

Computer Vision 2016 Spring HW#3 theory

2013-11415 Sanha Lee

Q1.

(a) Known part is like below.

for world point X , camera 1 has X_{c1} , camera 2 has X_{c2} .

for $M_1 = A_1[R_1|t_1]$ and $M_2 = A_2[R_2|t_2]$, $X_{c1} = R_1X + t_1$ (1), $X_{c2} = R_2X + t_2$ (2).

for length constant z_1 and z_2 , $p = \frac{1}{z_1}X_{c1}$ (3), $q = \frac{1}{z_2}X_{c2}$ (4).

Let's think about a projection of reference camera center to plane π . If the distance between the plane π and the center of camera 2 is d_2 , we can consider a normal vectors n_2 from camera 2 center to plane π , which's length is $\frac{1}{d_2}$.

Then, we can know that

$$n_2^T X_{c2} = 1 \quad (5).$$

Now, let's think about $\frac{z_1}{z_2}p$.

$$\frac{z_1}{z_2}p = \frac{1}{z_2}(R_1X + t_1) = \frac{1}{z_2}(R_1R_2^{-1}(X_{c2} - t_2) + t_1) \quad (\text{by (3) and (1)})$$

(R_1 must have inverse because it's rotation)

$$= \frac{1}{z_2}(R_1R_2^{-1}(z_2q - t_2) + t_1) = R_1R_2^{-1}q - \frac{1}{z_2}R_1R_2^{-1}t_2 + \frac{1}{z_2}t_1 \quad (\text{by (4)})$$

$$= (R_1R_2^{-1} - R_1R_2^{-1}t_2 \cdot n_2^T + t_1 \cdot n_2^T) q \quad (\text{by (5) and (4)})$$

Then, we can get the homography matrix $H = (R_1R_2^{-1} - R_1R_2^{-1}t_2 \cdot n_2^T + t_1 \cdot n_2^T)$.

This H satisfy the equation $\frac{z_1}{z_2}p = Hq$, which means it is a projective transformation of the plane.

Therefore, this H exists for any point X .

(b) Homography matrix is 3×3 , therefore we have 9 unknowns. However, because it shows homogeneous equality, we can re-scale it. Therefore, only 8 unknowns left.

$$\begin{bmatrix} wx'_i \\ wy'_i \\ w \end{bmatrix} = \begin{bmatrix} h_{00} & h_{01} & h_{02} \\ h_{10} & h_{11} & h_{12} \\ h_{20} & h_{21} & h_{22} \end{bmatrix} \begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix}$$

Because of above matrix equation, one pair of point makes two equations like below.

$$x'_i(h_{20}x_i + h_{21}y_i + h_{22}) = h_{00}x_i + h_{01}y_i + h_{02}$$

$$y'_i(h_{20}x_i + h_{21}y_i + h_{22}) = h_{10}x_i + h_{11}y_i + h_{12}$$

Therefore, to reveal 8 unknowns, 4 pairs of points are needed.

(c) From above equation, we can construct next equation using n pairs of points.

$$\mathbf{Ah} = \mathbf{0} \text{ when } \mathbf{A} = \begin{bmatrix} x_1 & y_1 & 1 & 0 & 0 & 0 & -x'_1x_1 & -x'_1y_1 & -x'_1 \\ 0 & 0 & 0 & x_1 & y_1 & 1 & -y'_1x_1 & -y'_1y_1 & -y'_1 \\ x_2 & y_2 & 1 & 0 & 0 & 0 & -x'_2x_2 & -x'_2y_2 & -x'_2 \\ 0 & 0 & 0 & x_2 & y_2 & 1 & -y'_2x_2 & -y'_2y_2 & -y'_2 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x_n & y_n & 1 & 0 & 0 & 0 & -x'_nx_n & -x'_ny_n & -x'_n \\ 0 & 0 & 0 & x_n & y_n & 1 & -y'_nx_n & -y'_ny_n & -y'_n \end{bmatrix}, \quad \mathbf{h} = \begin{bmatrix} h_{00} \\ h_{01} \\ h_{02} \\ h_{10} \\ h_{11} \\ h_{12} \\ h_{20} \\ h_{21} \\ h_{22} \end{bmatrix}, \quad \mathbf{0} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix},$$

Then, a least squares could be performed to $\|\mathbf{Ah} - \mathbf{0}\|^2$.

For \hat{h} which is unit vector and $h \equiv \hat{h}$, solution of least squares problem $\|\mathbf{A}\hat{h} - \mathbf{0}\|^2$ is

$\hat{h} = \text{eigenvector of } A^T A \text{ with smallest eigenvalue.}$