

2018-17119 박수현 컴퓨터비전 HW3 theory writeup

Question1: Camera model

(30 points)

Consider the camera model below.

$$\begin{array}{l} \text{image} \\ \left(\begin{array}{c} u \\ v \\ 1 \end{array} \right) = \mathbf{K} (\mathbf{R} | \mathbf{t}) \left(\begin{array}{c} X \\ Y \\ Z \\ 1 \end{array} \right) \\ \text{camera} \rightarrow \text{image} \quad \text{world} \rightarrow \text{camera} \end{array} \quad (1)$$

$$\mathbf{K} = \begin{pmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{pmatrix}, \mathbf{R} = \begin{pmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{pmatrix}, \mathbf{t} = \begin{pmatrix} t_1 \\ t_2 \\ t_3 \end{pmatrix}, \quad (2)$$

The coordinate $[u, v, 1]^T$ and $[X, Y, Z, 1]^T$ represent homogeneous coordinates in the image and world coordinate system, respectively. \mathbf{K} is the camera intrinsic matrix, \mathbf{R} is the rotation matrix, and \mathbf{t} is the translation vector. Note that the rotation matrix has some unique properties, that each row vectors and column vectors are unit vectors, and $\mathbf{R}^{-1} = \mathbf{R}^T$. (You can use these properties without any proof.)

- (a) What is the image coordinates of the world coordinate system's origin $O(0, 0, 0)$? [5 points]

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_1 \\ r_{21} & r_{22} & r_{23} & t_2 \\ r_{31} & r_{32} & r_{33} & t_3 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} t_1 \\ t_2 \\ t_3 \end{bmatrix} = \begin{bmatrix} ft_1 \\ ft_2 \\ t_3 \end{bmatrix}$$

Converting from homogeneous coordinates, $u = \frac{ft_1}{t_3}$, $v = \frac{ft_2}{t_3}$
 $\therefore \left(f\frac{t_1}{t_3}, f\frac{t_2}{t_3} \right)$

- (b) What is the coordinate of the camera in the world coordinate system? [5 points]

The camera coordinate frame is related to the world coordinate frame by a rotation (to align axes) and a translation (to align origins).

Let coordinates of a point in the camera frame $X_c = (x_c, y_c, z_c)$

coordinates of a point in the world frame $X_w = (x_w, y_w, z_w)$

coordinates of the camera center in the world frame $C = (c_1, c_2, c_3)$

$\Rightarrow X_c = R(X - C)$, and in matrix representation, $X_c = [R \mid -RC] \begin{bmatrix} x_w \\ 1 \end{bmatrix}$
 $(\because \text{rotation \& translation})$

As the extrinsic camera parameters in the model can be represented as a 4×4 3D transformation matrix $[R \mid t]$, $t = -RC$

$t = -RC \Leftrightarrow -t = RC \Leftrightarrow (R^{-1})(-t) = C \Leftrightarrow R^T(-t) = C$

$$C = (R^T)(-t) = \begin{bmatrix} r_{11} & r_{21} & r_{31} \\ r_{12} & r_{22} & r_{32} \\ r_{13} & r_{23} & r_{33} \end{bmatrix} \begin{bmatrix} -t_1 \\ -t_2 \\ -t_3 \end{bmatrix} = \begin{bmatrix} -r_{11}t_1 & -r_{21}t_2 & -r_{31}t_3 \\ -r_{12}t_1 & -r_{22}t_2 & -r_{32}t_3 \\ -r_{13}t_1 & -r_{23}t_2 & -r_{33}t_3 \end{bmatrix}$$

$$\begin{bmatrix} -r_{11}t_1 & -r_{21}t_2 & -r_{31}t_3 \\ -r_{12}t_1 & -r_{22}t_2 & -r_{32}t_3 \\ -r_{13}t_1 & -r_{23}t_2 & -r_{33}t_3 \end{bmatrix} \quad (3 \times 1 \text{ matrix})$$

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_1 \\ r_{21} & r_{22} & r_{23} & t_2 \\ r_{31} & r_{32} & r_{33} & t_3 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

(c) In the camera coordinate system, camera is looking at $[0, 0, 1]^T$ (i.e. $+z$). Convert this direction into the world coordinate system with a unit vector. [8 points]

Followed by Question 1 (b), $X_c = RX_w + T \iff X_w = R^T(X_c - t)$

Given $X_c = (0, 0, 1)$

$$R^T(X_c - t) = \begin{bmatrix} r_{11} & r_{21} & r_{31} \\ r_{12} & r_{22} & r_{32} \\ r_{13} & r_{23} & r_{33} \end{bmatrix} \begin{bmatrix} -t_1 \\ -t_2 \\ -t_3 \end{bmatrix}' = \begin{bmatrix} -r_{11}t_1 - r_{21}t_2 + r_{31}(1-t_3) \\ -r_{12}t_1 - r_{22}t_2 + r_{32}(1-t_3) \\ -r_{13}t_1 - r_{23}t_2 + r_{33}(1-t_3) \end{bmatrix} \quad \leftarrow V \text{ (a } 3 \times 1 \text{ matrix)}$$

To find a unit vector with the same vector as V , we divide V by its magnitude.

$$\|V\| = \sqrt{(-r_{11}t_1 - r_{21}t_2 + r_{31}(1-t_3))^2 + (-r_{12}t_1 - r_{22}t_2 + r_{32}(1-t_3))^2 + (-r_{13}t_1 - r_{23}t_2 + r_{33}(1-t_3))^2}$$

$$\therefore \frac{V}{\|V\|} = \frac{1}{\sqrt{(-r_{11}t_1 - r_{21}t_2 + r_{31}(1-t_3))^2 + (-r_{12}t_1 - r_{22}t_2 + r_{32}(1-t_3))^2 + (-r_{13}t_1 - r_{23}t_2 + r_{33}(1-t_3))^2}} \begin{bmatrix} -r_{11}t_1 - r_{21}t_2 + r_{31}(1-t_3) \\ -r_{12}t_1 - r_{22}t_2 + r_{32}(1-t_3) \\ -r_{13}t_1 - r_{23}t_2 + r_{33}(1-t_3) \end{bmatrix}$$

(d) The straight line in the world coordinate system can be expressed as $[d_1, d_2, d_3]^T t + [x_1, x_2, x_3]^T$ ($t \in \mathbf{R}$). Assume that this line is not parallel to the xy-plane and does not pass through the origin in the camera coordinate system. What is the expression of this line in the image? [12 points]

Let line $\ell = \begin{bmatrix} \text{direction} \\ d_1 \\ d_2 \\ d_3 \end{bmatrix} t + \begin{bmatrix} \text{point} \\ x_1 \\ x_2 \\ x_3 \end{bmatrix}, \quad t \in \mathbf{R}$

$$K(R(t))\ell = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_1 \\ r_{21} & r_{22} & r_{23} & t_2 \\ r_{31} & r_{32} & r_{33} & t_3 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix} t + \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} fr_{11} & fr_{12} & fr_{13} & ft_1 \\ fr_{21} & fr_{22} & fr_{23} & ft_2 \\ fr_{31} & fr_{32} & fr_{33} & ft_3 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix} t + \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

$$\therefore \left(\begin{bmatrix} fr_{11}d_1 + fr_{12}d_2 + fr_{13}d_3 + ft_1 \\ fr_{21}d_1 + fr_{22}d_2 + fr_{23}d_3 + ft_2 \\ fr_{31}d_1 + fr_{32}d_2 + fr_{33}d_3 + ft_3 \end{bmatrix} t + \begin{bmatrix} fr_{11}x_1 + fr_{12}x_2 + fr_{13}x_3 + ft_1 \\ fr_{21}x_1 + fr_{22}x_2 + fr_{23}x_3 + ft_2 \\ fr_{31}x_1 + fr_{32}x_2 + fr_{33}x_3 + ft_3 \end{bmatrix} \right) \div$$

$$\therefore \frac{1}{fr_{31}d_1 + fr_{32}d_2 + fr_{33}d_3 + ft_3} \begin{bmatrix} fr_{11}d_1 + fr_{12}d_2 + fr_{13}d_3 + ft_1 \\ fr_{21}d_1 + fr_{22}d_2 + fr_{23}d_3 + ft_2 \end{bmatrix} t$$

$$+ \frac{1}{fr_{31}x_1 + fr_{32}x_2 + fr_{33}x_3 + ft_3} \begin{bmatrix} fr_{11}x_1 + fr_{12}x_2 + fr_{13}x_3 + ft_1 \\ fr_{21}x_1 + fr_{22}x_2 + fr_{23}x_3 + ft_2 \end{bmatrix}, \quad fr \in \mathbf{R}$$

Question2: Camera Calibration

(10 points)

A camera is a mapping between the 3D world $\mathbf{X}_i = [X_i \ Y_i \ Z_i \ 1]^T$ and a 2D image $\mathbf{x}_i = [u_i \ v_i \ 1]^T$ (both are homogeneous coordinates). A camera projection matrix $\mathbf{P} = [m_{ij}]$ maps $\mathbf{x}_i = \mathbf{P}\mathbf{X}_i$. Given $n \geq 6$ pairs of corresponding points, we have an over-determined system of equations, $\mathbf{Ap} = \mathbf{0}$, where \mathbf{p} is the vectorized form of \mathbf{P} . Consider the SVD of \mathbf{A} as $\mathbf{A} = \mathbf{U}\Sigma\mathbf{V}^T$. Show that the solution $\mathbf{p} = \min_p \|\mathbf{Ap}\|, \|\mathbf{p}\| = 1$ is the last column of \mathbf{V} corresponding to smallest singular value of \mathbf{A} .

or smallest eigenvalue of

$$\mathbf{A}^T \mathbf{A}$$

$$\begin{matrix} \mathbf{x}_i \\ u \\ v \\ 1 \end{matrix} = \begin{matrix} \mathbf{P} \\ m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{matrix} \begin{matrix} \mathbf{X}_i \\ X \\ Y \\ Z \\ 1 \end{matrix} \quad (3)$$

Singular value

\mathbf{P} : 12×1 matrix ("flatten-out" vectorized form of \mathbf{P})

\mathbf{A} : $2n \times 12$ matrix ($2n \geq 12 \therefore n \geq 6$)

For background information, the SVD of \mathbf{A} is a decomposition of the form

$$\begin{matrix} 12 \\ \mathbf{A} \\ 2n \end{matrix} = \begin{matrix} 2n \\ \mathbf{U} \\ 2n \end{matrix} \quad \begin{matrix} 12 \\ \Sigma \\ 12 \end{matrix} \quad \begin{matrix} 12 \\ \mathbf{V}^T \\ 12 \end{matrix}$$

orthogonal diagonal orthogonal

↳ diagonal elements are singular values $\sigma_1, \dots, \sigma_{12}$; others are all zeros.

Let the solution be $\hat{\mathbf{p}}$, and I will prove that $\hat{\mathbf{p}}$ is the last column of \mathbf{V} corresponding to the smallest singular value of \mathbf{A} .

$$\begin{aligned} \hat{\mathbf{p}} &= \min_{\mathbf{p}} \|\mathbf{U}\Sigma\mathbf{V}^T \mathbf{p}\| \\ &= \min_{\mathbf{p}} \|\Sigma\mathbf{V}^T \mathbf{p}\| \quad \leftarrow \|\mathbf{U}\Sigma\mathbf{V}^T \mathbf{p}\| = \|\Sigma\mathbf{V}^T \mathbf{p}\| \end{aligned}$$

∴ An orthogonal $n \times m$ matrix \mathbf{Q} satisfies $\mathbf{Q}^T \mathbf{Q} = \mathbf{I}_m$, i.e., $\mathbf{Q}^T = \mathbf{Q}^{-1}$ by definition.

Note that \mathbf{U} and \mathbf{V}^T are orthogonal matrices:

$$\begin{aligned} \|\mathbf{U}\Sigma\mathbf{V}^T \mathbf{p}\| &= \sqrt{\mathbf{U}\Sigma\mathbf{V}^T \mathbf{p} \cdot \mathbf{U}\Sigma\mathbf{V}^T \mathbf{p}} = \sqrt{(\mathbf{U}\Sigma\mathbf{V}^T \mathbf{p})^T \mathbf{U}\Sigma\mathbf{V}^T \mathbf{p}} \\ &= \sqrt{(\Sigma\mathbf{V}^T \mathbf{p})^T \mathbf{U}^T \mathbf{U}\Sigma\mathbf{V}^T \mathbf{p}} = \sqrt{(\Sigma\mathbf{V}^T \mathbf{p})^T \Sigma\mathbf{V}^T \mathbf{p}} = \|\Sigma\mathbf{V}^T \mathbf{p}\| \\ &= \mathbf{I}_{2n} \end{aligned}$$

$$= \min_{\mathbf{p}} \|\Sigma\mathbf{y}\| \quad \leftarrow \text{substitute } \mathbf{y} = \mathbf{V}^T \mathbf{p} \quad (12 \times 1 \text{ matrix})$$

In practice, the singular values in Σ are in descending order, i.e., $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r$.
(cf. `numpy.linalg.svd` outputs Σ as vector with the singular values, within each vector sorted in descending order, as well)

Since $\hat{p} = \min_p \|\Sigma y\|$, in order for \hat{p} to contain the smallest singular value σ_{12} , the best option for y is $y = [0, 0, \dots, 0, 1]^T$.

Recalling that $y = V^T p$ from substitution,

$$p = V y \quad (\because V^T \text{ is orthogonal, so } V = (V^T)^{-1})$$

$$\text{and } \|p\| = 1 \quad (\because \|p\| = \|V y\| = \|y\| = 1, V \text{ is orthogonal})$$

Therefore, the solution \hat{p} where $\|p\| = 1$ is the last column of V , which corresponds to the smallest singular value.

Question3: Homography

(10 points)

We can use homographies to create a panoramic image from multiple views of the same scene. This is possible when there is no camera translation between the views (e.g. only rotation about the camera center). In this case, corresponding points from two views of the same scene can be related by a homography:

$$\mathbf{p}_1^i \equiv \mathbf{H} \mathbf{p}_2^i \quad (4)$$

where \mathbf{p}_1^i and \mathbf{p}_2^i denote the homogeneous coordinates (e.g., $\mathbf{p}_1^i \equiv (x, y, 1)^T$) of the 2D projection of the i -th point in images 1 and 2 respectively, and \mathbf{H} is a 3×3 matrix representing the homography. Given N point correspondences and using (4), derive a set of $2N$ independent linear equations in the form $\mathbf{A}\mathbf{h} = \mathbf{b}$ where \mathbf{h} is a 9×1 vector containing the unknown entries of \mathbf{H} .

(a) What are the expressions for \mathbf{A} and \mathbf{b} ? [3 points]

$$\text{Let } \mathbf{p}_1^i = (x_1, y_1, 1)^T, \mathbf{p}_2^i = (x_2, y_2, 1)^T \\ \mathbf{p}_1^i \equiv \mathbf{H} \mathbf{p}_2^i \iff \begin{bmatrix} x_1 \\ y_1 \\ 1 \end{bmatrix} \cong \begin{bmatrix} h_{00} & h_{01} & h_{02} \\ h_{10} & h_{11} & h_{12} \\ h_{20} & h_{21} & h_{22} \end{bmatrix} \begin{bmatrix} x_2 \\ y_2 \\ 1 \end{bmatrix}, \quad x_1 = \frac{h_{00}x_2 + h_{01}y_2 + h_{02}}{h_{20}x_2 + h_{21}y_2 + h_{22}}, \quad y_1 = \frac{h_{10}x_2 + h_{11}y_2 + h_{12}}{h_{20}x_2 + h_{21}y_2 + h_{22}}$$

$$\Rightarrow x_1(h_{20}x_2 + h_{21}y_2 + h_{22}) = h_{00}x_2 + h_{01}y_2 + h_{02}$$

$$y_1(h_{20}x_2 + h_{21}y_2 + h_{22}) = h_{10}x_2 + h_{11}y_2 + h_{12}$$

Using the linear equations, $\mathbf{p}_1^i \equiv \mathbf{H} \mathbf{p}_2^i$ can be expressed in the form $\mathbf{A}\mathbf{h} = \mathbf{b}$ where

$$\left[\begin{array}{ccccccccc} x_2 & y_2 & 1 & 0 & 0 & 0 & -x_1x_2 & -x_1y_2 & -x_1 \\ 0 & 0 & 0 & x_2y_2 & 1 & -y_1x_2 & -y_1y_2 & -y_1 \\ & & & & & \vdots & & & \\ x_2 & y_2 & 1 & 0 & 0 & 0 & -x_1x_2 & -x_1y_2 & -x_1 \\ 0 & 0 & 0 & x_2y_2 & 1 & -y_1x_2 & -y_1y_2 & -y_1 \\ & & & & & \vdots & & & \\ x_2^N & y_2^N & 1 & 0 & 0 & 0 & -x_1x_2^N & -x_1y_2^N & -x_1 \\ 0 & 0 & 0 & x_2y_2^N & 1 & -y_1x_2^N & -y_1y_2^N & -y_1^N \end{array} \right] \begin{bmatrix} h_{00} \\ h_{01} \\ h_{02} \\ h_{10} \\ h_{11} \\ h_{12} \\ h_{20} \\ h_{21} \\ h_{22} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{bmatrix}$$

$A \quad 2N \times 9$ $h \quad 9 \times 1$ $b \quad 2N \times 1$

(b) How many correspondences will be needed to solve for \mathbf{h} ? [4 points]

4 or more ($N \geq 4$)

\because while \mathbf{h} has 9 unknown entries, we can multiply all h_{ij} (i,j $\in \{0,1,2\}$) by non-zero scaling factor k without changing the equations. Therefore, the degree of freedom is 8. Since each point has x and y coordinates, with at least 4 correspondences, the problem would be defined as an overdetermined system of equations. As a result, we can apply the least squares technique.

(c) How can we get \mathbf{H} ? (Hint: Question 2) [3 points]

As soon as we find \mathbf{h} , we can recover \mathbf{H} by reshaping the matrix from 9×1 to 3×3 .

\mathbf{h} is a (non zero) solution to a least squares problem $\|\mathbf{A}\mathbf{h}\|^2$.

Followed by Question 2, considering the SVD of \mathbf{A} as $\mathbf{A} = \mathbf{U}\Sigma\mathbf{V}^T$, the solution \mathbf{h} for $\min \|\mathbf{A}\mathbf{h}\|$ is the last column of \mathbf{V} corresponding to the smallest singular value of \mathbf{A} .

Note that minimizing $\|\mathbf{A}\mathbf{h}\|^2$ and $\|\mathbf{A}\mathbf{h}\|$ should both yield the same solution vector \mathbf{h} .

In conclusion, we can get \mathbf{H} through the following steps:

- ① Construct a $2N \times 9$ matrix \mathbf{A} using coordinates of p_1 and p_2 , as presented in Question 3 (a).
- ② Calculate the singular value decomposition of matrix \mathbf{A} s.t. $\mathbf{A} = \mathbf{U}\Sigma\mathbf{V}^T$
- ③ Retrieve the last column of \mathbf{V} (= last row of \mathbf{V}^T) as \mathbf{h} .
- ④ Reshape \mathbf{h} (9×1 matrix) into \mathbf{H} (3×3 matrix).