import pandas as pd

from google.colab import drive
drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

data=pd.read_csv('/content/drive/MyDrive/Mall_Customers.csv')

data.tail()

\Rightarrow		CustomerID	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
	195	196	Female	35	120	79
	196	197	Female	45	126	28
	197	198	Male	32	126	74
	198	199	Male	32	137	18
	199	200	Male	30	137	83

data.shape

(200, 5)

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)
memory usage: 7.9+ KB

data.isnull().sum()

CustomerID 0
Gender 0
Age 0
Annual Income (k\$) 0
Spending Score (1-100) 0

dtype: int64

data.describe()

	CustomerID	Age	Annual Income (k\$)	Spending Score (1-100)
count	200.000000	200.000000	200.000000	200.000000
mean	100.500000	38.850000	60.560000	50.200000
std	57.879185	13.969007	26.264721	25.823522
min	1.000000	18.000000	15.000000	1.000000
25%	50.750000	28.750000	41.500000	34.750000
50%	100.500000	36.000000	61.500000	50.000000
75%	150.250000	49.000000	78.000000	73.000000
max	200.000000	70.000000	137.000000	99.000000

Kmean Clustering

data.columns

```
Index(['CustomerID', 'Gender', 'Age', 'Annual Income (k$)',
          'Spending Score (1-100)'],
         dtype='object')
x=data[{'Annual Income (k$)','Spending Score (1-100)'}]
    <ipython-input-14-368ec8eb8f7a>:1: FutureWarning: Passing a set as an indexer is deprecated and will raise in a future version. Use a li
      x=data[{'Annual Income (k$)','Spending Score (1-100)'}]
from sklearn.cluster import KMeans
k means=KMeans()
k_{means.fit(x)}
   'local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The de
   eans
   ns()
k_means=KMeans(n_clusters=5)
k_means.fit_predict(x)
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
     warnings.warn(
    array([4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3,
          4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 2,
          2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 1, 0, 1, 2, 1, 0, 1, 0, 1,
          2, 1, 0, 1, 0, 1, 0, 1, 0, 1, 2, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1,
          0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1,
          0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1,
          0, 1], dtype=int32)
    4
```

Elbow Method to find optimal number of Clusters

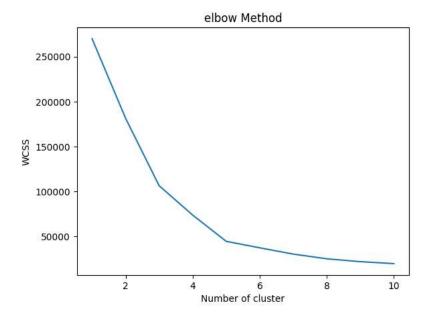
[269981.28,

181363.59595959593,

```
wcss=[]
for i in range(1,11):
 k_means=KMeans(n_clusters=i)
 k_{means.fit(x)}
 wcss.append(k_means.inertia_)
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
      warnings.warn(
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
       warnings.warn(
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
      warnings.warn(
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
      warnings.warn(
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/ kmeans.py:870: FutureWarning: The default value of `n init` will change from 10
       warnings.warn(
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
      warnings.warn(
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
       warnings.warn(
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
      warnings.warn(
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
      warnings.warn(
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/ kmeans.py:870: FutureWarning: The default value of `n init` will change from 10
       warnings.warn(
    - I
wcss
```

```
106348.37306211122,
73679.78903948836,
44448.4554479337,
37233.814510710006,
30241.34361793658,
24995.96978113596,
21818.114588452176,
19646.482018947238]
```

import matplotlib.pyplot as plt
plt.plot(range(1,11),wcss)
plt.title("elbow Method")
plt.xlabel("Number of cluster")
plt.ylabel("WCSS")
plt.show()



Model Training

 $x=data[{'Annual Income (k$)', 'Spending Score (1-100)'}]$

<ipython-input-21-368ec8eb8f7a>:1: FutureWarning: Passing a set as an indexer is deprecated and will raise in a future version. Use a li x=data[{'Annual Income (k\$)','Spending Score (1-100)'}]

x=data[{'Annual Income (k\$)','Spending Score (1-100)'}]

Х

	Spending Score (1-100)) Annual Income (k\$)
0	39	9 15
1	8′	1 15
2	6	5 16
3	77	7 16
4	40) 17
195	79	9 120
196	28	3 126
197	74	126
198	18	3 137
199	83	3 137
 195 196 197 198	 79 28 74 18	 9 120 3 126 4 126 3 137

200 rows × 2 columns

```
k_means=KMeans(n_clusters=5,random_state=42)
y_means=k_means.fit_predict(x)
```

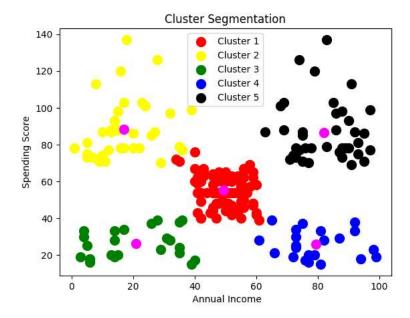
/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10 warnings.warn(

```
→
```

y_means

```
array([2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3,
```

```
plt.scatter(x.iloc[y_means==0,0],x.iloc[y_means==0,1],s=100,c='red',label="Cluster 1")
plt.scatter(x.iloc[y_means==1,0],x.iloc[y_means==1,1],s=100,c='yellow',label="Cluster 2")
plt.scatter(x.iloc[y_means==2,0],x.iloc[y_means==2,1],s=100,c='green',label="Cluster 3")
plt.scatter(x.iloc[y_means==3,0],x.iloc[y_means==3,1],s=100,c='blue',label="Cluster 4")
plt.scatter(x.iloc[y_means==4,0],x.iloc[y_means==4,1],s=100,c='black',label="Cluster 5")
plt.scatter(k_means.cluster_centers_[:,0],k_means.cluster_centers_[:,1],s=100,c='magenta')
plt.title("Cluster Segmentation")
plt.xlabel(" Annual Income")
plt.ylabel("Spending Score")
plt.legend()
plt.show()
```



k_means.predict([[13,70]])

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but KMeans was fitted wit warnings.warn(array([1], dtype=int32)

Save the Model

```
import joblib

joblib.dump(k_means,"customer_segmentation")

['customer_segmentation']

model=joblib.load("customer_segmentation")
```

4

model.predict([[15,39]])

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but KMeans was fitted wit warnings.warn(
array([2], dtype=int32)