INTRODUCTION TO ARTIFICIAL INTELLIGENCE

LAB ASSIGNMENT # 12

Adaptive Smoothing and HAC

PREPARED BY:

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PROBLEM STATEMENT 1:

Given the noisy image test_noisy.jpg (available in the laboratory_work folder). Use the adaptive smoothing (Ref. Diffusions and Confusions in Signal and Image Processing) process to remove the noise. Explain the selected values of the parameters in the algorithm.

Python Code:

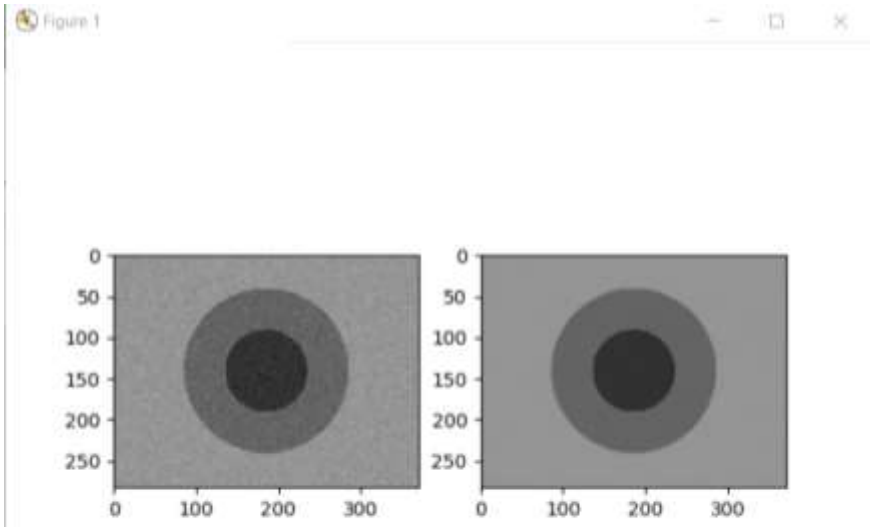
Commented [SY1]:

```
import numpy as np
import cv2
from matplotlib import pyplot as plt

def dist(i, j, k, l, kernel):
  return np.square(i - k) +  np.square(j - l) + np.square(kernel[i, j] -  kernel[k, l])

def get_sum_beta(i, j, dist_mat, gamma):
    m = dist_mat.shape[0]
    n = dist_mat.shape[1]
    s =0
    for k in range(m):
      for l in range(n):
          if i == k and l == j:
              s +=1 / (1 + np.power(dist_mat[i,j,k,l], gamma))
    return s

def get_beta(i, j , alpha, dist_mat, gamma):
    dist_sum = alpha / get_sum_beta(i, j, dist_mat,  gamma)
    return dist_sum

def get_wt(i, j, k, l, alpha, dist_mat, kernel):
    return get_beta(i, j, alpha, dist_mat, kernel) / (1 + np.power(dist_mat[i, j, k, l], gamma))

def create_distance_matrix(kernel):
    m,n = kernel.shape
    dist_mat = np.zeros((m,n,m,n))
    for i in range(m):
        for j in range(n):
            for k in range(m):
                for l in range( n):
                    dist_mat[i, j, k, l] = dist(i, j, k, l, kernel)
    return dist_mat

def create_weight_matrix(kernel, alpha, dist_mat):
    m,n = kernel.shape
```

```
    weight_mat = np.zeros((m,n,m,n))
    for i in range(m):
        for j in range(n):
            for k in range(m):
                for l in range( n):
                    weight_mat[i, j, k, l] = get_wt(i, j, k, l, alpha, dist_mat, kernel)
    return weight_mat

def adaptive_smoothing_images(kernel, alpha=0.2):
    dist_matrix = create_distance_matrix(kernel)
    weight_mat = create_weight_matrix(kernel, alpha, dist_mat)
    pass


kernel = np.array([[1,2,3],[4,5,6], [7,8,9]])
d = create_distance_matrix(kernel)
print(1 / d[0,0,:, :])



# Using cv2

img = cv2.imread('test_noisy.jpg')

dst = cv2.fastNlMeansDenoisingColored(img,None,10,10,7,21


plt.subplot(121),plt.imshow(img)
plt.subplot(122),plt.imshow(dst)
plt.show()
```

Output:

```
/usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:1: RuntimeWarning: divide by zero encountered in true_divide
  """Entry point for launching an IPython kernel.
array([[       inf, 0.5       , 0.125     ],
       [0.1       , 0.05555556, 0.03333333],
       [0.025     , 0.01851852, 0.01300000]])
```



PROBLEM STATEMENT 2:

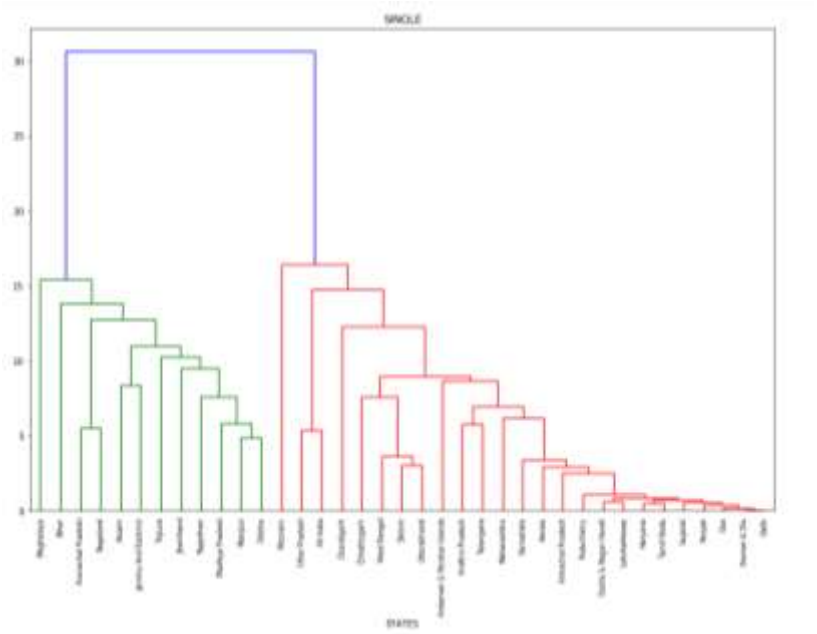
Use HAC with Euclidean/ Manhattan distance as a measure (Single link, complete link, Ward's distance, Group average, Centroid, Clusteroid) cluster the states of India based on the feature vector comprising of the following parameters (for one of the financial year values available in the data-set)
Percentage of schools with electricity
Percentage of schools with girls toilet
Percentage of schools with drinking water
Percentage of schools with boys toilet
For second problem you should get the data from the following:
https://data.gov.in/
As an example, a simple visualization based on percentage of schools with electricity is available at https://data.gov.in/major-indicator/percentage-schools-electricity

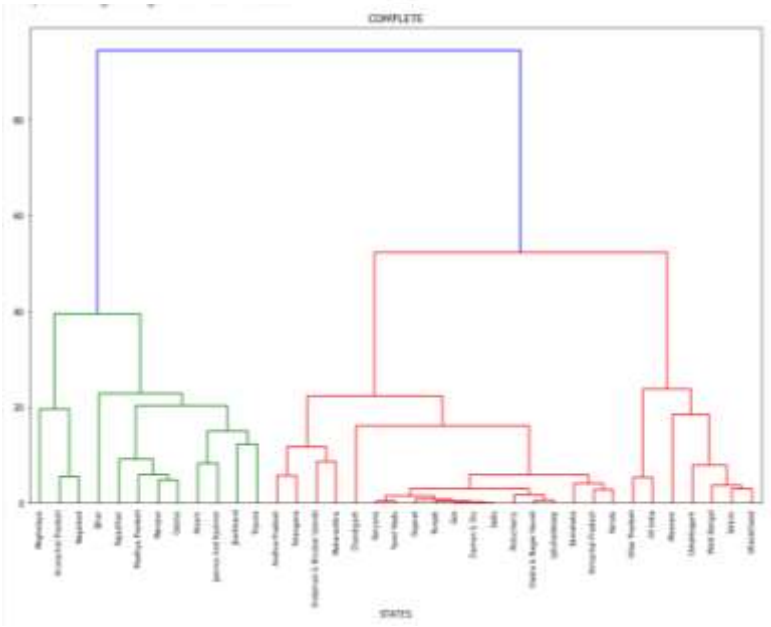
Code:

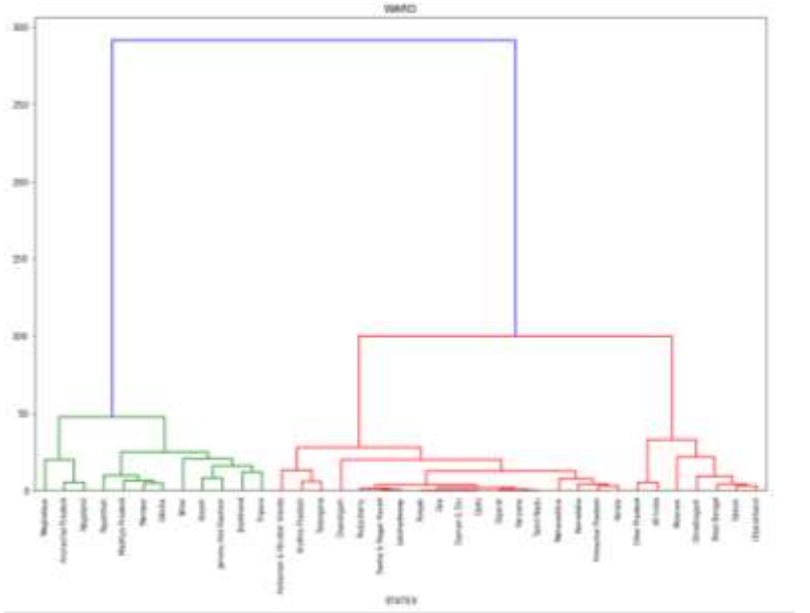
```
import pandas as pd
from scipy.cluster.hierarchy import dendrogram, linkage
import matplotlib.pyplot as plt
```

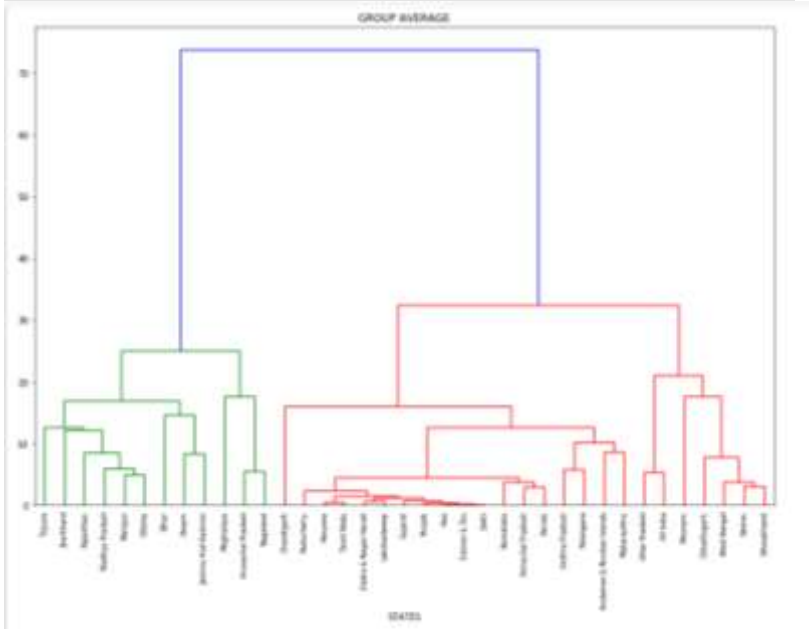
```
data = pd.read_csv('data.csv')
print(data.columns)

measures = ['single', 'complete','ward','average','centroid','median']
measureTitles = ['SINGLE', 'COMPLETE','WARD','GROUP AVERAGE','CENTROID','CLUSTEROID']

attribute=data.columns[1:]
for i in range(len(measures)):

    ytdist = data.set_index('State')[attribute]
    Z = hierarchy.linkage(ytdist, measures[i])

    labelList = data['State']
    fig, axes = plt.subplots(figsize=(15,10))
    dn = hierarchy.dendrogram(Z,orientation='top',ax=axes,distance_sort='descending',show_leaf_counts=False)
    labels = [data['State'][i] for i in dn['leaves']]
    axes.set_xticklabels(labels, rotation='vertical')
    plt.xlabel('STATES')
    plt.title(measureTitles[i])
    plt.show()
    plt.figure()
```

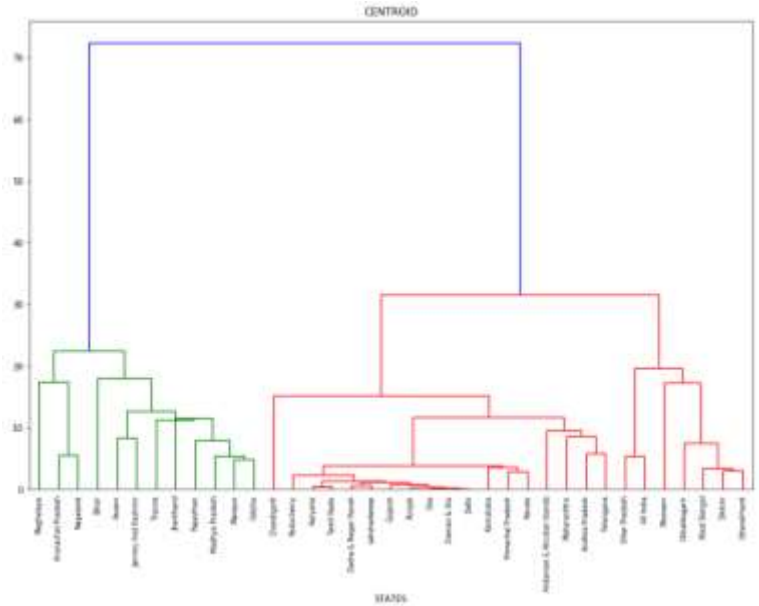
Outputs:

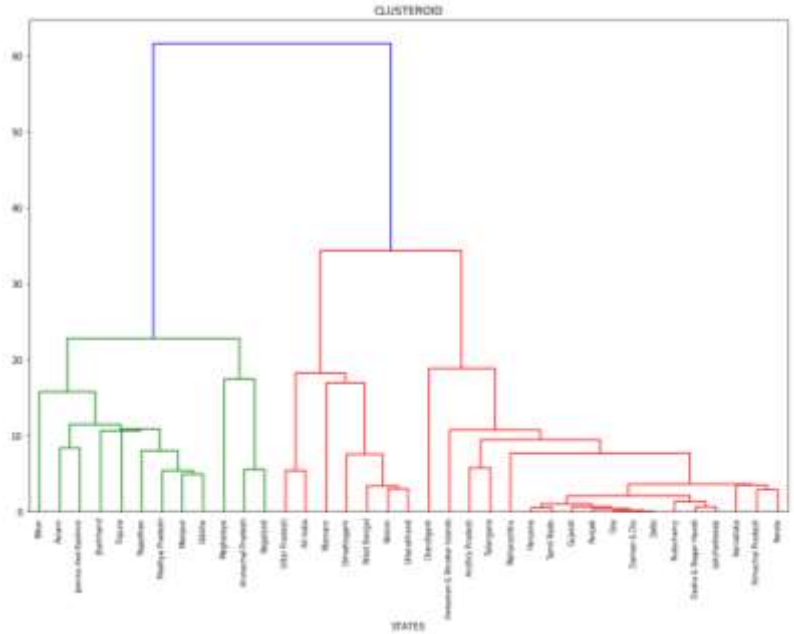

SINGLE
STATES


COMPLETE
STATES


WARD
STATES


GROUP AVERAGE
STATES


CENTROID
STATES


CLUSTEROID
STATES