

# CSE 152: Computer Vision

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## Lecture 15: Fundamental Matrix



# Agenda

- Why is stereo useful?
- Epipolar constraints
- **Fundamental matrix**

# Cross Product as Matrix Multiplication

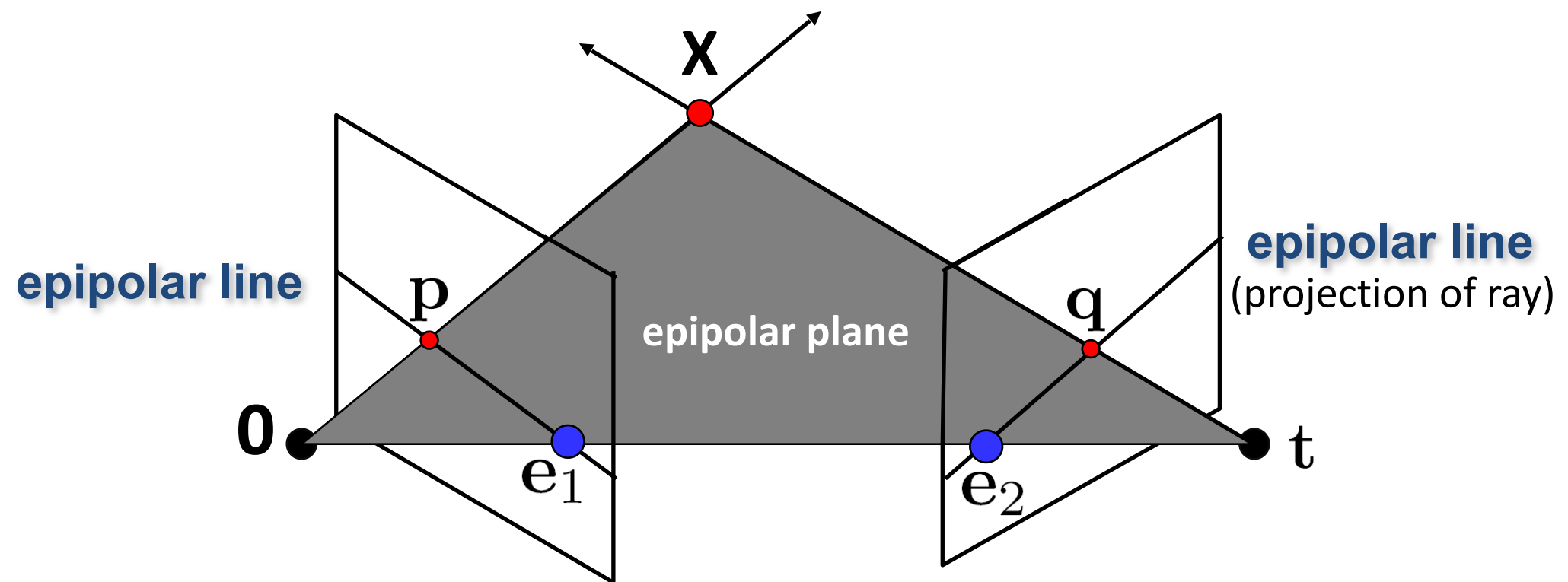
$$\mathbf{a} \times \mathbf{b} = \begin{bmatrix} 0 & -a_z & a_y \\ a_z & 0 & -a_x \\ -a_y & a_x & 0 \end{bmatrix} \begin{bmatrix} b_x \\ b_y \\ b_z \end{bmatrix} = [\mathbf{a}_\times] \mathbf{b}$$

$$[\mathbf{a}_\times] = -[\mathbf{a}_\times]^T$$

“skew-symmetric matrix”

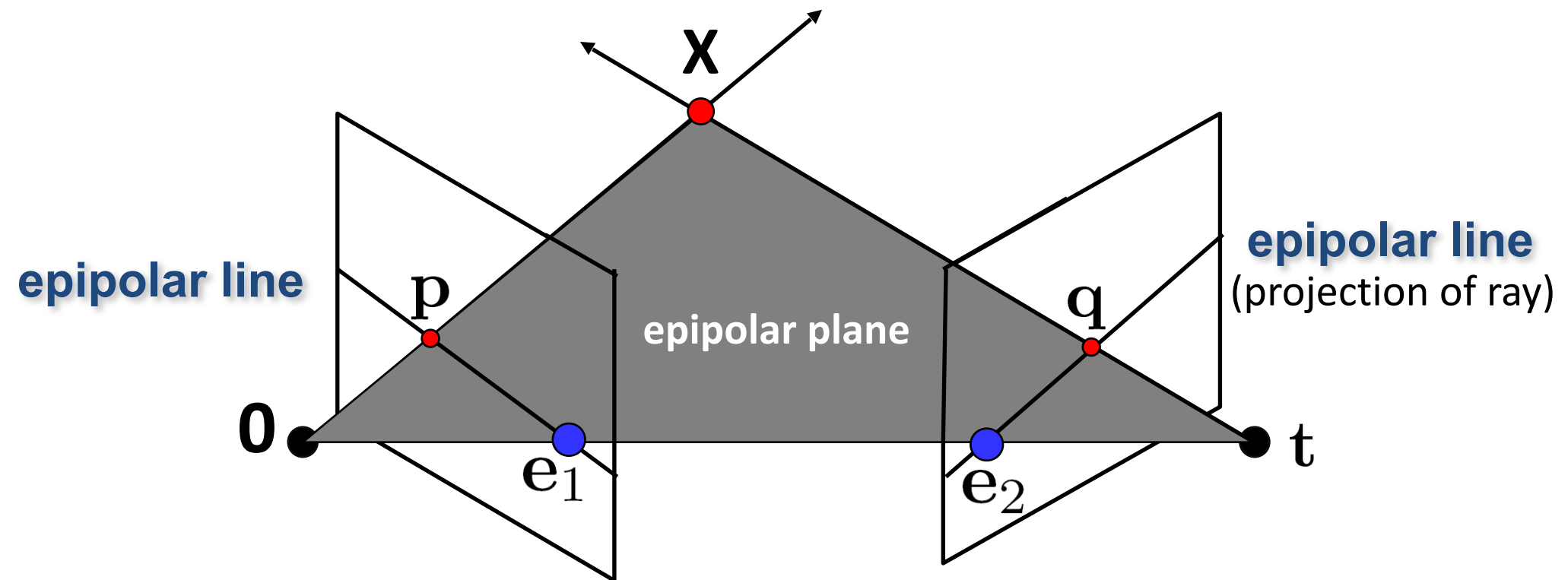
rank 2

# Essential Matrix



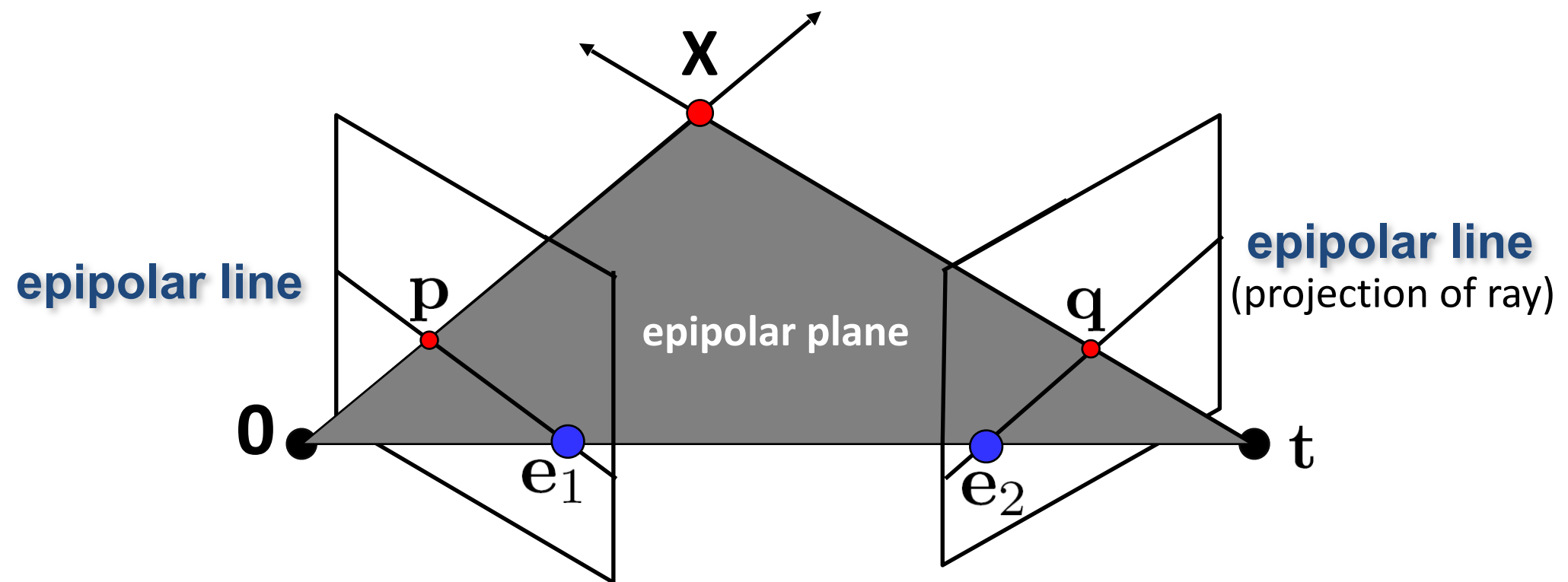
- Assume  $p$  and  $q$  in  $\mathbb{R}^3$  are two points on the (virtual) image plane of two cameras
- Denoted by the pinhole frame coordinate in the corresponding cameras

# Essential Matrix



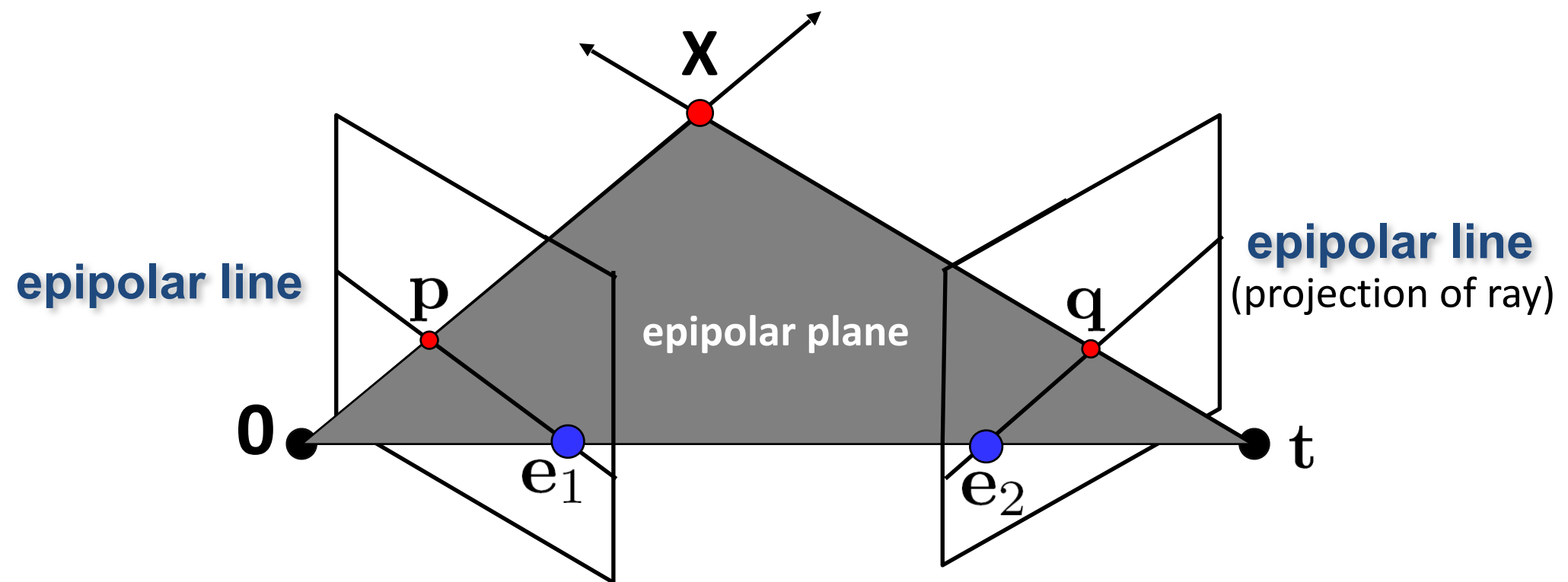
- Let camera 1 be  $[I, 0]$  and camera 2 be  $[R, t]$ .

# Essential Matrix



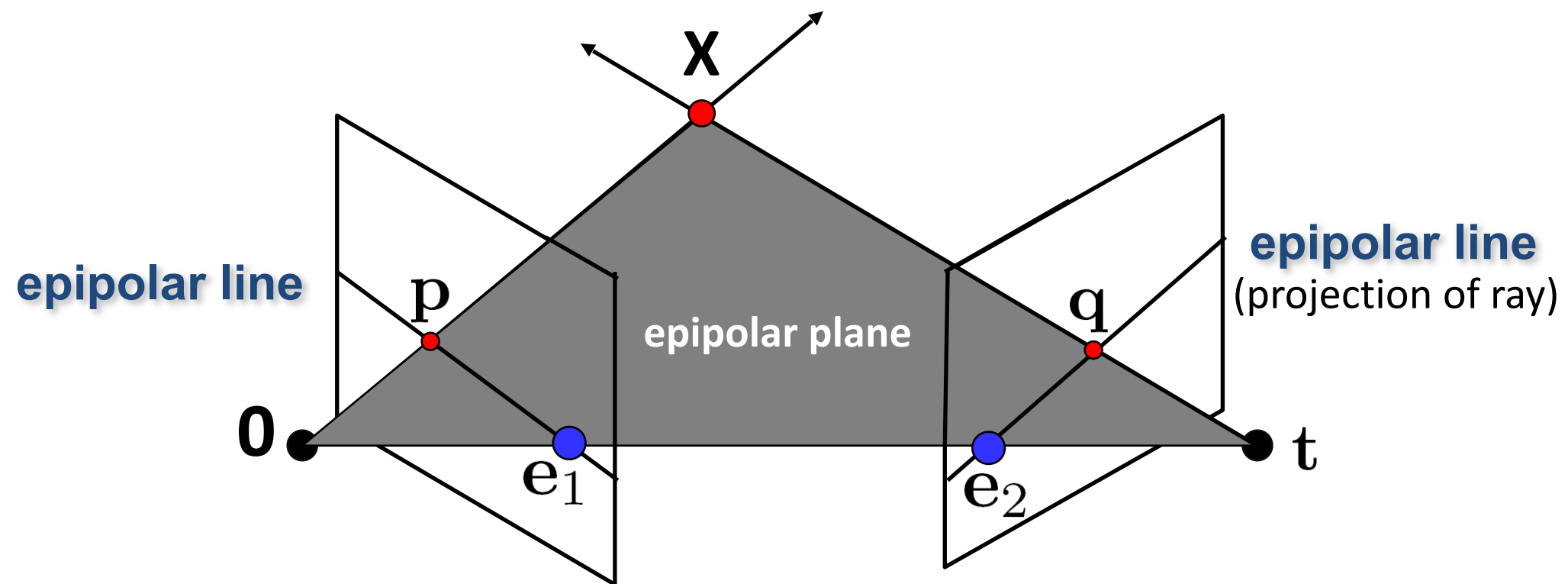
- Let camera 1 be  $[\mathbf{I}, \mathbf{0}]$  and camera 2 be  $[\mathbf{R}, \mathbf{t}]$ .
- In camera 1 pinhole frame, 3D point  $\mathbf{X}$  is given by  $\mathbf{X}_1 = \lambda_1 \mathbf{p}$

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- In camera 1 pinhole frame, 3D point  $\mathbf{X}$  is given by  $\mathbf{X}_1 = \lambda_1 \mathbf{p}$
- In camera 2 pinhole frame, 3D point  $\mathbf{X}$  is given by  $\mathbf{X}_2 = \lambda_2 \mathbf{q}$

# Essential Matrix

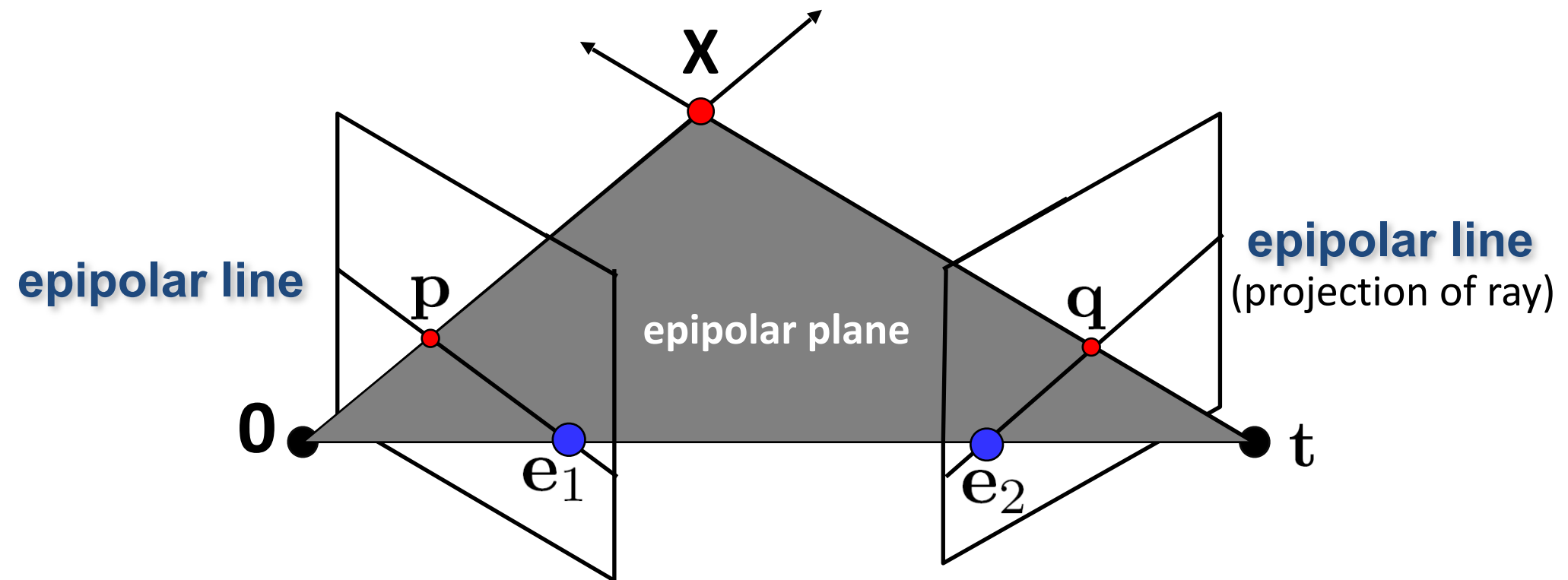


- Let camera 1 be  $[I, \mathbf{0}]$  and camera 2 be  $[R, \mathbf{t}]$ .
- In camera 1 pinhole frame, 3D point  $\mathbf{X}$  is given by  $\mathbf{X}_1 = \lambda_1 \mathbf{p}$
- In camera 2 pinhole frame, 3D point  $\mathbf{X}$  is given by  $\mathbf{X}_2 = \lambda_2 \mathbf{q}$
- Since camera 2 is related to camera 1 by rigid-body motion  $[R, \mathbf{t}]$

$$\begin{aligned} X_1 &= RX_2 + t \\ \lambda_1 p &= \lambda_2 Rq + t \end{aligned} \quad \text{(change of coordinate system)}$$

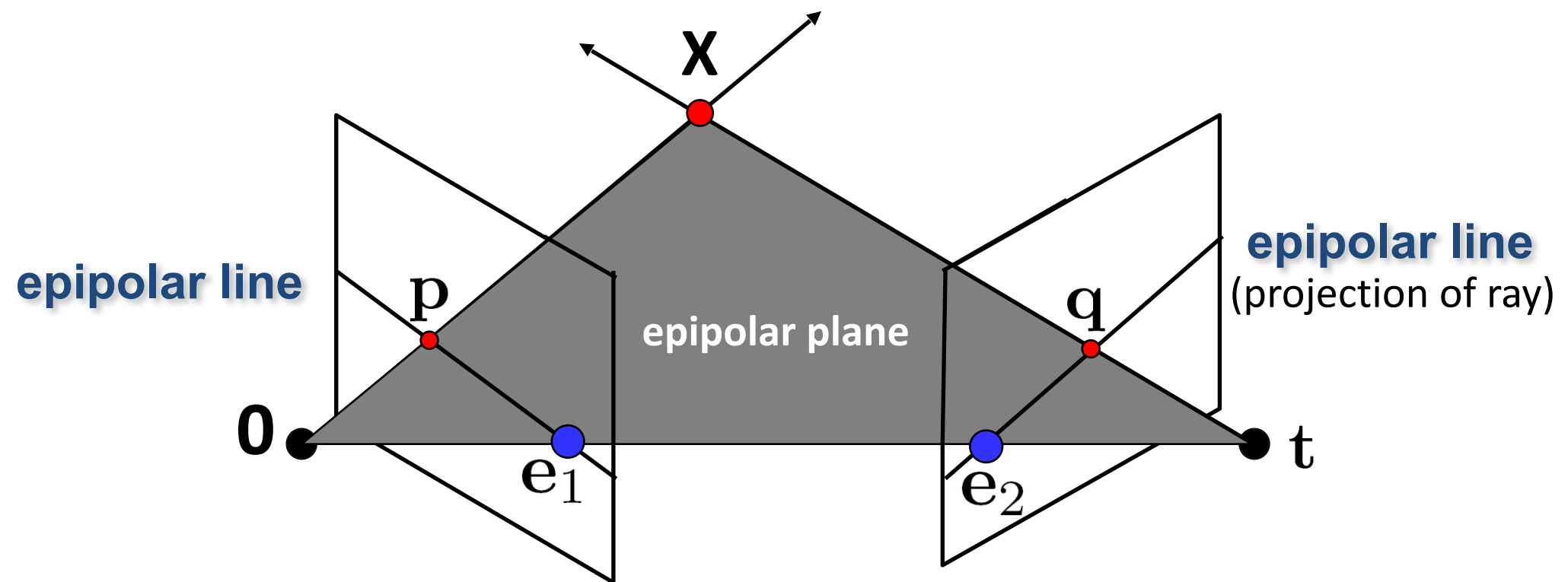


# Essential Matrix



- We have:  $\lambda_1 p = \lambda_2 Rq + t$

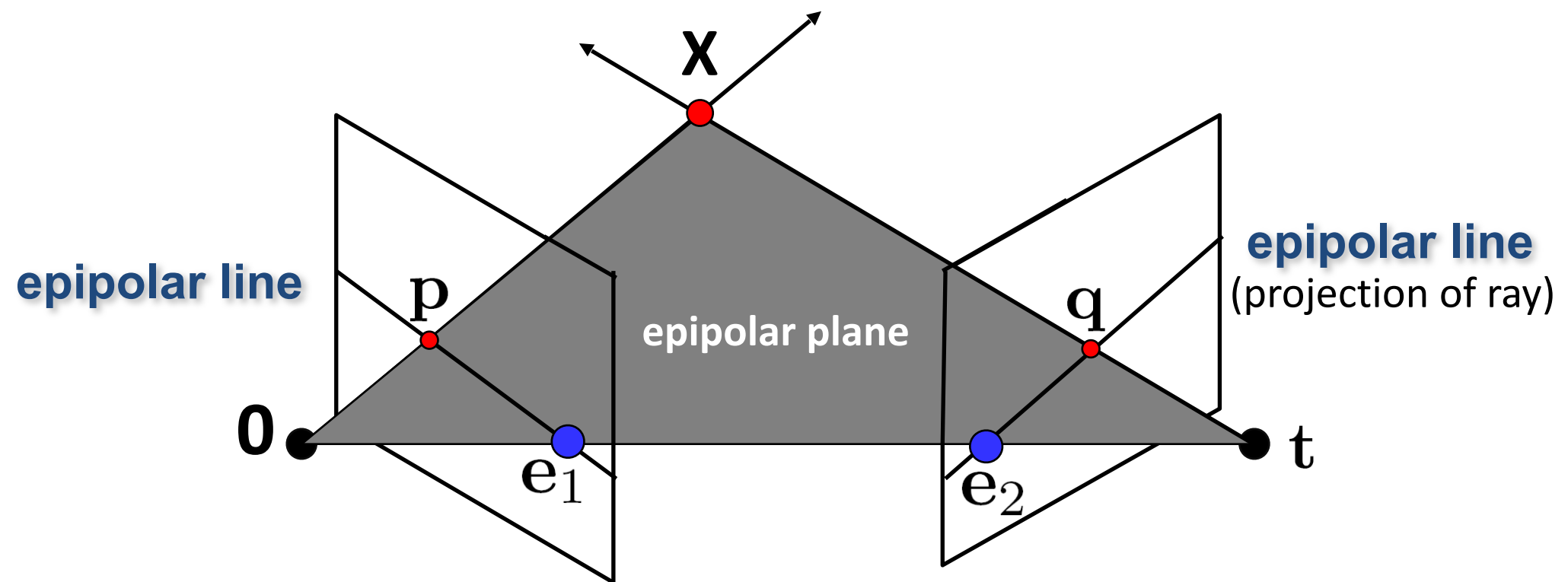
# Essential Matrix



- We have:  $\lambda_1 p = \lambda_2 Rq + t$
- Take cross-product with respect to  $t$ :

$$\lambda_1 [t]_{\times} p = \lambda_2 [t]_{\times} (Rq + t)$$

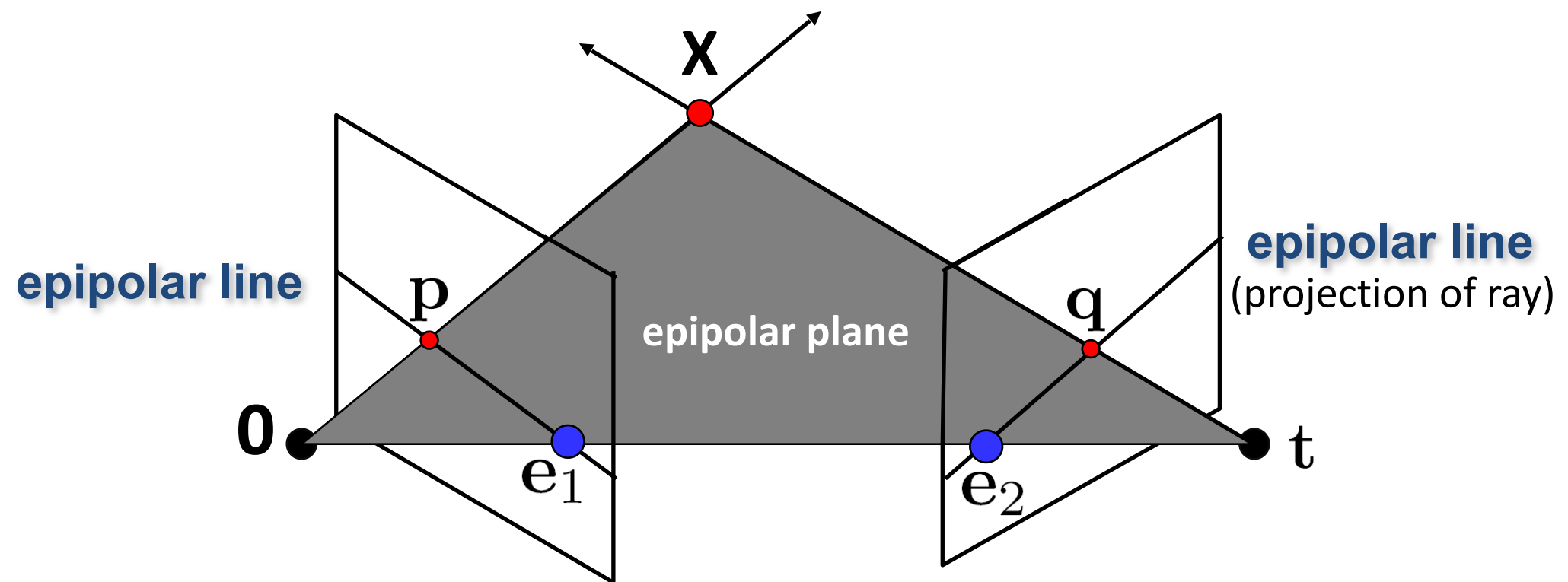
# Essential Matrix



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# Essential Matrix



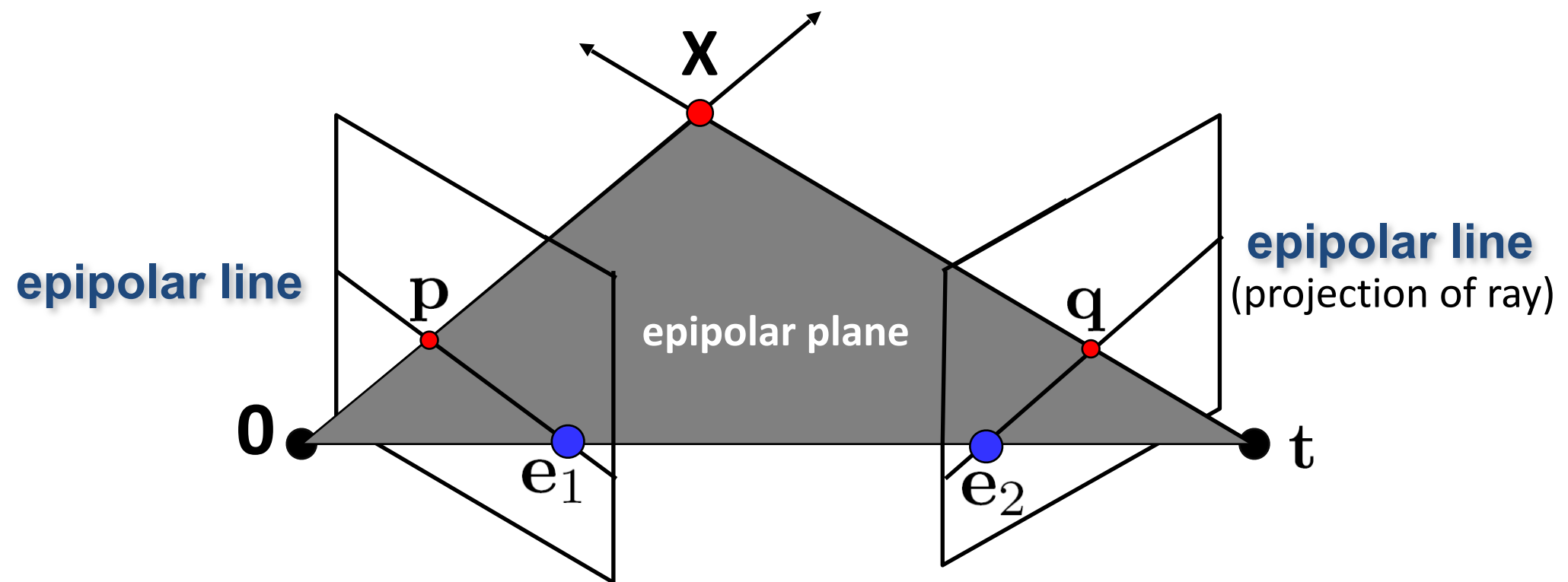
- We have:  $\lambda_1 p = \lambda_2 Rq + t$
- Take cross-product with respect to  $t$ :

$$\lambda_1 [t]_{\times} p = \lambda_2 [t]_{\times} Rq$$

- Take dot-product with respect to  $p$ :

$$0 = \lambda_2 p^T [t]_{\times} Rq$$

# Essential Matrix



- We have:  $p^T [t]_{\times} R q = 0$

- Define:  $\mathbf{E} = [t]_{\times} \mathbf{R}$

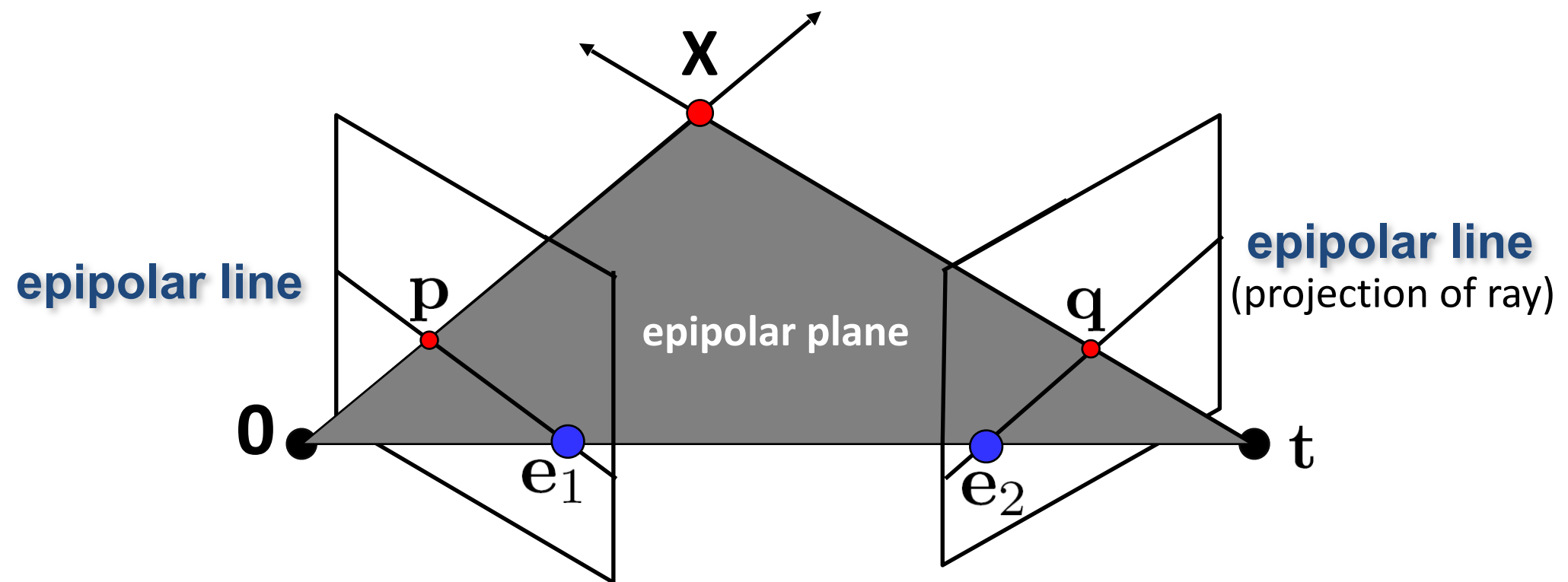
Essential matrix

- Then, we have:

$$\text{rank}(\mathbf{E})=2$$

$$p^T \mathbf{E} q = 0$$

# Fundamental Matrix



- Consider intrinsic camera matrices
- Then,  $\mathbf{p}$  and  $\mathbf{q}$  are in the pinhole frame and pixel counterparts are:

$$\mathbf{p}' = \mathbf{K}_1 \mathbf{p} \quad \mathbf{q}' = \mathbf{K}_2 \mathbf{q}$$

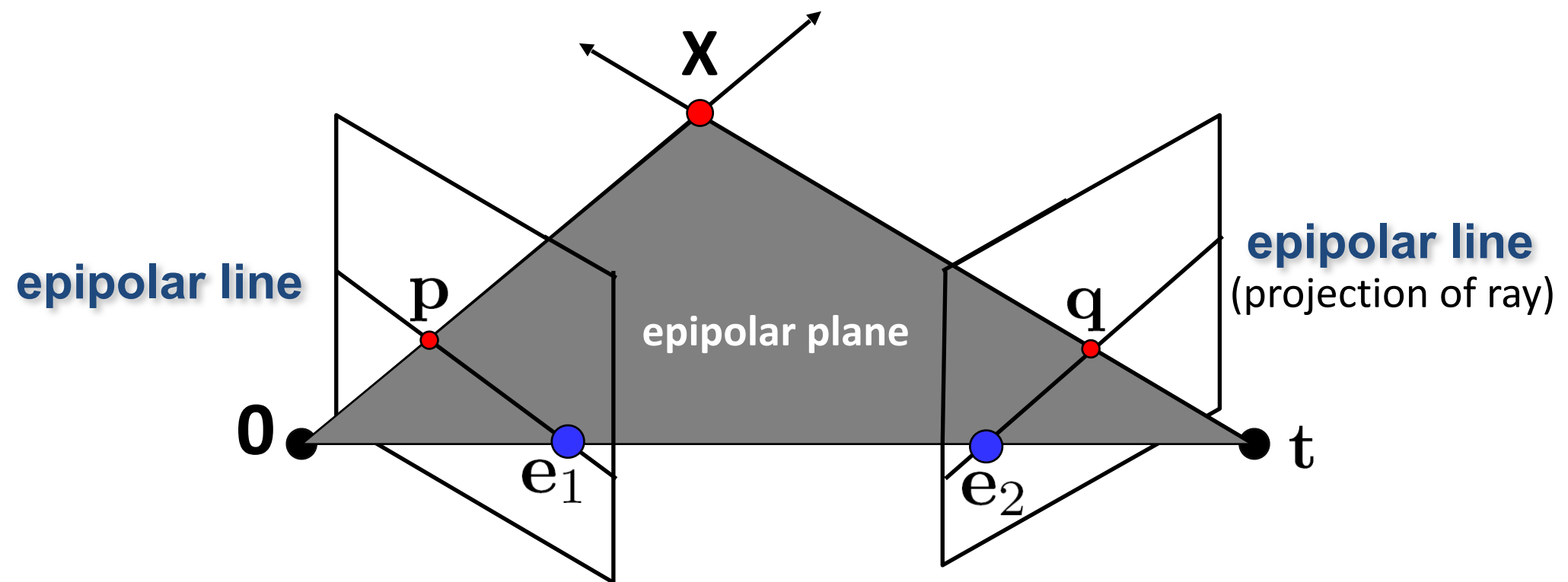
- Recall essential matrix constraint:

$$\mathbf{p}^T \mathbf{E} \mathbf{q} = 0$$

- Substituting, we have:

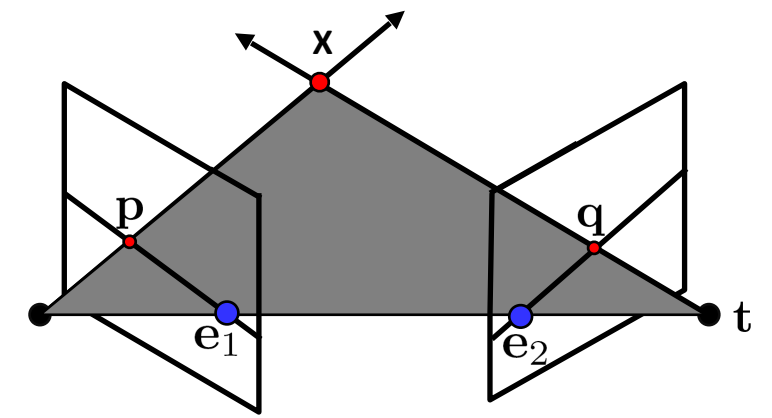
$$(\mathbf{K}_1^{-1} \mathbf{p}')^T \mathbf{E} (\mathbf{K}_2^{-1} \mathbf{q}') = 0$$

# Fundamental Matrix



- Essential matrix constraint in pixel space:  $(K_1^{-1}p')^T E (K_2^{-1}q') = 0$  .
- Rearranging:  $p'^T K_1^{-T} E K_2^{-1} q' = 0$
- Define:  $F = K_1^{-T} E K_2^{-1}$  ← Fundamental matrix  $\text{rank}(F)=2$
- Then, we have:  $p'^T F q' = 0$

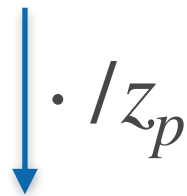
# Fundamental Matrix



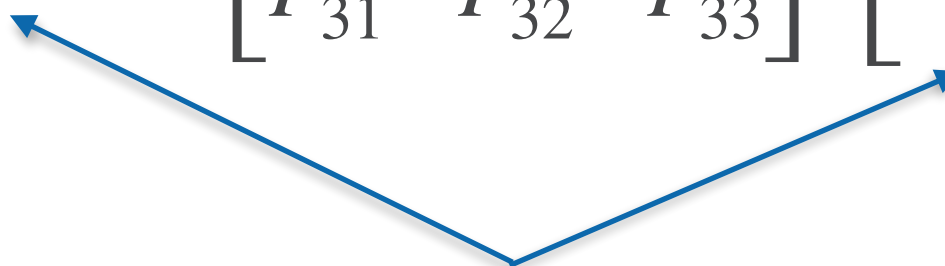
$$p'^T F q' = 0$$



$$[x_p, y_p, z_p] \begin{bmatrix} F_{11} & F_{12} & F_{13} \\ F_{21} & F_{22} & F_{23} \\ F_{31} & F_{32} & F_{33} \end{bmatrix} \begin{bmatrix} x_q \\ y_q \\ z_q \end{bmatrix} = 0$$



$$[x_p/z_p, y_p/z_p, 1] \begin{bmatrix} F_{11} & F_{12} & F_{13} \\ F_{21} & F_{22} & F_{23} \\ F_{31} & F_{32} & F_{33} \end{bmatrix} \begin{bmatrix} x_q/z_q \\ y_q/z_q \\ 1 \end{bmatrix} = 0$$



pixel coordinates



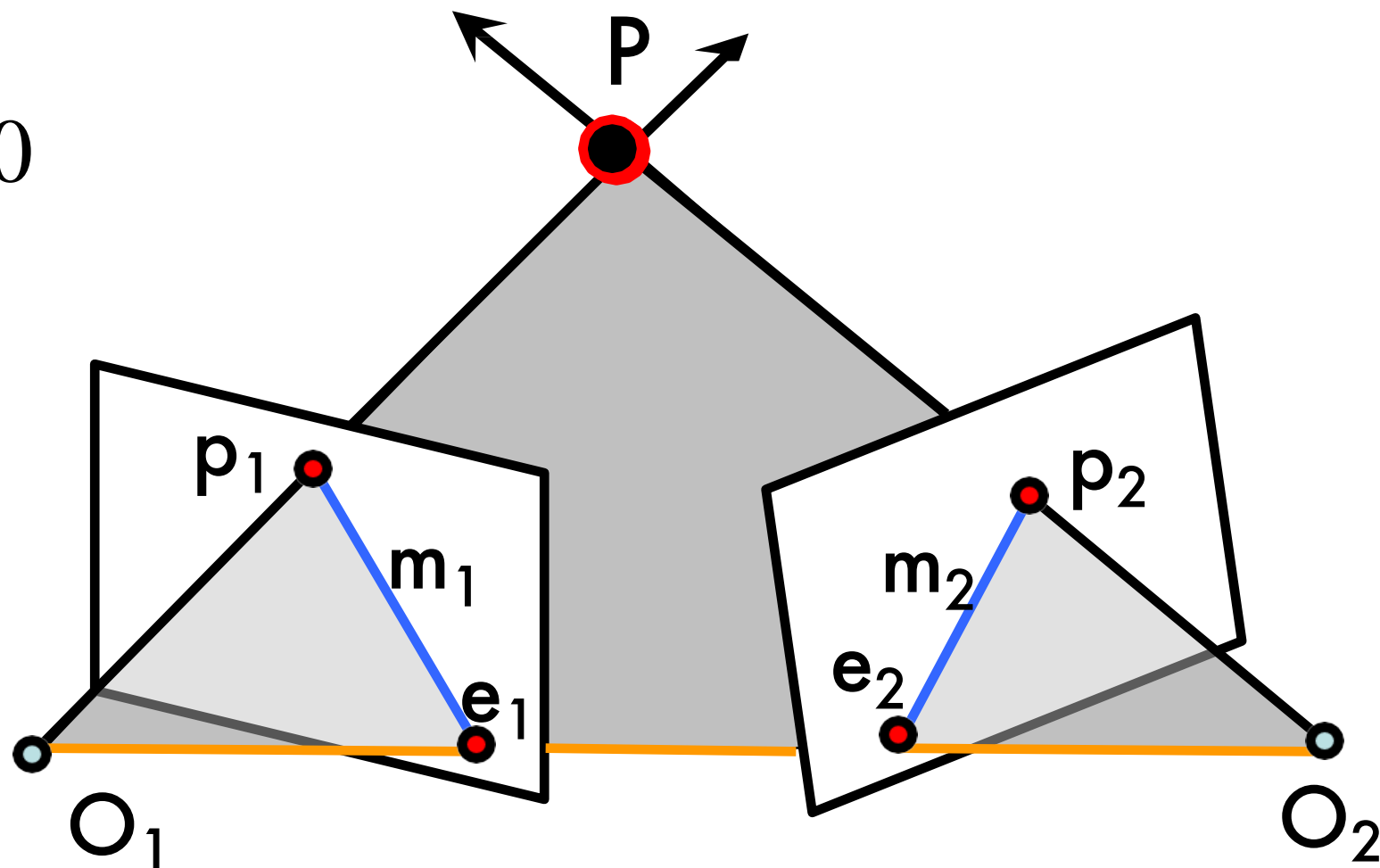
# Epipolar Constraint

$$p_1^T \cdot Fp_2 = 0$$

- $w_1 = Fp_2$  defines an equation  $w_1^T p_1 = 0$
- Note that,  $p_1$  is the corresponding point of  $p_2$  by the derivation of  $F$
- So  $w_1^T p_1 = 0$  is the line constraint that the corresponding point of  $p_2$  has to satisfy
- But the line that corresponds to  $p_2$  is the epipolar line, by the definition of epipolar line
- So,  $w_1 = Fp_2$  defines the epipolar line of  $p_2$

# Epipolar Constraint

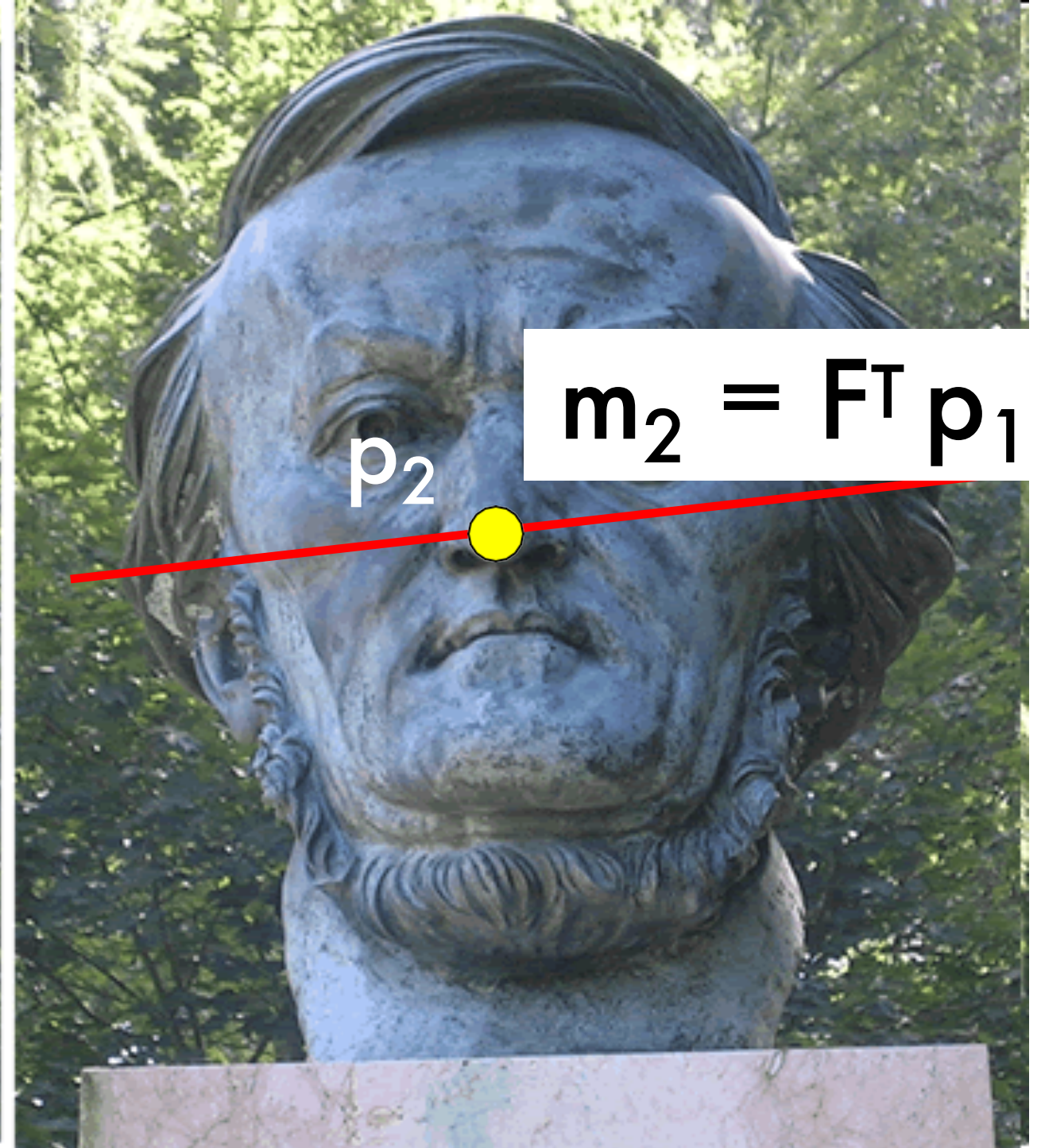
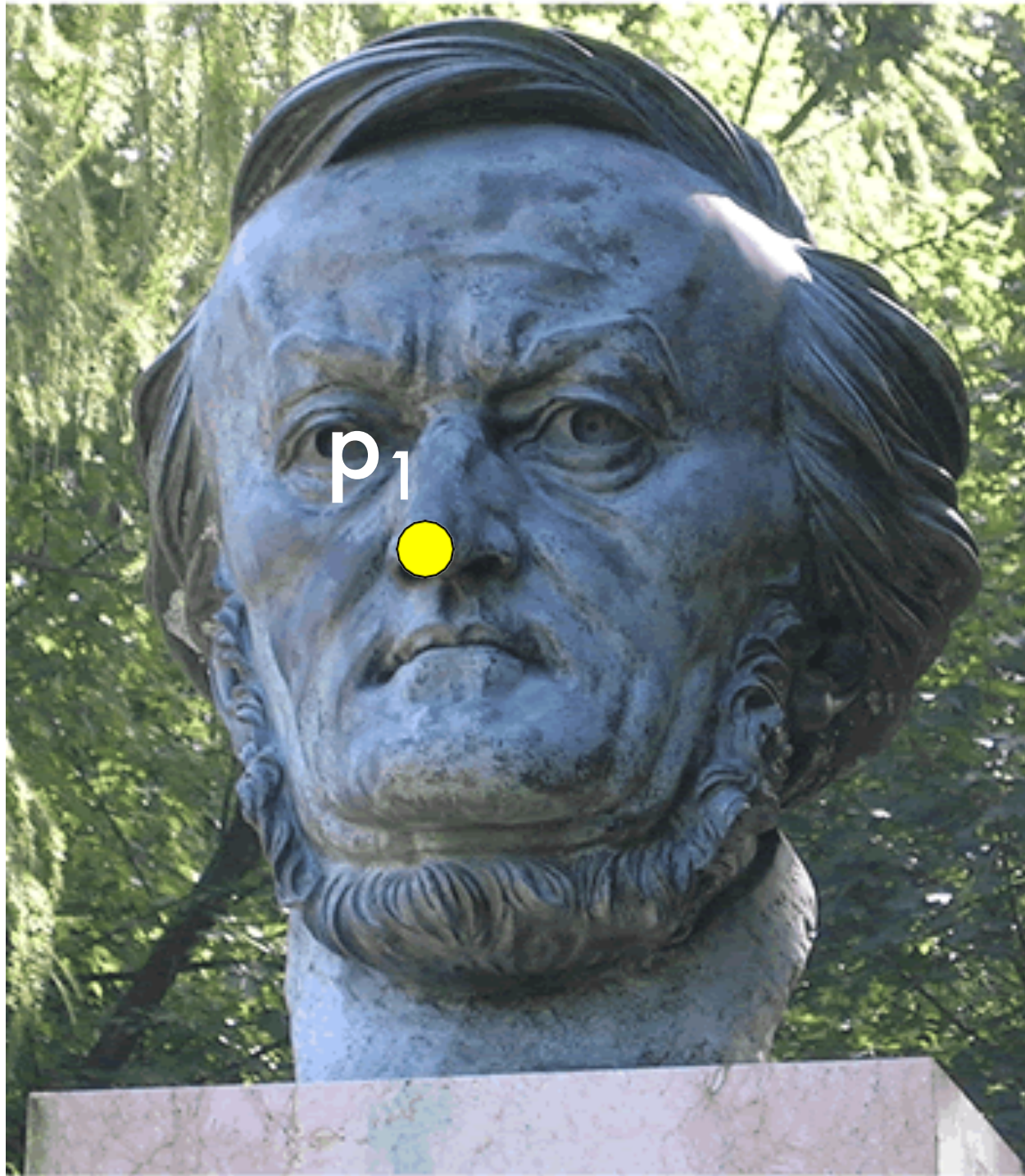
$$p_1^T \cdot F p_2 = 0$$



- $w_1 = F p_2$  defines an equation  $w_1^T p_1 = 0$ , the epipolar line  $m_1$  of  $p_2$
- $w_2 = F^T p_1$  defines an equation  $w_2^T p_2 = 0$ , the epipolar line  $m_2$  of  $p_1$
- $F$  is singular (rank two)
- $F e_2 = 0$       and       $F^T e_1 = 0$



# Why F is useful?



- Suppose  $F$  is known
- No additional information about the scene and camera is given
- Given a point on left image, we can compute the corresponding epipolar line in the second image



# Why $F$ is useful?

- $F$  captures information about the epipolar geometry of 2 views + camera parameters
- **MORE IMPORTANTLY:**  $F$  gives constraints on how the scene changes under view point transformation (without reconstructing the scene!)
- Powerful tool in:
  - 3D reconstruction
  - Multi-view object/scene matching