

GitHub: <https://github.com/r3ki3g/Image-processing-fundamentals/tree/main/assignment-2>

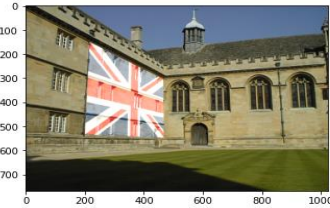
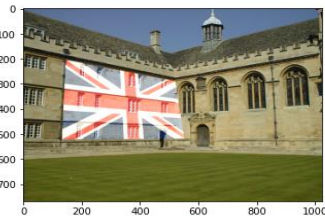
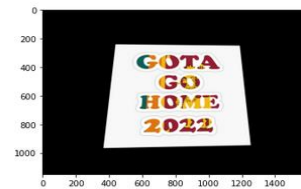
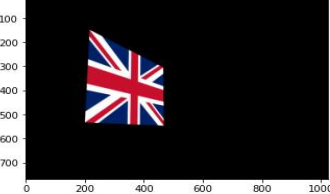
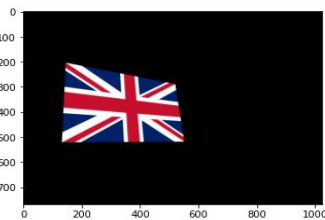
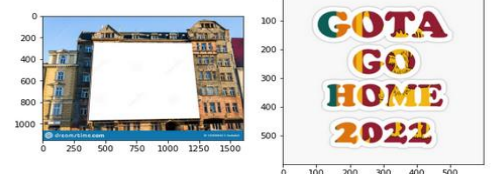
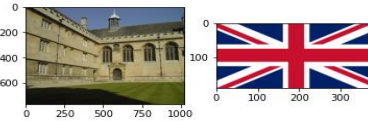
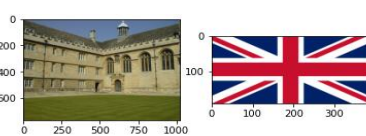
(All the codes and full python notebook is in the GitHub/)

Question 02

```
1 wadham = r"D:\ENTC\SEM_4\EN2550 - Fundamentals of Image Processing and Machine Vision\~images\ass2\wadham\001.jpg"
2 flag = r"D:\ENTC\SEM_4\EN2550 - Fundamentals of Image Processing and Machine Vision\~images\ass2\Flag_of_the_United_Kingdom
3
4 imgWadham = cv.cvtColor(cv.imread(wadham,1),cv.COLOR_BGR2RGB)
5 imgFlag = cv.cvtColor(cv.imread(flag,1),cv.COLOR_BGR2RGB)
6
7 fig, ax = plt.subplots(1,2)
8 ax[0].imshow(imgWadham)
9 ax[1].imshow(imgFlag)
10 plt.show()
11 X1 = np.array([[0,0],
12               [0,192],
13               [383,0],
14               [383,192]])
15
16 X2 = np.array([[146,205],
17               [133,522],
18               [521,290],
19               [552,519]])
20
21 H, mask = cv.findHomography(X1, X2, cv.RANSAC, 5.0)
22
23 w=1024
24 h=768
25
26 warped = cv.warpPerspective(imgFlag, H, (w, h))
27
28 plt.imshow(warped)
29 plt.show()
30
31 superImposed = cv.addWeighted(imgWadham,1,warped,0.7,1)
32
33 plt.imshow(superImposed)
34 plt.show()
```

Important points:

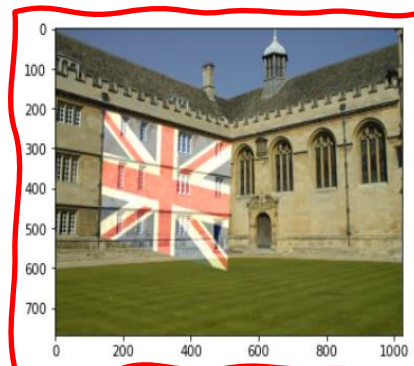
1. "Mouse click" and getting the coordinates was not possible because my notebook (Jupyter) froze every time I tried to do so. So I had to hardcode the coordinates.
2. First hardcoded the 4 corner coordinates of the flag then the 4 corner coordinates of the spot we need to super impose.
3. Then calculated the matrix using the cv.findHomography method and warped the flag as needed. Finally super imposed with fine weights using cv.addWeighted.



Experimented with finding the homographic matrix myself

However what I got was the affine transformation with [0 0 1] as the bottom row. Therefore the warping did not fit the wall.

```
In [43]: 1 X2 = np.array([[146],
2               [133],
3               [521]])
4 Y2 = np.array([[205],
5               [522],
6               [290]])
7 Ones2 = np.array([[1],
8               [1],
9               [1]])
10 M1 = np.array([[0,0,1],
11               [0,192,1],
12               [383,0,1]])
13 M1_inv = np.linalg.inv(M1)
14
15 Hr1 = M1_inv @ X2
16 Hr2 = M1_inv @ Y2
17 Hr3 = M1_inv @ Ones2
18 H = np.array([Hr1,Hr2,Hr3])
19
20 print(H)
21 w=1024
22 h=768
23
24 warped = cv.warpPerspective(imgFlag, H, (w, h))
25 superImposed = cv.addWeighted(imgWadham,1,warped,0.5,1)
26 plt.imshow(superImposed)
27 plt.show()
```



```
[[[ 9.79112272e-01]
  [-6.77083333e-02]
  [ 1.46000000e+02]]

 [[ 2.21932115e-01]
  [ 1.65104167e+00]
  [ 2.05000000e+02]]

 [[ 0.00000000e+00]
  [ 0.00000000e+00]
  [ 1.00000000e+00]]]
```

Question 03

```
14 def siftMatchFinder(img1,img2):
15     sift = cv.SIFT_create()
16     im1, des_1 = sift.detectAndCompute(img1, None)
17     im2, des_2 = sift.detectAndCompute(img2, None)
18     T1 = cv.BFMatcher(cv.NORM_L1, crossCheck = True)
19     Match = T1.match(des_1, des_2)
20     sortMatch = sorted(Match, key = lambda x:x.distance)
21     img4 = cv.drawMatches(img1, im1, img2, im2, sortMatch[:100], img2, flags =2)
22     fig, ax = plt.subplots(1,1, figsize = (18, 18))
23     ax.imshow(img4)
24     ax.axis('off')
25     plt.show()
26     return Match,[im1,im2]
```

It was impossible to find matches between 1.ppm and 5.ppm directly. So idea was to find the homography between 1.ppm and 2.ppm then 2.ppm and 3.ppm, and like so finally 4.ppm and 5.ppm. Then we can pre-multiply the homographies and get the composition of homography which is equivalent to homography between 1.ppm and 5.ppm

siftMatchFinder() function is defined to calculates sift matches between images. (It is used multiple times later in the code)

```
or corr in corres:
    pts1 = np.matrix([corr.item(0), corr.item(1), 1])
    pts2 = np.matrix([corr.item(2), corr.item(3), 1])

A2 = [0, 0, 0, -pts2.item(2) * pts1.item(0), -pts2.item(2) * pts1.item(1), -pts2.item(2) * pts1.item(2), pts2.item(2) * pts1.item(0), pts2.item(2) * pts1.item(1), pts2.item(2) * pts1.item(2)]
A1 = [-pts2.item(2) * pts1.item(0), -pts2.item(2) * pts1.item(1), -pts2.item(2) * pts1.item(2), 0, 0, 0, pts2.item(2) * pts1.item(0), pts2.item(2) * pts1.item(1), pts2.item(2) * pts1.item(2)]
```

Creating the data matrix to find the least error transform with RANSAC algorithm. Using matrix and Numpy library helps do the calculations fast and ends up letting us do much more iterations for the RANSAC.

```
6 def getDistanceBtwMat(corres, H):
7     pts1 = np.transpose(np.matrix([corres[0].item(0), corres[0].item(1), 1]))
8     es_pts1 = np.dot(H, pts1)
9     es_pts2 = (1/es_pts1.item(2)) * es_pts1
10
11     pts2 = np.transpose(np.matrix([corres[0].item(2), corres[0].item(3), 1]))
12     err = pts2 - es_pts2
13     return np.linalg.norm(err)
14
15 def HormographyFromRANSAC(corres, thresh):
16     final = None
17     max_inliers = []
18     for i in range(1000):
19         corres1 = corres[random.randrange(0, len(corres))]
20         corres2 = corres[random.randrange(0, len(corres))]
21         rand_four = np.vstack((corres1, corres2))
22         corres3 = corres[random.randrange(0, len(corres))]
23         rand_four = np.vstack((rand_four, corres3))
24         corres4 = corres[random.randrange(0, len(corres))]
25         rand_four = np.vstack((rand_four, corres4))
26         H = calc_Homography(rand_four)
27         inliers = []
28         for i in range(len(corres)):
29             d = getDistanceBtwMat(corres[i], H)
30             if d < 5:
31                 inliers.append(corres[i])
32         if len(inliers) > len(max_inliers):
33             max_inliers = inliers
34             final = H
35         if len(max_inliers) > (len(corres) * thresh):
36             break
37     return final, max_inliers
38
```

getDistanceBtwMat() function calculated the geometric distance between given vectors. It is used to see how the destination points are deviated from the required positions in the image.

By selecting the case of minimal eigen values, in each iteration we can calculate the homography matrix. The result is shown here:



```
69
70 M1,K1=siftMatchFinder(img1,img2)
71 corres_lis1=corr_list(M1,K1)
72 corres1 = np.matrix(corres_lis1)
73 H1, inliers1 = HormographyFromRANSAC(corres1, 0.6)
74
75 M2,K2=siftMatchFinder(img2,img3)
76 corres_lis2=corr_list(M2,K2)
77 corres2 = np.matrix(corres_lis2)
78 H2, inliers2 = HormographyFromRANSAC(corres2, 0.6)
79
80 M3,K3=siftMatchFinder(img3,img4)
81 corres_lis3=corr_list(M3,K3)
82 corres3 = np.matrix(corres_lis3)
83 H3, inliers3 = HormographyFromRANSAC(corres3, 0.6)
84
85 M4,K4=siftMatchFinder(img4,img5)
86 corres_lis4=corr_list(M4,K4)
87 corres4 = np.matrix(corres_lis4)
88 H4, inliers4 = HormographyFromRANSAC(corres4, 0.6)
89
90 H = H4 @ H3 @ H2 @ H1
91 print("hormography compositioned", H)
92
93 DST1 = cv.warpPerspective(img1, H, ((img5.shape[1]), img5.shape[0]))
```

```
hormography compositioned [[ 5.65738181e-01  5.79922504e-02  2.30593711e+02]
 [ 1.91483499e-01  1.13184499e+00 -1.56137084e+01]
 [ 4.04731077e-04 -4.21852199e-05  9.97146955e-01]]
```

Final stitched image and warped version are shown here



Question 01

```

1 #RANSAC circle finder
2 thres = 2
3 randomSamplingCount = 10000 #100000
4 # three sets of random points are selected from the given point set
5 #each time , one from each 3 groups will be selected (in the order) as the 3 poi
6 pointSet1_indexes = np.random.randint(0,len(X),(randomSamplingCount))
7 pointSet2_indexes = np.random.randint(0,len(X),(randomSamplingCount))
8 pointSet3_indexes = np.random.randint(0,len(X),(randomSamplingCount))
9 #points can be non-distinct -- fix it later -----
10
11 x1 = np.zeros(randomSamplingCount)
12 y1 = np.zeros(randomSamplingCount)
13 x2 = np.zeros(randomSamplingCount)
14 y2 = np.zeros(randomSamplingCount)
15 x3 = np.zeros(randomSamplingCount)
16 y3 = np.zeros(randomSamplingCount)
17 for i in range(randomSamplingCount):
18     x1[i] = X[pointSet1_indexes[i]][0]
19     y1[i] = X[pointSet1_indexes[i]][1]
20     x2[i] = X[pointSet2_indexes[i]][0]
21     y2[i] = X[pointSet2_indexes[i]][1]
22     x3[i] = X[pointSet3_indexes[i]][0]
23     y3[i] = X[pointSet3_indexes[i]][1]
24
25 print("debug x1 shape",x1.shape)
26
27 #These variables are needed to calculate the centre coords and the radii of cir
28 A2 = -2*(x1-x2)
29 B2 = -2*(y1-y2)
30 C2 = x1**2-x2**2+y1**2-y2**2
31 A3 = -2*(x1-x3)
32 B3 = -2*(y1-y3)

```

RANSAC algo for circle detection

```

35 x0Set = (B2*C3-B3*C2)/(B3*A2-B2*A3)
36 y0Set = (A2*C3-A3*C2)/(A3*B2-A2*B3)
37 rSet = np.sqrt((x1-x0Set)**2 + (y1-y0Set)**2)
38 countSet = np.zeros(len(x0Set))
39
40 print("debug x0Set shape",x0Set.shape)
41 #counting the inliers
42 for i in range(len(x0Set)):
43     x0 = x0Set[i]
44     y0 = y0Set[i]
45     r = rSet[i]
46     x = X[:,0]
47     y = X[:,1]
48     dSet = abs(np.sqrt((x-x0)**2+(y-y0)**2)-r) #d
49     close = dSet < thres
50     count = np.sum(close)
51     countSet[i] = count
52
53 mostMatchingIndex = np.argmax(countSet)
54 x0Best = x0Set[mostMatchingIndex]
55 y0Best = y0Set[mostMatchingIndex]
56 rBest = rSet[mostMatchingIndex]

```

Calculating inlier count for each random sample of size 3, then finding the maximum case as the solution

```

89 def bestFitCircleFinder(x,y):
90     x_m = np.mean(x)
91     y_m = np.mean(y)
92
93     def calc_R(xc, yc):
94         #calculate the distance of each 2D points from the center (xc, yc)
95         return np.sqrt((x-xc)**2 + (y-yc)**2)
96
97     def f_2(c):
98         # calculate the algebraic distance between the data points and the
99         Ri = calc_R(*c)
100         return Ri - np.mean(Ri)
101
102     center_estimate = x_m, y_m
103     center_2, ier = optimize.leastsq(f_2, center_estimate)
104
105     xc, yc = center_2
106     Ri = calc_R(*center_2)
107     R = np.mean(Ri)
108
109     return [(xc,yc,R)]
110
111 bestFitCircle = bestFitCircleFinder(X_inliers_x,X_inliers_y)
112 print("bestFitCircle",bestFitCircle)

```

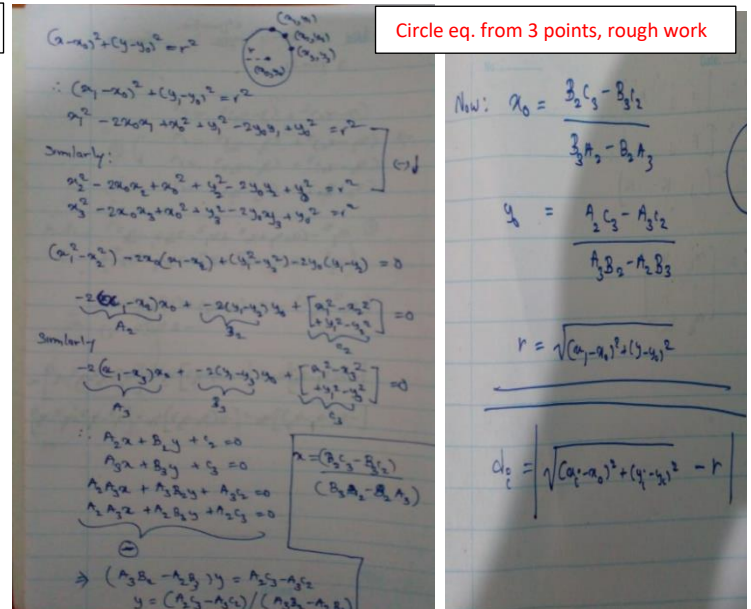
Best circle finder with scipy.optimize

Used Best circle finding code is from

https://scipy-cookbook.readthedocs.io/items/Least_Squares_Circle.html#Using-scipy.optimize.leastsq

Important points:

1. Instead of looping through a random index set, here I am using three random numpy vector s of indexes to select 3 points randomly for RANSAC. All the calculations are done as vectors to make the process faster.
2. 10,000 samples are taken for the RANSAC
3. Both RANSAC and best fit (least square) are almost overlapping for threshold of 2. Here shown is a most deviated instance I got.
4. Mistakenly did the line finding part (not asked in the assignment) and it is in the python notebook.(GITHUB)



```

debug countSet shape (10000,)
142
equation of the circle detected
(x-x0)**2 + (y-y0)**2 = r**2
(x - (0.8931490404885495))**2 + (y - (0.1119989989460876))**2 = 10.229327367079403**2
len(X_inliers_y) 65
bestFitCircle [0.2120519825223083, -0.058760650800690546, 10.152210574797733]

```

