

```
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 [80]: df=pd.read_excel(r'DataFinal17-20.xlsx',sheet_name='2018')
df1=df.drop(['Total'], axis=1)
df1=df1.iloc[:,2:15].values
#scaler = MinMaxScaler()
scaler= StandardScaler()
# transform data
df1 = scaler.fit_transform(df1)
```

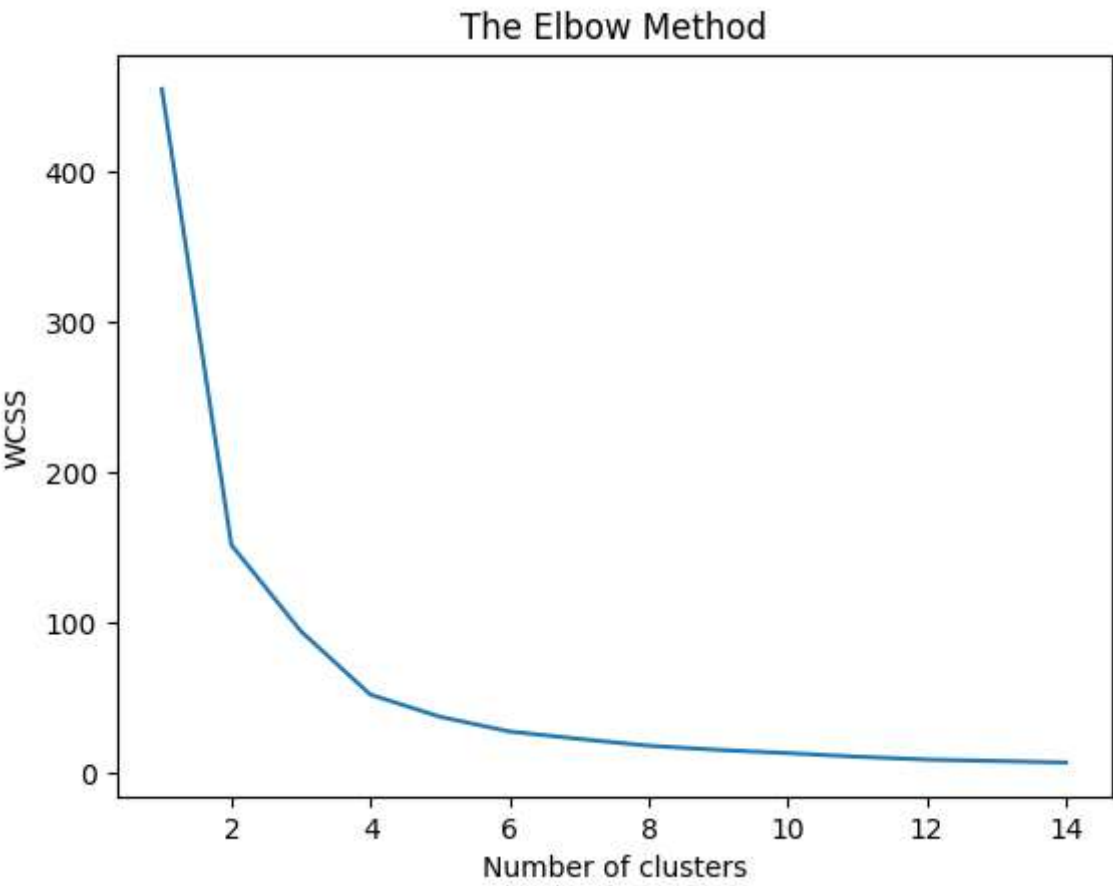
```
In [81]: from sklearn.cluster import KMeans

#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_)
```

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In [82]: wcss
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```
Out[82]: [455.00000000000006,
151.88198373485318,
94.4962974367791,
52.36778125280657,
37.66311578686052,
27.85608724129876,
22.956380978779134,
18.411323303516994,
15.654980301497902,
13.505706826278605,
11.131631283310846,
9.189929945241067,
8.259155435971405,
7.249011347468659]
```

```
In [83]: plt.plot(range(1, 15), wcss)
plt.title('The Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS')
plt.show()
```



```
In [84]: km=KMeans(n_clusters = 4, init = 'k-means++', random_state = 0)
y_kmeans=km.fit_predict(df1)
```

```
In [85]: y_kmeans
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```
Out[85]: array([0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 3, 0, 1, 1, 0, 1,
1, 1, 2, 1, 1, 1, 1, 1, 0, 2, 1, 1, 1])
```

```
In [86]: df['cluster']=y_kmeans
df.head()
```

Out[86]:

	SrNo	State/UT/District	Homicide/Murder	Causing death by negligence	Hurt	Assault on woman	Kidnapping and abduction	Human trafficking	Ra
0	1	Ahmednagar	308	799	1458	476	431	0	
1	2	Akola	104	146	1203	202	133	0	
2	3	Amravati	181	302	1325	419	236	1	
3	4	Aurangbad	193	532	1852	419	281	0	
4	5	Beed	173	318	944	249	145	7	

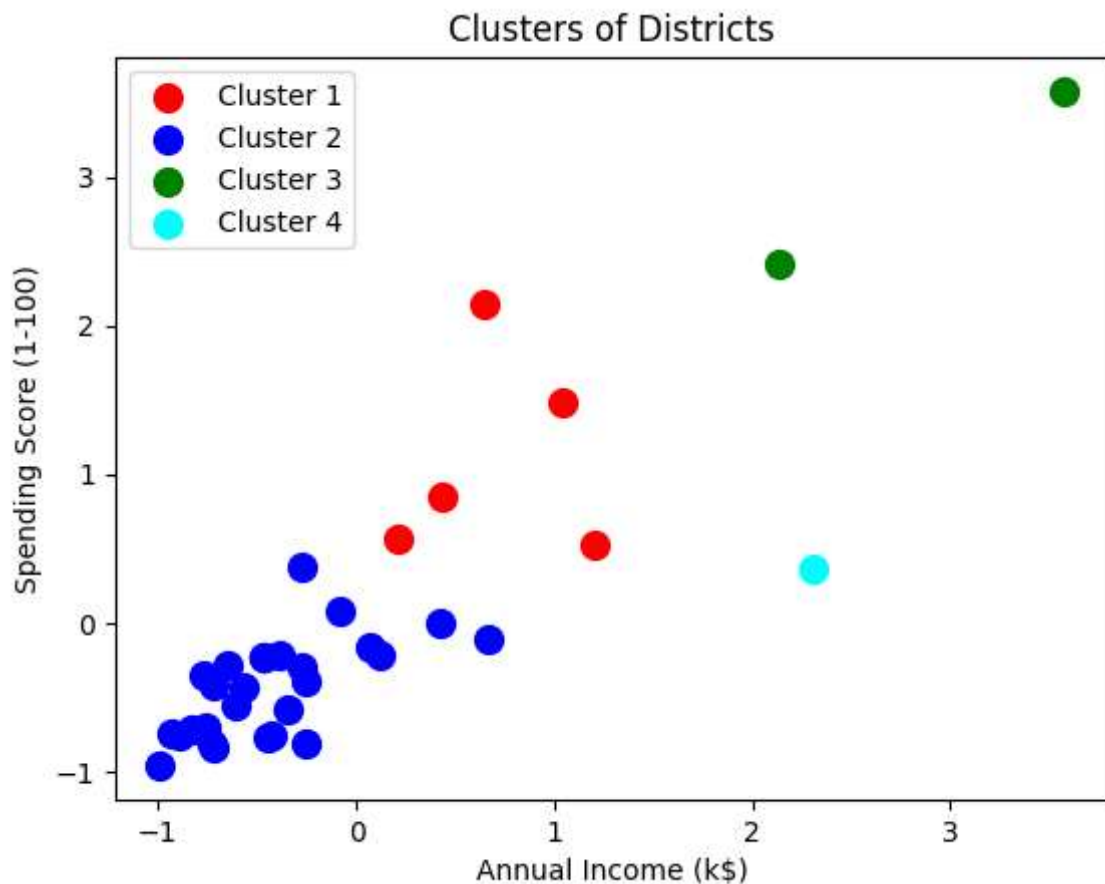
5 rows × 23 columns

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In [87]: #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 = 'Cluster 1')
plt.scatter(df1[y_kmeans == 1, 0], df1[y_kmeans == 1, 1], s = 100, c = 'blue', label = 'Cluster 2')
plt.scatter(df1[y_kmeans == 2, 0], df1[y_kmeans == 2, 1], s = 100, c = 'green', label = 'Cluster 3')
plt.scatter(df1[y_kmeans == 3, 0], df1[y_kmeans == 3, 1], s = 100, c = 'cyan', label = 'Cluster 4')
#plt.scatter(df1[y_kmeans == 4, 0], df1[y_kmeans == 4, 1], s = 100, c = 'magenta', label = 'Cluster 5')
#plt.scatter(df1[y_kmeans == 5, 0], df1[y_kmeans == 5, 1], s = 100, c = 'black', label = 'Cluster 6')
#plt.scatter(kmeans.cluster_centers_[0, 0], kmeans.cluster_centers_[0, 1], s = 300, c = 'red', label = 'Cluster 1')
plt.title('Clusters of Districts')
plt.xlabel('Annual Income (k$)')
plt.ylabel('Spending Score (1-100)')
plt.legend()
plt.show()
```



```
In [88]: from scipy.cluster.hierarchy import linkage
import scipy.cluster.hierarchy as sch # for creating dendrogram
```

```
In [89]: z = linkage(df1, method="complete", metric="euclidean")
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In [90]: 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()
```

