

*CSC I6716*  
*Spring 2004*



Topic 5 of Part 2  
Camera Models

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<http://www-cs.engr.ccnycuny.edu/~zhu/VisionCourse-2004.html>

- Closely Related Disciplines
  - Image processing – image to image
  - Pattern recognition – image to classes
  - 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 (Part 2)
  - Camera Geometric Model (1 lecture – this class- topic 5)
  - Camera Calibration (1 lecture – topic 6)
  - Stereo (2 lectures – topic 7)
  - Motion (2 lectures –topic 8)

- 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

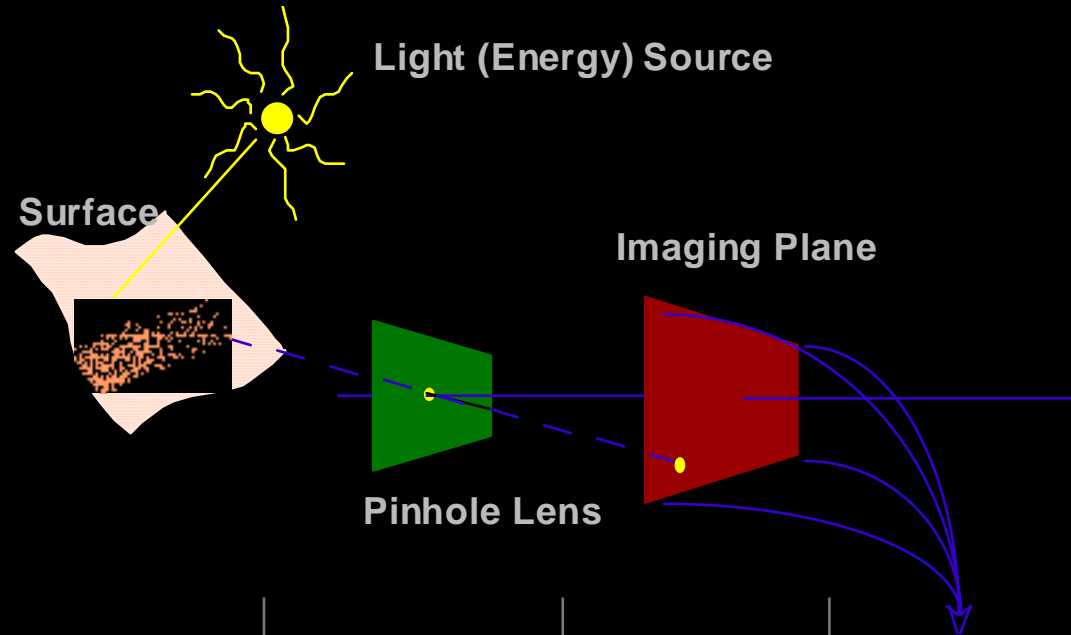
- Camera **Geometric** Models
  - Knowledge about 2D and 3D geometric transformations
  - Linear algebra (vector, matrix)
  - **This lecture is only about geometry**
- Goal

Build up relation between 2D images and 3D scenes

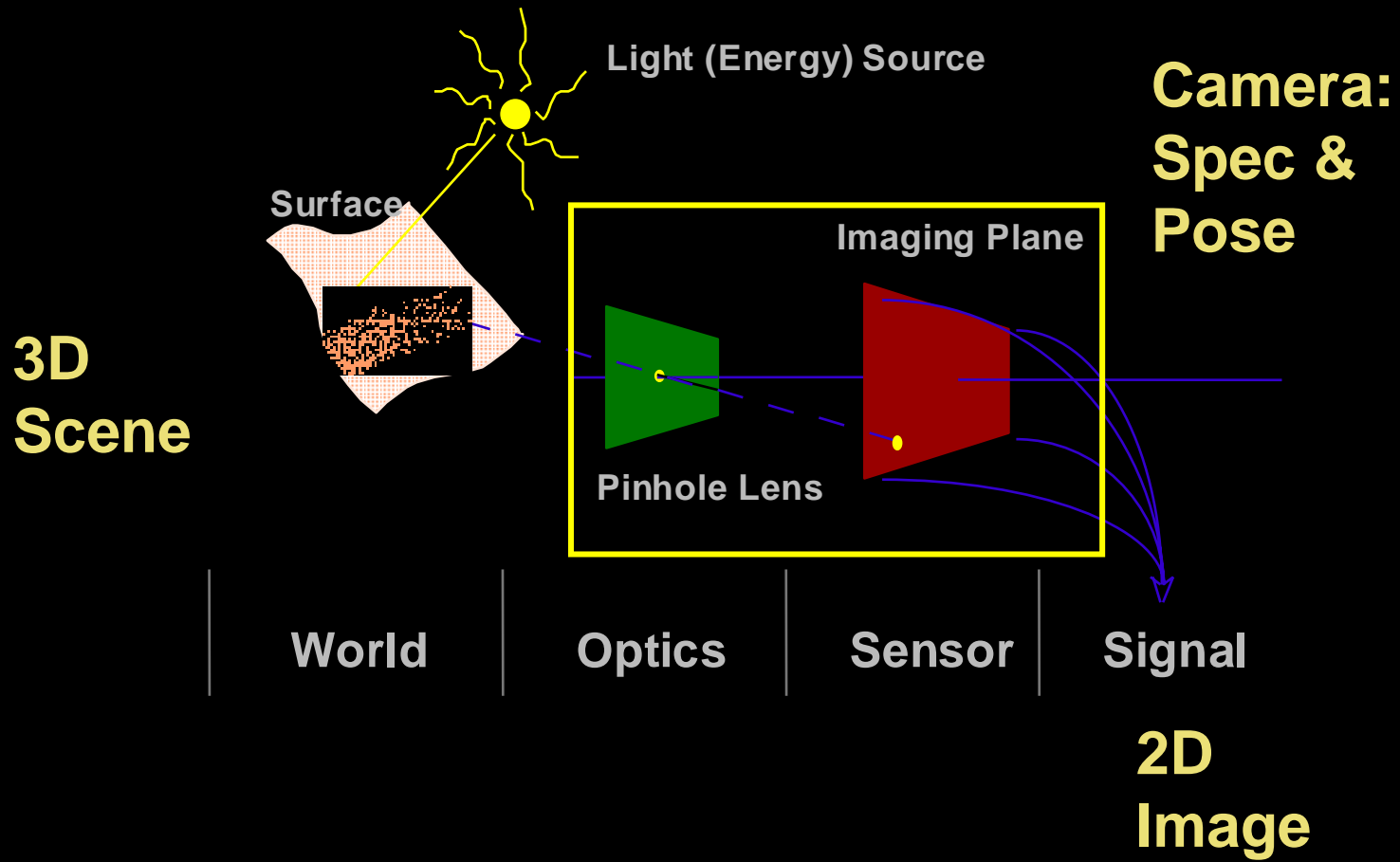
-3D Graphics (rendering): from 3D to 2D

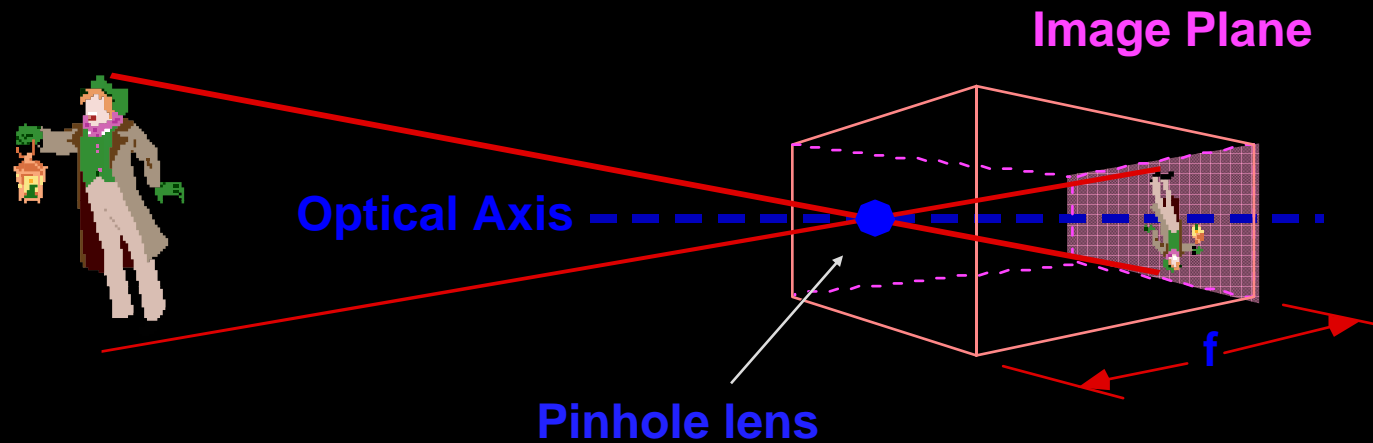
-3D Vision (stereo and motion): from 2D to 3D

-Calibration: Determining the parameters for mapping



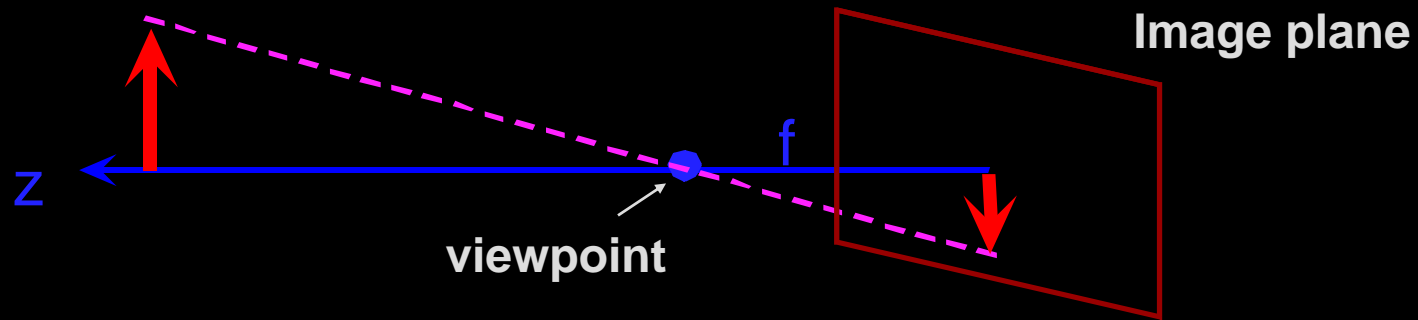
World	Optics	Sensor	Signal
		B&W Film	Silver Density
		Color Film	Silver density in three color layers
		TV Camera	Electrical





- 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

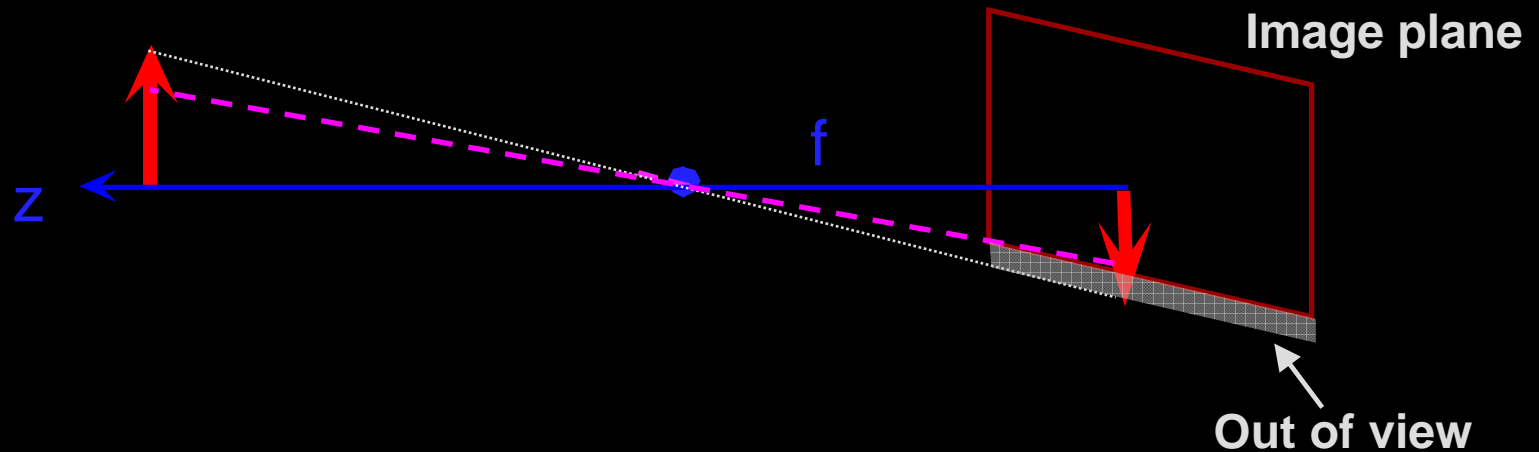
- Consider case with object on the optical axis:



- **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

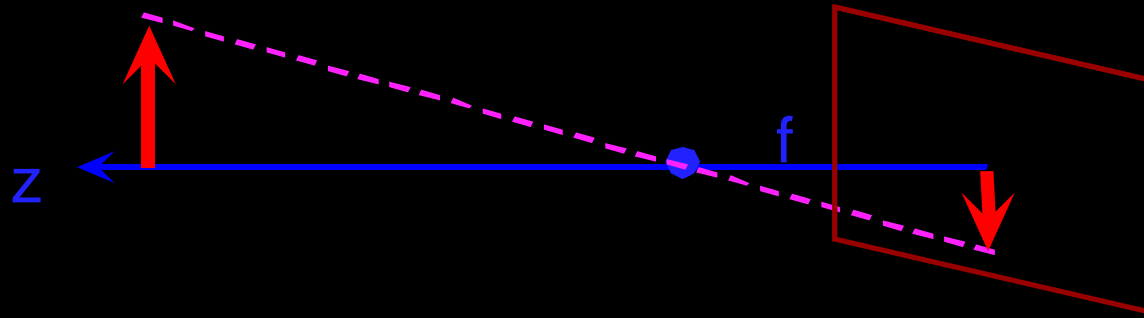


- Consider case with object on the optical axis:

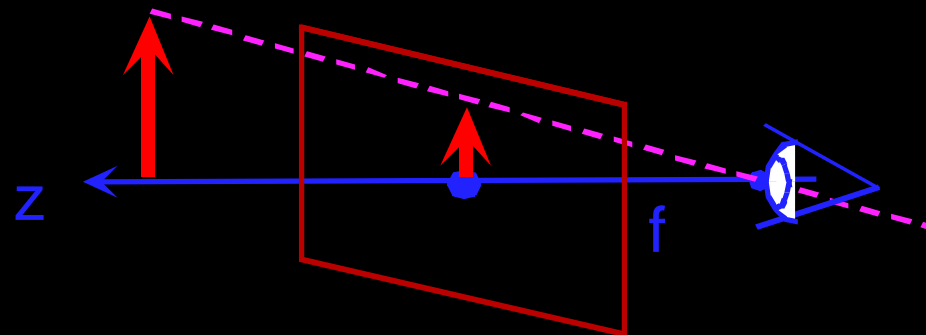


- **Optical axis**: the direction of imaging
- **Image plane**: a plane perpendicular to the optical axis
- **Center of Projection** (pinhole), focal point, viewpoint, , nodal point
- **Focal length**: distance from focal point to the image plane
- **FOV** : Field of View – viewing angles in horizontal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

- Consider case with object on the optical axis:



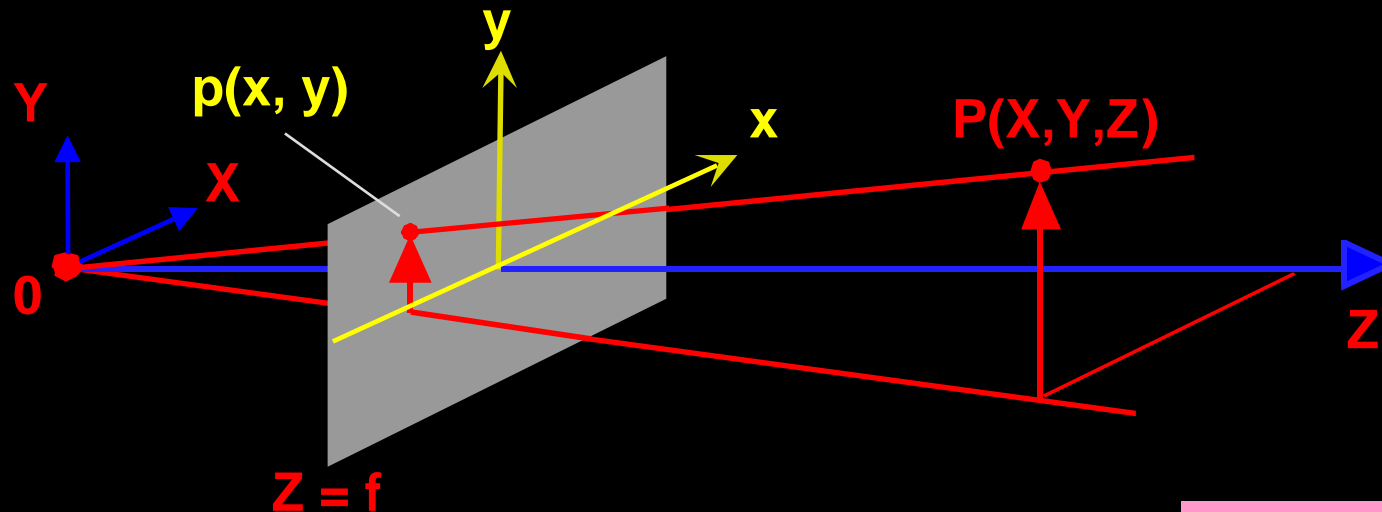
- More convenient with upright image:



Projection plane  $z = f$

- Equivalent mathematically

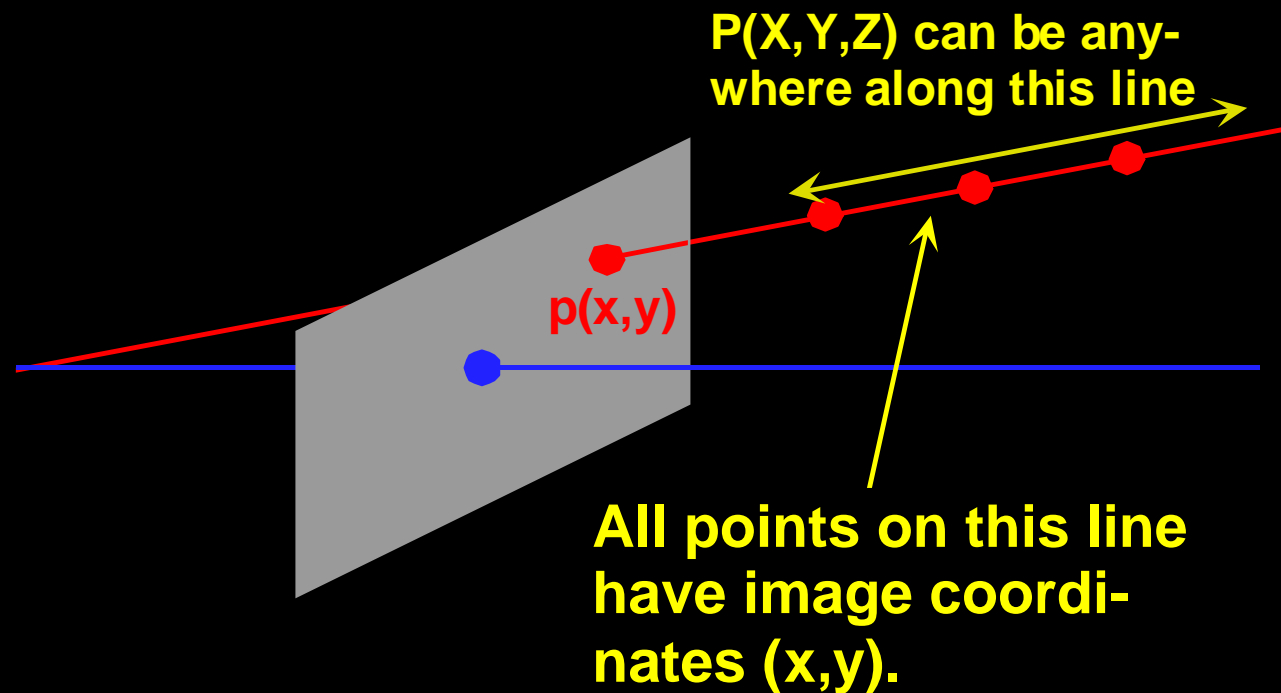
- Compute the image coordinates of  $p$  in terms of the world (camera) coordinates of  $P$ .



- 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

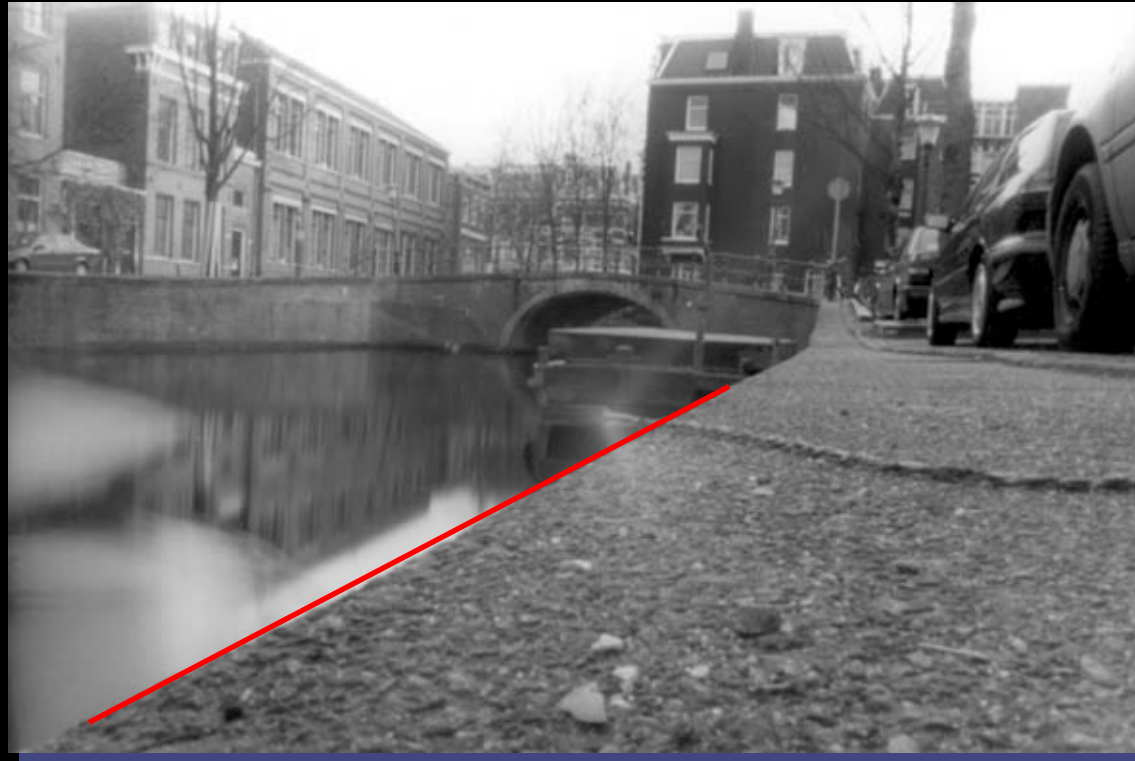


Photo by Robert Kosara, robert@kosara.net  
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## Amsterdam

- ✓ straight line
- × size
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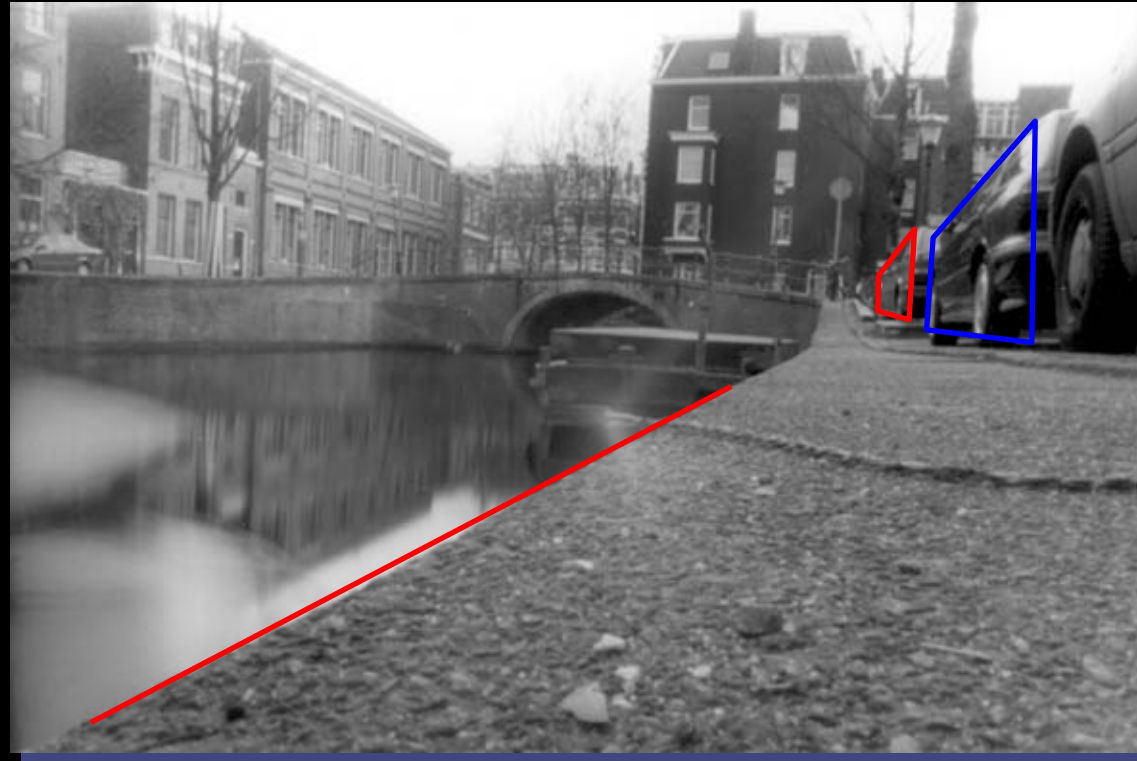


Photo by Robert Kosara, robert@kosara.net  
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## Amsterdam

- ✓ straight line
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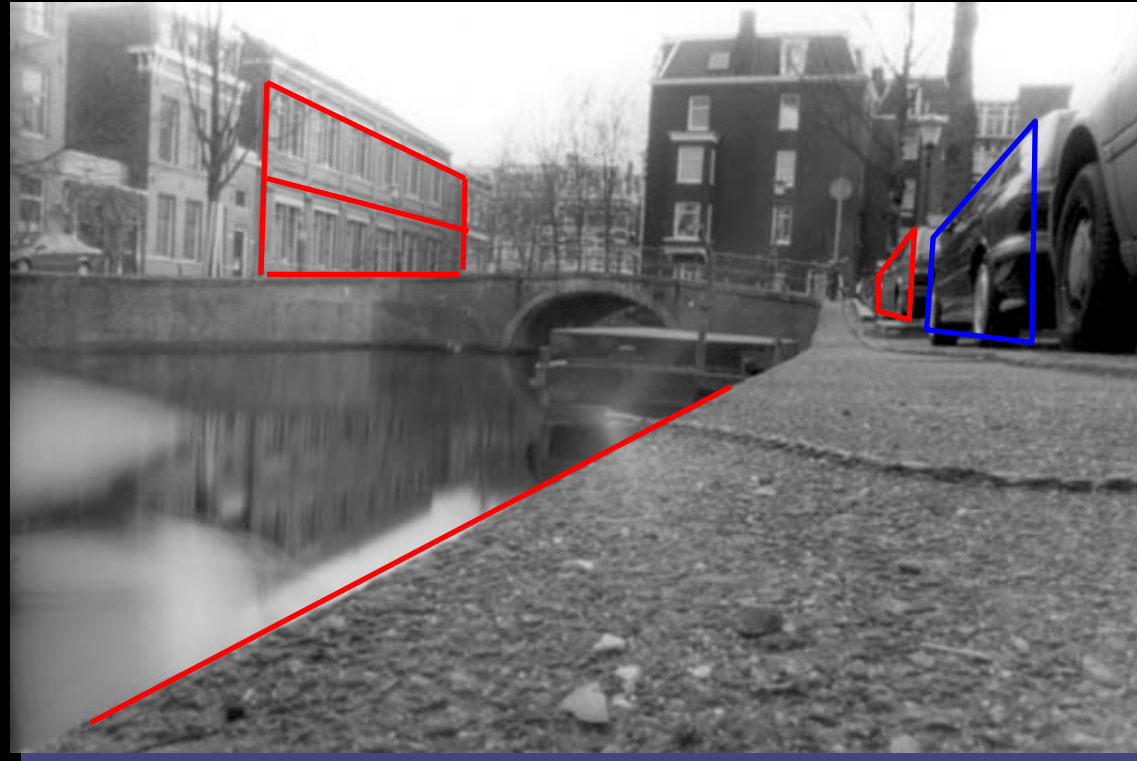


Photo by Robert Kosara, robert@kosara.net  
<http://www.kosara.net/gallery/pinholeamsterdam/pic01.html>



## Amsterdam

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

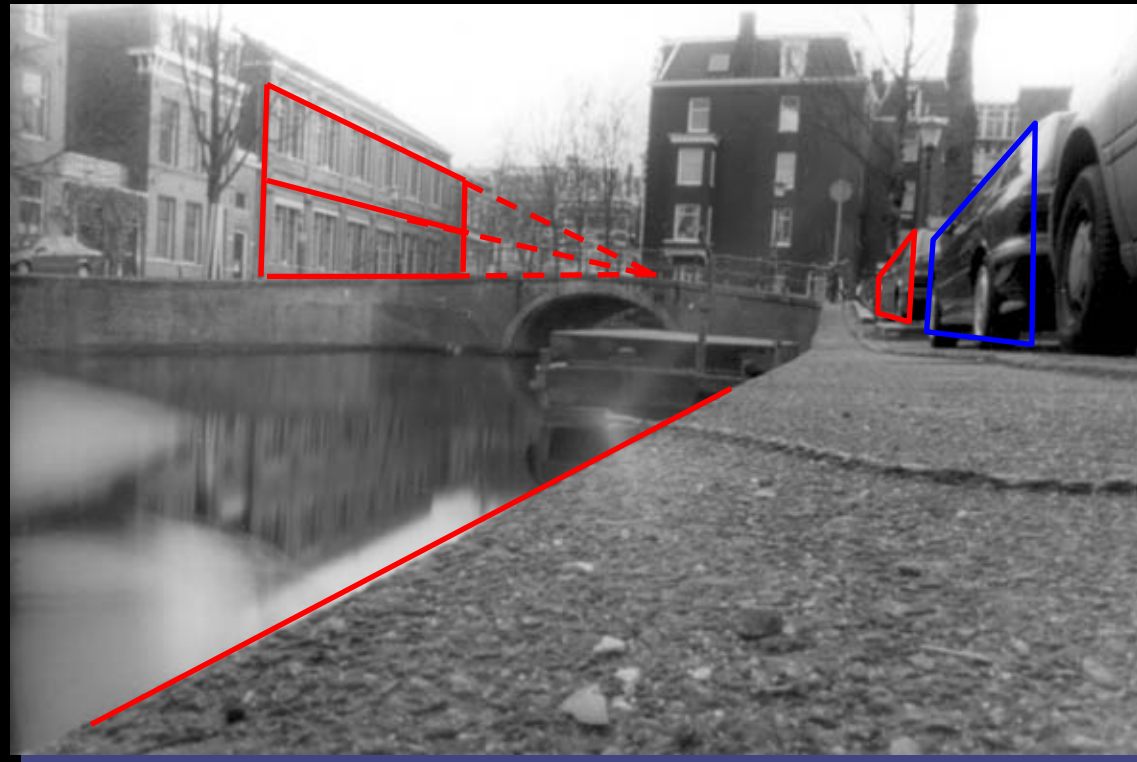


Photo by Robert Kosara, robert@kosara.net  
<http://www.kosara.net/gallery/pinholeamsterdam/pic01.html>

## Amsterdam

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

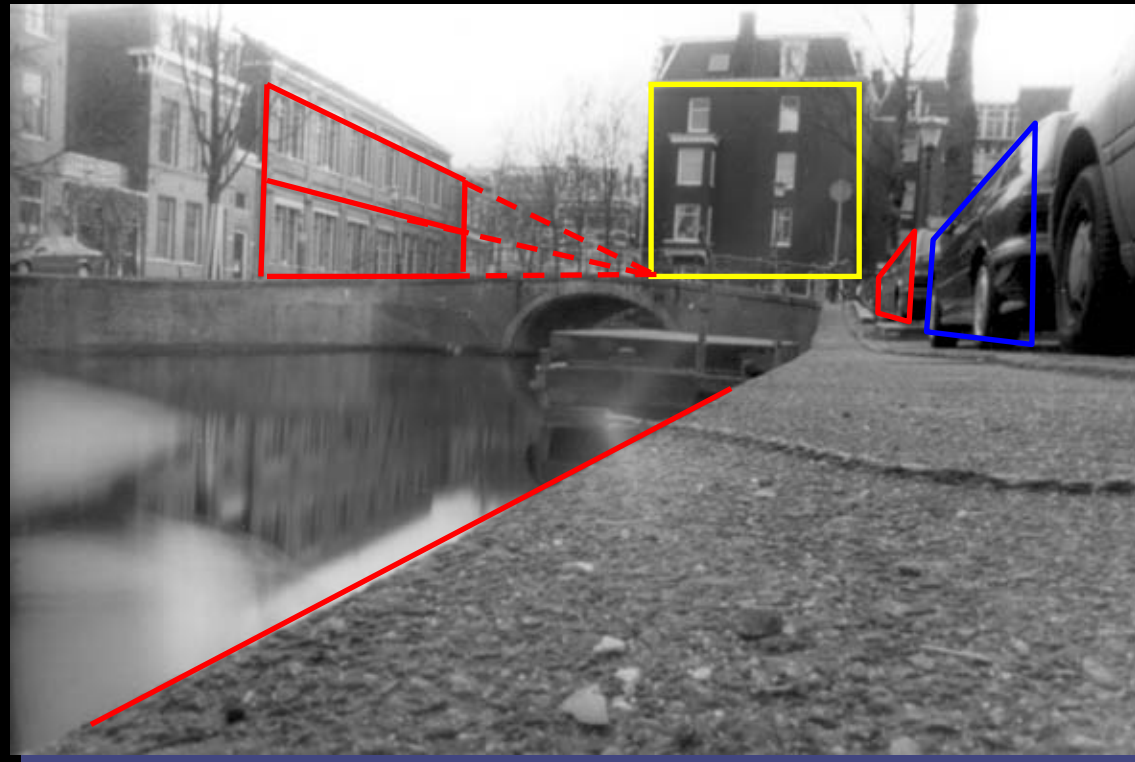
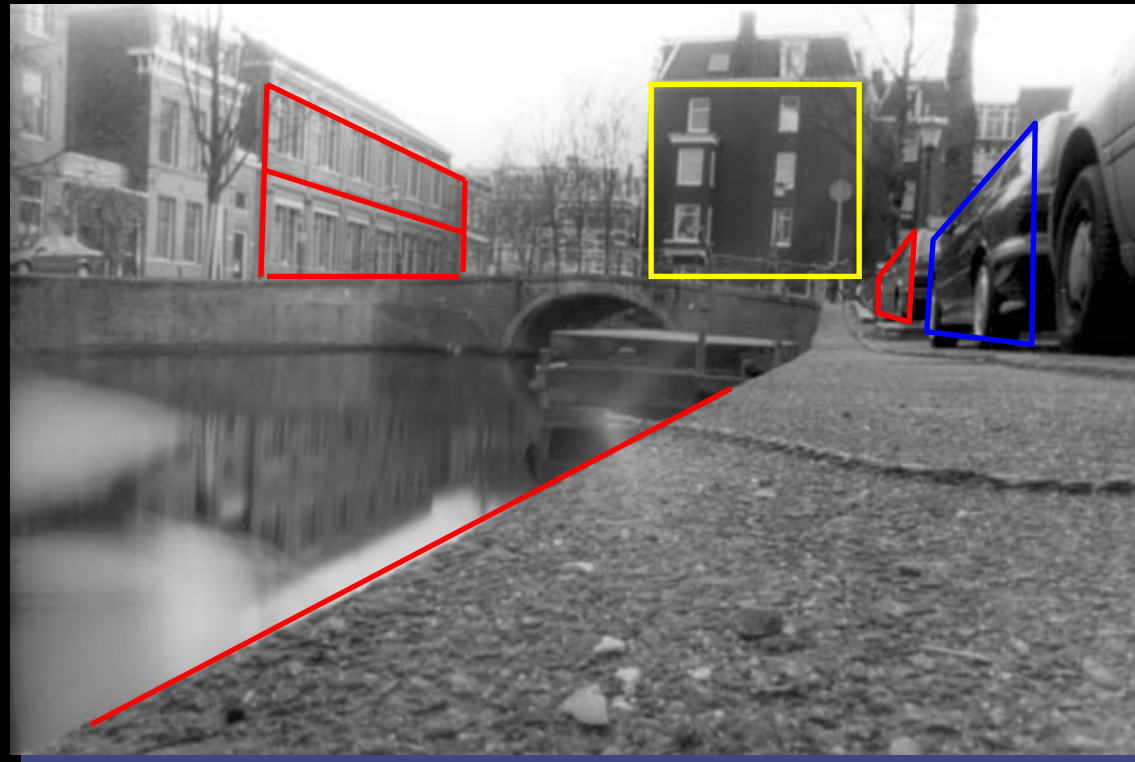


Photo by Robert Kosara, robert@kosara.net  
<http://www.kosara.net/gallery/pinholeamsterdam/pic01.html>

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

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



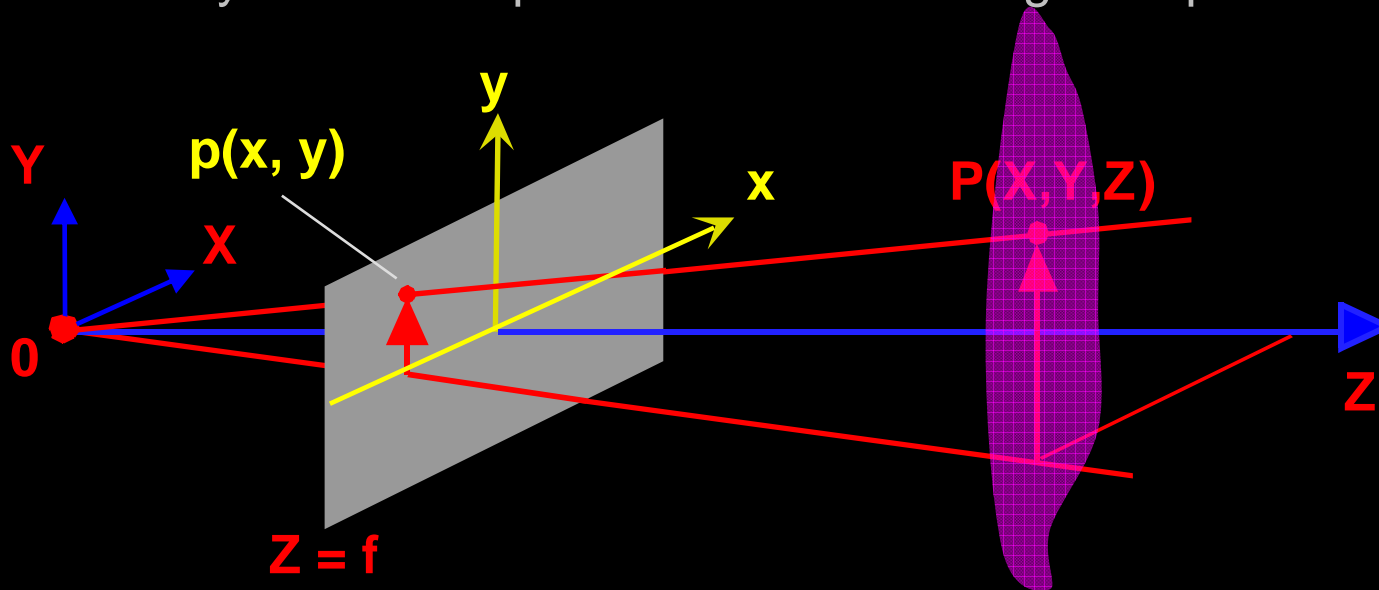
Markus Raetz, *Metamorphose II*, 1991-92, cast iron, 15 1/4 x 12 x 12 inches  
Fine Art Center University Gallery, Sep 15 – Oct 26

3D Computer Vision  
and Video Computing

It's real!



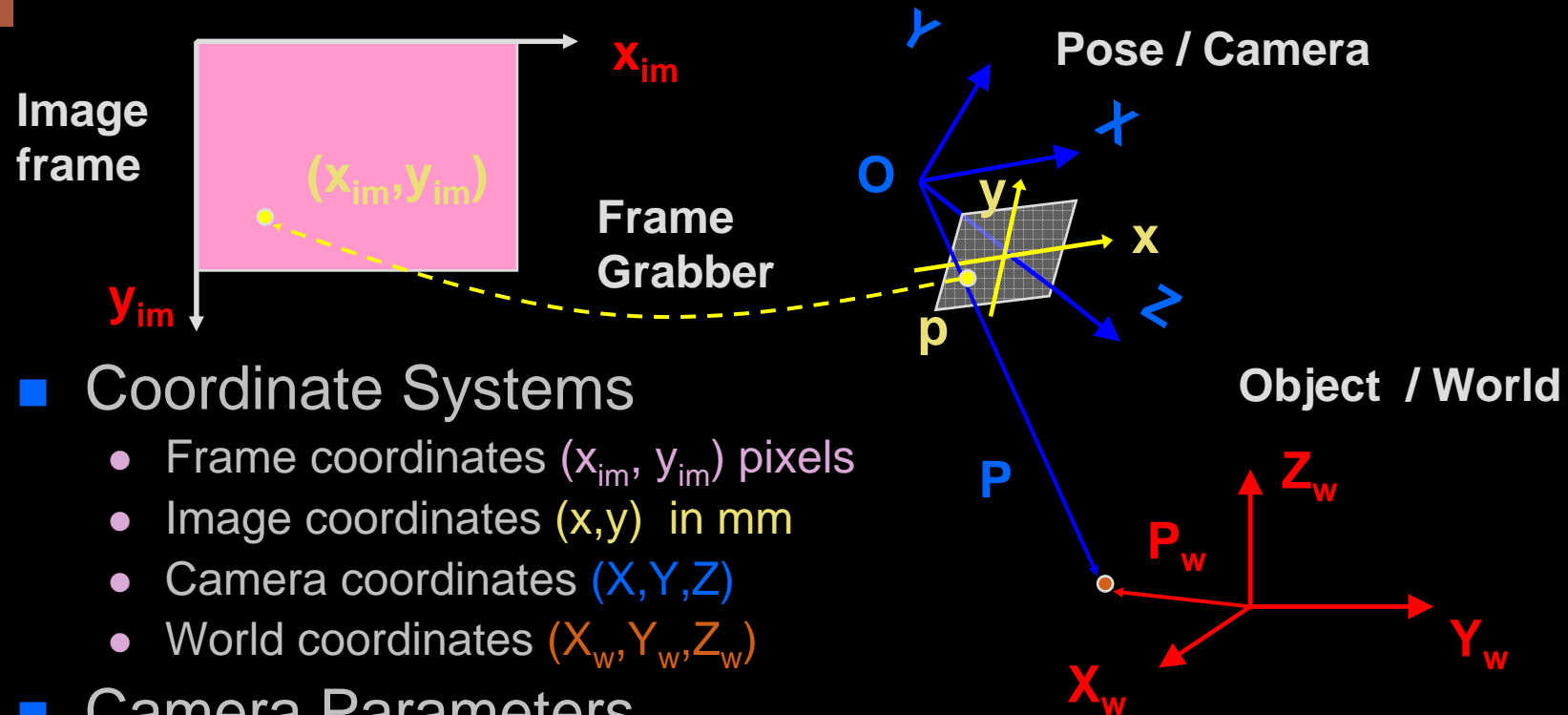
- Average depth  $\bar{Z}$  is much larger than the relative distance between any two scene points measured along the optical axis



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

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

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



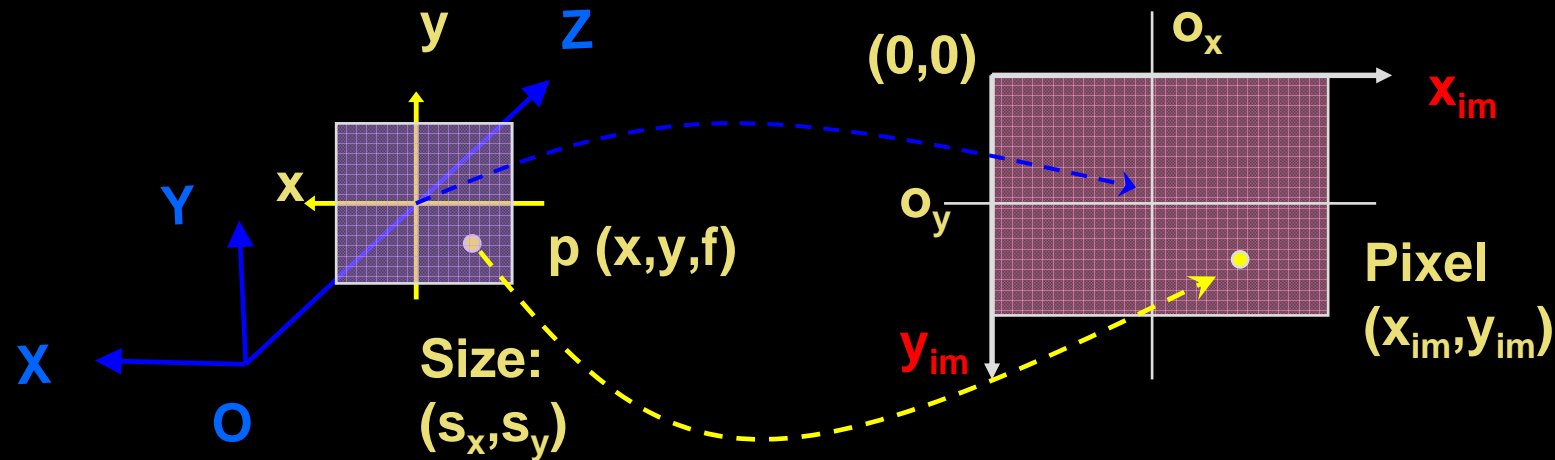
## ■ Coordinate Systems

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

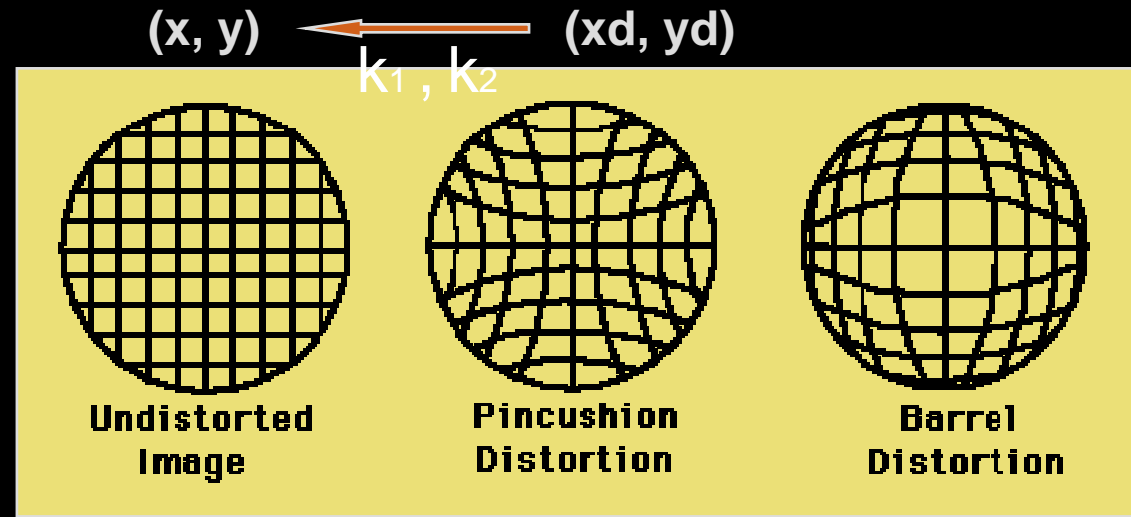
$$x = -(x_{im} - o_x) s_x$$

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

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

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

- Lens Distortions



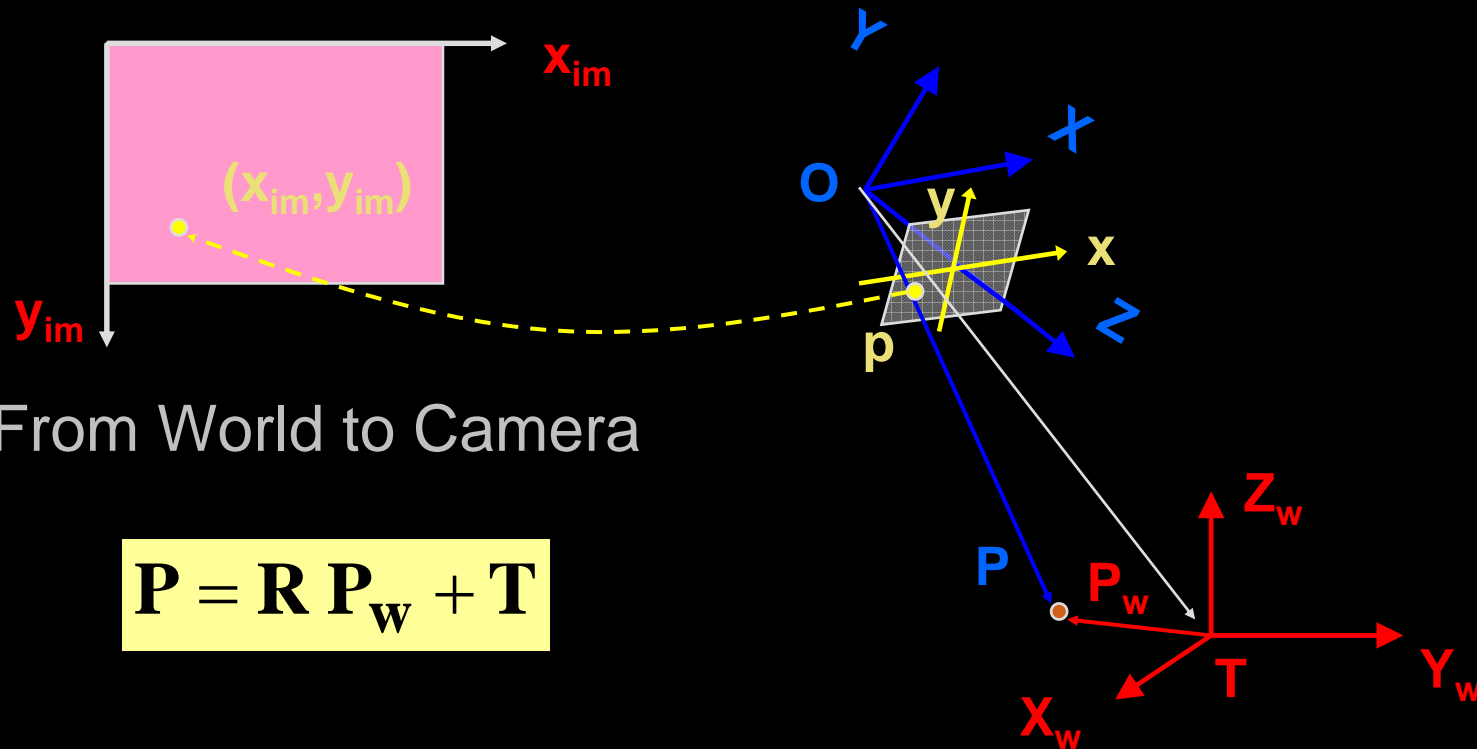
- Modeled as simple radial distortions

- $r^2 = x_d^2 + y_d^2$
- $(x_d, y_d)$  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

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

T: Transpose

- Scene point: 3D vector

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

- Translation: 3D vector

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

- Vector Operations

- Addition:

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

- Translation of a 3D vector

- Dot product ( a scalar):

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

- $\mathbf{a} \cdot \mathbf{b} = |\mathbf{a}| |\mathbf{b}| \cos \theta$

- Cross product (a vector)

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

- Generates a new vector that is orthogonal to both of them

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

## ■ 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) => 2+2 constraints (unit vectors) => 3 DOF ? (degrees of freedom)

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

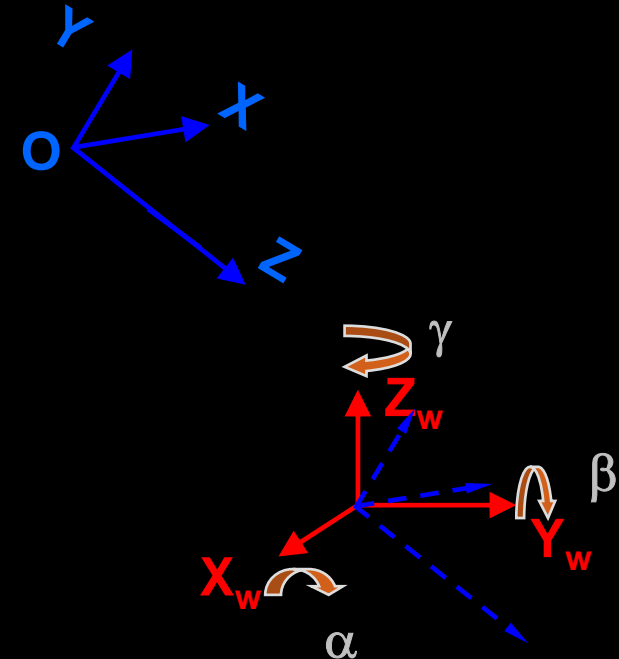
## ■ Matrix Operations

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

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



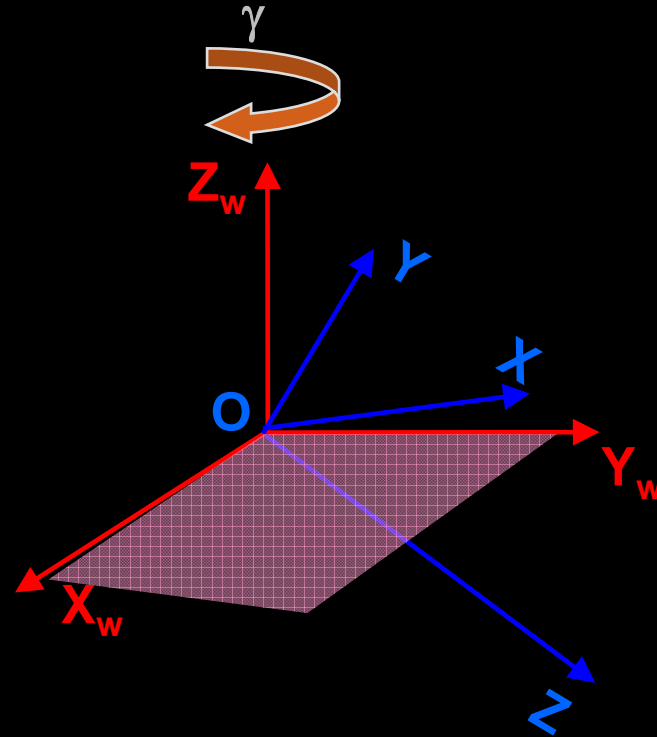
- Notes:
  - Only three rotations
  - Every time around one axis
  - Bring corresponding axes to each other
    - X<sub>w</sub> = X, Y<sub>w</sub> = Y, Z<sub>w</sub> = Z
  - First step (e.g.) Bring X<sub>w</sub> to X

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

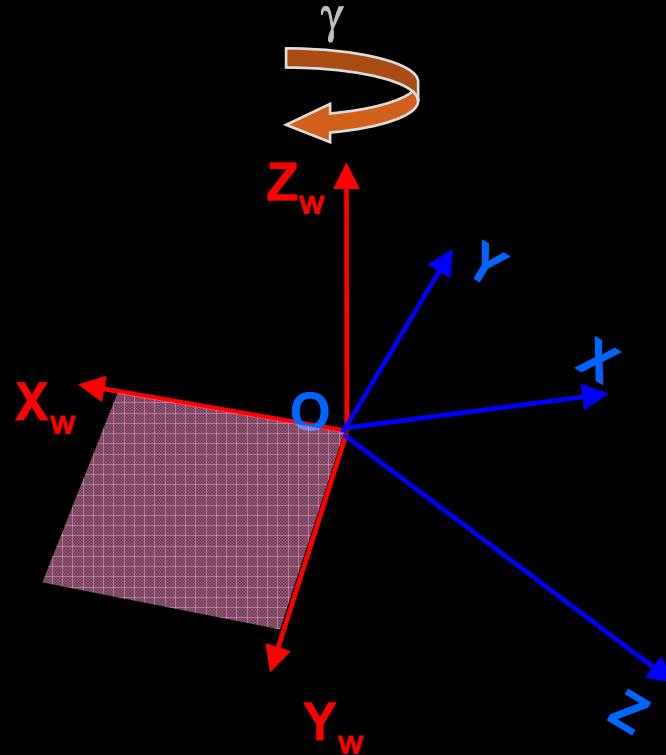
■ Rotation  $\gamma$  around the  $Z_w$  Axis

- 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  $Y_w$



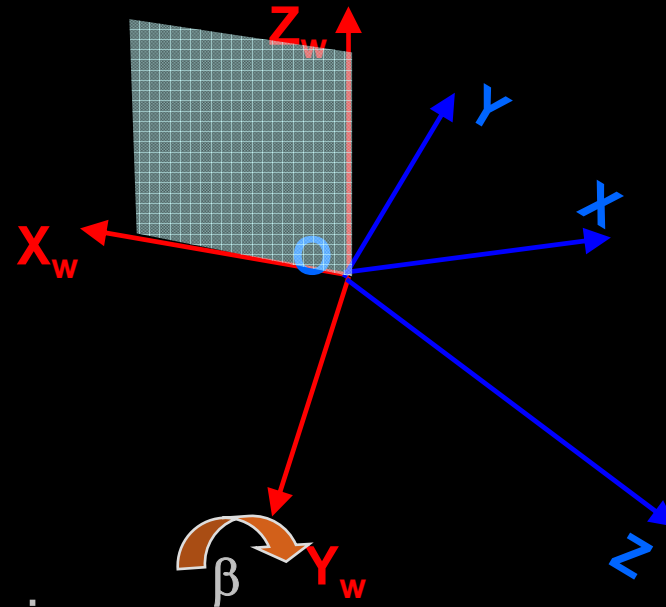
$$\mathbf{R}_\gamma = \begin{bmatrix} \cos \gamma & -\sin \gamma & 0 \\ \sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



- Rotation  $\gamma$  around the  $Z_w$  Axis
  - Rotate in  $X_w O Y_w$  plane so that
    - $Y_w \perp X \Rightarrow X$  in  $X_w O Z_w$  ( $\Leftarrow Y_w \perp X_w O Z_w$ )
    - $\Rightarrow Y_w$  in  $Y O Z$  ( $\Leftarrow X \perp Y O Z$ )
- $Z_w$  does not change

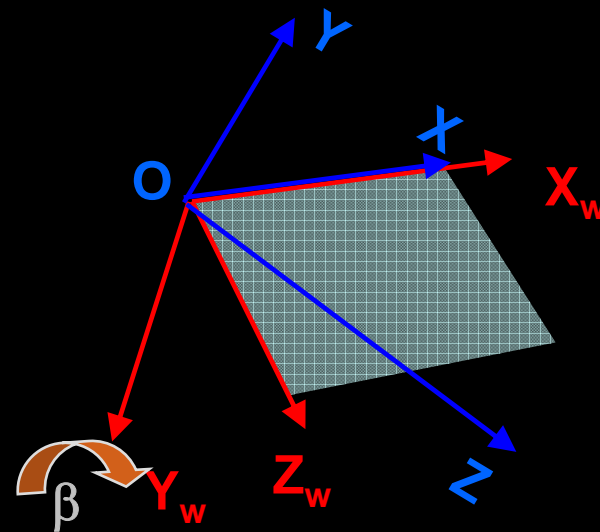


$$\mathbf{R}_\chi = \begin{bmatrix} \cos \beta & 0 & -\sin \beta \\ 0 & 1 & 0 \\ \sin \beta & 0 & \cos \beta \end{bmatrix}$$



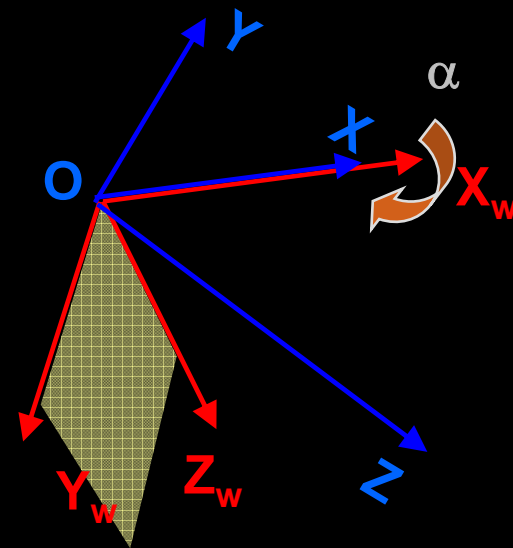
- Rotation  $\beta$  around the  $Y_w$  Axis
  - Rotate in  $X_w O Z_w$  plane so that
  - $X_w = X \Rightarrow Z_w$  in  $YOZ$  (&  $Y_w$  in  $YOZ$ )
- $Y_w$  does not change

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



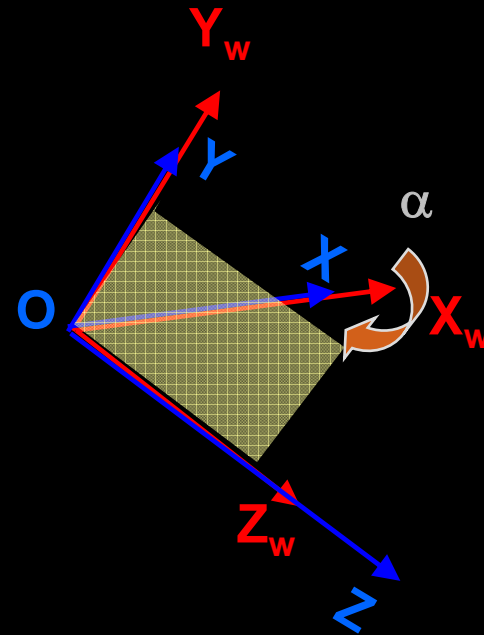
- Rotation  $\beta$  around the  $Y_w$  Axis
  - Rotate in  $X_w O Z_w$  plane so that
  - $X_w = X \Rightarrow Z_w$  in  $YOZ$  (&  $Y_w$  in  $YOZ$ )
- $Y_w$  does not change

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



- Rotation  $\alpha$  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  $\alpha$  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

Appendix A.9 of the textbook

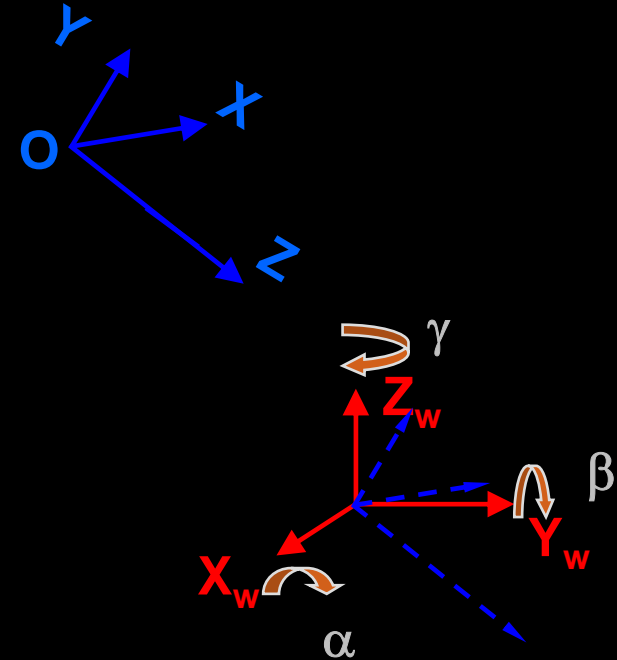
- Rotation around the Axes

- 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:  $\gamma, \beta, \alpha$
- Same  $\mathbf{R}$ , 6 different sets of  $\alpha, \beta, \gamma$
- $\mathbf{R}$  Non-linear function of  $\alpha, \beta, \gamma$**
- $\mathbf{R}$  is orthogonal**
- It's easy to compute angles from  $\mathbf{R}$**



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

## Appendix A.9 of the textbook

- According to Euler's Theorem, any 3D rotation can be described by a rotating angle,  $\theta$ , 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:  $\mathbf{P} = (X, Y, Z)^T$
- World:  $\mathbf{P}_w = (X_w, Y_w, Z_w)^T$
- Transform:  $\mathbf{R}, \mathbf{T}$

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

- Camera to Image

- Camera:  $\mathbf{P} = (X, Y, Z)^T$
- Image:  $\mathbf{p} = (x, y)^T$
- 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  $\delta Z$
- Define centroid vector  $\bar{\mathbf{P}}_w$
- 8 independent entries

$$\mathbf{Z} = \bar{\mathbf{Z}} = \mathbf{R}_3^T \bar{\mathbf{P}}_w + T_z$$

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

## ■ 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

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

- 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

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

## ■ 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

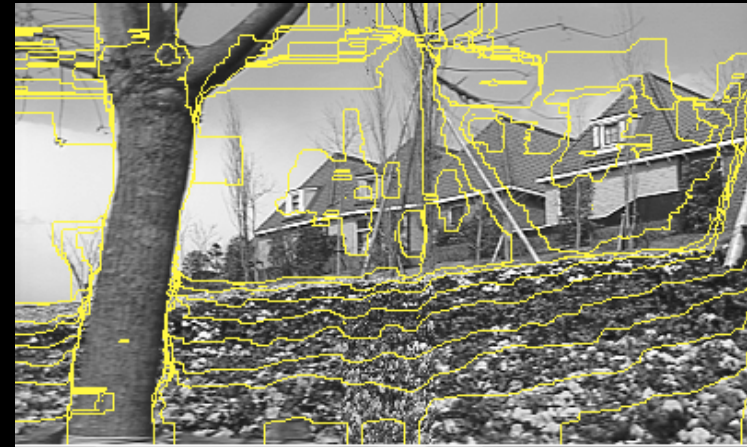
- Find intrinsic & extrinsic parameters
- Given image-world point pairs
- Probably a partially solved problem ?
- 11 independent entries
  - <-> 10 parameters:  $f_x, f_y, o_x, o_y, \alpha, \beta, \gamma, T_x, T_y, T_z$

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

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

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

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

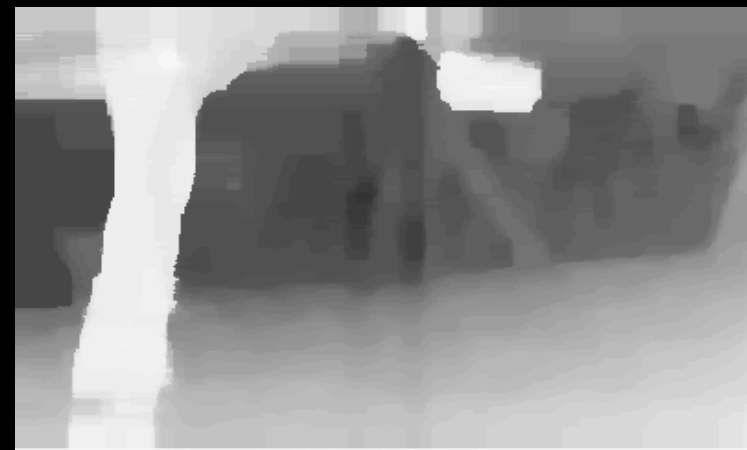


(1) Panoramic texture map

## Flower Garden Sequence

### Vision:

- Camera Calibration
- Motion Estimation
- 3D reconstruction

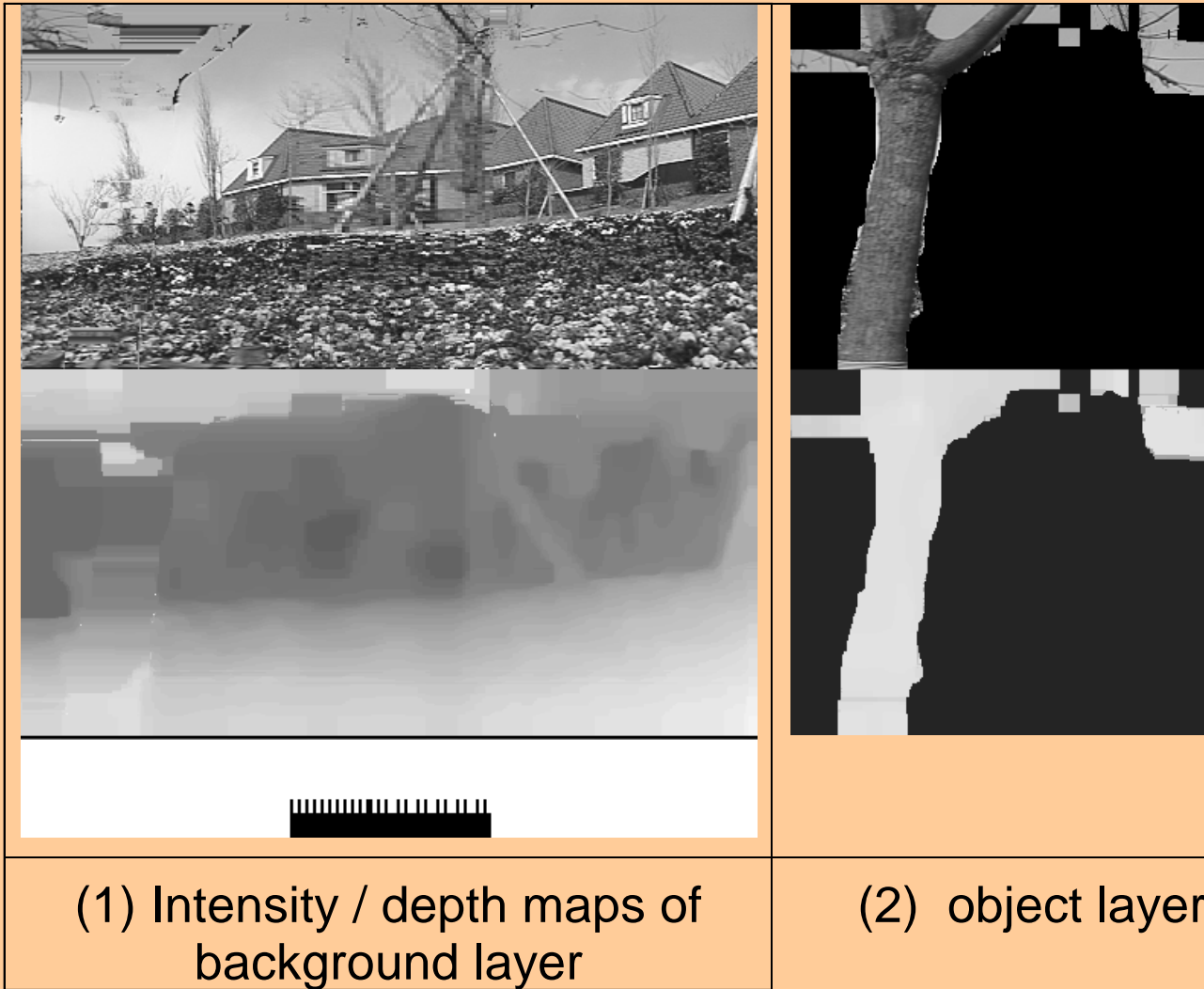


(2) panoramic depth map



# 3D Computer Vision

## Image-based 3D model of the FG sequence





## The Layered Panorama

The Layered Representation of the garden consists of two layers:

1. Background --- The meadow, house and the sky.
2. Foreground — The tree.

### Graphics:

- Virtual Camera
- Synthetic motion
- From 3D to 2D image



- 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)

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



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

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