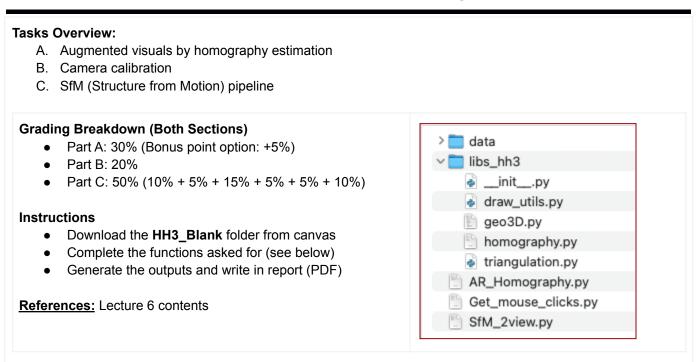
# EEL 4930/5934: <u>Autonomous Robots</u> HH3: Hands-on Homework #3 (Spring 2023)



Submission: [Through Canvas only; Due: April 4, 2023 by 11.59pm]

- A single zip file with a folder (code) and PDF (report)
  - Your completed code: do not add any more python files, just complete the functions
  - A PDF report: the generated outputs only (things that are asked in **blue color**; see below)
- Assignments more than 20 MB file size will get negative penalty (-10% to -50%)

## Part A: Augmented Visuals by Homography Estimation

#### Refer to Lecture-6: slide 16-24 and the following files

• Driver script: AR\_Homography.py

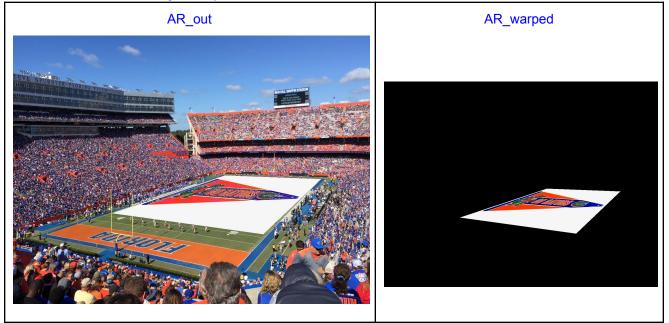
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- Library: libs\_hh3/homography.py
- Input data (see below): data/game\_fl.jpg and data/logo\_Gators.png



Your target is to augment the logo into the destination image by following the homography perspective transformation. To achieve this, please follow these steps

- Get the four points on the destination image using the Get\_mouse\_clicks.py
- The driver script AR\_Homography.py is already completed for you
- You need to complete the following library function (in libs\_hh3/homography.py) def computeHomography(Us, Vs)
- You can check your outputs compared to using cv2.findHomography function
- There will be two outputs as shown below; you should generate AR\_out and AR\_warped figures and show them in your report.



**Bonus point (+5):** There is a basic warping function implemented for you (in libs hh3/Homography.py)

def warp\_and\_augment(im\_logo, im\_dst, H)

You will see that it generates somewhat noisy output in some pixels. There are a few better ways to implement this function for much accurate warping. You will get bonus points (max: 5) if you can implement a better one! Please write the main ideas/intuitions of your algorithm briefly and show the comparative results in your report.

### Part B: Camera Calibration

In this part, you will calibrate a camera: cell-phone or any other camera that you have access to. Refer to <u>Lecture-6: slide 25-27</u> and find the **intrinsic camera parameters** as follows:

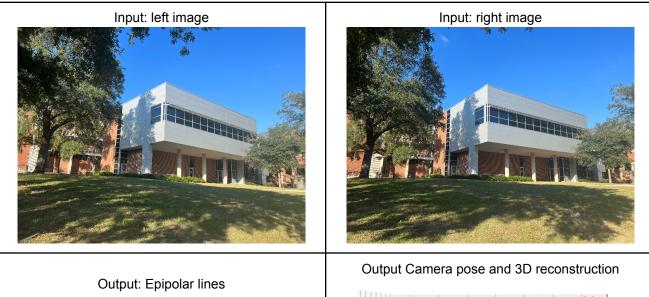
- Print a checkerboard and place it on a wall (a sample data/pattern.pdf is provided for you)
- Use any of the following libraries to calibrate your camera:
  - ROS, OpenCV, CalTech Matlab code
  - Or any other library of your choice
- Provide the intrinsic matrix **K** and your camera model in your report.
- Make sure your calibration is accurate! We will not have access to your camera for a comprehensive evaluation. However, if the **K** is inaccurate, your <u>Part C</u> results will be totally wrong.

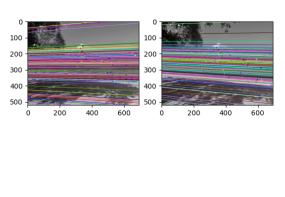
## Part C: SfM (Structure from Motion) pipeline

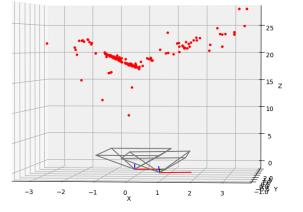
Refer to the Lecture-6 contents and the following files

- Driver script: SfM\_2view.py
- Library: libs hh3/geo3D.py
- Utility functions:
  - Some drawing functions are provided for you in libs\_hh3/draw\_utils.py
  - A few 3D triangulation algorithms are provided for you in libs\_hh3/triangulation.p
- Input data (for initial testing):
  - Images: data/uf\_left.png and data/right.png
  - Intrinsic matrix K: data/K\_iphone\_reduced.txt
- Data collection:
  - Use the same camera (you used in part B) to collect a 2-view image pair as the given input. Please try to choose a scene with a clear structure and identifiable object shapes.
  - Also use the intrinsic matrix **K** from your part B.

With the given inputs, the existing code will generate the following outputs







#### As you will see in the driver script: SfM 2view.py, the following steps are followed for the SfM

- 1. load\_image\_pair()
- 2. \_extract\_keypoints\_sift()
- 3. \_estimate\_fundamental\_matrix()
- 4. draw\_epipolar\_lines()
- 5. \_estimate\_essential\_matrix()
- 6. \_find\_camera\_matrices\_rt()
- 7. \_find\_projection\_matrices()
- 8. \_triangulate\_3d\_points()
- 9. plot\_point\_cloud()

You will need to implement parts of the colored functions #3, #5, #6, and #7. (see libs hh3/geo3D.py)

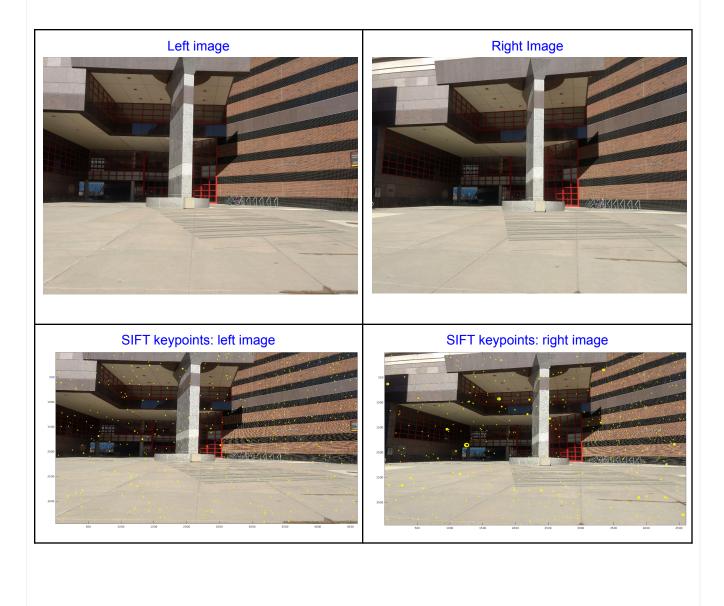
The ToDos are the following

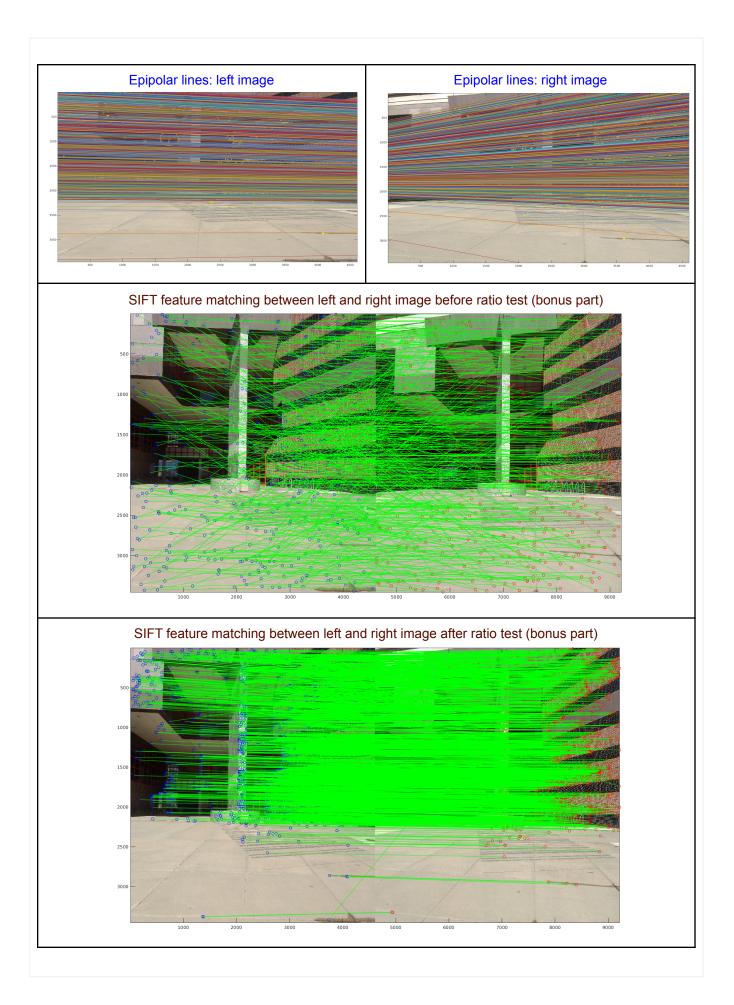
- Go through the implementations provided for you in libs\_hh3/geo3D.py
- Implement the \_estimate\_fundamental\_matrix() function [10 points]
- Implement the \_estimate\_essential\_matrix() function [5 points]
- Complete the find camera matrices rt() function [15 points]
- Implement the find projection matrices() function [5 points]

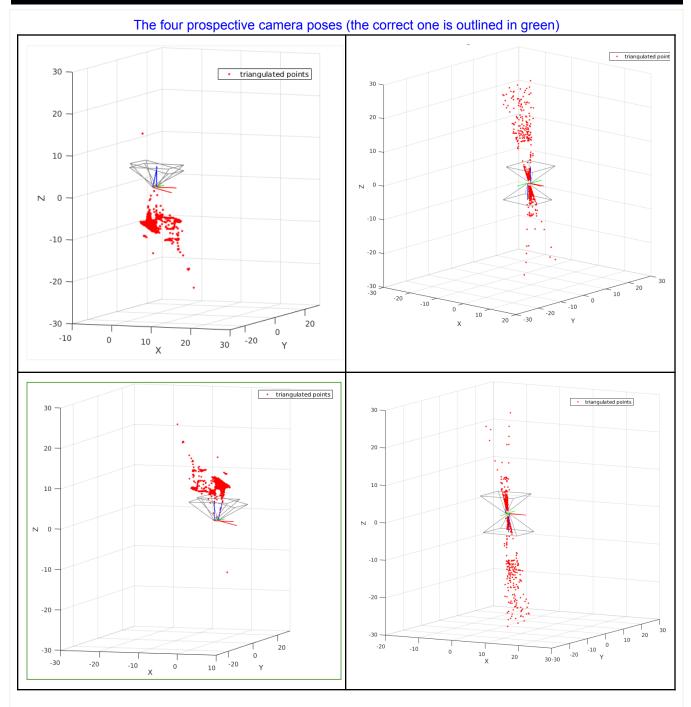
In addition, you will need to write a couple of functions to visualize

- The SIFT keypoints in each input image [5 points]
- SIFT keypoints matched between two images, before and after the 'ratio test' [Bonus: +5 points]
- The 4 camera poses estimated by the \_find\_camera\_matrices\_rt function [10 points]

For the report, show the generated outputs for the aforementioned cases. A sample example is given below:







Finally, you will **show the final 3D reconstruction** of your scene. Please adjust the scale and orientation of the plot so that the camera poses and the structure can be seen clearly. Also report the following calculations:

- Left projection matrix: P1
- Right projection matrix: P2
- Fundamental matrix: F
- Essential matrix: E

Remember, the assignment is due: <u>April 4, 2023 by 11.59pm</u>

Similar to the previous homework, your zipped submission folder (HH3\_GatorID.zip) should contain the completed code folder and a report PDF.