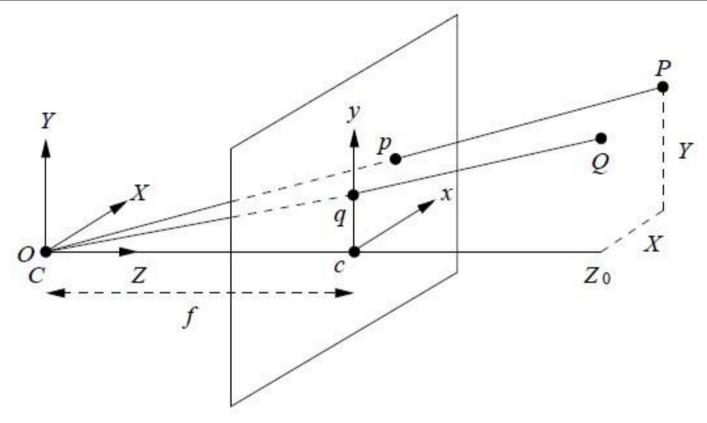
# Calibration and Stereo Vision Reminder Jesson

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## Scaled Orthographic Projection

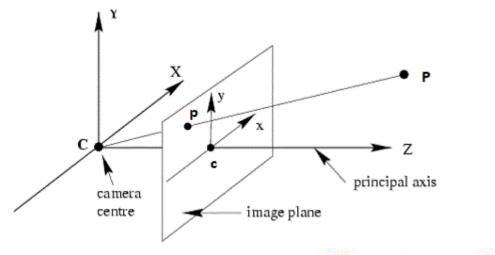


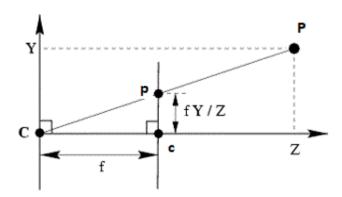
Scene depth << distance to camera. Z is the same for all scene points, say  $Z_0$ 

$$x = sX$$
,  $y = sY$ ,  $s = \frac{f}{Z_0}$  for all scene points.

#### Pinhole Camera Model

3D point  $P = (X, Y, Z)^T$  projects to 2D image point  $p = (x, y)^T$ .





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

Simplest form of perspective projection -> pinhole camera model => Pixel coordinates

#### Camera Intrinsic Parameters

Pinhole camera model

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

Camera sensor's pixels not exactly square

$$x = kf\frac{X}{Z}, \quad y = lf\frac{Y}{Z}$$

- *x*, *y* : coordinates (pixels)
- k, l : scale parameters (pixels/m)
- o *f* : focal length (m or mm)

#### **Intrinsic Parameters**

- $\circ$  f, k, l are not independent.
- Can rewrite as follows (in pixels):

$$f_x = kf$$
  $f_y = lf$ 

OSo,

$$x = f_x \frac{X}{Z}, \quad y = f_y \frac{Y}{Z}$$

- Image centre or principal point c may not be at origin.
- O Denote location of c in image plane as  $c_x$ ,  $c_y$ .
- Then,

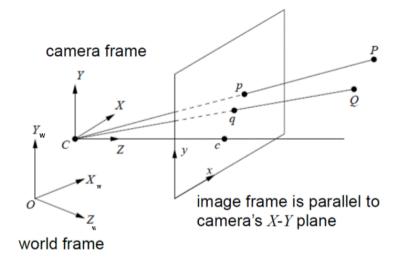
$$x = f_x \frac{X}{Z} + c_x, \quad y = f_y \frac{Y}{Z} + c_y$$

O Along optical axis, X = Y = 0, and  $x = c_x$ ,  $y = c_y$ .

#### **Extrinsic Parameters**

Translation matrix

$$\mathbf{T} = \begin{bmatrix} T_X \\ T_Y \\ T_Z \end{bmatrix}$$



Rotation matrix

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

#### **Rotation Matrix**

R may be decomposed into 3 rotations about the coordinate axes:

$$R = R_x R_v R_z$$

If rotations are performed in the above order:

- 1)  $\gamma$  = rotation angle about z-axis
- 2)  $\beta$  = rotation angle about (new) y-axis
- 3)  $\alpha$  = rotation angle about (new) x-axis

("tilt angle", "pan angle", and "nick angle" for the camera coordinate assignment shown before)

$$R_{\chi} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & \sin \alpha \\ 0 & -\sin \alpha & \cos \alpha \end{bmatrix}$$

$$R_y = \begin{bmatrix} \cos \beta & 0 & -\sin \beta \\ 0 & 1 & 0 \\ \sin \beta & 0 & \cos \beta \end{bmatrix}$$

$$R_z = \begin{bmatrix} \cos \gamma & \sin \gamma & 0 \\ -\sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

#### **Rotation Matrix**

By multiplying the 3 matrices  $R_X$ ,  $R_V$  and  $R_Z$ , one gets

$$\begin{split} \mathsf{R} = \left[ \begin{array}{ccc} \cos\beta\cos\gamma & \cos\beta\sin\gamma & -\sin\beta \\ \sin\alpha\sin\beta\cos\gamma - \cos\alpha\sin\gamma & \sin\alpha\sin\beta\sin\gamma + \cos\alpha\cos\gamma & \sin\alpha\cos\beta \\ \cos\alpha\sin\beta\cos\gamma + \sin\alpha\sin\gamma & \cos\alpha\sin\beta\sin\gamma - \sin\alpha\cos\gamma & \cos\alpha\cos\beta \end{array} \right] \end{split}$$

For formula manipulations, one tries to avoid the trigonometric functions and takes

Note that the coefficients of R are constrained:

$$RR^T = I$$
 (unit matrix)

#### **Rotation Matrix**

• Rotation matrix is orthonormal:

$$\mathbf{R}^{\mathsf{T}}\mathbf{R} = \mathbf{I}$$

- **⊙** So,  $\mathbf{R}^{-1} = \mathbf{R}^{\top}$ .
- In particular, let

$$\mathbf{R} = \left[ egin{array}{c} \mathbf{R}_1^ op \ \mathbf{R}_2^ op \ \mathbf{R}_3^ op \end{array} 
ight]$$

Then,

$$\mathbf{R}_i^{\top} \mathbf{R}_j = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$$

## Homogeneous coordinates

Is Perspective Projection a linear transformation?  $x=f\frac{X}{Z}, \quad y=f\frac{Y}{Z}$ • no—division by z is nonlinear

#### Trick: add one more coordinate:

$$(x,y) \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$
  $(x,y,z) \Rightarrow \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$ 

homogeneous image coordinates

homogeneous scene coordinates

#### Converting from homogeneous coordinates

$$\begin{bmatrix} x \\ y \\ w \end{bmatrix} \Rightarrow (x/w, y/w) \qquad \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} \Rightarrow (x/w, y/w, z/w)$$

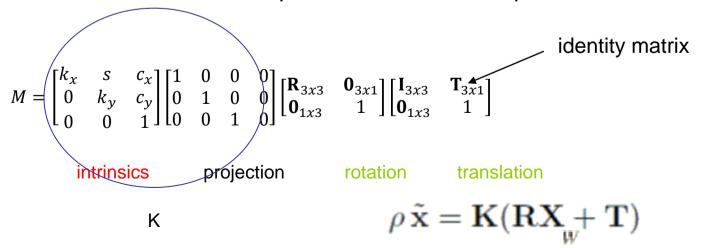
## Camera parameters

Summary - A camera is described by several parameters

- Translation T of the optical center from the origin of world coords
- Rotation R of the image plane
- focal length f, principle point (Cx, Cy), pixel size (k, l)

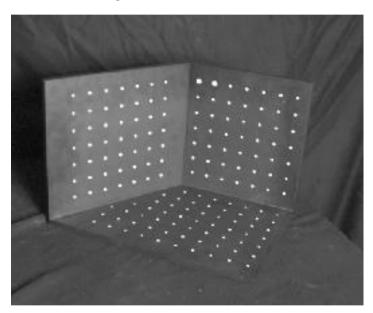
Projection equation - correspondence between image (u,v) and scene (X, Y,Z)

- The projection matrix models the cumulative effect of all parameters
  - Useful to decompose into a series of operations



## Estimating the projection matrix

- Place a known object in the scene
  - identify correspondence between image (ui,vi) and scene (Xi, Yi, Zi)



compute mapping from scene to image (previous slide)

$$\begin{bmatrix} u_i \\ v_i \\ 1 \end{bmatrix} \cong \begin{bmatrix} m_{00} & m_{01} & m_{02} & m_{03} \\ m_{10} & m_{11} & m_{12} & m_{13} \\ m_{20} & m_{21} & m_{22} & m_{23} \end{bmatrix} \begin{bmatrix} X_i \\ Y_i \\ Z_i \\ 1 \end{bmatrix}$$

CV

#### Direct linear calibration

$$\begin{bmatrix} u_i \\ v_i \\ 1 \end{bmatrix} \cong \begin{bmatrix} m_{00} & m_{01} & m_{02} & m_{03} \\ m_{10} & m_{11} & m_{12} & m_{13} \\ m_{20} & m_{21} & m_{22} & m_{23} \end{bmatrix} \begin{bmatrix} X_i \\ Y_i \\ Z_i \\ 1 \end{bmatrix}$$

$$u_{i} = \frac{m_{00}X_{i} + m_{01}Y_{i} + m_{02}Z_{i} + m_{03}}{m_{20}X_{i} + m_{21}Y_{i} + m_{22}Z_{i} + m_{23}}$$
$$v_{i} = \frac{m_{10}X_{i} + m_{11}Y_{i} + m_{12}Z_{i} + m_{13}}{m_{20}X_{i} + m_{21}Y_{i} + m_{22}Z_{i} + m_{23}}$$

$$u_i(m_{20}X_i + m_{21}Y_i + m_{22}Z_i + m_{23}) = m_{00}X_i + m_{01}Y_i + m_{02}Z_i + m_{03}$$
$$v_i(m_{20}X_i + m_{21}Y_i + m_{22}Z_i + m_{23}) = m_{10}X_i + m_{11}Y_i + m_{12}Z_i + m_{13}$$

$$\begin{bmatrix} X_{i} & Y_{i} & Z_{i} & 1 & 0 & 0 & 0 & -u_{i}X_{i} & -u_{i}Y_{i} & -u_{i}Z_{i} & -u_{i} \\ 0 & 0 & 0 & 0 & X_{i} & Y_{i} & Z_{i} & 1 & -v_{i}X_{i} & -v_{i}Y_{i} & -v_{i}Z_{i} & -v_{i} \end{bmatrix} \begin{bmatrix} m_{01} \\ m_{02} \\ m_{03} \\ m_{10} \\ m_{11} \\ m_{12} \\ m_{13} \\ m_{20} \\ m_{21} \\ m_{22} \\ m_{23} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

#### Direct linear calibration

$$\begin{bmatrix} X_1 & Y_1 & Z_1 & 1 & 0 & 0 & 0 & -u_1X_1 & -u_1Y_1 & -u_1Z_1 & -u_1 \\ 0 & 0 & 0 & 0 & X_1 & Y_1 & Z_1 & 1 & -v_1X_1 & -v_1Y_1 & -v_1Z_1 & -v_1 \\ \vdots & & & & & \vdots & & & & \\ X_n & Y_n & Z_n & 1 & 0 & 0 & 0 & -u_nX_n & -u_nY_n & -u_nZ_n & -u_n \\ 0 & 0 & 0 & 0 & X_n & Y_n & Z_n & 1 & -v_nX_n & -v_nY_n & -v_nZ_n & -v_n \end{bmatrix} \begin{bmatrix} m_{00} \\ m_{01} \\ m_{02} \\ m_{03} \\ m_{10} \\ m_{11} \\ m_{12} \\ m_{13} \\ m_{20} \\ m_{21} \\ m_{22} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{bmatrix}$$

Can solve for m<sub>ii</sub> by linear or non-linear methods.

• Estimate K, R, and T from M



Camera Calibration

#### SVD

The singular value decomposition of a matrix M is the factorization of M into the product of three matrices

$$M = UDV^T$$

where the columns of U and V are orthonormal and the matrix D is diagonal with positive real entries.

Since *U* and V have orthogonal columns:

$$U^T U = I$$
  
and  
 $V^T V = I$ 

#### SVD

$$\begin{bmatrix} X_1 & Y_1 & Z_1 & 1 & 0 & 0 & 0 & -u_1X_1 & -u_1Y_1 & -u_1Z_1 & -u_1 \\ 0 & 0 & 0 & 0 & X_1 & Y_1 & Z_1 & 1 & -v_1X_1 & -v_1Y_1 & -v_1Z_1 & -v_1 \\ \vdots & & & & & \vdots & & & & \\ X_n & Y_n & Z_n & 1 & 0 & 0 & 0 & -u_nX_n & -u_nY_n & -u_nZ_n & -u_n \\ 0 & 0 & 0 & 0 & X_n & Y_n & Z_n & 1 & -v_nX_n & -v_nY_n & -v_nZ_n & -v_n \end{bmatrix} \begin{bmatrix} m_{00} \\ m_{01} \\ m_{02} \\ m_{03} \\ m_{10} \\ m_{11} \\ m_{12} \\ m_{13} \\ m_{20} \\ m_{21} \\ m_{22} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{bmatrix}$$

$$M = UDV^T$$

Solution: the smallest element (null) of the diagonal of D (last row (NumPy) or column (Matlab) of V)

### Projection Equation

$$\rho \, \tilde{\mathbf{x}} = \mathbf{K} (\mathbf{R} \mathbf{X}_w + \mathbf{T})$$

Using the notation  $\mathbf{r}_i = (r_{i1} r_{i2} r_{i3})$ .

$$\mathbf{A} = \begin{pmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \\ 0 & 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} \mathbf{R} & \mathbf{t} \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} \mathbf{r}_1 & t_x \\ \mathbf{r}_2 & t_y \\ \mathbf{r}_3 & t_z \\ 0 & 1 \end{pmatrix}$$

$$\begin{pmatrix} su \\ sv \\ s \end{pmatrix} = \mathbf{K}\mathbf{A} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}_{w} \longrightarrow \begin{pmatrix} su \\ sv \\ s \end{pmatrix} = \mathbf{M} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}_{w}$$

## Projection Equation

$$\begin{pmatrix} su \\ sv \\ s \end{pmatrix} = \begin{pmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{pmatrix} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}_{w}$$

Using the notation  $\mathbf{m}_i = (m_{i1} \ m_{i2} \ m_{i3})$ .

$$\mathbf{M} = \begin{pmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{pmatrix} = \begin{pmatrix} \mathbf{m}_1 & m_{14} \\ \mathbf{m}_2 & m_{24} \\ \mathbf{m}_3 & m_{34} \end{pmatrix}$$

Using the notation  $\mathbf{r}_i$  defined for the matrix

$$M = KA = \begin{pmatrix} f_{x} \boldsymbol{r}_{1} + c_{x} \boldsymbol{r}_{3} & f_{x} t_{x} + c_{x} t_{z} \\ f_{y} \boldsymbol{r}_{2} + c_{y} \boldsymbol{r}_{3} & f_{y} t_{y} + c_{y} t_{z} \\ \boldsymbol{r}_{3} & t_{z} \end{pmatrix}$$

CV

 Expressing M in function of KA, and taking into account the orthogonality of the matrix R, the camera parameters can be estimated as:

$$\begin{cases} \mathbf{r}_{3} = \mathbf{m}_{3} \\ c_{x} = \mathbf{m}_{1} \cdot \mathbf{m}_{3} \\ c_{y} = \mathbf{m}_{2} \cdot \mathbf{m}_{3} \\ f_{x} = ||\mathbf{m}_{1} \wedge \mathbf{m}_{3}|| \\ f_{y} = ||\mathbf{m}_{2} \wedge \mathbf{m}_{3}|| \\ \mathbf{r}_{1} = 1/f_{x}(\mathbf{m}_{1} - c_{x}\mathbf{m}_{3}) \\ \mathbf{r}_{2} = 1/f_{y}(\mathbf{m}_{2} - c_{y}\mathbf{m}_{3}) \\ t_{x} = 1/f_{x}(\mathbf{m}_{14} - c_{x}\mathbf{m}_{34}) \\ t_{y} = 1/f_{y}(\mathbf{m}_{24} - c_{y}\mathbf{m}_{34}) \\ t_{z} = \mathbf{m}_{34} \end{cases}$$

With  $m_i = (m_{i1}, m_{i2}, m_{i3})$ 

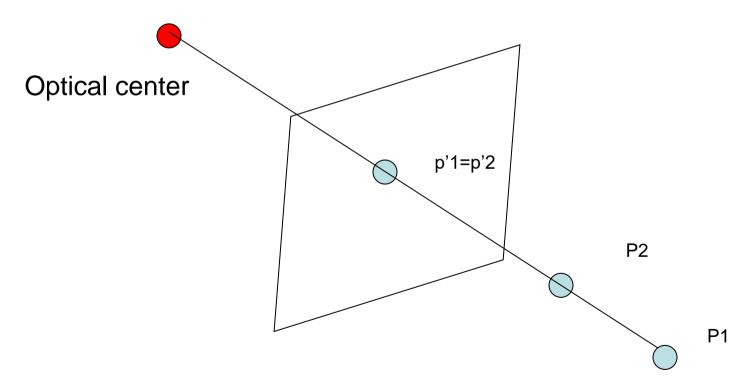
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With  $m_i = (m_{i1}, m_{i2}, m_{i3})$ 

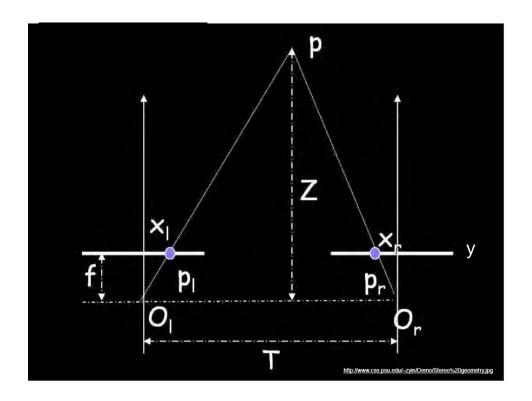
## Why multiple views?

- Structure and depth are inherently ambiguous from single views.
- => Many-to-one: any points along same ray map to same point in image



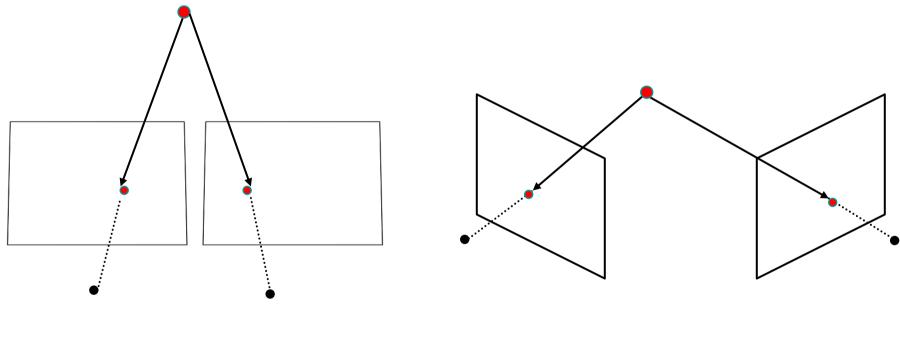
## Geometry for a simple stereo system

• First, assuming parallel optical axes, known camera parameters (i.e., calibrated cameras):



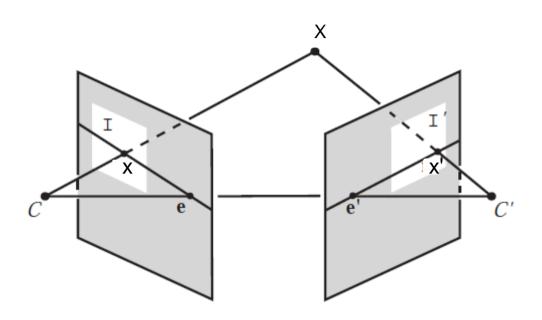
## General case, with calibrated cameras

The two cameras will not have parallel optical axes.



**VERSUS** 

# Triangulation



Having two points x (on left image) and x' (right image) projections of a scene point X -> How to estimate X?

From perspective projection we have:

$$x = MX$$
  $x' = M'X$ 

## **General Triangulation**

- Use the equations x = MX and x' = M'X to solve X
  - For the first camera:

$$\mathbf{M} = \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{bmatrix} = \begin{bmatrix} \boldsymbol{m}^{1\mathsf{T}} \\ \boldsymbol{m}^{2\mathsf{T}} \\ \boldsymbol{m}^{3\mathsf{T}} \end{bmatrix}$$

where  $m^{i\top}$  are the rows of M

- Eliminate the unknown scale sx = MX by forming a cross product  $x \times MX = 0$ 

$$x(\mathbf{m}^{3\mathsf{T}}\mathbf{X}) - (\mathbf{m}^{1\mathsf{T}}\mathbf{X}) = 0$$
$$y(\mathbf{m}^{3\mathsf{T}}\mathbf{X}) - (\mathbf{m}^{2\mathsf{T}}\mathbf{X}) = 0$$
$$x(\mathbf{m}^{2\mathsf{T}}\mathbf{X}) - y(\mathbf{m}^{1\mathsf{T}}\mathbf{X}) = 0$$

Rearrange as (first to equations only)

$$\begin{bmatrix} x \boldsymbol{m}^{3\mathsf{T}} - \boldsymbol{m}^{1\mathsf{T}} \\ y \boldsymbol{m}^{3\mathsf{T}} - \boldsymbol{m}^{2\mathsf{T}} \end{bmatrix} \mathbf{X} = 0$$

## **General Triangulation**

- Use the equations x = MX and x' = M'X to solve X
  - Similarly for the second camera:

$$\begin{bmatrix} x' \boldsymbol{m}'^{3T} - \boldsymbol{m}'^{1T} \\ y' \boldsymbol{m}'^{3T} - \boldsymbol{m}'^{2T} \end{bmatrix} \mathbf{X} = 0$$

Collecting together gives

$$AX = 0$$

where A is a  $4 \times 4$  matrix

$$A = \begin{bmatrix} x \boldsymbol{m}^{3T} - \boldsymbol{m}^{1T} \\ y \boldsymbol{m}^{3T} - \boldsymbol{m}^{2T} \\ x' \boldsymbol{m}'^{3T} - \boldsymbol{m}'^{1T} \\ y' \boldsymbol{m}'^{3T} - \boldsymbol{m}'^{2T} \end{bmatrix}$$

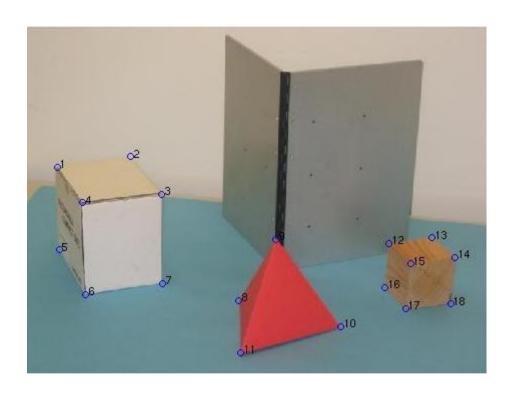
From which we can solve using SVD

$$A = UDV^T$$

Solution: the smallest element (null) of the diagonal of *D* (last row (NumPy) or column (Matlab) of *V*)

# **General Triangulation**

How do we check our reconstruction?





CV 29