CUSTOMER SEGMENTATION BASED ON THEIR BUYING BEHAVIOUR

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.cluster import KMeans
# loading the data from csv file to a Pandas DataFrame
customer_data = pd.read_csv('Mall_Customers.csv')
# first 5 rows in the dataframe
customer_data.head()
                                                                             1
        CustomerID Gender Age Annual Income (k$) Spending Score (1-100)
     0
                      Male
                            19
                                                15
                                                                        39
                 1
                      Male
     1
                 2
                            21
                                                15
                                                                        81
     2
                 3 Female
                            20
                                                16
                                                                         6
                                                                        77
                 4 Female
                            23
                                                16
      4
                 5 Female
                                                17
                                                                        40
# finding the number of rows and columns
customer_data.shape
     (200, 5)
# getting some informations about the dataset
customer_data.info()
     <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 200 entries, 0 to 199
    Data columns (total 5 columns):
     # Column
                                Non-Null Count Dtype
     0
         CustomerID
                                 200 non-null
                                                 int64
                                 200 non-null
                                                 obiect
         Gender
     1
                                 200 non-null
                                                 int64
         Age
         Annual Income (k$)
                                 200 non-null
                                                 int64
     3
         Spending Score (1-100) 200 non-null
                                               int64
    dtypes: int64(4), object(1)
    memory usage: 7.9+ KB
# checking for missing values
```

checking for missing values
customer_data.isnull().sum()

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

Double-click (or enter) to edit

Choosing the Annual Income Column & Spending Score column

X= customer_data.iloc[:,[3,4]].values
print(X)

[[15 391 [15 81] [16 6] 16 77] [17 [17 76] [18 61 [18 94] Γ 19 31 19 721 [19 14] 19 99]

[20 15]

```
[ 20 13]
       20
           79]
      [ 21 35]
       21
           66]
      [ 23
           291
      [ 23
           981
      [ 24
           35]
       24 73]
      [ 25
            5]
       25 73]
       28
           14]
      [ 28
           82]
       28
           32]
      [ 28
           61]
       29
           311
      [ 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]
      Γ 39
           36]
       39
           61]
      [ 39
           28]
      [ 39
           65]
       40
           55]
      [ 40 47]
      [ 40
           42]
      [ 40 42]
      [ 42 52]
      [ 42 60]
       43
           54]
      [ 43
           601
      [ 43
[ 43
           45]
           41]
      [ 44
# finding wcss value for different number of clusters
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_)
# plot an elbow graph
sns.set()
plt.plot(range(1,11), wcss)
plt.title('The Elbow Point Graph')
plt.xlabel('Number of Clusters')
plt.ylabel('WCSS')
plt.show()
```

plt.xlabel('Annual Income')
plt.ylabel('Spending Score')

The Elbow Point Graph

```
kmeans = KMeans(n_clusters=5, init='k-means++', random_state=0)
# return a label for each data point based on their cluster
Y = kmeans.fit_predict(X)
print(Y)
   2 0 2 0 2 0 2 0 2 0 2 0 2 0 2 0 2]
   /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change fr
     warnings.warn(
# plotting all the clusters and their Centroids
plt.figure(figsize=(8,8))
plt.scatter(X[Y==0,0], X[Y==0,1], s=50, c='green', label='Cluster 1')
\verb|plt.scatter(X[Y==1,0], X[Y==1,1], s=50, c='red', label='Cluster 2')|\\
plt.scatter(X[Y==2,0], X[Y==2,1], s=50, c='yellow', label='Cluster 3')
plt.scatter(X[Y==3,0], X[Y==3,1], s=50, c='violet', label='Cluster 4')
\verb|plt.scatter(X[Y==4,0], X[Y==4,1], s=50, c='blue', label='Cluster 5')|\\
# plot the centroids
plt.scatter(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], s=100, c='cyan', label='Centroids')
plt.title('Customer Groups')
```



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