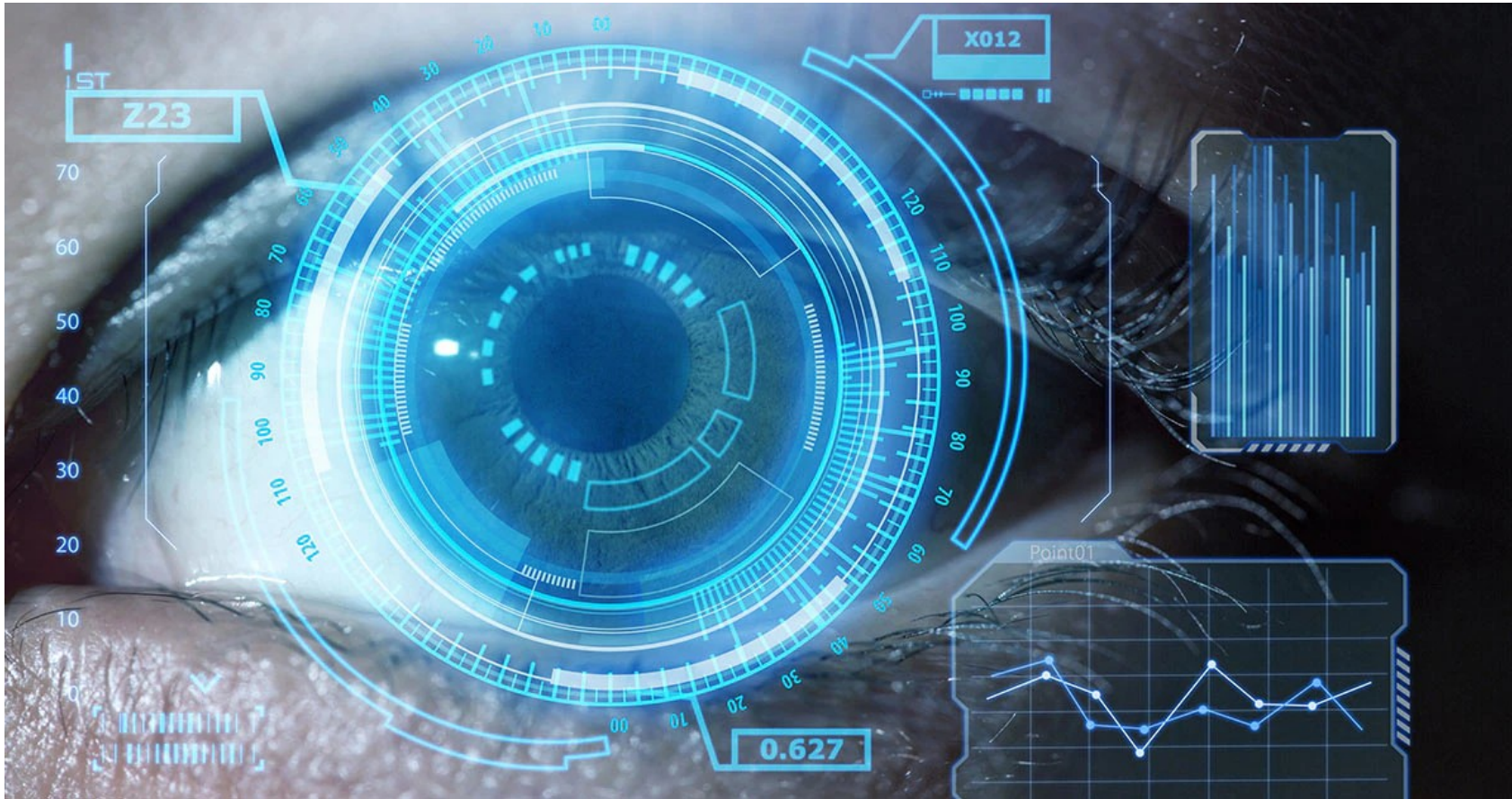


EECE 5550: Mobile Robotics



Lecture 8: Fundamentals of Computer Vision

Plan of the day

- Camera models
 - Image formation
 - Pinhole projection model
 - Homogeneous coordinates
- Feature extraction
 - Corner detectors
 - Feature descriptors

References

Today's lecture consists of selections from two lectures given by Margarita Chli in ETH Zurich's "Autonomous Mobile Robots" course (Spring 2021)



Perception II
Fundamentals of Computer Vision

Autonomous Mobile Robots

Margarita Chli
Roland Siegwart and Nick Lawrance

Autonomous Mobile Robots
Margarita Chli, Nick Lawrance and Roland Siegwart

Perception II | 1



Perception III
Fundamentals of Image Processing (incl. image features)

Autonomous Mobile Robots

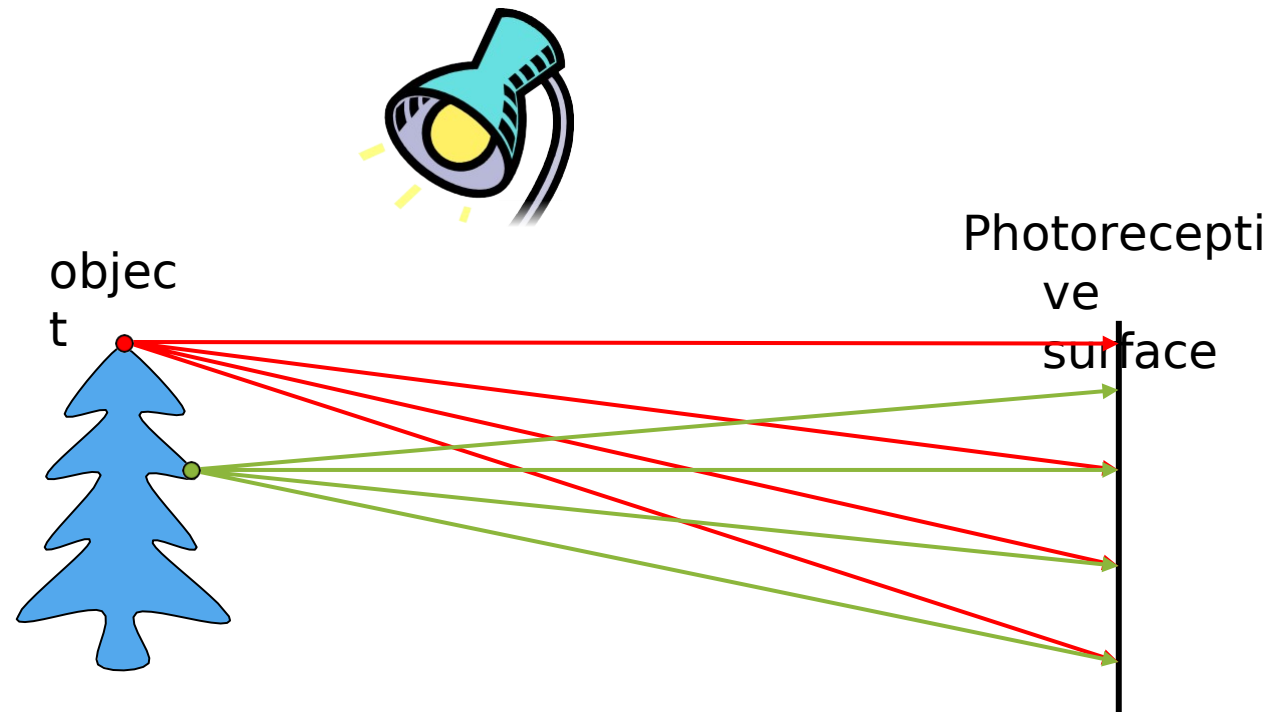
Margarita Chli
Roland Siegwart and Nick Lawrance

Autonomous Mobile Robots
Margarita Chli, Nick Lawrance and Roland Siegwart

Perception III | 1

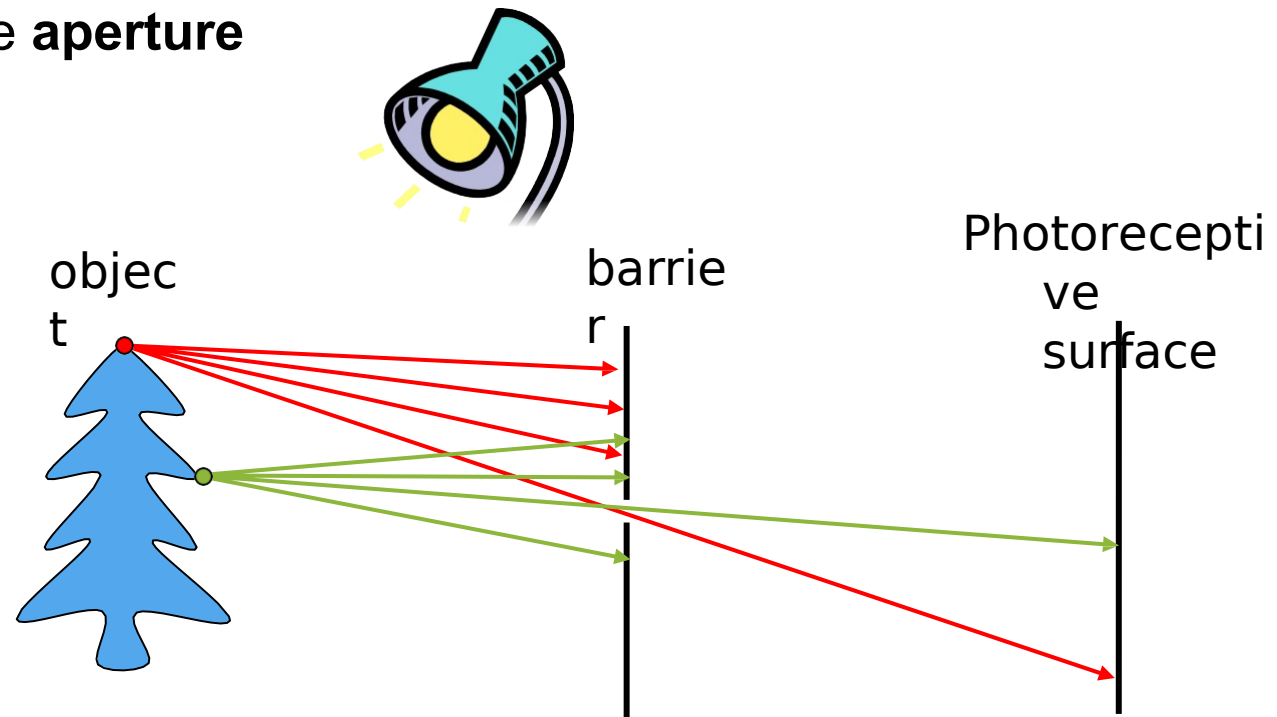
The camera | image formation

- If we place a piece of film in front of an object, do we get a reasonable image?

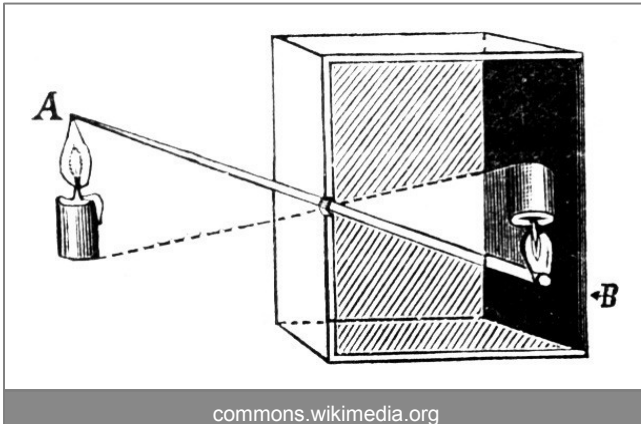


The camera | image formation

- If we place a piece of film in front of an object, do we get a reasonable image?
- Add a barrier to block off most of the rays
 - This reduces blurring
 - The opening is known as the **aperture**

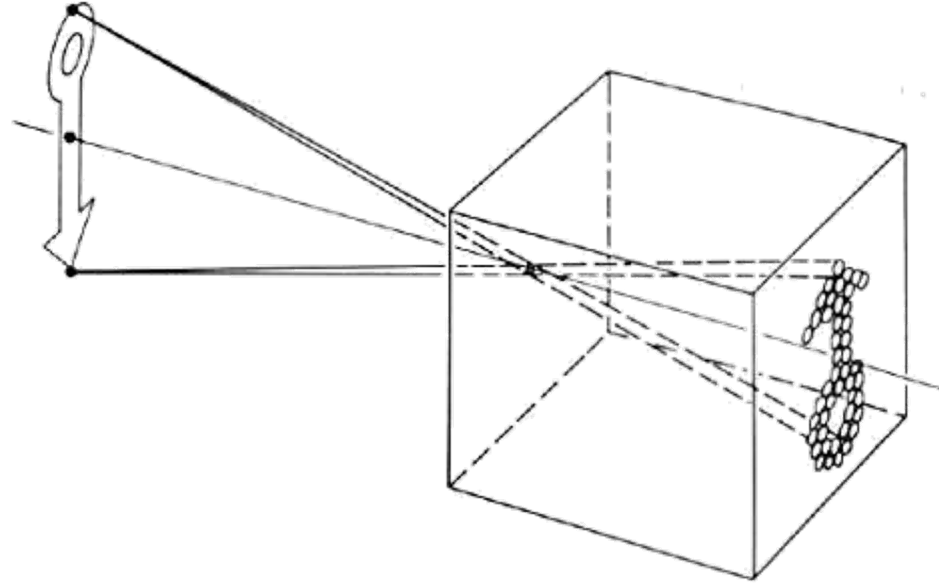


The camera | camera obscura



- Basic principle known to Mozi (470-390 BC), Aristotle (384-322 BC), Euclid (323-283 BC)
- Drawing aid for artists: described by Leonardo da Vinci (1452-1519)
- Image is inverted
- Depth of the room (box) is the effective focal length

The camera | the pinhole camera model

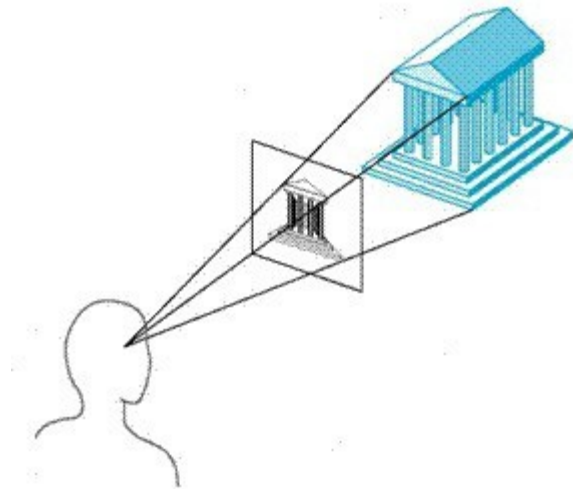


- Pinhole model:
 - Captures **beam of rays** – all rays through a single point (note: no lens!)
 - The point is called **Center of Projection** or **Optical Center**
 - The image is formed on the **Image Plane**
- We will use the pinhole camera model to describe how the image is formed

Based on slide by Steve Seitz

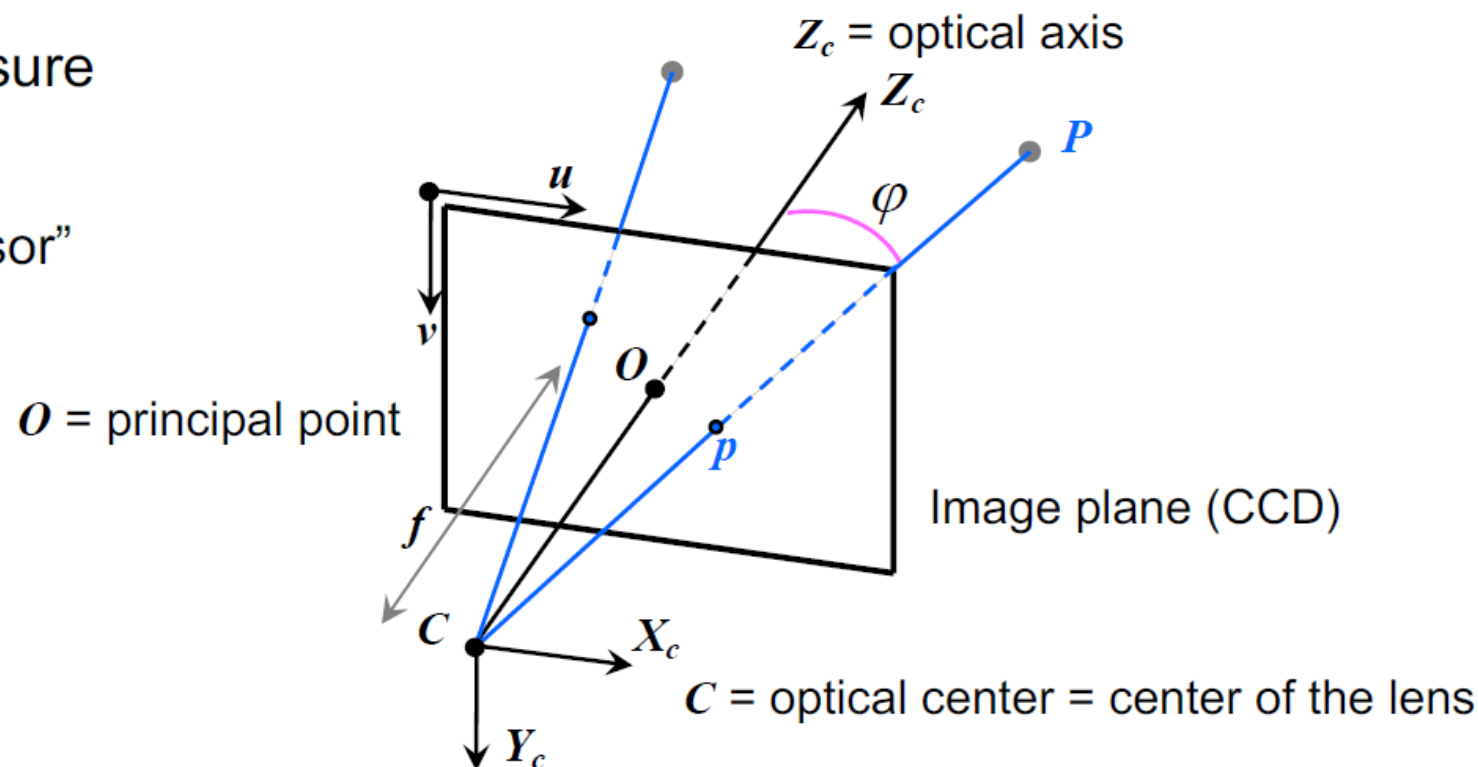
Perspective projection

How world points map to pixels in the image?



Perspective projection

- For convenience: the image plane is usually represented in front of C , such that the image preserves the same orientation (i.e. not flipped)
- A camera does not measure distances but angles!
⇒ a camera is a “bearing sensor”



Perspective projection | from world to pixel coordinates

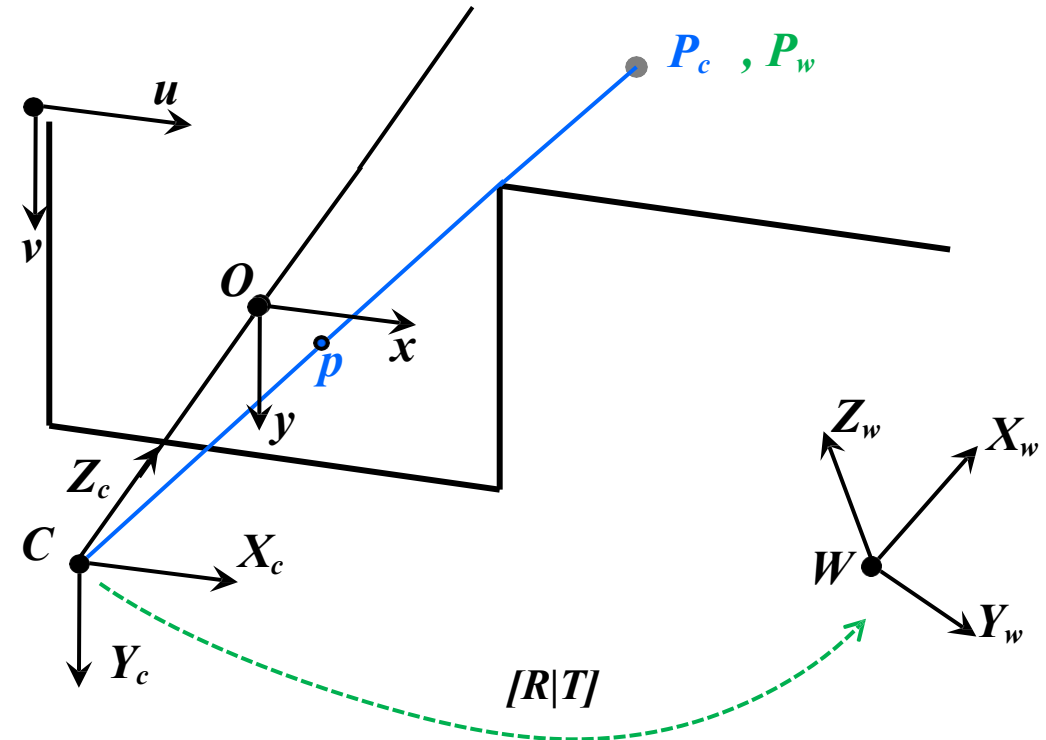
Find pixel coordinates (u,v) of point P_w in the world frame:

0. Convert world point P_w to camera point P_c

Find pixel coordinates (u,v) of point P_c in the camera frame:

1. Convert P_c to image-plane coordinates (x,y)

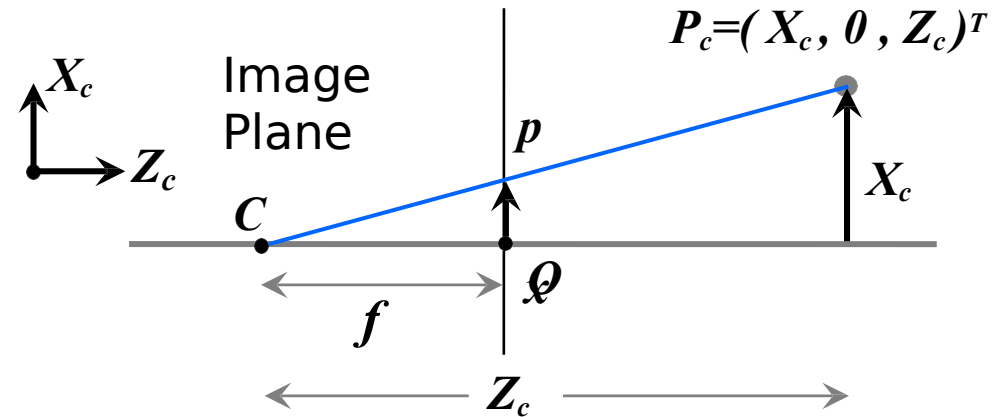
2. Convert P_c to (discretised) pixel coordinates (u,v)



Perspective projection | from camera frame to image plane

- The Camera point $P_c = (X_c, 0, Z_c)^T$ 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, in the general case:

$$\frac{y}{Y_c f} = \frac{Y_c}{Z_c} \Rightarrow y = \frac{fY_c}{Z_c}$$

1. Convert P_c to image-plane coordinates (x, y)

2. Convert P_c to (discretised) pixel coordinates (u, v)

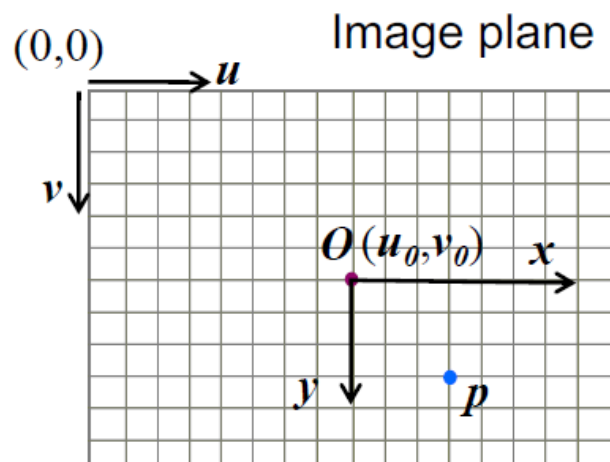
Perspective projection | from camera frame to pixel coords.



- Convert p from the local image plane coords (x,y) to the pixel coords (u,v) , we need to account for:
 - the pixel coords of the camera optical center $O = (u_0, v_0)$
 - scale factors k_u, k_v for the pixel-size in both dimensions

So: $u = u_0 + k_u x \Rightarrow u = u_0 + \frac{k_u f X_c}{Z_c}$

$$v = v_0 + k_v y \Rightarrow v = v_0 + \frac{k_v f Y_c}{Z_c}$$



- Use **Homogeneous Coordinates** for linear mapping from 3D to 2D, by introducing an extra element (scale):

$$p = \begin{pmatrix} u \\ v \end{pmatrix} \Rightarrow \tilde{p} = \begin{bmatrix} \lambda u \\ \lambda v \\ \lambda \end{bmatrix} = \lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} \text{ and similarly for the world coordinates. Note, usually } \lambda = 1$$

Perspective projection | from camera frame to pixel coords.

$$u = u_0 + \frac{k_u f X_c}{Z_c}$$

$$v = v_0 + \frac{k_v f Y_c}{Z_c}$$

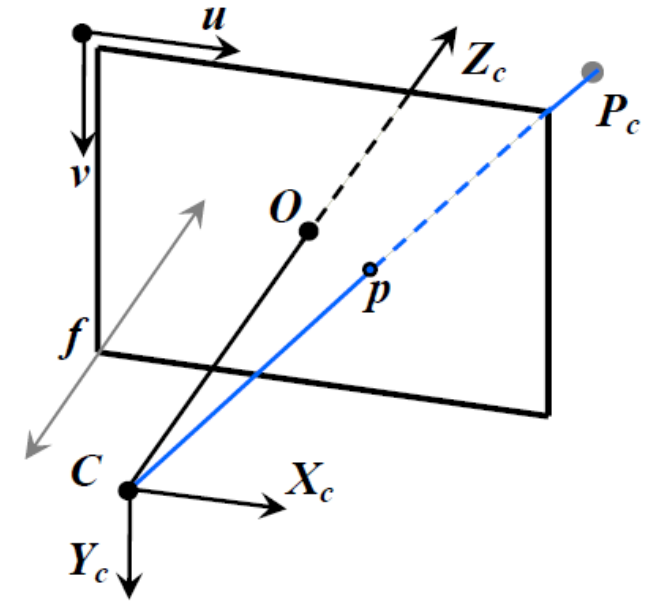
- Expressed in matrix form and homogeneous coordinates:

$$\begin{bmatrix} \lambda u \\ \lambda v \\ \lambda \end{bmatrix} = \begin{bmatrix} k_u f & 0 & u_0 \\ 0 & k_v f & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix}$$

- Or alternatively

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

Focal length in u -direction
 Focal length in v -direction
 “Calibration matrix”/
 “Matrix of Intrinsic Parameters”

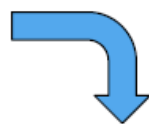


Perspective projection | from the world to the camera frame



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

in homogeneous coordinates



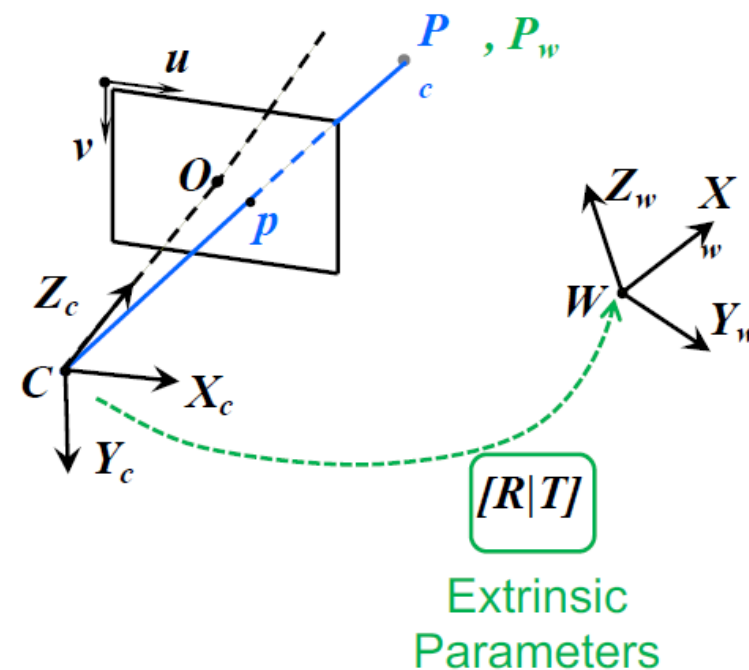
$$\begin{bmatrix} X_c \\ Y_c \\ Z_c \\ 1 \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_1 \\ r_{21} & r_{22} & r_{23} & t_2 \\ r_{31} & r_{32} & r_{33} & t_3 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix} = \begin{bmatrix} R & T \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

From the previous slide:

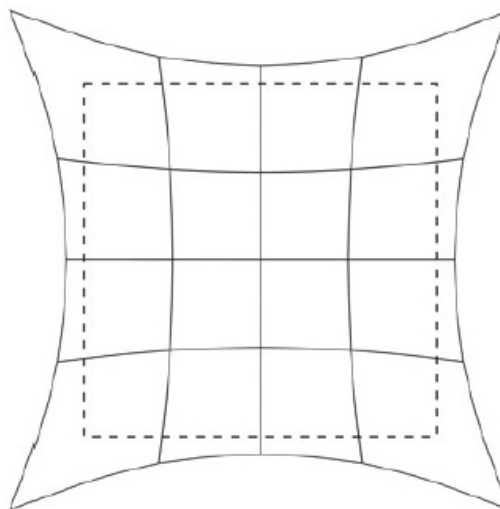
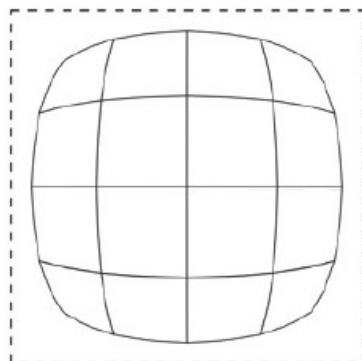
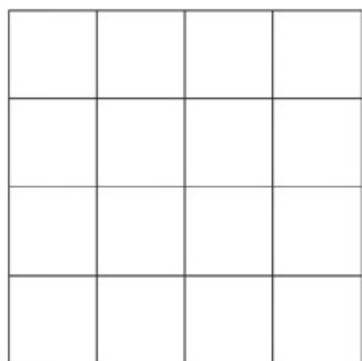
$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix}$$

Projection Matrix

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K \begin{bmatrix} R & T \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$



Perspective projection | radial distortion



No distortion



Barrel distortion



Pincushion distortion

- Amount of distortion is a non-linear fⁿ of distance from center of image

From ideal (u, v) to distorted pixel coordinates (u_d, v_d) :

- Simple quadratic distortion model:

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

where

Radial
Distortion
parameter

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

- Works well for most lenses

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