import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns import scipy.cluster.hierarchy as sch from sklearn.cluster import AgglomerativeClustering

-----read dataset-----data = pd.read_csv('Downloads/crime_data.csv') data

	Unnamed: 0	Murder	Assault	UrbanPop	Rape
0	Alabama	13.2	236	58	21.2
1	Alaska	10.0	263	48	44.5
2	Arizona	8.1	294	80	31.0
3	Arkansas	8.8	190	50	19.5
4	California	9.0	276	91	40.6
5	Colorado	7.9	204	78	38.7
6	Connecticut	3.3	110	77	11.1
7	Delaware	5.9	238	72	15.8
8	Florida	15.4	335	80	31.9
9	Georgia	17.4	211	60	25.8
10	Hawaii	5.3	46	83	20.2
11	Idaho	2.6	120	54	14.2
12	Illinois	10.4	249	83	24.0
13	Indiana	7.2	113	65	21.0

-----normalize-----

def norm_func(i):

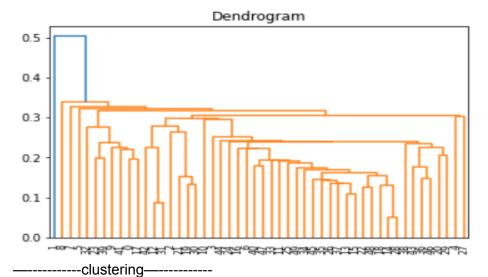
x = (i-i.min())/(i.max()-i.min())return (x)

-----divide-----

divide = norm_func(data.iloc[:,1:])
divide

	Murder	Assault	UrbanPop	Rape
0	0.746988	0.654110	0.440678	0.359173
1	0.554217	0.746575	0.271186	0.961240
2	0.439759	0.852740	0.813559	0.612403
3	0.481928	0.496575	0.305085	0.315245
4	0.493976	0.791096	1.000000	0.860465
5	0.427711	0.544521	0.779661	0.811370
6	0.150602	0.222603	0.762712	0.098191
7	0.307229	0.660959	0.677966	0.219638
8	0.879518	0.993151	0.813559	0.635659
9	1.000000	0.568493	0.474576	0.478036

Fext(0.5, 1.0, 'Dendrogram')



a = AgglomerativeClustering (n_clusters=4, affinity='euclidean', linkage='single')
a

AgglomerativeClustering(linkage='single', n_clusters=4)

------prediction-----predict = a.fit_predict(divide)
clusters = pd.DataFrame(predict,columns=['clusters'])
clusters

	clusters
0	0
1	3
2	0
3	0
4	0
5	0
6	0
7	1
8	2
9	0
10	0
11	0
12	0
13	0
14	0
15	0

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler

data = pd read_csv('Downloads/crime_data

data = pd.read_csv('Downloads/crime_data.csv')
data

	Unnamed: 0	Murder	Assault	UrbanPop	Rape
0	Alabama	13.2	236	58	21.2
1	Alaska	10.0	263	48	44.5
2	Arizona	8.1	294	80	31.0
3	Arkansas	8.8	190	50	19.5
4	California	9.0	276	91	40.6
5	Colorado	7.9	204	78	38.7
6	Connecticut	3.3	110	77	11.1
7	Delaware	5.9	238	72	15.8
8	Florida	15.4	335	80	31.9
9	Georgia	17.4	211	60	25.8
10	Hawaii	5.3	46	83	20.2
11	Idaho	2.6	120	54	14.2
12	Illinois	10.4	249	83	24.0
13	Indiana	7.2	113	65	21.0

------scale------

scaler = StandardScaler()

scaled = scaler.fit_transform(data.iloc[:,1:])

scaled

```
------function------
wss = []
for i in range (1,11):
  kmeans = KMeans (n_clusters=i)
  kmeans.fit(scaled)
  wss.append(kmeans.inertia_)
  -----plot-----
plt.plot(range(1,11),wss)
plt.title('Elbow Curve')
                           Elbow Curve
    200
    175
    150
    125
    100
     75
     50
     25
      ----clusters-
clusters = KMeans (n clusters=3)
clusters.fit(scaled)
KMeans(n clusters=3)
 clusters.labels
 array([1, 1, 1, 2, 1, 1, 0, 0, 1, 1, 0, 2, 1, 0, 2, 0, 2, 1, 2, 1, 0, 1,
        2, 1, 1, 2, 2, 1, 2, 0, 1, 1, 1, 2, 0, 0, 0, 0, 0, 1, 2, 1, 1, 0,
        2, 0, 0, 2, 2, 0])
data['New Clusters'] = clusters.labels
clusters.cluster_centers_
array([[-0.49440658, -0.3864845, 0.58167593, -0.26431024],
         [ 1.01513667, 1.02412028, 0.19959126, 0.85556386],
        [-0.88515915, -1.0213324, -0.94990286, -0.92016524]])
```

-----grouping-----

data.groupby('New Clusters').agg(['mean']).reset_index(drop=True)

	Murder	Assault	UrbanPop	Rape
	mean	mean	mean	mean
0	5.656250	138.875	73.875000	18.78125
1	12.165000	255.250	68.400000	29.16500
2	3.971429	86.500	51.928571	12.70000

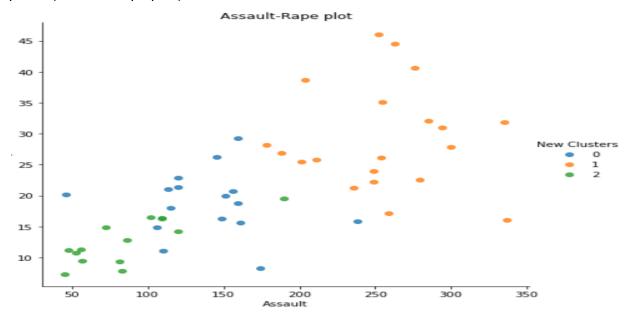
-----data-----

data

	Unnamed: 0	Murder	Assault	UrbanPop	Rape	New Clusters
0	Alabama	13.2	236	58	21.2	1
1	Alaska	10.0	263	48	44.5	1
2	Arizona	8.1	294	80	31.0	1
3	Arkansas	8.8	190	50	19.5	2
4	California	9.0	276	91	40.6	1
5	Colorado	7.9	204	78	38.7	1
6	Connecticut	3.3	110	77	11.1	0
7	Delaware	5.9	238	72	15.8	0

-----plot-----

sns.Implot('Assault','Rape',data=data,hue='New Clusters',fit_reg=False,size=6) plt.title('Assault-Rape plot')



-----dbscan-----

import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns

from sklearn.cluster import DBSCAN

from sklearn.preprocessing import StandardScaler

-----read data-----

data = pd.read_csv('Downloads/crime_data.csv')
data

	Unnamed: 0	Murder	Assault	UrbanPop	Rape
0	Alabama	13.2	236	58	21.2
1	Alaska	10.0	263	48	44.5
2	Arizona	8.1	294	80	31.0
3	Arkansas	8.8	190	50	19.5
4	California	9.0	276	91	40.6
5	Colorado	7.9	204	78	38.7
6	Connecticut	3.3	110	77	11.1
7	Delaware	5.9	238	72	15.8
8	Florida	15.4	335	80	31.9
9	Georgia	17.4	211	60	25.8
10	Hawaii	5.3	46	83	20.2
11	Idaho	2.6	120	54	14.2
12	Illinois	10.4	249	83	24.0
13	Indiana	7.2	113	65	21.0

_____divide_____

divide = data.iloc[:,1:]

divide

	Murder	Assault	UrbanPop	Rape
0	13.2	236	58	21.2
1	10.0	263	48	44.5
2	8.1	294	80	31.0
3	8.8	190	50	19.5
4	9.0	276	91	40.6
5	7.9	204	78	38.7
6	3.3	110	77	11.1
7	5.9	238	72	15.8
8	15.4	335	80	31.9
9	17.4	211	60	25.8
10	5.3	46	83	20.2
11	2.6	120	54	14.2
12	10.4	249	83	24.0

```
-----array-----
array = divide.values
array
array([[ 13.2, 236.,
                                         58.,
                                                      21.2],
                                         48.,
                                                      44.5],
                10., 263.
                                         80.,
                  8.1, 294.
                                                      31.],
                                         50.,
                  8.8, 190.
                                                      19.5],
                                          91.,
                           276.
                                                      40.6],
                                         78.,
                                                      38.7],
                  7.9, 204.
                                         77.,
                  3.3, 110.
                                                      11.1],
                                         72.,
                  5.9, 238.
                                                      15.8],
                                         80.,
                 15.4, 335.
                                                      31.9],
                 17.4, 211.
                                         60.,
                                                      25.8],
                                         83.,
                                                      20.2],
                  5.3,
                            46.
                  2.6, 120.
                                         54.
                                                      14.2],
                                         83.,
                                                      24. ],
21. ],
                10.4, 249.
                                          65.,
                   7.2, 113.
                                         57.,
                                                      11.3],
                  2.2,
                            56.
                  6. , 115.
                                         66.,
                                                      18.],
                                         52.,
                  9.7, 109.
                                                      16.3],
                                          66.,
                                                      22.2],
                 15.4, 249.
                                                       7.81.
                             83.
                                         51. .
   -----scaler-
scaler = StandardScaler()
scaled = scaler.fit transform(array)
scaled
array([[ 1.25517927,
                                       0.79078716, -0.52619514, -0.00345116],
                                       1.11805959, -1.22406668,
                0.51301858,
                                                                                     2.50942392],
                                      1.49381682,
                                                             1.00912225,
                                                                                    1.05346626],
               0.07236067,
             [ 0.23470832,
                                      0.23321191, -1.08449238, -0.18679398],
                0.28109336,
                                       1.2756352 , 1.77678094,
                                                                                     2.08881393],
                0.02597562,
                                      0.40290872,
                                                              0.86954794,
                                                                                     1.88390137],
             [-1.04088037, -0.73648418,
                                                             0.79976079, -1.09272319],
             [-0.43787481,
                                      0.81502956,
                                                             0.45082502, -0.58583422],
             [ 1.76541475,
                                       1.99078607,
                                                             1.00912225,
                                                                                    1.1505301 ]
                                                                                    0.49265293],
                2.22926518,
                                      0.48775713, -0.38662083,
             [-0.57702994, -1.51224105,
                                                              1.21848371, -0.11129987],
             [-1.20322802, -0.61527217, -0.80534376, -0.75839217],
             [ 0.60578867,
                                      0.94836277,
                                                             1.21848371,
                                                                                    0.29852525],
             [-0.13637203,
                                     -0.70012057, -0.03768506,
                                                                                    -0.0250209
             [-1.29599811, -1.39102904, -0.5959823 , -1.07115345],
             [-0.41468229, -0.67587817,
                                                             0.03210209, -0.34856705],
             [ 0.44344101, -0.74860538, -0.94491807, -0.53190987],
                                      0.94836277, 0.03210209,
             1.76541475,
                                                                                    0.10439756],
  -----dbscan–
dbscan = DBSCAN (eps=0.6,min samples=16)
dbscan.fit(scaled)
  DBSCAN(eps=0.6, min samples=16)
   ----labels-
dbscan.labels
-1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1,
           -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1],
          dtype=int64)
```

------dataframe------clusters = pd.DataFrame(dbscan.labels_,columns=['clusters']) clusters

clusters

	Clusters
0	-1
1	-1
2	-1
3	-1
4	-1
5	-1
6	-1
7	-1
8	-1
9	-1
10	-1
44	4