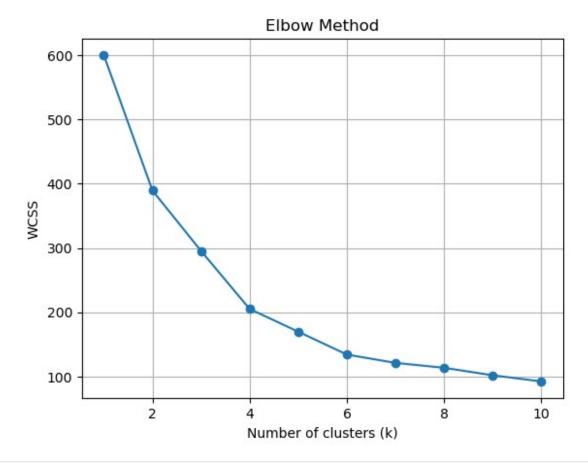
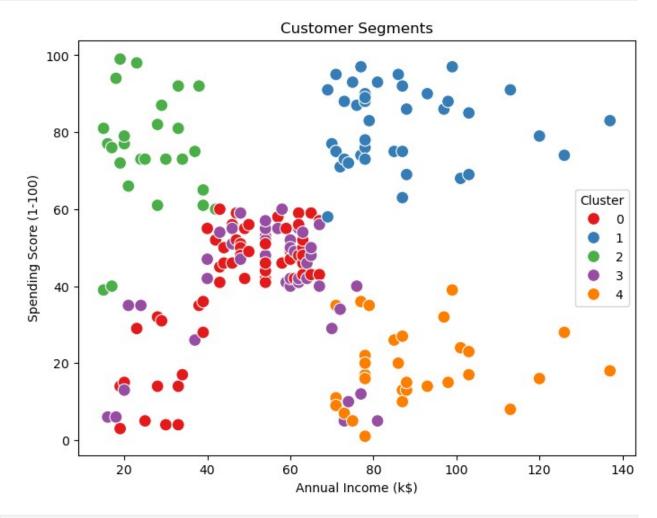
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
import os
warnings.filterwarnings("ignore")
os.environ["OMP NUM THREADS"] = "1"
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans
from sklearn.metrics import silhouette score
df = pd.read csv("C:/Users/de/Downloads/archive/Mall Customers.csv")
print(df.head())
   CustomerID Gender Age Annual Income (k$)
                                                 Spending Score (1-100)
0
            1
                 Male
                        19
                                             15
                                                                     39
            2
                 Male
                                             15
1
                        21
                                                                     81
2
            3 Female
                        20
                                             16
                                                                      6
3
            4 Female
                        23
                                             16
                                                                     77
            5 Female
4
                        31
                                             17
                                                                     40
print("Missing Values:\n", df.isnull().sum())
Missing Values:
CustomerID
                           0
Gender
                          0
                          0
Age
Annual Income (k$)
                          0
Spending Score (1-100)
                          0
dtype: int64
features = df[['Age', 'Annual Income (k$)', 'Spending Score (1-100)']]
scaler = StandardScaler()
scaled features = scaler.fit transform(features)
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.plot(range(1, 11), wcss, marker='o')
plt.title("Elbow Method")
plt.xlabel("Number of clusters (k)")
plt.ylabel("WCSS")
plt.grid(True)
plt.show()
```



```
for k in range(2, 11):
    kmeans = KMeans(n clusters=k, random state=42)
    kmeans.fit(scaled features)
    score = silhouette score(scaled features, kmeans.labels )
    print(f"Silhouette Score for k={k}: {score:.4f}")
Silhouette Score for k=2: 0.3355
Silhouette Score for k=3: 0.3579
Silhouette Score for k=4: 0.4040
Silhouette Score for k=5: 0.4085
Silhouette Score for k=6: 0.4311
Silhouette Score for k=7: 0.4101
Silhouette Score for k=8: 0.3674
Silhouette Score for k=9: 0.3744
Silhouette Score for k=10: 0.3619
k = 5
kmeans = KMeans(n clusters=k, random state=42)
clusters = kmeans.fit predict(scaled features)
df['Cluster'] = clusters
```

```
plt.figure(figsize=(8, 6))
sns.scatterplot(data=df, x='Annual Income (k$)', y='Spending Score (1-
100)', hue='Cluster', palette='Set1', s=100)
plt.title("Customer Segments")
plt.show()
```



```
cluster_summary = df.groupby('Cluster')[['Age', 'Annual Income (k$)',
'Spending Score (1-100)']].mean()
print("\nCluster Summary:\n", cluster_summary)
Cluster Summary:
                     Annual Income (k$) Spending Score (1-100)
Cluster
         55.275862
                              47.620690
                                                      41.706897
                                                      81.525000
1
         32.875000
                              86.100000
2
         25.769231
                              26.115385
                                                      74.846154
3
         26.733333
                              54.311111
                                                      40.911111
4
         44.387097
                              89.774194
                                                      18.483871
```

df.to_csv("segmented_customers.csv", index=False)