

CS534 Computer Vision Assignment-3

Camera Calibration and Augmented Reality

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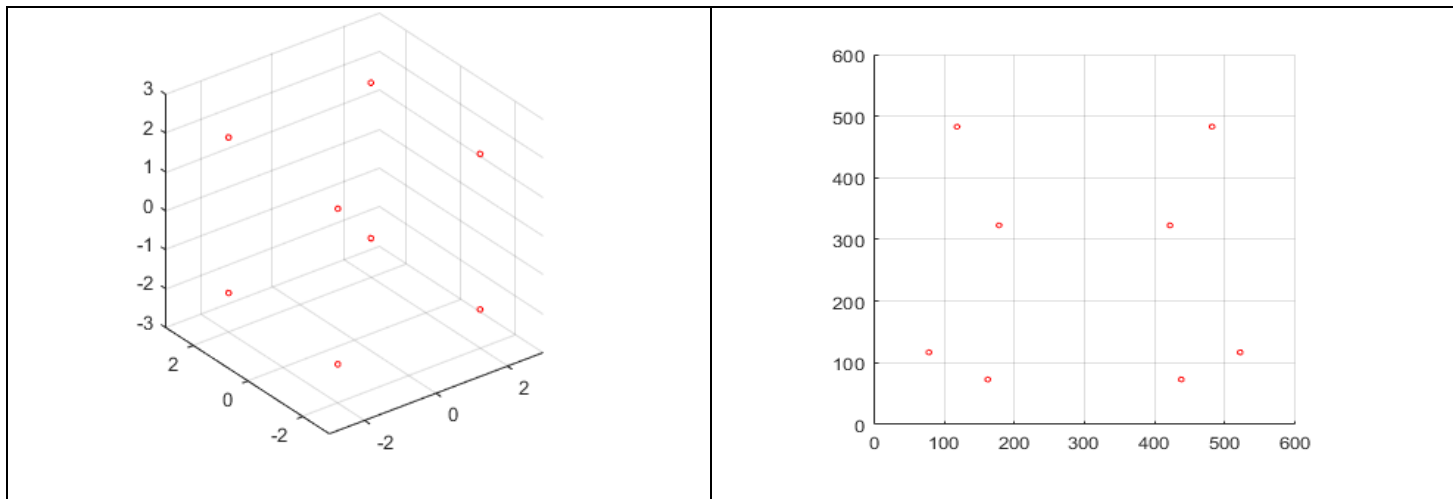
TASK 1: Camera Calibration using 3D calibration object

This involved camera calibration for a Robot vehicle.

1.1 Outputs

1.1.1 Drawing Image Points

Draw the image points, using small circles for each image point



1.1.2 Matrix P

Write a Matlab function that takes as argument the homogeneous coordinates of one cube corner and the homogeneous coordinates of its image, and returns 2 rows of the matrix P (slide 30 of the Camera Calibration pdf document). This matrix P will be used to compute the 12 elements of the projection matrix M such that $\lambda p_i = MP_i$

Code: *get_P_matrix.m*

1.1.3 Print Matrix P

Use this Matlab function to generate 2 rows of the matrix P for each cube corner and its image and obtain a matrix with 16 rows and 12 columns. Print matrix P.

Size(P)=16*12

```

par_i mix  % get_P_matrix.m  T
P_matrix = 2x12
    2     2     2     1     0     0     0     0    -844    -844    -844    -422
    0     0     0     0     2     2     2     1    -646    -646    -646    -323

P_matrix = 2x12
   -2     2     2     1     0     0     0     0     356    -356    -356    -178
    0     0     0     0    -2     2     2     1     646    -646    -646    -323

P_matrix = 2x12
   -2     2    -2     1     0     0     0     0     236    -236     236    -118
    0     0     0     0    -2     2    -2     1     966    -966     966    -483

P_matrix = 2x12
    2     2    -2     1     0     0     0     0    -964    -964     964    -482
    0     0     0     0     2     2    -2     1    -966    -966     966    -483

P_matrix = 2x12
    2    -2     2     1     0     0     0     0    -876     876    -876    -438
    0     0     0     0     2    -2     2     1    -146     146    -146    -73

P_matrix = 2x12
   -2    -2     2     1     0     0     0     0     324     324    -324    -162
    0     0     0     0    -2    -2     2     1     146     146    -146    -73

P_matrix = 2x12
   -2    -2    -2     1     0     0     0     0     156     156     156    -78
    0     0     0     0    -2    -2    -2     1     234     234     234    -117

P_matrix = 2x12
    2         -2         -2         1         0         0         0         0    -1044     1044     1044    -522
    0         0         0         0         2        -2        -2         1    -234      234      234    -117

```

1.1.4 Print Matrix M

Now we need to solve the system $Pm = 0$. Find the singular value decomposition of matrix P using matlab svd function. The last column vector of V obtained by svd(P) should be the 12 elements in row order of the projection matrix that transformed the cube corner coordinates into their images. Print the matrix M.

Q4: Print matrix M

```
M = get_M_matrix(P);
```

```
M_matrix = 3x4
```

```
0.1925    0.0283    0.0786    0.7346
0.0000    0.2044    0.0001    0.6120
0.0000    0.0001    0.0003    0.0024
```

```
disp(M)
```

```
0.1925    0.0283    0.0786    0.7346
0.0000    0.2044    0.0001    0.6120
0.0000    0.0001    0.0003    0.0024
```

1.1.5 Translational vector recovery

Now we need to recover the translation vector which is a null vector of M. Find the singular value decomposition of matrix $M = U\Sigma V^T$. The 4 elements of the last column of V are the homogeneous coordinates of the position of the camera center of projection in the frame of reference of the cube (as in slide 36). Print the corresponding 3 Euclidean coordinates of the camera center in the frame of reference of the cube.

Q5: Translational vector recovery

```
39 [m_U, m_Sigma, m_V_T] = svd(M);
40 t = m_V_T(:, end);
41 center = t(1: end - 1) / t(end)
```

```
center = 3x1
-0.0000
-2.9912
-8.2695
```

```
42 disp(center)
```

```
-0.0000
-2.9912
-8.2695
```

1.1.6 Print Matrix M'

Consider the 3x3 matrix M' composed of the first 3 columns of matrix M. Rescale the elements of this matrix so that its element m33 becomes equal to 1. Print matrix M'. Now let the rotation matrices be as defined in slide 38 where the axes e1, e2, e3 are the x, y, z axes respectively. Therefore M' can be written as $M_0 = KRT_zRT_yRT_x$

Q6: Print matrix M'

```
43 % Rescaling the elements of this matrix so that its element m33 becomes equal to 1
44 M_dash = M(:, 1:3) / M(3, 3) %3*4
```

```
M_dash = 3x3
    734.6289    107.8955    299.9999
         0         780.1442         0.2641
         0         0.3597         1.0000
```

1.1.7 Rotation Matrix Rx

```
45 [R_x, Theta_x, N_matrix] = get_Rx_Thetax_N(M_dash);
```

```
R_x = 3x3
    1.0000         0         0
         0     0.9410     0.3384
         0    -0.3384     0.9410
```

```
Theta_x = -19.7812
```

```
N_matrix = 3x3
    734.6289    -0.0000    318.8125
         0     734.0199    264.2723
         0         0         1.0627
```

1.1.8 Rotation angle Theta

```
[R_z, Theta_z] = get_Rz_Thetaz(N_matrix);
```

```
R_z = 3x3
    1.0000    0.0000         0
   -0.0000    1.0000         0
         0         0     1.0000
```

```
Theta_z = -7.2204e-05
```

1.1.9 Compute and Print Matrix K

```
K = M_dash / (R_x * R_y);
```

```
% Print K
```

```
K = K / K(3,3)
```

```
K = 3x3
```

```
896.6817 247.8504 299.9999
0.0011 896.1566 -321.9619
0.0000 0.8262 1.0000
```

```
focal_lengths = [K(1, 1), K(2, 2)]
```

```
focal_lengths = 1x2
```

```
896.6817 896.1566
```

```
image_center = [K(1, 3), K(2, 3)]
```

```
image_center = 1x2
```

```
299.9999 -321.9619
```

TASK 2: Camera Calibration using 2D calibration object

Executed separate code files for all the subfunctions

2.1 Outputs

2.1.1 Corner Extraction and Homography computation

images2.png, H is	images9.png, H is	images12.png, H is	images20.png, H is
Column 1	Column 1	Column 1	Column 1
0.9655	-1.0919	0.7037	-0.8559
0.0172	-0.1477	-0.1758	0.0113
-0.0000	-0.0005	-0.0005	0.0000
Column 2	Column 2	Column 2	Column 2
-0.0835	0.0345	-0.0492	0.2712
0.8809	-0.9309	0.8856	-0.3995
-0.0002	0.0001	-0.0002	0.0008
Column 3	Column 3	Column 3	Column 3
54.2980	-71.6419	75.9062	-122.6998
43.4384	-10.3829	59.9260	-57.9033
0.6033	-0.5191	0.6658	-0.6893

2.1.2 Computing Intrinsic and Extrinsic parameters

<p>B_matrix =</p> <pre> -0.0000 0.0000 0.0005 0.0000 -0.0000 0.0004 0.0005 0.0004 -1.0000 </pre>	<p>v_0 = 230.3654</p>
<p>lambda = -0.7692</p> <p>alpha = 723.3788</p> <p>beta = 702.7882</p> <p>gamma = 0.6580</p> <p>u_0 = 317.6932</p>	<p>A_matrix =</p> <pre> 723.3788 0.6580 317.6932 0 702.7882 230.3654 0 0 1.0000 </pre>
<p>image = images2.png</p> <p>R_matrix =</p> <pre> 0.9998 -0.0149 0.0027 0.0203 0.9864 0.1636 -0.0061 -0.1636 0.9865 </pre> <p>T_matrix =</p> <pre> -141.5014 -101.3666 449.8412 </pre> <p>R_T_R_matrix =</p> <pre> 1.0000 0.0061 -0.0000 0.0061 1.0000 0 -0.0000 0 1.0000 </pre>	<p>image = images9.png</p> <p>R_matrix =</p> <pre> -0.9241 -0.0119 -0.3818 -0.0271 -0.9941 0.1042 -0.3811 0.1078 0.9183 </pre> <p>T_matrix =</p> <pre> 93.1975 112.4254 -375.5904 </pre> <p>R_T_R_matrix =</p> <pre> 1.0000 -0.0031 0.0000 -0.0031 1.0000 -0.0000 0.0000 -0.0000 1.0000 </pre>

```
image = images12.png
```

```
R_matrix =
```

```
    0.9112    0.0035    0.4119
   -0.0544    0.9919    0.1141
   -0.4083   -0.1268    0.9041
```

```
T_matrix =
```

```
  -141.1402
  -100.1742
    501.5568
```

```
R_T_R_matrix =
```

```
    1.0000    0.0010    0.0000
    0.0010    1.0000    0.0000
    0.0000    0.0000    1.0000
```

```
image = images20.png
```

```
R_matrix =
```

```
   -0.9999    0.0071    0.0150
    0.0098   -0.7097    0.7044
    0.0114    0.7044    0.7096
```

```
T_matrix =
```

```
    111.9242
    120.8244
   -580.1506
```

```
R_T_R_matrix =
```

```
    1.0000   -0.0060    0.0000
   -0.0060    1.0000    0.0000
    0.0000    0.0000    1.0000
```

2.1.3 Projected Grid corners, improving accuracy

I) Projected Grid Corners

Figure 1: Projected grid corners



Figure 2: Projected grid corners

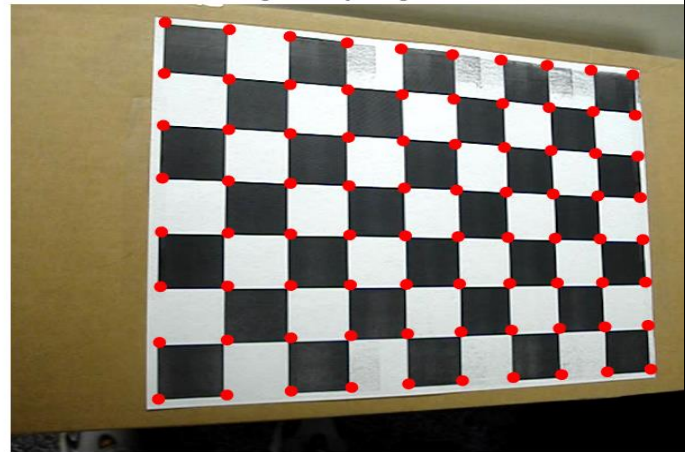


Figure 3: Projected grid corners



Figure 4: Projected grid corners



2) Harris corners

Figure 1: Harris corners



Figure 3: Harris corners

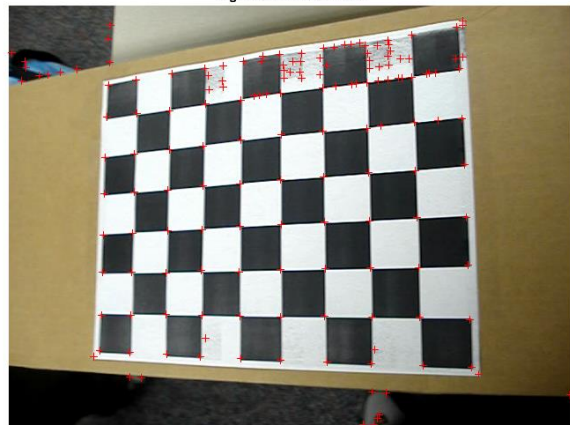


Figure 2: Harris corners

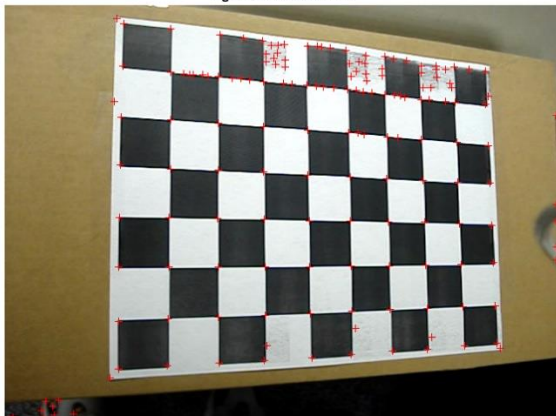


Figure 4: Harris corners



3) Grid Points:

Figure1: grid points



Figure2: grid points

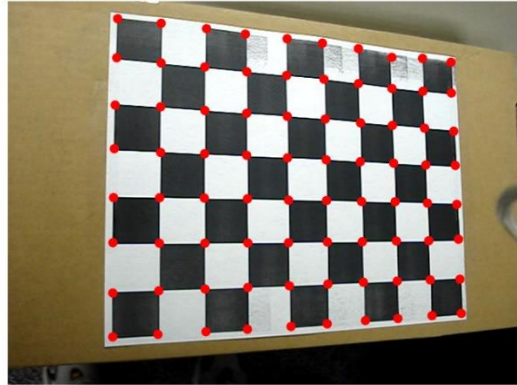


Figure3: grid points



Figure4: grid points



2.1.4 Updated H, K matrices

images2.png, H is

```
-0.9699    0.0873   -53.6111
-0.0176   -0.8914   -43.0744
-0.0000    0.0002   -0.6022
```

images9.png, H is

```
-1.1014    0.0357   -70.6884
-0.1500   -0.9475    -9.3528
-0.0005    0.0001   -0.5191
```

images12.png, H is

```
-0.7133    0.0517   -74.6938
 0.1782   -0.9021   -59.5219
 0.0005    0.0002   -0.6674
```

images20.png, H is

```
 0.8520   -0.2672   119.5679
-0.0072    0.4013    55.3491
 0.0000   -0.0008    0.6767
```

image : images2.png

R =

```
-0.9998    0.0142   -0.0043
-0.0180   -0.9864    0.1640
-0.0014    0.1640    0.9864
```

T =

```
147.8365
102.9665
-446.8938
```

image : images9.png

R =

```
-0.9243   -0.0094   -0.3815
-0.0261   -0.9952    0.0933
-0.3807    0.0970    0.9197
```

T =

```
99.1289
114.2336
-373.3122
```

image : images12.png

R =

```
-0.9153   -0.0063    0.4028
 0.0546   -0.9916    0.1158
 0.3991    0.1292    0.9079
```

T =

```
148.3980
101.9572
-496.2113
```

image : images20.png

R =

```
0.9999    0.0035    0.0044
-0.0097    0.7141    0.7000
 0.0034   -0.7000    0.7141
```

T =

```
-120.2781
-124.0846
 574.6801

721.3977   -0.6624   327.5222
 0  707.9265   234.6418
 0           0     1.0000
```

K_matrix =

```
721.3977   -0.6624   327.5222
 0  707.9265   234.6418
 0           0     1.0000
```

2.1.5 Err_Reprojection

(initial average is before improvement)

(the last average is after improvement)

```
image : images2.png
Sum err_reprojection:new corner-- approximate corner is=
  155.9689

Average err_reprojection:new corner-- approximate corner is=
  1.9496

Sum err_reprojection:new corner-- correct corner is=
  143.9755

Average err_reprojection:new corner-- correct corner is=
  1.7997

image : images9.png
Sum err_reprojection:new corner-- approximate corner is=
  156.8206

Average err_reprojection:new corner-- approximate corner is=
  1.9603

Sum err_reprojection:new corner-- correct corner is=
  149.1836

Average err_reprojection:new corner-- correct corner is=
  1.8648

image : images12.png
Sum err_reprojection:new corner-- approximate corner is=
  184.9190

Average err_reprojection:new corner-- approximate corner is=
  2.3115

Sum err_reprojection:new corner-- correct corner is=
  166.4552

Average err_reprojection:new corner-- correct corner is=
  2.0807
```

```
image : images20.png
Sum err_reprojection:new corner-- approximate corner is=
    90.2832

Average err_reprojection:new corner-- approximate corner is=
    1.1285

Sum err_reprojection:new corner-- correct corner is=
    135.8244

Average err_reprojection:new corner-- correct corner is=
    1.6978
```

2.1.6 Can this be done automatically?

This can be done automatically using Harris corner detector. a corner can be interpreted as the junction of two edges, where an edge is a sudden change in image brightness. In our case, the points we are interested in are dark so out of the many points reported by harris detector, the number of points reduce. Now, we also know the approximate distance between the points, hence detection of the four corners becomes even easier.

TASK 3 Augmented Reality 101

3.1 Outputs

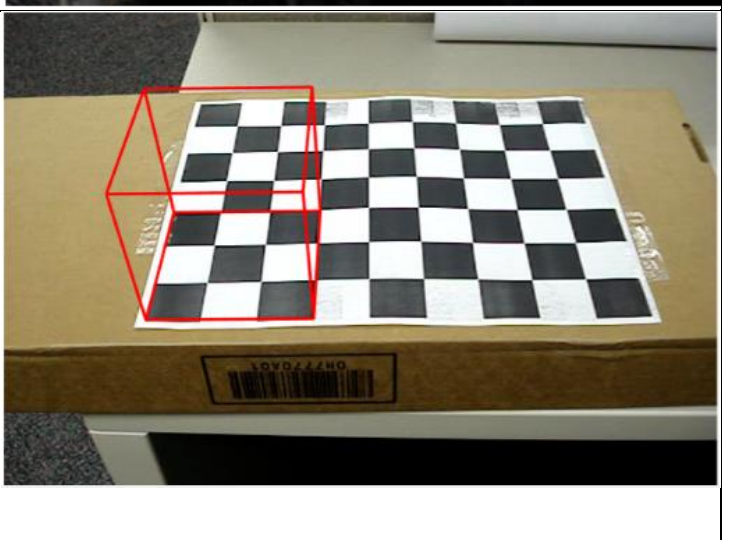
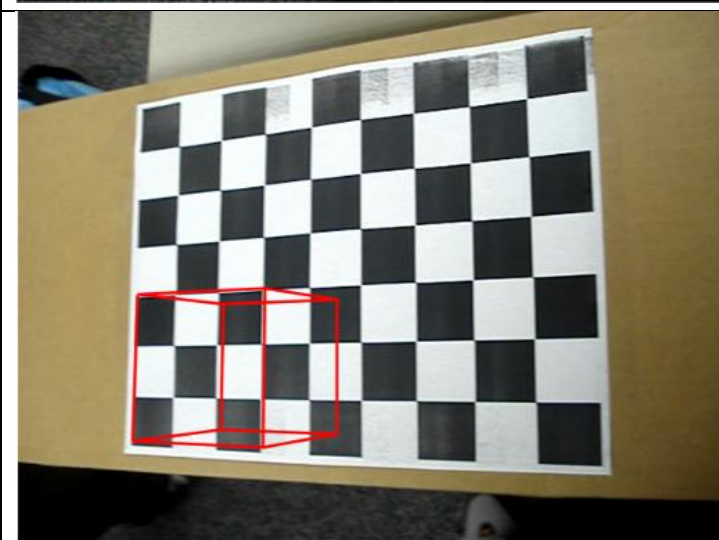
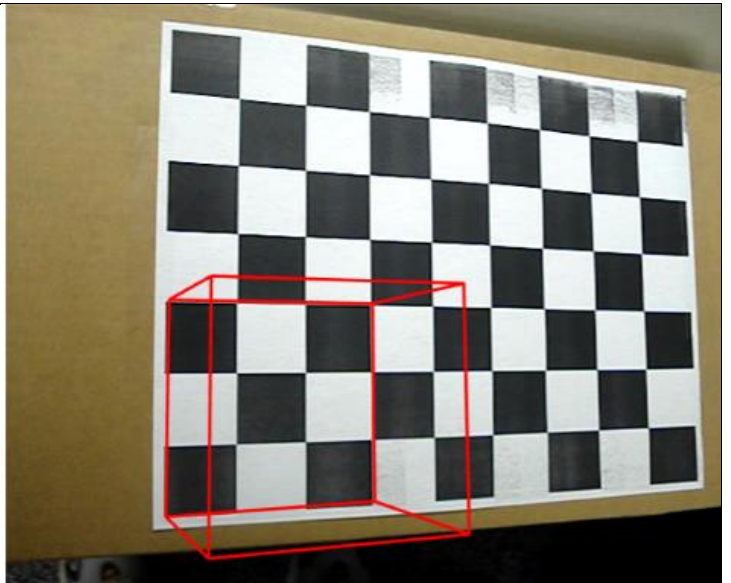
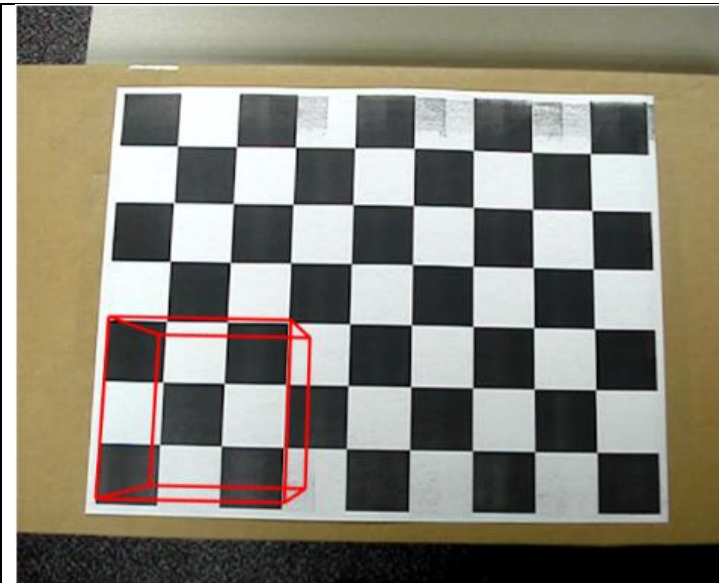
Output: 3.1.1 Image Augmentation

My RUID is 204003359 → So, I used 9.jpg as my clipart image





Output: 3.1.2 Object Augmentation



References

- 1) https://github.com/sumitd-archives/Augmented_Reality
- 2) [Computer Vision: Algorithms and Applications] by Richard Szeliski

3) Calibration by Viewing a Plane from Unknown Orientations - Zhang, ICCV99

Directions to execute the code

Execute the below files to run the code.

The folder Structure of the training and testing images is as follows:

Assignment 3

```
├── ...
├── Part 1
│   ├── part1.mlx
│   ├── Other support functions
│   └── Output
└── Part 2
    ├── part2.m
    ├── Other support functions
    ├── Output
├── Part 3
│   ├── part 3.mlx
│   ├── Other support functions
│   └── Output
```

Folder
code-matlab file(main)
matlab files

Folder
code-matlab file(main)
matlab files

Folder
Folder
code-matlab file(main)
matlab files

Folder