

ECE 5460 Image Processing AU2018

Project #1

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Linear Camera Calibration Methodology

Given a set of 2D and 3D points, the intrinsic and extrinsic parameters of cameras can be linearly estimated. In the homogenous system, suppose we have 2D point $p = (u, v, 1)^T$ on an image and its corresponding 3D point $P = (X, Y, Z, 1)^T$ in 3D space:

$$\begin{bmatrix} u_i \\ v_i \\ w_i \end{bmatrix} = M \begin{bmatrix} X_i \\ Y_i \\ Z_i \\ 1 \end{bmatrix}$$

where M is a 3x4 projection matrix with

$$x = \frac{u_i}{w_i} = \frac{m_{11}X_i + m_{12}Y_i + m_{13}Z_i + m_{14}}{m_{31}X_i + m_{32}Y_i + m_{33}Z_i + m_{34}} \quad y = \frac{v_i}{w_i} = \frac{m_{21}X_i + m_{22}Y_i + m_{23}Z_i + m_{24}}{m_{31}X_i + m_{32}Y_i + m_{33}Z_i + m_{34}}$$

Given N pairs of 3D and 2D coordinates, we have

$$\mathbf{A}\mathbf{m} = \mathbf{0}$$

where

$$\mathbf{A} = \begin{bmatrix} X_1 & Y_1 & Z_1 & 1 & 0 & 0 & 0 & 0 & -x_1X_1 & -x_1Y_1 & -x_1Z_1 & -x_1 \\ 0 & 0 & 0 & 0 & X_1 & Y_1 & Z_1 & 1 & -y_1X_1 & -y_1Y_1 & -y_1Z_1 & -y_1 \\ \vdots & \vdots & & & \vdots & & \vdots & & & \vdots & \vdots & \\ X_N & Y_N & Z_N & 1 & 0 & 0 & 0 & 0 & -x_NX_N & -x_NY_N & -x_NZ_N & -x_N \\ 0 & 0 & 0 & 0 & X_N & Y_N & Z_N & 1 & -y_NX_N & -y_NY_N & -y_NZ_N & -y_N \end{bmatrix}$$

and

$$\mathbf{m} = [m_{11} \quad m_{12} \quad \cdots \quad m_{33} \quad m_{34}]$$

Since \mathbf{A} has rank 11, the vector \mathbf{m} can be recovered from SVD related techniques as the column of \mathbf{V} corresponding to the zero (in practice the smallest) singular value of \mathbf{A} . This algorithm can be illustrated in details in the Matlab code.

As the multiplication of intrinsic camera matrix and extrinsic camera matrix, after divided by a scale factor $\sqrt{m_{31}^2 + m_{32}^2 + m_{33}^2}$ the expression of M can be written as

$$M = \begin{bmatrix} s_x f r_1 + c_0 r_3 & s_x f r_1 + c_0 r_3 \\ s_x f r_1 + c_0 r_3 & s_x f r_1 + c_0 r_3 \\ s_x f r_1 + c_0 r_3 & s_x f r_1 + c_0 r_3 \end{bmatrix}$$

Now let

$$q_1 = [m_{11} \quad m_{12} \quad m_{13}]^T \quad q_2 = [m_{21} \quad m_{22} \quad m_{23}]^T \quad q_3 = [m_{31} \quad m_{32} \quad m_{33}]^T$$

Then we have

$$\begin{aligned}
 r_3 &= q_3^T & t_z &= m_{34} & c_0 &= q_1^T q_3 & r_0 &= q_2^T q_3 \\
 s_x f &= \pm \sqrt{q_1^T q_1 - c_0^2} & s_y f &= \pm \sqrt{q_2^T q_2 - r_0^2} \\
 t_x &= (m_{14} - c_0 t_z) / (s_x f) & t_y &= (m_{24} - r_0 t_z) / (s_y f) \\
 r_1 &= (q_1^T - c_0 r_3) / (s_x f) & r_2 &= (q_2^T - r_0 r_3) / (s_y f)
 \end{aligned}$$

The above equations solve all the camera parameters.

Experimental Results

We randomly chose several points in the Fig. 1 and Fig. 2 as 2D references. The visual results are as follows

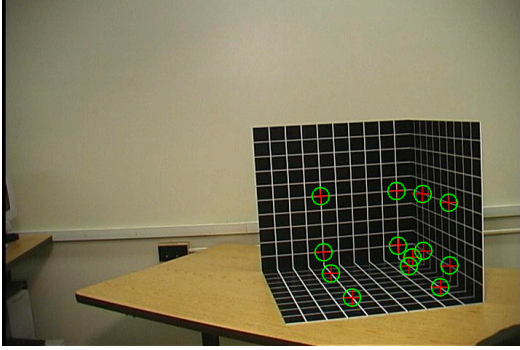


Figure 1 Reference points

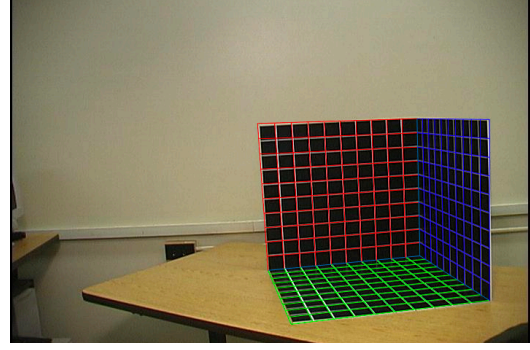


Figure 2 Calibration grid

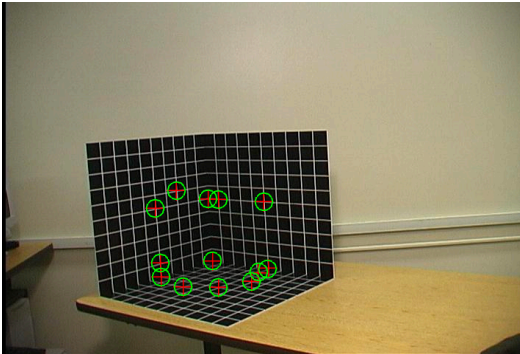


Figure 3 Reference points

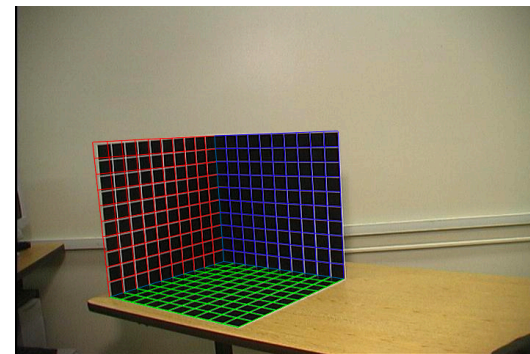


Figure 4 Calibration grid

Reference

- [1] <https://www.mathworks.com/matlabcentral/fileexchange/67132-camera-calibration-and-calculating-intrinsic-and-extrinsic-parameters>
- [2] <https://github.com/cdbunker/linearCameraCalibration>