

# CE7453: Photogrammetric Computer Vision

## Lecture 5

Central Perspective Model  
Homogenous Coordinate System  
Camera Matrix, Projection Matrix

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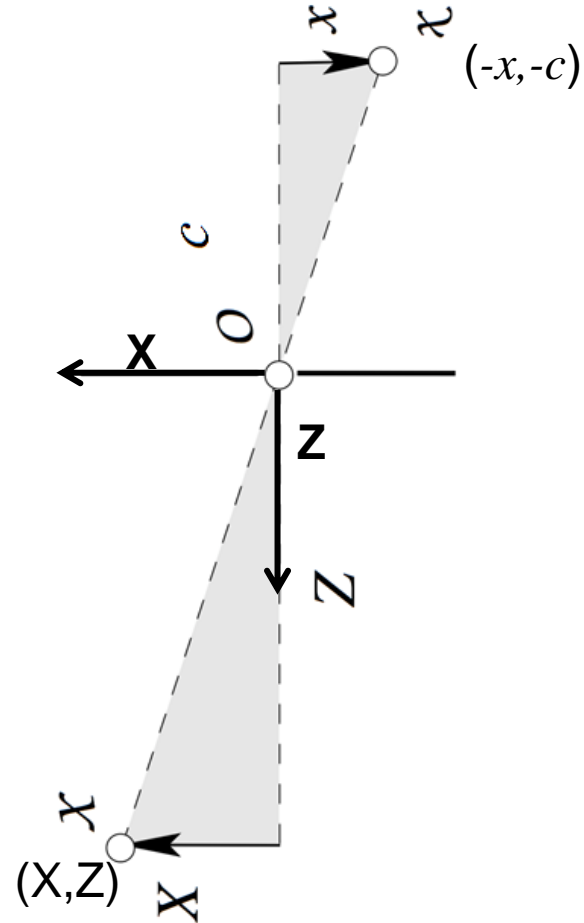
# Central Perspective Model

$$\frac{c}{Z} = \frac{x}{X} = m$$

$$x = \frac{c}{Z} X = mX$$

$c$  : camera constant

$m$  : map scale

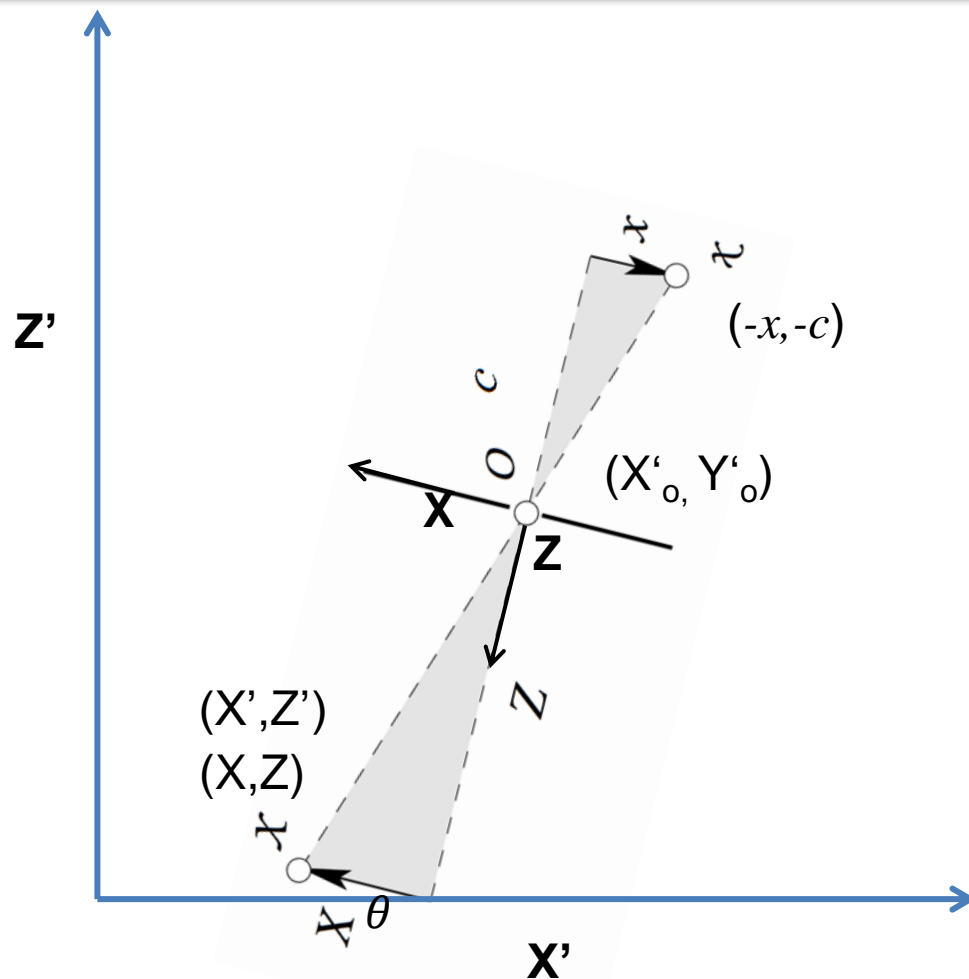


# Central Perspective Model

$$\begin{bmatrix} X \\ Z \end{bmatrix} = R \begin{bmatrix} X' - X'_o \\ Z' - Z'_o \end{bmatrix}$$

$$R = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$$

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



**X, Z: Camera Frame**

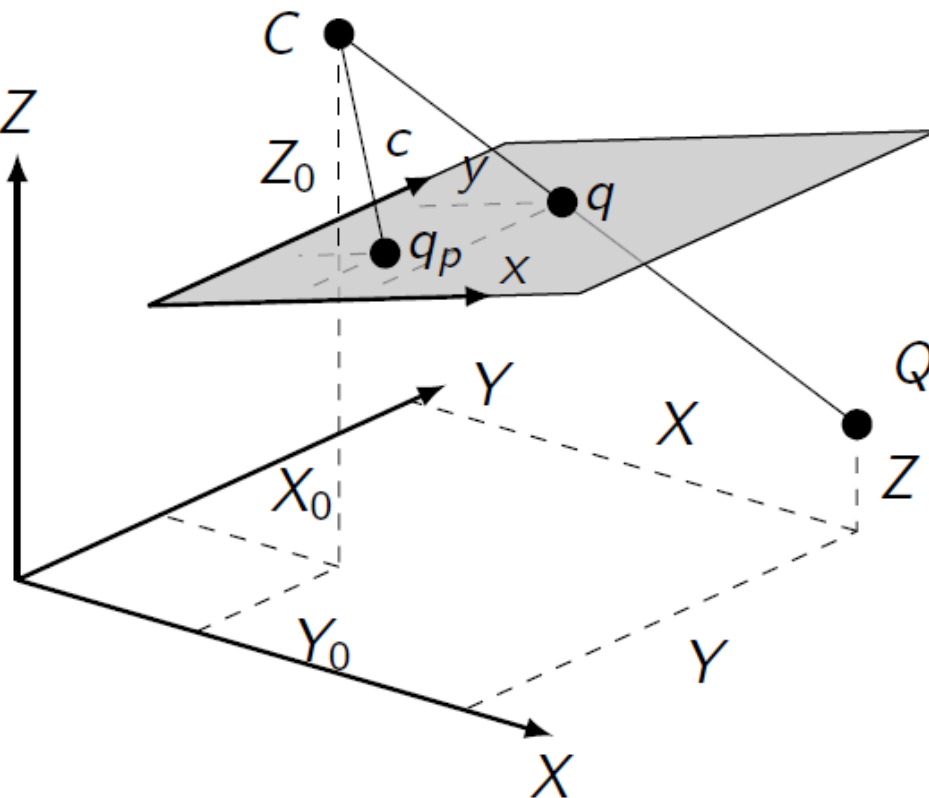
**X', Y': World Frame / Geodetic Frame**

# Co-linearity Equation – Cont.

The *collinearity equations*

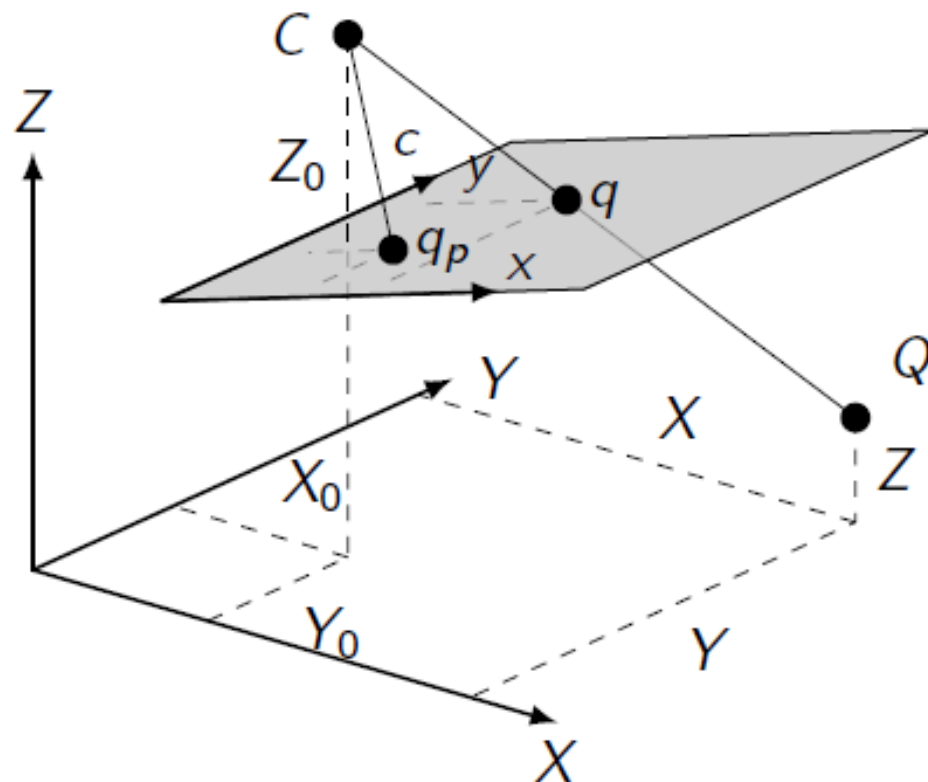
$$\begin{pmatrix} x - x_p \\ y - y_p \\ -c \end{pmatrix} = kR \begin{pmatrix} X - X_0 \\ Y - Y_0 \\ Z - Z_0 \end{pmatrix}$$

describe the relationship between the object point  $(X, Y, Z)^T$ , the position  $C = (X_0, Y_0, Z_0)^T$  of the camera center and the orientation  $R$  of the camera.



# Co-linearity Equation – Cont.

- The distance  $c$  is known as the *principal distance* or *camera constant*.
- The point  $q_p = (x_p, y_p)^T$  is called the *principal point*.
- The ray passing through the camera center  $C$  and the principal point  $q_p$  is called the *principal ray*.



# Co-linearity Equation – Cont.

■ From

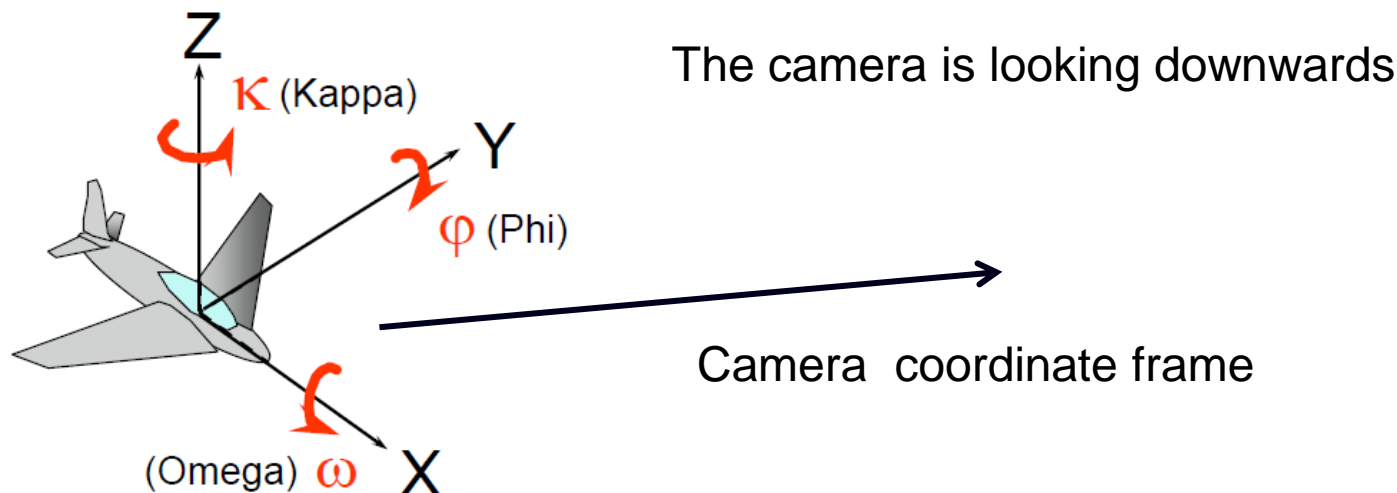
$$\begin{pmatrix} x - x_p \\ y - y_p \\ -c \end{pmatrix} = kR \begin{pmatrix} X - X_0 \\ Y - Y_0 \\ Z - Z_0 \end{pmatrix}, \text{ and } R = \begin{pmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{pmatrix},$$

We can get rid of  $k$  by ratioing the first and the third, the second and the third equation

$$x = x_p - c \frac{r_{11}(X - X_0) + r_{12}(Y - Y_0) + r_{13}(Z - Z_0)}{r_{31}(X - X_0) + r_{32}(Y - Y_0) + r_{33}(Z - Z_0)},$$
$$y = y_p - c \frac{r_{21}(X - X_0) + r_{22}(Y - Y_0) + r_{23}(Z - Z_0)}{r_{31}(X - X_0) + r_{32}(Y - Y_0) + r_{33}(Z - Z_0)}.$$

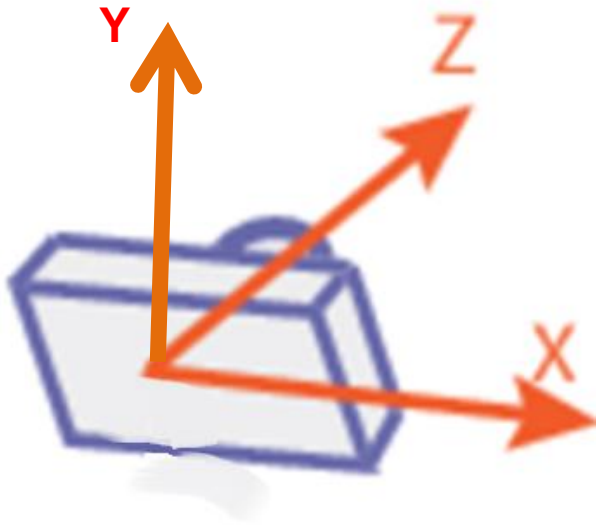
# Rotation Matrix

**angles  $\kappa$ ,  $\varphi$ ,  $\omega$**



- Right handed coordinate system
- Left handed coordinate system (invert Z direction)

# Rotation Matrix - Cont.



## Computer vision Convention

- Camera coordinate frame
- Left handed system – Z in a different direction as previous one



# Rotation Matrix - Cont.

$$R_x(\omega) \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos(\omega) & \sin(\omega) \\ 0 & -\sin(\omega) & \cos(\omega) \end{bmatrix}$$

$$R_y(\varphi) = \begin{bmatrix} \cos(\varphi) & 0 & -\sin(\varphi) \\ 0 & 1 & 0 \\ \sin(\varphi) & 0 & \cos(\varphi) \end{bmatrix}$$

$$R_z(\kappa) = \begin{bmatrix} \cos(\kappa) & \sin(\kappa) & 0 \\ -\sin(\kappa) & \cos(\kappa) & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

# Rotation Matrix - Cont.

$$R = R_z R_y R_x = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$$

Would changing the sequent result in a different matrix?

# Rotation Matrix - Cont.

$$\begin{pmatrix} \cos \phi \cos \kappa & \sin \omega \sin \phi \cos \kappa + \cos \omega \sin \kappa & -\cos \omega \sin \phi \cos \kappa + \sin \omega \sin \kappa \\ -\cos \phi \sin \kappa & -\sin \omega \sin \phi \sin \kappa + \cos \omega \cos \kappa & \cos \omega \sin \phi \sin \kappa + \sin \omega \cos \kappa \\ \sin \phi & -\sin \omega \cos \phi & \cos \omega \cos \phi \end{pmatrix}$$

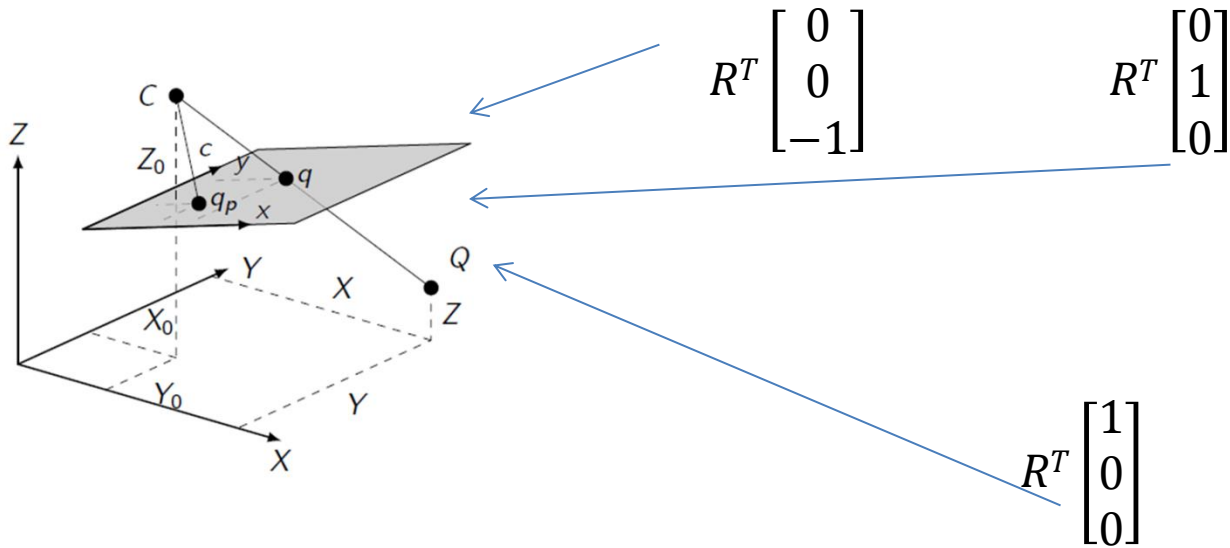
**Question: given a rotation matrix, how to compute the angles? –  
Any problem you foresee?**

$\omega = \text{atan}(-r_{32}/r_{33}); \quad \kappa = \text{atan}(-r_{21}/r_{11}); \quad \phi = \text{asin}(r_{13});$

# Rotation Matrix - Cont.

## ● Geometric meaning of Rotation Matrix

$$\begin{bmatrix} x - xp \\ y - yp \\ -c \end{bmatrix} = kR \begin{bmatrix} X - X_0 \\ Y - Y_0 \\ Z - Z_0 \end{bmatrix} \quad \longrightarrow \quad R^T \begin{bmatrix} x - xp \\ y - yp \\ -c \end{bmatrix} = k \begin{bmatrix} X - X_0 \\ Y - Y_0 \\ Z - Z_0 \end{bmatrix}$$



## Where are they in the figure?

# Rotation Matrix – Quaternion.

- Representing the rotation using a rotating vector  $W=(w_1, w_2, w_3)$  and an angle  $\theta$ .

*Rodrigues' rotation formula: the rotation then being*

$$R_w(\theta) = I_{3 \times 3} + \sin(\theta) \times S + [1 - \cos(\theta)] \times S^2$$

where

$$S = \begin{bmatrix} 0 & -w_3 & w_2 \\ w_3 & 0 & -w_1 \\ -w_2 & w_1 & 0 \end{bmatrix}$$

Full proof see here:

[https://en.wikipedia.org/wiki/Rodrigues%27\\_rotation\\_formula](https://en.wikipedia.org/wiki/Rodrigues%27_rotation_formula)

# Rotation Matrix – Quaternion – Cont.

- A rotation matrix can be represented by four parameters,  $\theta, w_1, w_2, w_3$ , where  $[w_1, w_2, w_3]$  is a unit vector
- Let's do a little bit of mathematical trick here:
- $q_1 = \sin(\theta/2)w_1, q_2 = \sin(\theta/2)w_2,$   
 $q_3 = \sin(\theta/2)w_3,$
- Then

$$\begin{aligned}q_1 &= w_1 \sin(\theta/2) \\ q_2 &= w_2 \sin(\theta/2) \\ q_3 &= w_3 \sin(\theta/2)\end{aligned}$$

Do a little bit of math by replacing these elements back to the previous *Rodrigues equation*

# Rotation Matrix – Quaternion – Cont.

Note:

$$\sin(\theta) = 2 \sin\left(\frac{\theta}{2}\right) \cos\left(\frac{\theta}{2}\right),$$

$$\cos(\theta) = \cos^2\left(\frac{\theta}{2}\right) - \sin^2\left(\frac{\theta}{2}\right)$$

$$\begin{aligned} S^2 &= S \times S \\ &= \begin{bmatrix} -w_3^2 - w_2^2 & w_1 w_2 & w_1 w_3 \\ w_1 w_2 & -w_3^2 - w_1^2 & w_2 w_3 \\ w_1 w_3 & w_2 w_3 & -w_2^2 - w_1^2 \end{bmatrix} \end{aligned}$$

# Rotation Matrix – Quaternion – Cont.

$$R_w(\theta) = I_{3 \times 3} + \sin(\theta) \times S + [1 - \cos(\theta)] \times S^2$$

$$S \times \sin(\theta) = 2 \sin\left(\frac{\theta}{2}\right) \cos\left(\frac{\theta}{2}\right) = 2 \sin\left(\frac{\theta}{2}\right) \cos\left(\frac{\theta}{2}\right) \begin{bmatrix} 0 & -q_3 & q_2 \\ q_3 & 0 & -q_1 \\ -q_2 & q_1 & 0 \end{bmatrix} / \sin\left(\frac{\theta}{2}\right)$$

$$\begin{aligned} S^2 \times [1 - \cos(\theta)] &= S^2 \left[ 1 - \cos^2\left(\frac{\theta}{2}\right) + \sin^2\left(\frac{\theta}{2}\right) \right] \\ &= 2 \sin^2\left(\frac{\theta}{2}\right) \begin{bmatrix} -w_3^2 - w_2^2 & w_1 w_2 & w_1 w_3 \\ w_1 w_2 & -w_3^2 - w_1^2 & w_2 w_3 \\ w_1 w_3 & w_2 w_3 & -w_2^2 - w_1^2 \end{bmatrix} \\ &= 2 \begin{bmatrix} -q_3^2 - q_2^2 & q_1 q_2 & q_1 q_3 \\ q_1 q_2 & -q_3^2 - q_1^2 & q_2 q_3 \\ q_1 q_3 & q_2 q_3 & -q_2^2 - q_1^2 \end{bmatrix} \end{aligned}$$



# Rotation Matrix – Quaternion – Cont.

- Then

$$R_w(\theta) = I + 2\cos\left(\frac{\theta}{2}\right) \begin{bmatrix} 0 & -q_3 & q_2 \\ q_3 & 0 & -q_1 \\ -q_2 & q_1 & 0 \end{bmatrix} +$$
$$2 \begin{bmatrix} -q_3^2 - q_2^2 & q_1q_2 & q_1q_3 \\ q_1q_2 & -q_3^2 - q_1^2 & q_2q_3 \\ q_1q_3 & q_2q_3 & -q_2^2 - q_1^2 \end{bmatrix}$$

$$\text{Let } q_0 = \cos\left(\frac{\theta}{2}\right)$$

# Rotation Matrix – Quaternion – Cont.

- Then

$$R_w(\theta) = \begin{bmatrix} 1 - 2q_3^2 - 2q_2^2 & 2q_1q_2 - 2q_0q_3 & 2q_1q_3 + 2q_0q_2 \\ 2q_1q_2 + 2q_0q_3 & 1 - 2q_3^2 - 2q_1^2 & 2q_2q_3 - 2q_0q_1 \\ 2q_1q_3 - 2q_0q_2 & 2q_2q_3 + 2q_0q_1 & 1 - 2q_2^2 - 2q_1^2 \end{bmatrix}$$

Where  $q_0 = \cos(\frac{\theta}{2})$ ,  $\mathbf{q} = [w_1, w_2, w_3] \sin(\frac{\theta}{2})$

We have  $q_0^2 + q_1^2 + q_2^2 + q_3^2 = 1$

This is just another parameterization of your rotation Matrix! – **Nothing special, you can even use all the elements in the matrix as your parameters**

# Rotation Matrix – Quaternion – Cont.

- Given a rotation matrix, how to get  $\mathbf{n} = [w_1, w_2, w_3]$  and  $\theta$ ?
- Tip 1: take the trace of the previous matrix to get  $\theta$ .
- Tip 2: make combinations of the elements to get  $\mathbf{n}$

$$\theta = \cos^{-1} \left( \frac{\text{trace}(\mathbf{R}) - 1}{2} \right), \hat{\mathbf{n}} = \frac{1}{2 \sin \theta} \begin{pmatrix} r_{32} - r_{23} \\ r_{13} - r_{31} \\ r_{21} - r_{12} \end{pmatrix}$$

# Rotation Matrix – Quaternion – Cont.

- Advantage, easy to formulate when given the rotation axis and angle, this is very common
  - Have vector  $v_1$ , want to rotate to  $v_2$
  - Need rotation vector  $\hat{r}$ , angle  $\theta$

$$\theta = \arccos(\hat{\mathbf{v}}_1 \bullet \hat{\mathbf{v}}_2)$$

$$\mathbf{r} = \mathbf{v}_1 \times \mathbf{v}_2$$

$$\mathbf{n} = \mathbf{r}/|\mathbf{r}|$$

$$q_0 = \cos\left(\frac{\theta}{2}\right), \mathbf{q} = \mathbf{n} \sin\left(\frac{\theta}{2}\right), \text{ Plug back}$$

# Rotation Matrix – Comparison.

Euler	Quaternion
<p>Advantage:</p> <ul style="list-style-type: none"><li>Minimal representation (3 parameters)</li><li>Easy interpretation</li></ul> <p>Disadvantages:</p> <ul style="list-style-type: none"><li>Many “alternative” Euler representations exist (XYZ, ZXZ, ZYX, ...)</li><li>Difficult to concatenate</li><li>Singularities (gimbal lock)</li></ul>	<p>Advantage:</p> <ul style="list-style-type: none"><li>Easy to represent rotating vectors</li><li>Inverse = easy to compute</li></ul> <p>Disadvantages:</p> <ul style="list-style-type: none"><li>One over-parameterization</li></ul>

# 3D to 2D relationship

$$x = x_p - c \frac{r_{11}(X - X_0) + r_{12}(Y - Y_0) + r_{13}(Z - Z_0)}{r_{31}(X - X_0) + r_{32}(Y - Y_0) + r_{33}(Z - Z_0)},$$
$$y = y_p - c \frac{r_{21}(X - X_0) + r_{22}(Y - Y_0) + r_{23}(Z - Z_0)}{r_{31}(X - X_0) + r_{32}(Y - Y_0) + r_{33}(Z - Z_0)}.$$

$r_{ij}$  can be parameterized by  $\omega, \varphi, \kappa$  or  $q_0, q_1, q_2, q_3$

Given any ground points  $X, Y, Z$ , to get its position in the image, you need to know,  $r_{ij}$ ,  $X_0, Y_0, Z_0$  of this image, and the principal points  $x_p$  and  $y_p$  ;  
 $x, y$ : image position in the actual films.

# 3D to 2D relationship

- Then you need to know pixel size, in order to navigate back to its pixel location:

$$x_{pix} = \frac{x}{psz_x}$$
$$y_{pix} = \frac{y}{psz_y} \text{ or } imgheight - \frac{y}{psz_y}$$

$psz_x$  and  $psz_y$ : pixel size of one cell in CCD

# 3D to 2D relationship

- Recall

$$\begin{bmatrix} x - xp \\ y - yp \\ -c \end{bmatrix} = kR \begin{bmatrix} X - X_0 \\ Y - Y_0 \\ Z - Z_0 \end{bmatrix}$$
$$\begin{bmatrix} x_{pix} \\ y_{pix} \\ -c \end{bmatrix} = \begin{bmatrix} \frac{1}{psz_x} & \frac{(-c)(psz_x)}{yp} \\ \frac{1}{psz_y} & \frac{(-c)(psz_y)}{xp} \\ 1 & \end{bmatrix} \begin{bmatrix} x - xp \\ y - yp \\ -c \end{bmatrix}$$



# 3D to 2D relationship

$$\begin{aligned} \begin{bmatrix} x_{pix} \\ y_{pix} \\ -c \end{bmatrix} &= \begin{bmatrix} \frac{1}{psz_x} & \frac{xp}{(-c)(psz_x)} \\ \frac{1}{psz_y} & \frac{yp}{(-c)(psz_y)} \\ & 1 \end{bmatrix} \begin{bmatrix} x - xp \\ y - yp \\ -c \end{bmatrix} \\ &= \begin{bmatrix} \frac{-c}{psz_x} & \frac{xp}{(psz_x)} \\ \frac{-c}{psz_y} & \frac{yp}{(psz_y)} \\ & -c \end{bmatrix} \begin{bmatrix} (x - xp)/(-c) \\ (y - yp)/(-c) \\ 1 \end{bmatrix} \end{aligned}$$

# 3D to 2D relationship – Camera matrix

$$\begin{bmatrix} x_{pix} \\ y_{pix} \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{-c}{psz_x} & \frac{xp}{(psz_x)} \\ \frac{-c}{psz_y} & \frac{yp}{(psz_y)} \\ 1 & \end{bmatrix} \begin{bmatrix} (x - xp)/(-c) \\ (y - yp)/(-c) \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} x_{pix} \\ y_{pix} \\ 1 \end{bmatrix} = \mathbf{K} \begin{bmatrix} (x - xp)/(-c) \\ (y - yp)/(-c) \\ 1 \end{bmatrix} = \mathbf{K} \mathbf{k} \mathbf{R} \begin{bmatrix} X - X_0 \\ Y - Y_0 \\ Z - Z_0 \end{bmatrix} / (-c)$$
$$= \mathbf{K} \lambda \mathbf{R} \begin{bmatrix} X - X_0 \\ Y - Y_0 \\ Z - Z_0 \end{bmatrix}$$

$\mathbf{K}$  is called camera matrix in computer vision

# 3D to 2D relationship – Projection Matrix

- Let's make it look even more compact:

$$\begin{bmatrix} x_{pix} \\ y_{pix} \\ 1 \end{bmatrix} = \lambda \mathbf{K} \begin{bmatrix} R & -RT \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}, \mathbf{T} = \begin{bmatrix} X_0 \\ Y_0 \\ Z_0 \end{bmatrix}$$

Let  $\mathbf{P} = \lambda \mathbf{K} \begin{bmatrix} R & -RT \end{bmatrix}$ , then  $\mathbf{P}$  is called Projection Matrix.

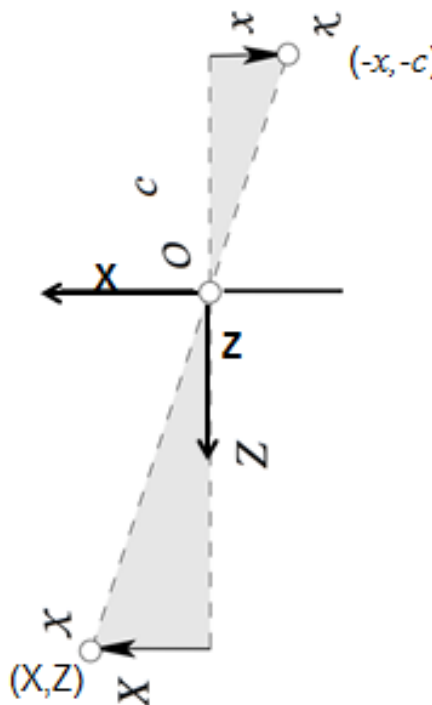
$$\begin{bmatrix} x_{pix} \\ y_{pix} \\ 1 \end{bmatrix} = \mathbf{P} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

# Projection Matrix

$$\begin{bmatrix} x_{pix} \\ y_{pix} \\ 1 \end{bmatrix} = \lambda \mathbf{K} \begin{bmatrix} R & -RT \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}, \mathbf{T} = \begin{bmatrix} X_0 \\ Y_0 \\ Z_0 \end{bmatrix}$$

$\mathbf{P} = \lambda \mathbf{K} \begin{bmatrix} R & -RT \end{bmatrix}$  – *Projection Matrix*

# Homogeneous Coordinates



$$\frac{c}{Z} = \frac{x}{X} = m$$

$$x = \frac{c}{Z} X = mX$$

Given  $X$  we are getting  $x$  through a scale, this is represented by scaling through  $Z$

For any  $(\bar{X}, \bar{Z}) = k(X, Z)$ , we get the same point  $x$  on the image, We want to just represent such points in the space as one point.

Purpose: Easy to represent; a image point  $x$ , can be represented in the space by associating to the a scale factor  $k[x, 1]^T$

$\begin{bmatrix} x \\ 1 \end{bmatrix} = \begin{bmatrix} kx \\ k \end{bmatrix}$ , this is defined under the homogeneous coordinate representation

# Homogeneous Coordinates – Cont.

## Motivation

- Cameras generate a projected image of the world
- **Euclidian geometry is suboptimal to describe the central projection**
- In Euclidian geometry, the math can get difficult
- Projective geometry is an alternative algebraic representation of geometric objects and transformations

# Homogeneous Coordinates – Cont.

- Math becomes simpler
- Projective geometry does not change the geometric relations
- Computations can also be done in Euclidian geometry (but more difficult)

# Homogeneous Coordinates – Cont.

- H.C. are a system of coordinates used in projective geometry
- Formulas involving H.C. are often simpler than in the Cartesian world
- Points at infinity can be represented using finite coordinates
- A single matrix can represent affine and projective transformations



# Homogeneous Coordinates – Cont.

## Definition

The representation  $\mathbf{x}$  of a geometric object is **homogeneous** if  $\mathbf{x}$  and  $\lambda\mathbf{x}$  represent the same object for  $\lambda \neq 0$

## Example

$$\mathbf{x} = \lambda \mathbf{x}$$

homogeneous

$$\mathbf{x} \neq \lambda \mathbf{x}$$

Euclidian

# Homogeneous Coordinates – Cont.

- H.C. use a  $n+1$  dimensional vector to represent the same ( $n$ -dim.) point
- Example for  $\mathbb{R}^2/\mathbb{P}^2$

$$\mathbf{x} = \begin{bmatrix} x \\ y \end{bmatrix} \quad \rightarrow \quad \mathbf{X} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

$$\mathbf{X} = \begin{bmatrix} u \\ v \\ w \end{bmatrix} = w \begin{bmatrix} u/w \\ v/w \\ 1 \end{bmatrix} = \begin{bmatrix} u/w \\ v/w \\ 1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

# Homogeneous Coordinates – Cont.

## Definition

The representation  $\mathbf{x}$  of a geometric object is **homogeneous** if  $\mathbf{x}$  and  $\lambda\mathbf{x}$  represent the same object for  $\lambda \neq 0$

## Example

$$\mathbf{x} = \begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} wx \\ wy \\ w1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} x \\ y \end{bmatrix}$$

homogeneous Euclidian

# Homogeneous Coordinates – Cont.

- Homogeneous Coordinates of a point  $\chi$  in the plane  $\mathbb{R}^2$  is a 3-dim. vector

$$\chi : \quad \mathbf{x} = \begin{bmatrix} u \\ v \\ w \end{bmatrix} \quad \text{with } |\mathbf{x}|^2 = u^2 + v^2 + w^2 \neq 0$$

- it corresponds to Euclidian coordinates

$$\chi : \quad \mathbf{x} = \begin{bmatrix} u/w \\ v/w \end{bmatrix} \quad \text{with } w \neq 0$$

# Homogeneous Coordinates – Cont.

The projective plane  $\mathbb{P}^2(\mathbb{R})$  or  $\mathbb{P}^2$  contains

- All points  $\mathcal{X}$  of the Euclidian plane  $\mathbb{R}^2$  with  $x = [x, y]^\top$  expressed through the 3-valued vector (e.g.,  $\mathbf{x} = [x, y, 1]^\top$ )
- and all points at infinity, i.e.,  
 $\mathbf{x} = [x, y, 0]^\top$
- except  $[0, 0, 0]^\top$

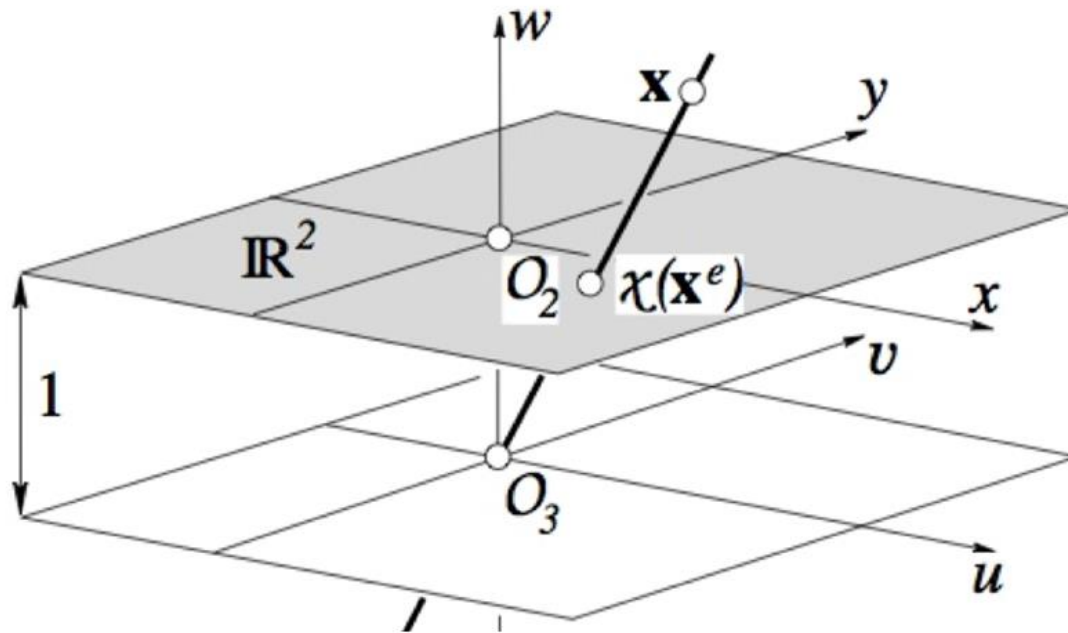
# Homogeneous Coordinates – Cont.

## From Homogeneous to Euclidian Coordinates

$$\begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} u/w \\ v/w \\ 1 \end{bmatrix} \rightarrow \begin{bmatrix} u/w \\ v/w \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix}$$

# Homogeneous Coordinates – Cont.

## From Homogeneous to Euclidian Coordinates



$$\begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} u/w \\ v/w \\ 1 \end{bmatrix} \rightarrow \begin{bmatrix} u/w \\ v/w \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix}$$

Image courtesy: Förstner

# Homogeneous Coordinates – Cont.

## 3D Points

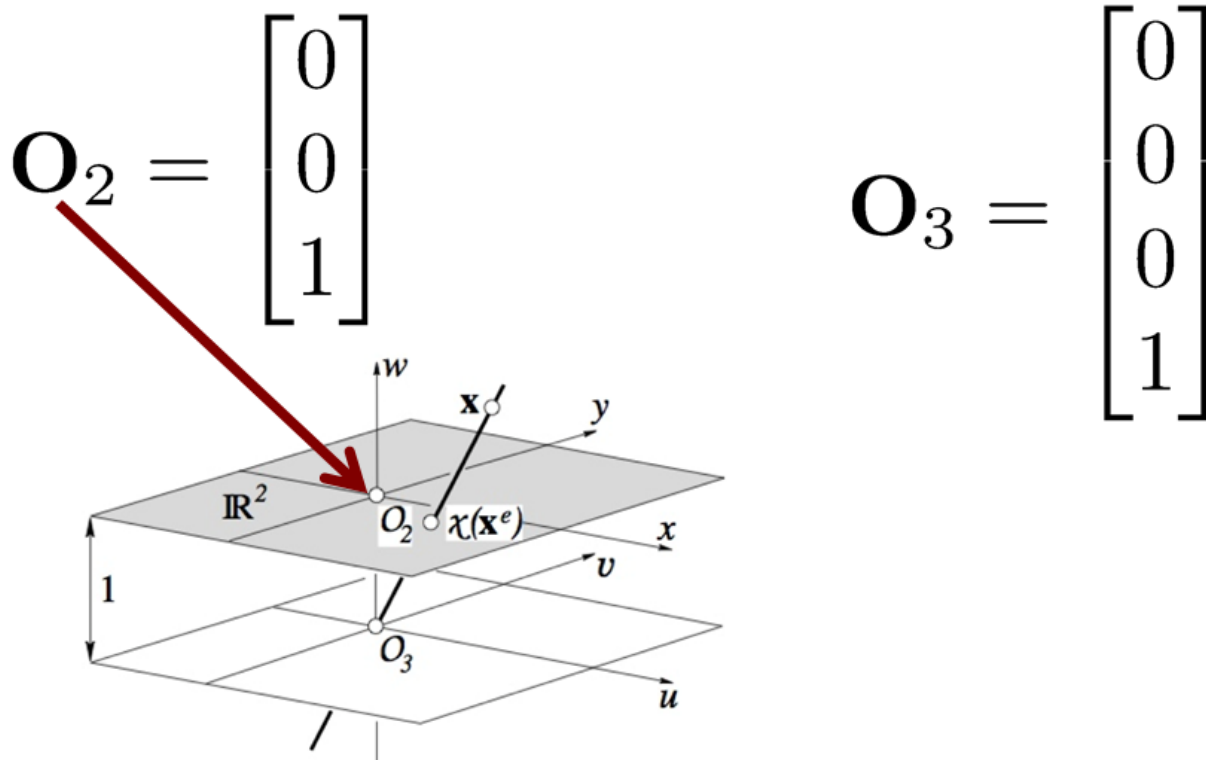
Analogous for points in 3D Euclidian space  $\mathbb{R}^3$

$$\begin{array}{ccc} \text{homogeneous} & & \text{Euclidian} \\ \mathbf{X} = \begin{bmatrix} U \\ V \\ W \\ T \end{bmatrix} = \begin{bmatrix} U/T \\ V/T \\ W/T \\ 1 \end{bmatrix} & \rightarrow & \begin{bmatrix} U/T \\ V/T \\ W/T \end{bmatrix} \end{array}$$



# Homogeneous Coordinates – Cont.

## Origin of the **Euclidian** Coordinate System in H.C.



# H.C. – Lines

## Representations of Lines

- Hesse normal form (angle  $\phi$ , distance  $d$ )

$$x \cos \phi + y \sin \phi - d = 0$$

- Intercept form

$$\frac{x}{x_0} + \frac{y}{y_0} = 1 \quad \text{or} \quad \frac{x}{x_0} + \frac{y}{y_0} - 1 = 0$$

- Standard form

$$ax + by + c = 0$$

**All form linear equations that are equal to zero**

# H.C. – Lines – Cont.

## Representations of Lines

$$\text{point } \mathbf{x} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

$$\text{Hesse } \mathbf{l} = \begin{bmatrix} \cos \phi \\ \sin \phi \\ -d \end{bmatrix}$$

$$\text{intercept } \mathbf{l} = \begin{bmatrix} 1 \\ x_0 \\ 1 \\ y_0 \\ -1 \end{bmatrix}$$

$$\text{standard } \mathbf{l} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

$$\mathbf{x} \cdot \mathbf{l} = \mathbf{x}^T \mathbf{l} = \mathbf{l}^T \mathbf{x} = 0$$

# H.C. – Lines – Cont.

## Definition

- Homogeneous Coordinates of a line  $\ell$  in the plane is a 3-dim. vector

$$\ell : \quad \mathbf{l} = \begin{bmatrix} l_1 \\ l_2 \\ l_3 \end{bmatrix} \quad \text{with } |\mathbf{l}|^2 = l_1^2 + l_2^2 + l_3^2 \neq 0$$

- it corresponds to Euclidian representation

$$l_1x + l_2y + l_3 = 0$$

$\mathbf{l} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$  does not correspond to any line and hence is excluded

## Test If a Point Lies on a Line

- A point

$$\mathbf{x} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

- lies on a line

$$\mathbf{l} = \begin{bmatrix} l_1 \\ l_2 \\ l_3 \end{bmatrix}$$

- if  $\mathbf{x} \cdot \mathbf{l} = 0$

## Intersecting Lines

- The intersection of two lines in H.C. is

$$\chi = \ell \cap m : \quad \mathbf{x} = \mathbf{l} \times \mathbf{m}$$

- **Simple way for computing the intersection of two lines using H.C.**

# H.C. – Lines – Cont.

- Line  $l$  between two points  $x, y$ :

Idea: both points line on that line, meaning:

$$x \cdot l = 0, y \cdot l = 0$$

We know that

$$(x \times y) \cdot x = 0$$

$$(x \times y) \cdot y = 0$$

Therefore:

$$l = x \times y$$

# H.C. Lines and points – Duality

- A point lies on a line if

$$\mathbf{x} \cdot \mathbf{l} = 0$$

- Intersection of two lines

$$\chi = \ell \cap m : \quad \mathbf{x} = \mathbf{l} \times \mathbf{m}$$

- A line through two given points

$$\ell = \chi \wedge y : \quad \mathbf{l} = \mathbf{x} \times \mathbf{y}$$



# Duality

- Without proof we give the definition of duality in the homogenous coordinate system formulation

*To any theorem of 2-dimension projective geometry, there corresponds a dual theorem, which may be derived by interchanging the roles of points and lines in the original theorem.*

# H.C. – Infinity

## Points at Infinity

- It is possible to **explicitly** model infinitively distant points **with finite coordinates**

$$\chi_{\infty} : \quad \mathbf{x}_{\infty} = \begin{bmatrix} u \\ v \\ 0 \end{bmatrix} \quad \mathbf{x}_{\infty} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \text{ (ideal point, all the lines intersect to this infinity point)}$$

- We can **maintain the direction** to that infinitively distant point
- Great tool when working with cameras as they are bearing-only sensors

# H.C. – Infinity – Cont.

## Intersection at Infinity

- All lines  $\ell$  with  $\ell \cdot \chi_\infty = 0$  pass through  $\chi_\infty$
- This means  $[u, v] \cdot [\cos \phi, \sin \phi] = 0$
- This hold for any line  $\mathbf{m} = [\cos \phi, \sin \phi, *]^T$   
i.e. for any line that is parallel to  $\ell$

$$\mathbf{l} \times \mathbf{m} = \begin{bmatrix} a \\ b \\ c \end{bmatrix} \times \begin{bmatrix} a \\ b \\ d \end{bmatrix} = \begin{bmatrix} bd - bc \\ ac - ad \\ ab - ab \end{bmatrix} = \begin{bmatrix} bd - bc \\ ac - ad \\ 0 \end{bmatrix}$$

**All parallel lines meet at one point at infinity!**

# H.C. – Infinity – Cont.

## Lines at Infinity

$l_{\infty} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ , all the points at infinity will lie on this line. (also called ideal line)

i.e.  $p_{\infty} = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix}$

The line at infinity is invariant of affine!

# H.C. – Infinity – Cont.



Image Courtesy: J. Jannene

## Analogous for 3D Objects

- 3D point

$$\mathbf{X} = \begin{bmatrix} U \\ V \\ W \\ T \end{bmatrix} = \begin{bmatrix} U/T \\ V/T \\ W/T \\ 1 \end{bmatrix} \rightarrow \begin{bmatrix} U/T \\ V/T \\ W/T \end{bmatrix}$$

- Plane

$$\mathbf{A} = \begin{bmatrix} A \\ B \\ C \\ D \end{bmatrix}$$

Similar properties in terms of infinity, can be extended



## Point on a Plane

- Via the scalar product, we can again test if a point lies on a plane

$$\mathbf{A} \cdot \mathbf{X} = \mathbf{A}^T \mathbf{X} = \mathbf{X}^T \mathbf{A} = 0$$

- which is based on

$$AX + BY + CZ + D = 0$$

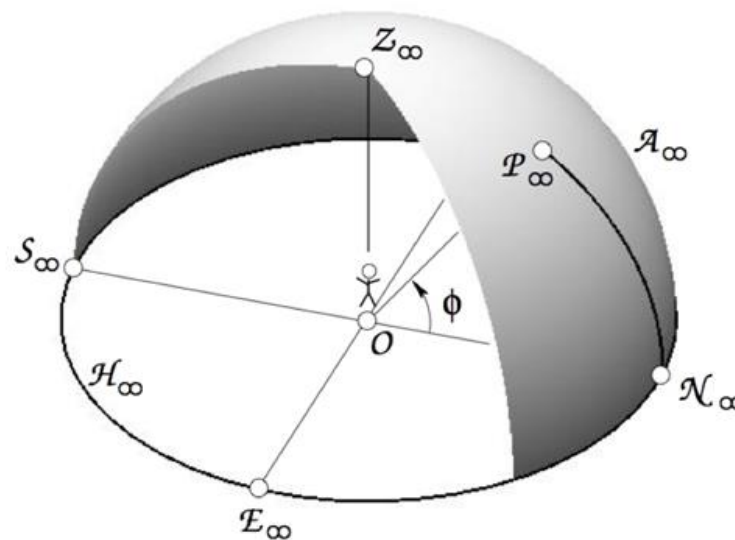
## 3D Objects at Infinity

- 3D point

$$\mathbf{P}_{\infty} = \begin{bmatrix} U \\ V \\ W \\ 0 \end{bmatrix}$$

- Plane

$$\mathbf{A}_{\infty} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$





# Projection Matrix

$$\begin{bmatrix} x_{pix} \\ y_{pix} \\ 1 \end{bmatrix} = \lambda \mathbf{K} \begin{bmatrix} R & -RT \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}, \mathbf{T} = \begin{bmatrix} X_0 \\ Y_0 \\ Z_0 \end{bmatrix}$$

$$\mathbf{P} = \lambda \mathbf{K} \begin{bmatrix} R & -RT \end{bmatrix} - \textit{Projection Matrix}$$

$$\begin{bmatrix} x_{pix} \\ y_{pix} \\ 1 \end{bmatrix} = \mathbf{K} \begin{bmatrix} R & -RT \end{bmatrix} \lambda \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} \quad \text{or understand as} \quad \frac{1}{\lambda} \begin{bmatrix} x_{pix} \\ y_{pix} \\ 1 \end{bmatrix} = \mathbf{K} \begin{bmatrix} R & -RT \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

This is represented as H.C. for the output, we can get the results by ignoring  $\lambda$  in the definition of  $\mathbf{P}$

$$\frac{1}{\lambda} \begin{bmatrix} x_{pix} \\ y_{pix} \\ 1 \end{bmatrix} = \mathbf{P} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

# Next Class

- Geometric Transformation
- RANSAC Algorithm
- Panorama – Assignment 2