

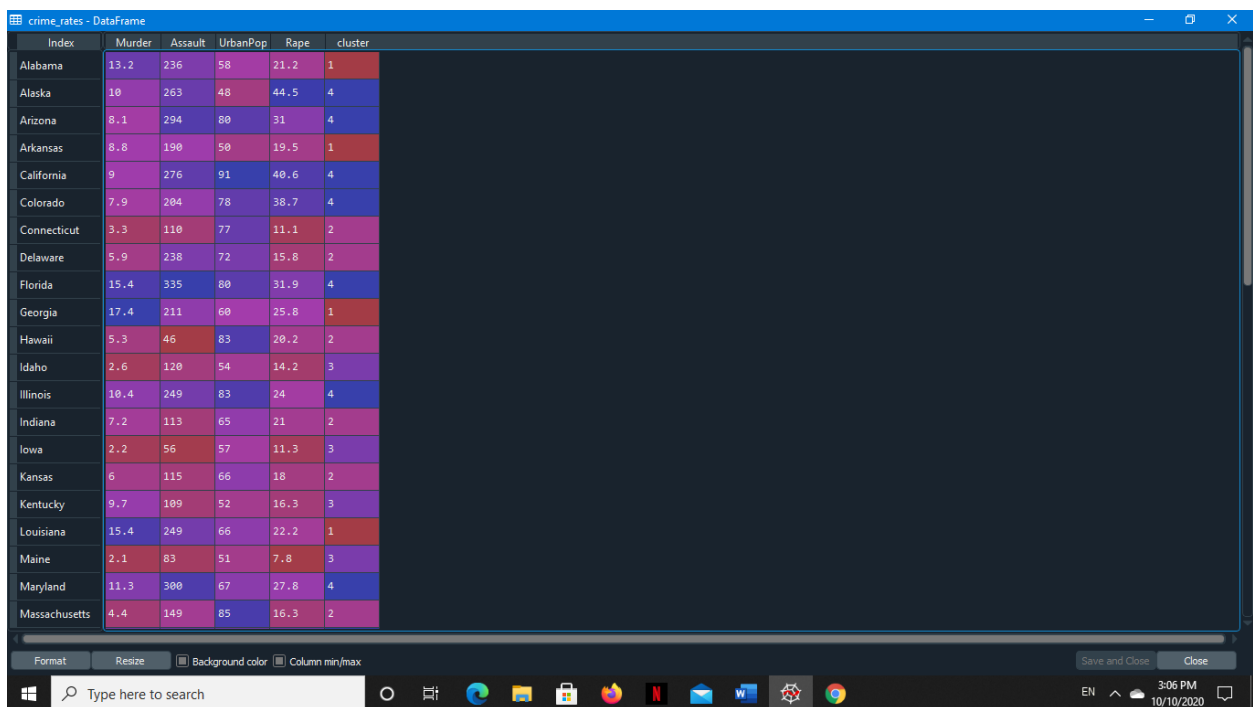
2.) Perform K-Means Clustering for the crime data and identify the number of clusters formed and draw inferences.

In this Assignment first we have to load the data set using some library and the data set is in the form of excel and load that dataset using Common Separated values (CSV).

```
import numpy as np # linear algebra
```

```
import pandas as pd
```

```
crime_rates=pd.read_csv("USArrests.csv", index_col=0)
```



Index	Murder	Assault	UrbanPop	Rape	cluster
Alabama	13.2	236	58	21.2	1
Alaska	10	263	48	44.5	4
Arizona	8.1	294	80	31	4
Arkansas	8.8	190	50	19.5	1
California	9	276	91	40.6	4
Colorado	7.9	204	78	38.7	4
Connecticut	3.3	110	77	11.1	2
Delaware	5.9	238	72	15.8	2
Florida	15.4	335	80	31.9	4
Georgia	17.4	211	60	25.8	1
Hawaii	5.3	46	83	20.2	2
Idaho	2.6	120	54	14.2	3
Illinois	10.4	249	83	24	4
Indiana	7.2	113	65	21	2
Iowa	2.2	56	57	11.3	3
Kansas	6	115	66	18	2
Kentucky	9.7	109	52	16.3	3
Louisiana	15.4	249	66	22.2	1
Maine	2.1	83	51	7.8	3
Maryland	11.3	300	67	27.8	4
Massachusetts	4.4	149	85	16.3	2

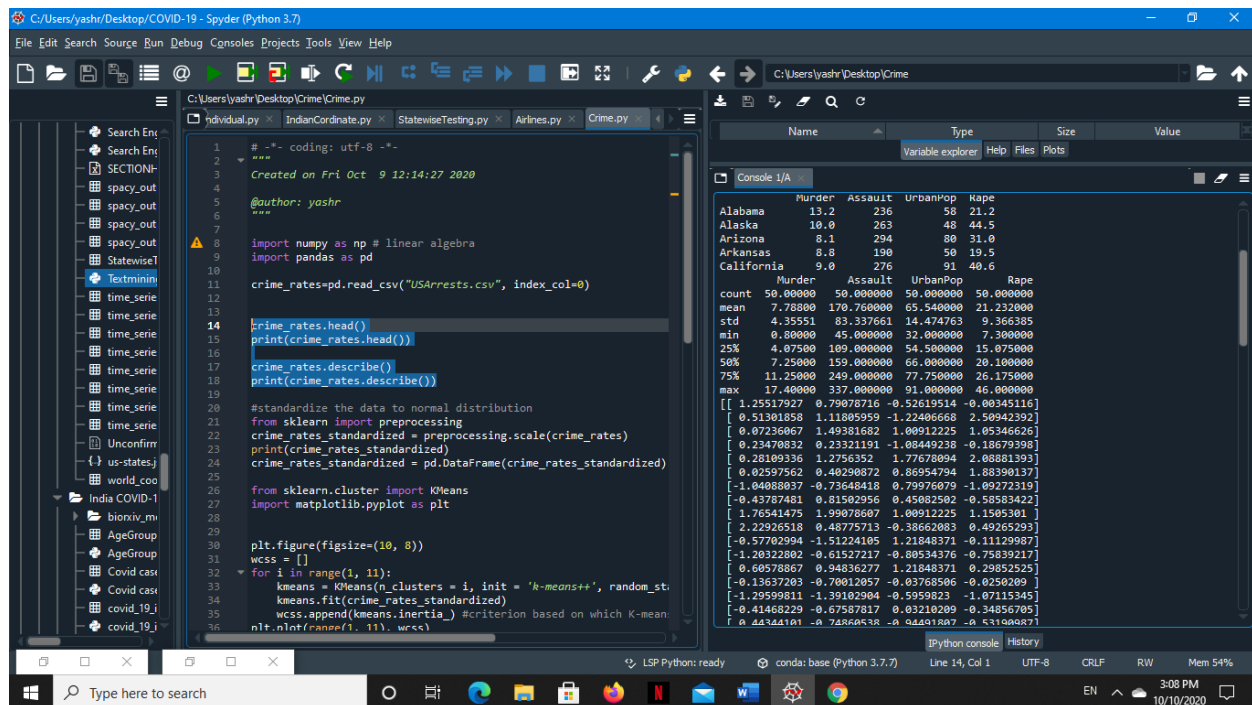
Now we have to perform some data preprocessing technique to access and clean the data set.

```
crime_rates.head()
```

```
print(crime_rates.head())
```

```
crime_rates.describe()
```

```
print(crime_rates.describe())
```



```
1  # -*- coding: utf-8 -*-
2  """
3  Created on Fri Oct 9 12:14:27 2020
4
5  @author: yashr
6  """
7
8  import numpy as np # linear algebra
9  import pandas as pd
10
11  crime_rates=pd.read_csv("USArrests.csv", index_col=0)
12
13
14  crime_rates.head()
15  print(crime_rates.head())
16
17  crime_rates.describe()
18  print(crime_rates.describe())
19
20  #standardize the data to normal distribution
21  from sklearn import preprocessing
22  crime_rates_standardized = preprocessing.scale(crime_rates)
23  print(crime_rates_standardized)
24  crime_rates_standardized = pd.DataFrame(crime_rates_standardized)
25
26
27  from sklearn.cluster import KMeans
28  import matplotlib.pyplot as plt
29
30  plt.figure(figsize=(10, 8))
31  wcss = []
32  for i in range(1, 11):
33      kmeans = KMeans(n_clusters = i, init = 'k-means++', random_st
34      kmeans.fit(crime_rates_standardized)
35      wcss.append(kmeans.inertia_) #Criterion based on which K-mean
36      plt.plot(range(1, 11), wcss)
```

	Murder	Assault	UrbanPop	Rape
Alabama	13.2	236	58	21.2
Alaska	10.0	263	48	44.5
Arizona	8.1	294	80	31.0
Arkansas	8.0	190	50	19.5
California	9.0	276	91	40.6

	Murder	Assault	UrbanPop	Rape
count	50.00000	50.00000	50.00000	50.00000
mean	7.78800	170.76000	65.54000	21.23200
std	4.35551	83.33766	14.47473	9.36638
min	0.80000	45.00000	32.00000	7.30000
25%	4.07500	109.00000	54.50000	15.07500
50%	7.25000	159.00000	66.00000	20.10000
75%	11.25000	249.00000	77.75000	26.17500
max	17.40000	337.00000	91.00000	46.00000

```
#standardize the data to normal distribution
```

```
from sklearn import preprocessing
```

```
crime_rates_standardized = preprocessing.scale(crime_rates)
```

```
print(crime_rates_standardized)
```

```
crime_rates_standardized = pd.DataFrame(crime_rates_standardized)
```

Now we are going to perform Clustering

```
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
```

```
plt.figure(figsize=(10, 8))
wcss = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters = i, init = 'k-means++', random_state = 42)
    kmeans.fit(crime_rates_standardized)
    wcss.append(kmeans.inertia_) #criterion based on which K-means clustering works
plt.plot(range(1, 11), wcss)
plt.title('The Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS')
plt.show()
```

crime_rates_standardized - DataFrame

Index	0	1	2	3
0	1.25518	0.790787	-0.526195	-0.00345116
1	0.513019	1.11806	-1.22407	2.50942
2	0.0723607	1.49382	1.00912	1.05347
3	0.234708	0.233212	-1.08449	-0.186794
4	0.281093	1.27564	1.77678	2.08881
5	0.0259756	0.402909	0.869548	1.8839
6	-1.04088	-0.736484	0.799761	-1.09272
7	-0.437875	0.81503	0.450825	-0.585834
8	1.76541	1.99079	1.00912	1.15053
9	2.22927	0.487757	-0.386621	0.492653
10	-0.57703	-1.51224	1.21848	-0.1113
11	-1.20323	-0.615272	-0.805344	-0.758392
12	0.605709	0.948363	1.21848	0.298525
13	-0.136372	-0.700121	-0.0376851	-0.0250209
14	-1.296	-1.39103	-0.595982	-1.07115
15	-0.414682	-0.675878	0.0321021	-0.348567
16	0.443441	-0.748605	-0.944918	-0.53191
17	1.76541	0.948363	0.0321021	0.104398
18	-1.31919	-1.06376	-1.01471	-1.44862
19	0.814521	1.56654	0.101809	0.70835
20	-0.785763	-0.263757	1.35806	-0.53191

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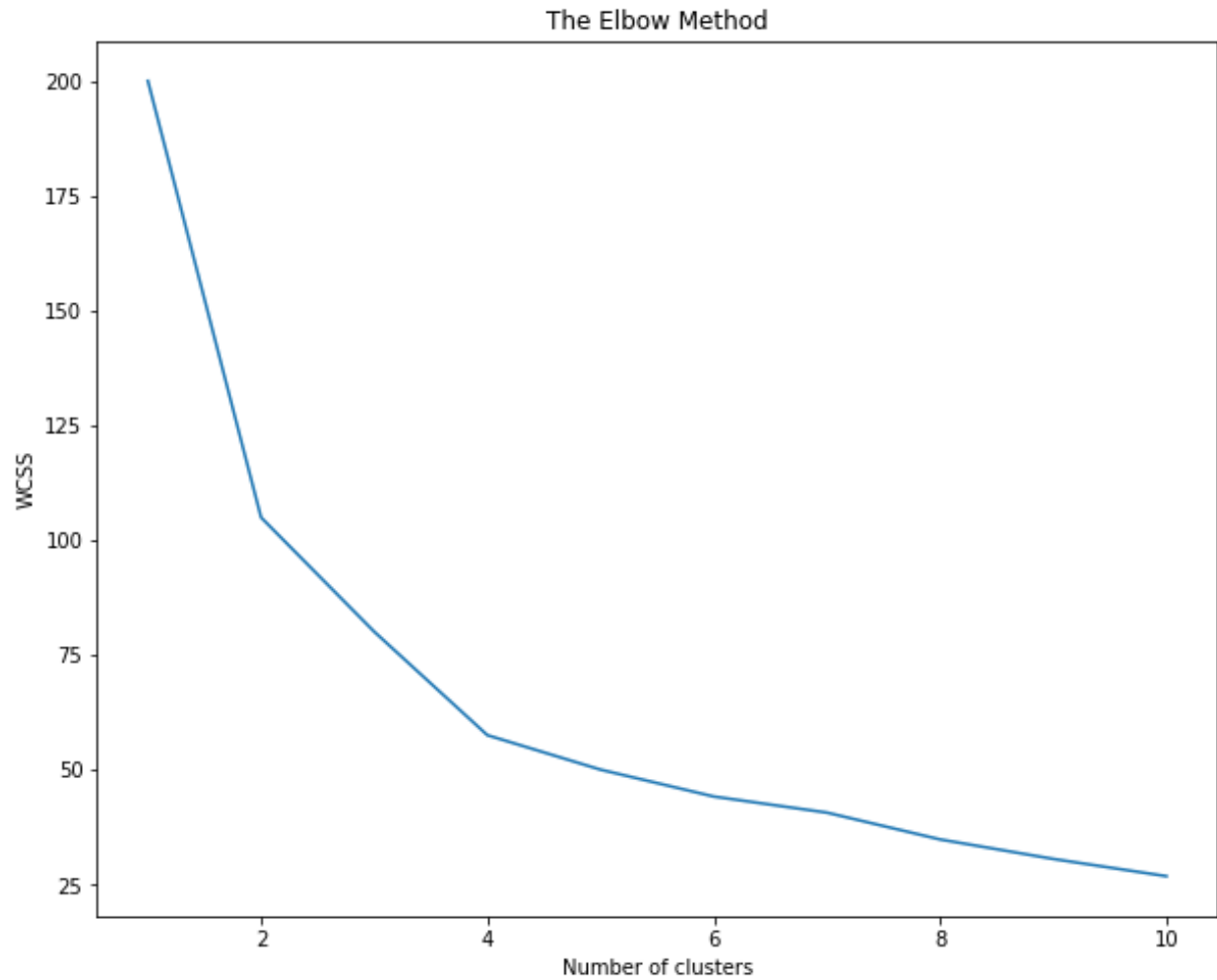
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wcss - List (10 elements)

Ind.	Type	Size	Value
0	float64	1	200.0
1	float64	1	104.96163315756871
2	float64	1	80.08569526137276
3	float64	1	57.55425863091105
4	float64	1	50.05119672966492
5	float64	1	44.20084360686918
6	float64	1	40.70266300421998
7	float64	1	34.85432218312674
8	float64	1	30.616017585613793
9	float64	1	26.856423116032715

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Fitting K-Means to the dataset

```
kmeans = KMeans(n_clusters = 4, init = 'k-means++', random_state = 42)
```

```
y_kmeans = kmeans.fit_predict(crime_rates_standardized)
```

```
y_kmeans
```

```
print(y_kmeans)
```

#beginning of the cluster numbering with 1 instead of 0

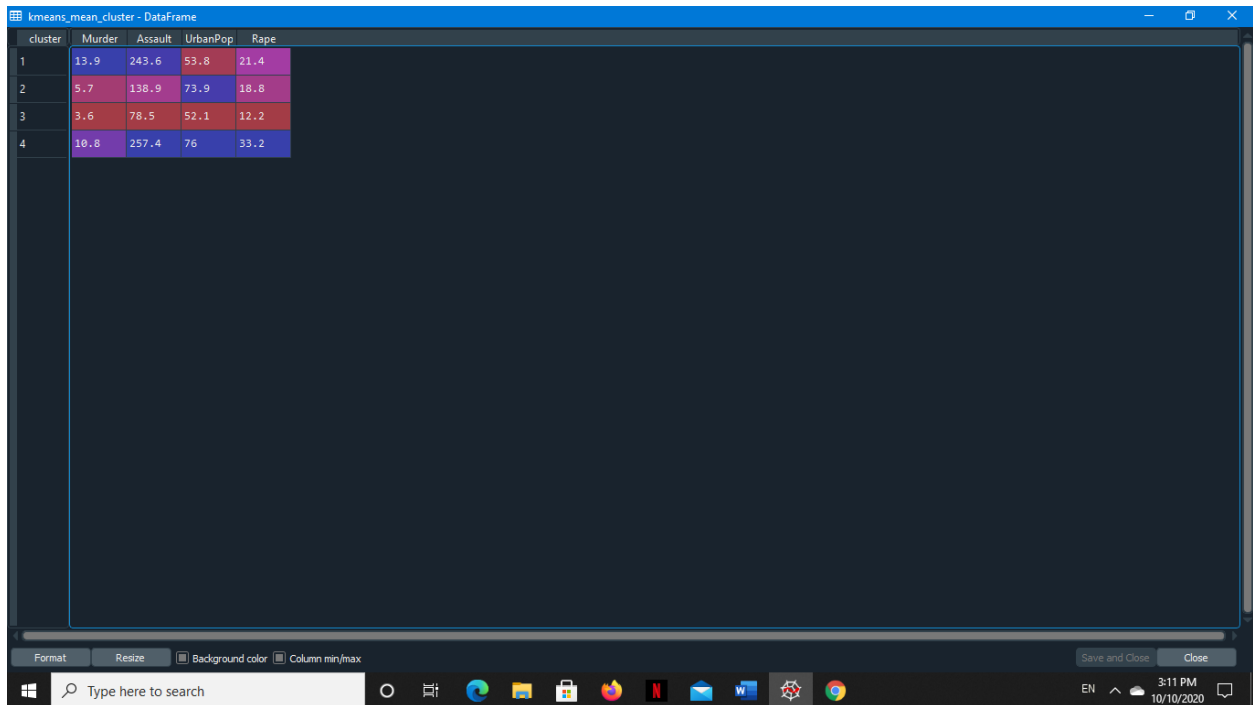
```
y_kmeans1=y_kmeans+1
```

```
# New list called cluster
cluster = list(y_kmeans1)

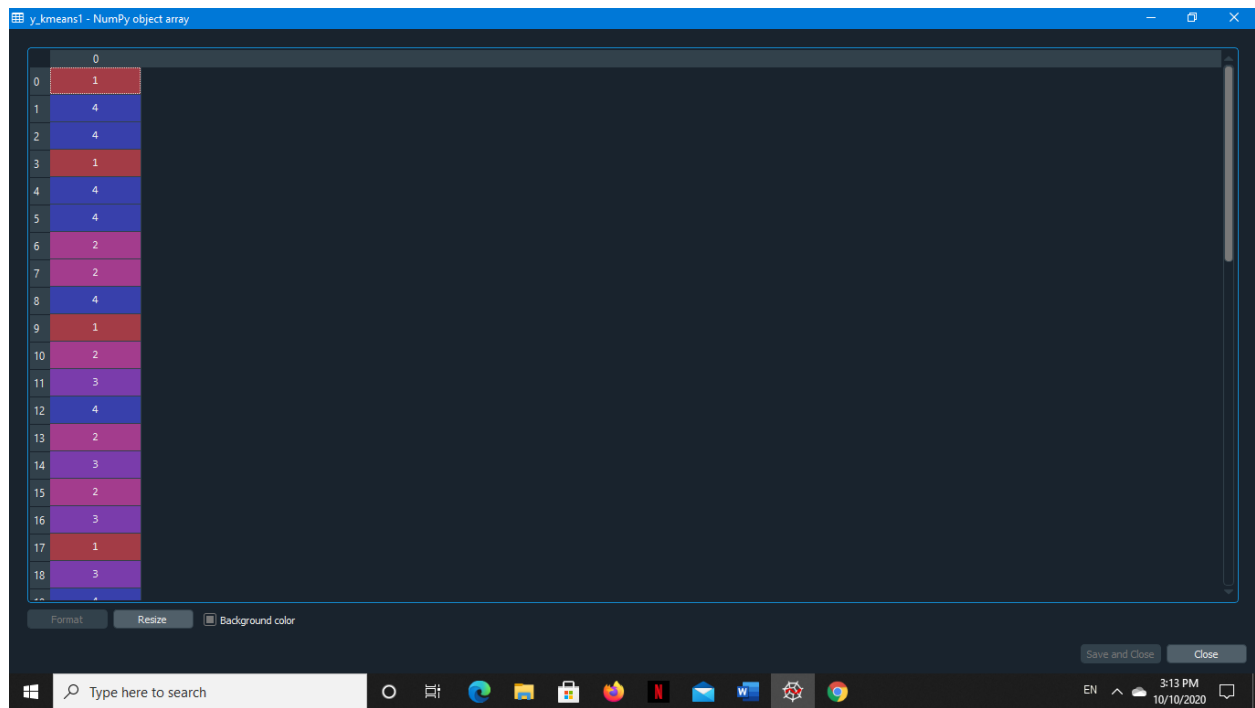
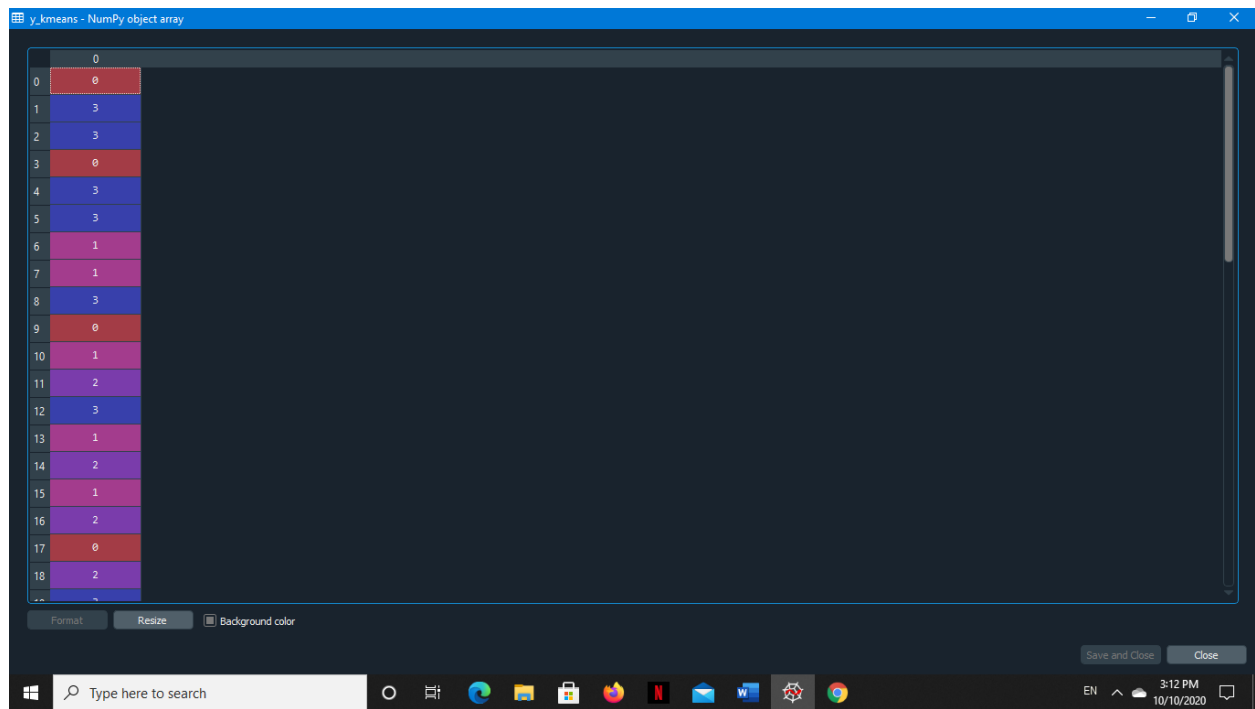
# Adding cluster to our data set
crime_rates['cluster'] = cluster

#Mean of clusters 1 to 4
kmeans_mean_cluster = pd.DataFrame(round(crime_rates.groupby('cluster').mean(),1))

kmeans_mean_cluster
print(kmeans_mean_cluster)
```



cluster	Murder	Assault	UrbanPop	Rape
1	13.9	243.6	53.8	21.4
2	5.7	138.9	73.9	18.8
3	3.6	78.5	52.1	12.2
4	10.8	257.4	76	33.2



import seaborn as sns

```
plt.figure(figsize=(12,6))
```

```
sns.scatterplot(x=crime_rates['Murder'], y=crime_rates['Assault'],hue=y_kmeans1)
```

```
crime_rates[crime_rates['cluster']==1]
```

```
print(crime_rates[crime_rates['cluster']==1])
```

