```
In [330...
         import numpy as np
         import skimage
         from skimage import transform
         from skimage import io
         import cv2
         import matplotlib.pyplot as plt
         from scipy.spatial import distance
         import scipy
         import random
         from PIL import Image
In [331...
         def get_sift(img):
             get the keypoints and compute their SIFT descriptors
             sift = cv2.xfeatures2d.SIFT create()
             kp, distance = sift.detectAndCompute(img, None)
             return kp, distance
         def plot_inlier_matches(ax, img1, img2, inliers):
             plot the match between two image according to the matched keypoints
             res = np.hstack([img1, img2])
             ax.set aspect('equal')
             ax.imshow(res, cmap='gray')
             ax.plot(inliers[:,0], inliers[:,1], '+r')
             ax.plot(inliers[:,2] + img1.shape[1], inliers[:,3], '+r')
             ax.plot([inliers[:,0], inliers[:,2] + img1.shape[1]],
                     [inliers[:,1], inliers[:,3]], 'r', linewidth=0.4)
             ax.axis('off')
In [332...
         ### Provided code - nothing to change here ###
         .....
         Harris Corner Detector
         Usage: Call the function harris(filename) for corner detection
         Reference
                    (Code adapted from):
```

```
filename: Path of image file
   threshold: (optional)Threshold for corner detection
   min distance : (optional)Minimum number of pixels separating
    corners and image boundary
    returns: filtered coords list. index zero is row and index 1 is columns.
   harrisim = compute_harris_response(im)
   filtered_coords = get_harris_points(harrisim,min_distance, threshold)
   plot harris points(im, filtered coords)
   return filtered coords
def gauss_derivative_kernels(size, sizey=None):
   """ returns x and y derivatives of a 2D
       gauss kernel array for convolutions """
   size = int(size)
   if not sizey:
        sizey = size
   else:
       sizey = int(sizey)
   y, x = mgrid[-size:size+1, -sizey:sizey+1]
   #x and y derivatives of a 2D gaussian with standard dev half of size
   # (ignore scale factor)
   gx = -x * exp(-(x**2/float((0.5*size)**2)+y**2/float((0.5*sizey)**2)))
   gy = -y * exp(-(x**2/float((0.5*size)**2)+y**2/float((0.5*sizey)**2)))
   return gx,gy
def gauss_kernel(size, sizey = None):
    """ Returns a normalized 2D gauss kernel array for convolutions """
   size = int(size)
   if not sizey:
       sizey = size
   else:
       sizey = int(sizey)
   x, y = mgrid[-size:size+1, -sizey:sizey+1]
   g = exp(-(x**2/float(size)+y**2/float(sizey)))
   return g / g.sum()
def compute_harris_response(im):
   """ compute the Harris corner detector response function
        for each pixel in the image"""
   #derivatives
   gx,gy = gauss_derivative_kernels(3)
   imx = signal.convolve(im,gx, mode='same')
   imy = signal.convolve(im,gy, mode='same')
   #kernel for blurring
   gauss = gauss kernel(3)
   #compute components of the structure tensor
   Wxx = signal.convolve(imx*imx,gauss, mode='same')
   Wxy = signal.convolve(imx*imy,gauss, mode='same')
   Wyy = signal.convolve(imy*imy,gauss, mode='same')
   #determinant and trace
   Wdet = Wxx*Wyy - Wxy**2
   Wtr = Wxx + Wyy
   return Wdet / Wtr
def get_harris_points(harrisim, min_distance=10, threshold=0.1):
```

```
""" return corners from a Harris response image
       min_distance is the minimum nbr of pixels separating
       corners and image boundary"""
   #find top corner candidates above a threshold
   corner threshold = max(harrisim.ravel()) * threshold
   harrisim t = (harrisim > corner threshold) * 1
   #get coordinates of candidates
   candidates = harrisim_t.nonzero()
   coords = [ (candidates[0][c], candidates[1][c]) for c in range(len(candidates[0]
   #...and their values
   candidate_values = [harrisim[c[0]][c[1]] for c in coords]
   #sort candidates
   index = argsort(candidate values)
   #store allowed point locations in array
   allowed locations = zeros(harrisim.shape)
   allowed locations[min distance:-min distance,min distance:-min distance] = 1
   #select the best points taking min distance into account
   filtered coords = []
   for i in index:
       if allowed locations[coords[i][0]][coords[i][1]] == 1:
           filtered coords.append(coords[i])
           allowed locations[(coords[i][0]-min distance):(coords[i][0]+min distance
               (coords[i][1]-min_distance):(coords[i][1]+min_distance)] = 0
   return filtered coords
def plot harris points(image, filtered coords):
   """ plots corners found in image"""
   figure()
   gray()
   imshow(image)
   plot([p[1] for p in filtered coords],[p[0] for p in filtered coords],'r*')
   axis('off')
   show()
### Provided code end
                                        ###
```

## Your implementations

```
In [333... def get_matches(img1, img2, percentile):
    kpoint1, dis1 = get_sift(img1)
    kpoint2, dis2 = get_sift(img2)
    kpoint1, kpoint2 = np.array(kpoint1), np.array(kpoint2)

dist = scipy.spatial.distance.cdist(dis1, dis2, 'sqeuclidean')

# Flatten the matrix and find the N-th percentile value
    threshold = np.percentile(dist, percentile)
    print(f"threshold is = {threshold}")

# Get the indices of elements below the percentile
    indices = np.argwhere(dist < threshold)
    print(f"num of matches allowed {indices.shape}")</pre>
```

```
# get putative under distance distances
    matches = []
    new_matches= []
    h,w = dist.shape
    for i in range(h):
        for j in range(w):
            if dist[i][j] <= 10000:</pre>
                new matches = list(kpoint1[i].pt+kpoint2[j].pt)
                matches.append(new matches)
    matches = np.array(matches)
    print(f"num matches {matches.shape}")
    return matches
def get residual(H, data):
    count = len(data)
    point1 = np.hstack((data[:,:2],np.ones(( count,1))))
    point2 = data[:,2:]
    est = np.zeros((count,2))
    for i in range(count):
        \# x, y = x/w, y/w
        tmp = np.dot(H,point1[i])/np.dot(H,point1[i])[-1]
        est[i] = tmp[:2]
    residual = np.linalg.norm(point2 - est, axis=1) ** 2
    return residual
def ransac(data, t, req_inliers, iterations=1000):
    best H = None
    best inliers = None
    max_inliers = 0
    best_residual = float('inf')
    for i in range(iterations):
        sub index = np.random.choice(len(data), 4)
        subset = data[sub_index]
        #get homography
        A = []
        for i in range(4):
            point1 = np.append(subset[i][:2],1)
            point2 = np.append(subset[i][2:],1)
            coord1 = [0,0,0,point1[0],point1[1],point1[2],-point2[1]*point1[0],-poi
            coord2 = [point1[0],point1[1],point1[2],0,0,0,-point2[0]*point1[0],-poi
            A.append(coord1)
            A.append(coord2)
        #SVD
        A = np.array(A)
        _, _, V = np.linalg.svd(A)
```

```
H = V[-1,:].reshape((3,3))
    #this is homography transformation
    H = H/H[2,2]
    if np.linalg.matrix rank(H) < 3:</pre>
        #H is degenerate
        continue
    # Get inliers
    errors = get_residual(H, data)
    index = np.where(errors < t)[0]</pre>
    inliers = data[index]
    # update the model if inliers more than required
    if len(inliers) >= req_inliers and len(inliers) >= max_inliers:
        curr_err = errors[index].sum() / len(inliers)
        if len(inliers) == max inliers:
            # check the residual
            if curr_err > best_residual:
                continue
        best_residual = errors[index].sum() / len(inliers)
        best inliers = inliers.copy()
        best H = H.copy()
        max_inliers = len(inliers)
return best H, best inliers, max inliers, best residual
```

```
In [334...
          import numpy as np
          from skimage.transform import ProjectiveTransform, warp
          from skimage import io
          def warp_images(img1, img2, H):
              Warp the right image onto the left image using the provided homography matrix.
              Args:
              - left_image: left color image.
              - right_image: right color image.
              - homography_matrix (numpy.array): 3x3 homography transformation matrix.
              Returns:
              - The composite panorama image.
              transform = ProjectiveTransform(H)
              # Calculate the boundary of warped image
              h, w, z = img1.shape
              raw = np.array([[0, 0], [0, h], [w, h], [w, 0]])
              trans = transform(raw)
              combine = np.vstack((raw, trans))
              min_dot = np.int32(combine.min(axis=0))
```

## Main functions

```
In [335...
          # Load images
          img1 = np.array(Image.open('Images/left.jpg'))
          img2 = np.array(Image.open('Images/right.jpg'))
          gray1 = Image.open('Images/left.jpg').convert("L")
          gray2 = Image.open('Images/right.jpg').convert("L")
          gray1 = cv2.cvtColor(np.array(gray1), cv2.COLOR_GRAY2BGR)
          gray2 = cv2.cvtColor(np.array(gray2), cv2.COLOR_GRAY2BGR)
In [336... # compute and display the initial SIFT matching result
          data = get matches(gray1, gray2, percentile=1)
         threshold is = 139373.71
         num of matches allowed (153575, 2)
         num matches (61, 4)
 In [ ]: # part (d) performn RANSAC to get the homography and inliers,
          # display the inlier matching, report the average residual
          H, ransac_match, best_inliers, best_model_errors = ransac(data, 100, 15)
          print("Average residual:", np.average(best_model_errors))
          print("Inliers:", best_inliers)
          fig, ax = plt.subplots(figsize=(20,10))
          plot_inlier_matches(ax, gray1, gray2, ransac_match)
```

Average residual: 18.13999979716613 Inliers: 42



```
In [338... H.shape
```

Out[338... (3, 3)

```
In [339... # part (e) warp images to stitch them together,
    # display and report the stitching results

im = warp_images(img2, img1, H)

plt.figure(figsize=(12, 8))
    plt.imshow(im)
    plt.show()

im = Image.fromarray(im)
    im.save("two_image.jpg", format="jpeg")
```

offset is (921, 173)

