

```

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

customer_data =pd.read_csv('/content/Mall_Customers.csv')

customer_data.head()

{"summary":{"\n  \"name\": \"customer_data\",\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          83,\n          39\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      }\n    ]\n  },\n  \"type\": \"dataframe\", \"variable_name\": \"customer_data\"}

```

```
customer_data.shape
```

```
(200, 5)
```

```
customer_data.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
```

```
customer_data.isnull().sum()
```

```
CustomerID                0
```

```
Gender                    0
```

```
Age                      0
```

```
Annual Income (k$)       0
```

```
Spending Score (1-100)   0
```

```
dtype: int64
```

```
x=customer_data.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]
[ 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]
[ 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]
```

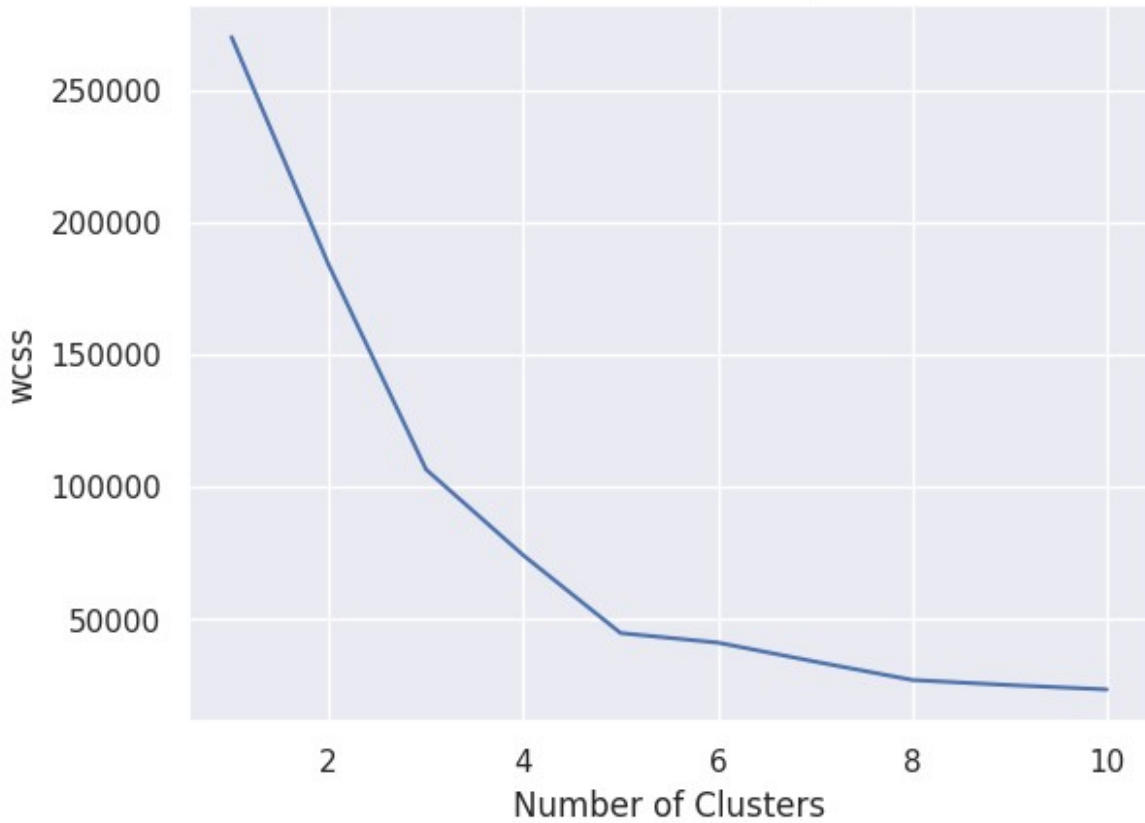
```
[ 71 11]
[ 71 75]
[ 71  9]
[ 71 75]
[ 72 34]
[ 72 71]
[ 73  5]
[ 73 88]
[ 73  7]
[ 73 73]
[ 74 10]
[ 74 72]
[ 75  5]
[ 75 93]
[ 76 40]
[ 76 87]
[ 77 12]
[ 77 97]
[ 77 36]
[ 77 74]
[ 78 22]
[ 78 90]
[ 78 17]
[ 78 88]
[ 78 20]
[ 78 76]
[ 78 16]
[ 78 89]
[ 78  1]
[ 78 78]
[ 78  1]
[ 78 73]
[ 79 35]
[ 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 Point Graph')
plt.xlabel('Number of Clusters')
plt.ylabel('wcss')
plt.show()
```

The Elbow Point Graph



```
kmeans=KMeans(n_clusters=5,init='k-means++',random_state=42)
y=kmeans.fit_predict(x)
print(y)
```

[illegible]

```
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='violet',label='cluster5')
plt.scatter(kmeans.cluster_centers_[0,0],kmeans.cluster_centers_[0,1],
```

```
s=300,c='cyan',label='Centroids')
plt.title('Clusters of Customers')
plt.xlabel('Annual Income')
plt.ylabel('Spending Score')
plt.legend()
plt.show()
```

