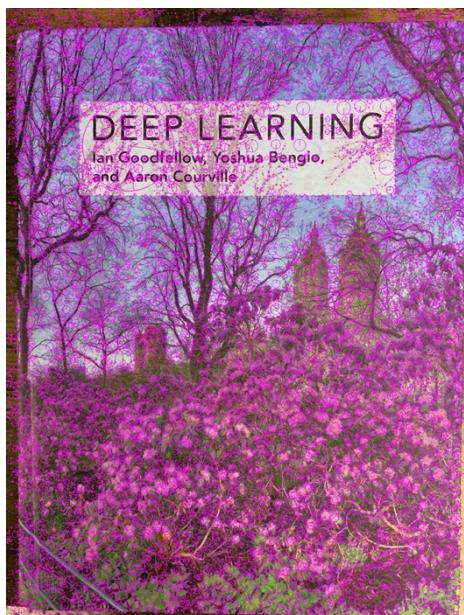
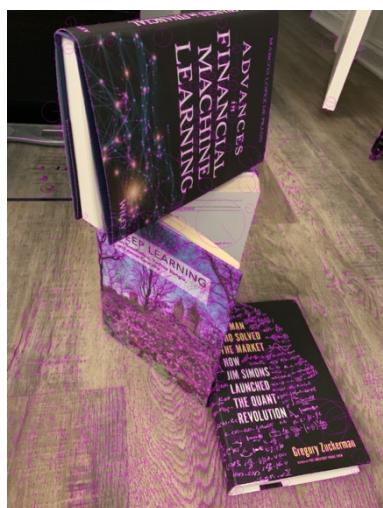
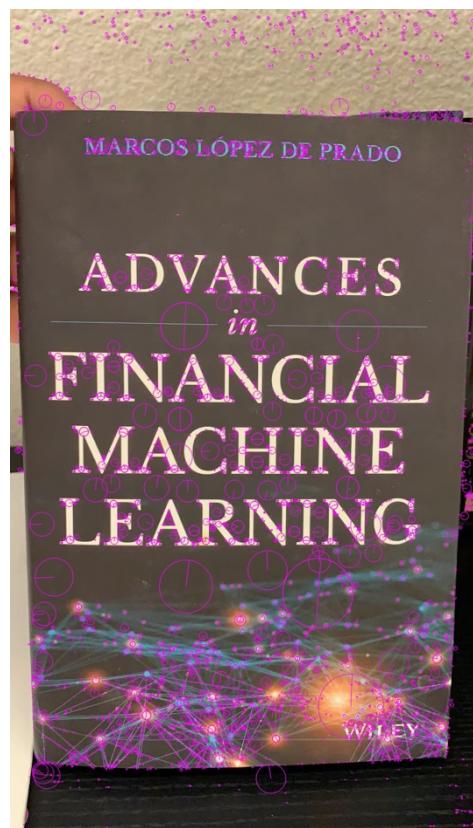
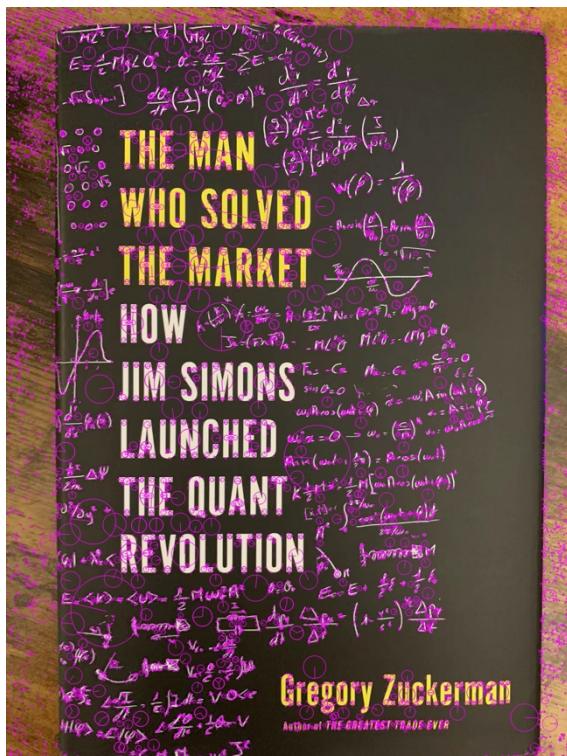


1. A brief description of the programs you implement.
  1. We iterate through each combination of source image and destination image
  2. Using SIFT operation find the key points and descriptors
  3. Draw the keypoints on the image ( use the arg  
`cv.DRAW_MATCHES_FLAGS_DRAW_RICH_KEYPOINTS` to get the directions)
  4. Use the bruteforce knn matches
  5. Filter out the good matches using the ratio test (0.75)
  6. Sort the good matches matches found based on the distance
  6. Draw the top 20 matches on the images
  7. Find homography matrix and mask using `findhomography` function with RANSAC
  8. Find the number of inliers from the mask obtained.
  9. Find distances between dest keypoints and projected points after applying the homography matrix operation for projection, sort the points according the differences (error)
  10. Draw the top 10 matches with min error.
2. Show the results of intermediate steps as listed in the descriptions above.
  - a. SIFT features: show the detected features overlaid on the images (*both the locations and directions, not 128-d vectors*). Also give out *the number of features detected* in each image.

No of Key Points found using SIFT in image `src_0.jpg` 5259  
No of Key Points found using SIFT in image `src_1.jpg` 42457  
No of Key Points found using SIFT in image `src_2.jpg` 13004

No of Key Points found using SIFT in image `dst_0.jpg` 49979  
No of Key Points found using SIFT in image `dst_1.jpg` 10625





- b. Graphically show the top-20 scoring matches found by the matcher before the RANSAC operation. Provide statistics of how many matches are found for each image pair.

Number of good matches found for src\_0.jpg dst\_0.jpg 219

Number of good matches found for src\_0.jpg dst\_1.jpg 529

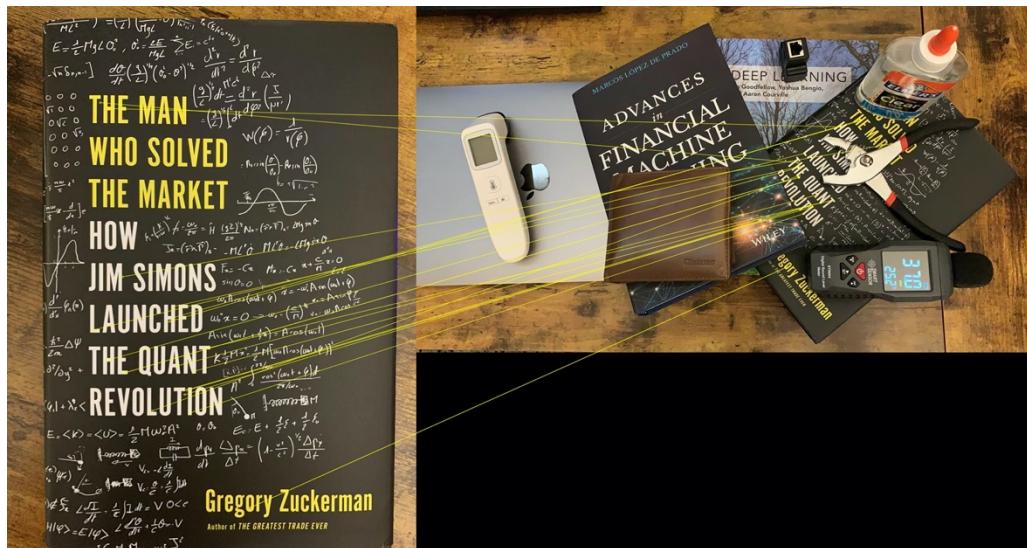
Number of good matches found for src\_1.jpg dst\_0.jpg 155

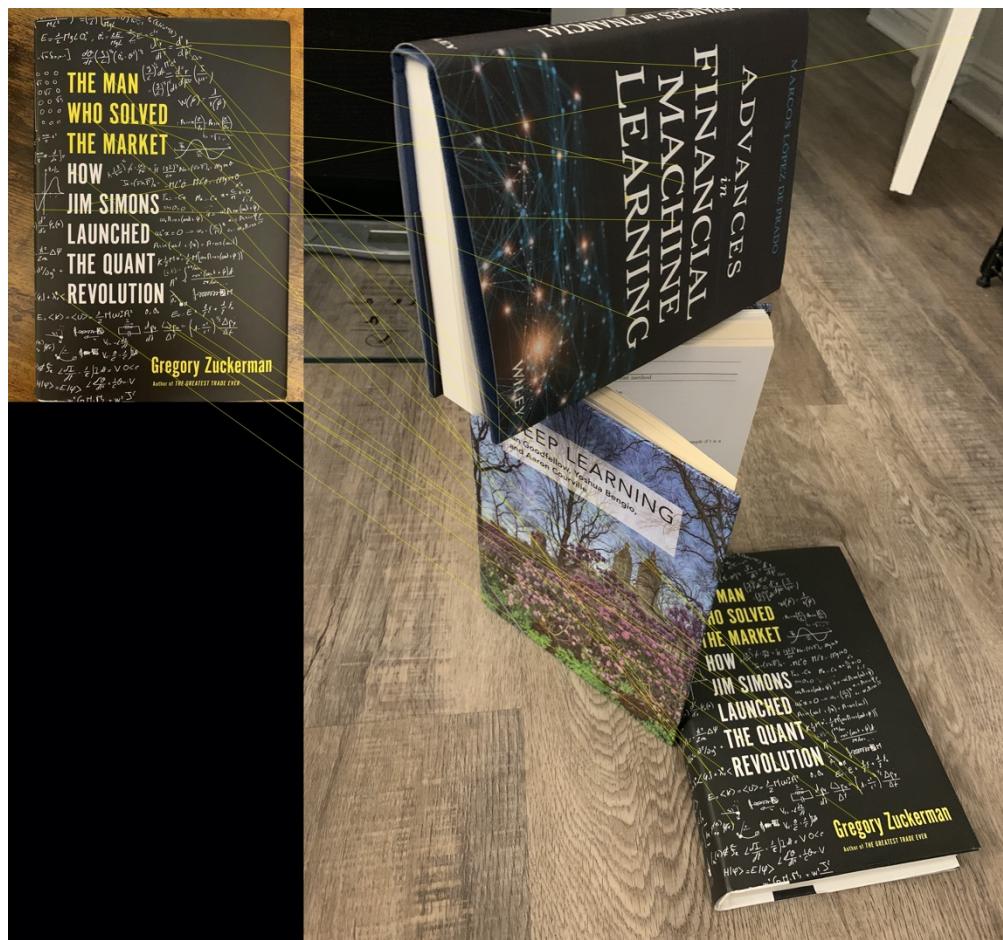
Number of good matches found for src\_1.jpg dst\_1.jpg 606

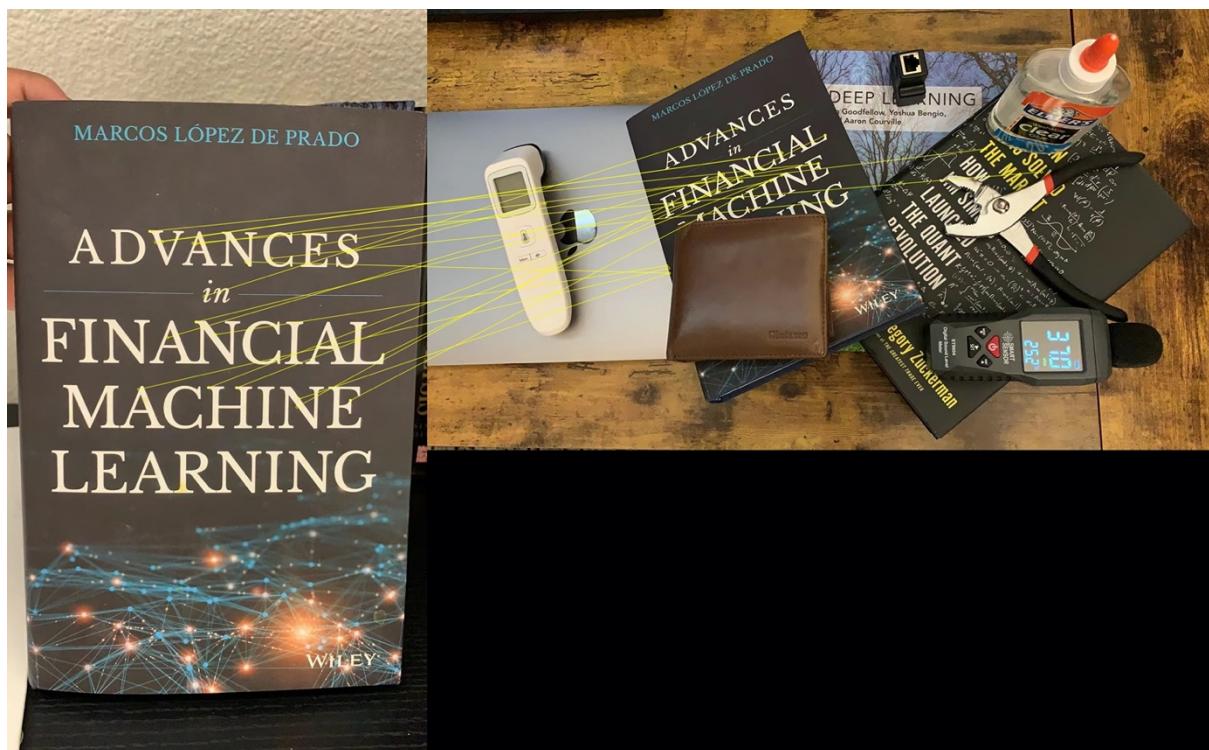
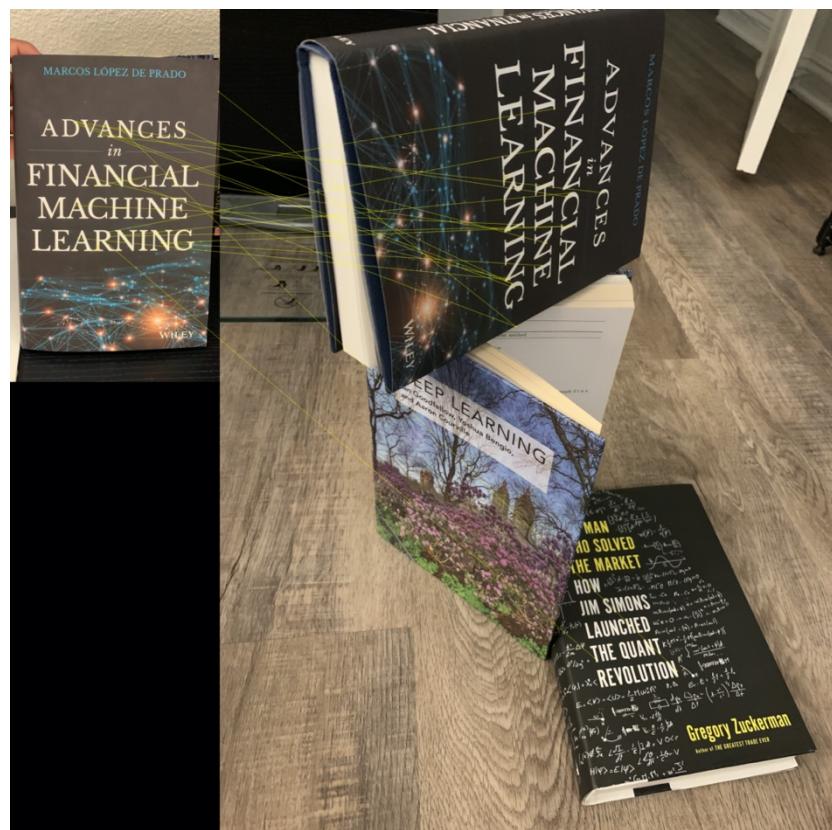
Number of good matches found for src\_2.jpg dst\_0.jpg 1241

Number of good matches found for src\_2.jpg dst\_1.jpg 1047









- c. Show total number of inlier matches after homography estimations. Also show top-10 matches that have the minimum error between the projected source keypoint and the destination keypoint. (Hint: check the mask value returned by the function estimating the homography)

No of inliers found for image combination src\_0.jpg dst\_0.jpg 48

No of inliers found for image combination src\_0.jpg dst\_1.jpg 243

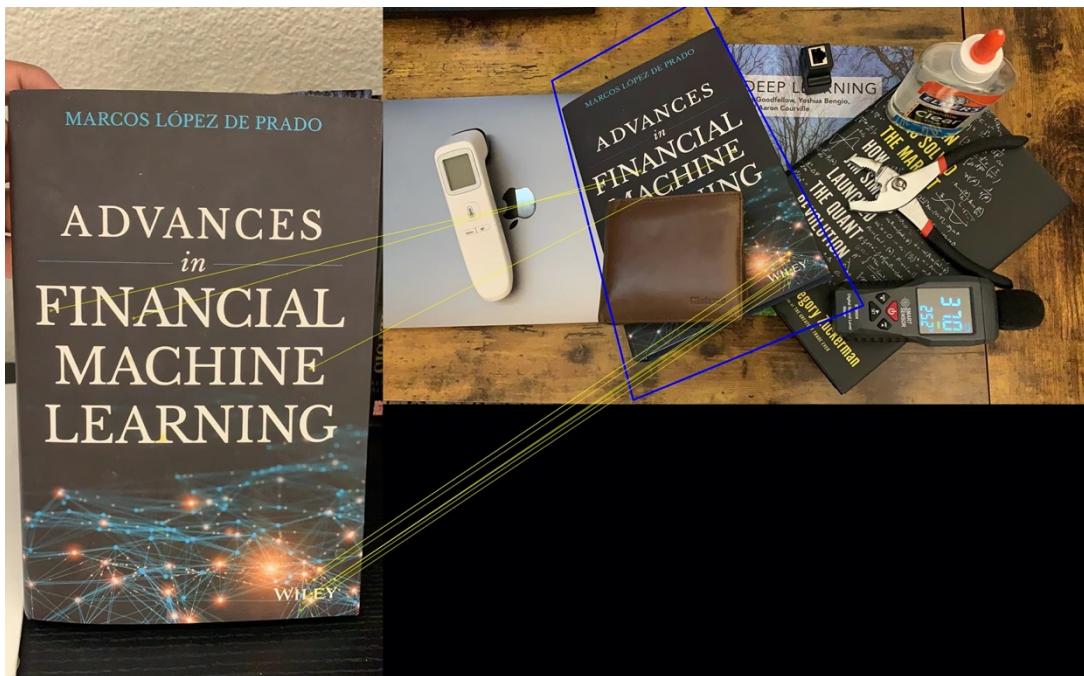
No of inliers found for image combination src\_1.jpg dst\_0.jpg 30

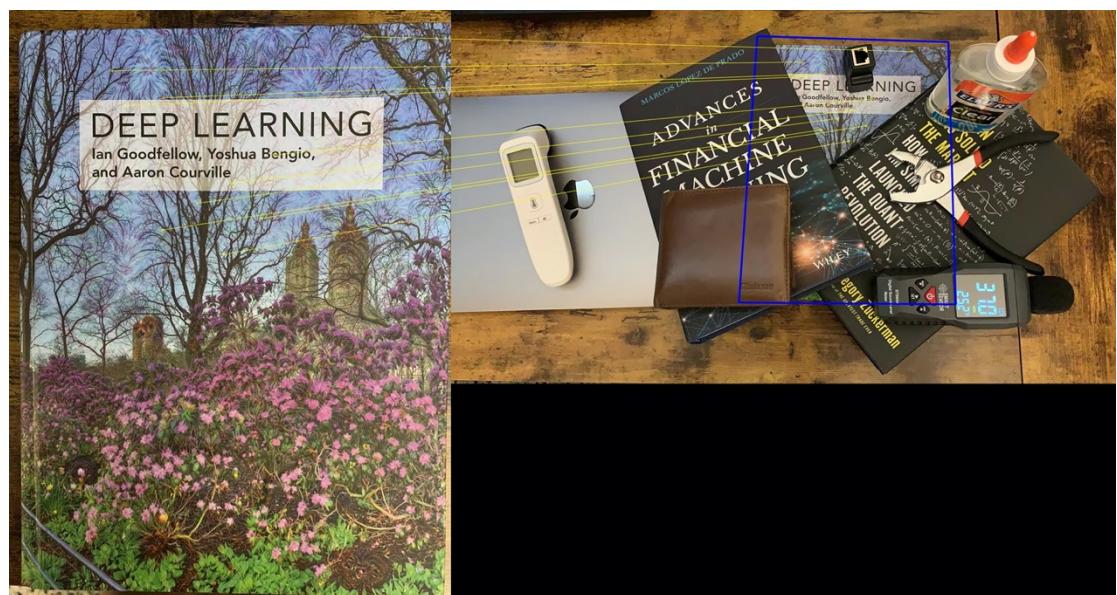
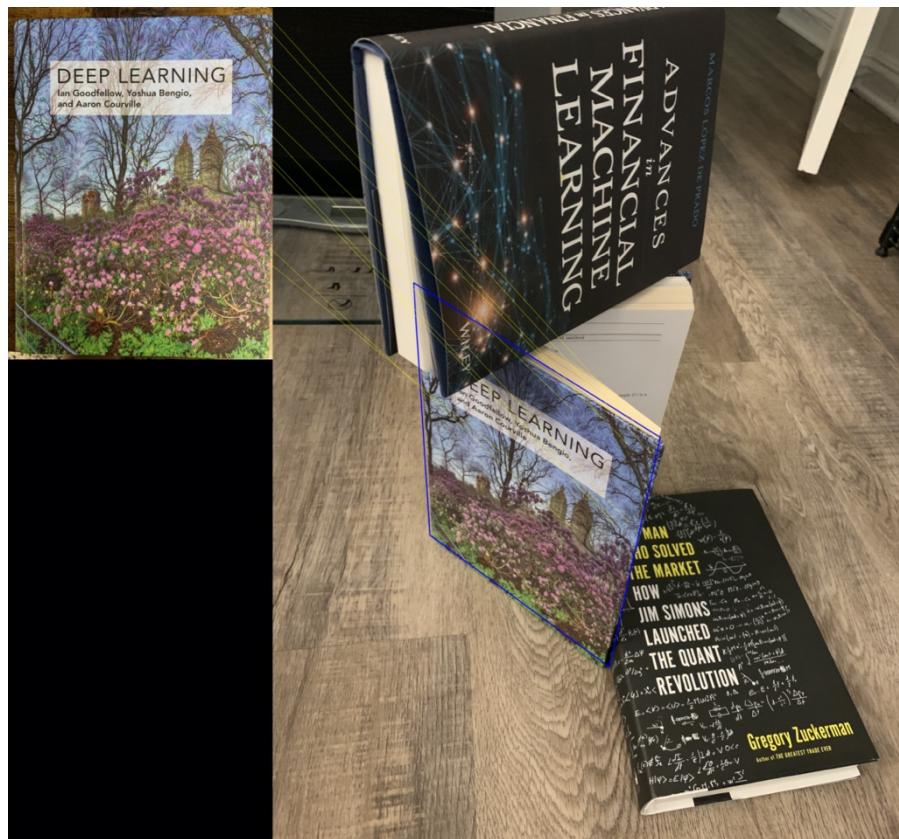
No of inliers found for image combination src\_1.jpg dst\_1.jpg 427

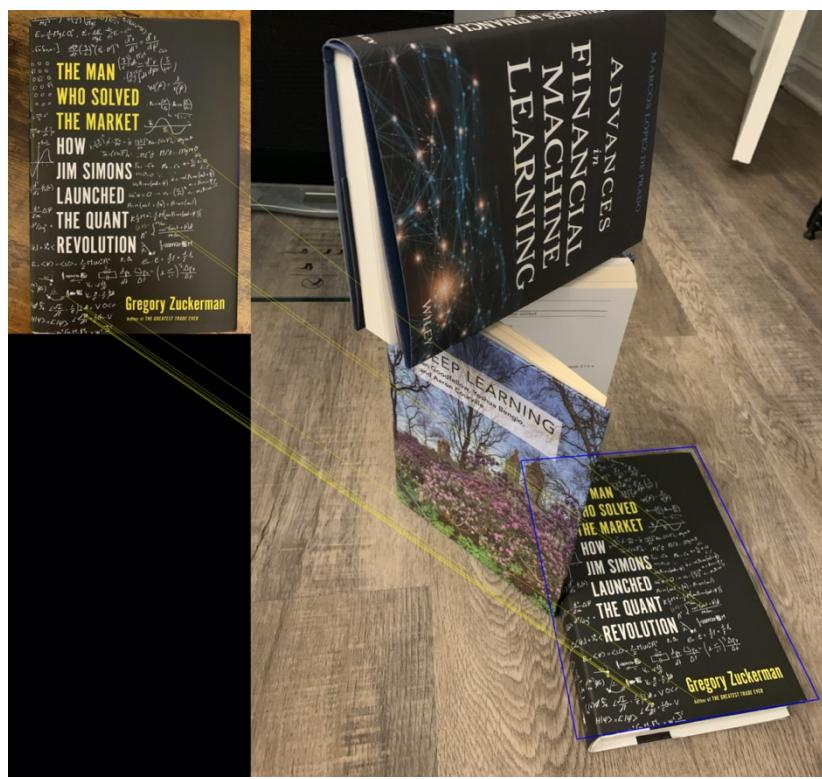
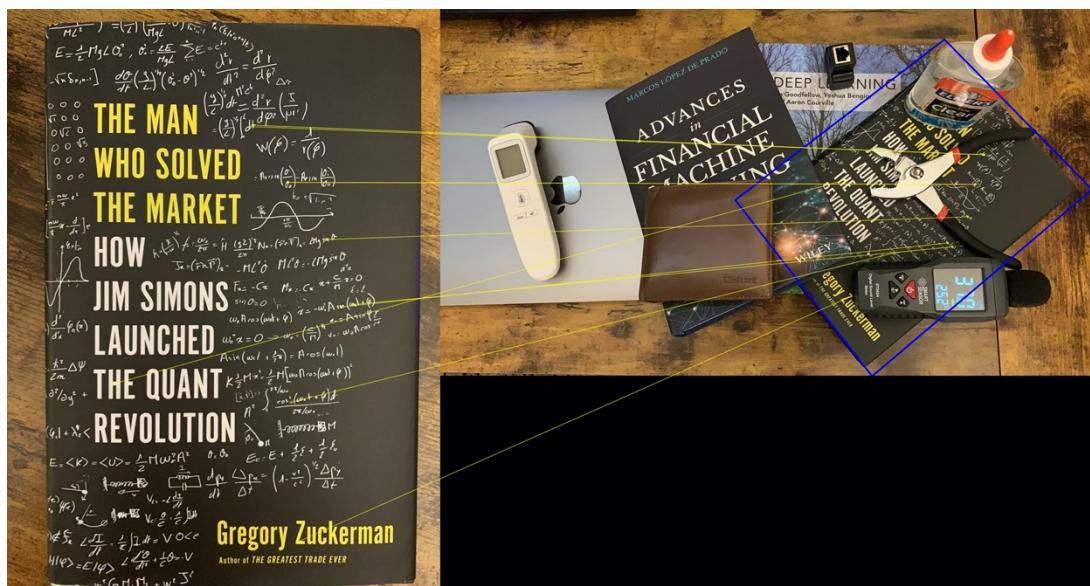
No of inliers found for image combination src\_2.jpg dst\_0.jpg 635

No of inliers found for image combination src\_2.jpg dst\_1.jpg 402









d. Output the computed homography matrix. (The built-in homography finder also applies a non-linear optimization step at the end; you can ignore or disable this step if you wish.)

Homography Matrix for src\_0.jpg dst\_0.jpg

```
[[ 3.50941630e-01 -1.20062098e+00  2.32703002e+03]
 [ 1.37673968e+00  7.79759844e-02 -5.42872173e+01]
 [ 1.98417755e-04 -2.74249648e-04  1.00000000e+00]]
```

Homography Matrix for src\_0.jpg dst\_1.jpg

```
[[ 3.45488480e-01  4.16982644e-03  4.59782936e+02]
 [-2.17602511e-01  2.92198177e-01  2.00664455e+02]
 [-5.64259651e-05 -1.71364131e-04  1.00000000e+00]]
```

Homography Matrix for src\_1.jpg dst\_0.jpg

```
[[ 7.84292397e-01  2.00187628e-01  6.79394639e+02]
 [ 4.20863442e-01  1.23376587e+00  1.33159558e+03]
 [-7.94313094e-05  2.03188167e-04  1.00000000e+00]]
```

Homography Matrix for src\_1.jpg dst\_1.jpg

```
[[ 4.52379694e-01 -8.93854911e-02  8.84559291e+02]
 [ 1.46736281e-02  3.93965833e-01  7.54379109e+01]
 [ 1.34278389e-05 -6.97134468e-05  1.00000000e+00]]
```

Homography Matrix for src\_2.jpg dst\_0.jpg

```
[[ 8.19927361e-01 -1.36280812e-01  1.41908453e+03]
 [ 4.10725717e-02  1.67992565e-01  2.36921943e+03]
 [ 4.29394197e-05 -1.80772576e-04  1.00000000e+00]]
```

Homography Matrix for src\_2.jpg dst\_1.jpg

```
[[ 2.11957612e-01 -3.92439093e-01  1.48597026e+03]
 [ 3.10456637e-01  2.58301888e-01  8.99926830e+01]
 [-6.69603689e-05 -3.47699306e-05  1.00000000e+00]]
```

3. An qualitative analysis of your test results: how well does the method work? Does it work equally well on the different examples? If not, why might the performance be better in one case then the other? Note that the main goal is to locate the objects in the given images. We could transform the entire object image and overlay on the target image, using the computed homography, but this is not asked for. It will suffice to make your judgment based on results of feature matching (after homography computation).

The algorithm works well to find the object in all cases. Since sufficient number of features get matched to locate the object.

However if we notice how all feature keypoints are being matched. There are more outliers in some combinations of images.

With respect to the orientation of the query image:

In cases where the face of the book in the query image is not fully visible in the train image. i.e the book is slightly moved in the 3D space about x ,y or z axis (roll,pitch,yaw) the feature keypoints get mismatched.

With respect to occlusion in the train image:

In cases where the keypoints are occluded as well we can see that there are mismatches