

Question 1 a)

The epipole of any epipolar line

For any point x , the epipolar line $\ell' = Fx$ contains the epipole e' .

$$\Rightarrow (e'^T) \ell' = 0$$

$$(e'^T) Fx = 0 \quad \forall x$$

$$\Rightarrow (e'^T F)x = 0 \Rightarrow e'^T F = 0$$

This shows that e' is the left null space of F .

Similarly it can be shown that e is the right null space of e .

$$\Rightarrow \boxed{Fe = 0}$$

Question 1 b) $x^T F x' = 0$

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

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

$\Rightarrow F^T$ is the fundamental matrix between I_2 and I_1 .

Reason: F contains information about relative orientation of I_2 wrt I_1 , and since orientations are relative $\rightarrow F_{12} \neq F_{21}$.

c) Question 1 c)

Fundamental matrix is used for uncalibrated camera pairs while essential matrix is used for calibrated camera pair.

$$x^T F x' = 0$$
$$x^T (K)^{-T} (R)^T S_b (R')^{-1} (K')^{-1} x' = 0$$

$\underbrace{\hspace{10em}}_F$

K, K' is the calibration matrix for Camera 1 and 2.

$$\underbrace{x^T}_{\downarrow} \underbrace{K^{-T}}_{E} \underbrace{R^T S_b R'^{-1}}_{\downarrow} \underbrace{K'^{-1} x'}_{\downarrow} = 0$$

$$\therefore E = R^T S_b R'^{-1}$$

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

Question 2 a)

(b)

Consider a world coordinate X and its image coordinate in I_1 is x and I_2 is x' .

$$x = P X \quad , \quad x' = P' X$$

Considering camera 1 as the base frame. $\Rightarrow R = I$ and $t = 0$ (by definition)

$$x = K [I | 0] \begin{bmatrix} x \\ 1 \end{bmatrix}$$

$$\Rightarrow x = K X$$

Camera 2's frame has only a rotation R from base frame

$$\therefore x' = K' [R | 0] \begin{bmatrix} x \\ 1 \end{bmatrix}$$

$$\Rightarrow x' = K' R X$$

$$x' = K' R \underbrace{K^{-1} x}_X$$

$$x' = (K' R X^{-1}) x$$

$$\text{Or } x = (K' R K^{-1})^{-1} x' = K R^{-1} K^{-1} x'$$

thus, the homography relation is given as:

$$x = [K R^{-1} K' | x'] \Rightarrow x = H^{-1} x'$$

homography matrix.

* If translation is involved, such a homography relation won't be involved.

$$x = k[I|0] \begin{bmatrix} x \\ 1 \end{bmatrix}$$

$$x = K[R] - X_0 \begin{bmatrix} x \\ 1 \end{bmatrix}$$

Translation:

$$\Rightarrow x' = K'R(\cancel{x} - X_0)$$

$$x' = K'R(K^{-1}x - X_0)$$

Clearly, this doesn't match the homography relation.

In this case, there is an ambiguous scaling factor which leads to subtraction of X_0 from K'^x and this leads to multiple possible x which relates to x' due to the unknown depth factor. Thus we get a line (epipolar) instead of a direct point x .

(Question 2 b)

There are 3 different methods to map pixels in image 1 to image 2.

- (i) Fundamental matrix
- (ii) Essential matrix
- (iii) Homography matrix

1) Fundamental matrix is used for a pair of uncalibrated cameras.

~~It has 11 parameters which describe the mapping.~~

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

fundamental matrix.

Assumption: The rotation and translation between the two viewing positions must be small.

This is because the smaller the translations and rotations are, the more number of common world points can be found.

2) Essential matrix :

We use essential matrix when we know the calibration matrix for the cameras, i.e. K and K' .

$$\boxed{x^T K^T E K' x' = 0}$$

Essential matrix.

Assumptions : (i) The cameras are calibrated and we know the calibration matrices K and K' .

(ii) The rotation angle and translation distance must be small.

3) Homography matrix :

This is used when there is only rotation involved and no translation.

$$n = Hx'$$

Assumptions : (i) Must be captured from same camera.

(ii) No translation involved.

(iii) Only rotation is involved.

(iv) Both images capture same planar surface.

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Question 2 c)

Stereo rectification has 2 homographies involved.

(i) Right camera

(ii) Left camera.

- * We use homography of left camera to make epipolar lines parallel and bring epipoles to infinity.
- * We use homography of right camera to align right camera with respect to left and then make epipolar lines // and bring epipoles to ∞ .

\Rightarrow These are called as homographies because we are dealing with images captured from same cameras and only rotations involved and no translation.

R is the rotation matrix of right camera w.r.t left camera.

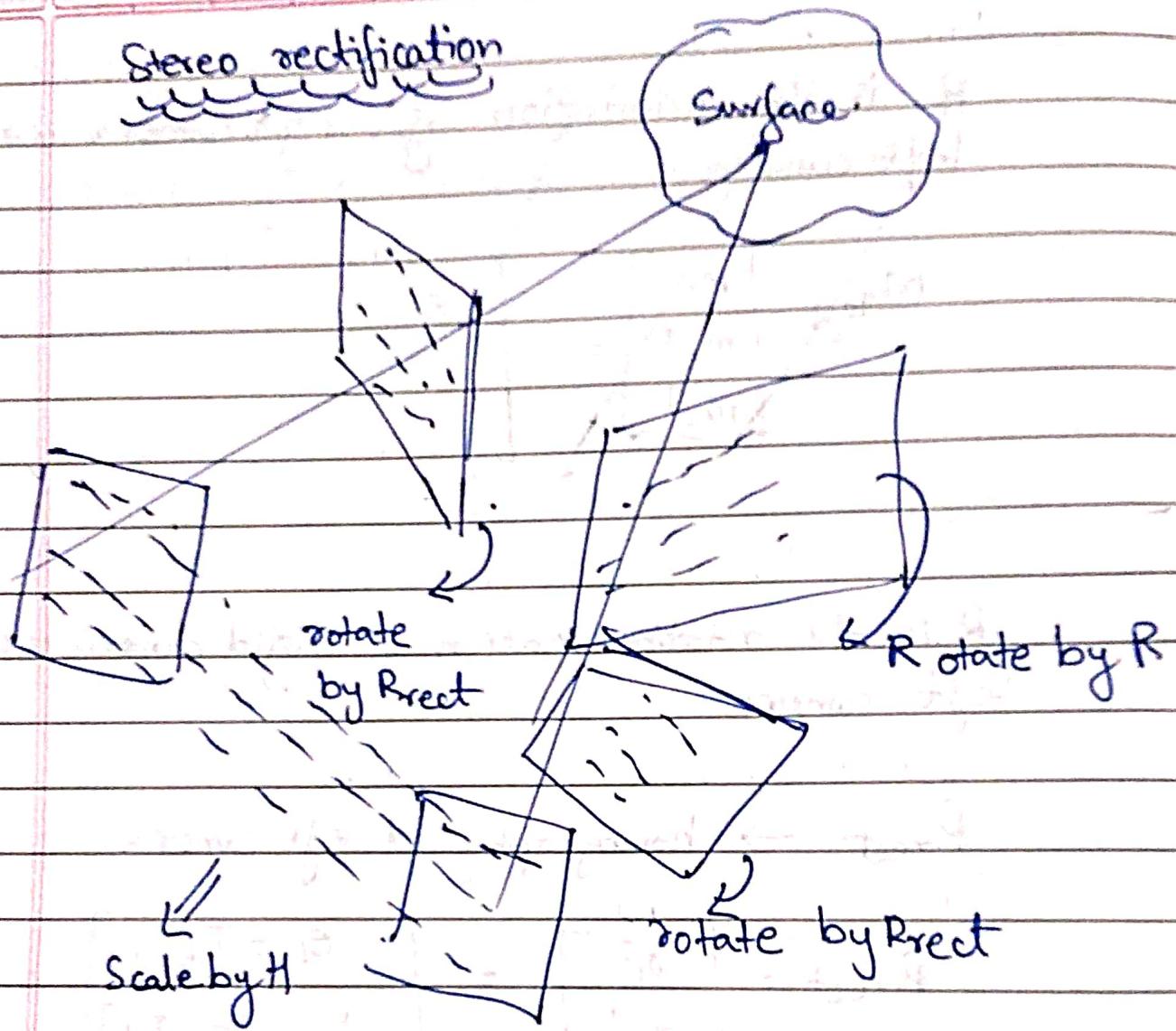
$R_{\text{rect}} \rightarrow$ homography of left camera.

$$R_{\text{rect}} = \begin{bmatrix} \gamma_1^T \\ \gamma_2^T \\ \gamma_3^T \end{bmatrix} = \begin{bmatrix} e_1 = T / \|T\| \\ 1 \\ \sqrt{T_x^2 + T_y^2} \end{bmatrix}$$

~~$\gamma_1, \gamma_2, \gamma_3$~~

homography of right camera = $R R_{\text{rect}}$

Stereo rectification



Output - Stereo rectified images

Question 3)

We use DLT for estimating the projection matrix P from known correspondences between the image coordinates and world points.

$$x = P X \quad \text{Projection matrix, } P$$

image ↗ World
coordinates. points.

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = P \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \quad \left\{ \begin{array}{l} \text{In homogeneous} \\ \text{coordinates} \end{array} \right\}$$

$$P = \begin{bmatrix} P_{11} & P_{12} & P_{13} & P_{14} \\ P_{21} & P_{22} & P_{23} & P_{24} \\ P_{31} & P_{32} & P_{33} & P_{34} \end{bmatrix}$$

$$x = \frac{P_{11}X + P_{12}Y + P_{13}Z + P_{14}}{P_{31}X + P_{32}Y + P_{33}Z + P_{34}}$$

$$y = \frac{P_{21}X + P_{22}Y + P_{23}Z + P_{24}}{P_{31}X + P_{32}Y + P_{33}Z + P_{34}}$$

$$P = K R \begin{bmatrix} I_3 & -X_0 \end{bmatrix}$$

$c, s, m,$
 α_H, γ_H
 5 parameters 3 rotation
 parameters 3 translation
 parameters

Since there are 11 unknown parameters for uncalibrated cameras, we need at least 6 points to estimate parameters.

\Rightarrow For each corresponding pair $\{x_i, \tilde{x}_i\}$ we have

$$\tilde{x}_i = P X_i \quad i = 1, 2, \dots, 6 \text{ [or more]}$$

$$\tilde{x}_i^o = \begin{bmatrix} P_{11} & P_{21} & P_{31} & P_{41} \\ P_{12} & P_{22} & P_{32} & P_{42} \\ P_{13} & P_{23} & P_{33} & P_{43} \\ P_{14} & P_{24} & P_{34} & P_{44} \end{bmatrix} \begin{matrix} \rightarrow A^T \\ \rightarrow B^T \\ \rightarrow C^T \end{matrix} X_i$$

$$A = \begin{bmatrix} P_{11} \\ P_{12} \\ P_{13} \\ P_{14} \end{bmatrix} \quad B = \begin{bmatrix} P_{21} \\ P_{22} \\ P_{23} \\ P_{24} \end{bmatrix} \quad C = \begin{bmatrix} P_{31} \\ P_{32} \\ P_{33} \\ P_{34} \end{bmatrix}$$

$$\Rightarrow x_i^o = \frac{A^T \tilde{x}_i^o}{C^T \tilde{x}_i^o} \quad y_i = \frac{B^T \tilde{x}_i^o}{C^T \tilde{x}_i^o}$$

$$x_i^o C^T \tilde{x}_i^o - A^T \tilde{x}_i^o = 0 \quad \dots \dots \text{eq 1}$$

$$y_i C^T \tilde{x}_i^o - B^T \tilde{x}_i^o = 0 \quad \dots \dots \text{eq 2}$$

We can write eq. 1, 2 as linear combination of A, B, C vectors by unrolling them into a long vector p.

$$P = \begin{bmatrix} A \\ B \\ C \end{bmatrix} = \begin{bmatrix} P_{11} \\ P_{12} \\ \vdots \\ P_{21} \\ \vdots \\ P_{31} \\ \vdots \\ P_{34} \end{bmatrix}_{12 \times 1}$$

$$\Rightarrow \text{eq } 1 = [-x_i, -y_i, -z_i, -1, 0, 0, 0, 0, x_i x_i, x_i y_i, x_i z_i, x_i] P$$

$$\text{eq } 2 = [0, 0, 0, 0, 0, -x_i, -y_i, -z_i, -1, y_i x_i, y_i y_i, y_i z_i, y_i] P$$

$$\text{Let } \text{eq } 1 = a_{x_i}^T P \text{ and } \text{eq } 2 = a_{y_i}^T P$$

We can stack all the 6 points into 12×12 matrix.

$$\begin{bmatrix} a_{x_1}^T \\ a_{y_1}^T \\ a_{x_2}^T \\ \vdots \\ a_{x_6}^T \\ a_{y_6}^T \end{bmatrix} P = 0 = M_{12 \times 12} P = 0$$

To solve a system of linear equations of form

$Ax = 0$, is equivalent to find null space of A . which can be found by solving for SVD of A and, choosing \mathbf{v} as singular vector belonging to singular value 0,

o We calculate the SVD of M and choose \mathbf{p} as the last column in V^T .

$$M_{12 \times 12} = U_{12 \times 12} S_{12 \times 12} V^T_{12 \times r_2}$$

Last column in V^T ie v_{12}

$$P = v_{12} = \begin{bmatrix} p_{11} \\ p_{12} \\ \vdots \\ p_{34} \end{bmatrix}$$

We convert \mathbf{p} to projection matrix P by reshaping it into 3×4 matrix.

$$p = \begin{bmatrix} p_{11} \\ p_{12} \\ \vdots \\ p_{34} \end{bmatrix} \quad P = \begin{bmatrix} p_{11} & p_{12} & \cdots & p_{14} \\ p_{21} & \cdots & \cdots & p_{24} \\ p_{31} & \cdots & \cdots & p_{34} \end{bmatrix}$$

FAILING CONDITION

This algorithm would fail if all points are on a plane.

Consider all points lying on some $Z = z$

$$\text{Consider } \begin{bmatrix} a^T \\ x_i^T \end{bmatrix} = \begin{bmatrix} -x_i & -y_i & -z_i \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ \vdots & & \end{bmatrix} \begin{bmatrix} -1 & 0 & 0 & 0 & 0 & x_i x_i & x_i y_i & z_i x_i \\ 0 & -x_i & -y_i & -z_i & 1 & y_i x_i & y_i y_i & z_i y_i \\ \vdots & & & & & & & \end{bmatrix}$$

These three columns can be easily converted into zero columns using some operations.

$$= \begin{bmatrix} -x_i & -y_i & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} -1 & 0 & 0 & 0 & 0 & x_i x_i & x_i y_i & 0 & x_i \\ 0 & -x_i & -y_i & 0 & 1 & y_i x_i & y_i y_i & 0 & y_i \\ \vdots & & & & & & & & \end{bmatrix}$$

M becomes a rank deficient matrix and thus no solution exists if all x_i lie on same plane.

We found that $M_p = 0$

but $M_p \neq 0$ because of errors/redundant observations.

$$\rightarrow M_p = \Delta K$$

we want p such that $\Delta K^T \Delta K$ is minimized.

$$\hat{P} = \underset{P}{\operatorname{arg\min}} \Delta K^T \Delta K$$

$$= \underset{P}{\operatorname{arg\min}} P^T M^T M P.$$

We can clearly see that this value will be minimized by a singular vector belonging to the smallest singular value.

$$\therefore P = V_{12} \quad \left\{ \text{last column of } V^T \right\}$$

corresponding to singular vector corresponding to smallest singular value?

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Question 4 a)

m observations are given to us and we know the relative ^{transformation} pose between any two pose.

Let the poses of robot be: $\{x_0, x_1, \dots, x_{m-1}\}$

We also have the control signals between observation i and j :

u_{ji} to go from i to j and

u_{ij} to go from j to i

Each of the pose is a ~~3x3~~ 3×1 vector

$$x_i = \begin{bmatrix} x_{xi} \\ x_{yi} \\ x_{zi} \end{bmatrix} \quad \{2D\text{ plane}\}$$

$$u_{ij} = [\Delta\phi_{ij} \ v_{ij}]^T$$

angle translation.

$$\hat{x}_j = f(x_i, u_{ij})$$

Our goal is to minimize the error between \hat{x}_j and x_j .

i.e. Cost function

$$\sum_{i=0}^{m-1} \sum_{\substack{j=0 \\ j \neq i}}^{m-1} \|f(\hat{x}_i, v_{ij}) - x_j\|^2$$

$f(\hat{x}_i, v_{ij})$ is a highly non linear function

we want to linearize it.

$$f(\hat{x}_i, v_{ij}) \approx f(x_i, v_{ij}) + F_{ij} \delta x_i$$

$\underbrace{\quad}_{\text{initial estimate}}$

$\overset{T}{\mathbf{J}}$ Jacobian.

Cost function =

$$\sum_{i=0}^{m-1} \sum_{\substack{j=0 \\ j \neq i}}^{m-1} \|f(x_i, v_{ij}) + F_{ij} \delta \hat{x}_i - x_j\|^2$$

$f(x_i, v_{ij}) - x_j$ is the odometry error
 a_{ij}

$$\therefore \text{Cost function} = \sum_{i=0}^{m-1} \sum_{\substack{j=0 \\ j \neq i}}^{m-1} \|F_{ij} \delta \hat{x}_i - a_{ij}\|^2$$

Minimizing this would give us optimized trajectory.

Question 4 b)

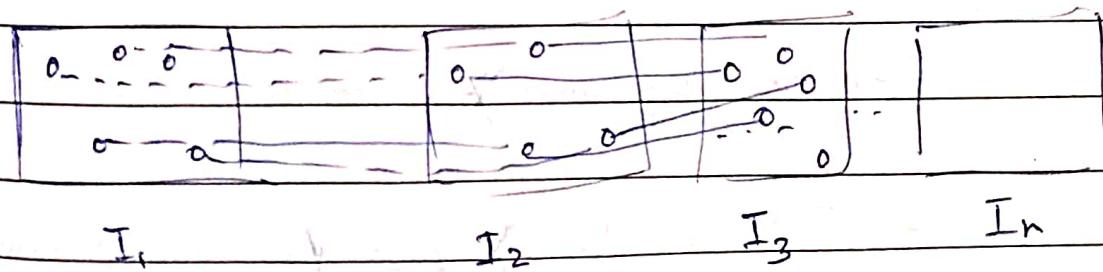
Given a set of images $\{I_1, I_2, \dots, I_n\}$, if we want to obtain the pose estimates and optimise the trajectory, this is structure from motion problem.

- 1) the first step would be to compute features in the images which are likely to be invariant across all images (or at least most of them)

This can be done using feature detectors such as

- SIFT
- ORB
- Hessian-Laplacian

Now we have features in the images.



- 2) Next step is to find the correspondances between these features as shown in the illustration above.

This can be done by matching the feature descriptors via approximate nearest neighbour.

3) Now that we have correspondance we need to compute the 3D world points and the projection matrix.

Initial guess :

No additional information given so can't use triangulation
as it need projection matrix.

Consider m scenes. and n points in total (features we have extracted)

P_j is projection matrix of j^{th} scene.

\mathbf{x}_{ij}^* is 3D word point of i^{th} feature in image in j^{th} view.
 \mathbf{x}_{ij} is the i^{th} feature in j^{th} view.

$$\mathbf{x}_{ij} = P_j \cdot \mathbf{x}_{ij}^*$$

Estimate P_j and \mathbf{x}_{ij}^* to minimize:

$f(P_1, P_2, \dots, P_m, \mathbf{x}_1^*, \mathbf{x}_2^*, \dots, \mathbf{x}_n^*) \Leftarrow \text{loss function.}$

This will be minimized for:

$$\underset{P_i, \mathbf{x}_i^*}{\operatorname{argmin}} \sum_{i=1}^m \sum_{j=1}^N (w_{ij} \mathbf{x}_{ij}) \| P_i \mathbf{x}_i^* - \mathbf{x}_{ij} \|_2^2$$

$$w_{ij} = \begin{cases} 1 & \text{if } j^{\text{th}} \text{ point seen in } i^{\text{th}} \text{ frame.} \\ 0 & \text{if not seen.} \end{cases}$$

$$\Rightarrow \underset{P_i, X_j}{\operatorname{argmin}} \sum_{i=1}^M \sum_{j=1}^N \left(\begin{bmatrix} P_{i1}X_j + P_{i2}Y_j + P_{i3}Z_j + P_{i4} \\ P_{i5}X_j + P_{i6}Y_j + P_{i7}Z_j + P_{i8} \end{bmatrix} - \begin{bmatrix} x_{ij} \\ y_{ij} \end{bmatrix} \right)^2 + \left(\begin{bmatrix} P_{i21}X_j + P_{i22}Y_j + P_{i23}Z_j + P_{i24} \\ P_{i31}X_j + P_{i32}Y_j + P_{i33}Z_j + P_{i34} \end{bmatrix} - \begin{bmatrix} x_{ij} \\ y_{ij} \end{bmatrix} \right)^2$$

This is reprojection error.

$$P_i X_j = \hat{x}_{ij} \quad \left\{ \begin{array}{l} \text{estimate of } j^{\text{th}} \text{ point in } i^{\text{th}} \\ \text{scene} \end{array} \right\}$$

$$\underset{X_j, P_i}{\operatorname{argmin}} \sum_{i=1}^M \sum_{j=1}^N \left\| \hat{x}_{ij} - x_{ij} \right\|^2$$

~~Using this, we will have an initial estimate for 3D world points for Bundle adjustment.~~

~~We didn't have projection matrix so we can't use triangulation for initial estimates.~~

4) This can be solved using LM-algorithm.

$$(J^T J + \lambda I) \Delta K = -J^T r(K)$$