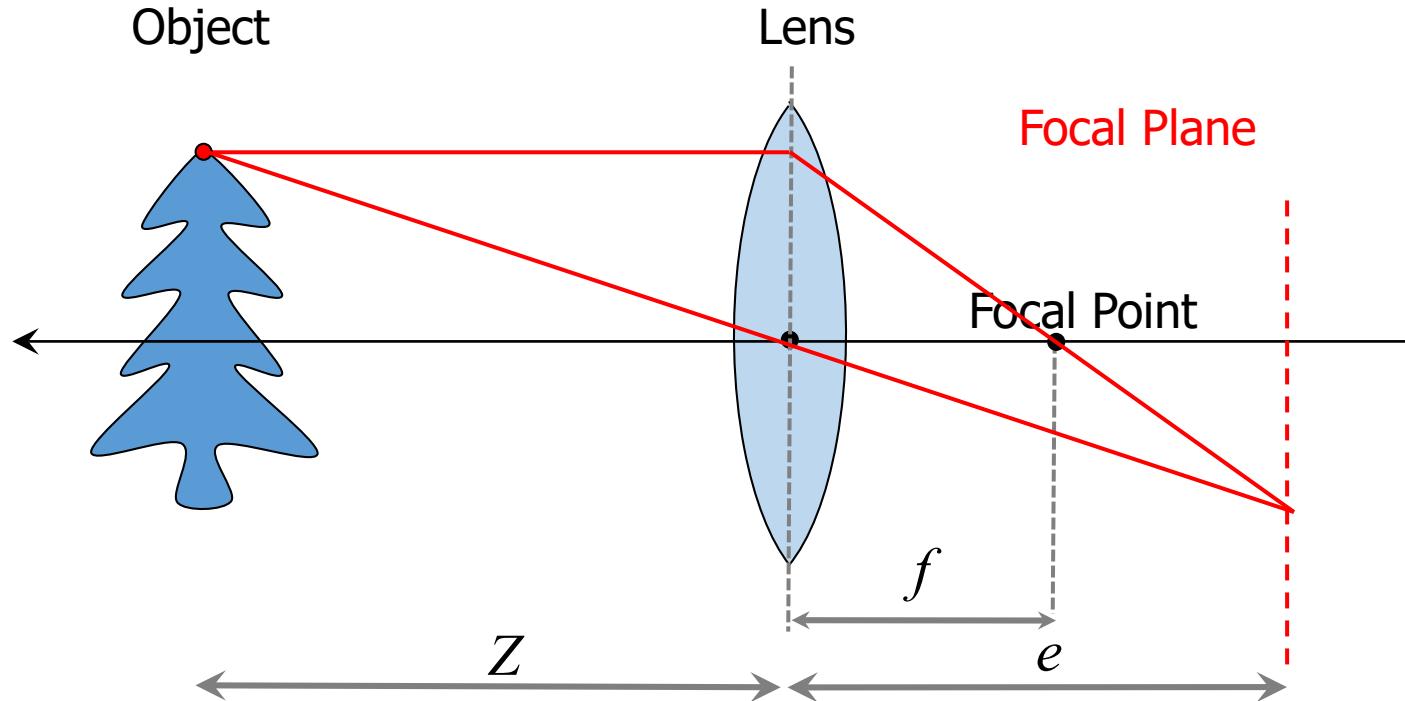
# Blur Circle

- The blur circle has radius:  $R = L\delta/(2e)$ 
  - Observe how both **small aperture ( $L$ )** and  $\delta$  yield a **small blur circle ( $R$ )**
  - To capture a sharp image, we must **adjust the camera settings** such that  $R$  remains **smaller than the 1 pixel**



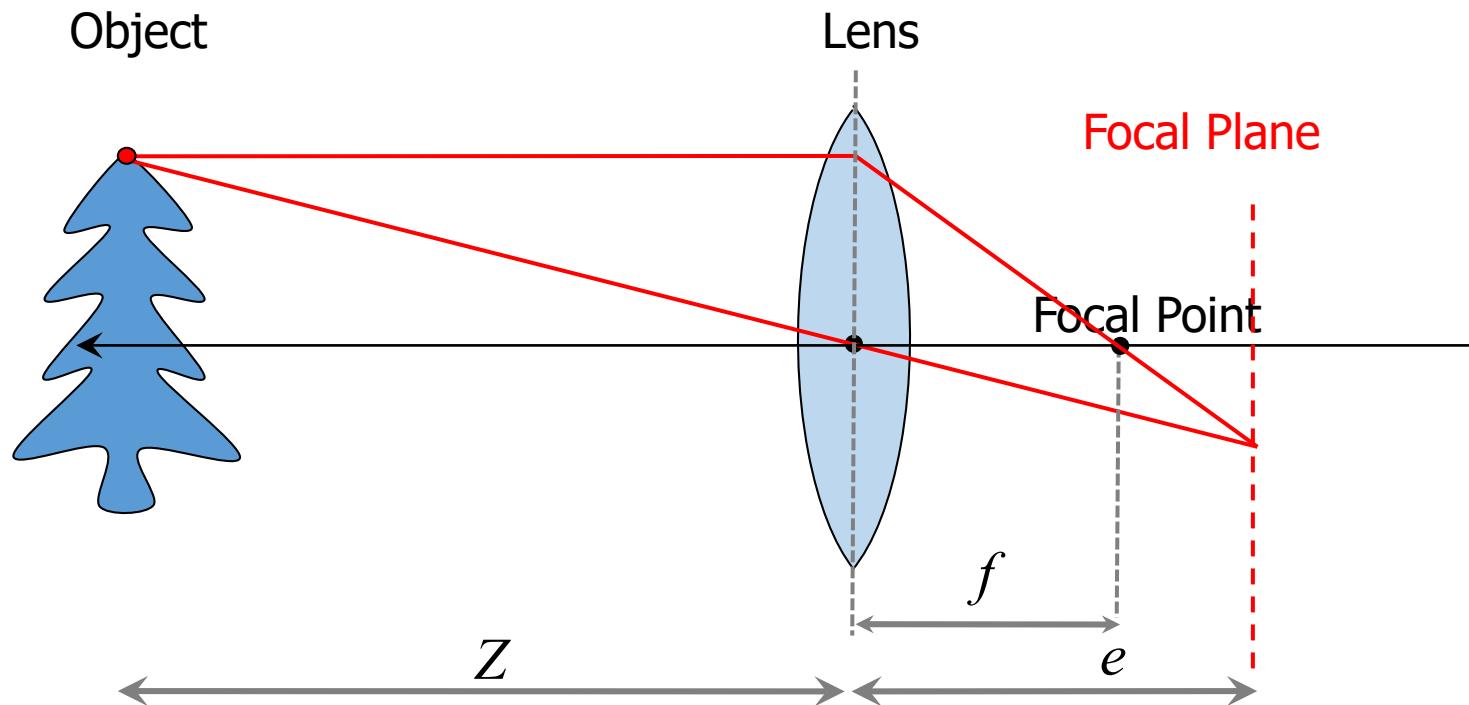
# The Pin-hole approximation

- What happens if  $z \gg f$  and  $z \gg L$  ?



# The Pin-hole approximation

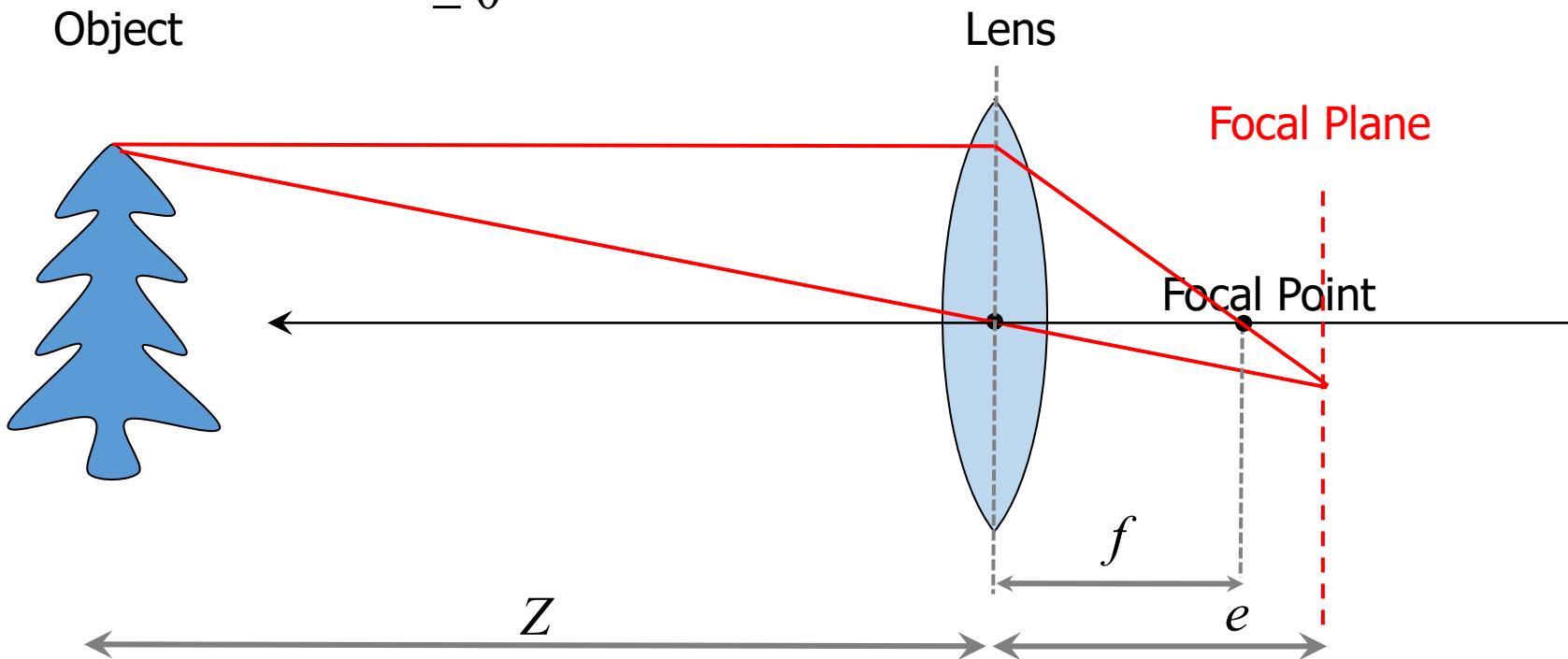
- What happens if  $z \gg f$  and  $z \gg L$  ?



# The Pin-hole approximation

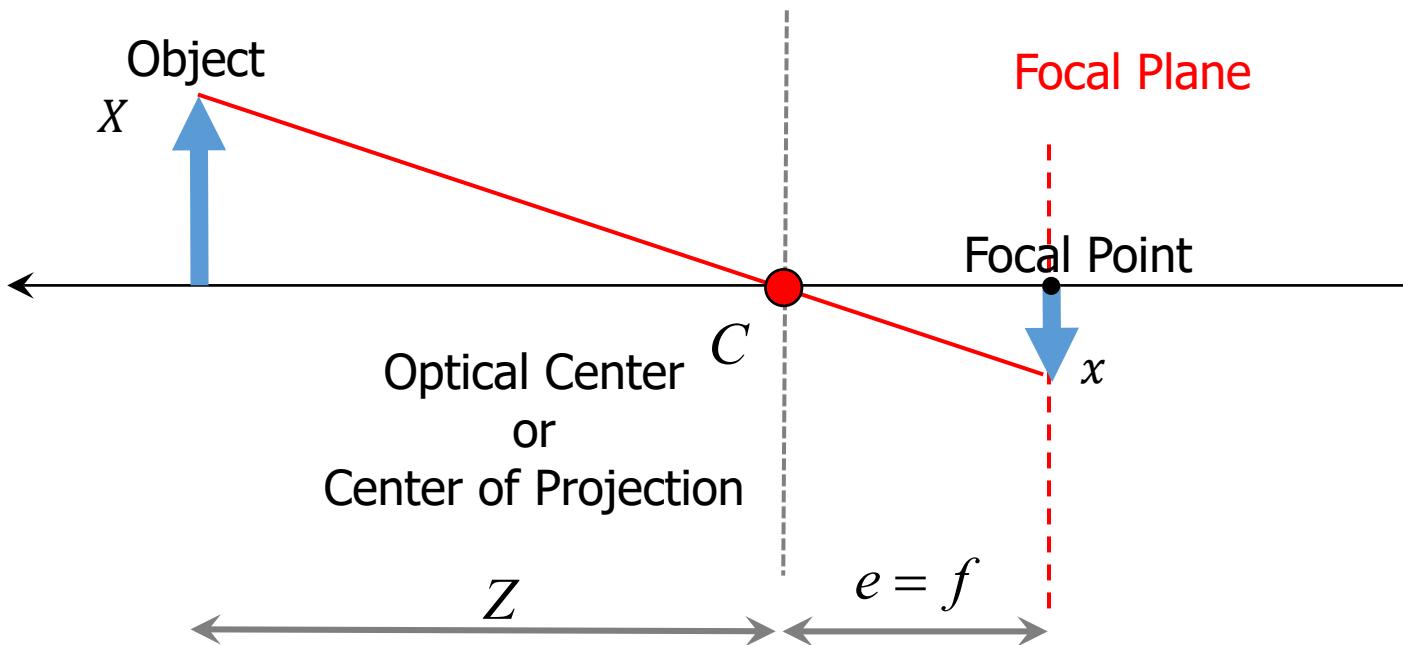
- What happens if  $z \gg f$  and  $z \gg L$  ?

- We observe that  $\frac{1}{f} = \frac{1}{z} + \frac{1}{e} \Rightarrow \frac{1}{f} \approx \frac{1}{e} \Rightarrow f \approx e$  focal plane approaches the focal point



# The Pin-hole approximation

- This is known as **Pinhole Approximation**
- The relation between the image and object becomes:  $-\frac{x}{X} = \frac{f}{Z} \Rightarrow x = -f \frac{X}{Z}$
- This is called **Perspective Projection**



# Perspective effects

Far away objects appear smaller, with **size inversely proportional to distance**



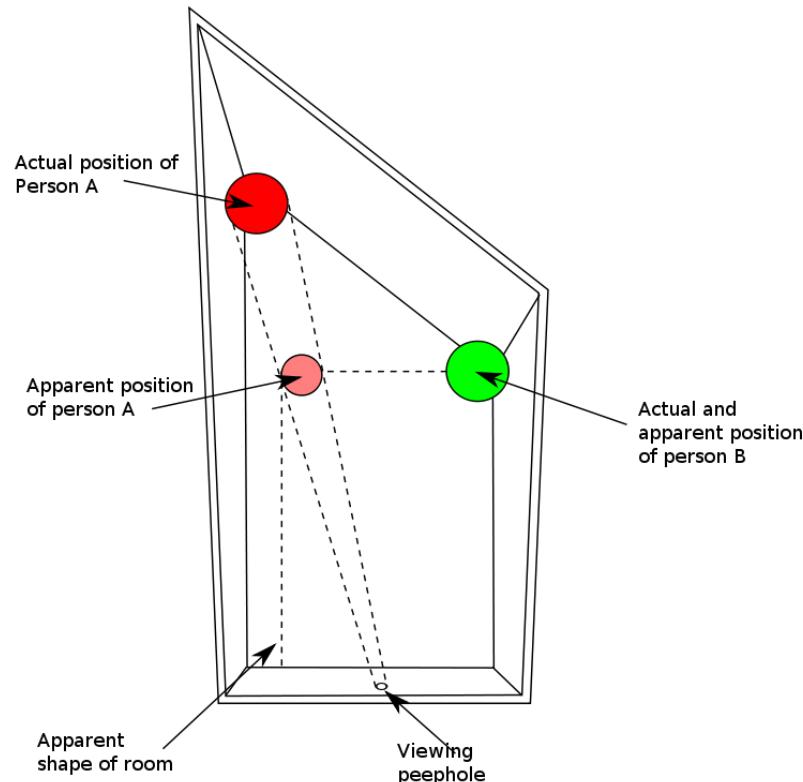
# Perspective can be used as a prior

Perspective gives us very strong depth cues and, thanks to experience (e.g., Manhattan world assumption), we **can perceive a 3D scene by viewing its 2D representation** (i.e. the image)



# Playing with Perspective: the Ames room

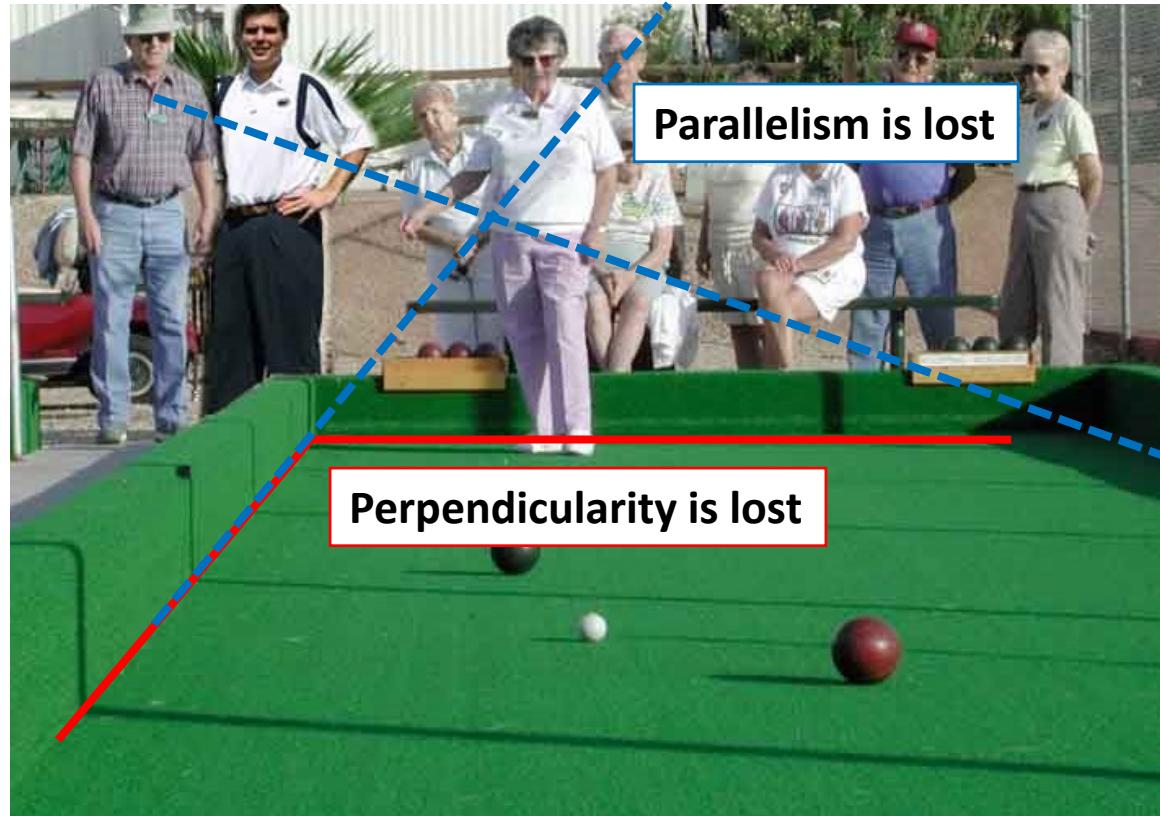
- However, our perception of a 3D scene can be fooled. An example is the Ames room (btw, check out the **Ames room** in the **Technorama** science museum in Winterthur)



Ames room documentary by neuroscientist  
Dr. V.S. Ramachandran:  
<https://youtu.be/Ttd0YjXF0no>

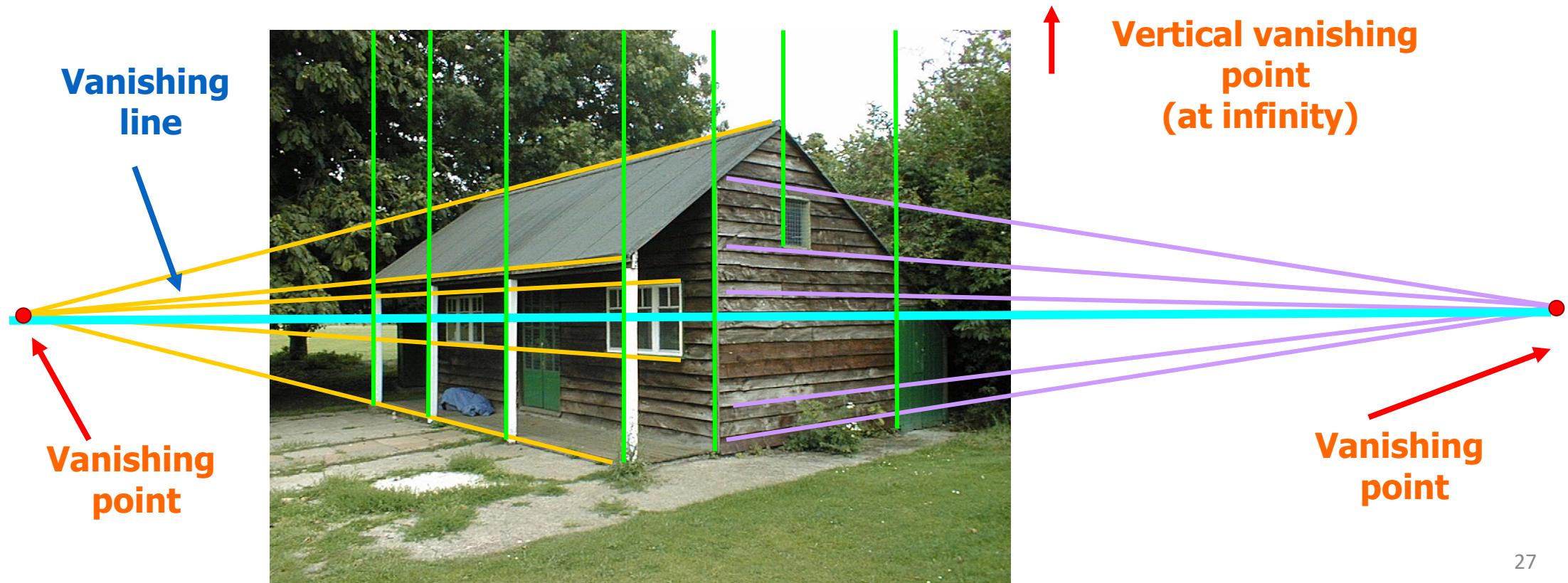
# Perspective Projection: what is preserved or lost?

- Straight lines are still **straight**
- Lengths and angles are **not preserved**



# Vanishing points and lines

- Parallel lines intersect at a “vanishing point” in the image
- Parallel planes intersect at a “vanishing line” in the image
- Notice that vanishing points can fall both inside or outside the image

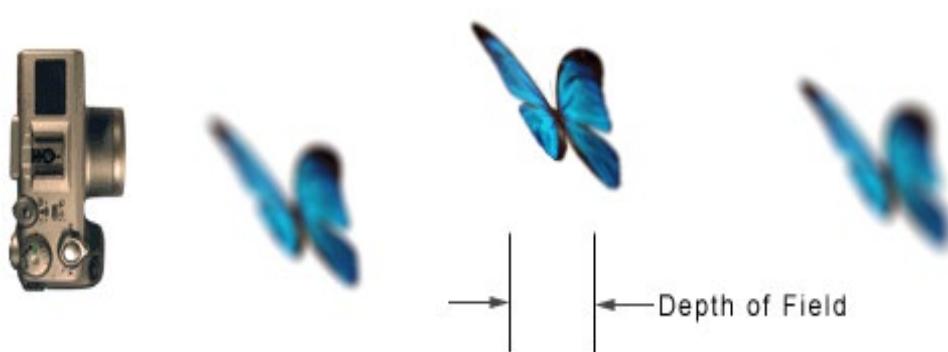


# Today's Outline

- Image Formation
- Other camera parameters
- Digital camera
- Perspective camera model
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# Focus and Depth of Field

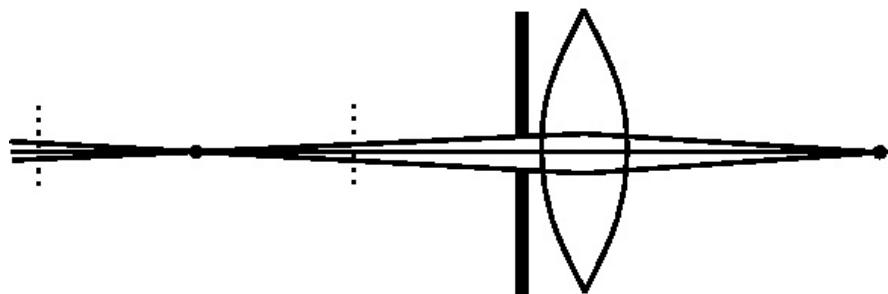
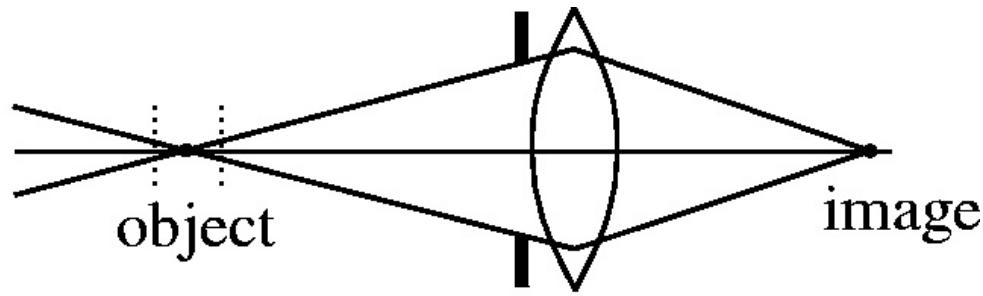
- **Depth of Field** is the distance between the **nearest and farthest objects** in a scene that appear **acceptably sharp** in an image



the depth of field will increase to infinity. For example, if your camera has a hyperfocal distance of 18 feet,

# Focus and Depth of Field

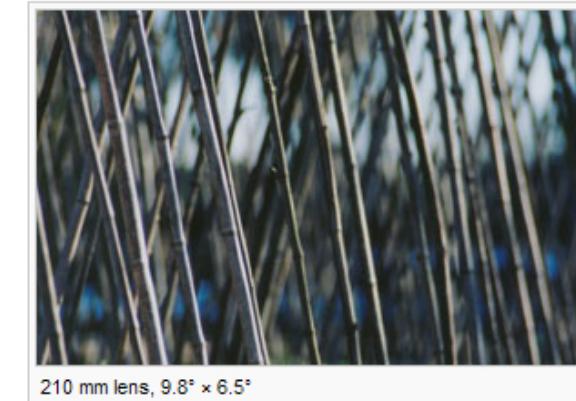
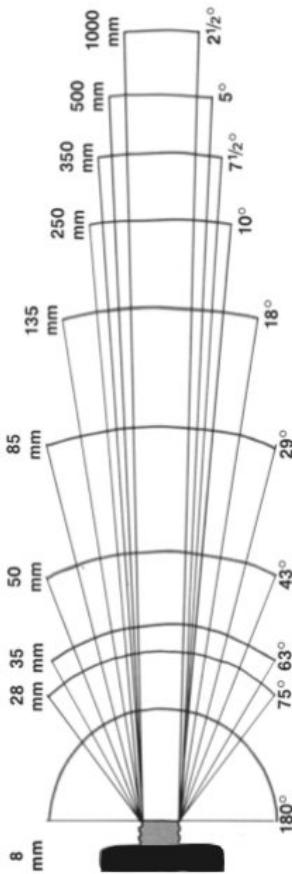
- A smaller aperture increases the depth of field but reduces the amount of light into the camera: recall the definition of blur circle (it reduces with aperture)



What is the depth of field of an ideal pinhole camera?

# Field of View (FOV)

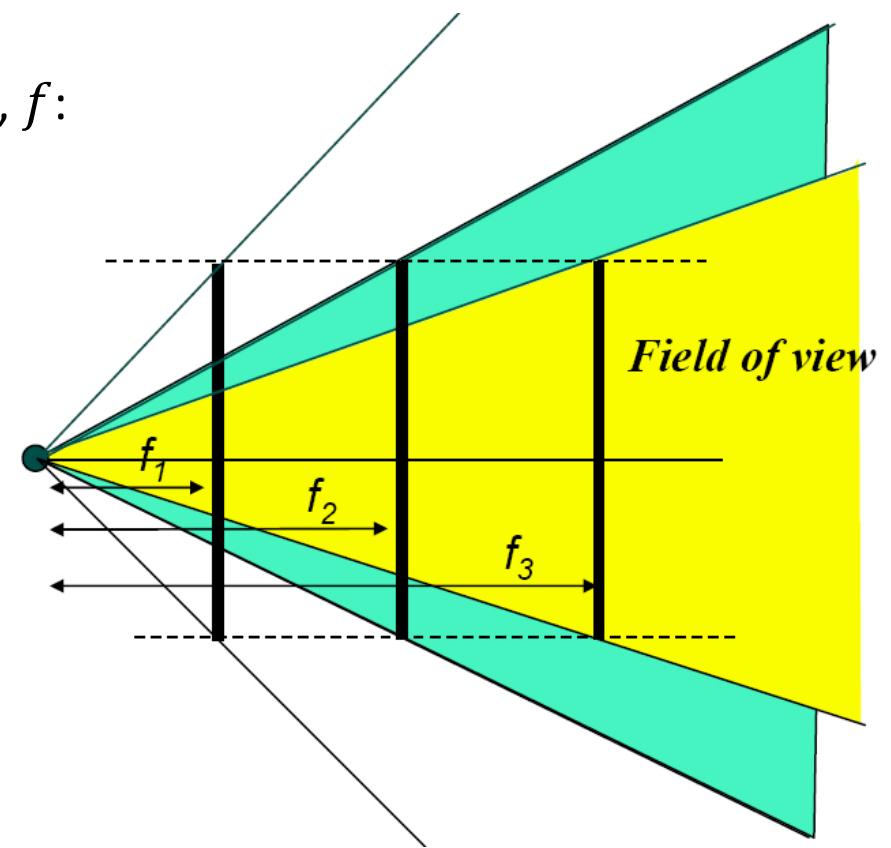
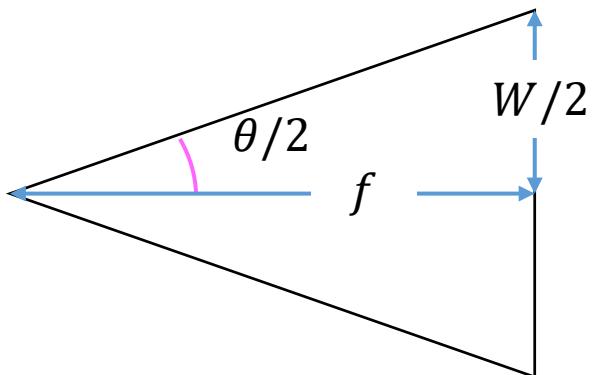
- The FOV is the angular portion of 3D scene seen by the camera



# Field of View (FOV)

- As focal length  $f$  gets smaller, image becomes more *wide angle*
- As focal length  $f$  gets larger, image becomes more *narrow angle*
- Relation between field of view,  $\theta$ , image size,  $W$ , and focal length,  $f$ :

$$\tan \frac{\theta}{2} = \frac{W}{2f} \rightarrow f = \frac{W}{2} \left[ \tan \frac{\theta}{2} \right]^{-1}$$

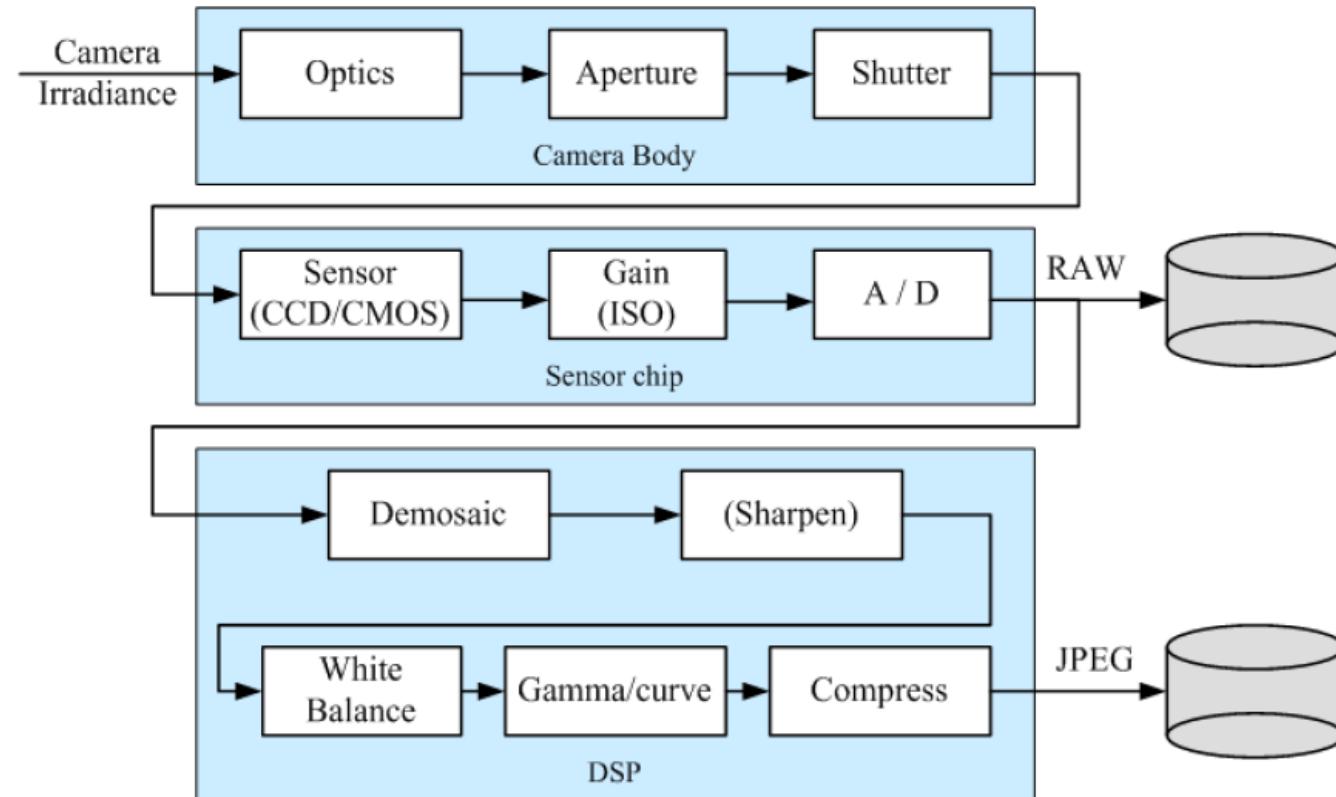


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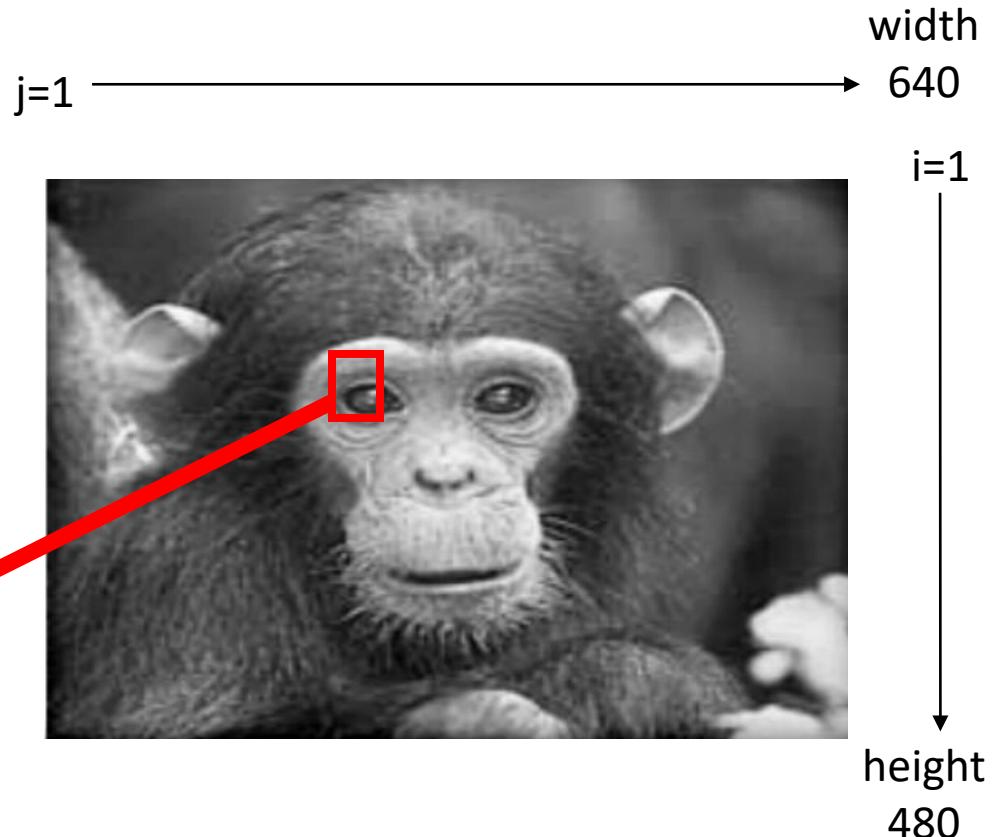
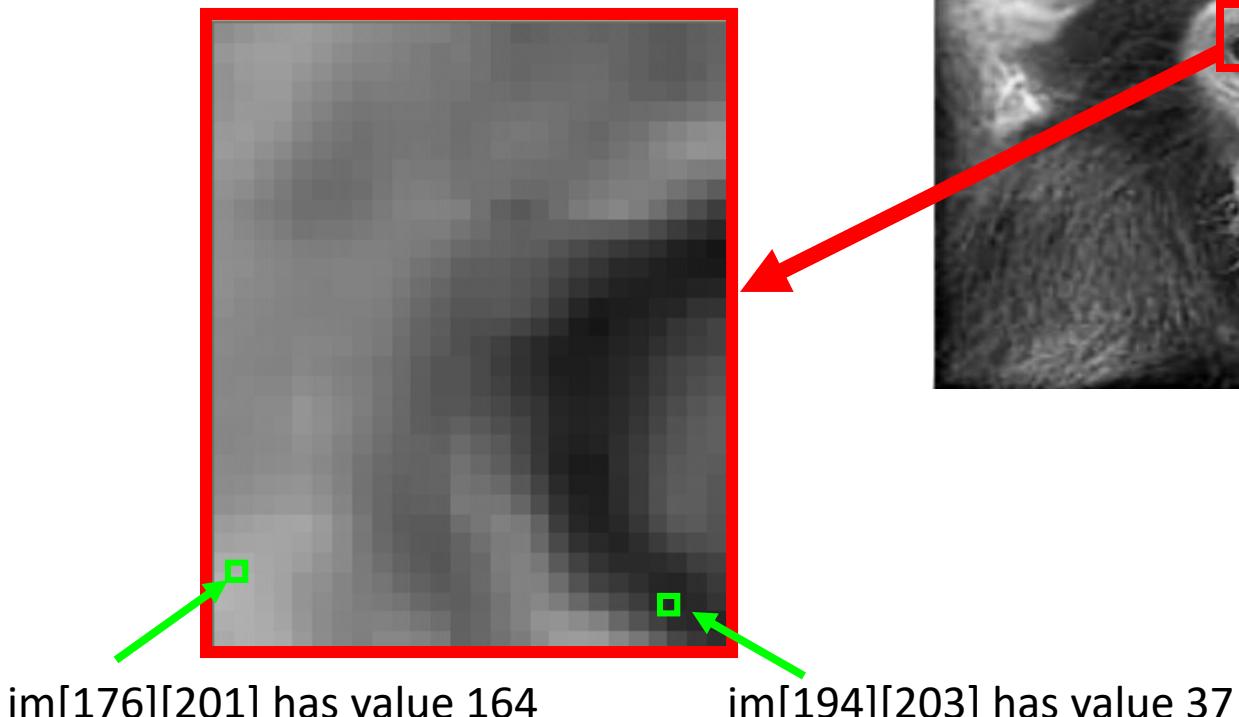
# Digital camera

- In a digital camera the **film** is an array of photodiodes (CCD or CMOS) that convert photons (light energy) into electrons



# Digital images

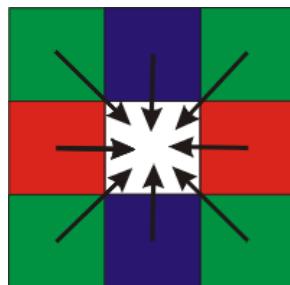
Pixel Intensity with 8 bits  
ranges between [0,255]



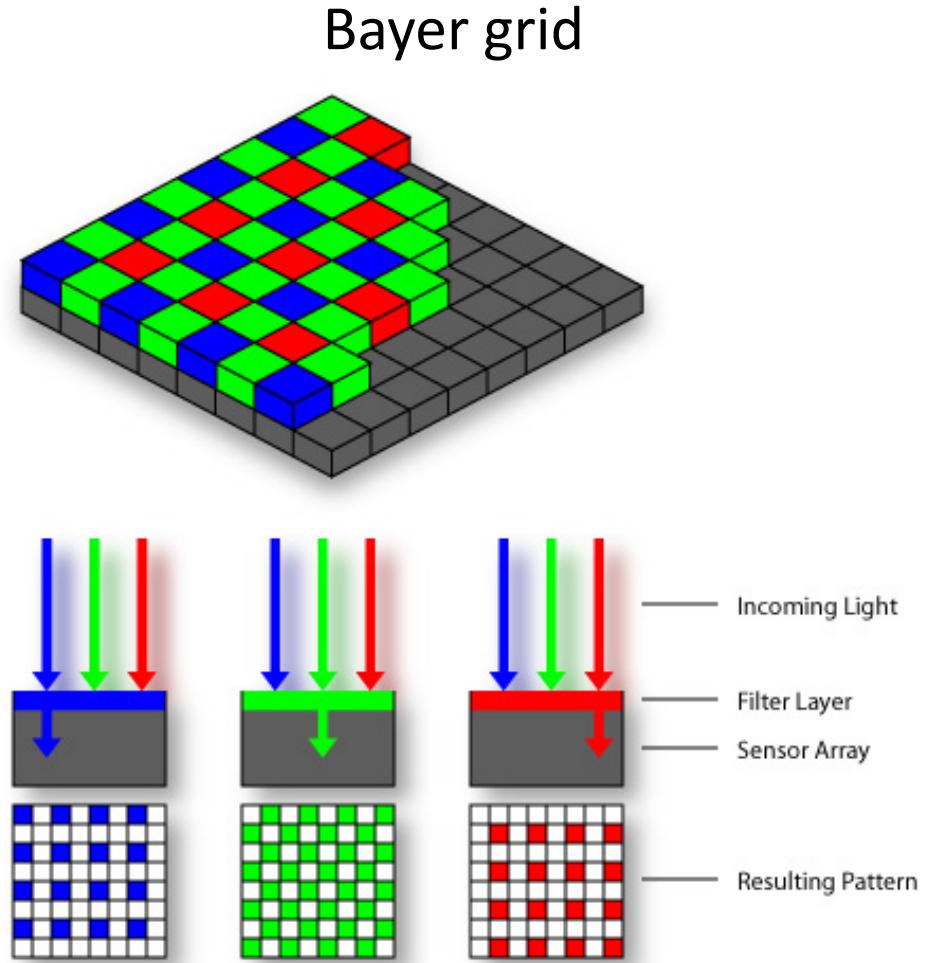
NB. Matlab coordinates: [rows, cols]; C/C++ [cols, rows]

# Color sensing in digital cameras

- The **Bayer pattern** (invented by Bayer in 1976, who worked at Kodak) places interleaved RGB filters over the pixel array
- The reason why the number of green filters is twice as that of red and blue ones is because the **luminance** signal is mostly **determined by green** values and the **human visual system** is much more **sensitive to spatial differences in luminance** than in chrominance.



For each pixel, the missing color components can be estimated from neighboring values by interpolation (**demosaicing**)



# Color sensing in digital cameras

RGB color space

... but there are also many other color spaces... (e.g., YUV)



R



G



B

# Rolling vs Global Shutter Camera

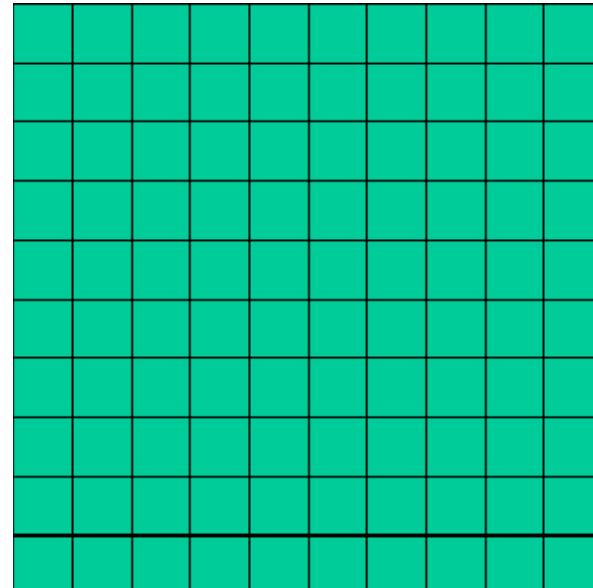
## Rolling Shutter

- **Rows of pixels are exposed and read at different times**, one after the other
- May **distort (skew) moving objects**
- **Cheap**



## Global Shutter

- **All pixels are exposed simultaneously**
- **No distortion** of moving objects
- More **costly**



# Rolling vs Global Shutter Camera

- Rolling Shutter cameras may distort (skew) moving objects
- Global Shutter cameras do not have this problem



Rolling shutter



Global shutter

# Cool application of rolling shutter cameras

- High speed wave-form capturing from a guitar



Recorded from an iPhone 4 (rolling shutter):  
<https://youtu.be/TKF6nFzpHBU>

# An example camera datasheet

## **mvBlueFOX-IGC / -MLC**

### Technical Details

#### Sensors

| mvBlueFOX-IGC<br>mvBlueFOX-MLC | Resolution<br>(H x V pixels) | Sensor size<br>(optical) | Pixel size<br>( $\mu\text{m}$ ) | Frame<br>rate | Sensor<br>technology | Readout type | ADC<br>resolution /<br>output<br>in bits | Sensor      |             |
|--------------------------------|------------------------------|--------------------------|---------------------------------|---------------|----------------------|--------------|--|-------------|-------------|
| -200w <sup>1,2</sup>           | G/C                          | 752 x 480                | 1/3"                            | 6 x 6         | 90                   | CMOS         | Global                                   | 10 → 10 / 8 | Aptina MT9V |
| -202b                          | G/C                          | 1280 x 960               | 1/3"                            | 3.75 x 3.75   | 24.6                 | CMOS         | Global                                   | 10 → 10 / 8 | Aptina MT9M |
| -202d <sup>1</sup>             | G/C                          | 1280 x 960               | 1/3"                            | 3.75 x 3.75   | 24.6                 | CMOS         | Rolling                                  | 10 → 10 / 8 | Aptina MT9M |
| -205 <sup>2</sup>              | G/C                          | 2592 x 1944              | 1/2.5"                          | 2.2 x 2.2     | 5.8                  | CMOS         | Global Reset                             | 10 → 10 / 8 | Aptina MT9P |

<sup>1</sup>High Dynamic Range (HDR) mode supported

<sup>2</sup>Software trigger supported



Nano-drone of my lab equipped  
with this camera

Sample: mvBlueFOX-IGC200wG means version with housing and 752 x 480 CMOS gray scale sensor.

mvBlueFOX-MLC200wG means single-board version without housing and with 752 x 480 CMOS gray scale sensor.

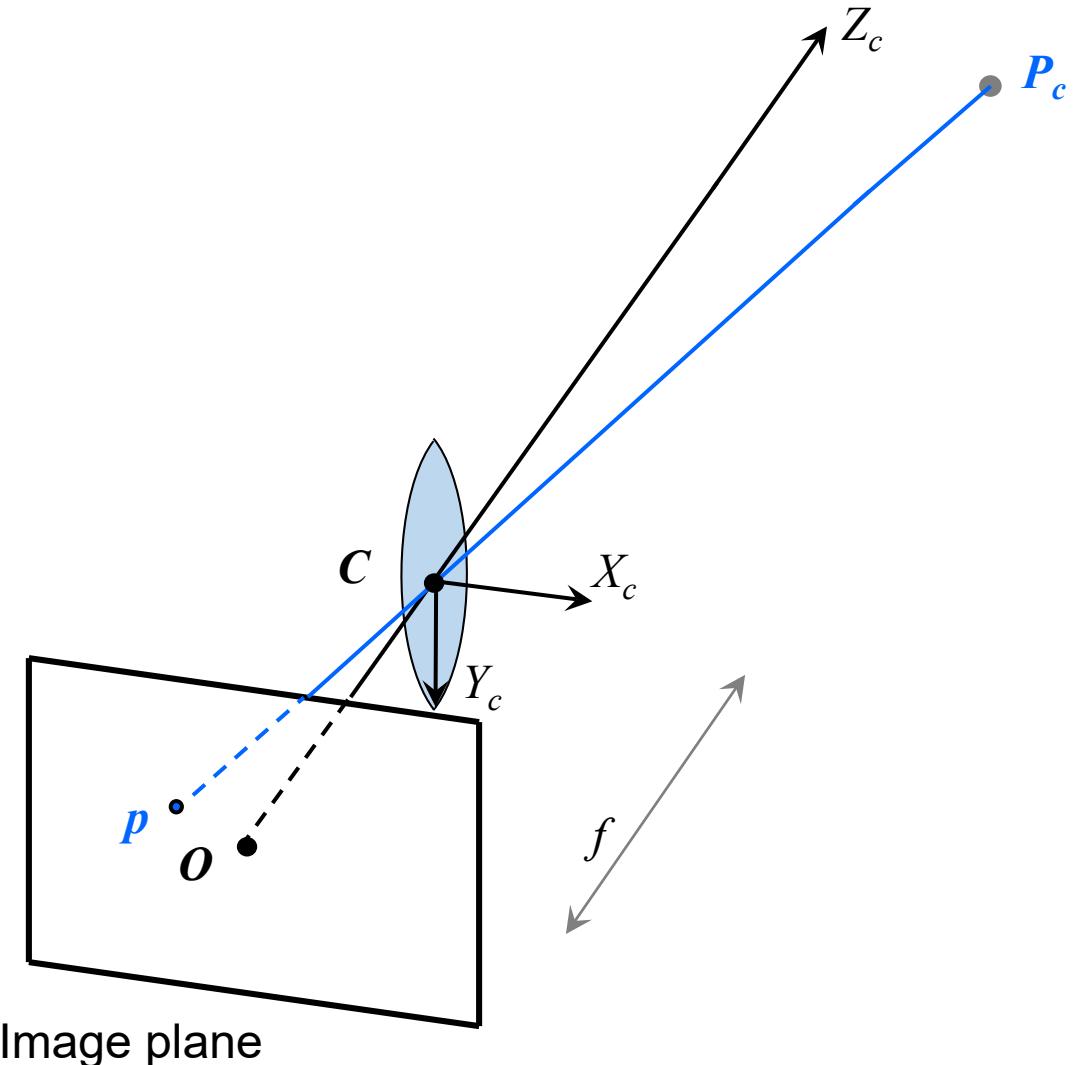
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# Perspective Camera

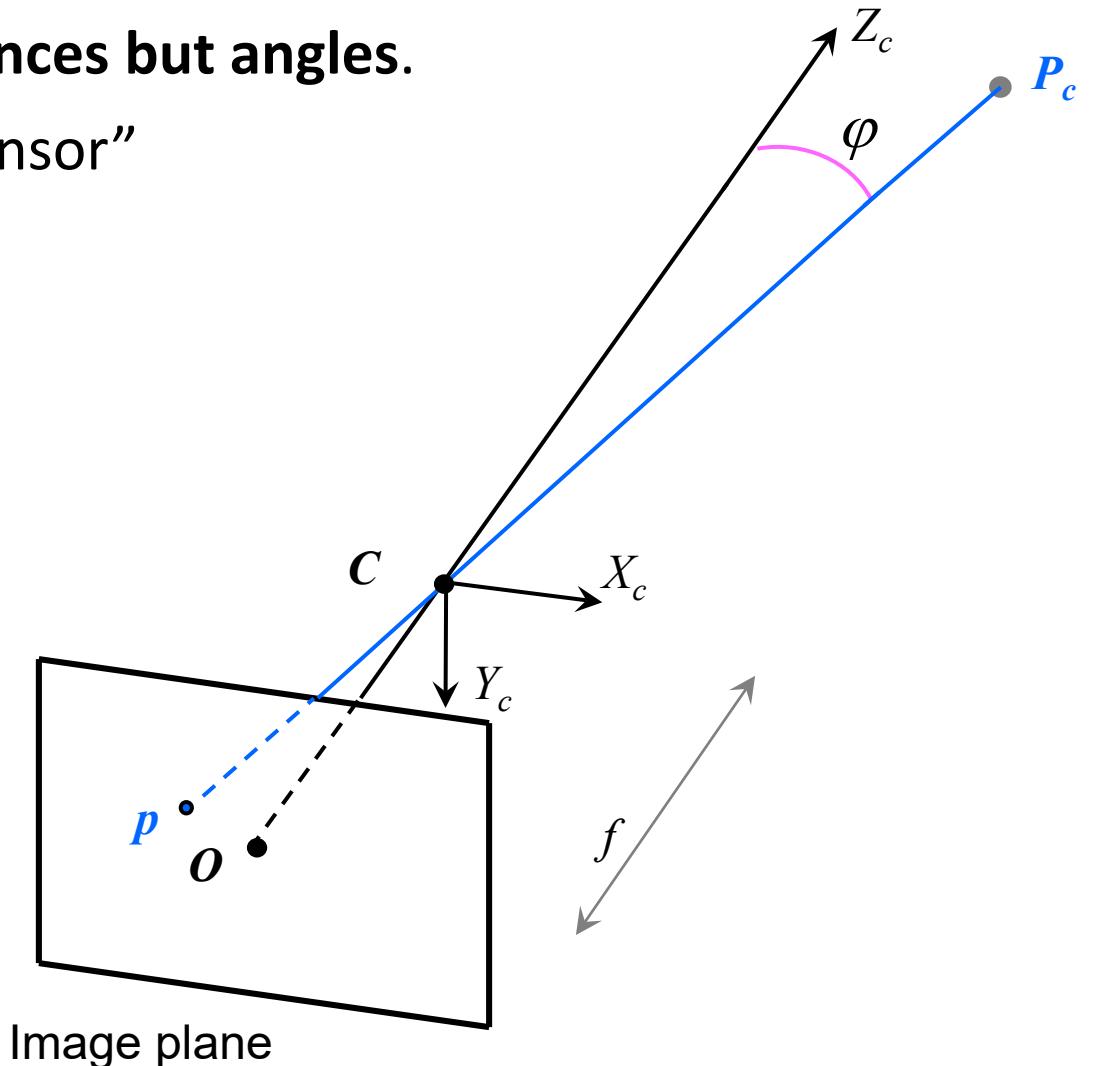
## Nomenclature:

- $C$  = optical center  
= center of the lens  
= center of projection
- $X_c, Y_c, Z_c$  = axes of the camera reference frame
- $Z_c$  = optical axis
- $O$  = principal point  
= intersection of optical axis and image plane



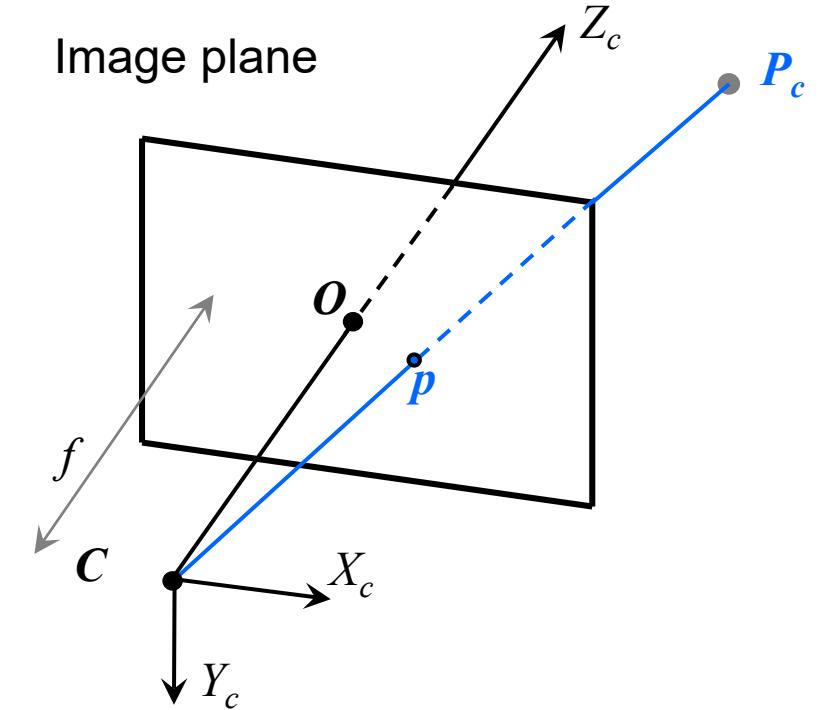
# Perspective Camera

- Note that a camera does not measure distances but angles.
- So, we call it a “bearing sensor” or “angle sensor”



# Perspective Camera

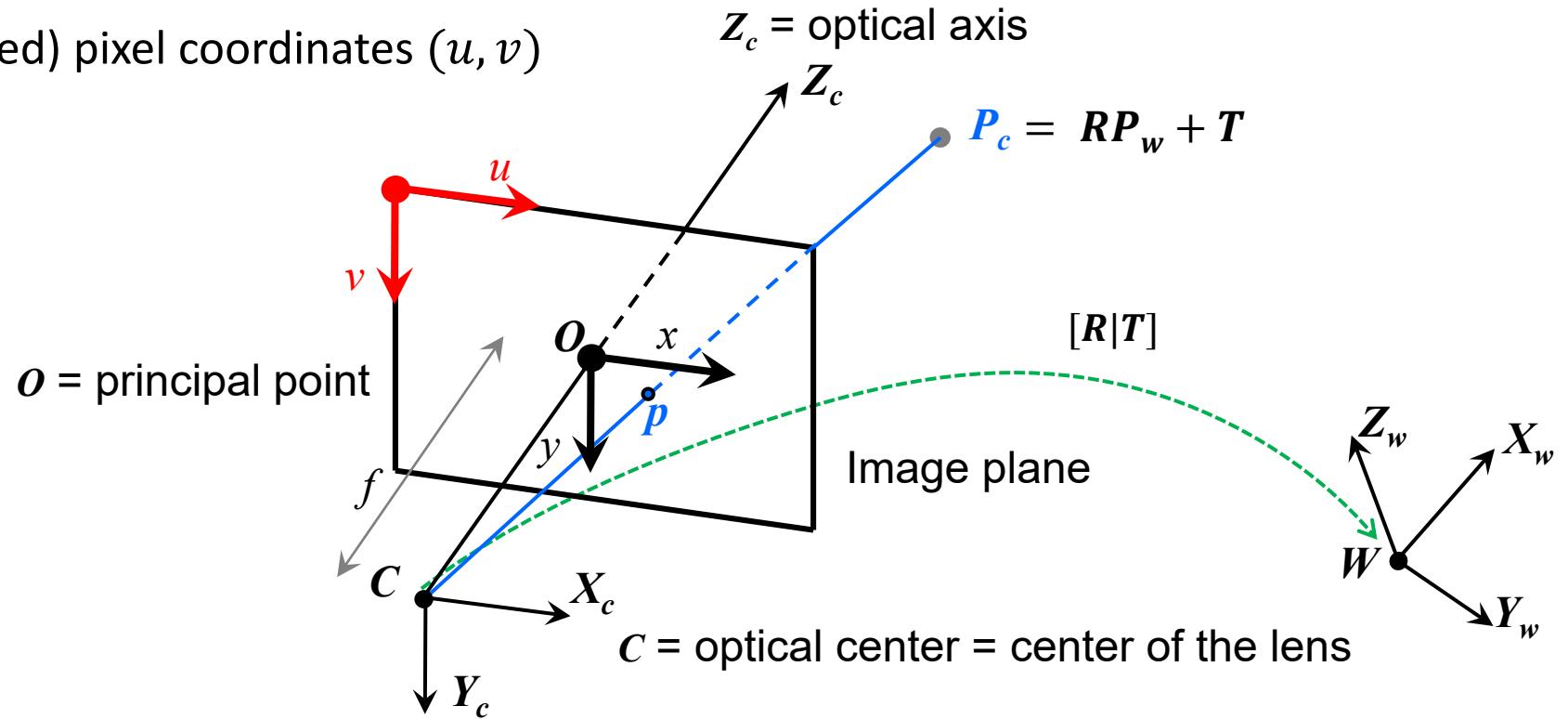
- For convenience, the **image plane** is usually **represented in front** of the lens,  $C$ , such that the image preserves the same orientation (i.e. not flipped)



# From World to Pixel coordinates

**Goal: Find pixel coordinates  $(u, v)$  of the projection of world point  $P_W$  onto the image plane:**

- Convert world point  $P_W$  to camera point  $P_C$  through rigid body transformation  $[R, T]$
- Convert  $P_C$  to image-plane coordinates  $(x, y)$
- Convert  $(x, y)$  to (discretized) pixel coordinates  $(u, v)$



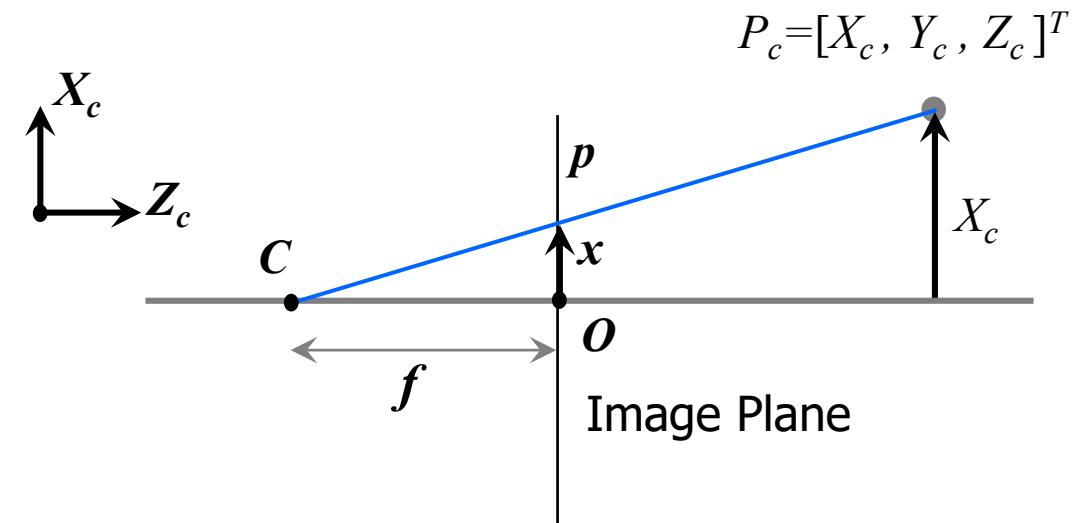
# Perspective Projection – 1/4

**From Camera frame to image plane coordinates ( $x, y$ )**

- The Camera point  $P_c = [X_c, Y_c, Z_c]^T$  projects to  $p = (x, y)$  onto the image plane

- From similar triangles:  $\frac{x}{f} = \frac{X_c}{Z_c} \Rightarrow x = \frac{fX_c}{Z_c}$

- Similarly, for  $y$ :  $\frac{y}{f} = \frac{Y_c}{Z_c} \Rightarrow y = \frac{fY_c}{Z_c}$



# Perspective Projection – 2/4

**From image plane coordinates  $(x, y)$  to pixel coordinates  $(u, v)$**

- Let  $O = (u_0, v_0)$  be the pixel coordinates of the camera optical center
- Let  $k_u, k_v$  be the pixel conversion factors (inverse of the pixel-size along  $x$  and  $y$ )

$$u = u_0 + k_u x \rightarrow u = u_0 + \frac{k_u f X_C}{Z_C}$$

$$v = v_0 + k_v y \rightarrow v = v_0 + \frac{k_v f Y_C}{Z_C}$$

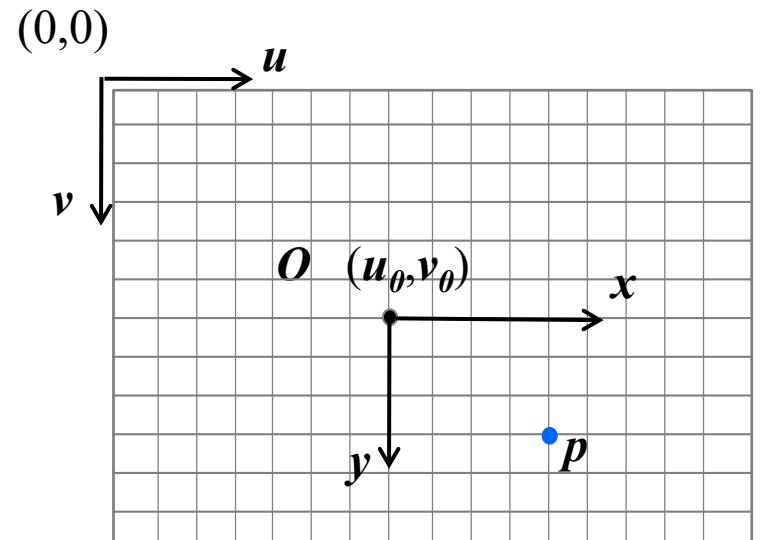


Image plane

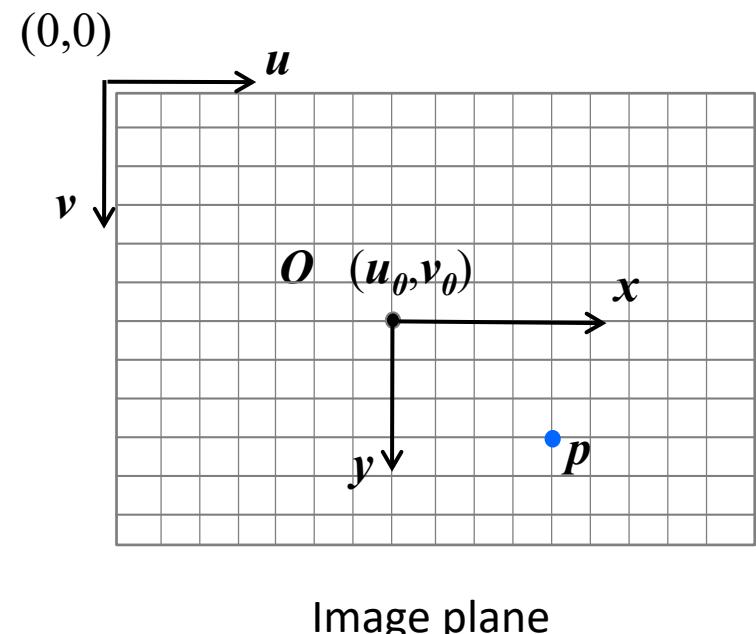
# Perspective Projection – 2/4

**From image plane coordinates  $(x, y)$  to pixel coordinates  $(u, v)$**

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$$u = u_0 + k_u x \rightarrow u = u_0 + \frac{\alpha_u x_c}{z_c}$$
$$v = v_0 + k_v y \rightarrow v = v_0 + \frac{\alpha_v y_c}{z_c}$$

Focal lengths  
(expressed in pixels)



# Perspective Projection – 3/4

- Use **Homogeneous Coordinates** for linear mapping from 3D to 2D:

$$p = \begin{pmatrix} u \\ v \end{pmatrix} \xrightarrow{\hspace{1cm}} \tilde{p} = \begin{bmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{bmatrix} = \lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix}$$

- Using matrix form and homogeneous coordinates, the perspective projection becomes:

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \alpha_u & 0 & u_0 \\ 0 & \alpha_v & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix}$$

This matrix is called “**Calibration matrix**” or “**Intrinsic Parameter Matrix**” and is often denoted as  $\mathbf{K}$

In the past it was common to assume a skew factor ( $K_{12} \neq 0$ ) to account for possible skew in the pixel manufacturing process. However, the camera manufacturing process today is so good that we can safely assume  $K_{12} = 0$  and  $\alpha_u = \alpha_v$  (i.e., square pixels).

# Perspective Projection – 3/4

- Use **Homogeneous Coordinates** for linear mapping from 3D to 2D:

$$p = \begin{pmatrix} u \\ v \end{pmatrix} \xrightarrow{\hspace{1cm}} \tilde{p} = \begin{bmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{bmatrix} = \lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix}$$

- Using matrix form and homogeneous coordinates, the perspective projection becomes:

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix}$$

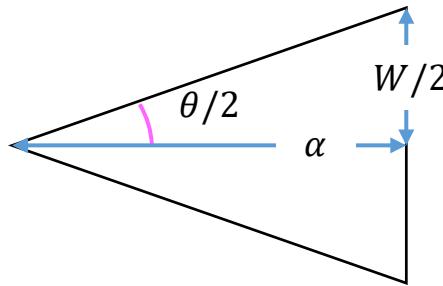
Compact form

# Exercise 1

- **Determine the Intrinsic Parameter Matrix ( $K$ )** of a digital camera with an image size  $640 \times 480$  pixels and a horizontal field of view of  $90^\circ$
- Assume square pixels and the principal point as the center of the diagonals
- What is the vertical field of view?
- What's the projection on the image plane of  $P_c = [1, 1, 2]^T$

# Exercise 1 - Solution

- Recall:



$$\tan \frac{\theta}{2} = \frac{W}{2\alpha} \rightarrow \alpha = \frac{W}{2} \left[ \tan \frac{\theta}{2} \right]^{-1} = \frac{640}{2 \tan \frac{\pi/2}{2}} = 320 \text{ [pixels]}$$

→  $\alpha_u = \alpha_v = \alpha = 320 \text{ pixels}$  (because pixels are square)

- Because the Principal Point is center of the diagonals →  $O = (u_0, v_0) = \left( \frac{640}{2}, \frac{480}{2} \right) = (320, 240)$

$$\rightarrow K = \begin{bmatrix} 320 & 0 & 320 \\ 0 & 320 & 240 \\ 0 & 0 & 1 \end{bmatrix}$$

- Vertical FOV:  $\theta_V = 2 \tan^{-1} \frac{H}{2\alpha} = 2 \tan^{-1} \frac{480}{2 \cdot 320} = 73.74^\circ$

# Exercise 1 - Solution

- What's the projection on the image plane of  $P_c = [1, 1, 2]^T$
- **Solution 1** (using Perspective Projection Equation in plane form)

$$u = u_0 + \frac{\alpha X_C}{Z_C} = 320 + \frac{320 \cdot 1}{2} = 480 \text{ [pixels]}$$

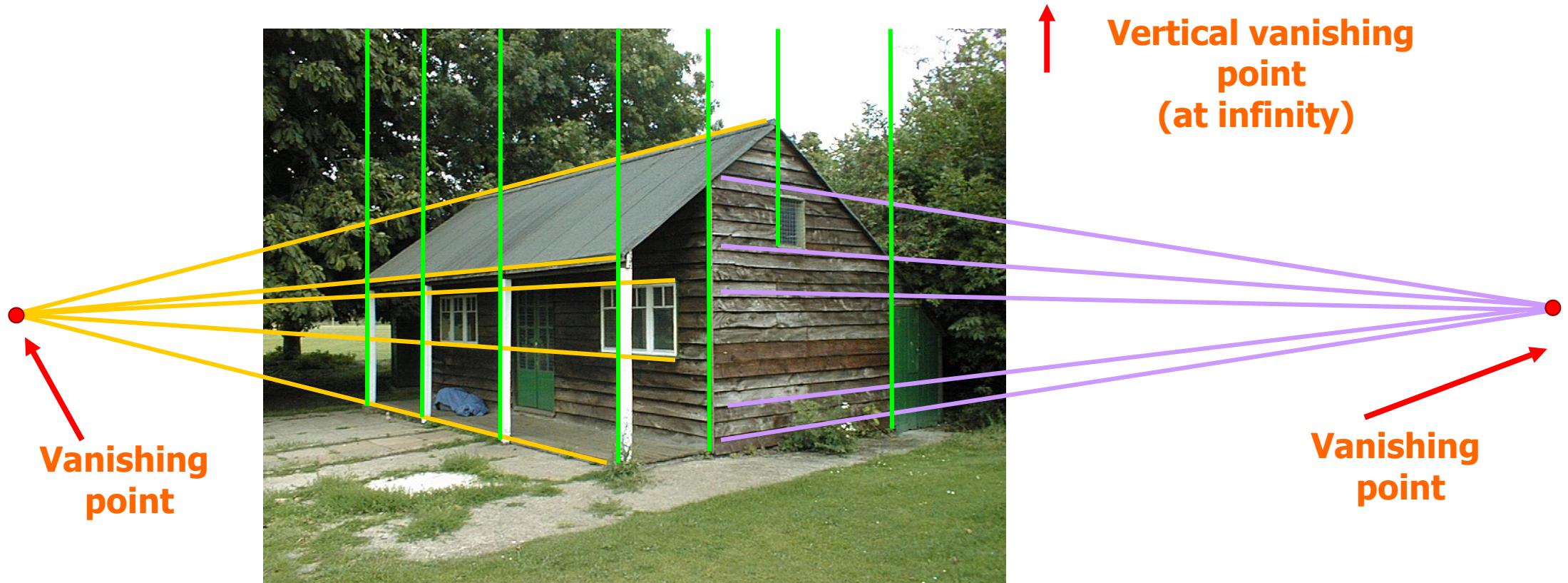
$$v = v_0 + \frac{\alpha Y_C}{Z_C} = 240 + \frac{320 \cdot 1}{2} = 400 \text{ [pixels]}$$

- **Solution 2** (using Perspective Projection Equation in matrix form)

$$\begin{bmatrix} \lambda u \\ \lambda v \\ \lambda \end{bmatrix} = \begin{bmatrix} 320 & 0 & 320 \\ 0 & 320 & 240 \\ 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 960 \\ 800 \\ 2 \end{bmatrix} \rightarrow \begin{aligned} u &= \frac{960}{2} = 480 \text{ [pixels]} \\ v &= \frac{800}{2} = 400 \text{ [pixels]} \end{aligned}$$

# Exercise 2

- Prove that world's parallel lines intersect at a vanishing point in the camera image



# Exercise 2 - Solution

- World's parallel lines with direction  $(l, m, n)$ , passing through  $(X_0, Y_0, Z_0)$ , are described by these equations:

$$X = X_0 + sl$$

$$Y = Y_0 + sm$$

$$Z = Z_0 + sn$$

- Let's consider the perspective projection equation in plane form:

$$u = u_0 + \alpha \frac{X}{Z} \quad , \quad v = v_0 + \alpha \frac{Y}{Z}$$

- Substitute line equations into the perspective projection equation and compute limit for  $s \rightarrow \infty$

$$\lim_{s \rightarrow \infty} u_0 + \alpha \frac{X_0 + sl}{Z_0 + sn} = u_0 + \alpha \frac{l}{n} \quad , \quad \lim_{s \rightarrow \infty} v_0 + \alpha \frac{Y_0 + sm}{Z_0 + sn} = v_0 + \alpha \frac{m}{n}$$

- These are the image coordinates of the vanishing point. **Notice that they only depend on the line direction** (indeed, they do not depend on  $(X_0, Y_0, Z_0)$ ).

# Exercise 2 - Solution

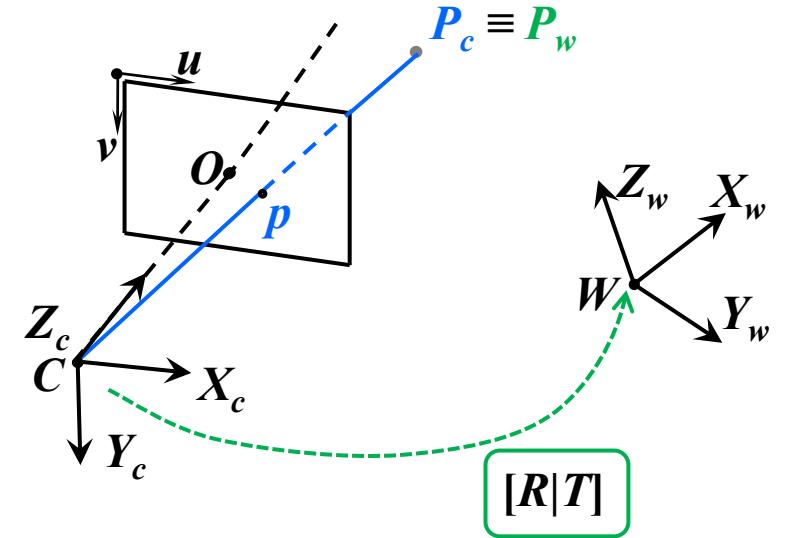
- You can verify that the solution satisfies this matrix equation. **Can you show it geometrically?**

$$\begin{bmatrix} \lambda u \\ \lambda v \\ \lambda \end{bmatrix} = \begin{bmatrix} \alpha & 0 & u_0 \\ 0 & \alpha & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} l \\ m \\ n \end{bmatrix}$$

# Perspective Projection – 4/4

From the World frame to the Camera frame

$$\begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \end{bmatrix} + \begin{bmatrix} t_1 \\ t_2 \\ t_3 \end{bmatrix}$$



Extrinsic Parameters

# Perspective Projection – 4/4

From the World frame to the Camera frame

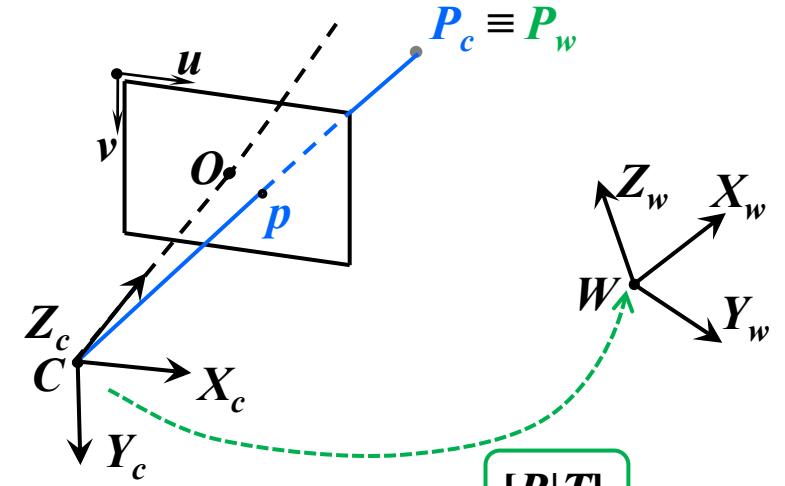
$$\begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_1 \\ r_{21} & r_{22} & r_{23} & t_2 \\ r_{31} & r_{32} & r_{33} & t_3 \end{bmatrix} \cdot \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix} = \begin{bmatrix} R & T \end{bmatrix} \cdot \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

By substituting the above one into the Perspective Projection Equation:

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix}$$

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K [R \mid T] \cdot \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

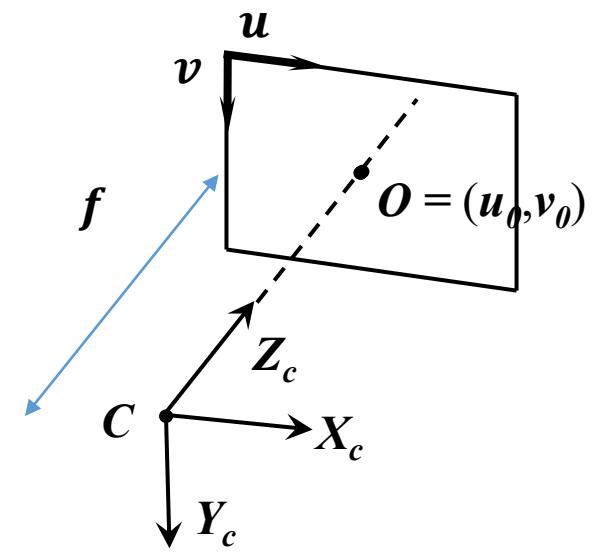
Projection Matrix (M)



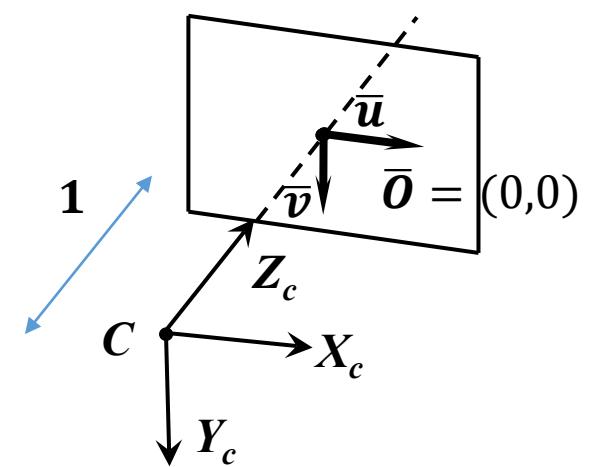
Extrinsic Parameters

Perspective Projection Equation

- It is often convenient to use normalized image coordinates onto a **virtual image plane with focal length equal to 1 unit** (adimensional) and **origin of the pixel coordinates at the principal point**.



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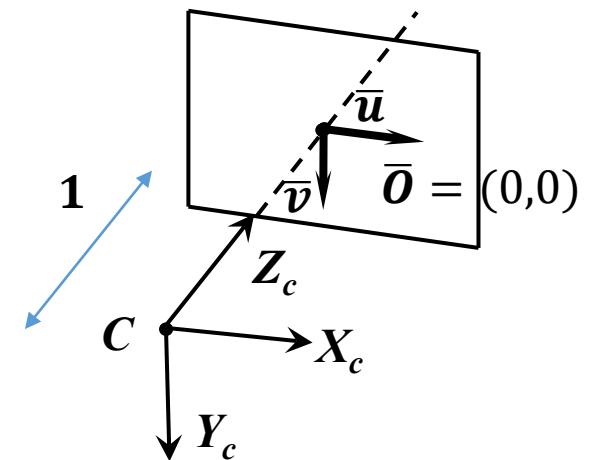
# Normalized image coordinates

- It is often convenient to use normalized image coordinates onto a **virtual image plane with focal length equal to 1 unit** (adimensional) and **origin of the pixel coordinates at the principal point**.
- To do this, we just pre-multiply both terms of the perspective projection equation in camera frame coordinates by  $K^{-1}$ :

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix} \Rightarrow \lambda K^{-1} \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K^{-1} \cdot K \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix} \Rightarrow \lambda K^{-1} \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix}$$

- We define the **unit-plane normalized image coordinates**  $(\bar{u}, \bar{v})$ :

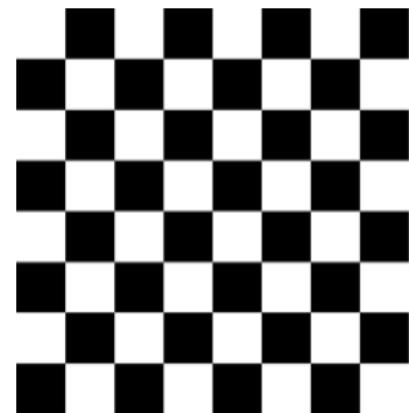
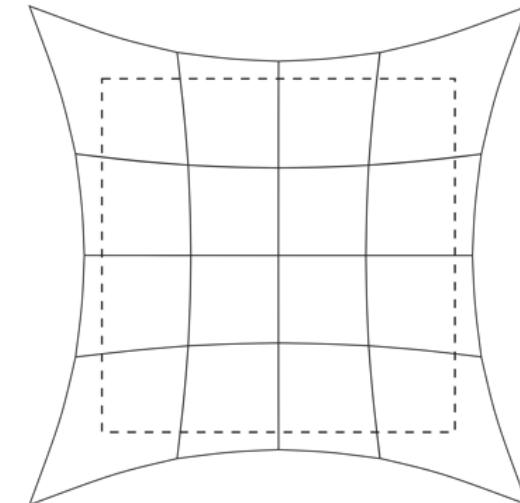
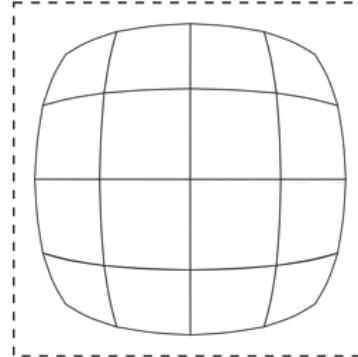
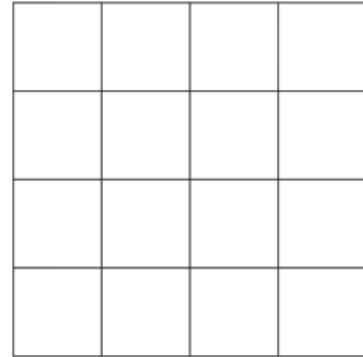
$$\begin{bmatrix} \bar{u} \\ \bar{v} \\ 1 \end{bmatrix} = K^{-1} \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{\alpha} & 0 & -\frac{u_0}{\alpha} \\ 0 & \frac{1}{\alpha} & -\frac{v_0}{\alpha} \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{u-u_0}{\alpha} \\ \frac{v-v_0}{\alpha} \\ 1 \end{bmatrix} \Rightarrow \lambda \begin{bmatrix} \bar{u} \\ \bar{v} \\ 1 \end{bmatrix} = \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix}$$



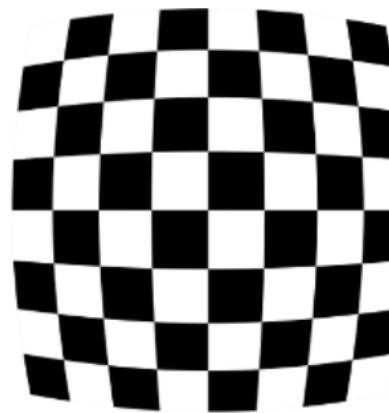
# Today's Outline

- Image Formation
- Other camera parameters
- Digital camera
- Perspective camera model
- Lens distortion

# Radial Distortion

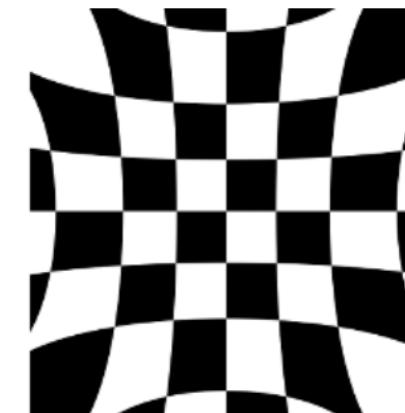


No distortion



Positive radial distortion  
(Barrel distortion)

$$k_1 > 0$$



Negative radial distortion  
(Pincushion distortion)

$$k_1 < 0$$

# Radial Distortion in the OpenCV and Matlab Camera Models

- The standard model of radial distortion is a transformation **from the ideal (non-distorted) coordinates  $(u, v)$  to the real (distorted) coordinates  $(u_d, v_d)$**
- For a given non distorted image point  $(u, v)$ , the amount of distortion is a nonlinear function of its distance  $r$  from the principal point. For most lenses, **this simple quadratic model of radial distortion is sufficient:**

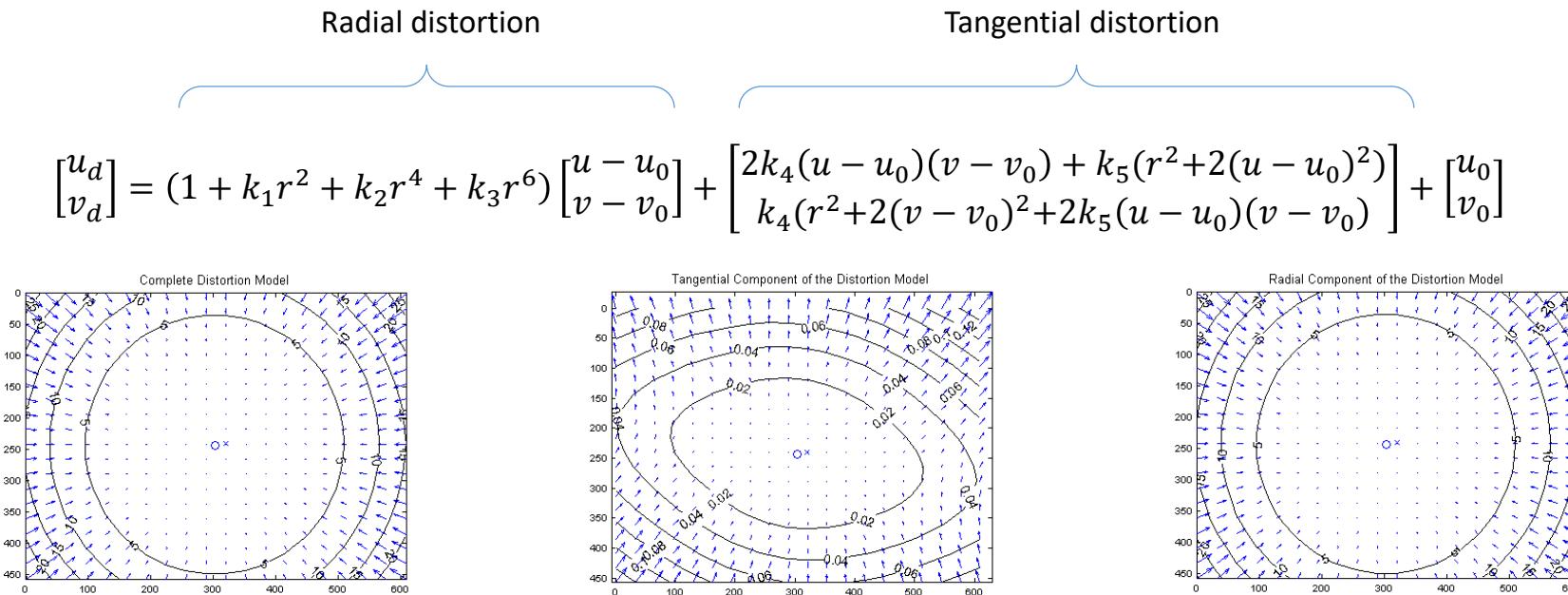
$$\begin{bmatrix} u_d \\ v_d \end{bmatrix} = (1 + k_1 r^2) \begin{bmatrix} u - u_0 \\ v - v_0 \end{bmatrix} + \begin{bmatrix} u_0 \\ v_0 \end{bmatrix}$$

where

$$r^2 = (u - u_0)^2 + (v - v_0)^2$$

# Radial & Tangential Distortion in the OpenCV and Matlab Camera Models

- **Radial Distortion:** Depending on the amount of distortion (and thus on the camera field of view), higher order terms can be introduced for the radial distortion
- **Tangential Distortion:** if the lens is misaligned (not perfectly parallel to the image sensor), a *non-radial* (tangential) distortion is introduced



The **left figure** shows the impact of the **complete distortion model (radial + tangential)** on each pixel of the image. Each arrow represents the effective displacement of a pixel induced by the lens distortion. Observe that points at the corners of the image are displaced by as much as 25 pixels. The **center figure** shows the impact of the **tangential component of distortion**. On this plot, the **maximum induced displacement is 0.14 pixel** (at the upper left corner of the image). Finally, the **right figure** shows the impact of the **radial component of distortion**. This plot is very similar to the full distortion plot, showing that the tangential component could very well be discarded in the complete distortion model. On the three figures, the **cross** indicates the **center of the image** (i.e., the center of the two diagonals), and the **circle** the location of the **principal point**.

# Summary: Perspective projection equations

- A world's 3D point  $P_w = (X_w, Y_w, Z_w)$  projects into the image point  $p = (u, v)$

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K[R | T] \cdot \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix} \quad \text{where} \quad K = \begin{bmatrix} \alpha & 0 & u_0 \\ 0 & \alpha & v_0 \\ 0 & 0 & 1 \end{bmatrix}$$

- To account for radial distortion, distorted pixel coordinates  $(u_d, v_d)$  can be obtained as:

$$\begin{bmatrix} u_d \\ v_d \end{bmatrix} = (1 + k_1 r^2) \begin{bmatrix} u - u_0 \\ v - v_0 \end{bmatrix} + \begin{bmatrix} u_0 \\ v_0 \end{bmatrix} \quad \text{where} \quad r^2 = (u - u_0)^2 + (v - v_0)^2$$

- For convenience, the projection of  $P_w$  into pixel coordinates, including lens distortion, is usually denoted:

$$\begin{bmatrix} u_d \\ v_d \end{bmatrix} = \pi(P_w, K, R, T)$$

# Summary (things to remember)

- This-lens equation
- From the thin lens to the pinhole camera
- Perspective Projection Equation
- Vanishing points and lines
- Intrinsic and extrinsic parameters ( $\mathbf{K}, \mathbf{R}, \mathbf{T}$ )
- Homogeneous coordinates
- Normalized image coordinates
- Radial distortion

# Readings

- Chapter 4 of Autonomous Mobile Robots book: [link](#)

# Understanding Check

Are you able to:

- Explain what a blur circle is?
- Derive the thin lens equation and perform the pinhole approximation?
- Explain how to build an Ames room?
- Derive a relation between the field of view and the focal length?
- Proof the perspective projection equation, including lens distortion and world-to-camera projection?
- Explain normalized image coordinates and their geometric explanation?
- Define vanishing points and lines?
- Prove that parallel lines intersect at vanishing points?