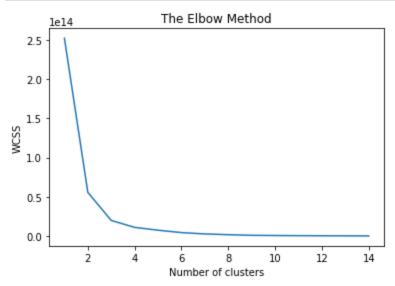
```
In [1]:
         from sklearn.cluster import KMeans
         import pandas as pd
         from sklearn.preprocessing import MinMaxScaler
          from matplotlib import pyplot as plt
          %matplotlib inline
In [2]:
         df=pd.read excel(r'DataFinal17-20.xlsx', sheet name='2020')
         df1=df.iloc[:,2:16]
         df1=df1.drop(['Total'], axis=1)
         df1.head()
                                                                                                    Offences
Out[2]:
                                                                                  Offence
                                                                                                   relating to
                              Causing
                                            Assault Kidnapping
                                                                                          Offences
                                                                                  against
                                                                                                   documents
                                                                  Human
            Homicide/Murder
                              death by Hurt
                                                                                           against
                                                on
                                                          and
                                                                         Rape
                                                               trafficking
                                                                                   public
                                                                                                        and
                            negligence
                                                     abduction
                                            woman
                                                                                          property
                                                                               tranquility
                                                                                                    property
                                                                                                      marks
                       254
                                      1826
                                                                           92
                                                                                     905
                                                                                                        220
         0
                                  620
                                               543
                                                          369
                                                                      0
                                                                                             3104
                                                                                                         86
         1
                                      1345
                                               245
                                                           76
                                                                       0
                                                                           28
                                                                                              946
                       114
                                  116
                                                                                     111
         2
                       201
                                  354
                                      2133
                                                          190
                                                                       0
                                                                           97
                                                                                             2198
                                                                                                        250
                                               461
                                                                                     144
         3
                       139
                                  386
                                      1869
                                               335
                                                          242
                                                                       0
                                                                           80
                                                                                     428
                                                                                             3821
                                                                                                        265
         4
                       206
                                  308 1233
                                               248
                                                           91
                                                                       0
                                                                           50
                                                                                     508
                                                                                             1002
                                                                                                        104
In [3]:
         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 [4]:
          WCSS
         [251991555908767.72,
Out[4]:
          55692833341133.74,
          19924824817824.41,
          11065108262943.742,
          7470720341378.248,
          4418872317352.032,
          2716133886907.053,
          1760690819680.5554,
          955934227092.7084,
          716673512599.725,
          528485592493.5999,
          406730790121.76666,
          301135915319.76666,
          194466396974.37378]
In [5]:
         plt.plot(range(1, 15), wcss)
         plt.title('The Elbow Method')
         plt.xlabel('Number of clusters')
```

```
plt.ylabel('WCSS')
plt.show()
```



```
In [6]: km=KMeans(n_clusters=6)
    y_pred=km.fit_predict(df1)
```

```
In [7]: y_pred
```

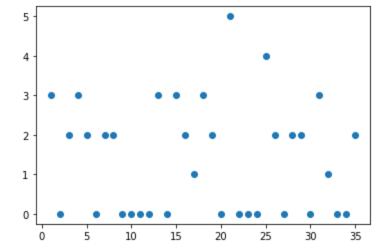
Out[7]: array([3, 0, 2, 3, 2, 0, 2, 2, 0, 0, 0, 0, 3, 0, 3, 2, 1, 3, 2, 0, 5, 0, 0, 0, 0, 4, 2, 0, 2, 2, 0, 3, 1, 0, 0, 2])

```
In [8]: df['cluster']=y_pred
    df.head()
```

Out[8]:		SrNo	District	Homicide/Murder	Causing death by negligence	Hurt	Assault on woman	Kidnapping and abduction	Human trafficking	Rape	Offence against public tranquility	
	0	1	Ahmednagar	254	620	1826	543	369	0	92	905	
	1	2	Akola	114	116	1345	245	76	0	28	111	
	2	3	Amravati	201	354	2133	461	190	0	97	144	
	3	4	Aurangbad	139	386	1869	335	242	0	80	428	
	4	5	Beed	206	308	1233	248	91	0	50	508	

5 rows × 23 columns

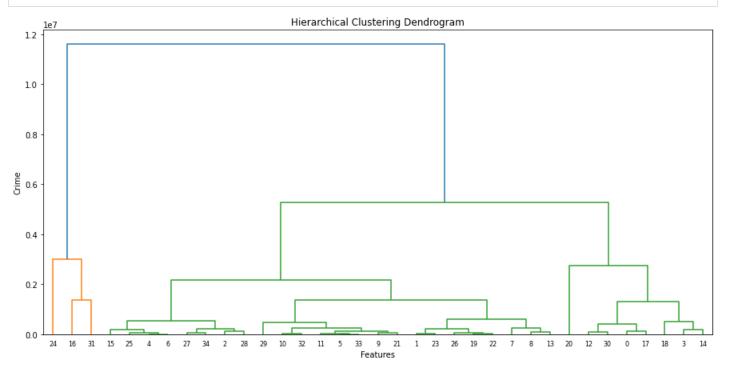
Out[9]: <matplotlib.collections.PathCollection at 0x1f22ef52f40>



```
In [10]: from scipy.cluster.hierarchy import linkage import scipy.cluster.hierarchy as sch # for creating dendrogram
```

```
In [11]: z = linkage(df1, method="complete", metric="euclidean")
```

```
In [12]:    plt.figure(figsize=(15,7))
    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()
```



```
In [13]: #df = pd.read_excel (r'Path where the Excel file is stored\File name.xlsx', sheet_name='yd
In []:
```

In []:		
In []:		