

```
In [53]: from sklearn.cluster import KMeans
import pandas as pd
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import StandardScaler
from matplotlib import pyplot as plt
%matplotlib inline
```

```
In [91]: df=pd.read_excel(r'DataFinal17-20.xlsx',sheet_name='2019')
df1=df.drop(['Total'], axis=1)
df1=df1.iloc[:,2:15].values
#scaler = MinMaxScaler()
scaler= StandardScaler()
# transform data
df1 = scaler.fit_transform(df1)
```

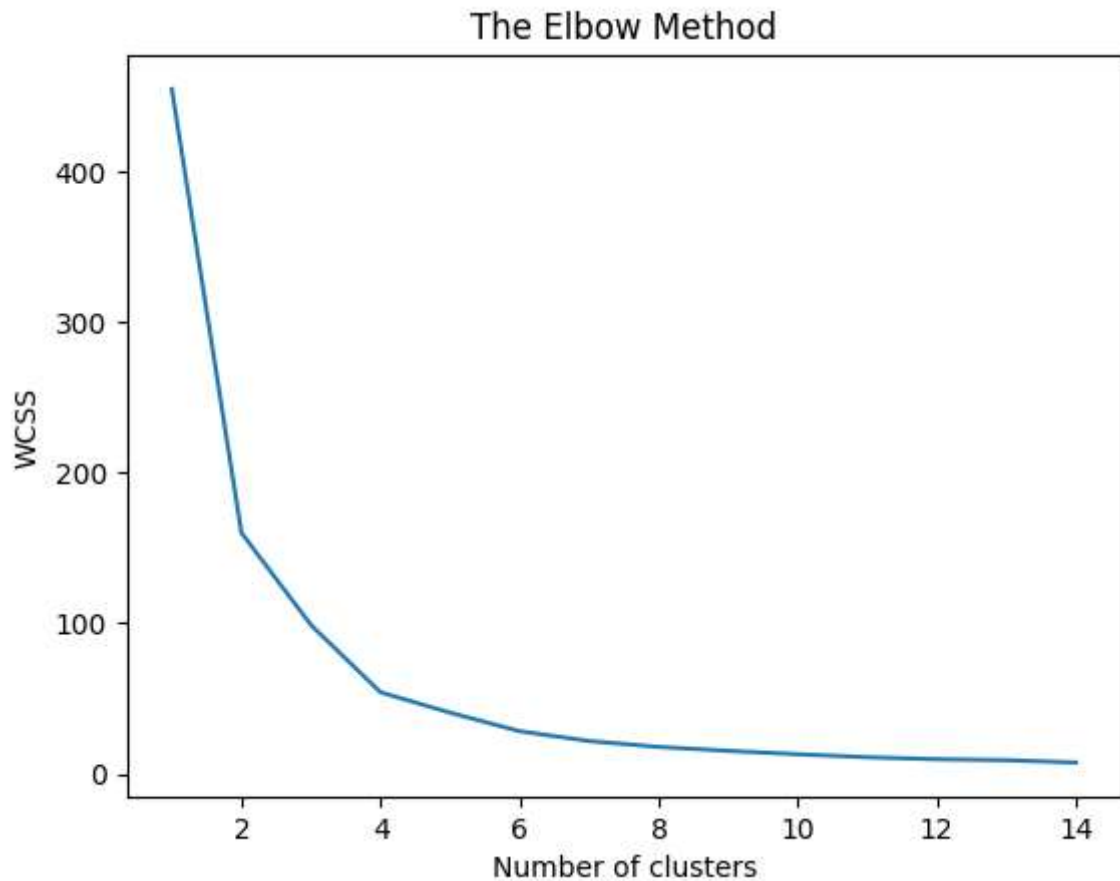
```
In [92]: from sklearn.cluster import KMeans

#create a list for the wcss parameter
wcss = []
#test with 14 clusters
for i in range(1, 15):
    kmeans = KMeans(n_clusters = i, init = 'k-means++', random_state =0)
    kmeans.fit(df1)
    wcss.append(kmeans.inertia_)
```

```
In [93]: wcss
```

```
Out[93]: [455.0,
160.16871868554736,
98.85558888393712,
54.114590464152684,
40.579311227790214,
28.250766109101185,
21.7378878789146,
17.81191386844559,
15.197416213239427,
12.971175003258992,
10.963050359064402,
9.647866972458212,
8.99326653752742,
7.4540376244609865]
```

```
In [94]: plt.plot(range(1, 15), wcss)
plt.title('The Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS')
plt.show()
```



```
In [95]: km=KMeans(n_clusters = 4, init = 'k-means++', random_state = 0)
y_kmeans=km.fit_predict(df1)
```

```
In [96]: y_kmeans
```

```
Out[96]: array([0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 0, 1, 1, 0, 1,
                1, 1, 3, 1, 1, 1, 1, 1, 0, 3, 1, 1, 1])
```

```
In [97]: df['cluster']=y_kmeans
df.head()
```

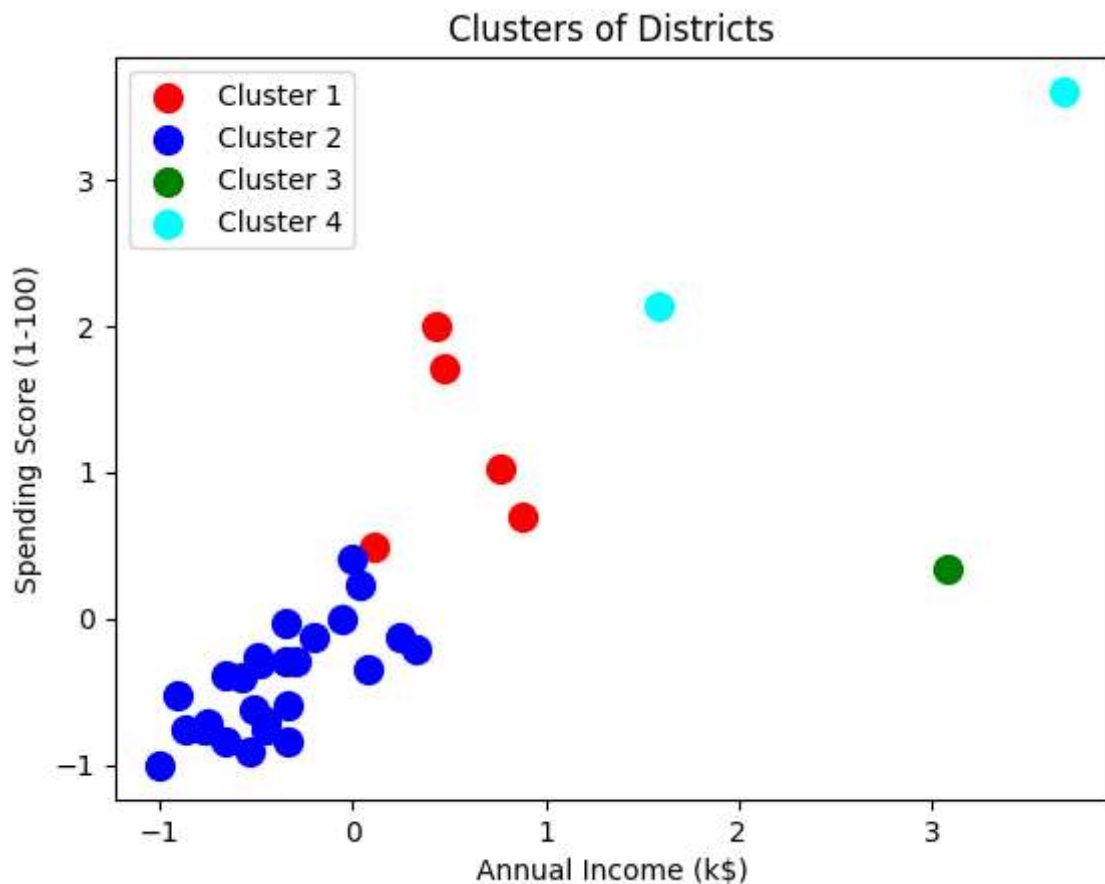
```
Out[97]:
```

	S. No	State/UT/District	Homicide/Murder(3,4,15,16)	Causing death by negligence(5 to 12)	Hurt(20,26)	Assault on woman(35)	Kidn: abducti
0	1	Ahmednagar	217	832	1362	486	
1	2	Akola	96	156	1167	197	
2	3	Amravati	187	316	1681	400	
3	4	Aurangbad	170	489	1718	344	
4	5	Beed	129	316	955	204	

5 rows × 23 columns

```
In [98]: #plt.scatter(df['SrNo'],df['cluster'])
```

```
#for col in df.columns:
#    print(col)
plt.scatter(df1[y_kmeans == 0, 0], df1[y_kmeans == 0, 1], s = 100, c = 'red', label =
plt.scatter(df1[y_kmeans == 1, 0], df1[y_kmeans == 1, 1], s = 100, c = 'blue', label =
plt.scatter(df1[y_kmeans == 2, 0], df1[y_kmeans == 2, 1], s = 100, c = 'green', label =
plt.scatter(df1[y_kmeans == 3, 0], df1[y_kmeans == 3, 1], s = 100, c = 'cyan', label =
#plt.scatter(df1[y_kmeans == 4, 0], df1[y_kmeans == 4, 1], s = 100, c = 'magenta', label =
#plt.scatter(df1[y_kmeans == 5, 0], df1[y_kmeans == 5, 1], s = 100, c = 'black', label =
#plt.scatter(kmeans.cluster_centers_[0, 0], kmeans.cluster_centers_[0, 1], s = 300, c =
plt.title('Clusters of Districts')
plt.xlabel('Annual Income (k$)')
plt.ylabel('Spending Score (1-100)')
plt.legend()
plt.show()
```



```
In [99]: from scipy.cluster.hierarchy import linkage
import scipy.cluster.hierarchy as sch # for creating dendrogram
```

```
In [100... z = linkage(df1, method="complete", metric="euclidean")
```

```
In [101... plt.figure(figsize=(15, 10))
plt.title('Hierarchical Clustering Dendrogram')
plt.xlabel('Features')
plt.ylabel('Crime')
sch.dendrogram(z,
    leaf_rotation=0., # rotates the x axis labels
    leaf_font_size=8., # font size for the x axis labels
)
plt.show()
```

