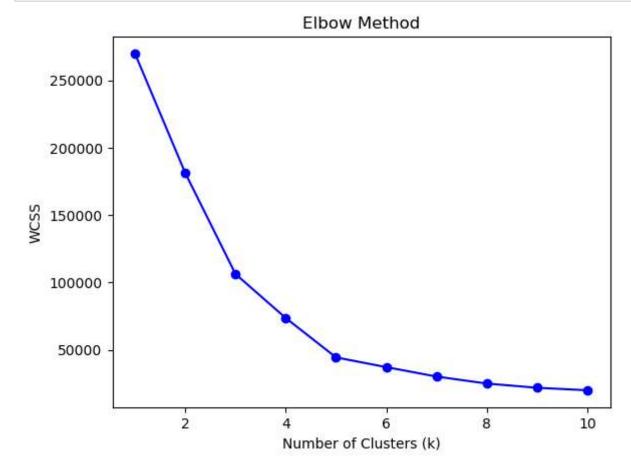
# Making k -means clustering project

```
In [19]: # Importing the Libraries
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

In [20]: # To Importing the dataset
dataset = pd.read_csv(r"C:\Users\HP\Downloads\Machine Learning\7 July, Hierarchical_clustering,K_means_cluster
x = dataset.iloc[:,[3,4]].values
```

## Use elbow method for finding the optimal number of clusters

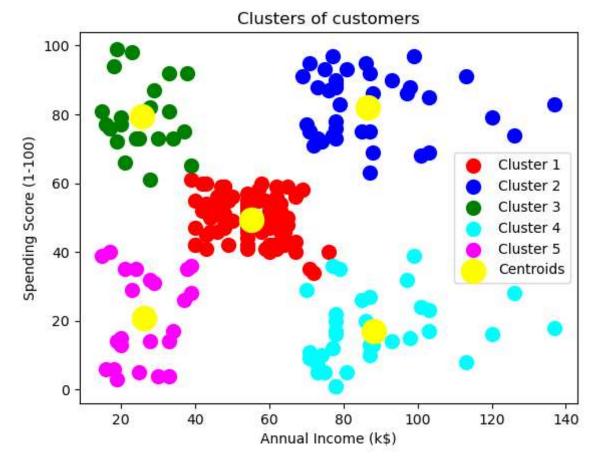
```
In [24]: import warnings
         # Ignore all Warnings:
         warnings.filterwarnings("ignore")
         from sklearn.cluster import KMeans
         # Assuming that you have stored data in 'x'
         # x should be a 2D array or matrix with shape (n_samples, n_features)
         # Initialize an empty list to store the WCSS values for different numbers of clusters
         # Define the range of cluster numbers from for try
         k_values = range (1,11) # Try cluster numbers from 1 to 10
         # Calculate WCSS for each cluster number
         for k in k values:
             kmeans = KMeans(n_clusters = k, random_state = 42)
             kmeans.fit(x)
             wcss.append(kmeans.inertia_) # Inertia is the WCSS value
         # To Plot the WCSS values against the number of clusters:
         plt.title('Elbow Method')
         plt.plot(k values,wcss, 'bo-')
         plt.xlabel('Number of Clusters (k)')
         plt.ylabel('WCSS')
         plt.show()
```



### Training the k-means model on the dataset

```
In [25]: kmeans = KMeans(n_clusters = 5, init = 'k-means++', random_state = 42)
y_kmeans = kmeans.fit_predict(x)
y_kmeans

Out[25]: array([4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2,
```



```
In [32]: dataset['Cluster'] = kmeans.labels_
    dataset.to_csv('modified_dataset.csv', index = False)
# Replace ('modified_dataset.csv' with the desire filename)
```

```
In [34]: dataset.to_csv('modified_dataset.csv', index = False)
```

#### In [35]: dataset.head()

#### Out[35]:

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

```
In [33]: dataset['kmeans'] = y_kmeans
  dataset.head()
```

Out[33]:

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

In [ ]: