# **Computer Vision: Ex3**

Or Tal

### Part one

## Compute the fundamental matrix using normalized 8 points algorithm, and draw the epipolar lines

Note: please see attached code files and sources,

after unpacking simply run from shell: python3 eight\_point\_alg.py

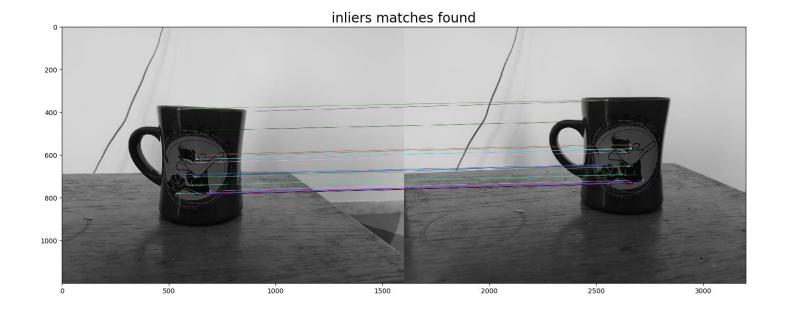
## input images:



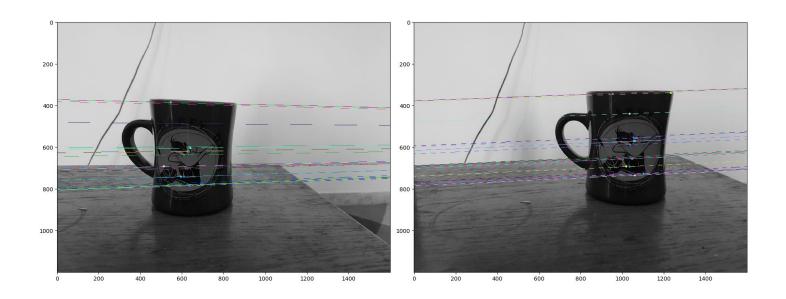


all matches found

200 
400 
800 
1000 
500 1000 1500 2000 2500 3000



our result: epipolar lines



#### 1. Show how the linear triangulation method extends to n > 2 images

The linear triangulation works for two images, hence in case of n > 2 images it should be about the same for the following point correspondence  $x_1, \dots, x_n$  (indexed by image) with  $P_1, \dots, P_n$  according camera matrices we could then describe the matrix A as following:

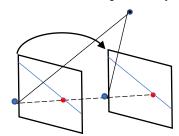
$$for \ \mathbf{x}_{i} = (x_{i}, y_{i})^{T} \to A = \begin{bmatrix} x_{1} \mathbf{p}_{1}^{3T} - \mathbf{p}_{1}^{1T} \\ y_{1} \mathbf{p}_{1}^{3T} - \mathbf{p}_{1}^{2T} \\ x_{2} \mathbf{p}_{2}^{3T} - \mathbf{p}_{2}^{1T} \\ y_{2} \mathbf{p}_{2}^{3T} - \mathbf{p}_{2}^{2T} \\ y_{2} \mathbf{p}_{2}^{3T} - \mathbf{p}_{2}^{2T} \\ \vdots \\ x_{n} \mathbf{p}_{n}^{3T} - \mathbf{p}_{n}^{1T} \\ y_{n} \mathbf{p}_{n}^{3T} - \mathbf{p}_{n}^{2T} \end{bmatrix}_{2n \times 4}$$
 where  $\mathbf{p}_{i}^{JT}$  is the j'th row of the i'th camera matrix.

The 3D homogeneous point **X** is the one satisfying  $AX = \mathbf{0}$ 

One way to find X would be computed by taking the last vector of V matrix in the SVD of A:  $A = UDV^T$ 

#### 2. Derive a method for triangulation in the case of pure translational motion of the cameras

We had previously seen that in case of pure translation, the fundamental matrix would be  $F = [e']_{\times}$ 

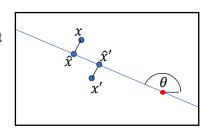


Due to that, we may conclude that the epipolar lines which correspond to the same point X in the world frame, should be the same, as the epipole remains at the same.

We may then conclude that we may parametrize the epipolar line by its relevant angle  $\theta$ where the epipole is its axis, and hence calculate the min geometric distance

$$c(x,x') = d(x,l(\theta))^{2} + d(x',l(\theta))$$

After doing so, we would estimate  $\hat{x}$ ,  $\hat{x}'$  matching x, x' on the epipolar line we just found, hence assuming the translation is accurate and the only distortion is in the measurements, the lines from camera centers via  $\hat{x}$ ,  $\hat{x}'$  would then lie on a single plane, therefor they would intersect at an estimated 3D point  $\hat{X}$ 



Purposed method for the triangulation in case of pure motion: (this assumes that the translation is known)

- 1. Assume that the first camera is positioned at the origin, and the second camera is a pure translation of it. Find the epipole in the image intersecting the line between camera centers (the translation) with the image plane. The 3D line would be represented by 4 parameters matching ax + by + cz + d = 0, and in this case, as it passes through the origin, we may assume that d = 0, moreover, we would know that z = 1 on the intersection plane, we could then have two constraints:  $\begin{cases} ax + by + c = 0\\ 1 + x^2 + y^2 = (ax)^2 + (by)^2 + c^2 \end{cases}$ Solving these will define the epipole in the image frame.
- 2. For the corresponding pts x, x', find the line  $l(\theta)$  that minimizes  $d(x, l)^2 + d(x', l)^2$

The line equation would match:  $tan\theta \cdot x - y + (e_y - e_x \cdot tan\theta)$ ; where  $\mathbf{e} = \begin{bmatrix} e_x \\ e_y \\ 1 \end{bmatrix}$  is the epipole

Hence: 
$$\boldsymbol{l} = (tan\theta, -1, (e_y - e_x \cdot tan\theta))^T = (a, b, c)^T$$

We could then compute the distance for  $\mathbf{x} = (x_0, y_0, 1) \Rightarrow d(\mathbf{x}, \mathbf{l}(\theta))^2 = \frac{(\mathbf{l}^T \mathbf{x})^2}{a^2 + b^2}$  (same for  $\mathbf{x}' = (x_1, y_1, 1)^T$ )

So, we would need to solve the minimization problem:  $\min_{\theta} \{d(x, l(\theta))^2 + d(x', l(\theta))^2\}$ 

A closed formula for this would be: (calculated using wolfram alpha for simplicity)

$$\theta \approx \tan^{-1} \left( \frac{x_0 (y_0 - e_y) - e_x (y_0 + y_1) + x_1 (y_1 - e_y) + 2e_x e_y}{x_0^2 - 2e_x (x_0 + x_1) + x_1^2 + 2e_x^2} \right) + \pi k; \ k \in \mathbb{Z}$$

3.  $\widehat{x}, \widehat{x}'$  could now be estimated, for  $\widehat{x} = (x, y)^T, x = (x_0, y_0)^T, l = (a, b, c)^T$ : (same for  $\widehat{x}'$ ;  $x' = (x_1, y_1)^T$ )

$$x = \frac{b(bx_0 - ay_0) - ac}{a^2 + b^2}, y = \frac{a(-bx_0 + ay_0) - bc}{a^2 + b^2}, z = 1$$

Shifting  $\widehat{x}'$  by the translation t, we could calculate the estimated world point  $\widehat{X}$  by building two 3D line equations, between the camera centers and  $\widehat{x}$ ,  $\widehat{x}' + t$  accordingly, and then calculate  $\widehat{X}$  by their intersection.