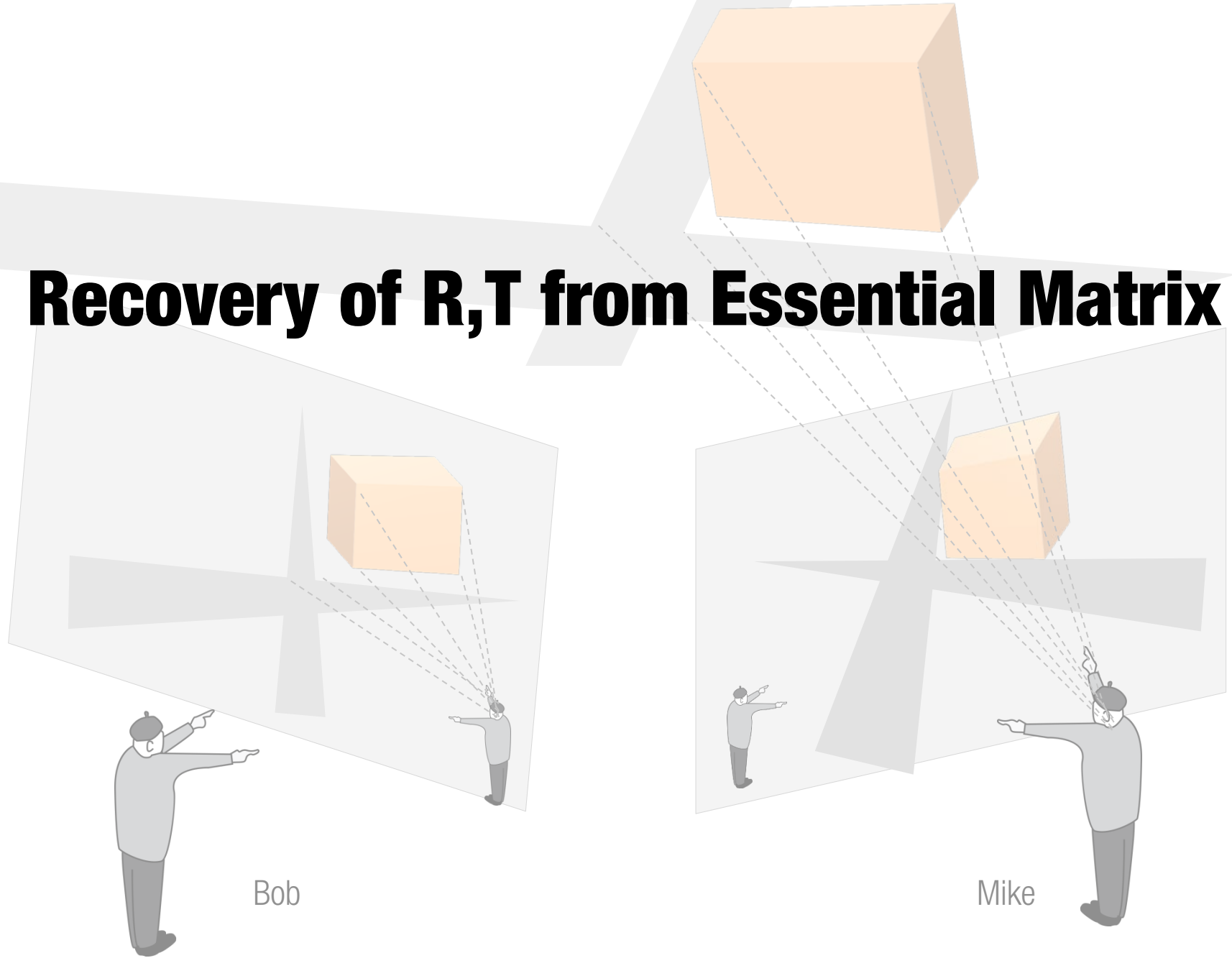
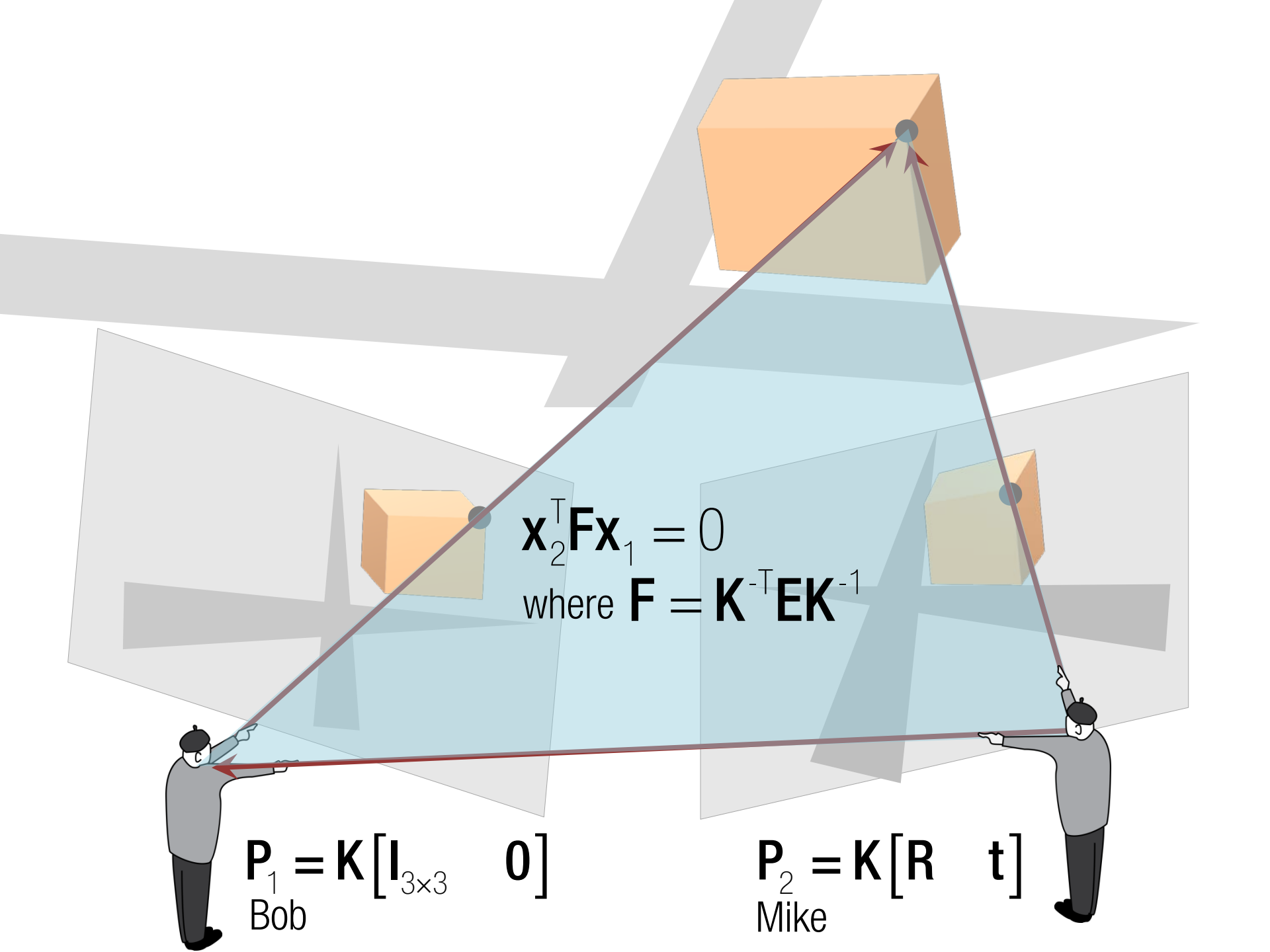


Recovery of R, T from Essential Matrix



Bob

Mike



$$\mathbf{x}_2^T \mathbf{F} \mathbf{x}_1 = 0$$

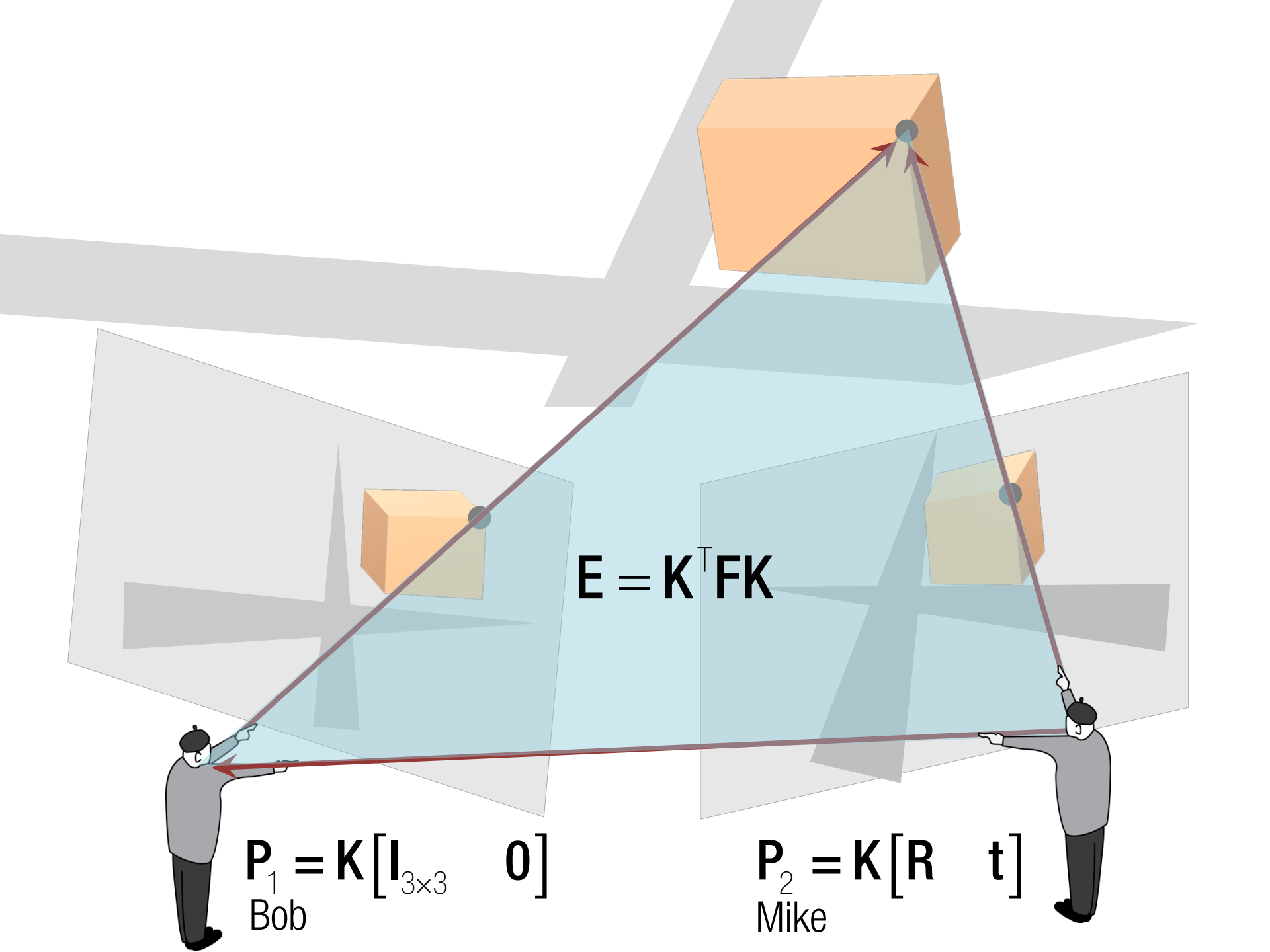
$$\text{where } \mathbf{F} = \mathbf{K}^{-T} \mathbf{E} \mathbf{K}^{-1}$$

$$\mathbf{P}_1 = \mathbf{K} \begin{bmatrix} \mathbf{I}_{3 \times 3} & \mathbf{0} \end{bmatrix}$$

Bob

$$\mathbf{P}_2 = \mathbf{K} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix}$$

Mike



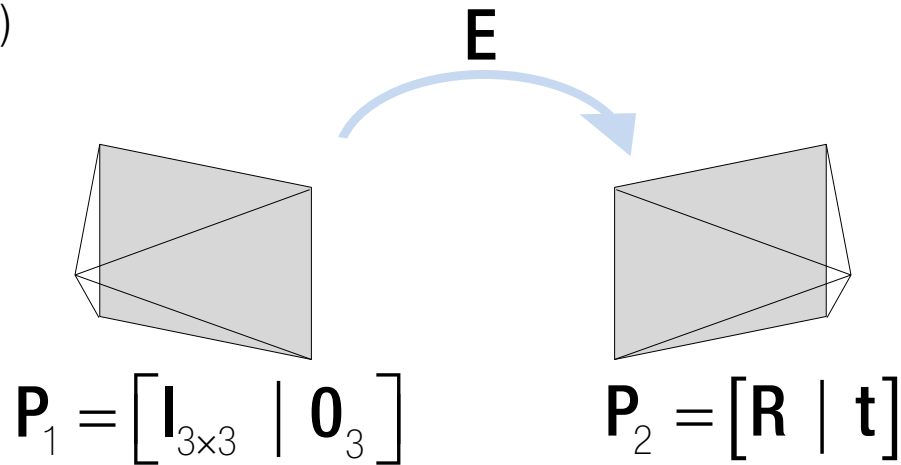
$$E = K^T F K$$

Bob $P_1 = K \begin{bmatrix} I_{3 \times 3} & 0 \end{bmatrix}$

Mike $P_2 = K \begin{bmatrix} R & t \end{bmatrix}$

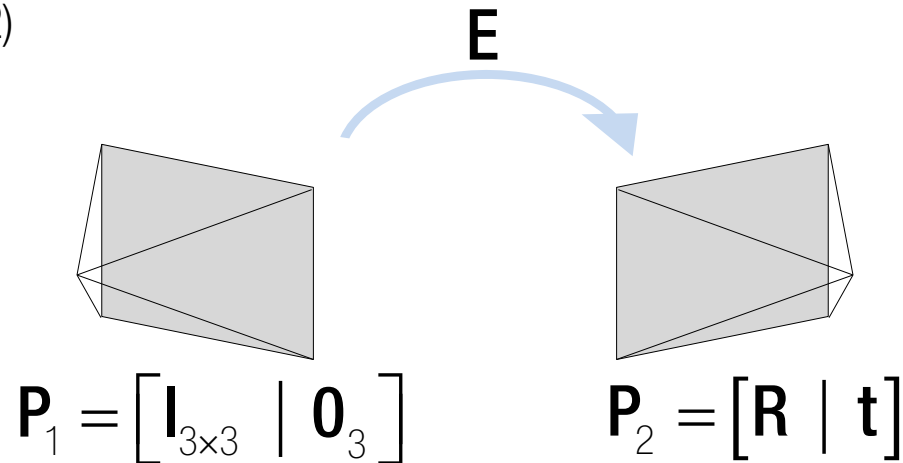
$\mathbf{E} = \begin{bmatrix} \mathbf{t} \end{bmatrix}_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix (rank 2)



$\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

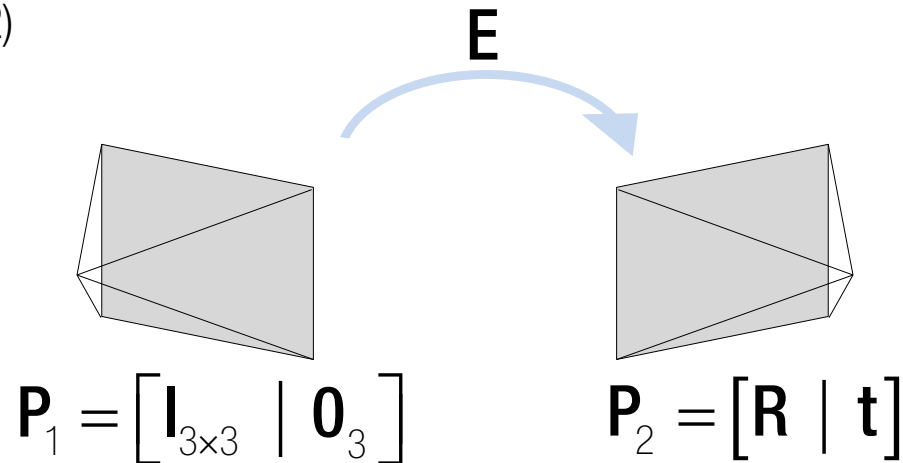
Essential matrix (rank 2)



\mathbf{t} : Epipole in image 2 because $\mathbf{P}_2 \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} = [\mathbf{R} \mid \mathbf{t}] \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} = \mathbf{t}$

$\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix (rank 2)

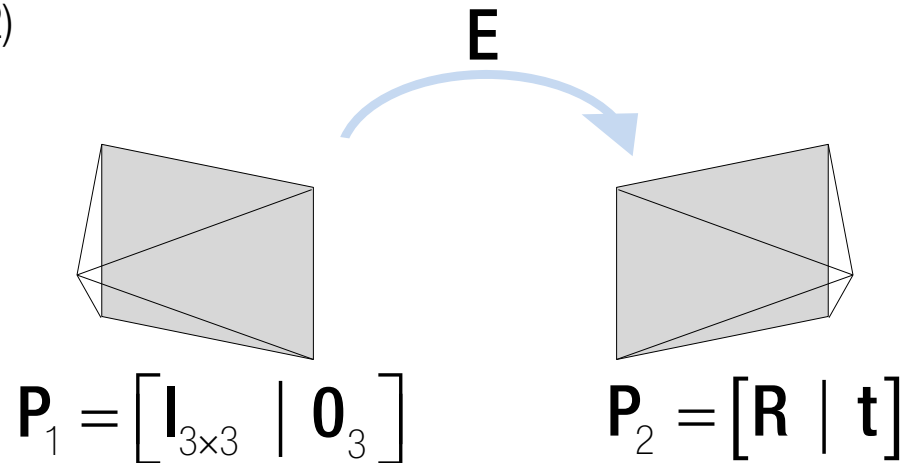


\mathbf{t} : Epipole in image 2 because $\mathbf{P}_2 \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} = [\mathbf{R} \mid \mathbf{t}] \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} = \mathbf{t}$

$\mathbf{t}^T \mathbf{E} = \mathbf{0}$: Left nullspace of the essential matrix is the epipole in image 2.

$\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix (rank 2)



\mathbf{t} : Epipole in image 2 because $\mathbf{P}_2 \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} = [\mathbf{R} \mid \mathbf{t}] \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} = \mathbf{t}$

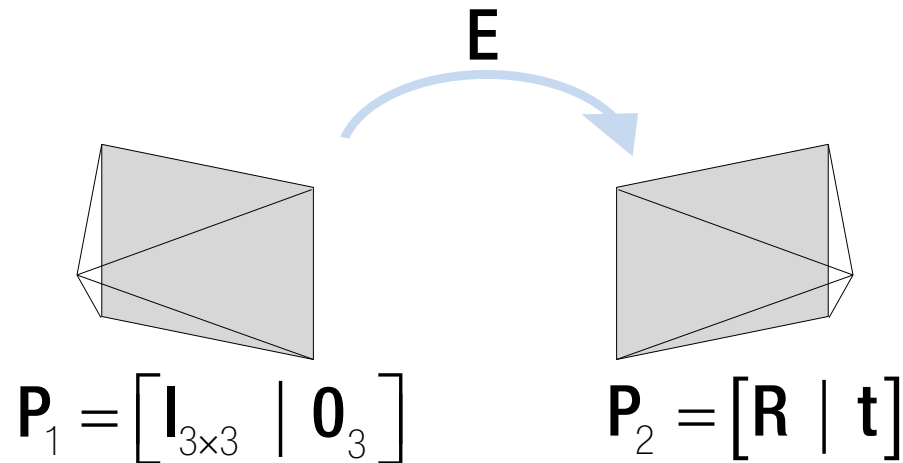
$\mathbf{t}^T \mathbf{E} = \mathbf{0}$: Left nullspace of the essential matrix is the epipole in image 2.

$\rightarrow \mathbf{t} = \mathbf{u}_3$, or $-\mathbf{u}_3$ where $\mathbf{U} = [\mathbf{u}_1 \quad \mathbf{u}_2 \quad \mathbf{u}_3]$ and $\mathbf{E} = \mathbf{U} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T$

Singular value decomposition (SVD)

$\mathbf{E} = \begin{bmatrix} \mathbf{t} \end{bmatrix}_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

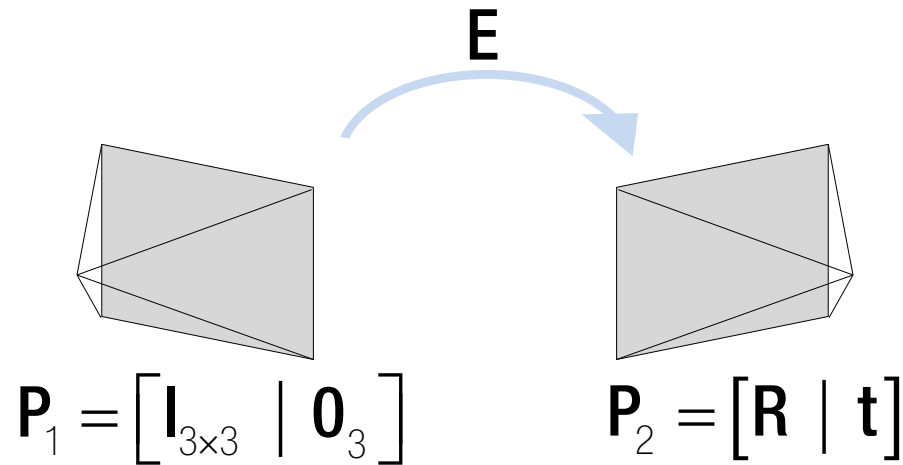
Essential matrix



$$\mathbf{E} = \mathbf{U} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T$$

$\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

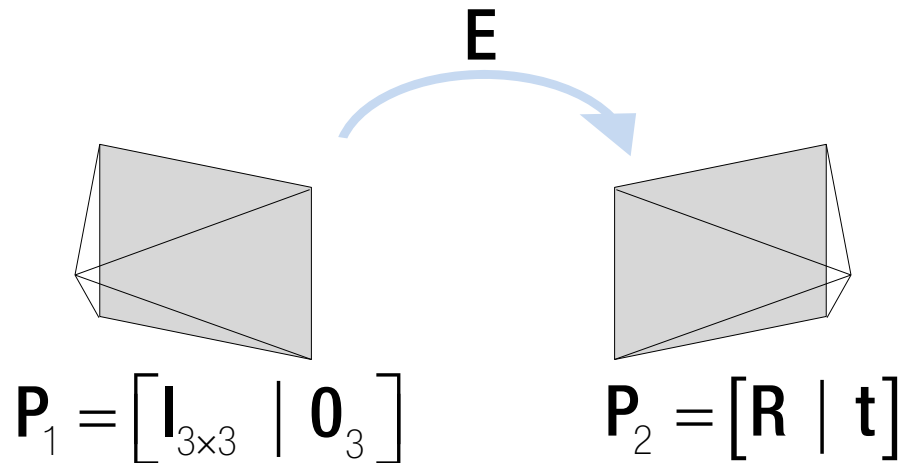
Essential matrix



$$\mathbf{E} = \mathbf{U} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T = [\mathbf{t}]_{\times} \mathbf{R}$$

$\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix

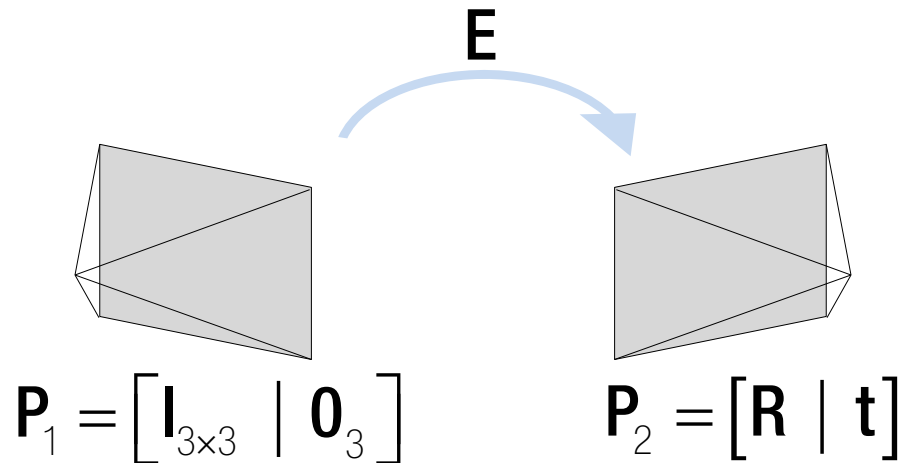


$$\mathbf{E} = \mathbf{U} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T = [\mathbf{t}]_{\times} \mathbf{R} = \left(\mathbf{U} \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{U}^T \right) (\mathbf{U} \mathbf{Y} \mathbf{V}^T)$$

where $\mathbf{U} = [\mathbf{u}_1 \quad \mathbf{u}_2 \quad \mathbf{t}]$

$\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix



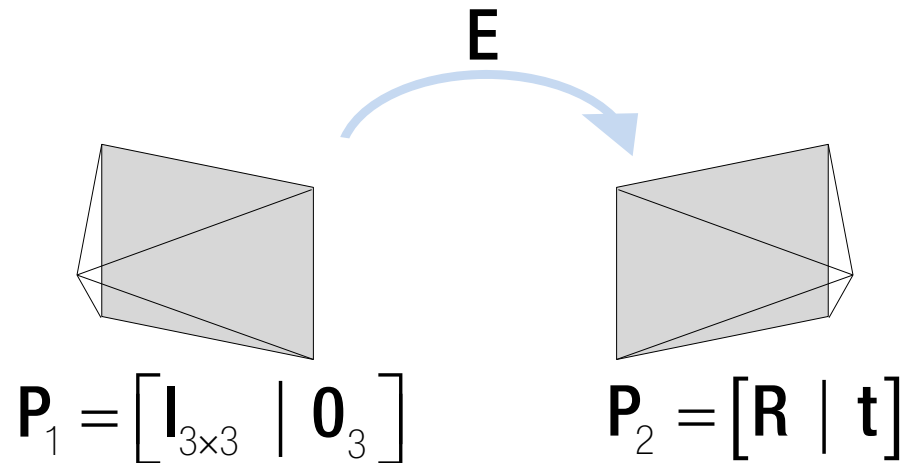
$$\mathbf{E} = \mathbf{U} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T = [\mathbf{t}]_{\times} \mathbf{R} = \left(\mathbf{U} \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{U}^T \right) (\mathbf{U} \mathbf{Y} \mathbf{V}^T)$$

How do we set \mathbf{Y} ?

where $\mathbf{U} = [\mathbf{u}_1 \quad \mathbf{u}_2 \quad \mathbf{t}]$

$\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix



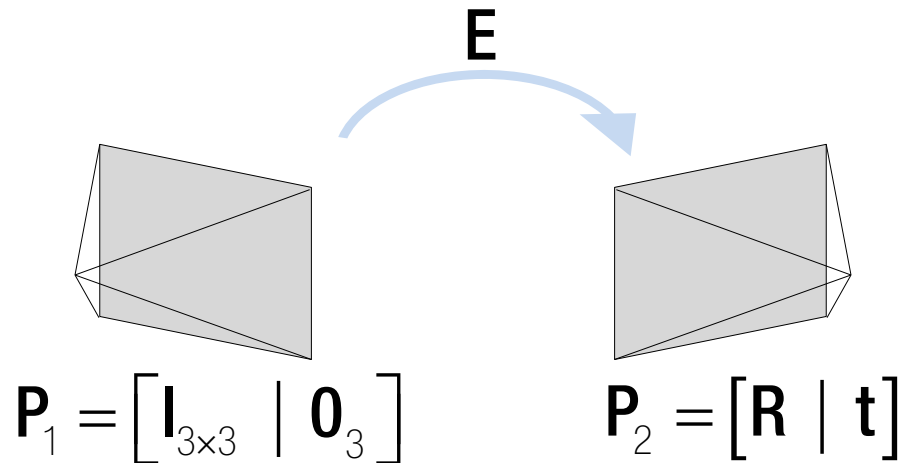
$$\mathbf{E} = \mathbf{U} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T = [\mathbf{t}]_{\times} \mathbf{R} = \mathbf{U} \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{Y} \mathbf{V}^T$$

How do we set \mathbf{Y} ?

$$\therefore \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{Y}$$

$\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix



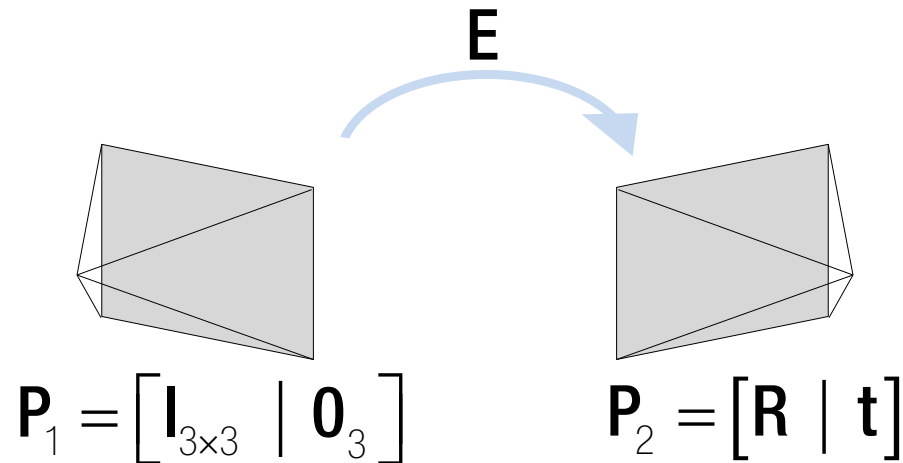
$$\mathbf{E} = \mathbf{U} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{V}^T = [\mathbf{t}]_{\times} \mathbf{R} = \mathbf{U} \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{Y} \mathbf{V}^T$$

How do we set \mathbf{Y} ?

$$\therefore \mathbf{Y} = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \text{ or } \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}^T$$

$\mathbf{E} = [\mathbf{t}]_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix



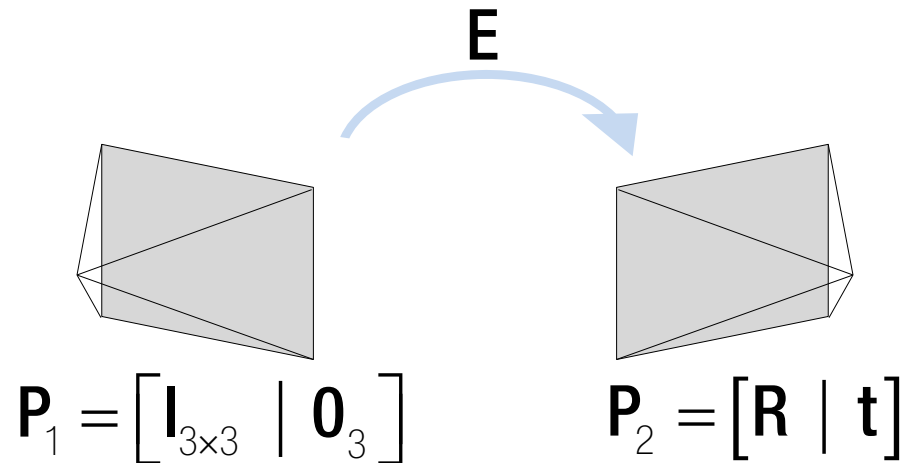
$$\mathbf{t} = \mathbf{u}_3, \text{ or } -\mathbf{u}_3$$

$$\mathbf{R} = \mathbf{U} \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \mathbf{V}^T, \text{ or } \mathbf{U} \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}^T \mathbf{V}^T$$

where $\mathbf{E} = \mathbf{U} \mathbf{D} \mathbf{V}^T$ $\mathbf{U} = [\mathbf{u}_1 \quad \mathbf{u}_2 \quad \mathbf{t}]$

$\mathbf{E} = \begin{bmatrix} \mathbf{t} \\ \times \end{bmatrix} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix



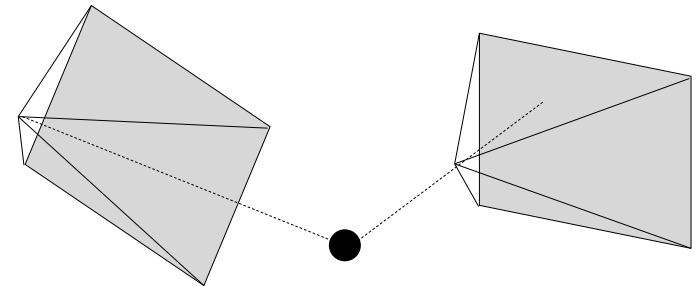
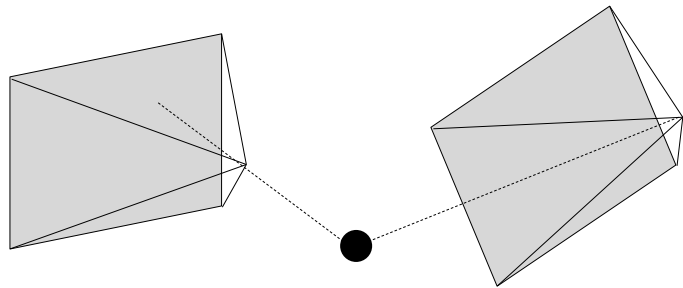
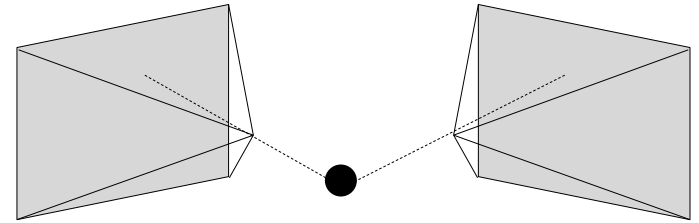
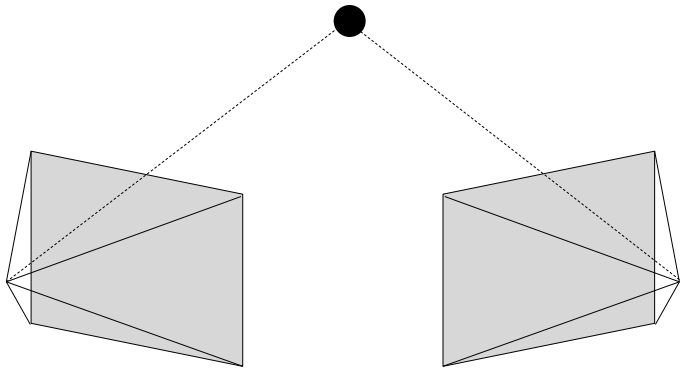
Four configurations:

$$\mathbf{P}_2 = \begin{bmatrix} \mathbf{U}\mathbf{Y}\mathbf{V}^T & | & \mathbf{u}_3 \end{bmatrix}, \text{ or } \begin{bmatrix} \mathbf{U}\mathbf{Y}^T\mathbf{V}^T & | & \mathbf{u}_3 \end{bmatrix}, \text{ or } \begin{bmatrix} \mathbf{U}\mathbf{Y}\mathbf{V}^T & | & -\mathbf{u}_3 \end{bmatrix}, \text{ or } \begin{bmatrix} \mathbf{U}\mathbf{Y}^T\mathbf{V}^T & | & -\mathbf{u}_3 \end{bmatrix}$$

$\mathbf{E} = \begin{bmatrix} \mathbf{t} \\ \times \end{bmatrix} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix

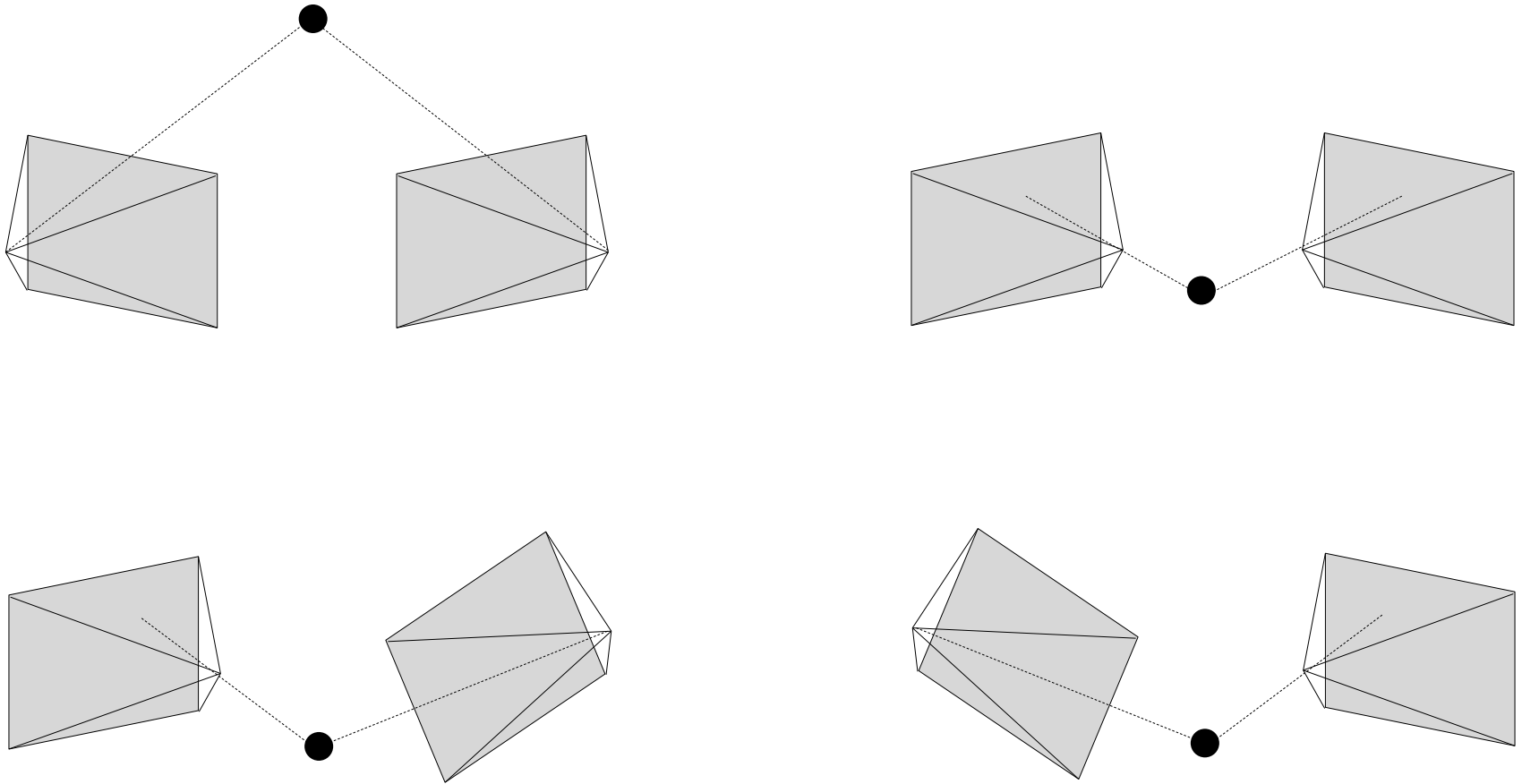
Four configurations:



$\mathbf{E} = \begin{bmatrix} \mathbf{t} \end{bmatrix}_{\times} \mathbf{R}$ How to decompose the essential matrix to rotation and translation?

Essential matrix

Four configurations: can be resolved by point triangulation.



Four possible reconstructions from E

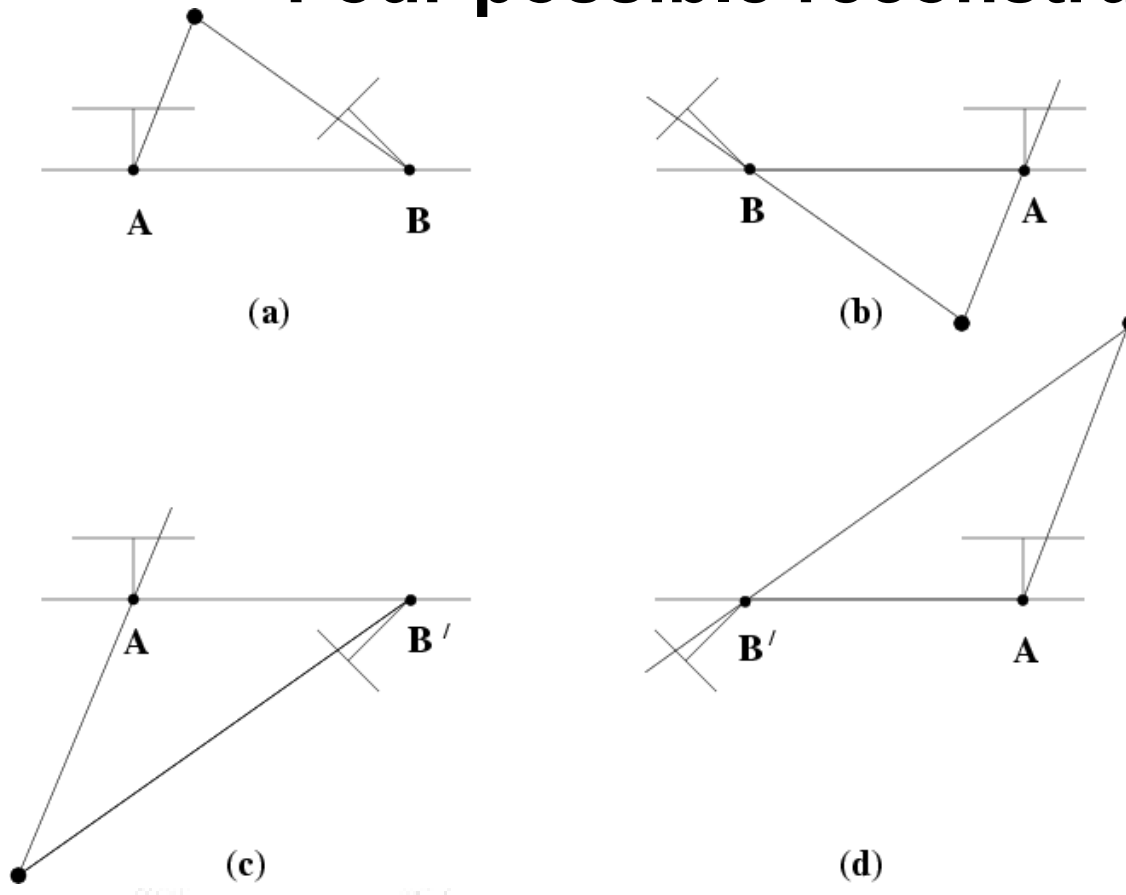


Fig. 9.12. The four possible solutions for calibrated reconstruction from E. Between the left and right sides there is a baseline reversal. Between the top and bottom rows camera B rotates 180° about the baseline. Note, only in (a) is the reconstructed point in front of both cameras.

Result 9.19. For a given essential matrix $E = U \text{diag}(1, 1, 0) V^T$, and first camera matrix $P = [I \mid \mathbf{0}]$, there are four possible choices for the second camera matrix P' , namely

$$P' = [UWV^T \mid +\mathbf{u}_3] \text{ or } [UWV^T \mid -\mathbf{u}_3] \text{ or } [UW^T V^T \mid +\mathbf{u}_3] \text{ or } [UW^T V^T \mid -\mathbf{u}_3].$$

(only one solution where points is in front of both cameras)

2.2 Camera Pose Extraction

Goal Given \mathbf{E} , enumerate four camera pose configurations, $(\mathbf{C}_1, \mathbf{R}_1)$, $(\mathbf{C}_2, \mathbf{R}_2)$, $(\mathbf{C}_3, \mathbf{R}_3)$, and $(\mathbf{C}_4, \mathbf{R}_4)$ where $\mathbf{C} \in \mathbb{R}^3$ is the camera center and $\mathbf{R} \in SO(3)$ is the rotation matrix, i.e., $\mathbf{P} = \mathbf{KR} \begin{bmatrix} \mathbf{I}_{3 \times 3} & -\mathbf{C} \end{bmatrix}$:

`[Cset Rset] = ExtractCameraPose(E)`

(INPUT) \mathbf{E} : essential matrix

(OUTPUT) \mathbf{Cset} and \mathbf{Rset} : four configurations of camera centers and rotations, i.e., $\mathbf{Cset}\{i\} = \mathbf{C}_i$ and $\mathbf{Rset}\{i\} = \mathbf{R}_i$.

There are four camera pose configurations given an essential matrix. Let $\mathbf{E} = \mathbf{UDV}^T$ and $\mathbf{W} = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$. The four configurations are enumerated below:

1. $\mathbf{C}_1 = \mathbf{U}(:, 3)$ and $\mathbf{R}_1 = \mathbf{UWV}^T$
2. $\mathbf{C}_2 = -\mathbf{U}(:, 3)$ and $\mathbf{R}_2 = \mathbf{UWV}^T$
3. $\mathbf{C}_3 = \mathbf{U}(:, 3)$ and $\mathbf{R}_3 = \mathbf{UW}^T\mathbf{V}^T$
4. $\mathbf{C}_4 = -\mathbf{U}(:, 3)$ and $\mathbf{R}_4 = \mathbf{UW}^T\mathbf{V}^T$.

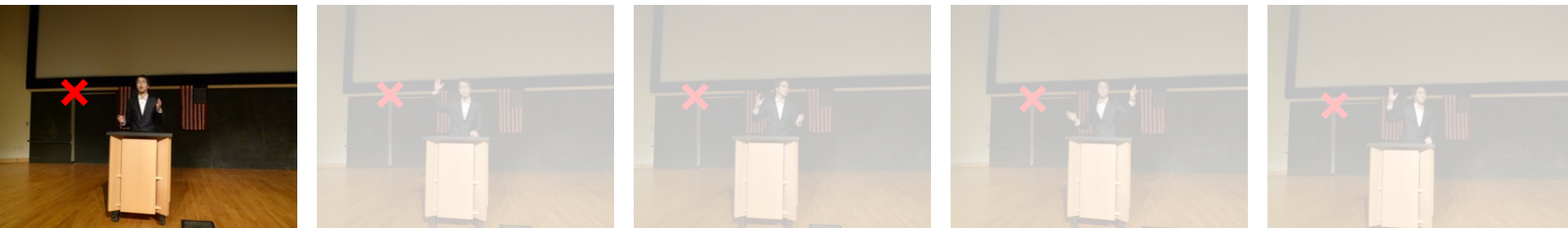
Note that the determinant of a rotation matrix is one. If $\det(\mathbf{R}) = -1$, the camera pose must be corrected, i.e., $\mathbf{C} \leftarrow -\mathbf{C}$ and $\mathbf{R} \leftarrow -\mathbf{R}$.

Point Triangulation

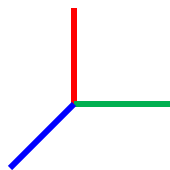


✗ 2D correspondences

Point Triangulation



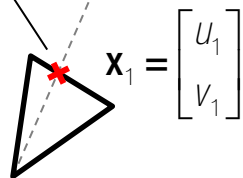
✗ 2D correspondences



3D point

$$\mathbf{x} = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

2D projection



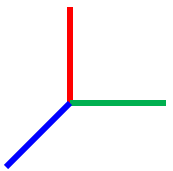
3D camera pose

$$\mathbf{P}_1 \in \mathbb{R}^{3 \times 4}$$

Point Triangulation



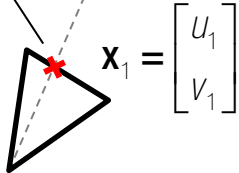
✗ 2D correspondences



3D point

$$\mathbf{x} = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

2D projection



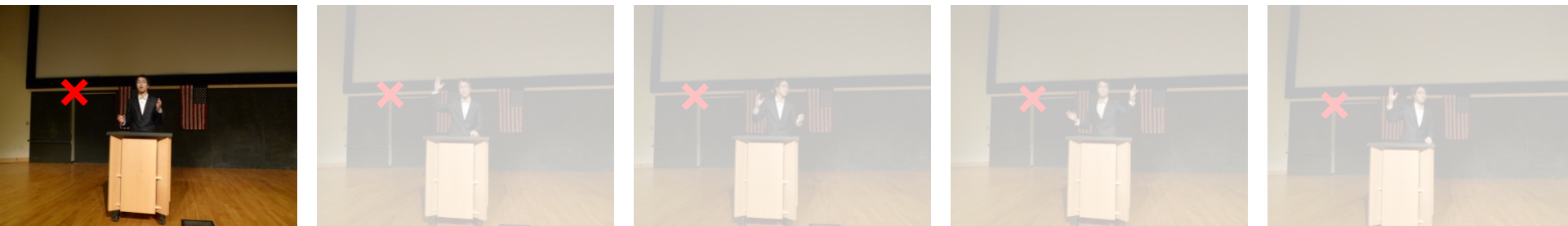
$$\mathbf{x}_1 = \begin{bmatrix} u_1 \\ v_1 \end{bmatrix}$$

3D camera pose

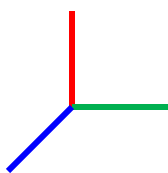
$$\mathbf{P}_1 \in \mathbb{R}^{3 \times 4}$$

$$\lambda \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix} = \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix}$$

Point Triangulation



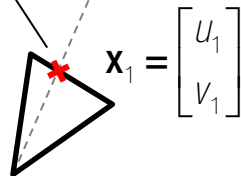
✗ 2D correspondences



3D point

$$\mathbf{x} = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

2D projection



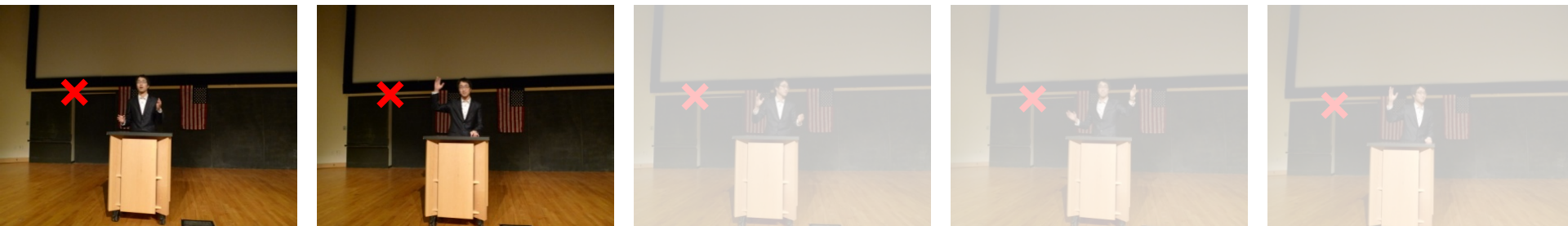
3D camera pose

$$\mathbf{P}_1 \in \mathbb{R}^{3 \times 4}$$

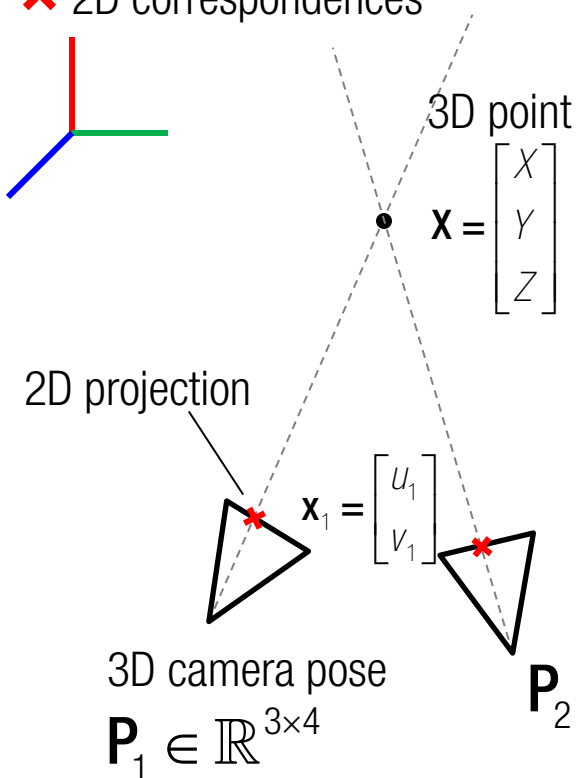
$$\lambda \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix} = \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} \longrightarrow \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix}_{\times} \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = \mathbf{0}$$

Cross product between two parallel vectors equals to zero.

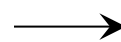
Point Triangulation



✗ 2D correspondences



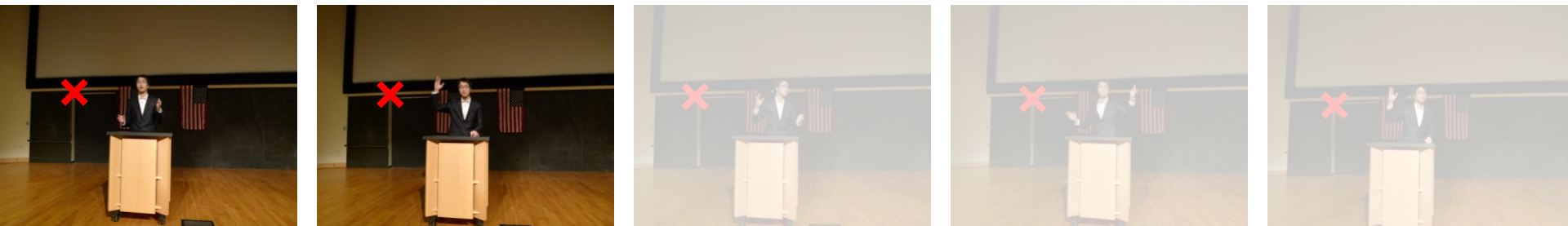
$$\lambda \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix} = \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix}$$



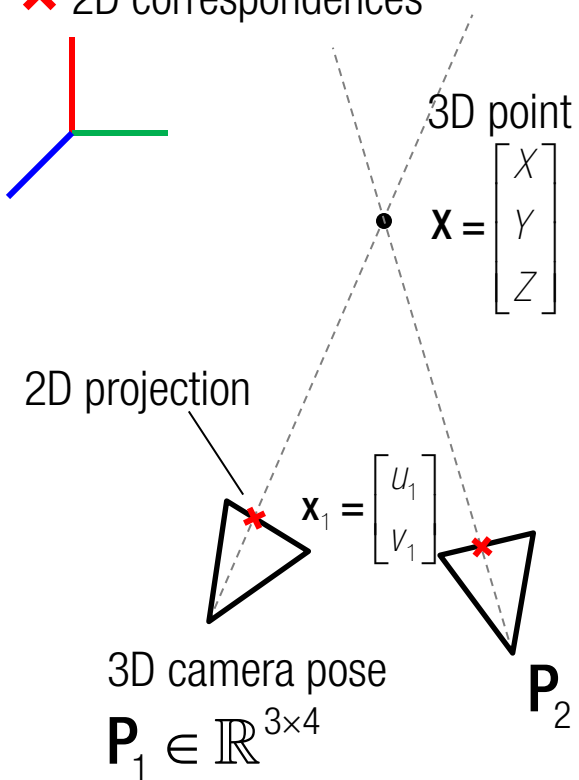
$$\begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix}_x \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

$$\begin{bmatrix} \mathbf{x}_2 \\ 1 \end{bmatrix}_x \mathbf{P}_2 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

Point Triangulation



✗ 2D correspondences



$$\lambda \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix} = \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} \longrightarrow \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix}_{\times} \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

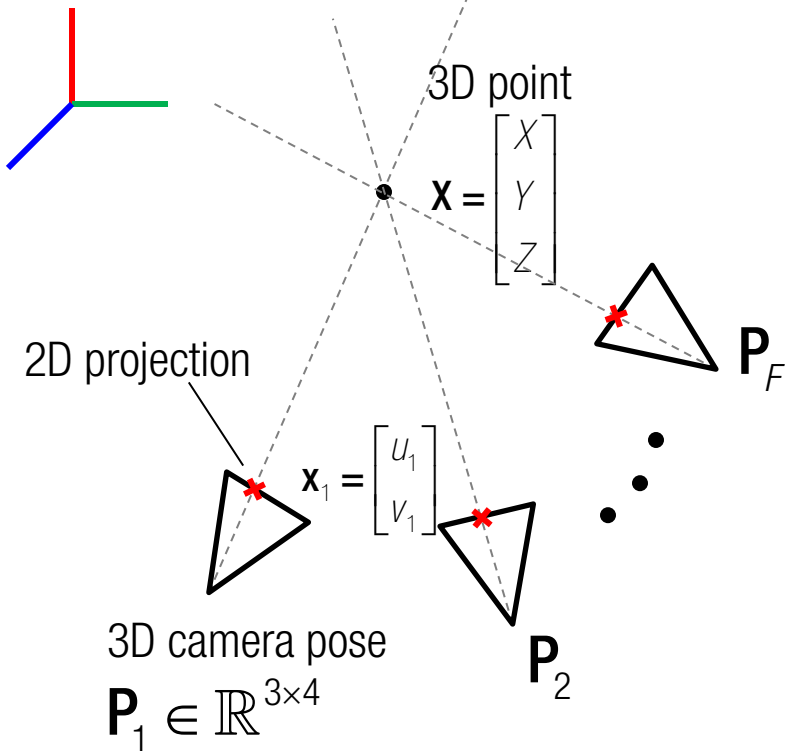
$$\begin{bmatrix} \mathbf{x}_2 \\ 1 \end{bmatrix}_{\times} \mathbf{P}_2 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

$$\begin{bmatrix} \mathbf{x}_1 \\ 1 \\ \mathbf{x}_2 \\ 1 \end{bmatrix}_{\times} \begin{bmatrix} \mathbf{P}_1 \\ \mathbf{P}_2 \end{bmatrix} \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

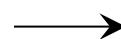
Point Triangulation



✗ 2D correspondences



$$\lambda \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix} = \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix}$$



$$\begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix}_x \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

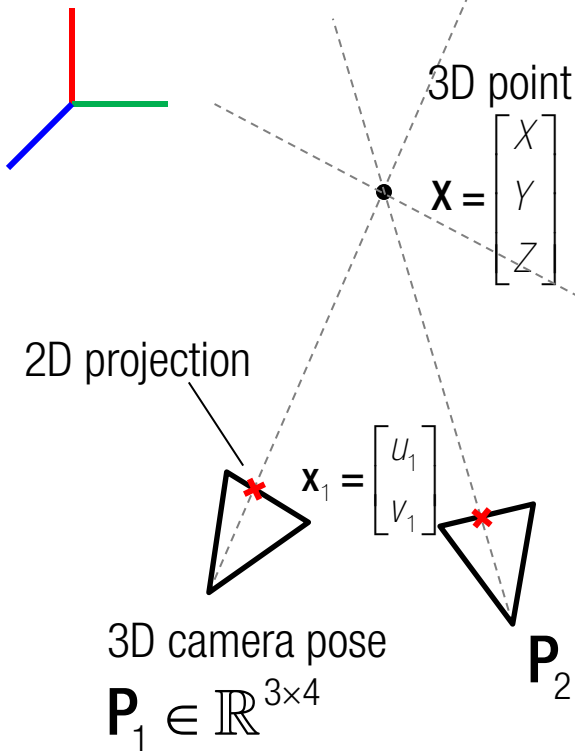
$$\begin{bmatrix} \mathbf{x}_2 \\ 1 \end{bmatrix}_x \mathbf{P}_2 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

$$\begin{bmatrix} \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix}_x \mathbf{P}_1 \\ \begin{bmatrix} \mathbf{x}_2 \\ 1 \end{bmatrix}_x \mathbf{P}_2 \\ \vdots \\ \begin{bmatrix} \mathbf{x}_F \\ 1 \end{bmatrix}_x \mathbf{P}_F \end{bmatrix} \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

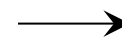
Point Triangulation



✗ 2D correspondences



$$\lambda \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix} = \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix}$$



$$\begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix}_x \mathbf{P}_1 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

$$\begin{bmatrix} \mathbf{x}_2 \\ 1 \end{bmatrix}_x \mathbf{P}_2 \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

$$3F \begin{bmatrix} \begin{bmatrix} \mathbf{x}_1 \\ 1 \end{bmatrix}_x \mathbf{P}_1 \\ \vdots \\ \begin{bmatrix} \mathbf{x}_F \\ 1 \end{bmatrix}_x \mathbf{P}_F \end{bmatrix} \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix} = 0$$

$$\text{rank} \left(\begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix}_x \mathbf{P} \right) = 2$$

Least squares if $F \geq 2$

Point Triangulation



$$P_1 = K_1 \begin{bmatrix} I_{3 \times 3} & 0_3 \end{bmatrix}$$

$$P_2 = K_2 \begin{bmatrix} I_{3 \times 3} & -C \end{bmatrix} \quad C = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

```
% Intrinsic parameter
K1 = [2329.558 0 1141.452; 0 2329.558 927.052; 0 0 1];
K2 = [2329.558 0 1241.731; 0 2329.558 927.052; 0 0 1];
```

```
% Camera matrices
P1 = K1 * [eye(3) zeros(3,1)];
C = [1;0;0];
P2 = K2 * [eye(3) -C];
```

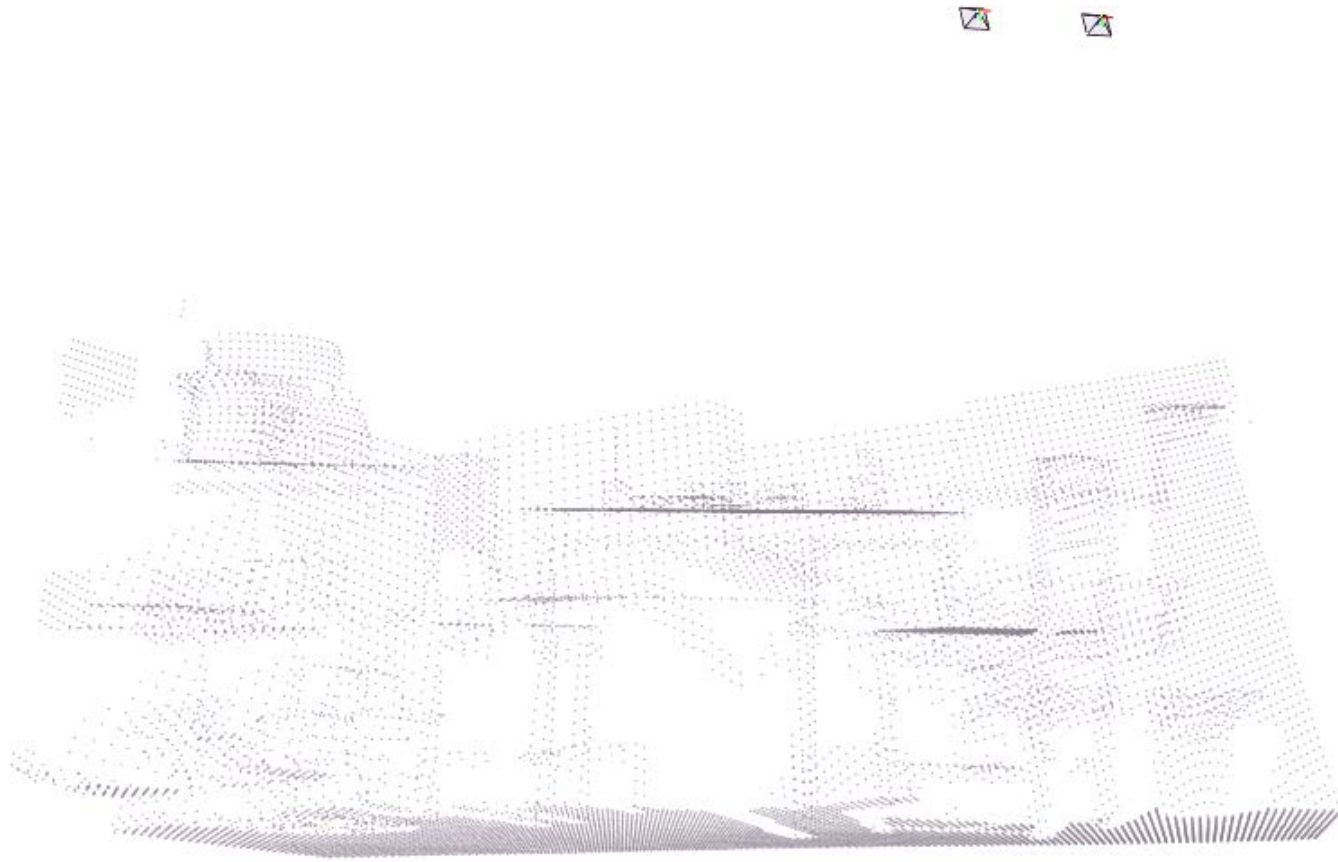
```
% Correspondences
x1 = [1382;986;1];
x2 = [1144;986;1];
skew1 = Vec2Skew(x1);
skew2 = Vec2Skew(x2);
```

```
% Solve
A = [skew1*P1; skew2*P2];
[u,d,v] = svd(A);
X = v(:,end)/v(end,end);
```

```
function skew = Vec2Skew(v)
skew = [0 -v(3) v(2); v(3) 0 -v(1); -v(2) v(1) 0];
```

```
X =
    0.7111
    0.1743
    6.8865
    1.0000
```

Point Triangulation



3.1 Linear Triangulation

Goal Given two camera poses, $(\mathbf{C}_1, \mathbf{R}_1)$ and $(\mathbf{C}_2, \mathbf{R}_2)$, and correspondences $\mathbf{x}_1 \leftrightarrow \mathbf{x}_2$, triangulate 3D points using linear least squares:

$\mathbf{X} = \text{LinearTriangulation}(\mathbf{K}, \mathbf{C}_1, \mathbf{R}_1, \mathbf{C}_2, \mathbf{R}_2, \mathbf{x}_1, \mathbf{x}_2)$

(INPUT) \mathbf{C}_1 and \mathbf{R}_1 : the first camera pose

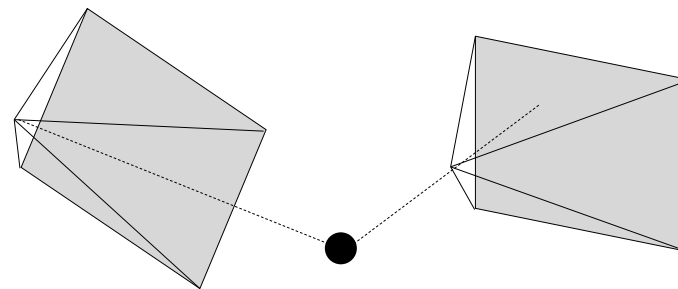
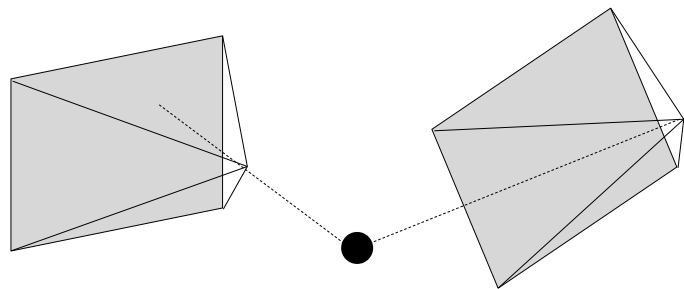
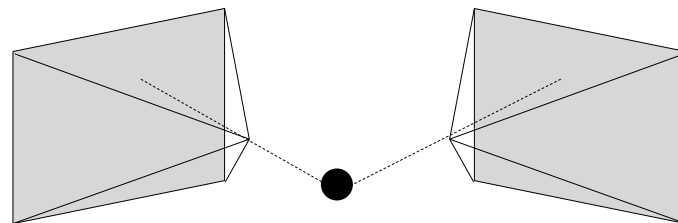
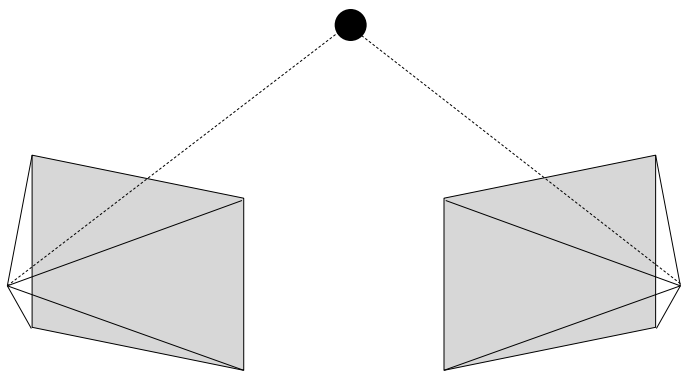
(INPUT) \mathbf{C}_2 and \mathbf{R}_2 : the second camera pose

(INPUT) \mathbf{x}_1 and \mathbf{x}_2 : two $N \times 2$ matrices whose row represents correspondence between the first and second images where N is the number of correspondences.

(OUTPUT) \mathbf{X} : $N \times 3$ matrix whose row represents 3D triangulated point.

Camera pose disambiguation via point triangulation

Four configurations:



3.2 Camera Pose Disambiguation

Goal Given four camera pose configuration and their triangulated points, find the unique camera pose by checking the *cheirality* condition—the reconstructed points must be in front of the cameras:

`[C R X0] = DisambiguateCameraPose(Cset, Rset, Xset)`

(INPUT) **Cset** and **Rset**: four configurations of camera centers and rotations

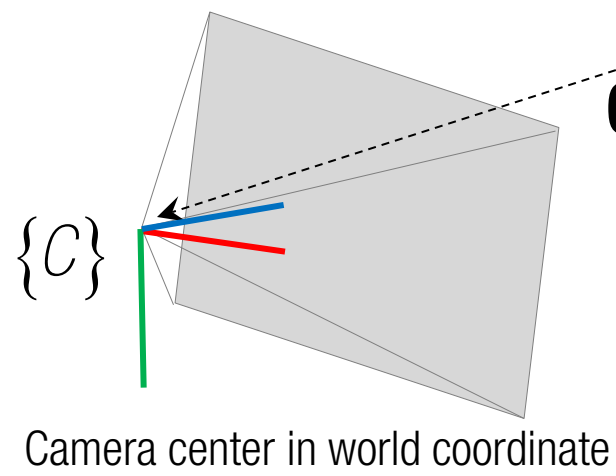
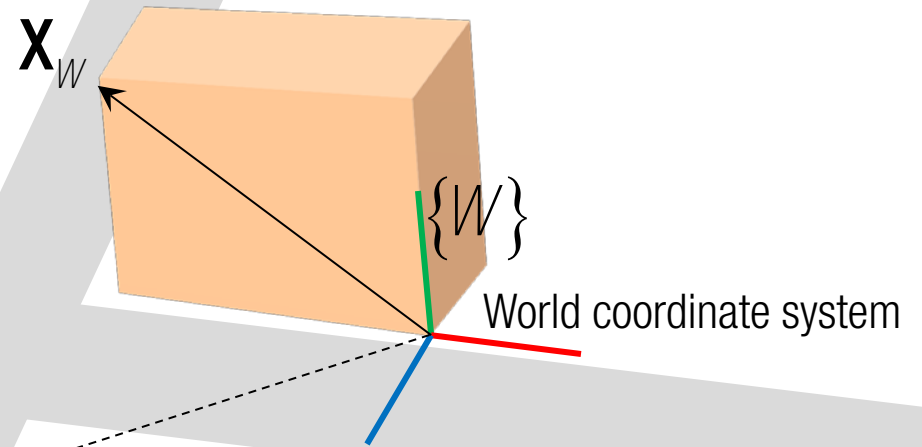
(INPUT) **Xset**: four sets of triangulated points from four camera pose configurations

(OUTPUT) **C** and **R**: the correct camera pose

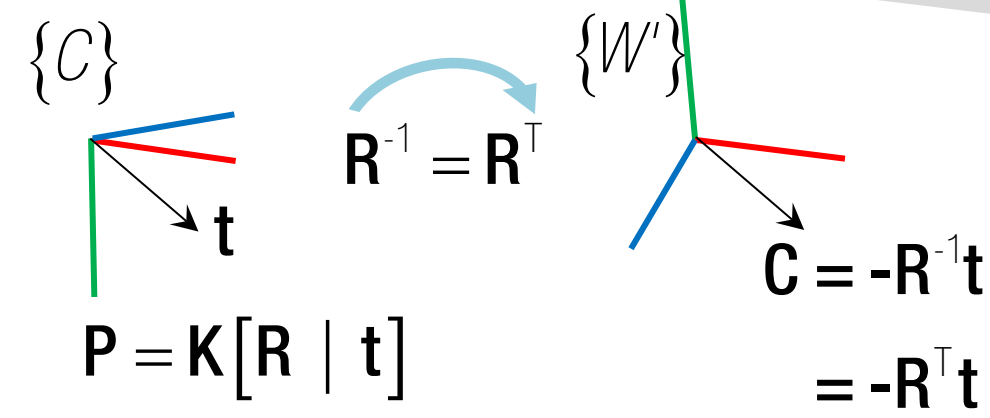
(OUTPUT) **X0**: the 3D triangulated points from the correct camera pose

The sign of the Z element in the camera coordinate system indicates the location of the 3D point with respect to the camera, i.e., a 3D point \mathbf{X} is in front of a camera if (\mathbf{C}, \mathbf{R}) if $\mathbf{r}_3(\mathbf{X} - \mathbf{C}) > 0$ where \mathbf{r}_3 is the third row of \mathbf{R} . Not all triangulated points satisfy this condition due to the presence of correspondence noise. The best camera configuration, $(\mathbf{C}, \mathbf{R}, \mathbf{X})$ is the one that produces the maximum number of points satisfying the cheirality condition.

Third person (world) perspective



$$\mathbf{C} = -\mathbf{R}^{-1}\mathbf{t}$$



$$\begin{aligned} \mathbf{P} &= \mathbf{K}[\mathbf{R} \mid \mathbf{t}] \\ &= \mathbf{K}[\mathbf{R} \mid -\mathbf{R}\mathbf{C}] \\ &= \mathbf{K}\mathbf{R}[\mathbf{I}_{3 \times 3} \mid \underline{-\mathbf{C}}] \end{aligned}$$

$$\begin{aligned} \mathbf{C} &= -\mathbf{R}^{-1}\mathbf{t} \\ &= -\mathbf{R}^T\mathbf{t} \end{aligned}$$

Camera center seen from world coordinate system