

EXPERIMENT NO: 10

## **Customer Segmentation using K-Means Clustering**

### **Aim:**

To segment mall customers into distinct groups based on their annual income and spending score using the K-Means clustering algorithm.

### **Algorithm:**

- 1. Load Data:** Import the Mall\_Customers.csv dataset using pandas.
- 2. Visualize Data:** Use pairplot() to observe relationships among features.
- 3. Select Features:** Extract Annual Income (k\$) and Spending Score (1-100) for clustering.
- 4. Apply K-Means:**
- 5. Initialize K-Means with 5 clusters.**
- 6. Fit the model and assign cluster labels to each customer.**
- 7. Visualize Clusters:**
- 8. Plot scatter points showing different clusters based on income and spending.**
- 9. Find Optimal Clusters:**
- 10. Calculate WCSS (Within-Cluster Sum of Squares) for cluster counts 1–9.**
- 11. Plot the Elbow Curve to determine the best number of clusters.**

## Program:

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

```
[2]: df=pd.read_csv("C:/Users/vijay/Downloads/Mall_Customers.csv")
df.info()
```

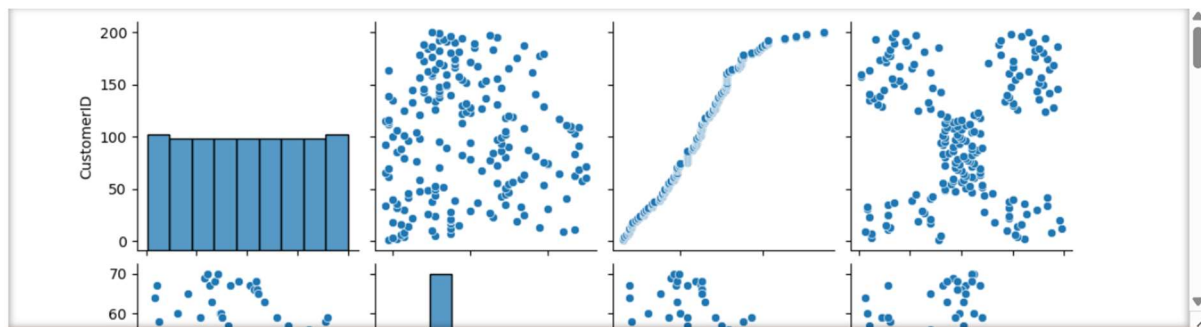
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   CustomerID                           200 non-null    int64
1   Gender                               200 non-null    object
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
```

```
[3]: df.head()
```

```
[3]:
```

	CustomerID	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40

```
[8]: p=sns.pairplot(df)
plt.show()
```



```
[9]: features=df.iloc[:,[3,4]].values
from sklearn.cluster import KMeans
model=KMeans(n_clusters=5)
model.fit(features)
KMeans(n_clusters=5)
```

C:\ProgramData\anaconda3\Lib\site-packages\sklearn\cluster\\_kmeans.py:1419: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

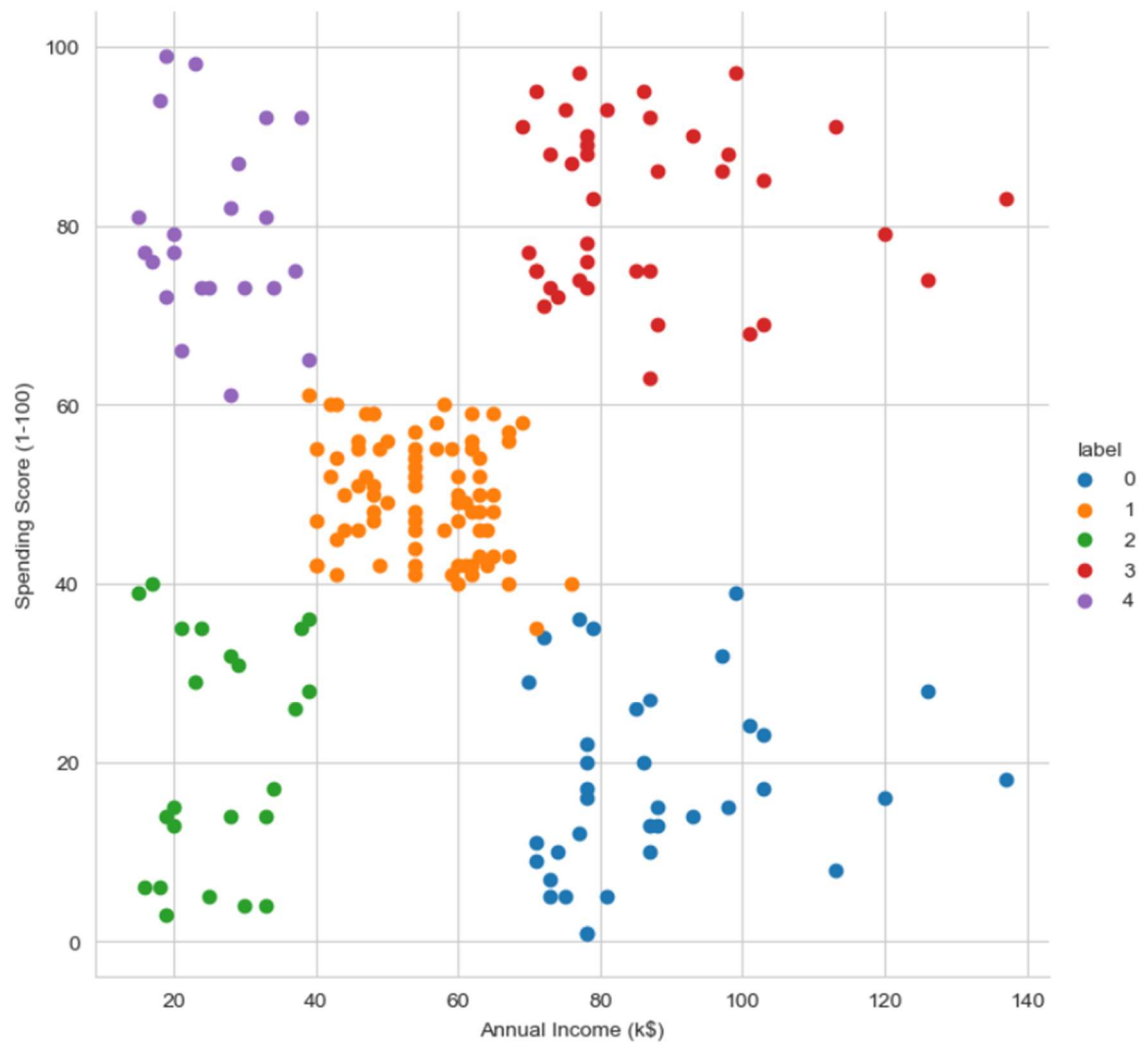
```
[9]: KMeans
KMeans(n_clusters=5)
```

```
[10]: Final=df.iloc[:,[3,4]]
Final['label']=model.predict(features)
Final.head()
```

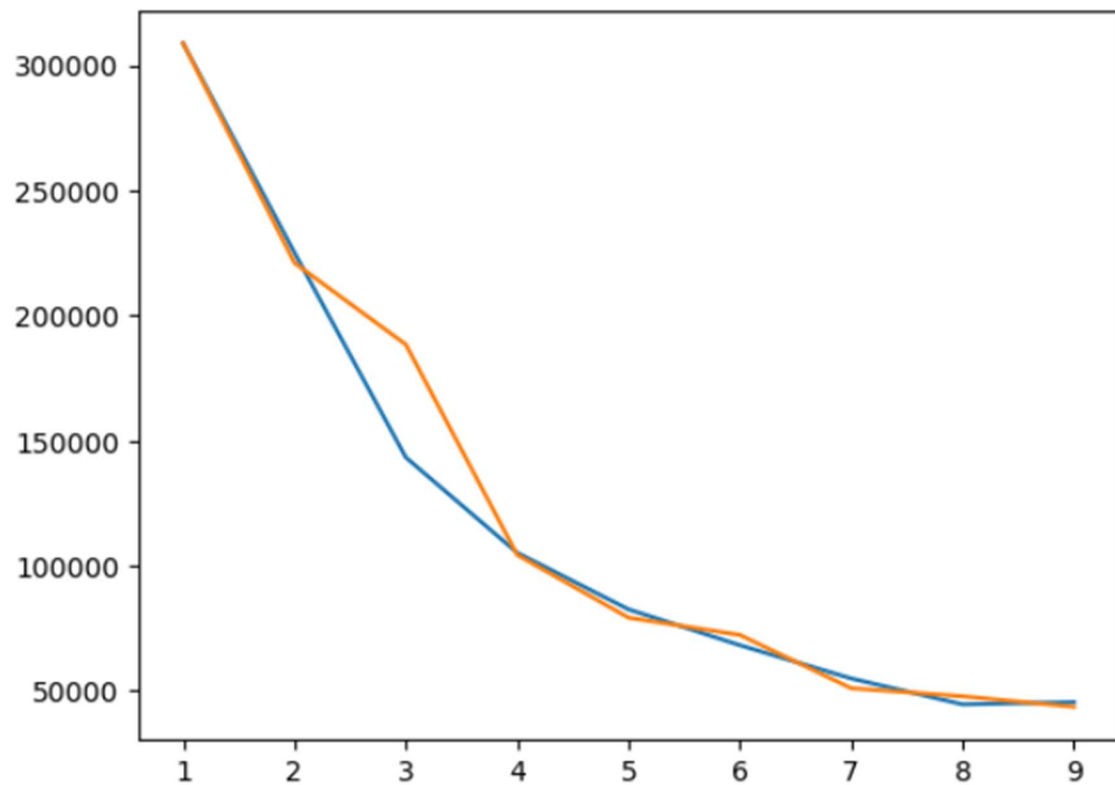
10]:

	Annual Income (k\$)	Spending Score (1-100)	label
0	15	39	2
1	15	81	4
2	16	6	2
3	16	77	4
4	17	40	2

```
16]: sns.set_style("whitegrid")
sns.FacetGrid(Final,hue="label",height=7) \
.map(plt.scatter,"Annual Income (k$)", "Spending Score (1-100)") \
.add_legend();
plt.show()
```



```
features_el=df.iloc[:,[2,3,4]].values
from sklearn.cluster import KMeans
wcss=[]
for i in range(1,10):
    model=KMeans(n_clusters=i)
    model.fit(features_el)
    wcss.append(model.inertia_)
plt.plot(range(1,10),wcss)
plt.show()
```



**Result:**

The K-Means algorithm groups customers into 5 clusters based on their spending habits and income. The Elbow Method helps confirm that 5 clusters provide the most balanced segmentation, clearly showing different customer behavior groups.