

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.cluster import KMeans

csd=pd.read_excel('/content/Mall_Customers.xlsx')

csd.head()

{"summary":{"\n  \"name\": \"csd\",\n  \"rows\": 200,\n  \"fields\": [\n    {\n      \"column\": \"CustomerID\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 57,\n        \"min\": 1,\n        \"max\": 200,\n        \"num_unique_values\": 200,\n        \"samples\": [\n          96,\n          16,\n          31\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      },\n      \"column\": \"Gender\",\n      \"properties\": {\n        \"dtype\": \"category\",\n        \"num_unique_values\": 2,\n        \"samples\": [\n          \"Female\",\n          \"Male\"\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      },\n      \"column\": \"Age\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 13,\n        \"min\": 18,\n        \"max\": 70,\n        \"num_unique_values\": 51,\n        \"samples\": [\n          55,\n          26\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      },\n      \"column\": \"Annual Income (k$)\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 26,\n        \"min\": 15,\n        \"max\": 137,\n        \"num_unique_values\": 64,\n        \"samples\": [\n          87,\n          101\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      },\n      \"column\": \"Spending Score (1-100)\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 25,\n        \"min\": 1,\n        \"max\": 99,\n        \"num_unique_values\": 84,\n        \"samples\": [\n          39,\n          83\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      }\n    }\n  ],\n  \"type\": \"dataframe\",\n  \"variable_name\": \"csd\"}

csd.isnull().sum()

CustomerID      0
Gender           0
Age             0
Annual Income (k$)  0
Spending Score (1-100)  0
dtype: int64

csd.shape

(200, 5)

```

```

csd.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
#   Column                Non-Null Count  Dtype
---  -
0   CustomerID            200 non-null    int64
1   Gender                200 non-null    object
2   Age                  200 non-null    int64
3   Annual Income (k$)    200 non-null    int64
4   Spending Score (1-100) 200 non-null    int64
dtypes: int64(4), object(1)
memory usage: 7.9+ KB

x=csd.iloc[:,[3,4]].values

print(x)

[[ 15  39]
 [ 15  81]
 [ 16   6]
 [ 16  77]
 [ 17  40]
 [ 17  76]
 [ 18   6]
 [ 18  94]
 [ 19   3]
 [ 19  72]
 [ 19  14]
 [ 19  99]
 [ 20  15]
 [ 20  77]
 [ 20  13]
 [ 20  79]
 [ 21  35]
 [ 21  66]
 [ 23  29]
 [ 23  98]
 [ 24  35]
 [ 24  73]
 [ 25   5]
 [ 25  73]
 [ 28  14]
 [ 28  82]
 [ 28  32]
 [ 28  61]
 [ 29  31]
 [ 29  87]
 [ 30   4]

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[ 30 73]
[ 33  4]
[ 33 92]
[ 33 14]
[ 33 81]
[ 34 17]
[ 34 73]
[ 37 26]
[ 37 75]
[ 38 35]
[ 38 92]
[ 39 36]
[ 39 61]
[ 39 28]
[ 39 65]
[ 40 55]
[ 40 47]
[ 40 42]
[ 40 42]
[ 42 52]
[ 42 60]
[ 43 54]
[ 43 60]
[ 43 45]
[ 43 41]
[ 44 50]
[ 44 46]
[ 46 51]
[ 46 46]
[ 46 56]
[ 46 55]
[ 47 52]
[ 47 59]
[ 48 51]
[ 48 59]
[ 48 50]
[ 48 48]
[ 48 59]
[ 48 47]
[ 49 55]
[ 49 42]
[ 50 49]
[ 50 56]
[ 54 47]
[ 54 54]
[ 54 53]
[ 54 48]
[ 54 52]
[ 54 42]
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[ 54 51]
[ 54 55]
[ 54 41]
[ 54 44]
[ 54 57]
[ 54 46]
[ 57 58]
[ 57 55]
[ 58 60]
[ 58 46]
[ 59 55]
[ 59 41]
[ 60 49]
[ 60 40]
[ 60 42]
[ 60 52]
[ 60 47]
[ 60 50]
[ 61 42]
[ 61 49]
[ 62 41]
[ 62 48]
[ 62 59]
[ 62 55]
[ 62 56]
[ 62 42]
[ 63 50]
[ 63 46]
[ 63 43]
[ 63 48]
[ 63 52]
[ 63 54]
[ 64 42]
[ 64 46]
[ 65 48]
[ 65 50]
[ 65 43]
[ 65 59]
[ 67 43]
[ 67 57]
[ 67 56]
[ 67 40]
[ 69 58]
[ 69 91]
[ 70 29]
[ 70 77]
[ 71 35]
[ 71 95]
[ 71 11]
```

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[ 71 75]
[ 71 9]
[ 71 75]
[ 72 34]
[ 72 71]
[ 73 5]
[ 73 88]
[ 73 7]
[ 73 73]
[ 74 10]
[ 74 72]
[ 75 5]
[ 75 93]
[ 76 40]
[ 76 87]
[ 77 12]
[ 77 97]
[ 77 36]
[ 77 74]
[ 78 22]
[ 78 90]
[ 78 17]
[ 78 88]
[ 78 20]
[ 78 76]
[ 78 16]
[ 78 89]
[ 78 1]
[ 78 78]
[ 78 1]
[ 78 73]
[ 79 35]
[ 79 83]
[ 81 5]
[ 81 93]
[ 85 26]
[ 85 75]
[ 86 20]
[ 86 95]
[ 87 27]
[ 87 63]
[ 87 13]
[ 87 75]
[ 87 10]
[ 87 92]
[ 88 13]
[ 88 86]
[ 88 15]
[ 88 69]
```

```
[ 93  14]
[ 93  90]
[ 97  32]
[ 97  86]
[ 98  15]
[ 98  88]
[ 99  39]
[ 99  97]
[101  24]
[101  68]
[103  17]
[103  85]
[103  23]
[103  69]
[113   8]
[113  91]
[120  16]
[120  79]
[126  28]
[126  74]
[137  18]
[137  83]]
```

```
wcss = []
```

```
for i in range(1,11):
```

```
    Kmeans = KMeans(n_clusters = i, init = 'k-means++', random_state=42)
    Kmeans.fit(x)
```

```
    wcss.append(Kmeans.inertia_)
```

```
sns.set()
```

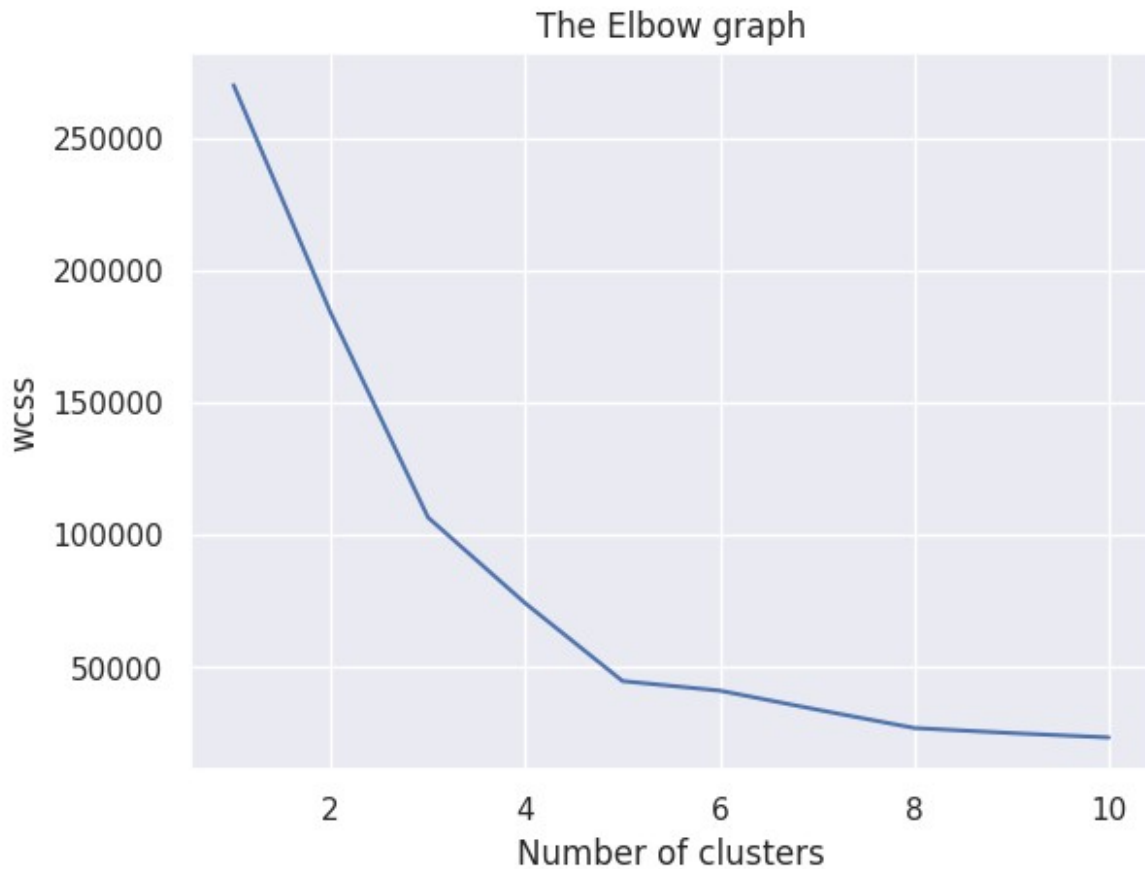
```
plt.plot(range(1,11),wcss)
```

```
plt.title('The Elbow graph')
```

```
plt.xlabel('Number of clusters')
```

```
plt.ylabel('wcss')
```

```
plt.show()
```



```
K=KMeans(n_clusters=5,init='k-means++',random_state=0)
y=K.fit_predict(x)
```

```
print(y)
```

```
[3 4 3 4 3 4 3 4 3 4 3 4 3 4 3 4 3 4 3 4 3 4 3 4 3 4 3 4 3 4 3
4 3
 4 3 4 3 4 3 0 3 4 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0
 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0
 0 0 0 0 0 0 0 0 0 0 0 1 2 1 0 1 2 1 2 1 0 1 2 1 2 1 2 1 2 1 0 1 2 1
2 1
 2 1 2 1 2 1 2 1 2 1 2 1 2 1 2 1 2 1 2 1 2 1 2 1 2 1 2 1 2 1 2
1 2
 1 2 1 2 1 2 1 2 1 2 1 2 1 2 1]
```

```
plt.figure(figsize=(8,8))
plt.scatter(x[y==0,0],x[y==0,1],s=50,c='green',label='cluster1')
plt.scatter(x[y==1,0],x[y==1,1],s=50,c='red',label='cluster2')
plt.scatter(x[y==2,0],x[y==2,1],s=50,c='yellow',label='cluster3')
plt.scatter(x[y==3,0],x[y==3,1],s=50,c='blue',label='cluster4')
plt.scatter(x[y==4,0],x[y==4,1],s=50,c='black',label='cluster5')
```

```
plt.scatter(K.cluster_centers_[0],K.cluster_centers_[1],s=300,c='cyan',label='centroids')
plt.title('Clusters of customers')
plt.show()
```

