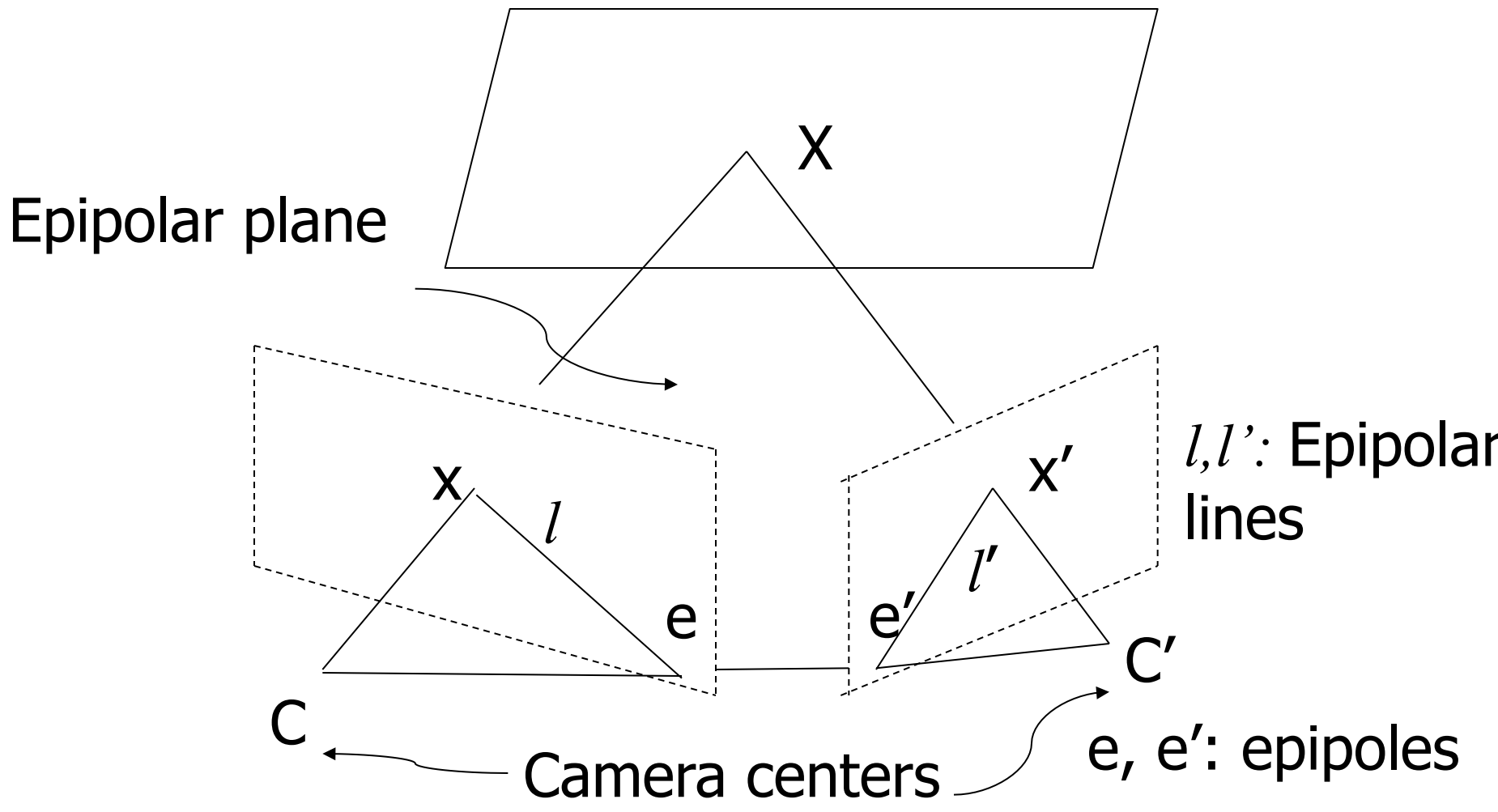




Stereo Geometry

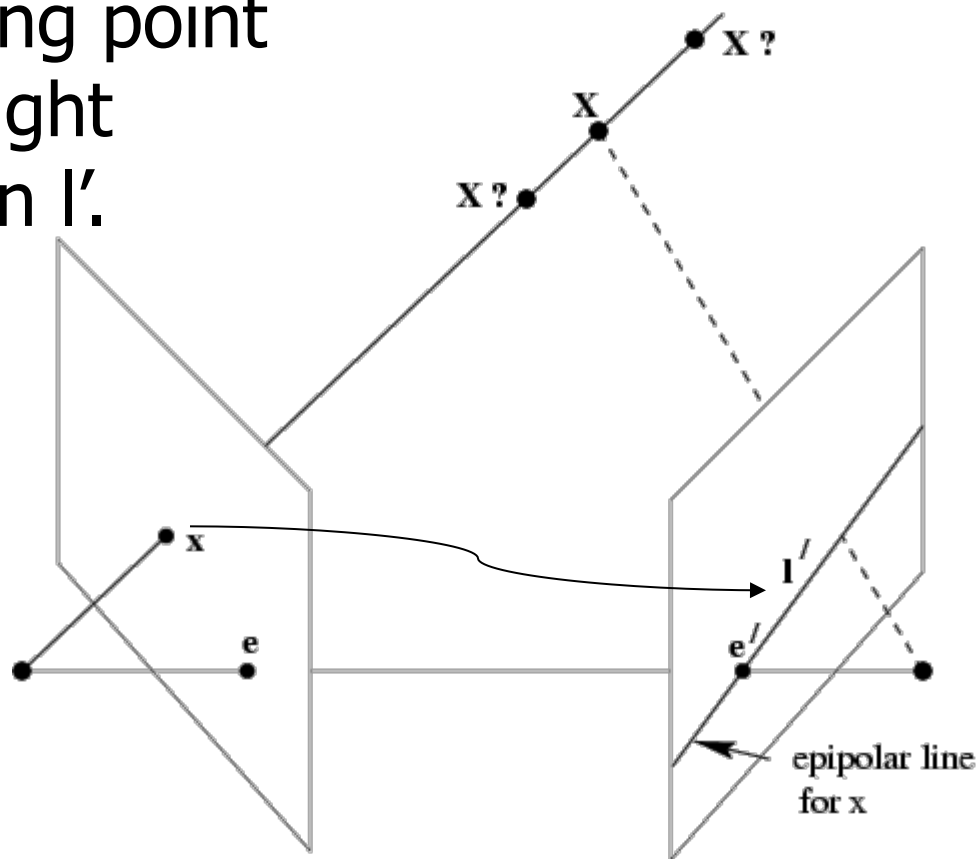
Jayanta Mukhopadhyay
Dept. of Computer Science and Engg.

Stereo Set-up

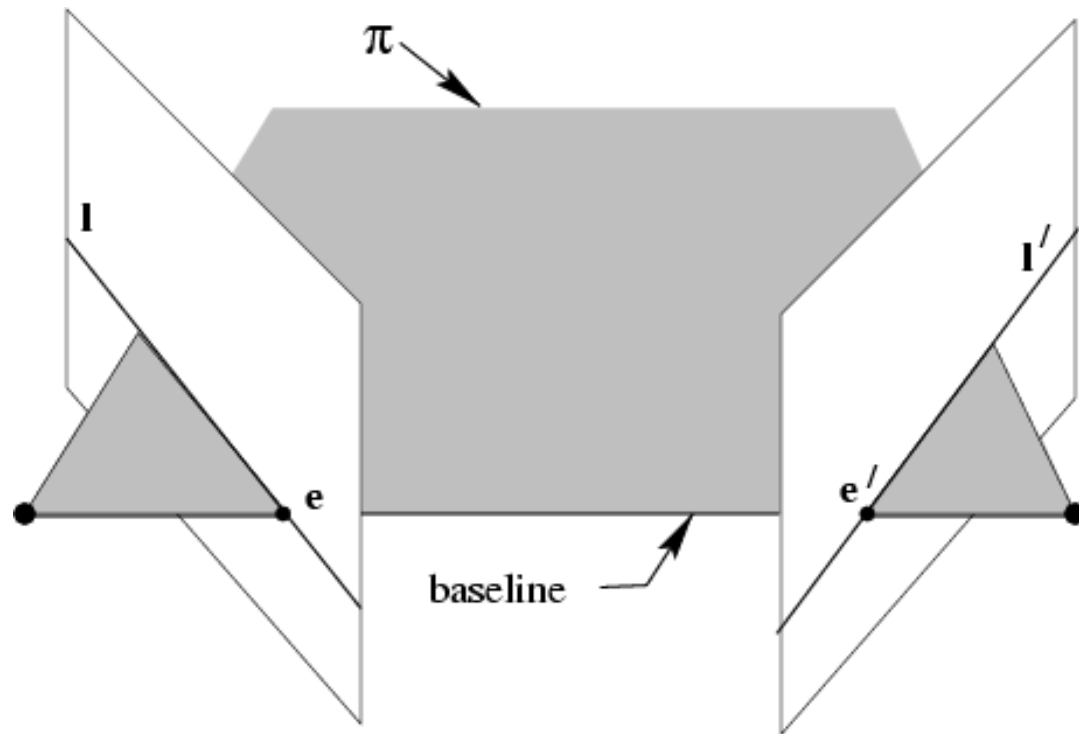


Epipolar geometry

Corresponding point
of x in the right
image lies on l' .



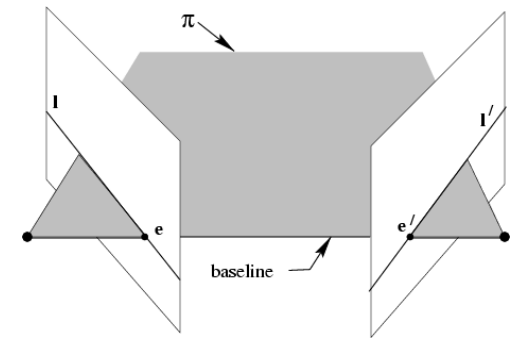
Epipolar geometry



All points on π project on l and l'

From Hartley and Zisserman, "Multiple view geometry in computer vision", Cambridge Univ. Press (2000)

Epipolar geometry



Epipoles e, e'

= intersection of baseline with image plane

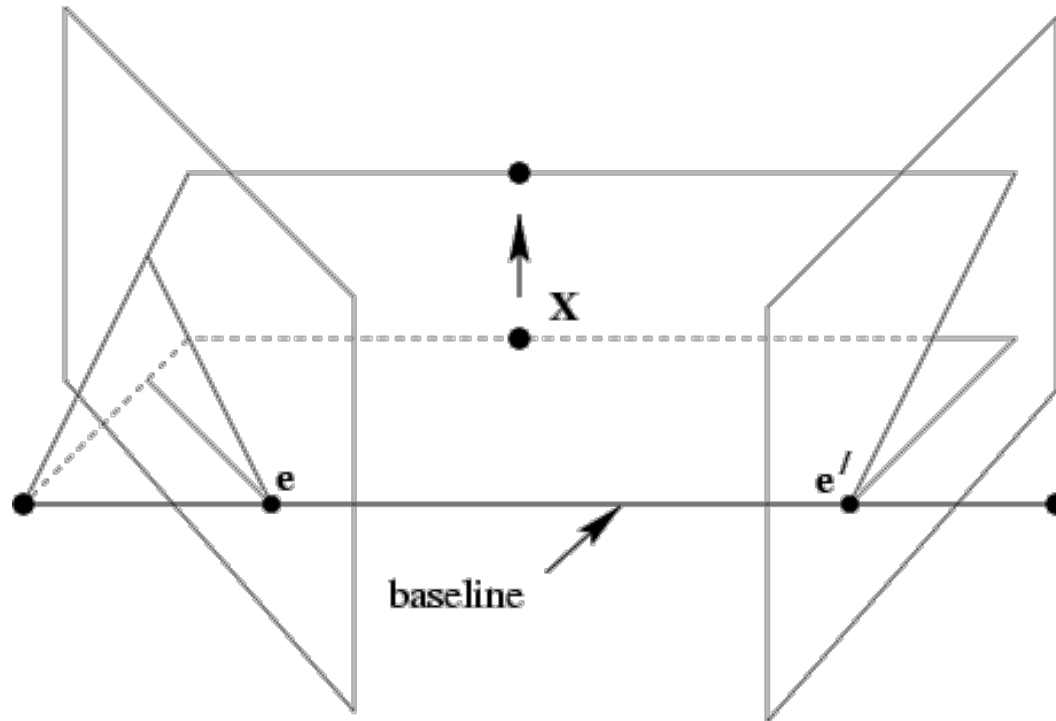
= projection of projection center in other image

= vanishing point of camera motion direction

An epipolar plane = plane containing baseline (1-D family)

An epipolar line = intersection of epipolar plane with image (always come in corresponding pairs)

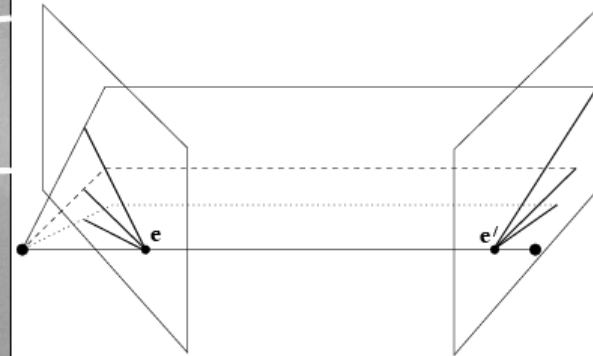
Epipolar geometry



Family of planes π and lines l and l'
Intersection in e and e'

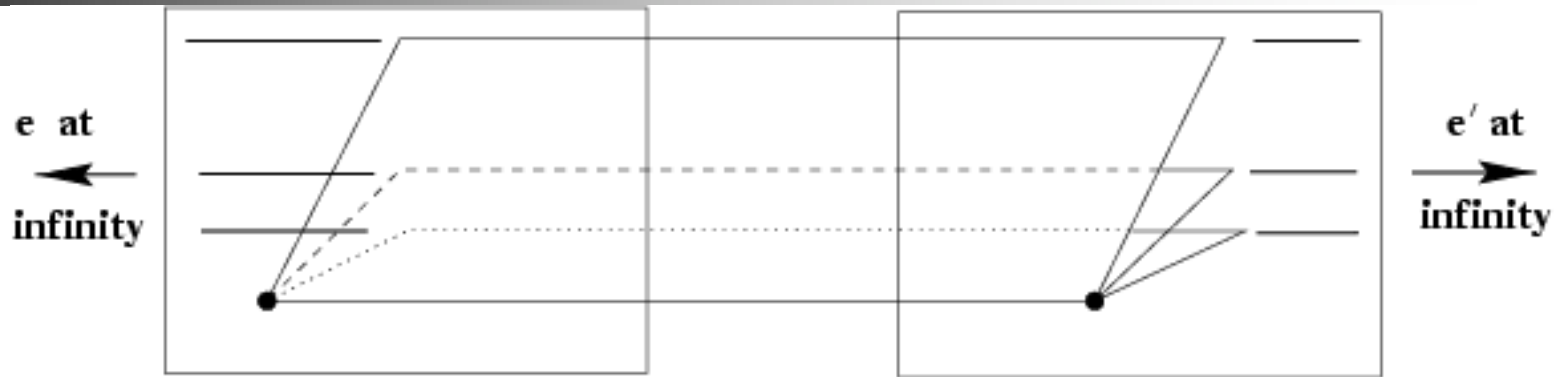
From Hartley and Zisserman, "Multiple view geometry in computer vision", Cambridge Univ. Press (2000)

Example: converging cameras



From Hartley and Zisserman, "Multiple view geometry in computer vision", Cambridge Univ. Press (2000)

Example: motion parallel with image plane

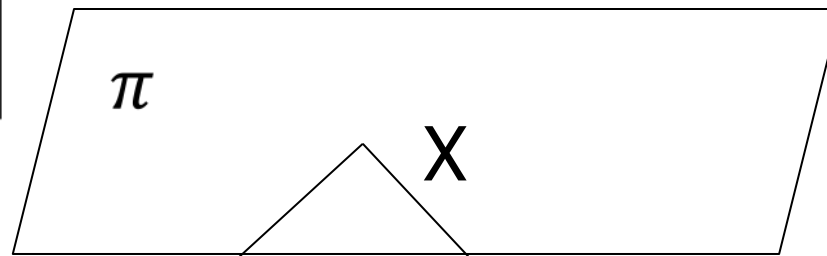


From Hartley and Zisserman, "Multiple view geometry in computer vision", Cambridge Univ. Press (2000)

Epipolar Geometry

$$[e']_{\times} = \begin{bmatrix} 0 & -e_z & e_y \\ e_z & 0 & -e_x \\ -e_y & e_x & 0 \end{bmatrix}$$

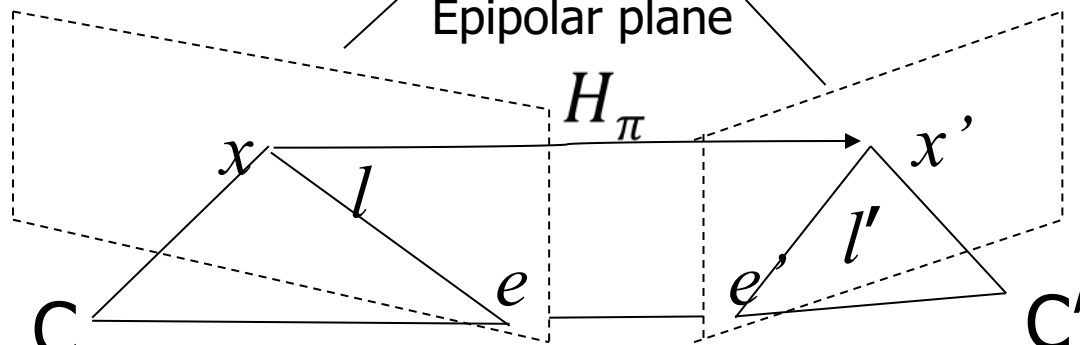
$$H_{\pi} = [K'R | Kt] \begin{bmatrix} K^{-1} \\ 0^T \end{bmatrix} \\ = K'RK^{-1}$$



$$x' = P'X \\ = P'P^+x$$

$$P^+ = \begin{bmatrix} K^{-1} \\ 0^T \end{bmatrix}$$

$$l' = e' \times x' \\ = [e']_{\times} x' \\ = [e']_{\times} H_{\pi} x \\ = Fx$$



$$P = K[I | 0]$$

$$P' = K'[R | t]$$

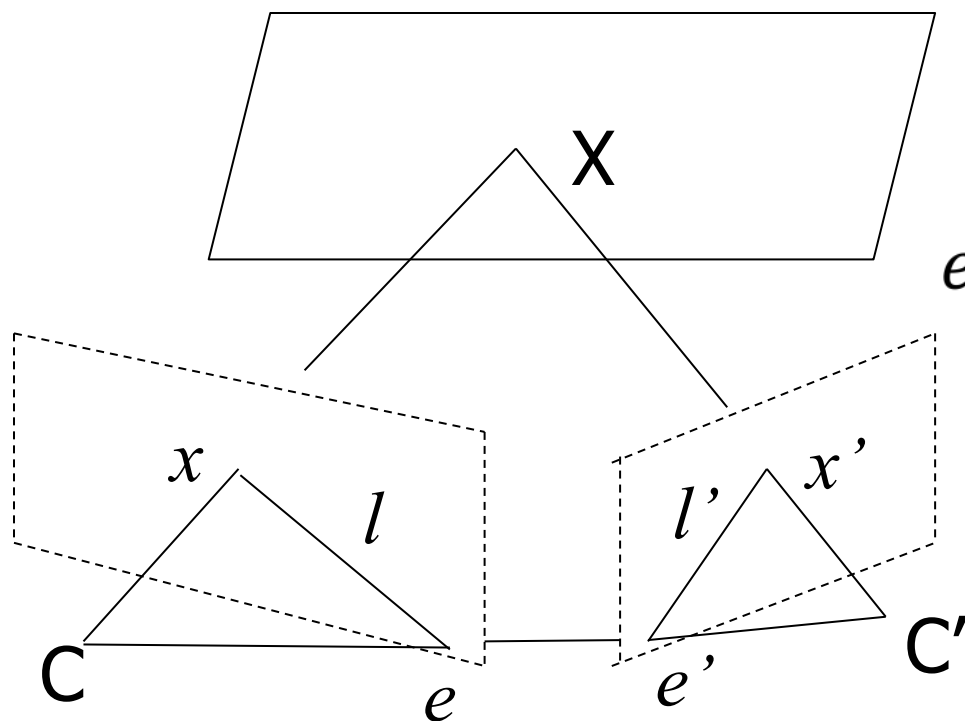
$$F = [e']_{\times} K'RK^{-1}$$

Coplanar: $X, x, x', C, C', e, e', l, l'$

Epipolar Geometry

Fundamental
Matrix: F

$$l' = Fx$$



$$e = PC'$$

$$e' = P'C$$

$$e = -KR^T t \equiv KR^T t$$

$$e' = K't$$

$$l' = Fx$$

$$\Rightarrow x'^T l' = 0$$

$$\Rightarrow x'^T Fx = 0$$

$$\Rightarrow x^T F^T x' = 0$$

$$C = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$P = K[I \mid 0]$$

$$P' = K'[R \mid t]$$

$$C' = \begin{bmatrix} -R^T t \\ 1 \end{bmatrix}$$

Fundamental and Essential Matrices

$$P = K[I \mid 0]$$

$$P' = K'[R \mid t]$$

$$F = [e']_{\times} P' P^+ \\ = [K' t]_{\times} K' R K^{-1}$$

Say, $P = [I \mid 0]$, i.e. $K = I$

$$P' = K'[R \mid t] = [K'R \mid K't] = [M \mid m]$$

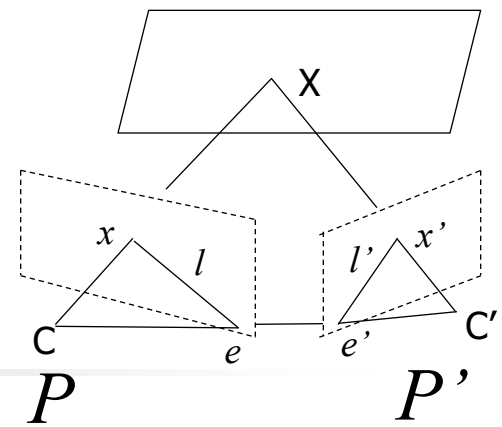
$$F = [m]_{\times} M$$

for $K = I$ and $K' = I$,

$$F = [t]_{\times} R$$

Essential Matrix (E)

Fundamental Matrix: Properties



$$x'^T F x = x^T F^T x' = 0, \forall (x', x)$$

Transpose:

If F is fundamental matrix of (P, P') , F^T for (P', P) .

Epipolar lines: For x , epipolar line $l' = Fx$.
For x' , epipolar line $l = F^T x'$.

F is a correlation.
→ Rank deficient.
→ Inverse does not exist.

Epipoles: $e'^T (Fx) = e^T (F^T x') = (Fe)^T x' = 0$

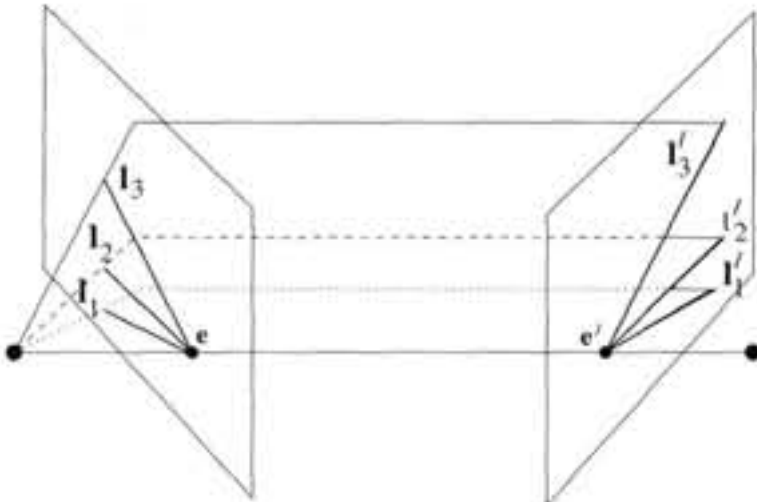
$e'^T F = 0 \rightarrow e'$ is **left NULL vector** of F .

$Fe = 0 \rightarrow e$ is the **right NULL vector** of F .

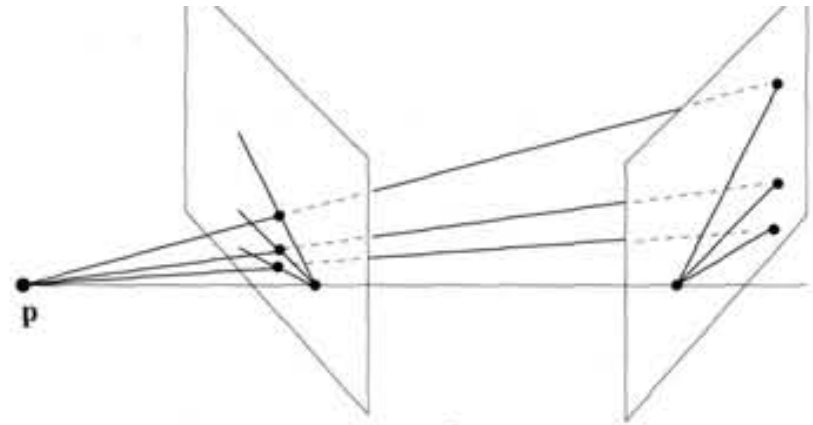
Rank deficient:

$\det(F) = 0$ and F is a projective element $\rightarrow 7$ d.o.f.

Epipolar line homography



l_i and l'_i form pencil of planes with axis the baseline



l_i and l'_i related by a perspectivity with centre at any point p on the baseline
→ 1D homography.

From Hartley and Zisserman, "Multiple view geometry in computer vision", Cambridge Univ. Press (2000)



Epipolar line homography

Let (l, l') be corresponding epipolar lines and k is any line not passing through epipole e .

Then, $l' = F [k]_x l$

Proof: $k \times l = p$ (point of intersection)

As p lies on l , $Fp = l'$

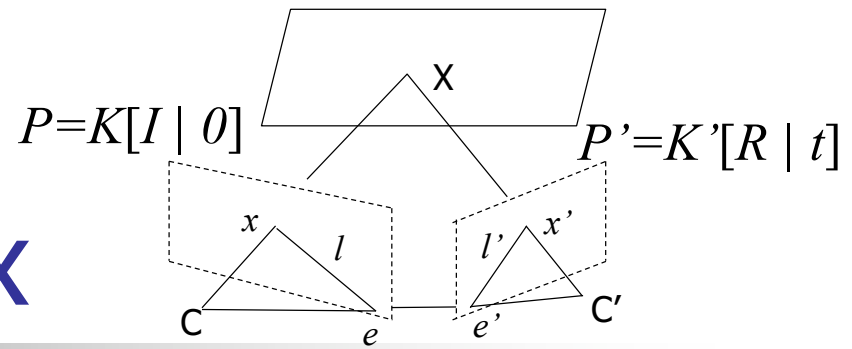
$\rightarrow l' = F k \times l = F [k]_x l \quad (Q.E.D.)$

As $e^T e > 0$,

the line e does not pass through the epipole e .

$\rightarrow l' = F [e]_x l$

Essential Matrix



Stereo geometry for calibrated cameras.

$$x_c = K^{-1}x \quad x_c'^T E x_c = 0$$

Coordinates in calibrated image planes.

$$E = K'^T F K$$

$$E = [t]_x R$$

$$F = K'^{-T} E K^{-1}$$

$$= K'^{-T} [t]_x R K^{-1} \quad 6 \text{ parameters}$$

$$\text{d.o.f.} = 5$$

$$x_c' = K'^{-1}x'$$

$$x_c'^T F x_c = 0$$

$$\rightarrow (K' x_c')^T F (K x_c) = 0$$

$$\rightarrow x_c'^T K'^T F K x_c = 0$$

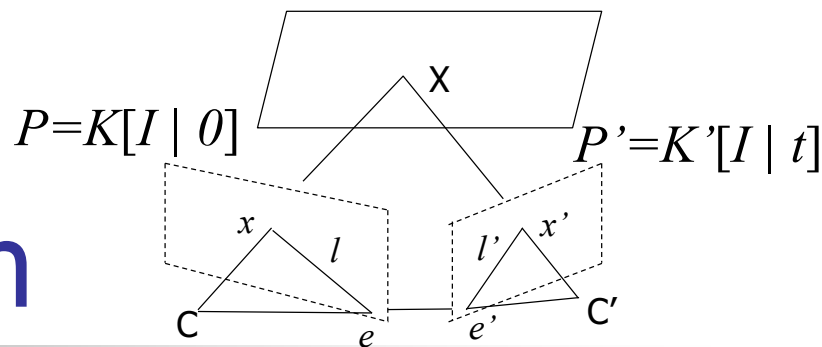
$$\rightarrow x_c'^T (K'^T F K) x_c = 0$$

$$E e_c = e_c' E = 0$$

$$\text{Rank: } 2$$

$$\det(E) = 0$$

Pure translation



$$F = [e']_x K' I K^{-1} \\ = [e']_x K' K^{-1}$$

camera translation $\parallel l'$
to x-axis, $e' = [1 \ 0 \ 0]^T$

$$\rightarrow F = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{bmatrix}$$

For $K = K'$, $F = [e']_x$ $[e']_x = \begin{bmatrix} 0 & -e_z & e_y \\ e_z & 0 & -e_x \\ -e_y & e_x & 0 \end{bmatrix}$

$$x \equiv PX = [K \mid 0] \begin{bmatrix} \tilde{X} \\ 1 \end{bmatrix} = K \tilde{X}$$

$$\rightarrow x'^T F x = 0 \\ \rightarrow y' = y$$

$$K^{-1} x \equiv \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} \quad \tilde{X} = Z K^{-1} x$$

$$K = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad Z = \frac{f t_x}{x' - x}$$

$$x' \equiv P' X = [K' \mid K' t] \begin{bmatrix} \tilde{X} \\ 1 \end{bmatrix} = K' \tilde{X} + K' t \\ = K' Z K^{-1} x + K' t \\ = Z \left(K' K^{-1} x + \frac{K' t}{Z} \right)$$

For $K = K'$, $x' \equiv Z \left(x + \frac{K t}{Z} \right)$

$$\rightarrow x' = x + \frac{K t}{Z}$$

General Motion of Camera

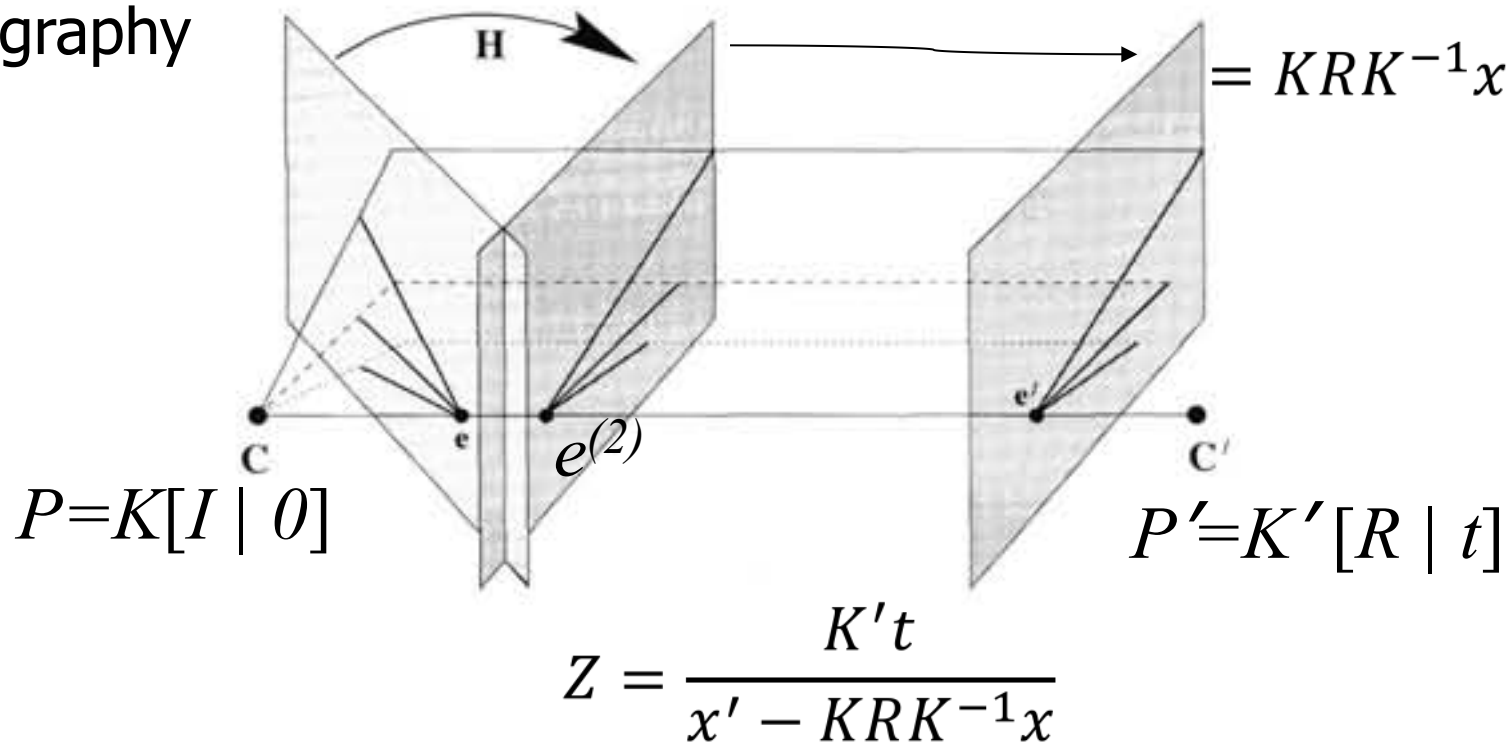
$$\begin{aligned} x^{(2)} &= K[R|0]X \\ &= KR[I|0]X \\ &= KRK^{-1}K[I|0]X \\ &= KRK^{-1}x \end{aligned}$$

Rotational
Homography

$$H = KRK^{-1}$$

||^l stereo

$$\begin{aligned} x' &= x^{(2)} + \frac{K't}{Z} \\ &= KRK^{-1}x + \frac{K't}{Z} \end{aligned}$$



From Hartley and Zisserman, "Multiple view geometry in computer vision", Cambridge Univ. Press (2000)

Estimation of Fundamental Matrix

$$F = \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix}$$

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

$$x' x f_{11} + x' y f_{12} + x' f_{13} + y' x f_{21} + y' y f_{22} + y' f_{23} + x f_{31} + y f_{32} + f_{33} = 0$$

$$[x' x, x' y, x', y' x, y' y, y', x, y, 1] [f_{11}, f_{12}, f_{13}, f_{21}, f_{22}, f_{23}, f_{31}, f_{32}, f_{33}]^T = 0$$

$$\begin{bmatrix} x'_1 x_1 & x'_1 y_1 & x'_1 & y'_1 x_1 & y'_1 y_1 & y'_1 & x_1 & y_1 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x'_n x_n & x'_n y_n & x'_n & y'_n x_n & y'_n y_n & y'_n & x_n & y_n & 1 \end{bmatrix} \mathbf{f} = 0$$

$$A \mathbf{f} = 0$$

- Solution up to scale.
- Minimum 8 point correspondences.
- Use of DLT (for 7 point correspondences from linear combination of smallest and second smallest eigen vectors).

Estimation of Fundamental Matrix

$$\begin{bmatrix}
 x_1 x_1' & y_1 x_1' & x_1' & x_1 y_1' & y_1 y_1' & y_1' & x_1 & y_1 & 1 \\
 x_2 x_2' & y_2 x_2' & x_2' & x_2 y_2' & y_2 y_2' & y_2' & x_2 & y_2 & 1 \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
 x_n x_n' & y_n x_n' & x_n' & x_n y_n' & y_n y_n' & y_n' & x_n & y_n & 1
 \end{bmatrix}
 \begin{bmatrix}
 f_{11} \\
 f_{12} \\
 f_{13} \\
 f_{21} \\
 f_{22} \\
 f_{23} \\
 f_{31} \\
 f_{32} \\
 f_{33}
 \end{bmatrix} = 0$$

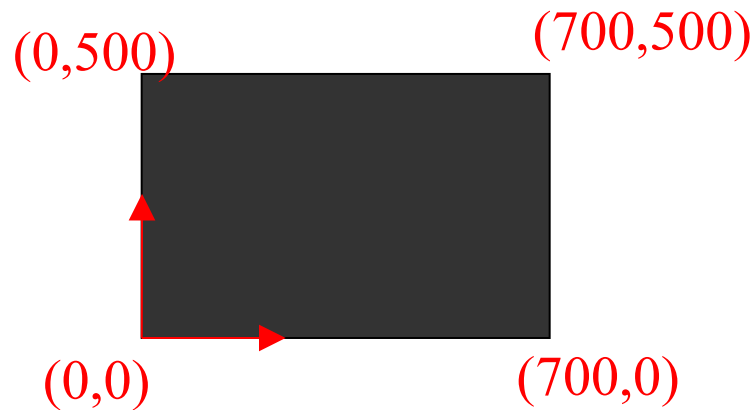
$\sim 10000 \quad \sim 10000 \quad \sim 100 \quad \sim 10000 \quad \sim 10000 \quad \sim 100 \quad \sim 100 \quad \sim 100 \quad 1$



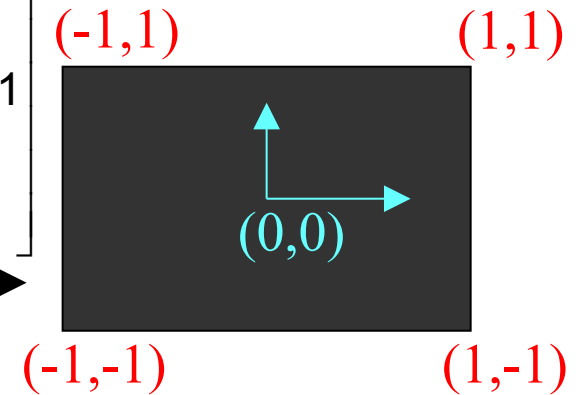
Orders of magnitude difference
 Between column of data matrix
 → least-squares yield poor results

The normalized 8-point algorithm

Transform image to
 $\sim [-1,1] \times [-1,1]$



$$\begin{bmatrix} \frac{2}{700} & 0 & -1 \\ \frac{2}{500} & -1 & 1 \end{bmatrix}$$



Least squares yields good results
(Hartley, PAMI '97)



The singularity constraint

$$\det F = 0 \quad \text{rank } F = 2$$

SVD from linearly computed F matrix (rank 3)

$$F = U \begin{bmatrix} \sigma_1 & & \\ & \sigma_2 & \\ & & \sigma_3 \end{bmatrix} V^T = U_1 \sigma_1 V_1^T + U_2 \sigma_2 V_2^T + U_3 \sigma_3 V_3^T$$

$$\min \|F - F'\|_F$$

Compute closest rank-2 approximation

$$F' = U \begin{bmatrix} \sigma_1 & & \\ & \sigma_2 & \\ & & 0 \end{bmatrix} V^T = U_1 \sigma_1 V_1^T + U_2 \sigma_2 V_2^T$$

The singularity constraint

Nonsingular
 F



Singular
 F



Non-singular F causes epipolar lines not converging.

From Hartley and Zisserman, "Multiple view geometry in computer vision", Cambridge Univ. Press (2000)



The singularity constraint for Essential Matrix

$$\det(E)=0$$

Estimate \hat{E} by any of the techniques used for F .

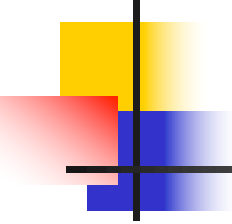
Perform SVD of \hat{E} .

$$\hat{E}=UDV^T \quad \text{Where } D=\text{diag}(a,b,c) \quad a \geq b \geq c$$

For essential matrix, two singular values are the same.

$$\Rightarrow \hat{E} = U\hat{D}V^T \quad \text{where } \hat{D} = \left(\frac{a+b}{2}, \frac{a+b}{2}, 0 \right)$$

The minimum case – 7 point correspondences


$$\begin{bmatrix} x'_1 x_1 & x'_1 y_1 & x'_1 & y'_1 x_1 & y'_1 y_1 & y'_1 & x_1 & y_1 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x'_7 x_7 & x'_7 y_7 & x'_7 & y'_7 x_7 & y'_7 y_7 & y'_7 & x_7 & y_7 & 1 \end{bmatrix} \mathbf{f} = 0$$

$$\mathbf{A} = \mathbf{U}_{7 \times 7} \text{diag}(\sigma_1, \dots, \sigma_7, 0, 0) \mathbf{V}_{9 \times 9}^T$$

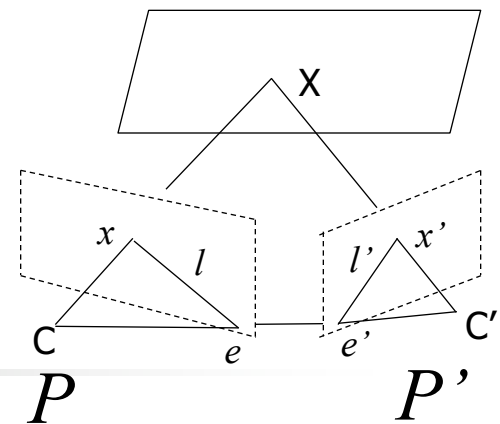
$\mathbf{F}_1, \mathbf{F}_2 \rightarrow$ Eigen vectors corresponding to two zero's. The solution is $\mathbf{F}_1 + \lambda \mathbf{F}_2$.

But $\mathbf{F}_1 + \lambda \mathbf{F}_2$ not automatically rank 2.

Solve for λ from $\det(\mathbf{F}_1 + \lambda \mathbf{F}_2) = 0$.

As it is a cubic polynomial, there are 1 or 3 solutions.

Parametric representation of F



Over parameterization: $F=[t]_x M \rightarrow \{t, M\} \rightarrow 12$ params.

Epipolar parameterization:

$$F = \begin{bmatrix} a & b & \alpha a + \beta b \\ c & d & \alpha c + \beta d \\ e & f & \alpha e + \beta f \end{bmatrix} \quad \{a, b, c, d, e, f, \alpha, \beta\}$$

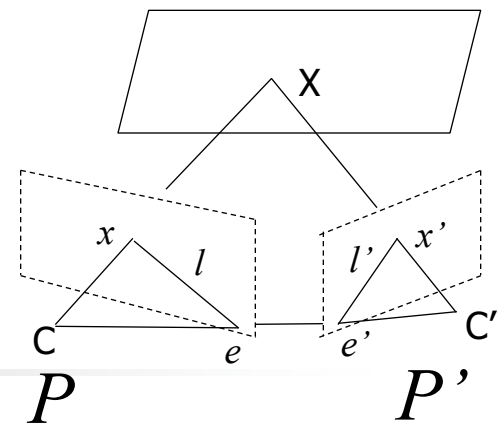
Right epipole: $e' = [\alpha \quad \beta \quad -1]^T$

Both epipoles as parameters $\{a, b, c, d, \alpha, \beta, \alpha', \beta'\}$

$$F = \begin{bmatrix} a & b & \alpha a + \beta b \\ c & d & \alpha c + \beta d \\ \alpha' a + \beta' c & \alpha' b + \beta' d & \alpha \alpha' a + \alpha \beta' c + \alpha' \beta b + \beta \beta' d \end{bmatrix}$$

Epipoles: $e' = [\alpha \quad \beta \quad -1]^T \quad e = [\alpha' \quad \beta' \quad -1]^T$

Retrieving the camera matrices from F



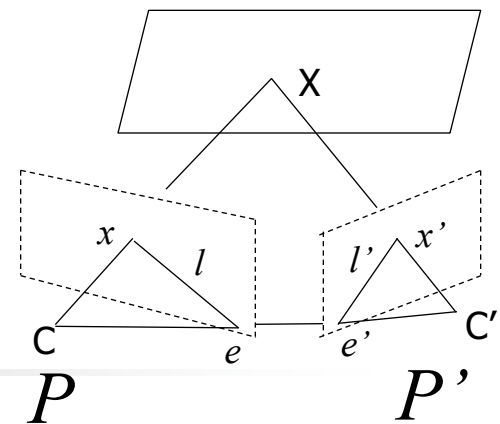
- F only depends on projective properties of P and P' .
- Independent of choice of world frame.
- $(P, P') \rightarrow F$ (unique)
- $F \rightarrow (P, P')$ (?)
- Given a homography H (4x4 non-singular matrix) in P^3 , if $(P, P') \rightarrow F$, then $(PH, P'H) \rightarrow F$.

Proof: $PX \leftrightarrow P'X$

$$\rightarrow (PH)(H^{-1}X) \leftrightarrow (P'H)(H^{-1}X)$$

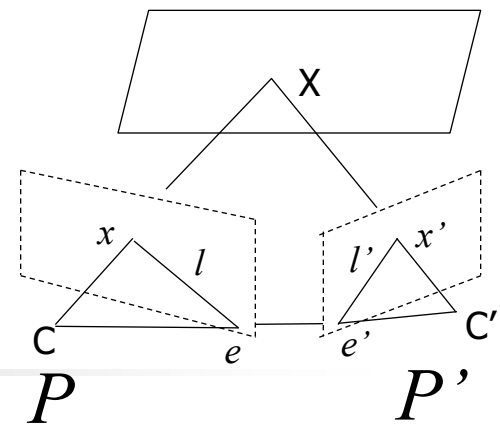
- F does not uniquely map to (P, P') .

Retrieving the camera matrices from F



- $P=[I \mid 0]$ & $P'=[M \mid m] \rightarrow F=[m]_x M$.
- If F derived from both (P_1, P_1') and (P_2, P_2') , there exists 4×4 H s.t. $P_2=P_1 H$ & $P_2'=P_1' H$.
- Suppose a rank 2 matrix F decomposed in two different ways, $F=[a]_x A=[b]_x B$. Then $a=k b$ and $B=k^{-1}(A+a v^T)$ for some non zero constants k and 3-vector v .
- d.o.f. of P + d.o.f. of $P'=22$
- d.o.f. of $H=15$
- d.o.f. of $F=22-15=7$

Retrieving the camera matrices from F



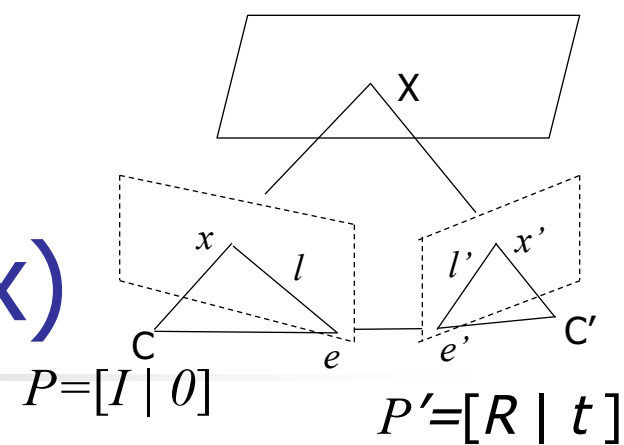
- F corresponds to (P, P') , iff $P'^T F P$ is *skew symmetric*.

Proof: For a skew symmetric matrix S , $X^T S X = 0$, for all X .

Now, $X^T P'^T F P X = (P' X)^T F (P X) = x'^T F x = 0$ (for any X in P^3 , as F is the fundamental matrix). ...

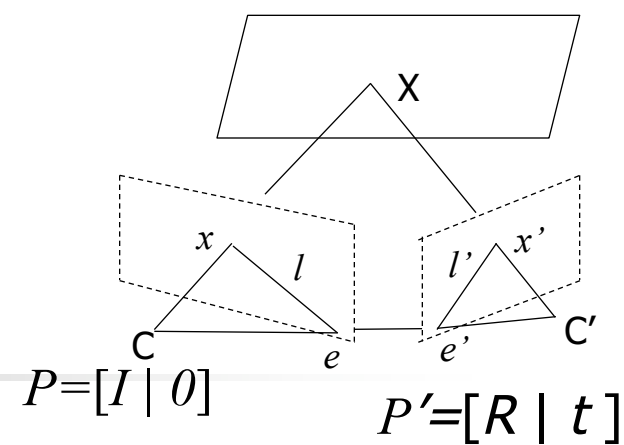
- F corresponds to $P = [I \mid 0]$ & $P' = [SF \mid e']$, where e' is the right epipole of F s.t. $e'^T F = 0$.
- A good choice of $S = [e']_x$.
- $F \rightarrow ([I \mid 0], [[e']_x F \mid e'])$
 $\leftrightarrow ([I \mid 0], [[e']_x F + e' v^T \mid ke'])$

The camera matrices from E (Essential matrix)



- E is an essential matrix iff two of its singular values are equal and the third one is zero.
- $E = [t]_x R$
- $[t]_x$ and R can be computed through decomposition of E s.t. $E = SR$, where S is a skew symmetric matrix and R is orthogonal.

Decomposition of E (Essential matrix)



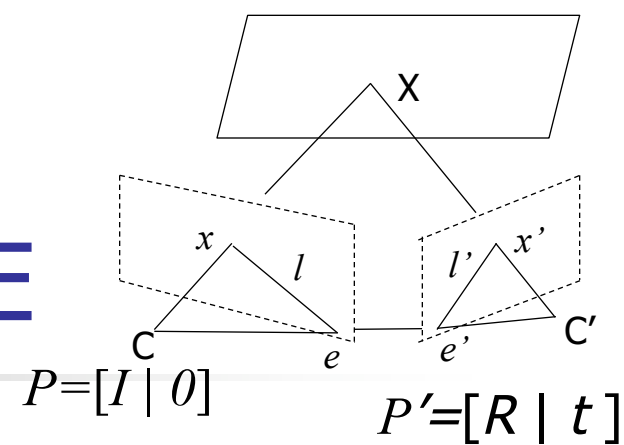
- SVD of $E = U \text{diag}(1, 1, 0) V^T$
- Two possible decomposition of $E = SR$
- $S = UZU^T$ and $R = UWV^T$ or UW^TV^T

$$Z = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$W = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- Any skew symmetric matrix S can be decomposed as $S = kUZU^T$
- W is orthogonal and $Z = \text{diag}(1, 1, 0)W$.

Camera matrices from E



- SVD of $E = U \text{diag}(1, 1, 0) V^T$
- Two possible decomposition of $E = SR$
- $S = UZU^T$ and $R = UWV^T$ or UW^TV^T

$$Z = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad P' = \begin{bmatrix} UWV^T & +u_3 \\ UW^TV^T & +u_3 \end{bmatrix} \text{ or } \begin{bmatrix} UWV^T & -u_3 \\ UW^TV^T & -u_3 \end{bmatrix} \quad W = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

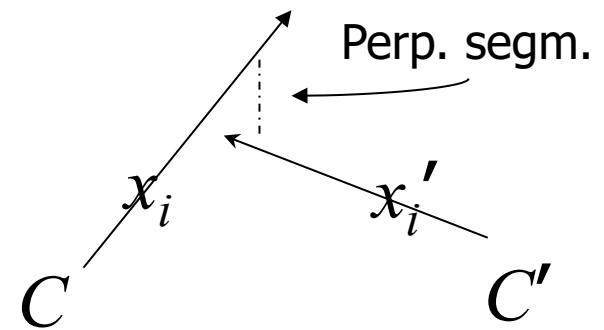
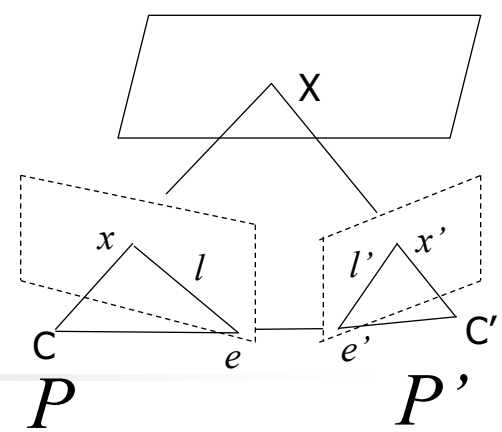
↑ Last column of U .

Out of the four only one is valid for viewing a point from both the cameras. It is sufficient to test a single point for the above.

Computing scene points (structure)

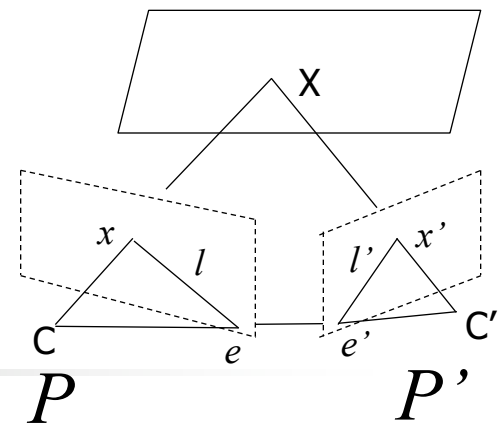
Given $x_i \leftrightarrow x_i'$, compute X .

1. Compute F .
2. Compute P and P' .
3. For each (x_i, x_i') compute X by triangulation.
 - i. Compute intersection of Cx_i and $C'x_i'$.
 - ii. Compute segment perpendicular to both.
 - iii. Get the mid-point.



Not projective invariant, i.e. $(PH, P'H)$ does not give $H^{-1}X$.

Minimizing Reprojection Error



Given $x_i \leftrightarrow x_i'$, compute X .

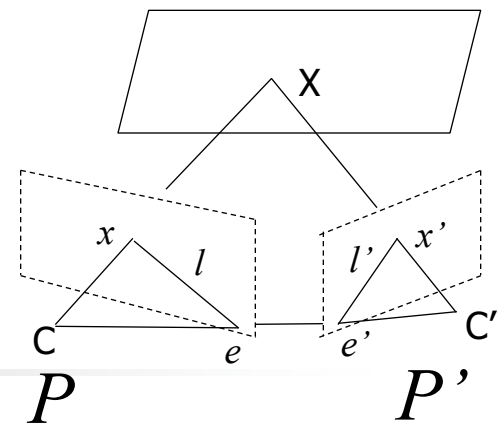
1. Estimate \hat{X} s.t. $P\hat{X} = \hat{x}$ and $P'\hat{X} = \hat{x}'$.
2. Minimize the reprojection error (E_{rp}).

$$E_{rp} = d(x, \hat{x})^2 + d(x', \hat{x}')^2$$

subject to $x'^T F x = 0$

Projective invariant.

Linear triangulation methods



Given $x_i \leftrightarrow x_i'$, compute X .

$$x \times PX = 0$$

$$x' \times P'X = 0$$

4 equations,
3 unknowns.

$$[A]_{4 \times 4} X = 0$$

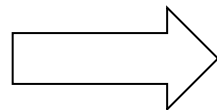
Minimize $\|AX\|$
subject to $\|X\|=1$.

Use DLT.

Not projective invariant.

Generalize to multiview correspondences.

$$\begin{matrix} x_1 & \leftrightarrow & x_2 & \leftrightarrow & x_3 \\ P_1 & & P_2 & & P_3 \end{matrix}$$



$$x_1 \times P_1 X = 0$$

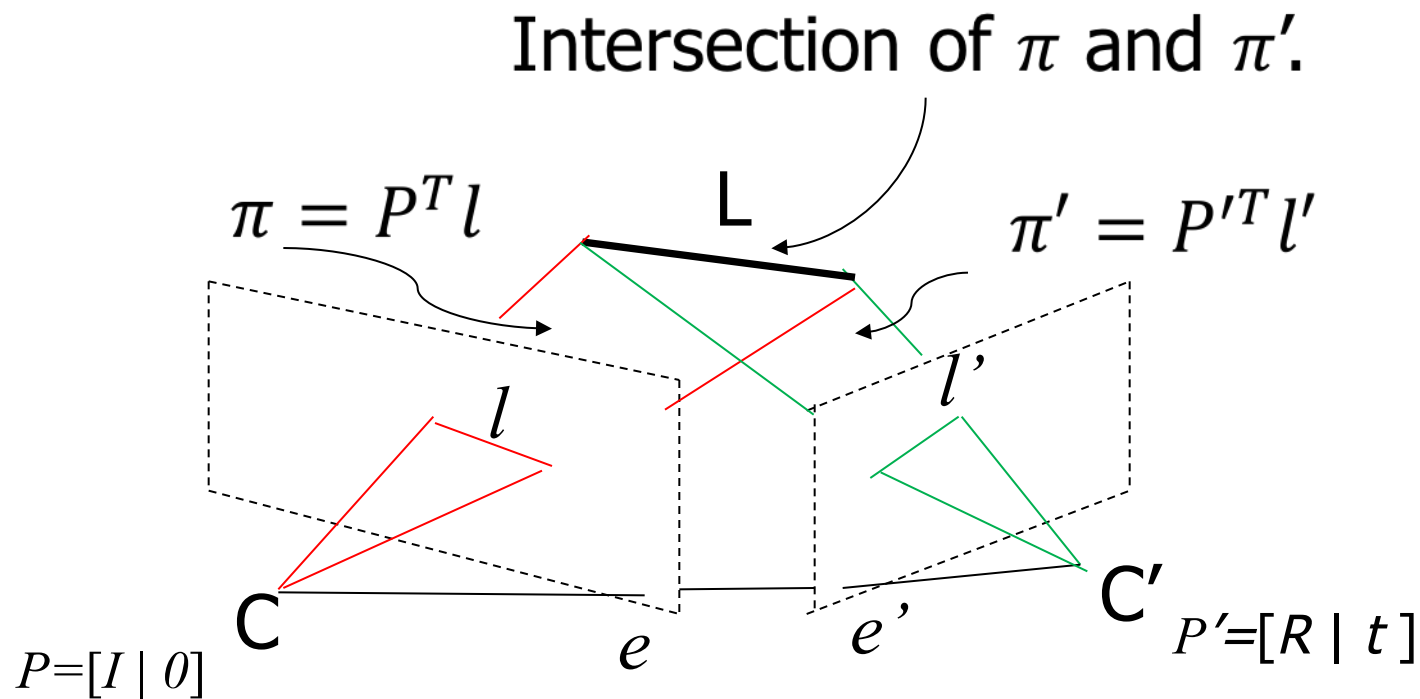
$$x_2 \times P_2 X = 0$$

$$x_3 \times P_3 X = 0$$

6 equations
3 unknowns.



Line reconstruction



A convenient way of representing 3D line.

$$L = \begin{bmatrix} \pi \\ \pi' \end{bmatrix}$$

Plane Induced Homography

Proof:

$$x' = P'X = [A|a]X$$

$$\text{Now, } x = PX = [I|0]X$$

$$\text{So any point in } \overrightarrow{CX} \text{ is } X = \begin{bmatrix} x \\ \rho \end{bmatrix}$$

$$\text{When it intersects } \pi, \pi^T \begin{bmatrix} x \\ \rho \end{bmatrix} = 0.$$

$$\Rightarrow v^T x + \rho = 0$$

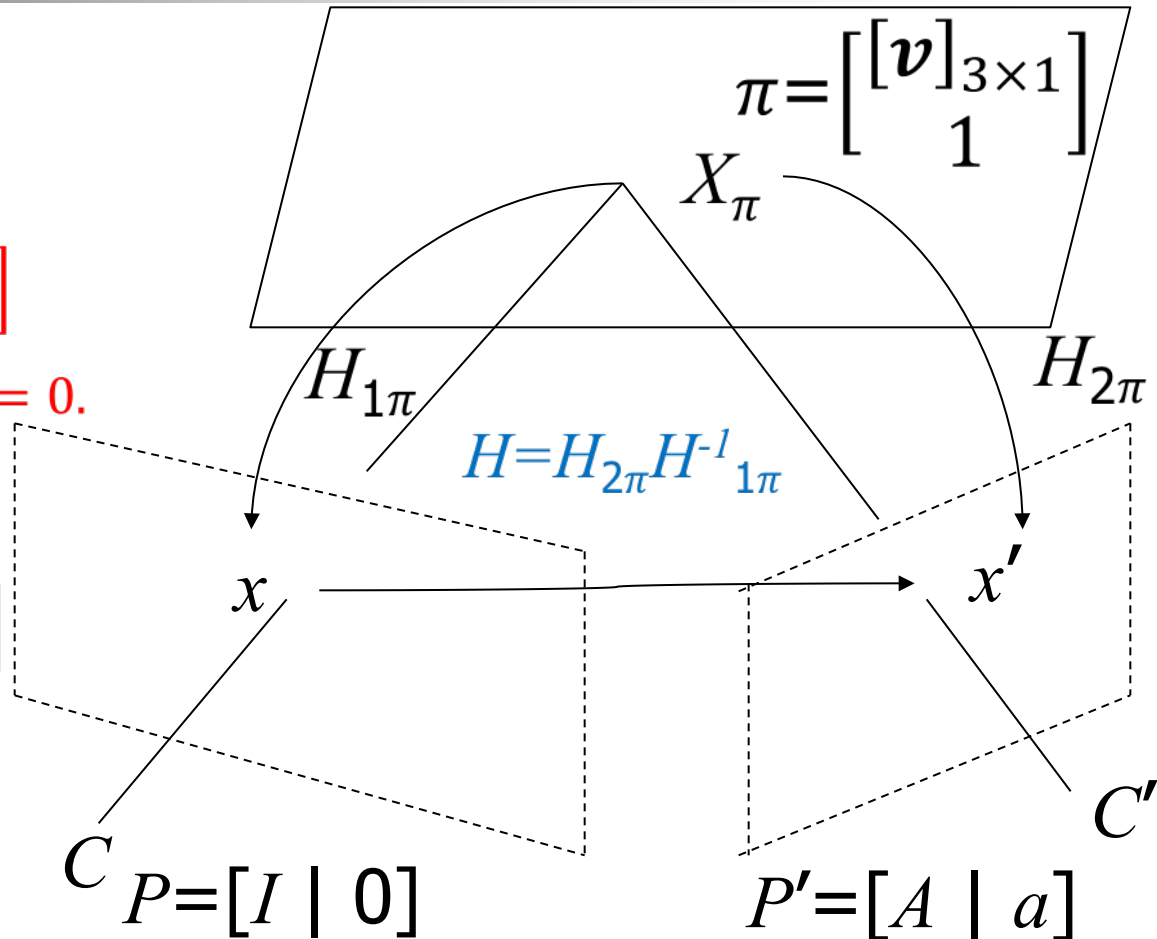
$$\Rightarrow \rho = -v^T x$$

$$\text{So, } x' = P'X = [A|a] \begin{bmatrix} x \\ -v^T x \end{bmatrix}$$

$$= Ax - a v^T x$$

$$= (A - a v^T)x$$

H

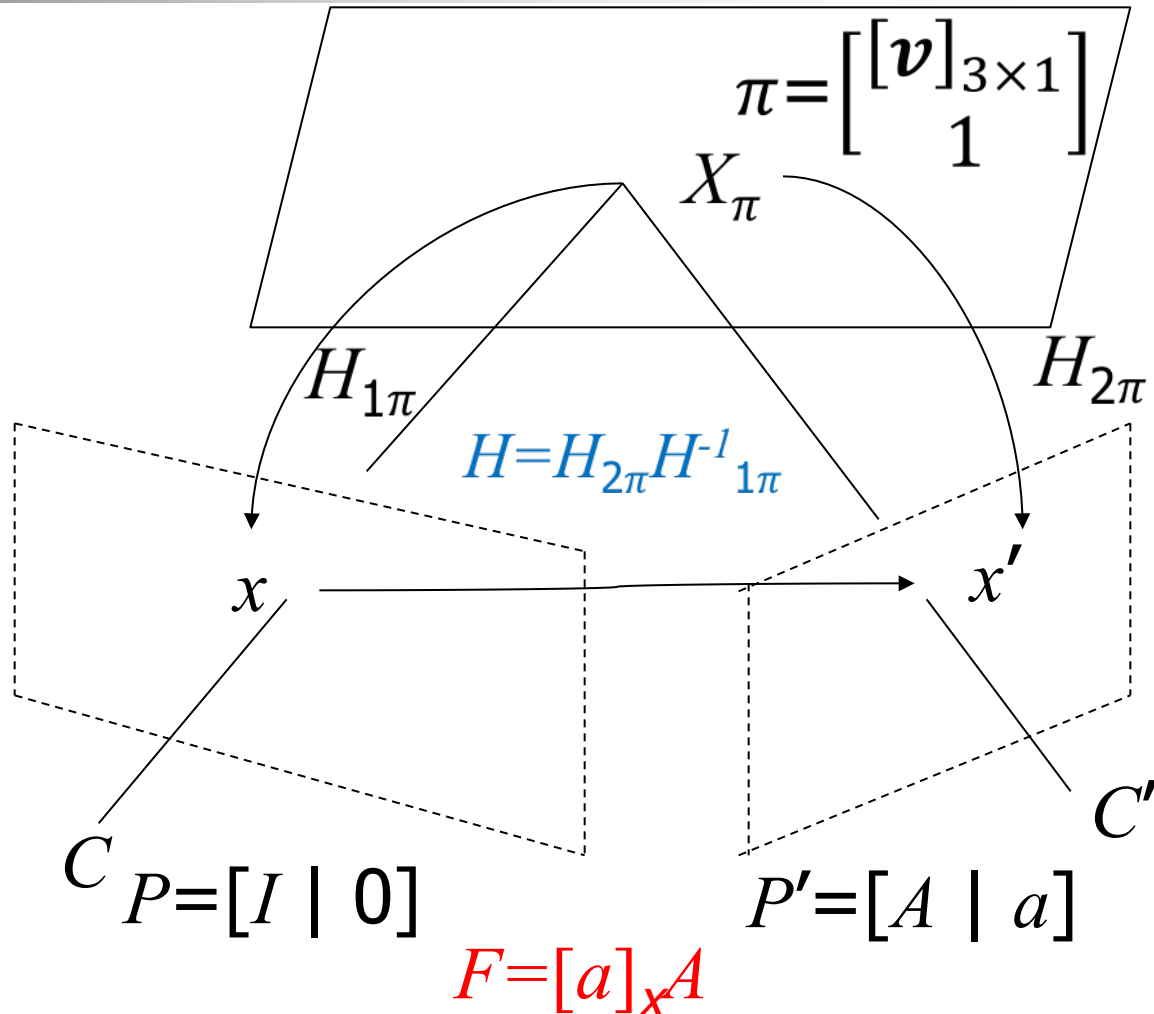


Plane Induced Homography

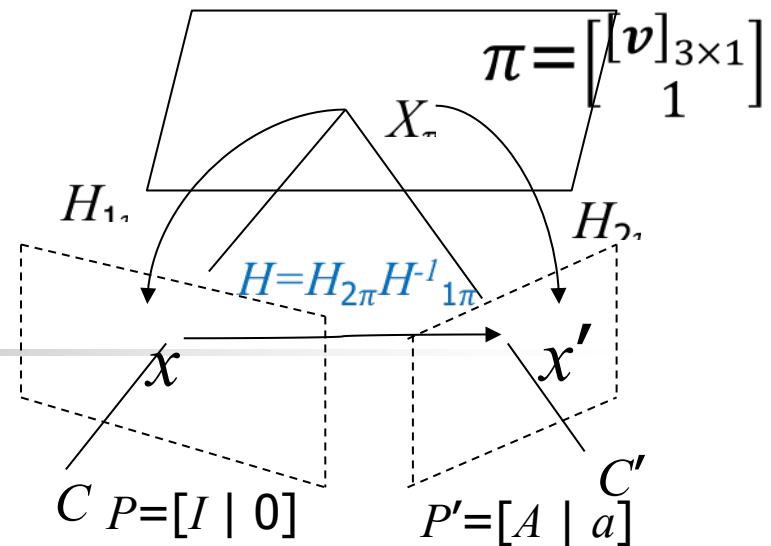
$$(A - a v^T)x$$

H

Given F , the three parameter family of homographies induced by a world plane $\pi(=[v^T \ 1])$ is $H=A - e' v^T$
Where $[e']_x A=F$.



Plane induced homography



A transformation H between two stereo images is plane induced homography if F is decomposed into $[e']_x H$. Hence, $P = [I | 0]$ & $P' = [H | e']$.

Plane at infinity

H is the transformation w.r.t. plane $[0 \ 0 \ 0 \ 1]^T$ in the camera coordinate.

Given $P = [I | 0]$, $P' = [A | a]$, & a plane induced homography H , the plane can be recovered by solving $kH = A - a\mathbf{v}^T$, (linear equations for unknowns k and \mathbf{v}).

Homography compatible stereo geometry

H is compatible iff $H^T F$ is skew symmetric, i.e.

$$H^T F + F^T H = 0$$

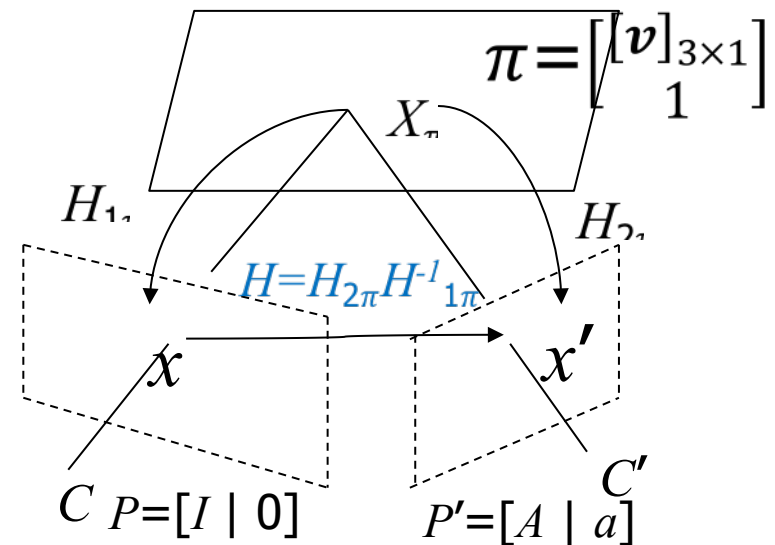
$$x'^T F x = 0$$

And, $x' = Hx$

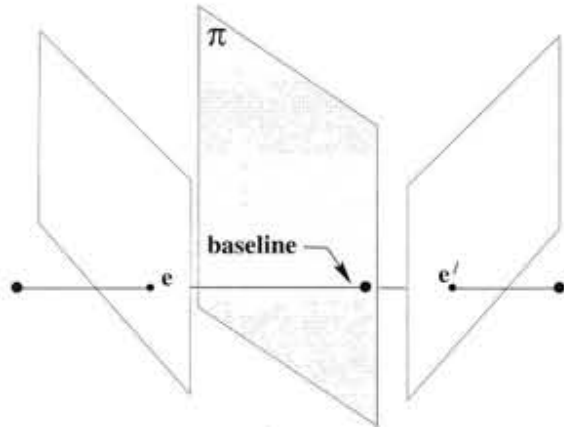
$$\Rightarrow (Hx)^T F x = 0$$

$$\Rightarrow x^T H^T F x = 0$$

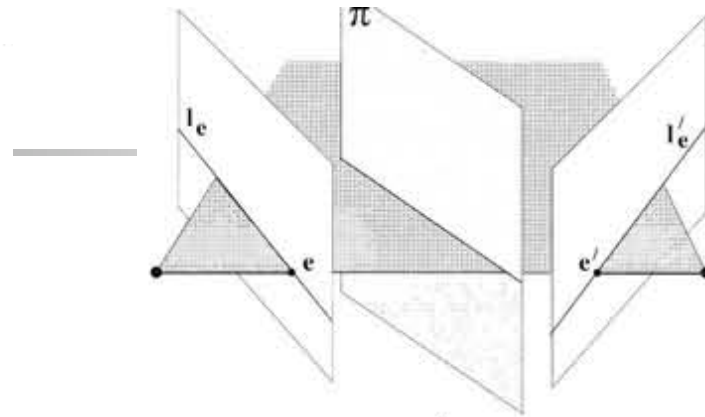
As this is true for all x ,
 $H^T F$ is skew symmetric.



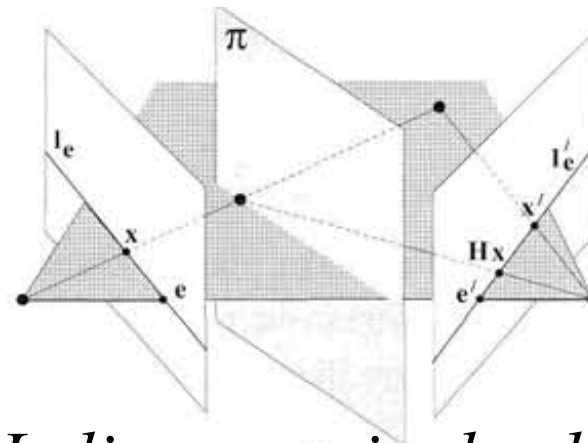
Plane induced H and epipolar constraints



Epipoles mapped by H , as $e' = He$, since they are images of the point on the plane where the baseline intersects it.

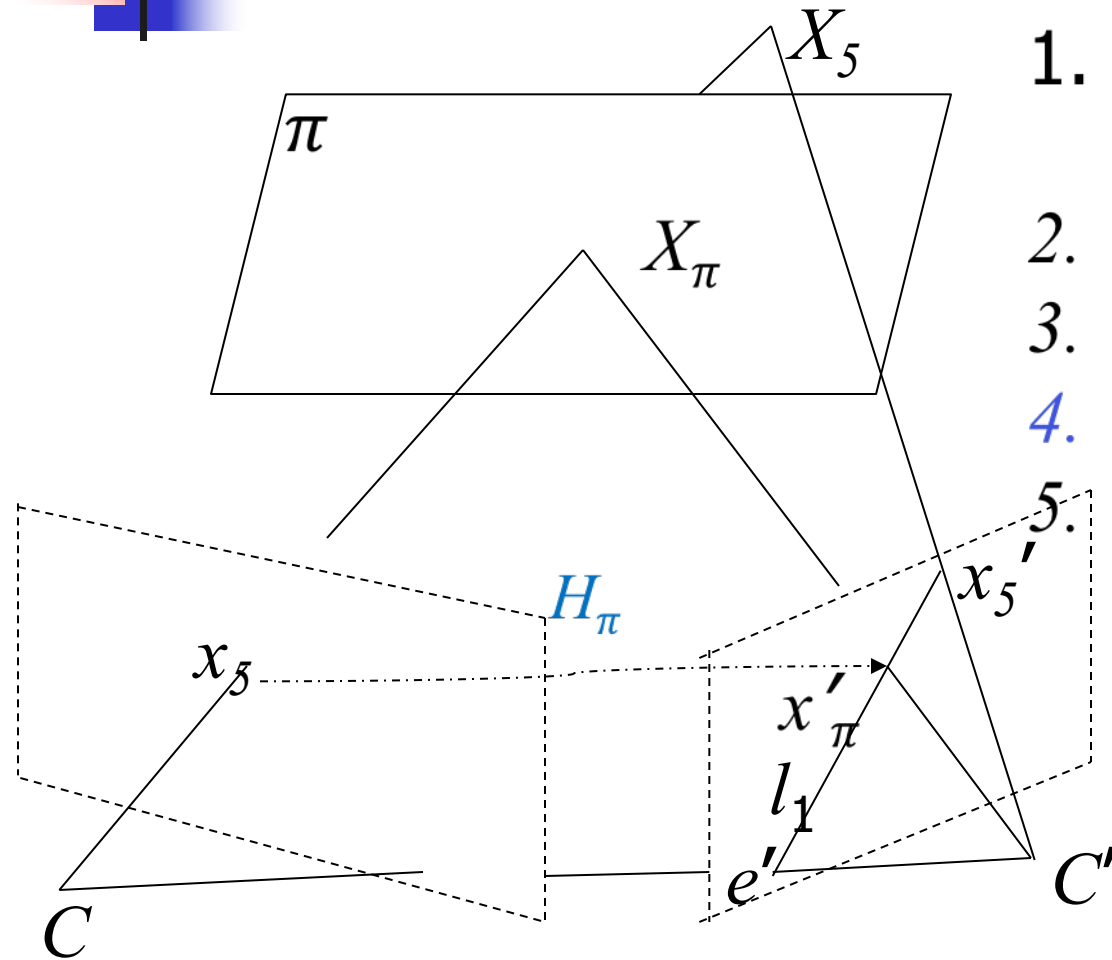


Epipolar lines are mapped by the homography as $H^T l'_e = l_e$.



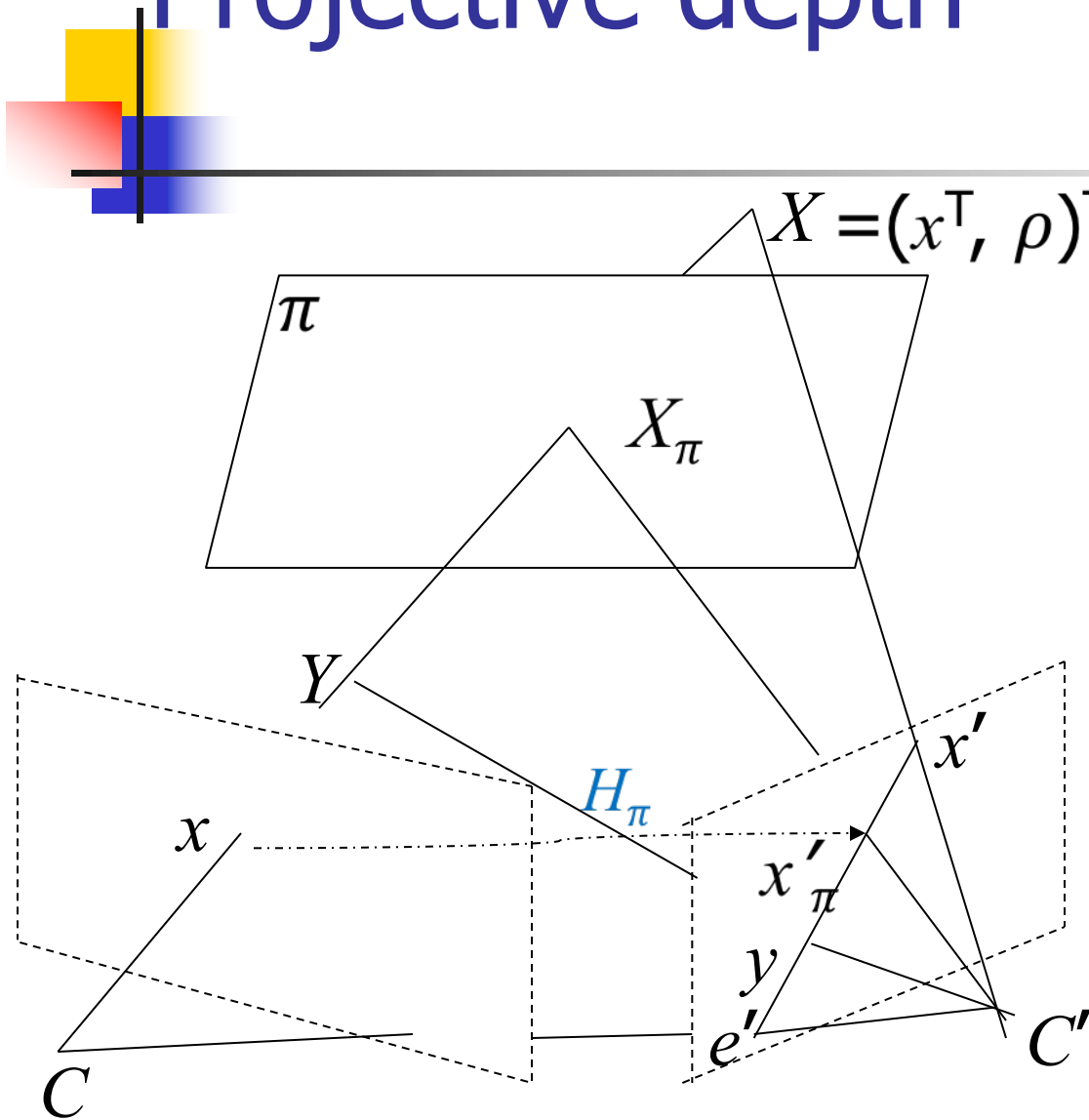
Hx lies on epipolar line l'_e so $l'_e = Fx = x' x (Hx)$.

Computing F from 6 points out of which 4 are coplanar



1. Use 4 coplanar points to compute H_π .
2. $l_1 = H_\pi(x_5) \times x_5'$
3. $l_2 = H_\pi(x_6) \times x_6'$
4. $e' = l_1 \times l_2$
5. $F = [e']_x H_\pi$.

Projective depth



Projective depth
w.r.t. the plane

$$x' = H_\pi x + \rho e'$$

For images of points
lying on the plane, $\rho=0$.

In front of the plane
towards camera
(say y), $\rho < 0$.

Behind the plane
away from camera
(say x'), $\rho > 0$.



Given F and 3 point correspondences, compute H .

First Method

1. Obtain $(P=[I|0], P'=[A|a])$ from F and construct 3 scene points, X_1, X_2 , & X_3 .
2. Obtain plane $(v^T, 1)^T$.
3. Compute $H=A-av^T$

Second Method

1. Obtain (e, e') from F .
2. Use 3 correspondences + (e, e') , to obtain H .

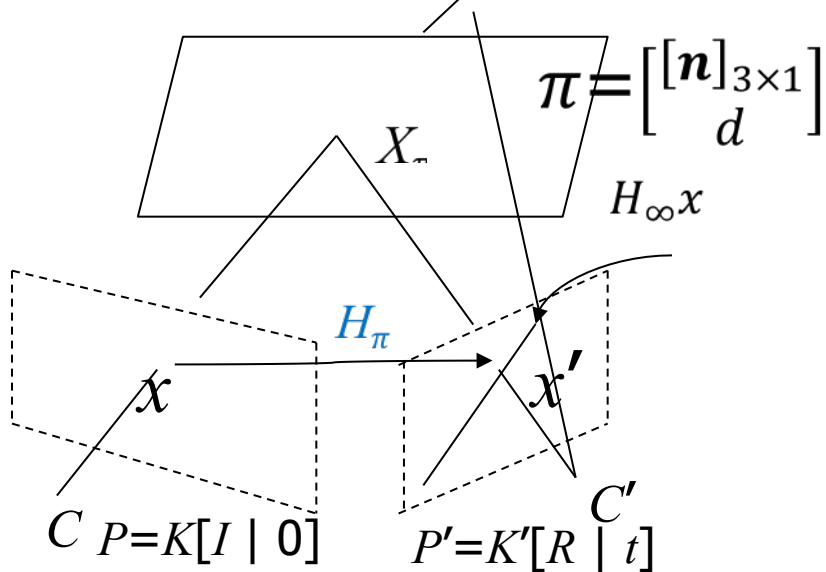
Any 3 points can bipartition the image space, w.r.t. the plane formed by them.

Infinite Homography

$$H = (A - a v^T)$$

For $P=[I | 0]$, $P'=[A | a]$
and $\text{plane}=[v^T \ 1]^T$

Homography induced by the plane at α .



$$H_\pi = (K'R - e'v^T)K^{-1}$$

where $v = \frac{n}{d}$ and $e' = K't$

$$H_\pi = K' \left(R - \frac{tn^T}{d} \right) K^{-1}$$

$$= K'RK^{-1} - K' \frac{tn^T}{d} K^{-1}$$

As $d \rightarrow \infty$, $H_\infty = K'RK^{-1}$

As $Z \rightarrow \infty$,
 x' is the image of point on π_∞ .

$$x' = K'RK^{-1}x + \frac{K't}{Z}$$

Vanishing point $\rightarrow H_\infty x + \frac{K't}{Z}$



H_α and Vanishing points

- H_α maps vanishing points between two images.
- H_α can be computed by identifying three non-collinear vanishing points given F or from 4 vanishing points.
- Let $P=[M | m]$, $P'=[M' | m']$, $X=[x_\alpha^T 0]^T$ (a point at infinity).

$$x=PX=M x_\alpha$$

$$x'=P'X=M' x_\alpha$$

$$x'=M'M^{-1}x \rightarrow H_\alpha=M'M^{-1}$$

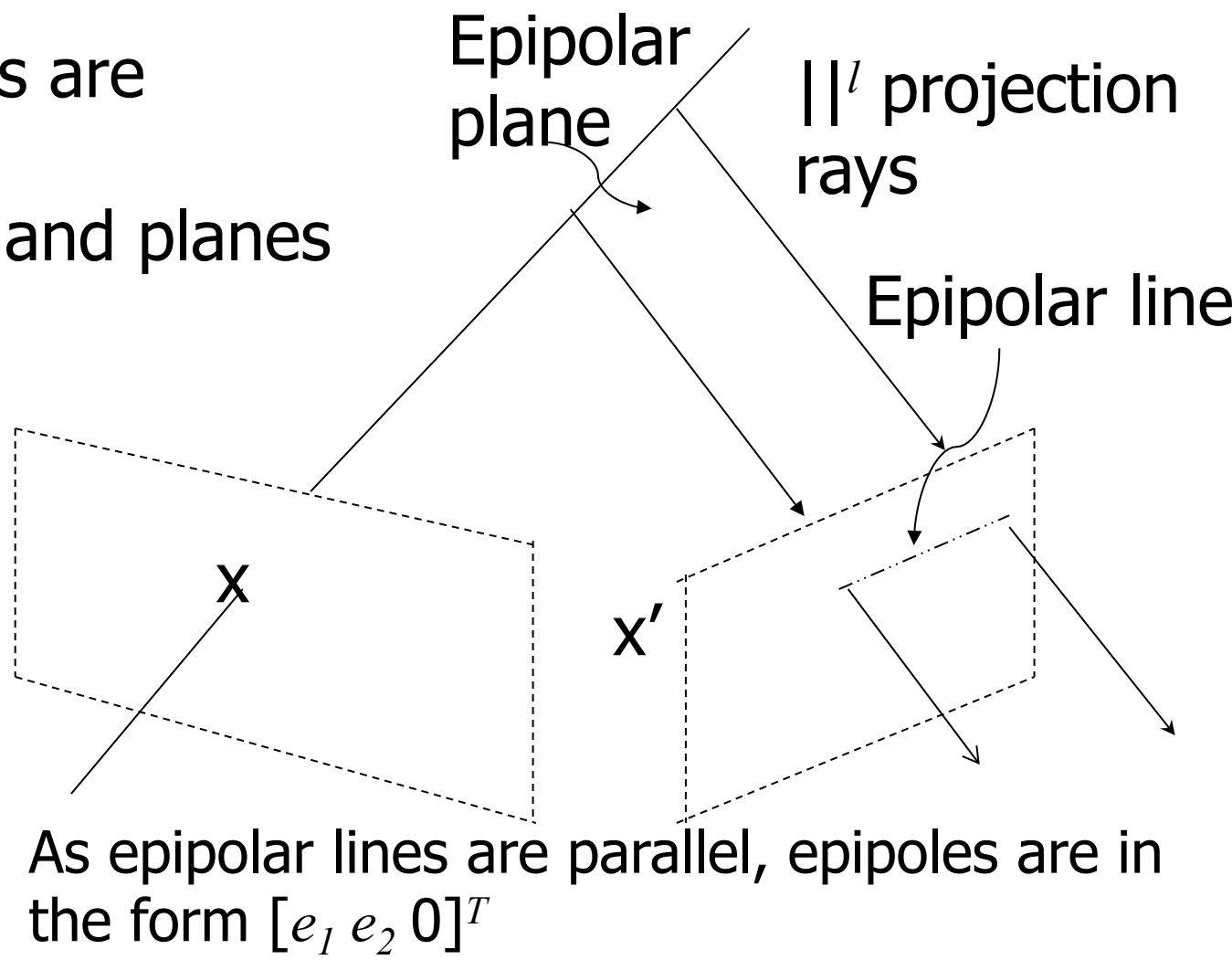
Affine epipolar geometry

- Projection rays are parallel.
- Epipolar lines and planes are parallel.

Form of F :

$$\begin{bmatrix} 0 & 0 & a \\ 0 & 0 & b \\ e & d & c \end{bmatrix}$$

a, b, c, d, e all non-zero.

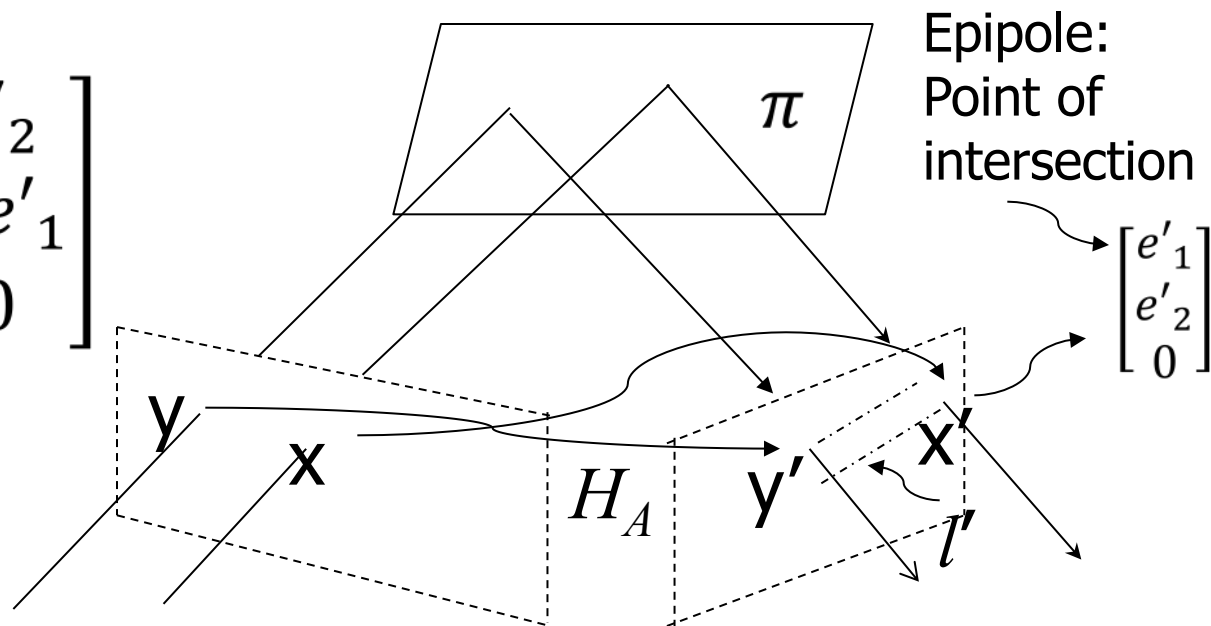


Affine stereo

$$\begin{bmatrix} 0 & 0 & a \\ 0 & 0 & b \\ e & d & c \end{bmatrix} \begin{matrix} l' = e' \times H_A x \\ = [e']_{\times} H_A x \\ \Rightarrow F_A = [e']_{\times} H_A \end{matrix}$$

$$[e']_{\times} = \begin{bmatrix} 0 & 0 & e'_2 \\ 0 & 0 & -e'_1 \\ -e'_2 & e'_1 & 0 \end{bmatrix} = \begin{bmatrix} \mathbf{0} & \mathbf{b} \\ -\mathbf{b}^T & 0 \end{bmatrix}$$

$$H_A = \begin{bmatrix} A & \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix}$$



F_A : d.o.f.: 4

$$F_A = \begin{bmatrix} 0 & \mathbf{b} \\ -\mathbf{b}^T A & -\mathbf{b}^T \mathbf{t} \end{bmatrix} = \begin{bmatrix} 0 & 0 & e'_2 \\ 0 & 0 & -e'_1 \\ e & d & c \end{bmatrix}$$

Left epipole: $[-d \ e \ 0]^T$

Right epipole: $[-b \ a \ 0]^T$



Estimating F_A

$$F_A = \begin{bmatrix} 0 & 0 & a \\ 0 & 0 & b \\ e & d & c \end{bmatrix}$$

Epipolar lines:

$$l' = F_A \mathbf{x} = [a \quad b \quad ex + dy + c]^T$$

$$l = F_A^T \mathbf{x}' = [e \quad d \quad ax' + by' + c]^T$$

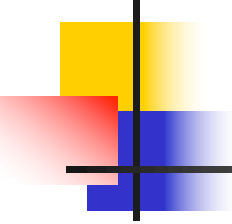
Point correspondence: *Reduced to a single linear equation.*

$$\mathbf{x}'^T F_A \mathbf{x} = 0 \Rightarrow ax' + by' + ex + dy + c = 0$$

$$[A]_{N \times 5} \mathbf{f}_{5 \times 1} = 0 \quad \text{Solve using DLT.}$$

Minimum 4 point correspondences required to get F_A .

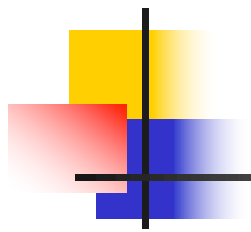
Singularity constraint is satisfied by the structure of F_A .



Estimating F_A (another approach)

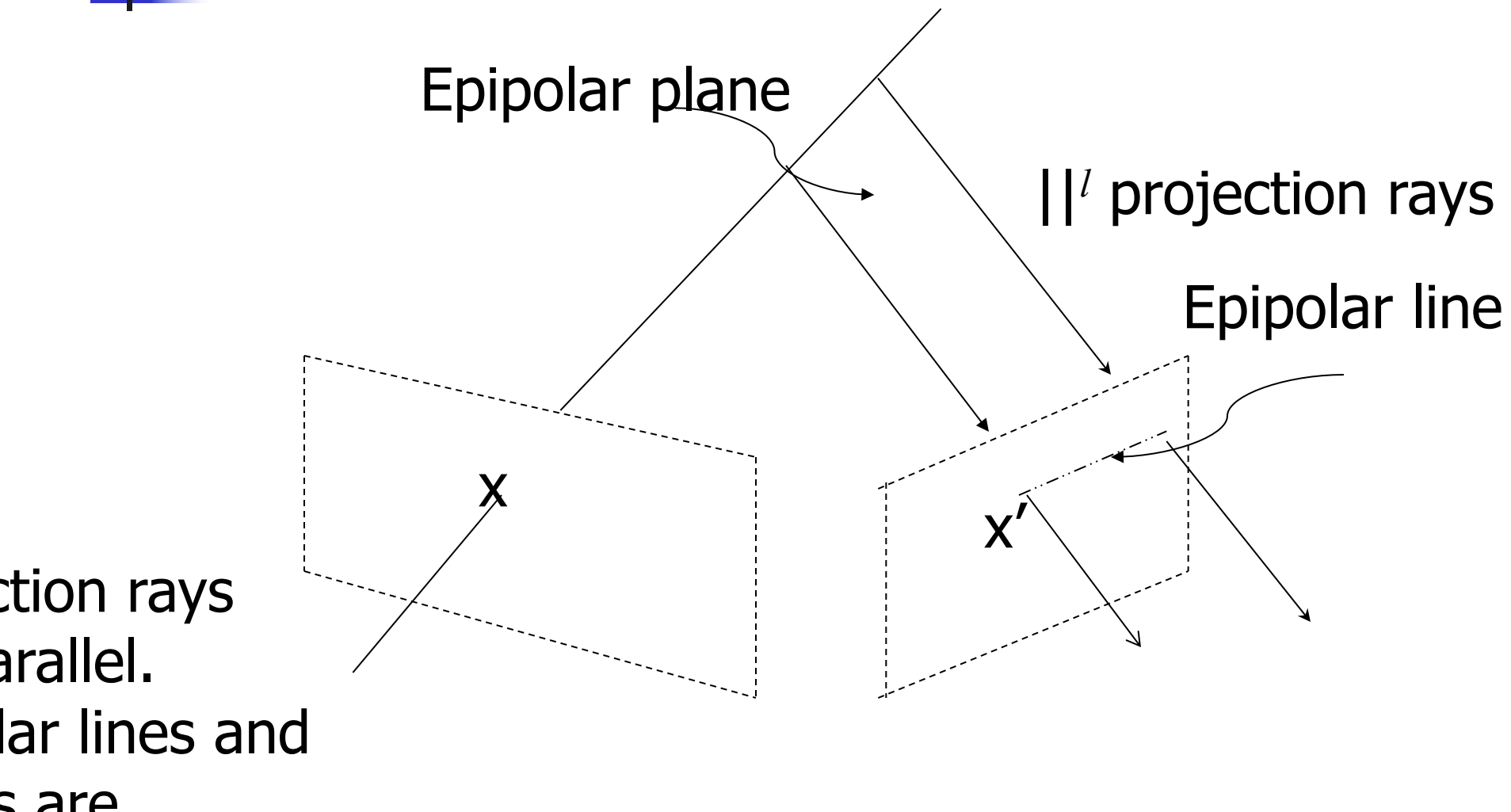
$$F_A = \begin{bmatrix} 0 & 0 & a \\ 0 & 0 & b \\ e & d & c \end{bmatrix}$$

1. Compute H_A using 3 point-correspondences.
2. $l' = H_A x_4' \times x_4'$ (say, $[l_1 \ l_2 \ l_3]^T$)
3. Get e' from l' as $[l_2 \ -l_1 \ 0]$
4. $F_A = [e']_{\times} H_A$

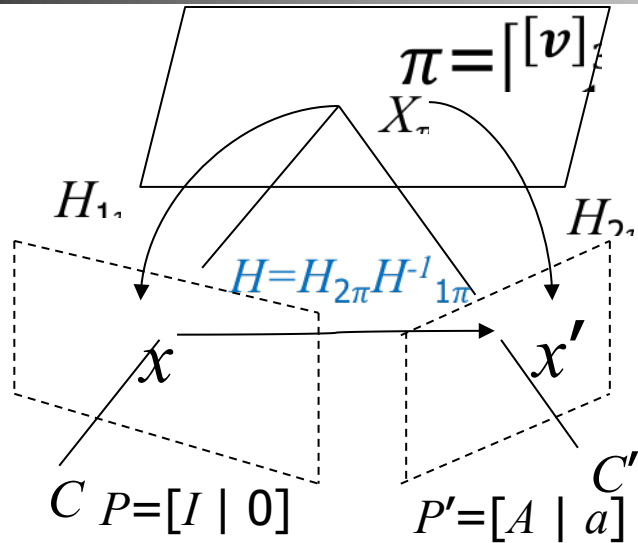


Thank you!

Affine epipolar geometry

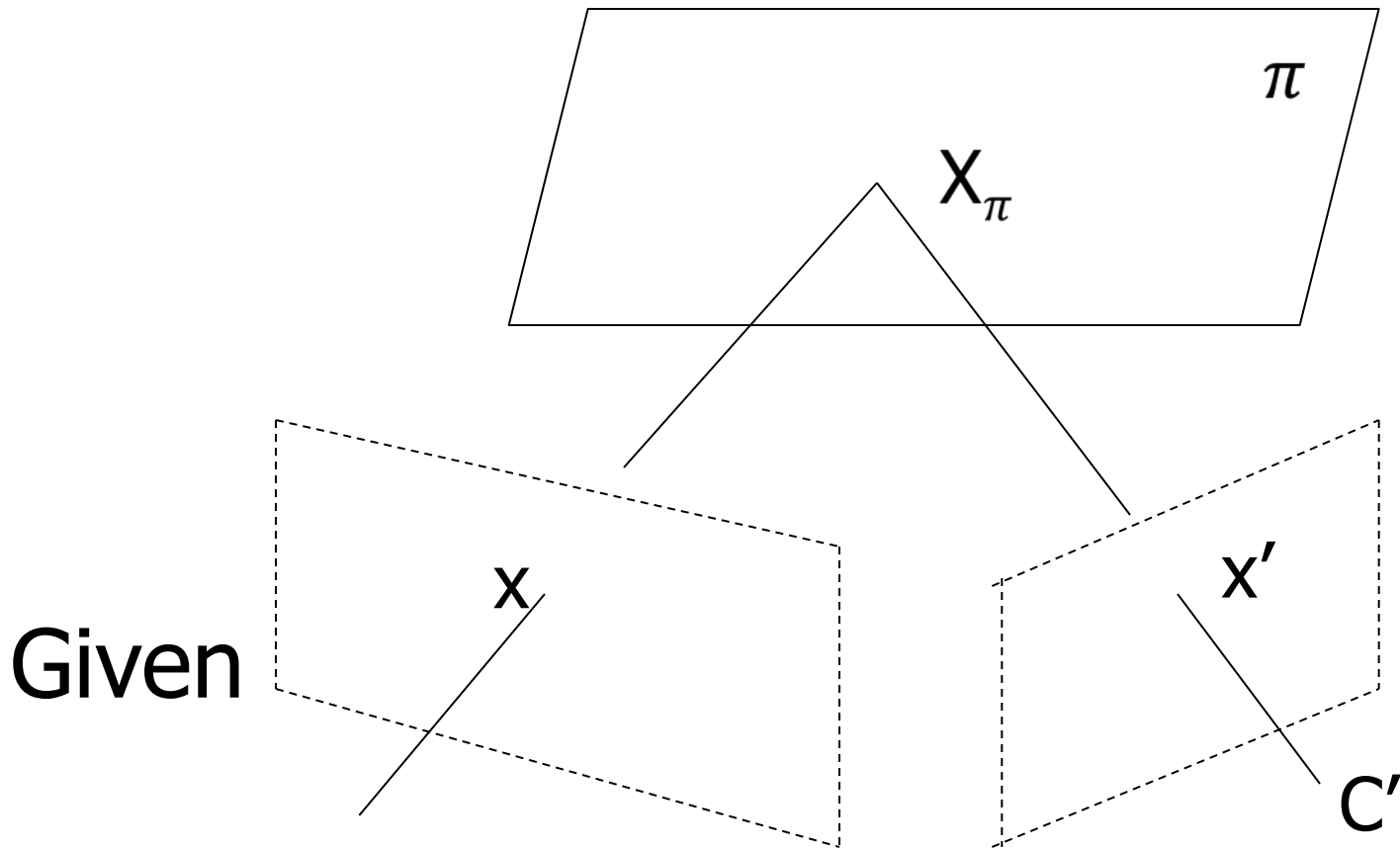


Plane Induced Homography

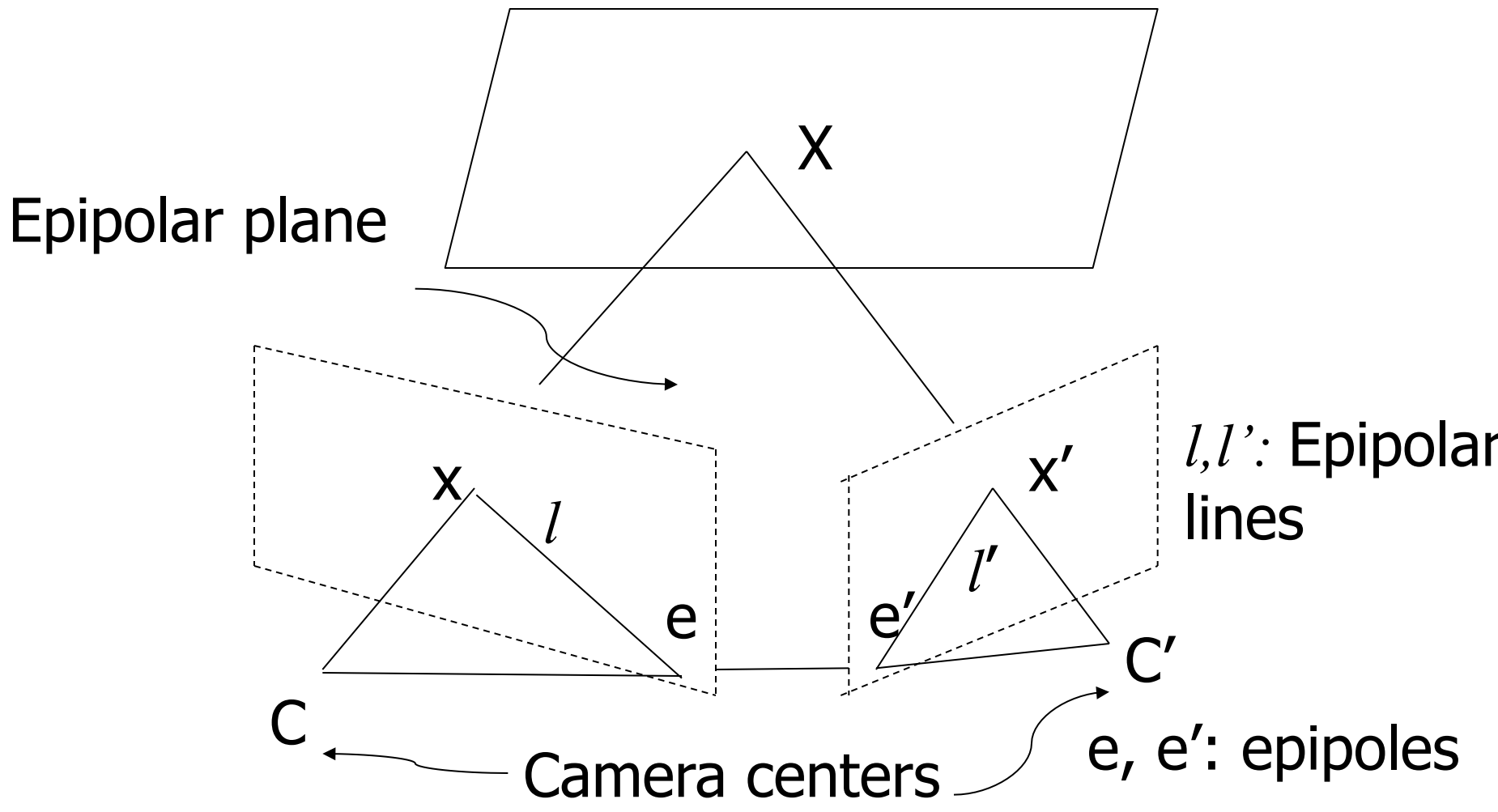




Plane Induced Homography

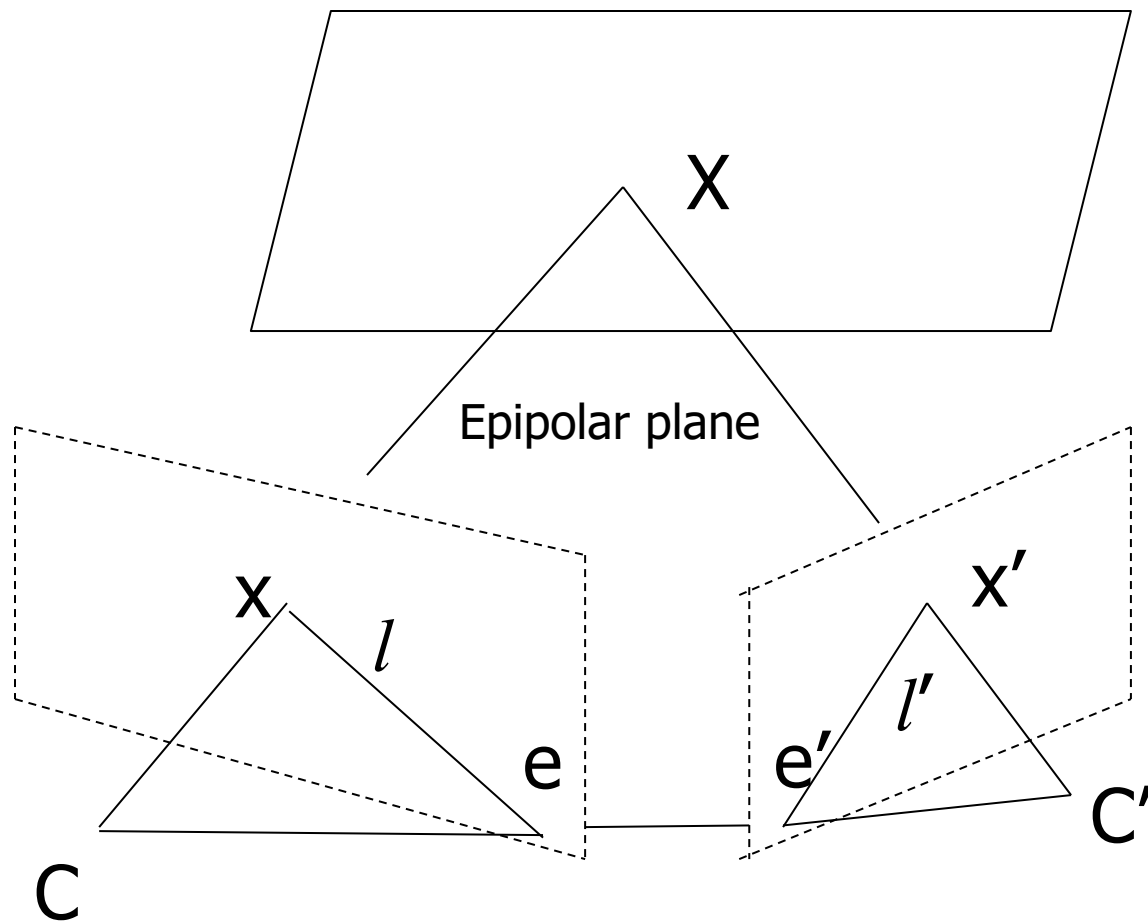


Stereo Set-up



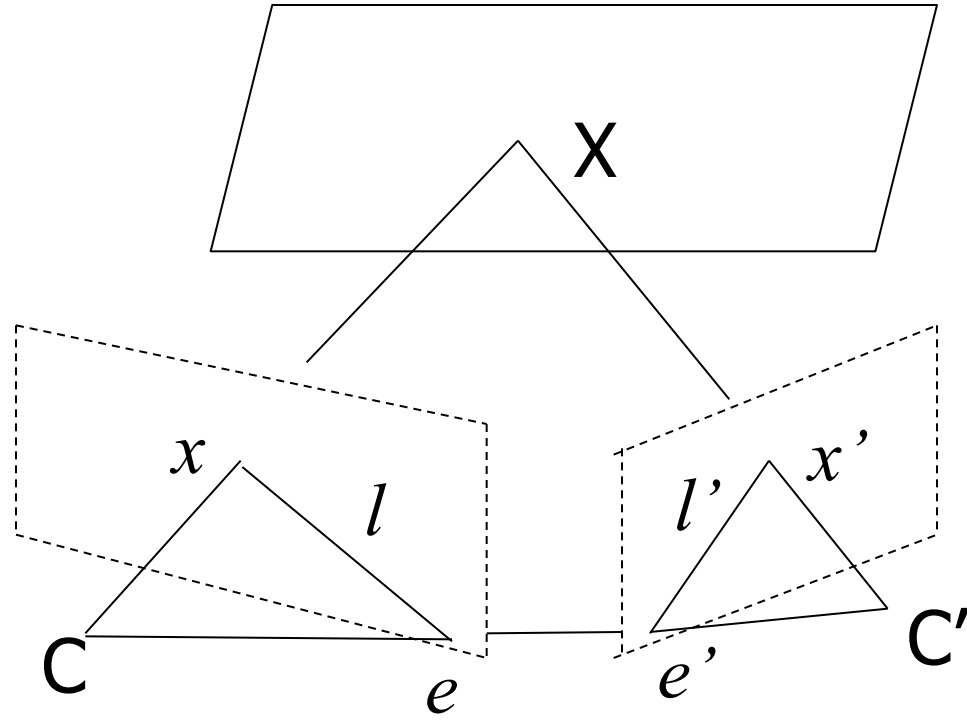


Epipolar Geometry

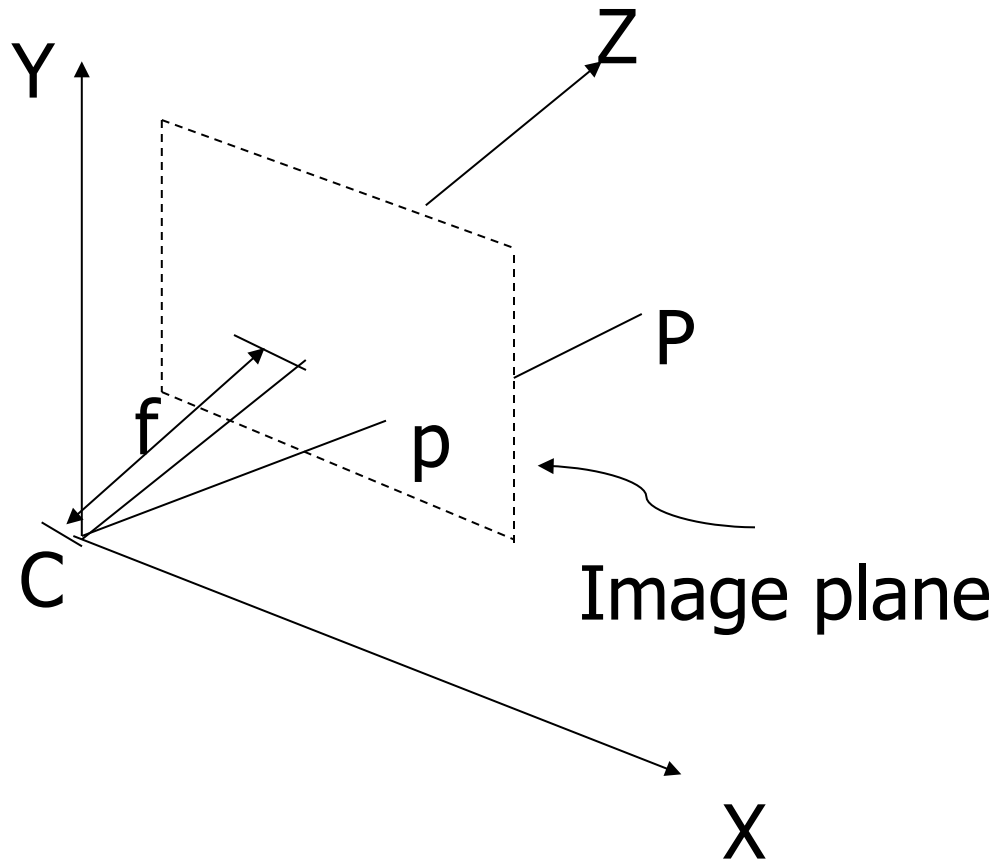




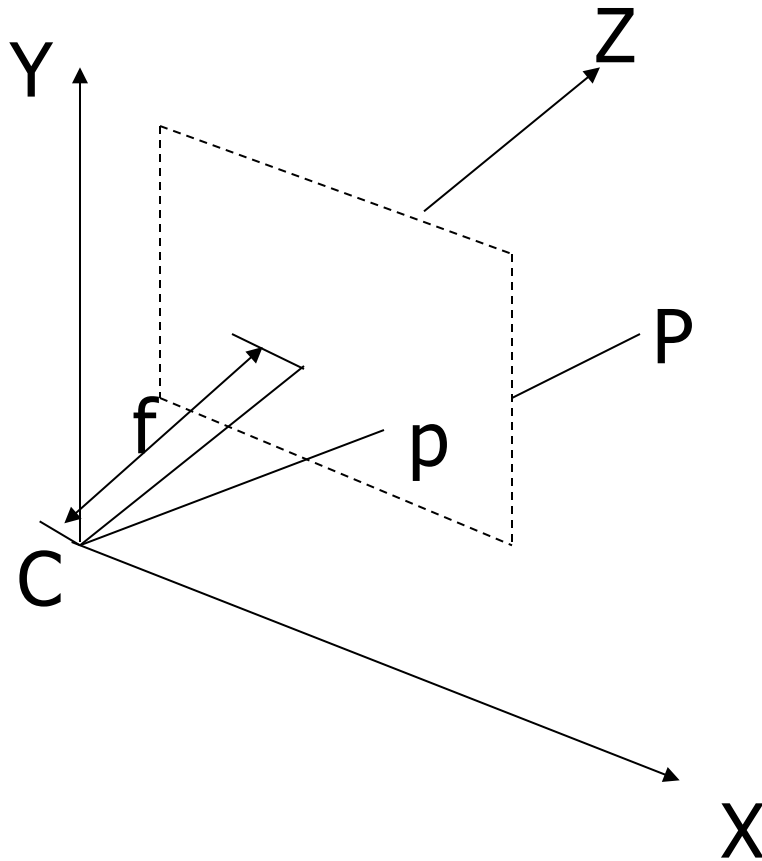
Epipolar Geometry



Pinhole camera

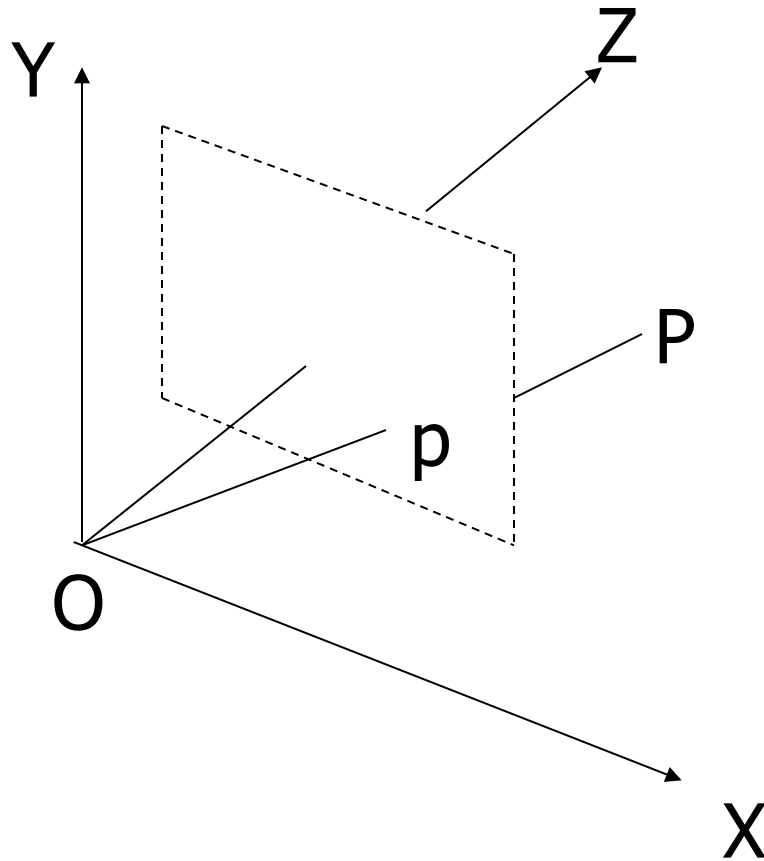


Pinhole camera





Pinhole camera





Computing vanishing line

- Identify groups of sets of parallel lines in a plane at different directions.
- Obtain their vanishing points.
- Get the line among them.