








OLLSCOIL NA GAILLIMHÉ  
UNIVERSITY OF GALWAY

Dr Waqar Shahid Qureshi  
[Waqarshahid.qureshi@universityofgalway.ie](mailto:Waqarshahid.qureshi@universityofgalway.ie)



# Revision from Transformation

# Hierarchy of 3D Transformations

Transformation	Matrix	# DoF	Preserves	Icon
translation	$\begin{bmatrix} \mathbf{I} &   & \mathbf{t} \end{bmatrix}_{3 \times 4}$	3	orientation	
rigid (Euclidean)	$\begin{bmatrix} \mathbf{R} &   & \mathbf{t} \end{bmatrix}_{3 \times 4}$	6	lengths	
similarity	$\begin{bmatrix} s\mathbf{R} &   & \mathbf{t} \end{bmatrix}_{3 \times 4}$	7	angles	
affine	$\begin{bmatrix} \mathbf{A} \end{bmatrix}_{3 \times 4}$	12	parallelism	
projective	$\begin{bmatrix} \tilde{\mathbf{H}} \end{bmatrix}_{4 \times 4}$	15	straight lines	

# 3D Rotation Matrices

## • Rotations about Principal Axes

$$\begin{bmatrix} \cos \theta & -\sin \theta & 0 & 0 \\ \sin \theta & \cos \theta & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos \gamma & -\sin \gamma & 0 \\ 0 & \sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \beta & 0 & \sin \beta & 0 \\ 0 & 1 & 0 & 0 \\ -\sin \beta & 0 & \cos \beta & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

About origin, in right-handed coordinate system, counter clockwise when looking towards origin from positive axis

- Rotation matrix is orthonormal with Determinant of +1 and 3 dof
- Inverse of a rotation matrix is its transpose
- Concatenation of Rotations is also a rotation
- IMP: A rotation matrix transforms its own rows onto the principal axes

- Any 3D rotation matrix can be described as rotation about an axis  $\mathbf{n}$  by an angle  $\theta$
- To rotate about given axis  $\mathbf{n}$  by  $\theta$ :
  - Rotate axes onto a principal axis
    - by composing appropriate matrix through cross products
  - Rotate about principal axes and then undo the earlier transformation
  - OR use Rodriguez formula
- To compute  $\mathbf{n}$  and  $\theta$  from a 3D rotation matrix
  - $\mathbf{n}$  is the eigenvector corresponding to the real eigenvalue of 1
  - $\theta$  can be computed by the other 2 eigenvalues, which are  $\cos \theta \pm i \sin \theta$
  - To disambiguate angle values, check for consistency with Rodriguez formula

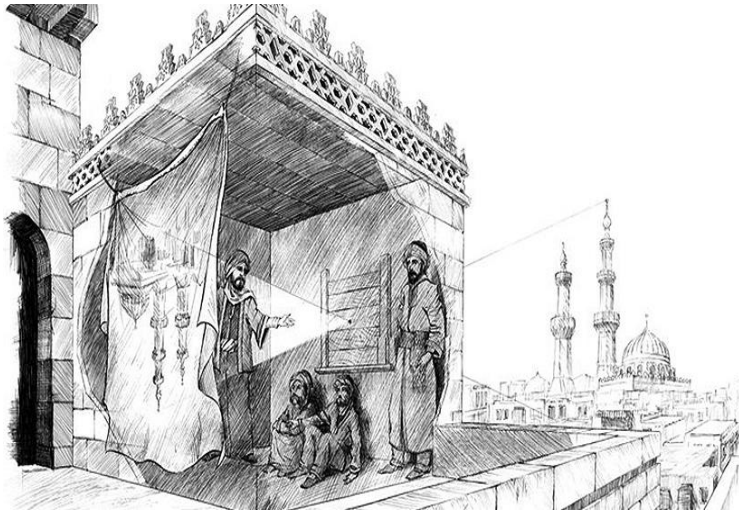


# Camera Model

Part-1



# Pinhole camera



First described by Ibn Al-Haytham

ابو علي، الحسن بن الحسن بن الهيثم  
in his 7-volume work

كتاب المناظر

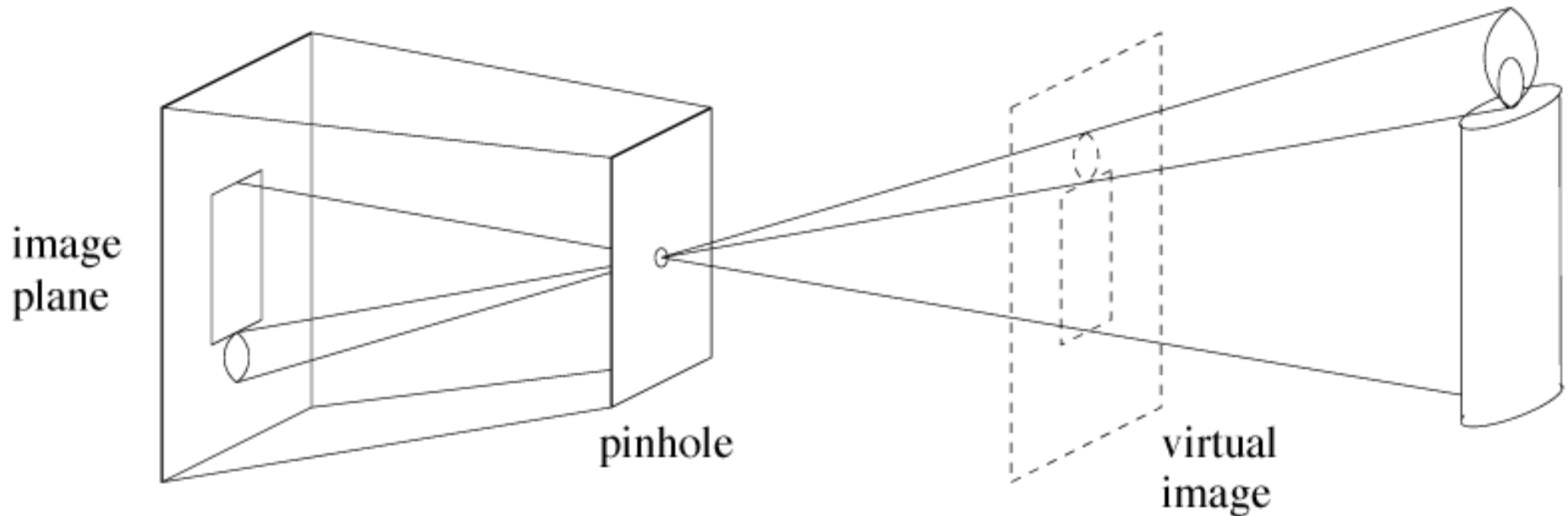
He termed it

بيت المظلم Al-beit Al-muzlim  
which was later translated into Latin as “camera  
obscura”

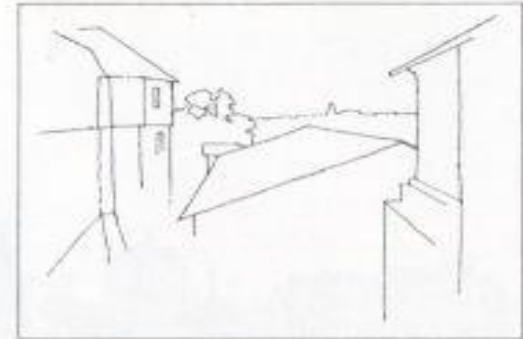
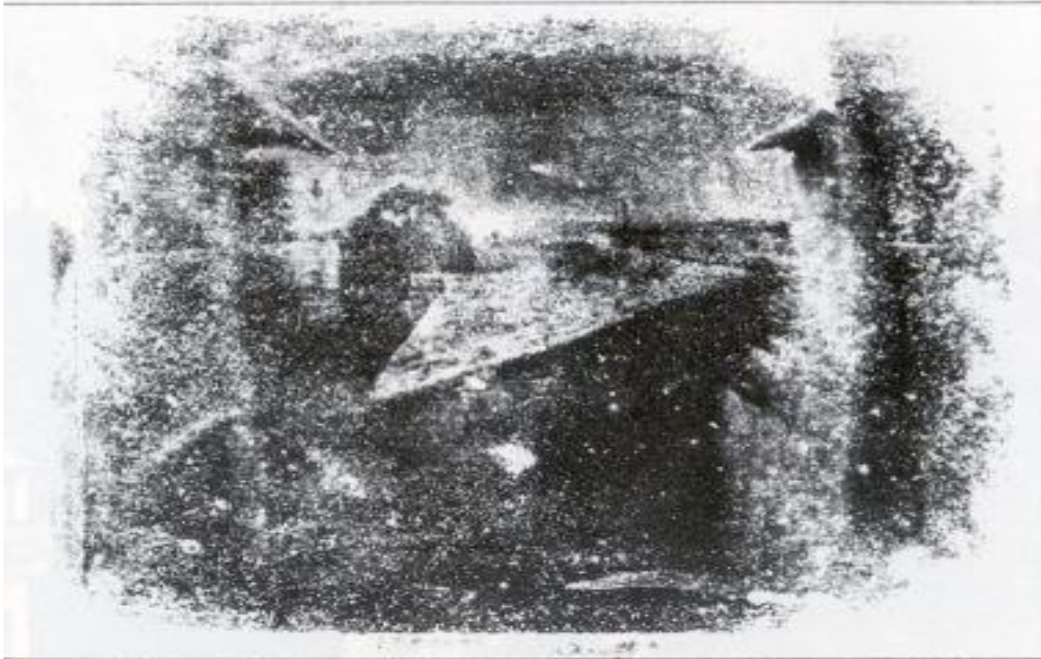


# Pinhole Camera

- Lens is assumed to be single point
- Infinitesimally small aperture
- Has infinite depth of field i.e. everything is in focus



# The first photograph on record



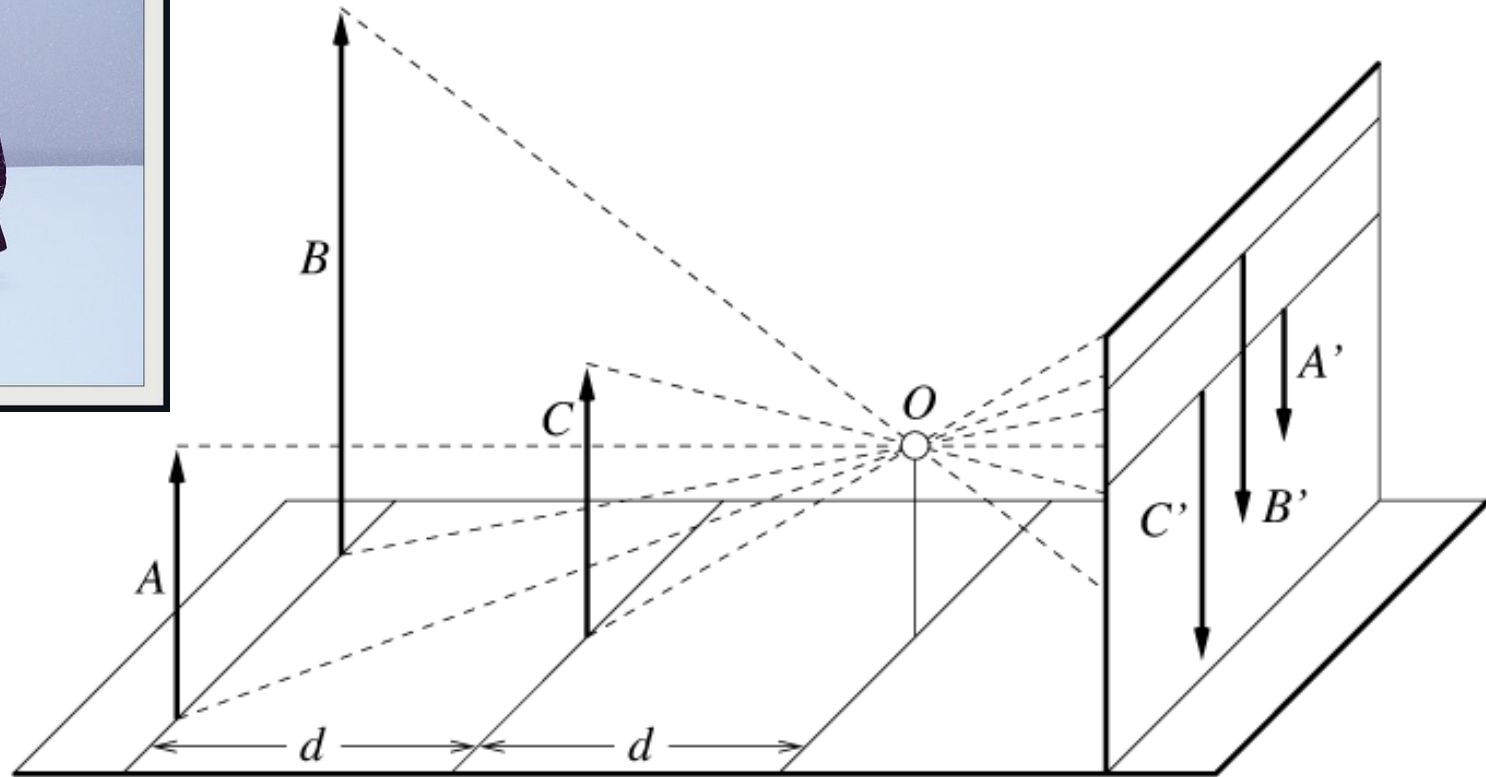
## THE FIRST PHOTOGRAPH

The world's first photograph was made in 1826 by Nicéphore Niépce from a window in his estate in France. For "film" Niépce used a sensitized pewter plate and he got a blurred image of the rooftops outlined above. This photograph is usually retouched to make it legible, but the version shown at left is what it really looks like.





# Pinhole Camera Properties: Distant objects are smaller



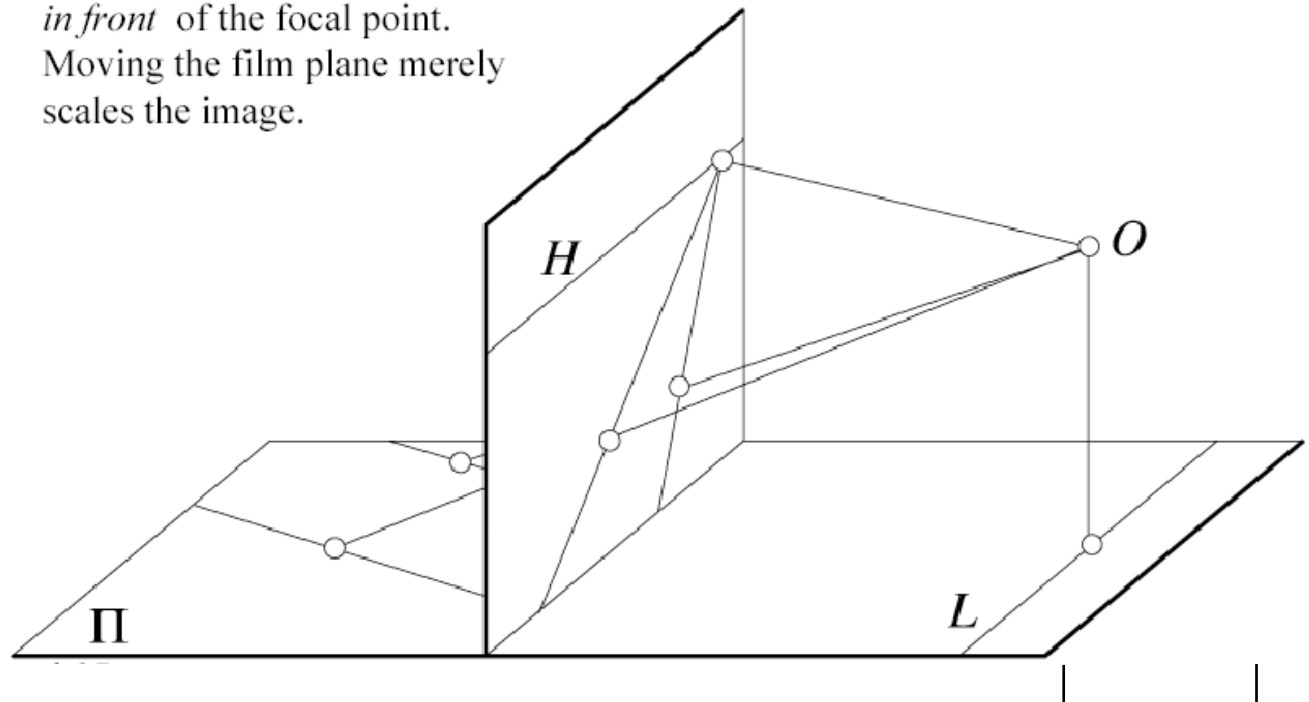
Slide Credit: Forsyth/Ponce <http://www.cs.berkeley.edu/~daf/bookpages/slides.html>  
and Khurram Shafique, Object Video



# Pinhole Camera Properties

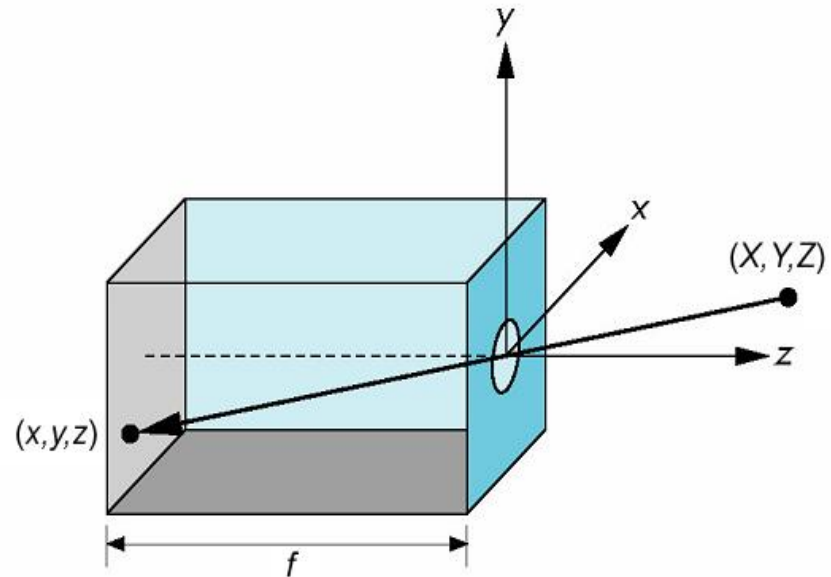
- Lines map to lines
- Polygons map to polygons
- Parallel lines meet

Common to draw film plane  
*in front* of the focal point.  
Moving the film plane merely  
scales the image.



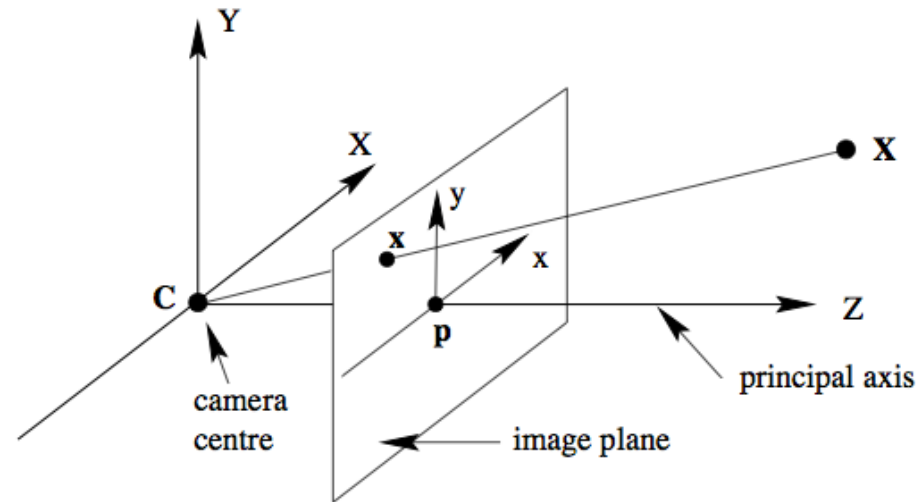
# Pinhole Camera

- Pinhole is considered the *center of projection, camera center or optical center*
- Let center of projection be at the origin of Euclidean space
- Plane  $Z = f$  is the *imaging plane, or focal plane*
- Line from camera center, perpendicular to imaging plane is the *principal axis or principal ray*



# Pinhole Camera in Canonical Configuration

- Camera center  $\mathbf{C}$  is at Euclidean origin
- Principal axis aligned with  $Z$ -axis
- *Principal point*  $\mathbf{p}$  is the point where principal axis cuts the imaging plane
- Imaging plane is often taken by convention to be in front of the camera



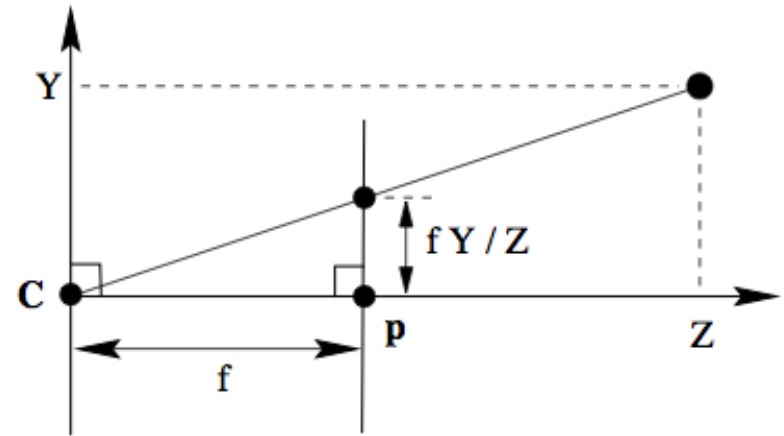
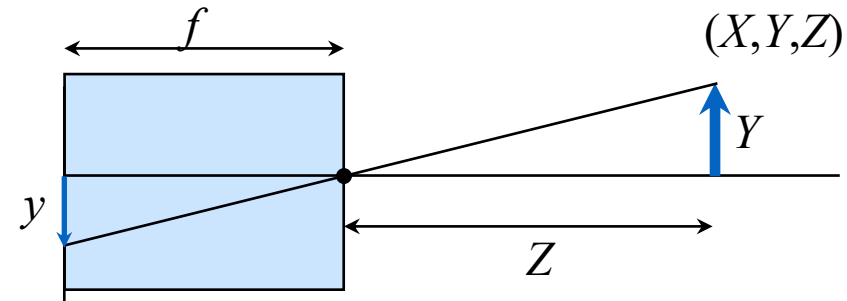
# Pinhole Camera in Canonical Configuration

- Camera maps point  $(X, Y, Z)^T$  to  $(x, y)^T$
- By similar triangles

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

$$y = \frac{fY}{Z}$$

- Similarly  $x = \frac{fX}{Z}$



$$(X, Y, Z)^T \mapsto \left( \frac{fX}{Z}, \frac{fY}{Z} \right)^T$$

$$\mathbb{R}^3 \mapsto \mathbb{R}^2$$

- Thus, the camera maps
- This mapping is from





# Central Projection

- We can write this as a matrix using the homogeneous coordinates

$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \frac{1}{f} & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

- Verify that since  $hx = X, \quad hy = Y, \quad h = \frac{Z}{f}$

- Hence  $x = \frac{fX}{Z}$   
 $y = \frac{fY}{Z}$



# Central Projection

- Camera in canonical view (centered at origin with optical axis aligned with world  $Z$  axis, image axes aligned with  $X$  and  $Y$ )

$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \frac{1}{f} & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

- Since any scaling of a homogeneous equation is valid, it is often written as

$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$



# Central Projection

- The camera can be more compactly written as

$$\mathbf{x} = \mathbf{P}\mathbf{X}$$

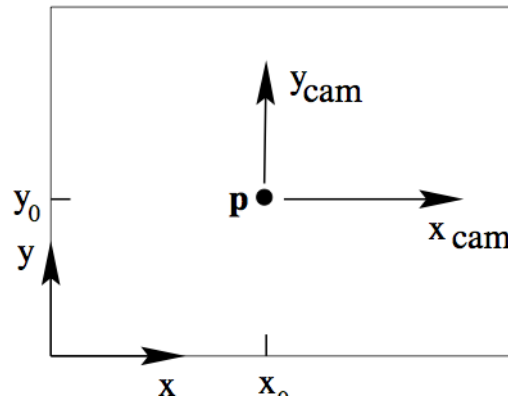
- where  $\mathbf{P}$  is a  $3 \times 4$  matrix that maps from  $\mathbb{P}^3 \mapsto \mathbb{P}^2$
- $\mathbf{P}$  may also be written as:

$$\mathbf{P} = \text{diag}(f, f, 1) [\mathbf{I} \mid \mathbf{0}]$$



# Principal Point Offset

- The expression  $P = \text{diag}(f, f, 1) [\mathbf{I} \mid \mathbf{0}]$  assumes that image origin is at the principal point.
- This may not be the case in general. For example:



- If the image coordinates of the principal point are  $(p_x, p_y)^T$ , then the camera mapping will be

$$(X, Y, Z)^T \mapsto \left( \frac{fX}{Z} + p_x, \frac{fY}{Z} + p_y \right)^T$$

# Central Projection with Principal Point Offset

- In matrix form, this mapping becomes

$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} f & 0 & p_x & 0 \\ 0 & f & p_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

- Sometimes, for convenience, it is also written as

$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} f & 0 & p_x \\ 0 & f & p_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

- Or  $\mathbf{x} = \mathbf{K} [\mathbf{I} \mid \mathbf{0}] \mathbf{X}$
- $\mathbf{K}$  is called the camera calibration matrix, which is a 3x3 matrix of internal camera parameters





# CCD Camera

- We have assumed same units for world & image coordinates
- In a CCD camera, image coordinates are measured in pixels
- Some CCD cameras also have non-square pixels
- We can convert to pixel units as

$$K = \begin{bmatrix} m_x f & 0 & x_0 \\ 0 & m_y f & y_0 \\ 0 & 0 & 1 \end{bmatrix}$$

where  $m_x$  and  $m_y$  are scale factors of pixels per unit length, needed to convert to pixel dimension

- $m_x = \text{\#of pixels in } x \text{ direction} / \text{size of CCD array in } x \text{ direction}$
- $m_y = \text{\#of pixels in } y \text{ direction} / \text{size of CCD array in } y \text{ direction}$
- $(x_0, y_0)$  is principal point offset in pixel dimensions

$$x_0 = m_x p_x, \quad y_0 = m_y p_y$$



## Example

If you had a camera where:

- The CCD array is 2000 pixels wide and 1500 pixels tall.
- The physical size of the CCD sensor is 20 mm by 15 mm.
- The focal length  $f = 50$  mm.

You would calculate:

- $m_x = \frac{2000 \text{ pixels}}{20 \text{ mm}} = 100 \text{ pixels/mm}$
- $m_y = \frac{1500 \text{ pixels}}{15 \text{ mm}} = 100 \text{ pixels/mm}$

Thus, your matrix  $K$  would have the scale factors  $m_x f$  and  $m_y f$ , and the principal point offset  $(x_0, y_0)$ .

# Pinhole camera in general view

- This is for the case when the camera's optical axis is aligned with the world z-axis
- What if that is not the case?

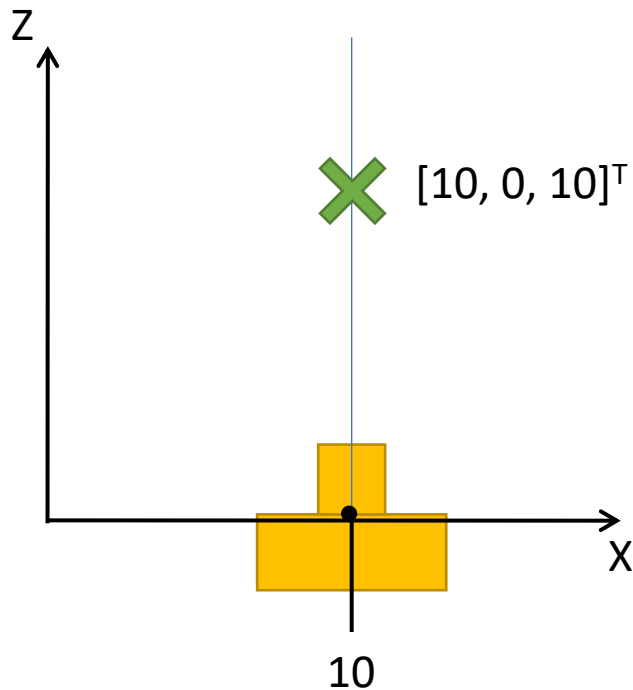
# Pinhole camera in general view

- If the camera center is at coordinates  $\mathbf{C}$  in the world, i.e. the camera is moved  $\mathbf{C}$  from the origin, we should move the world point by  $\mathbf{C}^{-1}$
- Then the perspective transform equation will be applicable
- Same holds for rotations



# Example

- Translation by 10 units to the right



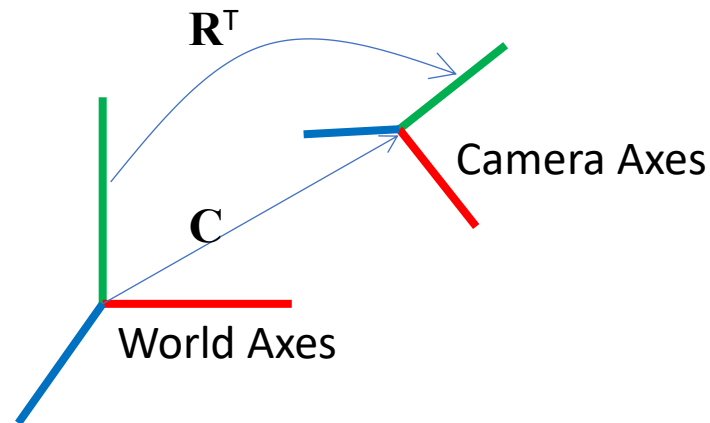
$$\begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & -10 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 10 \\ 0 \\ 10 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 10 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 10 \end{bmatrix}$$



# Pinhole camera in general view

- In general, the camera center is at a rotation of  $R^T$ , followed by a translation of  $C$  from the world origin
- Then



$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} m_x f & 0 & x_0 & 0 \\ 0 & m_y f & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \left( \begin{bmatrix} r_{11} & r_{12} & r_{13} & 0 \\ r_{21} & r_{22} & r_{23} & 0 \\ r_{31} & r_{32} & r_{33} & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & -C_x \\ 0 & 1 & 0 & -C_y \\ 0 & 0 & 1 & -C_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} \right)$$

# Pinhole camera in general view

$$\text{Canonical View} \begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} f & 0 & p_x \\ 0 & f & p_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

$$\text{General View} \begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} f & 0 & p_x \\ 0 & f & p_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_X \\ r_{21} & r_{22} & r_{23} & t_Y \\ r_{31} & r_{32} & r_{33} & t_Z \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

$$\mathbf{x} = \mathbf{K} [\mathbf{R} \mid \mathbf{T}] \mathbf{X}$$

$$\mathbf{x} = \mathbf{K} \mathbf{R} \left[ \mathbf{I}_{3 \times 3} \mid -\tilde{\mathbf{C}} \right] \mathbf{X}$$

$\mathbf{x}$  – image point

$\mathbf{X}$  – world point

$\mathbf{K}$  – 3x3 matrix of internal camera parameters

$[\mathbf{R} \mid \mathbf{T}]$  – 3x4 matrix of external camera parameters

$\mathbf{R}$  – rotation needed to align camera to world axes

$\mathbf{T}$  – Translation needed to bring camera to world origin

$\mathbf{T} = -\mathbf{R}\mathbf{C}$  where  $\mathbf{C}$  is the vector of camera center

# Camera Model Example

- Think that the camera was originally at the origin looking down Z axis
- Then it was translated by  $(r_1, r_2, r_3)^T$ , rotated by  $\phi$  along X,  $\theta$  along Z, then translated by  $(x_0, y_0, z_0)^T$
- This is the scenario in the figure on right

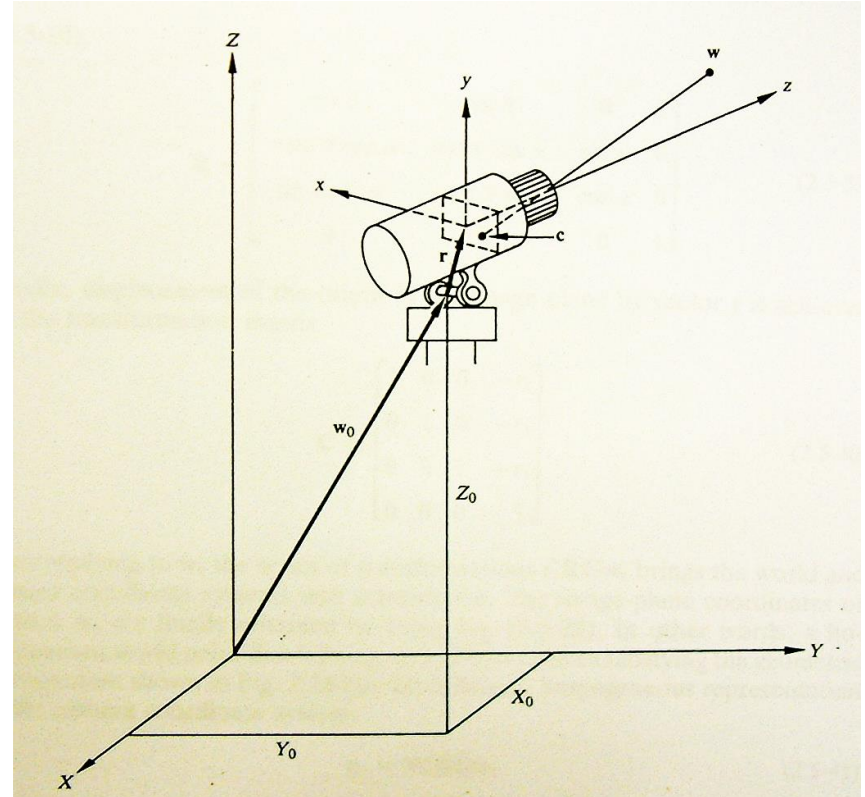
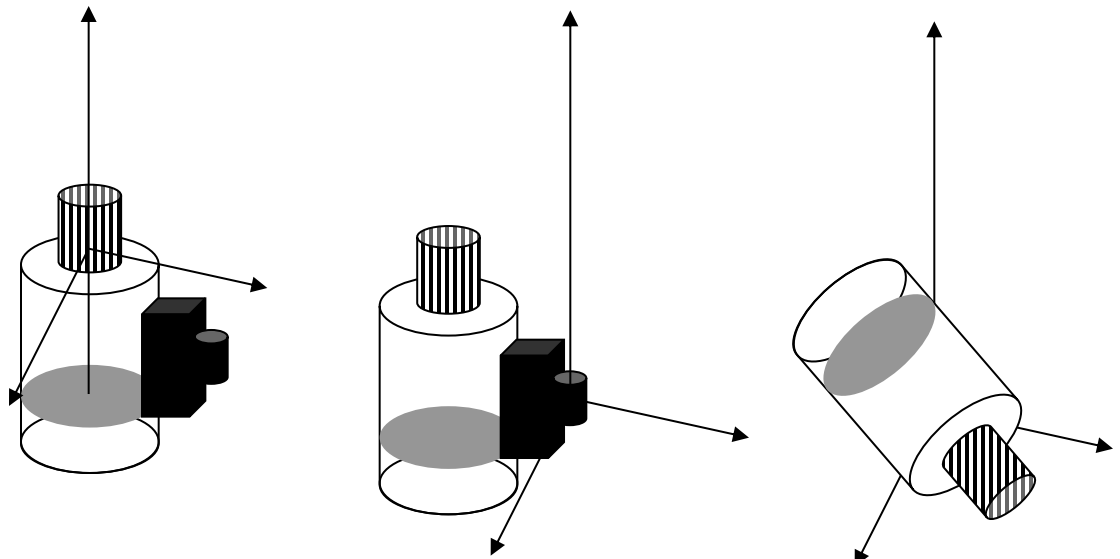


Figure Reference: Gonzales and Woods, "Digital Image Processing"

# Camera Model Example



$$\begin{matrix} \hat{x} \\ \hat{y} \\ \hat{z} \\ \hat{t} \end{matrix} \begin{bmatrix} m_x f & 0 & x_0 & 0 \\ 0 & m_y f & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{matrix} \hat{u} \\ \hat{v} \\ \hat{w} \\ \hat{t} \end{matrix} \left[ \begin{matrix} \hat{x} \\ \hat{y} \\ \hat{z} \\ \hat{t} \end{matrix} \begin{bmatrix} 1 & 0 & 0 & X_0 \\ 0 & 1 & 0 & Y_0 \\ 0 & 0 & 1 & Z_0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{matrix} \hat{u} \\ \hat{v} \\ \hat{w} \\ \hat{t} \end{matrix} \begin{bmatrix} \cos \theta & -\sin \theta & 0 & 0 \\ \sin \theta & \cos \theta & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{matrix} \hat{u} \\ \hat{v} \\ \hat{w} \\ \hat{t} \end{matrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos \phi & -\sin \phi & 0 \\ 0 & \sin \phi & \cos \phi & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{matrix} \hat{u} \\ \hat{v} \\ \hat{w} \\ \hat{t} \end{matrix} \begin{bmatrix} 1 & 0 & 0 & r_1 \\ 0 & 1 & 0 & r_2 \\ 0 & 0 & 1 & r_3 \\ 0 & 0 & 0 & 1 \end{bmatrix} \right]^{-1}$$

$$\begin{matrix} \hat{x} \\ \hat{y} \\ \hat{z} \\ \hat{t} \end{matrix} \begin{bmatrix} m_x f & 0 & x_0 & 0 \\ 0 & m_y f & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{matrix} \hat{u} \\ \hat{v} \\ \hat{w} \\ \hat{t} \end{matrix} \begin{bmatrix} 1 & 0 & 0 & -r_1 \\ 0 & 1 & 0 & -r_2 \\ 0 & 0 & 1 & -r_3 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{matrix} \hat{u} \\ \hat{v} \\ \hat{w} \\ \hat{t} \end{matrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos \phi & \sin \phi & 0 \\ 0 & -\sin \phi & \cos \phi & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{matrix} \hat{u} \\ \hat{v} \\ \hat{w} \\ \hat{t} \end{matrix} \begin{bmatrix} \cos \theta & \sin \theta & 0 & 0 \\ -\sin \theta & \cos \theta & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{matrix} \hat{u} \\ \hat{v} \\ \hat{w} \\ \hat{t} \end{matrix} \begin{bmatrix} 1 & 0 & 0 & -X_0 \\ 0 & 1 & 0 & -Y_0 \\ 0 & 0 & 1 & -Z_0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

# Camera Model Example

$$x = f \frac{(X - X_0) \cos \theta + (Y - Y_0) \sin \theta - r_1}{-(X - X_0) \sin \theta \sin \phi + (Y - Y_0) \cos \theta \sin \phi - (Z - Z_0) \cos \phi + r_3 + f},$$
$$y = f \frac{-(X - X_0) \sin \theta \cos \phi + (Y - Y_0) \cos \theta \cos \phi + (Z - Z_0) \sin \phi - r_2}{-(X - X_0) \sin \theta \sin \phi + (Y - Y_0) \cos \theta \sin \phi - (Z - Z_0) \cos \phi + r_3 + f}.$$

- This camera model is applicable in many situations
- For example, this is the typical surveillance camera scenario



# Aircraft Example

	OTTER	system_id
	TV	sensor_type
	0001	serial_number
9.400008152666640300e+08		image_time
3.813193746469612200e+01		vehicle_latitude
-7.734523185193877700e+01		vehicle_longitude
9.949658409987658800e+02		vehicle_height
9.995171174441039900e-01		vehicle_pitch
1.701626418113209000e+00		vehicle_roll
1.207010551753029400e+02		vehicle_heading
1.658968732990974800e-02		camera_focal_length
-5.361314389557259100e+01		camera_elevation
-7.232969433546705000e+00		camera_scan_angle
	480	number_image_lines
	640	number_image_samples



cameraMat = perspective\_transform \* gimbal\_rotation\_y \* gimbal\_rotation\_z \*  
gimbal\_translation \* vehicle\_rotation\_x \* vehicle\_rotation\_y \* vehicle\_rotation\_z \*  
vehicle\_translation ;

$$\mathbf{P} = \begin{bmatrix} m_x f & 0 & x_0 & 0 \\ 0 & m_y f & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \cos \omega & 0 & -\sin \omega & 0 \\ 0 & 1 & 0 & 0 \\ \sin \omega & 0 & \cos \omega & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \tau & \sin \tau & 0 & 0 \\ -\sin \tau & \cos \tau & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \varphi & 0 & -\sin \varphi & 0 \\ 0 & 1 & 0 & 0 \\ \sin \varphi & 0 & \cos \varphi & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos \beta & \sin \beta & 0 \\ 0 & -\sin \beta & \cos \beta & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \alpha & \sin \alpha & 0 & 0 \\ -\sin \alpha & \cos \alpha & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & -\Delta T_x \\ 0 & 1 & 0 & -\Delta T_y \\ 0 & 0 & 1 & -\Delta T_z \\ 0 & 0 & 0 & 1 \end{bmatrix}$$



$$\begin{aligned}
c(1,1) &= (\cos(c\_scn)*\cos(v\_rll)-\sin(c\_scn)*\sin(v\_pch)*\sin(v\_rll))*\cos(v\_hdg)-\sin(c\_scn)*\cos(v\_pch)*\sin(v\_hdg); \\
c(1,2) &= -(\cos(c\_scn)*\cos(v\_rll)-\sin(c\_scn)*\sin(v\_pch)*\sin(v\_rll))*\sin(v\_hdg)-\sin(c\_scn)*\cos(v\_pch)*\cos(v\_hdg); \\
c(1,3) &= -\cos(c\_scn)*\sin(v\_rll)-\sin(c\_scn)*\sin(v\_pch)*\cos(v\_rll); \\
c(1,4) &= -((\cos(c\_scn)*\cos(v\_rll)-\sin(c\_scn)*\sin(v\_pch)*\sin(v\_rll))*\cos(v\_hdg)-\sin(c\_scn)*\cos(v\_pch)*\sin(v\_hdg))*vx-(-\cos(c\_scn)*\cos(v\_rll)- \\
&\quad \sin(c\_scn)*\sin(v\_pch)*\sin(v\_rll))*\sin(v\_hdg)-\sin(c\_scn)*\cos(v\_pch)*\cos(v\_hdg))*vy-(-\cos(c\_scn)*\sin(v\_rll)-\sin(c\_scn)*\sin(v\_pch)*\cos(v\_rll))*vz;
\end{aligned}$$

$$\begin{aligned}
c(2,1) &= (-\sin(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(-\sin(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+\cos(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\cos(v\_hdg)+(- \\
&\quad \sin(c\_elv)*\cos(c\_scn)*\cos(v\_pch)-\cos(c\_elv)*\sin(v\_pch))*\sin(v\_hdg); \\
c(2,2) &= -(-\sin(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(-\sin(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+\cos(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\sin(v\_hdg)+(- \\
&\quad \sin(c\_elv)*\cos(c\_scn)*\cos(v\_pch)-\cos(c\_elv)*\sin(v\_pch))*\cos(v\_hdg); \\
c(2,3) &= \sin(c\_elv)*\sin(c\_scn)*\sin(v\_rll)+(-\sin(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+\cos(c\_elv)*\cos(v\_pch))*\cos(v\_rll); \\
c(2,4) &= -((-sin(c\_elv)*sin(c\_scn)*cos(v\_rll)+(-sin(c\_elv)*cos(c\_scn)*sin(v\_pch)+cos(c\_elv)*cos(v\_pch))*sin(v\_rll))*cos(v\_hdg)+(- \\
&\quad \sin(c\_elv)*\cos(c\_scn)*\cos(v\_pch)-\cos(c\_elv)*\sin(v\_pch))*\sin(v\_hdg))*vx-((-sin(c\_elv)*sin(c\_scn)*cos(v\_rll)+(- \\
&\quad \sin(c\_elv)*\cos(c\_scn)*sin(v\_pch)+cos(c\_elv)*cos(v\_pch))*sin(v\_rll))*sin(v\_hdg)+(-sin(c\_elv)*cos(c\_scn)*cos(v\_pch)- \\
&\quad \cos(c\_elv)*sin(v\_pch))*cos(v\_hdg))*vy-(-sin(c\_elv)*sin(c\_scn)*sin(v\_rll)+(-sin(c\_elv)*cos(c\_scn)*sin(v\_pch)+cos(c\_elv)*cos(v\_pch))*cos(v\_rll))*vz;
\end{aligned}$$

$$\begin{aligned}
c(3,1) &= (\cos(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+\sin(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\cos(v\_hdg)+(\cos(c\_elv)*\cos(c\_scn)*\cos(v\_pch)- \\
&\quad \sin(c\_elv)*\sin(v\_pch))*\sin(v\_hdg); \\
c(3,2) &= -(\cos(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+\sin(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\sin(v\_hdg)+(\cos(c\_elv)*\cos(c\_scn)*\cos(v\_pch)- \\
&\quad \sin(c\_elv)*\sin(v\_pch))*\cos(v\_hdg); \\
c(3,3) &= -\cos(c\_elv)*\sin(c\_scn)*\sin(v\_rll)+(\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+\sin(c\_elv)*\cos(v\_pch))*\cos(v\_rll); \\
c(3,4) &= - \\
&\quad ((\cos(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+\sin(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\cos(v\_hdg)+(\cos(c\_elv)*\cos(c\_scn)*\cos(v\_pch)- \\
&\quad \sin(c\_elv)*\sin(v\_pch))*\sin(v\_hdg))*vx-(- \\
&\quad \cos(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+\sin(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\sin(v\_hdg)+(\cos(c\_elv)*\cos(c\_scn)*\cos(v\_pch)- \\
&\quad \sin(c\_elv)*\sin(v\_pch))*\cos(v\_hdg))*vy-(-\cos(c\_elv)*\sin(c\_scn)*\sin(v\_rll)+(\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+\sin(c\_elv)*\cos(v\_pch))*\cos(v\_rll))*vz;
\end{aligned}$$

$$\begin{aligned}
c(4,1) &= \\
&\quad (1/fl*\cos(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(1/fl*\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+1/fl*\sin(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\cos(v\_hdg)+(1/fl*\cos(c\_elv)*\cos(c\_scn) \\
&\quad *\cos(v\_pch)-1/fl*\sin(c\_elv)*\sin(v\_pch))*\sin(v\_hdg); \\
c(4,2) &= - \\
&\quad (1/fl*\cos(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(1/fl*\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+1/fl*\sin(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\sin(v\_hdg)+(1/fl*\cos(c\_elv)*\cos(c\_scn) \\
&\quad *\cos(v\_pch)-1/fl*\sin(c\_elv)*\sin(v\_pch))*\cos(v\_hdg); \\
c(4,3) &= -1/fl*\cos(c\_elv)*\sin(c\_scn)*\sin(v\_rll)+(1/fl*\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+1/fl*\sin(c\_elv)*\cos(v\_pch))*\cos(v\_rll); \\
c(4,4) &= - \\
&\quad ((1/fl*\cos(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(1/fl*\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+1/fl*\sin(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\cos(v\_hdg)+(1/fl*\cos(c\_elv)*\cos(c\_scn) \\
&\quad *\cos(v\_pch)-1/fl*\sin(c\_elv)*\sin(v\_pch))*\sin(v\_hdg))*vx-(- \\
&\quad (1/fl*\cos(c\_elv)*\sin(c\_scn)*\cos(v\_rll)+(1/fl*\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+1/fl*\sin(c\_elv)*\cos(v\_pch))*\sin(v\_rll))*\sin(v\_hdg)+(1/fl*\cos(c\_elv)*\cos(c\_scn) \\
&\quad *\cos(v\_pch)-1/fl*\sin(c\_elv)*\sin(v\_pch))*\cos(v\_hdg))*vy-(- \\
&\quad 1/fl*\cos(c\_elv)*\sin(c\_scn)*\sin(v\_rll)+(1/fl*\cos(c\_elv)*\cos(c\_scn)*\sin(v\_pch)+1/fl*\sin(c\_elv)*\cos(v\_pch))*\cos(v\_rll))*vz+1;
\end{aligned}$$



# Pinhole camera in general view

Canonical View

$$\begin{bmatrix} m_x f & 0 & x_0 & 0 \\ 0 & m_y f & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} = \begin{bmatrix} hx \\ hy \\ h \end{bmatrix}$$

## Surveillance Camera Example (Small gimbal translation ignored)

$$\begin{bmatrix} m_x f & 0 & x_0 & 0 \\ 0 & m_y f & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos \phi & \sin \phi & 0 \\ 0 & -\sin \phi & \cos \phi & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos q & \sin q & 0 & 0 \\ -\sin q & \cos q & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & -X_0 \\ 0 & 1 & 0 & -Y_0 \\ 0 & 0 & 1 & -Z_0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

## Aircraft Example

$$\begin{bmatrix} m_x f & 0 & x_0 & 0 \\ 0 & m_y f & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \cos w & 0 & -\sin w & 0 \\ 0 & 1 & 0 & 0 \\ \sin w & 0 & \cos w & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos t & \sin t & 0 & 0 \\ -\sin t & \cos t & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos j & 0 & -\sin j & 0 \\ 0 & 1 & 0 & 0 \\ \sin j & 0 & \cos j & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos b & \sin b & 0 \\ 0 & -\sin b & \cos b & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos a & \sin a & 0 & 0 \\ -\sin a & \cos a & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & -DT_x \\ 0 & 1 & 0 & -DT_y \\ 0 & 0 & 1 & -DT_z \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Perspective Transform  
for Canonical View

Rotation needed to align  
camera with world axes

Translation by Inverse  
of Camera Center



# Summary: Perspective Camera Model

- The perspective camera model can be written as

$$\mathbf{x} = \mathbf{K} \left[ \mathbf{R} \mid -\mathbf{R}\tilde{\mathbf{C}} \right] \mathbf{X}$$

Image point  $\in \mathbb{P}^2$

3x3 matrix of internal camera parameters (intrinsic parameters)

3x4 matrix of external camera parameters (extrinsic parameters)

World point  $\in \mathbb{P}^3$

$$\mathbf{K} = \begin{bmatrix} m_x f & 0 & x_0 \\ 0 & m_y f & y_0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$x_0 = m_x p_x, \quad y_0 = m_y p_y$$



# Special Case 1: Perspective Camera Looking at a Plane

- Consider the case of camera looking at a plane
- This scenario occurs frequently in imaging applications



# Special Case 1:

## Perspective Camera Looking at a Plane

- Without loss of generality, we can assume that the plane has equation  $Z = 0$

$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} m_x f & 0 & x_0 \\ 0 & m_y f & y_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_1 & r_2 & r_3 & T_x \\ r_4 & r_5 & r_6 & T_y \\ r_7 & r_8 & r_9 & T_z \end{bmatrix} \begin{bmatrix} X \\ Y \\ 0 \\ 1 \end{bmatrix}$$

- The third column of  $[\mathbf{R} \mid \mathbf{T}]$  does not matter in this case and can be dropped. So we can rewrite the system as:

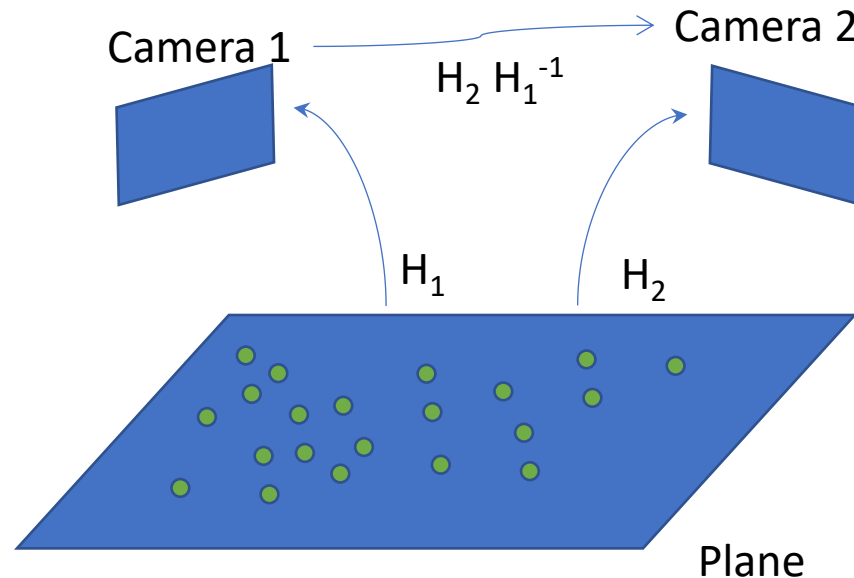
$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} a_1 & a_2 & a_3 \\ a_4 & a_5 & a_6 \\ a_7 & a_8 & a_9 \end{bmatrix} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$

- Conclusion: The 3D points lying on a plane are related to the image points by a homography.



# Special Case 1: Perspective Camera Looking at a Plane

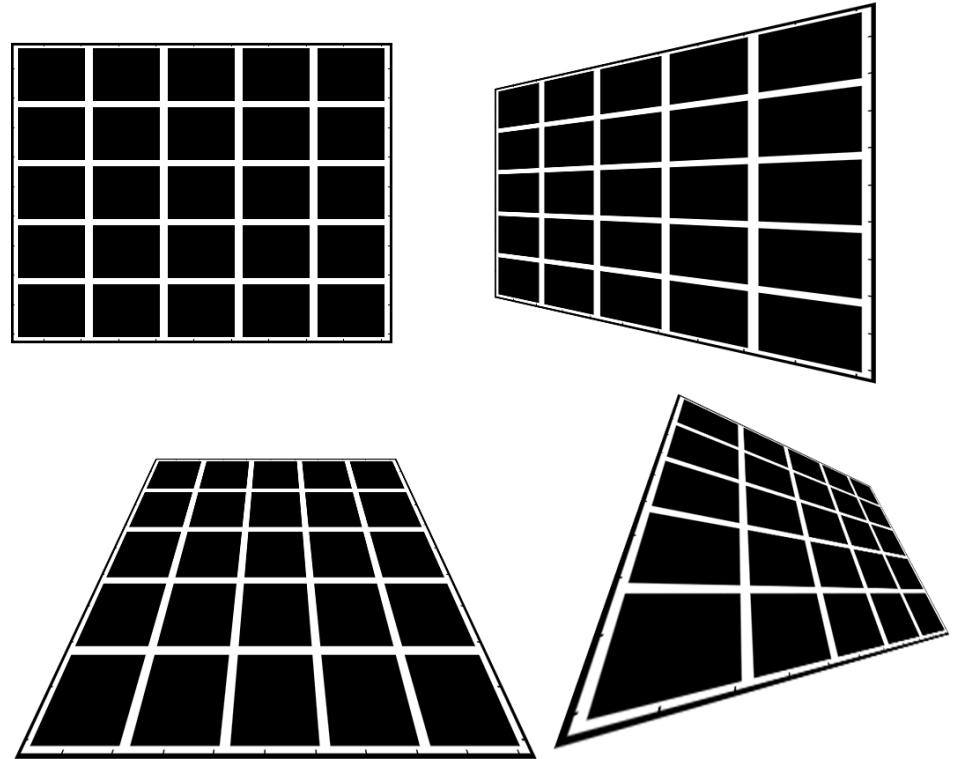
- Now consider relationship between *two* images of a plane



- Two images of a plane taken by a perspective camera are related by a homography

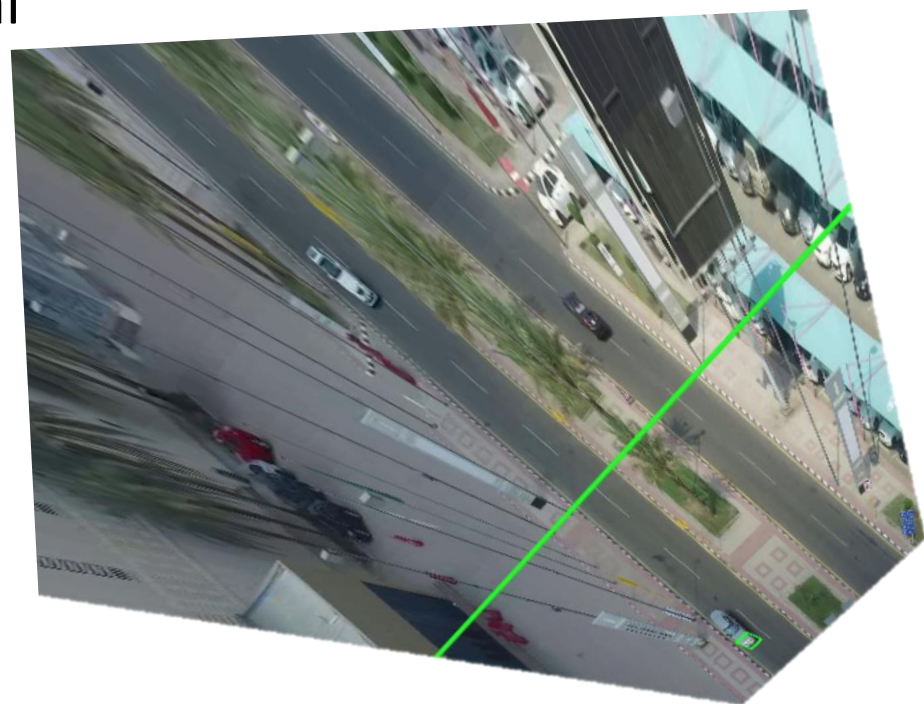
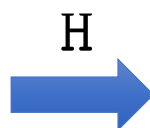
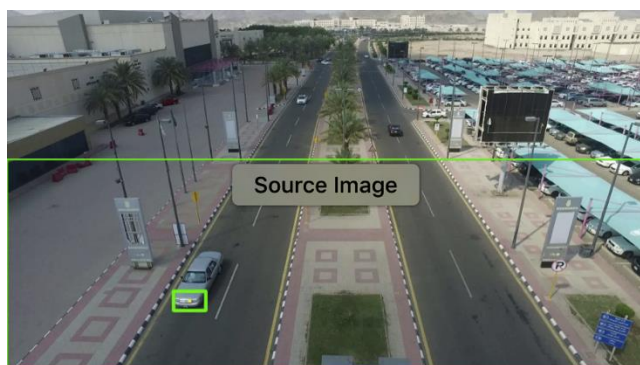
# Special Case 1: Perspective Camera Looking at a Plane

- It is no surprise that these projective transformations look like images of a plane taken from different angles



# Application: Rectification

- A perspective image of a plane can be transformed into one in which the plane is fronto-parallel, i.e. the optical axis coincides with the plane normal



Rectified image



H

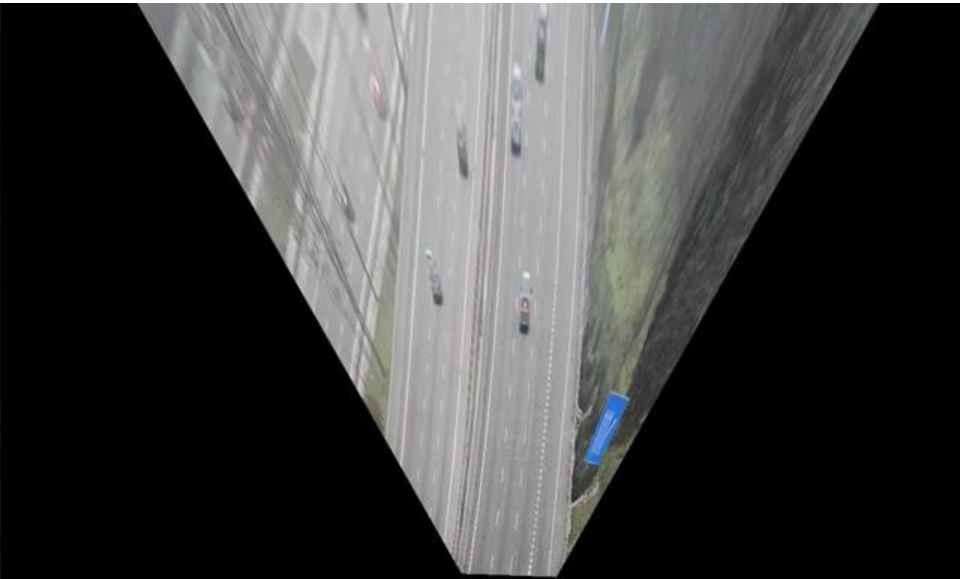


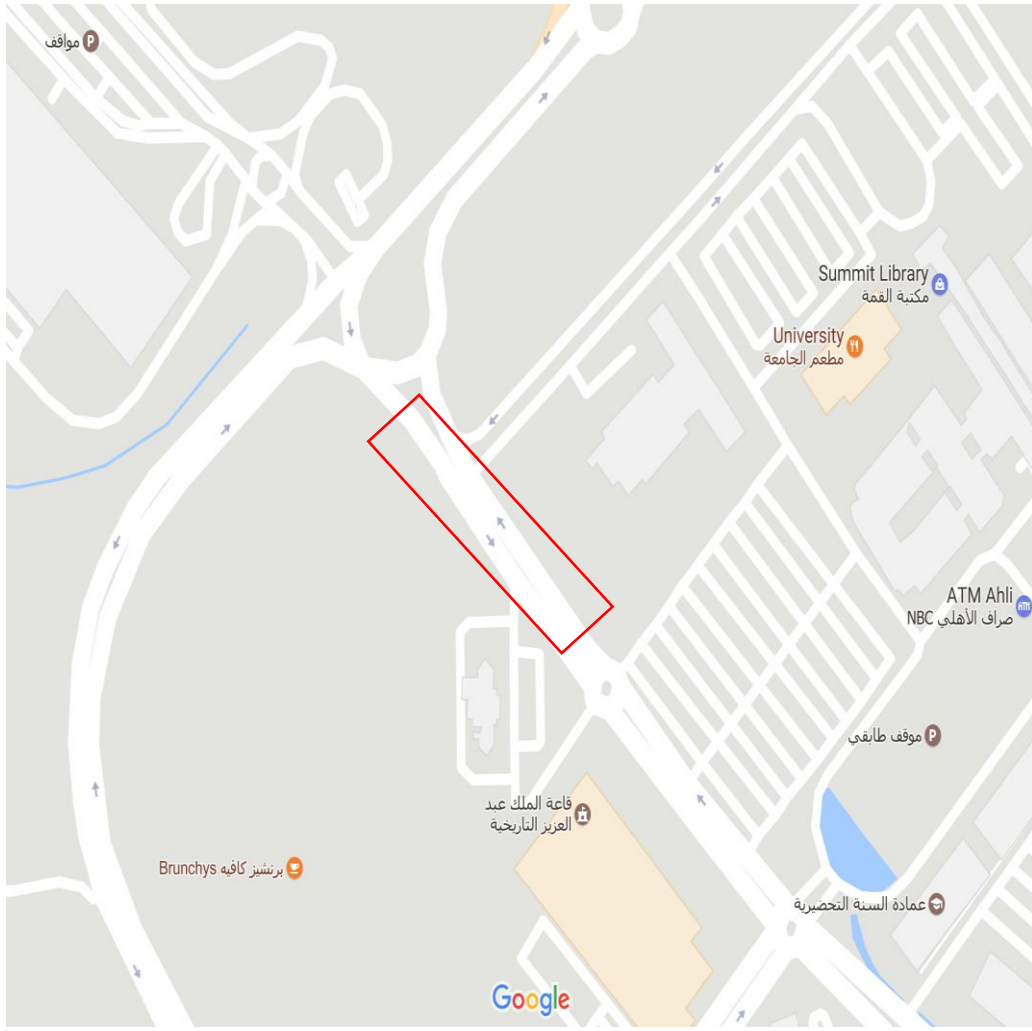
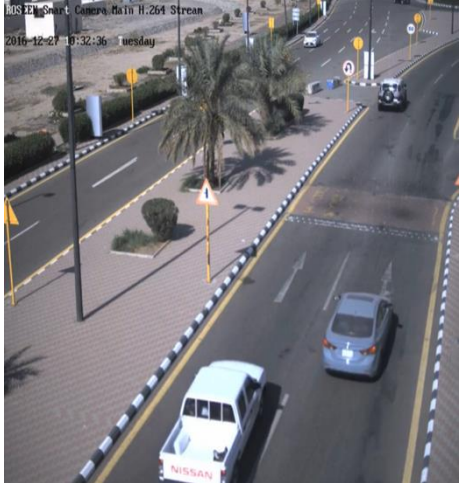
Satellite image





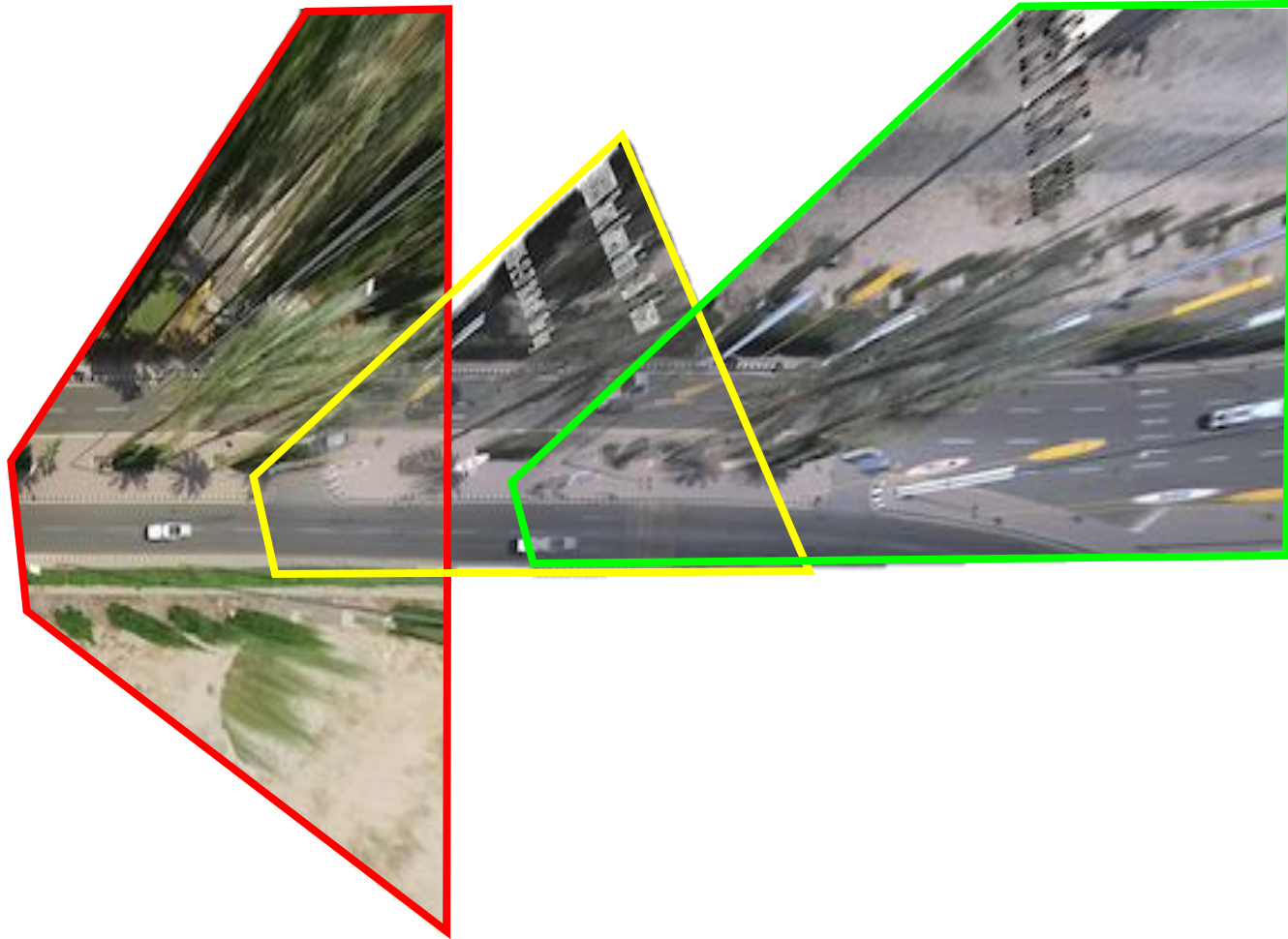
# Application: Rectification





Road Segment, covered by three cameras







Merged view of the road

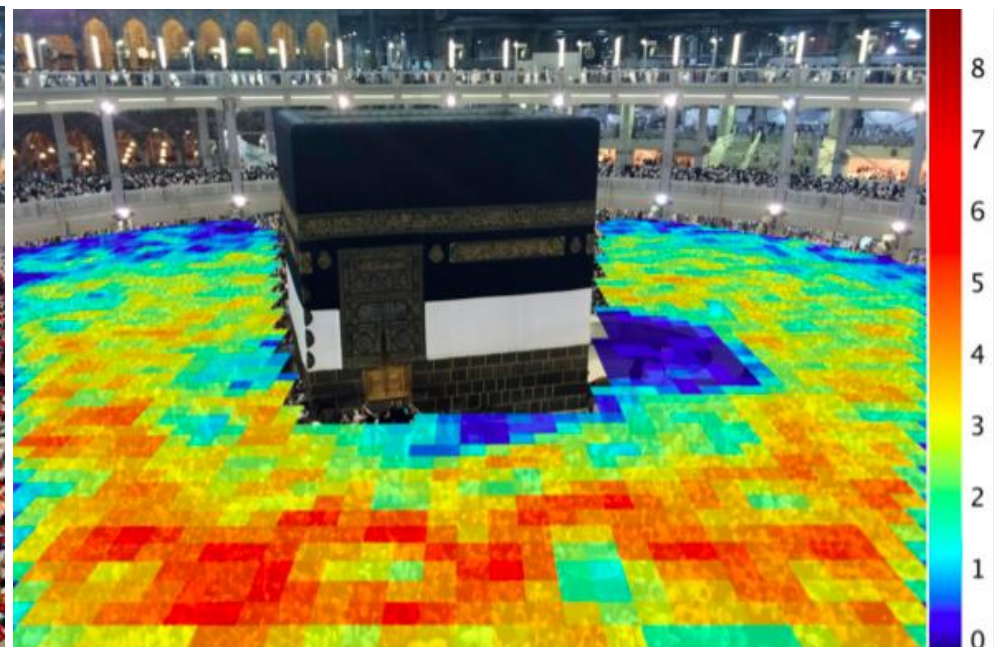






# Another Example of Rectification

- Measuring crowd density as persons per square meter



## Special Case 2: Rotation about Camera Center (Pure Rotation)

- Consider the case of camera that does not translate but only rotates about its optical center
- This scenario also occurs frequently in imaging applications



## Special Case 2:

### Rotation about Camera Center (Pure Rotation)

- $\mathbf{x}$  and  $\mathbf{x}'$  are images of a point  $\mathbf{X}$  before and after rotation of  $\mathbf{R}$  the camera respectively.

- Then

$$\mathbf{x} = \mathbf{K} [\mathbf{I} \mid \mathbf{0}] \mathbf{X}$$

$$\mathbf{x}' = \mathbf{K} [\mathbf{R} \mid \mathbf{0}] \mathbf{X}$$

$$= \mathbf{K}\mathbf{R} [\mathbf{I} \mid \mathbf{0}] \mathbf{X}$$

$$= \mathbf{K}\mathbf{R}\mathbf{K}^{-1}\mathbf{K} [\mathbf{I} \mid \mathbf{0}] \mathbf{X}$$

$$= \mathbf{K}\mathbf{R}\mathbf{K}^{-1}\mathbf{x}$$

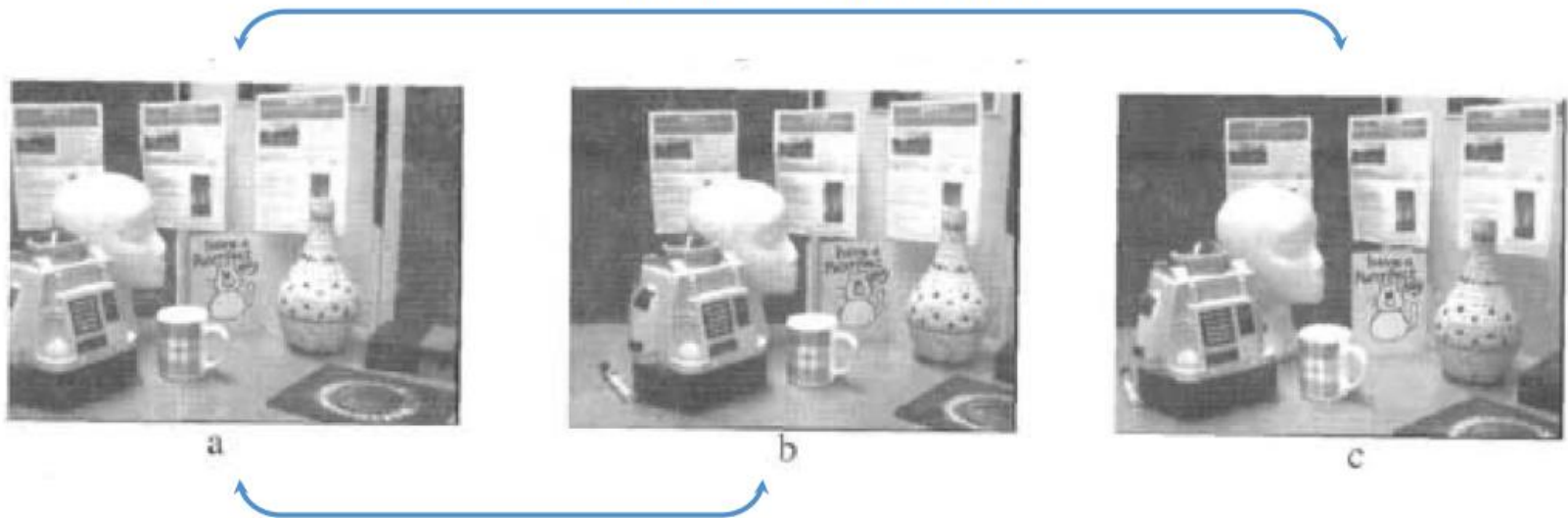
- Hence  $\mathbf{x}' = \mathbf{K}\mathbf{R}\mathbf{K}^{-1}\mathbf{x}$   $\mathbf{x}' = \mathbf{H}\mathbf{x}$

- This type of homography is called a *conjugate rotation* homography



# Special Case 2: Rotation about Camera Center (Pure Rotation)

Rotation + Translation



Pure Rotation

- The relative motion of objects in two images which are at different distances in the world is termed parallax
- A homography will not generate any parallax.
- Hence pure rotation of camera does not generate parallax



# Camera Calibration





# Recall: Perspective Camera Model

- The perspective camera model can be written as

$$\mathbf{x} = \mathbf{K} \left[ \mathbf{R} \mid -\mathbf{R}\tilde{\mathbf{C}} \right] \mathbf{X}$$

Image point  $\in \mathbb{P}^2$

3x3 matrix of internal camera parameters (intrinsic parameters)

3x4 matrix of external camera parameters (extrinsic parameters)

World point  $\in \mathbb{P}^3$

$$\mathbf{K} = \begin{bmatrix} m_x f & 0 & x_0 \\ 0 & m_y f & y_0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$x_0 = m_x p_x, \quad y_0 = m_y p_y$$



# Camera Calibration

- In general, the camera model looks like:

$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$
$$\mathbf{x} = \mathbf{P}\mathbf{X}$$

- $\mathbf{P}$  is a 3x4 matrix of rank 3
- Calibration is the process of finding the parameters  $[p_{11} \dots p_{34}]$
- If  $\mathbf{x}$  and  $\mathbf{X}$  are known, then we can solve for the unknown parameters in  $\mathbf{P}$



# Camera Calibration

- Camera model

$$\begin{bmatrix} hx \\ hy \\ h \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

- In inhomogeneous form

$$x = \frac{hx}{h} = \frac{p_{11}X + p_{12}Y + p_{13}Z + p_{14}}{p_{31}X + p_{32}Y + p_{33}Z + p_{34}} \quad y = \frac{hy}{h} = \frac{p_{21}X + p_{22}Y + p_{23}Z + p_{24}}{p_{31}X + p_{32}Y + p_{33}Z + p_{34}}$$

- Multiplying both sides by denominator and rearranging

$$p_{11}X + p_{12}Y + p_{13}Z + p_{14} - p_{31}Xx - p_{32}Yx - p_{33}Zx - p_{34}x = 0$$

$$p_{21}X + p_{22}Y + p_{23}Z + p_{24} - p_{31}Xy - p_{32}Yy - p_{33}Zy - p_{34}y = 0$$

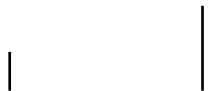


# Camera Calibration

$$p_{11}X + p_{12}Y + p_{13}Z + p_{14} - p_{31}Xx - p_{32}Yx - p_{33}Zx - p_{34}x = 0$$

$$p_{21}X + p_{22}Y + p_{23}Z + p_{24} - p_{31}Xy - p_{32}Yy - p_{33}Zy - p_{34}y = 0$$

- These equations have 12 unknowns
- Each correspondence between a world point and an image point yields two equations
- If 6 correspondences are known, we can solve for the unknowns



# Camera Calibration

$$p_{11}X + p_{12}Y + p_{13}Z + p_{14} - p_{31}Xx - p_{32}Yx - p_{33}Zx - p_{34}x = 0$$

$$p_{21}X + p_{22}Y + p_{23}Z + p_{24} - p_{31}Xy - p_{32}Yy - p_{33}Zy - p_{34}y = 0$$

- Separating out the known and the unknown terms

$$\begin{bmatrix} X_1 & Y_1 & Z_1 & 1 & 0 & 0 & 0 & 0 & -x_1X_1 & -x_1Y_1 & -x_1Z_1 & -x_1 \\ 0 & 0 & 0 & 0 & X_1 & Y_1 & Z_1 & 1 & -y_1X_1 & -y_1Y_1 & -y_1Z_1 & -y_1 \end{bmatrix} \begin{bmatrix} p_{11} \\ p_{12} \\ p_{13} \\ p_{14} \\ p_{21} \\ p_{22} \\ p_{23} \\ p_{24} \\ p_{31} \\ p_{32} \\ p_{33} \\ p_{34} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$



# Camera Calibration

- Given  $n$  correspondences...

$$\begin{bmatrix}
 X_1 & Y_1 & Z_1 & 1 & 0 & 0 & 0 & 0 & -x_1 X_1 & -x_1 Y_1 & -x_1 Z_1 & -x_1 \\
 0 & 0 & 0 & 0 & X_1 & Y_1 & Z_1 & 1 & -y_1 X_1 & -y_1 Y_1 & -y_1 Z_1 & -y_1 \\
 X_2 & Y_2 & Z_2 & 1 & 0 & 0 & 0 & 0 & -x_2 X_2 & -x_2 Y_2 & -x_2 Z_2 & -x_2 \\
 0 & 0 & 0 & 0 & X_2 & Y_2 & Z_2 & 1 & -y_2 X_2 & -y_2 Y_2 & -y_2 Z_2 & -y_2 \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
 X_n & Y_n & Z_n & 1 & 0 & 0 & 0 & 0 & -x_n X_n & -x_n Y_n & -x_n Z_n & -x_n \\
 0 & 0 & 0 & 0 & X_n & Y_n & Z_n & 1 & -y_n X_n & -y_n Y_n & -y_n Z_n & -y_n
 \end{bmatrix}
 \begin{bmatrix}
 p_{11} \\
 p_{12} \\
 p_{13} \\
 p_{14} \\
 p_{21} \\
 p_{22} \\
 p_{23} \\
 p_{24} \\
 p_{31} \\
 p_{32} \\
 p_{33} \\
 p_{34}
 \end{bmatrix}
 =
 \begin{bmatrix}
 0 \\
 0 \\
 0 \\
 0 \\
 \vdots \\
 0 \\
 0
 \end{bmatrix}$$

$$\mathbf{Ap} = \mathbf{0}$$

# Camera Calibration

- This system  $Ap = 0$  is a homogeneous system.
- A is rank deficient:  $\text{rank}(A) = 11$  (at most)
- Solution?
  - The null vector of A represents the p which is the solutions to the system  $Ap = 0$
  - How to find null space?
    1.  $\text{null}(A)$  in MATLAB, or
    2. Take SVD of A, as  $\text{svd}(C) = USV^T$ . The column of V corresponding to the singular value of zero represents the solution  
(in practice, you will have to take the smallest singular value)



# Camera Calibration: Summary

- Given a set of world points (in 3D coordinates) and their corresponding image points, we solve for the 3x4 camera matrix that relates them.
- This transforms into a problem of the form  $Ap = 0$ , which can be solved by finding the null vector of **A**.
- A more robust solution is through Direct Linear Transform, DLT (not covered in this class)





# Camera Calibration: Solving for Extrinsic and Intrinsic Parameters

- After finding  $\mathbf{p}$ , we end up with a 3x4 camera matrix relating world points to image points

$$\mathbf{x} = \mathbf{P}\mathbf{X}$$
$$\mathbf{P} = \mathbf{K}\mathbf{R} \left[ \mathbf{I} \mid -\tilde{\mathbf{C}} \right]$$

- How can I find camera rotation, translation and intrinsic parameters?
- Note that  $\mathbf{P}$  has 12 terms and 11 degrees of freedom.

$$\mathbf{P} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix}$$

# Camera Calibration: Solving for Extrinsic and Intrinsic Parameters

- Solving for Camera Center  $C$ :
- Consider  $P$  times  $C$

$$PC = KR \left[ \mathbf{I} \mid -\tilde{C} \right] C = ?$$

$$PC = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

- Hence, camera center  $C$  is a null vector of  $P$
- Note that  $PC = (0, 0, 0)^T$  is undefined in image plane, which is exactly what we should expect, since image of camera center is undefined.



# Camera Calibration: Solving for Extrinsic and Intrinsic Parameters

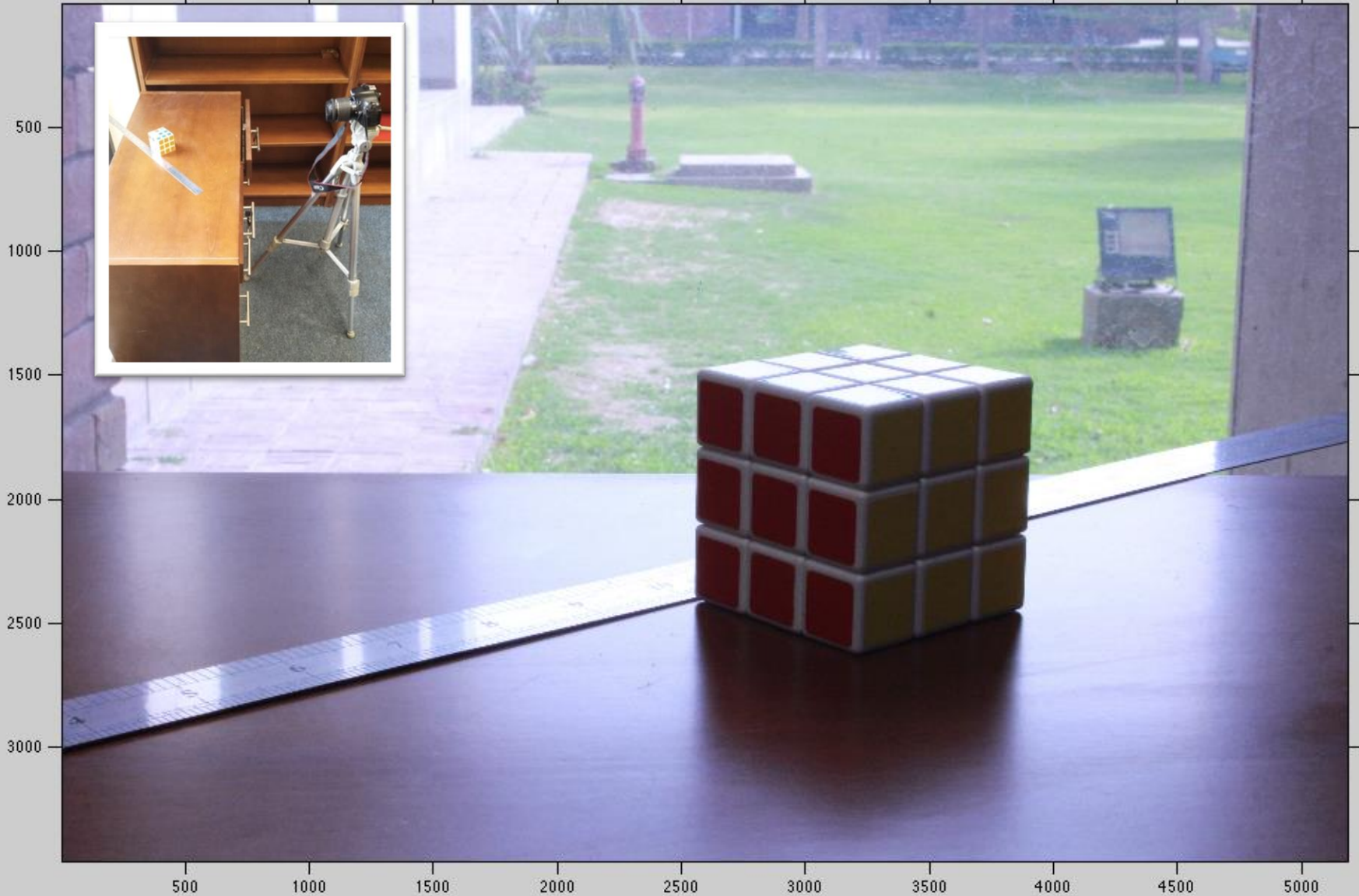
- Solving for **K** and **R**
- Note that  $\mathbf{P} = [\mathbf{KR} \mid -\mathbf{KRC}]$
- Hence first 3x3 block of P i.e.  $M_{3 \times 3} = \mathbf{KR}$
- K is an upper triangular matrix, R is an orthonormal matrix
- Solved through RQ decomposition  
RQ decomposition decomposes a matrix into an upper triangular matrix times an orthonormal matrix



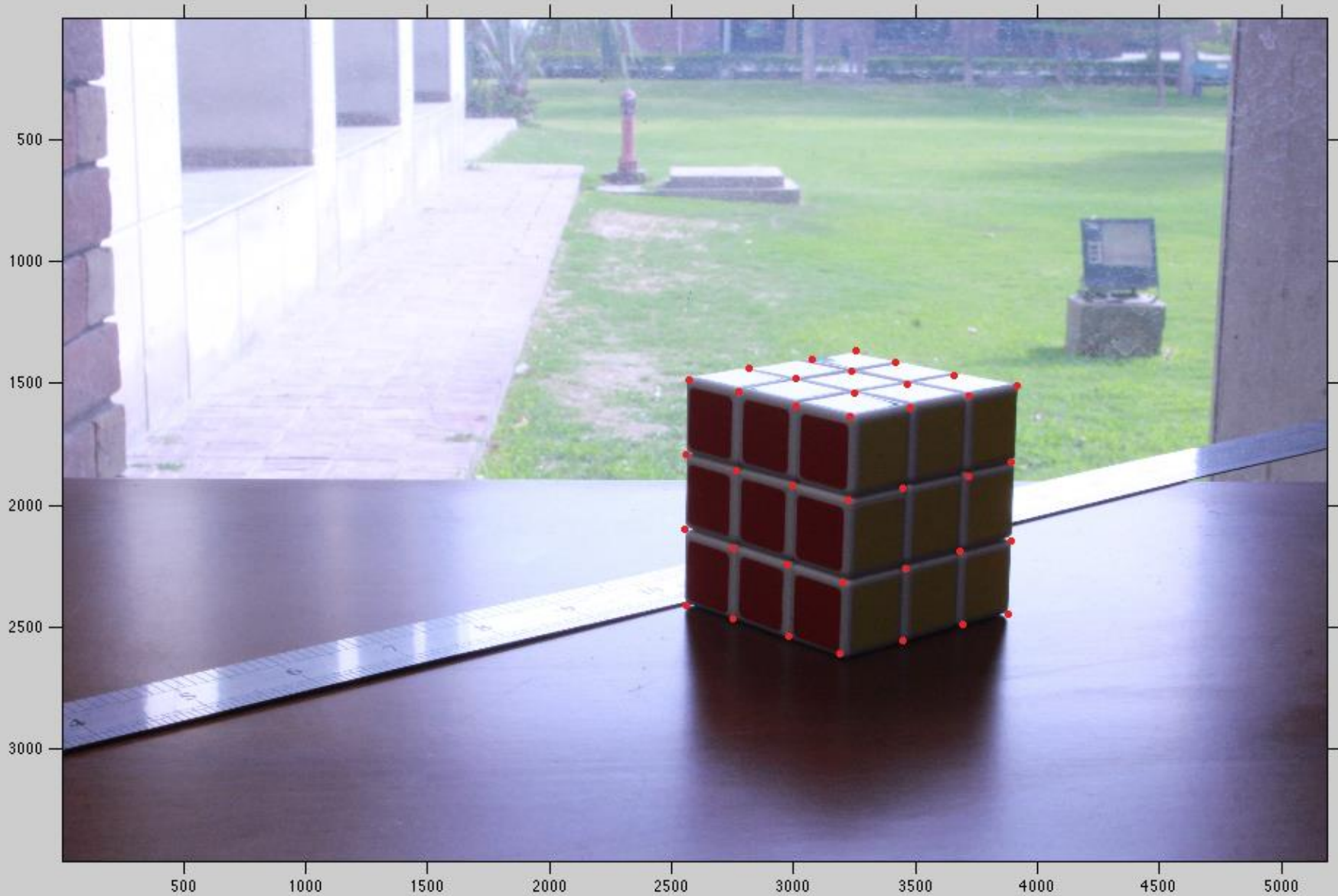
# Camera Calibration Example



# Take Image of a Calibration Target

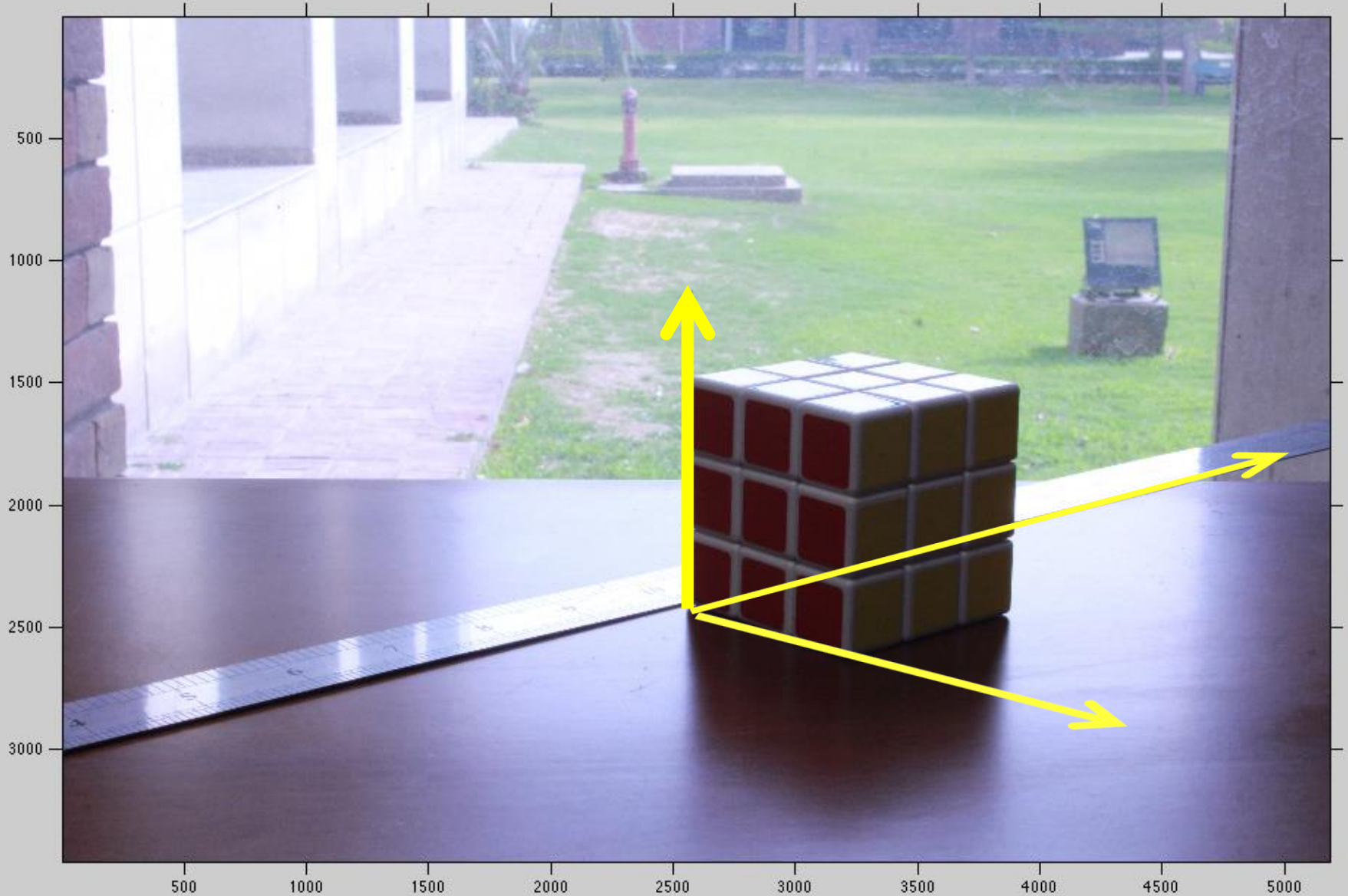


# Select Image Points





# Choose world coordinate system









# Find the solution

- Set up matrix A and find its right null vector through SVD
  - warning: `null(A)` is unlikely to work! (WHY?)
- Reshape into a 3x4 matrix

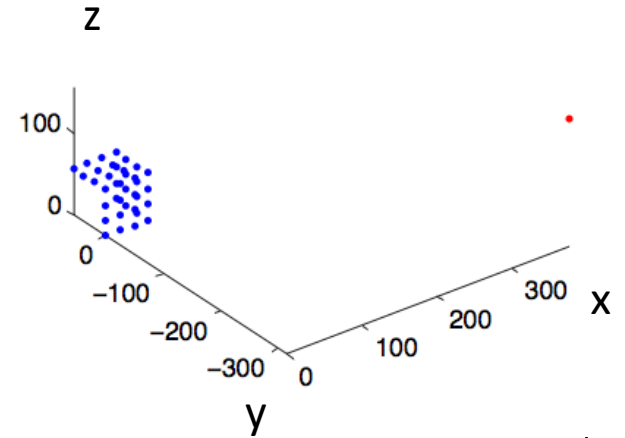
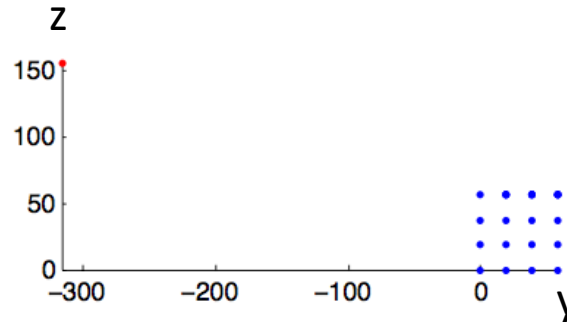
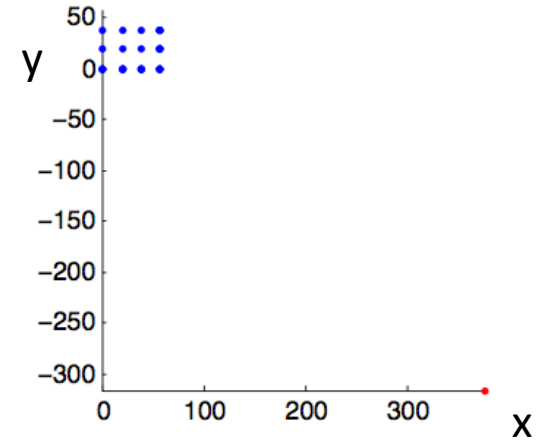
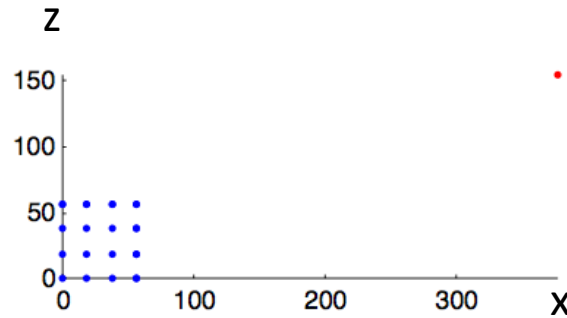
P =

```
-0.00010835    4.3034e-05    0.0047453    -0.68373
-0.0019211    -0.0044849    0.00023615    -0.7297
4.1144e-07    -3.5796e-07    1.1421e-07    -0.00028537
```



# Find camera center from P

$$C = \begin{pmatrix} 375.89 \\ -315.53 \\ 155.53 \\ 1 \end{pmatrix}$$



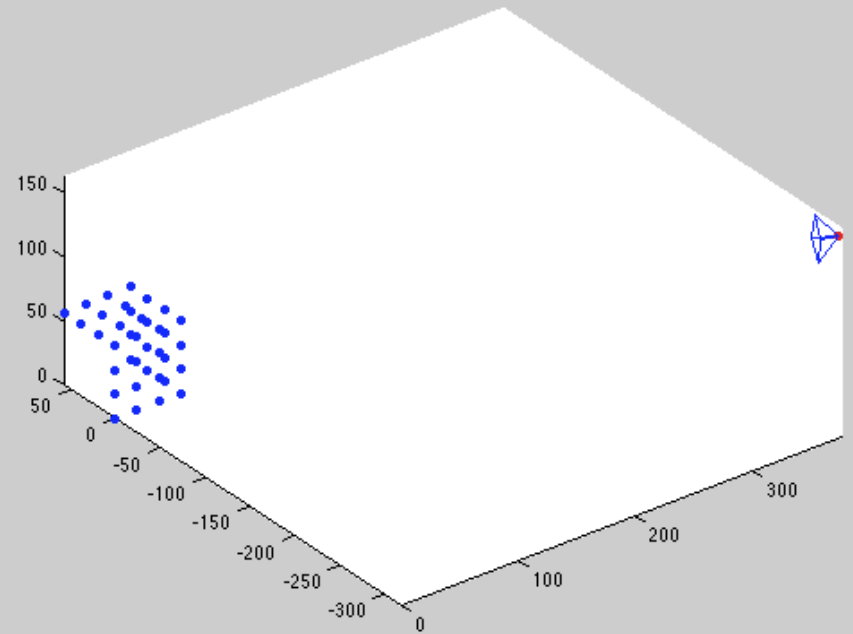
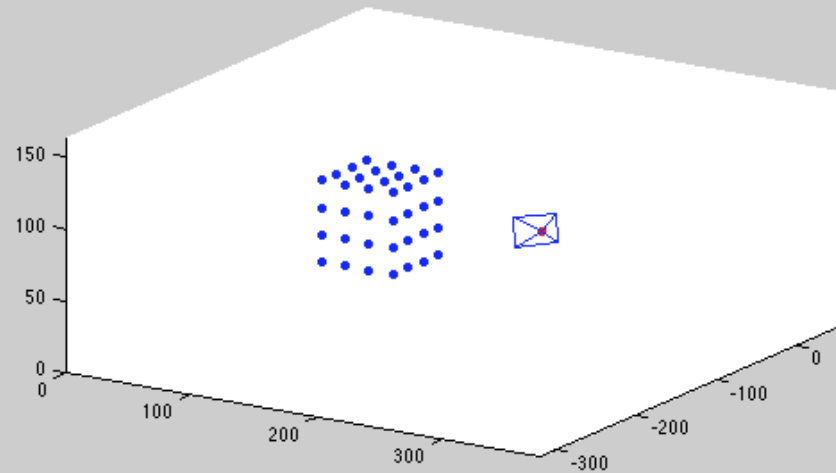
# Compute K and R by RQ Factorization

K =

-8376.2	66.191	1552.4
0	-8336.6	2712
0		1

R =

-0.16524	0.12231	0.97864
-0.65379	-0.75651	-0.01584
-0.73842	0.64244	-0.20497



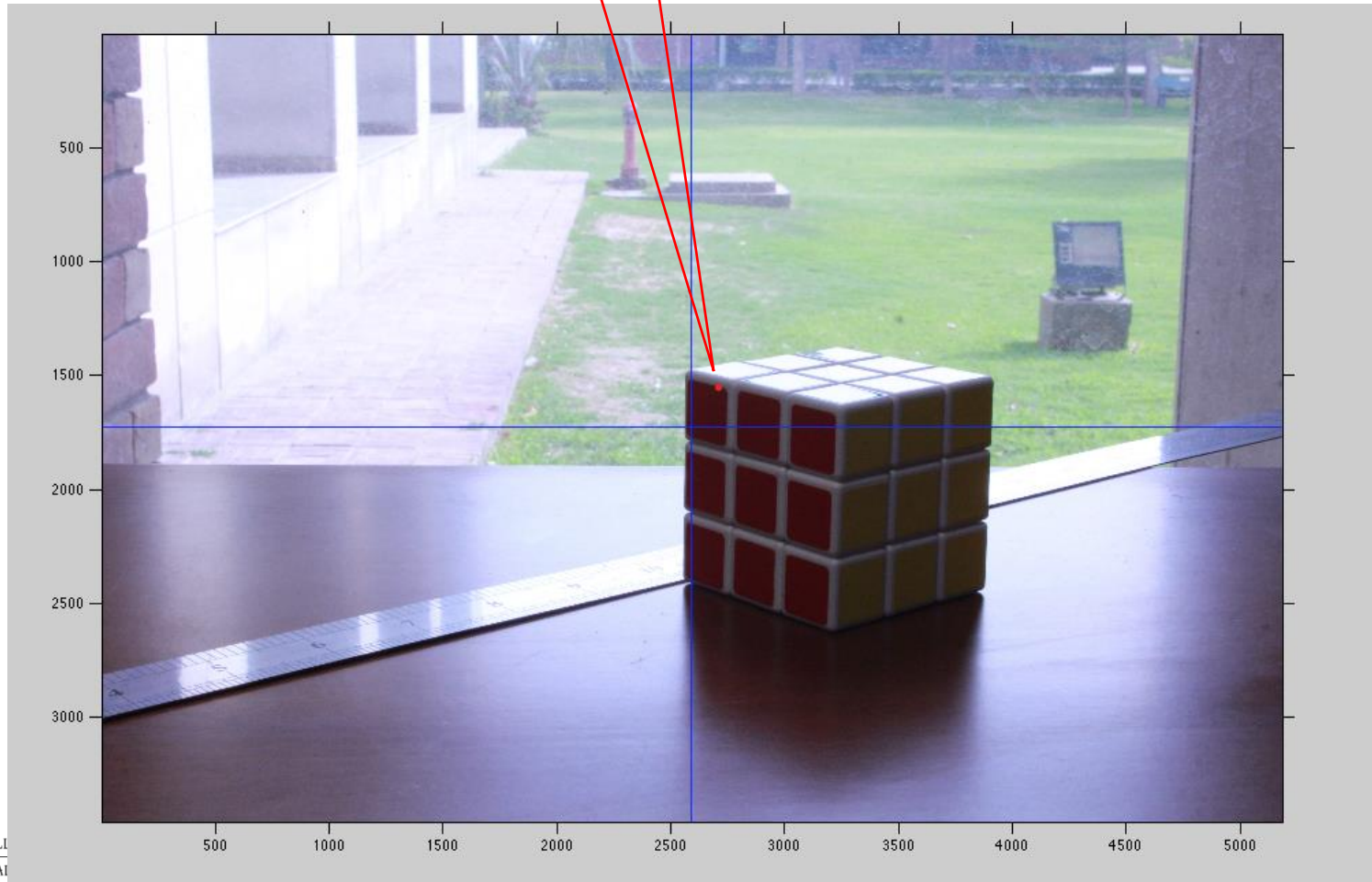
# Principal Point

$K =$

-8376.2	66.191	1552.4
0	-8336.6	2712
0		1

$R =$

-0.16524	0.12231	0.97864
-0.65379	-0.75651	-0.01584
-0.73842	0.64244	-0.20497



# Focal Length

K =

-8376.2	66.191	1552.4
0	-8336.6	2712
0		1

R =

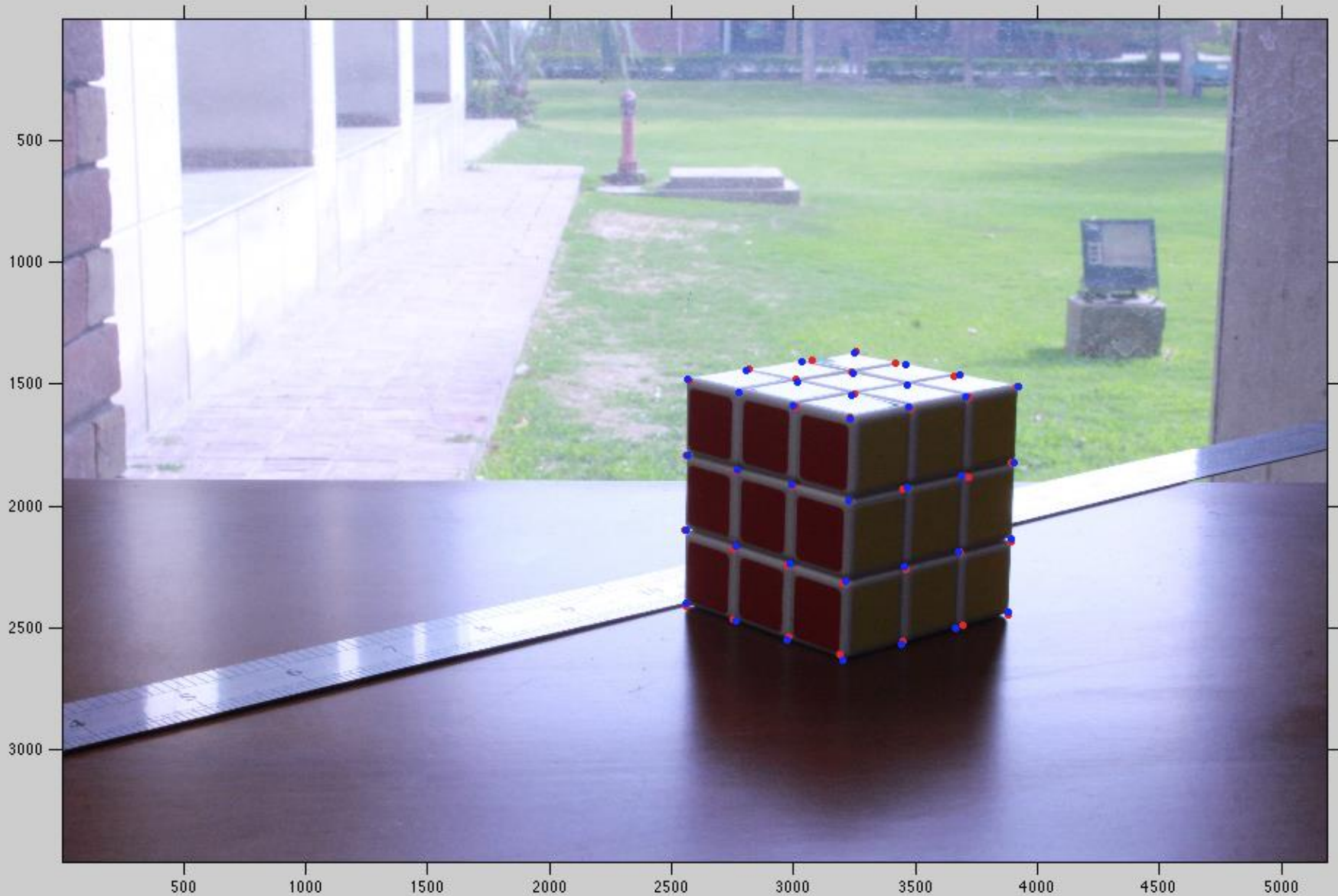
-0.16524	0.12231	0.97864
-0.65379	-0.75651	-0.01584
-0.73842	0.64244	-0.20497

$$m_x f = 8376.2 \quad m_y f = 8336.6$$

- Image was taken by Canon 600D
- Canon 600D uses APS-C format CMOS sensor\*
- Size of APS-C sensor is 23.6mm x 15.7mm
- Image size is 5184 x 3456
- Hence
- $m_x = 3456/15.7 = 220.13$  pix/mm
- $m_y = 5184/23.6 = 219.66$  pix/mm
- $f = [38.051, 37.952]$  mm
- Verification from EXIF data: 37mm
- Angle between x & y axis =  $90.455^\circ$



# Back-project World Points into Image



# Camera Anatomy: Column Vectors of P

- Let the columns of P be  $p_i$ ,  $i = 1, \dots, 4$
- Then  $p_1, p_2, p_3$  are the points at infinity of the world X, Y and Z axes respectively.
- Consider the point at infinity along X-axis:  $D = (1, 0, 0, 0)^T$

- This will be imaged at

$$PD = p_1$$

- Hence,
  - The first column of P is the image of the point at infinity along X-axis
  - The second column of P is the image of the point at infinity along Y-axis
  - The third column of P is the image of the point at infinity along Z-axis
- Similarly,  $p_4$  the fourth column of P is ...
- the image of world origin  $(0,0,0,1)^T$



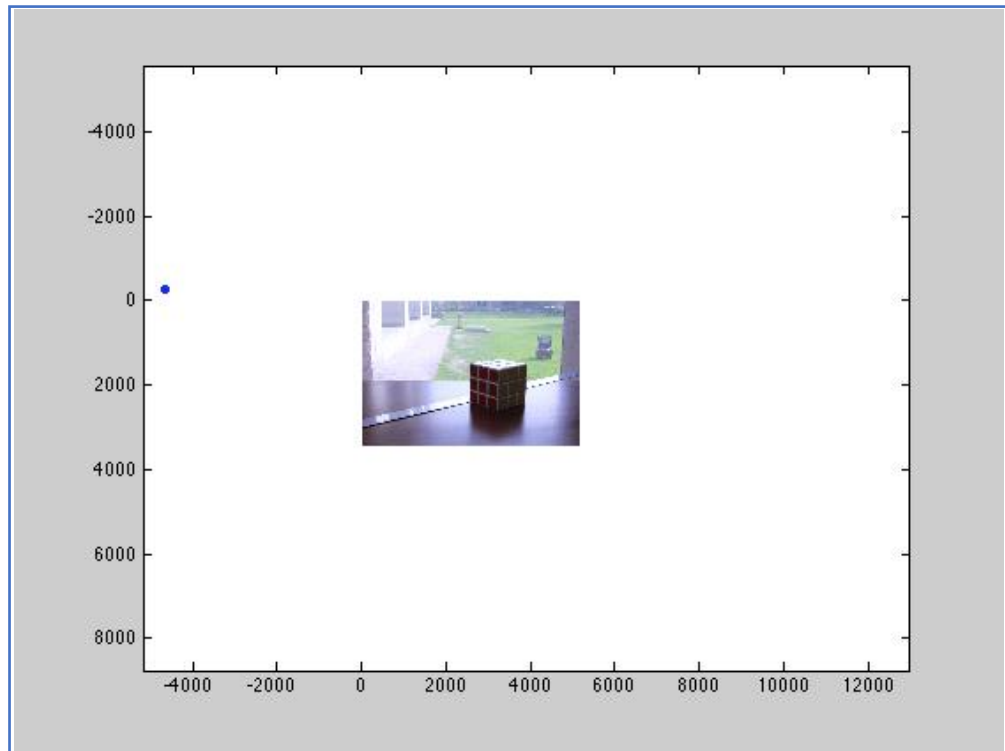


# Camera Anatomy: Column Vectors of P

P =

-0.00010835	4.3034e-05	0.0047453	-0.68373
-0.0019211	-0.0044849	0.00023615	-0.7297
4.1144e-07	-3.5796e-07	1.1421e-07	-0.00028537

-263.35  
-4669.2  
1

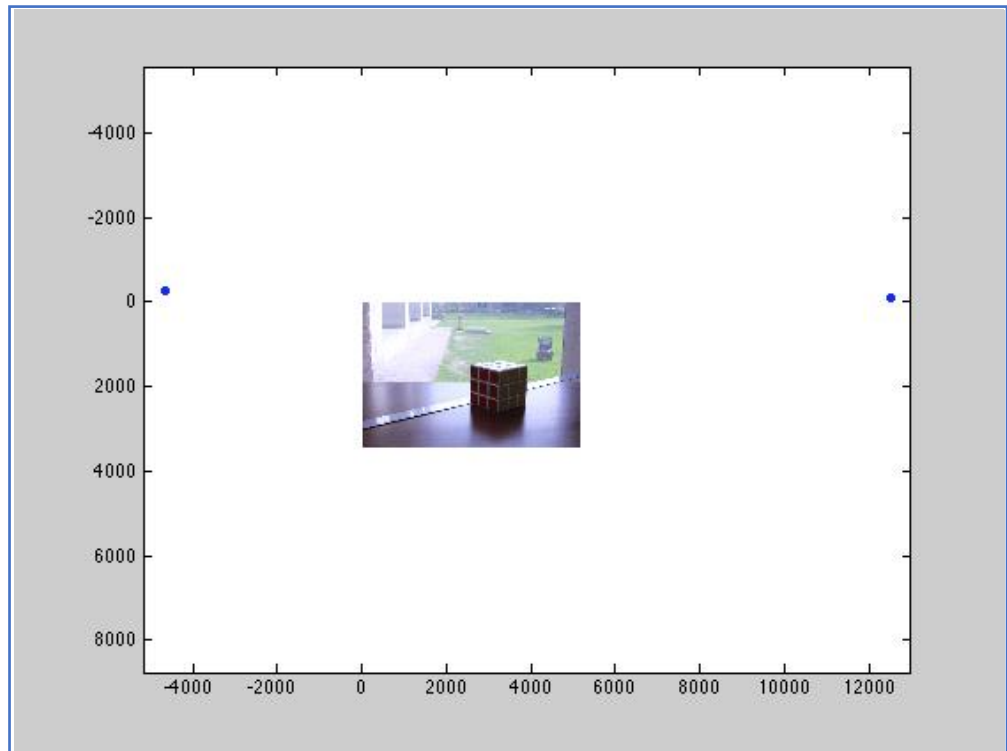


# Camera Anatomy: Column Vectors of P

P =

-0.00010835	4.3034e-05	0.0047453	-0.68373
-0.0019211	-0.0044849	0.00023615	-0.7297
4.1144e-07	-3.5796e-07	1.1421e-07	-0.00028537

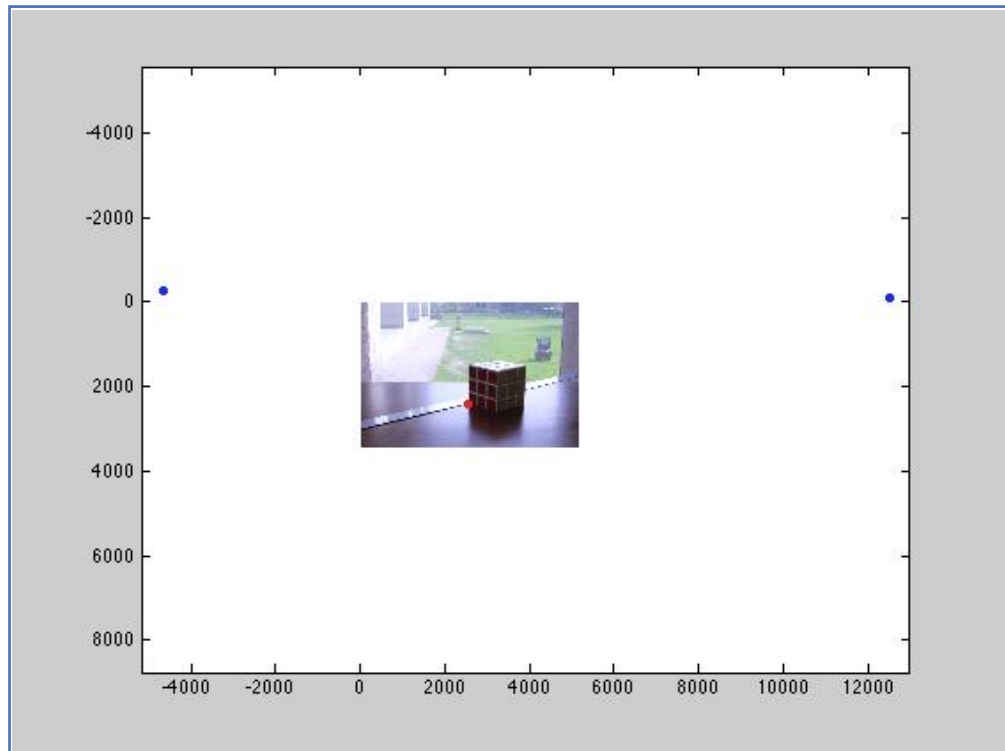
-263.35	-120.22
-4669.2	12529
1	1



# Camera Anatomy: Column Vectors of P

P =

-0.00010835	4.3034e-05	0.0047453	-0.68373	2396
-0.0019211	-0.0044849	0.00023615	-0.7297	2557
4.1144e-07	-3.5796e-07	1.1421e-07	-0.00028537	1

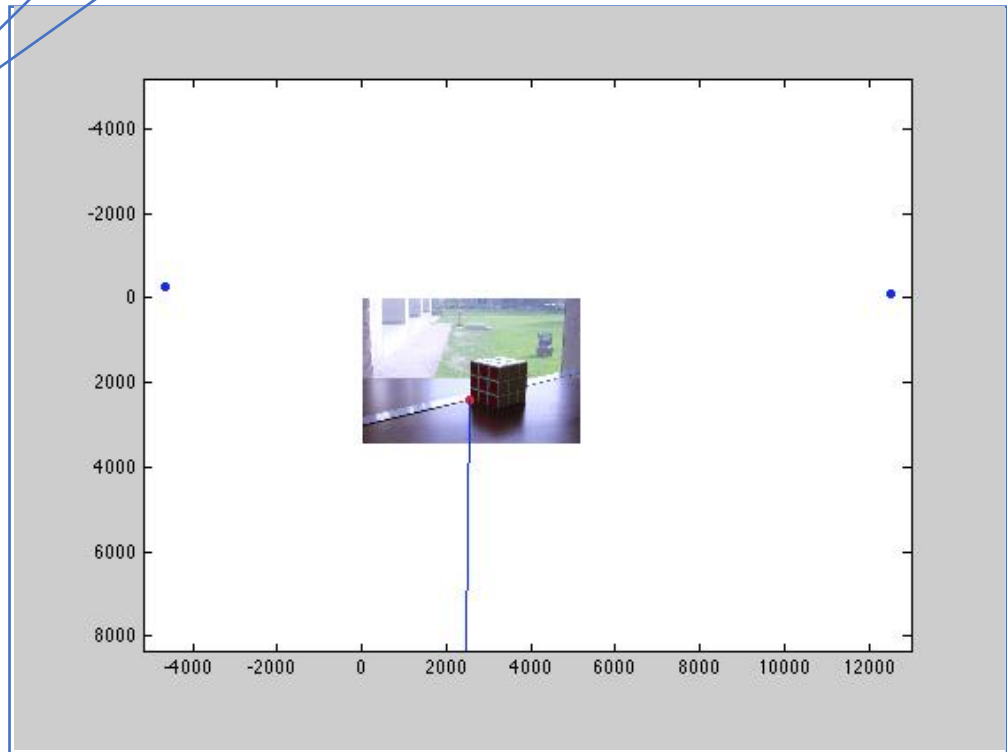


# Camera Anatomy: Column Vectors of P

P =

-0.00010835	4.3034e-05	0.0047453	-0.68373
-0.0019211	-0.0044849	0.00023615	-0.7297
4.1144e-07	-3.5796e-07	1.1421e-07	-0.00028537

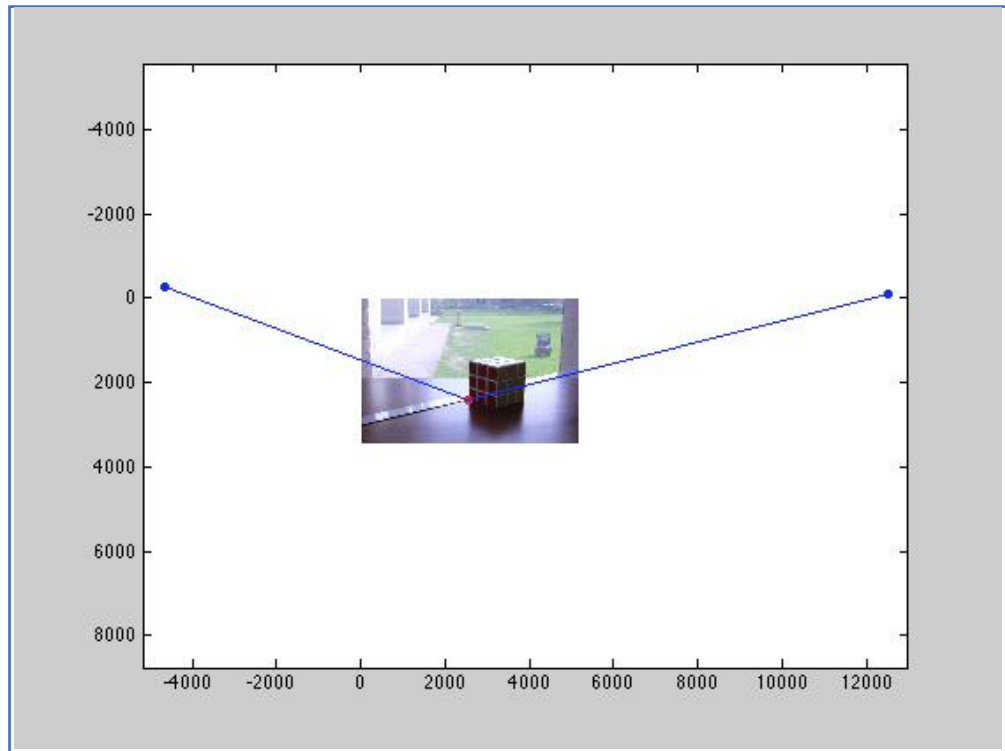
41550  
2067.7  
1



# Camera Anatomy: Column Vectors of P

P =

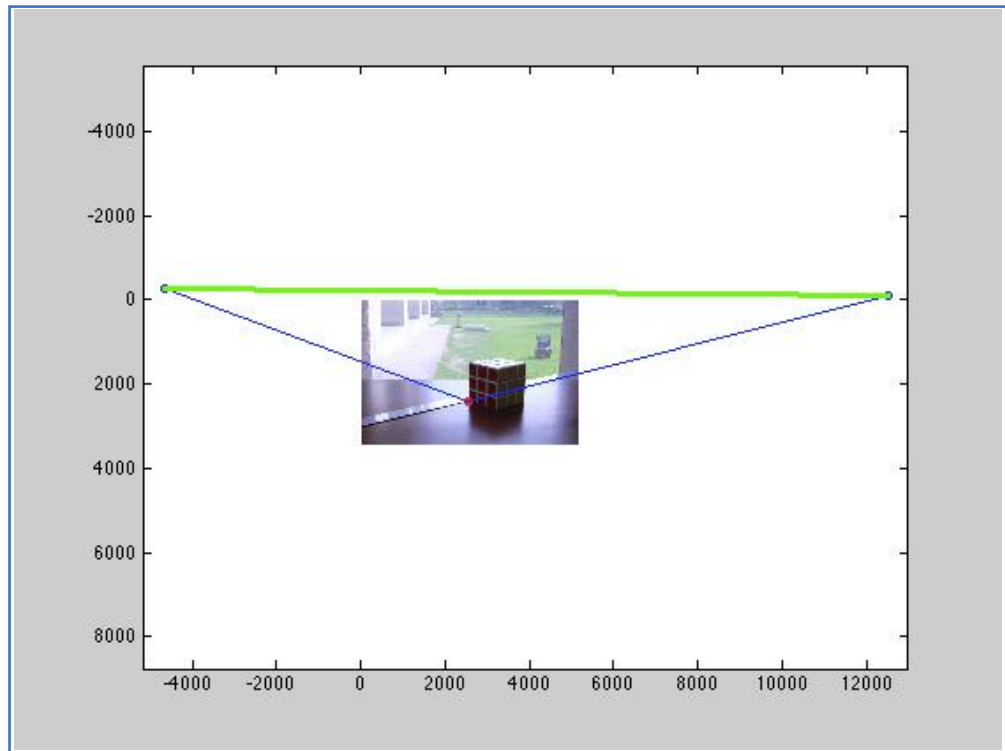
-0.00010835	4.3034e-05	0.0047453	-0.68373
-0.0019211	-0.0044849	0.00023615	-0.7297
4.1144e-07	-3.5796e-07	1.1421e-07	-0.00028537

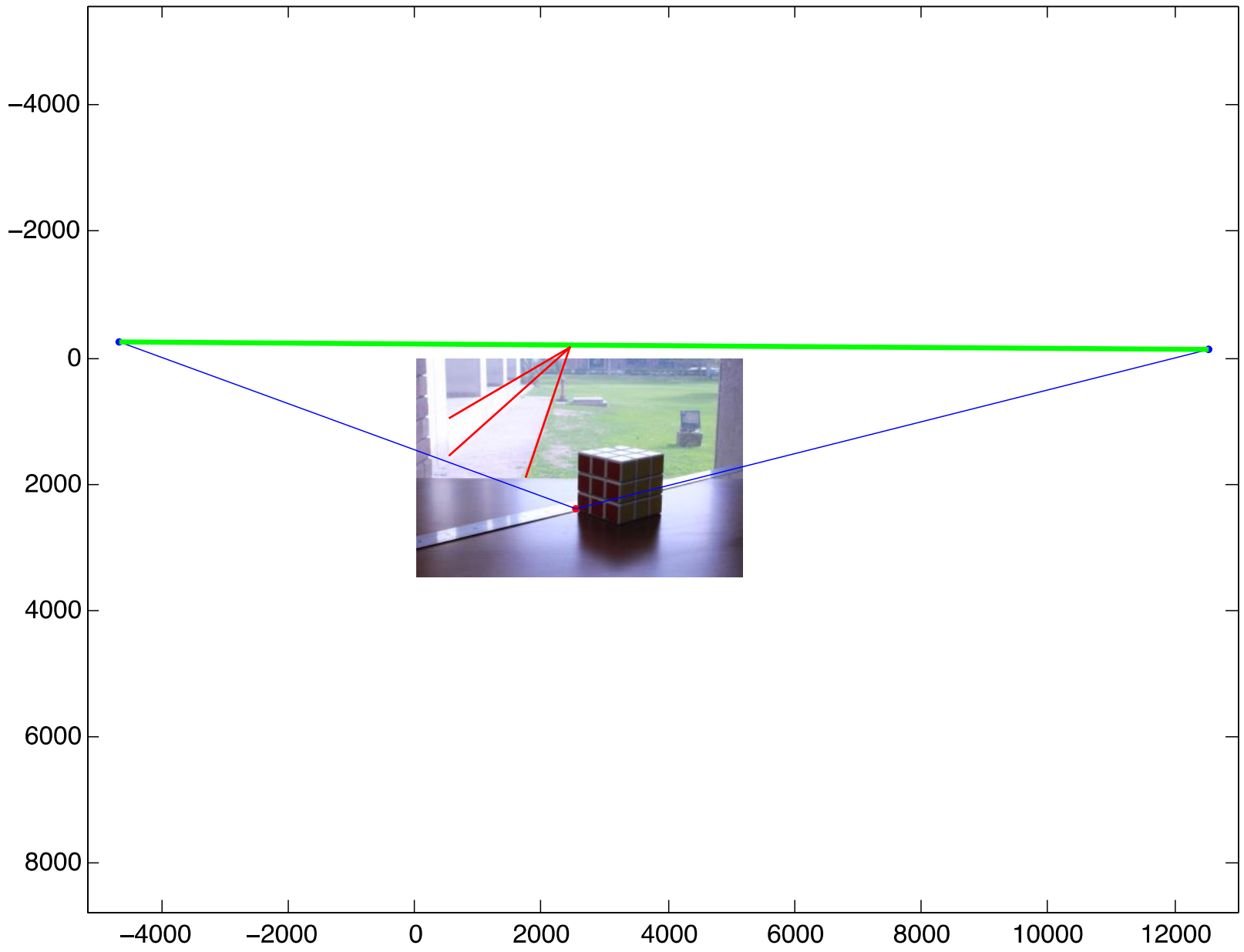


# Camera Anatomy: Column Vectors of $P$

$P =$

-0.00010835	4.3034e-05	0.0047453	-0.68373
-0.0019211	-0.0044849	0.00023615	-0.7297
4.1144e-07	-3.5796e-07	1.1421e-07	-0.00028537





# Camera Anatomy: Row Vectors of $\mathbf{P}$

- Row vectors are 4-vectors, which may be interpreted as planes.

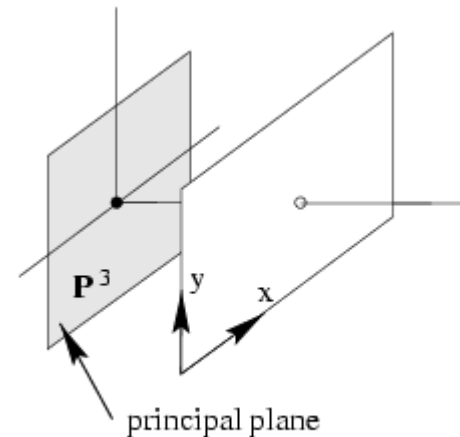
$$\mathbf{P} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix} = \begin{bmatrix} \mathbf{P}^{1\top} \\ \mathbf{P}^{2\top} \\ \mathbf{P}^{3\top} \end{bmatrix}$$





# Camera Anatomy: Row Vectors of $P$

- **Principal plane:** Plane through camera center, parallel to image plane, consisting of set of points  $X$  imaged on, line at infinity of image,
- i.e.  $PX = (x,y,0)^T$
- Thus, a point lies on principal plane iff  $P^3 X = 0$
- Thus,  $P^3$  is the vector representing the principal plane
- Also,  $C$  lies on  $P^3$  (verify)



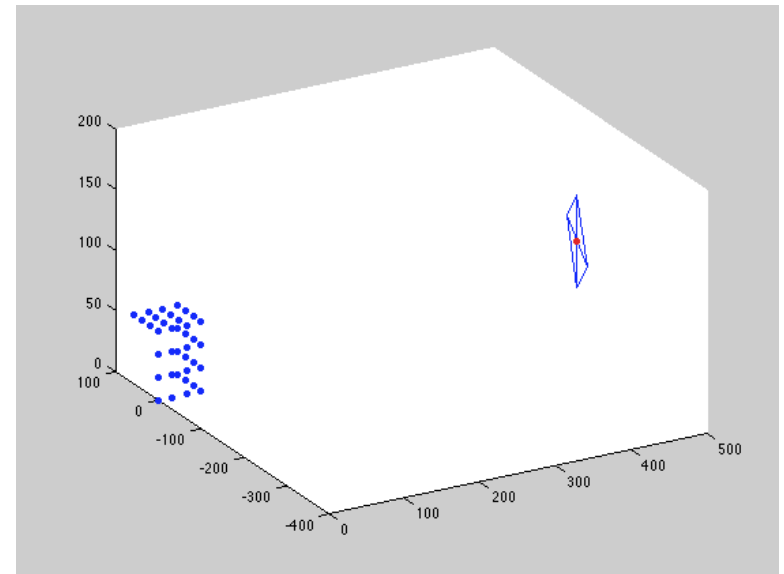
# Camera Anatomy: Row Vectors of $P$

$P =$

```
-0.00010835    4.3034e-05    0.0047453    -0.68373
-0.0019211    -0.0044849    0.00023615   -0.7297
4.1144e-07    -3.5796e-07   1.1421e-07   -0.00028537
```

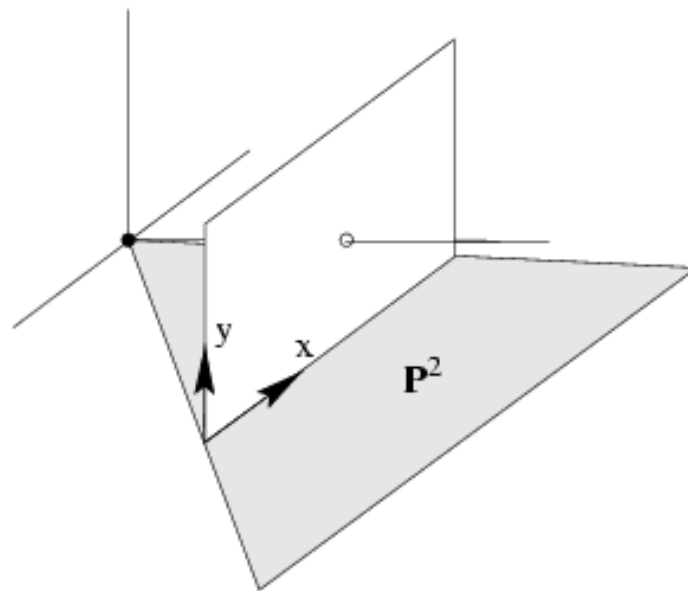
$$P^3 = [4.1144e-07 \quad -3.5796e-07 \quad 1.1421e-07 \quad -0.00028537]^T$$

Verify  $P^{3T}C = 1.3212e-14$



# Camera Anatomy: Row Vectors of $P$

- **Axis planes:** Consider the set of points  $X$  in  $P^1$
- They must satisfy  $P^{1T}X = 0$
- Hence, they will be imaged at  $PX = (0, y, w)^T$ , i.e. they are points on the  $y$ -axis of the image
- Also,  $C$  lies on  $P^1$  (verify)
- Hence,  $P^1$  is defined by the join of  $C$  and line  $x = 0$  in the image
- Similarly,  $P^2$  is defined by the join of  $C$  and line  $y = 0$



# Camera Anatomy: Row Vectors of $P$

- Axis planes are dependent on the choice of image axes, but principal plane is not.
- Intersection of planes  $P^1$  and  $P^2$  is the line joining the camera center and the image origin, i.e. the back projection of image origin
  - Not the optical axis in general...
- Camera center lies on all three planes  $P^1$ ,  $P^2$ ,  $P^3$



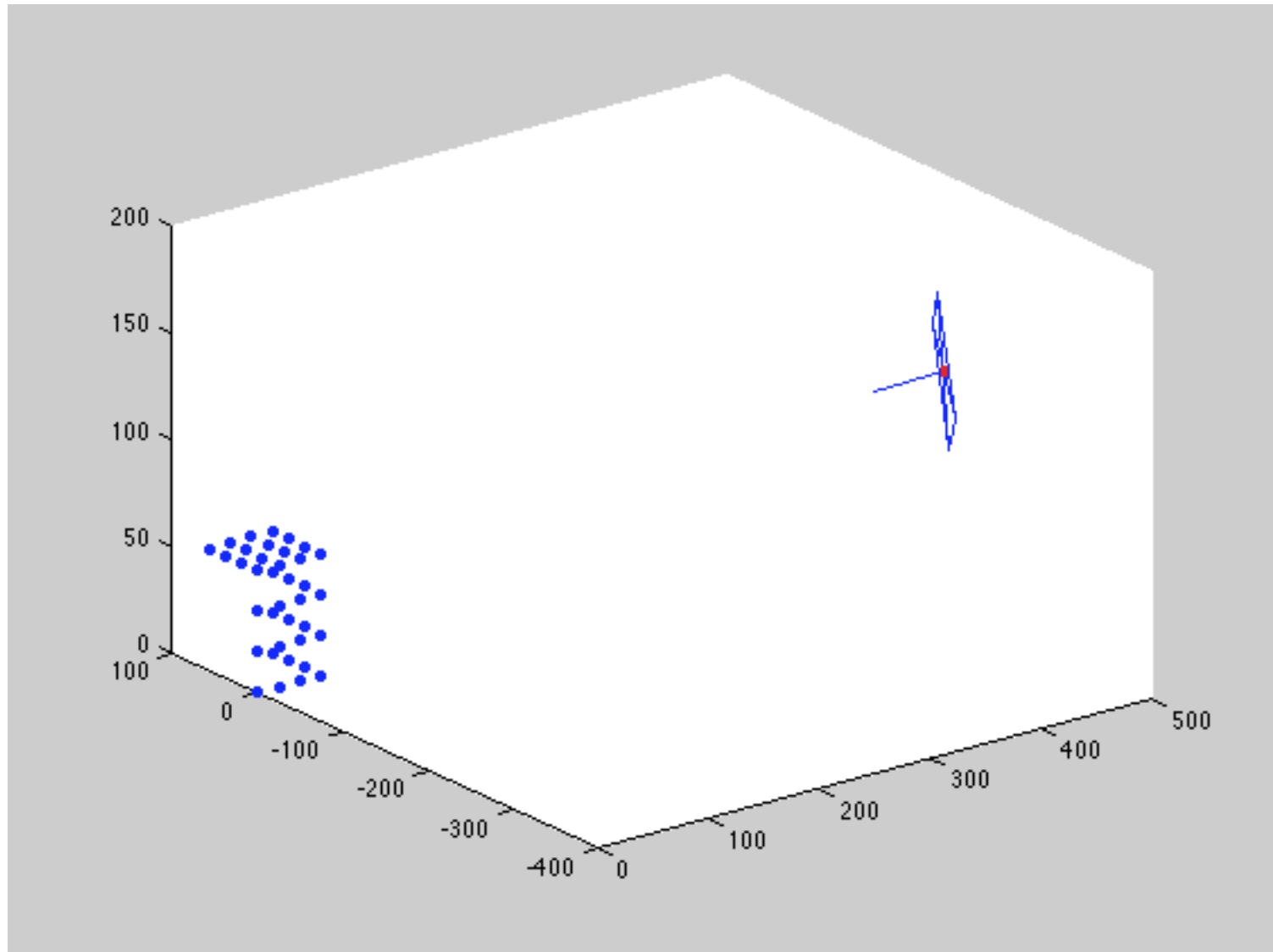
# Camera Anatomy: Principal Point and Principal Axis

- The principal axis is the ray through the camera center perpendicular to the principal plane  $P^3$ , which will intersect the imaging plane at the principal point.
- In general, the normal of a plane  $\pi = (\pi_1, \pi_2, \pi_3, \pi_4)^T$  is given by  $(\pi_1, \pi_2, \pi_3)^T$
- Thus principal axis is given by  $(p_{31}, p_{32}, p_{33})^T$
- Consider the point at infinity in the direction of principal axis, i.e.  $(p_{31}, p_{32}, p_{33}, 0)^T = \hat{P}^3$
- We can project this point back to the image to get the coordinates of the principal point as  $\mathbf{x}_0 = P\hat{P}^3$
- This involves only the first 3x3 block of P. Hence

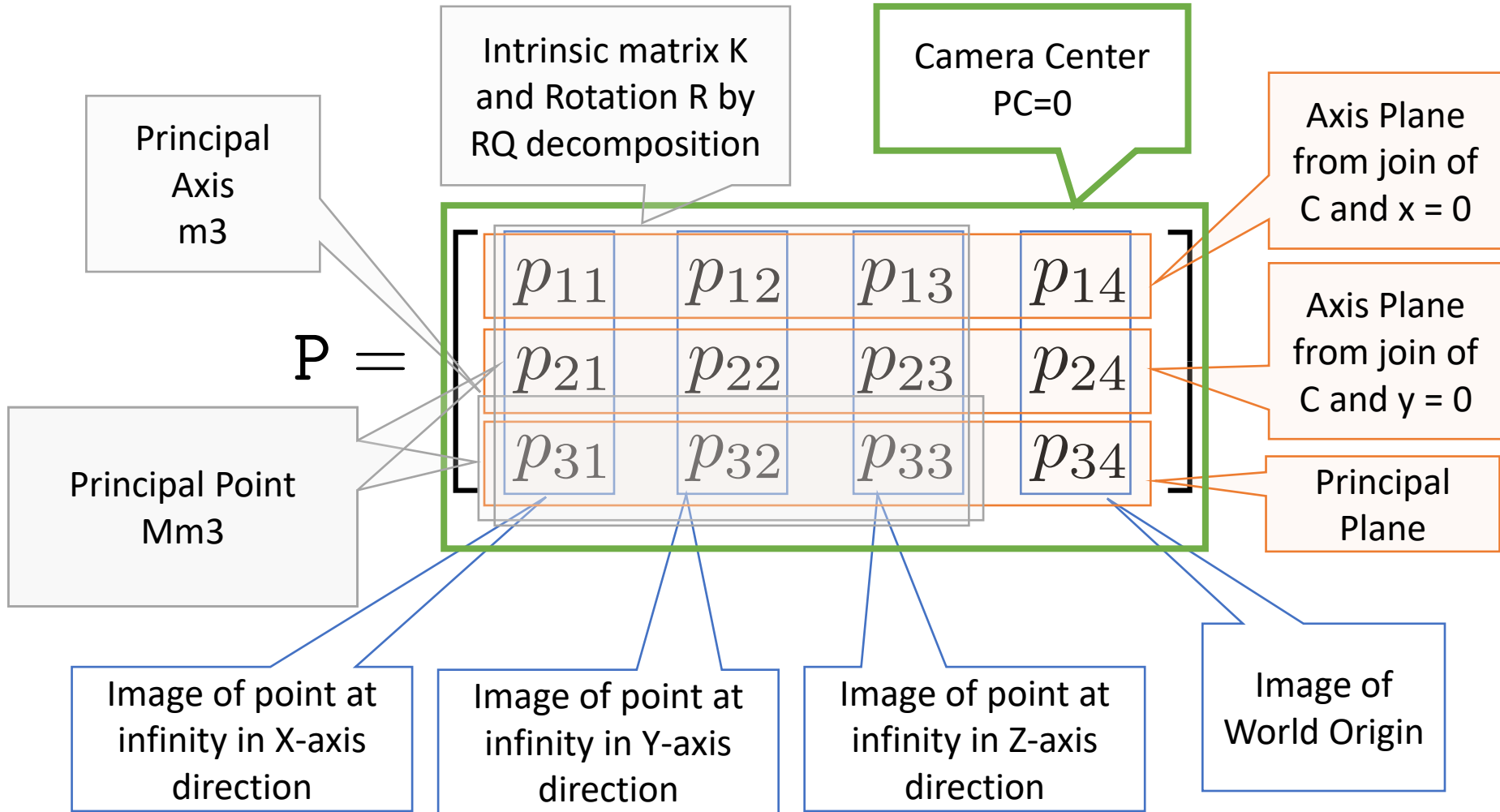
$$\mathbf{x}_0 = M\mathbf{m}^3$$

• where  $\mathbf{m}^{3T}$  is the last row of M, and M the first 3x3 block of P





# Summary: Camera Anatomy



# Action of Camera on Points, Lines and Planes





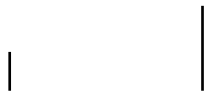
# Action of Camera on Points

- Action of camera on points is familiar to us

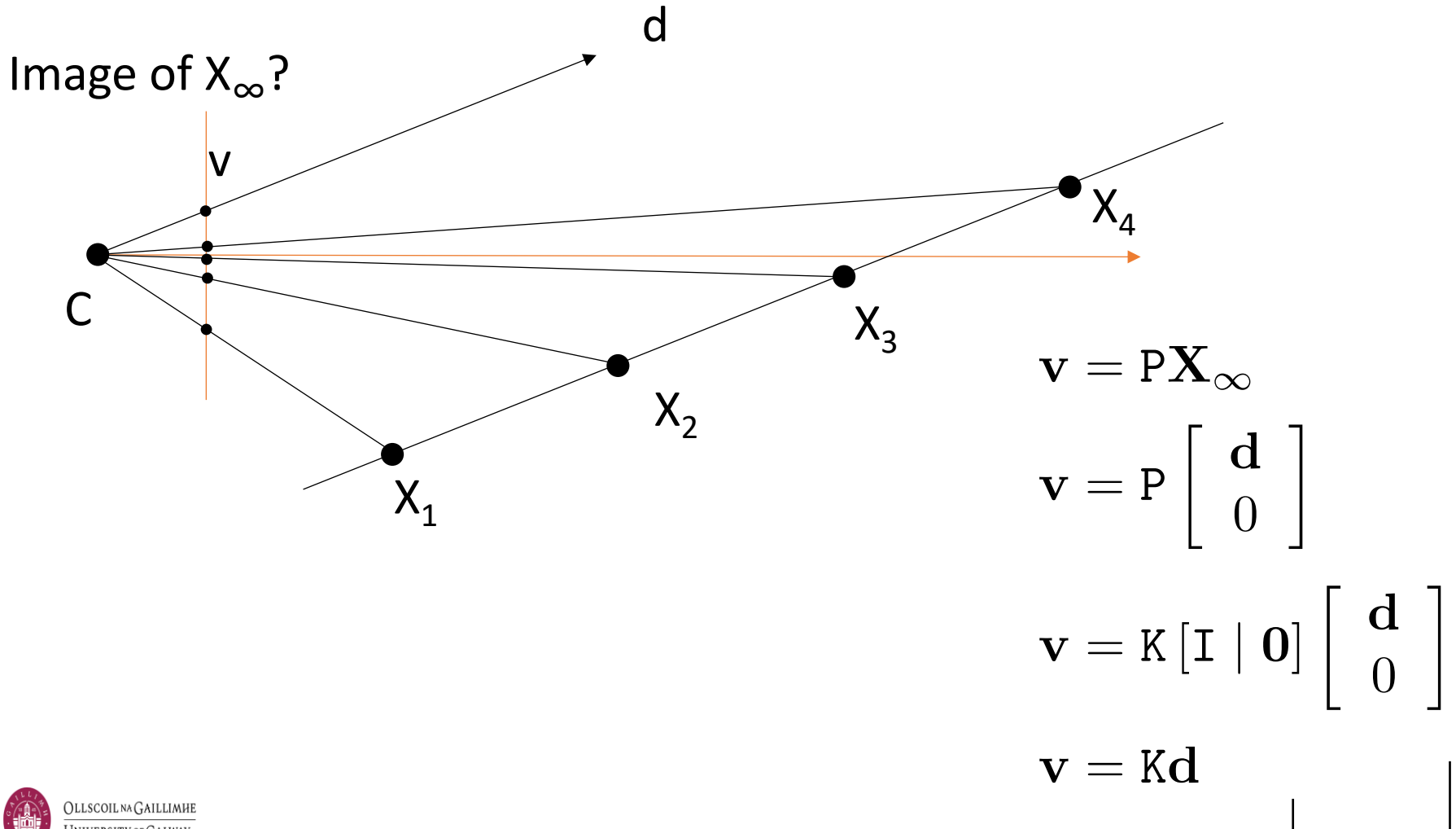
$$\mathbf{x} = K \left[ \mathbf{R} \mid -\mathbf{R}\tilde{\mathbf{C}} \right] \mathbf{X}$$

- In canonical view

$$\mathbf{x} = K \left[ \mathbf{I} \mid \mathbf{0} \right] \mathbf{X}$$

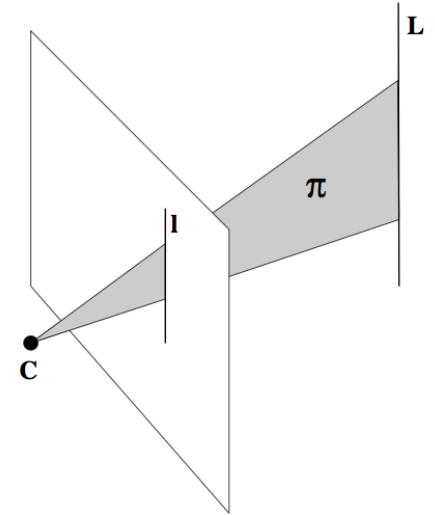


# Vanishing Points



# Action of Camera on Lines

- A line in 3-space will project to a line in the image.
- Geometric Proof: The join of the line  $L$  in 3-space and the camera center  $C$  forms a plane, and this plane will intersect the imaging plane in a line.
- Algebraic Proof:



Consider two points  $\mathbf{A}$  and  $\mathbf{B}$  in 3-space. The line formed by their join can be parameterized as  $\mathbf{X}(\mu) = \mathbf{A} + \mu\mathbf{B}$ .  $\mathbf{X}(\mu)$  will project in the image to

$$\begin{aligned}\mathbf{x}(\mu) &= P(\mathbf{A} + \mu\mathbf{B}) = P\mathbf{A} + \mu P\mathbf{B} \\ &= \mathbf{a} + \mu\mathbf{b}\end{aligned}$$

which is the line joining image points  $\mathbf{a}$  and  $\mathbf{b}$ .



# Back Projection of Points to Rays

- Given a point  $x$  in the image of a camera with matrix  $\mathbf{P}$ , we want to determine set of points in the world that map to this point
- i.e. the ray in the world that this point back-projects to.
- Note that camera center  $C$  always lies along the ray.
- Claim:  $\mathbf{P}^+x$  also lies along the ray, where  $\mathbf{P}^+ = \mathbf{P}^T(\mathbf{P}\mathbf{P}^T)^{-1}$
- Proof:

The image of  $\mathbf{P}^+x$  will be at

$$\begin{aligned} \mathbf{P}\mathbf{P}^+x & \\ &= \mathbf{P}\mathbf{P}^T(\mathbf{P}\mathbf{P}^T)^{-1}x \\ &= x \end{aligned}$$

Therefore  $\mathbf{P}^+x$  must be along the ray



# Back Projection of Points to Rays

- Since we know two points along the ray,  $\mathbf{C}$  and  $P^+x$
- Therefore, points along the ray can be written in parametric form as

$$\mathbf{X}(\lambda) = P^+ \mathbf{x} + \lambda \mathbf{C}$$



# Back Projection of Points to Rays

- In canonical view, relationship is even more simpler
- Consider the back-projected ray which is given by a direction  $\mathbf{d}$
- All points on this ray can be written as the join of camera center and the ray vector

$$\tilde{\mathbf{X}}(\lambda) = \mathbf{0} + \lambda \mathbf{d} = \lambda \mathbf{d}$$

- The image of such a point in canonical view will be at

$$\mathbf{x} = \mathbf{K} [\mathbf{I} \mid \mathbf{0}] \begin{bmatrix} \lambda \mathbf{d} \\ 1 \end{bmatrix}$$

$$\mathbf{x} = \mathbf{K} \mathbf{d}$$

- So, given  $\mathbf{x}$ , the ray direction in canonical view can be computed simply as

$$\mathbf{d} = \mathbf{K}^{-1} \mathbf{x}$$



# Back Projection of Lines

- Lines in image will back project to planes in the world.
- Result: The set of points in the world mapping to a line  $l$  via the camera matrix  $P$  is the plane  $P^T l$

- Proof

A point  $x$  lies on  $l$  iff  $x^T l = 0$ .

A world point  $X$  maps to a point  $PX$ , which will lie on  $l$  iff

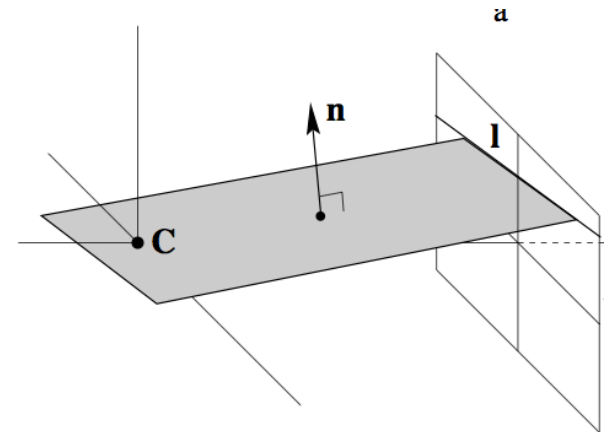
$$X^T P^T l = 0$$

Thus if  $P^T l$  is taken to represent a plane, then  $X$  lies on this plane iff  $X$  maps to a point on the line  $l$

Hence,  $P^T l$  is the plane which is the back projection of line  $l$ .

# Relationship between Image Line and Plane Normal

- Result: An image line  $l$  defines a plane through the camera center with normal direction  $n = K^T l$  in the camera canonical coordinate system



- Proof:

Points  $x$  on line  $l$  back-project to directions  $d = K^{-1}x$

Since these direction vectors lie on the plane, they are orthogonal to the plane normal  $n$ .

Thus,  $d^T n = 0$

$$x^T K^{-T} n = 0$$

Since points on  $l$  satisfy  $x^T l = 0$ , it follows that

$$l = K^{-T} n$$

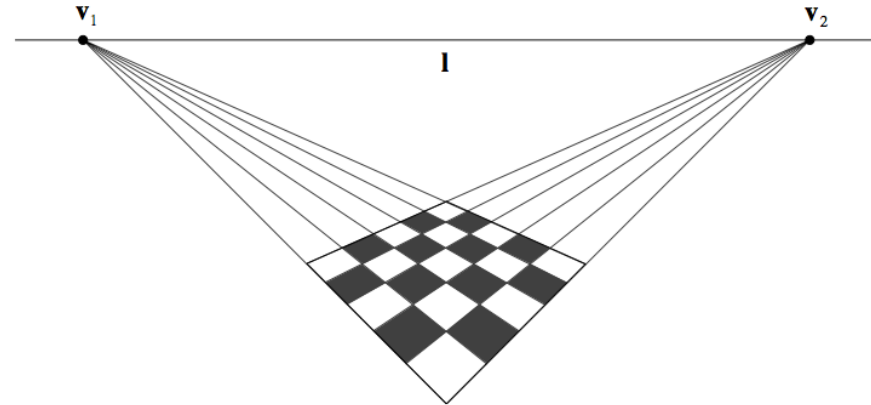
Hence  $n = K^T l$





# Vanishing Line

- The vanishing line of a plane is the image of the line at infinity of the plane
- The vanishing line will depend only on the *orientation* of the plane, and not on its absolute position
- Parallel planes have the same vanishing line



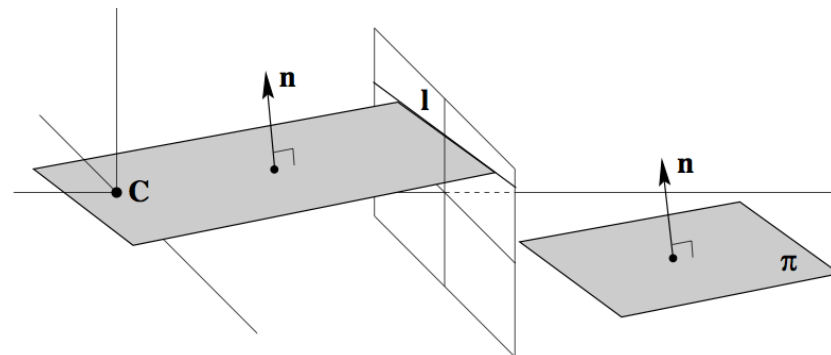
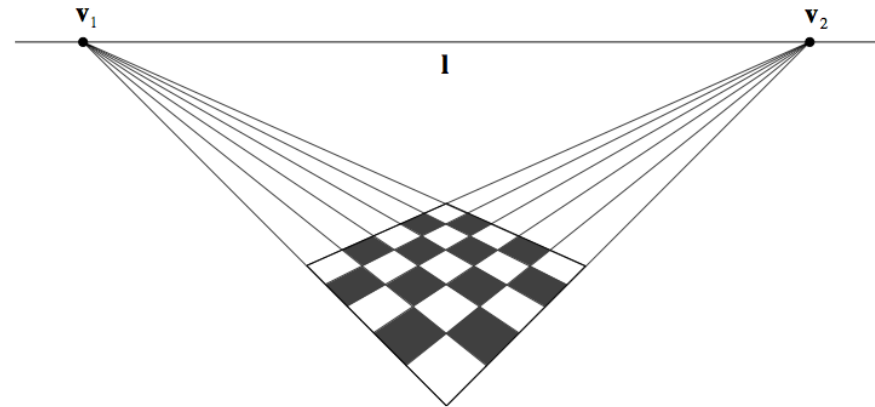
# Vanishing Line

- If camera calibration matrix  $K$  is known, a scene plane's vanishing line can be used to find the plane's orientation relative to the camera.
- In the canonical coordinate system, the orientation of the plane having the vanishing line  $l$  in the image is given by

$$n = K^T l$$

- The vanishing line as a function of plane normal is given by

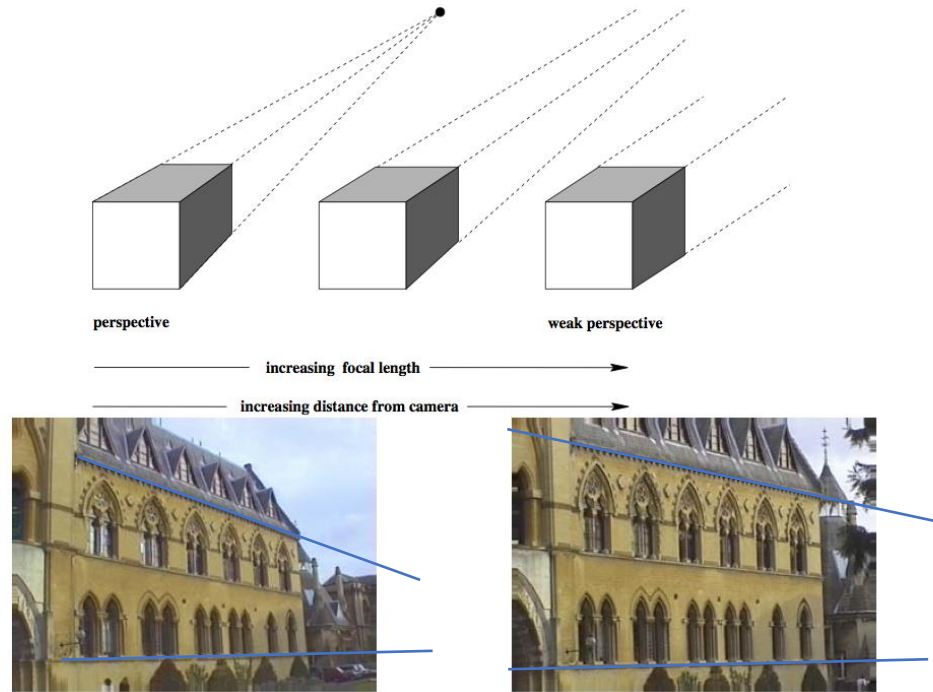
$$l = K^{-T} n$$



# Orthographic Cameras

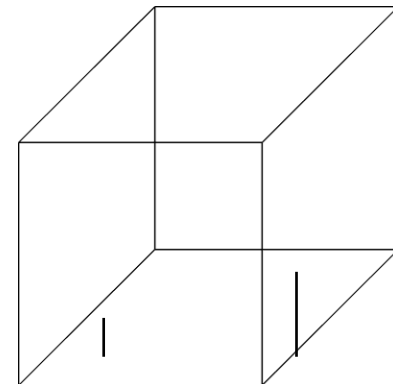
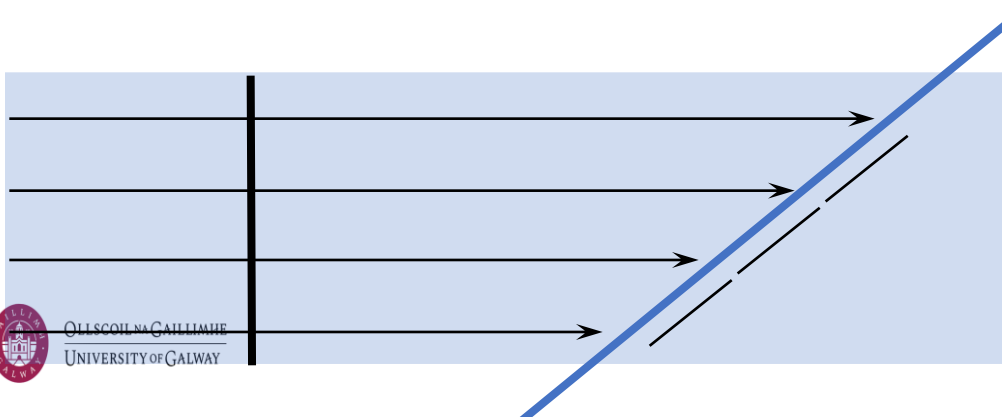
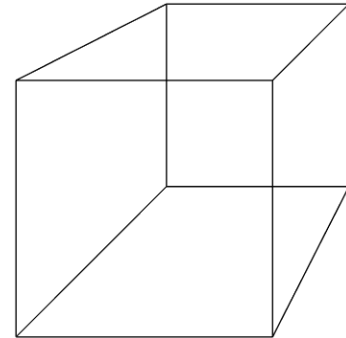
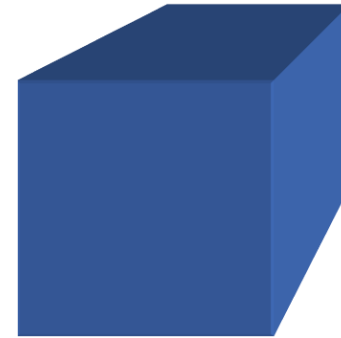
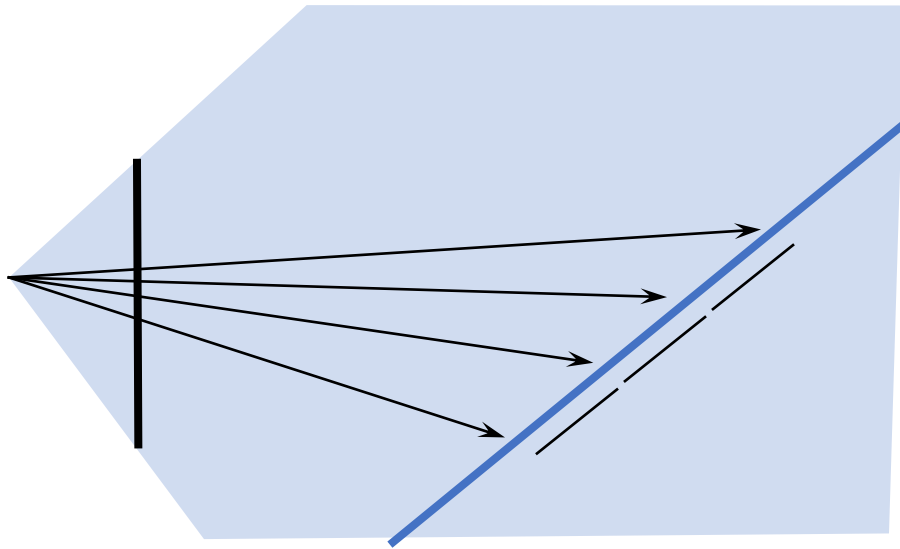
# Cameras at Infinity

- Consider the image sequence in which the camera moves away from an object but zooms in to keep the size of the object the same.



# Cameras at Infinity

- If the camera center moves back from the scene to infinity, then all rays entering the camera will be parallel



# Another Type of Camera: Orthographic Camera

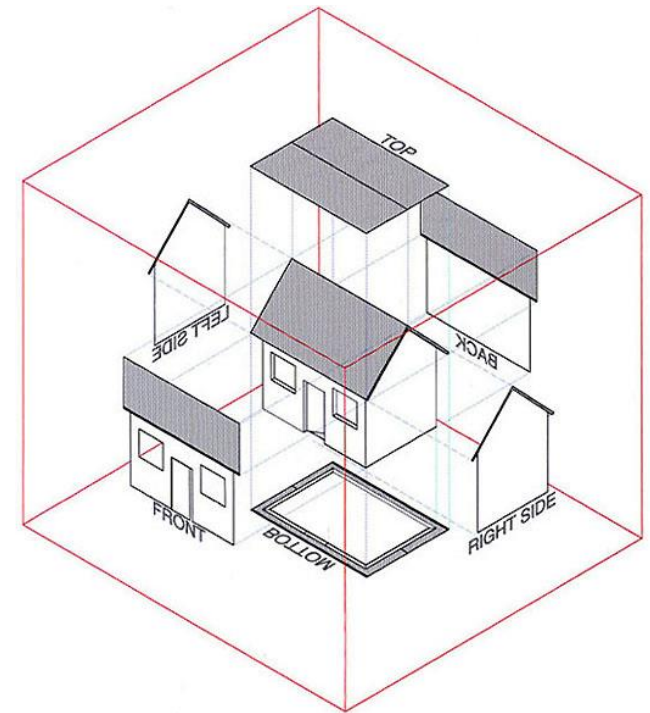
- Parallel Lines remain parallel and do not converge (also termed parallel projection)



The Colonnade  
401 Jefferson Ave., Scranton, PA  
SPRING 2007

JEFFERSON AVENUE ELEVATION

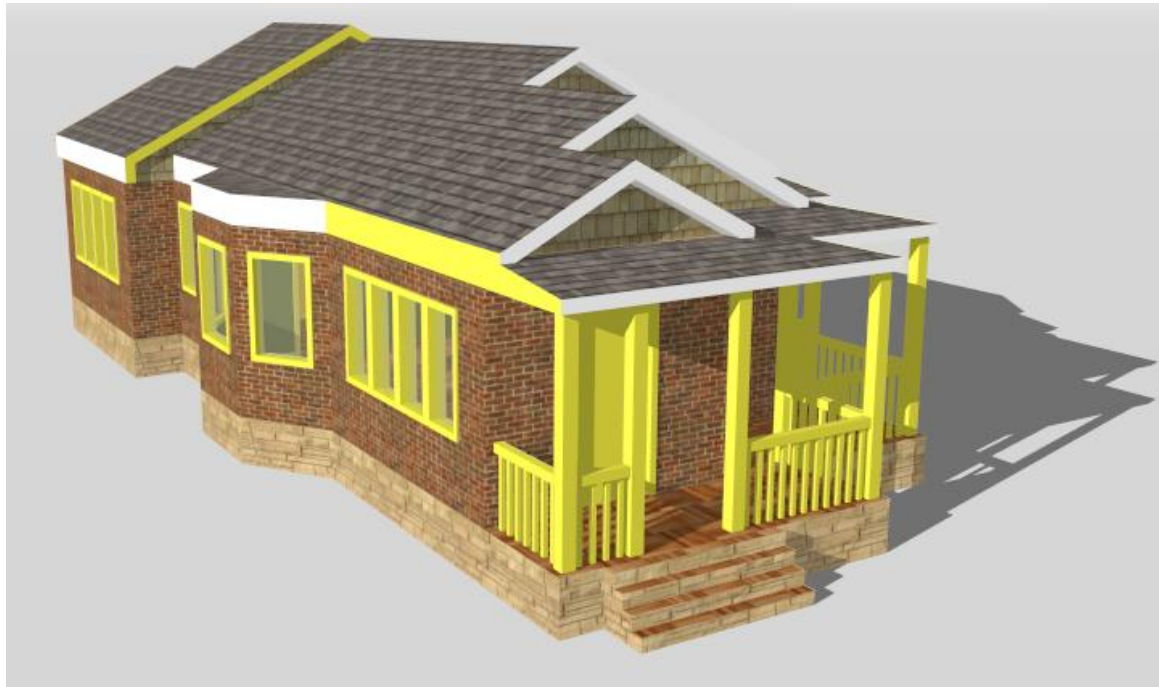
leonorimullerdavis



# Type of Projection?

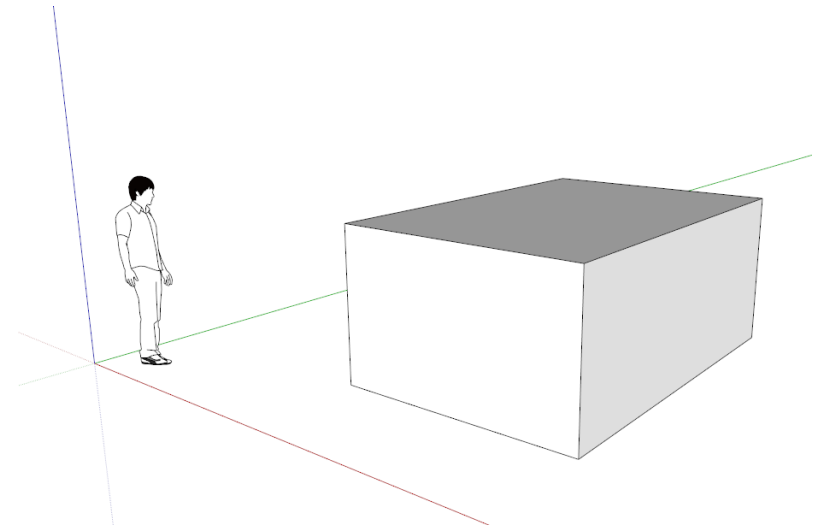
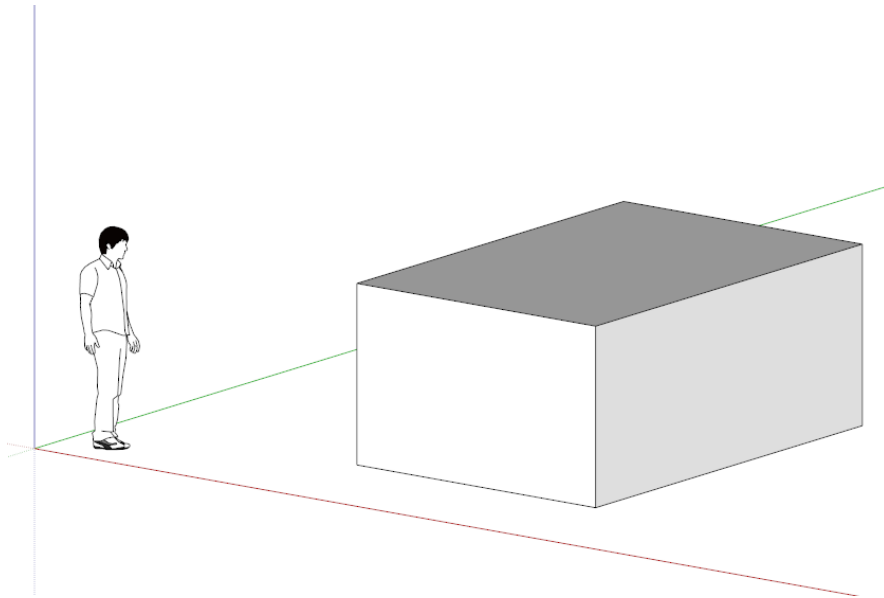


# Type of Projection?

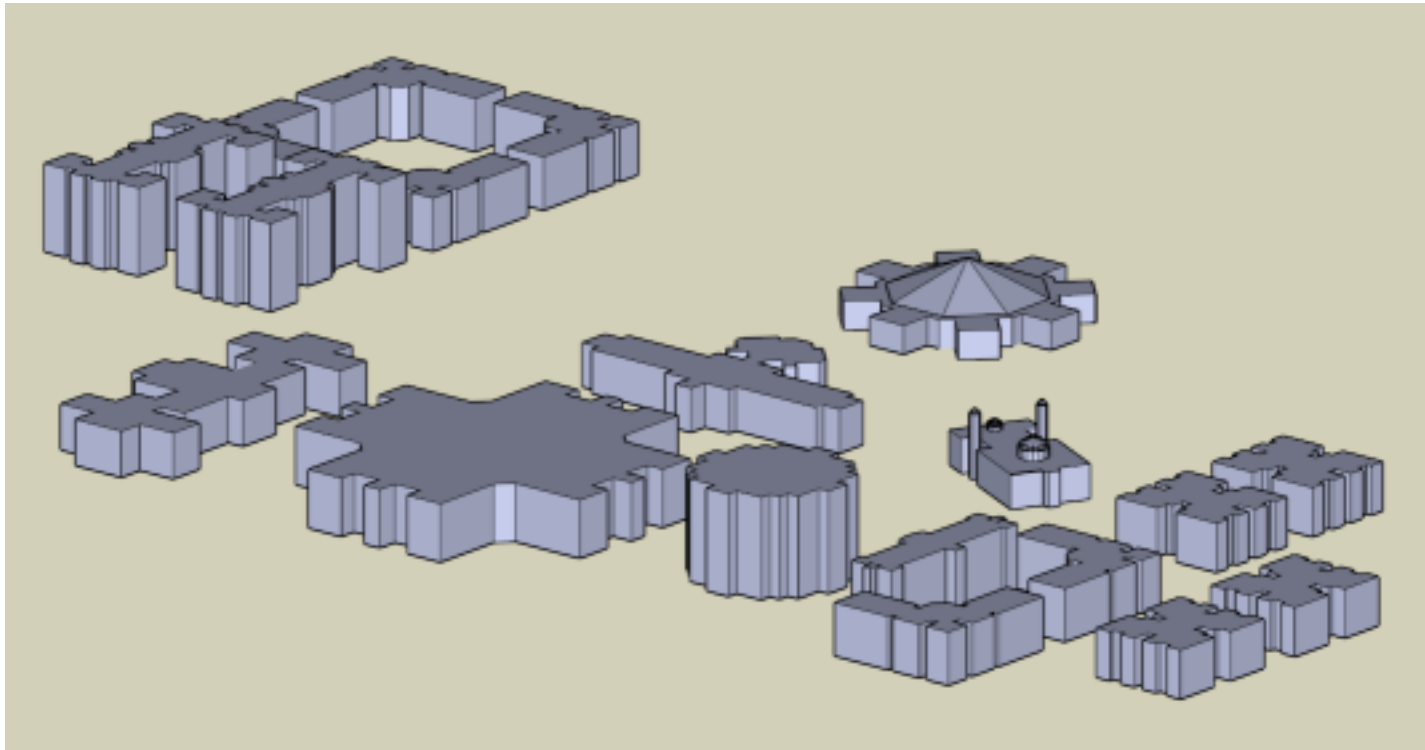




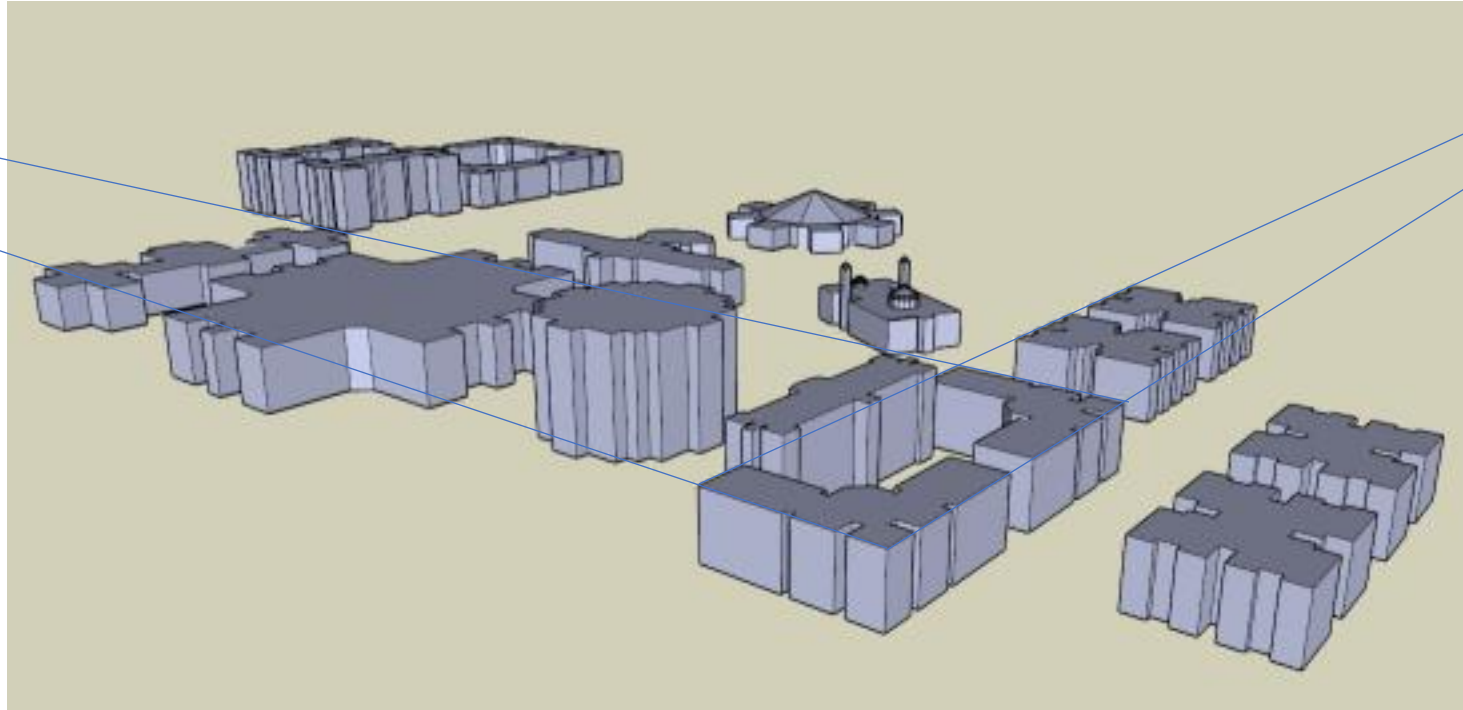
# Type of Projection?



# Type of Projection?

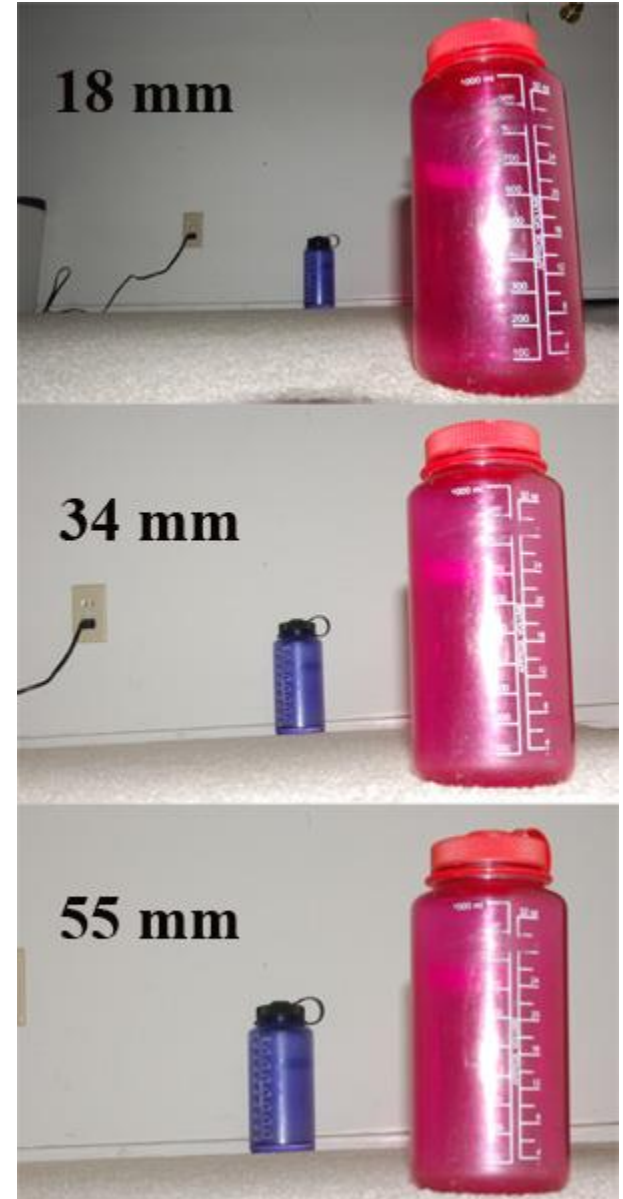


# Type of Projection?



# Perspective Distortion

- Perspective distortion: the effect that further away objects appear smaller in size
- As focal length increases (more zoom), perspective distortion becomes less
- Orthographic camera can be considered as being very far away so there is no variation in  $Z$ , and having very long focal length. Hence it has no perspective distortion.
- Equal lengths in the world will appear of equal size in the image



# Orthographic Projection

- In canonical view:
- The relationship between image coordinates and scene coordinates is:

$$x = X, \quad y = Y$$

- In matrix form

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

M (first 3x3 block of P) is now a singular matrix

Last row is  $[0, 0, 0, 1]^T$ , so this is also termed as an affine camera

Camera center is given by  $\text{null}(P) = [0 \ 0 \ 1 \ 0]^T$  is a point at infinity



# Orthographic Projection

- In general view:

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_X \\ r_{21} & r_{22} & r_{23} & t_Y \\ r_{31} & r_{32} & r_{33} & t_Z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

- This can also be written as

$$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_X \\ r_{21} & r_{22} & r_{23} & t_Y \\ r_{31} & r_{32} & r_{33} & t_Z \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Or 
$$\tilde{\mathbf{x}} = \mathbf{K}_o [\mathbf{R} \mid \mathbf{T}] \mathbf{X}$$

- Note that 3<sup>rd</sup> row of R|T does not matter

$$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} t_X \\ t_Y \end{bmatrix}$$

# Relationship between Orthographic and Perspective Projection

- Consider a pinhole camera which is very far away and zoomed into the scene.
- Since the depth variation of the scene is small compared to the distance of the camera, it may be approximated by a constant value.

- Hence

$$x = \frac{fX}{Z} \quad y = \frac{fY}{Z}$$

- Since  $\frac{f}{Z}$  is now a constant, we can write

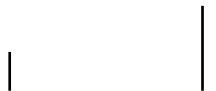
$$x = mX \quad y = mY$$

- which is scaled orthographic projection



# Some properties of Orthographic Projection

- Parallel lines remain parallel
- There is no perspective distortion. Equal lengths in the world appear as equal lengths in the image.
- Images of a plane taken by an orthographic camera are related by an affine transformation [Proof?]





# Practical Cases of Orthographic Camera

- A camera which is actually very far away compared to the depth variation in the scene can be approximated by an orthographic camera
- Orthographic projection is often used in computer games
- Scanner provides an orthographic projection of a document image
- Also used in some algorithms as an approximation for mathematical simplicity (e.g. Tomasi / Kanade Structure for Motion Algorithm)



# Satellite Cameras

- Satellite images are typically taken by a line scan pushbroom camera, which is orthographic along the direction of motion and perspective in the line-scan direction

