Project 2: Computer Vision and Image Processing

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1. Software Used

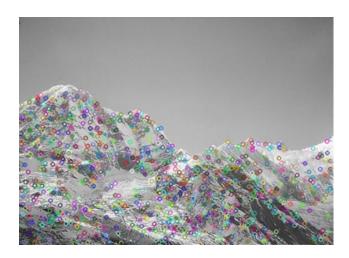
- Anaconda Navigator
- Jupyter Notebook
- Python 3.6
- 2. Libraries Used
- Num-py
- Open CV
- Matplot
- 3. Task

3.1 Image Features and Homography

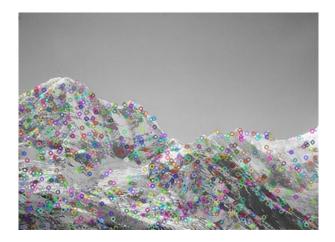
 $Reference: https://opencv-python-tutroals.readthedocs.io/en/latest/py_tutorials/py_feature2d/py_feature_homography/py_feature_homography.html$

3.1.1 Extract SIFT Features

Task1_sift1.png



Task1_sift2.png



3.1.2 Match the key-points using k-nearest neighbors

Task1_matches_knn.png



3.1.3 Homography matrix H

```
M, mask = cv2.findHomography(src_pts, dst_pts, cv2.RANSAC,5.0)
matchesMask = mask.ravel().tolist()
print(M)
```

Ransac Values

```
[[ 1.58930391e+00 -2.91559485e-01 -3.95969641e+02]
[ 4.49424740e-01 1.43111047e+00 -1.90614333e+02]
[ 1.21265321e-03 -6.28722024e-05 1.00000000e+00]]
```

3.1.4 Match Image for 10 random matches

```
points = []
for i in range(len(matchesMask)):
    if matchesMask[i] == 1:
        points.append(i)
points = points[:10]

img3 = cv2.drawMatches(img1,kp1,img2, kp2,[good[point] for point in points],None,flags=2)
plt.imsave(img3, 'task1_matches.jpg')
print(M)
```

Task1_matches.png



3.1.5 Warping Image

```
rows1, cols1 = temp1.shape[:2]
rows2, cols2 = temp2.shape[:2]

lp1 = np.float32([[0,0], [0,rows1], [cols1, rows1], [cols1,0]]).reshape(-1,1,2)
temp = np.float32([[0,0], [0,rows2], [cols2, rows2], [cols2,0]]).reshape(-1,1,2)

lp2 = cv2.perspectiveTransform(temp, M)
lp = np.concatenate((lp1, lp2), axis=0)

[x_min, y_min] = np.int32(lp.min(axis=0).ravel() = 0.5)
[x_max, y_max] = np.int32(lp.max(axis=0).ravel() + 0.5)

translation_dist = [-x_min, -y_min]
H_translation = np.array([[1, 0, translation_dist[0]], [0, 1, translation_dist[1]], [0,0,1]])

result = cv2.warpPerspective(temp1, H_translation.dot(M), (x_max = x_min, y_max = y_min))
result[translation_dist[1]:rows1+translation_dist[0]:cols1+translation_dist[0]] = temp2
```

Task1_pano.png



3.2 Epi-polar Geometry

 $Reference: https://docs.opencv.org/3.2.0/da/de9/tutorial_py_epipolar_geometry.html$

3.2.1 Extract SIFT Features and Match the key-points using k-nearest neighbors

```
import cv2
import numpy as np

img = cv2.imread('/Users/kamallala/Downloads/data/tsucuba_left.png', 0)

sift = cv2.xfeatures2d.SIFT_create()
kp = sift.detect(img,None)

cv2.drawKeypoints(img,kp)

cv2.imwrite('task2_sift1.jpg',img)
```

Task2_sift1.png



Task2_sift2.png



Task2_matches_knn.png

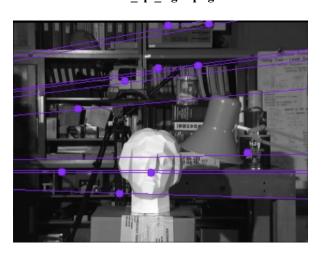


3.2.2 Fundamental Matrix F (with RANSAC)

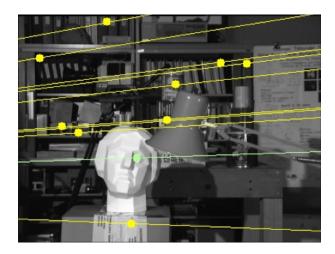
```
[[ 7.27595761e-11 -5.99488616e-04 5.31139374e-02]
[ 5.99607825e-04 1.90566061e-06 7.94587572e+11]
[-5.29851913e-02 -7.94587572e+11 1.00000000e+00]]
```

3.2.3 Computing Epiline

Task2_epi_right.png



Task2_epi_left.png



3.2.4 Disparity Map



Task2_disparity.png

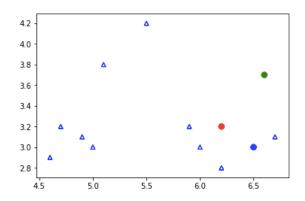
3.3 K-Means Clustering

3.3.1 Classifying Samples

```
======= Iteration : 0 ======== Clusters [[6.2 3.2]
[6.6 3.7]
[6.5 3. ]]
Assignments [0, 0, 2, 0, 1, 0, 0, 2, 0, 0]

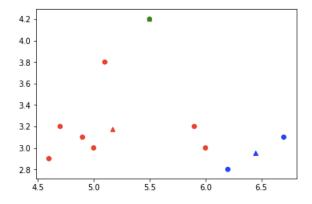
Assignments [0, 0, 2, 0, 1, 0, 0, 2, 0, 0]
['red', 'red', 'blue', 'red', 'green', 'red', 'red', 'red', 'red', 'red']
```

task3 iter1 a.png



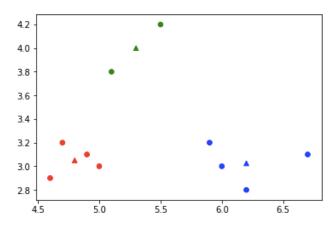
3.3.2 Compute Clusters for 1st Iteration

task3 iter1 b.png

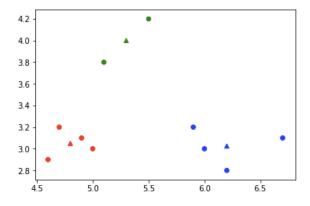


3.3.3 Compute Clusters for 2nd Iteration

task3 iter2 a.png



task3 iter2 b.png



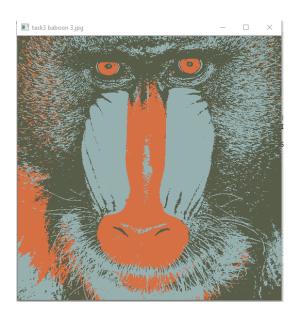
```
UBIT='ksehgal';
import random
import math
import numpy as np
np.random.seed(sum([ord(c) for c in UBIT]))
*matplotlib inline
from copy import deepcopy
import pandas as pd
from matplotlib import pyplot as plt
import warnings
#Euclidian Distance between two d-dimensional points
def eucldist(p0,p1):
      dist = 0.0
       for i in range(0,len(p0)):
             dist += (p0[i] - p1[i])**2
      return math.sqrt(dist)
        # d - Dimensionality of Datapoints
24
25
       d = len(datapoints[0])
       #Limit our iterations
Max_Iterations = 1000
i = 0
27
28
29
        cluster = [0] * len(datapoints)
       prev_cluster = [-1] * len(datapoints)
#Choose Centers for the Clusters
cluster_centers = []
31
32
34
35
       for i in range(0,k):
           cluster_centers= np.array([[6.2,3.2],[6.6,3.7],[6.5,3.0]])
36
            c1 x=cluster centers[:,:1]
            cl_y=cluster_centers[:,1:]
38
            f1=datapoints[:,:1]
39
            f2=datapoints[:,1:]
           plt.scatter(fl, f2,marker="^",facecolors='none',edgecolors='b', s=30)
colors=["red","green","blue"]
plt.scatter(cl_x, cl_y, marker=".", s=200, c=colors)
42
43
        while (cluster != prev_cluster) or (i > Max_Iterations):
           prev_cluster = list(cluster)
45
46
            #Update Point's Cluster
            for p in range(0,len(datapoints)):
49
50
               min_dist = float("inf")
#Check min distance against all centers
                for c in range(0,len(cluster_centers)):
52
                     dist = eucldist(datapoints[p],cluster_centers[c])
                    if (dist < min_dist):
    min_dist = dist</pre>
53
            cluster[p] = c # Reassign Point to new Cluster
print("====== Iteration :",i-3,"======")
print("Clusters",cluster_centers)
56
57
            print("Assignments", cluster)
```

Reference: https://gist.github.com/pmsosa/5454ade527adbee105dd51066ef30c5f

3.3.4 Image Quantization

K - Number of Clusters

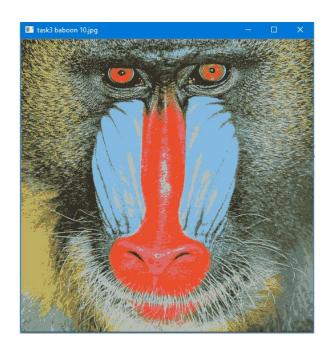
K=3 task3_baboon_3.png



K=5 task3_baboon_5.png



K=10 task3_baboon_10.jpg



K=20 task3_baboon_20.jpg

