## Assignment - 02

Name: Vakeesan.K

Index NO: 190643G

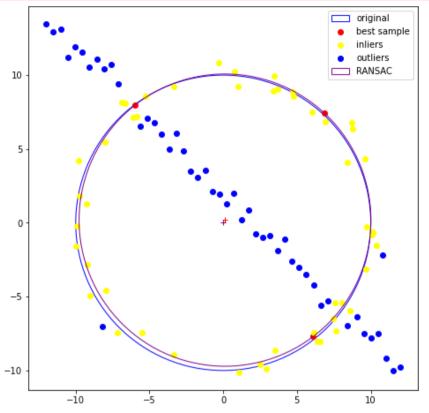
Github link: https://github.com/vakeesanvk/image\_processing\_assignment02

Q1)

```
In [6]: import numpy as np
         from scipy.optimize import minimize
         from scipy import linalg
         import matplotlib.pyplot as plt
         import math
         import random
         # np.random.seed ( 0 )
         N,r = 100,10
         half_n,s = N//2,r/16
         t = np.random.uniform(0,2*np.pi,half_n)
         n = s*np.random.randn (half_n)
        x , y = (r+n)*np.cos(t) , (r+n)*np.sin(t) #
        X_circ = np.hstack((x.reshape(half_n,1) ,y.reshape(half_n,1)))
        m, b = -1, 2
         x = np.linspace(-12, 12, half_n)
         y = m*x + b + s*np.random.randn(half_n) #line equation with random noise
         X_line = np.hstack((x.reshape(half_n , 1) , y.reshape(half_n ,1)))
         X = np.vstack (( X_circ , X_line ))
         plt.figure(figsize=(8,8))
         plt.plot(0,0,color='purple', marker='+')
         original_circle = plt.Circle((0, 0), radius=r, color='blue', fc='y', fill=False,label='original')
         plt.gca().add_patch(original_circle) #draw the circle without the noisy points
         def my_ransac_1(X,t): #t - threshold, X- data points
             num_iterations=math.inf
             iterations_done =0
             best_model = None
             best_sample= None
             #min abs error=math.inf #defining the absolute mean error when there is same number of inliers for more models
             prob_outlier = 0.5 #assume the outlier probability
             desired prob = 0.95
             inlier_max_count=0
            p=0.99 #ideal value
            while(num_iterations > iterations_done):
                 #estimate x,y,r from three radom points
                 cor_1,cor_2,cor_3= random.choices(X,k=3)
                 #fina a circle through the three points
                 x1,y1,x2,y2,x3,y3=cor_1[0],cor_1[1],cor_2[0],cor_2[1],cor_3[0],cor_3[1]
                 c,a,b,s = (x1-x2)**2 + (y1-y2)**2 , (x2-x3)**2 + (y2-y3)**2, (x3-x1)**2 + (y3-y1)**2, 2*(a*b + b*c + c*a) - (a*a + b*b + c*c)
                 px = (a*(b+c-a)*x1 + b*(c+a-b)*x2 + c*(a+b-c)*x3) / s #x_coordinate of the center of the circle
                 py = (a*(b+c-a)*y1 + b*(c+a-b)*y2 + c*(a+b-c)*y3) / s #y\_coordinate of the center of ther circle
                 ar,br,cr= a**0.5,b**0.5,c**0.5
                 r = ar*br*cr / ((ar+br+cr)*(-ar+br+cr)*(ar-br+cr)*(ar+br-cr))**0.5 #radius of the circle
                 #select the points within the donut region
                 inliers=[] #inlier set
                 #abs_error=0 #set the initial value to zero
                 for i in X: #take all the points from the all data and find the inliers
                     x_,y_=i[0],i[1] #x_coordinate , y_coordinate
                     \textbf{if}(abs(np.sqrt(pow(x\_-px,2)+pow(y\_-py,2))-r)<\textbf{t}): \textit{\#check the componets inside the donut region}
                         inliers.append([x_,y_])
                         \#abs\_error+=np.abs(np.sqrt(pow(x_-px,2)+pow(y_-py,2))-r)
                 #check whether those points set has maximum number of inliers
                 if ((inlier_max_count<=len(inliers)) ): #and (abs_error >= min_abs_error)
                     inlier_max_count=len(inliers)
                     #min_abs_error=abs_error
                     best_model=[px,py,r,cor_1,cor_2,cor_3]
                     best_sample=[[x1,y1],[x2,y2],[x3,y3]]
                     best_inliers=inliers
                 #loop condition
                 prob_outlier = 1 - len(inliers)/N
                     num_iterations = math.log(1 - desired_prob)/math.log(1 - (1 - prob_outlier)**3)
                 except ZeroDivisionError :
                     pass
                 iterations_done = iterations_done + 1
             return best_model,best_sample,best_inliers
         best_model,best_sample,inliers=my_ransac_1(X,1)
         a,b,r_ransac=best_model[0],best_model[1],best_model[2]
         x_{=}[x[0]  for x  in best_sample]
         y_=[y[1] for y in best_sample]
         plt.scatter(x_, y_, c='red', marker='o', label='best sample') #plot the best 3 points
         x_i = [x[0] \text{ for } x \text{ in inliers if } x \text{ not in best_sample}]
         y_i = [x[1] for x in inliers if x not in best_sample]
         plt.scatter(x_i, y_i, c='yellow', marker='o', label='inliers') #plot the inliers
         x_d, y_d=[],[]
         for j in X:
            if ((inliers-j).all()):
                 x_d.append(j[0])
                 y_d.append(j[1])
         plt.scatter(x_d, y_d, c='blue', marker='o', label='outliers') #plot the outliers
         circle = plt.Circle((a, b), radius=r_ransac, color='purple', fc='y', fill=False,label='RANSAC') #plot the estimated circle
         plt.gca().add_patch(circle)
         plt.plot(a,b,'r+')
```

```
# delete the comment to see the donut region
# circle1 = plt.Circle((a, b), radius=(r+threshold), color='g',linestyle='dashed', fc='y', fill=False) #show the donut region
# circle2 = plt.Circle((a, b), radius=(r-threshold), color='g',linestyle='dashed', fc='y', fill=False) #show the donut region
# plt.gca().add_patch(circle1)
# plt.gca().add_patch(circle2)
plt.axis('scaled'),plt.legend(),plt.show()
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_11212\952546335.py:47: RuntimeWarning: invalid value encountered in double_scalars
px = (a*(b+c-a)*x1 + b*(c+a-b)*x2 + c*(a+b-c)*x3) / s #x_coordinate of the center of the circle
C:\Users\USER\AppData\Local\Temp\ipykernel_11212\952546335.py:48: RuntimeWarning: invalid value encountered in double_scalars
py = (a*(b+c-a)*y1 + b*(c+a-b)*y2 + c*(a+b-c)*y3) / s #y_coordinate of the center of ther circle
C:\Users\USER\AppData\Local\Temp\ipykernel_11212\952546335.py:50: RuntimeWarning: invalid value encountered in double_scalars
r = ar*br*cr / ((ar+br+cr)*(-ar+br+cr)*(ar-br+cr))**0.5 #radius of the circle
```



Q2)

```
In [4]: import cv2 as cv
         import numpy as np
         import argparse
         def click_event(event, x, y, flags, params): #selecting the four points
            if event == cv.EVENT_LBUTTONDOWN:
                cv.circle(base_image_copy, (x,y), 4, (0,0,255), -1)
                 points.append([x, y])
                if len(points) <= 4:</pre>
                     cv.imshow('image', base_image_copy)
         points = []
         base_image = cv.imread(r"D:\01.Vakeesan_vk\01.academics\04. 4 th semester\8 - EN2550 - Fundamentals of image processing\img_process\assignment_02\003
         base_image_copy = base_image.copy()
         subject_image = cv.imread(r"D:\01.Vakeesan_vk\01.academics\04. 4 th semester\8 - EN2550 - Fundamentals of image processing\img_process\assignment_02\
         cv.imshow('image', base_image_copy)
         cv.setMouseCallback('image', click_event)
         cv.waitKey(0)
         cv.destroyAllWindows()
         # Let's sort the points in the following order
         # Top-Left, Top-Right, Bottom-Right, Bottom-Left
         # select the region being the contour
         sorted_pts = np.zeros((4, 2), dtype="float32")
         s = np.sum(points, axis=1) #find the sum of the points
         sorted_pts[0] = points[np.argmin(s)]
         sorted_pts[2] = points[np.argmax(s)]
         diff = np.diff(points, axis=1) #fin the difference of the points
         sorted_pts[1] = points[np.argmin(diff)]
         sorted_pts[3] = points[np.argmax(diff)]
         h base, w base, c base = base image.shape
         h_subject, w_subject = subject_image.shape[:2]
         pts1 = np.float32([[0, 0], [w_subject, 0], [w_subject, h_subject], [0, h_subject]])
         pts2 = np.float32(sorted_pts) #soted_pts
         # Get the transformation matrix and use it to get the warped image of the subject
         transformation_matrix = cv.getPerspectiveTransform(pts1, pts2)
         warped_img = cv.warpPerspective(subject_image, transformation_matrix, (w_base, h_base))
         # Create a mask
         mask = np.zeros(base_image.shape, dtype=np.uint8)
         roi_corners = np.int32(sorted_pts)
         # Fill in the region selected with white color
         filled_mask = mask.copy()
         cv.fillConvexPoly(filled_mask, roi_corners, (255, 255, 255))
         # Invert the mask color
         inverted_mask = cv.bitwise_not(filled_mask)
         # Bitwise AND the mask with the base image
        masked image = cv.bitwise and(base image, inverted mask)
         output = cv.addWeighted(warped_img, 0.5, base_image, 1, 0.0) #set the alpha=0.5 , beta=1
         plt.figure(figsize=(8,8))
         plt.imshow(cv.cvtColor(output,cv.COLOR_BGR2RGB))
```

```
100 - 200 - 200 400 600 800 1000
```

```
In [34]: import matplotlib.pyplot as plt
         import cv2 as cv
         import math
         import numpy as np
         %matplotlib inline
         def homography_finder(pairs):
             A = []
             for x1, y1, x2, y2 in pairs:
                 A.append([x1, y1, 1, 0, 0, 0, -x2 * x1, -x2 * y1, -x1])
                 A.append([0, 0, 0, x1, y1, 1, -y2 * x1, -y2 * y1, -y1])
             A = np.asarray(A)
             # Singular Value Decomposition (SVD)
             U, S, V = np.linalg.svd(A)
             # V has shape (9, 9) for any number of input pairs. V[-1] is the eigenvector
             # of (A^T)A with the smalles eigenvalue. Reshape into 3x3 matrix.
             H = np.reshape(V[-1], (3, 3))
             # Normalization
             H = (1 / H.item(8)) * H
             #H /= H[2][2] #both given normalization methods are same
             return H
         def my_ransac_2(point_map,t):
             num_iterations=math.inf
             iterations_done =0
             N=len(point_map)
             #min_abs_error=math.inf #same usage as in question 01
             prob_outlier = 0.5
             desired_prob = 0.95
             inlier_max_count=0
             p=0.99 #ideal value
             while(num_iterations > iterations_done):
                 # randomly choose 4 points from the matrix to compute the homography
                 pairs = [point_map[i] for i in np.random.choice(len(point_map), 4)]
                 H=homography_finder(pairs)
                 inliers=0
                 #abs_error=0
                 for pair in point_map:
                      # points in homogeneous coordinates
                     p1 = np.array([pair[0], pair[1], 1])
                     p2 = np.array([pair[2], pair[3], 1])
                     p2_estimate = np.dot(H, np.transpose(p1))
                     p2_estimate = (1 / p2_estimate[2]) * p2_estimate
                     if (np.linalg.norm(np.transpose(p2) - p2_estimate)<50): #lesser the value (here 50) taken time is higher</pre>
                          inliers+=1
                 #check the maximum inliers's model detected
                 if (inlier_max_count <=inliers): #and (abs_error >= min_abs_error)
                     inlier_max_count = inliers
                     homography = H
                     #if (bestInliers > (point_map * threshold)): break
                 #loop condition
                 prob_outlier = 1 - inliers/N
                 try: #avoid the zero division error
                     num_iterations = math.log(1 - desired_prob)/math.log(1 - (1 - prob_outlier)**3)
                 except ZeroDivisionError :
                 iterations_done = iterations_done + 1
             return homography
         def stitch_image(inputIm, refIm, H):
             H_inv = np.linalg.inv(H) #inverse of the homography
             inputH, inputW, c = inputIm.shape
             outputH, outputW, c = refIm.shape
             min_x,min_y, max_x, max_y = float("inf"),float("inf"),float("-inf")
             cornersi = [(0,0), (inputH, inputW), (0, inputW), (inputH, 0)]
             cornerso = [(0,0), (inputH, inputW), (0, inputW), (inputH, 0)]
             for i,j in cornersi:
                     x, y, w = np.matmul(H, [j, i, 1])
                     x,y = x/w, y/w
                     if x > max_x: max_x = int(x)
                     if x < min_x: min_x = int(x)</pre>
                     if y > max_y: max_y = int(y)
                     if y < min_y: min_y = int(y)</pre>
             warpIm = np.zeros((max_y - min_y,max_x - min_x, 3))
             for i in range(0, max_x - min_x):
                 for j in range (0, max_y - min_y):
                     x, y, w = np.matmul(H_inv, [i + min_x, j + min_y, 1])
                     x,y = int(x/w), int(y/w)
                     a,b, c = 0,0,0
```

```
if not (y < 0 \text{ or } y >= \text{inputH or } x < 0 \text{ or } x >= \text{inputW}):
                          a, b, c = inputIm[y, x, :]
                      warpIm[j, i, :] = [a/255, b/255, c/255]
              oldx,oldy,oldmx, oldmy = min_x, min_y, max_x, max_y
              for i,j in cornerso:
                  if j > max_x: max_x = int(j)
                  if j < min_x: min_x = int(j)</pre>
                  if i > max_y: max_y = int(i)
                  if i < min_y: min_y = int(i)</pre>
              mergeIm = np.zeros(((max_y - min_y),(max_x - min_x), 3))
              for i in range(min_x, max_x):
                  for j in range (min_y, max_y):
                      a,b,c = 0,0,0
                      if not (j < oldy or j >= oldmy or i < oldx or i >= oldmx):
                          a, b, c = warpIm[j - oldy, i - oldx, :]
                          if a == 0.0 or b == 0.0 or c == 0.0:
                              if not (j < 0 or j >= outputH or i < 0 or i >= outputW):
                                  a, b, c = refIm[j, i, :]/255
                      else:
                          if not (j < 0 or j >= outputH or i < 0 or i >= outputW):
                              a, b, c = refIm[j, i, :]/255
                      mergeIm[j - min_y, i- min_x, :] = [a, b, c]
             return mergeIm
          # read images
          img1 = cv.imread(r"D:\01.Vakeesan_vk\01.academics\04. 4 th semester\8 - EN2550 - Fundamentals of image processing\img_process\assignment_02\img1.ppm'
          img2 = cv.imread(r"D:\01.Vakeesan_vk\01.academics\04. 4 th semester\8 - EN2550 - Fundamentals of image processing\img_process\assignment_02\img5.ppm'
          sift = cv.xfeatures2d.SIFT_create()
          keypoints_1, descriptors_1 = sift.detectAndCompute(img1,None)
          keypoints_2, descriptors_2 = sift.detectAndCompute(img2,None)
          bf = cv.BFMatcher(cv2.NORM_L1, crossCheck=True)
          matches = bf.match(descriptors_1,descriptors_2)
          matches = sorted(matches, key = lambda x:x.distance) #sort the matching points
          point_map = np.array([[ keypoints_1[match.queryIdx].pt[0],
                                   keypoints_1[match.queryIdx].pt[1],
                                   keypoints_2[match.trainIdx].pt[0],
                                   keypoints_2[match.trainIdx].pt[1]] for match in matches])
          homography = my_ransac_2(point_map,0.1)
          original\_homography = np.array ([[6.2544644e-01\ ,\ 5.7759174e-02\ ,\ 2.2201217e+02],\ \textit{\#get it from H1toH5 file}) \\
                                        [ 2.2240536e-01 , 1.1652147e+00 , -2.5605611e+01],
                                        [4.9212545e-04 ,-3.6542424e-05 , 1.0000000e+00]])
          ori_stitch=stitch_image(img1,img2,original_homography)
          my_stitch=stitch_image(img1,img2,homography)
          img3 = cv2.drawMatches(img1, keypoints_1, img2, keypoints_2, matches[:50], img1, flags=2) # inliers can be added to show here
          print(" Original Homography \n",original_homography)
          print('')
          print(" Calclated Homography \n",homography)
          Original Homography
          [[ 6.2544644e-01 5.7759174e-02 2.2201217e+02]
           [ 2.2240536e-01 1.1652147e+00 -2.5605611e+01]
          [ 4.9212545e-04 -3.6542424e-05 1.0000000e+00]]
          Calclated Homography
          [[ 2.33005736e-01 8.11764668e-01 9.00173634e+01]
           [-3.79772324e-01 1.49766627e+00 2.34148751e+01]
           [-5.25269498e-04 8.34863292e-04 1.00000000e+00]]
In [33]: plt.imshow(cv.cvtColor(img3,cv.COLOR_BGR2RGB)),plt.title("SIFT featuring matching"),plt.show()
          fig,ax=plt.subplots(1,2,figsize=(12,6))
          ax[0].imshow(ori_stitch),ax[0].set_title("Stitching using given homography")
          ax[1].imshow(my_stitch),ax[1].set_title("Stitching using my_ransac function"),plt.show()
                           SIFT featuring matching
                        400
                              600
                                    800
                                          1000
                                               1200
                     Stitching using given homography
                                                                         Stitching using my ransac function
          100
                                                               100
          200
                                                               200
          300
          400
                                                               400
          500
                                                               500
          600
                                                               600
          700
                                                               700
                                                                       100
                                                                            200
                                                                                             500
```

Out[33]: (<matplotlib.image.AxesImage at 0x1cf1e3ec220>, Text(0.5, 1.0, 'Stitching using my\_ransac function'),

400

500

300

700

600

100

200