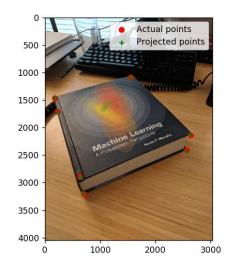
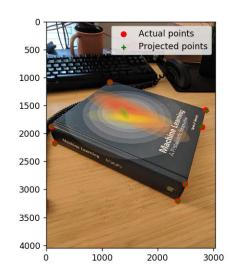
CS 4476/7646 Project 2

Tarun Pasumarthi tpasumarthi3@gatech.edu 903085537

<insert visualization of projected 3D points and actual 2D points for image provided by us here>



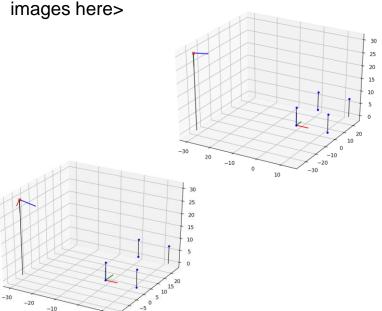


< insert the two images of your fiducial object here>

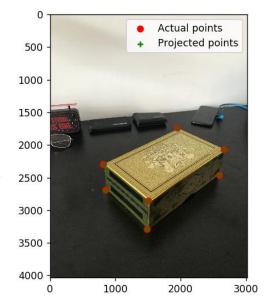


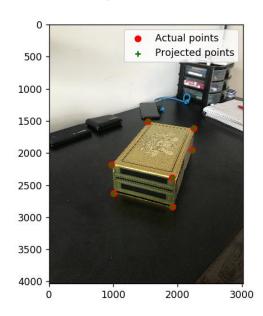


<insert visualization the initial guesses for
rotation matrix and camera center for the two
images here>

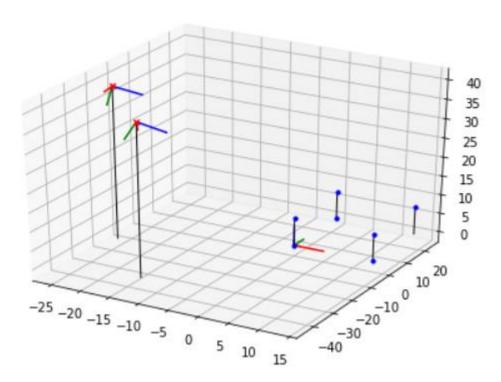


<insert visualization of projected 3D points and
actual 2D points for both the images you took>





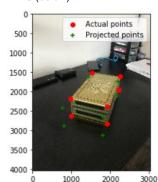
<insert visualization of both camera poses here>

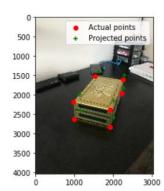


If you were to increase the xcoordinates or any other coordinates of the camera center t, the projected points should move down on the respective axis in the real world. So in the image this would be moving down the line that corresponds to the vanishing point of the given axis. Similarly, decreasing any of the coordinates of the center, should move the projected points up on the respective real-world axis, which in the image would be moving up the line corresponding to the vanishing point of that given axis.

X-axis of camera center t +3 (above)

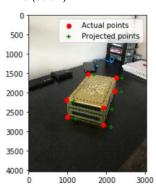
X-axis of camera center t -3 (below)

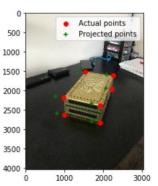




Y-axis of camera center t +5 (above)

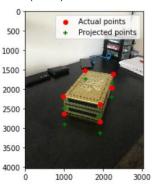
Y-axis of camera center t -5 (below)

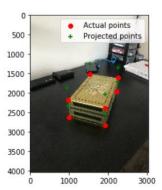




Z-axis of camera center t +5 (above)

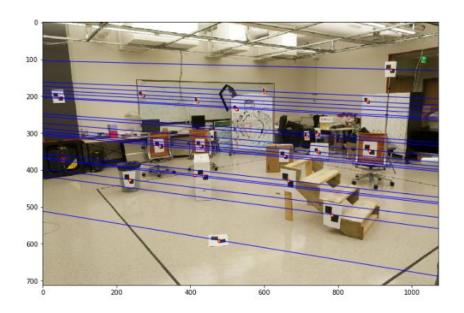
Z-axis of camera center t -5 (below)



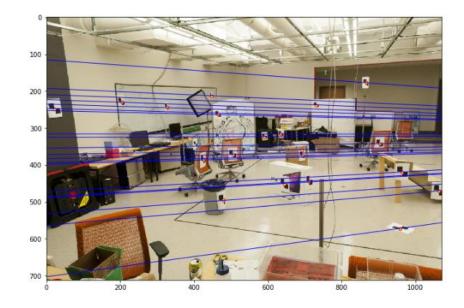


Part 2: Fundamental Matrix Estimation

Room: Left Image with Epipolar Lines



Room: Right Image with Epipolar Lines



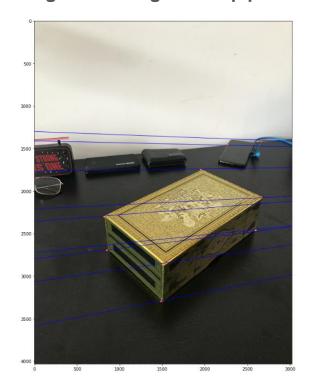
Part 2: Fundamental Matrix Estimation

Fundamental Matrix Estimation Result:

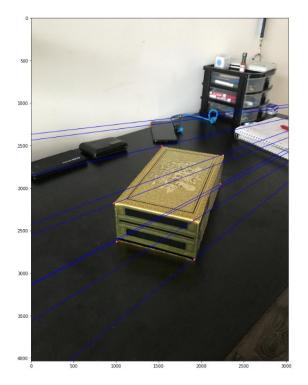
```
[[-0.00005608 0.00059327 -0.014601 ]
[ 0.00087223 -0.00015445 -2.39641427]
[-0.22260075 1.74869963 57.12943309]]
```

Part 2: Fundamental Matrix Estimation: Your Images

Your Image: Left Image with Epipolar Lines



Your Image: Right Image with Epipolar Lines



Part 2: Fundamental Matrix Estimation: Your Image

Fundamental Matrix Estimation Result:

```
[[ -0.00005366 -0.0005305 0.83333035]
[ 0.00125642 -0.00048652 -4.93357538]
[ -1.19089577 5.90673263 -468.75343629]]
```

Part 2: Reflection Questions

- 1. Epipoles are the where the camera center projects on the image plane of the other camera. So a rotated image or a zoomed image would have the same camera center compared to the original, and therefore we would not able to calculate a viable epipole, epipolar lines, and fundamental matrix.
- 2. Since the depth of the given point is not known, the point in a given image could correspond to any point on a given line on the other image. This is why a point in one image would correspond to a line in the other image.
- 3. When you have two images with camera centers in the images, you get two epipoles that are displayed within each of these images, so all the epipolar line converge at a point in each of the images, which corresponds to the epipoles.
- 4. Horizontal epipolar lines mean both cameras are facing outward and are side by side. The epipoles are infinitely far away.
- 5. The fundamental matrix is defined up to a scale as given x'Fx=0, and scalar a: x'(aF)x=0, x'Fx=a0, x'Fx=a0, so there are infinitely many matrices that satisfy the equation, but all with the same scale.
- 6. Since the epipole itself lies on the epipolar line, eF= 0, F has a null space which is not the zero vector. So it is rank deficient and is of rank 2.

Part 2: Extra Credit: Fundamental Matrix Song

Reflect on the Fundamental Matrix Song

Link here:

https://www.youtube.com/watch?v= DgGV3l82NTk I have learned that songs are a great way to capture some fundamental points (pun intended) of a given topic. Specifically I've leaned properties of the fundamental matrix, such as having 7 degrees of freedom, being of rank 2, x'Fx=0, null(F)=e, null(Fh)=e', all epipolar lines pass through epipoles, and three views allow trifocal tensors with additional properties. Furthermore, I aspire to be as cool as the creators of this song, one day.

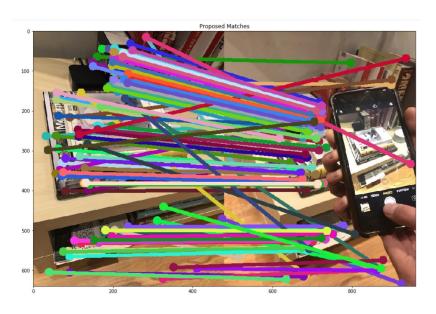
Part 3: RANSAC Iterations Questions

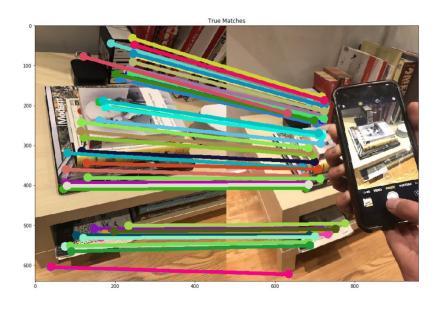
Delete the questions and type your answers to the three RANSAC Iterations questions from the jupyter notebook below:

- 14 iterations are needed
- 2. If we used 18 points, we would need 42 iterations
- 3. If our dataset had a 70% point correspondence accuracy, we would need 167 iterations

Part 3: RANSAC Inlier Matches

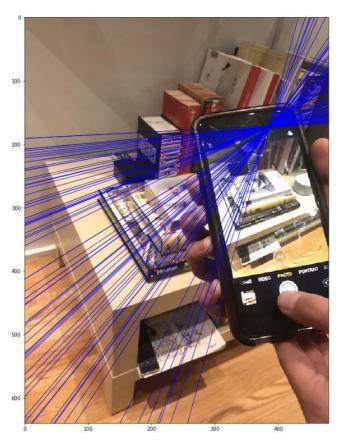
Paste two sets of images: displaying matches loaded from disk; and inlier matches after RANSAC





Part 3: RANSAC Implementation





Tests

<Provide a screenshot of the results when you run `pytest` from the unit tests directory with your final code implementation (note: we will re-run these tests).>

```
(base) C:\Users\tyco9\OneDrive\Documents\Spring 2020\CS 6476 Computer Vision\ps2 release v1\proj2 unit tests>pytest
 platform win32 -- Python 3.7.3, pytest-4.3.1, py-1.8.0, pluggy-0.9.0
rootdir: C:\Users\tyco9\OneDrive\Documents\Spring 2020\CS 6476 Computer Vision\ps2 release v1, inifile:
plugins: remotedata-0.3.1, openfiles-0.3.2, doctestplus-0.3.0, arraydiff-0.3
collected 20 items
part1 unit test.py .....
                                                                                                25%]
test essential matrix decomposition.py ...
test fundamental matrix.py
                                                                                               80%]
test_ransac.py ....
                                                                                              [100%]
                             ======== warnings summary ======
proj2 unit tests/test ransac.py::test ransac find inliers
 C:\Users\tyco9\OneDrive\Documents\Spring 2020\CS 6476 Computer Vision\ps2 release v1\proj2 unit tests\test ransac.py:4
6: DeprecationWarning: elementwise comparison failed; this will raise an error in the future.
   assert outliers not in inliers
  Docs: https://docs.pytest.org/en/latest/warnings.html
        (base) C:\Users\tyco9\OneDrive\Documents\Spring 2020\CS 6476 Computer Vision\ps2 release v1\proj2 unit tests>
```

Conclusions

In this project, I have learned how to map 3-d real world points into 2-d image coordinates using the projection matrix, I have learned the relation between the camera center, rotation matrix, intrinsic matrix, and projection matrix, I have learned how to derive the fundamental matrix using corresponding points among two images, I have learned the relation between the fundamental and essential matrix. I have leaned how to calculate the number of RANSAC iteration needed to produce an inlier set with a given probability, and I have learned how to estimate and fine tune a fundamental matrix using the RANSAC algorithm. This was a very informative assignment, and it has helped me grasp some of the concepts that were confusing during lecture. I did not face many challenges, as the assignment was clear and provided helpful instructions.