

Assignment 4

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1 Robust Homography Estimation and Stitching

E1:

记录:

Two cameras $P_1 = [A_1, t_1]$ $P_2 = [A_2, t_2]$ with same camera center. there is a homography H between two.

First, the expression for the camera center.

$$\begin{cases} P_1 \begin{bmatrix} C_1 \\ 1 \end{bmatrix} = A_1 C_1 + t_1 & C_1 = -A_1^{-1} t_1 \\ P_2 \begin{bmatrix} C_2 \\ 1 \end{bmatrix} = A_2 C_2 + t_2 & C_2 = -A_2^{-1} t_2 \end{cases}$$
$$C_1 = C_2 \Rightarrow -A_1^{-1} t_1 = -A_2^{-1} t_2$$
$$A_2 A_1^{-1} t_1 = A_2 A_2^{-1} t_2$$
$$A_2 A_1^{-1} t_1 = t_2$$

3D points $X_i = \begin{bmatrix} X_i \\ 1 \end{bmatrix}$ are projected with two cameras

$$P_1 \begin{bmatrix} X_i \\ 1 \end{bmatrix} = A_1 X_i + t_1 \quad P_2 \begin{bmatrix} X_i \\ 1 \end{bmatrix} = A_2 X_i + t_2$$
$$X_i = A_1^{-1} (P_1 X_i - A_1^{-1} t_1)$$
$$P_2 X_i = A_2 (A_1^{-1} P_1 X_i - A_1^{-1} t_1) + t_2$$

and $A_2 A_1^{-1} t_1 = t_2$

$$\Rightarrow P_2 X_i = \underline{A_2 A_1^{-1}} P_1 X_i$$

so $H = A_2 A_1^{-1}$

Figure 1

E2:

Degrees of freedom is 8.

Minimal number of point correspondences is 4.

Iterations of RANSAC is $(1 - 0.9^4)^n < 1 - 0.98 \quad n \geq 3.66 \quad n = 4$

CE1:

The number of SIFT features in a is 947.

The number of SIFT features in b is 865.

The number of matches is 204.

The number of inliers is 149.

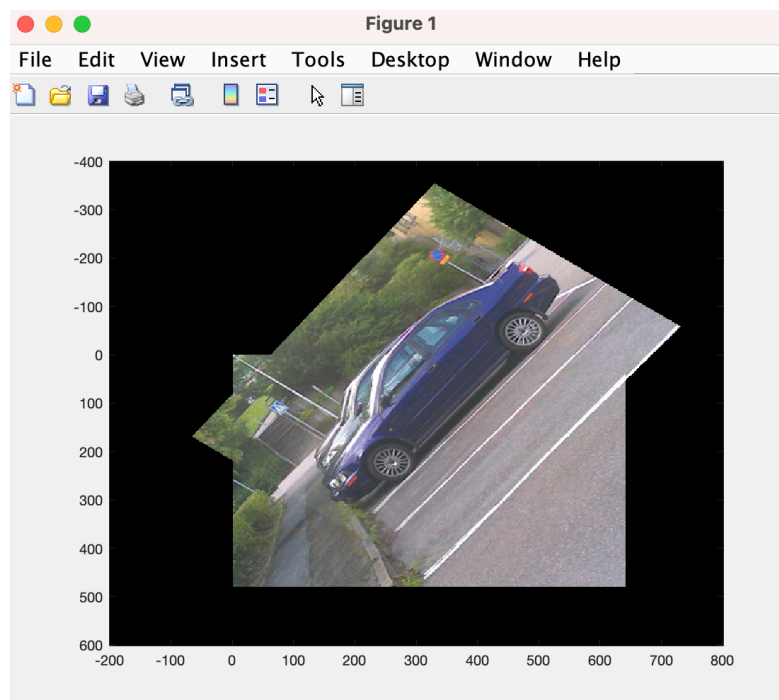


Figure 2

2 Robust Essential Matrix Estimation

E3:

Degrees of freedom is 5.

Minimal number of point correspondences is 5.

Iterations of RANSAC is $(1 - 0.9^5)^n < 1 - 0.98 \quad n \geq 4.38 \quad n = 5$

CE2:

The number of inliers = 1465

 RMS

0.4119

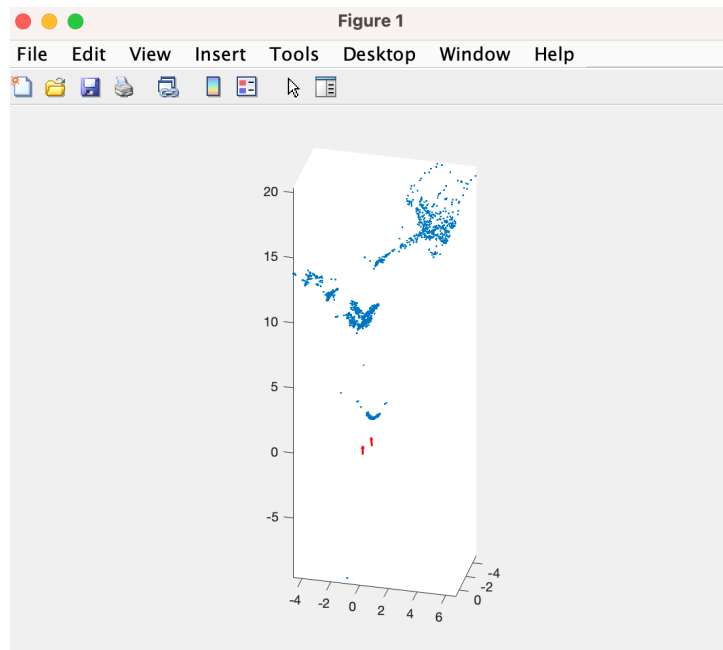


Figure 3

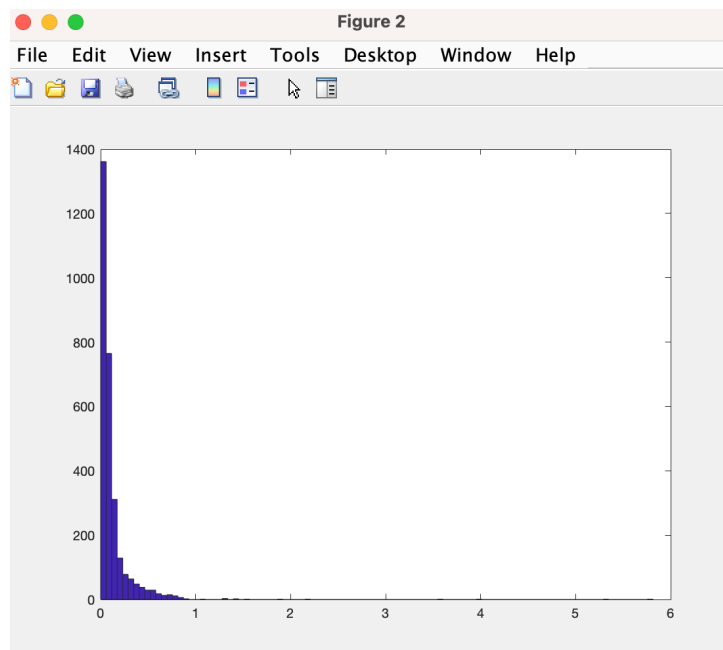


Figure 4

3 Calibrated Structure from Motion and Local Optimization

CE3:

The final RMS error is 0.3939.

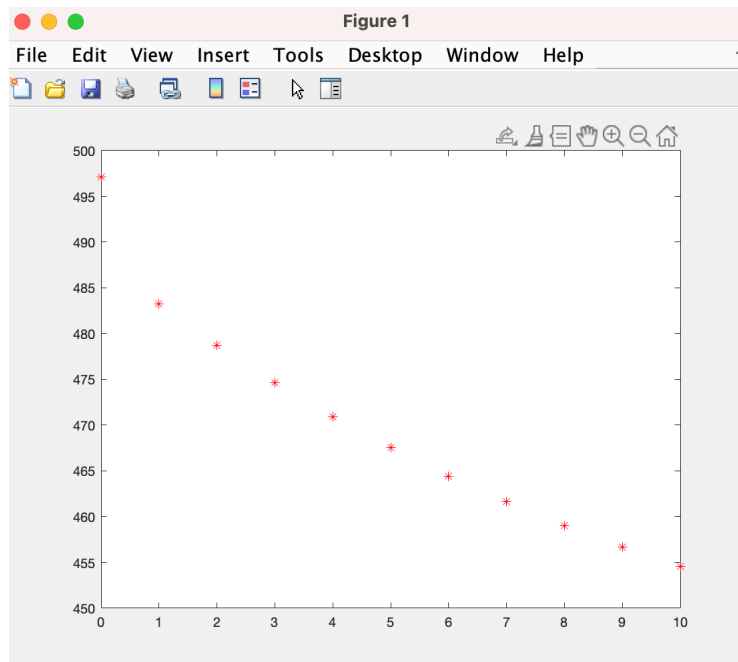


Figure 5

CE4:

The final RMS error is 0.2461.

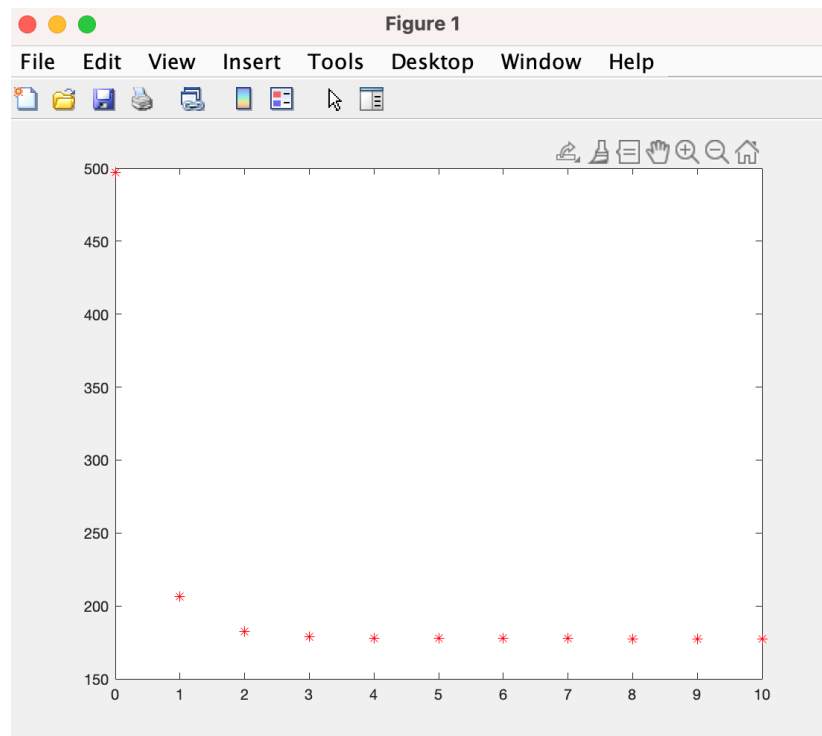


Figure 6

4 Dynamic Objects and Factorization

E4:

The matrix X can be rewritten as $X = C_k B + t$, where $C = [c_1 \ c_2 \ \dots \ c_m]$ is a matrix with the vectors c_{ik} as columns, $B = [B_1 \ B_2 \ \dots \ B_n]$ is a matrix with the vectors B_j as rows, and t is a vector of all ones. Since the row vectors B_1, \dots, B_n and 1 are linearly independent, the rank of X is $n+1$.

To minimize $\|A - X\|_F^2$, we can set the gradient of the objective function with respect to X to zero:

$$\nabla_X (\|A - X\|_F^2) = -2(A - X) = 0$$

This yields $X = A$. Substituting X in the equation for X , we get:

$$A = C_k B + t$$

Multiplying both sides by 1^T , we get:

$$tA = tC_k B + t t$$

Since $tB = [\text{sum}(B_1) \ \text{sum}(B_2) \ \dots \ \text{sum}(B_n)] = n\bar{B}$, where \bar{B} is the mean row vector of B , and $(1^T t) t = n$, we have:

$$tA = nC_k \bar{B} + n$$

Dividing both sides by n , we get:

$$\bar{A} = C_k \bar{B} + 1,$$

where \bar{A} is the row mean vector of A . Solving for t_i , we get:

$$t_i = \bar{A}_i - c_{ik} \bar{B}_k,$$

where \bar{A}_i is the row mean vector of $[a_{i1} \ a_{i2} \ \dots \ a_{in}]$ and \bar{B}_k is the mean row vector of B_k .

To find B_k and c_{ik} when t_i has been eliminated:

Compute the mean row vector \bar{B} of B .

Subtract \bar{B} from each row vector B_j to get the centered row vectors $\tilde{B}_j = B_j - \bar{B}$.

Compute the centered matrix $\tilde{C} = [\tilde{c}_1 \ \tilde{c}_2 \ \dots \ \tilde{c}_m]$, where $\tilde{c}_{ik} = c_{ik} - t_i \bar{B}_k$.

Solve for the centered matrix \tilde{C} by minimizing the objective function $\|\tilde{A} - \tilde{C} \tilde{B}\|_F^2$, where \tilde{A} is the centered matrix of A , and $\tilde{B} = [\tilde{B}_1 \ \tilde{B}_2 \ \dots \ \tilde{B}_n]$ is the matrix with the centered row vectors \tilde{B}_j as rows.

Once \tilde{C} and \tilde{B} have been found, we can compute B_k as $B_k = \tilde{B}_k + \bar{B}$, and c_{ik} as $c_{ik} = \tilde{c}_{ik} + t_i$.

Note that this method assumes that the noise in the measurements a_{ij} is uncorrelated and has zero mean.