

CV Project #2

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Introduction

In this project, we are asked to calculate the fundamental matrix given all the camera information as well as calculate it using 8-points method. The fundamental matrix is a 1 to n projection between image points on the two images. And for the second part we are asked to rectify the two cameras. After the rectification, the image points on the two images corresponding to the same 3D point will locate on the same image column of the rectified images. For the third part, we need to reconstruct 3D facial points from two images and all the camera information is known.

Calculating fundamental matrix F

There are two methods to calculate fundamental matrix F. The first method is calculating F by definition $W_l^{-T} S^T R W_r^{-1}$. Using the same method from project 1, we can calibrate the two cameras and we can get W_l, W_r, R, t . The second method is the 8-points method.

From the epipolar constraint that P_l, P_r and the baseline must be coplanar, we can derive

$$(t \times P_l) R P_r = 0 \quad (1)$$

And it can be written as

$$P_l^T S^T R P_r = 0 \quad (2)$$

where

$$S = \begin{bmatrix} 0 & -t_z & t_y \\ t_z & 0 & -t_x \\ -t_y & t_x & 0 \end{bmatrix} \quad (3)$$

and

$$P_l = \lambda_l W_l^{-1} U_l, P_r = \lambda_r W_r^{-1} U_r \quad (4)$$

Then Eq.2 can be written as

$$U_l^T W_l^{-T} S^T R W_r^{-1} U_r = 0 \quad (5)$$

and $W_l^{-T} S^T R W_r^{-1}$ is called fundamental matrix F .

From project 1 we know that the projection matrix P can be reconstructed from known 3D, 2D points, and we can also exclude outliers by RANSAC method. Then the intrinsic matrix W , rotation matrix R , translation vector t can be reconstructed from P . After knowing the intrinsic and extrinsic matrix, we can calculate the fundamental matrix by definition

$$F = W_l^{-T} S^T R W_r^{-1} \quad (6)$$

The result is following.
For the left camera

$$W_l = \begin{bmatrix} 1632 & 0 & 1056 \\ 0 & 1600 & 753.1 \\ 0 & 0 & 1 \end{bmatrix}, R_l = \begin{bmatrix} -0.5105 & 0.8598 & 0.0115 \\ 0.1345 & 0.0606 & 0.9891 \\ 0.8497 & 0.5065 & -0.1466 \end{bmatrix}$$

$$t_l = [9.8585 \quad -28.9584 \quad 122.3561]^T$$

For the right camera

$$W_r = \begin{bmatrix} 1620 & 0 & 1015 \\ 0 & 1602 & 726.9 \\ 0 & 0 & 1 \end{bmatrix}, R_r = \begin{bmatrix} -0.8567 & 0.5157 & 0.0132 \\ 0.1255 & 0.1839 & 0.9749 \\ 0.5003 & 0.8368 & -0.2223 \end{bmatrix}$$

$$t_r = [20.2861 \quad -17.0725 \quad 119.3243]^T$$

We can find that the intrinsic matrix for the two cameras are not the same. This is because the focal length f , sampling rate s_x, s_y , image origin c, r and distortion parameters are not the same for the two cameras.

In order to calculate F , we need to calculate relative rotation and translation between the two cameras. And in this report all the relative rotations and translations are the rotation and translation of the right camera relative to the left camera, denote as R_{lr} and t_{lr} .

$$P_{Cl} = R_l P_W + t_l, P_{Cr} = R_r P_W + t_r \quad (7)$$

And due to the coordinates in world frame are the same for the left and right camera, we can derive

$$P_{Cl} = R_l R_r^{-1} P_{Cr} + t_l - R_l R_r^{-1} t_r \quad (8)$$

$$R_{lr} = R_l R_r^{-1} \quad (9)$$

$$t_{lr} = t_l - R_{lr} t_r \quad (10)$$

Then from Eq.6 we can calculate the fundamental matrix F

$$F = \begin{bmatrix} -4.6381e-07 & 1.0462e-5 & -0.0097 \\ 1.6865e-06 & -2.4980e-06 & -0.0417 \\ -0.0026 & 0.0289 & 10.2901 \end{bmatrix}$$

We can also use the 8-points method to solve the fundamental matrix. The basic idea is that we can solve

$$U_l^T F U_r = 0 \quad (11)$$

linearly using at least 8 points from each image. Write fundamental matrix F as $F = [F_1 \quad F_2 \quad F_3]$, where F_1, F_2, F_3 are the first, second and third columns of F . Then we can rewrite Eq.11 as follow

$$[U_{li}^T F_1 \quad U_{li}^T F_2 \quad U_{li}^T F_3] U_{ri} = 0 \quad (12)$$

And furtherly we can rewrite Eq.12 as

$$\begin{bmatrix} U_{li}^T c_{ri} & U_{li}^T r_{ri} & U_{li}^T \end{bmatrix} \begin{bmatrix} F_1 \\ F_2 \\ F_3 \end{bmatrix} = 0 \quad (13)$$

If we have N points we can construct a system of linear equations

$$\begin{bmatrix} U_{l1}^T c_{r1} & U_{l1}^T r_{r1} & U_{l1}^T \\ \vdots & \vdots & \vdots \\ U_{lN}^T c_{rN} & U_{lN}^T r_{rN} & U_{lN}^T \end{bmatrix} \begin{bmatrix} F_1 \\ F_2 \\ F_3 \end{bmatrix} = 0 \quad (14)$$

And with at least 8 points we can solve F up to a scale factor. And there is no constraint on the scale factor[1]. However the fundamental matrix we calculated denotes as F_{cal} may not have the property of F that the rank of F is 2. Then we need to add the constraint by applying SVD to F_{cal} and force the third singular value to be 0 and reform the fundamental matrix denote as F^* . [2] And the result is as follow

$$F^* = \begin{bmatrix} -1.1545e-08 & 5.9903e-07 & -9.8898e-04 \\ 2.3489e-07 & -1.0073e-07 & -0.0027 \\ -3.4872e-04 & 0.0018 & 1 \end{bmatrix}$$

The fundamental matrix we get here is very different from the one we get by definition. The first reason is that F^* is up to a scale factor. the second reason is that F_{cal} does not meet the property of a fundamental matrix so we apply SVD to it and force the third singular value to be 0. And during this process there are errors intriduced. In order to see how large the error is, we define the error in the following way

$$E = U_l^T F U_r \quad (15)$$

And we can visualize it as Figure 1

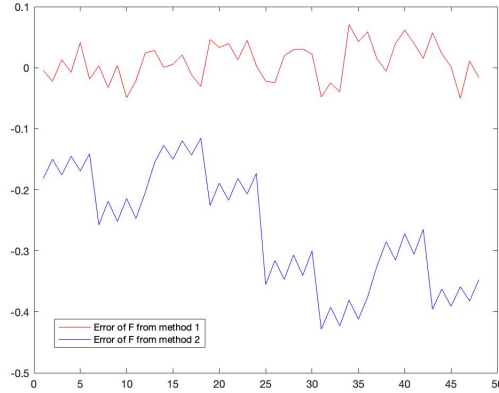


Figure 1: Errors of fundamental matrix

From Figure 1 we can see that the error of F^* is much bigger than F .

Rectification

In practice, the image plane of two cameras are often chosen to be coplanar and parallel to their base line. In this case the epipolar lines are parallel and conjugate epipolar lines are collinear. However it is hard to make the two image planes coplanar and parallel to the base line physically since the origins of the two camera are inside and the image plane is a virtual plane. The method we use to make the two image plane coplanar is called rectify, and it is done digitally.[1]

We can assume a rotation $R_{rec,l}$ of the left camera, and after the rotation, the new X axis and the relative translation t_{lr} are colinear. And for the right camera, we also apply a rotation $R_{rec,r}$ on it, and the new X axis of the right camera is colinear with t_{lr} . And we can write the following equation

$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = R_{rec,l} \frac{t_{lr}}{\|t_{lr}\|_2} \quad (16)$$

which means

$$R_1 = \frac{t_{lr}^T}{\|t_{lr}\|_2} \quad (17)$$

where R_1 is the first column of $R_{rec,l}$, then we can construct $R_{rec,l}$ that meets the properties of orthogonal matrix.

$$R_{rec,l} = \begin{bmatrix} t_x^* & t_y^* & t_z^* \\ t_y^*/\lambda & -t_x^*/\lambda & 0 \\ R_1 \times R_2 \end{bmatrix} \quad (18)$$

where $\begin{bmatrix} t_x^* & t_y^* & t_z^* \end{bmatrix} = \frac{t_{lr}^T}{\|t_{lr}\|_2}$, $\lambda = \sqrt{(t_x^*)^2 + (t_y^*)^2}$ and R_1, R_2 are the first and second column of $R_{rec,l}$. And for the right camera, since it has the same orientation with the left camera after rectifying, so the rectifying matrix for the right camera can be easily got as

$$R_{rec,r} = R_{rec,l} R_{lr} \quad (19)$$

And for the image points before and after rectifying, we have

$$\lambda_l^* \begin{bmatrix} c_l^* \\ r_l^* \\ 1 \end{bmatrix} = W_l \begin{bmatrix} X_{cl}^* \\ Y_{cl}^* \\ Z_{cl}^* \end{bmatrix} \quad (20)$$

$$\lambda_l \begin{bmatrix} c_l \\ r_l \\ 1 \end{bmatrix} = W_l \begin{bmatrix} X_{cl} \\ Y_{cl} \\ Z_{cl} \end{bmatrix} \quad (21)$$

$$\begin{bmatrix} X_{cl}^* \\ Y_{cl}^* \\ Z_{cl}^* \end{bmatrix} = R_{rec,l} \begin{bmatrix} X_{cl} \\ Y_{cl} \\ Z_{cl} \end{bmatrix} \quad (22)$$

Combine Eq.20, Eq.21 and Eq.22, we have

$$k_l \begin{bmatrix} c_l^* \\ r_l^* \\ 1 \end{bmatrix} = W R_{rec,l} W_l^{-1} \begin{bmatrix} c_l \\ r_l \\ 1 \end{bmatrix} \quad (23)$$

also we can derive the relationship for the right camera

$$k_r \begin{bmatrix} c_r^* \\ r_r^* \\ 1 \end{bmatrix} = W R_{rec,r} W_r^{-1} \begin{bmatrix} c_r \\ r_r \\ 1 \end{bmatrix} \quad (24)$$

where $W = \frac{1}{2}(W_l + W_r)$, because we need to make sure the intrinsic parameters are the same for both camera after rectifying. Then with Eq.22 and Eq.23 we can map the points before and after rectifying.

To generate the rectified image, we can use back forward mapping. The main idea of this method is that for each pixel on the rectified image, we can find the corresponding pixel on the original image using Eq.22 or Eq.23 and if the corresponding pixel on the original image is beyond the image then the rectified pixel is a black pixel. And from project 1 we know how to calculate the epipolar lines, then we can draw the images before and after the rectification and the facial points' corresponding epipolar lines.

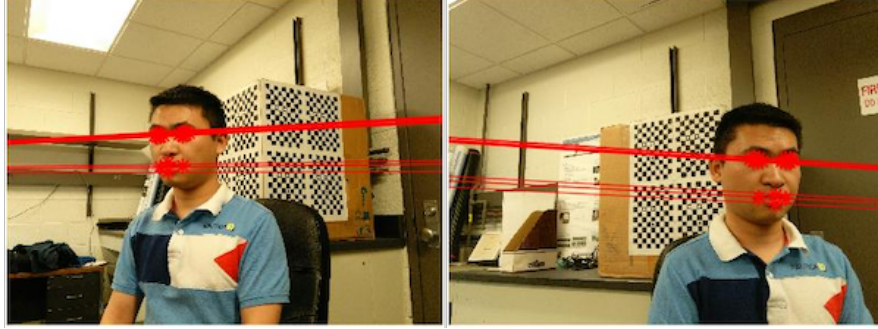


Figure 2: Left and right images before rectification

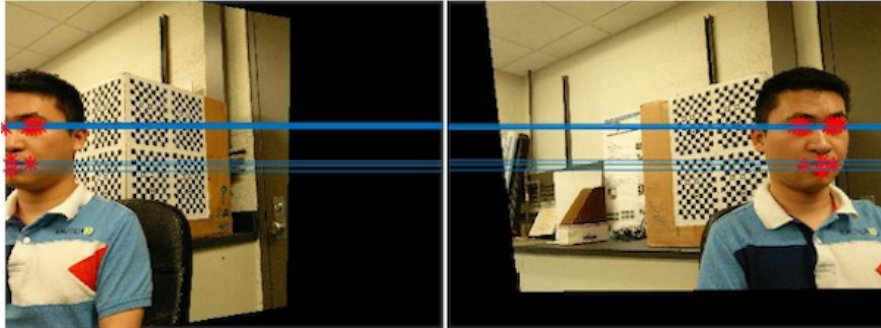


Figure 3: Left and right images after rectification

And from the Figure 2 and Figure 3, we can see that the result is pretty good, the epipolar lines before rectification are not parallel and the corresponding points between the two original images are not at the same image column. And after the rectification the epipolar lines are parallel and the corresponding points between two images are at the same image column. And after rectification, the point matching will be much easier.

Reconstruction

The major task for this part is to calculate the 3D coordinates of the facial points given their 2D coordinates in two images. We can use full reconstruction method since all the camera parameters are known. For each point we have

$$\lambda_l \begin{bmatrix} c_l \\ r_l \\ 1 \end{bmatrix} = \begin{bmatrix} P_{l1} & P_{l14} \\ P_{l2} & P_{l24} \\ P_{l3} & P_{l34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} \quad (25)$$

And we can rewrite Eq.25 as

$$\lambda_l c_l = P_1 \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + P_{14} \quad (26)$$

$$\lambda_l r_l = P_2 \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + P_{24} \quad (27)$$

$$\lambda_l = P_3 \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + P_{34} \quad (28)$$

Then substitute λ_l in Eq.26 and Eq.27 with Eq.28, we have

$$\begin{bmatrix} c_l P_{l3} - P_{l1} \\ r_l P_{l3} - P_{l2} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} P_{l14} - c_l P_{l34} \\ P_{l24} - r_l P_{l34} \end{bmatrix} \quad (29)$$

Also for the right image we have the similar equation, and combine all the equations we have

$$\begin{bmatrix} c_l P_{l3} - P_{l1} \\ r_l P_{l3} - P_{l2} \\ c_r P_{r3} - P_{r1} \\ r_r P_{r3} - P_{r2} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} P_{l14} - c_l P_{l34} \\ P_{l24} - r_l P_{l34} \\ P_{r14} - c_r P_{r34} \\ P_{r24} - r_r P_{r34} \end{bmatrix} \quad (30)$$

Then for each point we have 4 equations and 3 unknowns, we can use the least squares method to solve it. And the result is as follow.

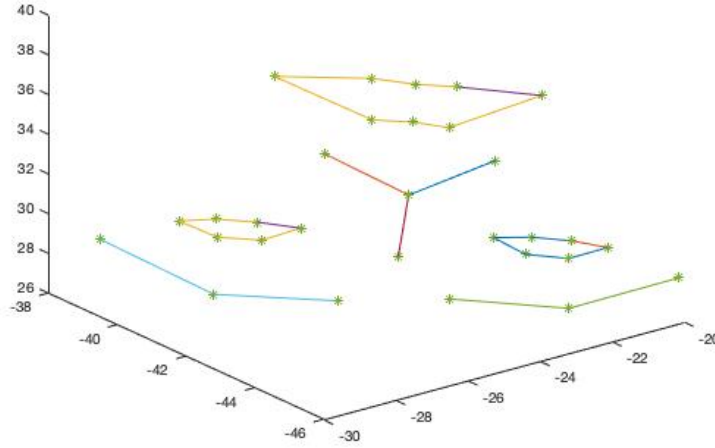


Figure 4: Reconstructed 3D facial points

And the width of the left eye, right eye and mouth are 2.4620, 2.3595, 5.3791.

Summary and conclusion

In this project we did camera calibration, fundamental matrix reconstruction from 8-points method, camera rectification and 3D full reconstruction.

In the first part, to reconstruct the fundamental matrix, we have two methods. If the points' 3D coordinates are known we can first calibrate the camera and then calculate the fundamental matrix by definition. And if the points' 3D coordinates are unknown, we can apply 8-points method to solve the fundamental matrix. However the result from 8-points method may not have the fundamental matrix's properties, so we need to apply the constraint additionally.

In the second part, we applied camera rectification to the two images to make the image plane coplanar and parallel to the base line. And after calculating the rectification matrix, we can use the back forward mapping to generate the rectified image. However the back forward mapping is very time consuming.

In the third part, we use the full reconstruction method to calculate the 3D facial points' coordinates.

References

- [1] Qiang Ji. RPI ECSE 6650 Computer Vision, Lecture Notes: 3D Reconstruction. URL:<https://www.ecse.rpi.edu/~qji/CV/3dreconstruction2.pdf>. Last visited on 2019/11/7.
- [2] Wikipedia contributors, "Eight-point algorithm" Wikipedia, the free encyclopedia, https://en.wikipedia.org/wiki/Eight-point_algorithm (*accessed November 7, 2019*).