Importing the Dependencies

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

Data Collection & Analysis

```
# 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()

	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

finding the number of rows and columns
customer_data.shape

(200, 5)

 $\mbox{\tt\#}$ getting some informations about the dataset $\mbox{\tt customer_data.info()}$

<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)							

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

memory usage: 7.9+ KB

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

Choosing the Annual Income Column & Spending Score column

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

```
[ 76 40]
[ 76
     87]
[ 77
     12]
77
     97]
[ 77
     36]
[ 77
     741
 78
     22]
Γ 78
     901
[ 78 17]
[ 78 88]
 78
     20]
 78
     761
[ 78
     16]
 78 89]
 78
      1]
78
     78]
[ 78
      1]
 78 73]
 79
     35]
 79
     83]
[ 81
      5]
 81 93]
 85
     26]
[ 85
     75]
Г 86
     201
[ 86
     95]
[ 87
     27]
[ 87
      63]
Γ 87
     131
[ 87
     75]
[ 87
     101
[ 87
      92]
88
     131
[ 88 86]
[ 88
     151
[ 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]]
```

Choosing the number of clusters

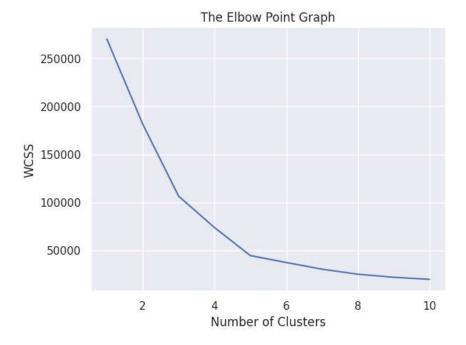
```
WCSS -> Within Clusters Sum of Squares
```

```
# 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_)
```

/usr/local/lib/python3.9/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will chang warnings.warn(

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 warnings.warn(
```

#·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()



Optimum Number of Clusters = 5

Training the k-Means Clustering Model

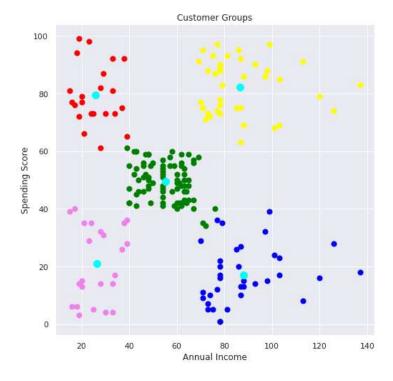
5 Clusters - 0, 1, 2, 3, 4

Visualizing all the Clusters

```
# 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')
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')
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')
plt.xlabel('Annual Income')
plt.ylabel('Spending Score')
plt.show()
```



✓ 0s completed at 2:30 PM

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