

CSE 344: Computer Vision  
Assignment 1

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Q1. Calibration toolbox from OpenCV used : [http://docs.opencv.org/3.0-beta/doc/tutorials/calib3d/camera\\_calibration/camera\\_calibration.html](http://docs.opencv.org/3.0-beta/doc/tutorials/calib3d/camera_calibration/camera_calibration.html)

a)

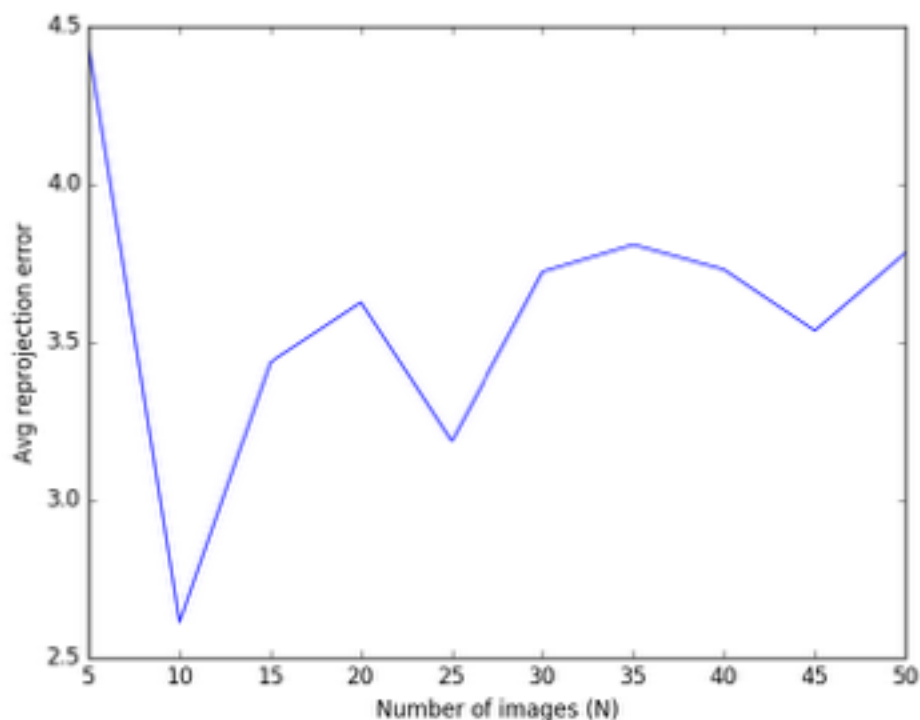
S1 : for 20 images with all DOFs covered

- Average re-projection error = 3.16271482714
- Estimated intrinsics:
  - $f_x = 3.8584530251284455e+03$
  - $f_y = 3.8584530251284455e+03$
  - $c_x = 1.1835000000000000e+03$
  - $c_y = 2.1035000000000000e+03$
- Estimated distortion coefficients: [-1.4186e-02, 1.8714e+00, 9.7445e-04, -4.6332e-03, -1.0061e+01]

S2: for first 20 images

- Average re-projection error for first 20 images: 4.58034775838
- Estimated intrinsics:
  - $f_x = 3.8311637596940427e+03$
  - $f_y = 3.8311637596940427e+03$
  - $c_x = 1.1835000000000000e+03$
  - $c_y = 2.1035000000000000e+03$
- Estimated distortion coefficients: [-8.5676e-03, 9.2993e-01, -1.8579e-02, 3.5245e-03, -1.6453e+00]

b) I used random sampling implemented in python's libraries to calibrate for different, randomly chosen sets of images with length  $N = 5, 10 \dots 50$ . The plot for Mean re-projection error vs  $N$  is shown below:



Q2.

Assumption: The rotation from car frame (V) to mount frame (M) is -30 degrees about the X axis. This is what I interpreted from the question and figure provided.

We have three transformations:

vTw: transformation from world frame (W) to car frame (V)

mTv: transformation from car frame (V) to mount frame (M)

cTm: transformation from mount frame (M) to camera frame (C)

These transformations are calculated one by one as explained in the code.

They are as follows:

vTw =

$$\begin{bmatrix} 0.8660254 & 0.5 & 0. & -1.19615242 \\ -0.5 & 0.8660254 & 0. & 9.92820323 \\ 0. & 0. & 1. & -1. \\ 0. & 0. & 0. & 1. \end{bmatrix}$$

mTv =

$$\begin{bmatrix} 1. & 0. & 0. & 0. \\ 0. & 0.8660254 & -0.5 & 2. \\ 0. & 0.5 & 0.8660254 & -3.46410162 \\ 0. & 0. & 0. & 1. \end{bmatrix}$$

cTm =

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 2 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

We want to transform a point from world frame P\_w to camera frame P\_c.

This is done by combining the three transformations above as follows:

- P\_w --> P\_v --> P\_m --> P\_c
- P\_c = cTm \* (P\_m)
- P\_c = cTm \* (mTv \* P\_v)
- P\_c = cTm \* (mTv \* (vTw \* P\_w))
- P\_c = cTm \* mTv \* vTw \* P\_w
- P\_c = cTw \* P\_w
- cTw = cTm \* mTv \* vTw

Transformation from world frame (W) to camera frame (C):

cTw =

$$\begin{bmatrix} 0.8660254 & 0.5 & 0. & -1.19615242 \\ -0.4330127 & 0.75 & -0.5 & 13.09807621 \\ -0.25 & 0.4330127 & 0.8660254 & 0.6339746 \\ 0. & 0. & 0. & 1. \end{bmatrix}$$

I used this matrix to transform all cube corners to the camera frame, and then projected these 3D points according to the perspective projection formula ( $f_x \cdot X/Z + c_x$ ,  $f_y \cdot Y/Z + c_y$ ) to get 2D pixel coordinates:

```
[[0 0 0 1]] --> [[-6096.44143844]] [[-6096.44143844]]
[[1 0 0 1]] --> [[-2133.85382338]] [[-2133.85382338]]
[[1 1 0 1]] --> [[ 1985.77308215]] [[ 1985.77308215]]
[[0 1 0 1]] --> [[-1333.93523668]] [[-1333.93523668]]
[[1 0 1 1]] --> [[ 164.47632413]] [[ 164.47632413]]
[[0 0 1 1]] --> [[-1893.36528927]] [[-1893.36528927]]
[[0 1 1 1]] --> [[-206.07773982]] [[-206.07773982]]
[[1 1 1 1]] --> [[ 1572.94858645]] [[ 1572.94858645]]
```

The normalised image coordinates were calculated with  $f = 1$  and standard procedure, to get these for cube corners:

```
[[0 0 0 1]] --> [[-1.88675135]] [[-1.88675135]]
[[1 0 0 1]] --> [[-0.85976266]] [[-0.85976266]]
[[1 1 0 1]] --> [[ 0.2079261]] [[ 0.2079261]]
[[0 1 0 1]] --> [[-0.65244678]] [[-0.65244678]]
[[1 0 1 1]] --> [[-0.26410162]] [[-0.26410162]]
[[0 0 1 1]] --> [[-0.79743495]] [[-0.79743495]]
[[0 1 1 1]] --> [[-0.36013857]] [[-0.36013857]]
[[1 1 1 1]] --> [[ 0.10093387]] [[ 0.10093387]]
```

Theory questions on further pages.

Q3 we wish to rotate a vector  $\vec{x}$  about an ~~axis~~  $\vec{u}$  by angle  $\theta$ .

$\vec{n}$  has 2 components, - parallel to  $\vec{u}$ , and perpendicular to  $\vec{u}$ .

$$\vec{n} = \underbrace{(\vec{u} \cdot \vec{n}) \vec{u}}_{\vec{n}_{||} \text{ (parallel)}} + \underbrace{\vec{n} - (\vec{u} \cdot \vec{n}) \vec{u}}_{\vec{n}_{\perp} \text{ (perpendicular)}}$$

The component  $\vec{n}_{||}$  does not change on rotation as it is about the rotation axis.

$\vec{n}_{\perp}$  rotates as follows:

$$|\vec{n}_{\perp}| = |\vec{n}_{\perp \text{ rot}}| \quad (\text{retain its magnitude})$$

$$\rightarrow \vec{n}_{\perp \text{ rot}} = \cos \theta \vec{n}_{\perp} + \sin \theta (\vec{u} \times \vec{n}_{\perp})$$

These represent 2 perpendicular axes in the plane orthogonal to  $\vec{u}$ .

The above comes from simple planar rotation.

$$\begin{aligned} \vec{n}_{\perp \text{ rot}} &= \cos \theta \vec{n}_{\perp} + \sin \theta (\vec{u} \times (\vec{n} - \vec{n}_{||})) \\ &= \cos \theta \vec{n}_{\perp} + \sin \theta (\vec{u} \times \vec{n}) - \sin \theta (\vec{u} \times \vec{n}_{||}) \\ &\quad \circ \text{ as } \vec{u} \parallel \vec{n}_{||} \end{aligned}$$

$$\text{we } \vec{x}_{\perp \text{rot}} = \cos\theta \vec{n}_{\perp} + \sin\theta (\hat{u} \times \vec{n})$$

$$\text{Now, } \vec{n}_{\text{rot}} = \vec{x}_{\perp \text{rot}} + \vec{x}_{\parallel \text{rot}}$$

$$= \vec{x}_{\parallel} + \cos\theta \vec{n}_{\perp} + \sin\theta (\hat{u} \times \vec{n})$$

$$= \vec{n}_{\parallel} + (\vec{n} - \vec{x}_{\parallel}) \cos\theta + \sin\theta (\hat{u} \times \vec{n})$$

$$= \cancel{\vec{x}_{\parallel} \cos\theta} + \dots$$

$$= \cos\theta \vec{n} + \sin\theta (\hat{u} \times \vec{n}) + (1 - \cos\theta) \vec{n}_{\parallel}$$

$$R\vec{n} = \cos\theta \vec{n} + \sin\theta (\hat{u} \times \vec{n}) + (1 - \cos\theta) (\hat{u} \cdot \vec{n}) \hat{u}$$

which is the desired result.

Q4.

$O$ : optical center in world frame.

$M$ : perspective projection matrix.

We know that  $M$  contains world to camera transformation.

~~The origin of the camera coordinates.~~

$$\Rightarrow M = K \cdot T$$

where  $T$  is 3D transform ( $W \rightarrow C$ )  
and  $K$  is intrinsic matrix.

$$\rightarrow M = K \times \begin{bmatrix} {}^C R_W & {}^C O_W \\ 0 & 1 \end{bmatrix}$$

$$\text{we have } O = \begin{pmatrix} {}^W O_C \\ 1 \end{pmatrix}$$

Now, if we calculate  $MO$ ,

$$MO = K \times \begin{bmatrix} {}^C R_W & {}^C O_W \\ 0 & 1 \end{bmatrix} \begin{bmatrix} {}^W O_C \\ 1 \end{bmatrix}$$

$$= K \left[ {}^C R_W {}^W O_C + {}^C O_W \right]$$

$$\left[ \because Tn = Rn + t \right]$$

-1.44



Now,  ${}^c R_w {}^w O_c$  is the rotated ~~vector~~ position vector of the camera origin, in the camera frame.

This will give us the reverse of ~~the~~ world origin in ~~the~~ camera frame  ${}^c O_w$ .

$${}^c R_w = -{}^c O_w$$

$$\neg MD = K (-{}^c O_w + {}^c O_w) \\ = 0.$$

$\therefore$  Hence proved.

This is intuitively correct as the optical center is the projection of the origin ~~in the~~ of the camera onto the image plane, which is by definition 0 vector.

