

# Computer Vision Lab 4: 3D Reconstruction

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# 1 Introduction

During the fourth lab session, the primary objective was to reconstruct a 3D scene from a pair of images taken from different camera positions. The computations involved calculating intrinsic and extrinsic parameters through Tsai method. Those parameters (for individual cameras) were then used to construct a global transformation matrix to get world co-ordinates from selected points in image. The process involved calibration within the stereo setup. The process is detailed in the coming sections.

The pair of images from stereo setup that were used in this lab session are presented in figure 1.

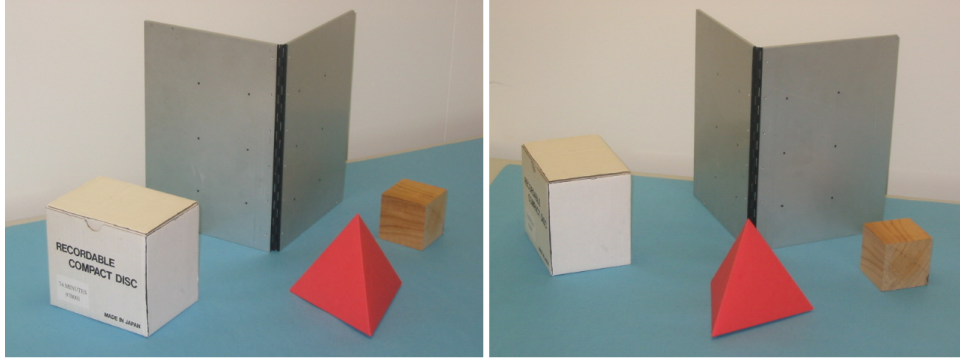


Figure 1: Images used (from stereo setup)

We reconstructed every element (2 cubes, 1 pyramid and a book-cover shaped object) from these images.

## 2 Verifying parameters of camera calibration matrix

The first step was to re-project 3D points into 2D space. A part of this code (generating calibration matrix) was already included in the main file of this exercise, which is `Let_me_reconstruction_3D.m`. The points used for calibration were also given in three files: `XYZ.mat`, `xyleft.mat` and `xyRight.mat`. The calibration was done using function `calibTSAI()`.

The projection matrix was calculated using  $M = K \times RT$ , where both  $K$  and  $RT$  were obtained from calibration. The resulting matrix was stored in `P1Tsai` and `P2Tsai`, corresponding to left and right camera poses respectively. These matrices were then multiplied with the given 3D points (in `XYZ.mat`) to produce 2D projected coordinates. Since there was an extra dimension and so to exclude the scaling, the first and second elements of the generated 2D coordinates were divided by the last element (Z-co-ordinate) of every point. The resulting positions of these points in the images are presented in figure 2. Figure 2 confirms the accuracy of calibration parameters.

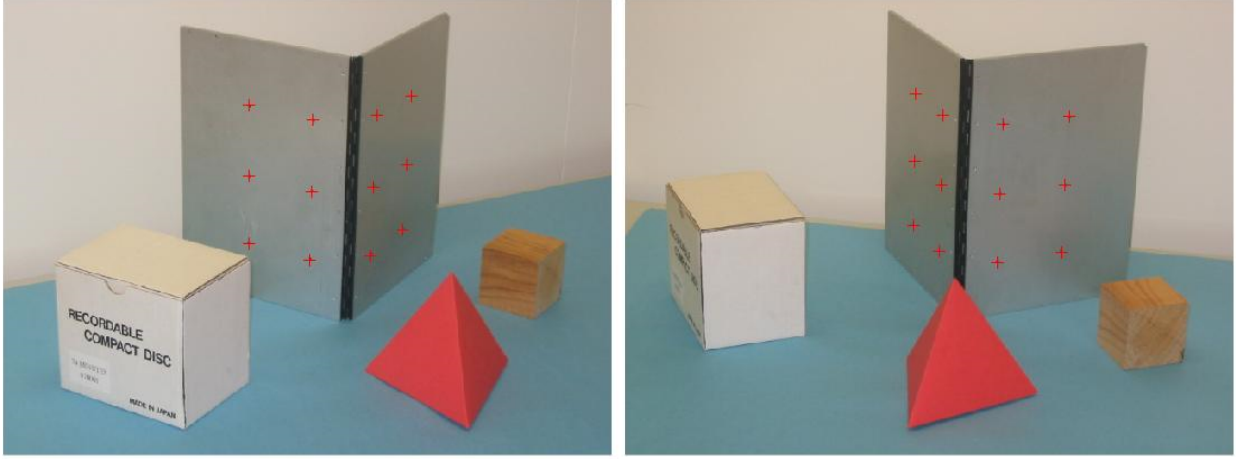


Figure 2: Reprojected points from 3D to 2D (in red)

### 3 Selecting points for projection to 3D

We selected 4 point from corners of pyramid, 8 points for both the cube and the rectangular box, and 6 points to represent book-cover-shaped large object. Please note that another way to represent the book-cover-shaped object could be to select 4 points and not project all the corners. However, we imagined that it would be a better visual representation if we do it for the complete shape.

In the code provided in `Let_me_reconstruction_3D.m`, there was already code written for selecting the points. However that code did not show the selected point over the image, therefore liberty was taken in slightly modifying it to show the points that were selected. Selected points are presented in figure 3.

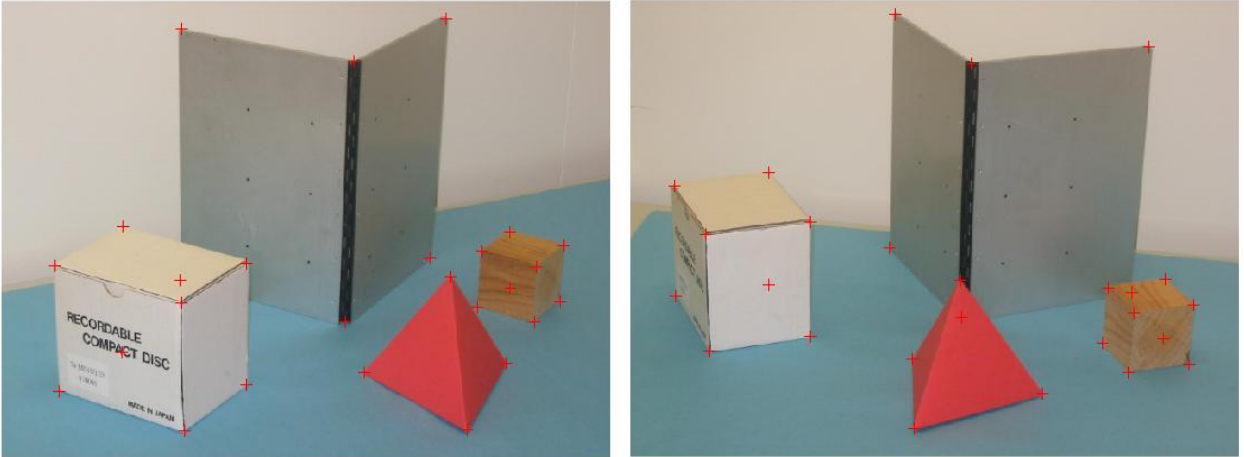


Figure 3: Selected points for each of the object (In red)

Selected point for each object were saved in a separate file for later use. The points for pyramid are stored in `pyrl.mat` and `pyrr.mat`, the points for cube is stored in `cubel.mat` and `cuber.mat`, for rectangular box in `rectl.mat` and `rectr.mat` and for book-cover-shaped object, the points are stored in `bookl.mat` and `bookr.mat`. (l for the left image and r for the right image)

The co-ordinates of selected points (In homogeneous) are given below:

### Pyramid:

```
>> Ppyrlpxhom % In left image
Ppyrlpxhom =
    473.0557    381.8368    490.0413    531.5616
    290.1986    390.8539    451.2471    380.7883
         1.0000         1.0000         1.0000         1.0000
>> Ppyrrpxhom % In right image
Ppyrrpxhom =
    347.2366    296.9089    298.7962    434.0518
    295.8604    377.6429    451.2471    414.1304
         1.0000         1.0000         1.0000         1.0000
```

### Cube:

```
>> Pcubelpxhom % In left image
Pcubelpxhom =
    505.1396    563.6455    560.5000    501.3650    534.7071    591.9548    588.1802    535.3362
    263.7765    279.5039    338.0098    322.9115    243.0164    258.1147    316.6206    302.1514
         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000
>> Pcuberpxhom % In right image
Pcuberpxhom =
    503.2523    530.3034    524.6415    496.9613    563.0164    593.2130    588.1802    561.1291
    303.4096    329.2025    391.4830    361.2864    294.6022    321.6533    383.3047    356.2536
         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000
```

### Rectangular box:

```
>> Prectlpxhom % In left image
Prectlpxhom =
    56.5944    189.9626    193.1081    60.9980    128.9404    258.5341    258.5341    127.0531
    278.2457    317.8788    452.5052    410.9849    237.9836    276.3585    403.4358    369.4646
         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000
>> Prectrpxhom % In right image
Prectrpxhom =
    46.5288    80.5000    83.6455    48.4161    146.5550    189.3336    189.9626    145.9260
    196.4633    247.4201    369.4646    311.5878    181.9941    233.5799    354.3663    299.6350
         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000         1.0000
```

### Bookcover shaped object:

```
>> Pbooklpxhom % In left image
Pbooklpxhom =
    189.3336    371.1422    467.3938    451.0374    361.7058    188.7045
```

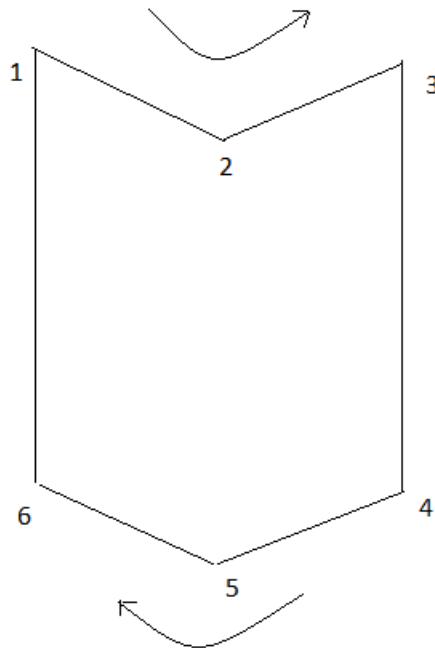
```

30.3820    63.7241    19.0583    270.6966    338.0098    293.9731
 1.0000     1.0000     1.0000     1.0000     1.0000     1.0000
>> Pbookrpxhom % In right image
Pbookrpxhom =
279.2942   359.8185   546.0308   527.7870   349.1239   273.0033
 16.5419    67.4987    50.5131   307.8132   334.2353   260.6311
 1.0000     1.0000     1.0000     1.0000     1.0000     1.0000

```

#### Note:

If you wish to do manual selection of the points instead of using the saved points, it is important to remember that the objects need to be selected in following order: First the pyramid, then cubes (cubes in any order) and lastly, the large bookcover shaped object. It is also important to remember that while selecting the points for bookcover shaped object, the points need to be selected clockwise. This is to avoid connecting points which do not share an edge. For assistance see figure 4.



#### Order of selection:

First: Pyramid (4 corners)

Second: Cube or rectangle (8 corners)

Third: Cube or rectangle (8 corners)

Fourth: Bookcover shaped object (6 corners)

(There can be another representation with 4 points for bookcover shaped object but with 6 points, it looks better)

Figure 4: Recommended order of selecting points for 3D visualization later

## 4 Calculating Camera co-ordinates

Two matrices  $K1_{sf}$  and  $K2_{sf}$  were computed from individual camera calibration matrices, these are  $3 \times 3$  matrices and are obtained by removing last columns of  $K1$  and  $K2$  respectively.

We obtained camera coordinates for each of above selected points by using  $K1_{sf}^{-1}$  multiplied by ar-

ray of selected points (presented in previous section). The points are stored in separate files named for ease as `cam_pyr_l.mat`, `cam_pyr_r.mat` etc.

## 5 3D Reconstruction

### 5.1 Global rotation and translation matrices

To calculate the global rotation and translation matrices, following equations were used:

$$R = R_r R_l' \quad (1)$$

and

$$T = T_l - RT_r \quad (2)$$

Where R and T are global rotation and translation matrices respectively.

### 5.2 Calculating 3D camera points

To compute 3D camera coordinates, we used algebraic and matrix operations for the calibration matrix. Essentially, the following equations were used:

$$P = P_1 + 0.5 * (P_2 - P_1) \quad (3)$$

Where

$$P_1 = a_0 P_l \quad (4)$$

and

$$P_2 = T + b_0 R' P_r \quad (5)$$

Here,  $a_0$  and  $b_0$  are components of a matrix  $abc$  that is computed by  $A^{-1}T$ , and  $A$  is given by,  $[p_l - R'P_r \ P_l \times R'P_r]$ .  $P_l$  and  $P_r$  are the points in (2D) camera coordinates respectively for the left and right images.

This was followed by saving the final point P as a homogeneous co-ordinate by adding an extra dimension with 1's.

### 5.3 Calculating 3D world points

We used following equation for calculating

$$X_c = RTX_c \implies X_w = RT^{-1}X_c \quad (6)$$

The 3D points were then plotted in a 3D graph and the results are shown in figure 5.

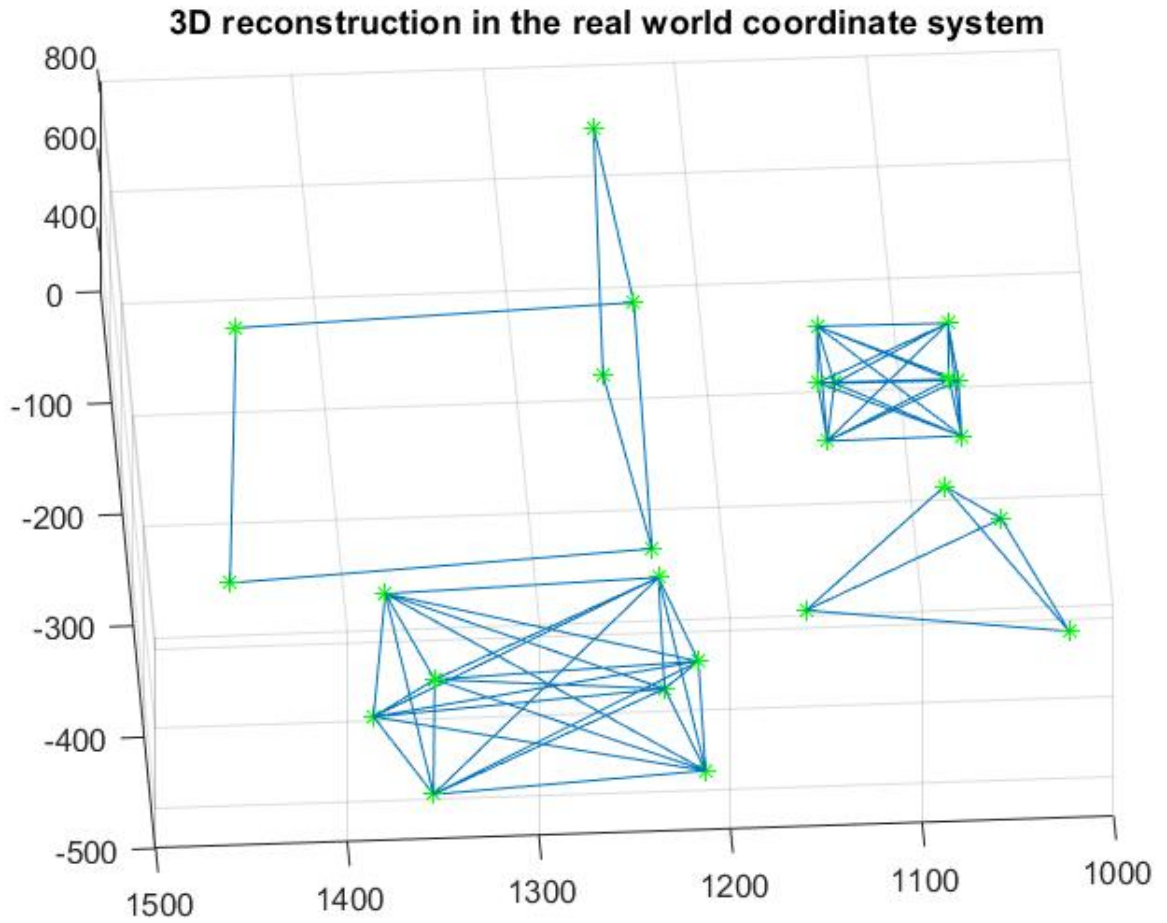


Figure 5: Caption

**Note:** The cube and rectangular box are presented with lines for diagonals as well. This was more intuitive. However, if needed the lines can easily be removed.

## 6 Conclusion

We were successful in our task of representing a 3D scene with the input of two images from different camera poses. Further, this lab exercise also completed the necessary tools for basics of Computer Vision - corner detection, calibrating camera, learning about epipolar geometry and 3D reconstruction.