

Problem 1: Homography Estimation

1. Sample k Correspondences

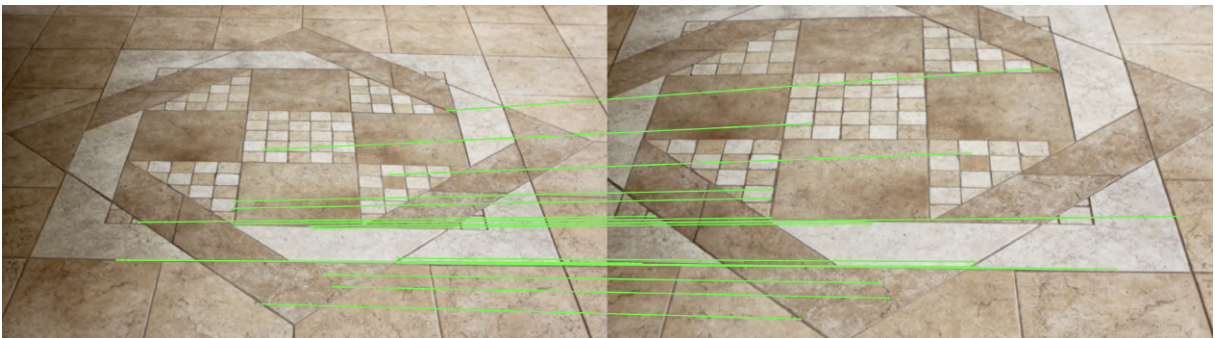
Image Pair 1 - Result



(a) keypoint sample, $k=4$

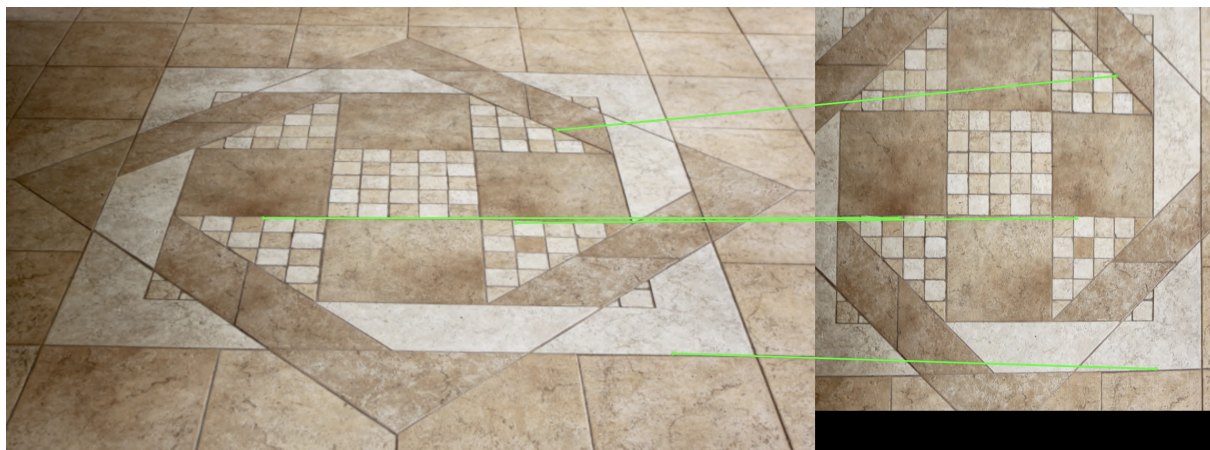


(b) keypoint sample, $k=8$

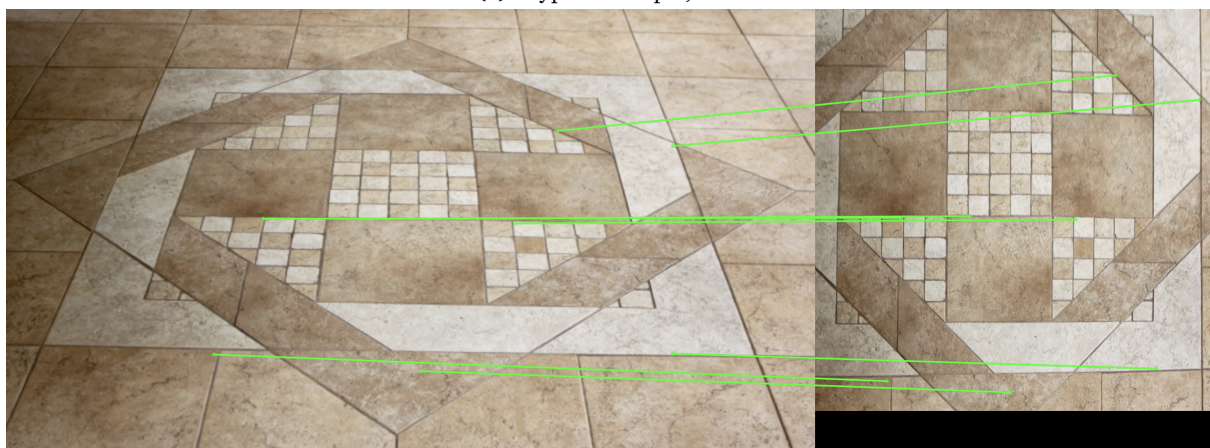


(c) keypoint sample, $k=20$

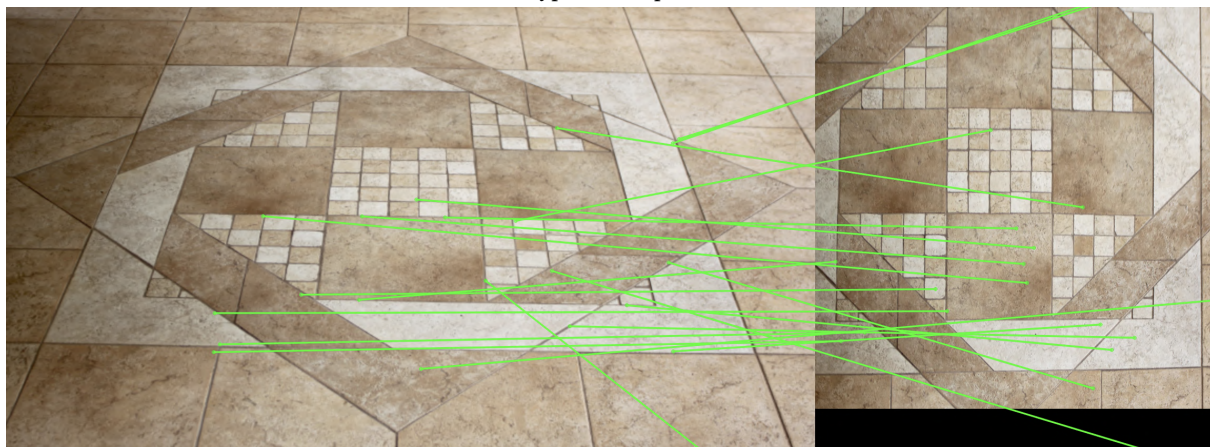
Image Pair 2 - Result



(a) keypoint sample, $k=4$



(b) keypoint sample, $k=8$



(c) keypoint sample, $k=20$

2. Methodology

1. Feature selection

- Use SIFT to find the local features.
- Use BFMatcher to find the correspondences.

2. Remove the outliers

- Ratio test: Remove the the keypoints that greater than 0.75.
- Random sample consensus(RANSAC): Use linear regressor to further remove the outliers. In the following section, we will see that RANSAC dramatically decrease the error.

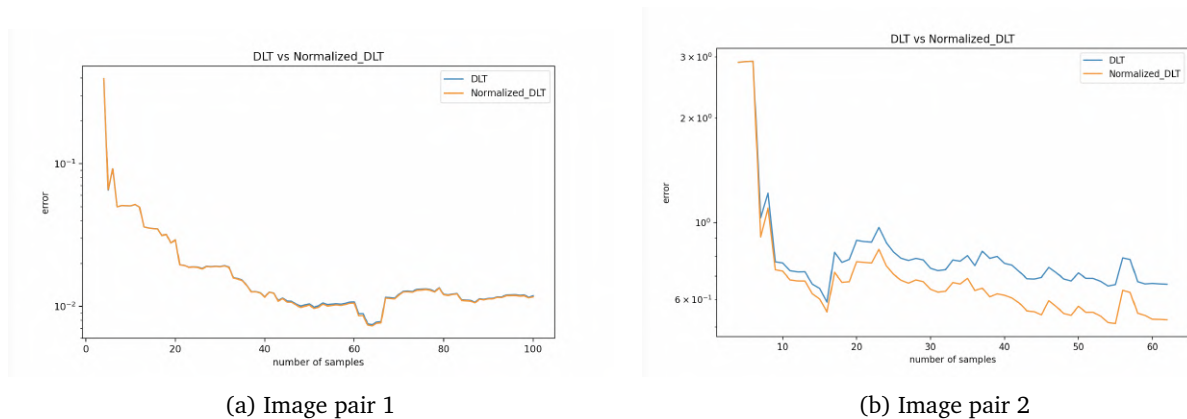
3. Homography estimation

- Direct linear transform(DLT): DLT is used to calculate the transformation matrix H.
- Normalized DLT: Normalize the keypoints (translate + scale) can be helpful in the derivation of the transformation matrix H.

4. Find the reprojection error

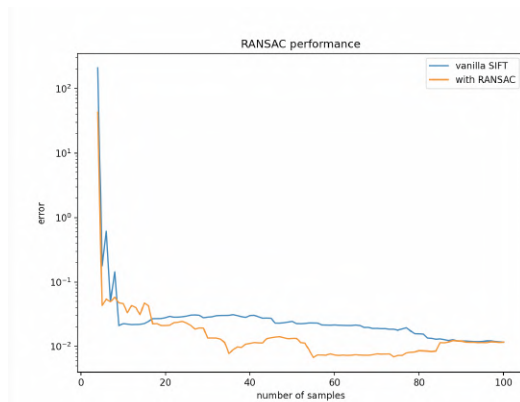
- Use groundtruth data to calculate the error(2-norm MSE).

3. Comparison between DLT and Normalized-DLT methods

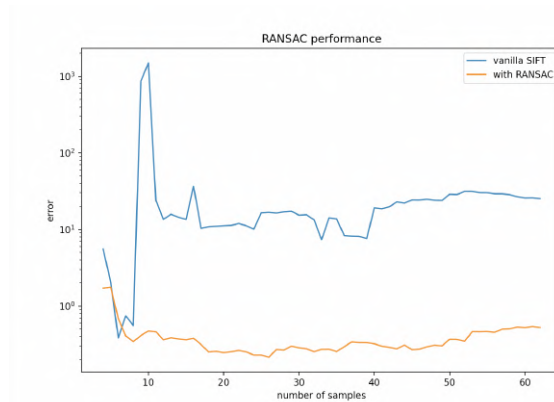


- As we can see, DLT method is almost as good as Normalized-DLT in the first image pair. However, normalized-DLT method is way better than DLT in the second pair images which is hugely different in the perspective.
- Visualization plots are shown as above in the log scale.

4. BONUS method: Random sample consensus(RANSAC)



(a) Image pair 1



(b) Image pair 2

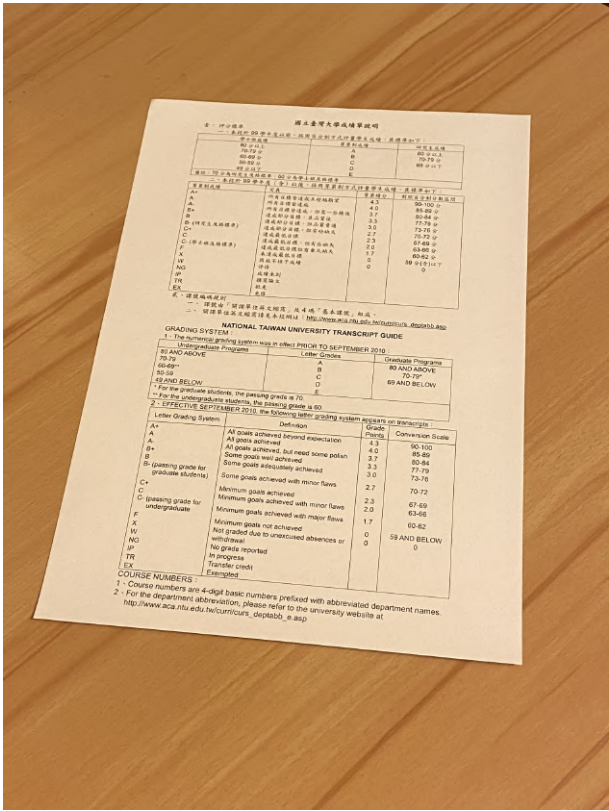
- We can manually remove some outliers or choose some points that is easy for the feature detector to predict correspondences. However, the result(error) is still not good enough by doing that. To further improve the performance, I use RANSAC with a linear regressor to remove the outliers.
- As we can see, RANSAC can dramatically decrease the error(error < 10 for $k = 4$, error < 1 for $k > 4$), especially in the second image pair which outliers presents frequently.
- Visualization plots are shown as above in the log scale.

5. Further Discussion

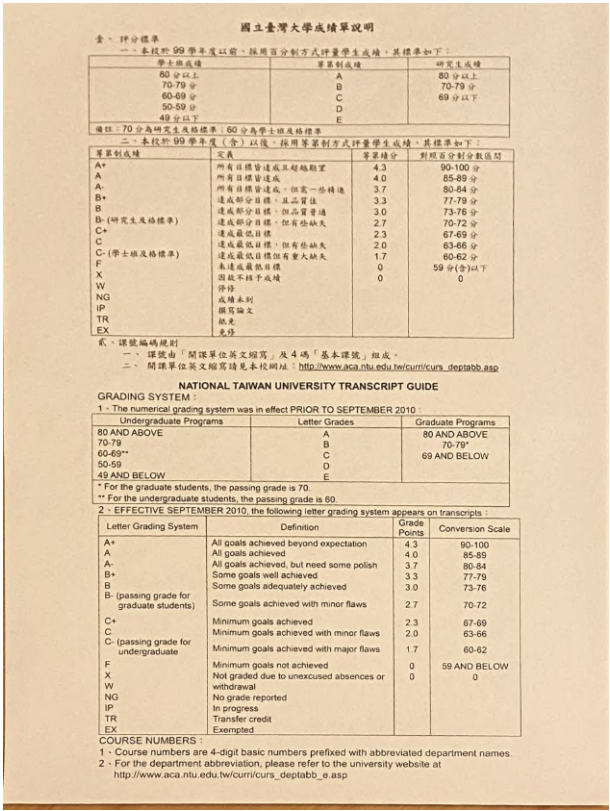
- As we see in the previous paragraph, DLT+SIFT can obtain good result for the application that perspective of two images is close enough(e.g. Image pair 1). The result converges easy by choosing more keypoints.
- On the other hand, if the perspective is hugely different(e.g. Image pair 2), technique like RANSAC can be very helpful. Remove the outlier not only improve the performance directly, it also makes Normalized-DLT more robust. Since outlier can be harmful to the process of normalization. (Normalization is useless if there are many outliers.)

Problem 2: Document rectification

1. Result



(a) source image



(b) output image

2. Methodology

1. Preprocess the image: gray, blur etc.
2. Use Canny edge detector to find the edges.
3. Use dilate, erode to minimize the noise.
4. Find the contour of the input image.
5. Find the corners of the input image.
6. Correspond the corners of the input image to the corners of result image.
7. Find the homography. (In the same way as problem1)
8. Transform the image, and do bilinear interpolation.

3. How to Run the code

1. For problem 1, use the following line to execute the code.
`python 1.py images/1-0.png images/1-2.png groundtruth_correspondences/correspondence_02.npy`
2. For problem 2, use 'python 2.py' to execute the code, the program will automatically choose 4-corners and the output will present in the folder with the name tmp.jpg.
3. Environment

```
python==3.7
numpy==1.19.5
opencv_python==4.6.0.66
scikit_image==0.15.0
skimage==0.0
```

4. Other people's contribution

1. Thanks to classmate 信承 to give me the idea that draw the matching points to visualize the correspondence. That helps me a lot in choosing good matching points.
2. Thanks to friend(not in this class) 雅晴 to give me the idea on how to find the corner of image.