

# Exercise 4: Camera calibration

02504 Computer vision

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February 22, 2023

These exercises will take you through:

**Direct linear transform (DLT)**, linear algorithm for camera calibration and **checkerboard calibration**, and bundle adjustment from Zhang (2000).

You should be able to perform camera calibration using both methods.

## Mathematical exercises: Direct linear transform (DLT)

In this section consider the 3D points

$$\mathbf{Q}_{ijk} = \begin{bmatrix} i \\ j \\ k \end{bmatrix}, \quad (1)$$

where  $i \in \{0, 1\}$ ,  $j \in \{0, 1\}$ , and  $k \in \{0, 1\}$ . Consider also a camera with  $f = 1000$  and a resolution of  $1920 \times 1080$ . Furthermore, the camera is transformed such that

$$\mathbf{R} = \begin{bmatrix} \sqrt{1/2} & -\sqrt{1/2} & 0 \\ \sqrt{1/2} & \sqrt{1/2} & 0 \\ 0 & 0 & 1 \end{bmatrix} \text{ and } \mathbf{t} = \begin{bmatrix} 0 \\ 0 \\ 10 \end{bmatrix}. \quad (2)$$

### Exercise 4.1

Find the projection matrix  $\mathcal{P}$  and the projections  $\mathbf{q}$ .

### Exercise 4.2

Write a function `pest` that uses  $\mathbf{Q}$  and  $\mathbf{q}$  to estimate  $\mathcal{P}$  with the DLT. Do not normalize your points to begin with.

Use the estimated projection matrix  $\mathbf{P}_{est}$  to project the points  $\mathbf{Q}$ , giving you the reprojected points  $\mathbf{q}_{est}$ .

What is the overall reprojection error  $\sqrt{\sum \frac{1}{n} \|\mathbf{q}_{est} - \mathbf{q}\|_2^2}$  (RMSE)?

Does normalizing your points change the results?

## Programming exercises: Checkerboard calibration

Here we will perform camera calibration with checkerboards. We do not yet have the ability to detect checkerboards, so for now we will define the points ourselves.

### Exercise 4.3

Define a function `checkerboard_points(n, m)` that returns the 3D points

$$\mathbf{Q}_{ij} = \begin{bmatrix} i - \frac{n-1}{2} \\ j - \frac{m-1}{2} \\ 0 \end{bmatrix}, \quad (7)$$

where  $i \in \{0, \dots, n-1\}$  and  $j \in \{0, \dots, m-1\}$ . The points should be returned as a  $3 \times (n \cdot m)$  matrix and their order does not matter. These points lie in the  $z = 0$  plane by definition.

### Exercise 4.4

Let  $\mathbf{Q}_\Omega$  define a set of corners on a checkerboard. Then define three sets of checkerboard points  $\mathbf{Q}_a$ ,  $\mathbf{Q}_b$ , and  $\mathbf{Q}_c$ , where

$$\mathbf{Q}_a = \mathcal{R}\left(\frac{\pi}{10}, 0, 0\right) \mathbf{Q}_\Omega, \quad (8)$$

$$\mathbf{Q}_b = \mathcal{R}(0, 0, 0) \mathbf{Q}_\Omega, \quad (9)$$

$$\mathbf{Q}_c = \mathcal{R}\left(-\frac{\pi}{10}, 0, 0\right) \mathbf{Q}_\Omega, \quad (10)$$

$$(11)$$

where

$$\mathcal{R}(\theta_x, \theta_y, \theta_z) = \begin{bmatrix} \cos(\theta_z) & -\sin(\theta_z) & 0 \\ \sin(\theta_z) & \cos(\theta_z) & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos(\theta_y) & 0 & \sin(\theta_y) \\ 0 & 1 & 0 \\ -\sin(\theta_y) & 0 & \cos(\theta_y) \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos(\theta_x) & -\sin(\theta_x) \\ 0 & \sin(\theta_x) & \cos(\theta_x) \end{bmatrix}. \quad (12)$$

Recall that you can compute  $\mathcal{R}$  with `scipy` as follows:

```
from scipy.spatial.transform import Rotation
R = Rotation.from_euler('xyz', [\theta_x, \theta_y, \theta_z]).as_matrix()
```

The points should look like [Figure 1](#).

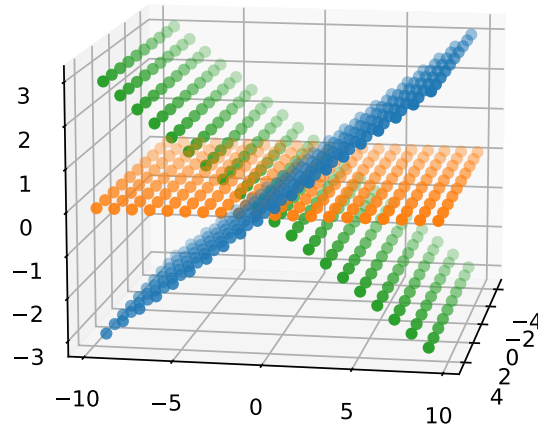


Figure 1: A 3D plot of  $\mathbf{Q}_a$ ,  $\mathbf{Q}_b$ , and  $\mathbf{Q}_c$ . In this case  $n = 10$ ,  $m = 20$ .

Using the projection matrix from [Exercise 4.1](#), project all the checkerboard points to the image plane, obtaining:  $\mathbf{q}_a$ ,  $\mathbf{q}_b$ , and  $\mathbf{q}_c$ .

We will now go through the method outlined in Zhang's method<sup>1</sup> step by step.

## Exercise 4.5

Define a function `estimateHomographies(Q_omega, qs)` which takes the following input arguments:

- **Q\_omega**: an array original un-transformed checkerboard points in 3D, for example  $\mathbf{Q}_\Omega$ .
- **qs**: a list of arrays, each element in the list containing  $\mathbf{Q}_\Omega$  projected to the image plane from different views, for example **qs** could be  $[\mathbf{q}_a, \mathbf{q}_b, \mathbf{q}_c]$ .

The function should return the homographies that map from **Q\_omega** to each of the entries in **qs**. The homographies should work as follows:

$$\mathbf{q} = H\tilde{\mathbf{Q}}_\Omega, \quad (13)$$

where  $\tilde{\mathbf{Q}}_\Omega$  is  $\mathbf{Q}_\Omega$  without the  $z$ -coordinate, in homogeneous coordinates. Remember that we need multiple orientations of checkerboards e.g. rotated and translated.

Use your function `hest` from week 2 to estimate the individual homographies. You should return a list of homographies; one homography for each checkerboard orientation.

Test your function using  $\mathbf{Q}_\Omega$ ,  $\mathbf{q}_a$ ,  $\mathbf{q}_b$ , and  $\mathbf{q}_c$ . Check that the estimated homographies are correct with [Equation 13](#).

## Exercise 4.6

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<sup>1</sup>Zhang, Zhengyou. "A flexible new technique for camera calibration." IEEE Transactions on pattern analysis and machine intelligence 22.11 (2000): 1330-1334.

Now, define a function `estimate_b(Hs)` that takes a list of homographies `Hs` and returns the vector `b`. Form the matrix `V`. This is the coefficient matrix used to estimate `b` using SVD.

Test your function with the homographies from previous exercise. See if you get the same result as by constructing  $\mathbf{B}_{\text{true}} = \mathbf{K}^{-\text{T}} \mathbf{K}^{-1}$ , and converting this into `btrue`.

Is `b` a scaled version of `btrue`?

Suggestions for debugging:

- Check that  $\mathbf{v}_{11} \cdot \mathbf{b}_{\text{true}} = \mathbf{h}_1^{\text{T}} \mathbf{B}_{\text{true}} \mathbf{h}_1$
- Be aware that  $\mathbf{v}_{\alpha\beta}$  use 1-indexing, while your code might not.

## Exercise 4.7

Next, define a function `estimateIntrinsics(Hs)` that takes a list of homographies `Hs` and returns a camera matrix `K`. Use your `estimate_b` from the previous exercise. From `b`, estimate the camera matrix `K` (they use `A` in the paper). Find the solution in Appendix B from the paper.

Test your function with the homographies from [Exercise 4.5](#). Do you get the original camera matrix?

## Exercise 4.8

Now, define a function `Rs, ts = estimateExtrinsics(K, Hs)` that takes the camera matrix `K` and the homographies `Hs` and returns the rotations `Rs` and translations `ts` of each checkerboard. Use the formulas given in the paper but you do not need to bother with Appendix C — we can live with the error.

What kind of rotations do you get, and are they valid?

Join the functions to make a larger function `K, Rs, ts = calibratecamera(qs, Q)` that finds the camera intrinsics and extrinsics from the checkerboard correspondences `q` and `Q`.

# Solutions

## Answer of exercise 4.1

The camera matrix is

$$\begin{bmatrix} 1000 & 0 & 960 \\ 0 & 1000 & 540 \\ 0 & 0 & 1 \end{bmatrix}. \quad (3)$$

The projection matrix is

$$\mathcal{P} = \begin{bmatrix} 707.11 & -707.11 & 960 & 9600 \\ 707.11 & 707.11 & 540 & 5400. \\ 0 & 0 & 1 & 10 \end{bmatrix}, \quad (4)$$

and the projections are

$$\mathbf{q}_{000} = \begin{bmatrix} 960 \\ 540 \end{bmatrix}, \quad \mathbf{q}_{001} = \begin{bmatrix} 960 \\ 540 \end{bmatrix}, \quad \mathbf{q}_{010} = \begin{bmatrix} 889.29 \\ 610.71 \end{bmatrix}, \quad \mathbf{q}_{011} = \begin{bmatrix} 895.72 \\ 604.28 \end{bmatrix}, \quad (5)$$

$$\mathbf{q}_{100} = \begin{bmatrix} 1030.71 \\ 610.71 \end{bmatrix}, \quad \mathbf{q}_{101} = \begin{bmatrix} 1024.28 \\ 604.28 \end{bmatrix}, \quad \mathbf{q}_{110} = \begin{bmatrix} 960 \\ 681.42 \end{bmatrix}, \quad \text{and } \mathbf{q}_{111} = \begin{bmatrix} 960. \\ 668.56 \end{bmatrix}. \quad (6)$$

## Answer of exercise 4.2

The projection matrix and the reprojections are identical to the original within machine precision. The reprojection error on my system is approximately  $10^{-11}$  without normalization and  $10^{-14}$  with normalization.