

ENPM673 – Perception for Autonomous Robots

Project 4

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Due date: 26th April 2023, 11:59 PM

Submission guidelines:

- This homework is to be done and submitted individually.
- Your submission on ELMS/Canvas must be a **zip file & a pdf file**, following the naming convention
- YourDirectoryID_proj4.zip & YourDirectoryID_proj4.pdf. If your email ID is abc@umd.edu or abc@terpmail.umd.edu, then your Directory ID is abc. Remember, this is your directory ID and NOT your UID.
- Please submit only the python script(s) you used to compute the results, the PDF report you generate for the project and a detailed README.md file which includes the steps to run your code and any non-standard libraries used. The zip file should contain only the source code and related files.
The report should not be inside the zip file.
- Include sample outputs in your report.
- For each section of the homework, explain briefly what you did, and describe any interesting problems you encountered and/or solutions you implemented.
- Disallowed function (In general, you are not allowed to use in-built functions unless you are instructed otherwise):
 - any inbuilt function that lets you compute the fundamental and essential matrices directly.
 - any other inbuilt function that directly computes the disparity or stereo correspondences.

Project Description:

In this project, we are going to implement the concept of Stereo Vision. We will be given 3 different datasets, each of them containing 2 images of the same scenario but taken from two different camera angles. By comparing the information about a scene from 2 vantage points, we can obtain the 3D information by examining the relative positions of objects.

Dataset Preparation:

Please use this [link](#) to download the 3 Datasets that you will try to implement stereo vision on, individually. Each folder contains a `calib.txt` file

A brief explanation of the terms used in the `txt` files:

cam0,1: camera matrices for the rectified views, in the form $[f \ 0 \ c_x; 0 \ f \ c_y; 0 \ 0 \ 1]$,

where f : focal length in pixels; c_x, c_y : principal point

doffs: x-difference of principal points, $doffs = c_{x1} - c_{x0}$ (here always $= 0$)

baseline: camera baseline in mm

width, height: image size

ndisp: a conservative bound on the number of disparity levels; the stereo algorithm MAY utilize this bound and search from $d = 0 \dots ndisp-1$.

vmin, vmax: a tight bound on minimum and maximum disparities, used for color visualization; the stereo algorithm MAY NOT utilize this information

Pipeline for creating a Stereo Vision System

1. Calibration

- First, we need to compare the two images in each dataset and select a set of matching features. You can use any inbuilt function for feature matching.
- Estimate the Fundamental matrix using the features obtained in the previous step. Refer to section 3.2.2 in this [link](#) to get an overall understanding of Fundamental matrix estimation. You can use the inbuilt SVD function to solve for the fundamental matrix. You have to use the RANSAC method to estimate the fundamental matrix.
- Estimate Essential matrix E from the Fundamental matrix F by accounting for the calibration parameters. You should implement the functions to estimate the Essential matrix and also to recover the rotation/translational matrices.
- Decompose E into a translation T and rotation R .

2. Rectification:

- Apply perspective transformation to make sure that the epipolar lines are horizontal for both images.
- You can use inbuilt functions for this purpose.
- Print the H1 and H2, homography matrices for both left and right images that will rectify the images.
- Plot the epipolar lines on both images along with features points

3. Correspondence:

- For each epipolar line, apply the concept of the matching window (discussed in class such as SSD or Cross Correlation).
- Calculate Disparity.
- Rescale the disparity to be from 0-255 and save the resulting image.
- You need to save the disparity as a grayscale and color image using heat map conversion.

4. Compute Depth Image:

- Using the disparity information obtained above, compute the depth information for each image pixel. The resulting depth image has the same dimensions as the disparity image but it has depth information instead.
- You need to save the depth image as a gray scale and color using heat map conversion.

References:

The dataset used for this project is [MiddleBury Stereo Dataset](#)

D. Scharstein, R. Szeliski and R. Zabih, "A taxonomy and evaluation of dense two-frame stereo correspondence algorithms," Proceedings IEEE Workshop on Stereo and Multi-Baseline Vision (SMBV 2001), Kauai, HI, USA, 2001, pp. 131-140, doi: 10.1109/SMBV.2001.988771.