

CSC I6716
Fall 2010



Topic 1 of Part II Camera Models

■ Closely Related Disciplines

- Image Processing – images to images
- Computer Graphics – models to images
- Computer Vision – images to models
- Photogrammetry – obtaining accurate measurements from images

■ What is 3-D (three dimensional) Vision?

- Motivation: making computers see (the 3D world as humans do)
- Computer Vision: 2D images to 3D structure
- Applications : robotics / VR /Image-based rendering/ 3D video

■ Lectures on 3-D Vision Fundamentals

- Camera Geometric Models (3 lectures)
- Camera Calibration (3 lectures)
- Stereo (4 lectures)
- Motion (4 lectures)

- Geometric Projection of a Camera
 - Pinhole camera model
 - Perspective projection
 - Weak-Perspective Projection
- Camera Parameters
 - Intrinsic Parameters: define mapping from 3D to 2D
 - Extrinsic parameters: define viewpoint and viewing direction
 - Basic Vector and Matrix Operations, Rotation
- Camera Models Revisited
 - Linear Version of the Projection Transformation Equation
 - Perspective Camera Model
 - Weak-Perspective Camera Model
 - Affine Camera Model
 - Camera Model for Planes
- Summary

■ Camera **Geometric** Models

- Knowledge about 2D and 3D geometric transformations
- Linear algebra (vector, matrix)
- **This lecture is only about geometry**

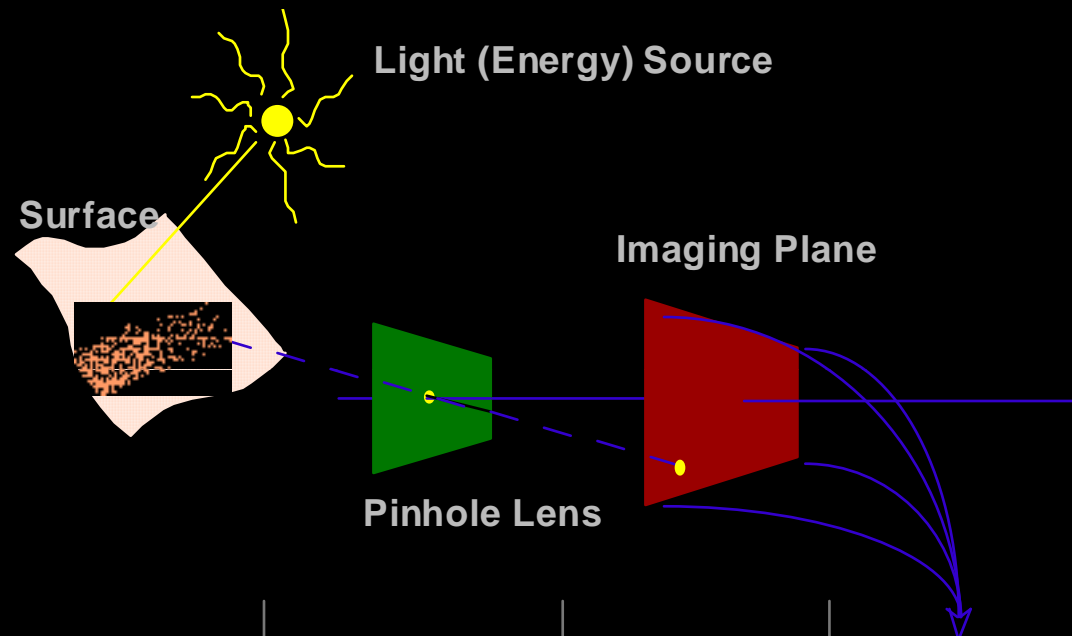
■ Goal

Build up relation between 2D images and 3D scenes

-3D Graphics (rendering): from 3D to 2D

-3D Vision (stereo and motion): from 2D to 3D

-Calibration: Determining the parameters for mapping



World

Optics

Sensor

Signal

B&W Film

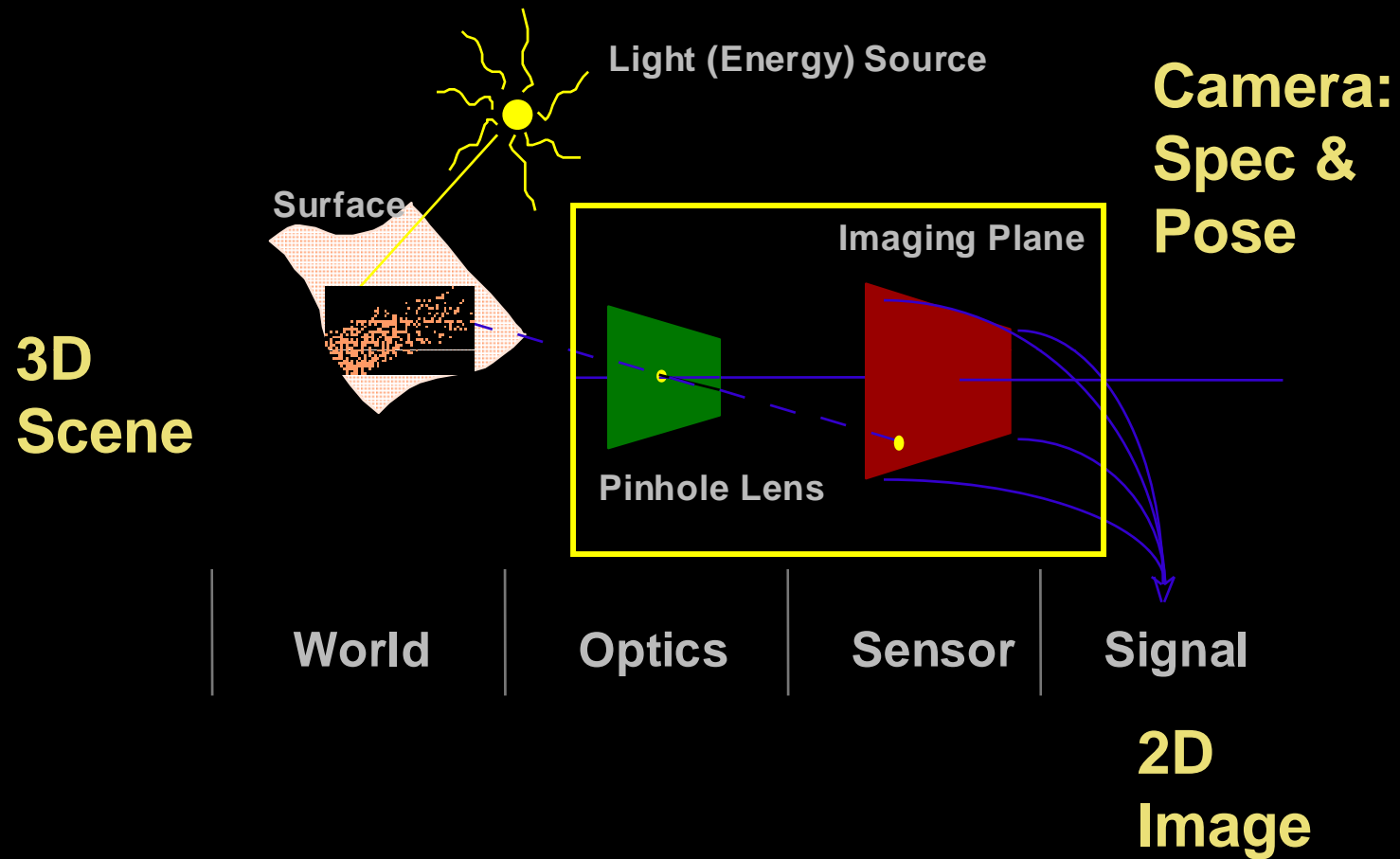
Silver Density

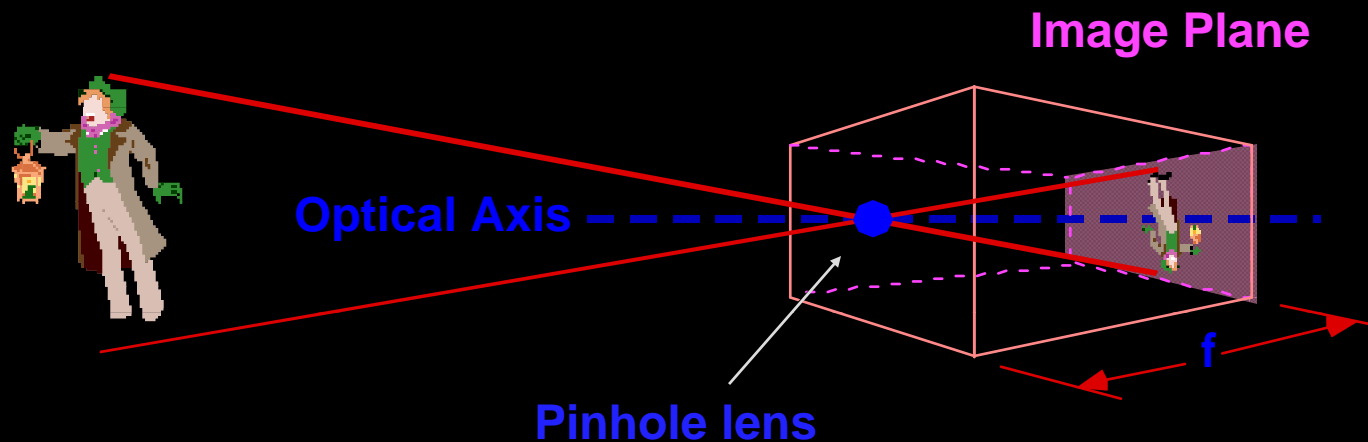
Color Film

Silver density
in three color
layers

TV Camera

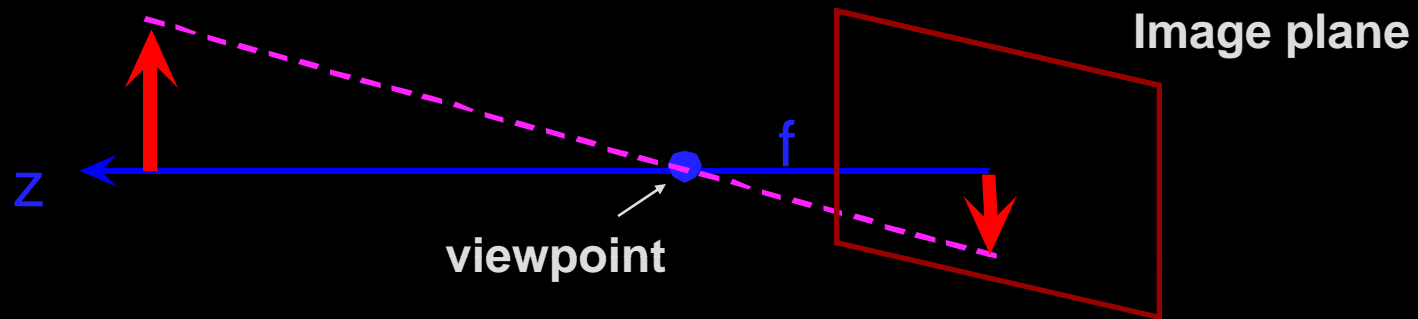
Electrical





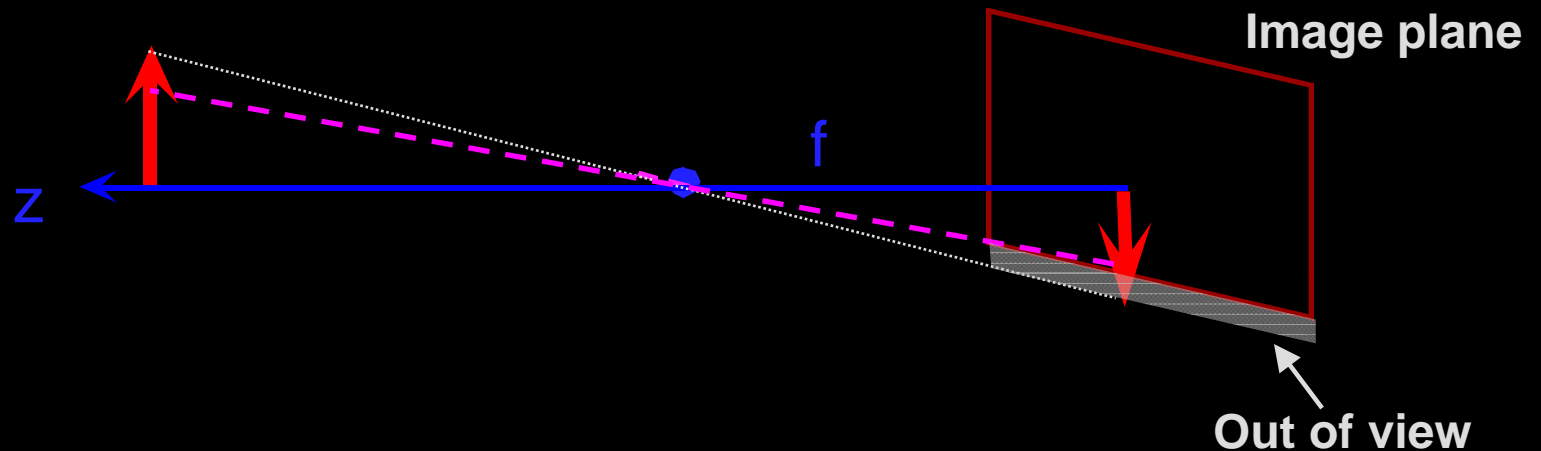
- Pin-hole is the basis for most graphics and vision
 - Derived from physical construction of early cameras
 - Mathematics is very straightforward
- 3D World projected to 2D Image
 - Image inverted, size reduced
 - Image is a 2D plane: No direct depth information
- Perspective projection
 - f called the focal length of the lens
 - given image size, change f will change FOV and figure sizes

- Consider case with object on the optical axis:



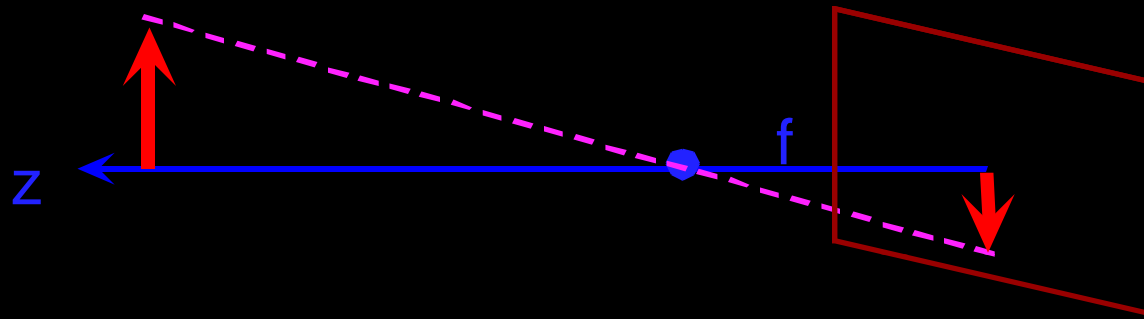
- **Optical axis**: the direction of imaging
- **Image plane**: a plane perpendicular to the optical axis
- **Center of Projection** (pinhole), focal point, viewpoint, nodal point
- **Focal length**: distance from focal point to the image plane
- **FOV** : Field of View – viewing angles in horizontal and vertical directions

- Consider case with object on the optical axis:

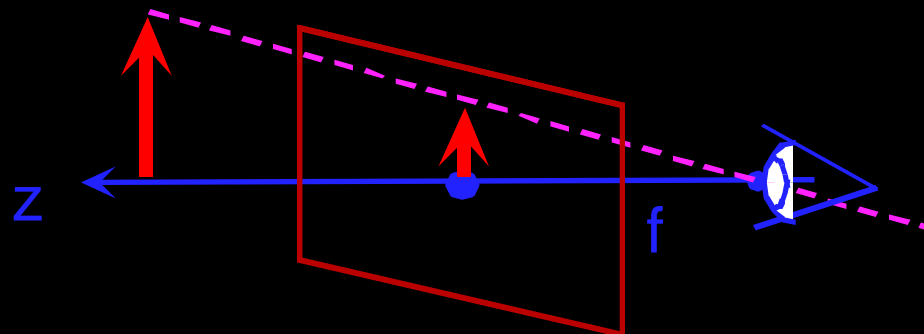


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- **Center of Projection** (pinhole), focal point, viewpoint, , nodal point
- **Focal length**: distance from focal point to the image plane
- **FOV** : Field of View – viewing angles in horizontal and vertical directions
- Increasing f will enlarge figures, but decrease FOV

- Consider case with object on the optical axis:



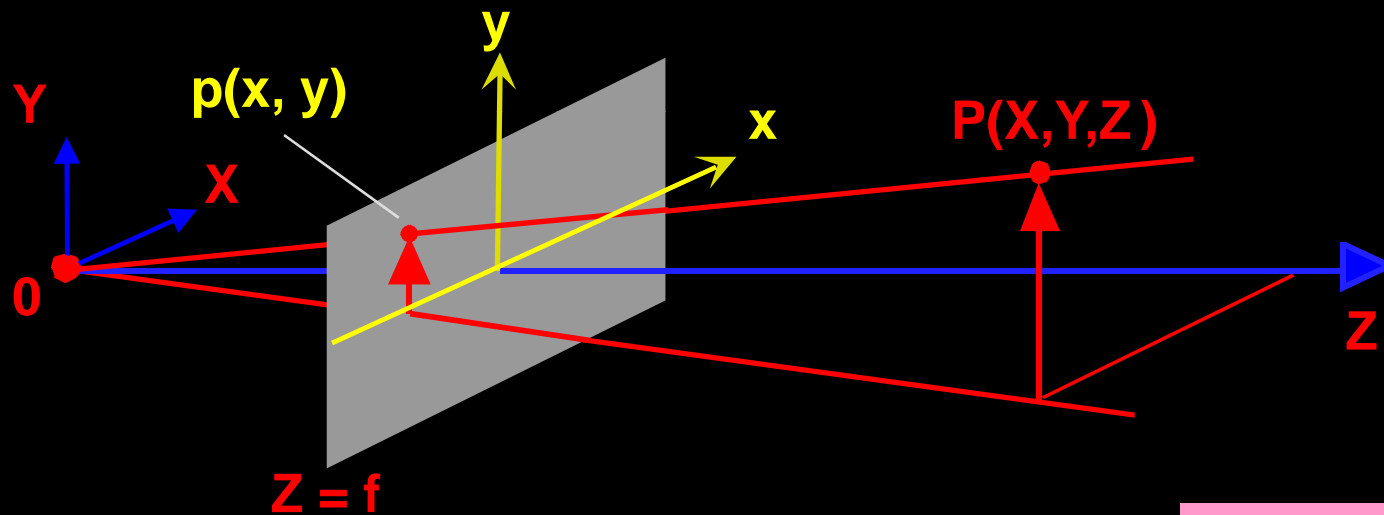
- More convenient with upright image:



Projection plane $z = f$

- Equivalent mathematically

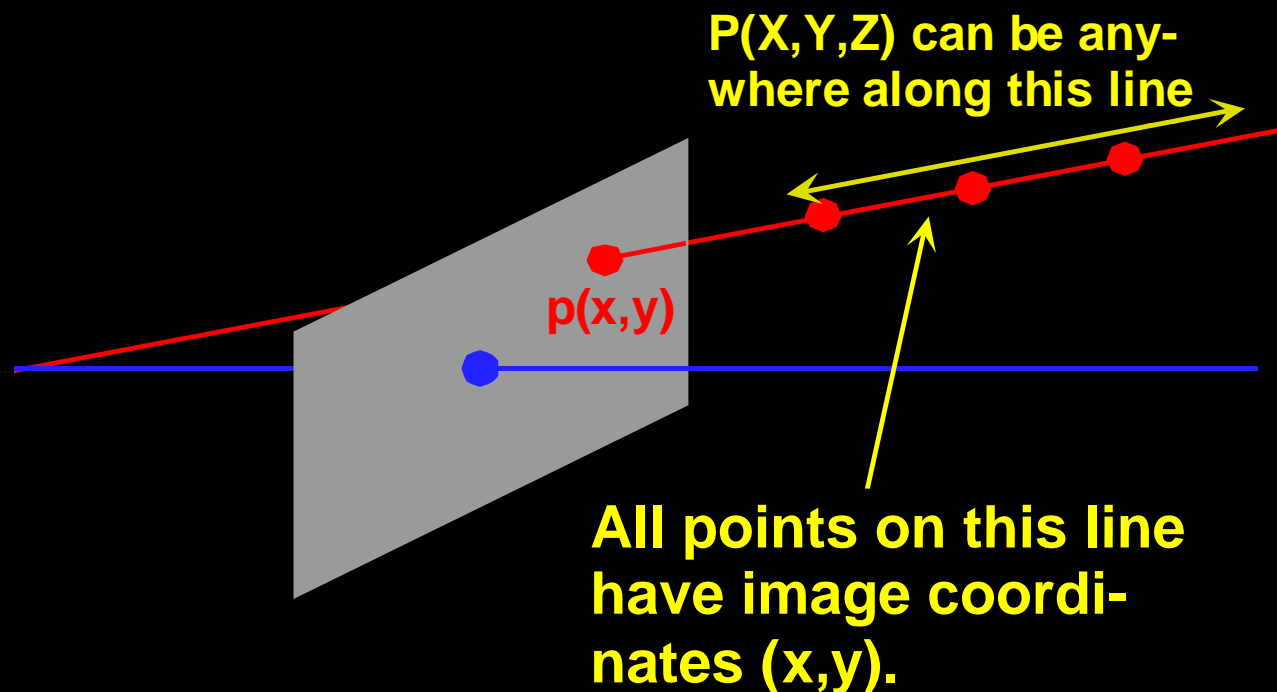
- Compute the image coordinates of p in terms of the world (camera) coordinates of P .



- Origin of camera at center of projection
- Z axis along optical axis
- Image Plane at $Z = f$; $x // X$ and $y // Y$

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

- Given a center of projection and image coordinates of a point, it is not possible to recover the 3D depth of the point from a single image.



In general, at least two images of the same point taken from two different locations are required to recover depth.

Amsterdam : **what do you see in this picture?**

- straight line
- size
- parallelism/angle
- shape
- shape of planes

- depth



Photo by Robert Kosara, robert@kosara.net

<http://www.kosara.net/gallery/pinholeamsterdam/pic01.html>

Amsterdam

- ✓ straight line
- size
- parallelism/angle
- shape
- shape of planes

- depth

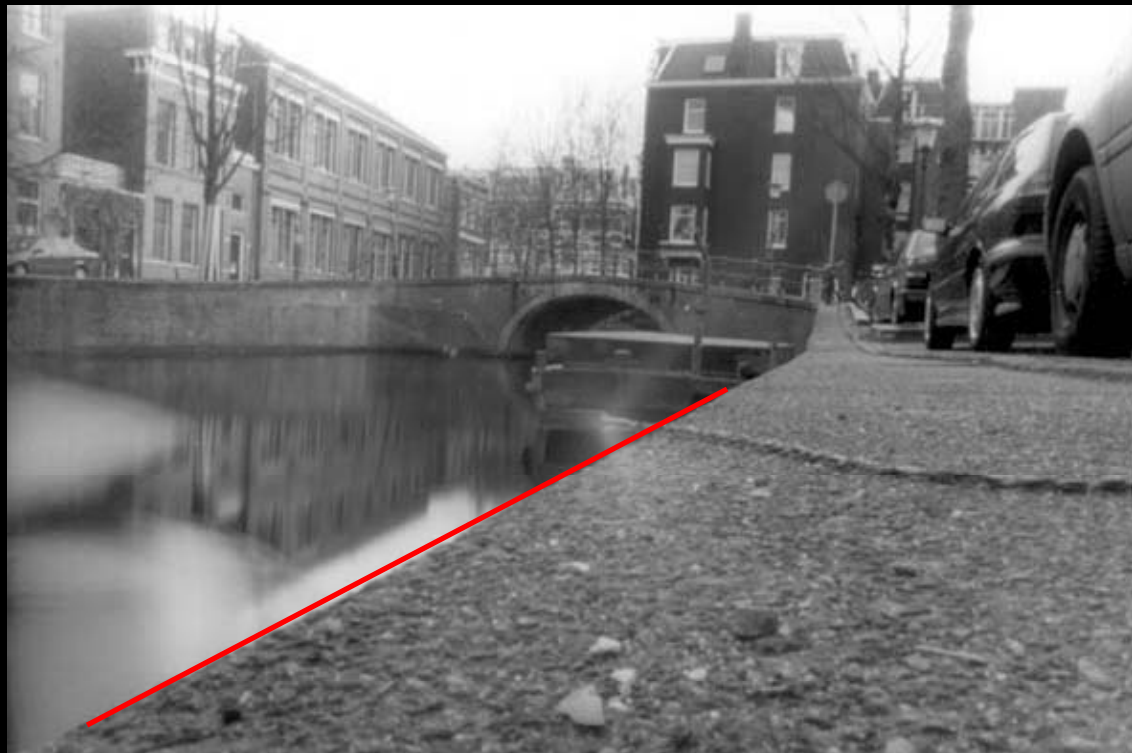


Photo by Robert Kosara, robert@kosara.net
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Amsterdam

- ✓ straight line
- × size
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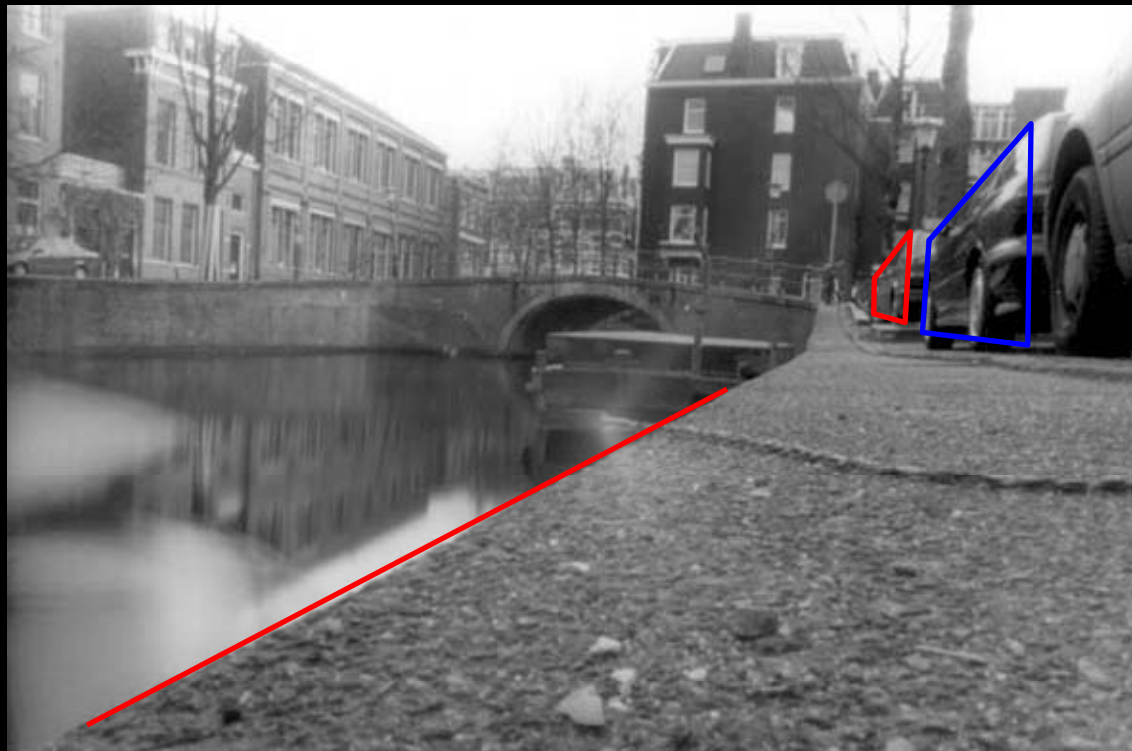


Photo by Robert Kosara, robert@kosara.net
<http://www.kosara.net/gallery/pinholeamsterdam/pic01.html>

Amsterdam

- ✓ straight line
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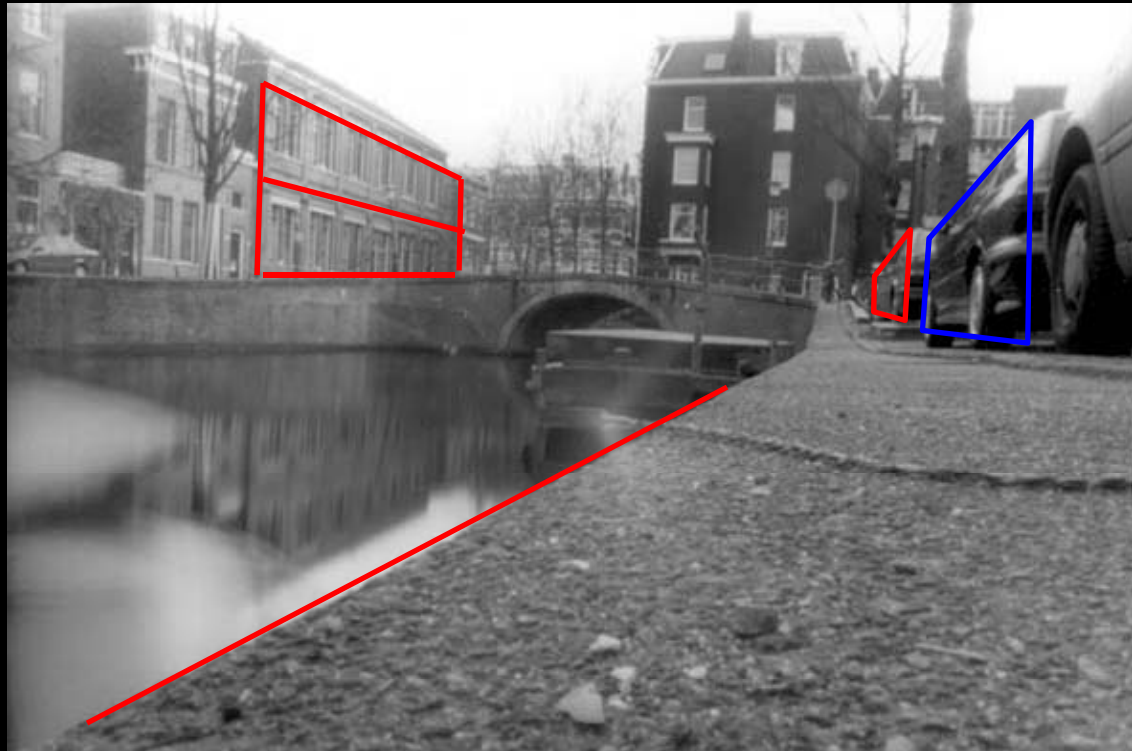


Photo by Robert Kosara, robert@kosara.net
<http://www.kosara.net/gallery/pinholeamsterdam/pic01.html>

Amsterdam

- ✓ straight line
- × size
- × parallelism/angle
- × shape
- shape of planes
- depth

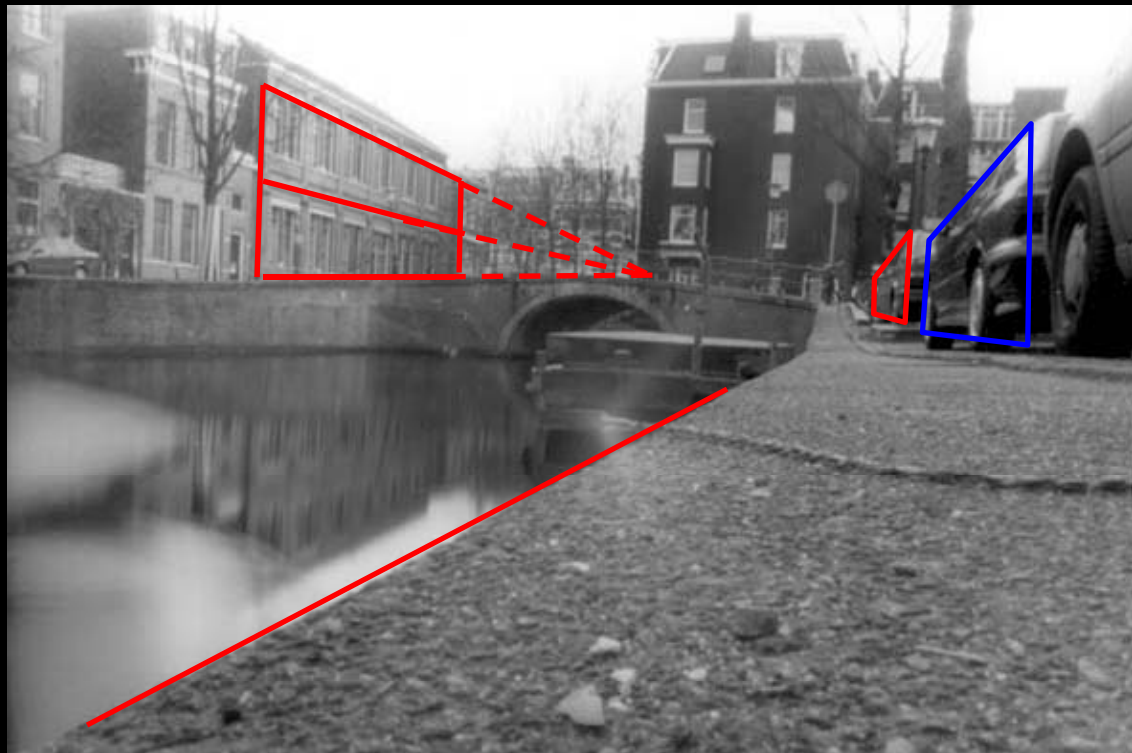


Photo by Robert Kosara, robert@kosara.net
<http://www.kosara.net/gallery/pinholeamsterdam/pic01.html>

Amsterdam

- ✓ straight line
- × size
- × parallelism/angle
- × shape
- shape of planes
- ✓ parallel to image
- depth

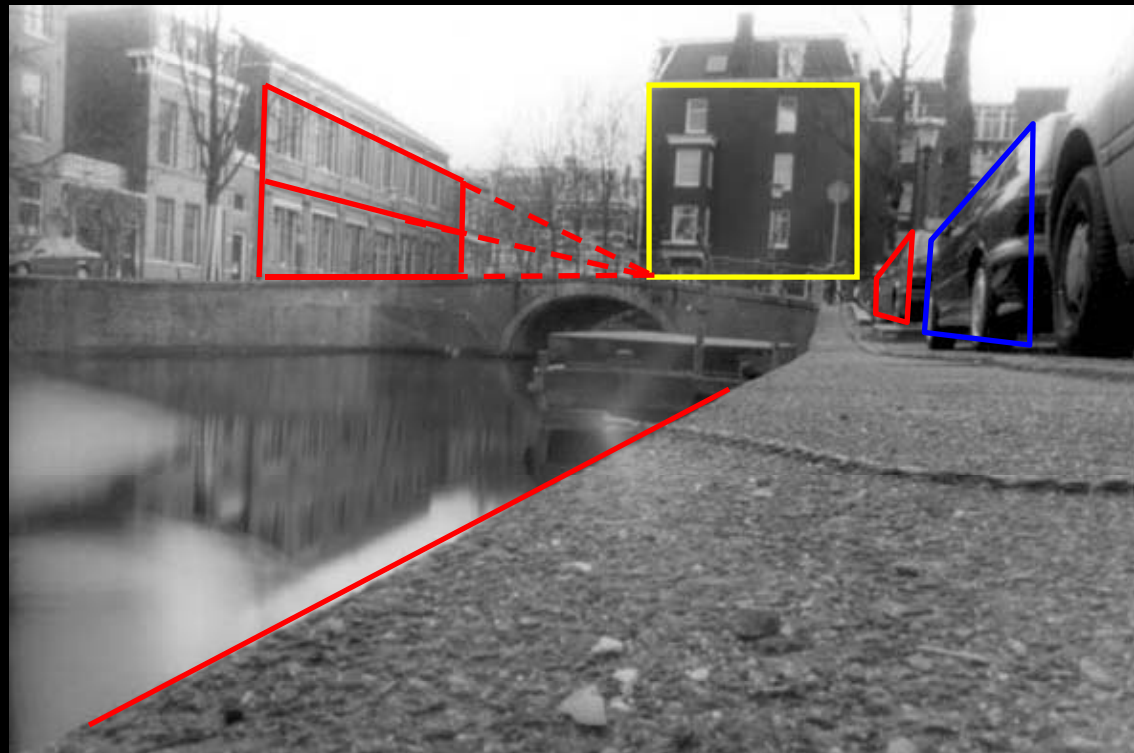
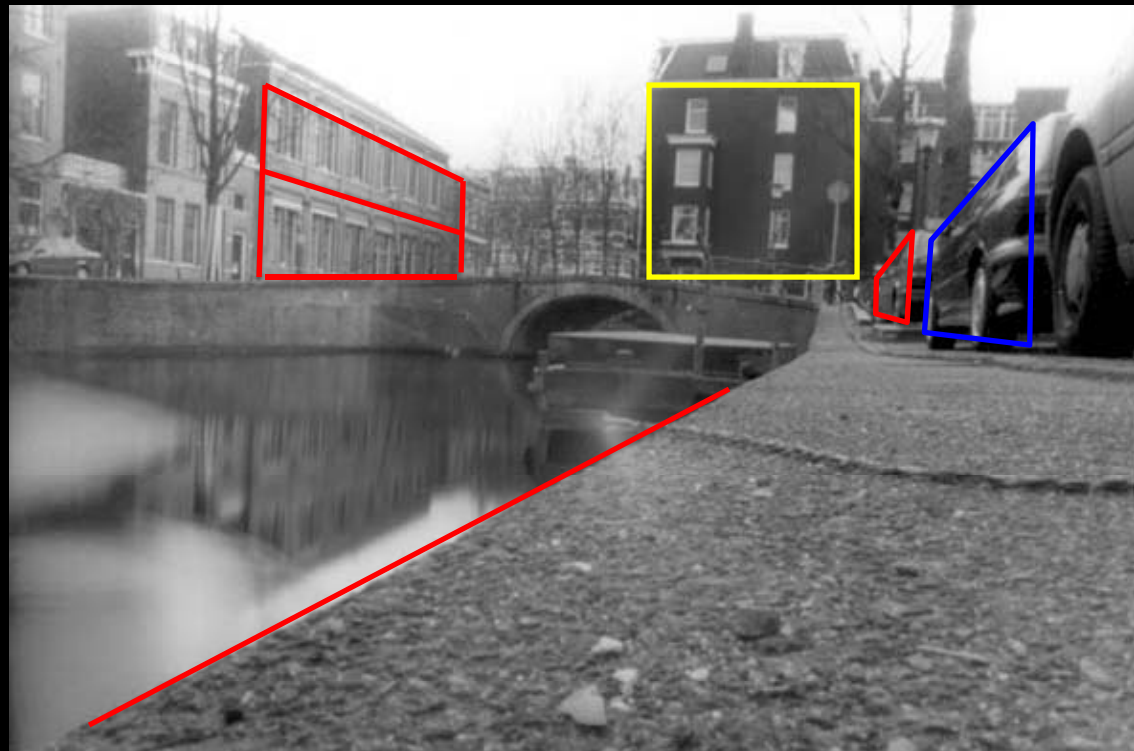


Photo by Robert Kosara, robert@kosara.net
<http://www.kosara.net/gallery/pinholeamsterdam/pic01.html>

Amsterdam: what do you see?

- ✓ straight line
- × size
- × parallelism/angle
- × shape
- shape of planes
- ✓ parallel to image
- Depth ?
 - stereo
 - motion
 - size
 - structure ...



- We see spatial shapes rather than individual pixels
- Knowledge: top-down vision belongs to human
- Stereo & Motion most successful in 3D CV & application
- You can see it but you don't know how...

3D Computer Vision

and Video Computing **Yet other pinhole camera images**

Rabbit or Man?



Markus Raetz, *Metamorphose II*, 1991-92, cast iron, 15 1/4 x 12 x 12 inches
Fine Art Center University Gallery, Sep 15 – Oct 26

3D Computer Vision

and Video Computing **Yet other pinhole camera images**

2D projections are not the “same” as the real object as we usually see everyday!



Markus Raetz, *Metamorphose II*, 1991-92, cast iron, 15 1/4 x 12 x 12 inches
Fine Art Center University Gallery, Sep 15 – Oct 26

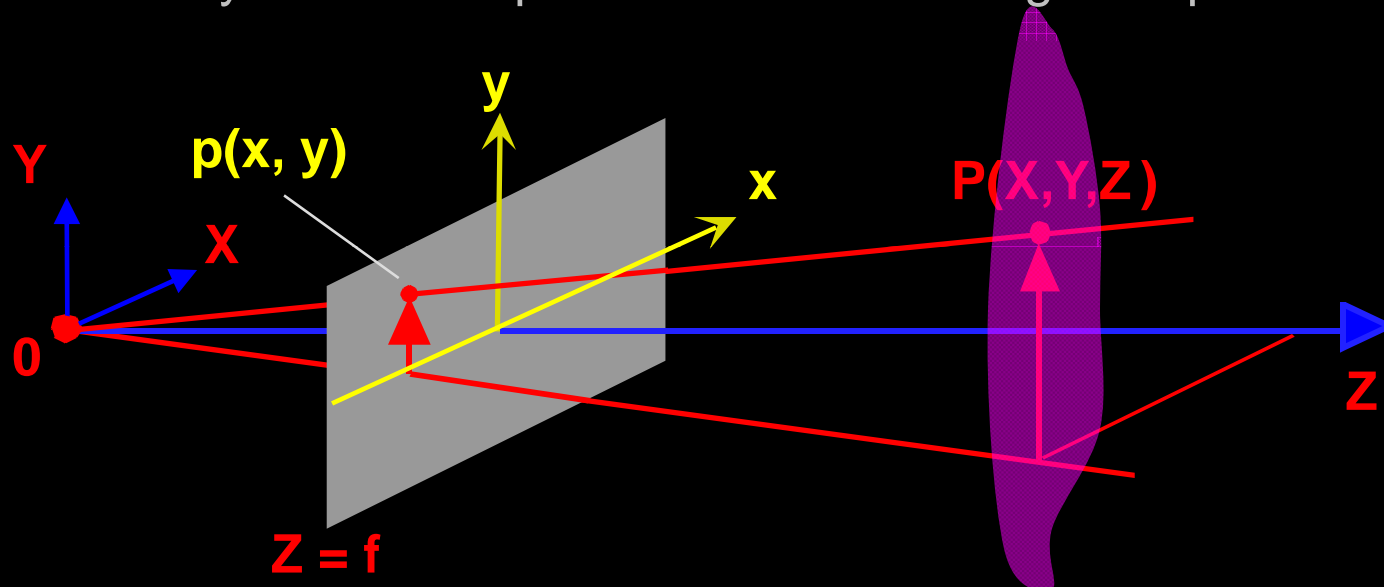
3D Computer Vision

and Video Computing

It's real!



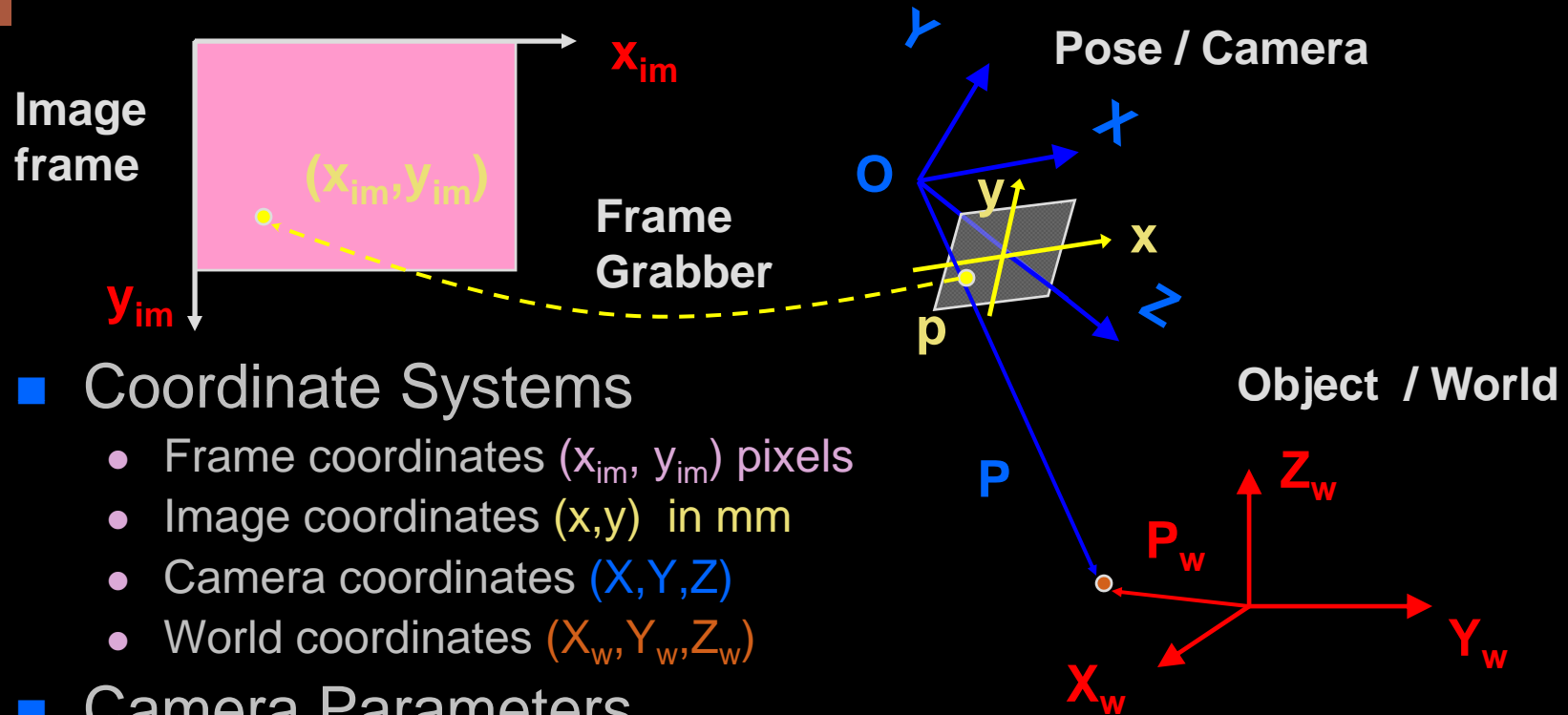
- Average depth \bar{Z} is much larger than the relative distance between any two scene points measured along the optical axis



- A sequence of two transformations
 - Orthographic projection : parallel rays
 - Isotropic scaling : f/\bar{Z}
- Linear Model
 - Preserve angles and shapes

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

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

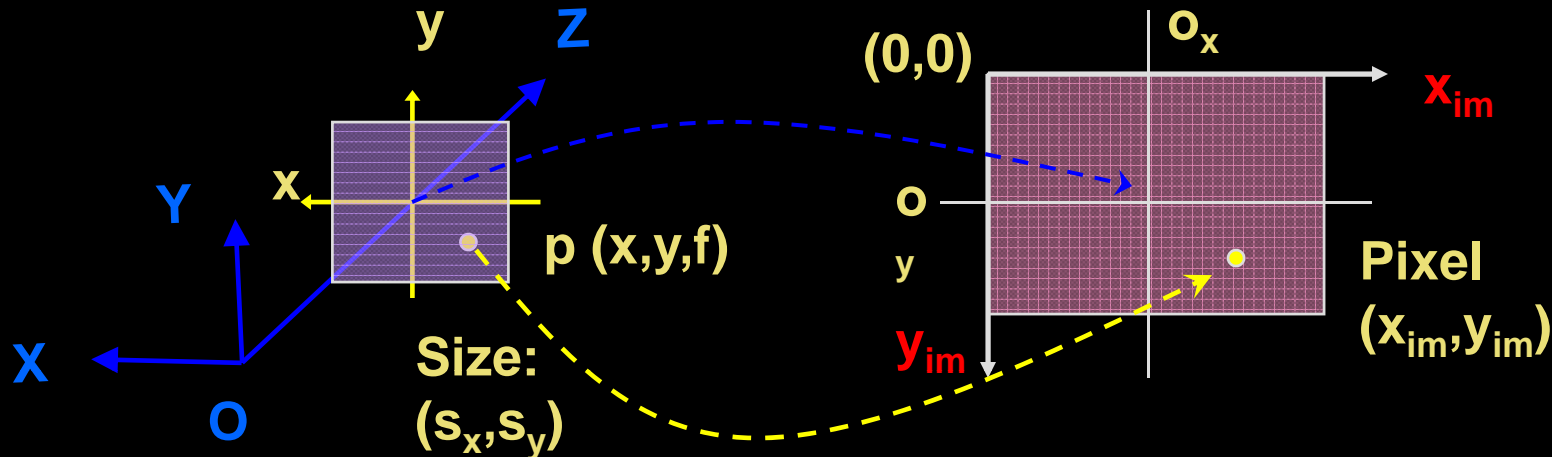


■ Coordinate Systems

- Frame coordinates (x_{im}, y_{im}) pixels
- Image coordinates (x, y) in mm
- Camera coordinates (X, Y, Z)
- World coordinates (X_w, Y_w, Z_w)

■ Camera Parameters

- Intrinsic Parameters (of the camera and the frame grabber): link the **frame coordinates** of an image point with its corresponding **camera coordinates**
- Extrinsic parameters: define the location and orientation of the **camera coordinate system** with respect to the **world coordinate system**



- From image to frame
 - Image center
 - Directions of axes
 - Pixel size
- From 3D to 2D
 - Perspective projection
- Intrinsic Parameters
 - (o_x, o_y) : image center (in pixels)
 - (s_x, s_y) : effective size of the pixel (in mm)
 - f : focal length

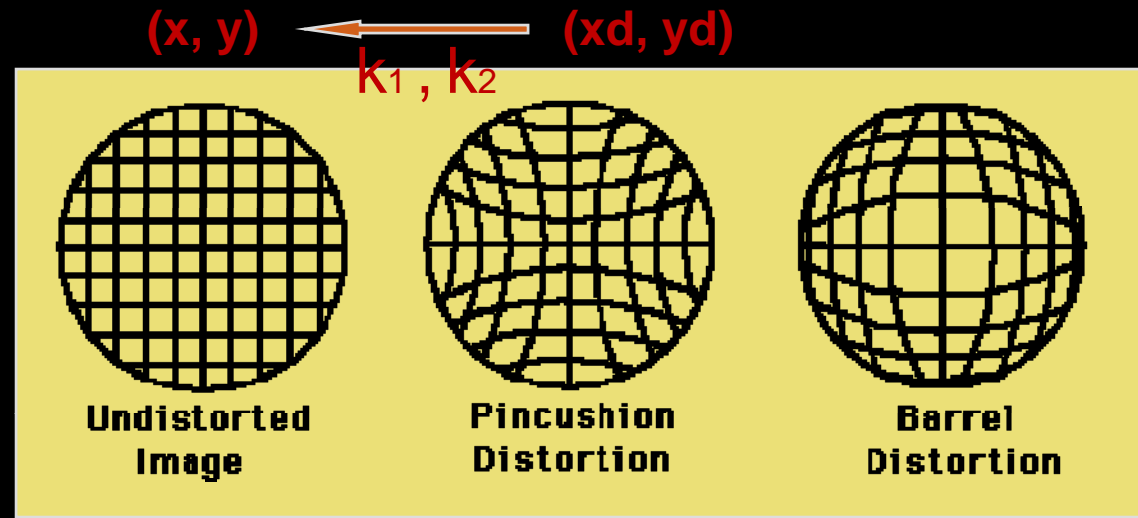
$$x = -(x_{im} - o_x)s_x$$

$$y = -(y_{im} - o_y)s_y$$

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

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

■ Lens Distortions



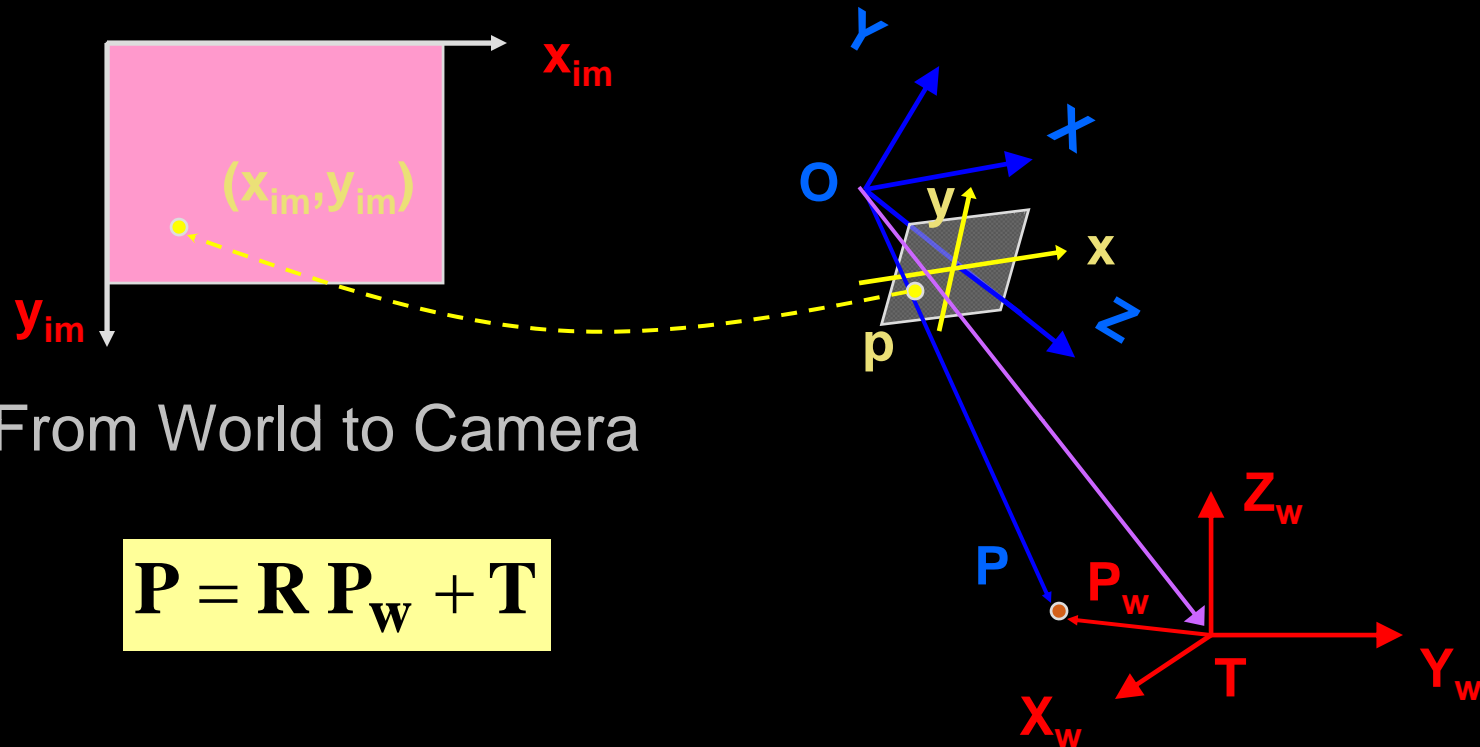
■ Modeled as simple radial distortions

- $r^2 = x_d^2 + y_d^2$
- (x_d, y_d) distorted points
- k_1, k_2 : distortion coefficients

$$x = x_d (1 + k_1 r^2 + k_2 r^4)$$

$$y = y_d (1 + k_1 r^2 + k_2 r^4)$$

- A model with $k_2 = 0$ is still accurate for a CCD sensor of 500x500 with ~5 pixels distortion on the outer boundary



■ From World to Camera

$$\mathbf{P} = \mathbf{R} \mathbf{P}_w + \mathbf{T}$$

■ Extrinsic Parameters

- A 3-D translation vector, T , describing the relative locations of the origins of the two coordinate systems (*what's it?*)
- A 3x3 rotation matrix, R , an orthogonal matrix that brings the corresponding axes of the two systems onto each other

■ A point as a 2D/ 3D vector

- Image point: 2D vector
- Scene point: 3D vector
- Translation: 3D vector

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

$$\mathbf{P} = (X, Y, Z)^T$$

$$\mathbf{T} = (T_x, T_y, T_z)^T$$

T: Transpose

■ Vector Operations

- Addition:
 - Translation of a 3D vector
- Dot product (a scalar):
 - $a \cdot b = |a||b|\cos\theta$
- Cross product (a vector)
 - Generates a new vector that is orthogonal to both of them

$$\mathbf{P} = \mathbf{P}_w + \mathbf{T} = (X_w + T_x, Y_w + T_y, Z_w + T_z)^T$$

$$c = \mathbf{a} \cdot \mathbf{b} = \mathbf{a}^T \mathbf{b}$$

$$\mathbf{c} = \mathbf{a} \times \mathbf{b}$$

$$\underline{\mathbf{a}} \times \underline{\mathbf{b}} = (a_2 b_3 - a_3 b_2) \underline{\mathbf{i}} + (a_3 b_1 - a_1 b_3) \underline{\mathbf{j}} + (a_1 b_2 - a_2 b_1) \underline{\mathbf{k}}$$

■ Rotation: 3x3 matrix

- Orthogonal :

$$\mathbf{R}^{-1} = \mathbf{R}^T, \text{ i.e. } \mathbf{R}\mathbf{R}^T = \mathbf{R}^T\mathbf{R} = \mathbf{I}$$

$$\mathbf{R} = (r_{ij})_{3 \times 3} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} = \begin{bmatrix} \mathbf{R}_1^T \\ \mathbf{R}_2^T \\ \mathbf{R}_3^T \end{bmatrix}$$

- 9 elements => 3+3 constraints (orthogonal/cross) => 2+2 constraints (unit vectors) => 3 DOF ? (degrees of freedom, orthogonal/dot)

- How to generate R from three angles? (next few slides)

■ Matrix Operations

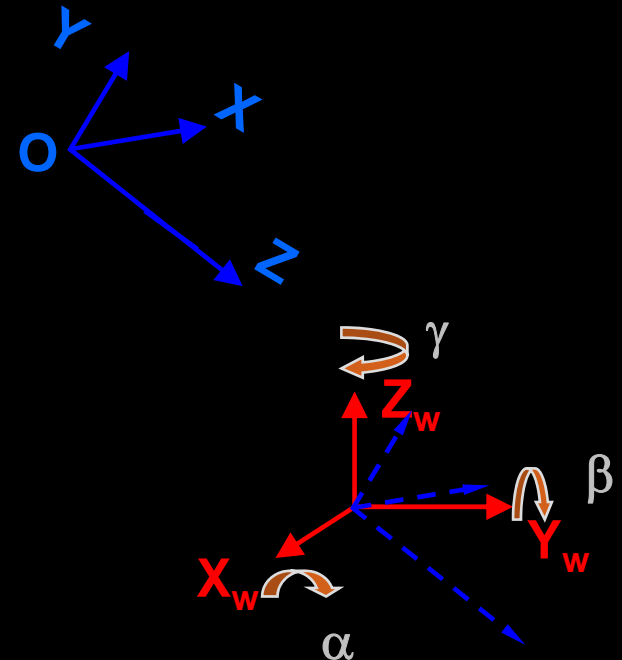
- $\mathbf{R} \mathbf{P}_w + \mathbf{T} = ?$ - Points in the World are projected on three new axes (of the camera system) and translated to a new origin

$$\mathbf{P} = \mathbf{R}\mathbf{P}_w + \mathbf{T} = \begin{pmatrix} r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x \\ r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y \\ r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z \end{pmatrix} = \begin{bmatrix} \mathbf{R}_1^T \mathbf{P}_w + T_x \\ \mathbf{R}_2^T \mathbf{P}_w + T_y \\ \mathbf{R}_3^T \mathbf{P}_w + T_z \end{bmatrix}$$

- Rotation around the Axes
 - Result of three consecutive rotations around the coordinate axes

$$\mathbf{R} = \mathbf{R}_\alpha \mathbf{R}_\beta \mathbf{R}_\gamma$$

- Notes:
 - Only three rotations
 - Every time around one axis
 - Bring corresponding axes to each other
 - $X_w = X, Y_w = Y, Z_w = Z$
 - First step (e.g.) Bring X_w to X



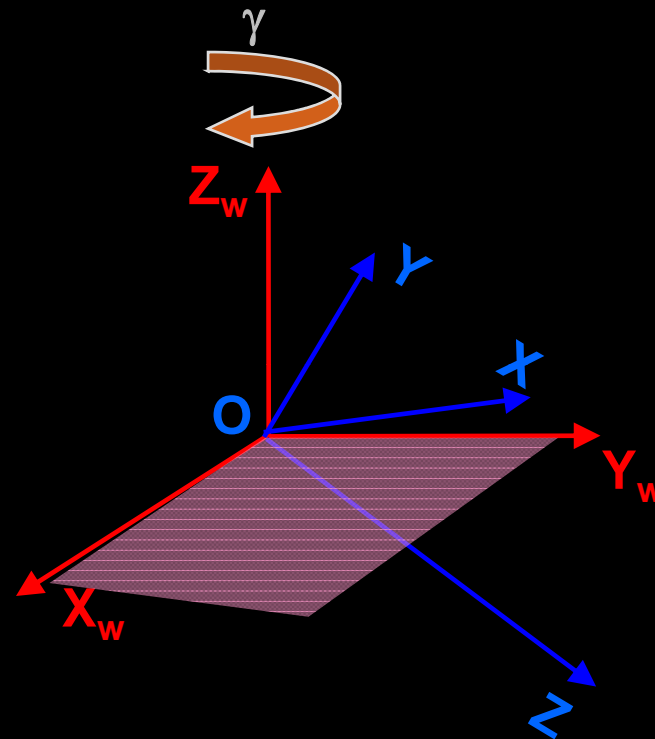


$$\mathbf{R}_\gamma = \begin{bmatrix} \cos \gamma & -\sin \gamma & 0 \\ \sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

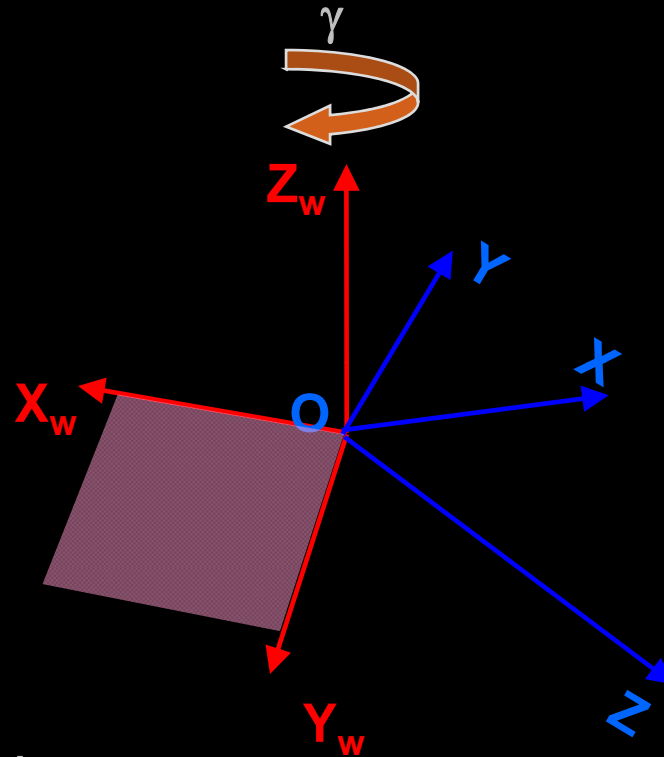
■ Rotation γ around the Z_w Axis

- Rotate in $X_w O Y_w$ plane
- Goal: Bring X_w to X
- But X is not in $X_w O Y_w$
- $Y_w \perp X \Rightarrow X$ in $X_w O Z_w$ ($\Leftarrow Y_w \perp X_w O Z_w$)
 $\Rightarrow Y_w$ in $Y O Z$ ($\Leftarrow X \perp Y O Z$)

■ Next time rotation around Y_w

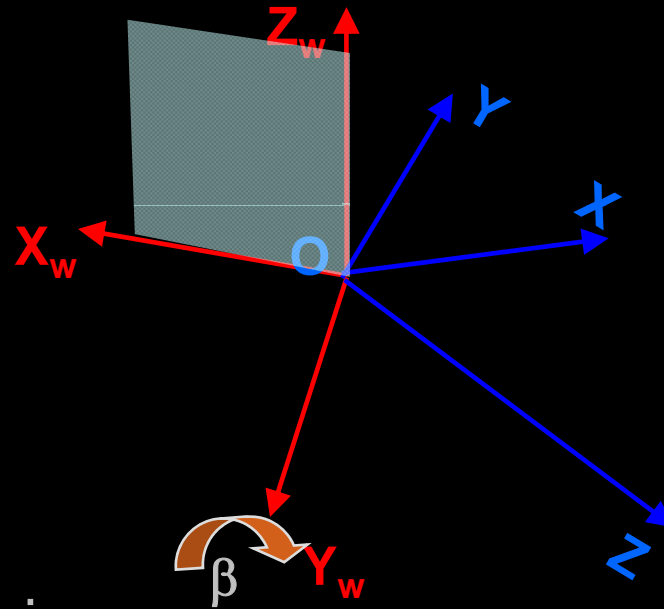


$$\mathbf{R}_\gamma = \begin{bmatrix} \cos \gamma & -\sin \gamma & 0 \\ \sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



- Rotation γ around the Z_w Axis
 - Rotate in X_wOY_w plane so that
 - $Y_w \perp X \Rightarrow X$ in X_wOZ_w ($\Leftarrow Y_w \perp X_wOZ_w$)
 $\Rightarrow Y_w$ in YOZ ($\Leftarrow X \perp YOZ$)
- Z_w does not change

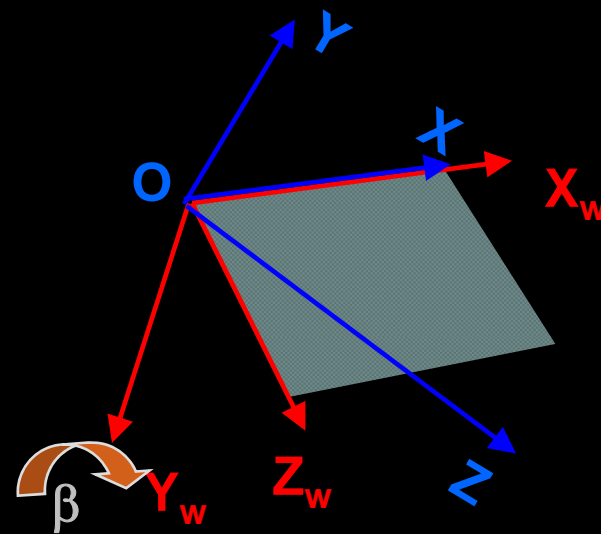
$$\mathbf{R}_\gamma = \begin{bmatrix} \cos \beta & 0 & -\sin \beta \\ 0 & 1 & 0 \\ \sin \beta & 0 & \cos \beta \end{bmatrix}$$



- Rotation β around the Y_w Axis
 - Rotate in $X_w O Z_w$ plane so that
 - $X_w = X \Rightarrow Z_w$ in YOZ (& Y_w in YOZ)
- Y_w does not change

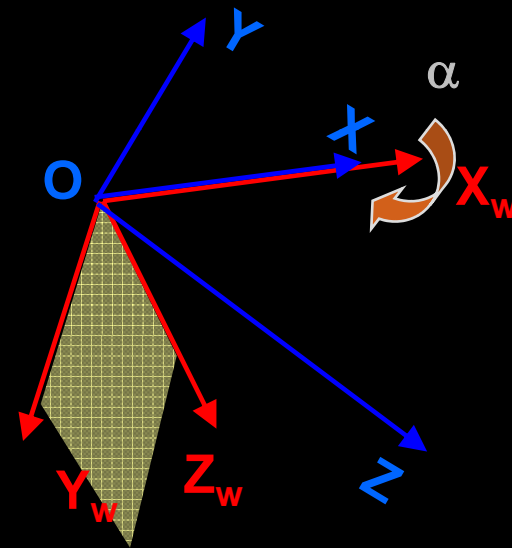


$$\mathbf{R}_\beta = \begin{bmatrix} \cos \beta & 0 & -\sin \beta \\ 0 & 1 & 0 \\ \sin \beta & 0 & \cos \beta \end{bmatrix}$$



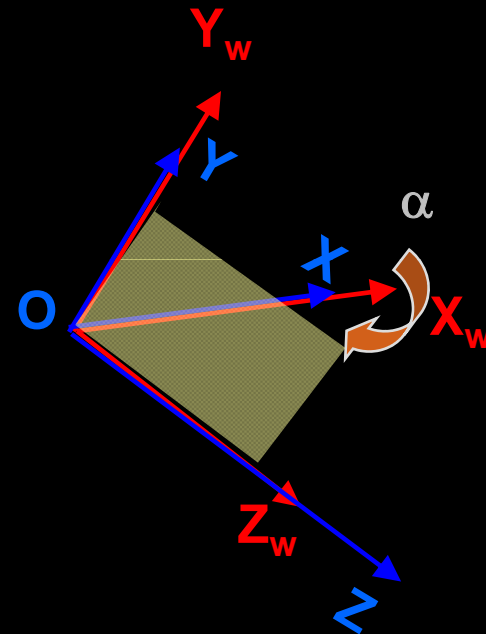
- Rotation β around the Y_w Axis
 - Rotate in X_wOZ_w plane so that
 - $X_w = X \Rightarrow Z_w$ in YOZ (& Y_w in YOZ)
- Y_w does not change

$$\mathbf{R}_\alpha = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & -\sin \alpha \\ 0 & \sin \alpha & \cos \alpha \end{bmatrix}$$



- Rotation α around the $X_w(X)$ Axis
 - Rotate in Y_wOZ_w plane so that
 - $Y_w = Y, Z_w = Z$ (& $X_w = X$)
- X_w does not change

$$\mathbf{R}_\alpha = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & -\sin \alpha \\ 0 & \sin \alpha & \cos \alpha \end{bmatrix}$$



- Rotation α around the $X_w(X)$ Axis
 - Rotate in Y_wOZ_w plane so that
 - $Y_w = Y, Z_w = Z$ (& $X_w = X$)
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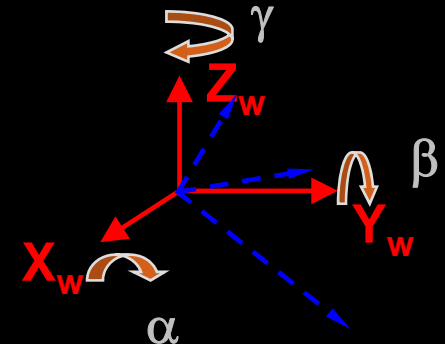
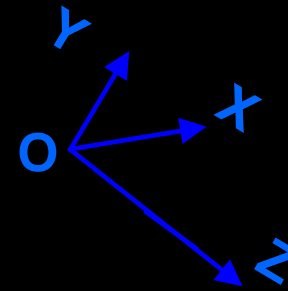
■ Rotation around the Axes

- Result of three consecutive rotations around the coordinate axes

$$\mathbf{R} = \mathbf{R}_\alpha \mathbf{R}_\beta \mathbf{R}_\gamma$$

■ Notes:

- Rotation directions
- The order of multiplications matters: γ, β, α
- Same R, 6 different sets of α, β, γ
- **R Non-linear function of α, β, γ**
- **R is orthogonal**
- **It's easy to compute angles from R**



$$\mathbf{R} = \begin{bmatrix} \cos \beta \cos \gamma & -\cos \beta \sin \gamma & -\sin \beta \\ -\sin \alpha \sin \beta \cos \gamma + \cos \alpha \sin \gamma & \sin \alpha \sin \beta \sin \gamma + \cos \alpha \cos \gamma & -\sin \alpha \cos \beta \\ \cos \alpha \sin \beta \cos \gamma + \sin \alpha \sin \gamma & -\cos \alpha \sin \beta \sin \gamma + \sin \alpha \cos \gamma & \cos \alpha \cos \beta \end{bmatrix}$$

Appendix A.9 of the textbook

- According to Euler's Theorem, any 3D rotation can be described by a rotating angle, θ , around an axis defined by an unit vector $\mathbf{n} = [n_1, n_2, n_3]^T$.
- Three degrees of freedom – why?

$$\mathbf{R} = I \cos \theta + \begin{bmatrix} n_1^2 & n_1 n_2 & n_1 n_3 \\ n_2 n_1 & n_2^2 & n_2 n_3 \\ n_3 n_1 & n_3 n_2 & n_3^2 \end{bmatrix} (1 - \cos \theta) + \begin{bmatrix} 0 & -n_3 & n_2 \\ n_3 & 0 & -n_1 \\ -n_2 & n_1 & 0 \end{bmatrix} \sin \theta$$

■ World to Camera

- Camera: $P = (X, Y, Z)^T$
- World: $P_w = (X_w, Y_w, Z_w)^T$
- Transform: R, T

$$P = RP_w + T = \begin{pmatrix} r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x \\ r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y \\ r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z \end{pmatrix} = \begin{bmatrix} \mathbf{R}_1^T \mathbf{P}_w + T_x \\ \mathbf{R}_2^T \mathbf{P}_w + T_y \\ \mathbf{R}_3^T \mathbf{P}_w + T_z \end{bmatrix}$$

■ Camera to Image

- Camera: $P = (X, Y, Z)^T$
- Image: $p = (x, y)^T$
- Not linear equations

$$(x, y) = \left(f \frac{X}{Z}, f \frac{Y}{Z} \right)$$

■ Image to Frame

- Neglecting distortion
- Frame $(x_{im}, y_{im})^T$

$$x = -(x_{im} - o_x)s_x$$

$$y = -(y_{im} - o_y)s_y$$

■ World to Frame

- $(X_w, Y_w, Z_w)^T \rightarrow (x_{im}, y_{im})^T$
- Effective focal lengths
 - $f_x = f/s_x, f_y = f/s_y$
 - Three are not independent

$$x_{im} - o_x = -f_x \frac{r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

$$y_{im} - o_y = -f_y \frac{r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

Linear Matrix Equation of perspective projection

Projective Space

- Add fourth coordinate
 - $P_w = (X_w, Y_w, Z_w, 1)^T$
- Define $(x_1, x_2, x_3)^T$ such that
 - $x_1/x_3 = X_{im}, x_2/x_3 = y_{im}$

$$\begin{pmatrix} x_{im} \\ y_{im} \end{pmatrix} = \begin{pmatrix} x_1 / x_3 \\ x_2 / x_3 \end{pmatrix}$$

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \mathbf{M}_{int} \mathbf{M}_{ext} \begin{pmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{pmatrix}$$

3x4 Matrix \mathbf{M}_{ext}

- Only extrinsic parameters
- World to camera

$$\mathbf{M}_{ext} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & T_x \\ r_{21} & r_{22} & r_{23} & T_y \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix} = \begin{bmatrix} \mathbf{R}_1^T & T_x \\ \mathbf{R}_2^T & T_y \\ \mathbf{R}_3^T & T_z \end{bmatrix}$$

3x3 Matrix \mathbf{M}_{int}

- Only intrinsic parameters
- Camera to frame

$$\mathbf{M}_{int} = \begin{bmatrix} -f_x & 0 & o_x \\ 0 & -f_y & o_y \\ 0 & 0 & 1 \end{bmatrix}$$

Simple Matrix Product! Projective Matrix $\mathbf{M} = \mathbf{M}_{int} \mathbf{M}_{ext}$

- $(X_w, Y_w, Z_w)^T \rightarrow (x_{im}, y_{im})^T$
- Linear Transform from projective space to projective plane
- \mathbf{M} defined up to a scale factor – 11 independent entries

■ Perspective Camera Model

- Making some assumptions
 - Known center: $O_x = O_y = 0$
 - Square pixel: $S_x = S_y = 1$
- 11 independent entries \leftrightarrow 7 parameters

$$\mathbf{M} = \begin{bmatrix} -fr_{11} & -fr_{12} & -fr_{13} & -fT_x \\ -fr_{21} & -fr_{22} & -fr_{23} & -fT_y \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix}$$

■ Weak-Perspective Camera Model

- Average Distance $\bar{Z} \gg$ Range δZ
- Define centroid vector $\bar{\mathbf{P}}_w$
- $\mathbf{Z} = \bar{\mathbf{Z}} = \mathbf{R}_3^T \bar{\mathbf{P}}_w + T_z$
- 8 independent entries

$$\mathbf{M}_{wp} = \begin{bmatrix} -fr_{11} & -fr_{12} & -fr_{13} & -fT_x \\ -fr_{21} & -fr_{22} & -fr_{23} & -fT_y \\ 0 & 0 & 0 & \mathbf{R}_3^T \bar{\mathbf{P}}_w + T_z \end{bmatrix}$$

■ Affine Camera Model

- Mathematical Generalization of Weak-Pers
- Doesn't correspond to physical camera
- But simple equation and appealing geometry
 - Doesn't preserve angle BUT parallelism
- 8 independent entries

$$\mathbf{M}_{af} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & b_1 \\ a_{21} & a_{22} & a_{23} & b_2 \\ 0 & 0 & 0 & b_3 \end{bmatrix}$$

■ Planes are very common in the Man-Made World

$$n_x X_w + n_y Y_w + n_z Z_w = d \iff \mathbf{n}^T \mathbf{P}_w = d$$

- One more constraint for all points: Z_w is a function of X_w and Y_w

■ Special case: Ground Plane

- $Z_w=0$
- $\mathbf{P}_w = (X_w, Y_w, 0, 1)^T$
- 3D point \rightarrow 2D point

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{bmatrix} -fr_{11} & -fr_{12} & -fr_{13} & -fT_x \\ -fr_{21} & -fr_{22} & -fr_{23} & -fT_y \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix} \begin{pmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{pmatrix} = \begin{pmatrix} \\ \\ 0 \\ 1 \end{pmatrix}$$

■ Projective Model of a Plane

- 8 independent entries

■ General Form ?

- 8 independent entries

■ A Plane in the World

$$n_x X_w + n_y Y_w + n_z Z_w = d \iff \mathbf{n}^T \mathbf{P}_w = d$$

- One more constraint for all points: Z_w is a function of X_w and Y_w

■ Special case: Ground Plane

- $Z_w = 0$
- $\mathbf{P}_w = (X_w, Y_w, 0, 1)^T$
- 3D point \rightarrow 2D point

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{bmatrix} -fr_{11} & -fr_{12} & -fr_{13} & -fT_x \\ -fr_{21} & -fr_{22} & -fr_{23} & -fT_y \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix} \begin{pmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{pmatrix} = 0$$

■ Projective Model of $Z_w = 0$

- 8 independent entries

■ General Form ?

- 8 independent entries

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{bmatrix} -fr_{11} & -fr_{12} & -fT_x \\ -fr_{21} & -fr_{22} & -fT_{23} \\ r_{31} & r_{32} & T_z \end{bmatrix} \begin{pmatrix} X_w \\ Y_w \\ 1 \end{pmatrix}$$

■ A Plane in the World

$$n_x X_w + n_y Y_w + n_z Z_w = d \iff \mathbf{n}^T \mathbf{P}_w = d$$

- One more constraint for all points: Z_w is a function of X_w and Y_w

■ Special case: Ground Plane

- $Z_w=0$
- $\mathbf{P}_w = (X_w, Y_w, 0, 1)^T$
- 3D point \rightarrow 2D point

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{bmatrix} -fr_{11} & -fr_{12} & -fr_{13} & -fT_x \\ -fr_{21} & -fr_{22} & -fr_{23} & -fT_y \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix} \begin{pmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{pmatrix}$$

■ Projective Model of $Z_w=0$

- 8 independent entries

■ General Form ?

- $n_z = 1$

$$Z_w = d - n_x X_w - n_y Y_w$$

- 8 independent entries

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{bmatrix} -f(r_{11} - n_x r_{13}) & -f(r_{12} - n_y r_{13}) & -f(dr_{13} + T_x) \\ -f(r_{21} - n_x r_{23}) & -f(r_{22} - n_y r_{23}) & -f(dr_{23} + T_y) \\ (r_{31} - n_x r_{33}) & (r_{32} - n_y r_{33}) & (dr_{33} + T_z) \end{bmatrix} \begin{pmatrix} X_w \\ Y_w \\ 1 \end{pmatrix}$$

■ 2D $(x_{im}, y_{im}) \rightarrow$ 3D (X_w, Y_w, Z_w) ?

■ Graphics /Rendering

- From 3D world to 2D image
 - Changing viewpoints and directions
 - Changing focal length
- Fast rendering algorithms

■ Vision / Reconstruction

- From 2D image to 3D model
 - Inverse problem
 - Much harder / unsolved
- Robust algorithms for matching and parameter estimation
- Need to estimate camera parameters first

■ Calibration

- Find intrinsic & extrinsic parameters
- Given image-world point pairs
- Probably a partially solved problem ?
- 11 independent entries
 - <-> 10 parameters: $f_x, f_y, o_x, o_y, \alpha, \beta, \gamma, T_x, T_y, T_z$

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \mathbf{M}_{\text{int}} \mathbf{M}_{\text{ext}} \begin{pmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} x_{im} \\ y_{im} \end{pmatrix} = \begin{pmatrix} x_1 / x_3 \\ x_2 / x_3 \end{pmatrix}$$

$$\mathbf{M}_{\text{int}} = \begin{bmatrix} -f_x & 0 & o_x \\ 0 & -f_y & o_y \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{M}_{\text{ext}} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & T_x \\ r_{21} & r_{22} & r_{23} & T_y \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix}$$

- Geometric Projection of a Camera
 - Pinhole camera model
 - Perspective projection
 - Weak-Perspective Projection
- Camera Parameters (10 or 11)
 - Intrinsic Parameters: $f, o_x, o_y, s_x, s_y, k_1$: 4 or 5 independent parameters
 - Extrinsic parameters: R, T – 6 DOF (degrees of freedom)
- Linear Equations of Camera Models (without distortion)
 - General Projection Transformation Equation : 11 parameters
 - Perspective Camera Model: 11 parameters
 - Weak-Perspective Camera Model: 8 parameters
 - Affine Camera Model: generalization of weak-perspective: 8
 - Projective transformation of planes: 8 parameters

- Determining the value of the extrinsic and intrinsic parameters of a camera

Calibration (Ch. 6)

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \mathbf{M}_{\text{int}} \mathbf{M}_{\text{ext}} \begin{pmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{pmatrix}$$



$$\mathbf{M}_{\text{int}} = \begin{bmatrix} -f_x & 0 & o_x \\ 0 & -f_y & o_y \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{M}_{\text{ext}} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & T_x \\ r_{21} & r_{22} & r_{23} & T_y \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix}$$