$\textbf{16-822} \hbox{: Geometry-based Methods in Vision (S16)} \quad \hbox{Released: Nov-21., Due: Dec 7th}$

Homework 3

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1 Theory (60 Points)

1.1 Affine cameras (35 Points)

In class, we have mostly focused on perspective cameras. For this part of the homework, we consider cameras with center lying on the plane at infinity, especially *affine cameras*. An affine camera is one that has a camera matrix of the form M = [A | b] in which A is singular and the last row of M is of the form (0,0,0,1). The following questions explore affine cameras a little more in depth.

- What happens to points on a plane perpendicular to the image plane that also passes through the world origin?
- Given a set of points X_i in 3D, and their projection x_i in the image, what is the projection of the centroid of X_i 's? (no equations, just 1 short line.)
- Show that an affine camera maps parallel world lines to parallel image lines; i.e., when using an affine camera, parallel lines in 3D project to parallel lines in image plane.¹
- Show that for parallel lines mapped by an affine camera, the ratio of lengths on line segments is an invariant.

For the following questions, consider a system of 2 affine cameras, with camera matrix M and M': (Last row of each camera matrix is of the form (0,0,0,1).)

- Show that epipolar planes (and lines) are parallel.
- Show that the affine fundamental matrix F defined by two affine cameras is invariant to an affine transformation of the world coordinates.

1.2 Fundamental matrices between three cameras (25 Points)

In this problem we consider the relationship between 3 images. Let us denote by F_{ij} the fundamental matrix between image i and j (from 1 to 3), and by e_{ij} the epipole generated by image j in image i. We will now show that the 3 fundamental matrices induced by 3 cameras are not independent.

- Draw the 3 cameras with e_{ij} labelled.
- Show that: $F_{ij}e_{ik} = e_{jk} \times e_{ji}$ for all triplets. (this can be shown purely geometrically, without involved algebra)

Actually, the converse is also true; if the camera preserves parallelism of lines, then it is an affine cameras.

• Based on these 3 quadratic relations, give a counting argument showing that the 3-camera geometry is characterized by 18 degrees of freedom. And thus conclude that the 3 fundamental matrices are not independent.

2 Having fun by learning from implementing! (40 Points)

In this part of the homework, we will explore 3D reconstruction. Especially, we will study a full pipeline of a typical 3D reconstruction system using Structure from Motion (SfM) and Stereo Matching. Slight different from the previous homeworks, this time you do not implement everything from scratch. Instead, you will delve into one of the existing packages for this purpose (Here we use the *SFMedu* system from the computer vision group at Princeton University) by analyzing the results and modifying some steps of particular interest and importance. For you, this is also a very good way to understand the material taught in the class (I can say this from first hand experience).

2.1 What you are supposed to do:

You need to both implement and answer questions as follows:



Figure 1: An illustration of the SFMedu system: Structure from Motion (SfM) and Stereo Reconstruction.

• Run the SFMedu system

I have already included the SFMedu package in ./code/ folder. Run the SFMedu system by running the SFMedu2.m script in MATLAB. Make sure that it runs successfully and produces a dense.ply point cloud file. Install MeshLab from http://meshlab.sourceforge.net, and open the point cloud in MeshLab and rotate around the model. Take a screen capture of the point cloud and put it in the report as the answer to this question. An illustration is shown in Figure 1.

• List the Major Steps

Read the code carefully, especially SFMedu2.m. Summarize the major steps for SfM and dense stereo matching, and write it down in the report as answer to this questions.

• Principal Point

What is the assumption for the principal points in this system? Under what condition will this assumption get violated?

• Data Structure

Graph is a data structure used in the code for bundle adjustment. It has 5 major fields. What are their names? What is the data in each field? How is the data organized? Why we need these 5 fields?

• Reconstruct Your Own Images

Capture/Find two sequences of images (at least 5 images per set) on some interesting objects or scenes, run the system on these images. If it fails, figure out why it fails, and change the code or retake good images until it works. Submit the input images, as well as three screen captures of your reconstruction per image set in the report. Explain all the changes or things you have to do to make it work. Which step do you think is the most unstable one? Explain why. List suggestion for a normal user without any computer vision background about how to take good images to make it work.

• Compute the Reprojection Errors

Write a function

function printReprojectionError(graph)

that takes a graph (as defined in the SFMedu system), and prints out the current reprojection error, so that you can insert the function into many steps for debugging purpose, to check if the reconstruction errors are getting smaller. A call to this function is at Line 160 of SFMedu2.m. Submit the function as code. Also, plot the reprojection errors in the report for the image sequence provided and the two sets of images you captured in the report PDF file.

• Visualize the Reprojection Points

Write a function

function visualizeReprojection(graph, frames)

that takes a graph (as defined in the SFMedu system), and draw the 3D keypoint point cloud projected onto each image, as well as their observed location. Figure 2 shows an example of the output for your function. You are required to write a function to produce the same kind of visualization. Each observed keypoint is represented by a red \times , and the reprojection of its 3D estimated location is represented by a green +, and these two points are connected by a blue line. For the 3D points that are projected to the image but not observed from the image, it should be shown as a yellow \circ . A call to this function is at Line 162 of SFMedu2.m. Submit the function as code, and also include the visualization results of the image sequence provided and the two sets of images you captured in the report PDF file.

• Levenberg-Marquardt

bundleAdjustment.m uses matlab function lsqnonlin to minimize the objective function via the Levenberg-Marquardt algorithm:



Figure 2: An example result produced by function visualizeReprojection(graph, frames).

```
[vec,resnorm,residuals,exitflag] =
lsqnonlin(@(x) reprojectionResidual(graph.ObsIdx,graph.ObsVal,px,py,f,x),
[Mot(:);Str(:)], [], [], options);
```

Please read all the code and figure out what is the objective function it is optimizing for. Write down the math equation of this objective function in your report.

The Matlab implementation Isquonlin of Levenberg-Marquardt is not suitable for very large scale problem. Read the document of Isquonlin and explain why this statement is true in your report.

• Motion Adjustment Only

Bundle adjustment typically optimizes the Structure (the 3D location of the points) and the Motion (the camera parameters) together, as the above lines of code. But we can also fix the Structure, to adjust the Motion only, i.e., not allow the 3D points to move, but only changing the rotaiton and location of the cameras to optimize the objective function. With some good initialization, this can be used as a fine adjustment to solve camera resectioning task as well (http://en.wikipedia.org/wiki/Camera_resectioning). Fill in Line 34 in bundleAdjustment.m.

Please write down the math equation for the objective function, and also change the above line of code to do this. Submit the code and illustrate the results.

• Structure Adjustment Only

In the same way of thinking, we can also fix the Motion, to adjust Structure only, i.e., not allow the cameras to move, but only allows the 3D points to change their locations. Similarly, with some good initialization, this can be used as a fine adjustment to do "triangulation" for reconstructing the 3D point locations. Fill in Line 30 in bundleAdjustment.m.

Please write down the math equation for the objective function, and also change the above line of code to do this. Submit the code and illustrate the results.

Although this can be used to do triangulation, there is certain drawback. What is that? Answer this in the report.

• Smarter Graph Merging

The current reconstruction is merging the sequence sequentially, by adding one camera into the mergedGraph in each step and do bundle adjustment again. A better graph merging will help to make the system more robust. For example, we can merge the graphs that have the maximal number of keypoint correspondences, instead of sequentially picking one. Change the code to implement this. Submit the code and illustrate the results.

• Intrinsics

Change the code to optimize for the full intrinsics matrix. Submit the code and illustrate your results.

2.2 What you have to submit:

You should submit three files:

- A report that contains all the images (both the images you captured and your result images) and answers to the questions.
- Your CODE folder that contains all source code (including YOUR readme) for your system, and your "SFMedu2.m" script that takes no parameter as input and runs directly in Matlab to generate the results reported in your PDF file.
- Your IMAGE folder that contains all the input, intermediate and output images.

2.3 Tips:

- Make sure to keep "sanity checks" in your code while you are debugging, by checking an equation that you know should be true. (e.g., $l^TCm = 0$). Keep such code commented when you submit, so that I can see and appreciate it!
- Normalize coordinates of points before doing anything. This is important!
- Remember the transformations that you are estimating are up to a scale.
- Keep in mind all the things Martial mentioned in class!
- Start early...

• Have fun!!