Lecture: Hands-on AI Based 3D Vision Assignment 02

Due: May 27th, 12:00am

1 Epipolar Geometry Conventions

In epipolar geometry, there are two cameras C_0 and C_1 , viewing the same scene from different viewpoints. C_0 's coordinate system is assumed to be aligned with the world coordinate system.

In the lecture we used the following conventions for deriving the epipolar constraint:

- R is the rotation matrix that maps vectors from camera 1's frame to camera 0's frame.
- \bullet **t** is the position of camera 1 expressed in world frame (= camera ='s frame)

It follows, that the projection matrices of the cameras are

$$\mathbf{P}_0 = \mathbf{K}_0 \begin{bmatrix} \mathbf{I} & \mathbf{0} \end{bmatrix} \qquad \qquad \mathbf{P}_1 = \mathbf{K}_1 \begin{bmatrix} \mathbf{R}^T & -\mathbf{R}^T \mathbf{t} \end{bmatrix}$$

In this scenario, the epipolar constraint can be written as

$$\mathbf{u}_0^T(\mathbf{K}_0^{-T}[\mathbf{t}]_{\times}\mathbf{R}\mathbf{K}_1^{-1})\mathbf{u}_1 = 0$$

or in terms of normalized coordinates

$$\mathbf{x}_0^T \underbrace{([\mathbf{t}]_{\times} \mathbf{R})}_{\mathbf{E}} \mathbf{x}_1 = 0$$

In other sources (such as Hartley-Zisserman), you will find different conventions. Here $\hat{\mathbf{R}}$ and $\hat{\mathbf{t}}$ denote camera 1's extrinsics, i.e.

$$\mathbf{\hat{P}}_1 = \mathbf{K}_1 \begin{bmatrix} \mathbf{\hat{R}} & \mathbf{\hat{t}} \end{bmatrix}$$

It follows that $\hat{\mathbf{R}} = \mathbf{R}^T$ and $\hat{\mathbf{t}} = -\mathbf{R}^T \mathbf{t}$. They then derive the epipolar constraint as

$$\mathbf{x}_1^T \underbrace{([\hat{\mathbf{t}}]_{\times} \hat{\mathbf{R}})}_{\hat{\mathbf{r}}} \mathbf{x}_0 = 0.$$

Note the swapping of \mathbf{x}_0 and \mathbf{x}_1 !

(a) Show that the two conventions are equivalent, i.e. $\mathbf{E} = \hat{\mathbf{E}}^T$.

Hint: You may use the fact that $[\mathbf{R}\mathbf{v}]_{\times} = \mathbf{R}[\mathbf{v}]_{\times}\mathbf{R}^T$ for rotation matrices

Hartley and Zisserman derive the following theorem (cf. Result 9.19)

Theorem 1 For a given essential matrix $\hat{\mathbf{E}} = \hat{\mathbf{U}} \operatorname{diag}(1, 1, 0) \hat{\mathbf{V}}^T$, the extrinsics of camera C_1 are any combination of:

•
$$\hat{\mathbf{t}} = \pm \hat{\mathbf{u}}_3$$

•
$$\hat{\mathbf{R}} = \hat{\mathbf{U}} \mathbf{W} \hat{\mathbf{V}}^T$$
 or $\hat{\mathbf{U}} \mathbf{W}^T \hat{\mathbf{V}}^T$

where

$$\mathbf{W} = \begin{bmatrix} & -1 \\ 1 & & \\ & & 1 \end{bmatrix}$$

(b) Show that from this we can derive the theorem for our convention, as shown in the lecture:

Theorem 2 For a given essential matrix $\mathbf{E} = \mathbf{U} \operatorname{diag}(1, 1, 0) \mathbf{V}^T$, the extrinsics of camera C_1 are any combination of:

•
$$\mathbf{t} = \pm \mathbf{u}_3$$

•
$$\mathbf{R} = \mathbf{U}\mathbf{W}\mathbf{V}^T$$
 or $\mathbf{U}\mathbf{W}^T\mathbf{V}^T$