

CS655000 Computer Vision Homework 2

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Part 1. Camera Calibration

- a. Compute the projection matrix from a set of 2D-3D point correspondences by using the least-squares (eigenvector) method for each image.

Chessboard 1	$\begin{bmatrix} [-1.20998210e-01 & 2.43149939e-02 & 3.70139071e-02 & -2.80793151e-01] \\ [6.73781113e-03 & -2.00491352e-02 & 1.46634752e-01 & -9.39465030e-01] \\ [-1.65686547e-05 & 2.01612051e-04 & 1.76893633e-04 & -3.10810712e-03] \end{bmatrix}$
Chessboard 2	$\begin{bmatrix} [-1.11019282e-01 & -3.84602121e-02 & 3.00658422e-02 & -5.07214623e-01] \\ [1.10579817e-02 & -1.30618413e-02 & 1.32845786e-01 & -8.42660119e-01] \\ [-1.07631674e-04 & 1.49557766e-04 & 1.37512268e-04 & -2.80725147e-03] \end{bmatrix}$

- b. Decompose the two computed projection matrices from (A) into the camera intrinsic matrices K, rotation matrices R and translation vectors t by using the Gram-Schmidt process. Any QR decomposition functions are allowed. The bottom right corner of intrinsic matrix K should be normalized to 1. Also, the focal length in K should be positive.

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output.txt - 記事本
檔案(F) 編輯(E) 格式(O) 檢視(V) 說明
***** pic 1 *****
# ----- begin compute the matrix P ----- #
----- projection matrix -----
[[-1.20998210e-01  2.43149939e-02  3.70139071e-02 -2.80793151e-01]
 [ 6.73781113e-03 -2.00491352e-02  1.46634752e-01 -9.39465030e-01]
 [-1.65686547e-05  2.01612051e-04  1.76893633e-04 -3.10810712e-03]]
# ----- begin decompose P ----- #
----- K -----
[[441.79419439  1.97692207 186.31607276]
 [ 0.          461.45465546 301.67533168]
 [ 0.          0.          1.          ]]
----- R -----
[[-0.99360011 -0.10867566  0.0307965 ]
 [ 0.09464302 -0.65215767  0.7521523 ]
 [-0.06165647  0.75025328  0.65826932]]
----- t -----
[ 2.51264954 -0.01472351 -11.56611198]]
# ----- begin Reprojection ----- #
chessboard_1 error : 0.7071067811865476
***** pic 2 *****
# ----- begin compute the matrix P ----- #
----- projection matrix -----
[[-1.11019282e-01 -3.84602121e-02  3.00658422e-02 -5.07214623e-01]
 [ 1.10579817e-02 -1.30618413e-02  1.32845786e-01 -8.42660119e-01]
 [-1.07631674e-04  1.49557766e-04  1.37512268e-04 -2.80725147e-03]]
# ----- begin decompose P ----- #
----- K -----
[[489.80440977 11.66276711 195.44556311]
 [ 0.          507.47836703 286.10939159]
 [ 0.          0.          1.          ]]
----- R -----
[[-0.80757977 -0.58968593  0.00924243]
 [ 0.35870077 -0.4786834   0.8013713 ]
 [-0.46813318  0.65048652  0.59809583]]
----- t -----
[ 0.37612932 -0.33836009 -12.20985892]]
# ----- begin Reprojection ----- #
chessboard_2 error : 0.9204467514322717
***** finish *****

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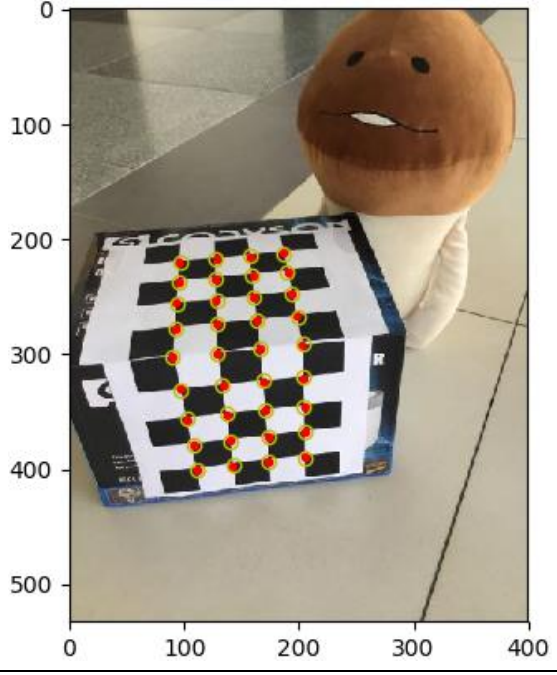
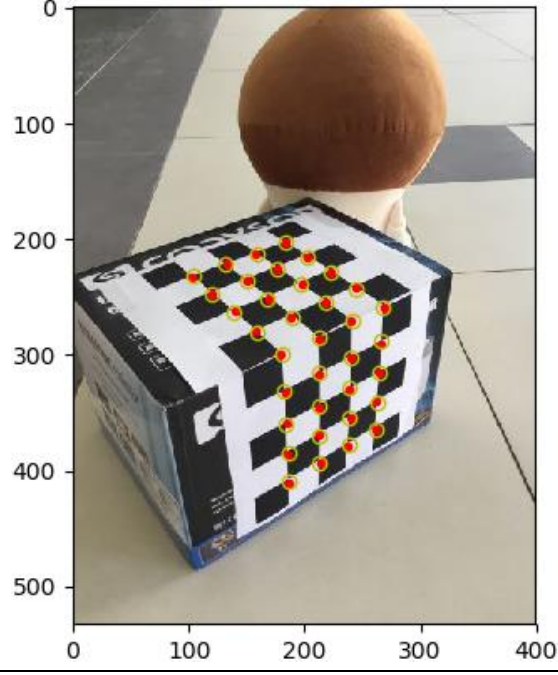
Chessboard 1

Chessboard 2

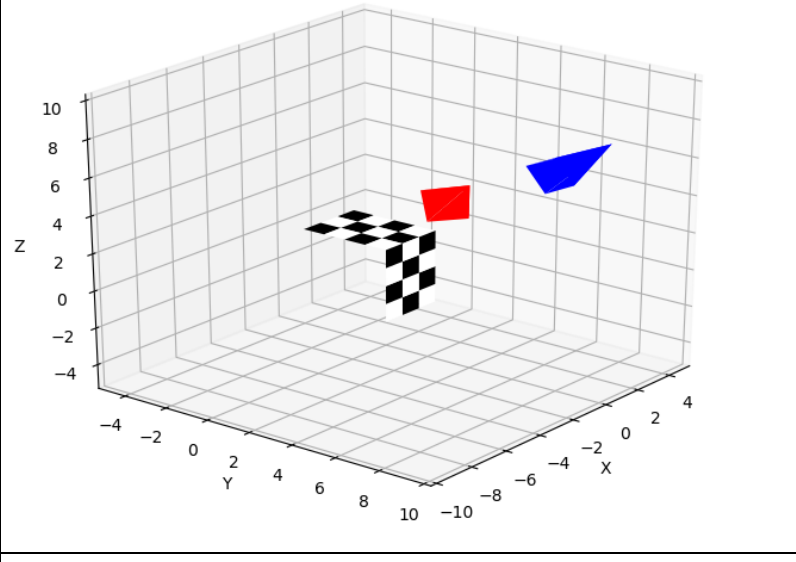
The decompose results show in the output.txt.

- c. Re-project 2D points on each of the chessboard images by using the computed intrinsic matrix, rotation matrix and translation vector. Show the results (2

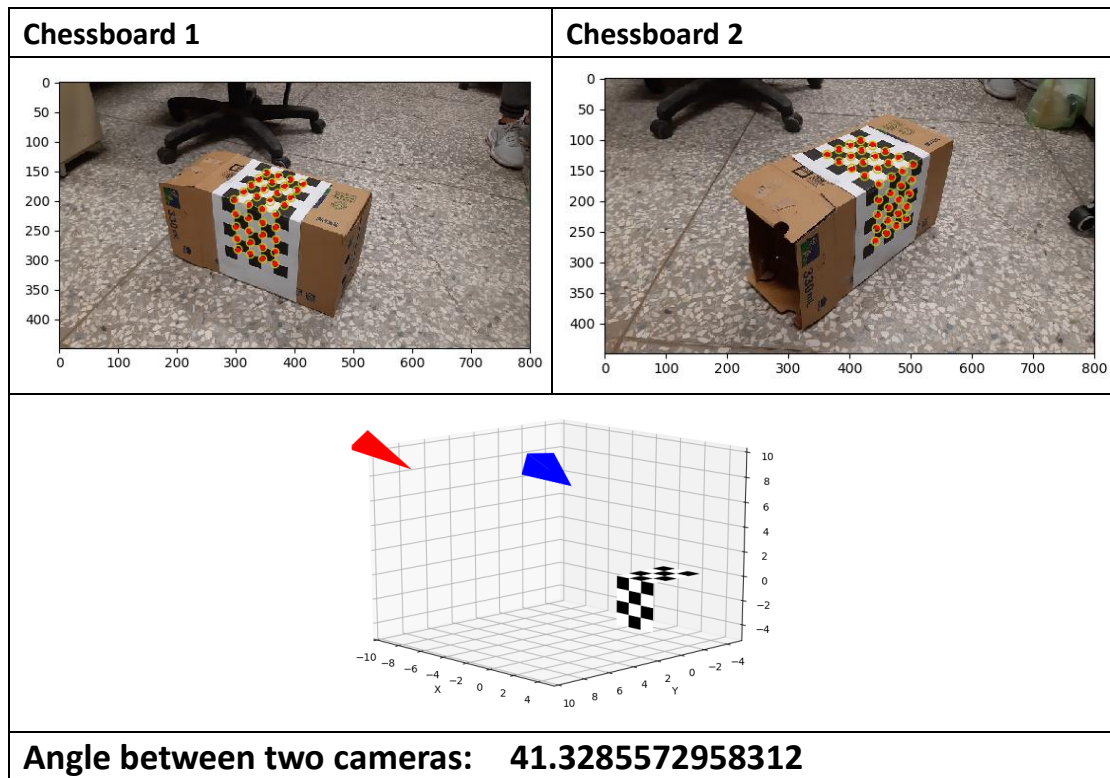
images) and compute the point re-projection root-mean-squared errors.

Chessboard 1	Chessboard 2
	
error : 0.7071067811865476	error : 0.9204467514322717

d. Plot camera poses for the computed extrinsic parameters (R , t) and then compute the angle between the two camera pose vectors.

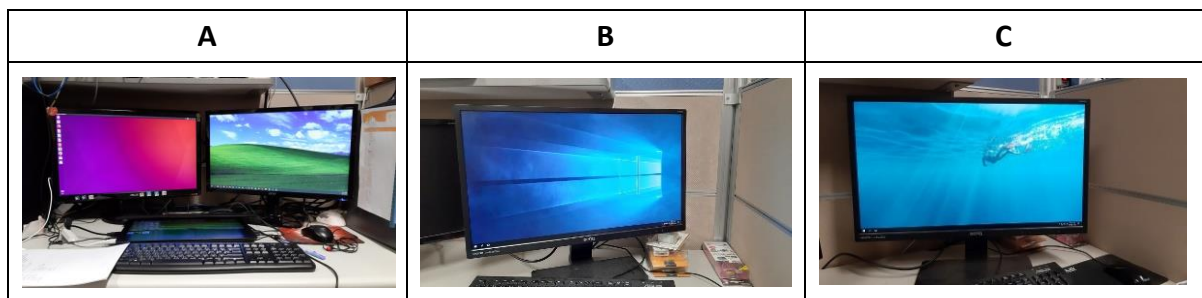
camera poses	
angle between the two camera pose vectors	Angle between two cameras: 24.4114323693213

e. E. (Bonus) (10%) Print out two “chessboard.png” in the attached file and paste them on a box. Take two pictures from different angles. For each image, perform the steps above (A ~ D).



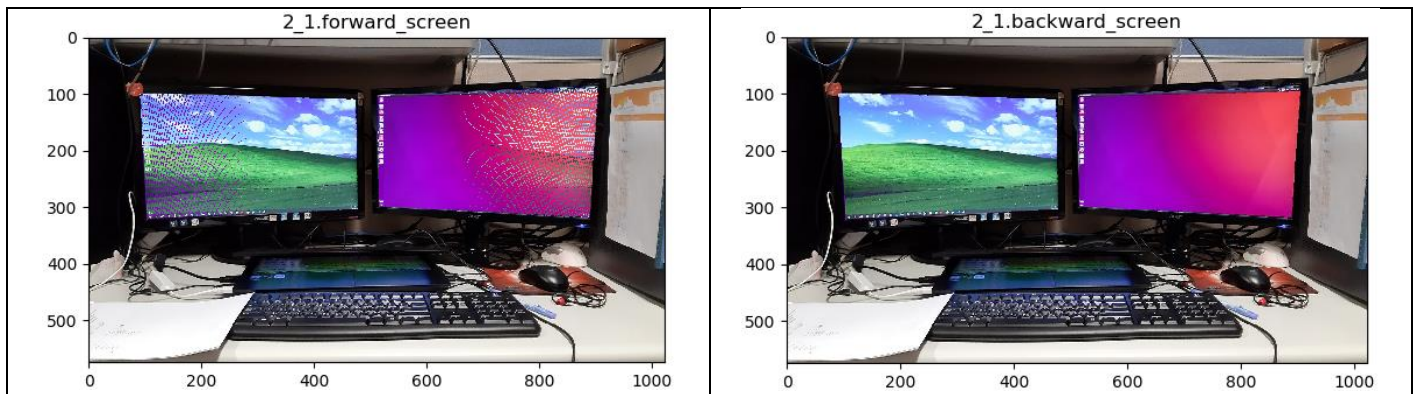
Part 2. Homography transformation

- a. Shoot three images A, B and C. Image A has to contain two objects. Image B and C should contain one object separately. Like the images shown above. (3 images)



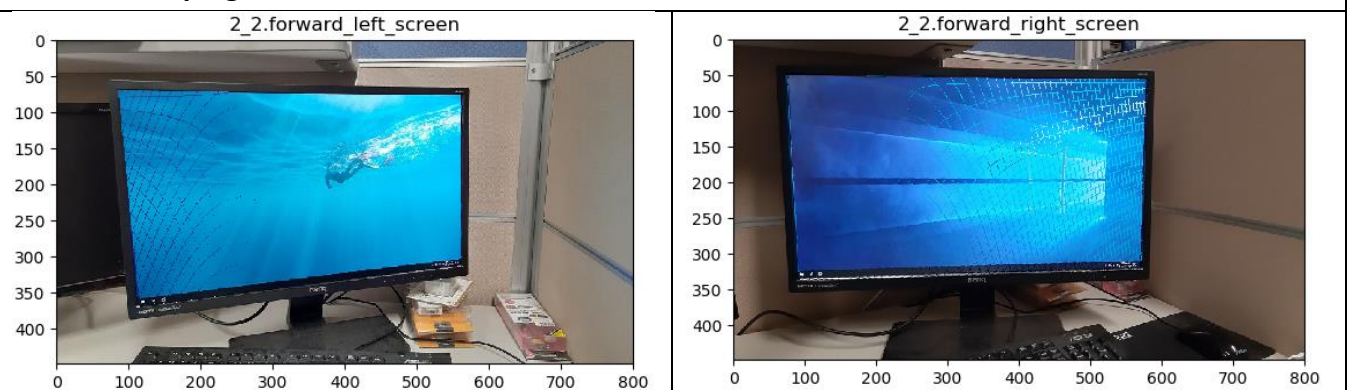
- b. Compute homography transformation between the two objects in image A. Use both backward and forward warping to switch them, like what example 1 shows. (2 images)

Forward warping	Backward warping
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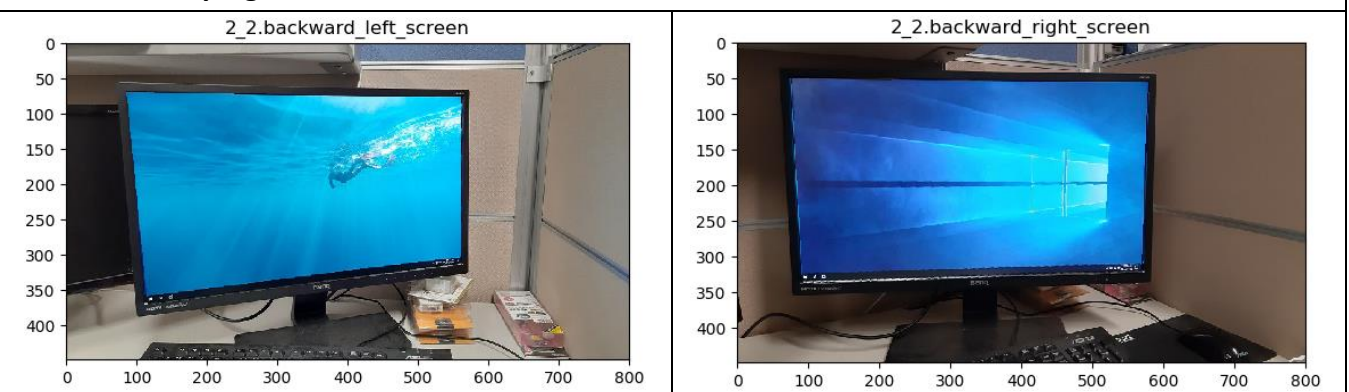


- c. Compute homography transformation between the object in image B and the object in image C. Use both backward and forward warping to switch them. Example 2 gives some illustration. (4 images)

Forward warping



Backward warping



- d. Discuss the difference between forward and backward warping based on your results.

使用Forward warping會無法完全填滿target的pixel位子。因為Forward是從source的pixel位置推算從target投影過來的pixel位置；另外因為投影計算的位置，會產生些許誤差，造成target位置無法被完全填滿。

Backward warping較能完全填滿target的pixel位子。因為Backward是直接從target的pixel位置推算從source投影過來的pixel位置，所以較能填滿。