```
In [16]: 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 [17]: 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 [18]:
        ### Provided code - nothing to change here ###
         .....
         Harris Corner Detector
         Usage: Call the function harris(filename) for corner detection
         Reference
                    (Code adapted from):
                     http://www.kaij.org/blog/?p=89
                     Kai Jiang - Harris Corner Detector in Python
         ....
         # Usage:
         #harris('./path/to/image.jpg')
         def harris(im, min_distance = 10, threshold = 0.1):
```

```
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 [19]: 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 [20]: 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

Inliers: 309

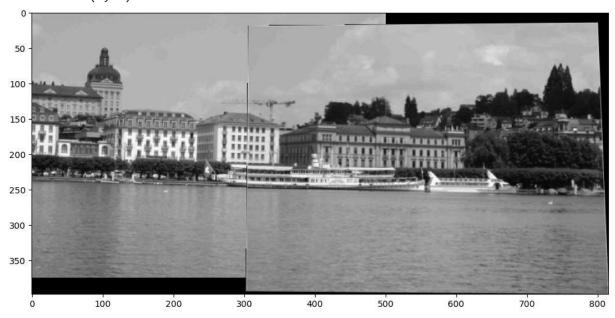
```
In [21]: # Load images
         img left = np.array(Image.open('pier/pier1.JPG'))
         img center = np.array(Image.open('pier/pier2.JPG'))
         img_right = np.array(Image.open('pier/pier3.JPG'))
         gray_l = Image.open('pier/pier1.JPG').convert("L")
         gray c = Image.open('pier/pier2.JPG').convert("L")
         gray_r = Image.open('pier/pier3.JPG').convert("L")
         gray_1 = cv2.cvtColor(np.array(gray_1), cv2.COLOR_GRAY2BGR)
         gray_c = cv2.cvtColor(np.array(gray_c), cv2.COLOR_GRAY2BGR)
         gray_r = cv2.cvtColor(np.array(gray_r), cv2.COLOR_GRAY2BGR)
In [22]: # compute and display the center and right matching
         matches = get_matches(gray_r, gray_c, percentile=1)
        threshold is = 127807.0
        num of matches allowed (17068, 2)
        num matches (383, 4)
In [23]: # performn RANSAC to get the homography for center and right
         H, ransac_match, best_inliers, best_model_errors = ransac(matches, 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, gray_c, gray_r, ransac_match)
        Average residual: 6.748763206930363
```

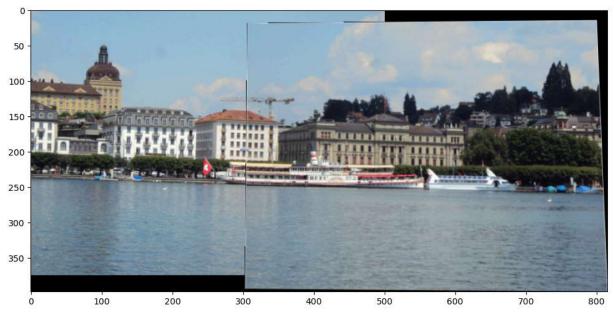


```
In [24]: H.shape
Out[24]: (3, 3)
In [25]: # warp right to center
    im_right_side = warp_images(gray_c, gray_r, H)
    color_im_right_side = warp_images(img_center, img_right, H)
    plt.figure(figsize=(12, 8))
    plt.imshow(im_right_side)
    plt.show()

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

offset is (0, 0) offset is (0, 0)





```
In [26]: # compute the left matching with center+ right
    matches_l_and_CR = get_matches(gray_l, im_right_side, percentile=10)

    threshold is = 207780.0
    num of matches allowed (376462, 2)
    num matches (293, 4)

In [27]: H, ransac_match, best_inliers, best_model_errors = ransac(matches_l_and_CR, 50, 35, print("Average residual:", np.average(best_model_errors))
    print("Inliers:", best_inliers)

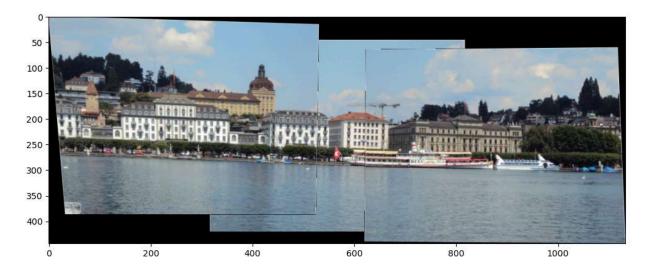
Average residual: 1.111529478845029
    Inliers: 276
```

```
In [28]: im = warp_images(color_im_right_side, img_left, H)

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

im = Image.fromarray(im)
im.save("3_image.jpg", format="jpeg")
```

offset is (317, 46)



Extra Credit

```
In [29]: # Apply your stitching code to your own images.
         # Load images
         img_left = np.array(Image.open('myImages/left.jpg'))
         img_center = np.array(Image.open('myImages/center.jpg'))
         img right = np.array(Image.open('myImages/right.jpg'))
         gray_l = Image.open('myImages/left.jpg').convert("L")
         gray c = Image.open('myImages/center.jpg').convert("L")
         gray_r = Image.open('myImages/right.jpg').convert("L")
         gray_1 = cv2.cvtColor(np.array(gray_1), cv2.COLOR_GRAY2BGR)
         gray c = cv2.cvtColor(np.array(gray c), cv2.COLOR GRAY2BGR)
         gray_r = cv2.cvtColor(np.array(gray_r), cv2.COLOR_GRAY2BGR)
         # compute and display the center and right matching
         matches = get_matches(gray_r, gray_c, percentile=1)
        threshold is = 108011.15
        num of matches allowed (8865, 2)
        num matches (95, 4)
In [30]: # performn RANSAC to get the homography for center and right
         H, ransac match, best inliers, best model errors = ransac(matches, 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, gray_c, gray_r, ransac_match)
```

Average residual: 17.767663685789575

Inliers: 84



```
In []: # warp right to center

im_right_side = warp_images(gray_c, gray_r, H)

color_im_right_side = warp_images(img_center, img_right, H)

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

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

color_im_right_side.save("extra_credit_2images.jpg", format="jpeg")
```

offset is (0, 26) offset is (0, 26)





```
In [32]: # compute the left matching with center+ right
         matches_l_and_CR = get_matches(gray_l, im_right_side, percentile=10)
         H, ransac_match, best_inliers, best_model_errors = ransac(matches_l_and_CR, 50, 35,
         print("Average residual:", np.average(best_model_errors))
         print("Inliers:", best_inliers)
        threshold is = 198813.0
        num of matches allowed (149651, 2)
        num matches (93, 4)
        Average residual: 9.83039580466553
        Inliers: 91
 In [ ]: im = warp_images(color_im_right_side, img_left, H)
         plt.figure(figsize=(12, 8))
         plt.imshow(im)
         plt.show()
         im = Image.fromarray(im)
         im.save("extra_credit_3images.jpg", format="jpeg")
        offset is (397, 0)
```

