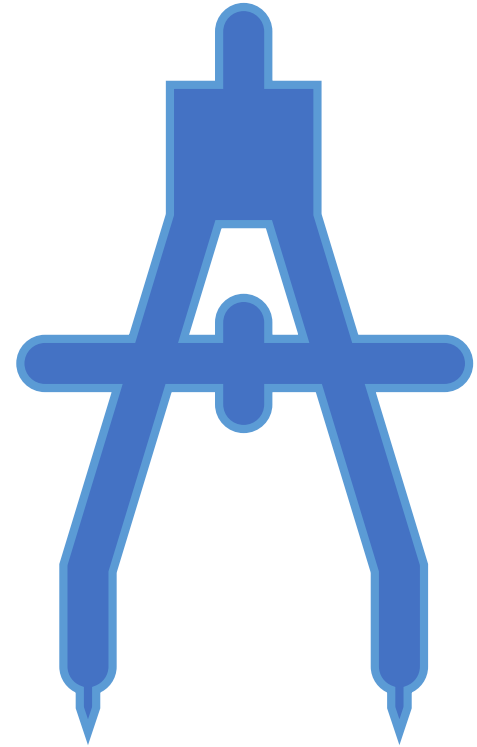


# Epipolar Geometry

Samuel Cheng

Slide credit: James Tompkin, Naoh  
Snaveley





# Structure from motion



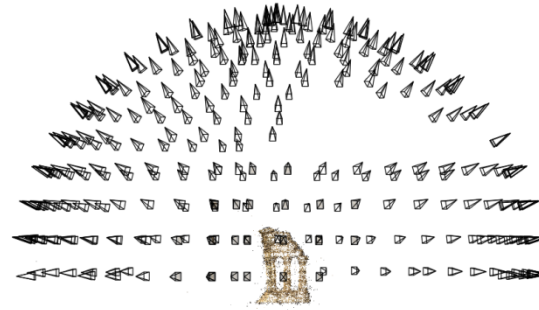
# Structure from motion

- Given many images, how can we
  - a) figure out where they were all taken from?
  - b) build a 3D model of the scene?

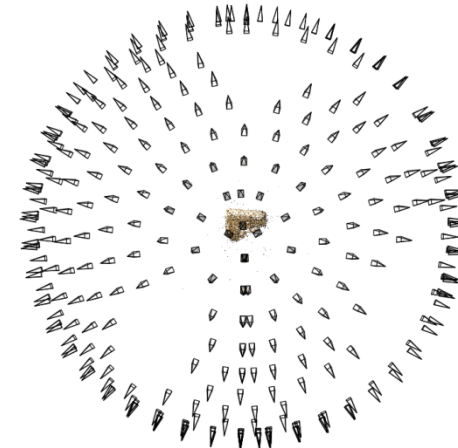


This is (roughly) the **structure from motion** problem

# Structure from motion



Reconstruction (side)



(top)

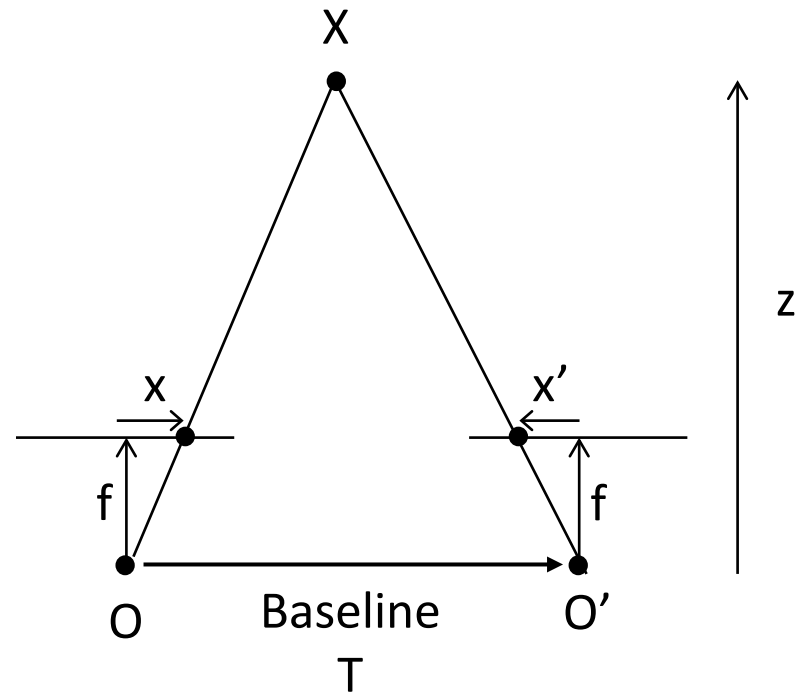
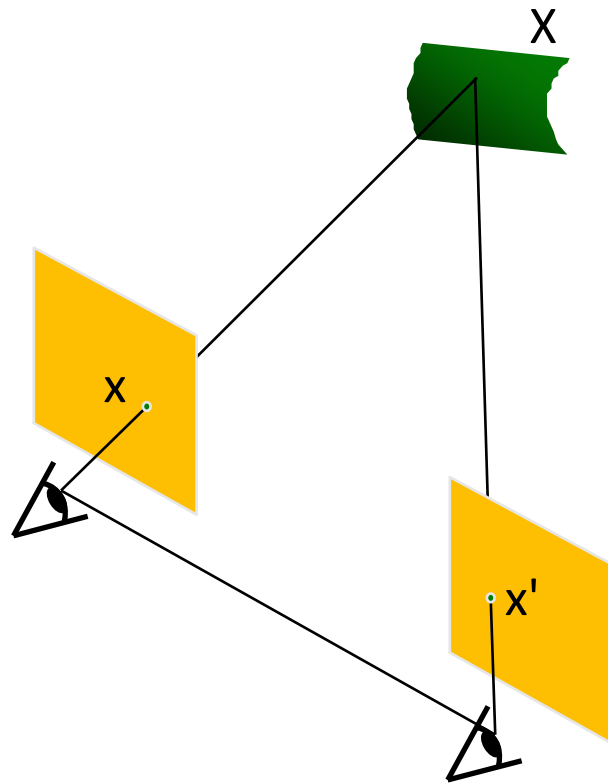
- Input: images with points in correspondence  
 $p_{i,j} = (u_{i,j}, v_{i,j})$
- Output
  - structure: 3D location  $\mathbf{x}_i$  for each point  $p_i$
  - motion: camera parameters  $\mathbf{R}_j$ ,  $\mathbf{t}_j$  possibly  $\mathbf{K}_j$
- Objective function: minimize *reprojection error*

# What we've seen so far...

- 2D transformations between images
  - Translations, affine transformations, homographies...
- 3D coordinates to 2D coordinates
  - Camera matrix
- Today: epipolar geometry and fundamental matrices

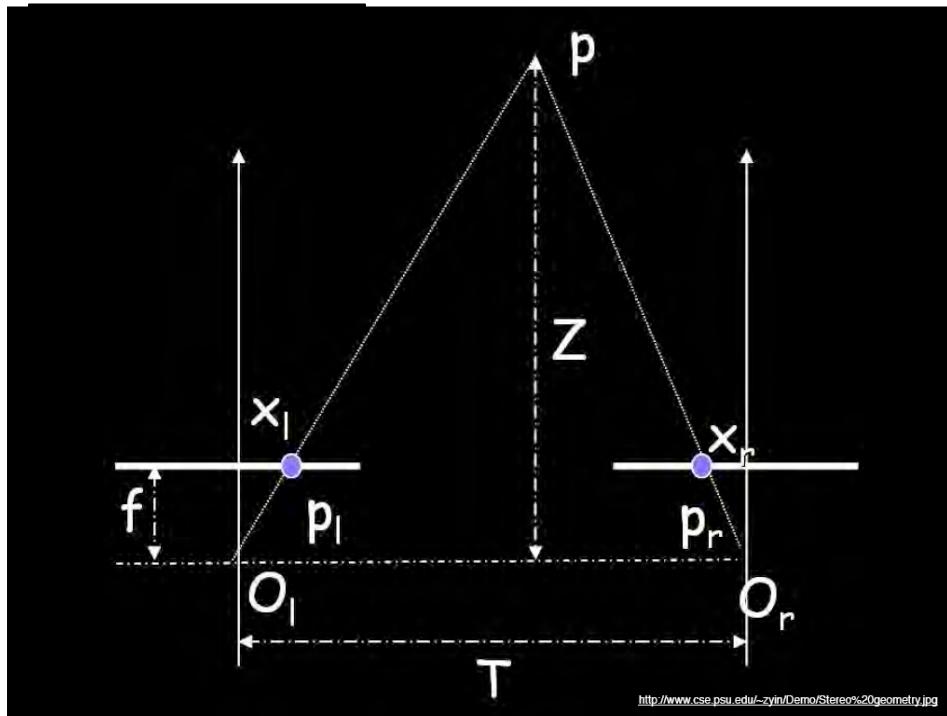
# Depth from disparity

- Goal: recover depth by finding image coordinate  $x'$  that corresponds to  $x$



# Geometry for a simple stereo system

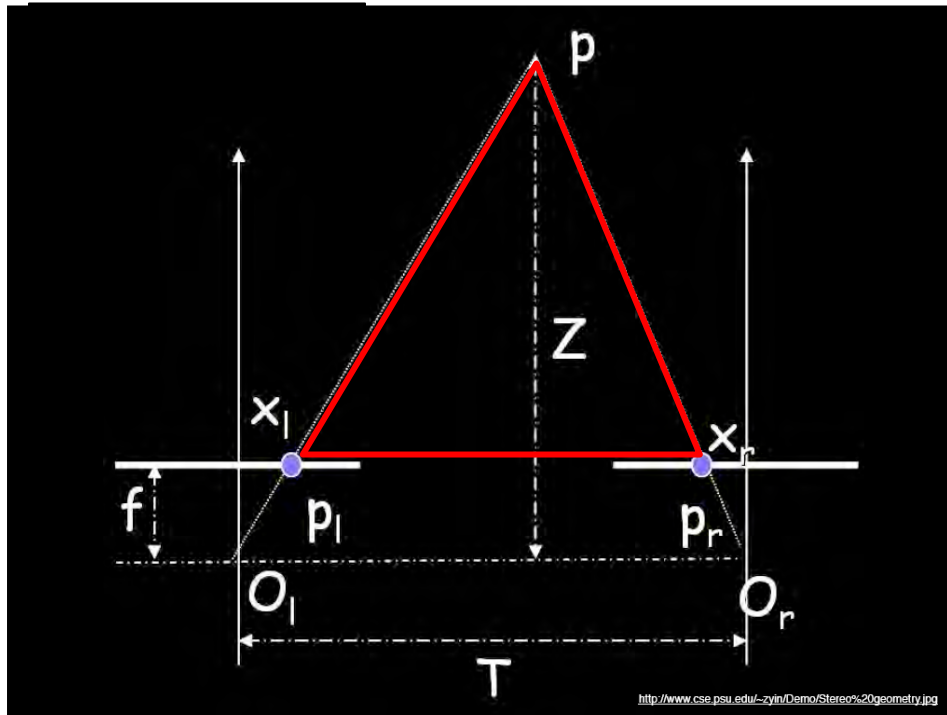
- Assume parallel optical axes, known camera parameters (i.e., calibrated cameras). **What is expression for  $Z$ ?**





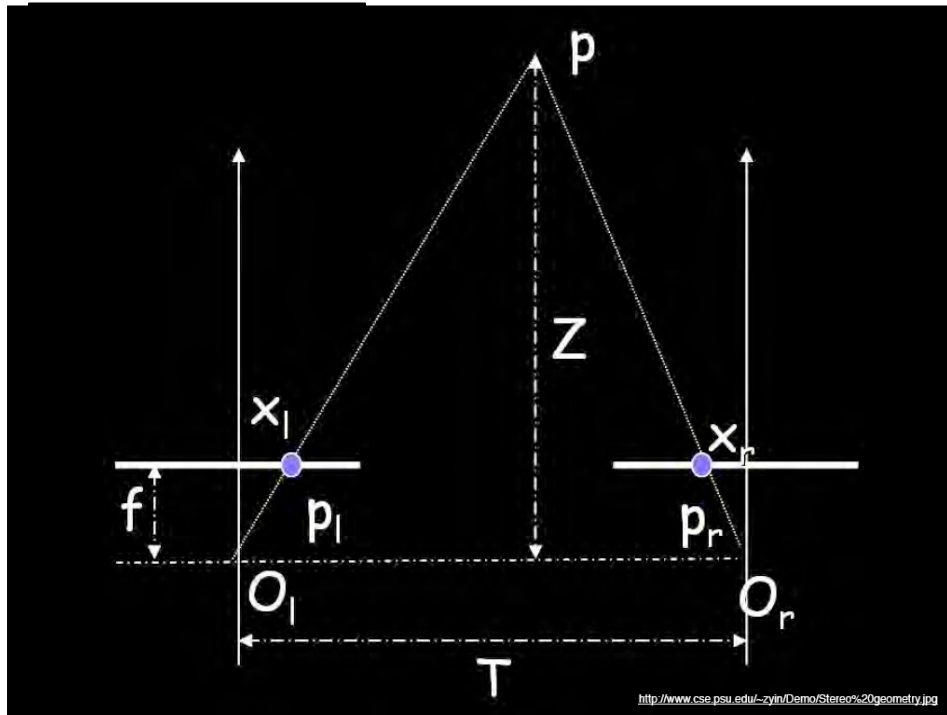
# Geometry for a simple stereo system

- Assume parallel optical axes, known camera parameters (i.e., calibrated cameras). **What is expression for  $Z$ ?**



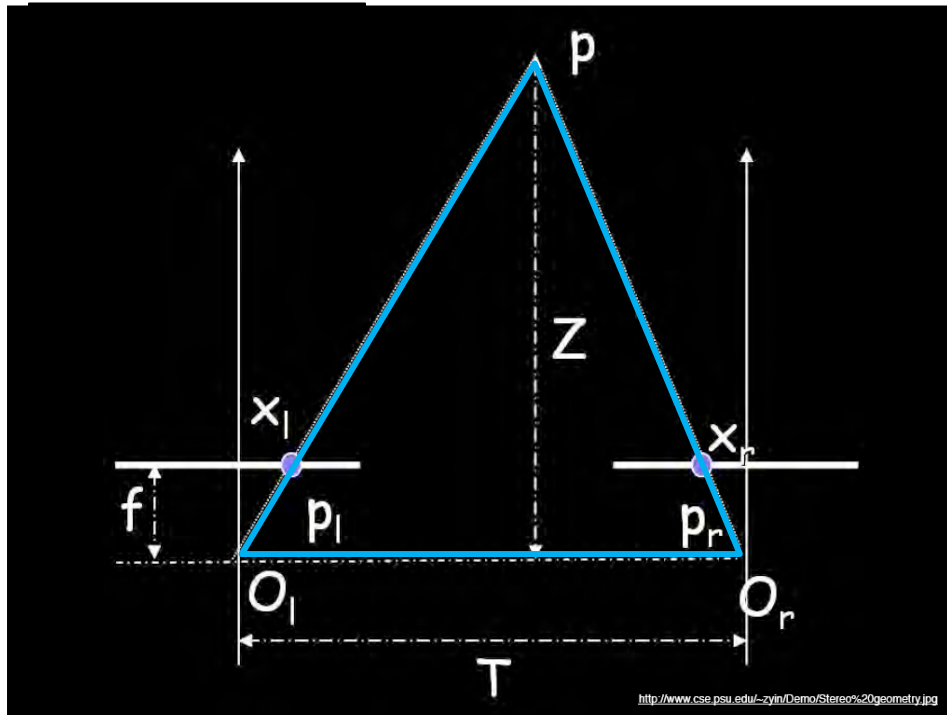
# Geometry for a simple stereo system

- Assume parallel optical axes, known camera parameters (i.e., calibrated cameras). **What is expression for  $Z$ ?**



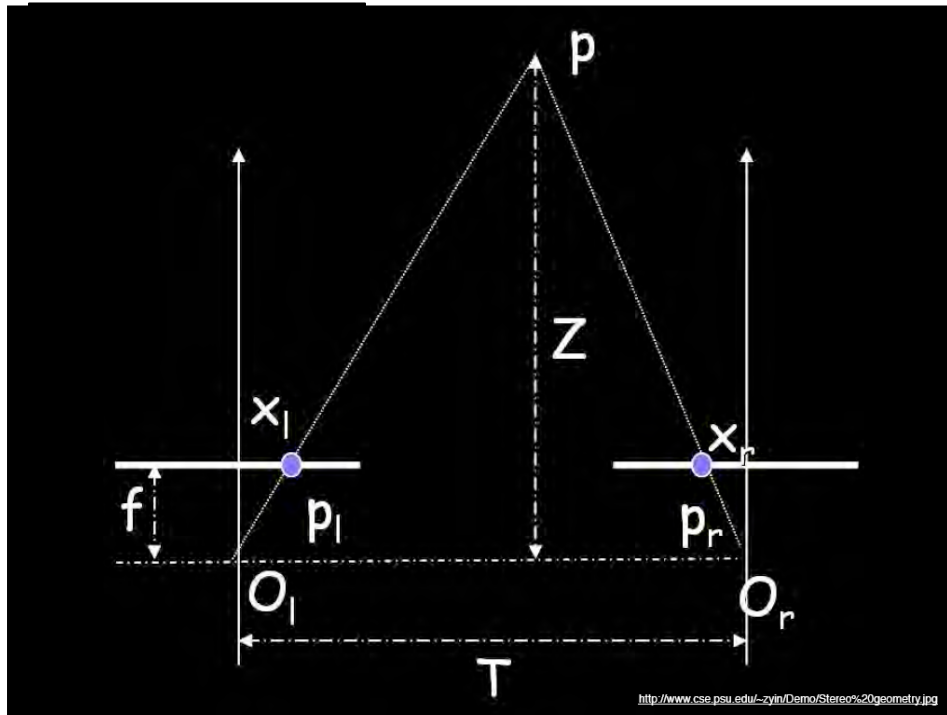
# Geometry for a simple stereo system

- Assume parallel optical axes, known camera parameters (i.e., calibrated cameras). **What is expression for  $Z$ ?**



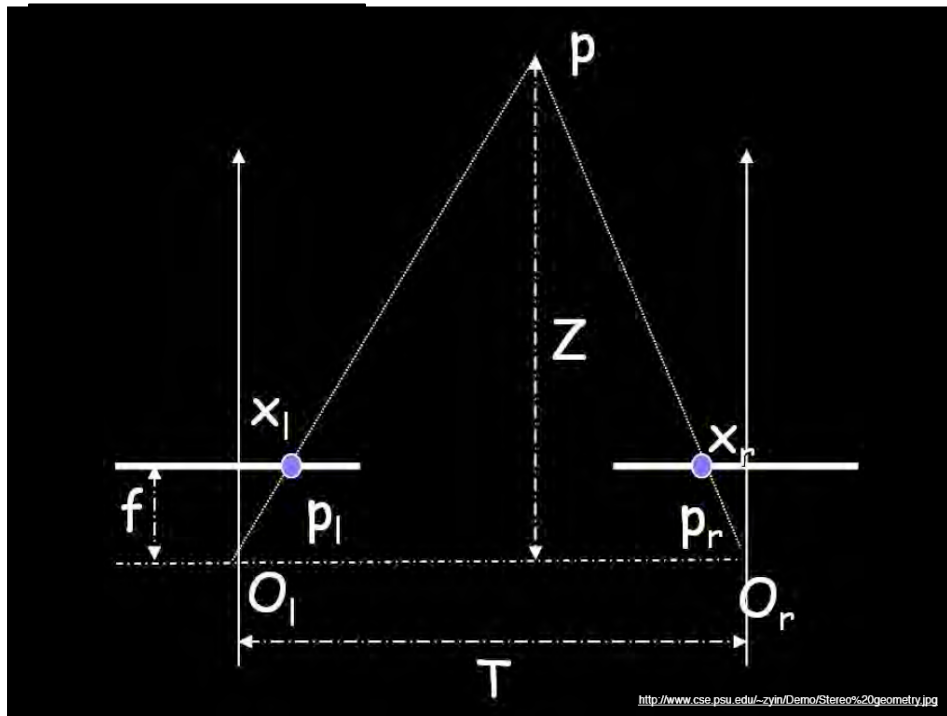
# Geometry for a simple stereo system

- Assume parallel optical axes, known camera parameters (i.e., calibrated cameras). **What is expression for  $Z$ ?**



# Geometry for a simple stereo system

- Assume parallel optical axes, known camera parameters (i.e., calibrated cameras). **What is expression for Z?**



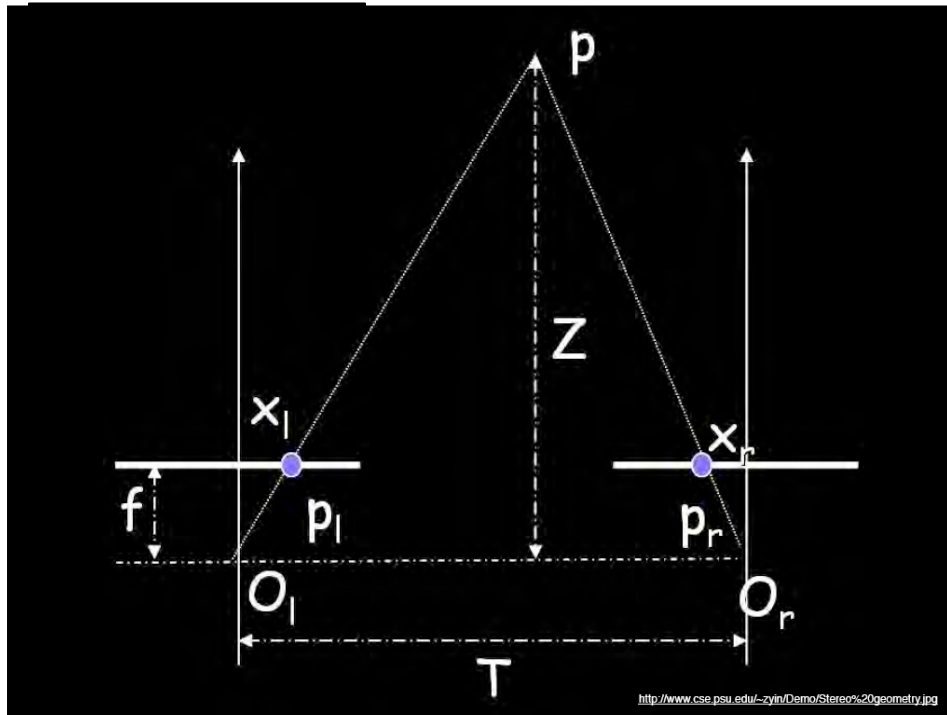
Similar triangles  $(p_l, P, p_r)$  and  $(O_l, P, O_r)$ :

$$\frac{T + x_r - x_l}{Z - f} = \frac{T}{Z}$$

$$Z = f \frac{T}{x_l - x_r}$$

# Geometry for a simple stereo system

- Assume parallel optical axes, known camera parameters (i.e., calibrated cameras). **What is expression for Z?**



Similar triangles ( $p_l, P, p_r$ ) and ( $O_l, P, O_r$ ):

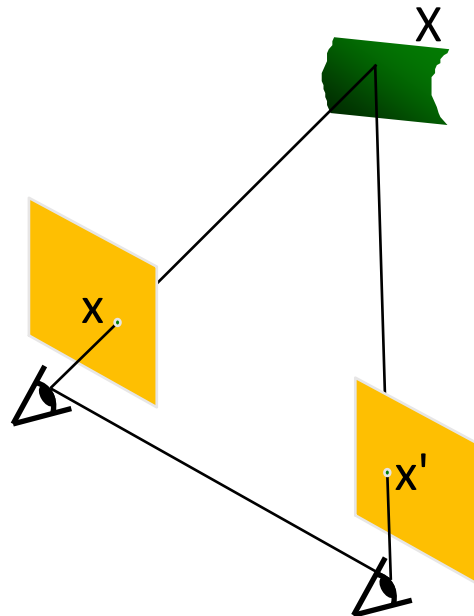
$$\frac{T + x_r - x_l}{Z - f} = \frac{T}{Z}$$

$$Z = f \frac{T}{x_l - x_r}$$

disparity  $\rightarrow$   $x_l - x_r$

# Depth from disparity

- Goal: recover depth by finding image coordinate  $x'$  that corresponds to  $x$
- Sub-Problems
  1. Calibration: How do we recover the relation of the cameras (if not already known)?
  2. Correspondence: How do we search for the matching point  $x'$ ?



# Depth from disparity

image  $I(x,y)$



Disparity map  $D(x,y)$

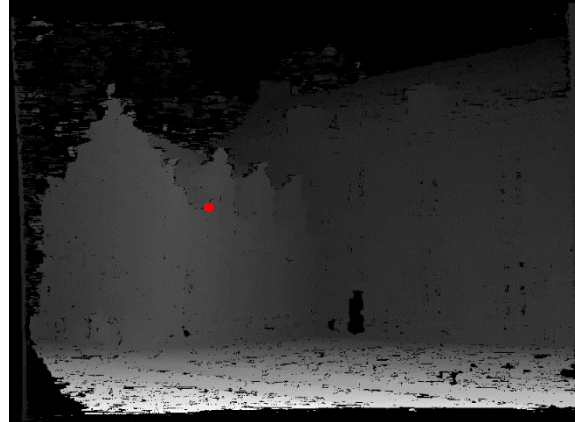
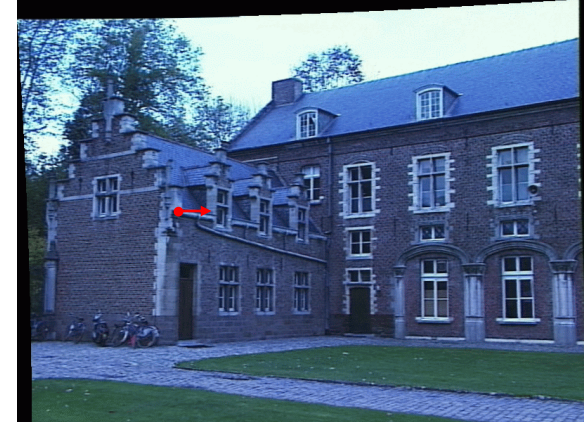


image  $I'(x',y')$



$$(x',y')=(x+D(x,y), y)$$



# Depth from disparity

image  $I(x,y)$



Disparity map  $D(x,y)$

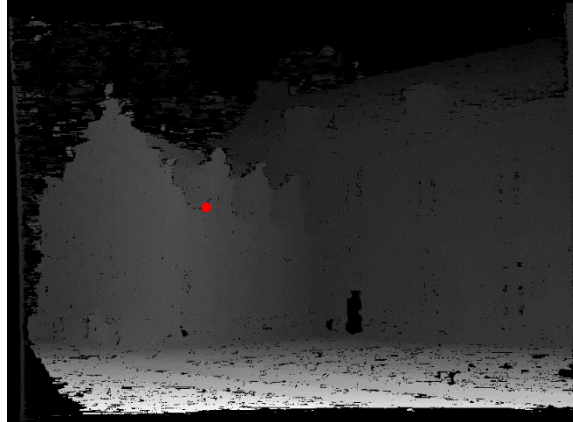


image  $I'(x',y')$



$$(x',y')=(x+D(x,y), y)$$

If we could find the **corresponding points** in two images, we could **estimate relative depth**...

# What do we need to know?

1. Calibration for the two cameras.
  1. Intrinsic matrices for both cameras (e.g.,  $f$ )
  2. Baseline distance  $T$  in parallel camera case
  3.  $R, t$  in non-parallel case
2. Correspondence for every pixel.

Correspondence for every pixel.  
Where do we need to search?



Correspondence for every pixel.  
Where do we need to search?





Correspondence for every pixel.  
Where do we need to search?



Correspondence for every pixel.  
Where do we need to search?

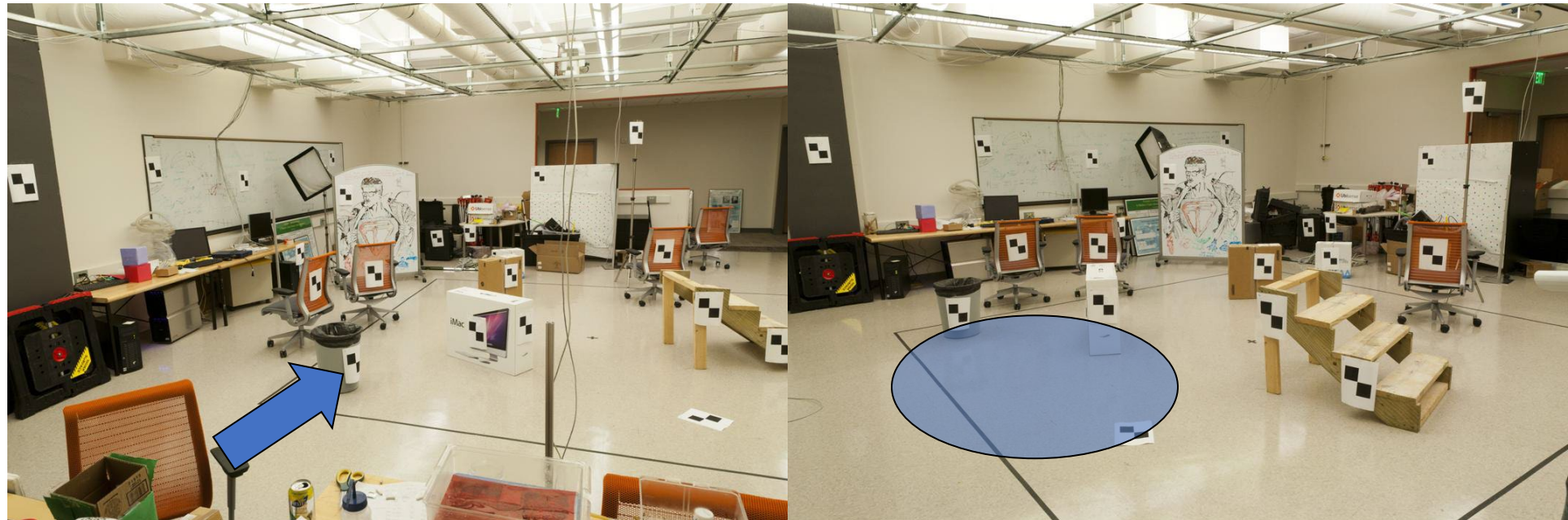




Correspondence for every pixel.  
Where do we need to search?



Correspondence for every pixel.  
Where do we need to search?

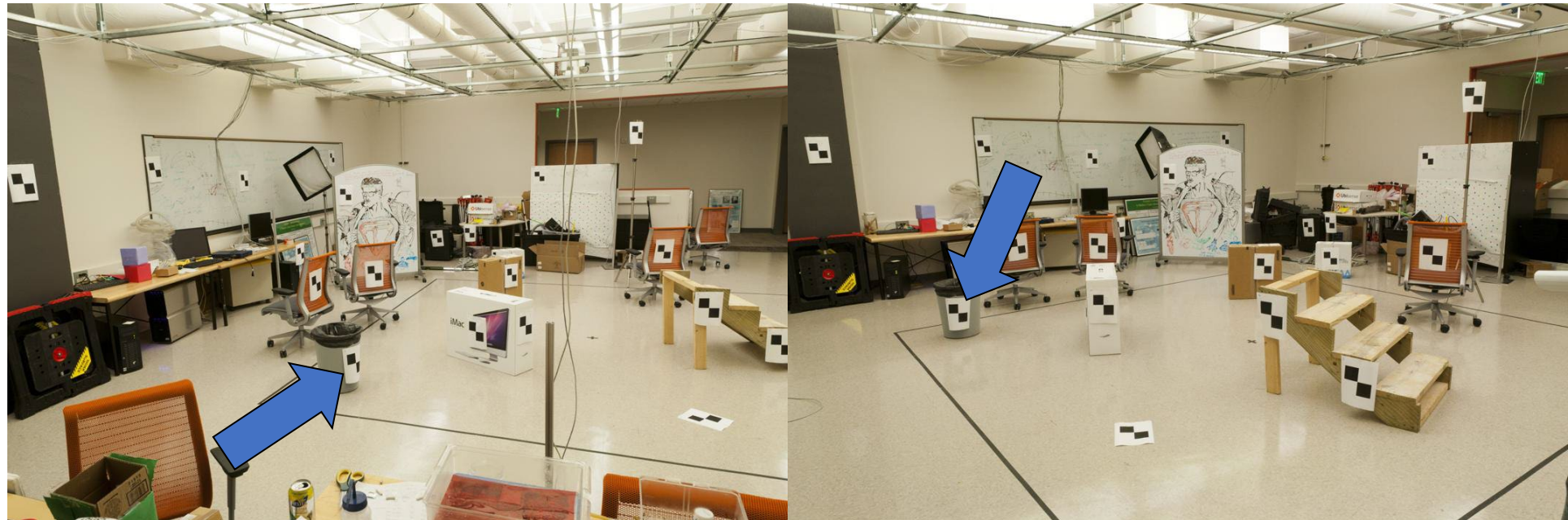




Correspondence for every pixel.  
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Correspondence for every pixel.  
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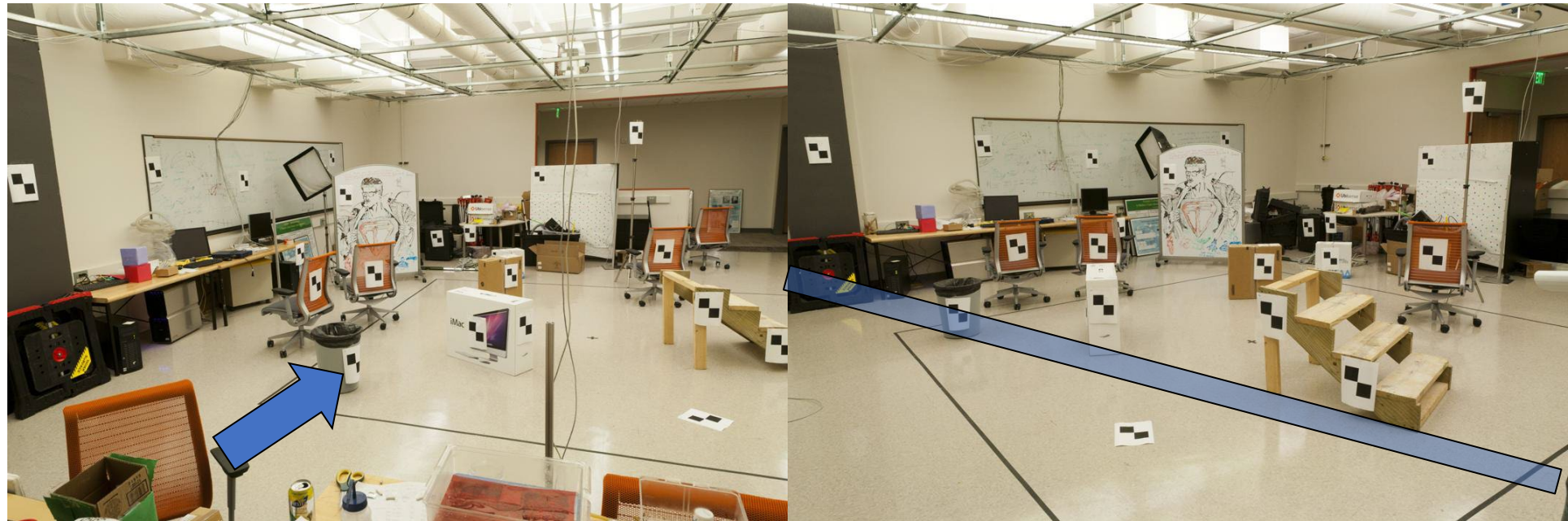




Correspondence for every pixel.  
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Correspondence for every pixel.  
Where do we need to search?

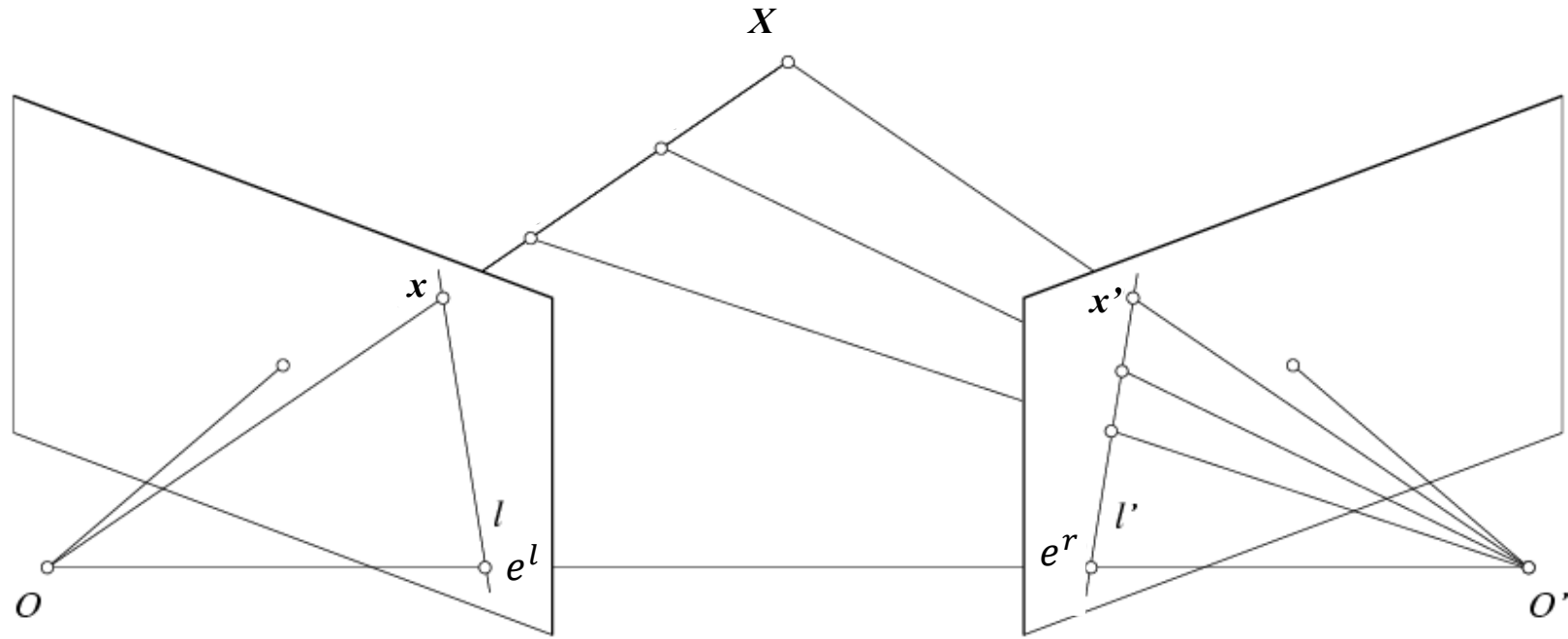


Wouldn't it be nice to know  
where matches can live?

*Epipolar geometry*

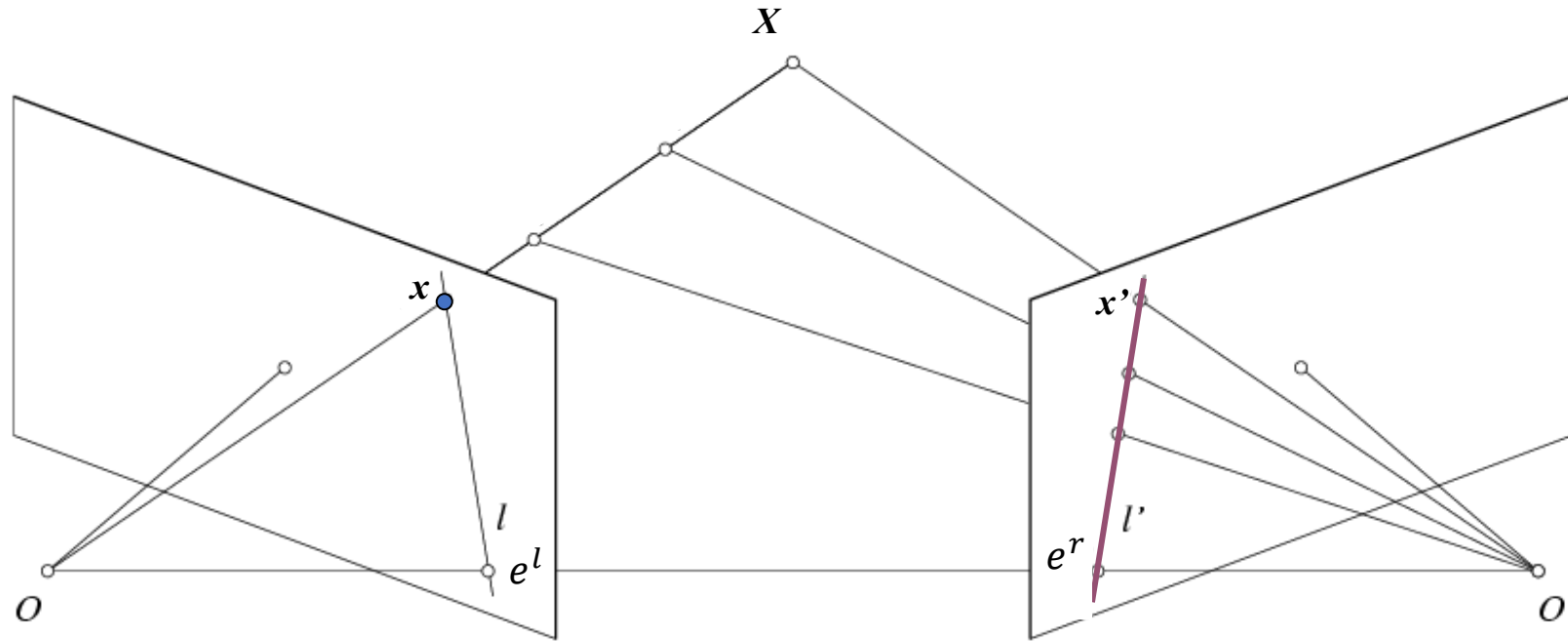
Constrains 2D search to 1D

# Key idea: Epipolar constraint



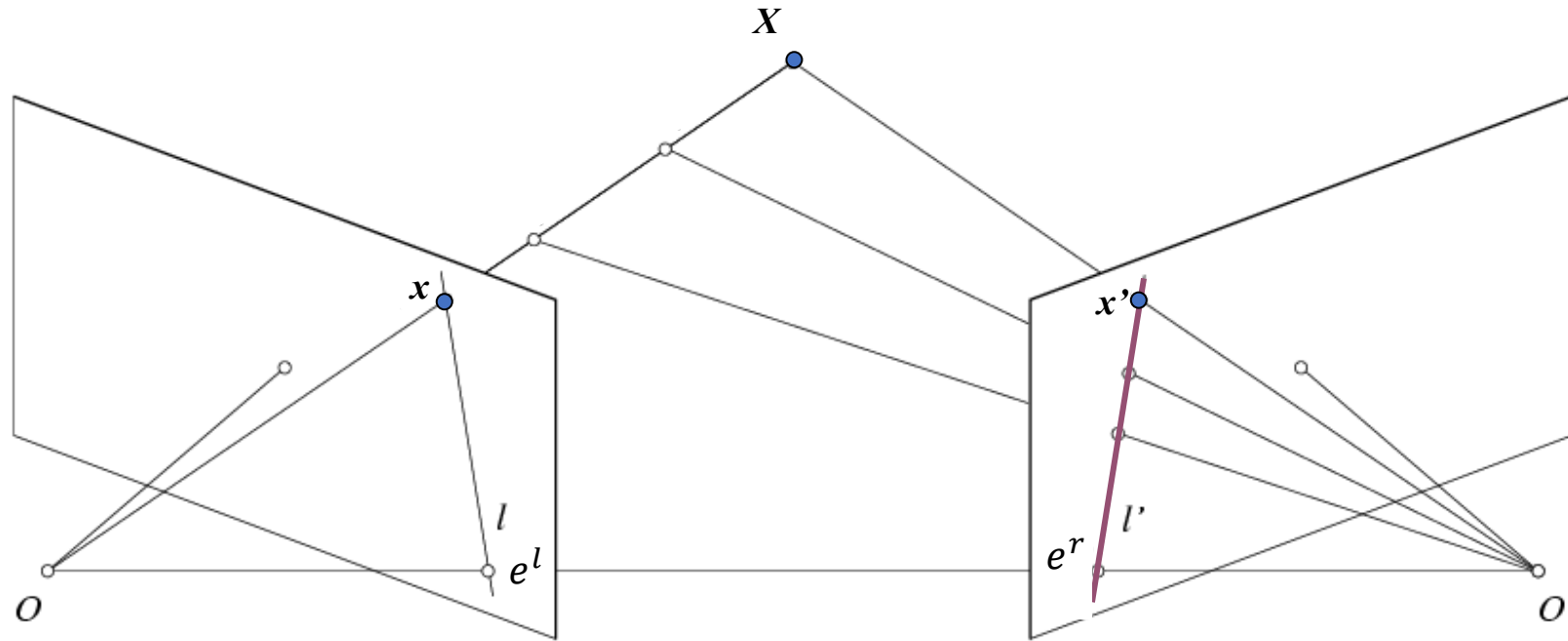


# Key idea: Epipolar constraint



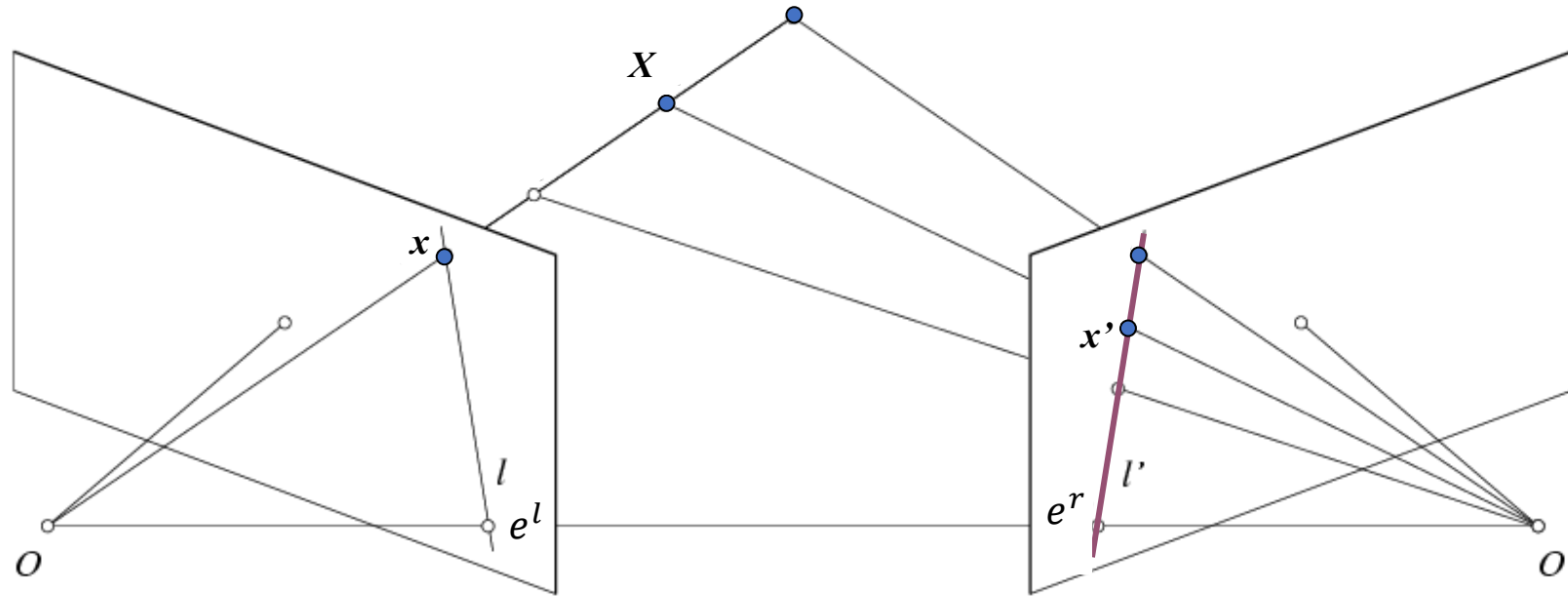
Potential matches for  $x$  have to lie on the corresponding line  $l'$ .

# Key idea: Epipolar constraint



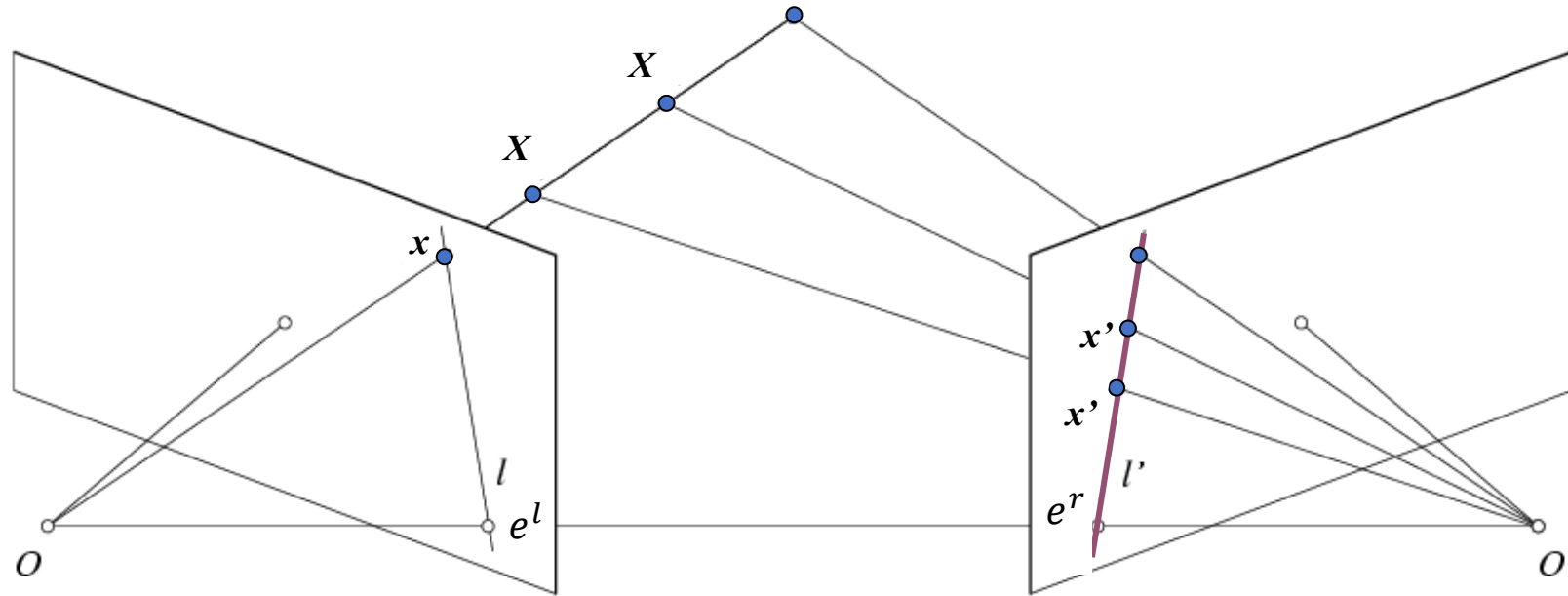
Potential matches for  $x$  have to lie on the corresponding line  $l'$ .

# Key idea: Epipolar constraint



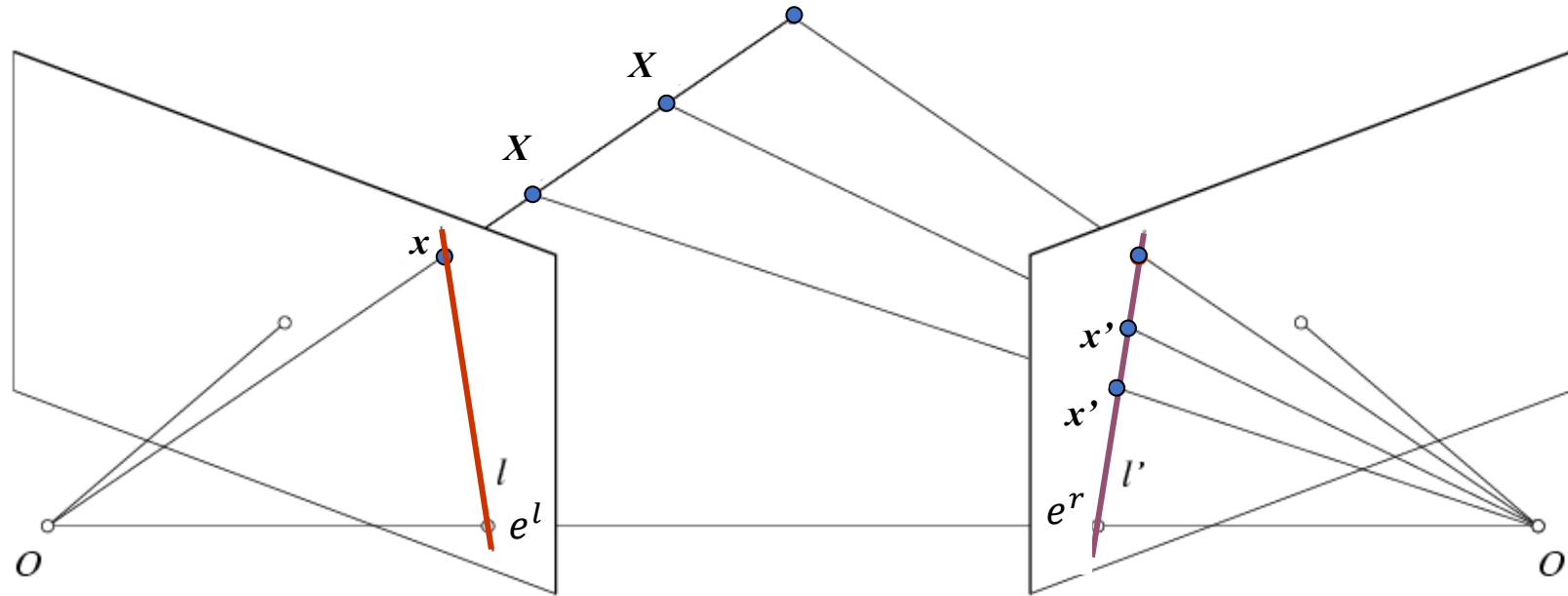
Potential matches for  $x$  have to lie on the corresponding line  $l'$ .

# Key idea: Epipolar constraint



Potential matches for  $x$  have to lie on the corresponding line  $l'$ .

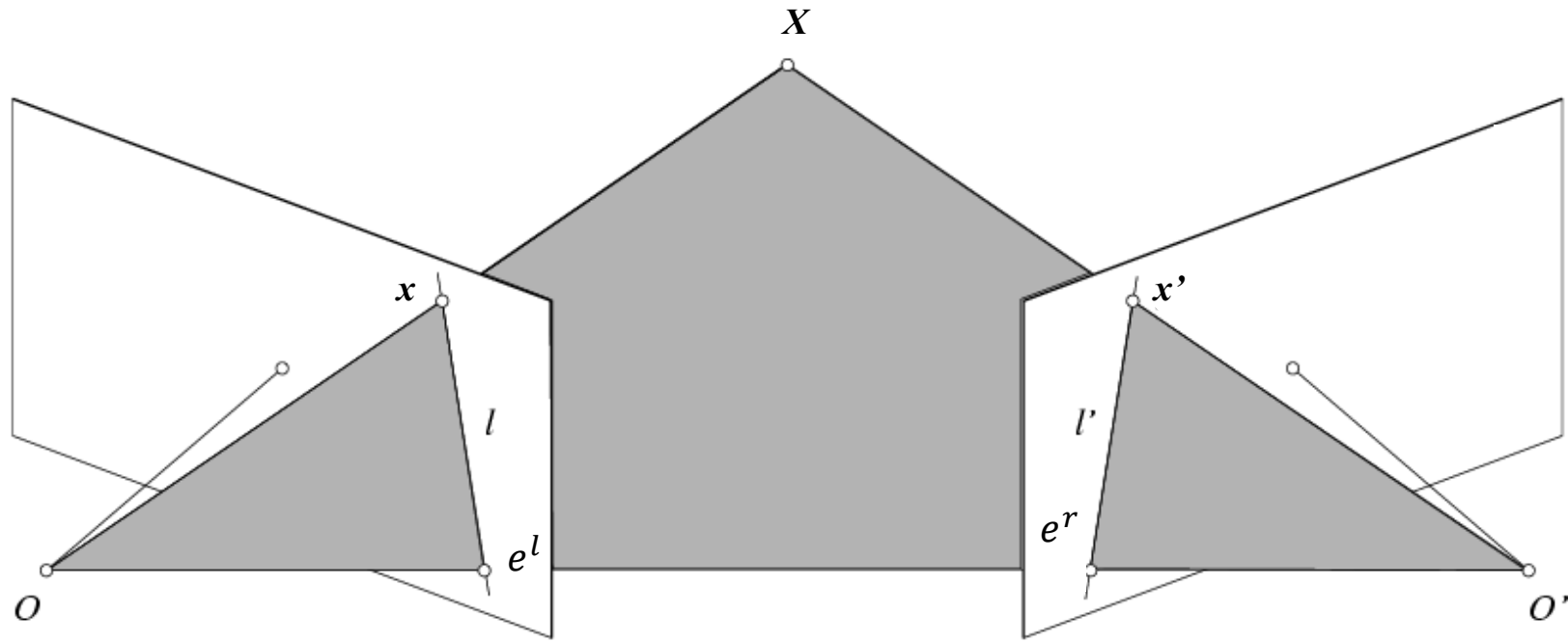
# Key idea: Epipolar constraint



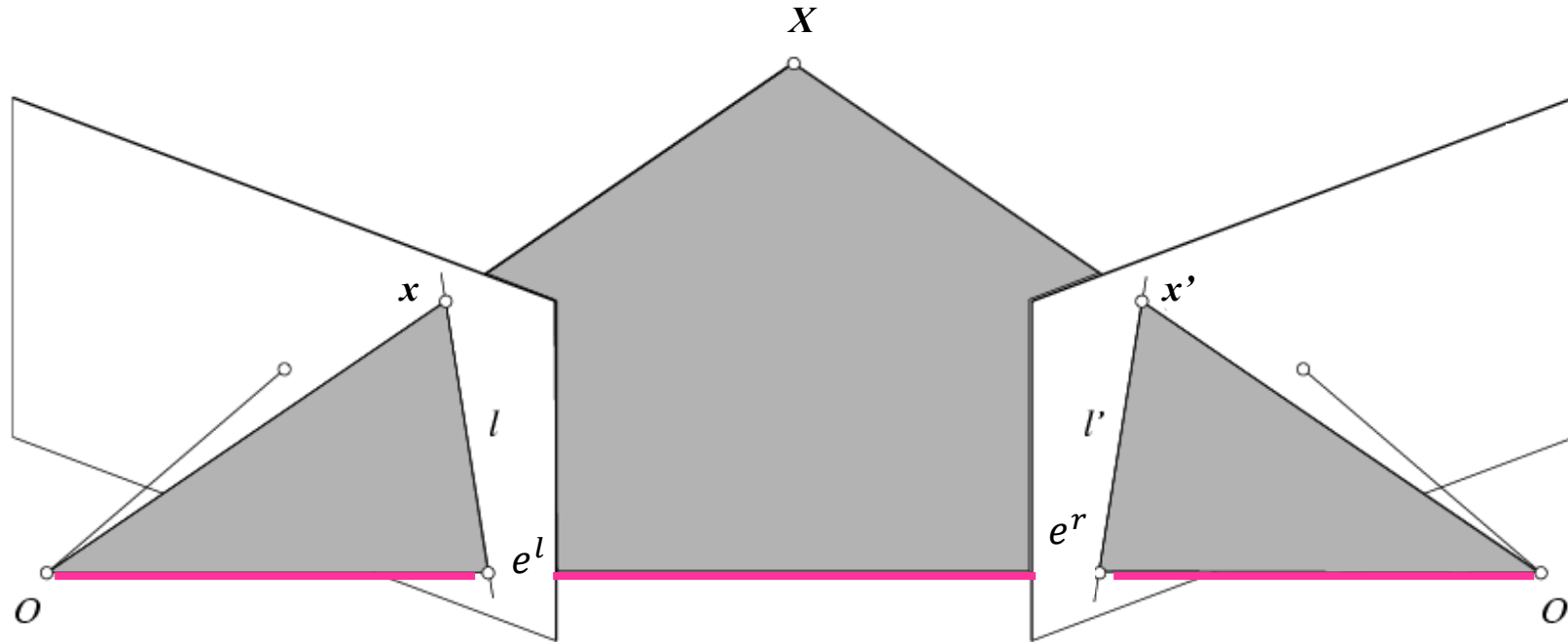
Potential matches for  $x'$  have to lie on the corresponding line  $l$ .

Potential matches for  $x$  have to lie on the corresponding line  $l'$ .

# Epipolar geometry: notation

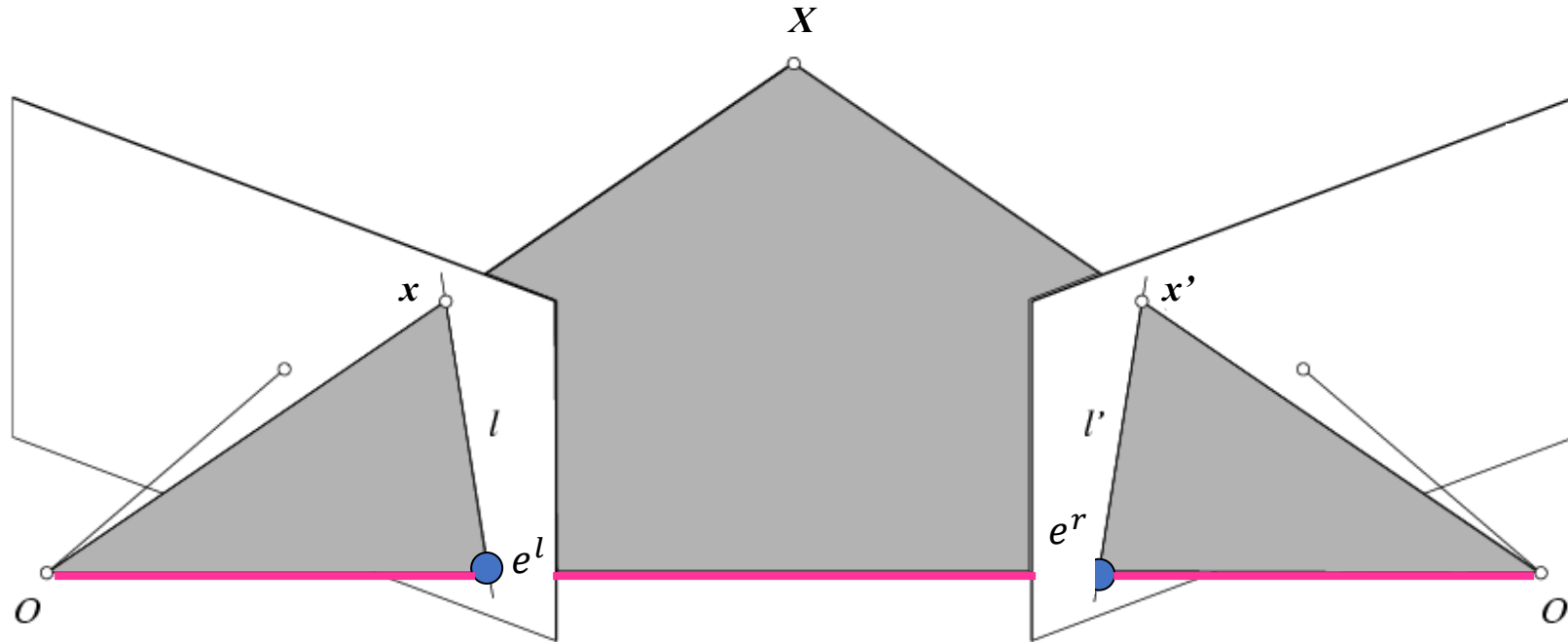


# Epipolar geometry: notation



- **Baseline** – line connecting the two camera centers

# Epipolar geometry: notation



- **Baseline** – line connecting the two camera centers

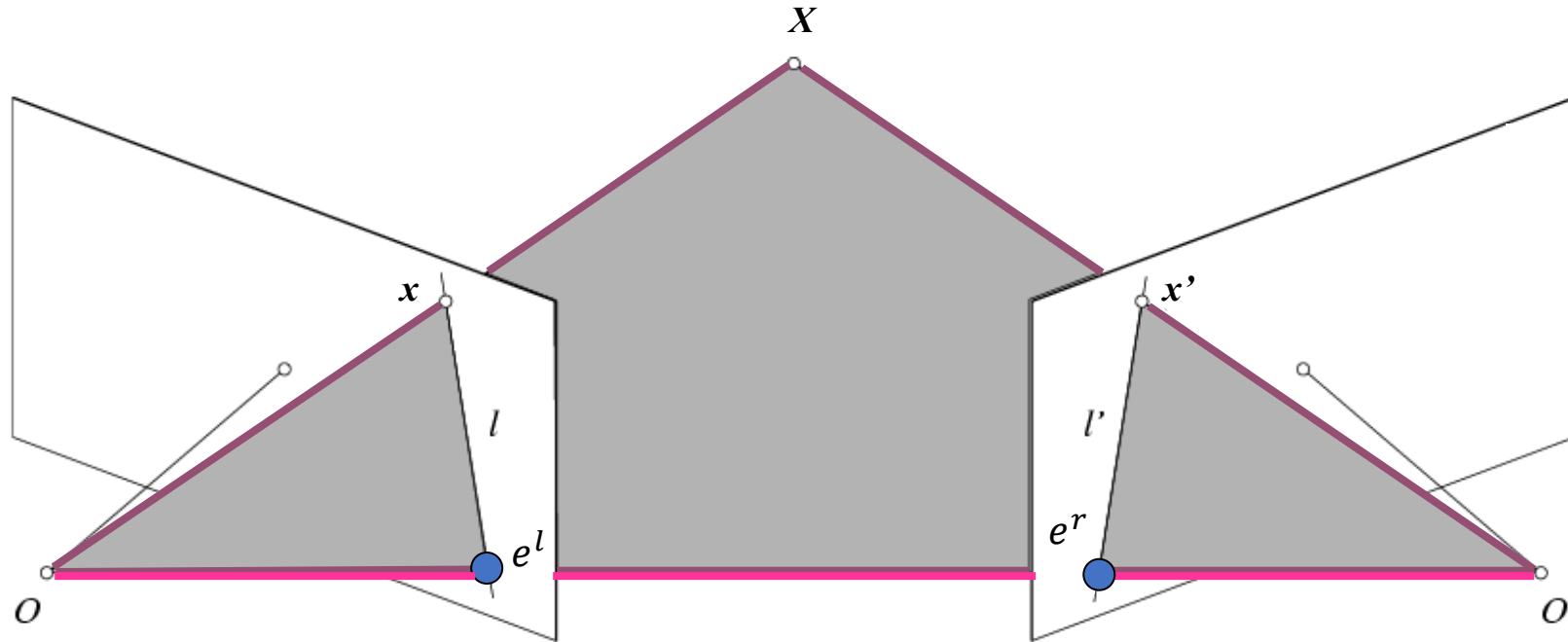
- **Epipoles**

- = intersections of baseline with image planes

- = projections of the other camera center

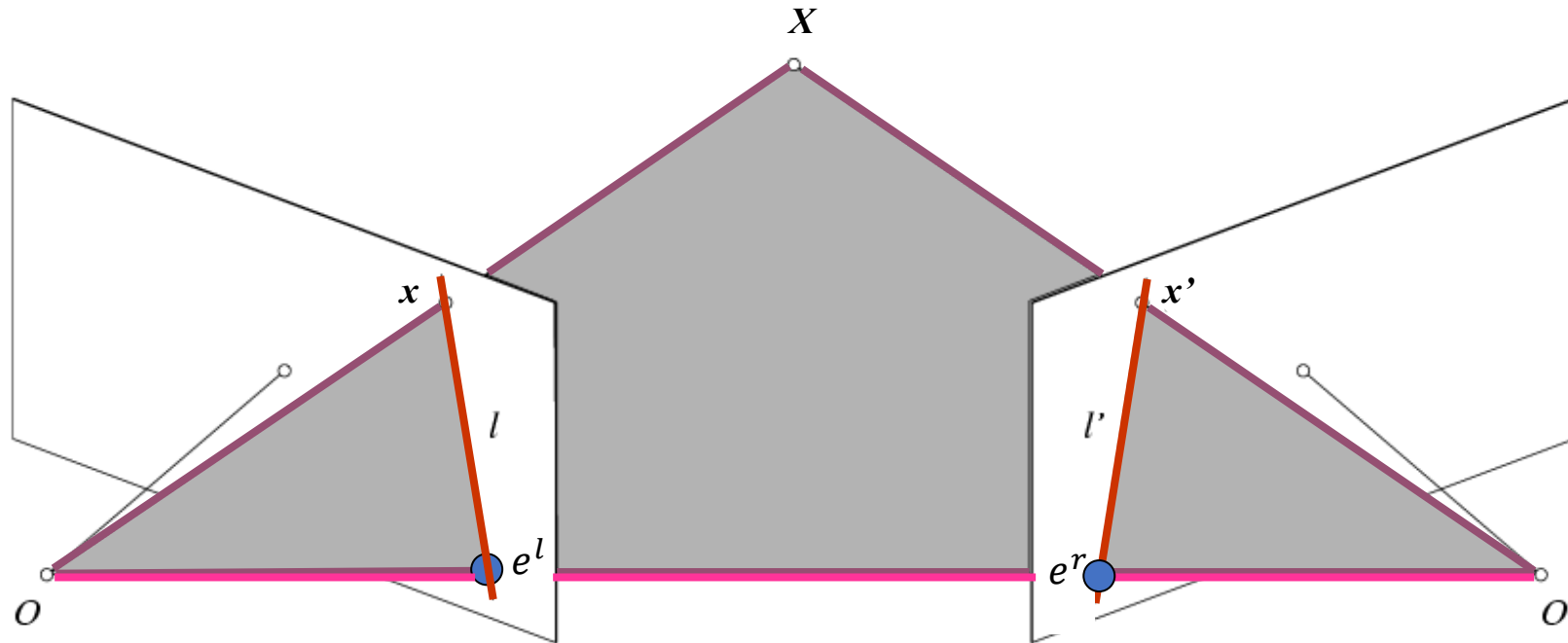


# Epipolar geometry: notation



- **Baseline** – line connecting the two camera centers
- **Epipoles**  
= intersections of baseline with image planes  
= projections of the other camera center
- **Epipolar Plane** – plane containing baseline (1D family)

# Epipolar geometry: notation



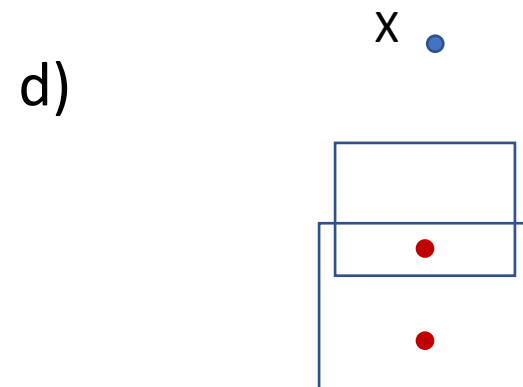
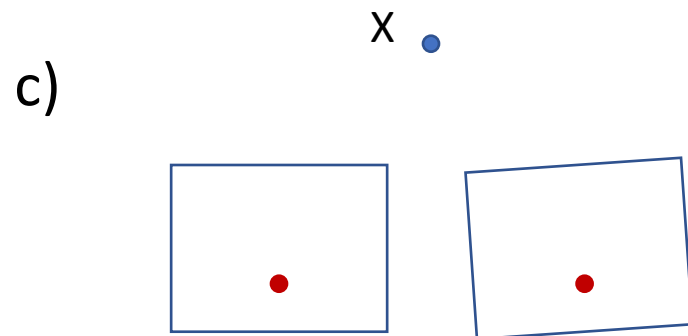
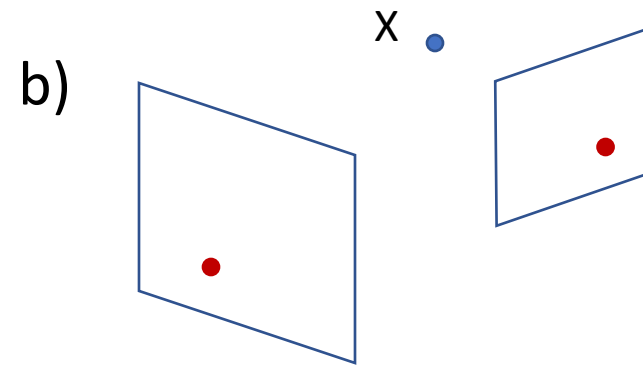
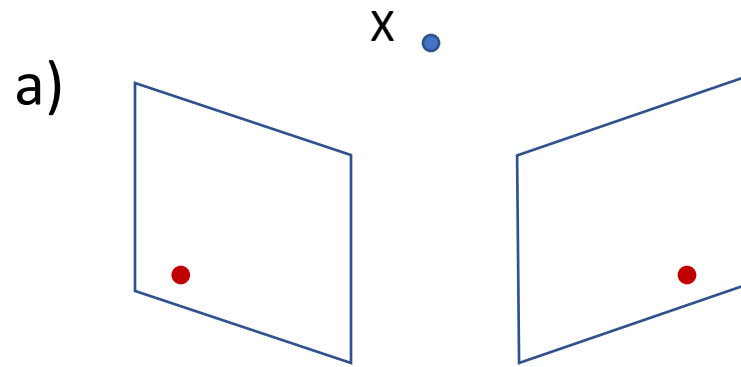
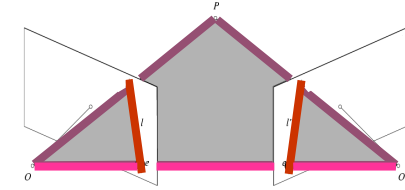
- **Baseline** – line connecting the two camera centers
- **Epipoles**  
= intersections of baseline with image planes  
= projections of the other camera center
- **Epipolar Plane** – plane containing baseline (1D family)
- **Epipolar Lines** - intersections of epipolar plane with image planes (always come in corresponding pairs)

# Think Pair Share

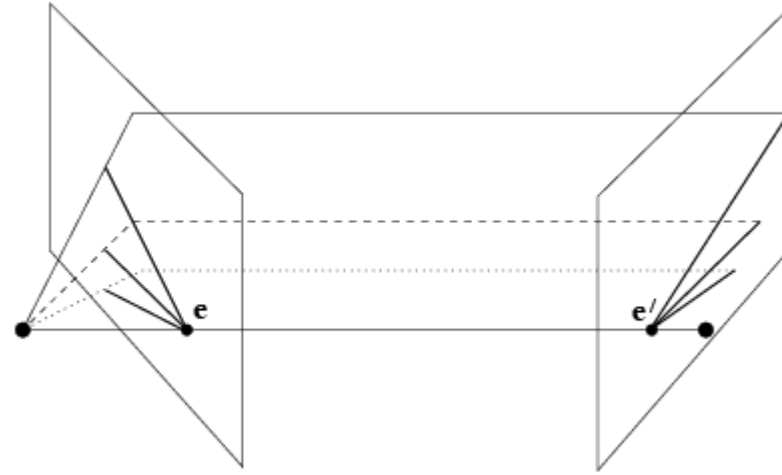
Where are the epipoles?

What do the epipolar lines look like?

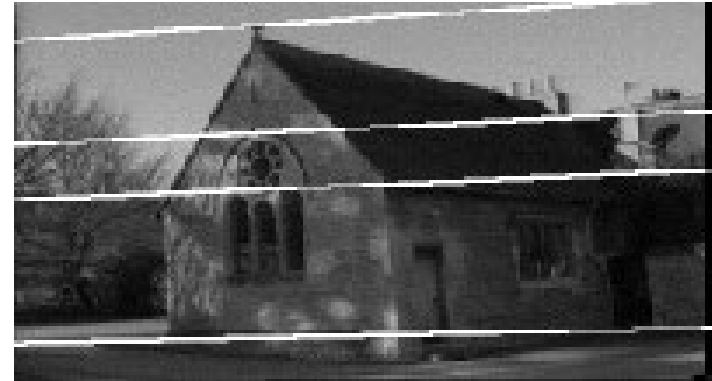
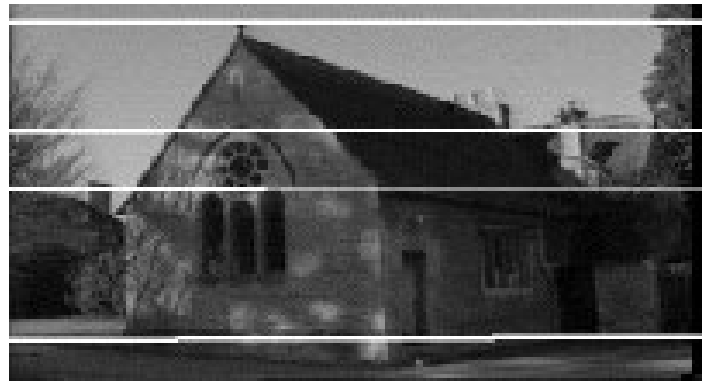
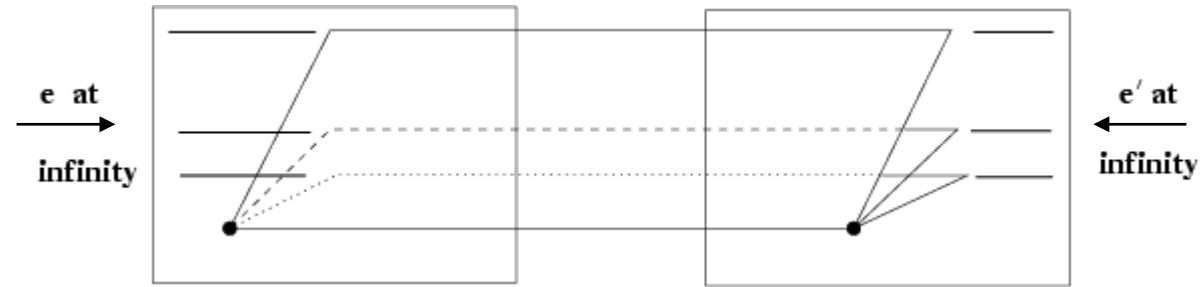
● = camera center



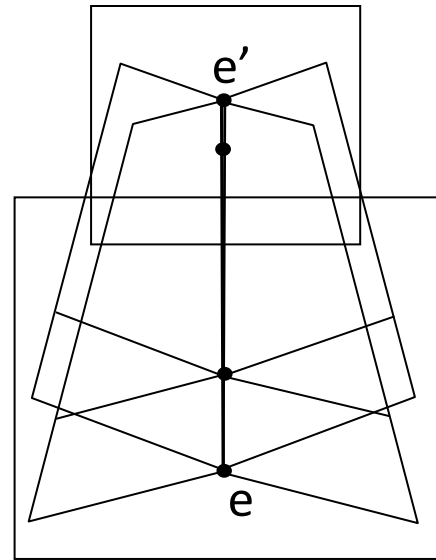
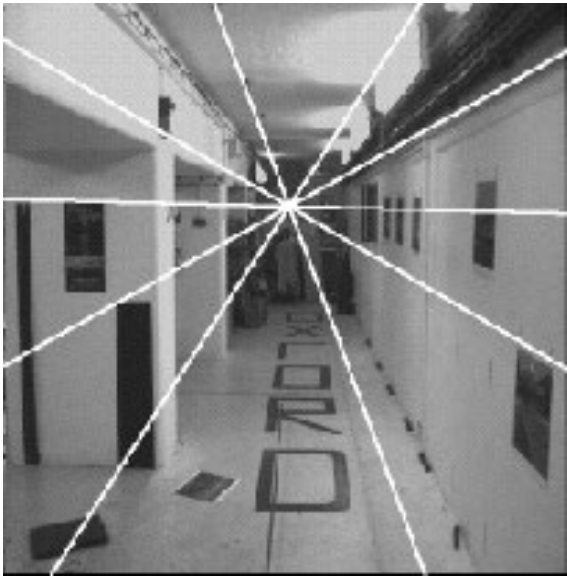
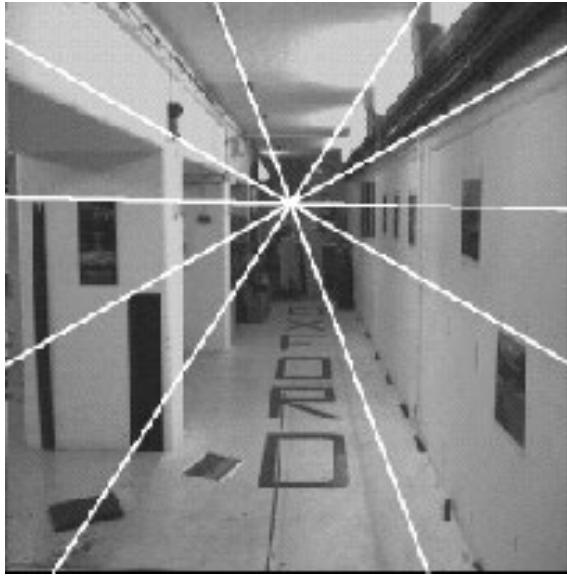
# Example: Converging cameras



# Example: Motion parallel to image plane



# Example: Forward motion

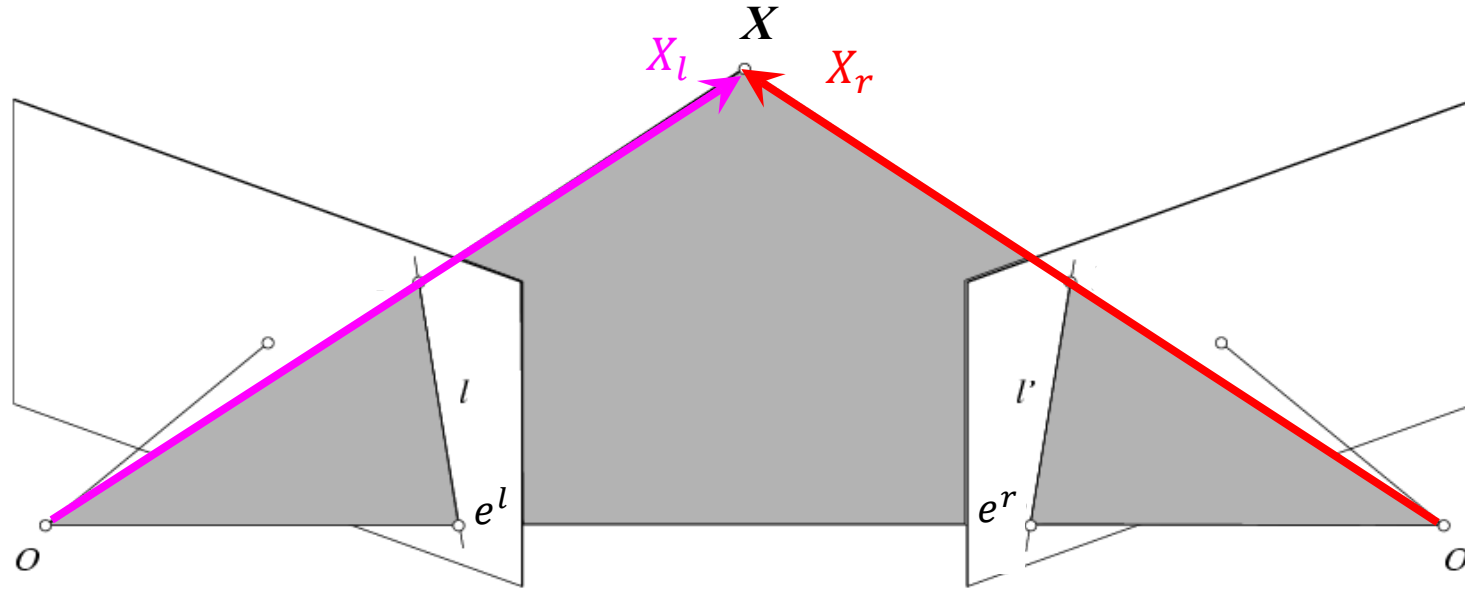


Epipole has same coordinates in both images.  
Points move along lines radiating from e:  
“Focus of expansion”

# How to find epipolar line of a point?

- A little bit more math, but cool math
- Essential matrix and fundamental matrix

# Essential matrix



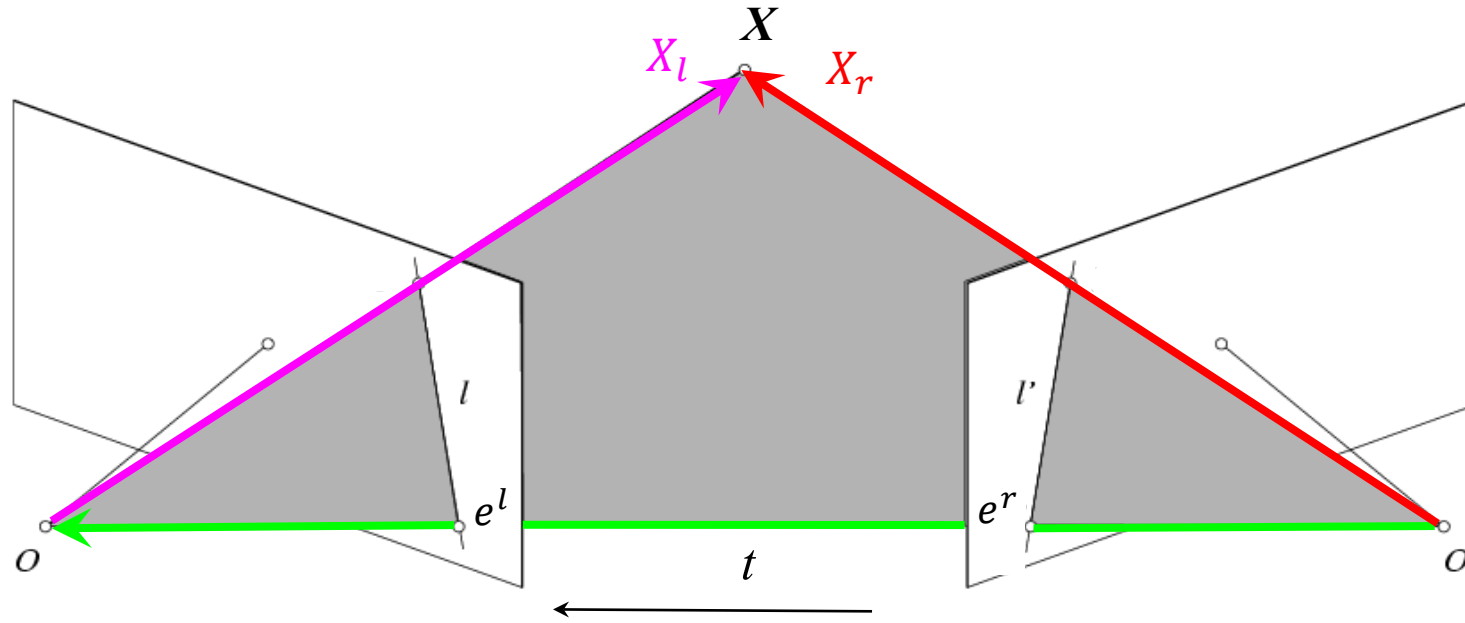
$X_l$ :  $X$  in terms of Cartesian coordinate of left camera

$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

There exists  $E$  such that  $X_l^T E X_r = 0$



# Essential matrix



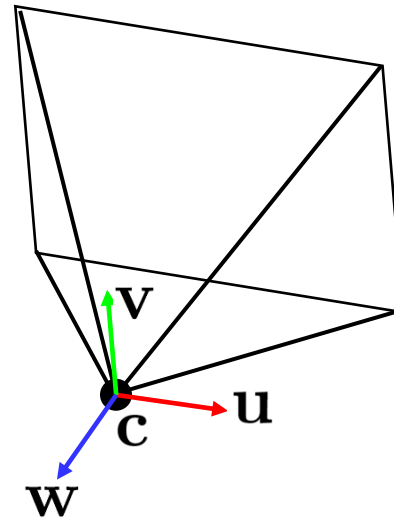
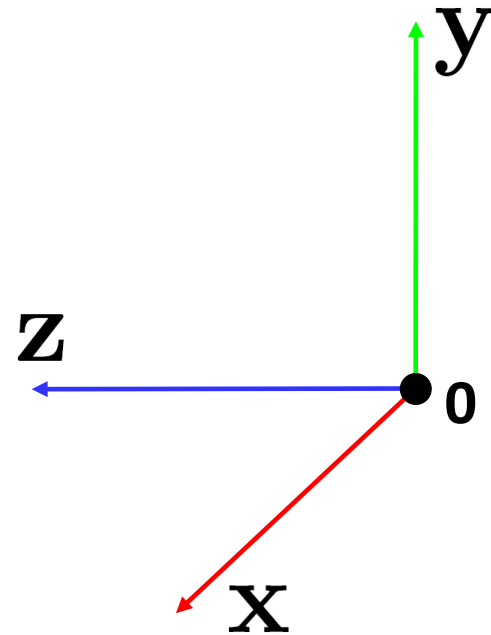
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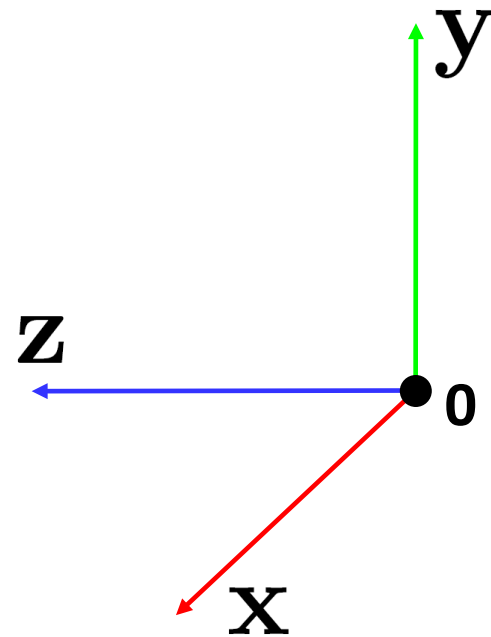
# Review: change of coordinate

- How do we change coordinate from one camera to another?
  - (Center of projection at the origin, x-axis points right, y-axis points up, z-axis points backwards)

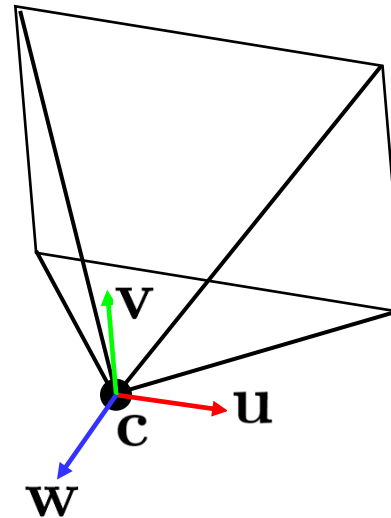


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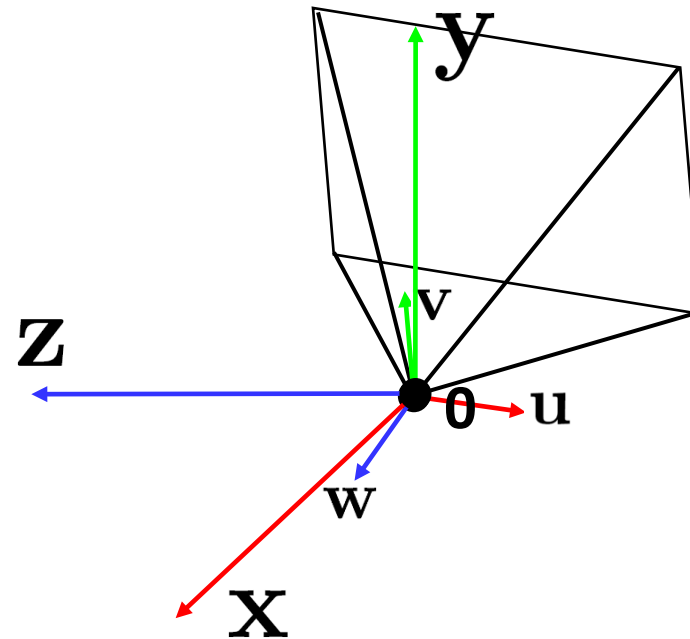


Step 1: Translate by  $-c$



# Review: change of coordinate

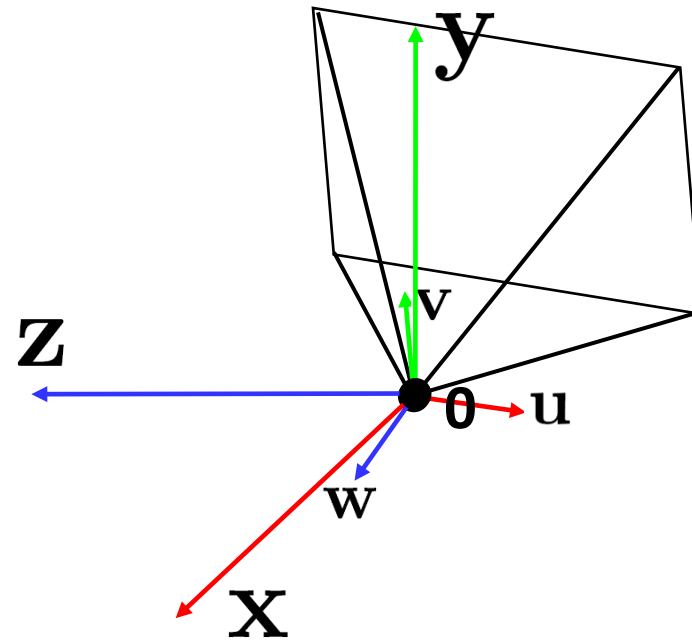
- How do we change coordinate from one camera to another?
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- How do we change coordinate from one camera to another?
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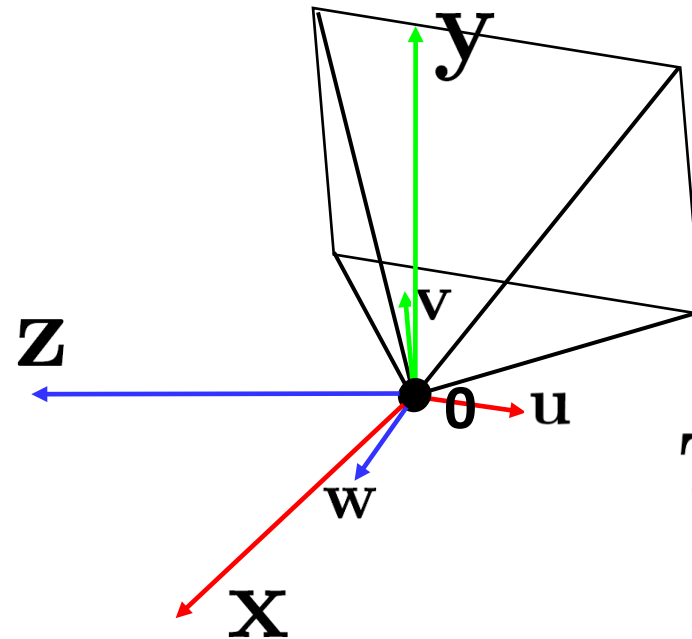


Step 1: Translate by  $-c$

How do we represent translation as a matrix multiplication?

# Review: change of coordinate

- How do we change coordinate from one camera to another?
  - (Center of projection at the origin, x-axis points right, y-axis points up, z-axis points backwards)



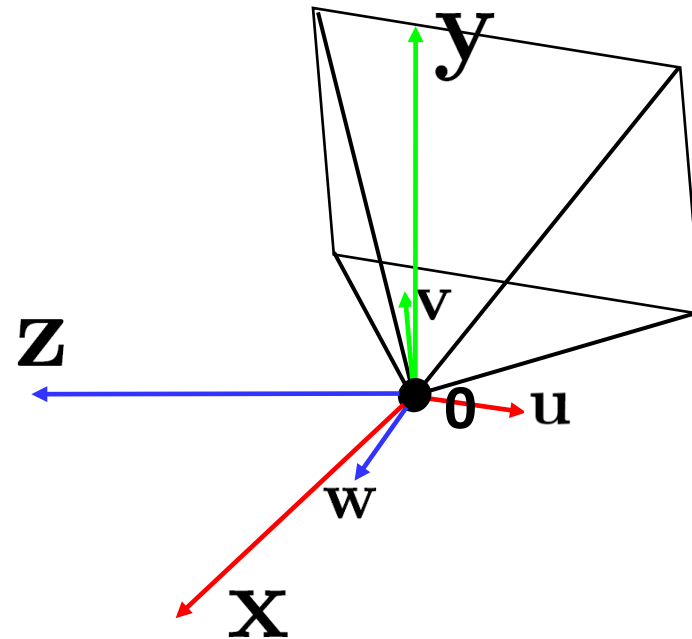
Step 1: Translate by  $-\mathbf{c}$

How do we represent translation as a matrix multiplication?

$$\mathbf{T} = \begin{bmatrix} \mathbf{I}_{3 \times 3} & -\mathbf{c} \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

# Review: change of coordinate

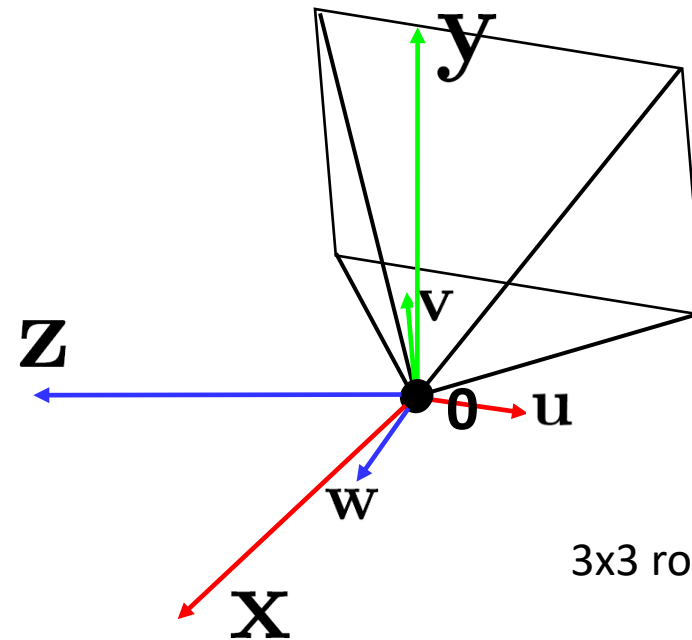
- How do we change coordinate from one camera to another?
  - (Center of projection at the origin, x-axis points right, y-axis points up, z-axis points backwards)



$$\mathbf{R} = \begin{bmatrix} \mathbf{u}^T \\ \mathbf{v}^T \\ \mathbf{w}^T \\ 1 \end{bmatrix}$$

# Review: change of coordinate

- How do we change coordinate from one camera to another?
  - (Center of projection at the origin, x-axis points right, y-axis points up, z-axis points backwards)



Step 1: Translate by  $-c$   
Step 2: Rotate by  $\mathbf{R}$

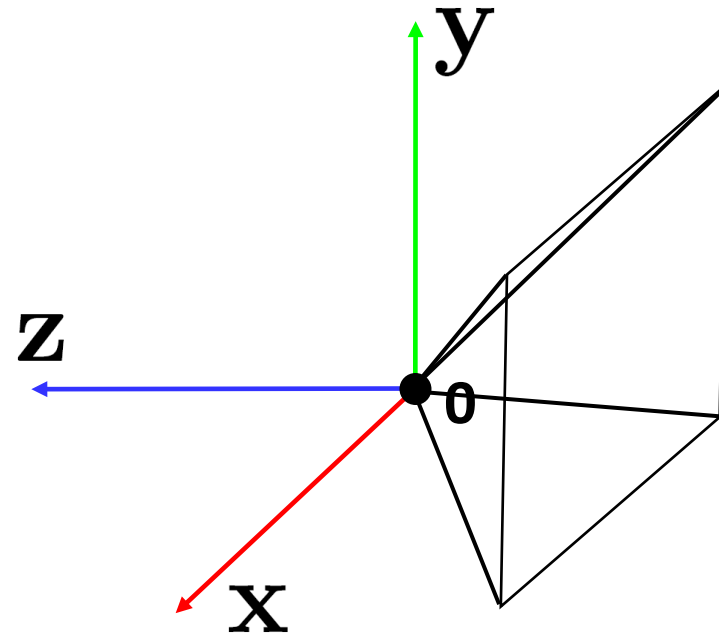
3x3 rotation matrix

$$\mathbf{R} = \begin{bmatrix} \mathbf{u}^T \\ \mathbf{v}^T \\ \mathbf{w}^T \\ 1 \end{bmatrix}$$



# Review: change of coordinate

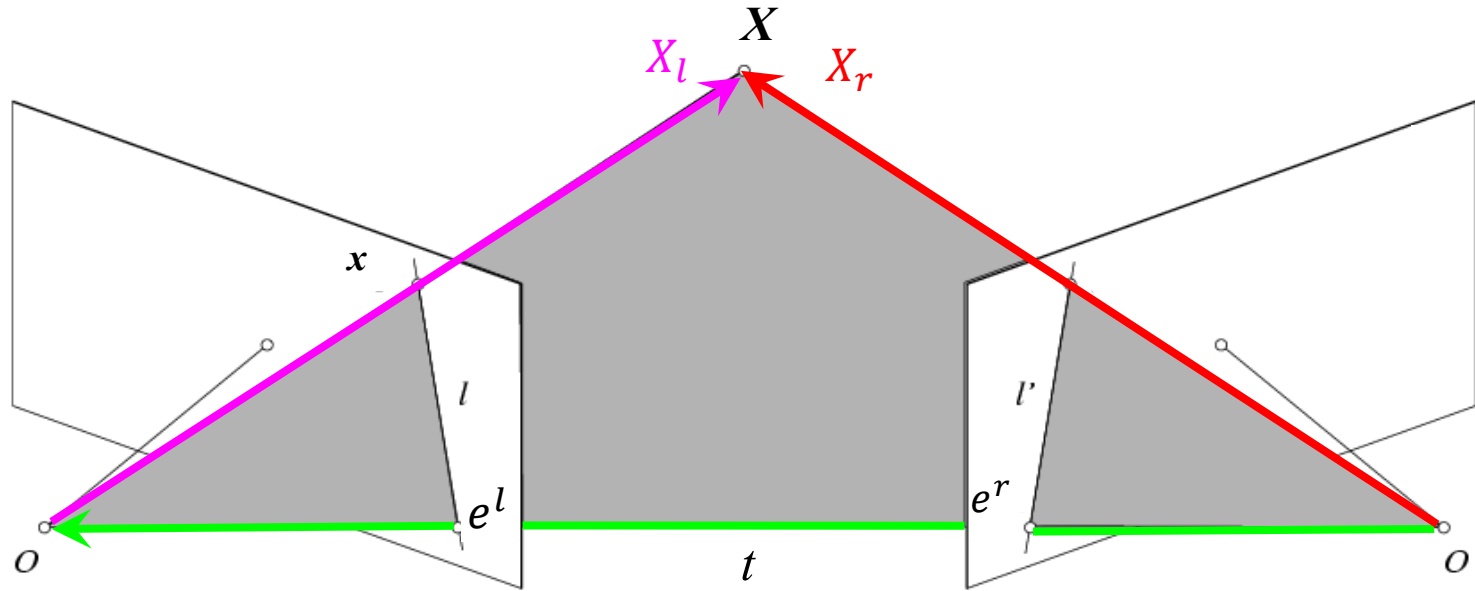
- How do we change coordinate from one camera to another?
  - (Center of projection at the origin, x-axis points right, y-axis points up, z-axis points backwards)



Step 1: Translate by  $-c$   
Step 2: Rotate by  $\mathbf{R}$

$$\mathbf{R} = \begin{bmatrix} \mathbf{u}^T \\ \mathbf{v}^T \\ \mathbf{w}^T \\ 1 \end{bmatrix}$$

# Essential matrix (Longuet-Higgins, 1981)



Let  $R$  be rotation from left to right camera, then  $X_l = R(X_r - t)$

# Cross product in matrix representation

- Let  $\mathbf{u} = [u_1, u_2, u_3]^T$ ,  $\mathbf{v} = [v_1, v_2, v_3]^T$

$$\mathbf{u} \times \mathbf{v} = \begin{vmatrix} i & j & k \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix} = \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix} i - \begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix} j + \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} k$$

# Cross product in matrix representation

- Let  $\mathbf{u} = [u_1, u_2, u_3]^T$ ,  $\mathbf{v} = [v_1, v_2, v_3]^T$

$$\begin{aligned}\mathbf{u} \times \mathbf{v} &= \begin{vmatrix} i & j & k \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix} = \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix} i - \begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix} j + \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} k \\ &= \left[ \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix}, -\begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix}, \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} \right]^T\end{aligned}$$

# Cross product in matrix representation

- Let  $\mathbf{u} = [u_1, u_2, u_3]^T$ ,  $\mathbf{v} = [v_1, v_2, v_3]^T$

$$\begin{aligned}\mathbf{u} \times \mathbf{v} &= \begin{vmatrix} i & j & k \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix} = \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix} i - \begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix} j + \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} k \\ &= \left[ \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix}, -\begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix}, \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} \right]^T \\ &= \begin{bmatrix} u_2 v_3 - u_3 v_2 \\ -u_1 v_3 + u_3 v_1 \\ u_1 v_2 - u_2 v_1 \end{bmatrix}\end{aligned}$$

# Cross product in matrix representation

- Let  $\mathbf{u} = [u_1, u_2, u_3]^T$ ,  $\mathbf{v} = [v_1, v_2, v_3]^T$

$$\begin{aligned}\mathbf{u} \times \mathbf{v} &= \begin{vmatrix} i & j & k \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix} = \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix} i - \begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix} j + \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} k \\ &= \left[ \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix}, -\begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix}, \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} \right]^T \\ &= \begin{bmatrix} u_2 v_3 - u_3 v_2 \\ -u_1 v_3 + u_3 v_1 \\ u_1 v_2 - u_2 v_1 \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & -u_3 & u_2 \\ u_3 & 0 & -u_1 \\ -u_2 & u_1 & 0 \end{bmatrix}}_{[\mathbf{u}]_{\times}} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}\end{aligned}$$



# Cross product in matrix representation

- Let  $\mathbf{u} = [u_1, u_2, u_3]^T$ ,  $\mathbf{v} = [v_1, v_2, v_3]^T$

$$\begin{aligned}\mathbf{u} \times \mathbf{v} &= \begin{vmatrix} i & j & k \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix} = \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix} i - \begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix} j + \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} k \\ &= \left[ \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix}, -\begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix}, \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} \right]^T \\ &= \begin{bmatrix} u_2 v_3 - u_3 v_2 \\ -u_1 v_3 + u_3 v_1 \\ u_1 v_2 - u_2 v_1 \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & -u_3 & u_2 \\ u_3 & 0 & -u_1 \\ -u_2 & u_1 & 0 \end{bmatrix}}_{[\mathbf{u}]_{\times}} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} \\ &= [\mathbf{u}]_{\times} \mathbf{v}\end{aligned}$$

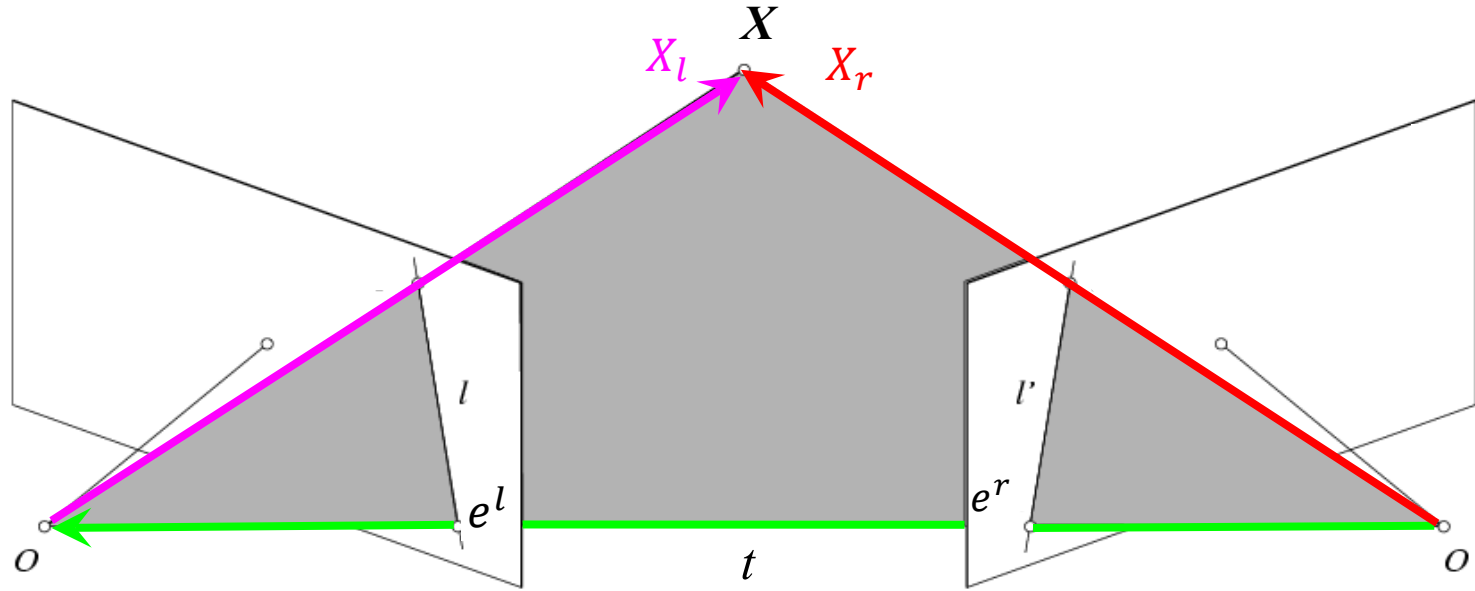
# Cross product in matrix representation

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N.B.  $\begin{bmatrix} 0 & -u_3 & u_2 \\ u_3 & 0 & -u_1 \\ -u_2 & u_1 & 0 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$ . Therefore,  $\text{rank}([\mathbf{u}]_{\times}) \leq 2$

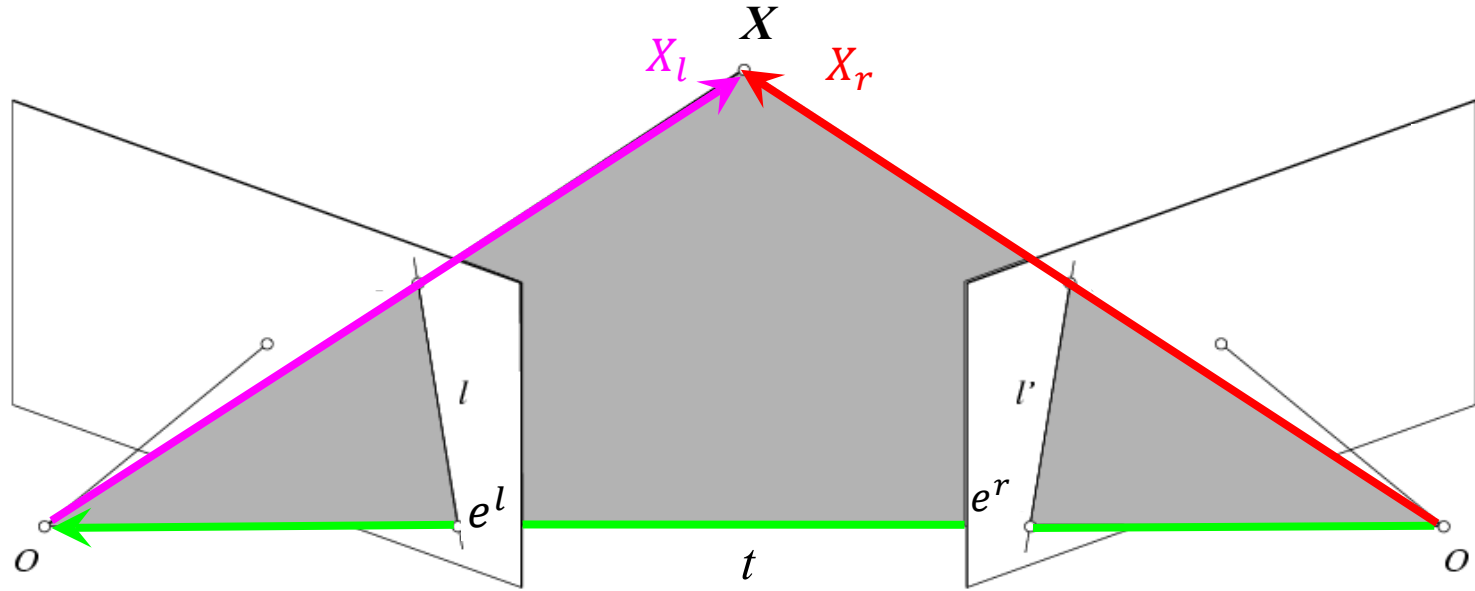
# Essential matrix (Longuet-Higgins, 1981)



Let  $R$  be rotation from right to left camera, then  $X_l = R(X_r - t)$

Let  $E$  be  $R[t]_{\times}$ ,  $X_l^T E X_r$

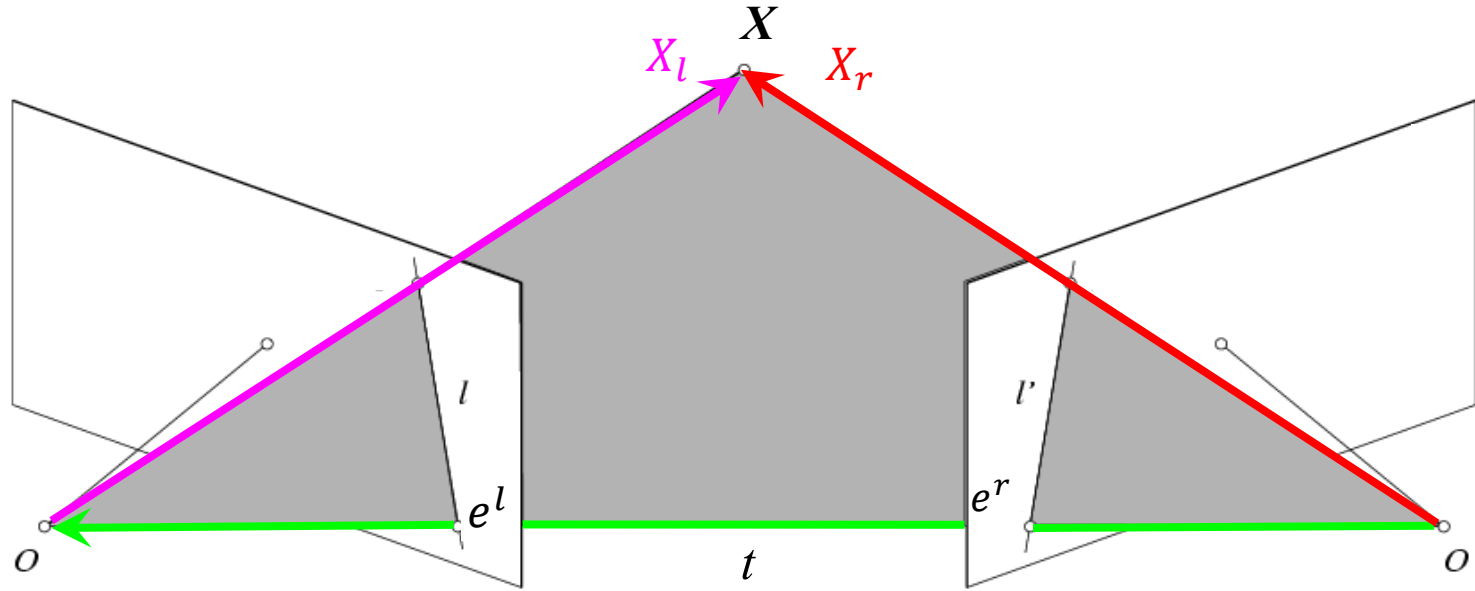
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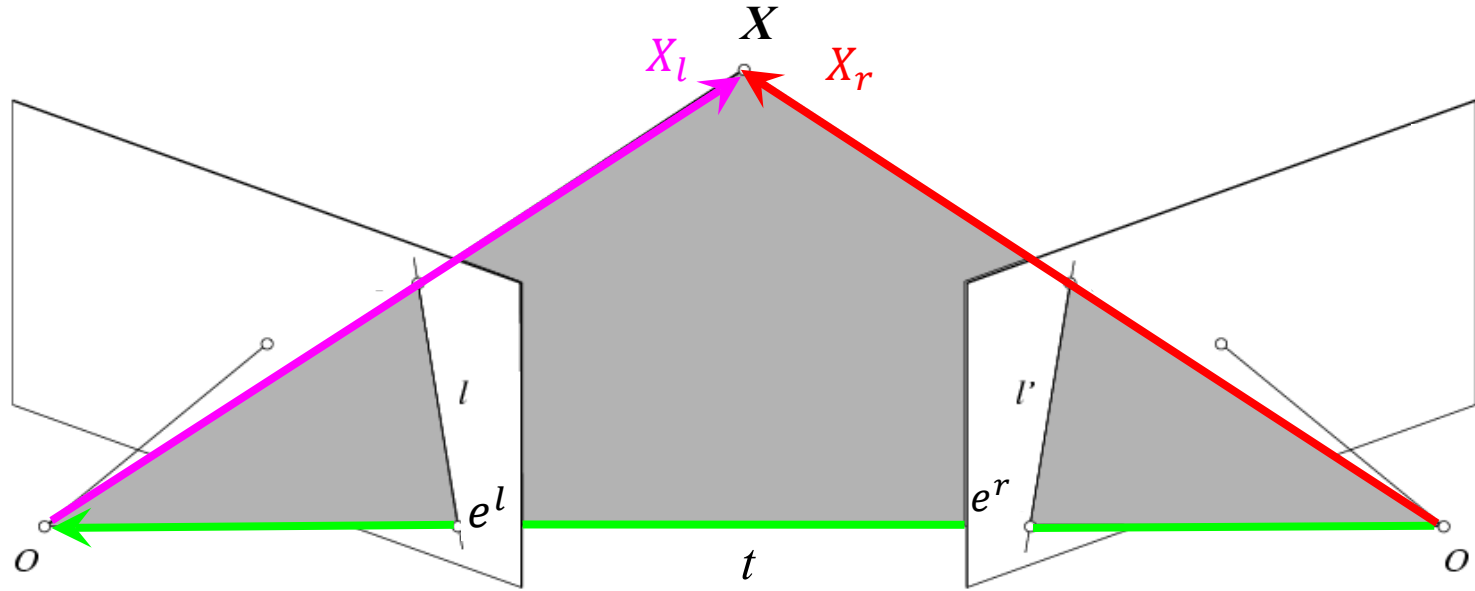
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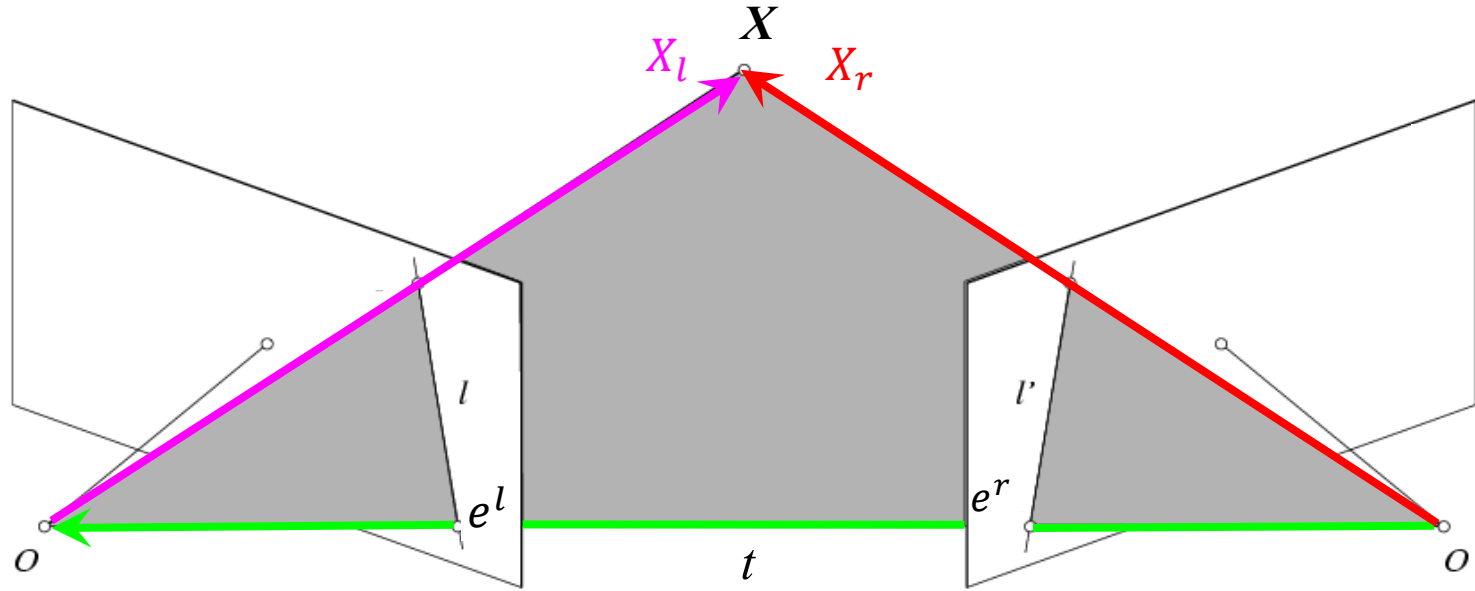


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$$\begin{aligned} \text{Let } E \text{ be } R[t]_{\times}, \quad X_l^T E X_r &= X_l^T (R[t]_{\times} X_r) = (X_r - t)^T R^T R[t]_{\times} X_r \\ &= (X_r - t) \cdot (t \times X_r) \end{aligned}$$



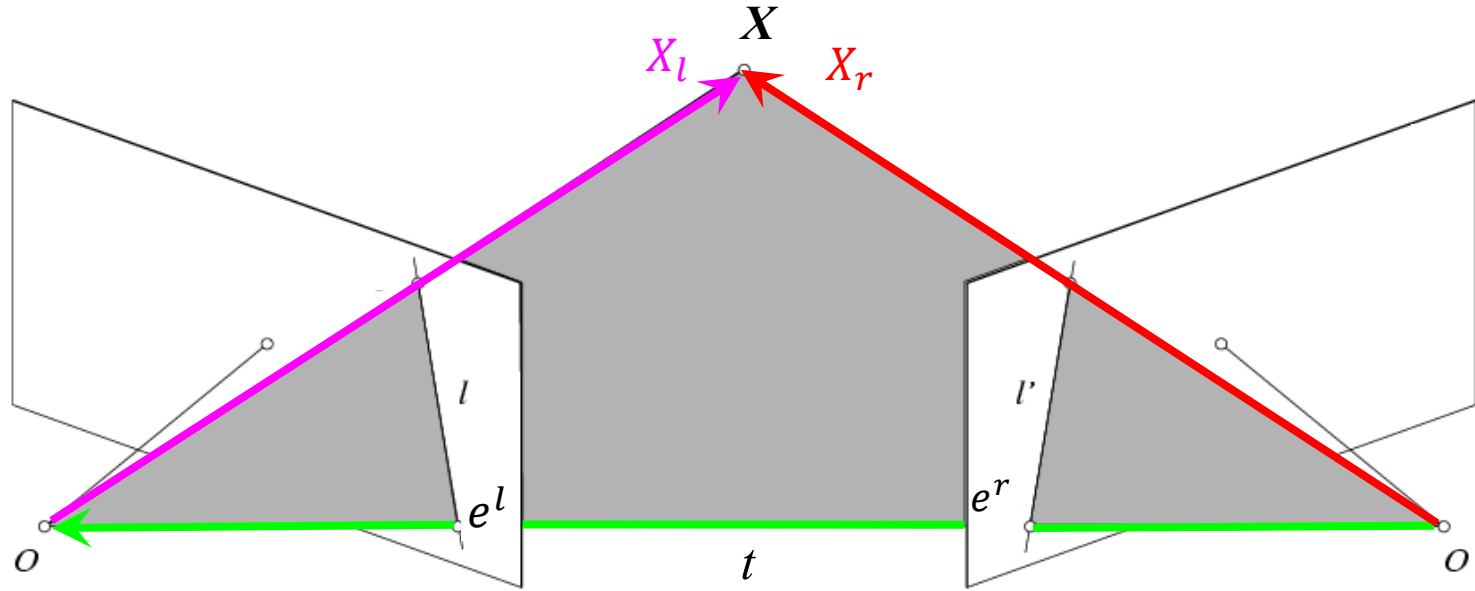
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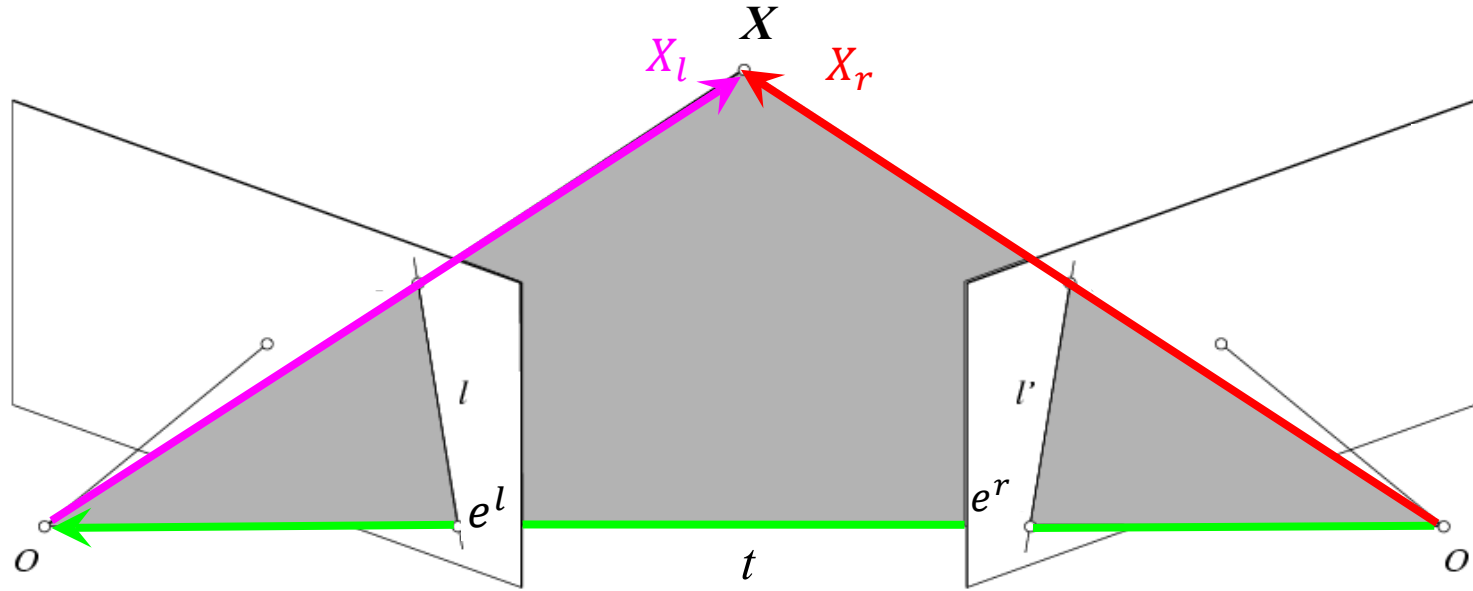
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# Essential matrix (Longuet-Higgins, 1981)

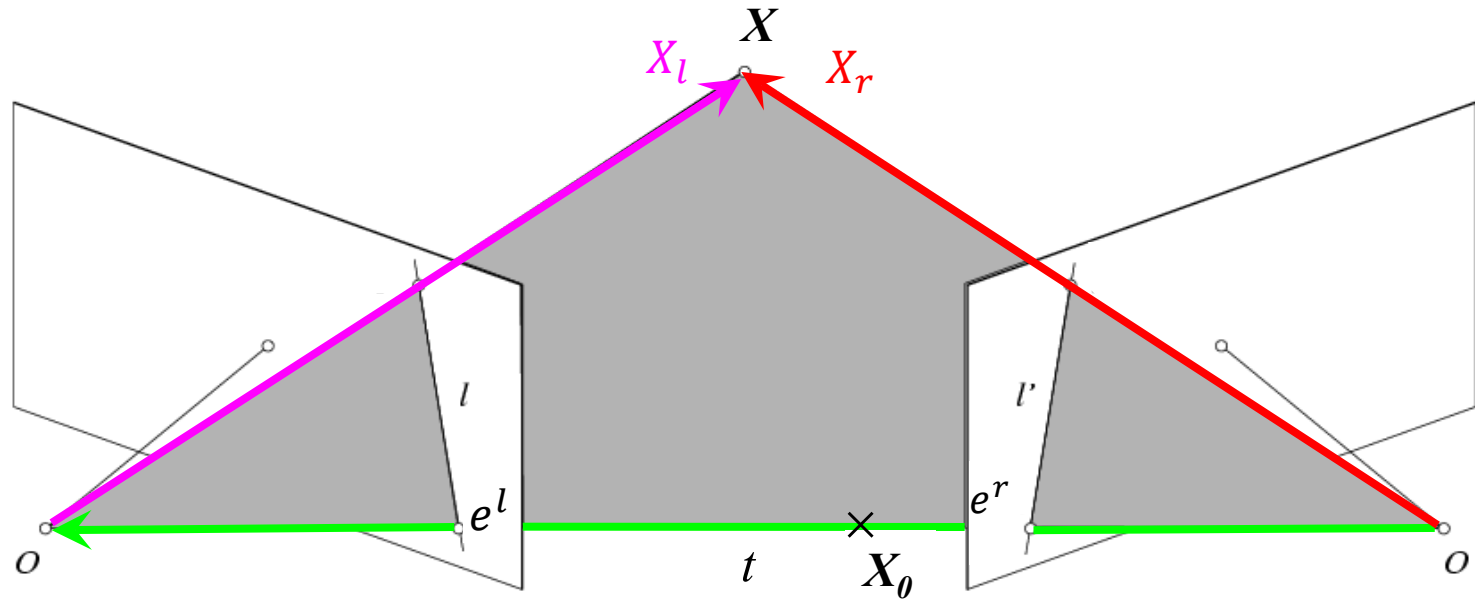


Let  $R$  be rotation from right to left camera, then  $X_l = R(X_r - t)$

$$\begin{aligned}
 \text{Let } E \text{ be } R[t]_{\times}, \quad X_l^T E X_r &= X_l^T (R[t]_{\times} X_r) = (X_r - t)^T R^T R[t]_{\times} X_r \\
 &= (X_r - t) \cdot (t \times X_r) \\
 &= X_r \cdot (t \times X_r) - t \cdot (t \times X_r) = 0
 \end{aligned}$$

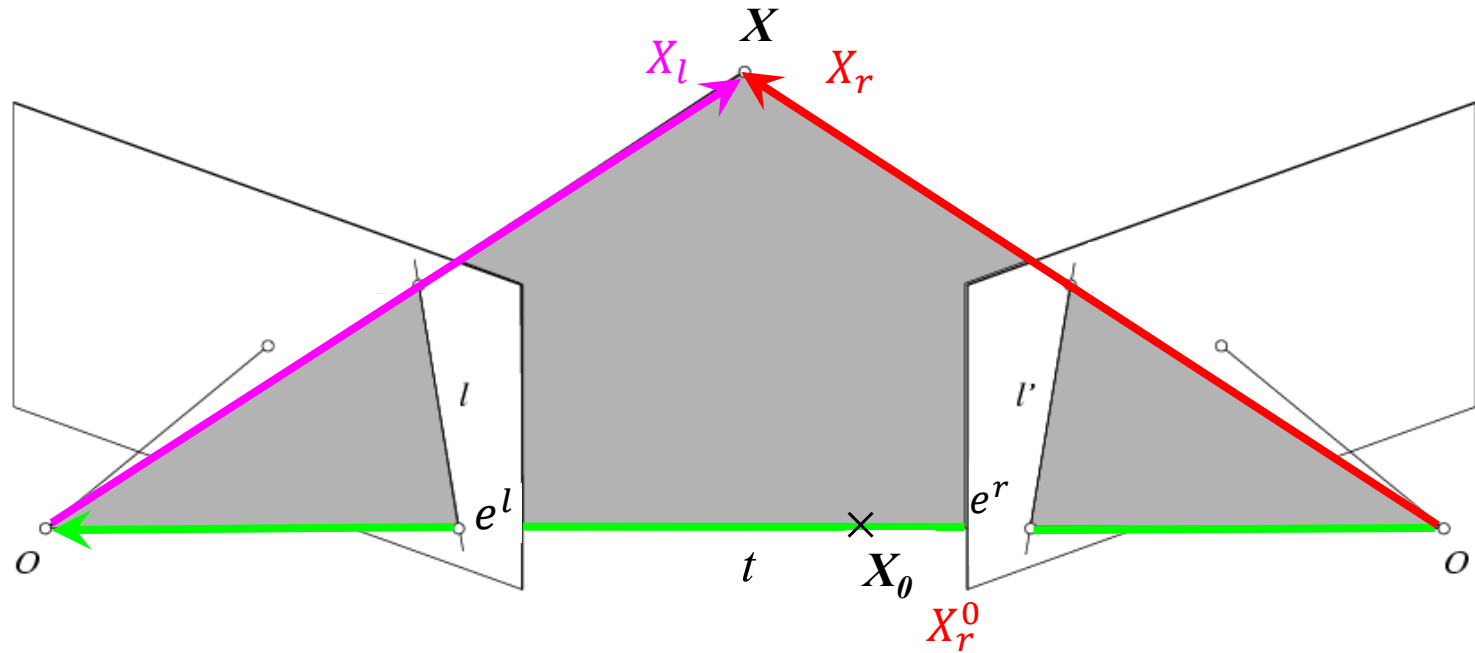
Note that  $\text{rank}(E) \leq \text{rank}([t]_{\times}) \leq 2$

# Essential matrix (Longuet-Higgins, 1981)



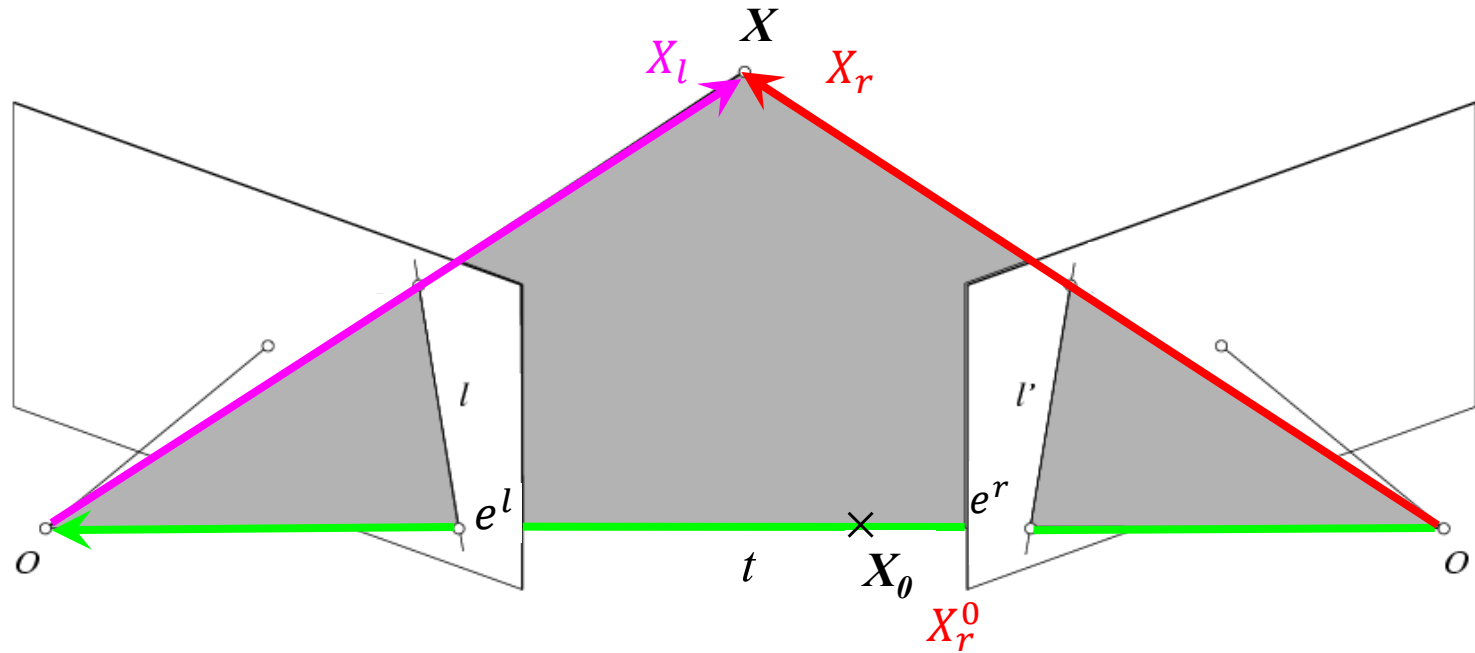
Consider any  $X_0$  lies between the two centers of projection

# Essential matrix (Longuet-Higgins, 1981)



Consider any  $X_0$  lies between the two centers of projection

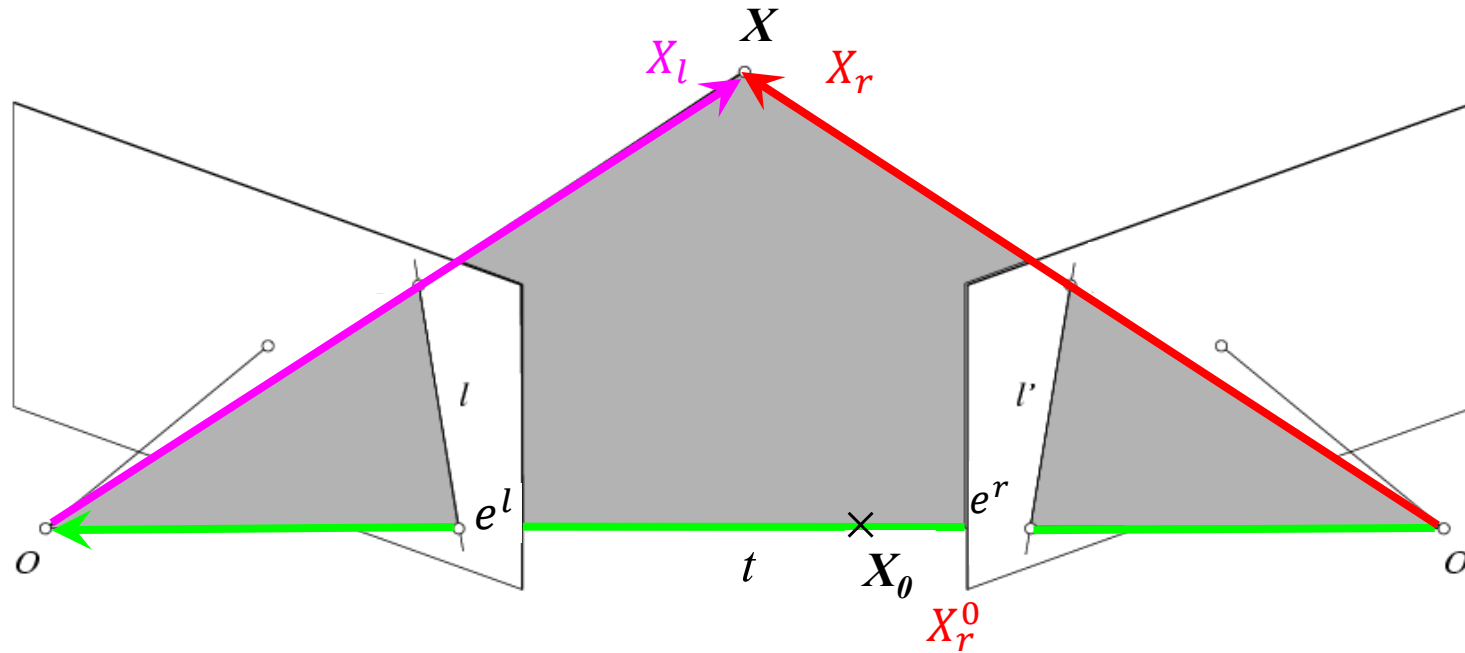
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Consider any  $X_0$  lies between the two centers of projection

$$EX_r^0 = (R[t]_{\times})X_r^0$$

# Essential matrix (Longuet-Higgins, 1981)

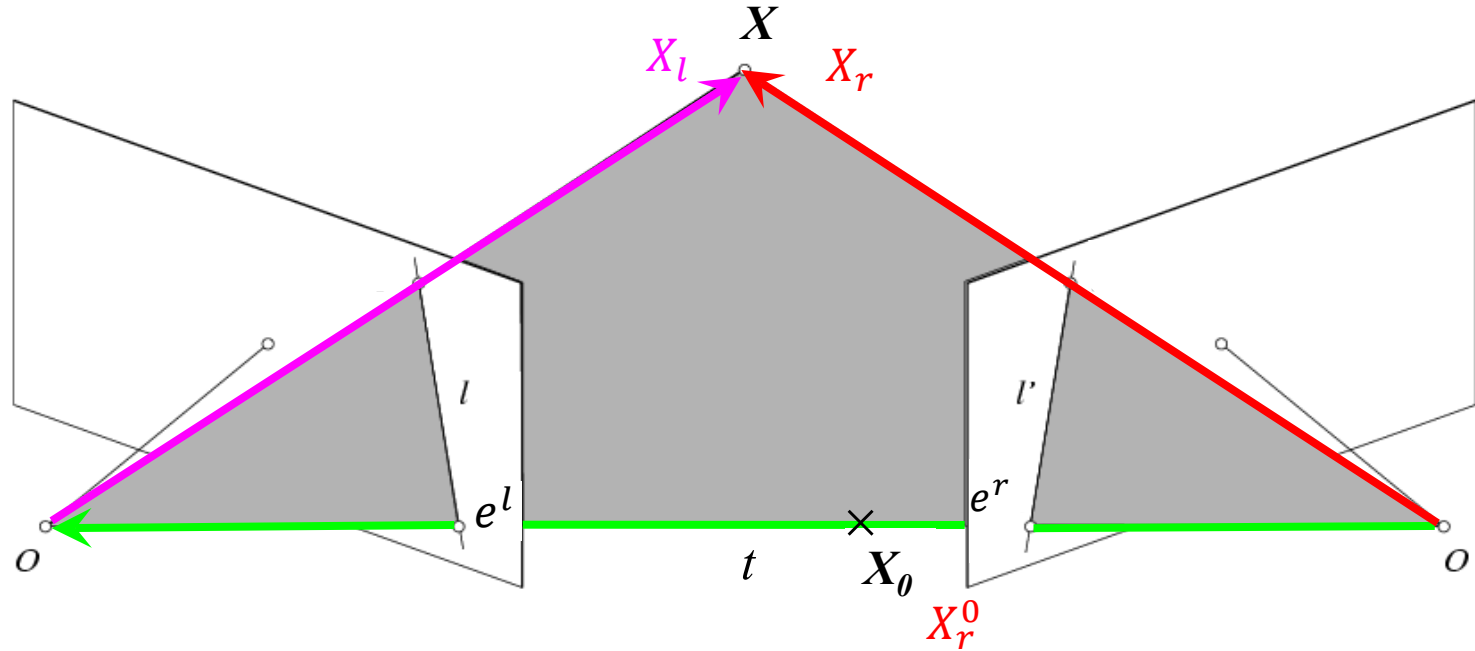


Consider any  $X_0$  lies between the two centers of projection

$$EX_r^0 = (R[t]_{\times})X_r^0 = R(t \times X_r^0)$$



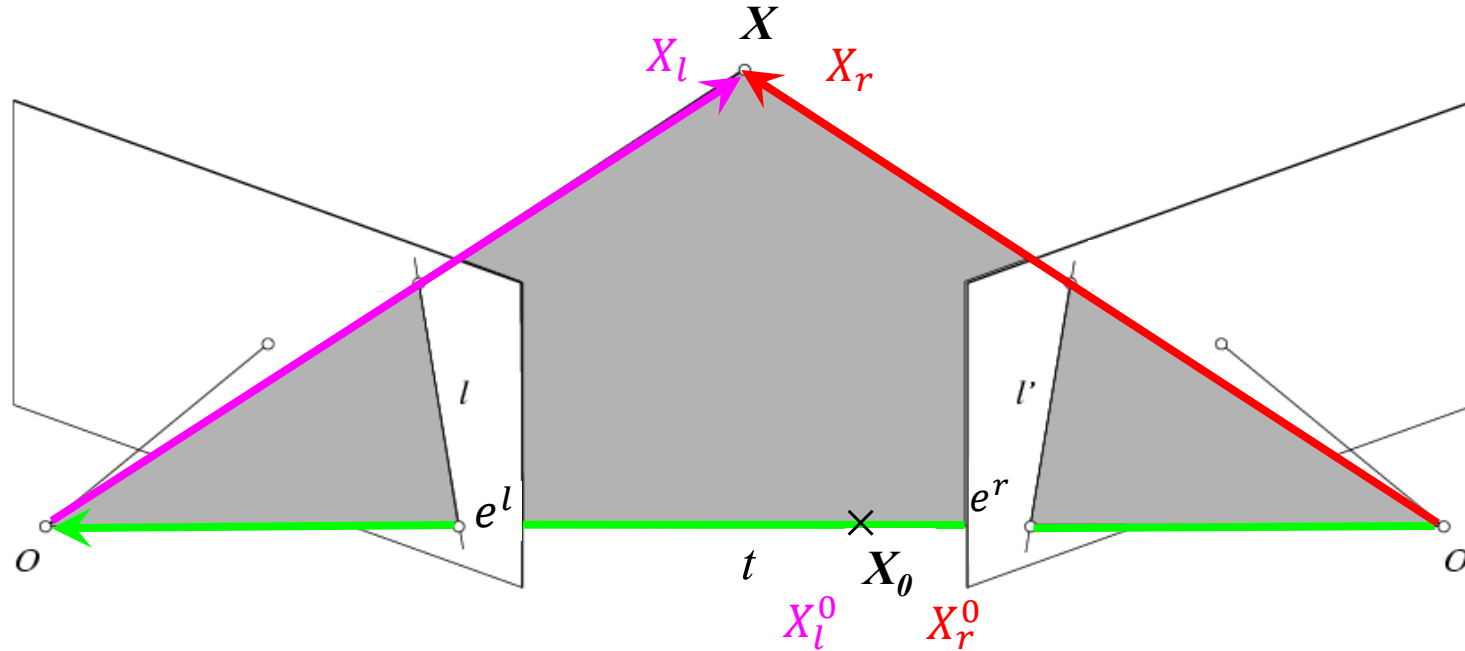
# Essential matrix (Longuet-Higgins, 1981)



Consider any  $X_0$  lies between the two centers of projection

$$EX_r^0 = (R[t]_{\times})X_r^0 = R(t \times X_r^0) = 0$$

# Essential matrix (Longuet-Higgins, 1981)

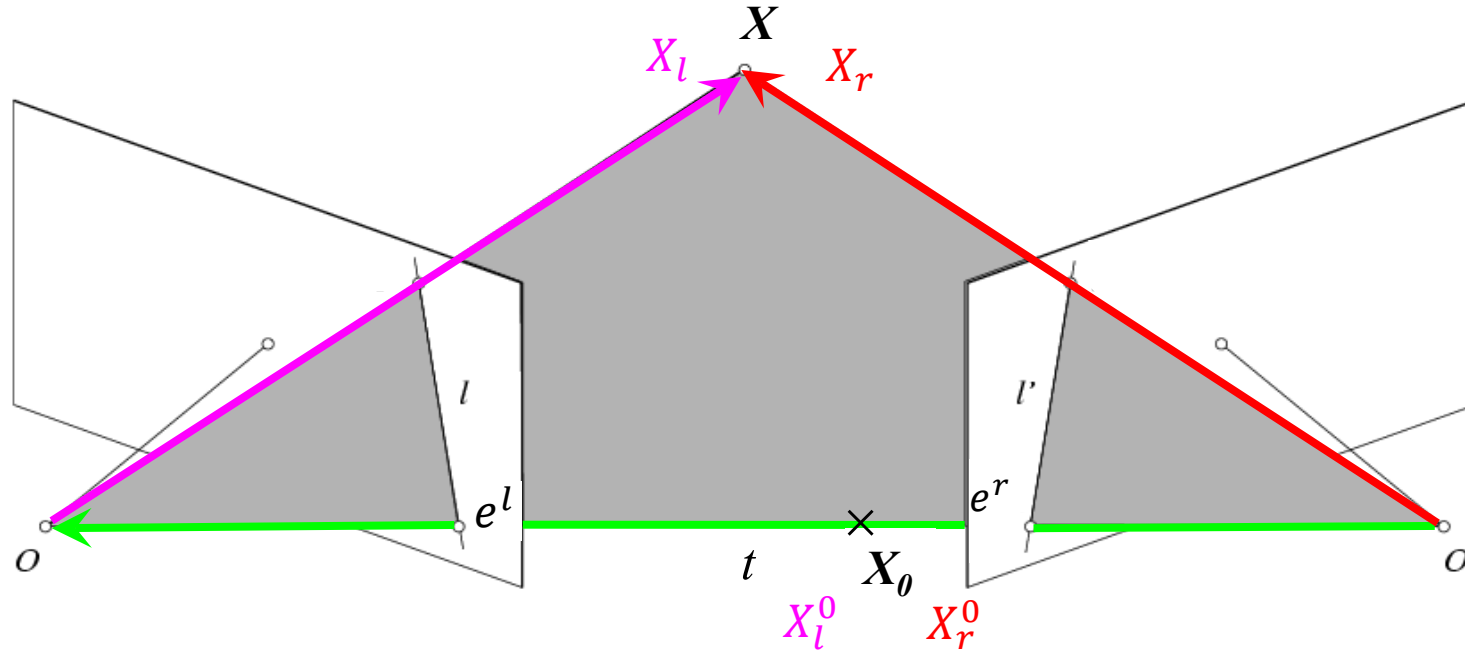


Consider any  $X_0$  lies between the two centers of projection

$$E X_r^0 = (R[t]_{\times}) X_r^0 = R(t \times X_r^0) = 0$$

$$E^T X_l^0 = (R[t]_{\times})^T R(X_l^0 - t)$$

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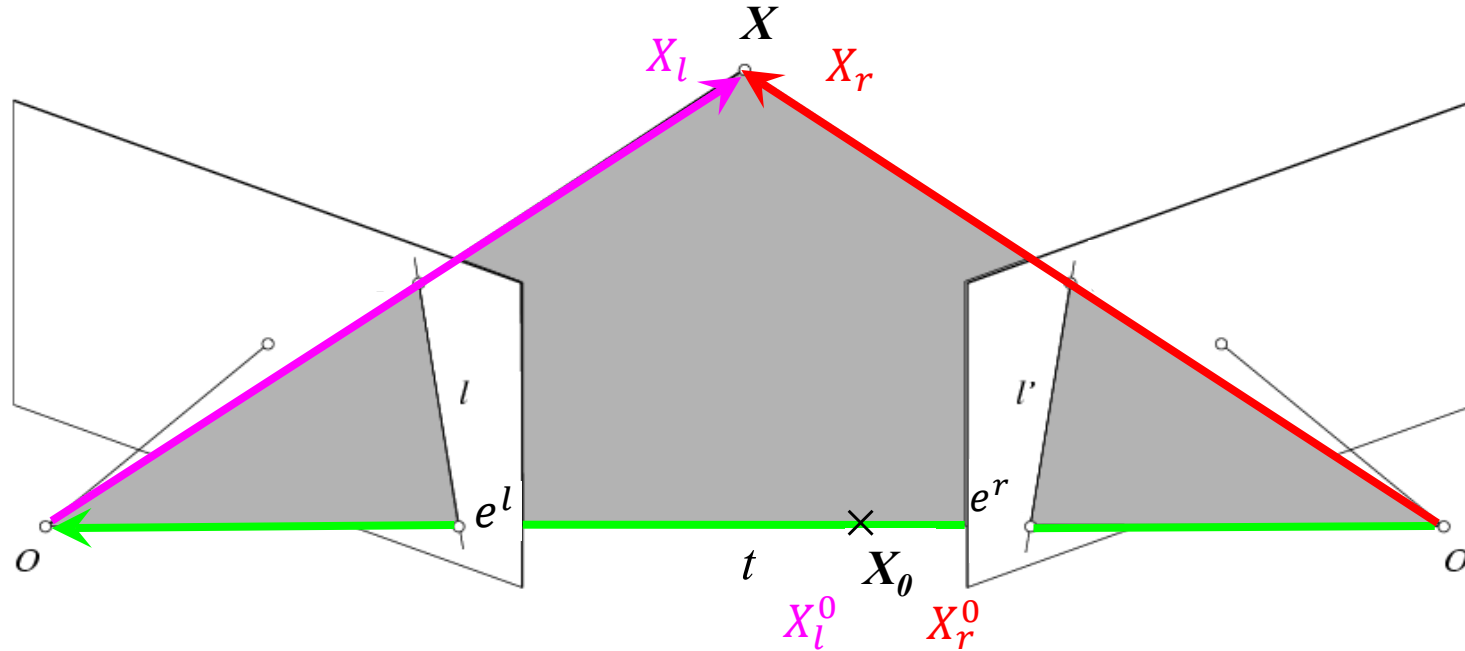


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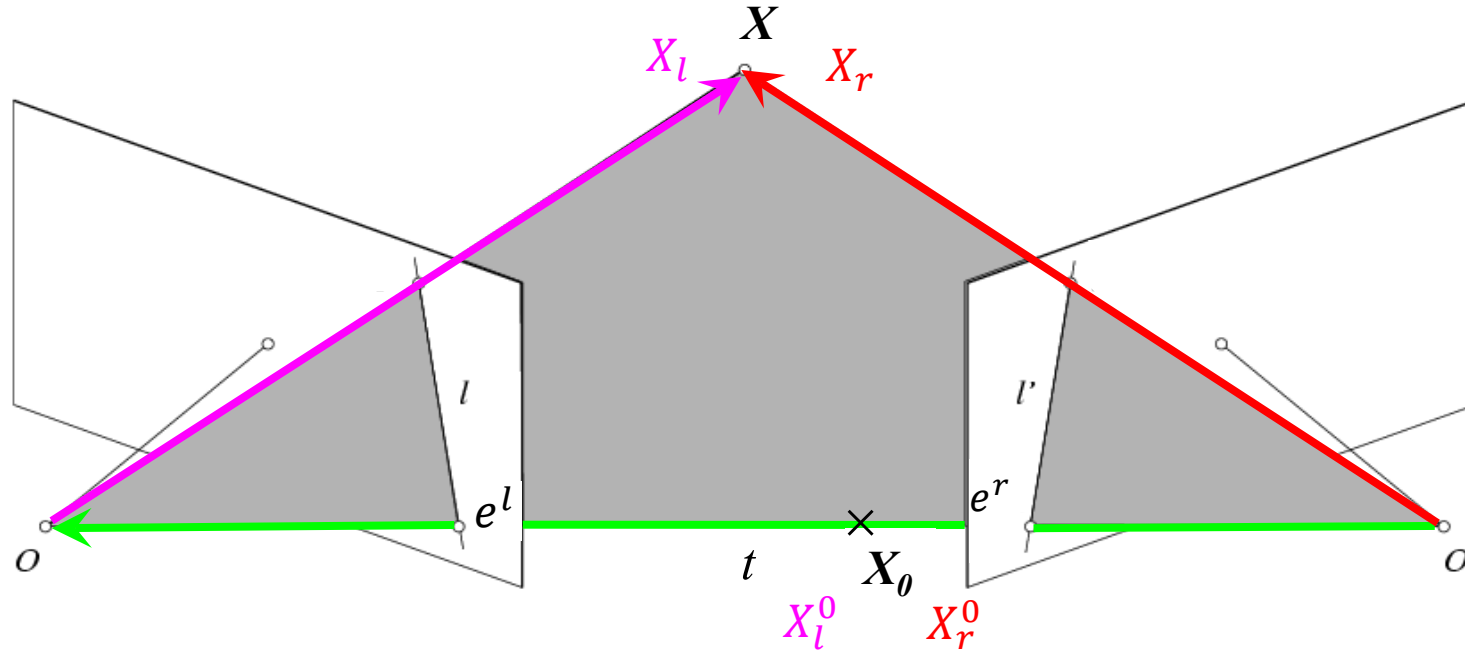
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$$\underbrace{\begin{bmatrix} 0 & -t_3 & t_2 \\ t_3 & 0 & -t_1 \\ -t_2 & t_1 & 0 \end{bmatrix}}_{[t]_{\times}}$$

# Essential matrix (Longuet-Higgins, 1981)



Consider any  $X_0$  lies between the two centers of projection

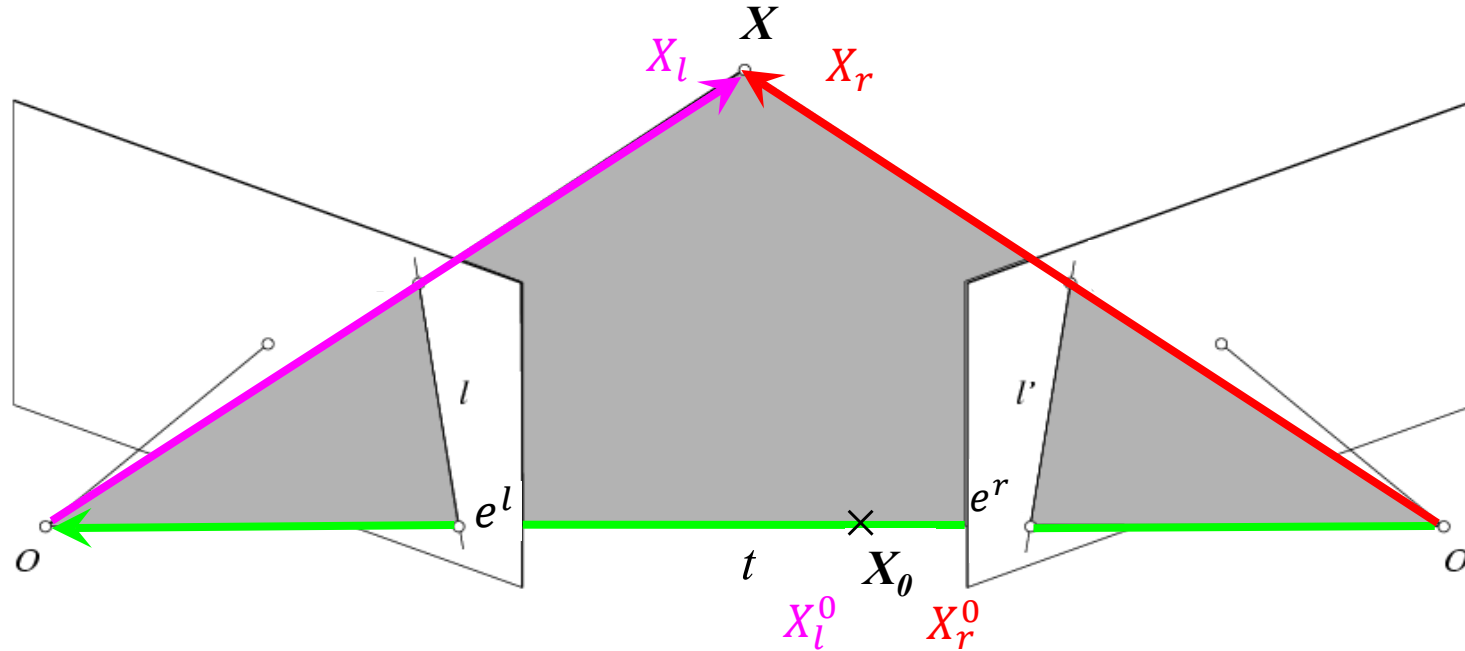
$$E X_r^0 = (R[t]_{\times}) X_r^0 = R(t \times X_r^0) = 0$$

$$E^T X_l^0 = (R[t]_{\times})^T R(X_r^0 - t) = ([t]_{\times})^T R^T R(X_r^0 - t)$$

$$= [-t]_{\times}(X_r^0 - t)$$

$$\underbrace{\begin{bmatrix} 0 & -t_3 & t_2 \\ t_3 & 0 & -t_1 \\ -t_2 & t_1 & 0 \end{bmatrix}}_{[t]_{\times}}$$

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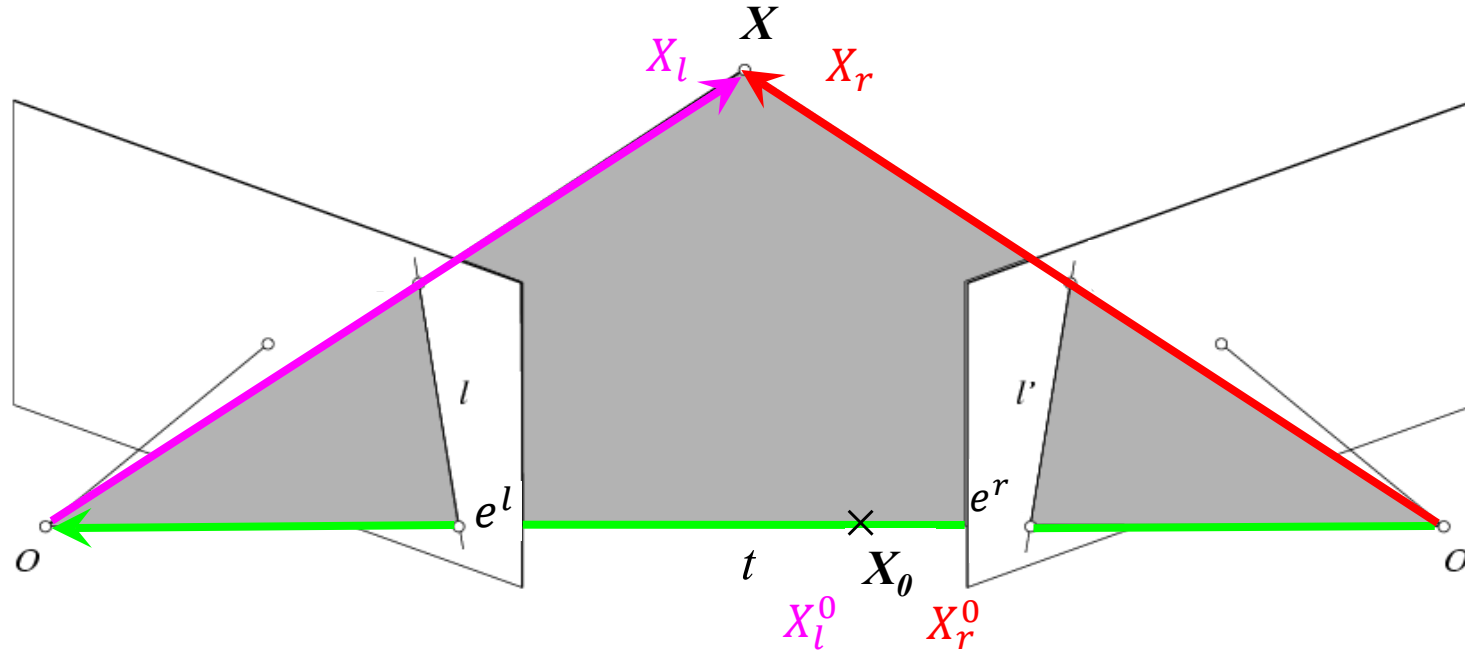
$$E X_r^0 = (R[t]_{\times}) X_r^0 = R(t \times X_r^0) = 0$$

$$E^T X_l^0 = (R[t]_{\times})^T R(X_r^0 - t) = ([t]_{\times})^T R^T R(X_r^0 - t)$$

$$= [-t]_{\times}(X_r^0 - t) = -t \times (X_r^0 - t)$$

$$\underbrace{\begin{bmatrix} 0 & -t_3 & t_2 \\ t_3 & 0 & -t_1 \\ -t_2 & t_1 & 0 \end{bmatrix}}_{[t]_{\times}}$$

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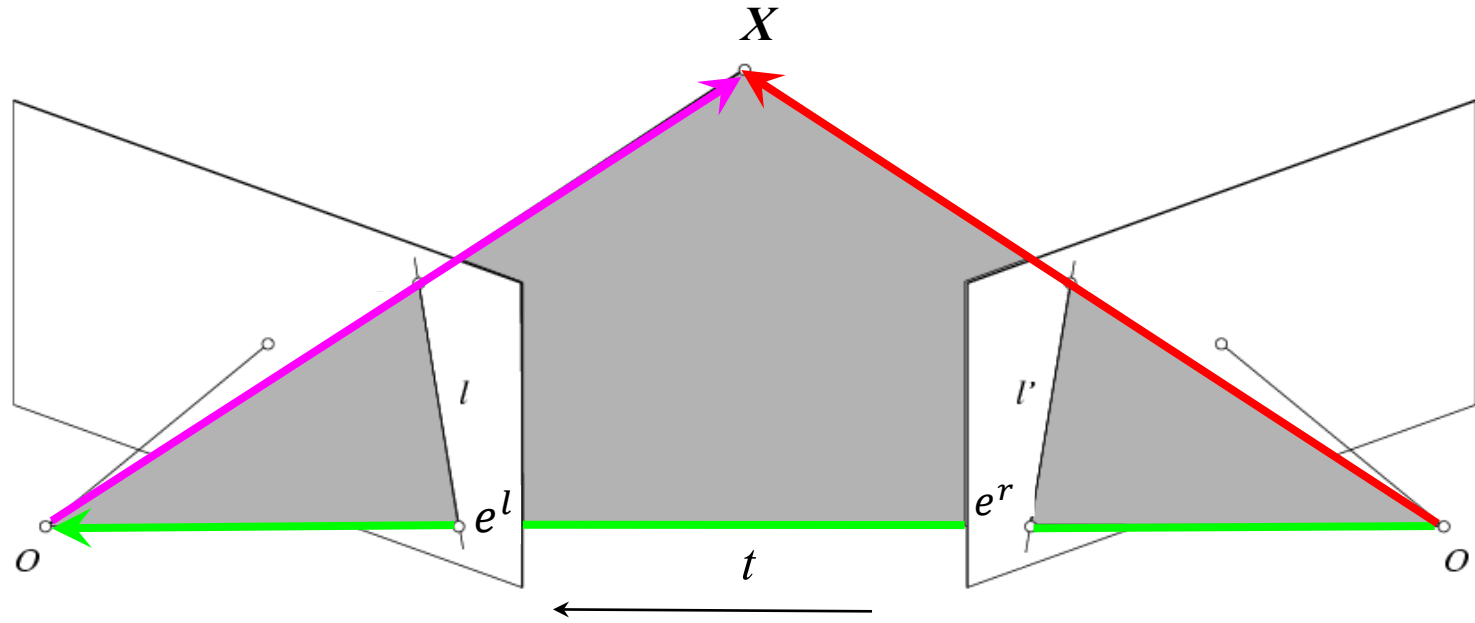
$$E X_r^0 = (R[t]_{\times}) X_r^0 = R(t \times X_r^0) = 0$$

$$E^T X_l^0 = (R[t]_{\times})^T R(X_r^0 - t) = ([t]_{\times})^T R^T R(X_r^0 - t)$$

$$= [-t]_{\times}(X_r^0 - t) = -t \times (X_r^0 - t) = 0$$

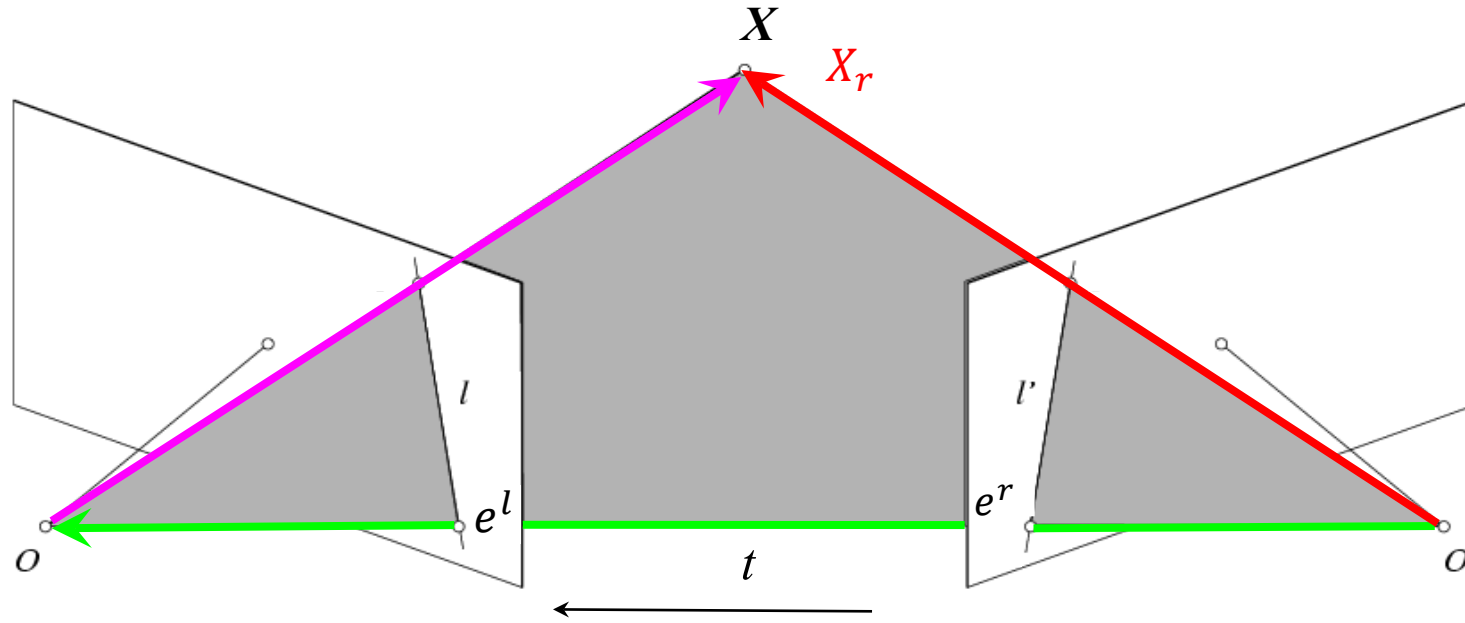
$$\underbrace{\begin{bmatrix} 0 & -t_3 & t_2 \\ t_3 & 0 & -t_1 \\ -t_2 & t_1 & 0 \end{bmatrix}}_{[t]_{\times}}$$

# Fundamental matrix (Faugeras and Luong, 1992)



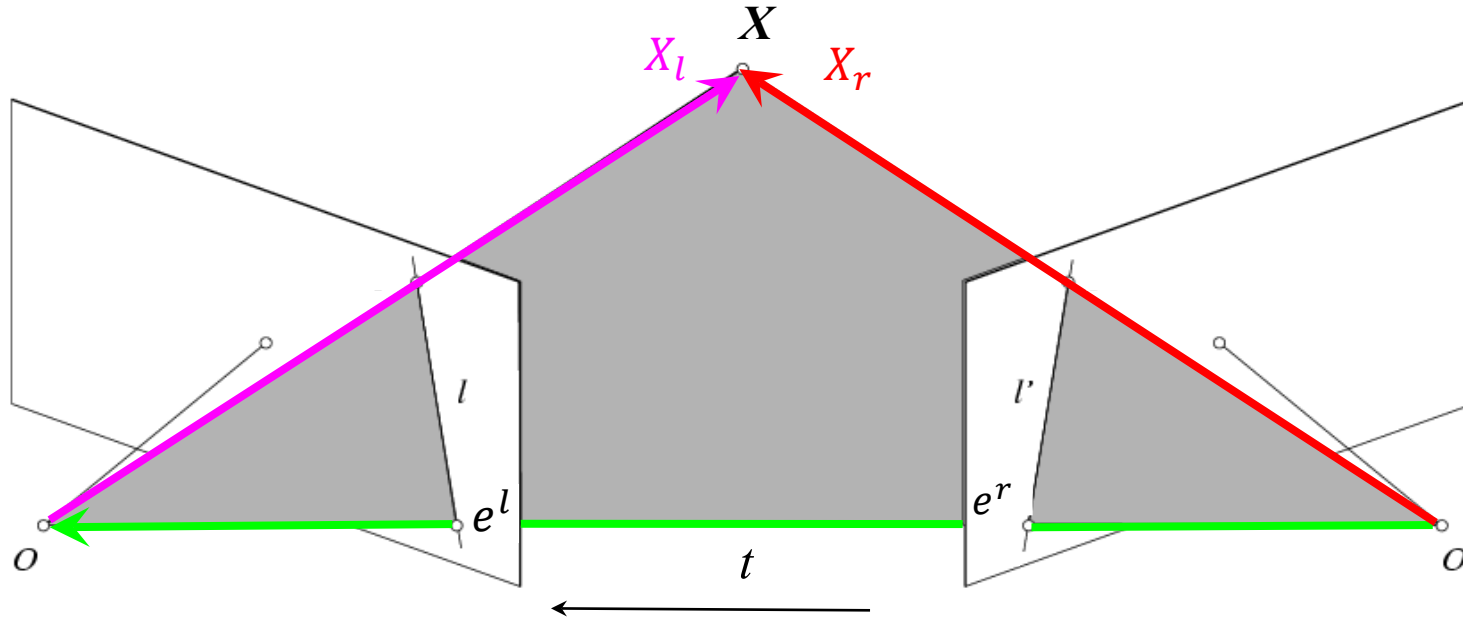


# Fundamental matrix (Faugeras and Luong, 1992)



$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

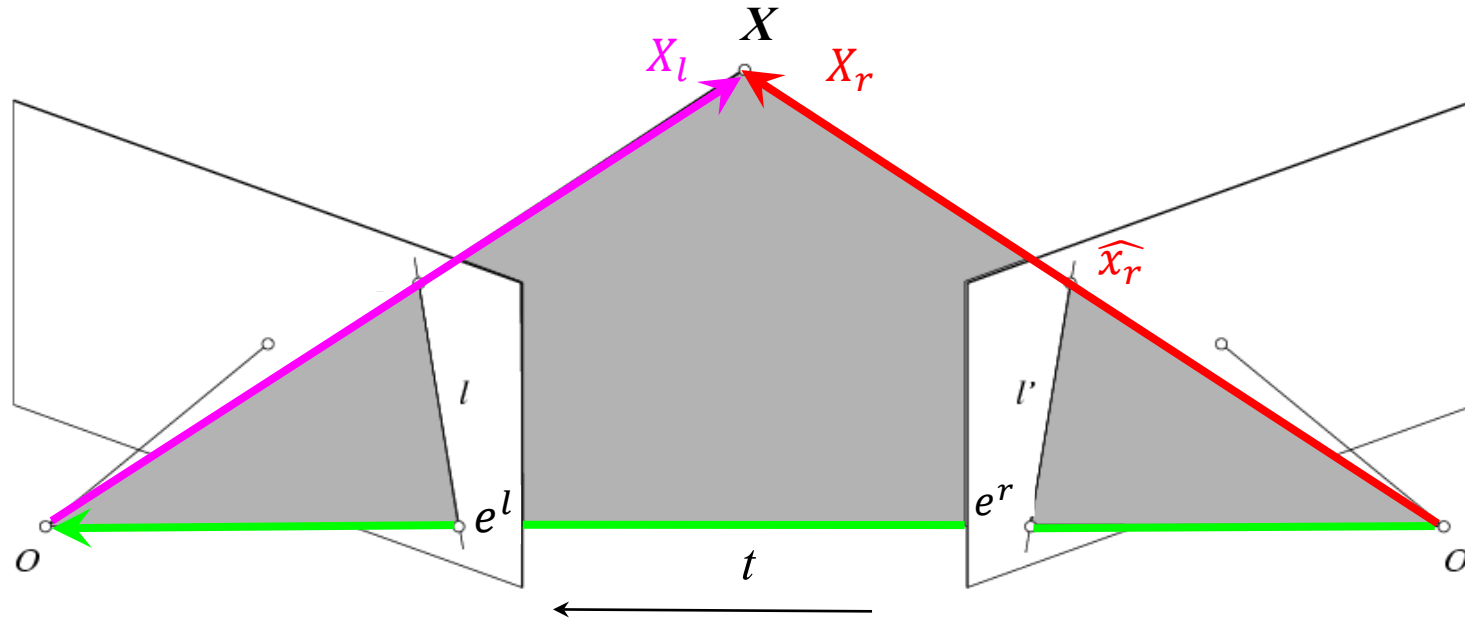
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$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

$X_l$ :  $X$  in terms of Cartesian coordinate of left camera

# Fundamental matrix (Faugeras and Luong, 1992)

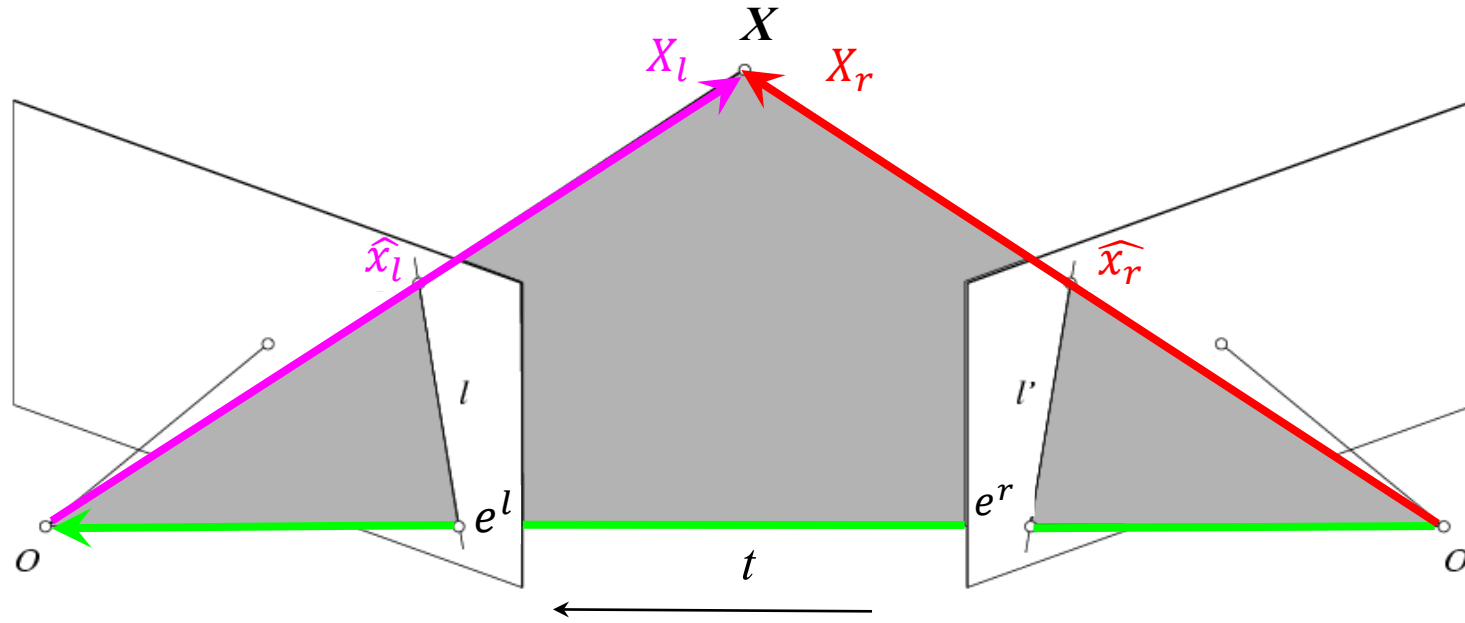


$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

$X_l$ :  $X$  in terms of Cartesian coordinate of left camera

$\hat{x}_r$ :  $X$ 's homogeneous coordinate of right view

# Fundamental matrix (Faugeras and Luong, 1992)



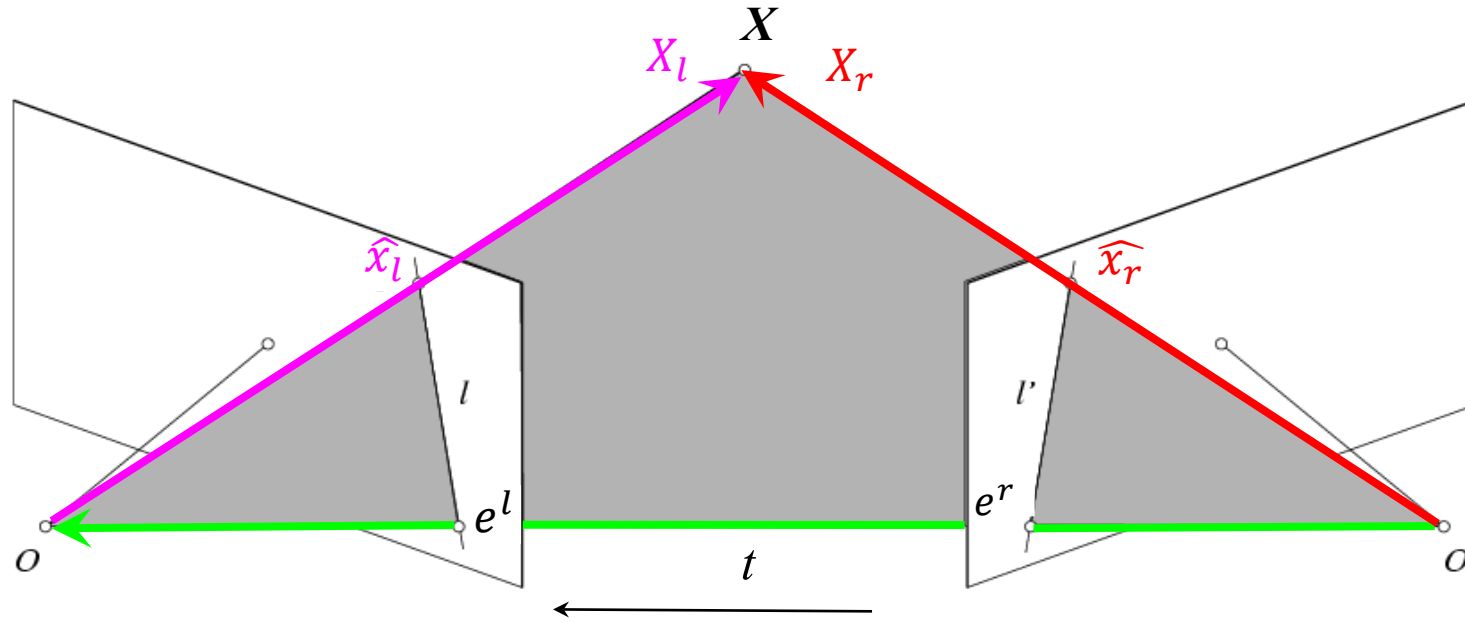
$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

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$\hat{x}_l$ :  $X$ 's homogeneous coordinate of left view

There exists  $F$  such that  $\hat{x}_l^T F \hat{x}_r = 0$

# Recall from last time

$$\mathbf{x} = \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}$$



$$w \begin{bmatrix} U \\ V \\ 1 \end{bmatrix} = \begin{bmatrix} -d_i & s & U_0 \\ 0 & -\alpha d_i & V_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

# Recall from last time

$$\mathbf{x} = \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}$$



$$w \begin{bmatrix} U \\ V \\ 1 \end{bmatrix} = \begin{bmatrix} -d_i & s & U_0 \\ 0 & -\alpha d_i & V_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Assume that  $\mathbf{t} = \mathbf{0}$  (the origin of camera coordinate = COP)  
and  $\mathbf{R} = I$  (camera coordinate is perfectly aligned)

# Recall from last time

$$\mathbf{x} = \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}$$



$$w \begin{bmatrix} U \\ V \\ 1 \end{bmatrix} = \begin{bmatrix} -d_i & s & U_0 \\ 0 & -\alpha d_i & V_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

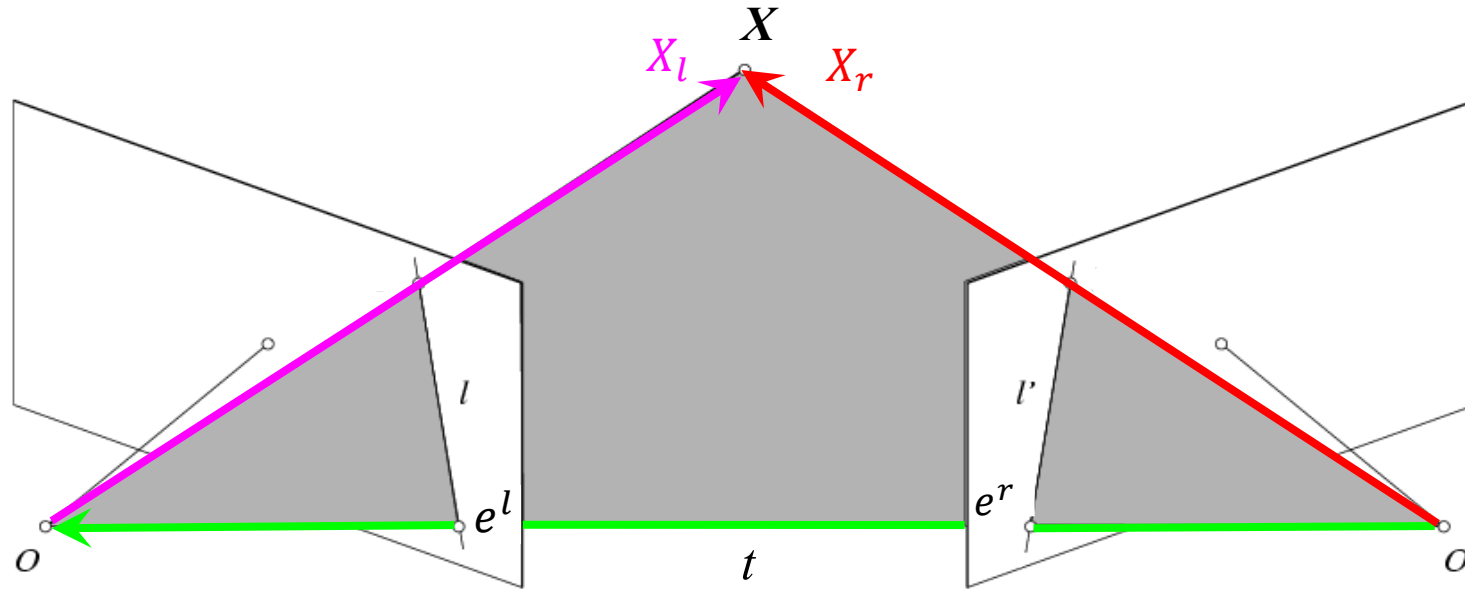
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and  $\mathbf{R} = I$  (camera coordinate is perfectly aligned)

$$\underbrace{\mathbf{x}}_{\substack{\text{2d projection} \\ \text{location in} \\ \text{homogenous} \\ \text{coordinate}}} = \mathbf{K} \underbrace{\mathbf{X}}_{\substack{\text{3d location in} \\ \text{Cartesian} \\ \text{coordinate}}}$$



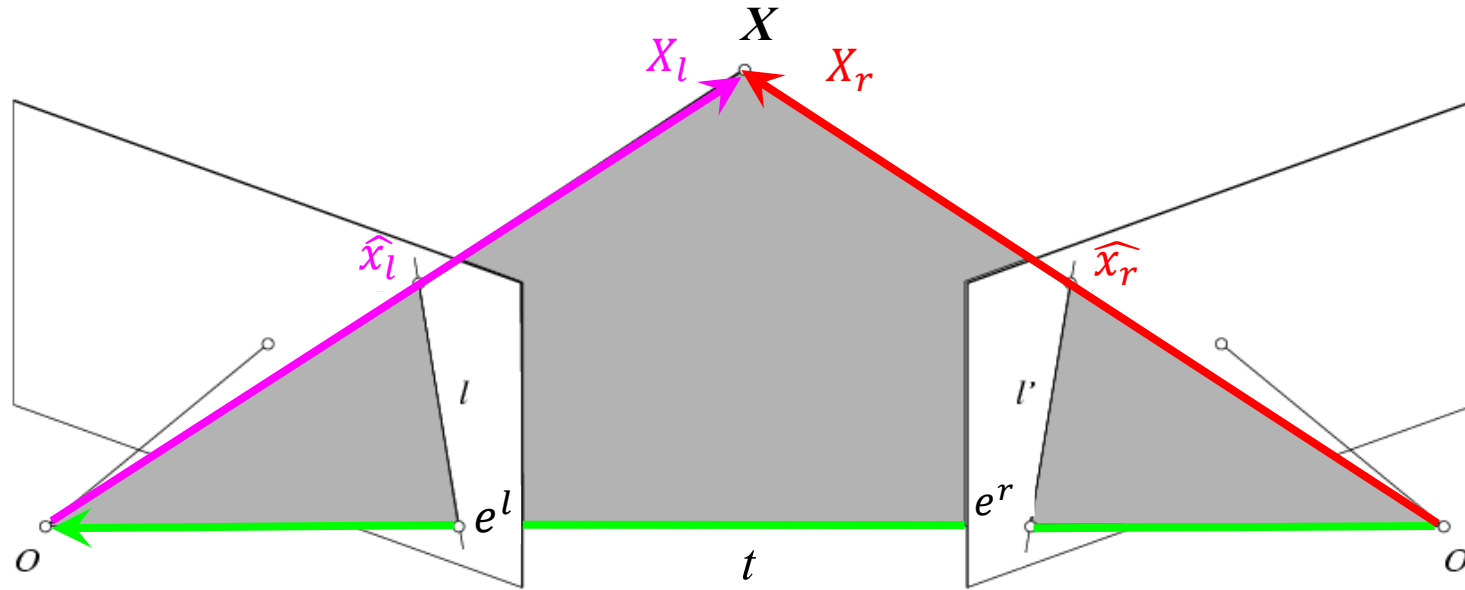
# Fundamental matrix



$X_l$ :  $X$  in terms of Cartesian coordinate of left camera

$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

# Fundamental matrix



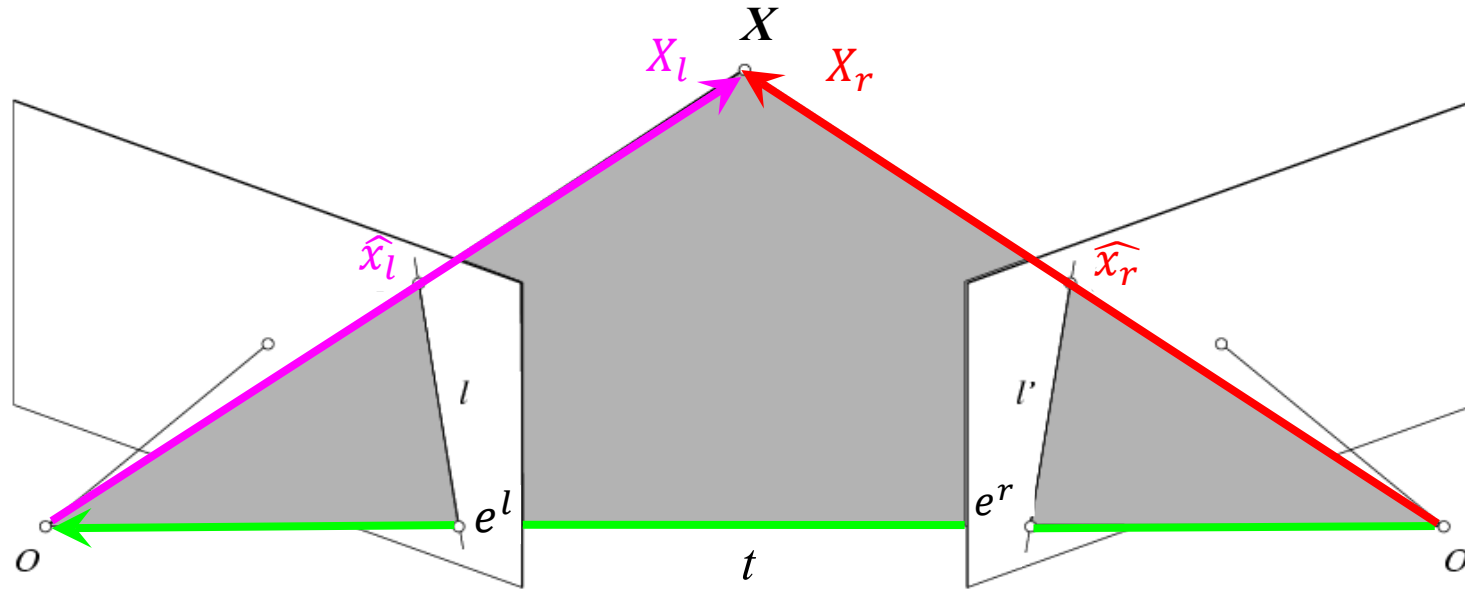
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$\hat{x}_l = K_l X_l$ :  $X$ 's homogeneous coordinate of left view

$\hat{x}_r = K_r X_r$ :  $X$ 's homogeneous coordinate of right view

# Fundamental matrix



$X_l$ :  $X$  in terms of Cartesian coordinate of left camera

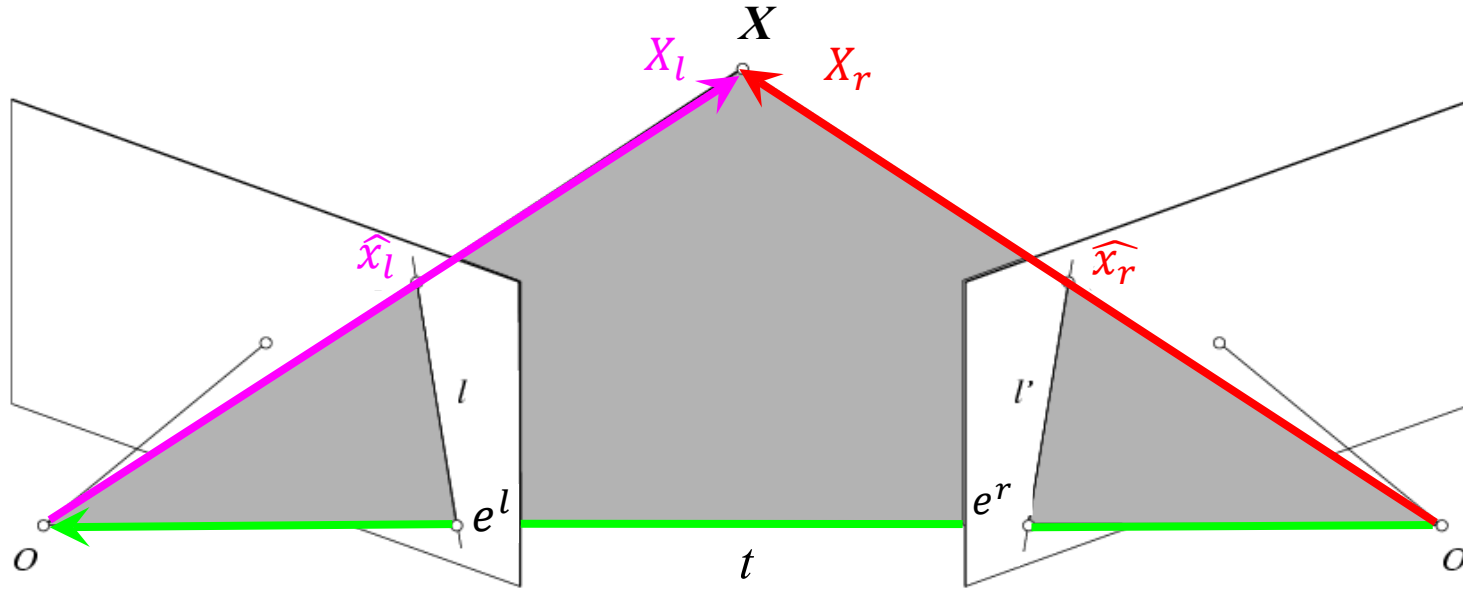
$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

$\hat{x}_l = K_l X_l$ :  $X$ 's homogeneous coordinate of left view

$\hat{x}_r = K_r X_r$ :  $X$ 's homogeneous coordinate of right view

$$\hat{x}_l^T \underbrace{(K_l^T)^{-1} E K_r^{-1}}_F \hat{x}_r$$

# Fundamental matrix



$X_l$ :  $X$  in terms of Cartesian coordinate of left camera

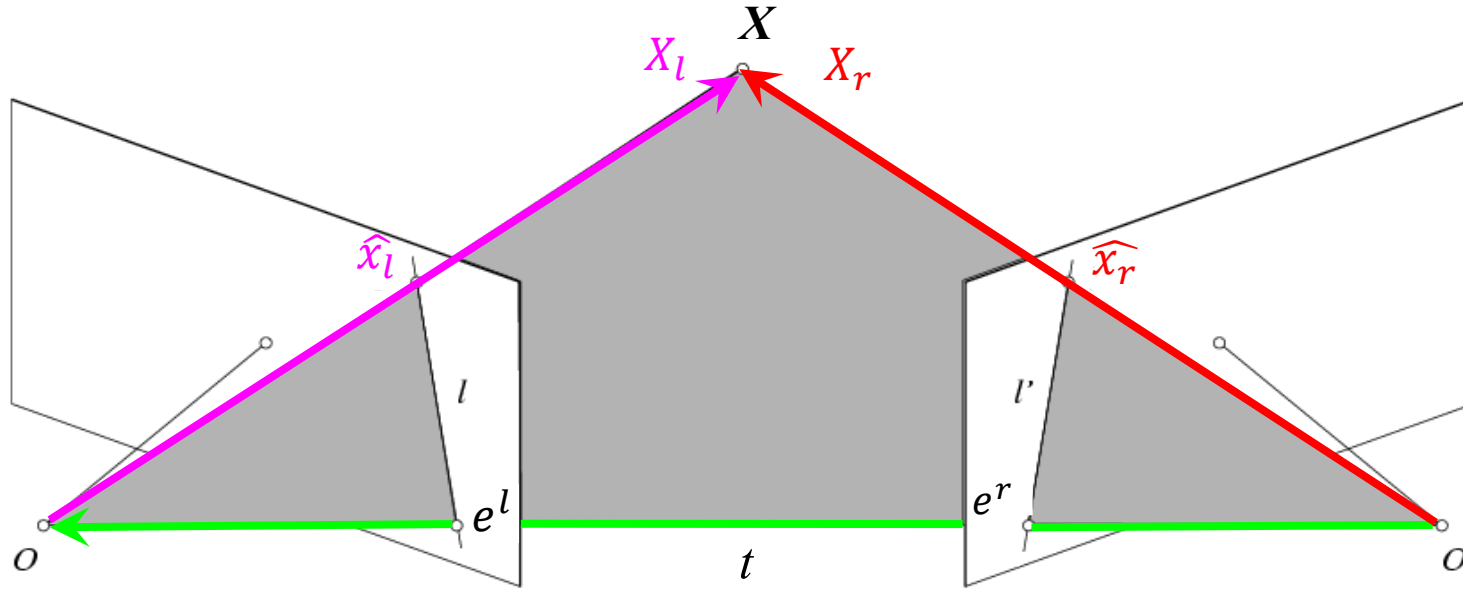
$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

$\hat{x}_l = K_l X_l$ :  $X$ 's homogeneous coordinate of left view

$\hat{x}_r = K_r X_r$ :  $X$ 's homogeneous coordinate of right view

$$\hat{x}_l^T \underbrace{(K_l^T)^{-1} E K_r^{-1}}_F \hat{x}_r = \underbrace{X_l^T K_l^T}_{\hat{x}_l} \underbrace{(K_l^T)^{-1} E K_r^{-1}}_F \underbrace{K_r X_r}_{\hat{x}_r}$$

# Fundamental matrix



$X_l$ :  $X$  in terms of Cartesian coordinate of left camera

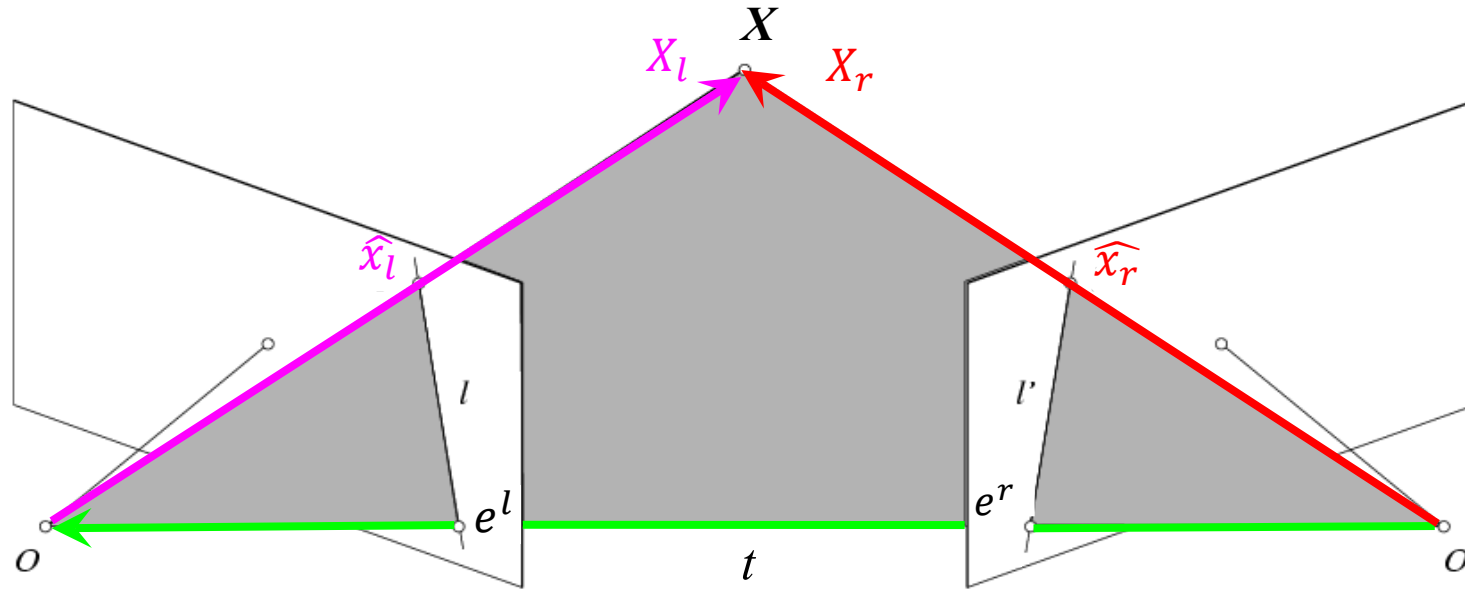
$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

$\hat{x}_l = K_l X_l$ :  $X$ 's homogeneous coordinate of left view

$\hat{x}_r = K_r X_r$ :  $X$ 's homogeneous coordinate of right view

$$\hat{x}_l^T \underbrace{(K_l^T)^{-1} E K_r^{-1}}_F \hat{x}_r = \underbrace{X_l^T K_l^T}_{\hat{x}_l} (K_l^T)^{-1} E K_r^{-1} \underbrace{K_r X_r}_{\hat{x}_r} = X_l^T E X_r$$

# Fundamental matrix



$X_l$ :  $X$  in terms of Cartesian coordinate of left camera

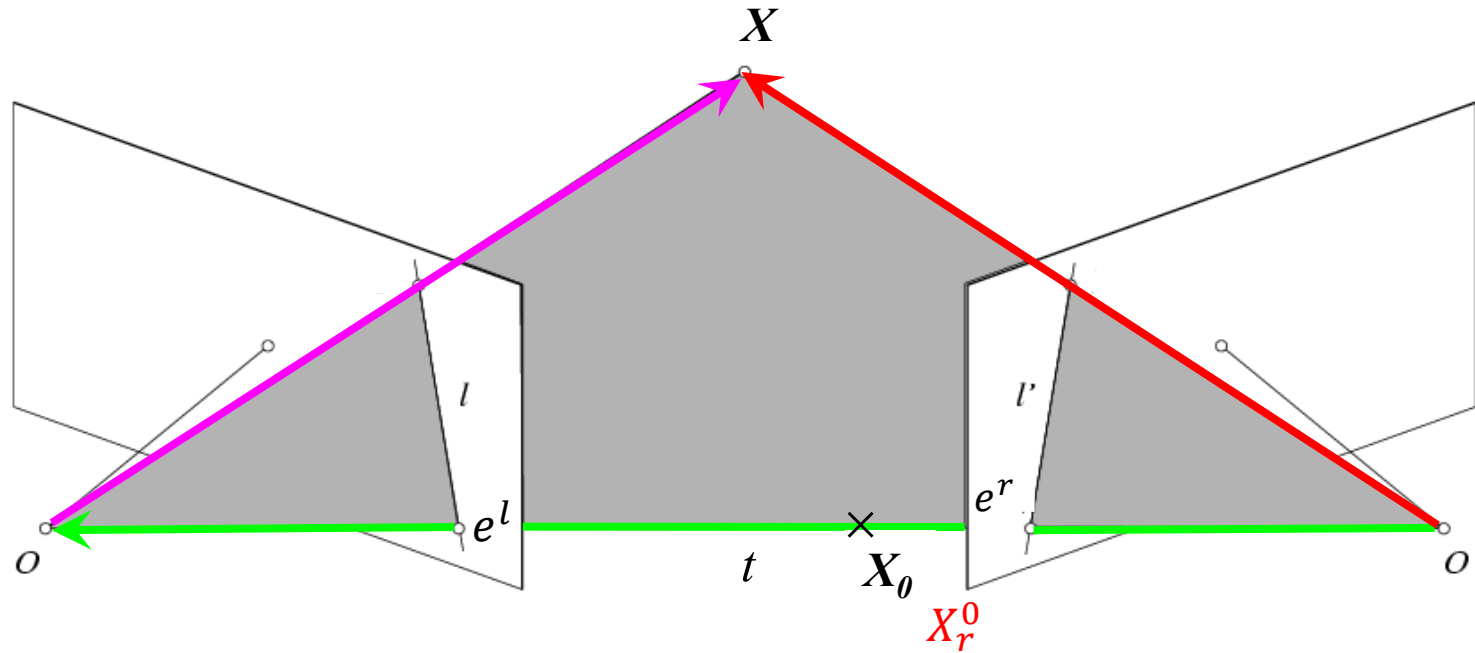
$X_r$ :  $X$  in terms of Cartesian coordinate of right camera

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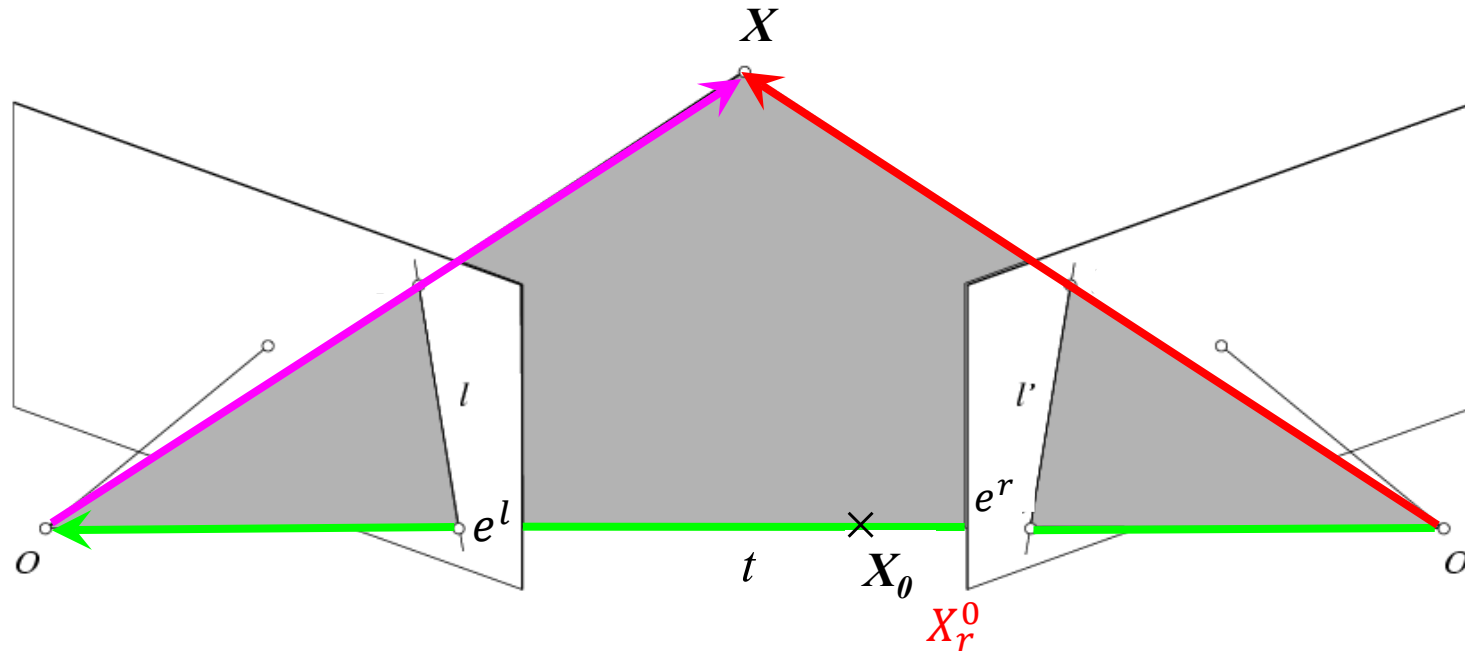
$$\hat{x}_l^T \underbrace{(K_l^T)^{-1} E K_r^{-1}}_F \hat{x}_r = \underbrace{X_l^T K_l^T}_{\hat{x}_l} (K_l^T)^{-1} E K_r^{-1} \underbrace{K_r X_r}_{\hat{x}_r} = X_l^T E X_r = 0$$

# Epipoles and fundamental matrix



Consider any  $X_0$  lie between the two centers of projection

# Epipoles and fundamental matrix

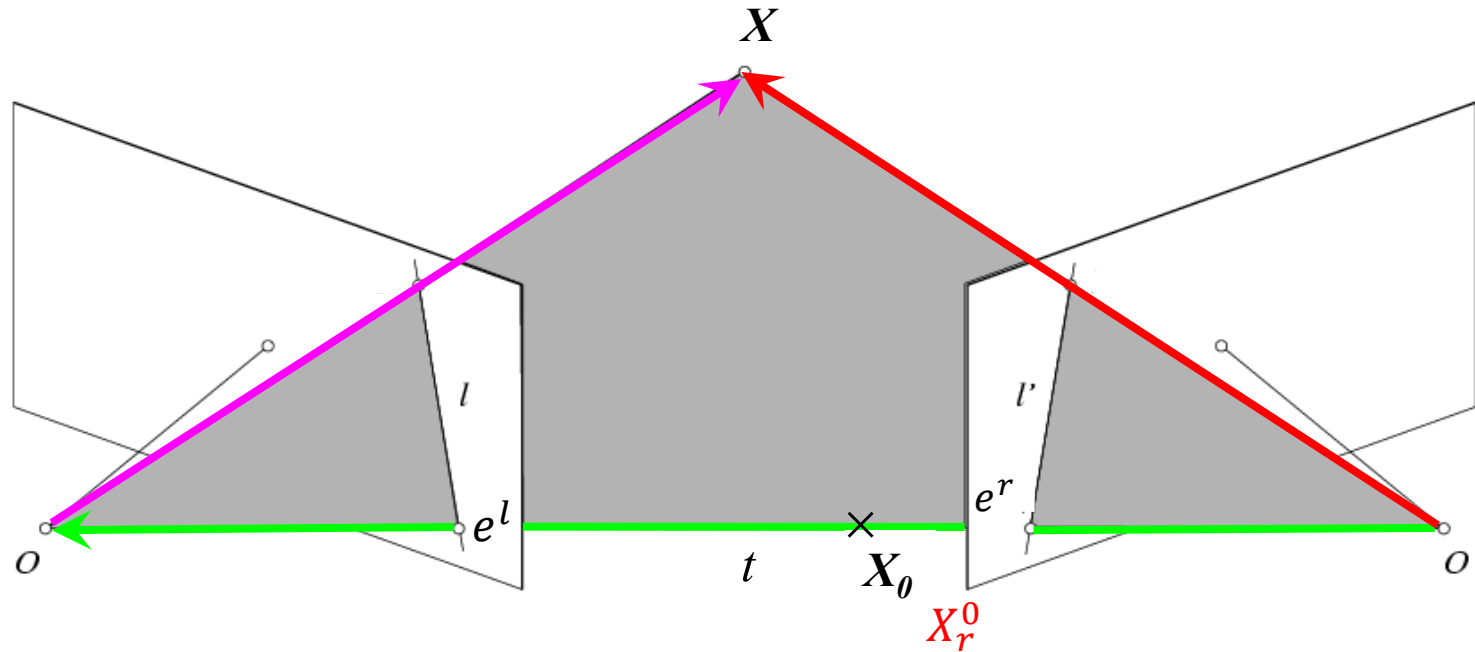


Consider any  $X_0$  lie between the two centers of projection

$$F e^r = \underbrace{(K_l^T)^{-1} E K_r^{-1}}_F e^r = (K_l^T)^{-1} E K_r^{-1} \underbrace{K_r X_r^0}_{e^r} = (K_l^T)^{-1} E X_r^0$$



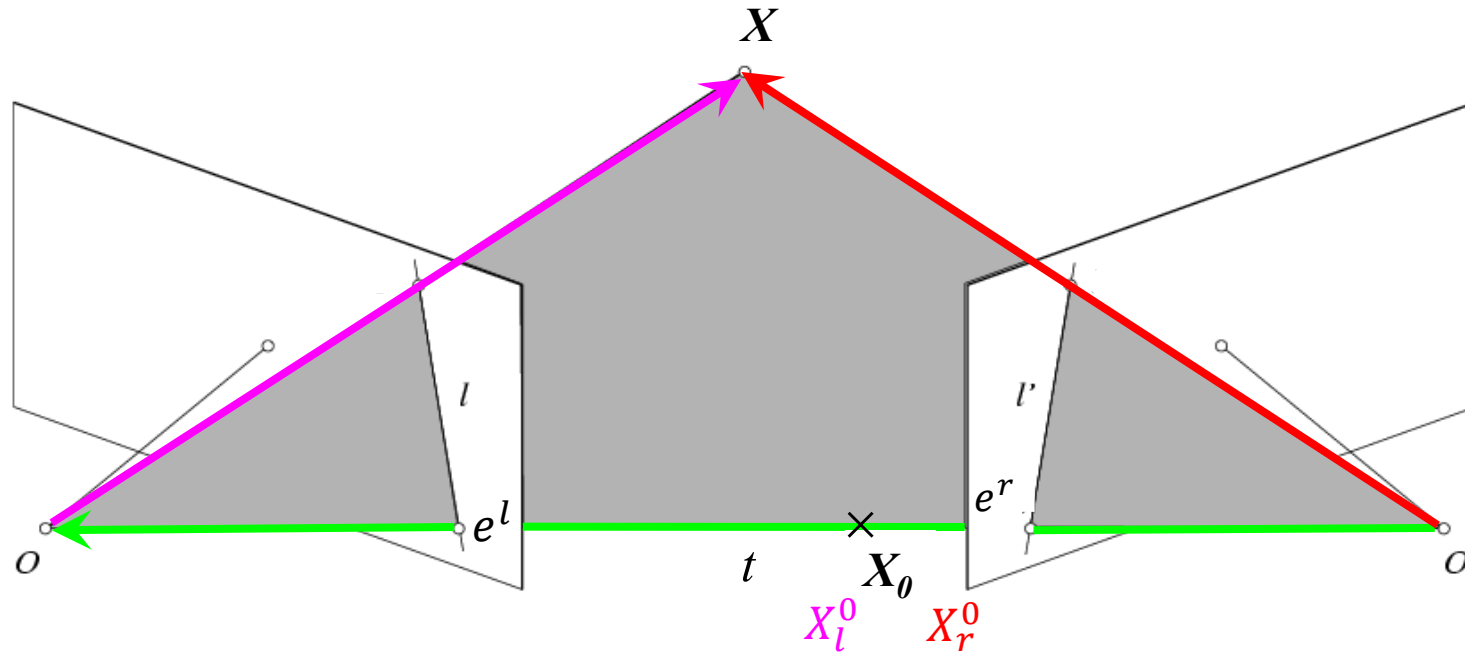
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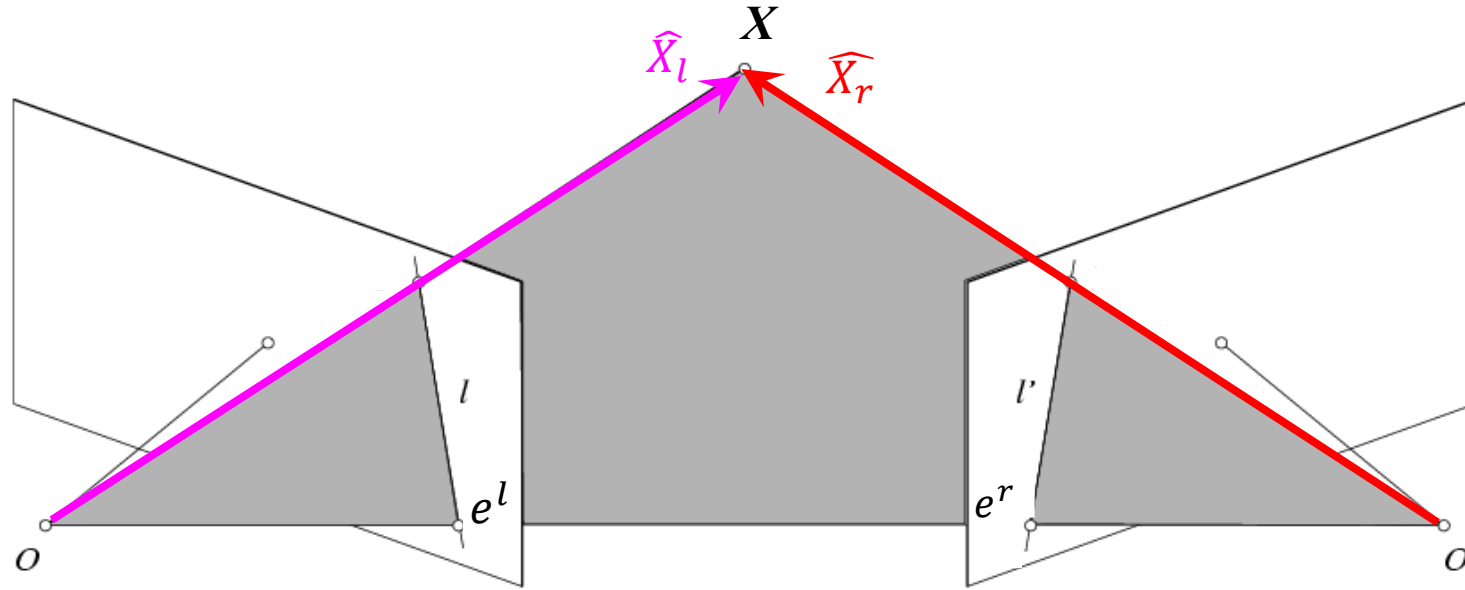


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# Review: Essential matrix

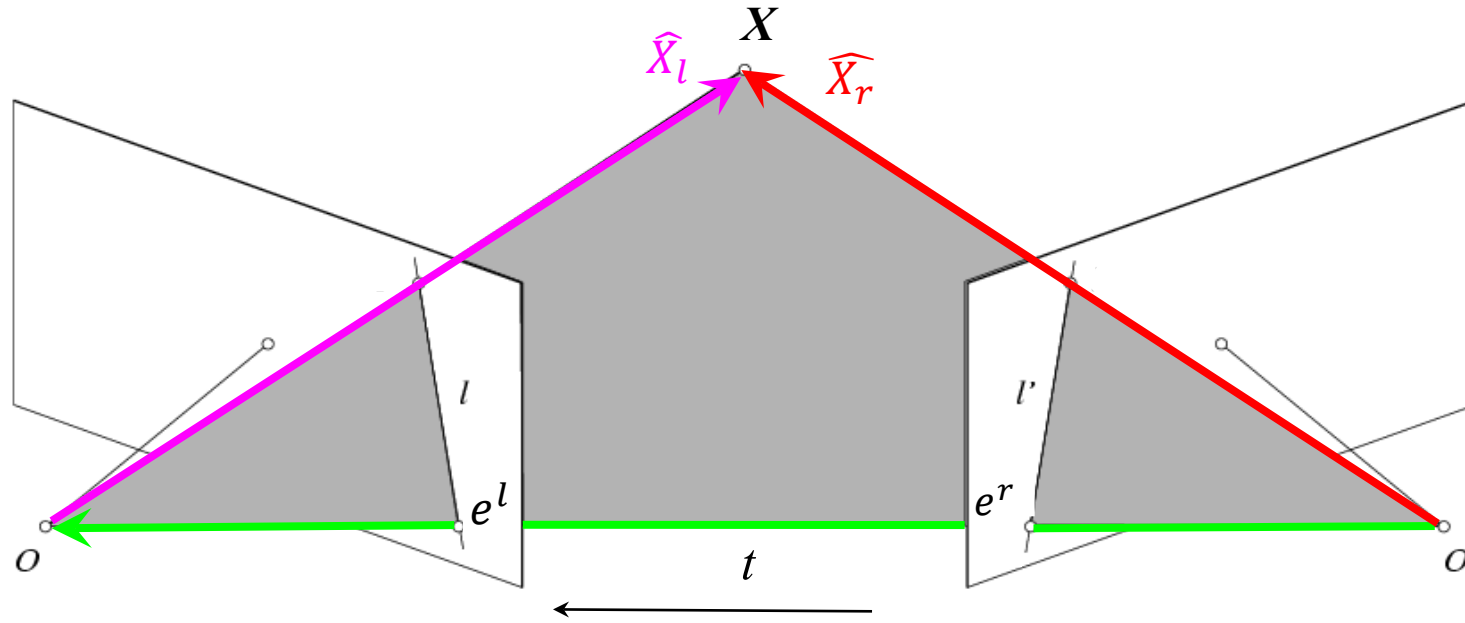


$\hat{X}_l$  :  $X$  in terms of Cartesian coordinate of left camera

$\hat{X}_r$  :  $X$  in terms of Cartesian coordinate of right camera

There exists  $E$  such that  $\hat{X}_l^T \underbrace{E}_{R[t]_x} \hat{X}_r = 0$

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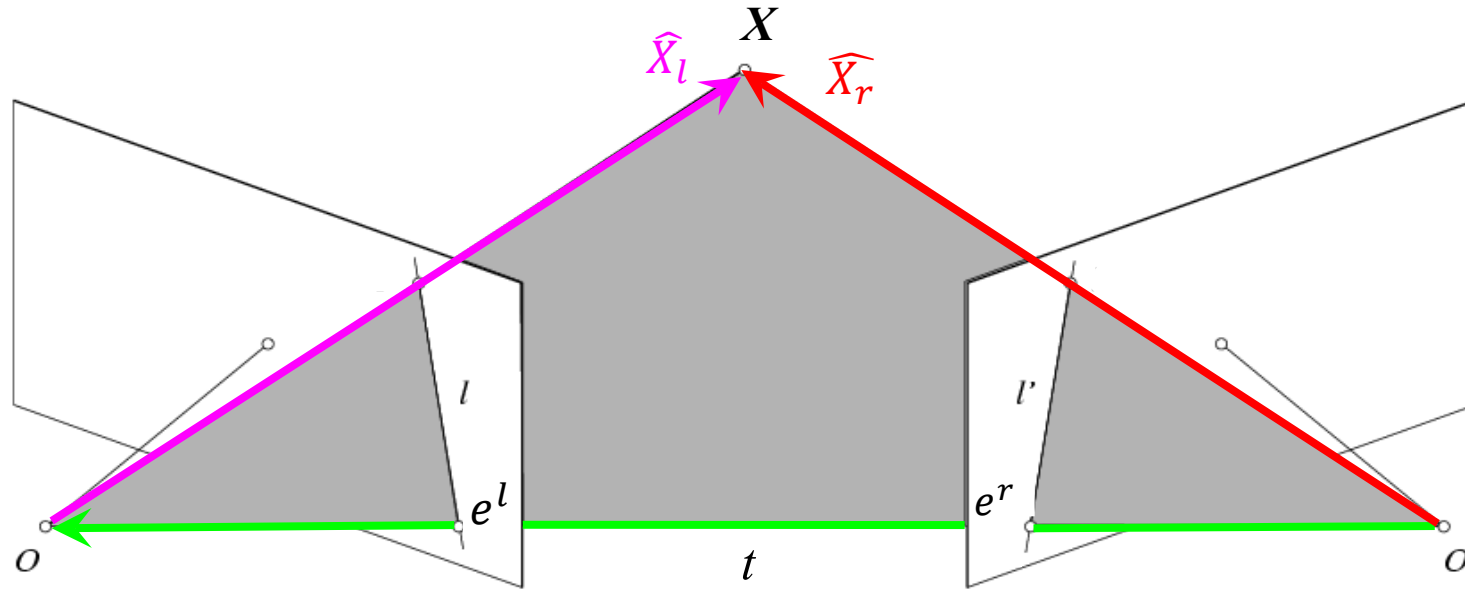


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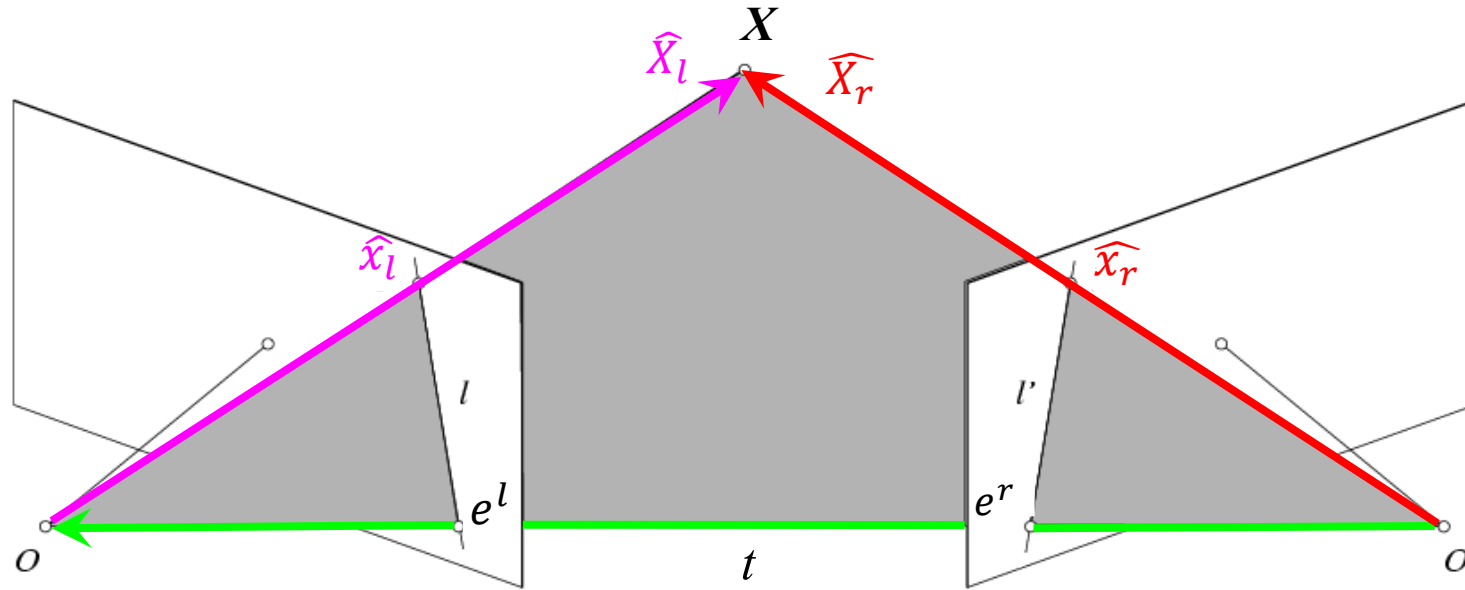
# Review: Fundamental matrix



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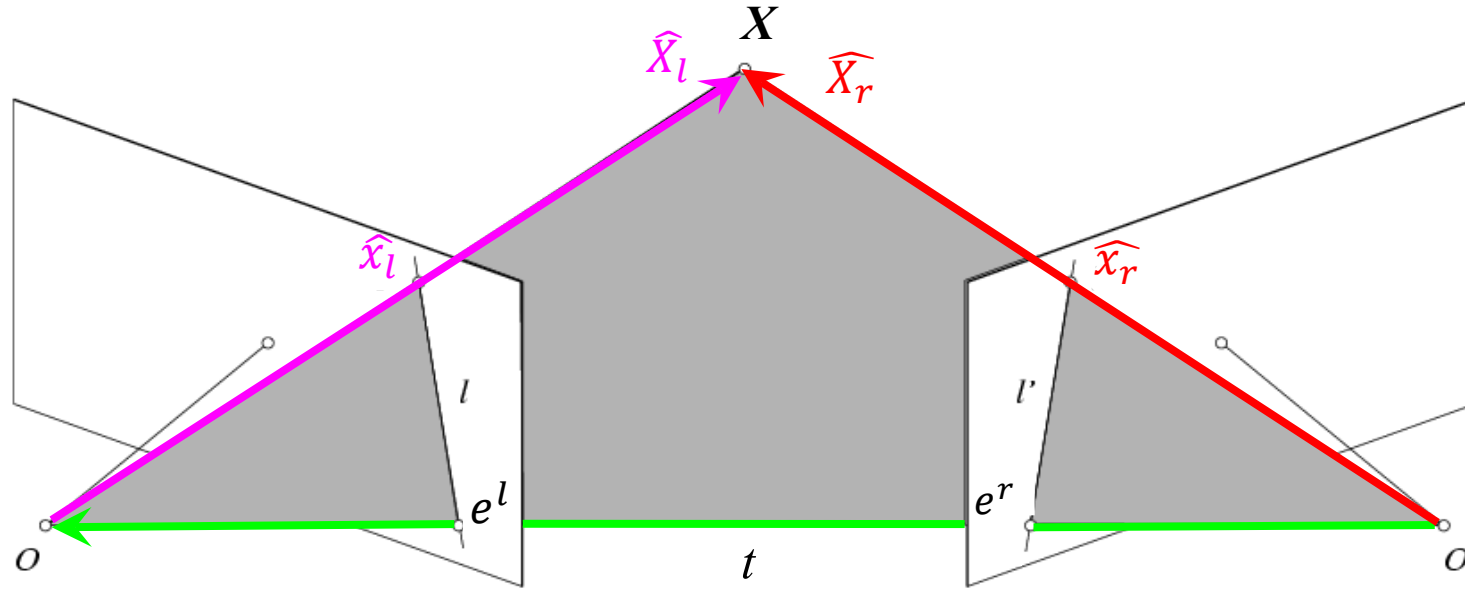
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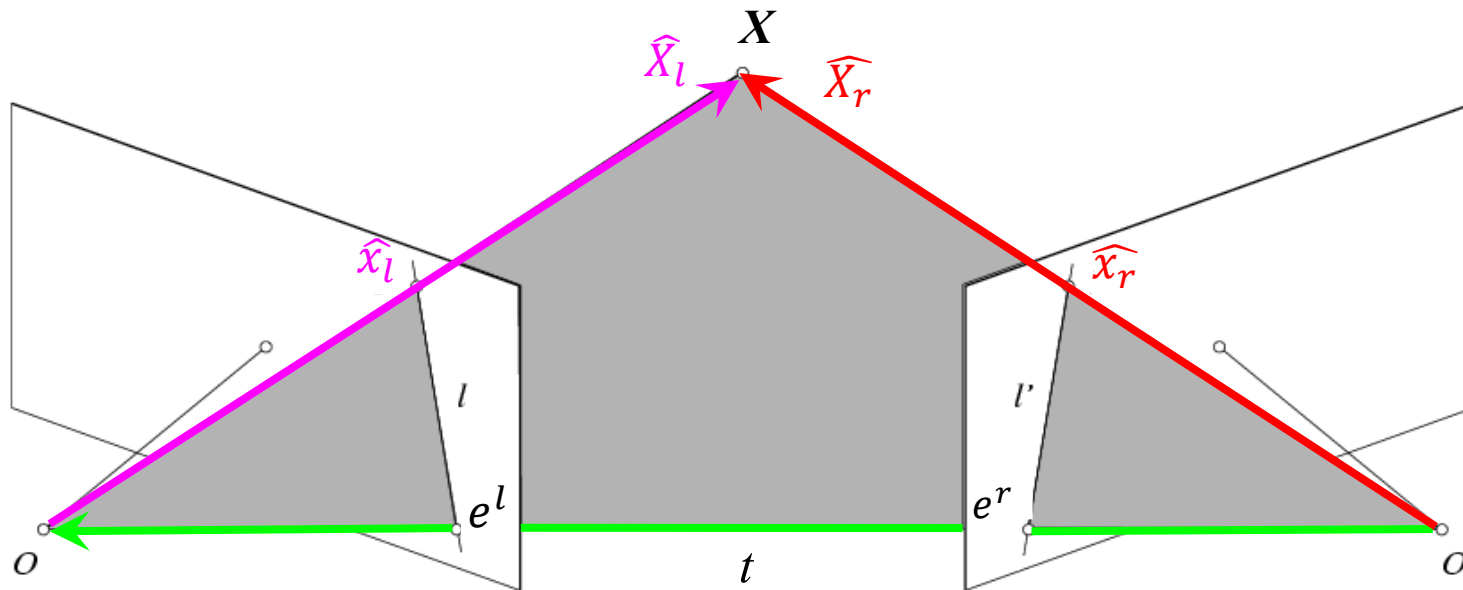
$\hat{X}_r$ :  $X$  in terms of Cartesian coordinate of right camera

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$\hat{x}_r = K_r \hat{X}_r$ :  $X$ 's homogeneous coordinate of right view

$$\hat{x}_l^T \underbrace{(K_l^T)^{-1} E K_r^{-1}}_F \hat{x}_r$$

# Review: Fundamental matrix



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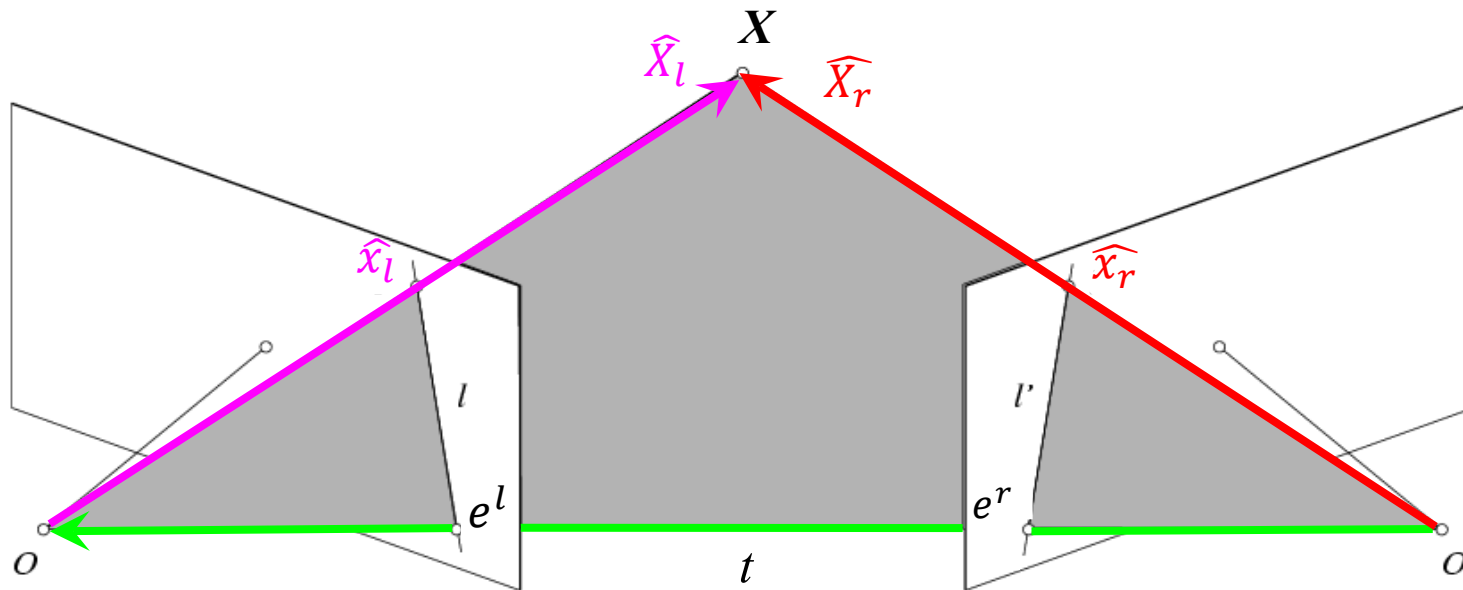
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# Review: Fundamental matrix



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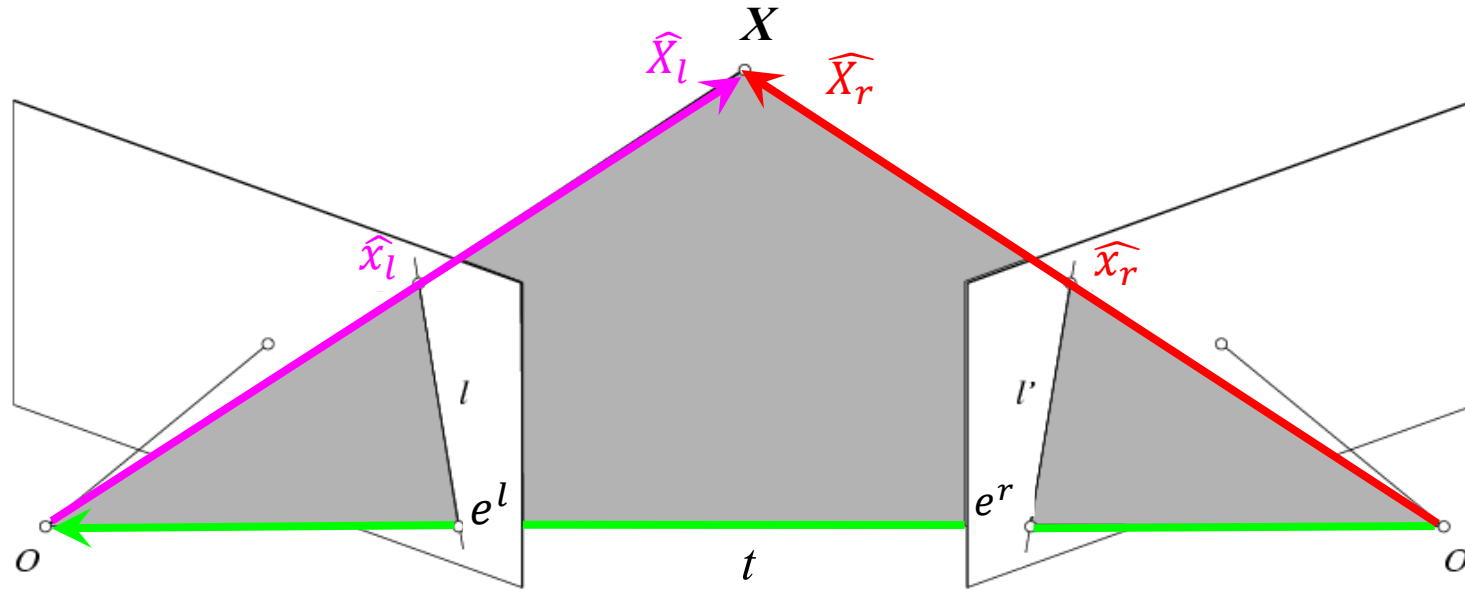
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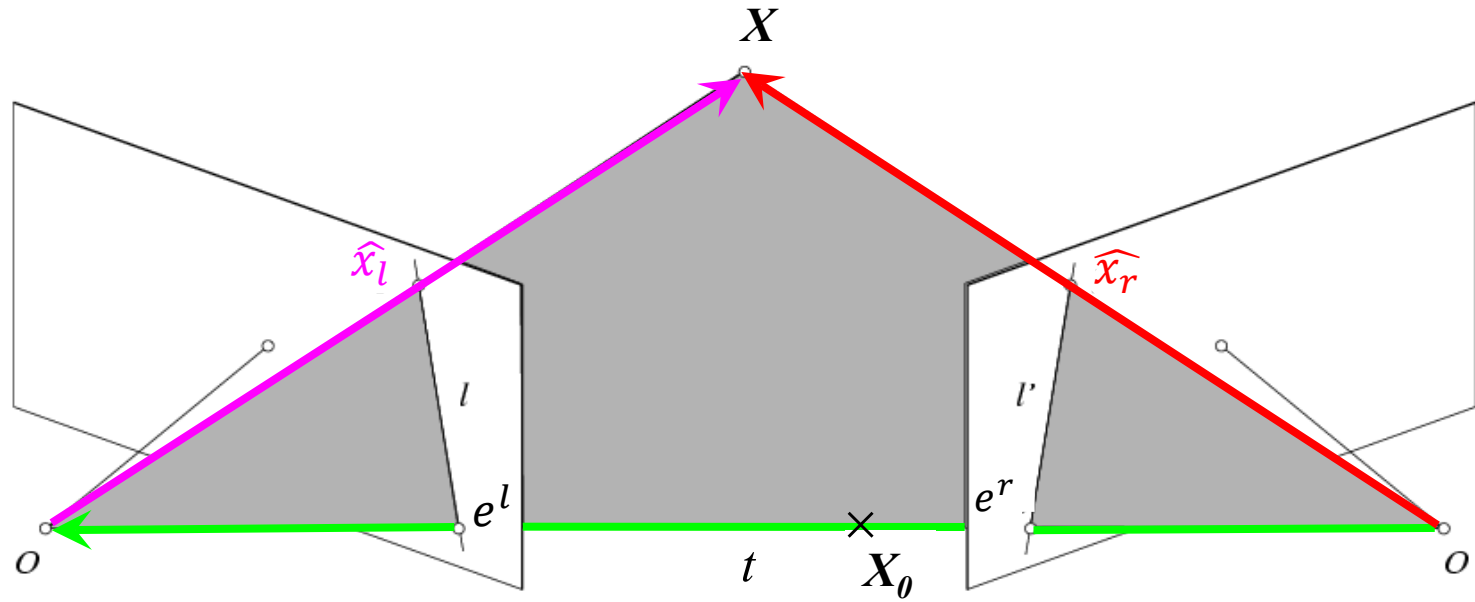
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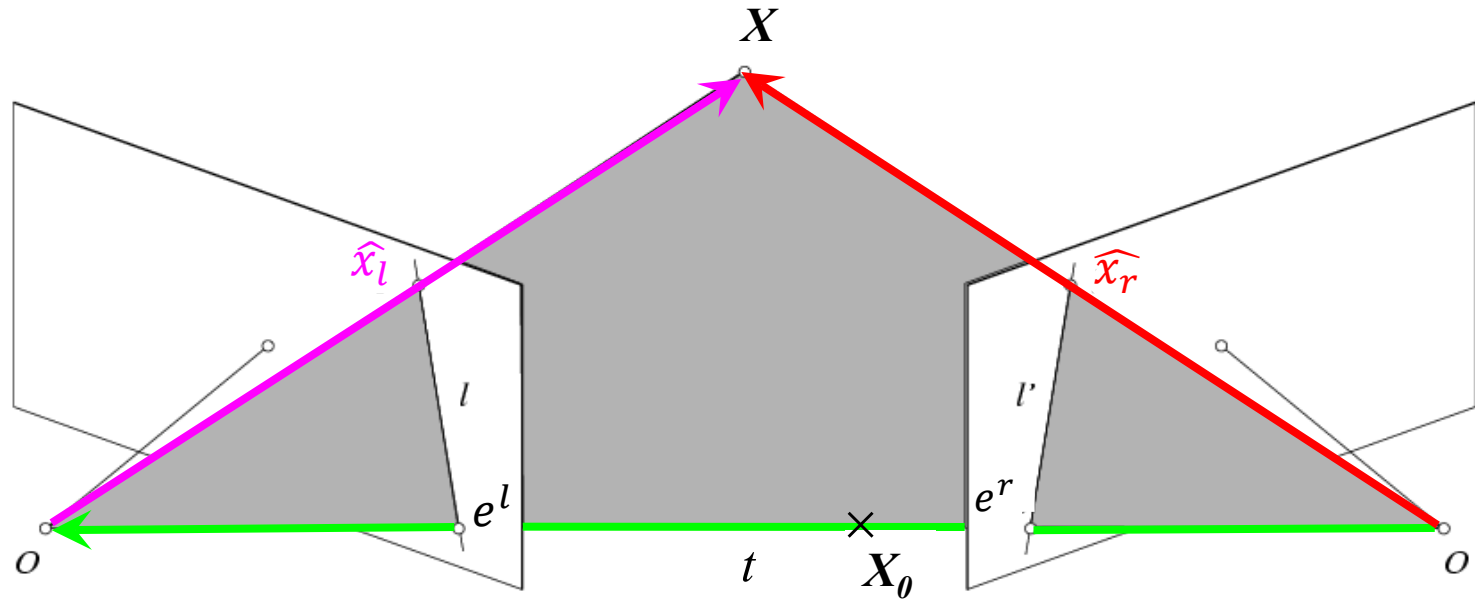
$$\hat{x}_l^T \underbrace{F \hat{x}_r}_{\text{a line}} = 0 \Rightarrow F \hat{x}_r \text{ is a line passing through } \hat{x}_l$$

# Epipoles and fundamental matrix



$$\hat{x}_l^T \underbrace{F \hat{x}_r}_{\text{A line passing through } \hat{x}_l} = 0$$

# Epipoles and fundamental matrix



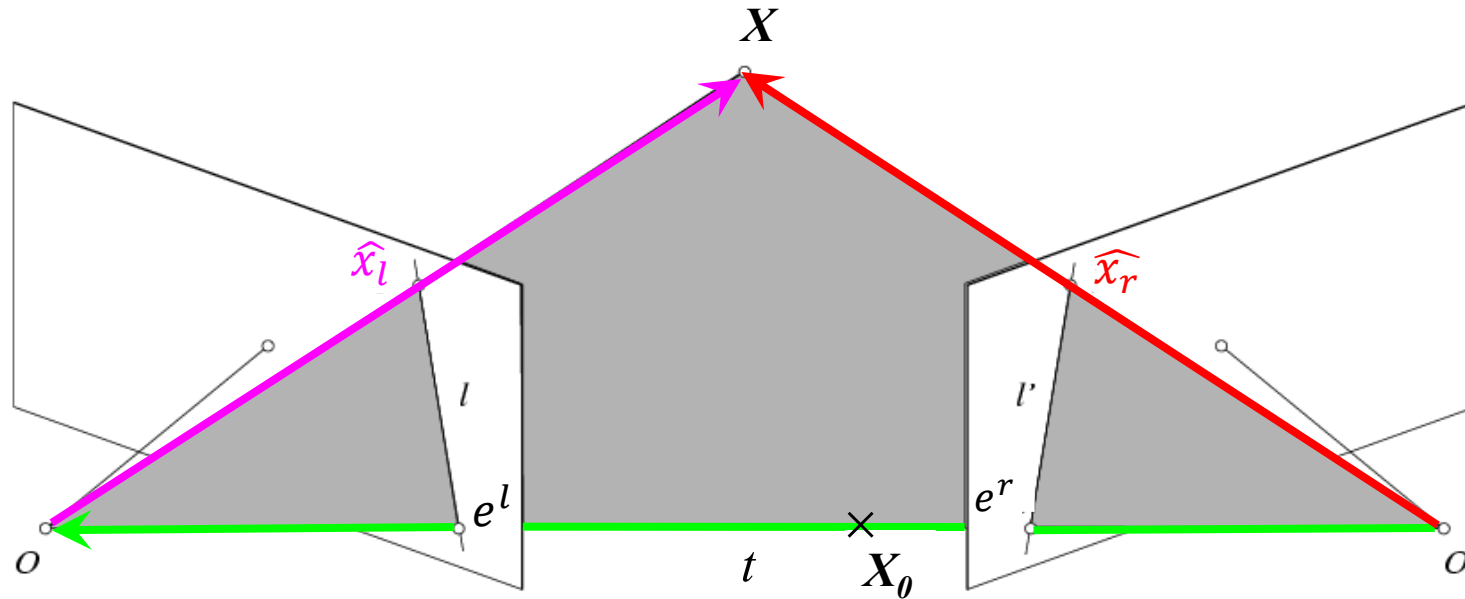
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A line passing through  $\hat{x}_l$

and recall that  $F e^r = F^T e^l = 0$



# Epipoles and fundamental matrix

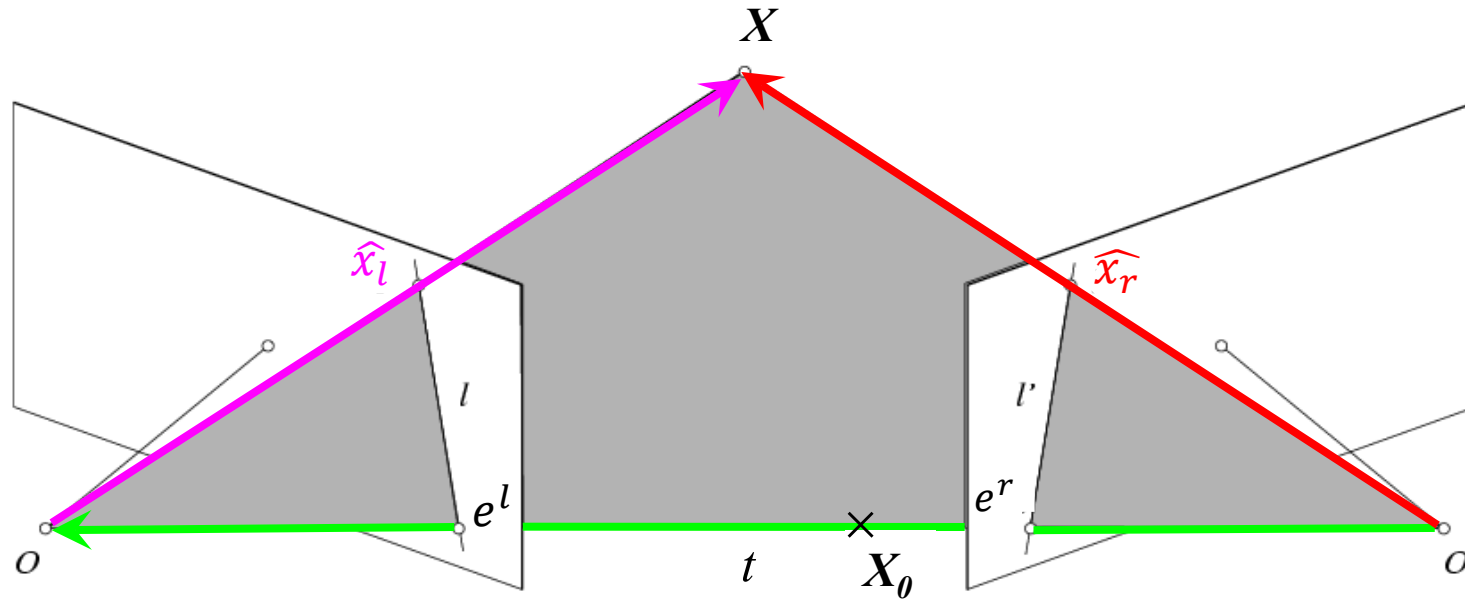


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# Epipoles and fundamental matrix

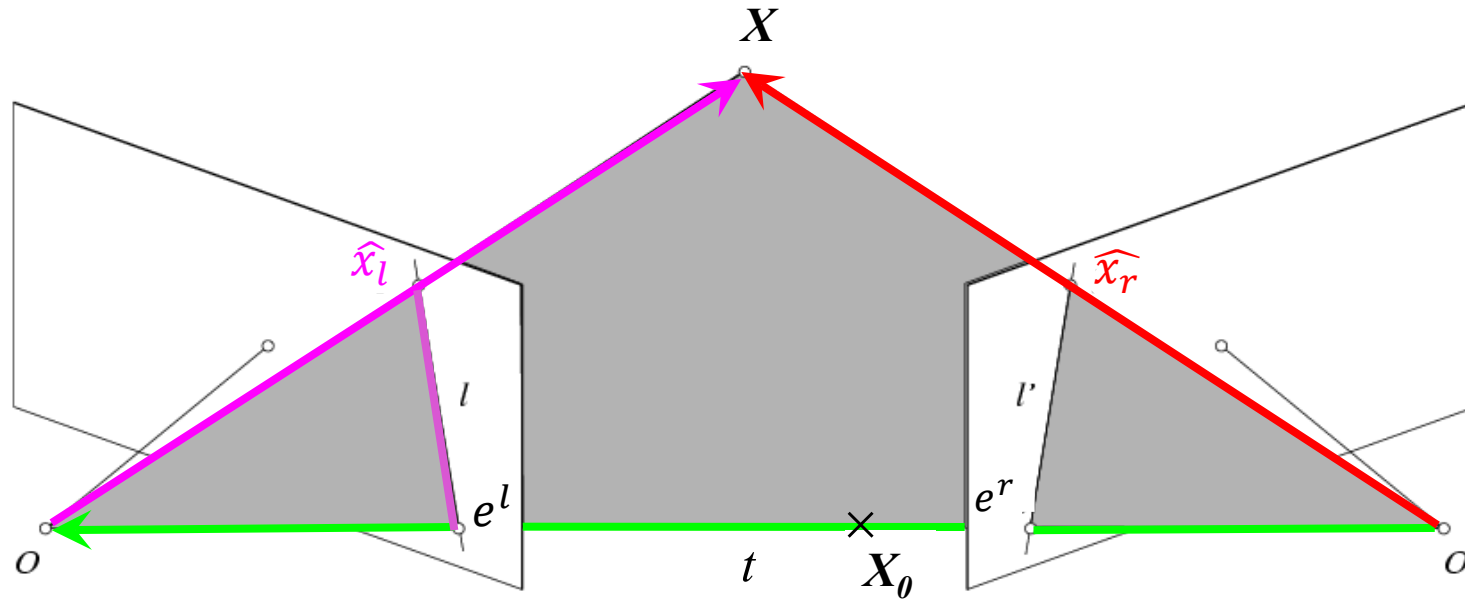


$$\hat{x}_l^T \underbrace{F \hat{x}_r}_{\text{A line passing through } \hat{x}_l} = 0 \quad \text{and recall that } F e^r = F^T e^l = 0$$

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$F \hat{x}_r$  passes through not just  $\hat{x}_l$  but also the epipole  $e$

# Epipoles and fundamental matrix



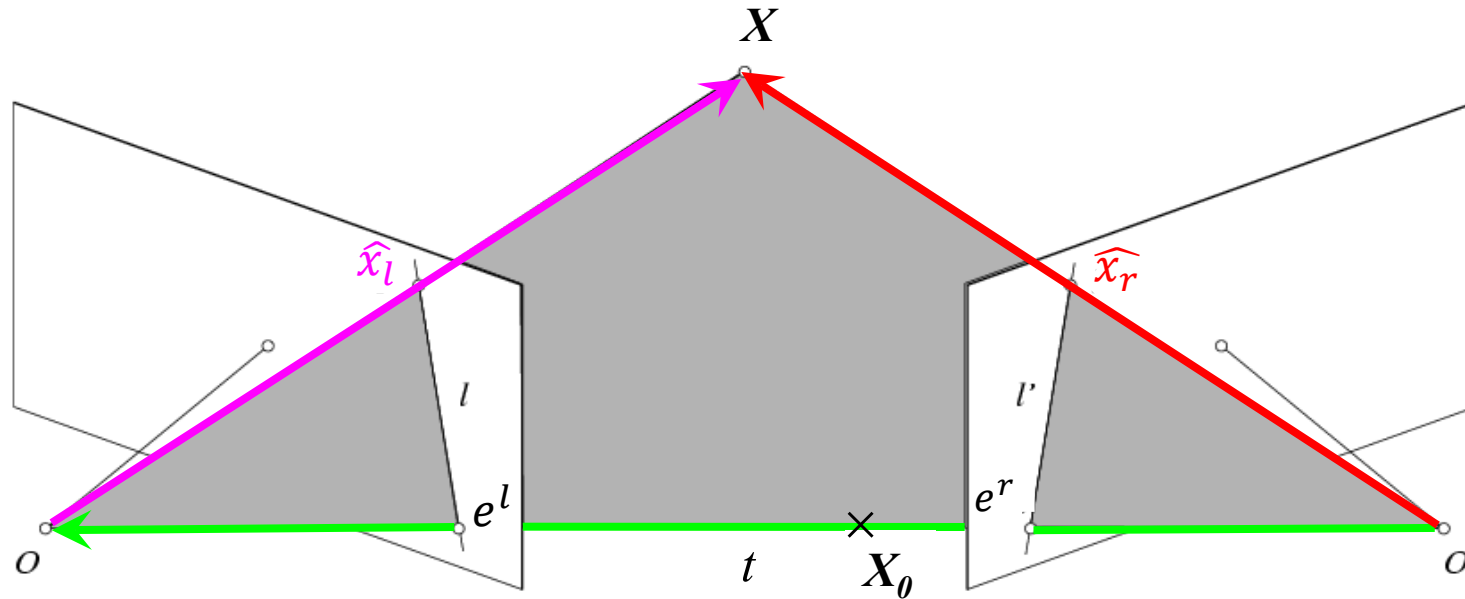
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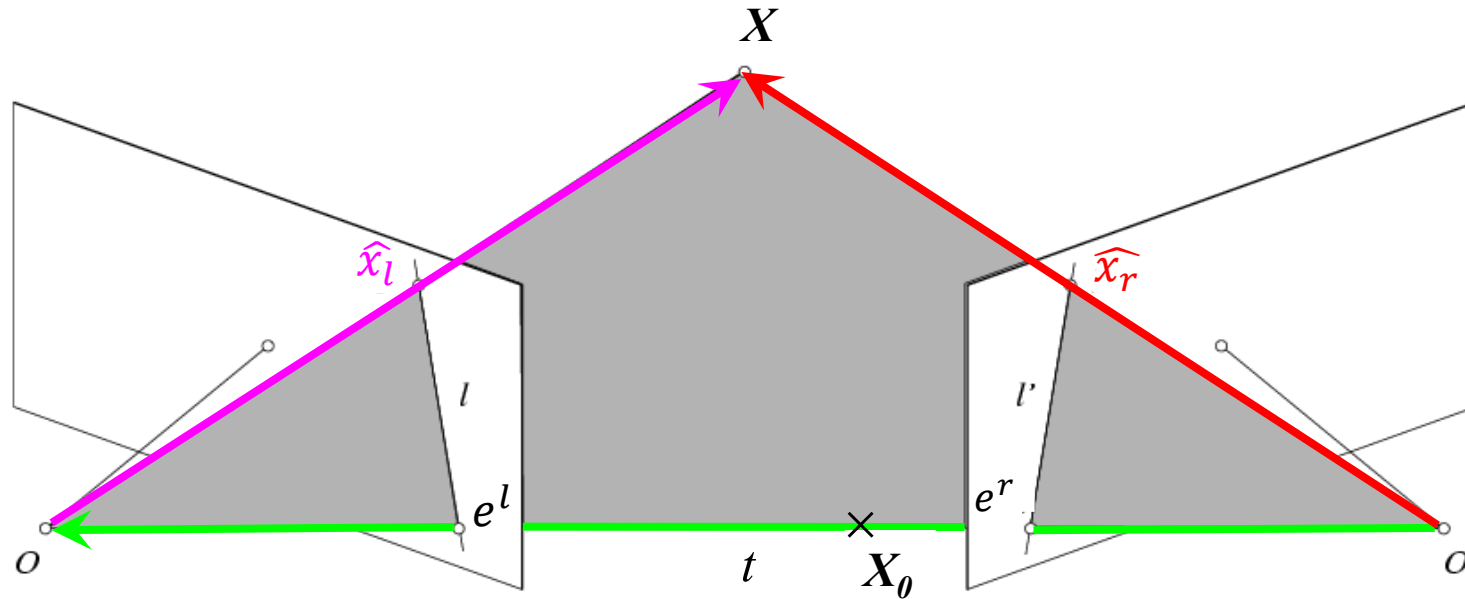
Thus, it is actually the epipolar line  $l$

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Similarly,  $\hat{x}_r^T \underbrace{F^T \hat{x}_l}_{\text{A line passing through } \hat{x}_r} = 0$  and recall that  $F e^r = F^T e^l = 0$

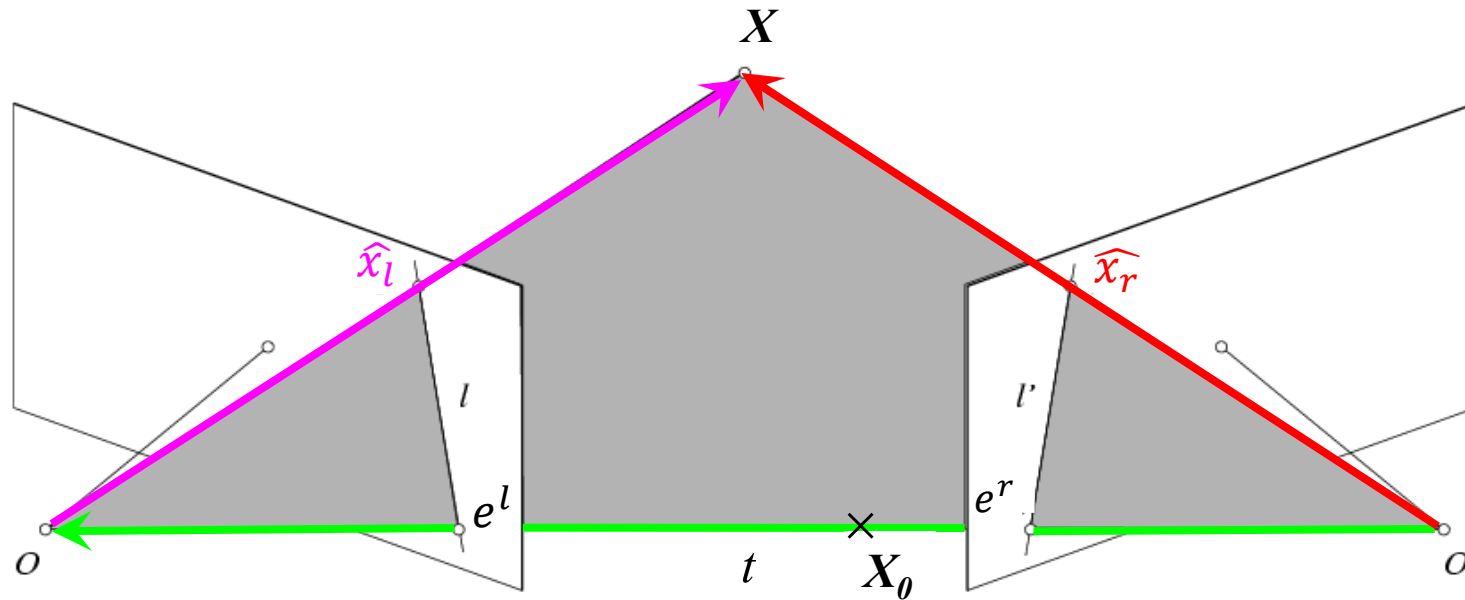
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Similarly,  $\hat{x}_r^T \underbrace{F^T \hat{x}_l}_{\text{A line passing through } \hat{x}_r} = 0$  and recall that  $F e^r = F^T e^l = 0$

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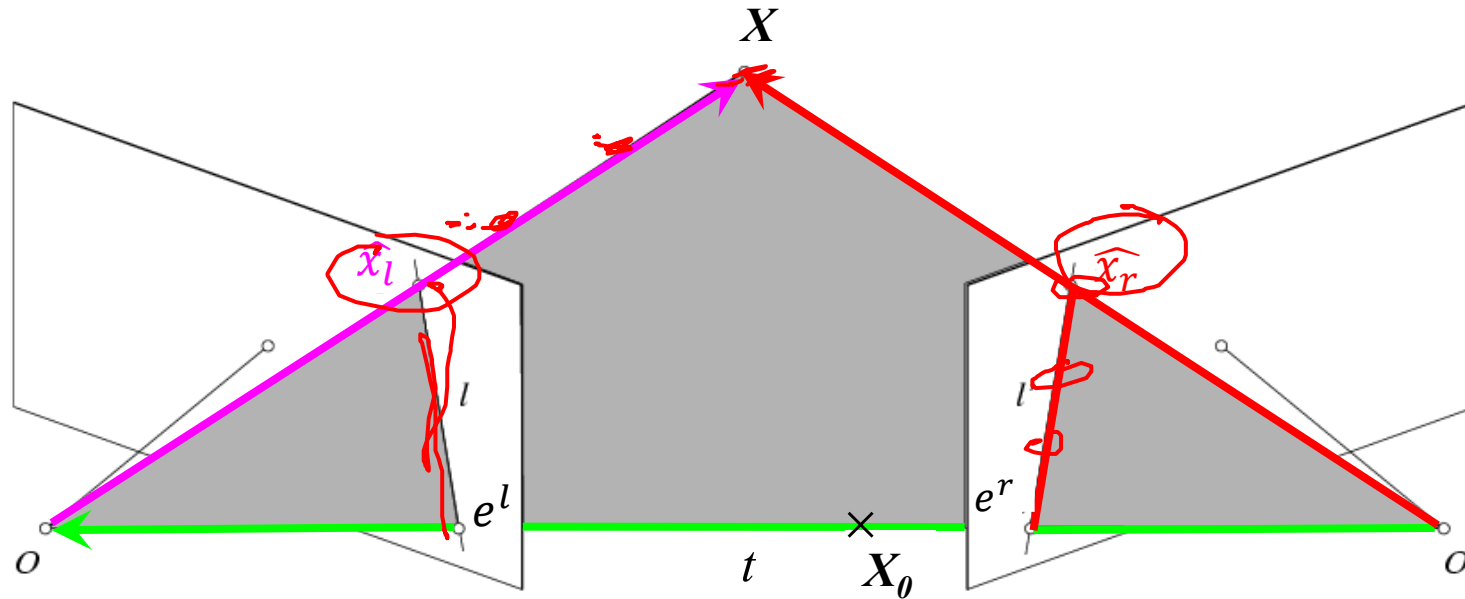


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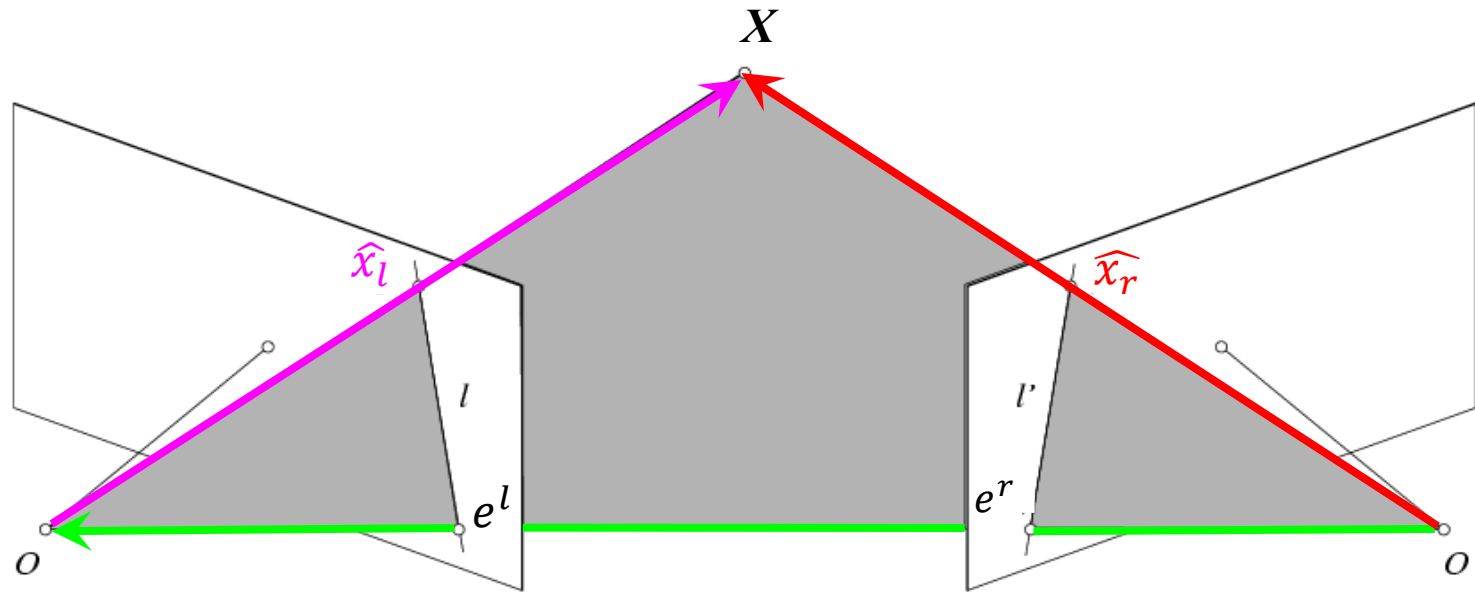
Similarly,  $\hat{x}_r^T \underbrace{F^T \hat{x}_l}_{\substack{\text{A line passing} \\ \text{through } \hat{x}_r}} = 0$  and recall that  $F e^r = F^T e^l = 0$

$$\Rightarrow e^r F^T \hat{x}_l = (F e^r)^T \hat{x}_l = 0$$

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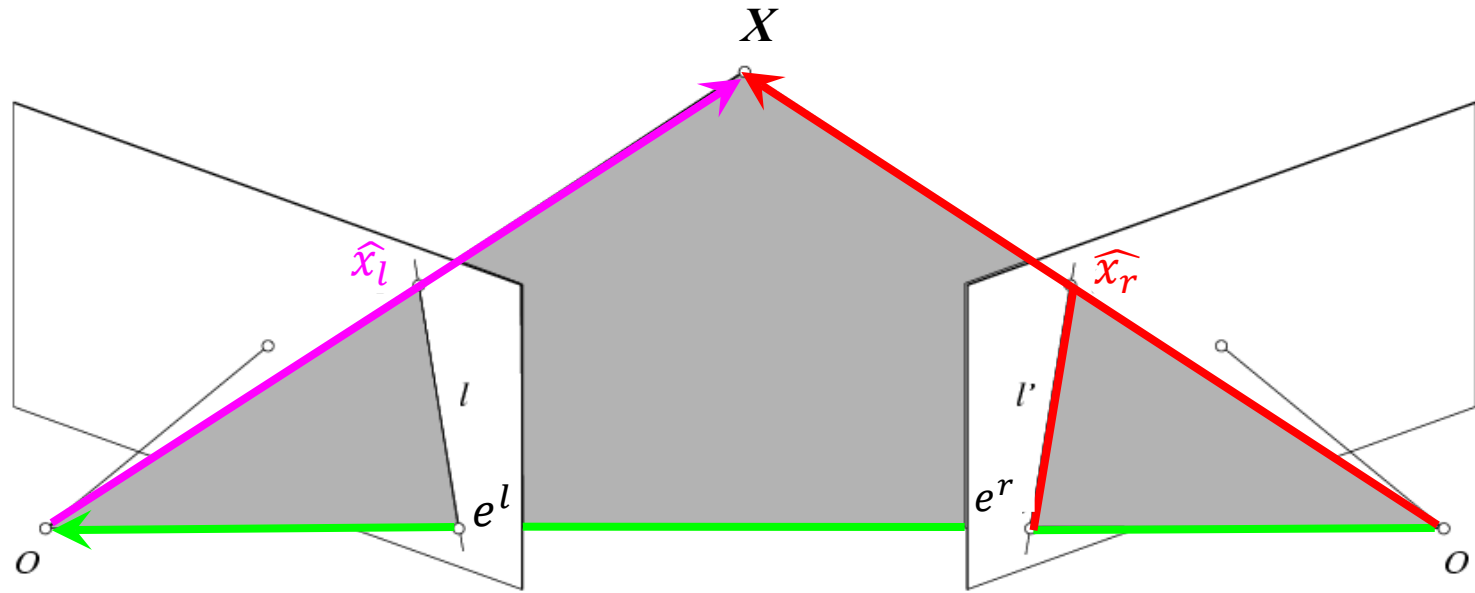
Thus, it is actually the epipolar line  $l'$

# Epipoles and fundamental matrix



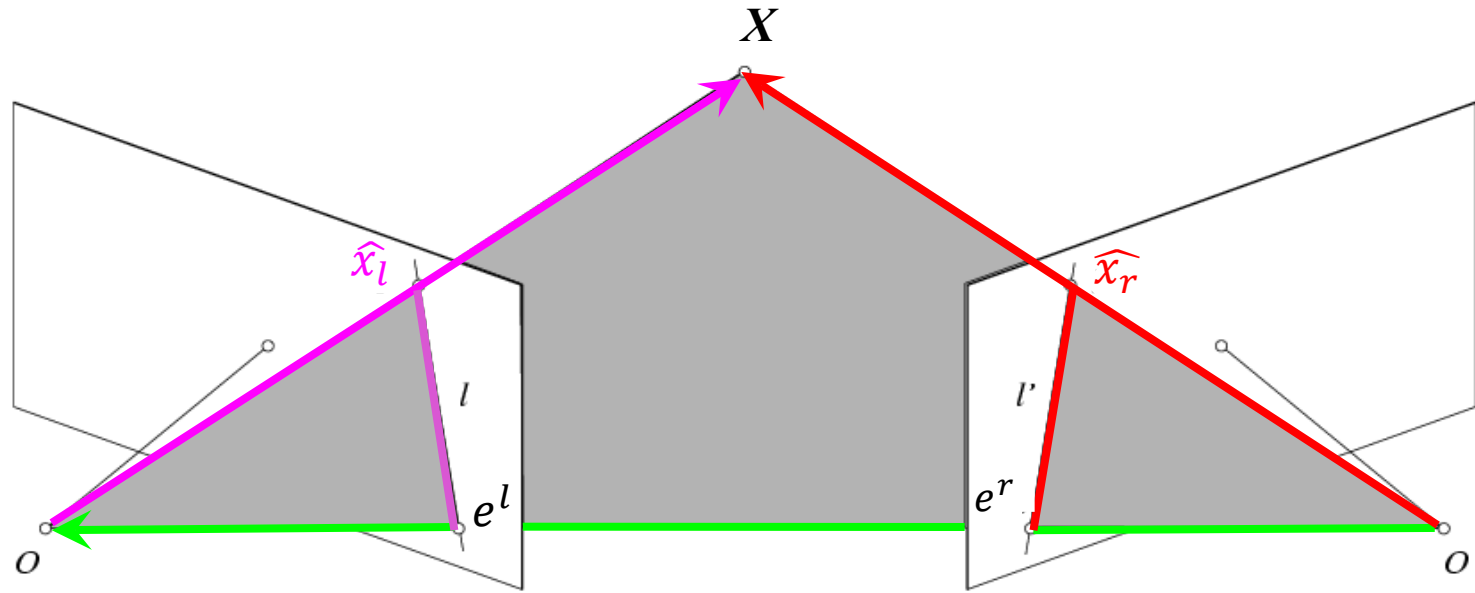


# Epipoles and fundamental matrix



$F\hat{x}_r$  is the epipolar line  $l$

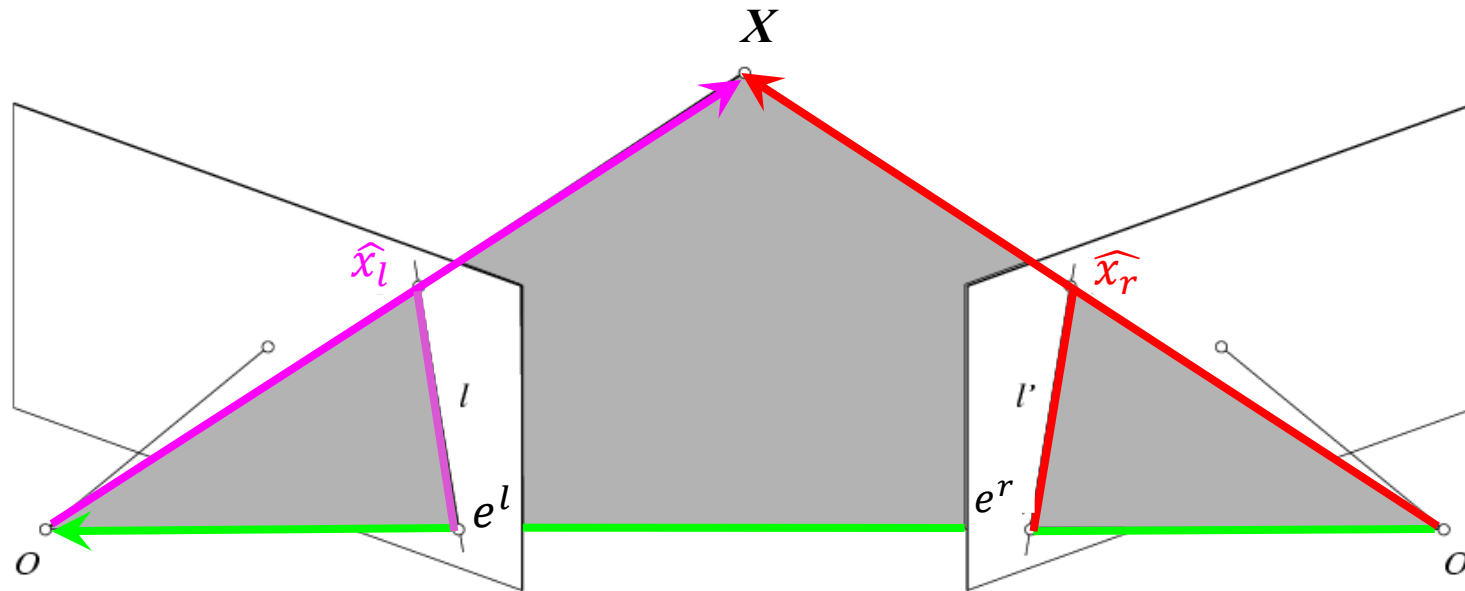
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# Epipoles and fundamental matrix

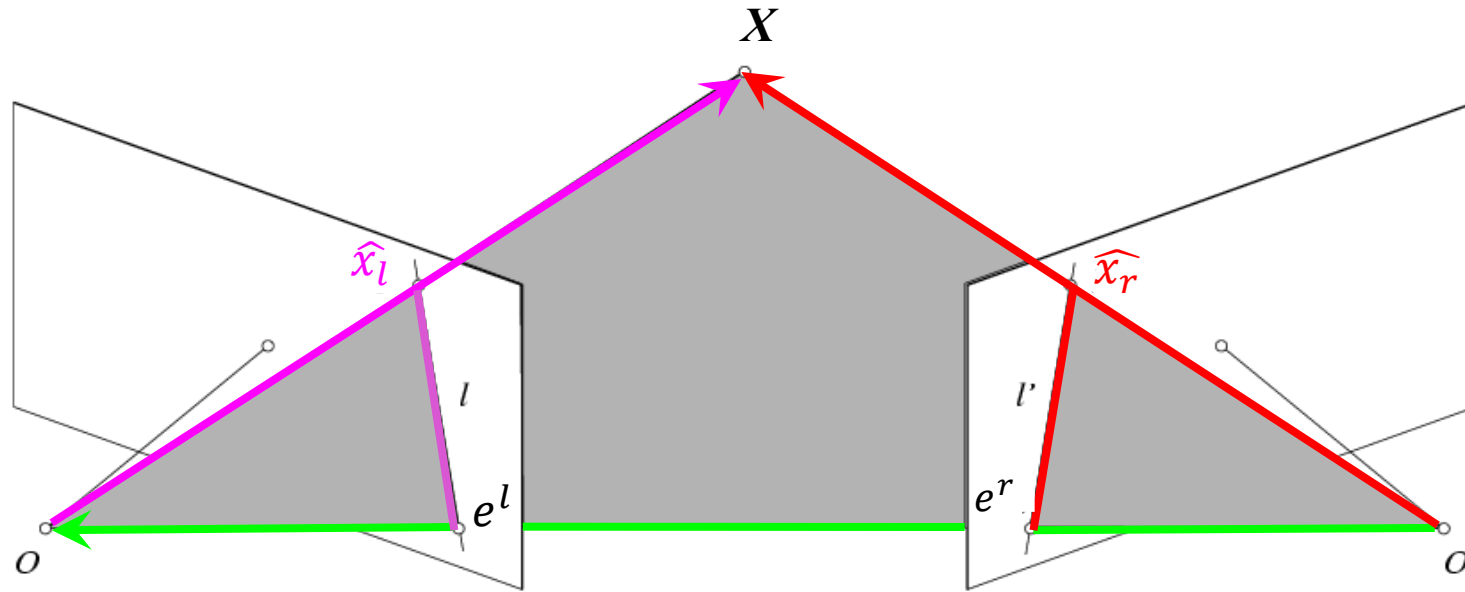


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Given  $\hat{x}_r$  on the right view, we just need to search  $F\hat{x}_r$  on the left

# Epipoles and fundamental matrix



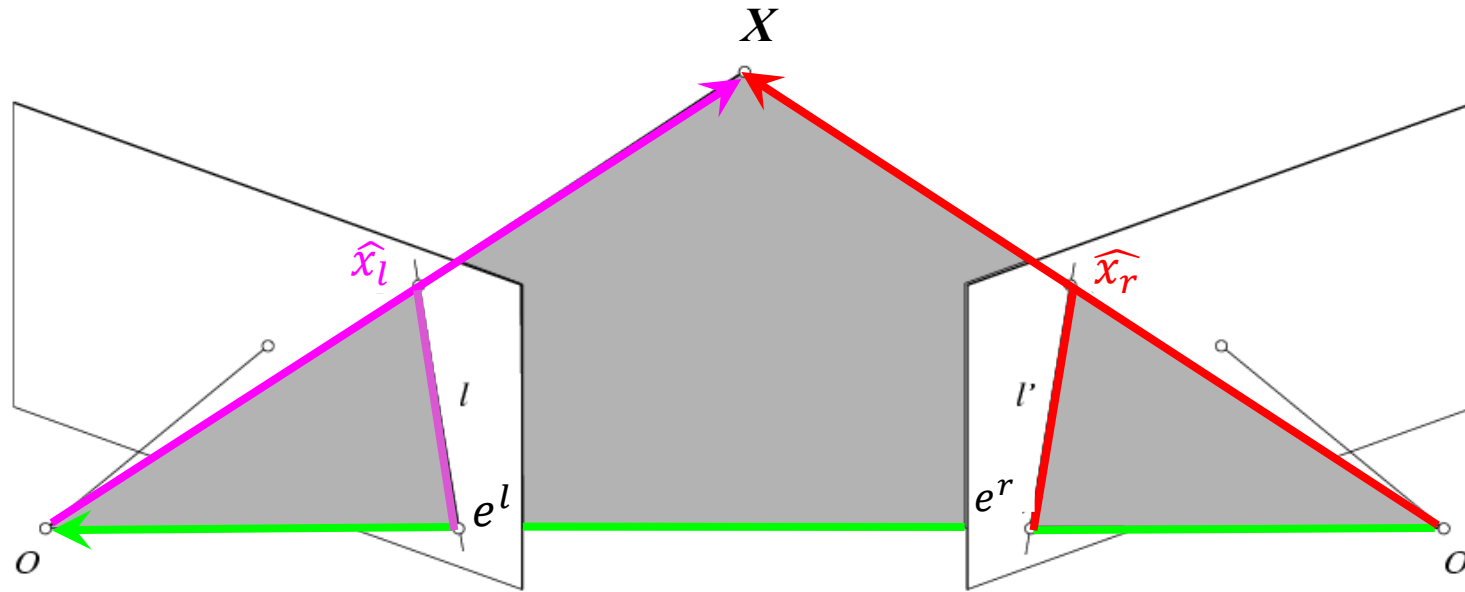
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# Epipoles and fundamental matrix



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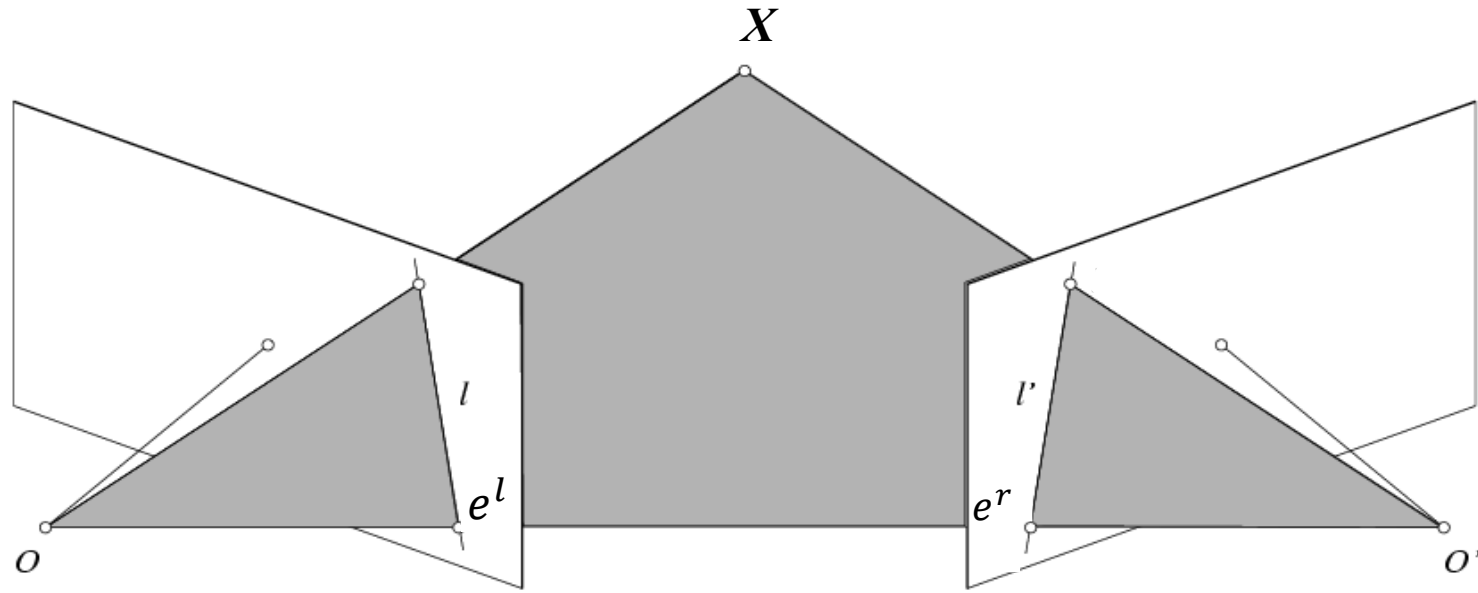
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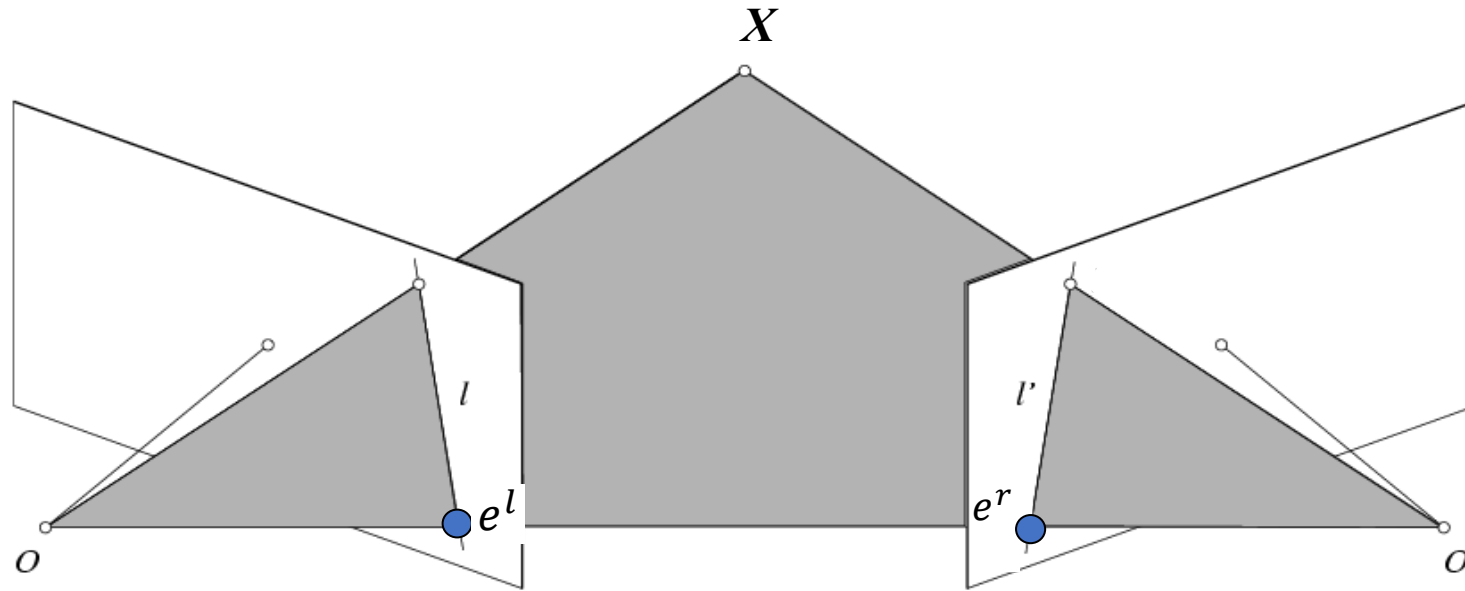
Given  $\hat{x}_l$  on the left view, we just need to search  $F^T\hat{x}_l$  on the right

**This restrict our search space from 2D to 1D**

# Fundamental matrix

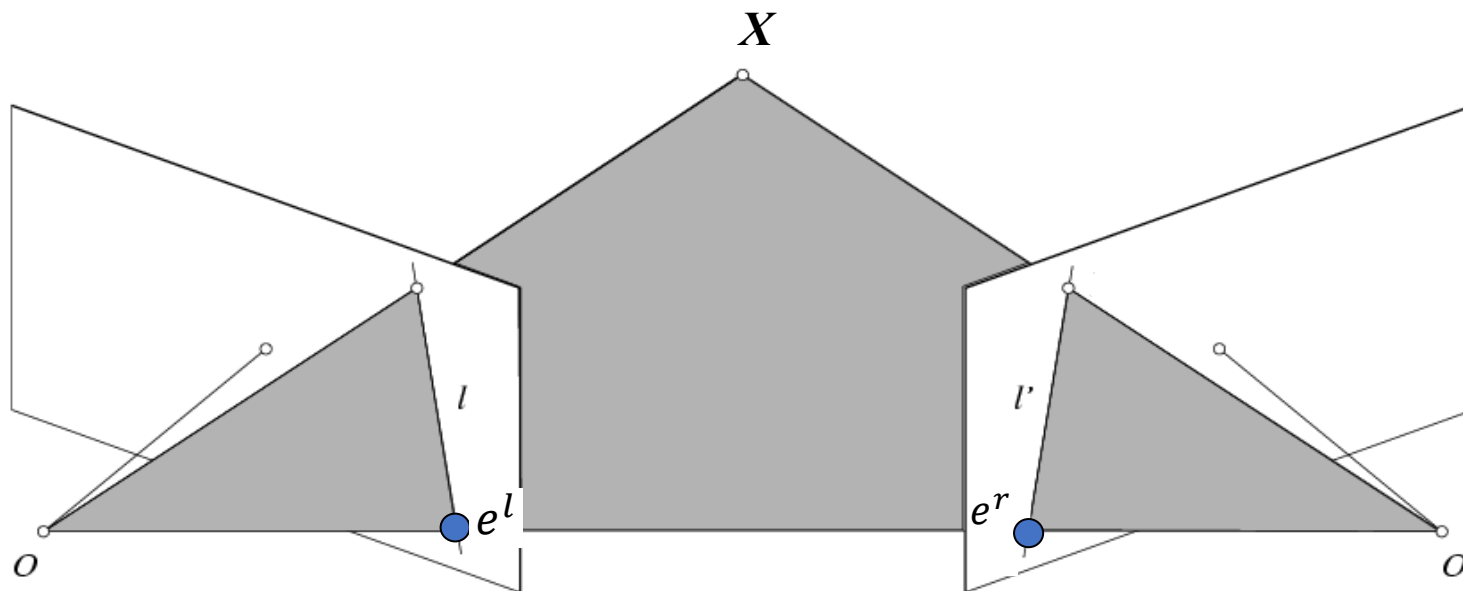


# Fundamental matrix



- $F e^r = 0$  and  $F^T e^l = 0$  (nullspaces of  $F = e^r$ ; nullspace of  $F^T = e^l$ )

# Fundamental matrix



- $F e^r = 0$  and  $F^T e^l = 0$  (nullspaces of  $F = e^r$ ; nullspace of  $F^T = e^l$ )
- $F$  is singular (rank two):  $\det(F)=0$
- $F$  has seven degrees of freedom: 9 entries but defined up to scale,  $\det(F)=0$



# Remark

- Since  $F = K_r^{-T} E K_l'^{-1}$

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- This is sometimes how essential matrix is introduced

# Estimating the Fundamental Matrix

- 8-point algorithm
  - Least squares solution using SVD on equations from 8 pairs of correspondences
  - Enforce  $\det(F)=0$  constraint using SVD on F

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**Note:** estimation of F (or E) is degenerate for a planar scene.

# 8-point algorithm

1. Solve a system of homogeneous linear equations
  - a. Write down the system of equations

$$\mathbf{x}^T F \mathbf{x}' = 0$$

$$uu'f_{11} + uv'f_{12} + uf_{13} + vu'f_{21} + vv'f_{22} + vf_{23} + u'f_{31} + v'f_{32} + f_{33} = 0$$

$$A\mathbf{f} = \begin{bmatrix} u_1u_1' & u_1v_1' & u_1 & v_1u_1' & v_1v_1' & v_1 & u_1' & v_1' & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ u_nu_n' & u_nv_n' & u_n & v_nu_n' & v_nv_n' & v_n & u_n' & v_n' & 1 \end{bmatrix} \begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ \vdots \\ f_{33} \end{bmatrix} = \mathbf{0}$$

# 8-point algorithm

1. Solve a system of homogeneous linear equations
  - a. Write down the system of equations
  - b. Solve  $\mathbf{f}$  from  $\mathbf{A}\mathbf{f}=\mathbf{0}$  using SVD

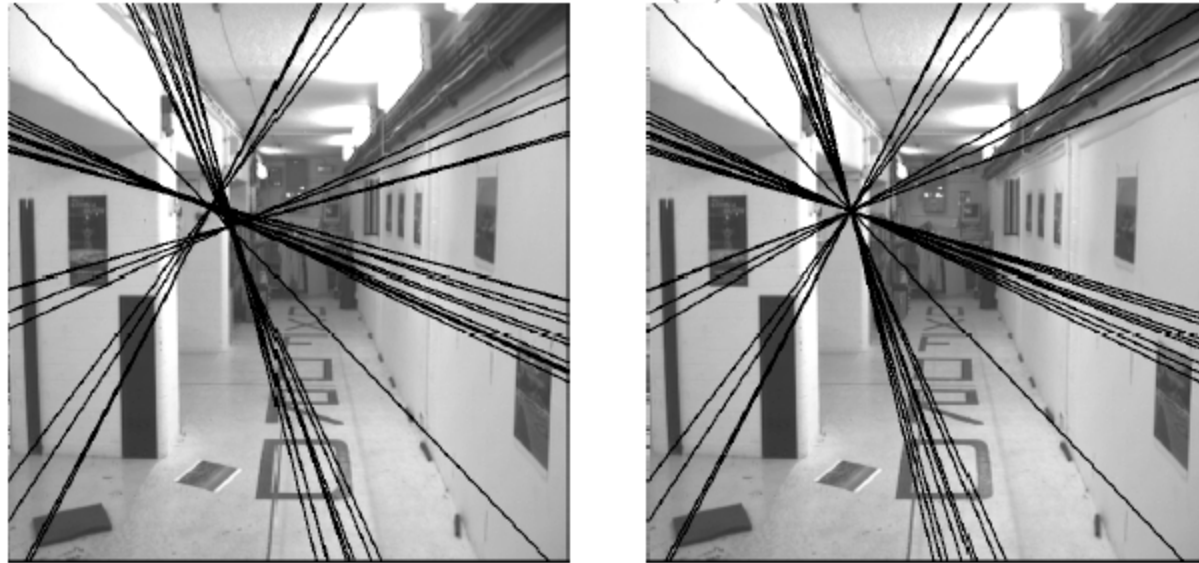
Matlab:

```
[U, S, V] = svd(A);  
f = V(:, end);  
F = reshape(f, [3 3])';
```



# Need to enforce singularity constraint

Fundamental matrix has rank 2 :  $\det(\mathbf{F}) = 0$ .



**Left :** Uncorrected  $\mathbf{F}$  – epipolar lines are not coincident.

**Right :** Epipolar lines from corrected  $\mathbf{F}$ .

# 8-point algorithm

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f = V(:, end);  
F = reshape(f, [3 3])';
```

2. Resolve  $\det(\mathbf{F}) = 0$  constraint using SVD

Matlab:

```
[U, S, V] = svd(F);  
S(3,3) = 0;  
F = U*S*V';
```

# Problem with eight-point algorithm

$$\begin{bmatrix} u'u & u'v & u' & v'u & v'v & v' & u & v \end{bmatrix} \begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ f_{22} \\ f_{23} \\ f_{31} \\ f_{32} \end{bmatrix} = -1$$

# Problem with eight-point algorithm

250906.36	183269.57	921.81	200931.10	146766.13	738.21	272.19	198.81
2692.28	131633.03	176.27	6196.73	302975.59	405.71	15.27	746.79
416374.23	871684.30	935.47	408110.89	854384.92	916.90	445.10	931.81
191183.60	171759.40	410.27	416435.62	374125.90	893.65	465.99	418.65
48988.86	30401.76	57.89	298604.57	185309.58	352.87	846.22	525.15
164786.04	546559.67	813.17	1998.37	6628.15	9.86	202.65	672.14
116407.01	2727.75	138.89	169941.27	3982.21	202.77	838.12	19.64
135384.58	75411.13	198.72	411350.03	229127.78	603.79	681.28	379.48

$$\begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ f_{22} \\ f_{23} \\ f_{31} \\ f_{32} \end{bmatrix} = -\mathbf{1}$$

# Problem with eight-point algorithm

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48988.86	30401.76	57.89	298604.57	185309.58	352.87	846.22	525.15
164786.04	546559.67	813.17	1998.37	6628.15	9.86	202.65	672.14
116407.01	2727.75	138.89	169941.27	3982.21	202.77	838.12	19.64
135384.58	75411.13	198.72	411350.03	229127.78	603.79	681.28	379.48

$$\begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ f_{22} \\ f_{23} \\ f_{31} \\ f_{32} \end{bmatrix} = -\mathbf{1}$$

- Poor numerical conditioning
- Can be fixed by rescaling the data

# Problem with eight-point algorithm

250906.36	183269.57	921.81	200931.10	146766.13	738.21	272.19	198.81
2692.28	131633.03	176.27	6196.73	302975.59	405.71	15.27	746.79
416374.23	871684.30	935.47	408110.89	854384.92	916.90	445.10	931.81
191183.60	171759.40	410.27	416435.62	374125.90	893.65	465.99	418.65
48988.86	30401.76	57.89	298604.57	185309.58	352.87	846.22	525.15
164786.04	546559.67	813.17	1998.37	6628.15	9.86	202.65	672.14
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- If we know the calibration matrices of the two cameras, we can estimate the essential matrix:  $E = K^T F K'$
- The essential matrix gives us the relative rotation and translation between the cameras, or their extrinsic parameters.

# From $E$ to get back $R$ and $t$

- If we decompose  $E$  using svd to  $USV^T$
- Let's define  $W = \begin{pmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
- One can verify that  $[t]_{\times} = VWSV^T$  and  $R = UW^TV^T$  is a pair of valid estimate of  $R$  and  $t$  such that  $R[t]_{\times} = UW^TV^TVWSV^T = USV^T = E$ 
  - How do we get  $t$  back from  $[t]_{\times}$ ?
- Note that there are more solutions
  - E.g., Replace  $W$  by  $W^T$
  - Note that only one solution will satisfy chirality (reconstructed points in front of the camera)

# Compute 3D point location

- Given  $R$ ,  $t$ , and observed points in the two views. We can construct two lines that should contain the 3D point. And the point should be just the interception
- Denote two lines as  $a_1 + \lambda_1 b_1$  and  $a_2 + \lambda_2 b_2$ . We can solve the interception by solving

- $\min_{\lambda, \mu} \|a_1 + \lambda_1 b_1 - a_2 - \lambda_2 b_2\|^2$

$$\Rightarrow \begin{cases} \lambda_1 \|b_1\|^2 + (a_1 - a_2)^\top b_1 - \lambda_2 b_1^\top b_2 = 0 \\ \lambda_2 \|b_2\|^2 - (a_1 - a_2)^\top b_2 - \lambda_1 b_1^\top b_2 = 0 \end{cases}$$

- The final point will be

$$\frac{a_1 + \lambda_1 b_1 + a_2 + \lambda_2 b_2}{2}$$

# Caveat

- Note that many textbooks use the convention of  $E = [t]_{\times}R$  rather than  $E = R[t]_{\times}$  as derived here
  - Both are correct just they first rotate and then translate the axis rather than first translate and then rotate the axis as shown earlier
  - But consequently, the  $t$  there is different as below
- Note that for any  $y$ ,  $[R t]_{\times}y = R ([t]_{\times}R^{\top}y)$  since crossing  $R t$  with  $y$  is the same as crossing  $t$  with  $R^{\top}y$  and then rotate by  $R$ 
  - $\Rightarrow [R t]_{\times} = R [t]_{\times}R^{\top}$
  - $\Rightarrow E = R[t]_{\times} = (R[t]_{\times}R^{\top})R = [Rt]_{\times}R$
  - Therefore, for the alternative formulation  $E = [t]_{\times}R$ , we have the translation there actually equal to  $Rt$  instead in our notations

# Summary

- $X_l^T E X_r = 0$
- $E = R[t]_{\times}$
- $F e^r = F^T e^l = 0$
- $\hat{x}_l^T F \hat{x}_r = 0$
- $F^T \hat{x}_l = l'$
- $F \hat{x}_r = l$
- Estimating the fundamental matrix is a kind of “weak calibration”

