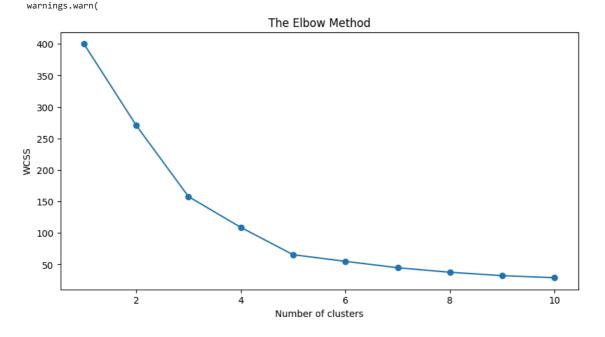
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
dataset = pd.read_csv('/content/Mall_Customers.csv')
dataset.head()
∓
         CustomerID Gender Age Annual Income (k$) Spending Score (1-100)
                                                                                \blacksquare
      0
                  1
                       Male
                              19
                                                   15
                                                                           39
                                                                                ılı.
      1
                  2
                       Male
                              21
                                                  15
                                                                           81
                                                  16
                                                                            6
      2
                  3 Female
                              20
      3
                  4 Female
                              23
                                                   16
                                                                           77
      4
                                                   17
                                                                           40
                  5 Female
                              31
             Generate code with dataset
                                           View recommended plots
 Next steps:
dataset.describe()
dataset.info()
<class 'pandas.core.frame.DataFrame'>
     RangeIndex: 200 entries, 0 to 199
     Data columns (total 5 columns):
                                  Non-Null Count Dtype
      # Column
      0 CustomerID
                                   200 non-null
                                                   int64
                                   200 non-null
          Gender
                                                   object
      1
                                   200 non-null
                                                   int64
      2
         Age
      3
          Annual Income (k$)
                                   200 non-null
                                                   int64
          Spending Score (1-100) 200 non-null
                                                   int64
     dtypes: int64(4), object(1)
     memory usage: 7.9+ KB
print(dataset.head(10))
print(dataset.shape)
        CustomerID Gender Age Annual Income (k$) Spending Score (1-100)
\overline{\rightarrow}
                 1
                      Male
                             19
                                                  15
                                                                           39
     1
                 2
                      Male
                             21
                                                  15
                                                                           81
     2
                 3
                    Female
                             20
                                                  16
                                                                           6
                 4
                    Female
                             23
                                                                           77
                 5
                             31
                    Female
                                                  17
                                                  17
                                                                           76
                 6
                    Female
                    Female
                             35
                                                  18
                                                                           6
     6
                    Female
                             23
                                                                           94
                 8
                                                  18
     8
                 9
                      Male
                             64
                                                  19
                                                                           3
                   Female
                                                                           72
     9
                10
                             30
                                                  19
     (200, 5)
X = dataset.iloc[:, [3, 4]].values
scaler = StandardScaler()
scaled_features = scaler.fit_transform(X)
wcss = []
for i in range(1,11):
    kmeans = KMeans(n_clusters=i, init='k-means++', random_state=42)
    kmeans.fit(scaled_features)
    wcss.append(kmeans.inertia_)
plt.figure(figsize=(10,5))
plt.plot(range(1,11), wcss, marker='o')
plt.title('The Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS')
plt.show()
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change fr
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change fr
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    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change fr
```



```
kmeans = KMeans(n_clusters=5, init='k-means++', random_state=42)
y_kmeans = kmeans.fit_predict(scaled_features)

dataset['Cluster'] = y_kmeans

plt.figure(figsize=(15, 5))
plt.scatter(scaled_features[y_kmeans == 0, 0], scaled_features[y_kmeans == 0, 1], s=70, c='orange', label='Cluster 1: High income and low plt.scatter(scaled_features[y_kmeans == 1, 0], scaled_features[y_kmeans == 1, 1], s=70, c='red', label='Cluster 2: Average income and ave plt.scatter(scaled_features[y_kmeans == 2, 0], scaled_features[y_kmeans == 2, 1], s=70, c='yellow', label='Cluster 3: High income and high plt.scatter(scaled_features[y_kmeans == 3, 0], scaled_features[y_kmeans == 3, 1], s=70, c='black', label='Cluster 4: Low income and high plt.scatter(scaled_features[y_kmeans == 4, 0], scaled_features[y_kmeans == 4, 1], s=70, c='blue', label='Cluster 5: Low income and low splt.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[:, 1], s=200, c='brown', label='Centroids')
plt.slabel('ANNUAL INCOME')
plt.ylabel('SPENDINGS SCORE')
plt.legend()
plt.show()
```

/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change fr warnings.warn(

