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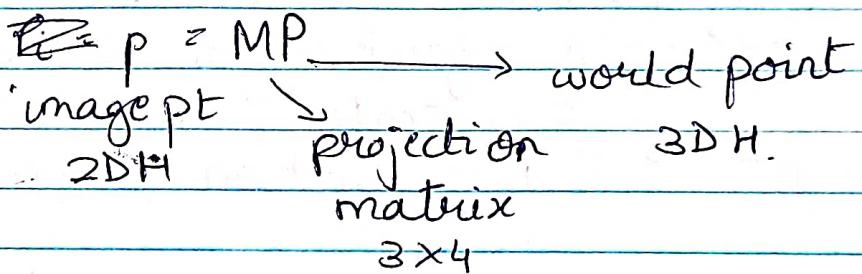
## Assignment 5

### Q1. Camera Calibration

Q1a)

$$p = MP$$

In forward projection, we have a given world point  $P$  and translation matrix  $M$  and we have to convert it to image point  $p$  ( ~~$3D H - 2D H$~~ ).



In camera calibration, we estimate the intrinsic and extrinsic parameters of the camera.

Given corresponding points in real world and image, find parameters  
 $\{P_i\}_{i=1}^m \leftrightarrow \{p_i\}_{i=1}^m$

In reconstruction, we are given an image point & we have to compute the world point from it. While doing 3DH to 2DH, we lose a lot of info so reconstructing 3DH from 2DH is difficult.

Therefore, according to the above definitions, forward projection is the easiest whereas reconstruction is the most difficult.

Q2

Q4.b)

Ans. For camera calibration, we need ~~two~~ corresponding world & image points which will be our input for the calibration algorithm.

Given corresponding  $\{P_i\}_{i=1}^m$  pixels  $\longleftrightarrow \{P_i\}_{i=1}^m$  m meters  
points.

Then we need to find the intrinsic and extrinsic parameters like

$K^*$ :  $f \rightarrow$  focal length,  $u_0, v_0 \rightarrow$  optical center  
 $s \rightarrow$  scale,  $k_u, k_v \rightarrow$  skew

$R^*, T^*$ :  $R \rightarrow$  rotation,  $T \rightarrow$  translation

(Q1c)

Ans. 1) First step in non-coplanar calibration is estimating the projection matrix

$$P_i' = MP_i$$

$$\begin{bmatrix} x_i' \\ y_i' \\ w_i' \end{bmatrix} = \begin{bmatrix} -m_1^T \\ -m_2^T \\ -m_3^T \end{bmatrix} P_i$$

For m points

$$\begin{bmatrix} P_1^T & 0 & -x_1 P_1^T \\ 0 & P_1^T & -y_1 P_1^T \\ \vdots & & \\ P_m^T & 0 & -x_m P_m^T \\ 0 & P_m^T & -y_m P_m^T \end{bmatrix} \begin{bmatrix} m_1 \\ m_2 \\ m_3 \\ \vdots \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{bmatrix}$$

$$Ax = 0$$

To solve  $Ax = 0$ , use SVD

2) Next find parameters from M

$$M \in K^* [R^* | T^*] = \hat{SM}$$

$$[K^* R^* | K^* T^*] = \hat{SM}$$

$$\begin{bmatrix} \alpha_u R_1^T + \alpha_v R_2^T + \alpha_w R_3^T \\ \alpha_v R_2^T + \alpha_u R_3^T \\ \alpha_w R_3^T \end{bmatrix} [K^* T^*] = \hat{SM} = \begin{bmatrix} -a_1^T \\ -a_2^T \\ -a_3^T \end{bmatrix} [b]$$

$$\alpha = \frac{1}{|a_3|} \quad u_0 = \frac{|S|^2}{|S|^2 a_1 a_3}$$

$$v_0 = |S|^2 a_2 a_3 \quad \alpha_v = \sqrt{|S|^2 a_2 a_3 - v_0^2}$$

$$s = \frac{1}{\alpha_v} |s|^4 (a_1 \times a_3) \cdot (a_2 \times a_3)$$

$$\alpha_u = \sqrt{|s|^2 a_1 \cdot a_1 - s^2 - u_0^2}$$

$$K^* = \begin{bmatrix} \alpha_u & s & u_0 \\ 0 & \alpha_v & v_0 \\ 0 & 0 & 1 \end{bmatrix}.$$

$$e = \text{sgn}(b_3)$$

$$T^* = e |s| (K^*)^{-1} b$$

$$u_3 = \cancel{s} |s| a_3$$

$$u_1 = \frac{1}{\alpha_v} |s|^2 a_2 \times a_3$$

$$u_2 = u_3 \times u_1$$

$$R^* = [r_1^T \ u_2^T \ u_3^T]^T$$

Q 8.C)

Ans. Steps in non - coplanar calibration algorithm :-

- ① Find projection matrix M
- ② Find parameters (extrinsic, intrinsic) from M.

Q1. d)

Ans.

$$P \cancel{DF} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 0 & 3 & 4 \\ 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 18 \\ 14 \\ 9 \end{bmatrix}$$

$$\therefore x = \frac{18}{7} \quad y = \frac{14}{7}$$

$\therefore$  2D image co-ordinates are

$$\begin{bmatrix} 18 & 7 \\ 2 & \end{bmatrix}$$

(3)

Q.1.e)

Ans.

$$(1, 2, 3) \leftrightarrow (100, 200)$$

Projection matrix  $M = ?$   
 ~~$A = ?$~~

$$\begin{bmatrix} P_1^T & 0 & -x_1 P_1^T \\ 0 & P_1^T & -y_1 P_1^T \\ \vdots & & \\ P_m^T & 0 & -x_m P_m^T \\ 0 & P_m^T & -y_m P_m^T \end{bmatrix} \begin{bmatrix} m_1 \\ m_2 \\ m_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad Ax = 0$$

substituting the given points in above matrix  $A$

$$\begin{bmatrix} 1 & 2 & 3 & 1 & 0 & 0 & 0 & 0 & -100 & -200 & -300 & -100 \\ 0 & 0 & 0 & 0 & 1 & 2 & 3 & 1 & -200 & -400 & -600 & -200 \end{bmatrix}$$

Q.1.f)

Ans. To find a unique solution for  $M$ , we need atleast 6 points. There are 11 unknown parameters which are  $\alpha_1, \alpha_2, \alpha_3, b, g_1, g_2, g_3, T^*, k^*, s, \alpha_u, \alpha_v, m_0, v_0, s$ . We need atleast 12 equations for this so we need 6 points to get 12 equations.

We estimate the projection matrix as mentioned below:-

$$\begin{bmatrix} P_1^T & 0 & -x_1 P_1^T \\ 0 & P_1^T & -y_1 P_1^T \\ \vdots & & \\ P_m^T & 0 & -x_m P_m^T \\ 0 & P_m^T & -y_m P_m^T \end{bmatrix} \begin{bmatrix} m_1 \\ m_2 \\ \vdots \\ m_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad Ax = 0$$

(4)

Then, to solve this, we use SVD

$$A = UDV^T$$

Solution = column of  $V$  belonging to zero singular value

$$\hat{x} = \begin{bmatrix} \hat{m}_1 \\ \hat{m}_2 \\ \hat{m}_3 \end{bmatrix} \Rightarrow \hat{M} = \begin{bmatrix} \hat{m}_1^T \\ \hat{m}_2^T \\ \hat{m}_3^T \end{bmatrix}$$

The solution is not unique if  $A\hat{x} = 0$

We also have  $A(S\hat{x}) = 0 \Rightarrow S\hat{x}$  is a solution.

Q 1.9)

"Ans."  $\hat{M} = \begin{bmatrix} \hat{m}_1^T \\ \hat{m}_2^T \\ \hat{m}_3^T \end{bmatrix}$  3x3 matrix

The solution for  $\hat{M}$  is not unique  
So, we find  $S$  so that

$$M = K^* \begin{bmatrix} R^* | T^* \end{bmatrix} = S\hat{M} \quad S = ?$$

We find parameters &  $S$  from  $M$

$$\begin{bmatrix} K^* \begin{bmatrix} R^* | T^* \end{bmatrix} \\ K^* \begin{bmatrix} R^* | K^* T^* \end{bmatrix} \end{bmatrix} = S\hat{M}$$

$$\left[ \begin{array}{c|c} \alpha_1 \mathbf{r}_1^T + \beta_1 \mathbf{r}_2^T + \gamma_1 \mathbf{r}_3^T & K^* T^* \\ \alpha_2 \mathbf{r}_1^T + \beta_2 \mathbf{r}_2^T + \gamma_2 \mathbf{r}_3^T & \\ \hline & M_3^T \end{array} \right] = S\hat{M} = S \left[ \begin{array}{c|c} -a_1^T & \\ -a_2^T & \\ -a_3^T & \\ \hline b & \end{array} \right]$$

Unknowns      Unknowns

(4a)

Ans<sup>g</sup>)  
contd.

To extract parameters using orthogonality of  
 $\mathbf{g}_1, \mathbf{g}_2, \mathbf{g}_3$

$$\mathbf{g}_1 \cdot \mathbf{g}_2 = 0 \quad \mathbf{g}_2 \cdot \mathbf{g}_3 = 0 \quad \mathbf{g}_1 \cdot \mathbf{g}_3 = 0$$

$$\mathbf{g}_1 \times \mathbf{g}_2 = \mathbf{g}_3 \quad \mathbf{g}_2 \times \mathbf{g}_3 = \mathbf{g}_1 \quad \mathbf{g}_3 \times \mathbf{g}_1 = \mathbf{g}_2$$

(5)

Q1. h)  
Ans.Given  $\{P_i\}_{i=1}^m \leftrightarrow \{\tilde{P}_i\}_{i=1}^m$  and estimated

$$M = \begin{bmatrix} -m_1^T & - \\ m_2^T & - \\ -m_3^T & - \end{bmatrix}$$

The quality of fit is assessed using :-

$$E = \frac{1}{m} \sum_i \left( \left\| x_i - \frac{m_1^T \tilde{P}_i}{m_3^T \tilde{P}_i} \right\|^2 + \left\| y_i - \frac{m_2^T \tilde{P}_i}{m_3^T \tilde{P}_i} \right\|^2 \right)$$

Distance between known &amp; predicted positions.

Q1.i)

Ans. ① For planar camera calibration, we first need to estimate 2D homography (projective map) between calibration plane and image (for several images)

② Estimate intrinsic parameters

③ Compute extrinsic parameters for view of interest.

Planar camera calibration solves 2D H points whereas non-planar solves 3D H points. For planar, points should be on the same plane whereas for non-planar, points will be on different planes. For planar, single image with different views is used for camera calibration whereas 3D image is used for calibration in non-planar.

(6)

Q1. j)

Ans.

Homography (H) 2D projective map is used for transformation from one plane to other and it is  $3 \times 3$ . Projection matrix is used to project 3D H points to get 2D H image points and it is  $3 \times 4$ .

For ~~per~~

$$\underline{P}_i^* = K^* [R^* | T^*] \underline{P}_i$$

2DH                     $3 \times 4$                     3DH

$$\begin{bmatrix} u_i \\ v_i \\ w_i \end{bmatrix} = K^* \begin{bmatrix} 1 & 1 & 1 \\ a_1 & a_2 & a_3 & T^* \end{bmatrix} \begin{bmatrix} x_i \\ y_i \\ 0 \\ 1 \end{bmatrix}$$

↳ To get a 2D projective map we assume  $\underline{P}_i^*_{\underline{z}} = 0 \Leftrightarrow \underline{P}_i^* = (x_i, y_i, 0)$   
 Z coordinate is  $> 0$

$$\therefore P_i^* = K^*(x_i a_1 + y_i a_2 + T^*)$$

$$\begin{bmatrix} u_i \\ v_i \\ w_i \end{bmatrix} = K^* \begin{bmatrix} a_1 & a_2 & T^* \end{bmatrix} \begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix}$$

$3 \times 3$

2D projection                    2DH

map. Homography

$$\begin{bmatrix} u_i \\ v_i \\ w_i \end{bmatrix} = H \underline{P}_i^* = \begin{bmatrix} h_1^T \\ h_2^T \\ h_3^T \end{bmatrix} \underline{P}_i^*$$

$$x_i = \frac{u_i}{w_i} = \frac{h_1^T \underline{P}_i^*}{h_3^T \underline{P}_i^*} \Rightarrow h_1^T \underline{P}_i^* - x_i h_3^T \underline{P}_i^* = 0$$

$$y_i = \frac{v_i}{w_i} = \frac{h_2^T \underline{P}_i^*}{h_3^T \underline{P}_i^*} \Rightarrow h_2^T \underline{P}_i^* - y_i h_3^T \underline{P}_i^* = 0.$$

Q2. Camera calibration 2.

Ans a)

$$(1, 2) \leftrightarrow (3, 4, 5)$$

Projection matrix  $\mathbf{M} = ?$

$$\begin{bmatrix} P_1^T & 0 & -x_1 P_1^T \\ 0 & P_1^T & -y_1 P_1^T \\ \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} m_1 \\ m_2 \\ m_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 3 & 4 & 5 & 1 & 0 & 0 & 0 & 0 & -3 & -4 & -5 & -1 \\ 0 & 0 & 0 & 0 & 3 & 4 & 5 & 1 & -6 & -8 & -10 & -2 \end{bmatrix} \begin{bmatrix} m_1 \\ m_2 \\ \vdots \\ m_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Q2.b)

Ans.

$$\mathbf{M} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \end{bmatrix}$$

$$S \hat{\mathbf{M}} = K^* [R^* | T^*]$$

$$S \hat{\mathbf{M}} = [K^* R^* | K^* T^*]$$

$$\left[ \begin{array}{c|c} \alpha_u \mathbf{r}_1^T + \alpha_v \mathbf{r}_2^T + \alpha_w \mathbf{r}_3^T & K^* T^* \\ \alpha_v \mathbf{r}_1^T + \alpha_u \mathbf{r}_2^T + \alpha_w \mathbf{r}_3^T & \\ \mathbf{r}_3^T & \end{array} \right] \equiv S \left[ \begin{array}{c|c} -\mathbf{a}_1^T & \\ -\mathbf{a}_2^T & \\ -\mathbf{a}_3^T & \\ \hline & b \end{array} \right]$$

$$\alpha_w = |S|^2 \mathbf{a}_1 \cdot \mathbf{a}_3$$

$$\alpha_v = |S|^2 \mathbf{a}_2 \cdot \mathbf{a}_3$$

$$\mathbf{M} = \left[ \begin{array}{c|c} -\mathbf{a}_1^T & \\ -\mathbf{a}_2^T & \\ -\mathbf{a}_3^T & \\ \hline & b \end{array} \right]$$

$$\therefore \begin{aligned} \alpha_1^T &= \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \\ \alpha_2^T &= \begin{bmatrix} 2 & 3 & 4 \end{bmatrix} \\ \alpha_3^T &= \begin{bmatrix} 3 & 4 & 5 \end{bmatrix} \end{aligned}$$

$$|S|^2 = \frac{1}{|\alpha_3|^2} \quad |\alpha_3|^2 = (\sqrt{3^2 + 4^2 + 5^2})^2$$

$$|\alpha_3|^2 = (\sqrt{9 + 16 + 25})^2$$

$$|\alpha_3|^2 = (\sqrt{50})^2 = 50$$

$$S^2 = \frac{1}{50}$$

$$\begin{aligned} \therefore U_0 &= S^2 \alpha_1 \cdot \alpha_3 \\ &= \frac{1}{50} (1 \ 2 \ 3) \cdot (3 \ 4 \ 5) \\ &= \frac{1}{50} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 3 \\ 4 \\ 5 \end{bmatrix} \\ &= \frac{1}{50} (3 + 8 + 15) \end{aligned}$$

$$U_0 = \frac{26}{50}$$

$$\begin{aligned} V_0 &= S^2 \alpha_2 \cdot \alpha_3 \\ &= \frac{1}{50} \begin{bmatrix} 2 \\ 3 \\ 4 \end{bmatrix} \cdot \begin{bmatrix} 3 \\ 4 \\ 5 \end{bmatrix} \\ &= \frac{1}{50} (6 + 12 + 20) \end{aligned}$$

$$V_0 = \frac{38}{50}$$

$$\therefore (U_0, V_0) = \left( \left( \frac{26}{50} \right), \left( \frac{38}{50} \right) \right)$$

(Q2.c)

Ans.

$(1, 2) \leftrightarrow (3, 4, 5)$   
image      P world.

$$M = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \end{bmatrix} \rightarrow m_1^T$$
$$\rightarrow m_2^T$$
$$\rightarrow m_3^T$$

$$E = \frac{1}{m} \sum_i \left( \left\| x_i - \frac{m_1^T p_i}{m_3^T p_i} \right\|^2 + \left\| y_i - \frac{m_2^T p_i}{m_3^T p_i} \right\|^2 \right)$$

$m = 1$  as number of point is 1.

$$m_1^T p = \begin{bmatrix} 1 & 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \\ 5 \\ 1 \end{bmatrix}$$
$$= 3 + 8 + 15 + 4 = 30$$

$$m_2^T p = \begin{bmatrix} 2 & 3 & 4 & 5 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \\ 5 \\ 1 \end{bmatrix}$$
$$= 6 + 12 + 20 + 5 = 43$$

$$m_3^T p = \begin{bmatrix} 3 & 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \\ 5 \\ 1 \end{bmatrix}$$
$$= 9 + 16 + 25 + 6 = 56.$$

$$\begin{aligned}
 E &= \left\| 1 - \frac{30}{56} \right\|^2 + \left\| 2 - \frac{43}{56} \right\|^2 \\
 &= \left\| \frac{56 - 30}{56} \right\|^2 + \left\| \frac{112 - 43}{56} \right\|^2 \\
 &= \left\| \frac{26}{56} \right\|^2 + \left\| \frac{69}{56} \right\|^2 \\
 &= (0.4642)^2 + (1.2321)^2 \\
 &\approx 0.2155 + 1.5181 \\
 E &\approx 1.7336
 \end{aligned}$$

(Q2, d)  $I + Q$ ,

$$I = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$Q = \begin{bmatrix} 5 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$I + Q = \begin{bmatrix} 6 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

i.e.,  $R^*$

$$T^* = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 1 \end{bmatrix}$$

$$\begin{aligned} T^* &= -RT^R \\ \therefore T &= -(R^*)^T T^* \end{aligned}$$

$\therefore$  Rotation &  
Translation  $\equiv$

$$\text{w.r.t World} \quad - \begin{bmatrix} 6 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 1 \end{bmatrix} \quad 4 \times 4 \quad 4 \times 1$$

$$= \begin{bmatrix} 6 \\ 2 \\ 3 \\ 1 \end{bmatrix} = \begin{bmatrix} -6 \\ -2 \\ -3 \\ -1 \end{bmatrix}$$

Q2.c) ~~is~~  $(1, 2) \leftrightarrow (3, 4, 0)$ .  
 Image world.

world point =  $\begin{bmatrix} 3 \\ 4 \\ 1 \end{bmatrix}$

$$\begin{bmatrix} P_1^{*T} & 0^T - x_1 P_1^{*T} \\ 0 & P_1^{*T} - y_1 P_1^{*T} \\ \vdots & \\ P_m^{*T} & 0^T - x_m P_m^{*T} \\ 0 & P_m^{*T} - y_m P_m^{*T} \end{bmatrix} \begin{bmatrix} h_1 \\ h_2 \\ h_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

2D H.

$$\begin{bmatrix} 3 & 4 & 0 & 0 & 0 & 0 & -3 & -4 & 0 \\ 0 & 0 & 0 & 3 & 4 & 0 & -6 & -8 & 0 \end{bmatrix} \begin{bmatrix} h_1 \\ h_2 \\ h_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Q. 3.

### Multiple view geometry

a)

Ans a)

Dense stereo matching is finding corresponding points for each pixel from two or more images at slightly different viewpoints. This produces more points and is correlation, SSD based. Sparse stereo matching is feature based and can handle large disparities. So, advantage of sparse stereo is that it can be used in frames with very high disparity between each frame. Advantage of dense approach is that it would work for frames with very less disparities, but would fail for frames with high disparity. The disadvantage of sparse stereo approach is that it would not work on frames with little or no difference between them. The disadvantage of dense approach is that it would fail for frames with high disparities.

Ans. b)

Correlation can be calculated as

$$\Phi(w_1, w_2) = \sum_i w_1(x_i, y_i) \cdot w_2(x_i, y_i)$$

SSD can be calculated as

$$\Phi(w_1, w_2) = - \sum_i (w_1(x_i, y_i) - w_2(x_i, y_i))^2$$

where high value means good correspondence. To remove dependency on absolute value, we normalize window values (subtract mean and divide by standard deviation).

NCC computes normalized cross correlation of two windows whereas SSD computes the

sum of square distance between the windows.

If the search space is allowed to be the entire image, there might be mismatching points since the chances of error are more.

The risk is that the chances of error making increases as there will be a lot of options corresponding to ~~a~~ point in an image.

The search space is hence reduced by only considering epipolar line instead of whole image.

(Q3.4)

$$\text{Ans: } z = f \frac{I}{d}$$

$$x_1 = (100, 200)$$

$$x_2 = (103, 200)$$

$$f = 10$$

$$T = 100$$

$$\text{depth } z = \frac{f T}{d}$$

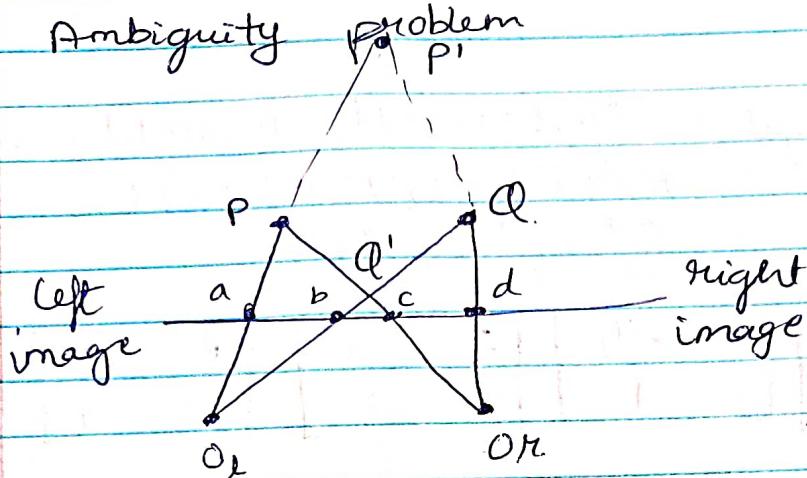
$$d = 103 - 100 = 3$$

$$\therefore \text{depth } z = \frac{10 \times 100}{3}$$

$$\text{depth } z = \frac{1000}{3} = 333.33$$

Q3.d

Ans(d) Ambiguity problem



A small change / error in correspondence can make large errors in reconstruction so, even if we make a mistake with two points, the incorrect points are very far from the original points.

Ideally

Ideally, if correspondence is done correctly then points (a) & (c) are matched & resulting P point is obtained. same is done, when (b) & (d) are matched and we get Q. But if we make a mistake & match (a) with (d) & (b) with (c), we get wrong points P' & Q' during reconstruction.

Correct points : P, Q

Incorrect : P', Q'

(Q 3.c) Rotation Translation

Ans e)  $R_L, T_L$

$R_R, T_R$

$$M_{\text{left} \leftarrow \text{right}} = M_{\text{left} \leftarrow \text{world}} M_{\text{world} \leftarrow \text{right}}$$

$$M_{L \leftarrow R} = R_L^T T(-T_L) T(T_R) R_R$$

$$= R_L^T T_L^{-1} T_R R_R$$

$$= \begin{bmatrix} R_L^T & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} I & -T_L \\ 0 & 1 \end{bmatrix} \begin{bmatrix} I & T_R \\ 0 & 1 \end{bmatrix} \begin{bmatrix} R_R & 0 \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} R_L^T & -R_L^T T_L \\ 0 & 1 \end{bmatrix} \begin{bmatrix} R_R & T_R \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} R_L^T R_R & R_L^T (T_R - T_L) \\ 0 & 1 \end{bmatrix}$$

$$R = R_L^T R_R$$

$$T = R_L^T (T_R - T_L)$$

## Multiple view geometry 2

(Q 4.a)

Ans.

$$f = 10 \text{ mm}$$

$$T = 20$$

$$\text{disparity} = 30 \text{ mm}$$

$$z = f \frac{T}{d}$$

$$\therefore z = 10 \times \frac{20}{30}$$

$$z = 6.66 \text{ mm}$$

(Q 4.b)

Ans.

$$A = (1, 2, 3)$$

$$B = (2, 3, 4)$$

$$A \times B = [A]_x B \equiv \begin{bmatrix} 0 & -a_z & a_y \\ a_z & 0 & -a_x \\ -a_y & a_x & 0 \end{bmatrix}$$

Rank 2 skew symmetric matrix

$$\text{Matrix} = \begin{bmatrix} 0 & -3 & 2 \\ 3 & 0 & -1 \\ -2 & 1 & 0 \end{bmatrix}$$

Above matrix if multiplied by B will result in the cross product  $A \times B$ .

Q 4.C)  
Ans.

$$F = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 3 & 4 & 5 \end{bmatrix}$$

$$P_L = (1, 2)$$

$$P_R = (2, 3)$$

$$P_L^T F P_L = [2 \ 3 \ 1] \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 3 & 4 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

$$= [11 \ 17 \ 23] \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

$$= 11 + 34 + 23$$

$$P_R^T F P_L = 68$$

Q4. a)

$$P_L = (1, 2)$$

$$P_R = (2, 3)$$

∴ Fundamental matrix

$$= \bar{P}_R^T (K_a^{*\top} + K_e^{*\top -1}) \bar{P}_L$$

$$P_R^T F P_L = 0.$$

↓ fundamental matrix

$$[x_i, y_i, 1] \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix} \begin{bmatrix} x_i' \\ y_i' \\ 1 \end{bmatrix} = 0.$$

$$[2, 3, 1] \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = 0.$$