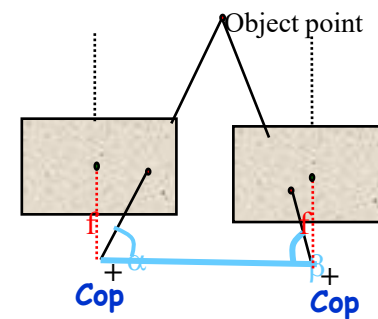


Geometry

Class 5

Last Class

- Traingulation in a rectified image
- Correspondence
- From 3D to 2D: projection matrix



Stereo Main Issues

- Correspondence
 - Geometric constraints
 - Local information
 - Global information
- Calibration
- Reconstruction (triangulation)

3D to 2D: $\tilde{p} = M\tilde{P}$

- Object + image are in the camera's coordinate system: $M = M_{int}$

$$M_{int} = \begin{pmatrix} s_x f & 0 & o_x & 0 \\ 0 & s_y f & o_y & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

- Object – general coordinate system:

$$M = M_{int} M_{ext}$$

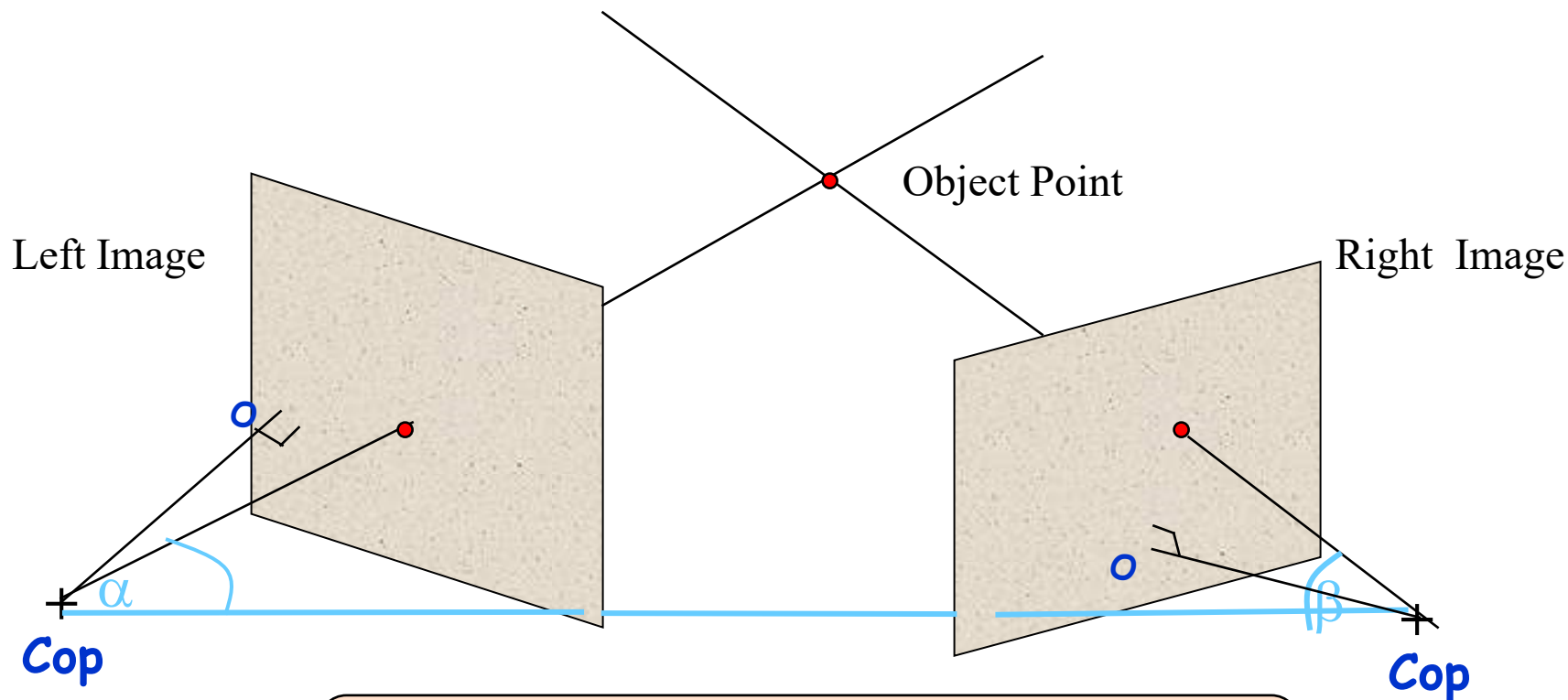
- $M_{ext} = \begin{pmatrix} R & -RT \\ 0 & 0 & 0 & 1 \end{pmatrix}$

Next

- Strereo – general case
- Epipolar geometry
- Rectification
- Other type of stereo images
- Homography
- RANSAC

Back to Stereo

General Case Triangulation



Possible to use trigonometry but
projective geometry is nicer & easier

Triangulation

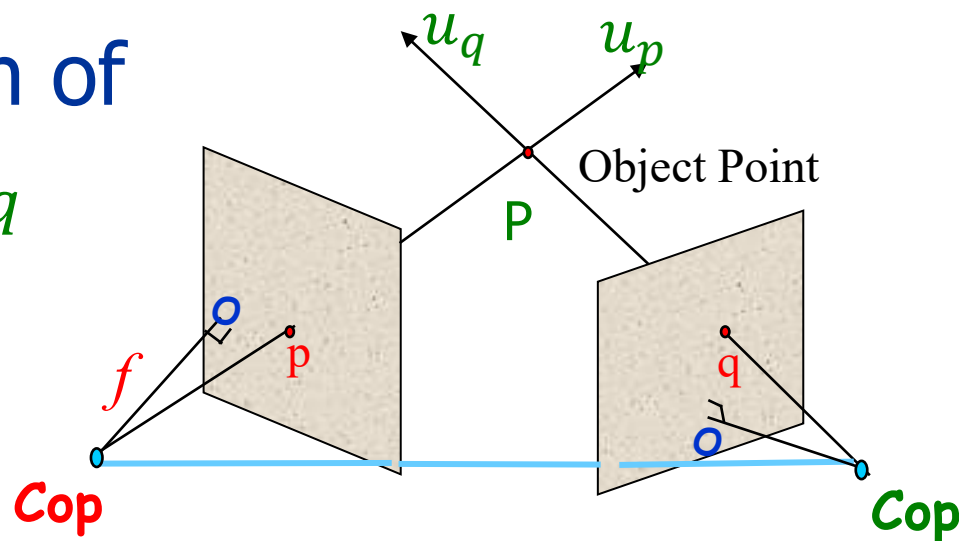
- We are looking for P such that:

$$\tilde{p} = M_L \tilde{P} \text{ and } \tilde{q} = M_R \tilde{P}$$

How to compute?

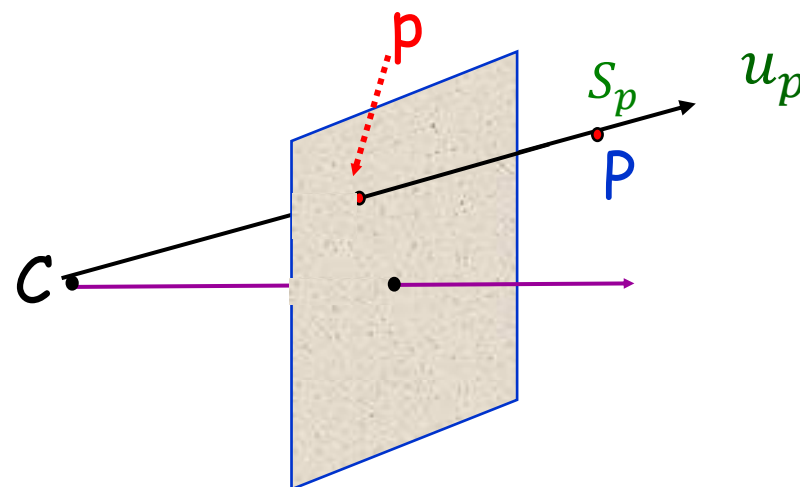
- P is the intersection of two lines u_p and u_q

How to compute?



Back Projection Ray

- Given M and p
- Let S_p be a point in 3D
s.t. $\tilde{p} = M \tilde{S}_p$
- The ray u_p is given by:



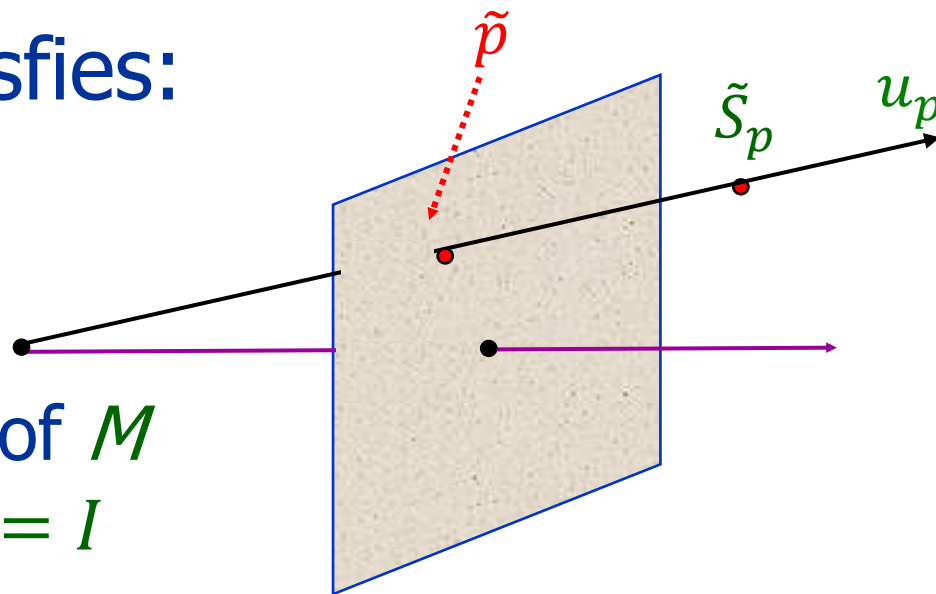
- $u_p = \{P_\lambda \mid P_\lambda = C + \lambda(S_p - C)\}$
- Denote $u_p(\lambda) = C + \lambda \overrightarrow{CS_p}$ a point on u_p
where $\overrightarrow{CS_p} = S_p - C$

Note: S_p and C are in Euclidean coordinates

Back Projection Ray

- Given M and p
- A point, $S_p \in u_p$ satisfies:

- $\tilde{p} = M \tilde{S}_p$
- $\tilde{S}_p = M^+ \tilde{p}$
- M^+ : pseudo inverse of M
 $MM^+ = I$
- $M^+ = M^T (MM^T)^{-1}$
- $MM^T (MM^T)^{-1} \tilde{p} = \tilde{p}$



Pseudoinverse

- Let A be a $n \times m$ matrix where $n < m$
- Define $A^+ = A^T (A A^T)^{-1}$
- $AA^+ = I$

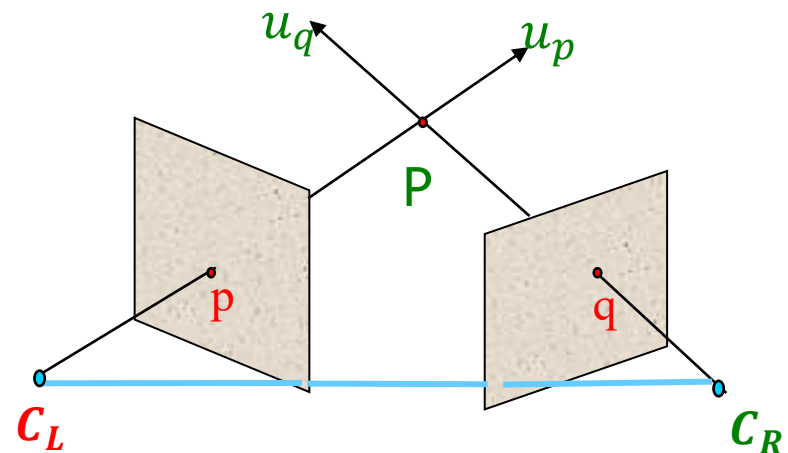
Triangulation

- We are looking for P such that:

$$\tilde{p} = M_L \tilde{P} \text{ and } \tilde{q} = M_R \tilde{P}$$

- P is the intersection of the lines u_p and u_q

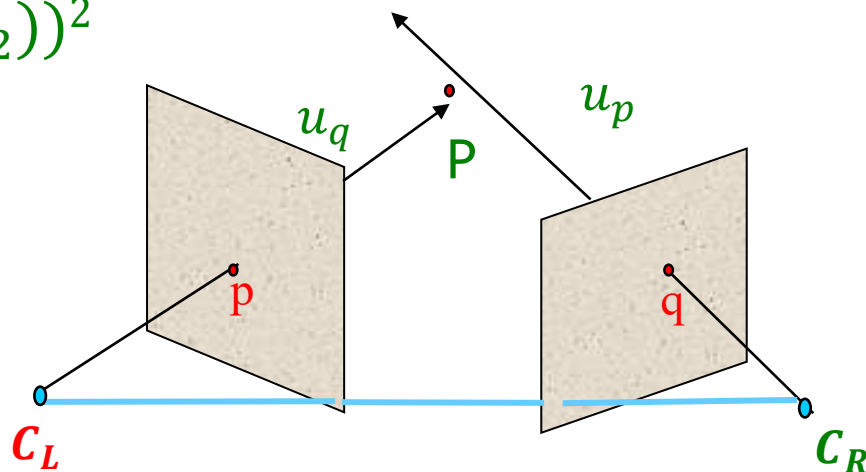
- $P = u_p(\lambda_1) = u_q(\lambda_2)$



Triangulation

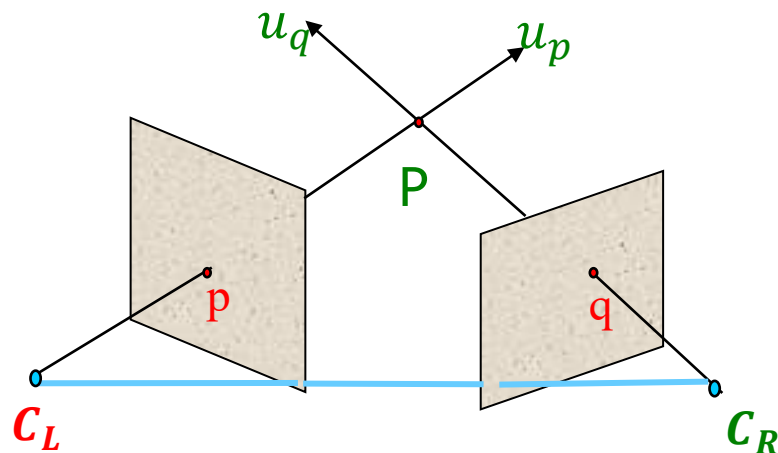
When u_p and u_q do not intersect,
we are looking for P that minimizes:

- $d(P, u_p(\lambda_1))^2 + d(P, u_q(\lambda_2))^2$
- Other possibilities...



Triangulation

- We look for λ_1 and λ_2 such that
 - $u_p(\lambda_1) = u_q(\lambda_2)$
 - It follows that $P = u_p(\lambda_1)$
- Some algebraic manipulation:
 - $$\begin{aligned} u_p(\lambda_1) &= C_L + \lambda_1 \overrightarrow{C_L S_p} \\ &= C_L + \lambda_1 \vec{v}_p \end{aligned}$$
 - $$\begin{aligned} u_q(\lambda_2) &= C_R + \lambda_2 \overrightarrow{C_R S_q} \\ &= C_R + \lambda_2 \vec{v}_q \end{aligned}$$
 - $$C_L + \lambda_1 \vec{v}_p = C_R + \lambda_2 \vec{v}_q$$



Denote $u_p(\lambda) = C + \lambda \overrightarrow{CS}$ a point on u_p where $\overrightarrow{CS} = S_p - C$

Triangulation

- $C_L - C_R = \lambda_2 \vec{v}_r - \lambda_1 \vec{v}_L$
- In matrix notation: $(-v_L, v_R) \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} = (C_L - C_R)$
- We search for (λ_1, λ_2) that minimizes:

$$\left\| (-v_p, v_q) \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} - (C_L - C_R) \right\|$$

- In python use: `(np.linalg.lstsq)`
- $P_L = u_p(\lambda_1), P_R = u_q(\lambda_2)$

- $P = \frac{P_L + P_R}{2}$



Triangulation: Another Minimization

- Minimize the distance of the projected point to the two image.
- Choose \tilde{P} that minimizes:

$$\min_{\tilde{P}} (|\tilde{p} - M_L \tilde{P}| + |\tilde{p} - M_R \tilde{P}|)$$

The COP

- Can we compute C from M ?
- **Claim:** The camera center, C is the null vector of M
- **Proof idea:**
 - Note that the degree of M is 3
 - Hence there exists a C s.t. $M\tilde{C} = (0,0,0)^T$
 - We will show: that C is the center of projection

Cont.

- We have to show that for any given 3D point, all points on the ray that connects it to the COP project to the same image point:
 - Consider a 3D point A and
 - Let $\tilde{a} = M \tilde{A}$
 - C is the center of projection of M if all points on the ray $P(\lambda) = C + \lambda(A - C)$ projects to \tilde{a}

Algebra ...

- $P(\lambda) = C + \lambda(A - C)$
- $P(\lambda) = (1 - \lambda)C + \lambda A$

Cartesian coordinates

- $\tilde{P}(\lambda) \cong \begin{pmatrix} (1 - \lambda)C_x + \lambda A_x \\ (1 - \lambda)C_y + \lambda A_y \\ (1 - \lambda)C_z + \lambda A_z \\ 1 \end{pmatrix}$

Homogenous coordinates

$$\cong \begin{pmatrix} (1 - \lambda)C_x + \lambda A_x \\ (1 - \lambda)C_y + \lambda A_y \\ (1 - \lambda)C_z + \lambda A_z \\ 1 - \lambda + \lambda \end{pmatrix} = (1 - \lambda)\tilde{C} + \lambda\tilde{A}$$

Cont.

It follows that:

- $M \tilde{P}(\lambda) = M(1 - \lambda)\tilde{C} + \lambda\tilde{A} = M\tilde{A} = \tilde{a}$

Q.E.D.

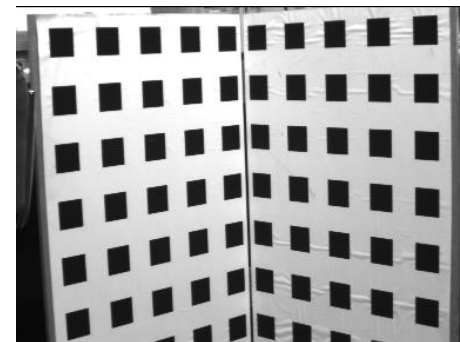
Camera Calibration

Camera Calibration

Find the camera matrices:

- Intrinsic parameters
- Extrinsic parameters

General Idea



Assumption:

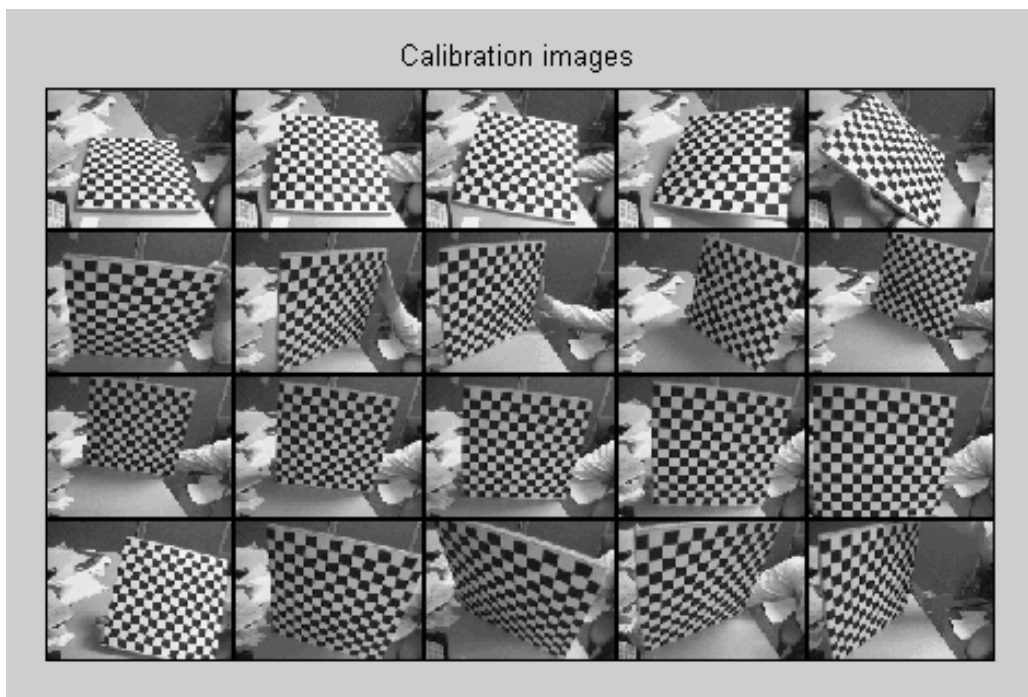
- Known 3D pattern $\{P_i\}$
- Known correspondence between image and object points $\{p_i\}$

Solution:

- Solve the linear equations: $\tilde{p}_i = M \tilde{P}_i$
 - Note: $\tilde{u} \cong \tilde{v}$ in the projective space: $\tilde{u} \times \tilde{v} = 0$

Calibration based on a planar surface

- Using several images allow to compute M and their location with respect to each other

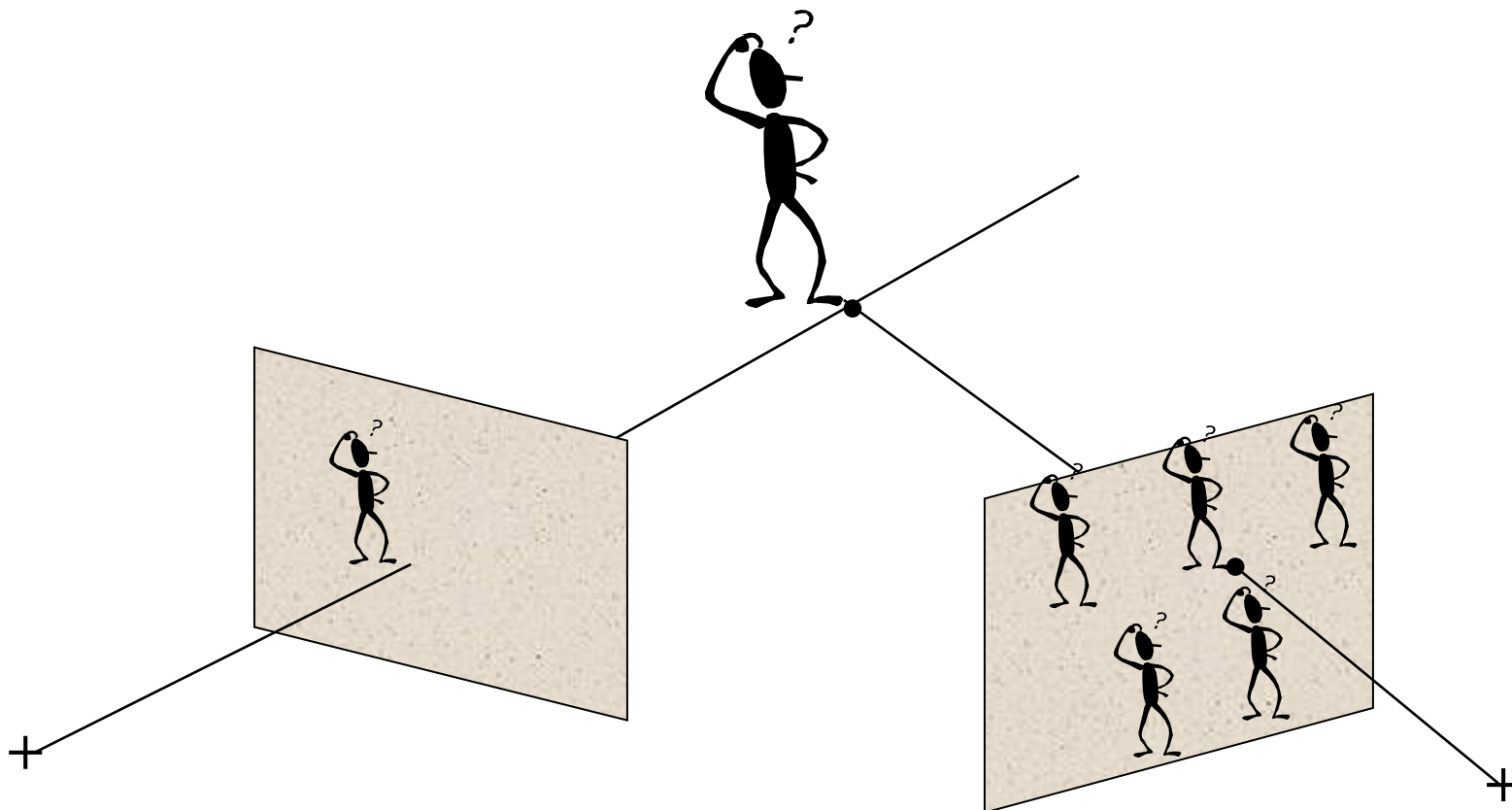


So far

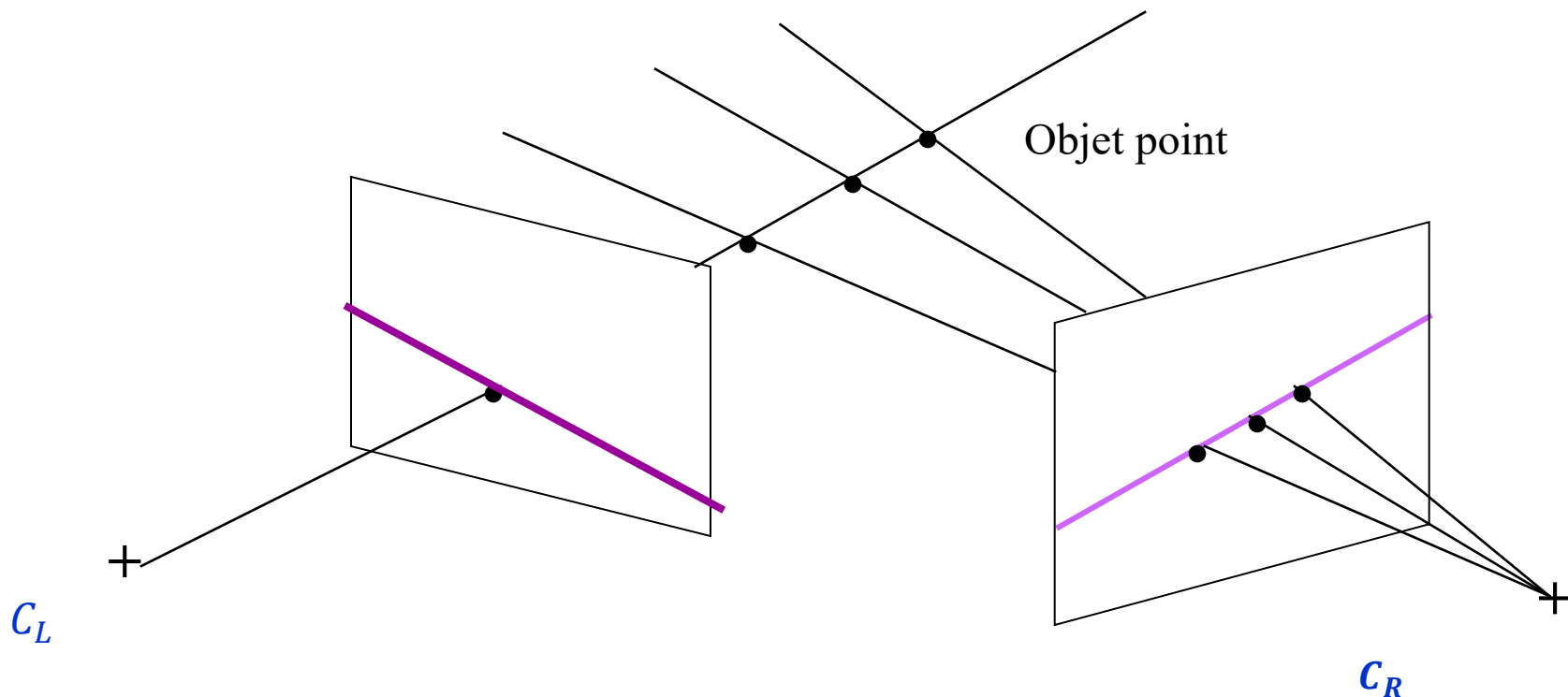
- Using geometry & algebra to compute:
 - Projection matrix
 - Traingulation
 - Calibration
- Next: correspondence

Epipolar Geometry

Correspondence: Geometry

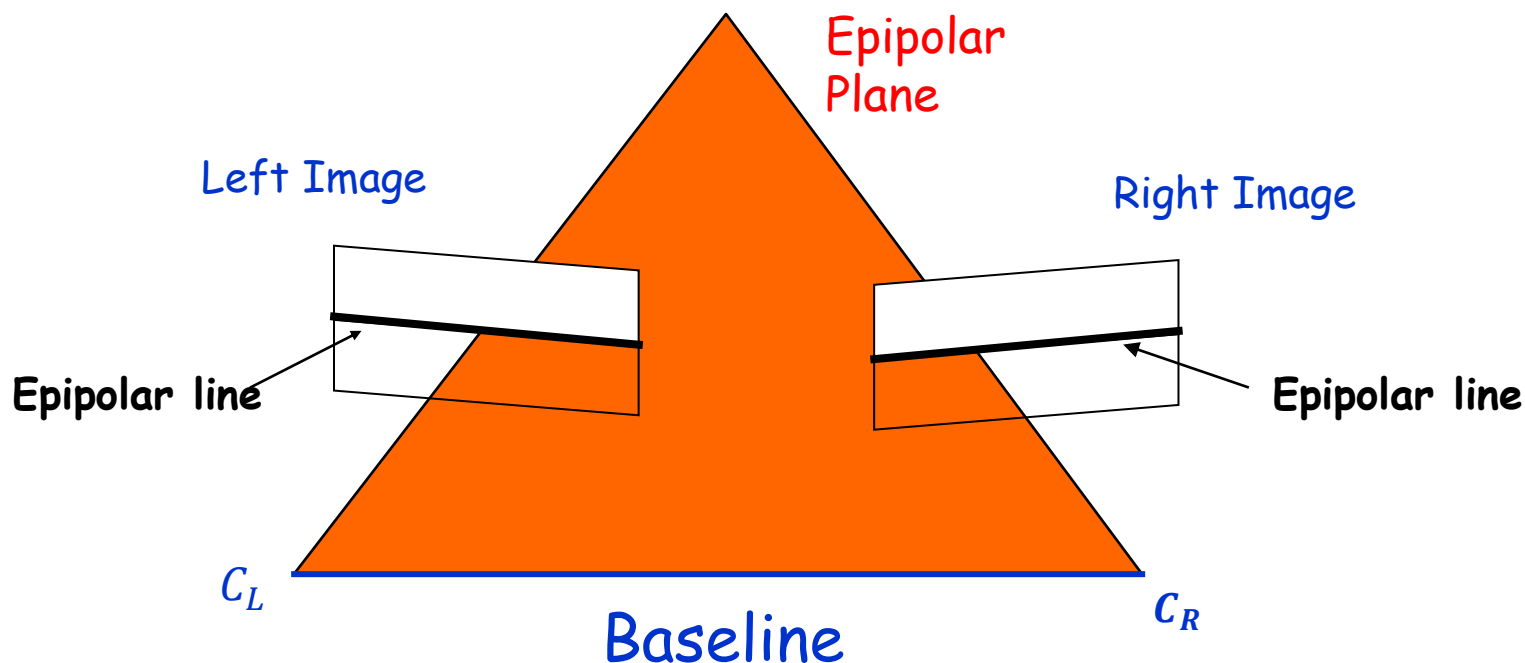


Epipolar Lines



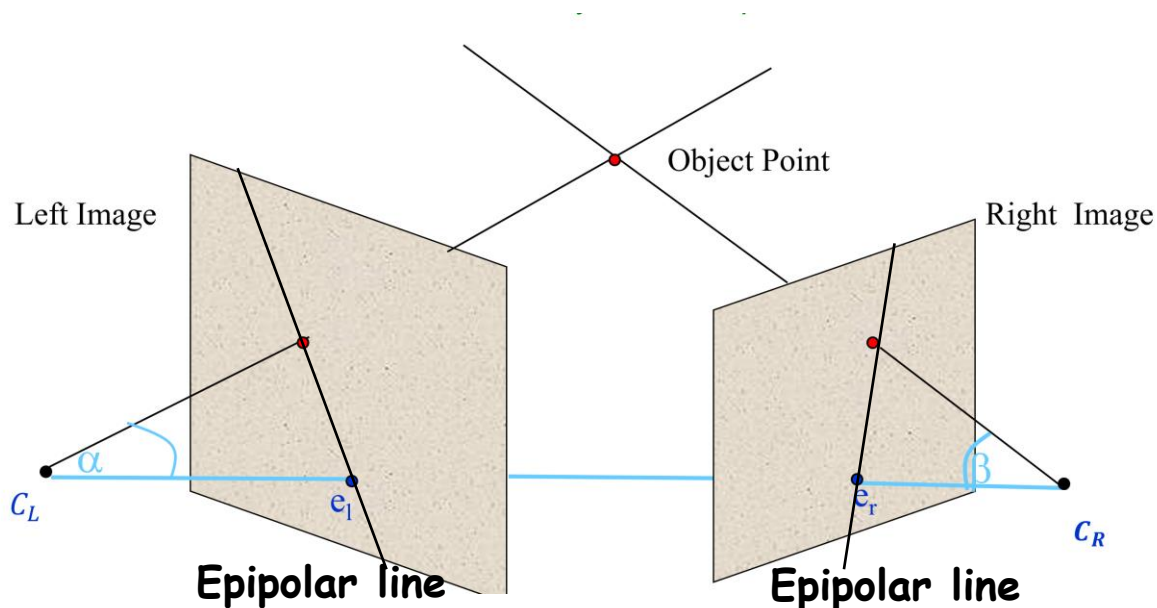
Epipolar Geometry

A pair of corresponding points must lay on corresponding epipolar lines

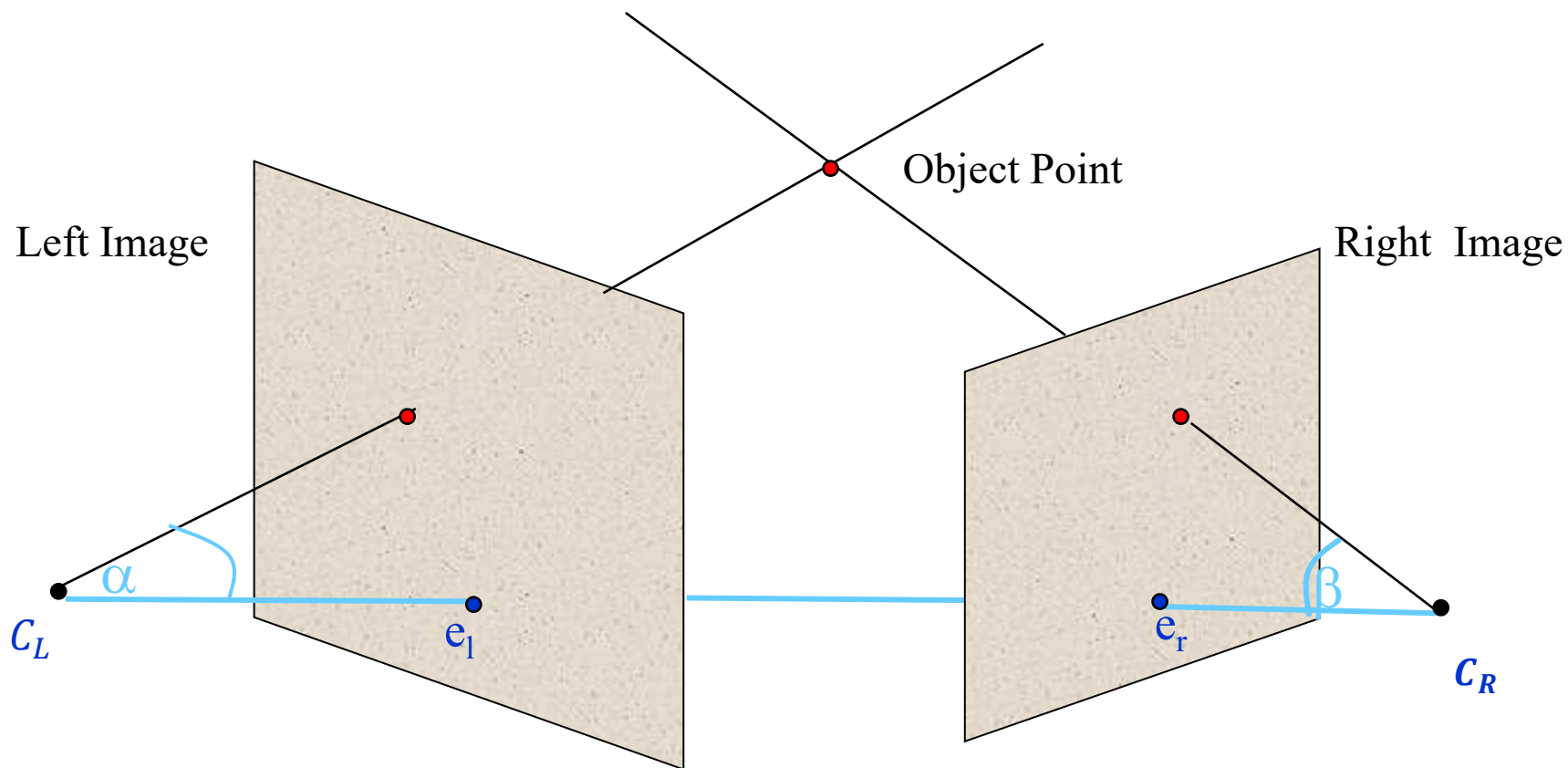


Epipolar Geometry

A pair of corresponding points must lay on corresponding epipolar lines

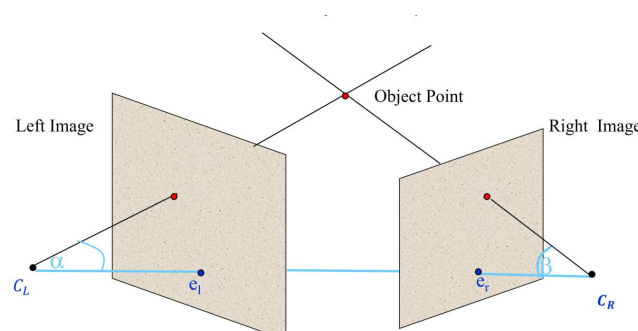


The epipole points are e_l and e_r



Epipole

- The projection of the *cop* of the other camera
- Except for the epipole, only one epipolar line goes through any image point
- All the epipolar lines go through the epipole

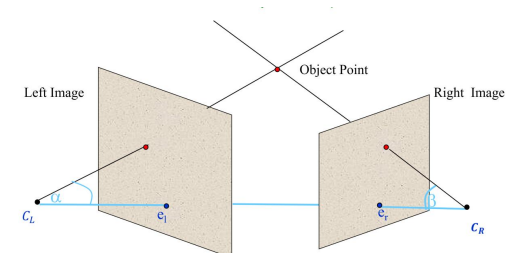


The Fundamental Matrix

That's the way to learn it

Fundamental Matrix

- F is a 3×3 matrix of rank 2
- Let p_L and p_R be corresponding image points: $\tilde{p}_R^T F \tilde{p}_L = 0$
- A point in one image defines the epipolar line in the other image:
 - $\tilde{\ell}_R = F \tilde{p}_L$ and $\tilde{\ell}_L = \tilde{p}_R^T F$
- F is based on intrinsic and extrinsic parameters

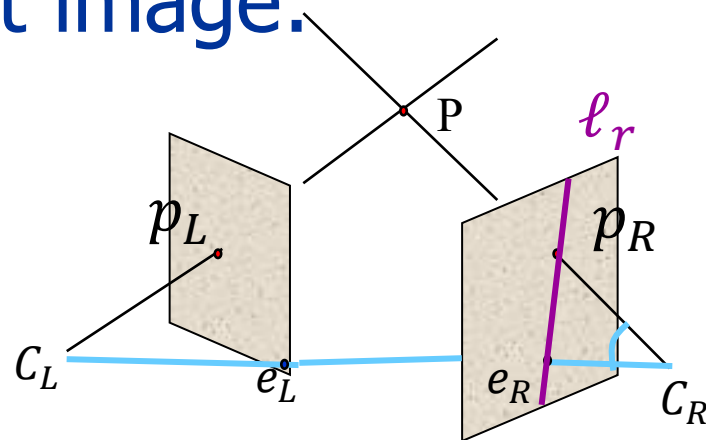


Determining the Epipolar Lines

- Depends on the intrinsic and extrinsic parameters
- Algebraic description homogenous coordinates:
 - A line is given by $\tilde{\ell}$
 - A point \tilde{p} lay on a line $\tilde{\ell}$ if $\tilde{p} \cdot \tilde{\ell} = 0$

Computing F

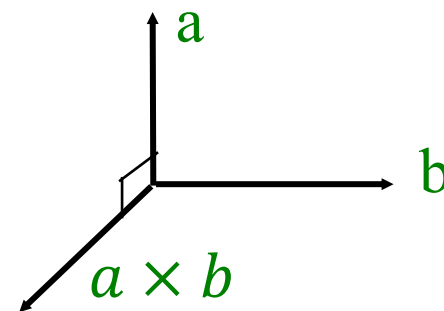
- Let M_L and M_R be the projection matrices
- A point in 3D that projects to p_L :
 - $\tilde{P} = M_L^+ \tilde{p}_l$
- Its projection to the right image:
 - $\tilde{p}_r = M_R \tilde{P}$
- A right image line:
 - $\ell_r = e_r \times (M_R M_L^+ \tilde{p}_l)$



Vector Product

- Let $a = (a_x, a_y, a_z)$ and $b = (b_x, b_y, b_z)$
- Vector (Cross product):
 - $a \times b = (a_y b_z - b_y a_z, -a_x b_z + b_x a_z, a_x b_y - b_x a_y)$

$$a \times b = \begin{pmatrix} 0 & -a_z & a_y \\ a_z & 0 & -a_x \\ -a_y & a_x & 0 \end{pmatrix} b$$



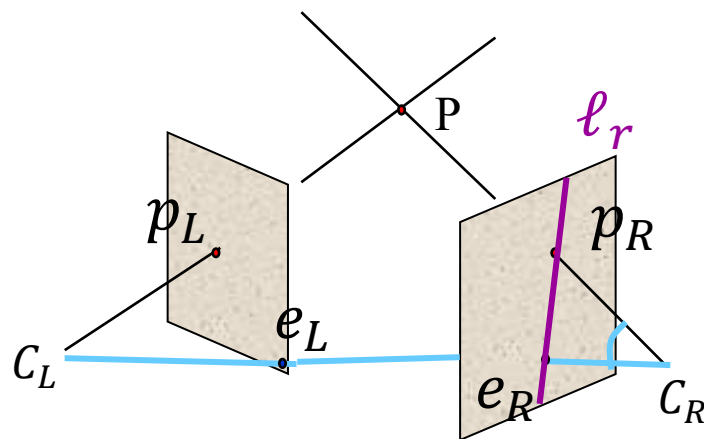
$$a \times b = [a]_{\times} b$$

$$[a]_{\times}$$

Computing F

All points, q , on the line defined by e_r and p_R satisfy:

- $\tilde{p}_R^T (e_r \times (M_R M_L^+ \tilde{p}_L)) = 0$
- $\tilde{p}_R^T ([e_r]_{\times} M_R M_L^+ \tilde{p}_L) = 0$
- $\ell_R = [e_r]_{\times} M_R M_L^+ \tilde{p}_L$
- $F = [e_r]_{\times} M_R M_L^+$



Summary Fundamental Matrix

- If p_L and p_R are corresponding image points then:
 - $\tilde{p}_R^T F \tilde{p}_L = 0$
- Epipolar lines:
 - $\tilde{\ell}_R = F \tilde{p}_L$ and $\tilde{\ell}_L = \tilde{p}_R^T F$
- The epipoles:
 - $\tilde{e}_R^T F = 0$ and $F \tilde{e}_L = 0$

A Question

- What is kF ?

Next

- More on Epipolar Geometry
- Uncalibrated pairs
- Other stereo pairs
- Special cases - Homography
 - planar surfaces, camera rotation
- More than 2 images:
 - Structure from motion