import pandas as pd

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

data.head()

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

Next steps:

Generate code with data



View recommended plots

data.tail()

→		CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)	
	195	196	Female	35	120	79	ıl.
	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

 \rightarrow (200, 5)

```
10 random numbers using numpy
*/ Generate
                                                                                Q
                                                                                         Close
```

print("number of rows", data.shape[0]) print("number of columns", data.shape[1])

number of rows 200 number of columns 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	Genre	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()

dtype: int64

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

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

data.columns

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

from sklearn.cluster import KMeans

k_means=KMeans()
k_means.fit(x)

```
\rightarrow
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning
      warnings.warn(
     ▼ KMeans
     KMeans()
                                                                                   k means=KMeans(n clusters=5)
k_means.fit_predict(x)
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning
      warnings.warn(
    array([0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3,
           0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 7, 3, 7, 7,
           7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 2, 7, 7, 2, 7, 7, 7, 7,
           2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 1, 2, 1, 2, 1, 5, 1, 5, 1,
           2, 1, 5, 1, 5, 1, 5, 1, 5, 1, 2, 1, 5, 1, 2, 1, 5, 1, 5, 1, 5, 1,
           5, 1, 5, 1, 5, 1, 2, 1, 5, 1, 5, 1, 5, 1, 5, 1, 5, 1, 5, 1, 5, 1,
           5, 1, 5, 1, 4, 6, 4, 6, 4, 6, 4, 6, 4, 6, 4, 6, 4, 6, 4, 6, 4, 6,
           4, 6], dtype=int32)
wcss=[]
for i in range(1,11):
k means= KMeans(n_clusters=i)
k_means.fit_predict(x)
wcss.append(k_means.inertia_)
→ /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/ kmeans.py:870: FutureWarning
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/ kmeans.py:870: FutureWarning
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning
      warnings.warn(
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/ kmeans.py:870: FutureWarning
      warnings.warn(
```

wcss

```
[269981.28,

181363.59595959593,

106348.37306211122,

73679.78903948836,

44448.4554479337,

37233.814510710006,

30259.65720728547,

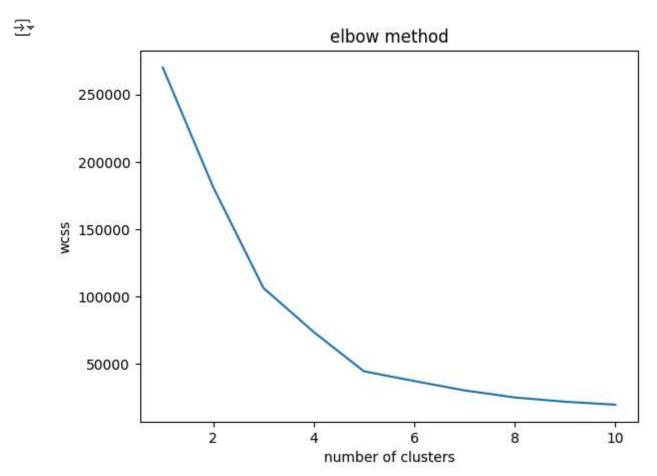
25018.781613414067,

21862.092672182895,

19641.456216651564]
```

import matplotlib.pyplot as plt

```
plt.plot(range(1,11),wcss)
plt.title("elbow method")
plt.xlabel("number of clusters")
plt.ylabel("wcss")
plt.show()
```



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

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

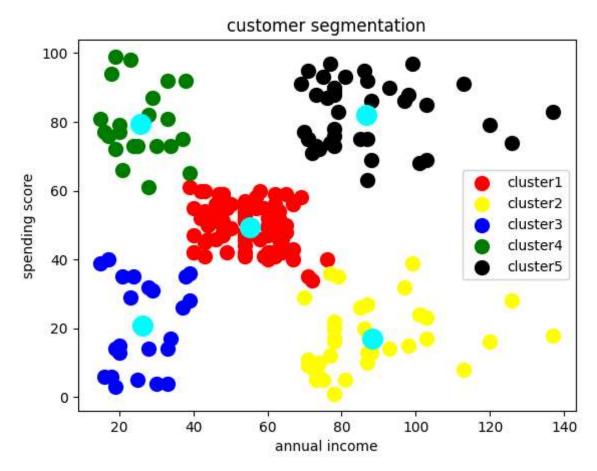
/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning warnings.warn(

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,
                                                                                                            2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3,
                                                                                                            0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                                                                                                            0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
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                                                                                                            0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                                                                                                                                                                                                                                                                                                         0, 0, 0,
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                                                                                                                                                                                                                  0, 0,
                                                                                                                                                                                                                                                       0, 4,
                                                                                                            1, 4, 1, 4, 1, 4, 0, 4, 1, 4, 1, 4, 1, 4, 1, 4,
                                                                                                            1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4,
                                                         4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1,
                                         1, 4], dtype=int32)
```

```
plt.scatter(x.iloc[Y_means==0,0],x.iloc[Y_means==0,1],s=100,c='red',label='cluster1
plt.scatter(x.iloc[Y_means==1,0],x.iloc[Y_means==1,1],s=100,c='yellow',label='cluster
plt.scatter(x.iloc[Y_means==2,0],x.iloc[Y_means==2,1],s=100,c='blue',label='cluster
plt.scatter(x.iloc[Y_means==3,0],x.iloc[Y_means==3,1],s=100,c='green',label='cluster
plt.scatter(x.iloc[Y_means==4,0],x.iloc[Y_means==4,1],s=100,c='black',label='cluster
plt.scatter(k_means.cluster_centers_[:,0],k_means.cluster_centers_[:,1],s=200,c='cy
plt.title("customer segmentation")
plt.xlabel("annual income")
plt.ylabel("spending score")
plt.legend()
plt.show()
```





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



/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not warnings.warn(array([2], dtype=int32)



Start coding or generate with AI.