EE16A: Homework 2

Problem 6: Image Stitching

This section of the notebook continues the image stiching problem. Be sure to have a figures folder in the same directory as the notebook. The figures folder should contain the files:

```
Berkeley_banner_1.jpg
Berkeley_banner_2.jpg
stacked_pieces.jpg
lefthalfpic.jpg
righthalfpic.jpg
```

Note: This structure is present in the provided HW2 zip file.

Run the next block of code before proceeding

```
In [1]:
        import numpy as np
        import numpy.matlib
        import matplotlib.pyplot as plt
        from mpl toolkits.mplot3d import Axes3D
        from numpy import pi, cos, exp, sin
        import matplotlib.image as mpimg
        import matplotlib.transforms as mtransforms
        %matplotlib inline
        #Loading images
        image1=mpimg.imread('figures/Berkeley_banner_1.jpg')
        image1=image1/255.0
        image2=mpimg.imread('figures/Berkeley banner 2.jpg')
        image2=image2/255.0
        image stack=mpimg.imread('figures/stacked pieces.jpg')
        image_stack=image_stack/255.0
        image1 marked=mpimg.imread('figures/lefthalfpic.jpg')
        image1 marked=image1 marked/255.0
        image2 marked=mpimg.imread('figures/righthalfpic.jpg')
        image2 marked=image2 marked/255.0
        def euclidean transform 2to1(transform mat,translation,image,position,LL,UL):
            new position=np.round(transform mat.dot(position)+translation)
            new_position=new_position.astype(int)
            if (new position>=LL).all() and (new position<UL).all():</pre>
                 values=image[new_position[0][0],new_position[1][0],:]
            else:
                values=np.array([2.0,2.0,2.0])
            return values
        def euclidean transform 1to2(transform mat,translation,image,position,LL,UL):
            transform mat inv=np.linalg.inv(transform mat)
            new position=np.round(transform mat inv.dot(position-translation))
            new_position=new_position.astype(int)
            if (new position>=LL).all() and (new position<UL).all():</pre>
                 values=image[new_position[0][0],new_position[1][0],:]
            else:
                 values=np.array([2.0,2.0,2.0])
            return values
```

We will stick to a simple example and just consider stitching two images (if you can stitch two pictures, then you could conceivably stitch more by applying the same technique over and over again).

Daniel decided to take an amazing picture of the Campanile overlooking the bay. Unfortunately, the field of view of his camera was not large enough to capture the entire scene, so he decided to take two pictues and stich them together.

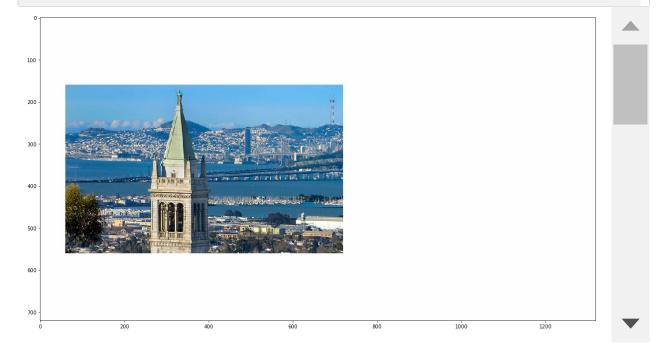
The next block will display the two images.

```
In [2]: plt.figure(figsize=(20,40))
    plt.subplot(311)
    plt.imshow(image1)

    plt.subplot(312)
    plt.imshow(image2)

    plt.subplot(313)
    plt.imshow(image_stack)

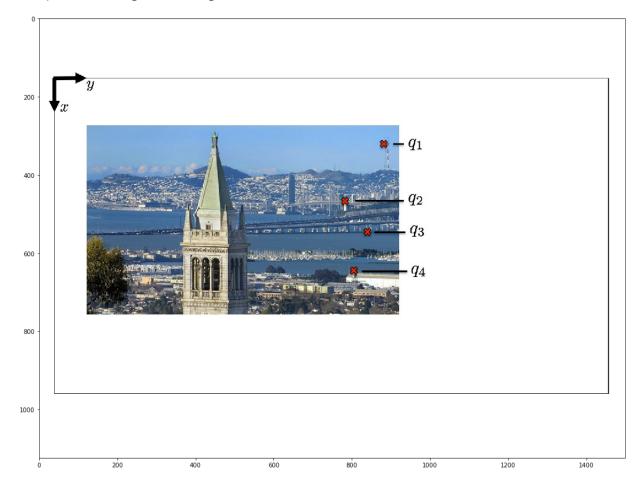
    plt.show()
```

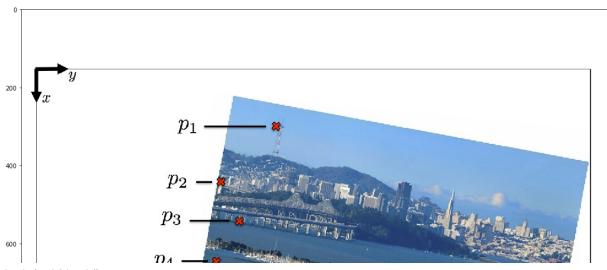


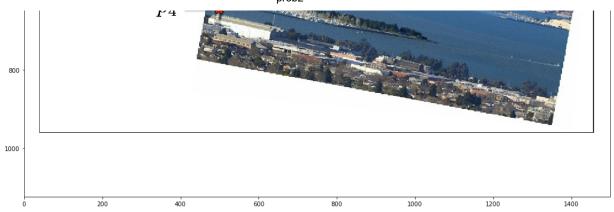
Once you apply Marcela's algorithm on the two images you get the following result (run the next block):

```
In [3]: plt.figure(figsize=(20,30))
    plt.subplot(211)
    plt.imshow(image1_marked)
    plt.subplot(212)
    plt.imshow(image2_marked)
```

Out[3]: <matplotlib.image.AxesImage at 0x20e84c4e400>







As you can see Marcela's algorithm was able to find four common points between the two images. These points expressed in the coordinates of the first image and second image are

$$\vec{p_1} = \begin{bmatrix} 200 \\ 700 \end{bmatrix} \qquad \vec{p_2} = \begin{bmatrix} 310 \\ 620 \end{bmatrix} \qquad \vec{p_3} = \begin{bmatrix} 390 \\ 660 \end{bmatrix} \qquad \vec{p_4} = \begin{bmatrix} 46 \\ 63 \end{bmatrix}$$

$$\vec{q_1} = \begin{bmatrix} 162.2976 \\ 565.8862 \end{bmatrix} \qquad \vec{q_2} = \begin{bmatrix} 285.4283 \\ 458.7469 \end{bmatrix} \qquad \vec{q_3} = \begin{bmatrix} 385.2465 \\ 498.1973 \end{bmatrix} \qquad \vec{q_4} = \begin{bmatrix} 465.789 \\ 455.013 \end{bmatrix}$$

It should be noted that in relation to the image the positive x-axis is down and the positive y-axis is right. This will have no bearing as to how you solve the problem, however it helps in interpreting what the numbers mean relative to the image you are seeing.

Using the points determine the parameters R_{11} , R_{12} , R_{21} , R_{22} , T_x , T_y that map the points from the first image to the points in the second image by solving an appropriate system of equations. Hint: you do not need all the points to recover the parameters.

```
In [4]: # Note that the following is a general template for solving for 6 unknowns from
        # You do not have to use the following code exactly.
        # All you need to do is to find parameters R_11, R_12, R_21, R_22, T_2, T_2.
        # If you prefer finding them another way it is fine.
         # fill in the entries
         A = np.array([[200,700,0,0,1,0],
                       [0,0,200,700,0,1],
                       [310,620,0,0,1,0],
                       [0,0,310,620,0,1],
                       [390,660,0,0,1,0],
                       [0,0,390,660,0,1]])
         # fill in the entries
         b = np.array([[162.2976],[565.8862],[285.4283],[458.7469],[385.2465],[498.1973]]
         A = A.astype(float)
         b = b.astype(float)
         # solve the linear system for the coefficiens
         z = np.linalg.solve(A,b)
         #Parameters for our transformation
         R 11 = z[0,0]
         R_{12} = z[1,0]
         R 21 = z[2,0]
         R_22 = z[3,0]
         T_x = z[4,0]
         T_y = z[5,0]
```

Stitch the images using the transformation you found by running the code below.

Note that it takes about 40 seconds for the block to finish running on a modern laptop.

```
In [5]: matrix_transform=np.array([[R_11,R_12],[R_21,R_22]])
        translation=np.array([T_x,T_y])
        #Creating image canvas (the image will be constructed on this)
        num row, num col, blah=image1.shape
        image_rec=1.0*np.ones((int(num_row),int(num_col),3))
        #Reconstructing the original image
        LL=np.array([[0],[0]]) #lower limit on image domain
        UL=np.array([[num_row],[num_col]]) #upper limit on image domain
        for row in range(0,int(num_row)):
            for col in range(0,int(num col)):
                #notice that the position is in terms of x and y, so the c
                position=np.array([[row],[col]])
                if image1[row,col,0] > 0.995 and image1[row,col,1] > 0.995 and image1[row
                     temp = euclidean_transform_2to1(matrix_transform, translation, image2,
                     image_rec[row,col,:] = temp
                 else:
                     image rec[row,col,:] = image1[row,col,:]
        plt.figure(figsize=(20,20))
        plt.imshow(image_rec)
        plt.axis('on')
        plt.show()
```

