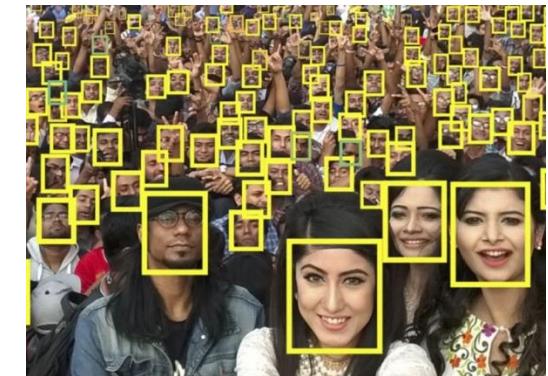
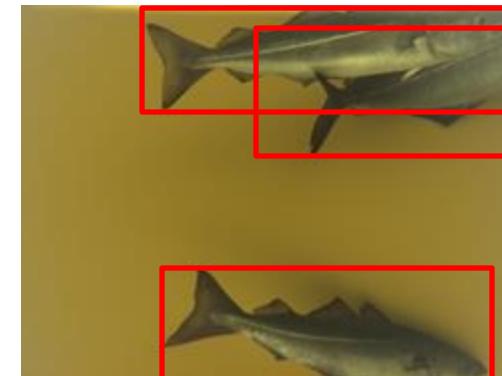
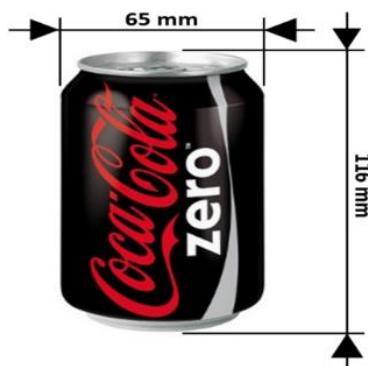
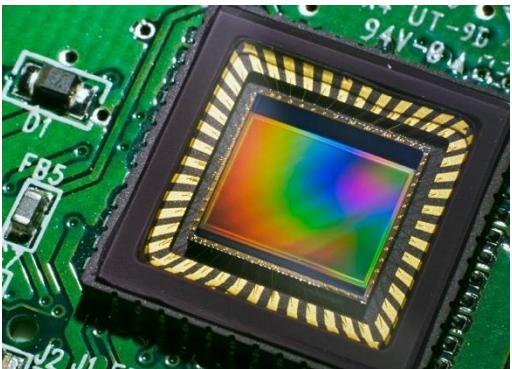




Multiple View Geometry

Image Formation and Camera Modeling





Objectives of this class

- Obtaining a first overview of how the image is formed in a camera
- Analysing the projection of a 3D scene in the image plane
- Knowing the *pinhole* model
- Understanding the equations that describe the geometry of the image formation process



Contents

Camera Modeling and Calibration

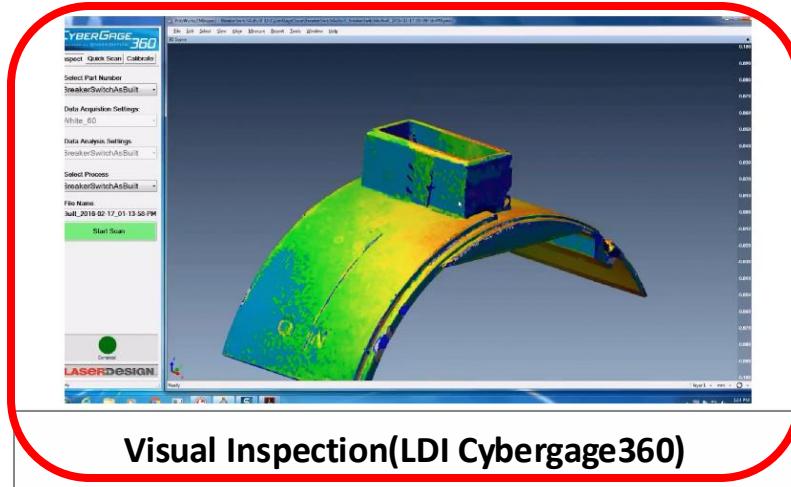
- 1 Introduction
- 2 *Pinhole* Model
- 3 Calibration



Introduction

Obtaining 3D information

- Applications





Introduction

Obtaining 3D information

- Applications

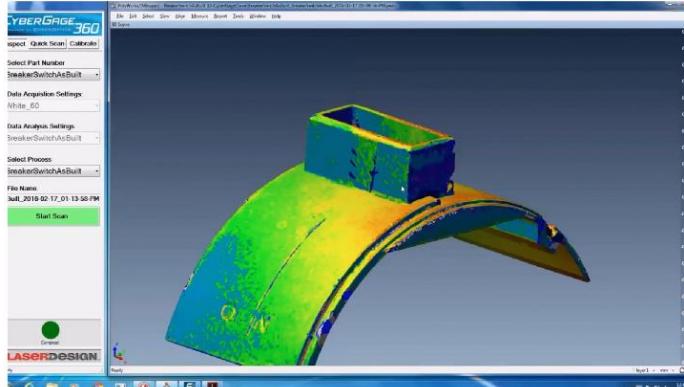




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Obtaining 3D information

- Applications



Visual Inspection(LDI Cybergage360)



Motion Modeling(Vicon + MotionBuilder)



Industrial Manipulation(Intel RealSense)



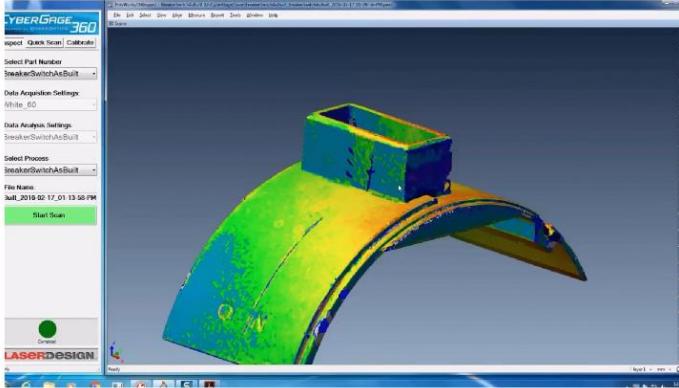
Autonomous Driving(Google car)



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Industrial Manipulation(Intel RealSense)

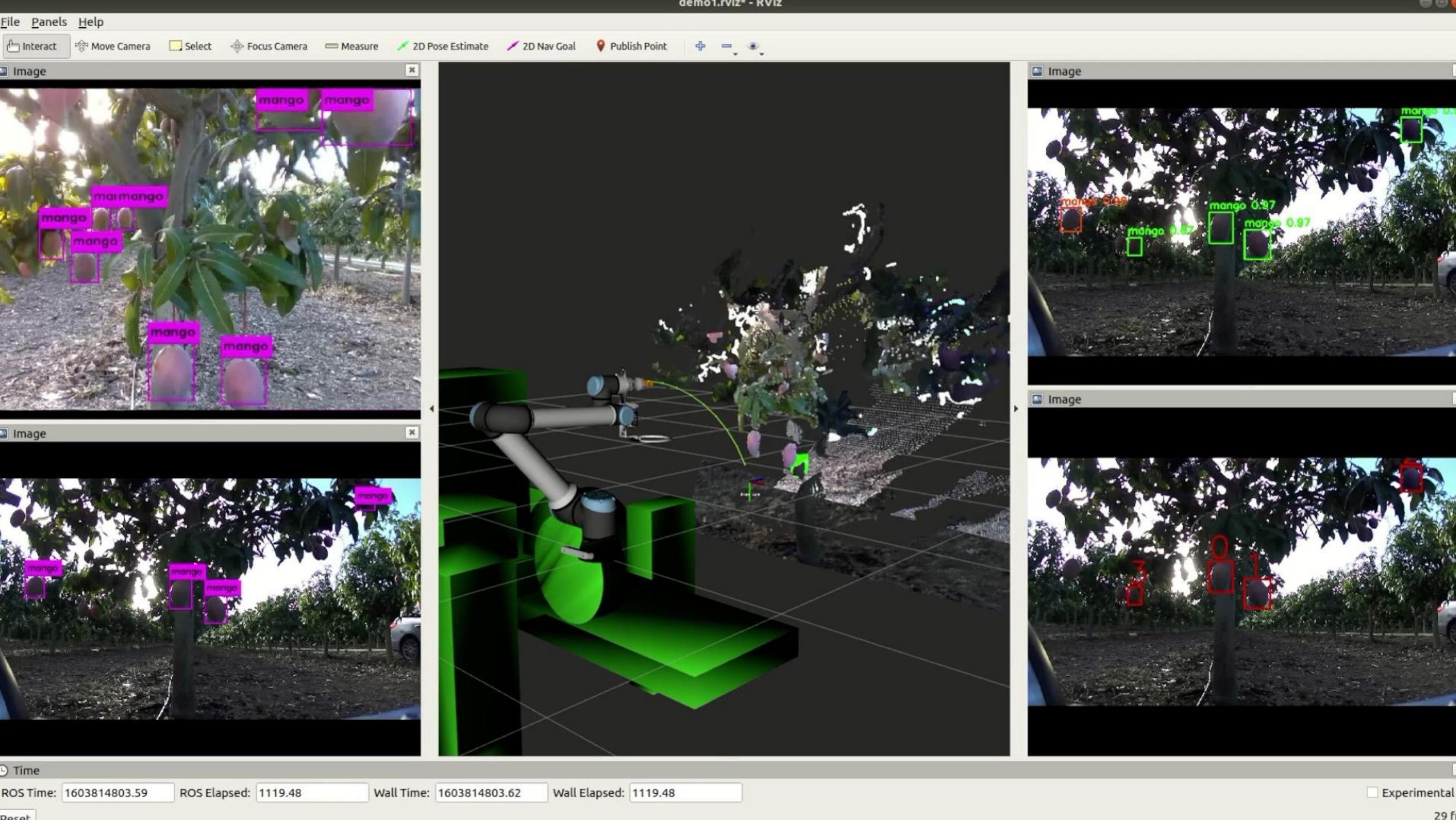


Motion Modeling(Vicon + MotionBuilder)



Autonomous Driving(Google car)



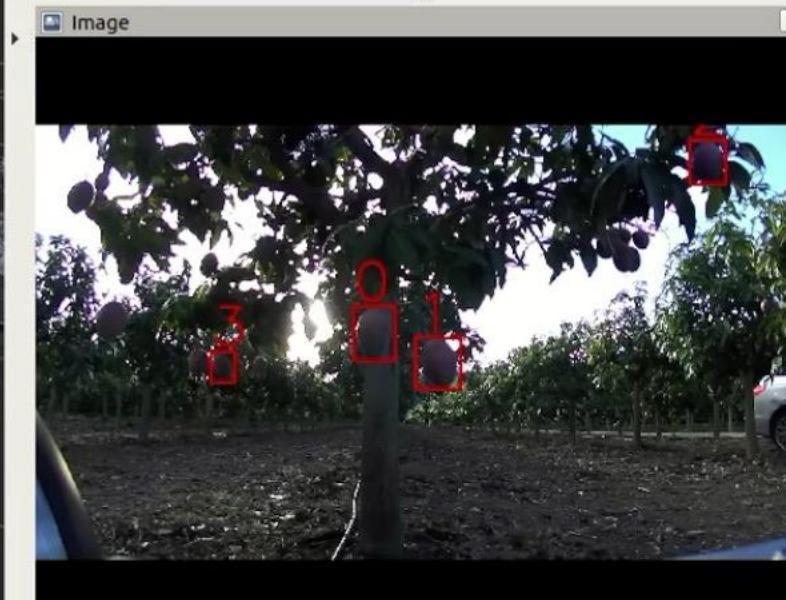
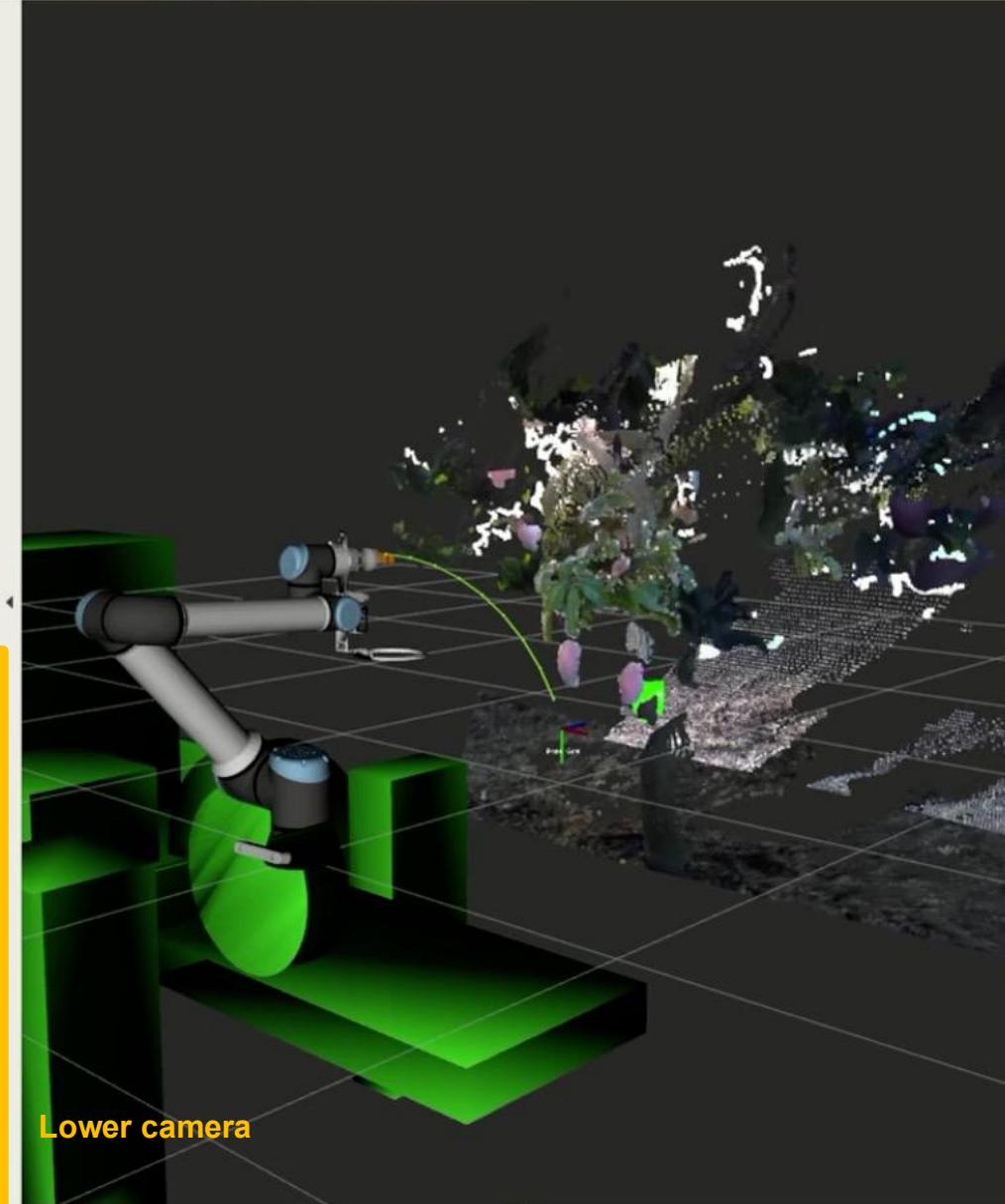
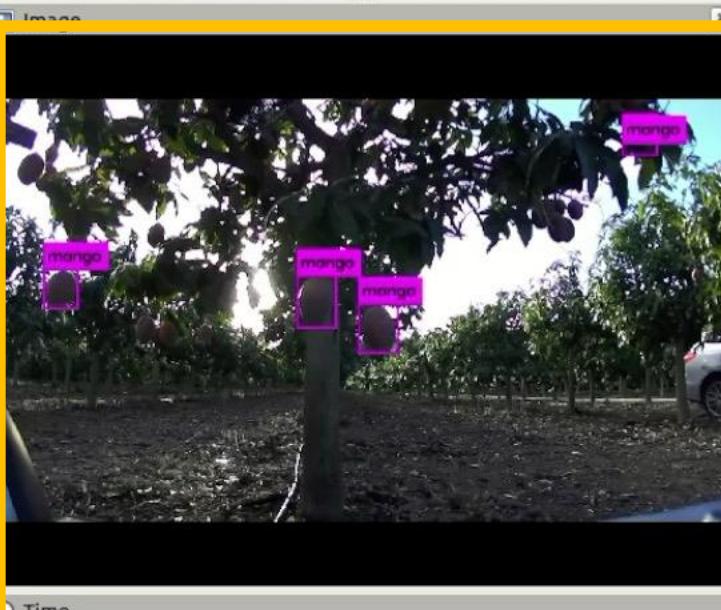
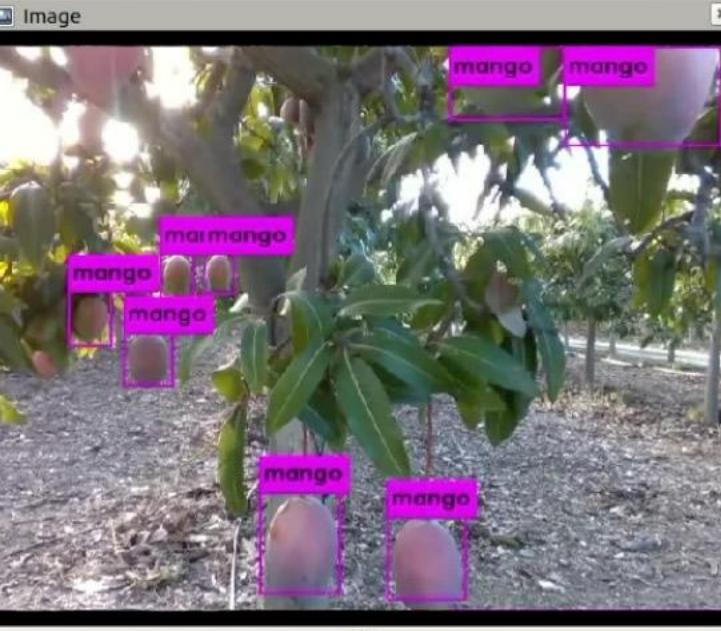


Interact Move Camera Select Focus Camera Measure 2D Pose Estimate 2D Nav Goal Publish Point

+

-

•



Interact

Move Camera

Select

Focus Camera

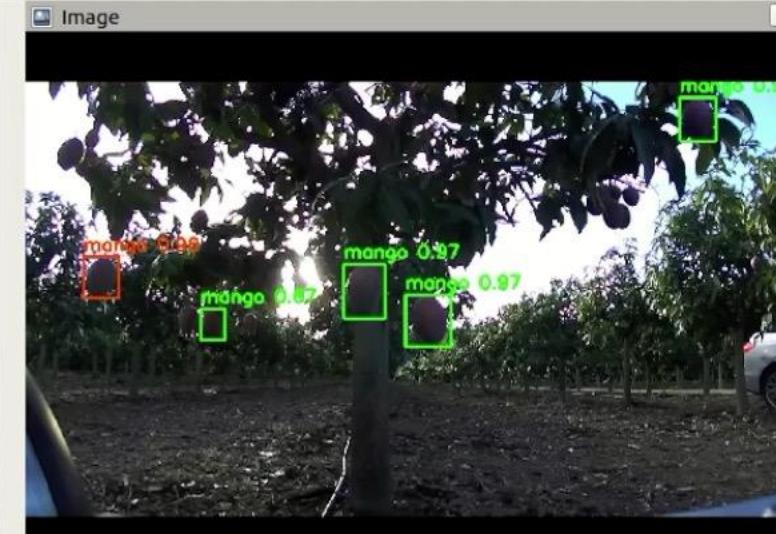
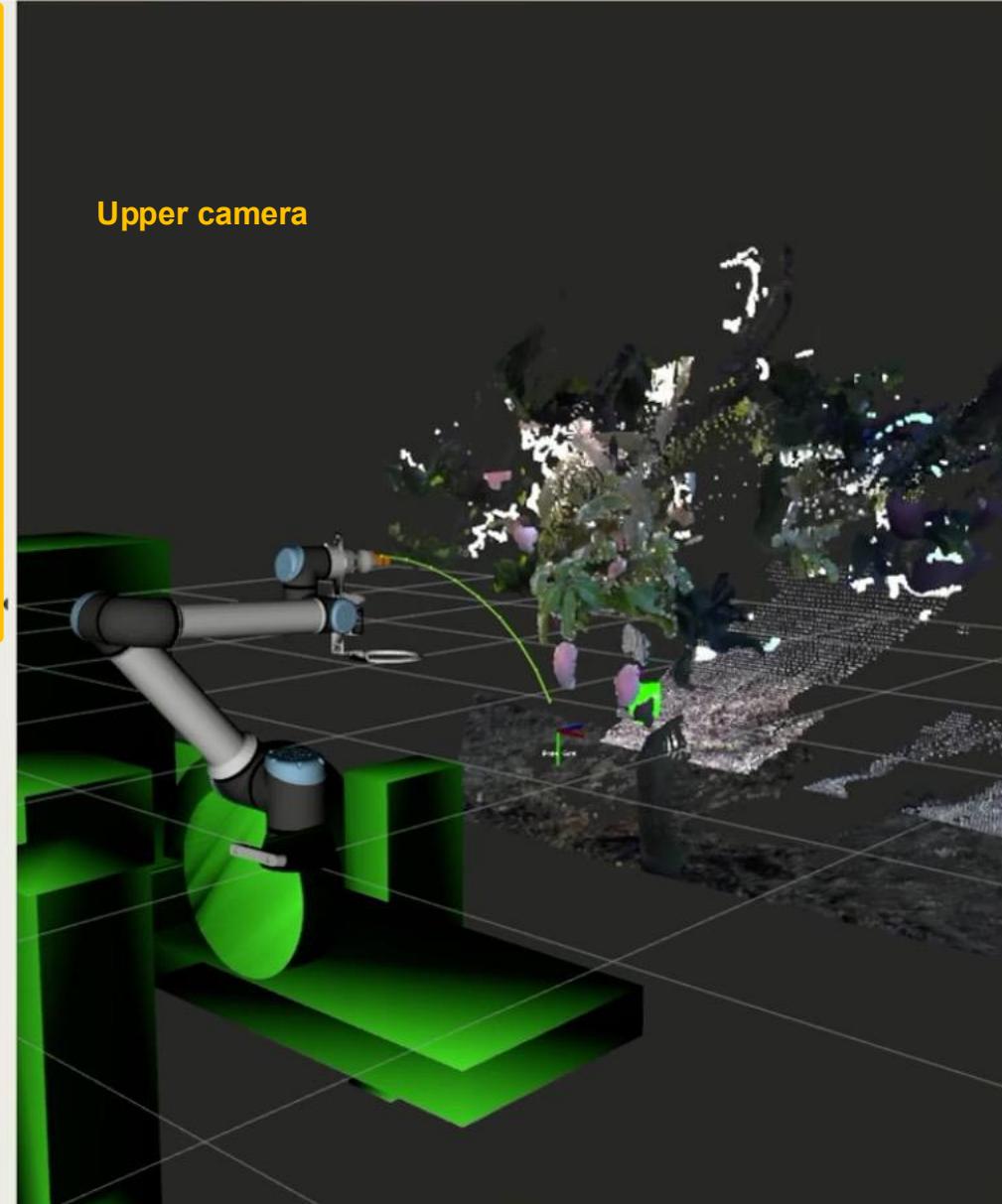
Measure

2D Pose Estimate

2D Nav Goal

Publish Point

+ - ⏪



Time

ROS Time: 1603814803.59

ROS Elapsed: 1119.48

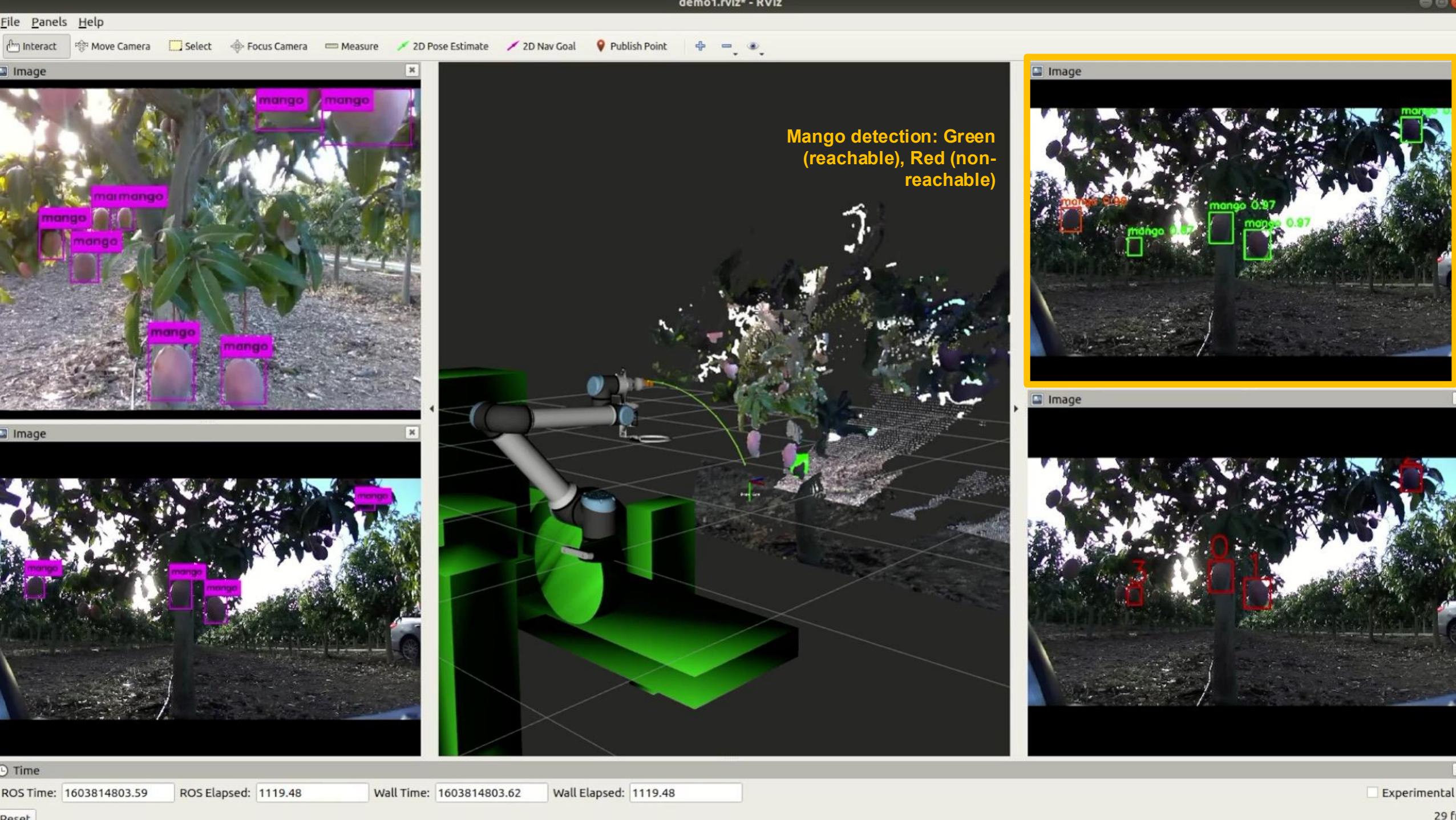
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Wall Elapsed: 1119.48

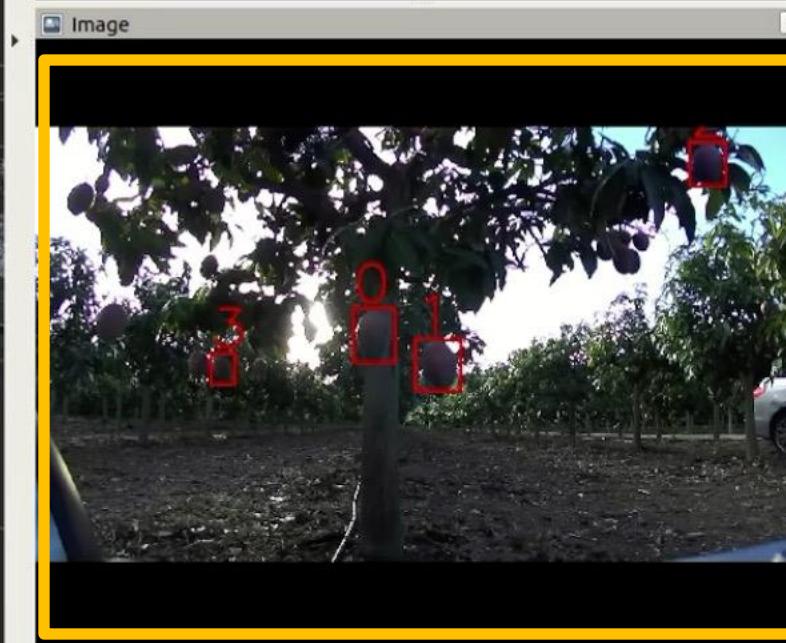
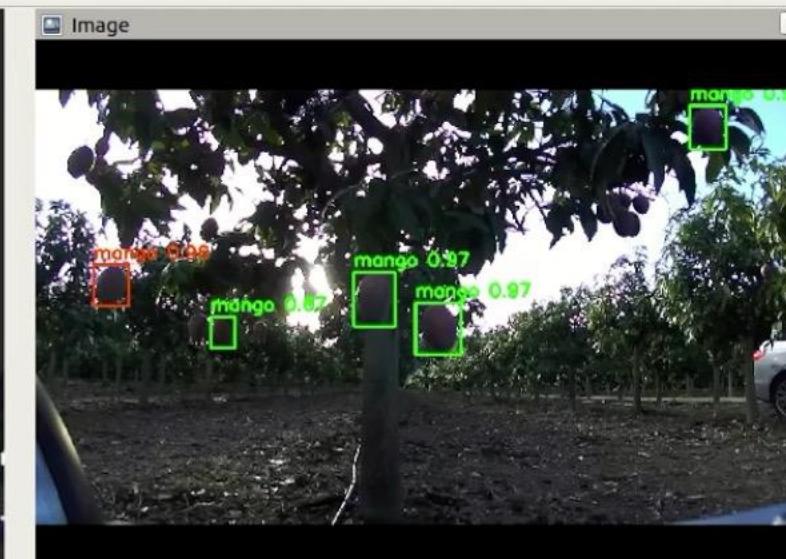
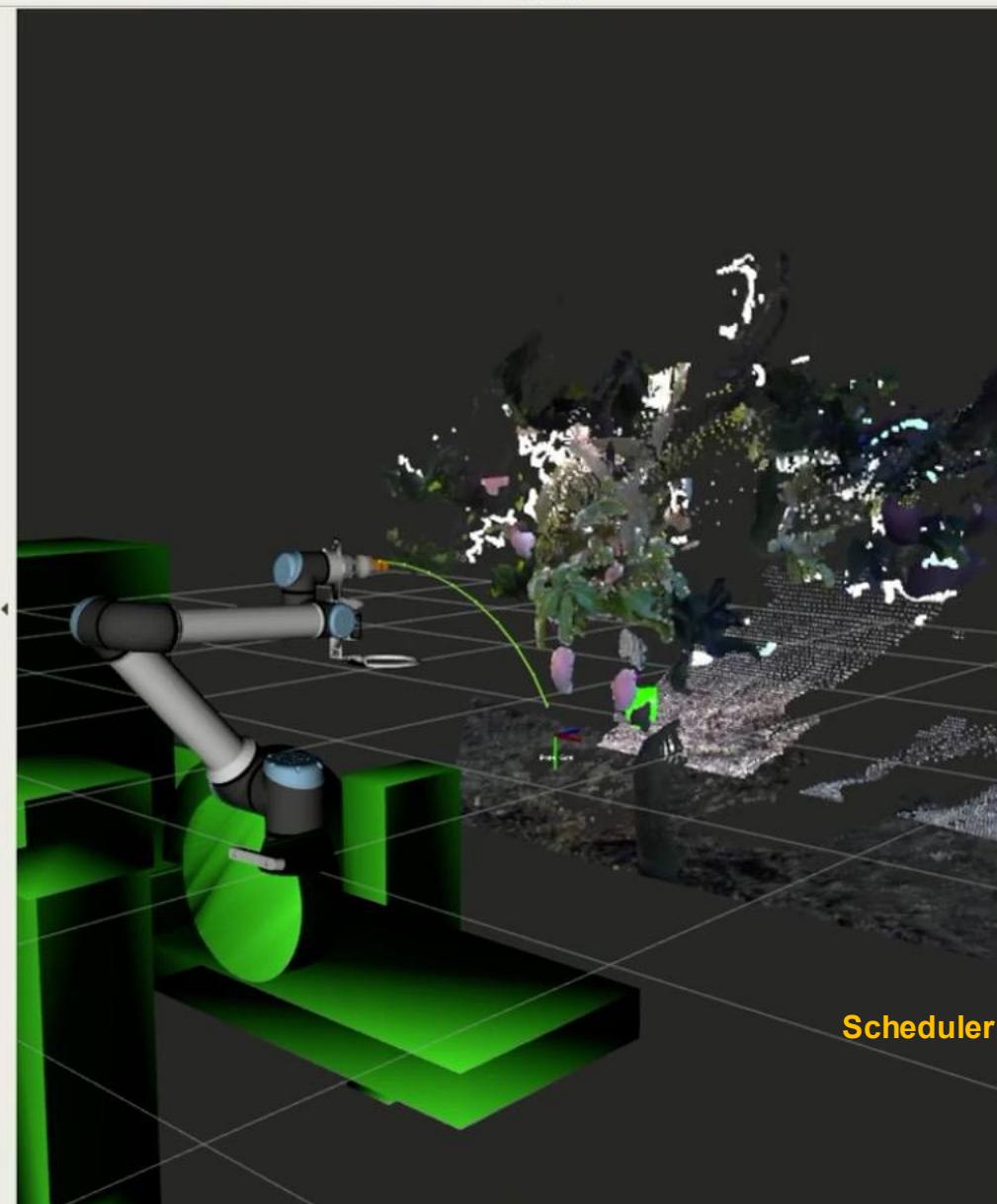
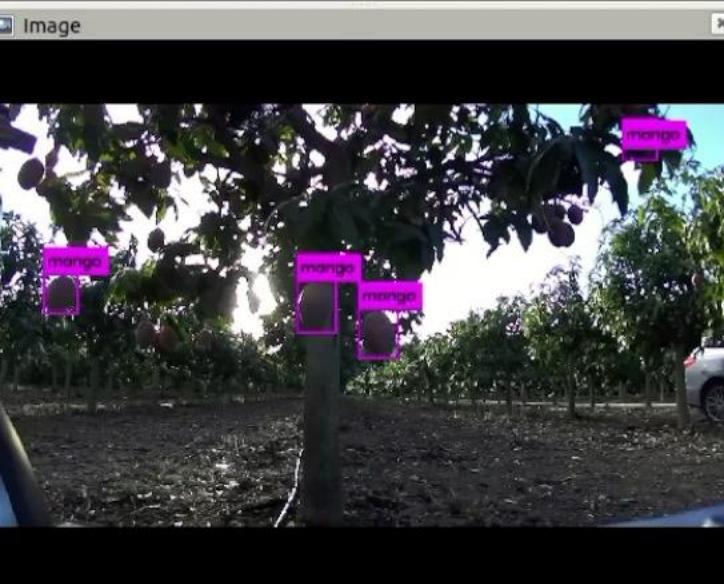
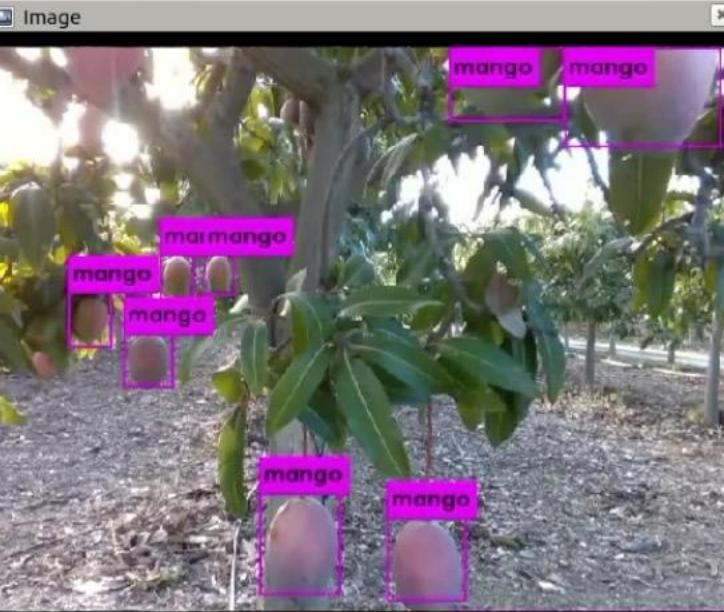
Experimental

Reset

29 fp



File Panels Help

 Interact Move Camera Select Focus Camera Measure 2D Pose Estimate 2D Nav Goal Publish Point + - 

Time

ROS Time: 1603814803.59

ROS Elapsed: 1119.48

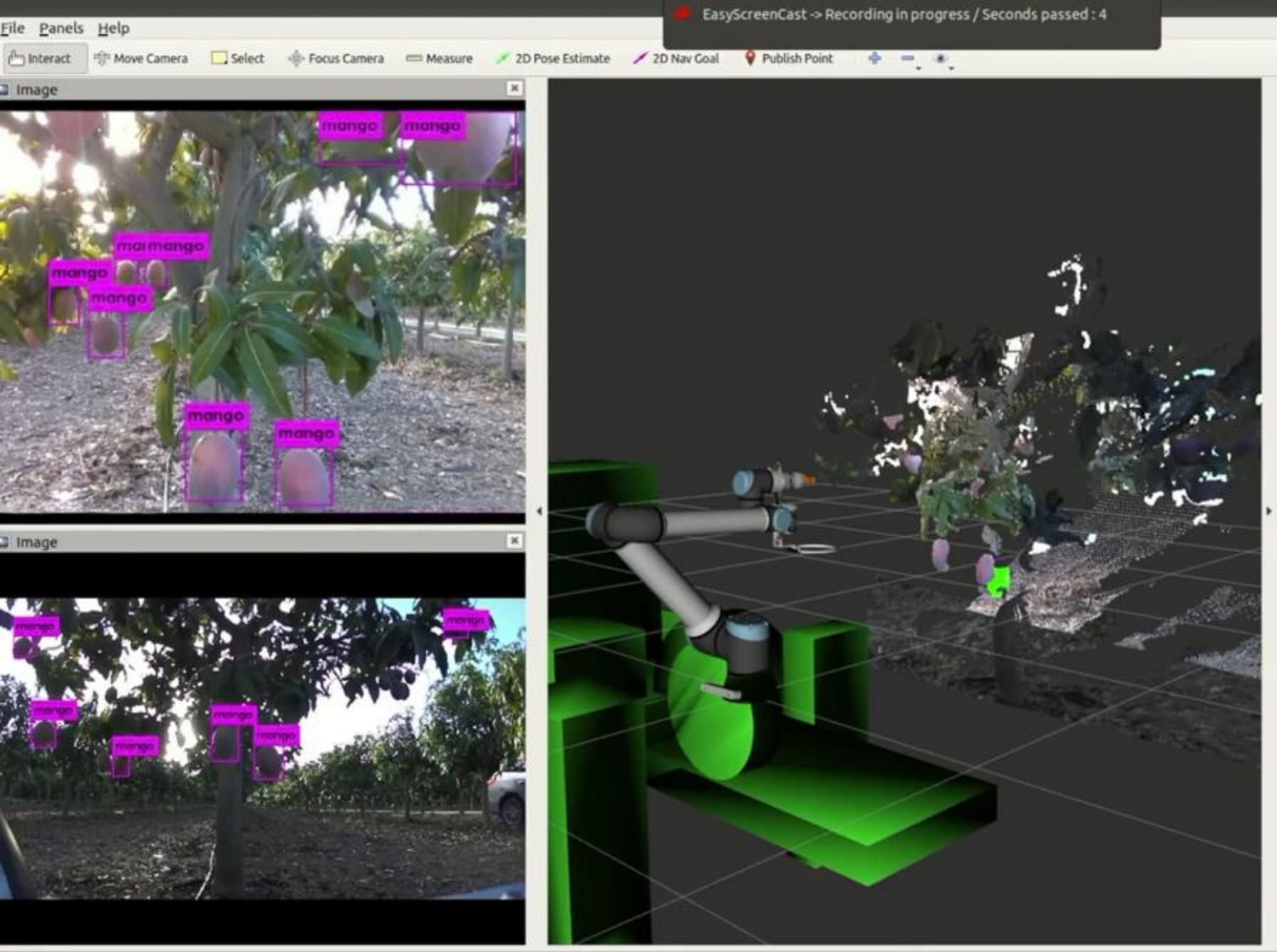
Wall Time: 1603814803.62

Wall Elapsed: 1119.48

Reset

Experimental

29 fp



Time

ROS Time: 1603814798.59 ROS Elapsed: 1114.48 Wall Time: 1603814798.63 Wall Elapsed: 1114.44

Experimental

Reset

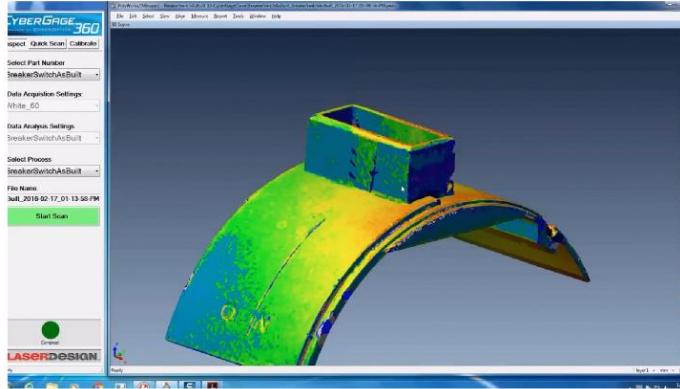
29 fp



Introduction

Obtaining 3D information

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Visual Inspection(LDI Cybergage360)

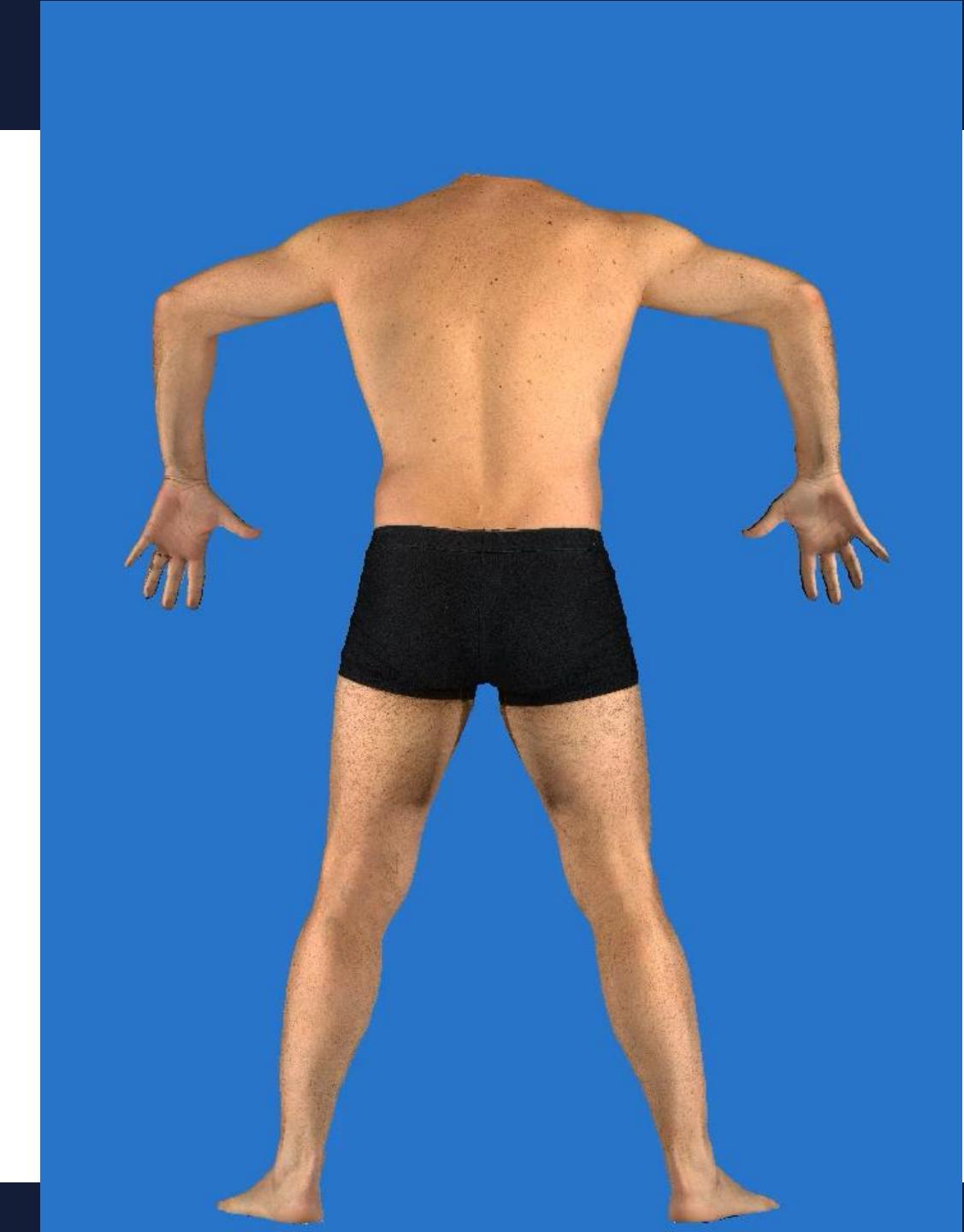


Industrial Manipulation(Intel RealSense)



Autonomous Driving(Google car)



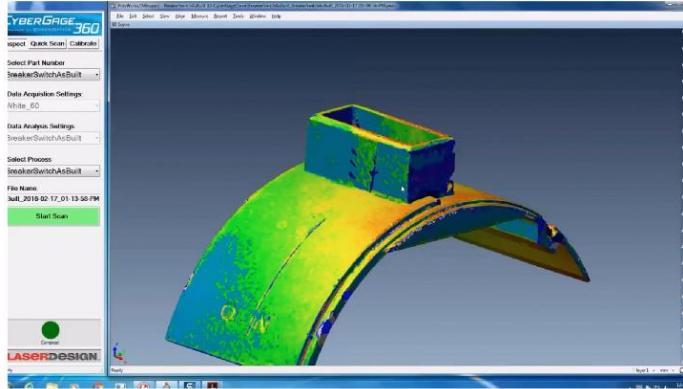




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Visual Inspection(LDI Cybergage360)



Industrial Manipulation(Intel RealSense)



Motion Modeling(Vicon + MotionBuilder)







Renaissance: first perspective paintings

Year 1415: Fillipo Brunelleschi

Year 1426: Masaccio

Il Pagamento del tributo. Capella Brancacci, Florence



https://it.wikipedia.org/wiki/Pagamento_del_tributo#/media/File:Masaccio7.jpg



Renaissance: first perspective paintings

Year 1470: Piero della Francesca

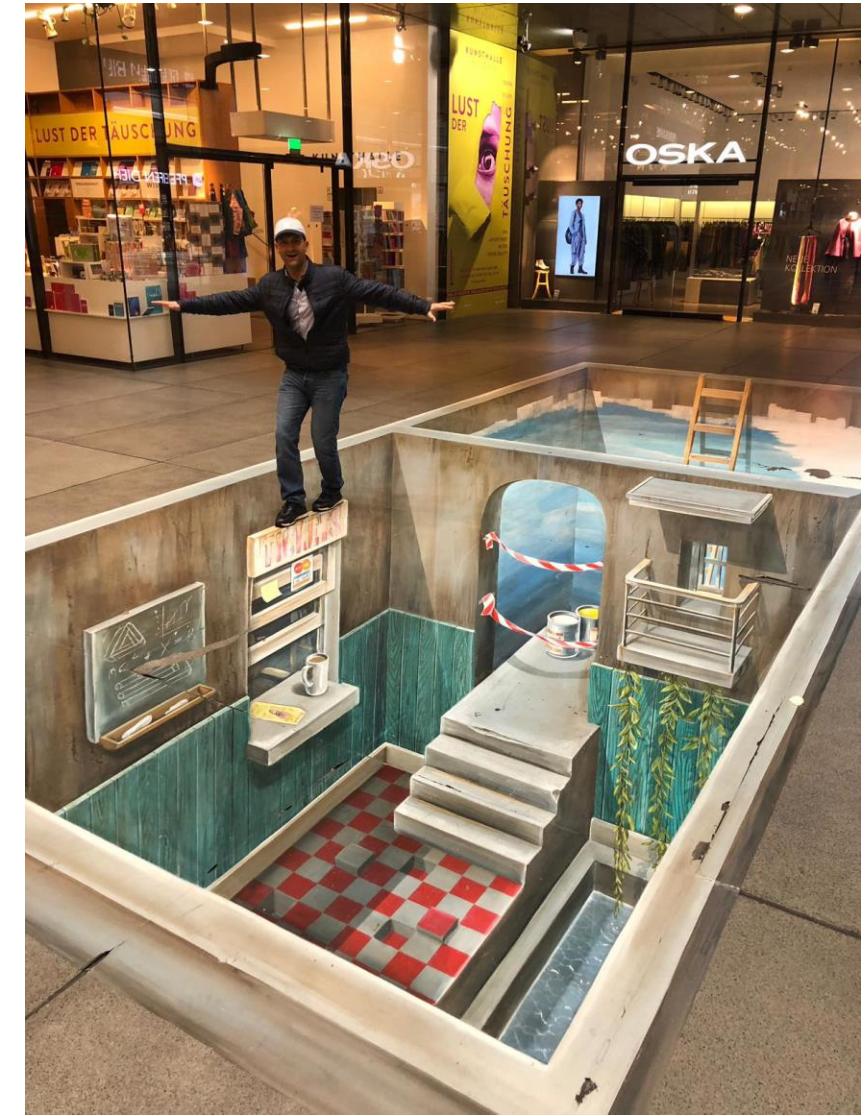
"The Ideal City"



<https://www.nytimes.com/2012/05/09/arts/09iht-conway09.html>



Perspective Projection





Perspective Projection



<http://www.julianbeever.net/>



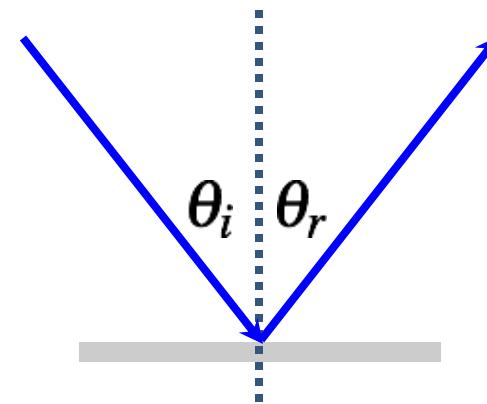
Perspective Projection



<http://www.julianbeever.net/>



Light Reflection

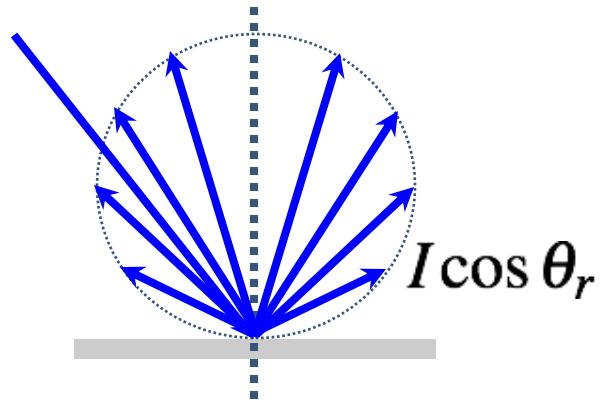


- **Specular reflection**
 - angle of incidence equals angle of reflection





Light Reflection



- **Lambertian reflection**
 - diffuse/matte surface
 - brightness invariant to observer's angle of view

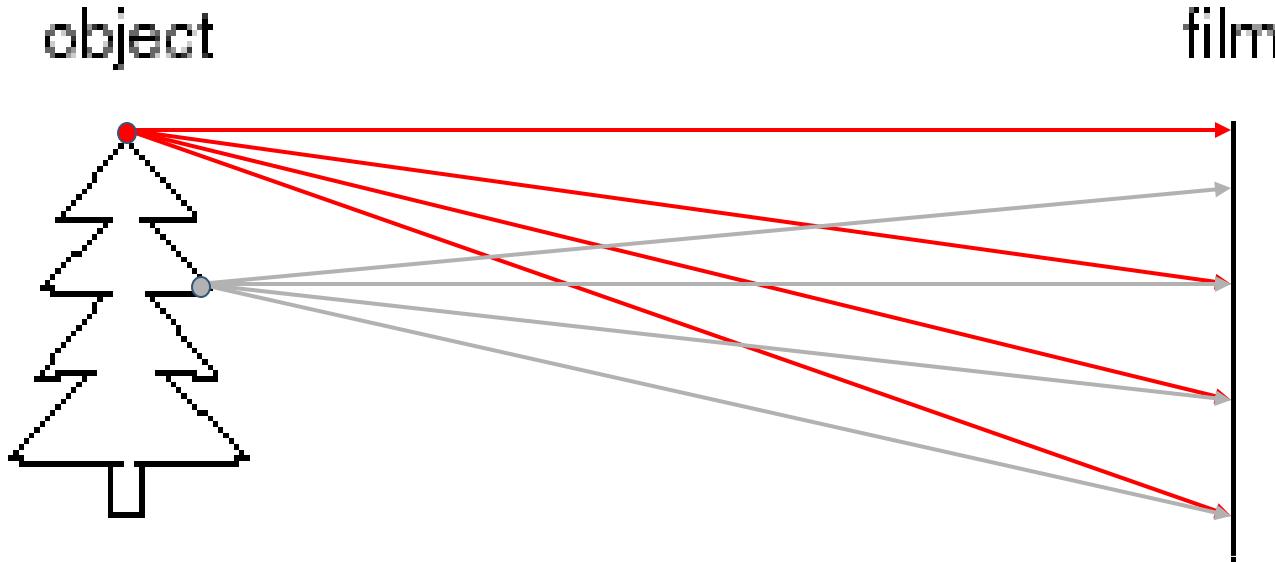


Johann Heinrich
Lambert 1728-1777





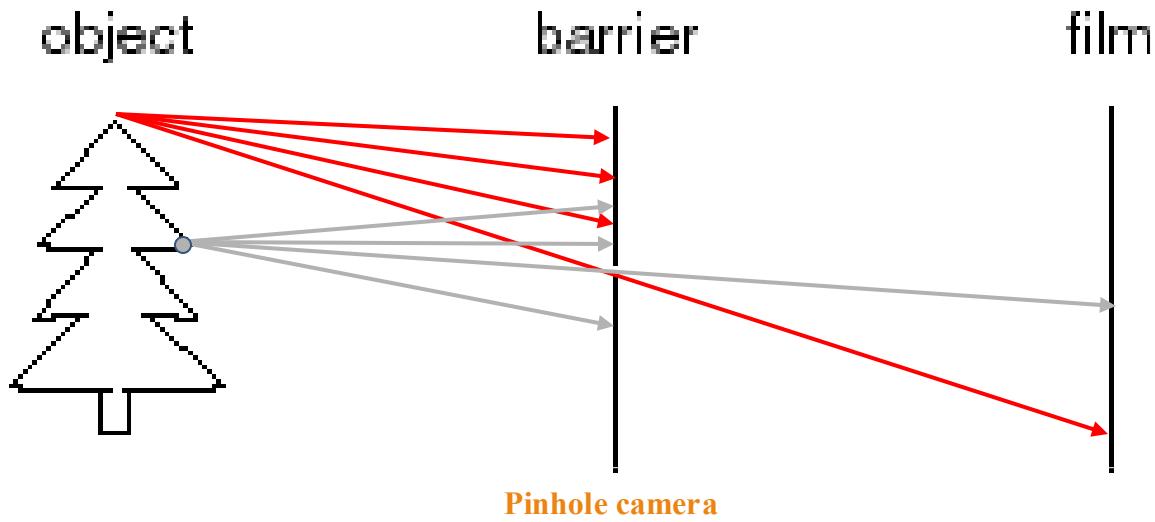
Pinhole Camera



- We place a photosensitive film in front of the object. Would this produce a reasonable image?
- What should be do?



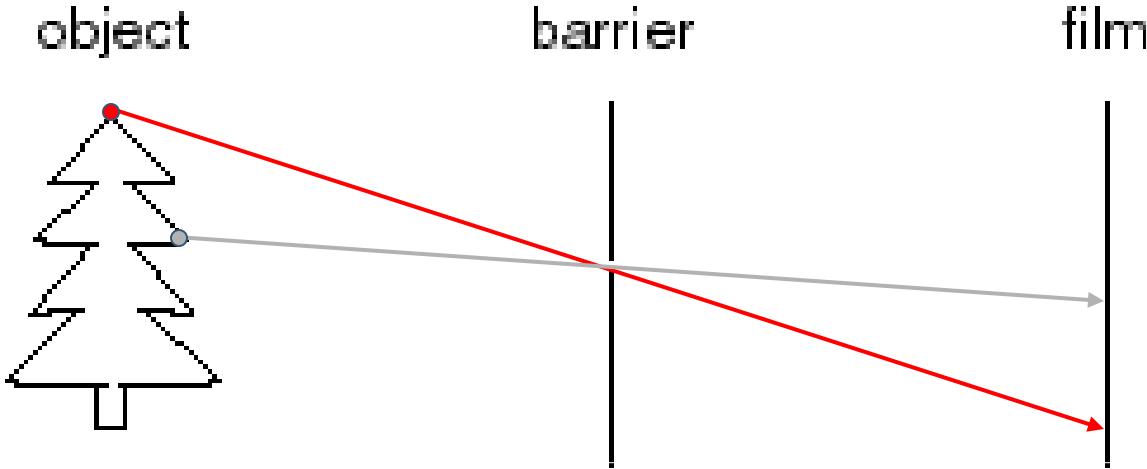
Pinhole Camera



- Add a barrier with a small pinhole



Pinhole Camera

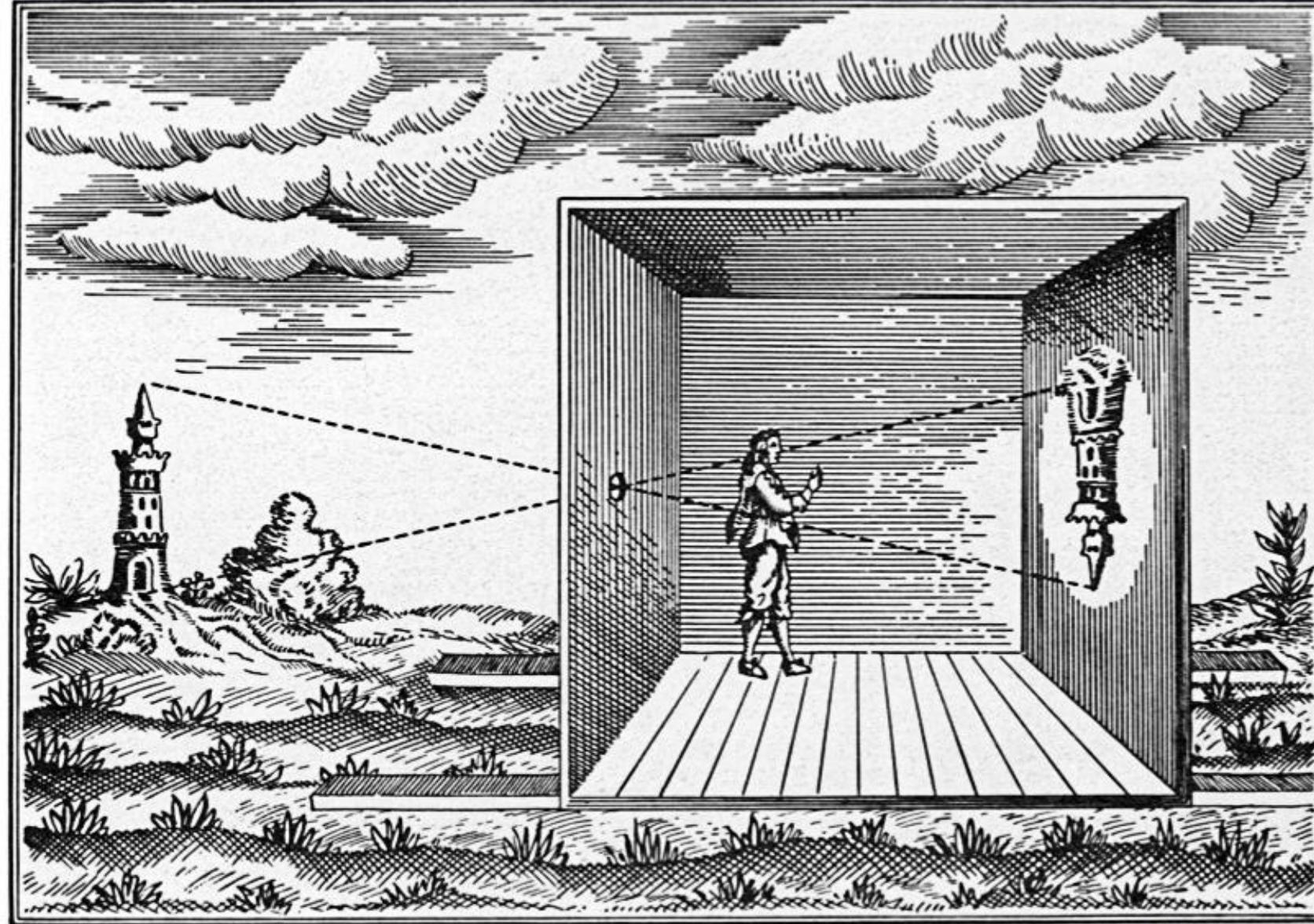


- Pinhole Model
- Camera obscura

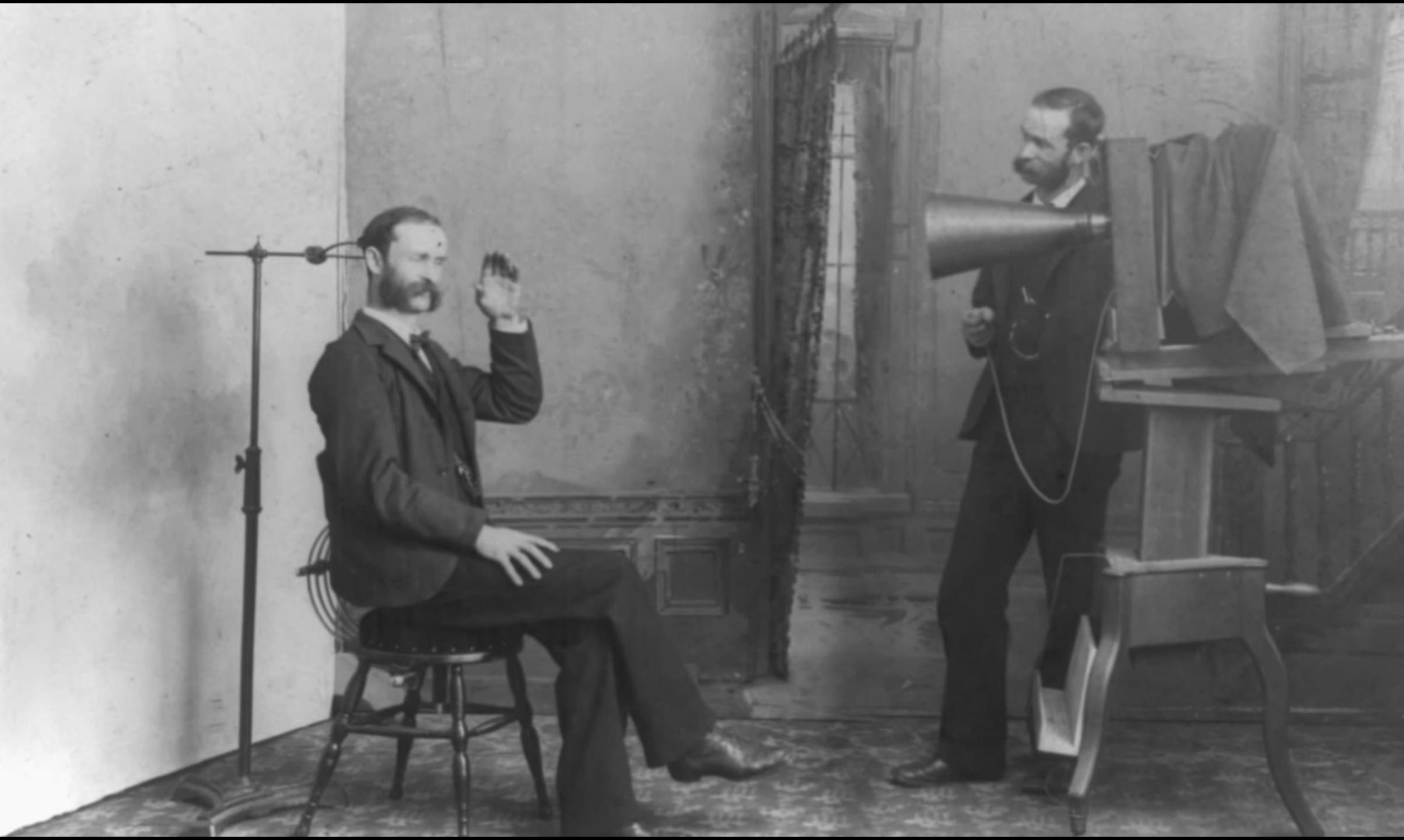
Slide by Steve Seitz



Pinhole Camera

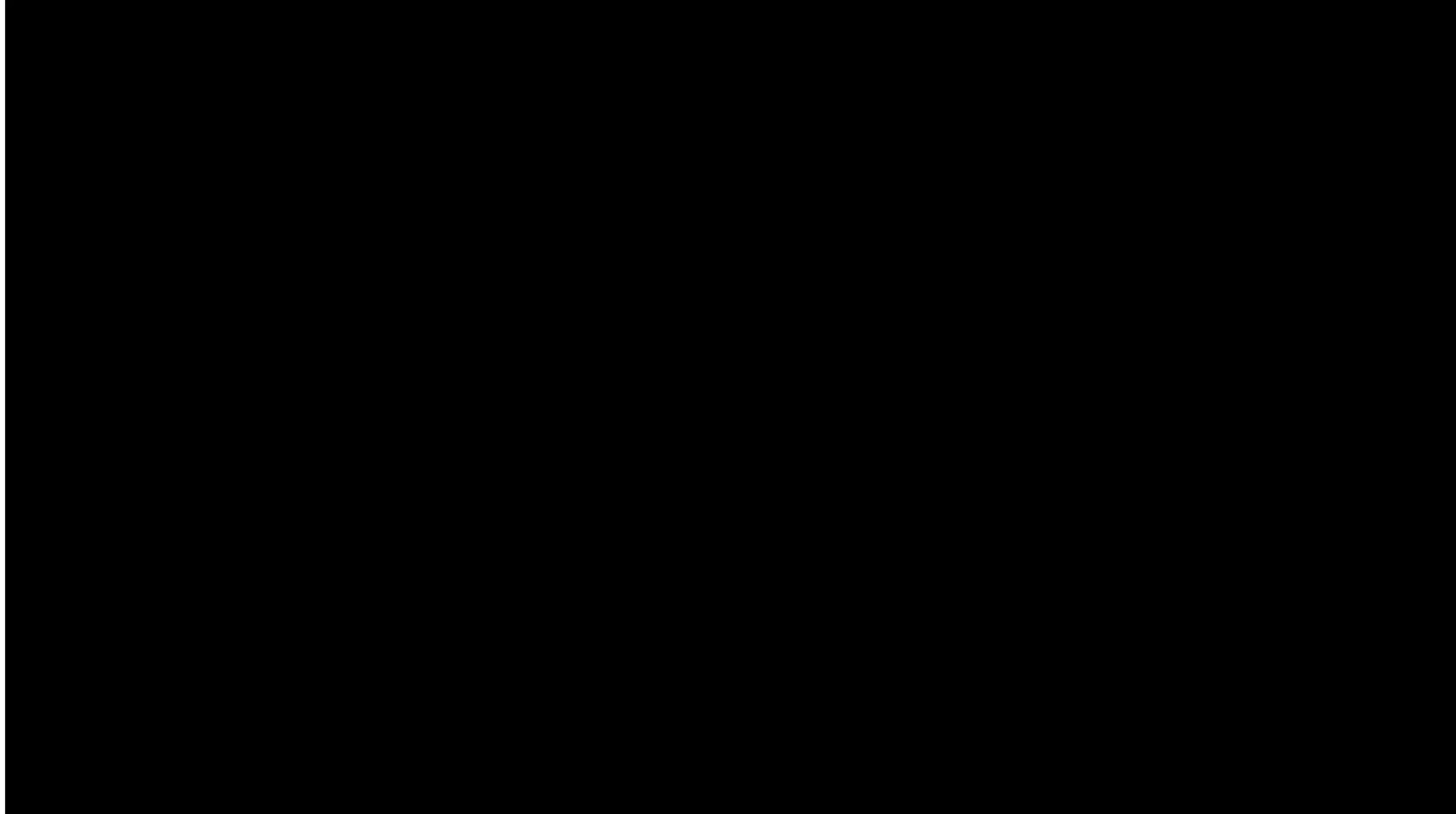






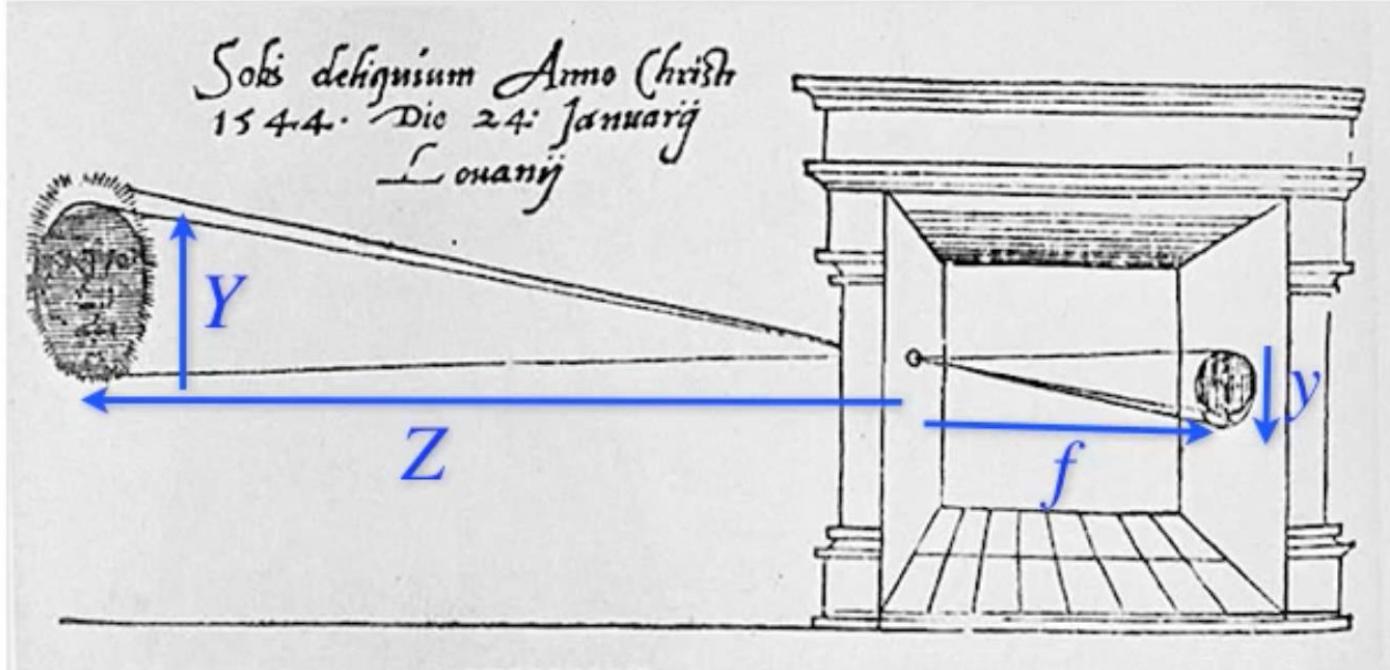


Pinhole Camera





Pinhole Camera: equations

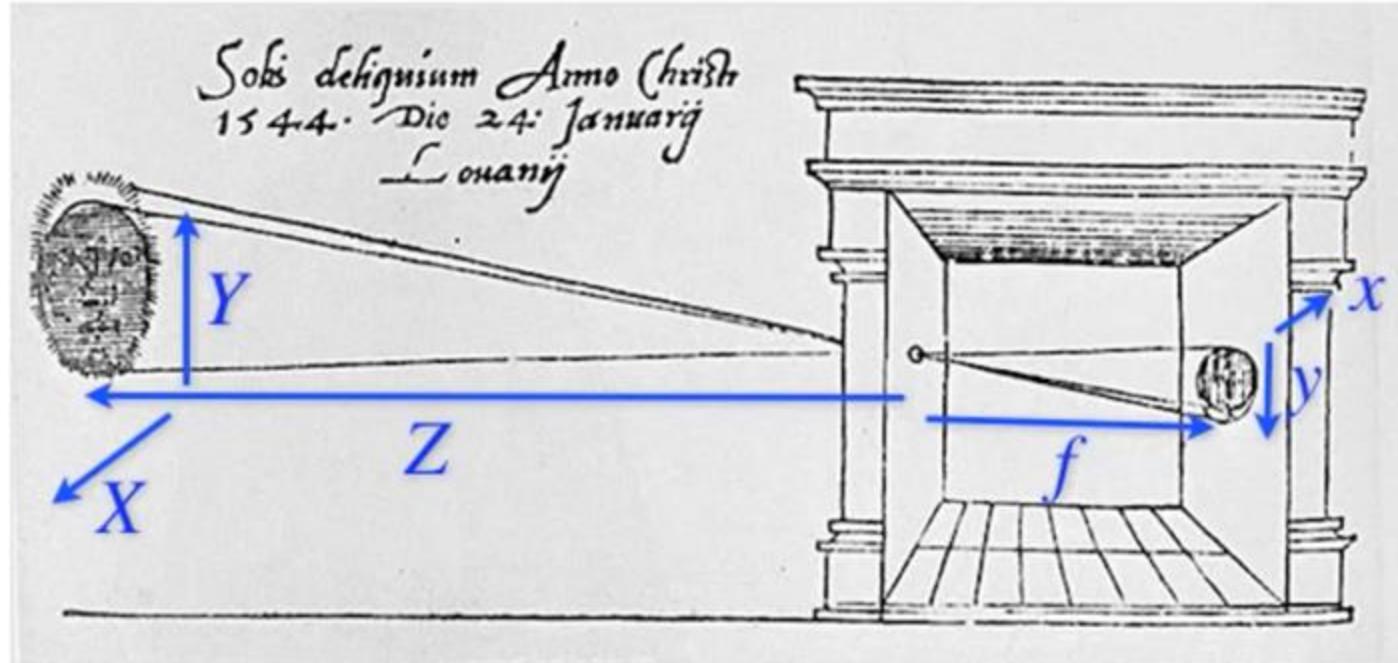


$$\frac{Y}{Z} = \frac{y}{f}$$

- Image formation is the mapping of scene points to the image plane



Pinhole Camera: equations



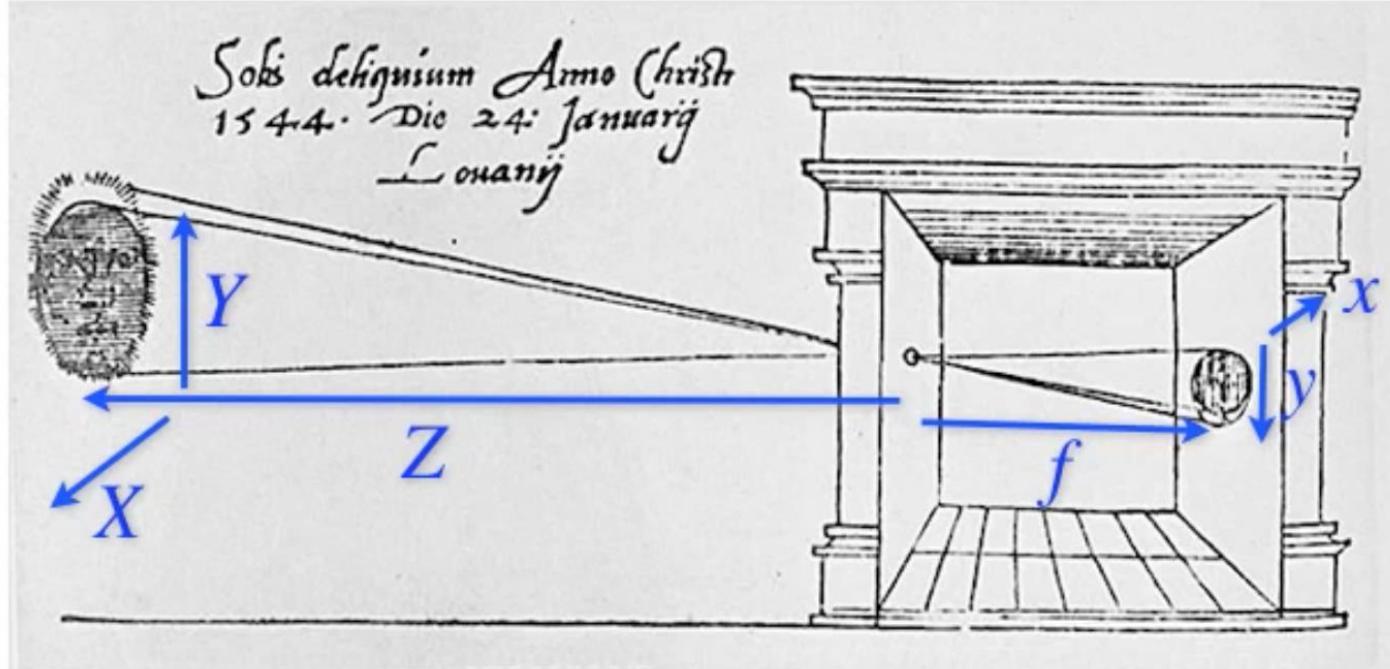
$$\frac{Y}{Z} = \frac{y}{f}$$

$$\frac{X}{Z} = \frac{x}{f}$$

- Image formation is the mapping of scene points to the image plane



Pinhole Camera: equations



$$\frac{Y}{Z} = \frac{y}{f}$$

$$\frac{X}{Z} = \frac{x}{f}$$

$$x = \frac{fX}{Z}, y = \frac{fY}{Z} \quad (X, Y, Z) \mapsto (x, y)$$
$$\mathbb{R}^3 \mapsto \mathbb{R}^2$$



Forced Perspective

2013

Kenzie Saunders | CC BY-ND 2.0



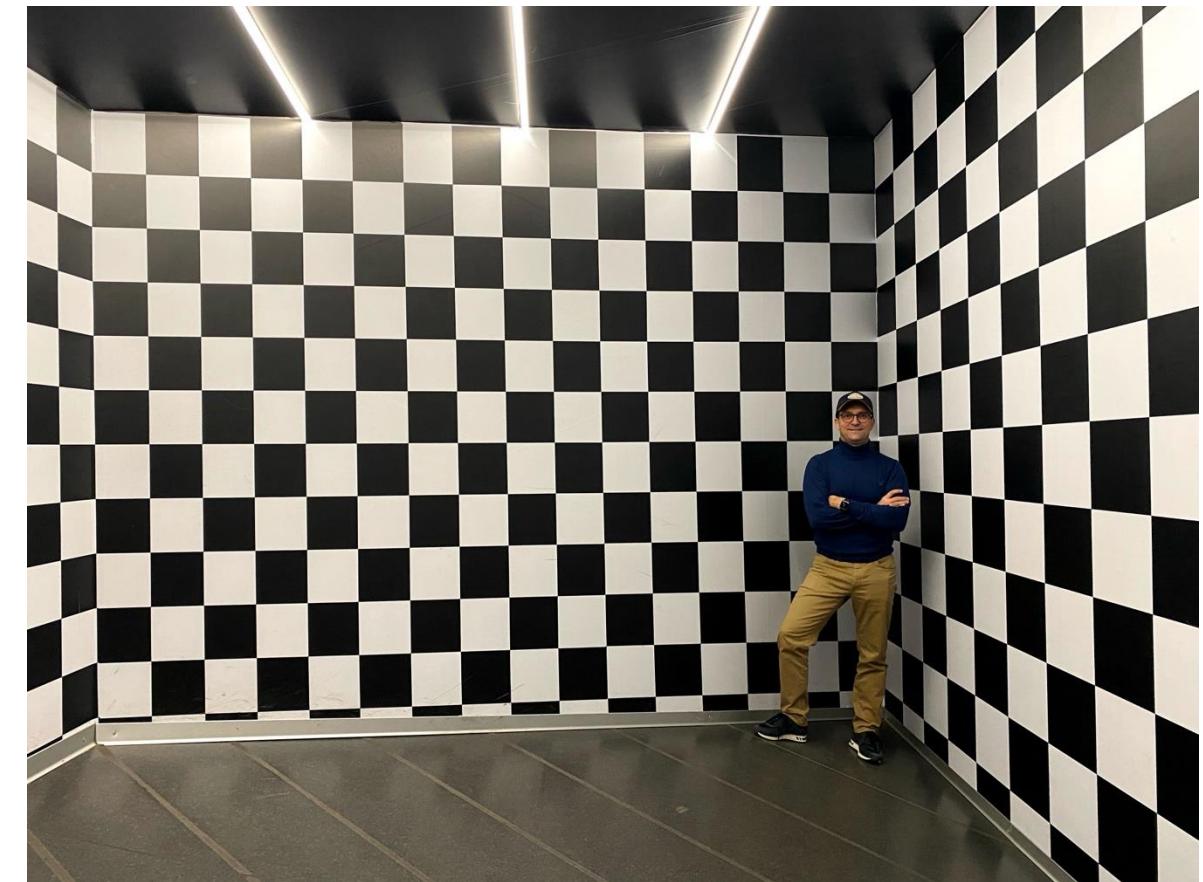
On Hands 2011

Seongbin Im | CC BY-SA 2.0





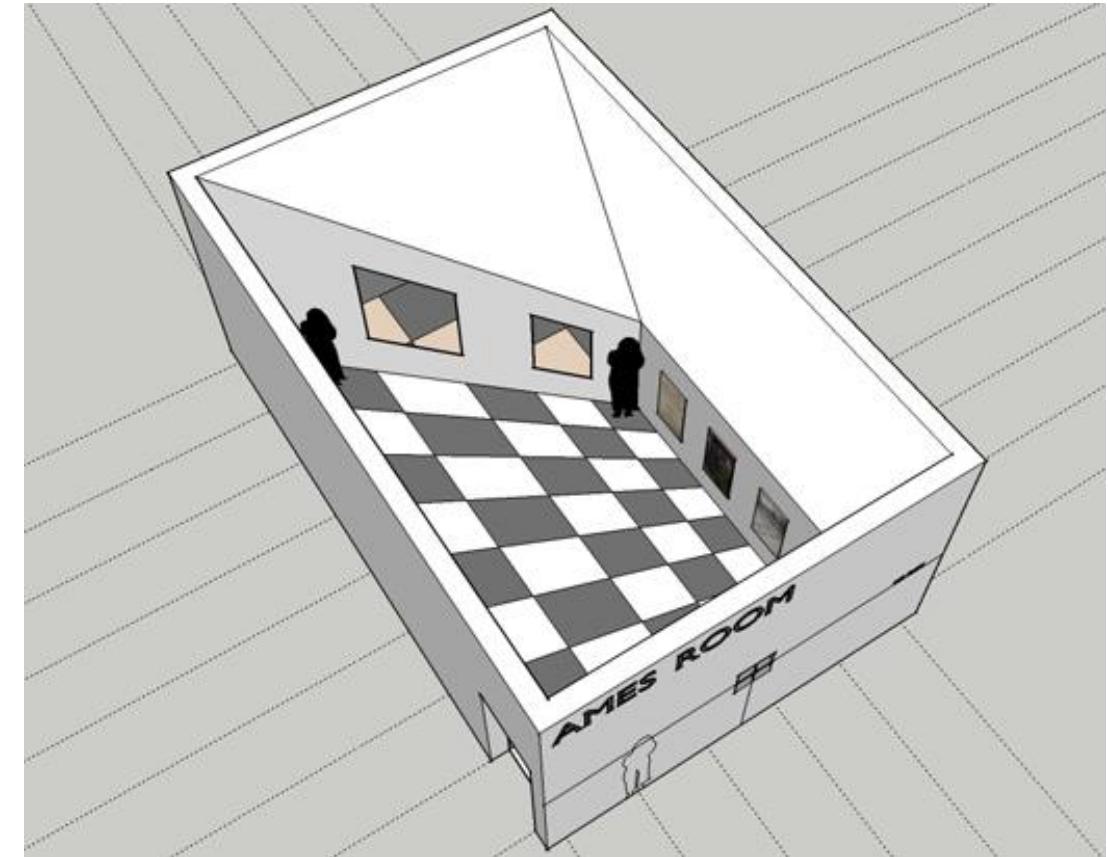
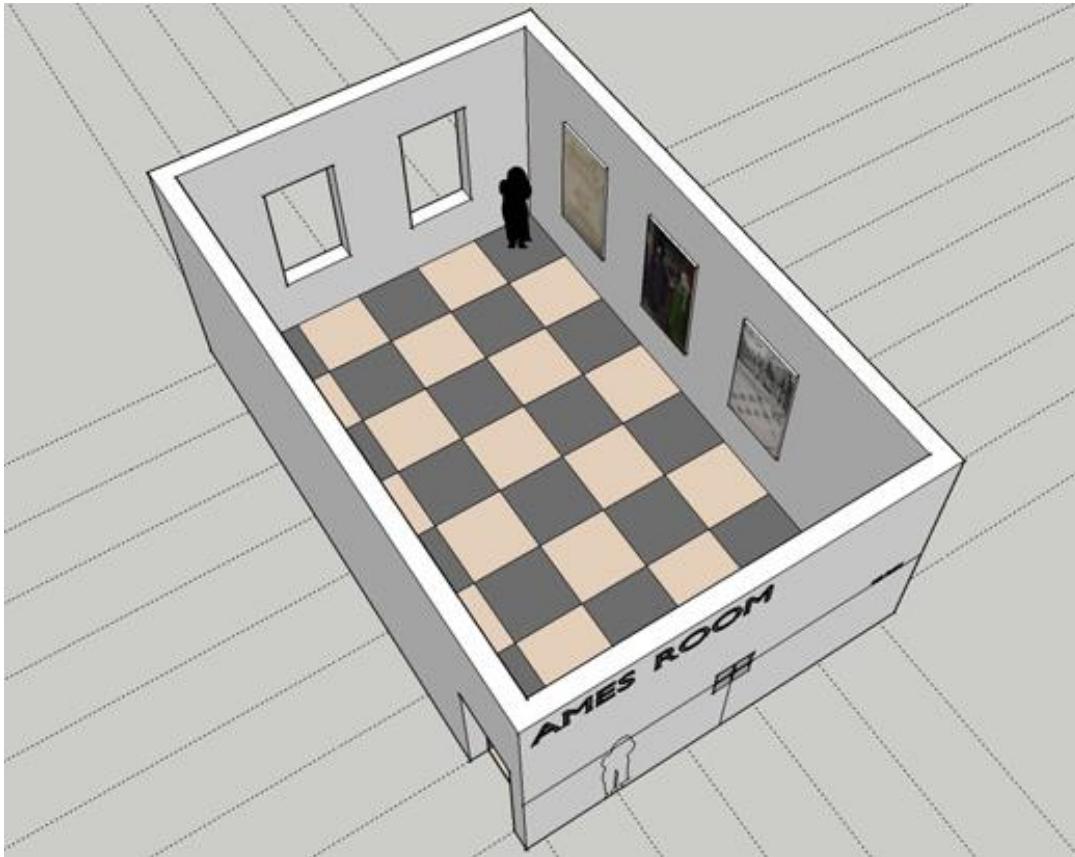
Ames Room







Ames Room



<http://perspectiveresources.blogspot.com/2011/06/how-to-construct-ames-room.html>

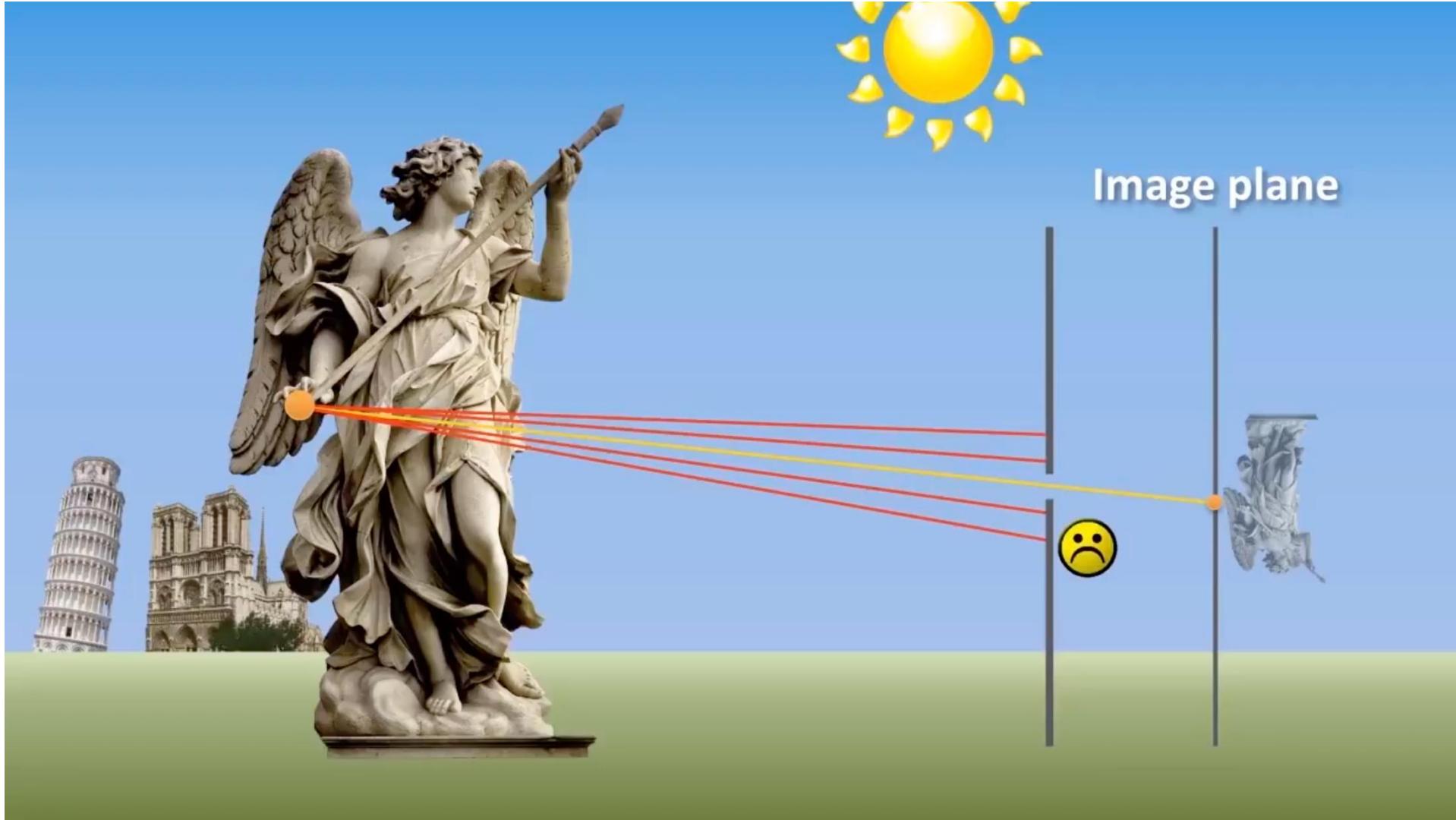


Pinhole Model





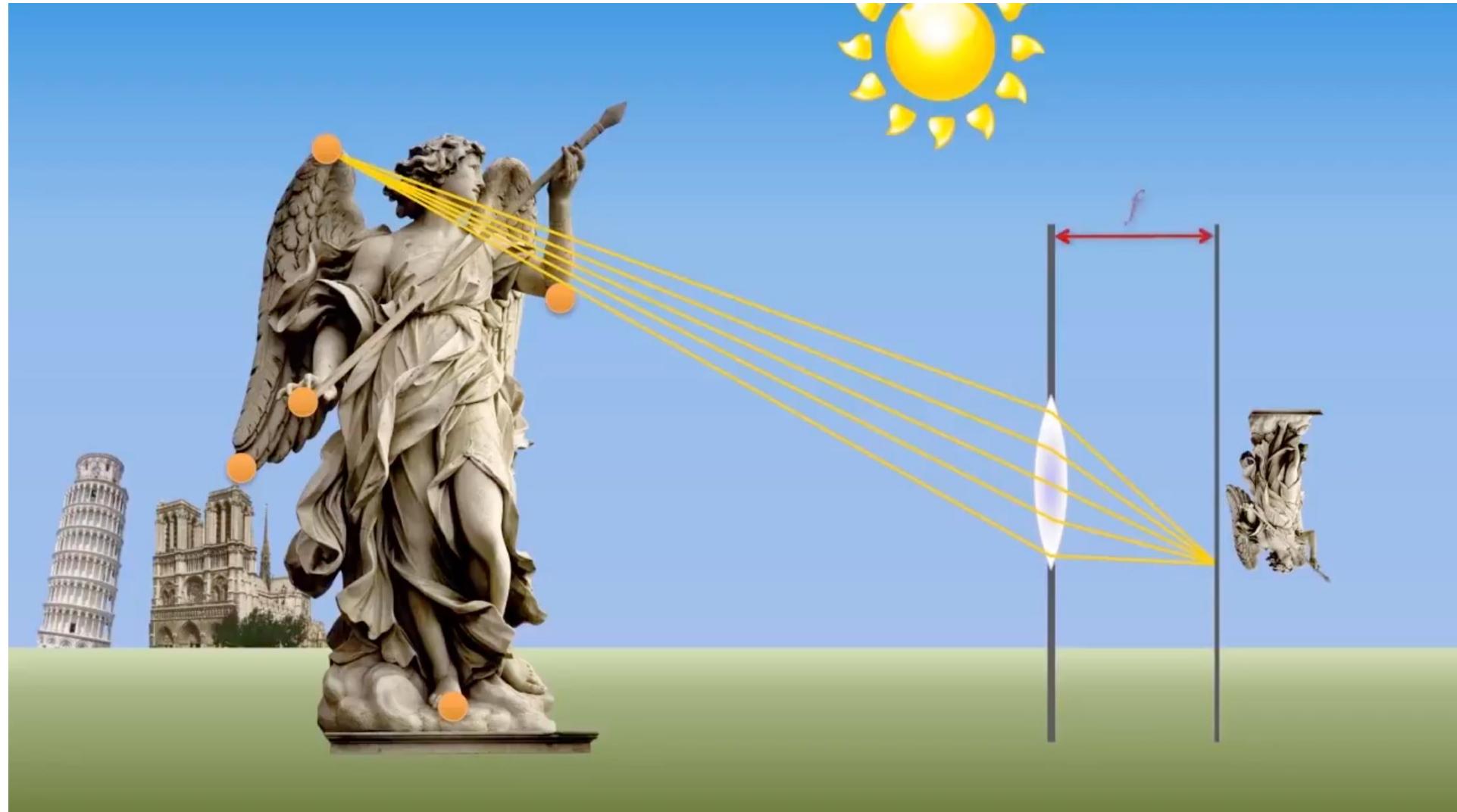
Pinhole Model

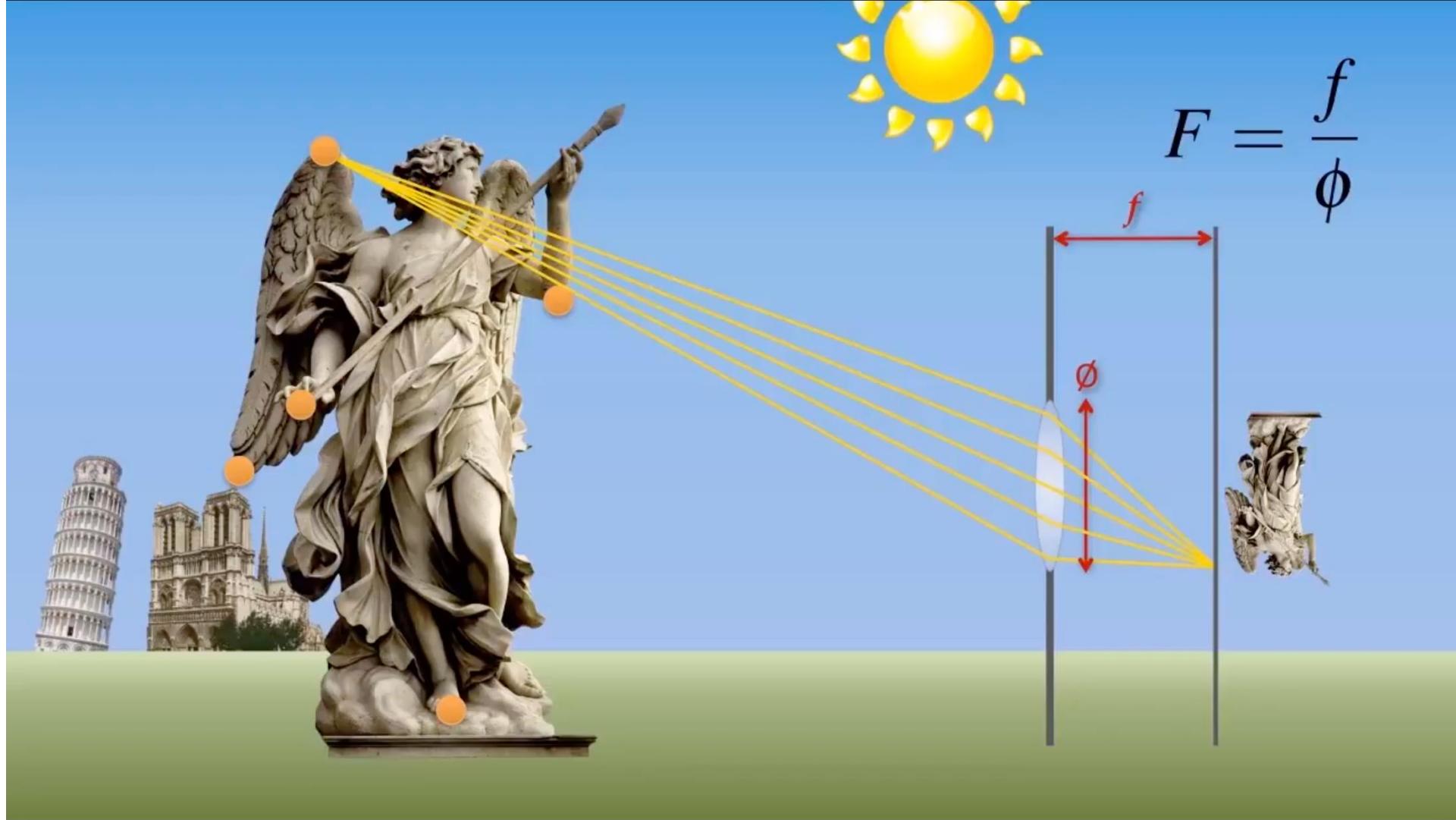


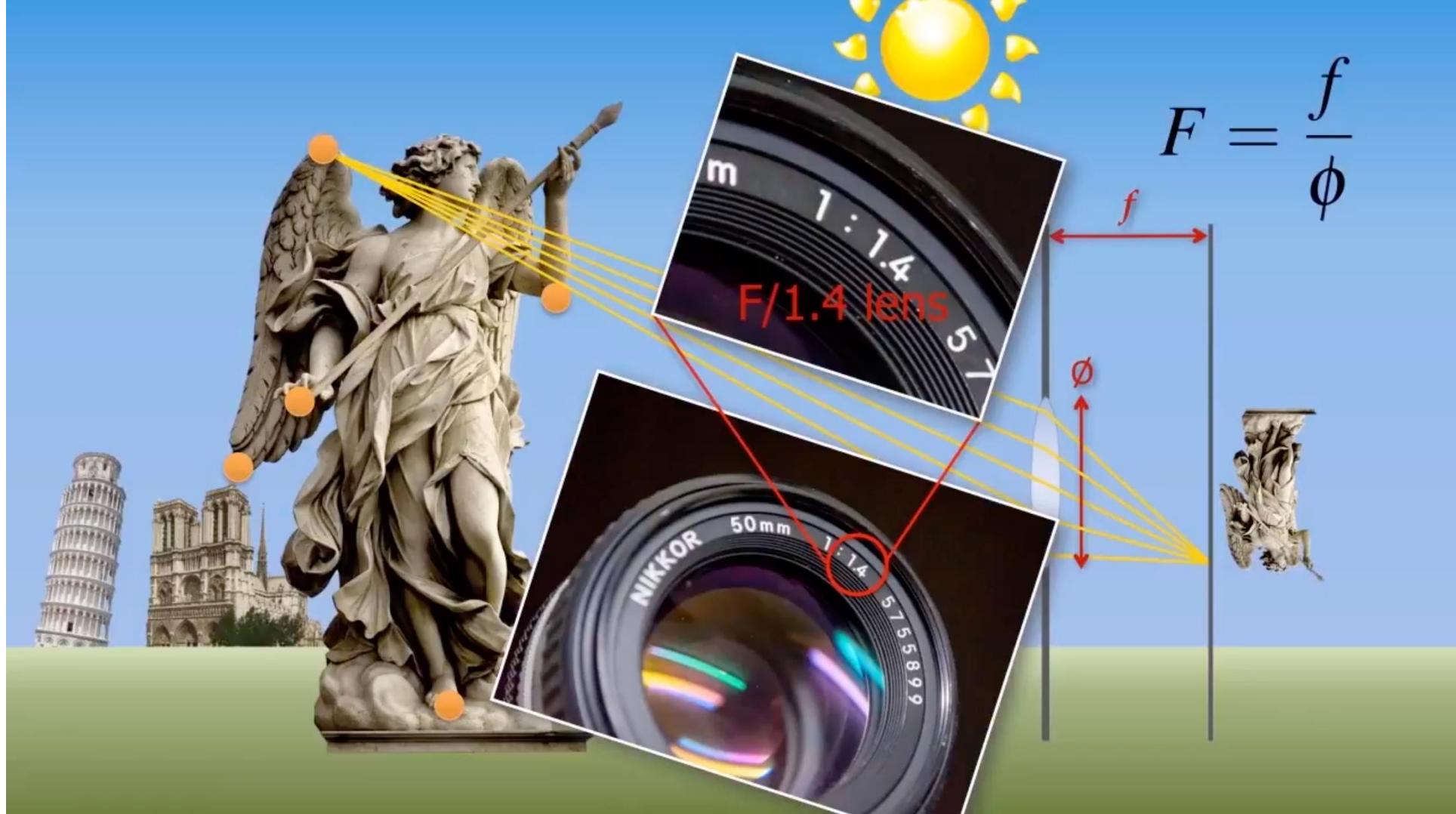


Cameras use LENSES



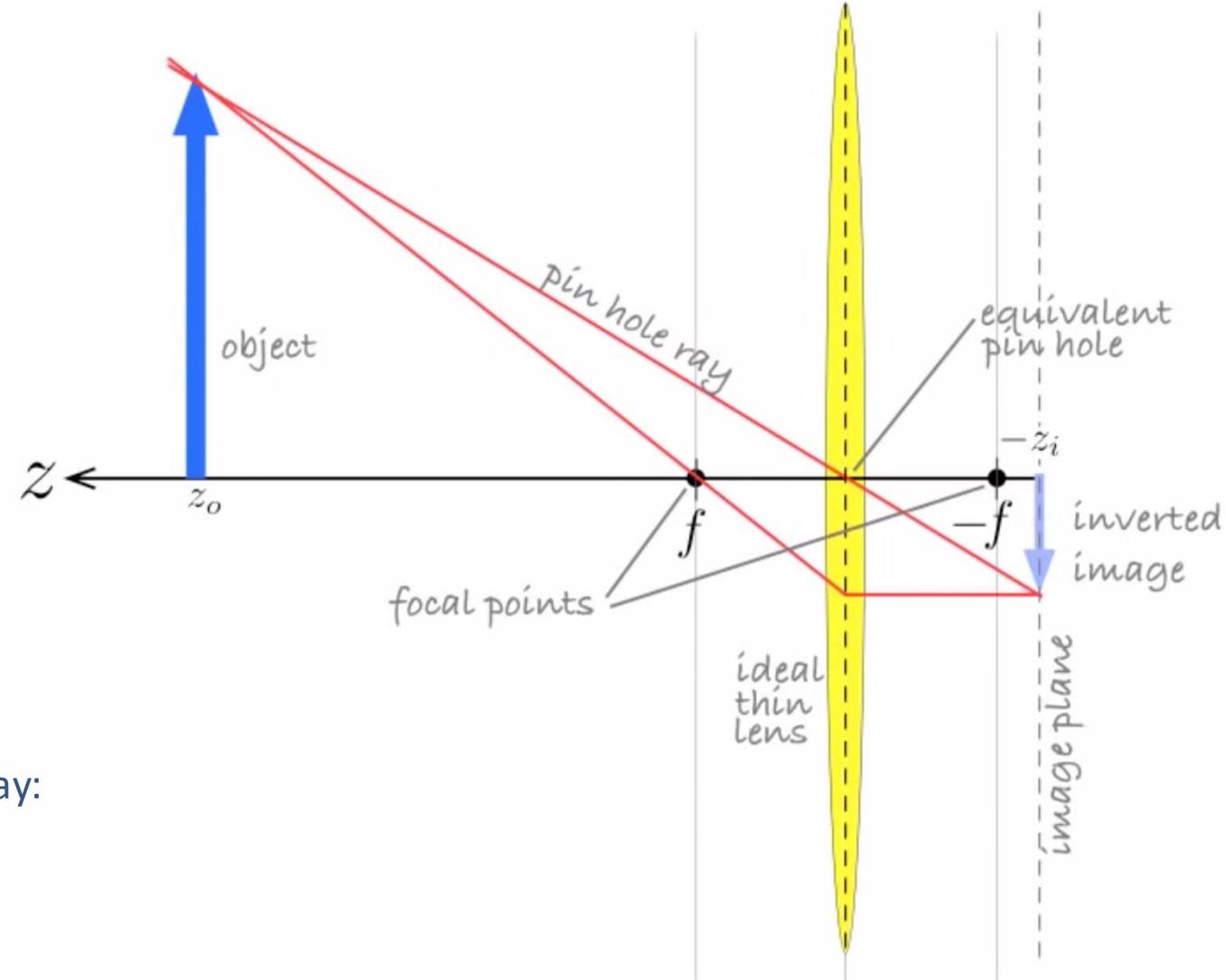








Thin Lens Model



$$\frac{1}{z_0} + \frac{1}{z_i} = \frac{1}{f}$$

Focusing of objects far away:

$$z_0 \rightarrow \infty$$

$$z_i \rightarrow f$$



Click to add a caption

Like

Share



Price to pay when projecting the world in 2D



$$(X, Y, Z) \mapsto (x, y)$$
$$\mathbb{R}^3 \mapsto \mathbb{R}^2$$



Homogeneous Coordinates

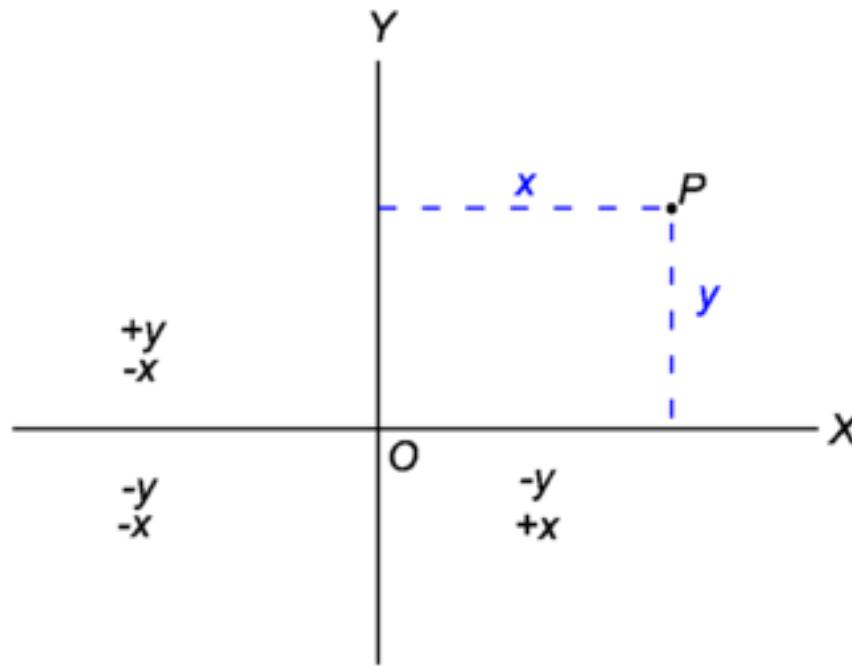
Cartesian Coord. → Homogeneous Coord.

$$\mathbf{p} = (x, y)^T$$

$$\mathbf{p} \in \mathbb{R}^2$$

$$\tilde{\mathbf{p}} = (x, y, 1)^T$$

$$\mathbf{p} \in \mathbb{P}^2$$



Homogeneous Coord. → Cartesian C.

$$\tilde{\mathbf{p}} = (\tilde{x}, \tilde{y}, \tilde{z})^T$$

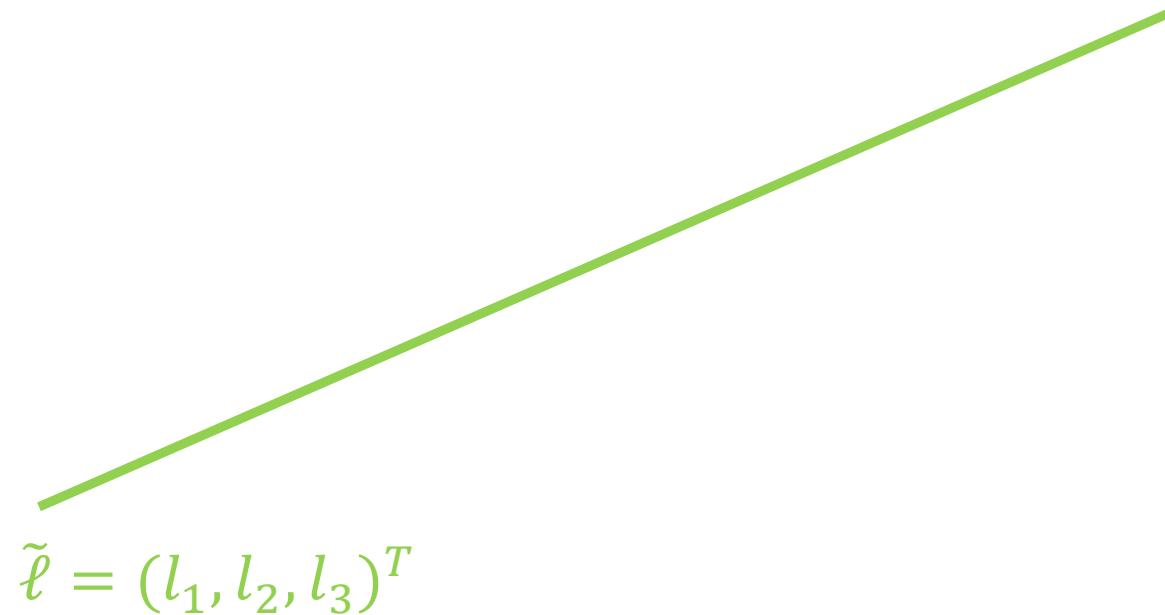
$$x = \frac{\tilde{x}}{\tilde{z}}$$

$$y = \frac{\tilde{y}}{\tilde{z}}$$

$$\mathbf{p} = (x, y)^T$$



Eq. of a Line in homogeneous Coordinates



“Point equation of a line”

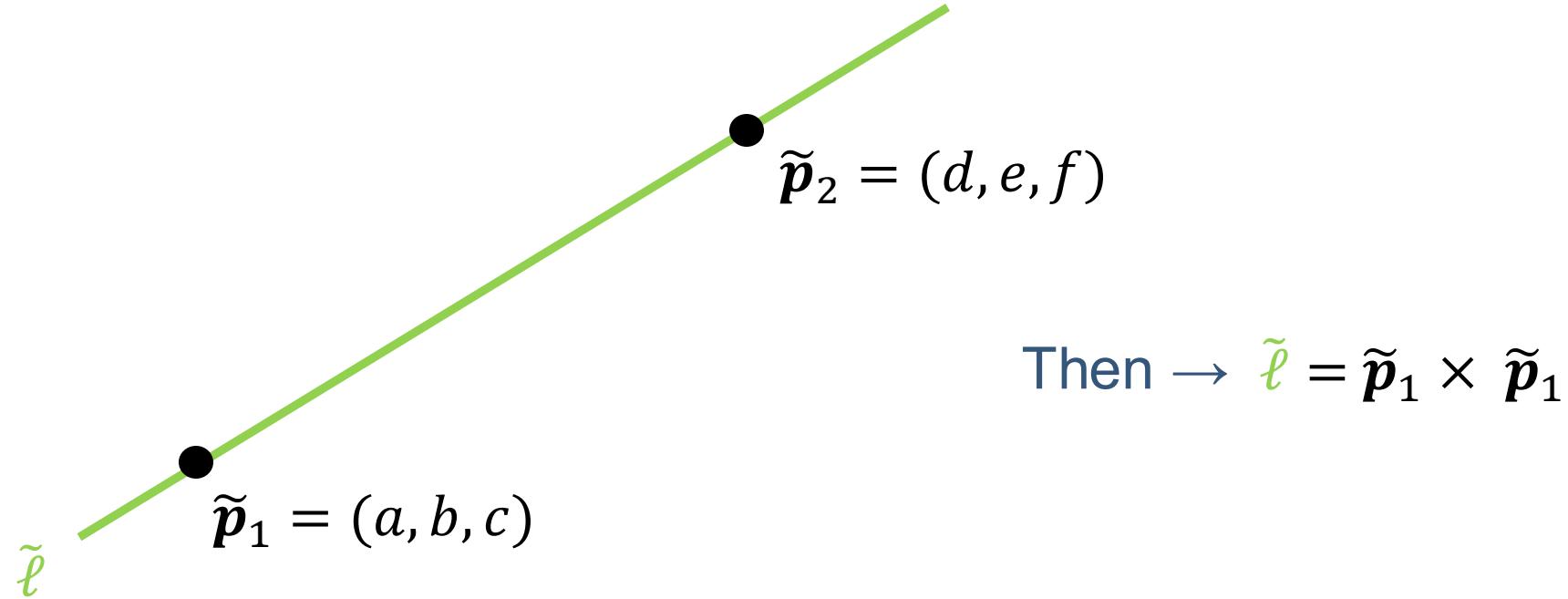
$$\text{Then } \rightarrow \tilde{\ell}^T \tilde{p} = 0$$

$$l_1 \tilde{x} + l_2 \tilde{y} + l_3 \tilde{z} = 0$$

Instead of $\rightarrow y = mx + c$

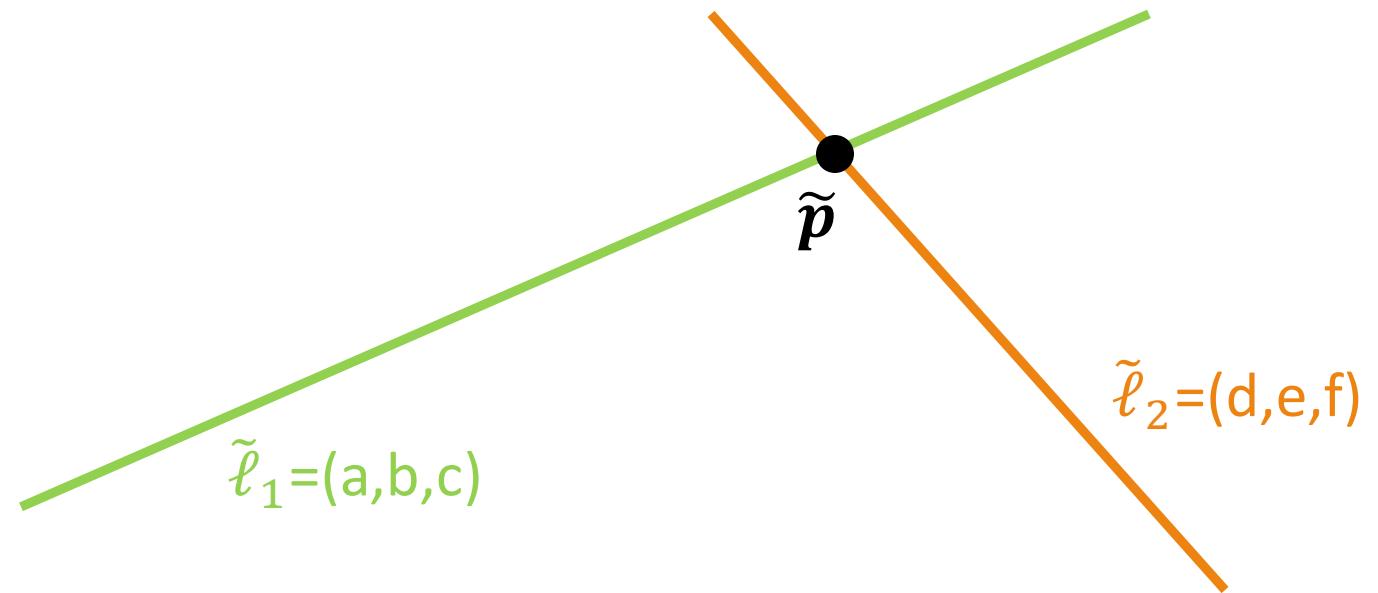


Equació d'una línia que passa per dos punts





Intersection of two Lines

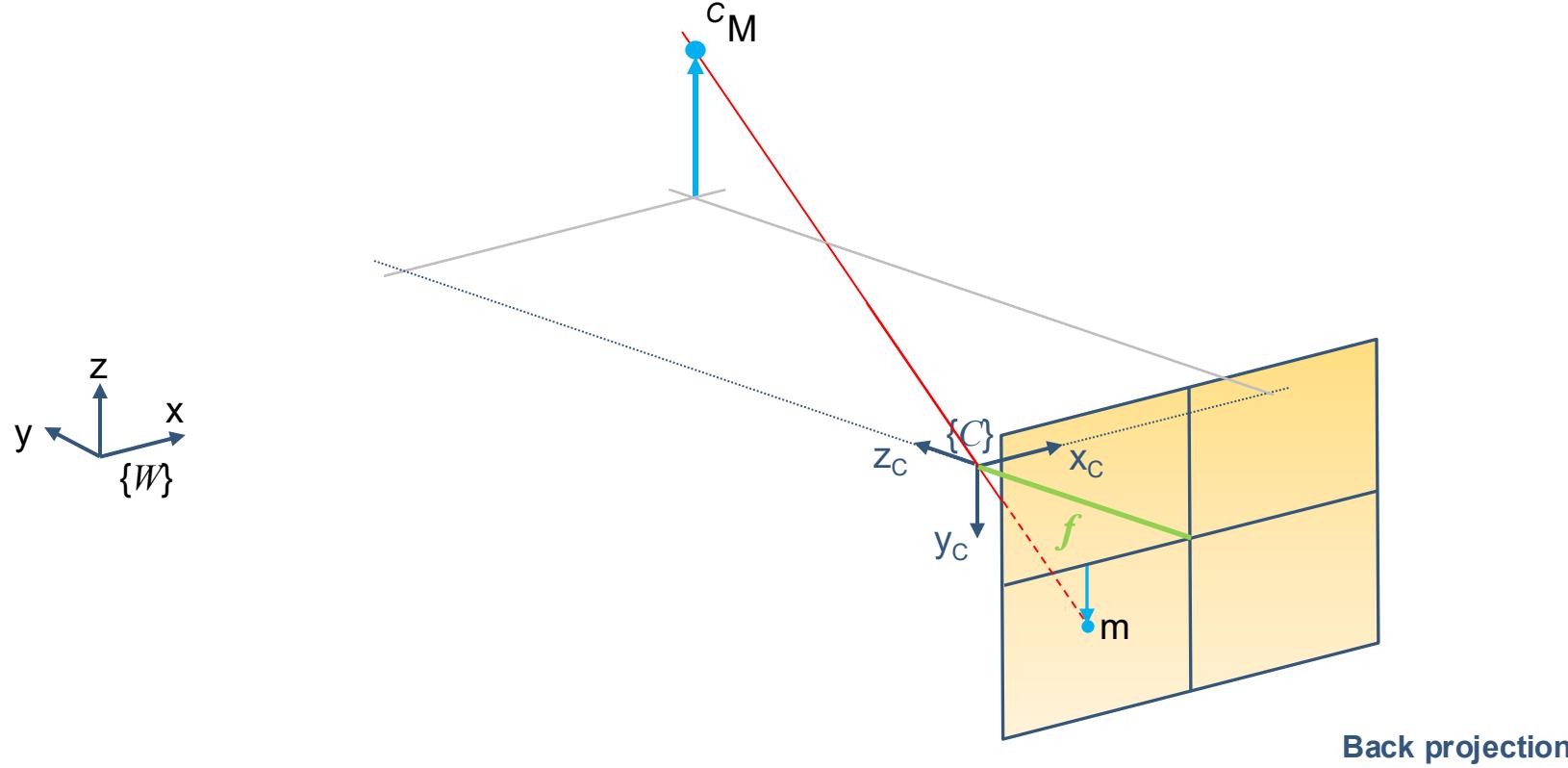


$$\tilde{p} = \tilde{\ell}_1 \times \tilde{\ell}_2$$

“Line equation of a point”

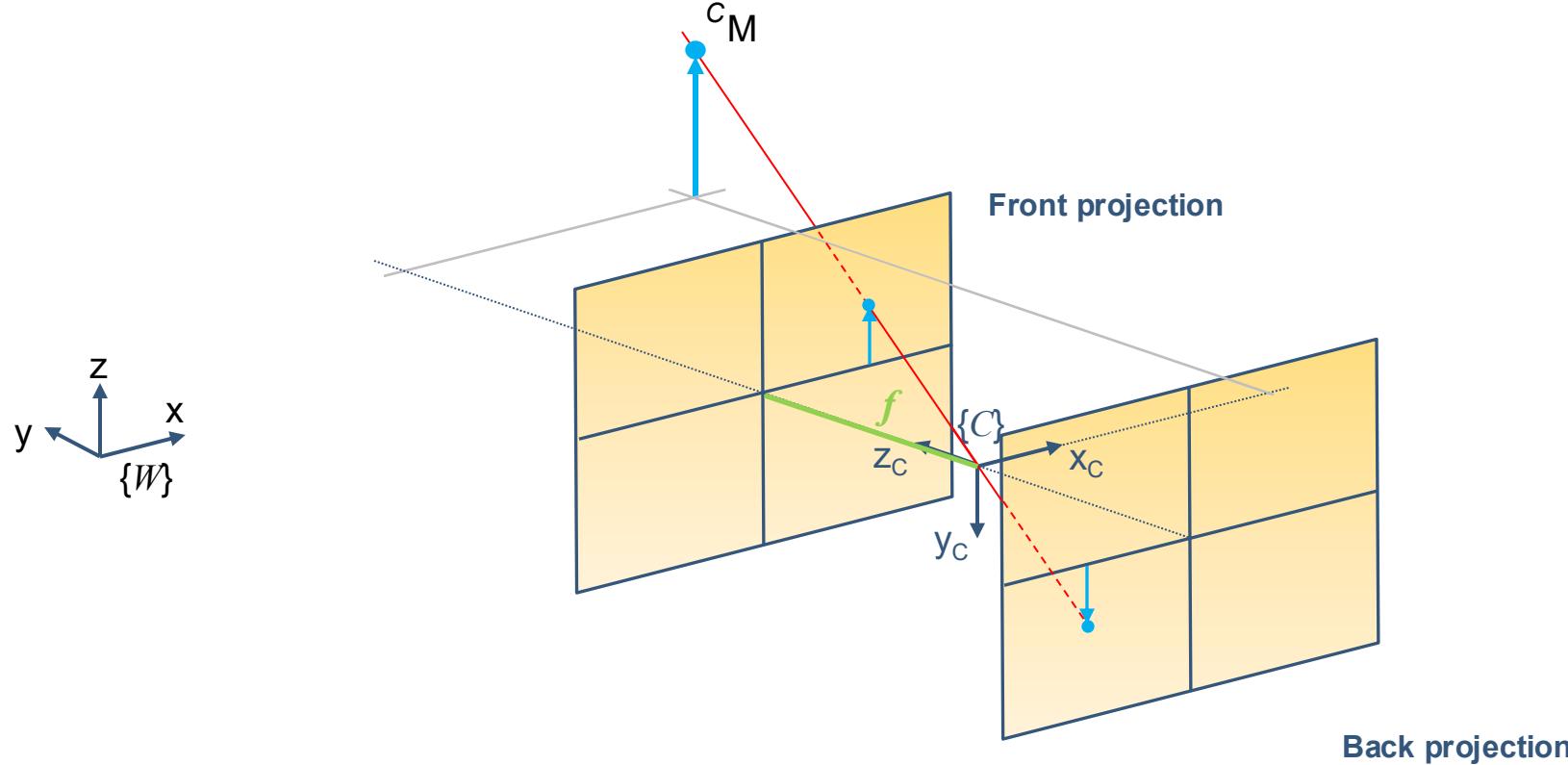


Projecting a point onto the image plane



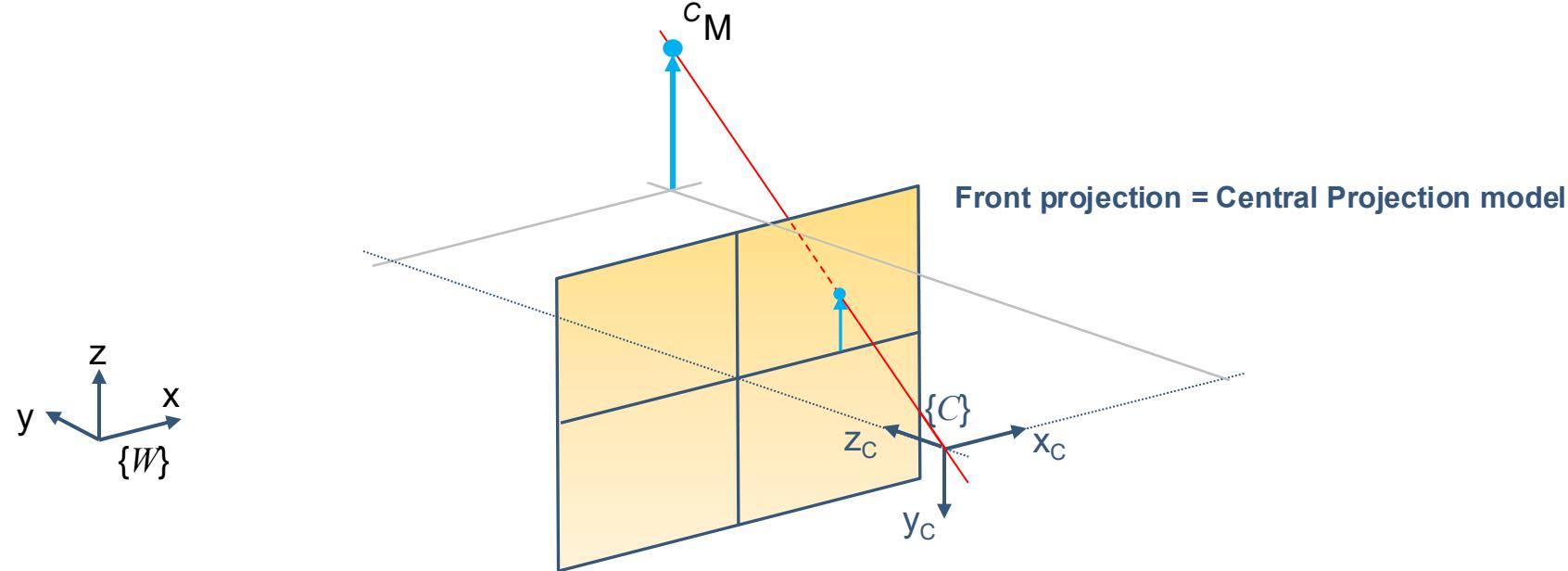


Projecting a point onto the image plane



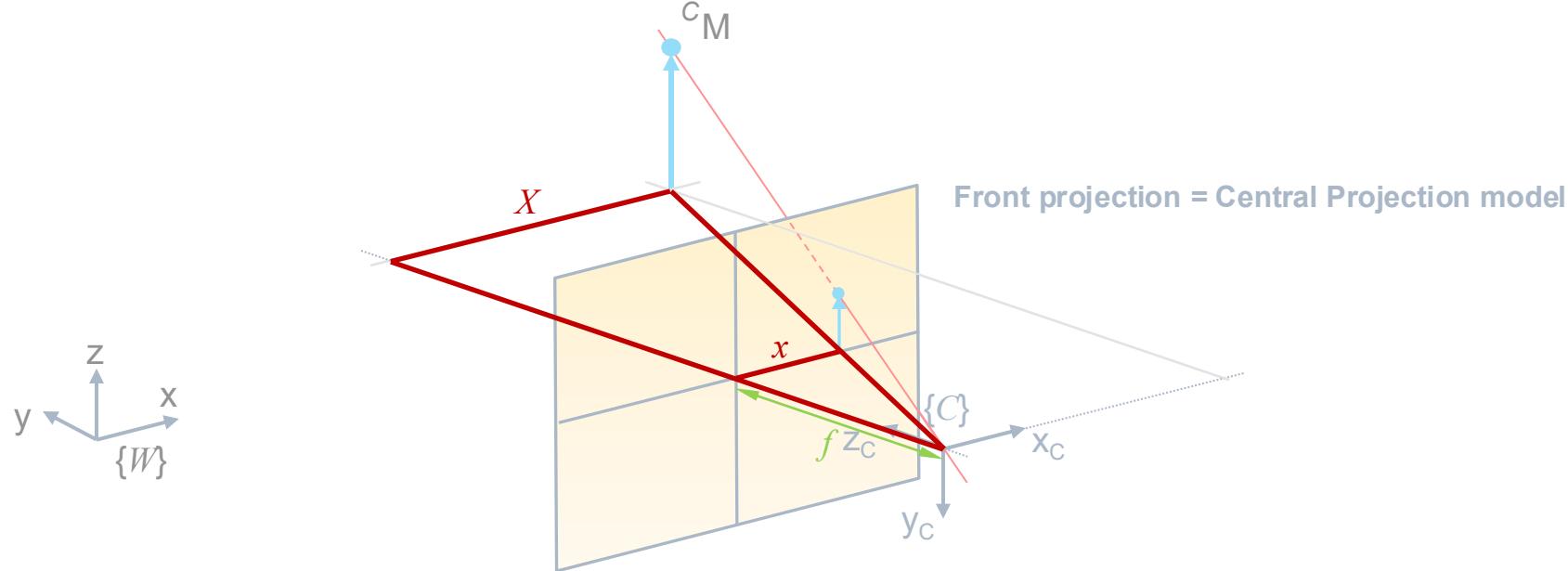


Projecting a point onto the image plane



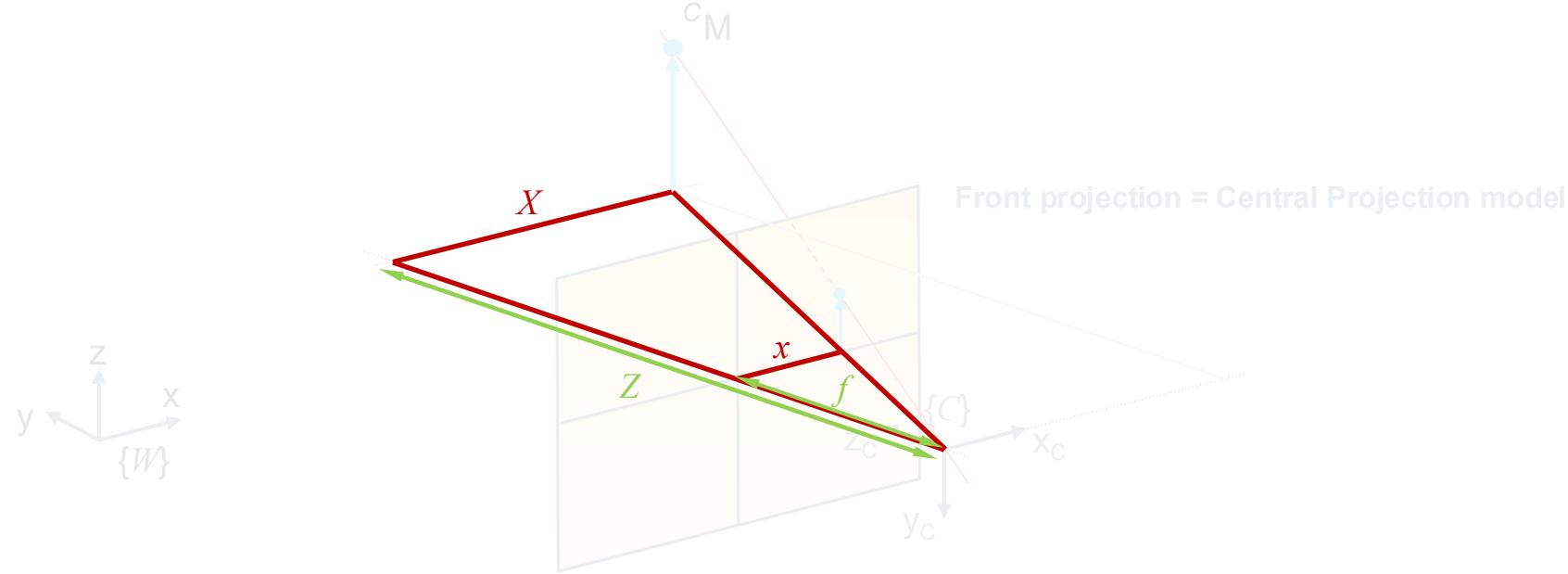


Projecting a point onto the image plane





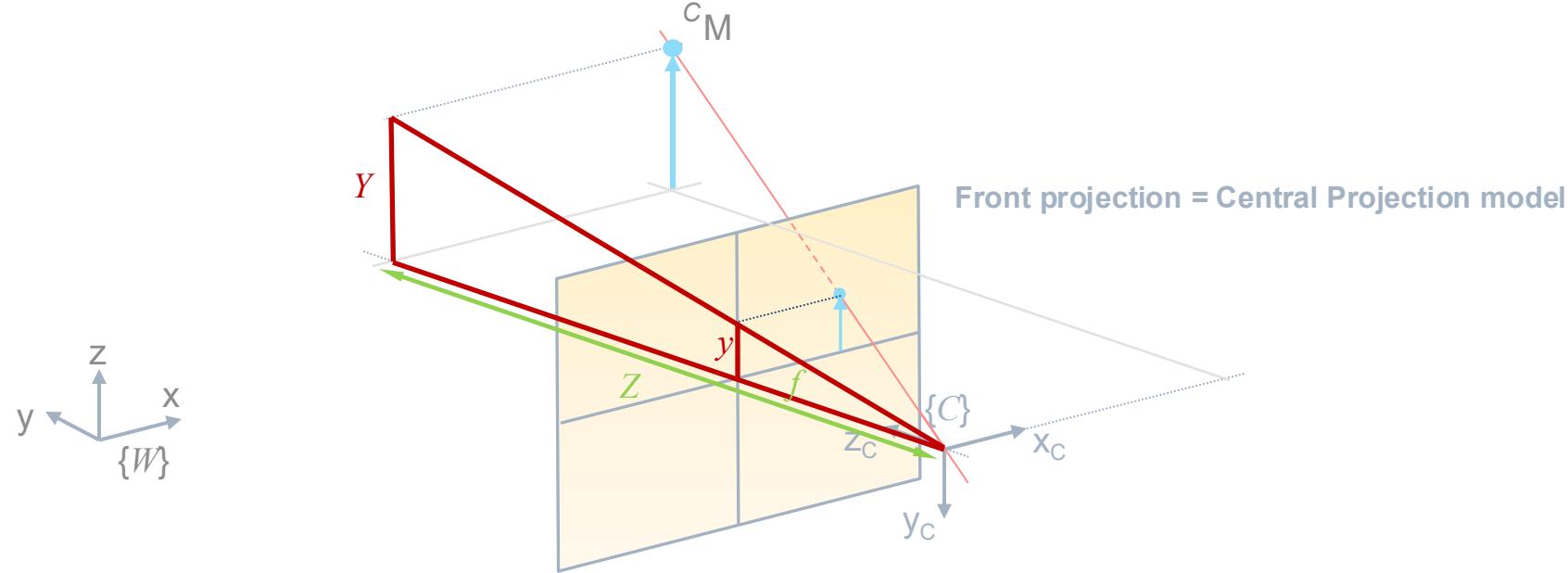
Projecting a point onto the image plane



$$\frac{x}{f} = \frac{X}{Z} \quad \rightarrow \quad x = \frac{f \cdot X}{Z}$$



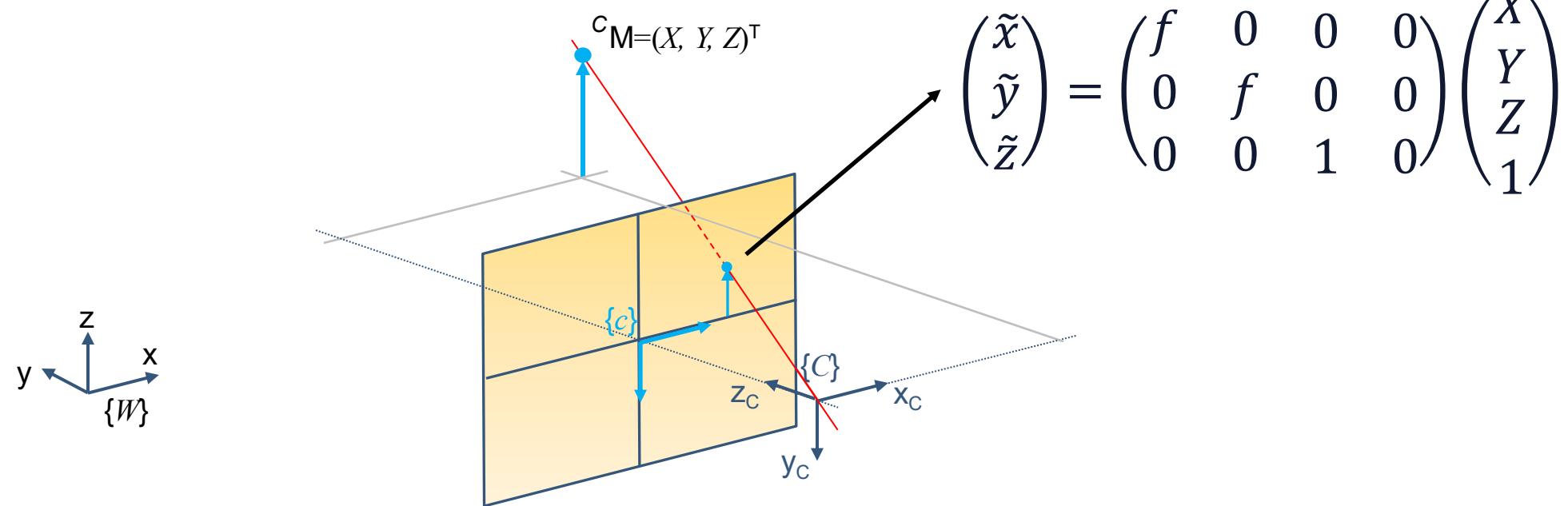
Projecting a point onto the image plane



$$\frac{y}{f} = \frac{Y}{Z} \quad \rightarrow \quad y = \frac{f \cdot Y}{Z}$$



Projecting a point onto the image plane





Perspective Projection

EXERCISE

Given a point M at position (20, 30, 4)
and a camera with focal distance $f = 2$

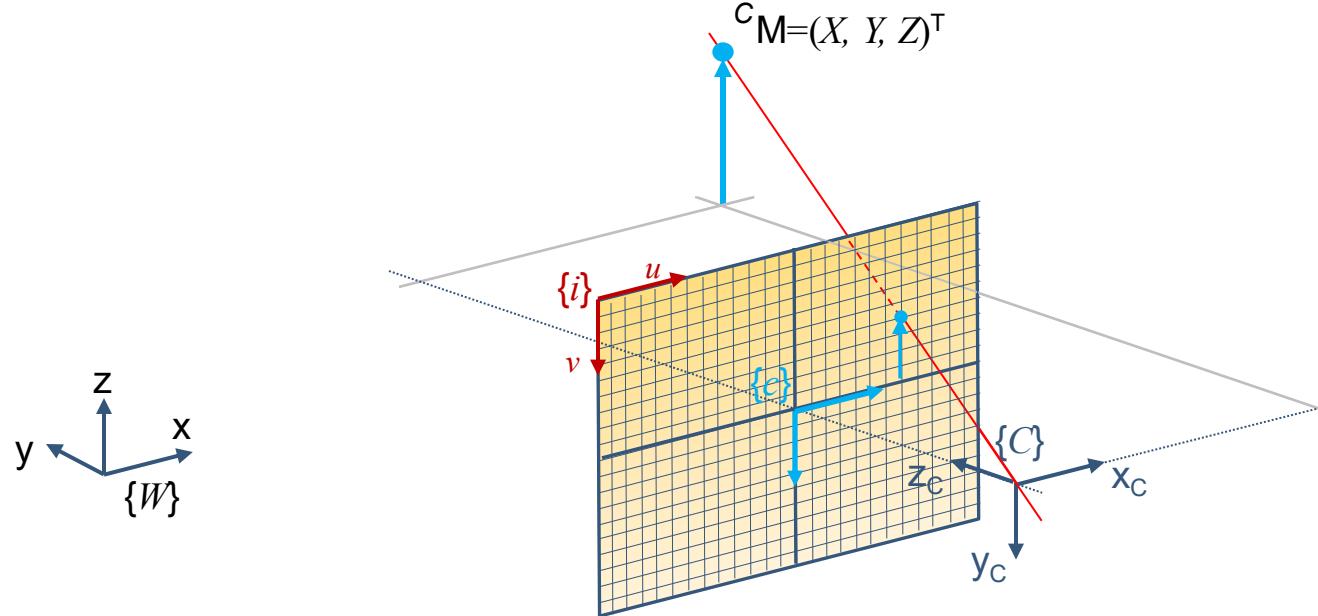
Where does M project onto the image?

(all the values are in mm)

Provide the results in cartesian coordinates



Projecting a point onto the image plane



- We should go from $\{c\}$ to $\{i\}$
- $\{c\}$ is in meters and $\{i\}$ in pixels
- Every pixel has size (ρ_u, ρ_v)

$$i_u = \frac{c_x}{\rho_u} + u_0$$

$$i_v = \frac{c_y}{\rho_v} + v_0$$

$$\begin{pmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{pmatrix} = \begin{pmatrix} \frac{1}{\rho_u} & 0 & u_0 \\ 0 & \frac{1}{\rho_v} & v_0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \tilde{x} \\ \tilde{y} \\ \tilde{z} \end{pmatrix}$$



Iphone 12 Pro vs 13 Pro

Camera	Apple iPhone 13 Pro Max	Apple iPhone 12 Pro Max
Main camera	Sony IMX703, 1.9 µm, 26 mm	Sony IMX603, 1.7 µm, 26 mm
Ultra-wide-angle camera	Sony IMX772, 1µm, 13 mm	Sony IMX372, 1 µm, 13 mm
Telephoto camera	Sony IMX713, 1µm, 77 mm	Sony IMX613, 1 µm, 65 mm
ToF		Sony IMX590
Selfie camera		Sony IMX514, 1µm





Projection Matrix

- Complete Model :

$$\begin{pmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{pmatrix} = \boxed{\begin{pmatrix} \frac{1}{\rho_u} & 0 & u_0 \\ 0 & \frac{1}{\rho_v} & v_0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}} \begin{pmatrix} \mathbf{R} & \mathbf{t} \end{pmatrix}^{-1} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}$$

Intrinsic Parameters: K

$$K = \begin{bmatrix} \alpha_x & \gamma & u_0 & 0 \\ 0 & \alpha_y & v_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

- Projection Matrix:

$$\begin{pmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{pmatrix} = \begin{pmatrix} P_{11} & P_{12} & P_{13} & P_{14} \\ P_{21} & P_{22} & P_{23} & P_{24} \\ P_{31} & P_{32} & P_{33} & P_{34} \end{pmatrix} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}$$



EXERCISE 1

Given a point M at 3D position (-2 1.3 5) (in meters)

Camera values:

$$f=0.012 \text{ m}; \text{ principal point } (512,512); \rho_u = \rho_v = 10\mu\text{m}$$

$$\begin{pmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{pmatrix} = \begin{pmatrix} \frac{1}{\rho_u} & 0 & u_0 \\ 0 & \frac{1}{\rho_v} & v_0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} \mathbf{R} & \mathbf{t} \end{pmatrix}^{-1} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}$$

The camera is placed at $\{W\}$.

Where does M project?



EXERCISE 2

Given a point M at 3D position (5150 -150 6600) (in mm)

Camera values: main camera iphone 13.

$f = 26 \text{ mm}$; principal point (2000, 1500); $\rho_u = \rho_v = 1.9 \mu\text{m}$

$$\begin{pmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{pmatrix} = \begin{pmatrix} \frac{1}{\rho_u} & 0 & u_0 \\ 0 & \frac{1}{\rho_v} & v_0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} \mathbf{R} & \mathbf{t} \end{pmatrix}^{-1} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}$$

The camera is placed at $\{W\}$ + offset 5.5 meters in the X direction.

Where does M project in the image in pixels?



Projection Matrix (P)

- If we multiply to the right by an arbitrary value, the equality is maintained (homogeneous coordinates!)

$$\begin{pmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{pmatrix} = \lambda \begin{pmatrix} P_{11} & P_{12} & P_{13} & P_{14} \\ P_{21} & P_{22} & P_{23} & P_{24} \\ P_{31} & P_{32} & P_{33} & P_{34} \end{pmatrix} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}$$

- Therefore, we can scale the matrix so that $P_{34} = 1$, and only 11 parameters need to be estimated

$$\begin{pmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{pmatrix} = \begin{pmatrix} P_{11} & P_{12} & P_{13} & P_{14} \\ P_{21} & P_{22} & P_{23} & P_{24} \\ P_{31} & P_{32} & P_{33} & 1 \end{pmatrix} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}$$



Camera Calibration

- What is Camera Calibration?
 - Finding the parameters, internal to the camera, that affect the imaging process OR finding the projection matrix that encodes those parameters
 - Intrinsic camera parameters:
 - Focal length
 - Principal point
 - Scaling factors for row pixels
 - Lens distortion
 - Extrinsic camera parameters:
 - 3D position
 - 3D orientation



Camera Calibration

Hall Method to compute P

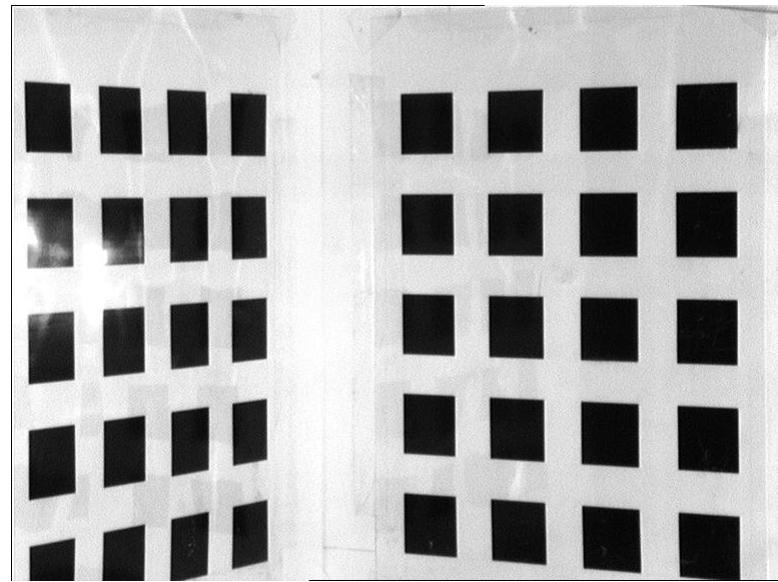
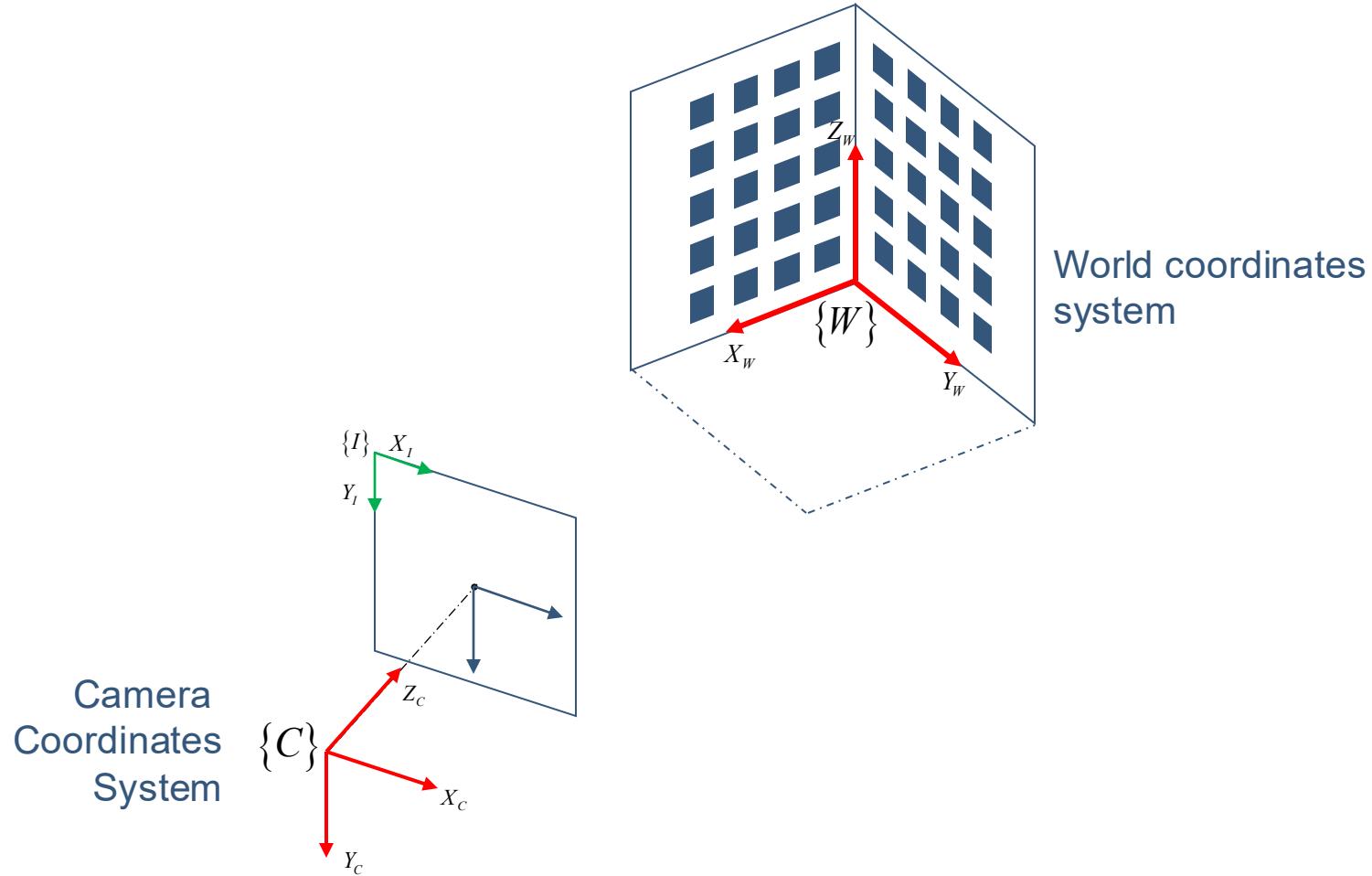


Image of the calibration pattern



Computing the projection Matrix

- Start with P

$$\begin{pmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{pmatrix} = \begin{pmatrix} P_{11} & P_{12} & P_{13} & P_{14} \\ P_{21} & P_{22} & P_{23} & P_{24} \\ P_{31} & P_{32} & P_{33} & 1 \end{pmatrix} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}$$

- Change of nomenclature:

$$\begin{pmatrix} s^I X_u \\ s^I Y_u \\ s \end{pmatrix} = \begin{pmatrix} A_{11} & A_{12} & A_{13} & A_{14} \\ A_{21} & A_{22} & A_{23} & A_{24} \\ A_{31} & A_{32} & A_{33} & 1 \end{pmatrix} \begin{pmatrix} {}^W X_w \\ {}^W Y_w \\ {}^W Z_w \\ 1 \end{pmatrix}$$



Computing the projection Matrix

$$\begin{pmatrix} s^I X_u \\ s^I Y_u \\ s \end{pmatrix} = \begin{pmatrix} A_{11} & A_{12} & A_{13} & A_{14} \\ A_{21} & A_{22} & A_{23} & A_{24} \\ A_{31} & A_{32} & A_{33} & 1 \end{pmatrix} \begin{pmatrix} {}^W X_w \\ {}^W Y_w \\ {}^W Z_w \\ 1 \end{pmatrix}$$

$${}^I X_u = \frac{A_{11} {}^W X_w + A_{12} {}^W Y_w + A_{13} {}^W Z_w + A_{14}}{A_{31} {}^W X_w + A_{32} {}^W Y_w + A_{33} {}^W Z_w + 1}$$

$${}^I Y_u = \frac{A_{21} {}^W X_w + A_{22} {}^W Y_w + A_{23} {}^W Z_w + A_{24}}{A_{31} {}^W X_w + A_{32} {}^W Y_w + A_{33} {}^W Z_w + 1}$$

$$A_{11} {}^W X_w - A_{31} {}^I X_u {}^W X_w + A_{12} {}^W Y_w - A_{32} {}^I X_u {}^W Y_w + A_{13} {}^W Z_w - A_{33} {}^I X_u {}^W Z_w + A_{14} = {}^I X_u$$

$$A_{21} {}^W X_w - A_{31} {}^I Y_u {}^W X_w + A_{22} {}^W Y_w - A_{32} {}^I Y_u {}^W Y_w + A_{23} {}^W Z_w - A_{33} {}^I Y_u {}^W Z_w + A_{24} = {}^I Y_u$$



Computing the projection Matrix

- For every 3D-2D point, we obtain 2 equations:

$$\begin{aligned} A_{11} {}^W X_w - A_{31} {}^I X_u {}^W X_w + A_{12} {}^W Y_w - A_{32} {}^I X_u {}^W Y_w + A_{13} {}^W Z_w - A_{33} {}^I X_u {}^W Z_w + A_{14} &= {}^I X_u \\ A_{21} {}^W X_w - A_{31} {}^I Y_u {}^W X_w + A_{22} {}^W Y_w - A_{32} {}^I Y_u {}^W Y_w + A_{23} {}^W Z_w - A_{33} {}^I Y_u {}^W Z_w + A_{24} &= {}^I Y_u \end{aligned}$$

- Place all the unknowns in a column vector:

$$A = (A_{11} \quad A_{12} \quad A_{13} \quad A_{14} \quad A_{21} \quad A_{22} \quad A_{23} \quad A_{24} \quad A_{31} \quad A_{32} \quad A_{33})^T$$

- And we obtain a linear system of equations $QA = B$
- REMEMBER: we have 11 unknowns, therefore we need at least six 3D-2D pairs, which will generate two rows on matrix Q:

$$\begin{aligned} Q_{2i-1} &= \left({}^W X_{wi} \quad {}^W Y_{wi} \quad {}^W Z_{wi} \quad 1 \quad 0 \quad 0 \quad 0 \quad 0 \quad -{}^I X_{ui} {}^W X_{wi} \quad -{}^I X_{ui} {}^W Y_{wi} \quad -{}^I X_{ui} {}^W Z_{wi} \right) \\ Q_{2i} &= \left(0 \quad 0 \quad 0 \quad 0 \quad {}^W X_{wi} \quad {}^W Y_{wi} \quad {}^W Z_{wi} \quad 1 \quad -{}^I Y_{ui} {}^W X_{wi} \quad -{}^I Y_{ui} {}^W Y_{wi} \quad -{}^I Y_{ui} {}^W Z_{wi} \right) \end{aligned}$$



Computing the projection Matrix

- REMEMBER: we have 11 unknowns, therefore we need at least six 3D-2D pairs, which will generate two rows on matrix Q:

$$Q_{2i-1} = \begin{pmatrix} {}^W X_{wi} & {}^W Y_{wi} & {}^W Z_{wi} & 1 & 0 & 0 & 0 & 0 & -{}^I X_{ui} {}^W X_{wi} & -{}^I X_{ui} {}^W Y_{wi} & -{}^I X_{ui} {}^W Z_{wi} \end{pmatrix}$$

$$Q_{2i} = \begin{pmatrix} 0 & 0 & 0 & 0 & {}^W X_{wi} & {}^W Y_{wi} & {}^W Z_{wi} & 1 & -{}^I Y_{ui} {}^W X_{wi} & -{}^I Y_{ui} {}^W Y_{wi} & -{}^I Y_{ui} {}^W Z_{wi} \end{pmatrix}$$

$$B_{2i-1} = \begin{pmatrix} {}^I X_{ui} \end{pmatrix}$$

$$B_{2i} = \begin{pmatrix} {}^I Y_{ui} \end{pmatrix}$$

- If we have more than 6 pairs, we can do least squares:

$$A = Q^{-1}B \quad \longrightarrow$$

$$A = (Q^t Q)^{-1} Q^t B$$



Extracting K and R from Projection Matrix P

We have the projection matrix P and need to extract:

■ **K** - Intrinsic matrix (upper triangular) ■ **R** - Rotation matrix (orthogonal)

$$\begin{array}{|c|c|c|c|} \hline P_{11} & P_{12} & P_{13} & P_{14} \\ \hline P_{21} & P_{22} & P_{23} & P_{24} \\ \hline P_{31} & P_{32} & P_{33} & P_{34} \\ \hline \end{array} = \begin{array}{|c|c|c|} \hline K_{11} & K_{12} & K_{13} \\ \hline 0 & K_{22} & K_{23} \\ \hline 0 & 0 & 1 \\ \hline \end{array} \times \begin{array}{|c|c|c|} \hline R_{11} & R_{12} & R_{13} \\ \hline R_{21} & R_{22} & R_{23} \\ \hline R_{31} & R_{32} & R_{33} \\ \hline \end{array}$$



Extracting K and R from Projection Matrix P

$$\begin{array}{|c|c|c|c|} \hline P_{11} & P_{12} & P_{13} & P_{14} \\ \hline P_{21} & P_{22} & P_{23} & P_{24} \\ \hline P_{31} & P_{32} & P_{33} & P_{34} \\ \hline \end{array} = \begin{array}{|c|c|c|} \hline K_{11} & K_{12} & K_{13} \\ \hline 0 & K_{22} & K_{23} \\ \hline 0 & 0 & 1 \\ \hline \end{array} \times \begin{array}{|c|c|c|} \hline R_{11} & R_{12} & R_{13} \\ \hline R_{21} & R_{22} & R_{23} \\ \hline R_{31} & R_{32} & R_{33} \\ \hline \end{array}$$

K - Intrinsic Matrix

- Upper triangular
- Positive diagonal elements
- $K(3,3) = 1$ (normalized)
- Contains focal lengths, principal point, skew

R - Rotation Matrix

- Orthogonal: $R \cdot R^T = I$
- $\det(R) = +1$
- Columns are unit vectors
- Represents camera orientation



Summary

- Why modelling a camera is important. Applications.
- Intuition about the *pinhole* model
- Description of the projection equations 3D-2D
- Camera calibration procedure
- Point Projection equation using 4×3 matrix P