

```
In [1]: import scipy.cluster.hierarchy as sch
import numpy as np
import pandas as pd
from matplotlib import pyplot as plt
import seaborn as sns
```

```
In [2]: crime=pd.read_csv("C:\\Users\\Admin\\Downloads\\assignment 4\\crime_data.csv")
crime.head()
```

Out[2]:

	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

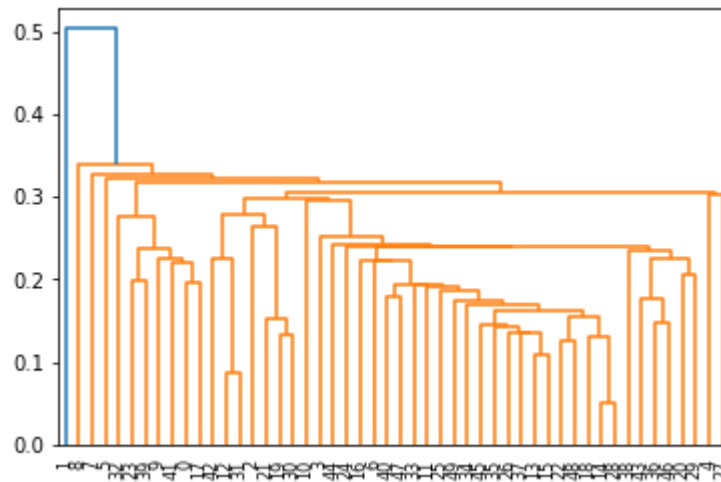
```
In [3]: crime.shape
```

Out[3]: (50, 5)

```
In [4]: def norm_func(i):
x= (i-i.min())/(i.max()-i.min())
return(x)
```

```
In [5]: df_norm =norm_func(crime.iloc[:,1:])
```

```
In [6]: dendrogram=sch.dendrogram(sch.linkage(df_norm,method='single'))
```



```
In [7]: from sklearn.cluster import AgglomerativeClustering
```

```
In [8]: hc=AgglomerativeClustering(n_clusters=2, affinity='euclidean',linkage='single')
```

```
In [9]: y_hc=hc.fit_predict(df_norm)  
Clusters=pd.DataFrame(y_hc,columns=['Clusters'])
```

```
In [10]: Clusters
```

Out[10]:

Clusters	
0	0
1	1
2	0
3	0
4	0
5	0
6	0
7	0
8	0
9	0
10	0
11	0
12	0
13	0
14	0
15	0
16	0
17	0
18	0
19	0
20	0
21	0
22	0
23	0
24	0

Clusters	
25	0
26	0
27	0
28	0
29	0
30	0
31	0
32	0
33	0
34	0
35	0
36	0
37	0
38	0
39	0
40	0
41	0
42	0
43	0
44	0
45	0
46	0
47	0
48	0
49	0

```
In [13]: crime['h_clusterid']=Clusters
```

In [14]: crime

Out[14]:

	Unnamed: 0	Murder	Assault	UrbanPop	Rape	h_clusterid
0	Alabama	13.2	236	58	21.2	0
1	Alaska	10.0	263	48	44.5	1
2	Arizona	8.1	294	80	31.0	0
3	Arkansas	8.8	190	50	19.5	0
4	California	9.0	276	91	40.6	0
5	Colorado	7.9	204	78	38.7	0
6	Connecticut	3.3	110	77	11.1	0
7	Delaware	5.9	238	72	15.8	0
8	Florida	15.4	335	80	31.9	0
9	Georgia	17.4	211	60	25.8	0
10	Hawaii	5.3	46	83	20.2	0
11	Idaho	2.6	120	54	14.2	0
12	Illinois	10.4	249	83	24.0	0
13	Indiana	7.2	113	65	21.0	0
14	Iowa	2.2	56	57	11.3	0
15	Kansas	6.0	115	66	18.0	0
16	Kentucky	9.7	109	52	16.3	0
17	Louisiana	15.4	249	66	22.2	0
18	Maine	2.1	83	51	7.8	0
19	Maryland	11.3	300	67	27.8	0
20	Massachusetts	4.4	149	85	16.3	0
21	Michigan	12.1	255	74	35.1	0
22	Minnesota	2.7	72	66	14.9	0
23	Mississippi	16.1	259	44	17.1	0
24	Missouri	9.0	178	70	28.2	0

	Unnamed: 0	Murder	Assault	UrbanPop	Rape	h_clusterid
25	Montana	6.0	109	53	16.4	0
26	Nebraska	4.3	102	62	16.5	0
27	Nevada	12.2	252	81	46.0	0
28	New Hampshire	2.1	57	56	9.5	0
29	New Jersey	7.4	159	89	18.8	0
30	New Mexico	11.4	285	70	32.1	0
31	New York	11.1	254	86	26.1	0
32	North Carolina	13.0	337	45	16.1	0
33	North Dakota	0.8	45	44	7.3	0
34	Ohio	7.3	120	75	21.4	0
35	Oklahoma	6.6	151	68	20.0	0
36	Oregon	4.9	159	67	29.3	0
37	Pennsylvania	6.3	106	72	14.9	0
38	Rhode Island	3.4	174	87	8.3	0
39	South Carolina	14.4	279	48	22.5	0
40	South Dakota	3.8	86	45	12.8	0
41	Tennessee	13.2	188	59	26.9	0
42	Texas	12.7	201	80	25.5	0
43	Utah	3.2	120	80	22.9	0
44	Vermont	2.2	48	32	11.2	0
45	Virginia	8.5	156	63	20.7	0
46	Washington	4.0	145	73	26.2	0
47	West Virginia	5.7	81	39	9.3	0
48	Wisconsin	2.6	53	66	10.8	0
49	Wyoming	6.8	161	60	15.6	0

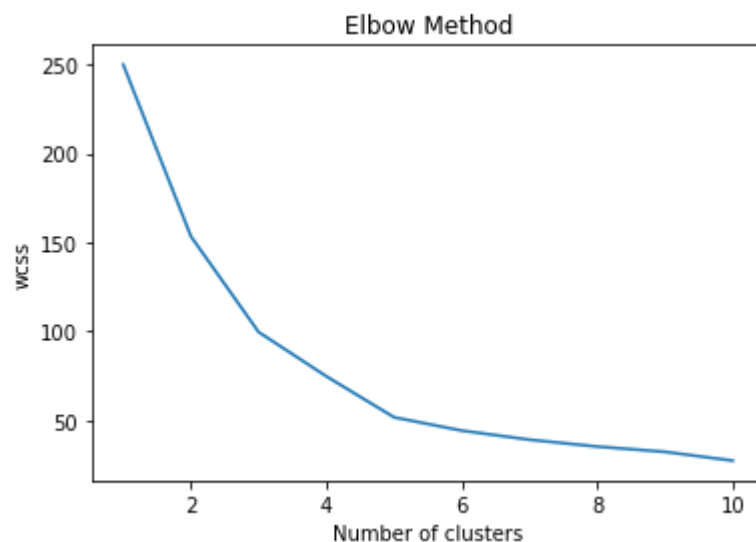
```
In [16]: from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
scaled_crime_df=scaler.fit_transform(crime.iloc[:,1:])
```

```
In [17]: wcss=[]
for i in range(1, 11):
    kmeans=KMeans(n_clusters=i,random_state=0)
    kmeans.fit(scaled_crime_df)
    wcss.append(kmeans.inertia_)

plt.plot(range(1,11),wcss)
plt.title('Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('wcss')
plt.show()
```

C:\Users\Admin\anaconda3\lib\site-packages\sklearn\cluster\\_kmeans.py:881: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

```
warnings.warn(
```





```
In [19]: from sklearn.cluster import KMeans  
clusters_new=KMeans(4,random_state=42)  
clusters_new.fit(scaled_crime_df)
```

```
Out[19]: KMeans(n_clusters=4, random_state=42)
```

```
In [21]: clusters_new.labels_
```

```
Out[21]: array([3, 2, 3, 1, 3, 3, 0, 0, 3, 3, 0, 1, 3, 0, 1, 0, 1, 3, 1, 3, 0, 3,  
                1, 3, 3, 1, 1, 3, 1, 0, 3, 3, 3, 1, 0, 0, 0, 0, 0, 3, 1, 3, 3, 0,  
                1, 0, 0, 1, 1, 0])
```

```
In [22]: crime['clusterid_new']=clusters_new.labels_
```

In [23]: crime

Out[23]:

	Unnamed: 0	Murder	Assault	UrbanPop	Rape	h_clusterid	clusterid_new
0	Alabama	13.2	236	58	21.2	0	3
1	Alaska	10.0	263	48	44.5	1	2
2	Arizona	8.1	294	80	31.0	0	3
3	Arkansas	8.8	190	50	19.5	0	1
4	California	9.0	276	91	40.6	0	3
5	Colorado	7.9	204	78	38.7	0	3
6	Connecticut	3.3	110	77	11.1	0	0
7	Delaware	5.9	238	72	15.8	0	0
8	Florida	15.4	335	80	31.9	0	3
9	Georgia	17.4	211	60	25.8	0	3
10	Hawaii	5.3	46	83	20.2	0	0
11	Idaho	2.6	120	54	14.2	0	1
12	Illinois	10.4	249	83	24.0	0	3
13	Indiana	7.2	113	65	21.0	0	0
14	Iowa	2.2	56	57	11.3	0	1
15	Kansas	6.0	115	66	18.0	0	0
16	Kentucky	9.7	109	52	16.3	0	1
17	Louisiana	15.4	249	66	22.2	0	3
18	Maine	2.1	83	51	7.8	0	1
19	Maryland	11.3	300	67	27.8	0	3
20	Massachusetts	4.4	149	85	16.3	0	0
21	Michigan	12.1	255	74	35.1	0	3
22	Minnesota	2.7	72	66	14.9	0	1
23	Mississippi	16.1	259	44	17.1	0	3
24	Missouri	9.0	178	70	28.2	0	3

	Unnamed: 0	Murder	Assault	UrbanPop	Rape	h_clusterid	clusterid_new
25	Montana	6.0	109	53	16.4	0	1
26	Nebraska	4.3	102	62	16.5	0	1
27	Nevada	12.2	252	81	46.0	0	3
28	New Hampshire	2.1	57	56	9.5	0	1
29	New Jersey	7.4	159	89	18.8	0	0
30	New Mexico	11.4	285	70	32.1	0	3
31	New York	11.1	254	86	26.1	0	3
32	North Carolina	13.0	337	45	16.1	0	3
33	North Dakota	0.8	45	44	7.3	0	1
34	Ohio	7.3	120	75	21.4	0	0
35	Oklahoma	6.6	151	68	20.0	0	0
36	Oregon	4.9	159	67	29.3	0	0
37	Pennsylvania	6.3	106	72	14.9	0	0
38	Rhode Island	3.4	174	87	8.3	0	0
39	South Carolina	14.4	279	48	22.5	0	3
40	South Dakota	3.8	86	45	12.8	0	1
41	Tennessee	13.2	188	59	26.9	0	3
42	Texas	12.7	201	80	25.5	0	3
43	Utah	3.2	120	80	22.9	0	0
44	Vermont	2.2	48	32	11.2	0	1
45	Virginia	8.5	156	63	20.7	0	0
46	Washington	4.0	145	73	26.2	0	0
47	West Virginia	5.7	81	39	9.3	0	1
48	Wisconsin	2.6	53	66	10.8	0	1
49	Wyoming	6.8	161	60	15.6	0	0

```
In [25]: clusters_new.cluster_centers_
```

```
Out[25]: array([[ -0.49440658, -0.3864845 ,  0.58167593, -0.26431024, -0.14285714],  
                [ -0.88515915, -1.0213324 , -0.94990286, -0.92016524, -0.14285714],  
                [  0.51301858,  1.11805959, -1.22406668,  2.50942392,  7.          ],  
                [  1.04156394,  1.01917611,  0.27452063,  0.7685186 , -0.14285714]])
```

```
In [26]: crime.groupby('clusterid_new').agg(['mean']).reset_index()
```

```
Out[26]:
```

	clusterid_new	Murder	Assault	UrbanPop	Rape	h_clusterid
		mean	mean	mean	mean	mean
0	0	5.656250	138.875000	73.875000	18.781250	0.0
1	1	3.971429	86.500000	51.928571	12.700000	0.0
2	2	10.000000	263.000000	48.000000	44.500000	1.0
3	3	12.278947	254.842105	69.473684	28.357895	0.0

In [27]: crime

Out[27]:

	Unnamed: 0	Murder	Assault	UrbanPop	Rape	h_clusterid	clusterid_new
0	Alabama	13.2	236	58	21.2	0	3
1	Alaska	10.0	263	48	44.5	1	2
2	Arizona	8.1	294	80	31.0	0	3
3	Arkansas	8.8	190	50	19.5	0	1
4	California	9.0	276	91	40.6	0	3
5	Colorado	7.9	204	78	38.7	0	3
6	Connecticut	3.3	110	77	11.1	0	0
7	Delaware	5.9	238	72	15.8	0	0
8	Florida	15.4	335	80	31.9	0	3
9	Georgia	17.4	211	60	25.8	0	3
10	Hawaii	5.3	46	83	20.2	0	0
11	Idaho	2.6	120	54	14.2	0	1
12	Illinois	10.4	249	83	24.0	0	3
13	Indiana	7.2	113	65	21.0	0	0
14	Iowa	2.2	56	57	11.3	0	1
15	Kansas	6.0	115	66	18.0	0	0
16	Kentucky	9.7	109	52	16.3	0	1
17	Louisiana	15.4	249	66	22.2	0	3
18	Maine	2.1	83	51	7.8	0	1
19	Maryland	11.3	300	67	27.8	0	3
20	Massachusetts	4.4	149	85	16.3	0	0
21	Michigan	12.1	255	74	35.1	0	3
22	Minnesota	2.7	72	66	14.9	0	1
23	Mississippi	16.1	259	44	17.1	0	3
24	Missouri	9.0	178	70	28.2	0	3

	Unnamed: 0	Murder	Assault	UrbanPop	Rape	h_clusterid	clusterid_new
25	Montana	6.0	109	53	16.4	0	1
26	Nebraska	4.3	102	62	16.5	0	1
27	Nevada	12.2	252	81	46.0	0	3
28	New Hampshire	2.1	57	56	9.5	0	1
29	New Jersey	7.4	159	89	18.8	0	0
30	New Mexico	11.4	285	70	32.1	0	3
31	New York	11.1	254	86	26.1	0	3
32	North Carolina	13.0	337	45	16.1	0	3
33	North Dakota	0.8	45	44	7.3	0	1
34	Ohio	7.3	120	75	21.4	0	0
35	Oklahoma	6.6	151	68	20.0	0	0
36	Oregon	4.9	159	67	29.3	0	0
37	Pennsylvania	6.3	106	72	14.9	0	0
38	Rhode Island	3.4	174	87	8.3	0	0
39	South Carolina	14.4	279	48	22.5	0	3
40	South Dakota	3.8	86	45	12.8	0	1
41	Tennessee	13.2	188	59	26.9	0	3
42	Texas	12.7	201	80	25.5	0	3
43	Utah	3.2	120	80	22.9	0	0
44	Vermont	2.2	48	32	11.2	0	1
45	Virginia	8.5	156	63	20.7	0	0
46	Washington	4.0	145	73	26.2	0	0
47	West Virginia	5.7	81	39	9.3	0	1
48	Wisconsin	2.6	53	66	10.8	0	1
49	Wyoming	6.8	161	60	15.6	0	0

```
In [28]: from sklearn .cluster import DBSCAN
```

```
In [32]: crime.drop(['Unnamed: 0'],axis=1,inplace=True)
```

```
In [33]: array=crime.values
```

In [34]: array

```

Out[34]: array([[ 13.2, 236. , 58. , 21.2, 0. , 3. ],
 [ 10. , 263. , 48. , 44.5, 1. , 2. ],
 [ 8.1, 294. , 80. , 31. , 0. , 3. ],
 [ 8.8, 190. , 50. , 19.5, 0. , 1. ],
 [ 9. , 276. , 91. , 40.6, 0. , 3. ],
 [ 7.9, 204. , 78. , 38.7, 0. , 3. ],
 [ 3.3, 110. , 77. , 11.1, 0. , 0. ],
 [ 5.9, 238. , 72. , 15.8, 0. , 0. ],
 [ 15.4, 335. , 80. , 31.9, 0. , 3. ],
 [ 17.4, 211. , 60. , 25.8, 0. , 3. ],
 [ 5.3, 46. , 83. , 20.2, 0. , 0. ],
 [ 2.6, 120. , 54. , 14.2, 0. , 1. ],
 [ 10.4, 249. , 83. , 24. , 0. , 3. ],
 [ 7.2, 113. , 65. , 21. , 0. , 0. ],
 [ 2.2, 56. , 57. , 11.3, 0. , 1. ],
 [ 6. , 115. , 66. , 18. , 0. , 0. ],
 [ 9.7, 109. , 52. , 16.3, 0. , 1. ],
 [ 15.4, 249. , 66. , 22.2, 0. , 3. ],
 [ 2.1, 83. , 51. , 7.8, 0. , 1. ],
 [ 11.3, 300. , 67. , 27.8, 0. , 3. ],
 [ 4.4, 149. , 85. , 16.3, 0. , 0. ],
 [ 12.1, 255. , 74. , 35.1, 0. , 3. ],
 [ 2.7, 72. , 66. , 14.9, 0. , 1. ],
 [ 16.1, 259. , 44. , 17.1, 0. , 3. ],
 [ 9. , 178. , 70. , 28.2, 0. , 3. ],
 [ 6. , 109. , 53. , 16.4, 0. , 1. ],
 [ 4.3, 102. , 62. , 16.5, 0. , 1. ],
 [ 12.2, 252. , 81. , 46. , 0. , 3. ],
 [ 2.1, 57. , 56. , 9.5, 0. , 1. ],
 [ 7.4, 159. , 89. , 18.8, 0. , 0. ],
 [ 11.4, 285. , 70. , 32.1, 0. , 3. ],
 [ 11.1, 254. , 86. , 26.1, 0. , 3. ],
 [ 13. , 337. , 45. , 16.1, 0. , 3. ],
 [ 0.8, 45. , 44. , 7.3, 0. , 1. ],
 [ 7.3, 120. , 75. , 21.4, 0. , 0. ],
 [ 6.6, 151. , 68. , 20. , 0. , 0. ],
 [ 4.9, 159. , 67. , 29.3, 0. , 0. ],
 [ 6.3, 106. , 72. , 14.9, 0. , 0. ],
 [ 3.4, 174. , 87. , 8.3, 0. , 0. ],
 [ 14.4, 279. , 48. , 22.5, 0. , 3. ],
 [ 3.8, 86. , 45. , 12.8, 0. , 1. ],

```



```
[ 13.2, 188. , 59. , 26.9, 0. , 3. ],  
[ 12.7, 201. , 80. , 25.5, 0. , 3. ],  
[ 3.2, 120. , 80. , 22.9, 0. , 0. ],  
[ 2.2, 48. , 32. , 11.2, 0. , 1. ],  
[ 8.5, 156. , 63. , 20.7, 0. , 0. ],  
[ 4. , 145. , 73. , 26.2, 0. , 0. ],  
[ 5.7, 81. , 39. , 9.3, 0. , 1. ],  
[ 2.6, 53. , 66. , 10.8, 0. , 1. ],  
[ 6.8, 161. , 60. , 15.6, 0. , 0. ]])
```

```
In [35]: stscaler=StandardScaler().fit(array)  
x=stscaler.transform(array)
```

In [36]: x

```

Out[36]: array([[ 1.25517927e+00,  7.90787158e-01, -5.26195142e-01,
                  -3.45115891e-03, -1.42857143e-01,  1.19947008e+00],
                 [ 5.13018579e-01,  1.11805959e+00, -1.22406668e+00,
                  2.50942392e+00,  7.00000000e+00,  4.20593403e-01],
                 [ 7.23606675e-02,  1.49381682e+00,  1.00912225e+00,
                  1.05346626e+00, -1.42857143e-01,  1.19947008e+00],
                 [ 2.34708319e-01,  2.33211909e-01, -1.08449238e+00,
                  -1.86793976e-01, -1.42857143e-01, -3.58283270e-01],
                 [ 2.81093362e-01,  1.27563520e+00,  1.77678094e+00,
                  2.08881393e+00, -1.42857143e-01,  1.19947008e+00],
                 [ 2.59756242e-02,  4.02908724e-01,  8.69547941e-01,
                  1.88390137e+00, -1.42857143e-01,  1.19947008e+00],
                 [-1.04088037e+00, -7.36484178e-01,  7.99760786e-01,
                  -1.09272319e+00, -1.42857143e-01, -1.13715994e+00],
                 [-4.37874809e-01,  8.15029561e-01,  4.50825016e-01,
                  -5.85834225e-01, -1.42857143e-01, -1.13715994e+00],
                 [ 1.76541475e+00,  1.99078607e+00,  1.00912225e+00,
                  1.15053010e+00, -1.42857143e-01,  1.19947008e+00],
                 [ 2.22926518e+00,  4.87757131e-01, -3.86620834e-01,
                  4.92652934e-01, -1.42857143e-01,  1.19947008e+00],
                 [-5.77029938e-01, -1.51224105e+00,  1.21848371e+00,
                  -1.11299875e-01, -1.42857143e-01, -1.13715994e+00],
                 [-1.20322802e+00, -6.15272167e-01, -8.05343759e-01,
                  -7.58392170e-01, -1.42857143e-01, -3.58283270e-01],
                 [ 6.05788665e-01,  9.48362772e-01,  1.21848371e+00,
                  2.98525246e-01, -1.42857143e-01,  1.19947008e+00],
                 [-1.36372027e-01, -7.00120574e-01, -3.76850632e-02,
                  -2.50209021e-02, -1.42857143e-01, -1.13715994e+00],
                 [-1.29599811e+00, -1.39102904e+00, -5.95982296e-01,
                  -1.07115345e+00, -1.42857143e-01, -3.58283270e-01],
                 [-4.14682287e-01, -6.75878172e-01,  3.21020909e-02,
                  -3.48567050e-01, -1.42857143e-01, -1.13715994e+00],
                 [ 4.43441014e-01, -7.48605379e-01, -9.44918067e-01,
                  -5.31909867e-01, -1.42857143e-01, -3.58283270e-01],
                 [ 1.76541475e+00,  9.48362772e-01,  3.21020909e-02,
                  1.04397557e-01, -1.42857143e-01,  1.19947008e+00],
                 [-1.31919063e+00, -1.06375661e+00, -1.01470522e+00,
                  -1.44862395e+00, -1.42857143e-01, -3.58283270e-01],
                 [ 8.14521360e-01,  1.56654403e+00,  1.01889245e-01,
                  7.08350366e-01, -1.42857143e-01,  1.19947008e+00],
                 [-7.85762633e-01, -2.63757335e-01,  1.35805802e+00,

```

```
-5.31909867e-01, -1.42857143e-01, -1.13715994e+00],  
[ 1.00006153e+00,  1.02108998e+00,  5.90399324e-01,  
  1.49564599e+00, -1.42857143e-01,  1.19947008e+00],  
[-1.18003550e+00, -1.19708982e+00,  3.21020909e-02,  
 -6.82898069e-01, -1.42857143e-01, -3.58283270e-01],  
[ 1.92776240e+00,  1.06957478e+00, -1.50321530e+00,  
 -4.45630894e-01, -1.42857143e-01,  1.19947008e+00],  
[ 2.81093362e-01,  8.77574958e-02,  3.11250707e-01,  
  7.51489853e-01, -1.42857143e-01,  1.19947008e+00],  
[-4.14682287e-01, -7.48605379e-01, -8.75130913e-01,  
 -5.21124995e-01, -1.42857143e-01, -3.58283270e-01],  
[-8.08955155e-01, -8.33453786e-01, -2.47046526e-01,  
 -5.10340124e-01, -1.42857143e-01, -3.58283270e-01],  
[ 1.02325405e+00,  9.84726376e-01,  1.07890940e+00,  
  2.67119700e+00, -1.42857143e-01,  1.19947008e+00],  
[-1.31919063e+00, -1.37890783e+00, -6.65769450e-01,  
 -1.26528114e+00, -1.42857143e-01, -3.58283270e-01],  
[-8.99869840e-02, -1.42545325e-01,  1.63720664e+00,  
 -2.62288077e-01, -1.42857143e-01, -1.13715994e+00],  
[ 8.37713882e-01,  1.38472601e+00,  3.11250707e-01,  
  1.17209984e+00, -1.42857143e-01,  1.19947008e+00],  
[ 7.68136317e-01,  1.00896878e+00,  1.42784517e+00,  
  5.25007549e-01, -1.42857143e-01,  1.19947008e+00],  
[ 1.20879423e+00,  2.01502847e+00, -1.43342815e+00,  
 -5.53479610e-01, -1.42857143e-01,  1.19947008e+00],  
[-1.62069341e+00, -1.52436225e+00, -1.50321530e+00,  
 -1.50254831e+00, -1.42857143e-01, -3.58283270e-01],  
[-1.13179506e-01, -6.15272167e-01,  6.60186478e-01,  
  1.81185843e-02, -1.42857143e-01, -1.13715994e+00],  
[-2.75527157e-01, -2.39514933e-01,  1.71676399e-01,  
 -1.32869618e-01, -1.42857143e-01, -1.13715994e+00],  
[-6.69800025e-01, -1.42545325e-01,  1.01889245e-01,  
  8.70123440e-01, -1.42857143e-01, -1.13715994e+00],  
[-3.45104722e-01, -7.84968982e-01,  4.50825016e-01,  
 -6.82898069e-01, -1.42857143e-01, -1.13715994e+00],  
[-1.01768785e+00,  3.92726915e-02,  1.49763233e+00,  
 -1.39469959e+00, -1.42857143e-01, -1.13715994e+00],  
[ 1.53348953e+00,  1.31199880e+00, -1.22406668e+00,  
  1.36752172e-01, -1.42857143e-01,  1.19947008e+00],  
[-9.24917763e-01, -1.02739300e+00, -1.43342815e+00,  
 -9.09380373e-01, -1.42857143e-01, -3.58283270e-01],  
[ 1.25517927e+00,  2.08969507e-01, -4.56407988e-01,  
  6.11286522e-01, -1.42857143e-01,  1.19947008e+00],
```

```
[ 1.13921666e+00,  3.66545121e-01,  1.00912225e+00,
  4.60298320e-01, -1.42857143e-01,  1.19947008e+00],
 [-1.06407289e+00, -6.15272167e-01,  1.00912225e+00,
  1.79891658e-01, -1.42857143e-01, -1.13715994e+00],
 [-1.29599811e+00, -1.48799864e+00, -2.34066115e+00,
 -1.08193832e+00, -1.42857143e-01, -3.58283270e-01],
 [ 1.65130754e-01, -1.78908928e-01, -1.77259372e-01,
 -5.73755169e-02, -1.42857143e-01, -1.13715994e+00],
 [-8.78532720e-01, -3.12242140e-01,  5.20612170e-01,
  5.35792421e-01, -1.42857143e-01, -1.13715994e+00],
 [-4.84259852e-01, -1.08799901e+00, -1.85215107e+00,
 -1.28685088e+00, -1.42857143e-01, -3.58283270e-01],
 [-1.20322802e+00, -1.42739264e+00,  3.21020909e-02,
 -1.12507780e+00, -1.42857143e-01, -3.58283270e-01],
 [-2.29142114e-01, -1.18302923e-01, -3.86620834e-01,
 -6.07403968e-01, -1.42857143e-01, -1.13715994e+00]])
```

```
In [37]: dbscan=DBSCAN(eps=0.8,min_samples=6)
         dbscan.fit(x)
```

```
Out[37]: DBSCAN(eps=0.8, min_samples=6)
```

```
In [38]: dbscan.labels_
```

```
Out[38]: array([-1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1,  0, -1,  0, -1,
               -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1,
                0,  0, -1,  0, -1, -1, -1, -1, -1, -1, -1,  0, -1, -1, -1,  0],
              dtype=int64)
```

```
In [39]: c1=pd.DataFrame(dbscan.labels_,columns=['clusters'])
```

In [40]: c1

Out[40]:

clusters	
0	-1
1	-1
2	-1
3	-1
4	-1
5	-1
6	-1
7	-1
8	-1
9	-1
10	-1
11	-1
12	-1
13	0
14	-1
15	0
16	-1
17	-1
18	-1
19	-1
20	-1
21	-1
22	-1
23	-1
24	-1

clusters	
25	-1
26	-1
27	-1
28	-1
29	-1
30	-1
31	-1
32	-1
33	-1
34	0
35	0
36	-1
37	0
38	-1
39	-1
40	-1
41	-1
42	-1
43	-1
44	-1
45	0
46	-1
47	-1
48	-1
49	0

```
In [42]: pd.concat([crime,c1],axis=1)
```

```
Out[42]:
```

	Murder	Assault	UrbanPop	Rape	h_clusterid	clusterid_new	clusters
0	13.2	236	58	21.2	0	3	-1
1	10.0	263	48	44.5	1	2	-1
2	8.1	294	80	31.0	0	3	-1
3	8.8	190	50	19.5	0	1	-1
4	9.0	276	91	40.6	0	3	-1
5	7.9	204	78	38.7	0	3	-1
6	3.3	110	77	11.1	0	0	-1
7	5.9	238	72	15.8	0	0	-1
8	15.4	335	80	31.9	0	3	-1
9	17.4	211	60	25.8	0	3	-1
10	5.3	46	83	20.2	0	0	-1
11	2.6	120	54	14.2	0	1	-1
12	10.4	249	83	24.0	0	3	-1
13	7.2	113	65	21.0	0	0	0
14	2.2	56	57	11.3	0	1	-1
15	6.0	115	66	18.0	0	0	0
16	9.7	109	52	16.3	0	1	-1
17	15.4	249	66	22.2	0	3	-1
18	2.1	83	51	7.8	0	1	-1
19	11.3	300	67	27.8	0	3	-1
20	4.4	149	85	16.3	0	0	-1
21	12.1	255	74	35.1	0	3	-1
22	2.7	72	66	14.9	0	1	-1
23	16.1	259	44	17.1	0	3	-1

	<b>Murder</b>	<b>Assault</b>	<b>UrbanPop</b>	<b>Rape</b>	<b>h_clusterid</b>	<b>clusterid_new</b>	<b>clusters</b>
<b>24</b>	9.0	178	70	28.2	0	3	-1
<b>25</b>	6.0	109	53	16.4	0	1	-1
<b>26</b>	4.3	102	62	16.5	0	1	-1
<b>27</b>	12.2	252	81	46.0	0	3	-1
<b>28</b>	2.1	57	56	9.5	0	1	-1
<b>29</b>	7.4	159	89	18.8	0	0	-1
<b>30</b>	11.4	285	70	32.1	0	3	-1
<b>31</b>	11.1	254	86	26.1	0	3	-1
<b>32</b>	13.0	337	45	16.1	0	3	-1
<b>33</b>	0.8	45	44	7.3	0	1	-1
<b>34</b>	7.3	120	75	21.4	0	0	0
<b>35</b>	6.6	151	68	20.0	0	0	0
<b>36</b>	4.9	159	67	29.3	0	0	-1
<b>37</b>	6.3	106	72	14.9	0	0	0
<b>38</b>	3.4	174	87	8.3	0	0	-1
<b>39</b>	14.4	279	48	22.5	0	3	-1
<b>40</b>	3.8	86	45	12.8	0	1	-1
<b>41</b>	13.2	188	59	26.9	0	3	-1
<b>42</b>	12.7	201	80	25.5	0	3	-1
<b>43</b>	3.2	120	80	22.9	0	0	-1
<b>44</b>	2.2	48	32	11.2	0	1	-1
<b>45</b>	8.5	156	63	20.7	0	0	0
<b>46</b>	4.0	145	73	26.2	0	0	-1
<b>47</b>	5.7	81	39	9.3	0	1	-1
<b>48</b>	2.6	53	66	10.8	0	1	-1
<b>49</b>	6.8	161	60	15.6	0	0	0



In [ ]: