

## ECE 5554: CV: HW-4

$\Rightarrow$

1. a) Essential Matrix

$$E = [t]_x R$$

$t_x$ : Skew Symmetric matrix

$R$ : Rotation Matrix, which is  $I$  - Identity matrix  
our case

$t$  translation is  $(F, 0, f)^T$  as camera translates by  $F$  in  $x'$  and  $z'$  directions not in  $y'$

$R = I$  is the identity matrix

$$[t]_x = \begin{bmatrix} 0 & -f & 0 \\ f & 0 & -F \\ 0 & F & 0 \end{bmatrix}$$

$$E = [t]_x$$

$$E = \begin{bmatrix} 0 & -f & 0 \\ f & 0 & -F \\ 0 & F & 0 \end{bmatrix}$$



## → Finding Epipole

Epipole in 2<sup>nd</sup> image is projection of the first camera center into the second image. It lies at point  $e$ , where  $Ee=0$

$(F, 0, F)^T \rightarrow$  The camera translated by

epipole ( $e$ ) in the second image  
 $e = (F, 0, F)^T$

Normalizing to image coordinates divide by last coordinate

$$e_{\text{image}} = \left( \frac{F}{F}, 0, \frac{F}{F} \right)^T = (1, 0, 1)^T$$

$$e_{\text{image}} = (1, 0) \text{ in normalized coordinates}$$



$$2.0) \quad E = [A]_x R$$

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

$$R = \begin{bmatrix} \cos 30^\circ & -\sin 30^\circ & 0 \\ \sin 30^\circ & \cos 30^\circ & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$[A]_x = \begin{bmatrix} 0 & -3 & 2 \\ 3 & 0 & -1 \\ -2 & 1 & 0 \end{bmatrix}$$

$$3) \quad E = [A]_{xc} R$$

$$= \begin{bmatrix} -1.5 & -2.598 & 2 \\ 2.598 & -1.5 & -1 \\ -1.232 & 1.866 & 0 \end{bmatrix}$$



b) Rank of  $E$ :  
Using (SVD) Singular Value Decomposition

$$\Sigma = [3.7417, 3.7417, 0]$$

Rank of  $E$  is 2 as there are two significant non-zero values.

```
[1] import numpy as np

# Define the translation vector t
t = np.array([1, 2, 3])

# Define the rotation matrix R
theta = np.radians(30) # Convert angle to radians
R = np.array([
    [np.cos(theta), -np.sin(theta), 0],
    [np.sin(theta), np.cos(theta), 0],
    [0, 0, 1]
])

# Compute the skew-symmetric matrix [t]_x
t_cross = np.array([
    [0, -t[2], t[1]],
    [t[2], 0, -t[0]],
    [-t[1], t[0], 0]
])

# Compute the essential matrix E = [t]_x * R
E = np.dot(t_cross, R)

# Perform Singular Value Decomposition (SVD) to compute the rank of E
U, S, Vt = np.linalg.svd(E)
rank_E = np.sum(S > 1e-10) # Count non-zero singular values

E, S, rank_E
```

```
⇒ (array([[ -1.5          , -2.59807621,  2.          ],
          [ 2.59807621, -1.5          , -1.          ],
          [-1.23205081,  1.8660254   ,  0.          ]]),
   array([3.74165739e+00, 3.74165739e+00, 1.98024198e-17]),
   2)
```