

ASSIGNMENT - 4

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Review Questions:

1. Camera Calibration:

a. Given the equation $p = MP$

Forward Projection: Given a world point P project it onto an image using a projection matrix M .

Calibration: Given image points $\{p_i\}_{i=1}^m$ and world points $\{P_i\}_{i=1}^m$, find the camera intrinsic and extrinsic parameters $(\alpha_u, \alpha_v, s, u_0, v_0, R^*, T^*, K)$.

Reconstruction: Given several images with $N \times \{p_i\}_{i=1}^m$ and the projection matrix M reconstruct the 3D object coordinates.

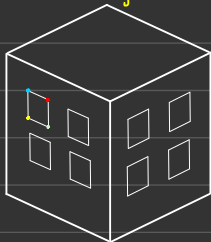
Reconstruction is the most difficult out of the three discussed above. Forward projection is the simplest.

b. The necessary inputs for camera calibration are, the points $\{p_i\}_{i=1}^m$ which are the image coordinates and points $\{P_i\}_{i=1}^m$ which are the corresponding world coordinates.

We need atleast 6 point pairs in order to do a 3D $\{p_i\}_{i=1}^m$ camera calibration or at least 4 points to perform planar camera calibration.

C. Given $\{p_i\}_{i=1}^m \longleftrightarrow \{P_i\}_{i=1}^m$ find camera parameters:

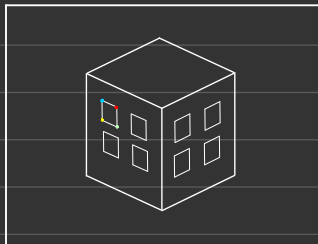
3D- object



$\{P_i\}$

3DH

2D- Image.



$\{P_i\}$

2DH

Steps:

- Find the projection matrix M , (estimate).
- Find parameters from M .

To estimate the projection matrix:

- Find the point pair matrix in the form of

$$\begin{array}{c}
 \text{A} \qquad \qquad \qquad \text{x} \qquad = \qquad \text{0} \\
 \left[\begin{array}{ccc}
 P_1^T & 0 & -x_1 P_1^T \\
 0 & P_1^T & -y_1 P_1^T \\
 \vdots & \vdots & \vdots \\
 P_m^T & 0 & -x_m P_m^T \\
 0 & P_m^T & -y_m P_m^T
 \end{array} \right] \begin{bmatrix} m_1 \\ m_2 \\ m_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{bmatrix} \\
 \begin{array}{ccc}
 2m \times 12 & 12 \times 1 & 2m \times 1
 \end{array}
 \end{array}$$

Number of corresponding points.

- Then solve $Ax = 0$ using SVD to obtain \hat{m}

- Find β so that

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

- Find the solution for the below equation

$$\underbrace{\begin{bmatrix} \alpha_u x_1^T + s x_2^T + u_0 x_3^T \\ \alpha_v x_1^T + v_0 x_3^T \\ x_3^T \end{bmatrix}}_{K^* R^*} \underbrace{K^* T^*}_{3 \times 4} = \beta \underbrace{\begin{bmatrix} -a_1^T & | & b \\ -a_2^T & | & b \\ -a_3^T & | & b \end{bmatrix}}_{\text{KNOWN}} = \beta \hat{M}$$

UNKNOWNS

d. Given

$$M = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 0 & 3 & 4 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

$$P_i = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 1 \end{bmatrix}$$

$$\beta_i = M P_i$$

$$\beta_i = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 0 & 3 & 4 \\ 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 1 \end{bmatrix} = \begin{bmatrix} 1+4+9+4 \\ 1+0+9+4 \\ 1+2+3+1 \end{bmatrix} = \begin{bmatrix} 18 \\ 14 \\ 7 \end{bmatrix}$$

The 2D point is $(\frac{18}{7}, 2)$

e. given $\{p_i\} \leftrightarrow \{P_i\}$
 $(100, 200) \leftrightarrow (1, 2, 3)$

The first two lines must be of the form:

$$= \begin{bmatrix} P_i^T & 0 & -x_i P_i^T \\ 0 & P_i^T & -y_i P_i^T \end{bmatrix} \quad P_i^T = [1, 2, 3, 1]$$

2×12

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

f. The minimum number of points needed to be able to find a unique solution for M is 6 because we have 11 unknowns. Hence we need 12 equations \Rightarrow 6 points.

Once we have the necessary points form a matrix in the form of

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

$2m \times 12$ 12×1 $2m \times 1$

Number of corresponding points.

And solve for $Ax = 0$ using SVD.

g. Once we have an estimate of the projection matrix u , we use the orthogonality of $\lambda_1, \lambda_2, \lambda_3$ and the orthogonality of a_1, a_2 and a_3 in the below equations

$$(1) \quad \alpha_u \lambda_1^T + \beta \lambda_2^T + u_0 \lambda_3^T = \beta a_1^T$$

$$(2) \quad \alpha_v \lambda_1^T + v_0 \lambda_3^T = \beta a_2^T$$

$$(3) \quad \lambda_3^T = \beta a_3^T$$

$$(4) \quad K^* T^* = \beta b$$

Some of the properties of orthogonality is stated below.

$$\lambda_1 \cdot \lambda_2 = 0$$

$$\lambda_2 \cdot \lambda_3 = 0$$

$$\lambda_1 \cdot \lambda_3 = 0$$

$$\lambda_2 \cdot \lambda_2 = 1$$

$$\lambda_1 \cdot \lambda_1 = 1$$

$$\lambda_1 \times \lambda_2 = \lambda_3$$

$$\lambda_2 \times \lambda_3 = \lambda_1$$

$$\lambda_3 \times \lambda_1 = \lambda_2$$

$$\lambda_3 \cdot \lambda_3 = 1$$

h.

Given $\{p_i\}_{i=1}^m \longleftrightarrow \{P_i\}$ and estimated $u = \begin{bmatrix} m_1^T \\ m_2^T \\ m_3^T \end{bmatrix}$
 assess the quality of fit using:

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

The above equation provides distance between known and predicted positions

Eg: $(7.3, 12.1, 15.3) \longleftrightarrow (5, 3)$ KNOWN \rightarrow $(6, 2)$ PREDICTED USING u \rightarrow L_1 \rightarrow ERROR

i. The below points show the principles and the approach for planar camera calibration.

- Estimate 2D homography (projective map) between calibration plane and image (for several images). → obtain $\hat{H}_1, \hat{H}_2, \hat{H}_3$
- Estimate intrinsic parameters. → using $\hat{H}_1, \hat{H}_2, \hat{H}_3$
- Compute extrinsic parameters for view of interest. (R, T) ←

For planar camera calibration we project using a 2D projective map as shown below.

$$\begin{aligned} \underset{2DH}{P_i^1} &= \underbrace{K^*}_{3 \times 3} \underbrace{[R^* | T^*]}_{3 \times 4} \underset{3DH}{P_i} \\ &= \underset{2 \times 2}{K^*} \begin{bmatrix} x_1 & x_2 & x_3 & T^* \end{bmatrix} \begin{bmatrix} X_i \\ Y_i \\ 0 \\ 1 \end{bmatrix} \end{aligned} \quad \left. \vphantom{\begin{aligned} \underset{2DH}{P_i^1} &= \underbrace{K^*}_{3 \times 3} \underbrace{[R^* | T^*]}_{3 \times 4} \underset{3DH}{P_i} \\ &= \underset{2 \times 2}{K^*} \begin{bmatrix} x_1 & x_2 & x_3 & T^* \end{bmatrix} \begin{bmatrix} X_i \\ Y_i \\ 0 \\ 1 \end{bmatrix} } \right\} \text{NON CO-PLANAR MODEL}$$

$$P_i^1 = \underbrace{K^* \begin{bmatrix} x_1 & x_2 & T^* \end{bmatrix}}_{\text{2D projective map}} \underbrace{\begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix}}_{\text{2DH}} \quad \left. \vphantom{\begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix}} \right\} \text{PLANAR MODEL}$$

j. We make a basic assumption that there are no coordinates present in the z plane \Rightarrow we assume that $z = 0$. We project the world plane point $(x_i, y_i, 1)$ to image coordinates $(x_i, y_i, 1)$.

with homography (2D projective map) = $K^* [x_1 \ x_2 \ 1^T]^*$
 instead of the projection matrix $M = K^* [x_1 \ x_2 \ x_3 \ 1^T]^*$

2. Camera Calibration 2.

a. given $\{p_i\} \leftrightarrow \{P_i\}$
 $(1, 2) \leftrightarrow (3, 4, 5)$

The first two lines must be of the form:

$$= \begin{bmatrix} P_i^T & 0 & -x_i P_i^T \\ 0 & P_i^T & -y_i P_i^T \end{bmatrix} \quad P_i^T = [3 \ 4 \ 5 \ 1]$$

2×12

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

b.

$$\overbrace{\begin{bmatrix} \alpha_u x_1^T + s x_2^T + u_0 x_3^T \\ \alpha_v x_1^T + v_0 x_3^T \\ x_3^T \end{bmatrix}}^{K^* R^*} K^{*T*} = f \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \end{bmatrix} = f \hat{M}$$

3×4 3×3 3×1

KNOWN

UNKNOWN

$$\begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \end{bmatrix} = \begin{bmatrix} -a_1^T & - \\ -a_2^T & - \\ -a_3^T & - \end{bmatrix} \quad b$$

$$|p| = \frac{1}{|a_3|}$$

$$u_0 = |p|^2 a_1 \cdot a_3$$

$$v_0 = |p|^2 a_2 \cdot a_3$$

$$|p|^2 = \frac{1}{3^2 + 4^2 + 5^2} = 0.02$$

$$u_0 = 0.02 (3 + 8 + 15) = 0.02 \times 26 = 0.52$$

$$v_0 = 0.02 (6 + 12 + 20) = 0.02 \times 38 = 0.76$$

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

Number of corresponding points.

here $m = 1$ $x_i = 1$ $y_i = 2$

$$m_1 = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$$

$$3 + 8 + 15 + 4$$

$$m_2 = \begin{bmatrix} 2 \\ 3 \\ 4 \\ 5 \end{bmatrix}$$

$$6 + 12 + 20 + 5$$

$$m_3 = \begin{bmatrix} 3 \\ 4 \\ 5 \\ 6 \end{bmatrix}$$

$$9 + 16 + 25 + 6$$

$$p_i = \begin{bmatrix} 3 \\ 4 \\ 5 \\ 1 \end{bmatrix}$$

$$E = \frac{1}{1} \left(\left(1 - \frac{30}{56} \right)^2 + \left(2 - \frac{43}{56} \right)^2 \right)$$

$$= 0.2155 + 1.5181 = 1.7336$$

Validation:

$$p_i = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \\ 5 \\ 1 \end{bmatrix} = \begin{bmatrix} 3+8+15+4 \\ 6+12+20+5 \\ 9+16+25+6 \end{bmatrix} = \begin{bmatrix} 30 \\ 43 \\ 56 \end{bmatrix}$$

actual point = (0.53, 0.76)

Yes the points are off.

d.

Given $R^* = I + Q$ $I = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$

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

$$R^* = \begin{bmatrix} 6 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad T^* = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 1 \end{bmatrix}$$

$$R = (R^*)^T$$

$$T = -(R^*)^T T^*$$

$$R = \begin{bmatrix} 6 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad T = - \begin{bmatrix} 6 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 1 \end{bmatrix} = \begin{bmatrix} -6 \\ -2 \\ -3 \\ -1 \end{bmatrix}$$

$$3DM \rightarrow 3I$$

e.

$$A = \begin{bmatrix} p_1^{*T} & 0 & -x_1 p_1^{*T} \\ 0 & p_1^{*T} & -y_1 p_1^{*T} \\ \vdots & \vdots & \vdots \\ p_m^{*T} & 0 & -x_m p_m^{*T} \\ 0 & p_m^{*T} & -y_m p_m^{*T} \end{bmatrix}$$

$2m \times 9$

$$p^{*T} = [3 \ 4 \ 1]$$

$$x_1 = 1 \quad y_1 = 2.$$

$$\begin{bmatrix} 3 & 4 & 1 & 0 & 0 & 0 & -3 & -4 & -1 \\ 0 & 0 & 0 & 3 & 4 & 1 & -6 & -8 & -2 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix}$$

The above matrix shows the first two rows.

3. Multiple view geometry:

a. The major difference between sparse and dense is that sparse is feature based and dense is correlation and SSD based.

The advantages of sparse stereo matching is that it can handle large disparities.

The advantages of dense stereo matching is that it produces more points. Disadvantage is that occlusion or ambiguous points can arise where it becomes difficult to correlate.

b.

10	11	10
13	11	10
...

11	13	10
10	13	11
...

10	11	10
13	11	10
...

100	50	10
70	100	10
...

lower similarity with higher correlation.
only because of higher value #s.

Normalize window values (Subtract mean and divide by standard deviation).

Zero mean normalized cross-correlation (ZNCC):

$$\psi(w_1, w_2) = \sum_i \frac{(w_1(x_i, y_i) - \mu_{w_1})(w_2(x_i, y_i) - \mu_{w_2})}{\sigma_{w_1} \sigma_{w_2}}$$

Zero mean normalized SSD (ZNSSD)

$$\psi(w_1, w_2) = \sum_i \left(\frac{w_1(x_i, y_i) - \mu_{w_1}}{\sigma_{w_1}} - \frac{w_2(x_i, y_i) - \mu_{w_2}}{\sigma_{w_2}} \right)^2$$

If we run the entire image as search space we may frequently observe:

- Object points not visible in both views (occlusion)
- Ambiguous points (multiple coordinates)
- Uniform regions (inside points cannot be distinguished).

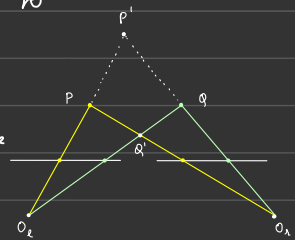
The search space can be reduced to a line by using axis aligned stereo where we will only need to search along the epipolar lines.

c. Given corresponding points
 $A(100, 200)$ $B(103, 200)$ $Z = ?$
 given $f : 10$ $T = 100$

we have: $Z = f \frac{T}{d}$ $d = -x_1 + x_2 = -100 + 103 = 3$

$$Z = 10 \times \frac{100}{3} = 333.34$$

d. Consider point P and Q projected on to two images as shown in the figure. O_L and O_R are two points produced by tracing the image points to the world point (origine).



While reconstruction if O_L and O_R are projected incorrectly P' and Q' will be the reconstructed points which are severely incorrect in correspondence to the original points.

e. calibration: $\left\{ \begin{array}{l} R_L, T_L \\ R_R, T_R \end{array} \right\}$ Rotation / Translation of left / right camera with respect to world

$$M_{\text{Right} \leftarrow \text{Left}} = M_{\text{Right} \leftarrow \text{World}} M_{\text{World} \leftarrow \text{Left}}$$

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

$$M_{\text{Right} \leftarrow \text{Left}} = R_R^T T_R^{-1} T_L R_L = TR$$

4. Multiple view geometry 2:

a. Given = $[f | 11: 10 \text{ mm} | 23]$ $\begin{bmatrix} \pi \\ 2 \end{bmatrix} = \begin{bmatrix} 20 \text{ mm} \\ 11 + 34 + 23 = 68 \end{bmatrix}$ $\text{parity} = 30 \text{ mm}$

we have $Z = \frac{1}{f} \frac{T}{d} = 10 \times \frac{20}{30} = 6.67 \text{ mm}$.

b. $A \times B = [A]_{\text{matrix}} \times B_{\text{vector}} \equiv \begin{bmatrix} 0 & -a_z & a_y \\ a_z & 0 & -a_x \\ -a_y & a_x & 0 \end{bmatrix} B$ Rank 2 skew symmetric matrix

given $a_x = 1 \quad a_y = 2 \quad a_z = 3$

\Rightarrow vector $A_x \equiv \begin{bmatrix} 0 & -3 & 2 \\ 3 & 0 & -1 \\ -2 & 1 & 0 \end{bmatrix}$

c. Given $f = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 3 & 4 & 5 \end{bmatrix}$ $P_1^T = [2 \ 3 \ 1]$ $P_2^T = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$

$P_1^T F P_2 = [2 \ 3 \ 1] \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 3 & 4 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$

$= \begin{bmatrix} 2+6+3 & 4+9+4 & 6+12+5 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$

$$= \begin{bmatrix} 11 & 17 & 23 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = 11 + 34 + 23 = 68$$

$$\text{Initially } p_1^T F p_1 = 0$$

d. Given $p_1 = \begin{pmatrix} x_1 & y_1 \\ 1 & 2 \end{pmatrix}$ $p_1' = \begin{pmatrix} x_1' & y_1' \\ 2 & 3 \end{pmatrix}$

$$\begin{bmatrix} x_1 x_1' & x_1 y_1' & x_1 & y_1 x_1' & y_1 y_1' & y_1 & x_1' & y_1' & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x_m x_m' & x_m y_m' & x_m & y_m x_m' & y_m y_m' & y_m & x_m' & y_m' & 1 \end{bmatrix} \begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ \vdots \\ f_{31} \\ f_{32} \\ f_{33} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$\text{KNOWN } m \times 9$
 \uparrow
 $\text{UNKNOWN } 9 \times 1$
 $m \times 1$

Required row \nearrow

$$\begin{bmatrix} 2 & 3 & 1 & 4 & 6 & 2 & 2 & 3 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x_m x_m' & x_m y_m' & x_m & y_m x_m' & y_m y_m' & y_m & x_m' & y_m' & 1 \end{bmatrix} \begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ \vdots \\ f_{31} \\ f_{32} \\ f_{33} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$\text{KNOWN } m \times 9$
 \uparrow
 $\text{UNKNOWN } 9 \times 1$
 $m \times 1$

END