

## import Libery

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

## Upload Data

```
cust_data=pd.read_csv('Mall_Customers.csv')
cust_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

#check shape
cust_data.shape

(200, 5)

#check null value
cust_data.isnull().sum()

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

cust_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

#seprate columns
cust_data_1=cust_data.iloc[:,3:5].values

scaler = StandardScaler()
X_scaled = scaler.fit_transform(cust_data_1)
```

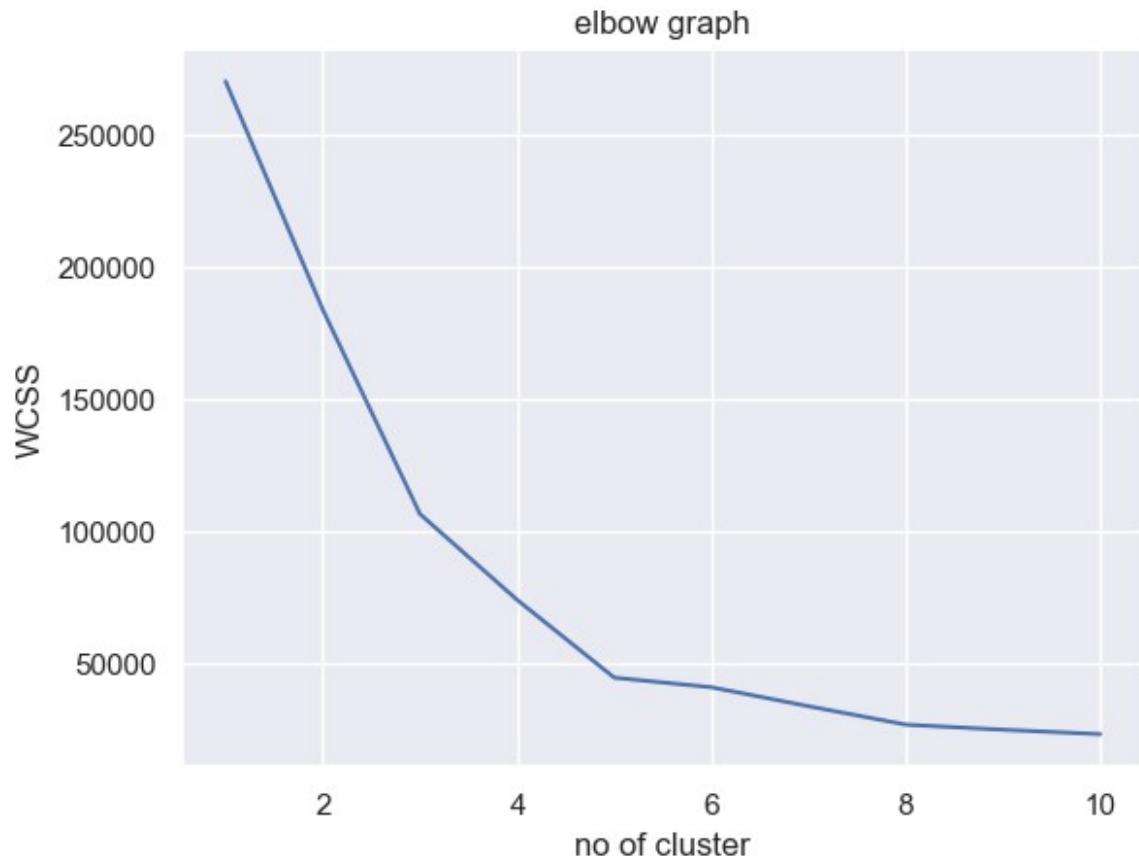
### choosing no of cluster

```
#finding wcss vales for diffrent no of cluster
wcss=[]

for i in range(1,11):
    xkmeans=KMeans( n_clusters=i,init='k-means++',random_state=42)
    xkmeans.fit(cust_data_1)

    wcss.append(xkmeans.inertia_)

#plot elbow graph
sns.set()
plt.plot(range(1,11),wcss)
plt.title('elbow graph')
plt.xlabel('no of cluster')
plt.ylabel('WCSS')
plt.show()
```



in above we consider optimal no of point is 5 because after that there is no any significant point

## Training the kmeans clustering model

```
xkmeans=KMeans(n_clusters=5,init='k-means++',random_state=0)

#label for each data point based on there cluster
y=xkmeans.fit_predict(cust_data_1)
y

array([3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4,
       4,
       3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3,
       0,
       3, 4, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 2, 1, 0, 1, 2, 1,
       2,
       1,
```

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

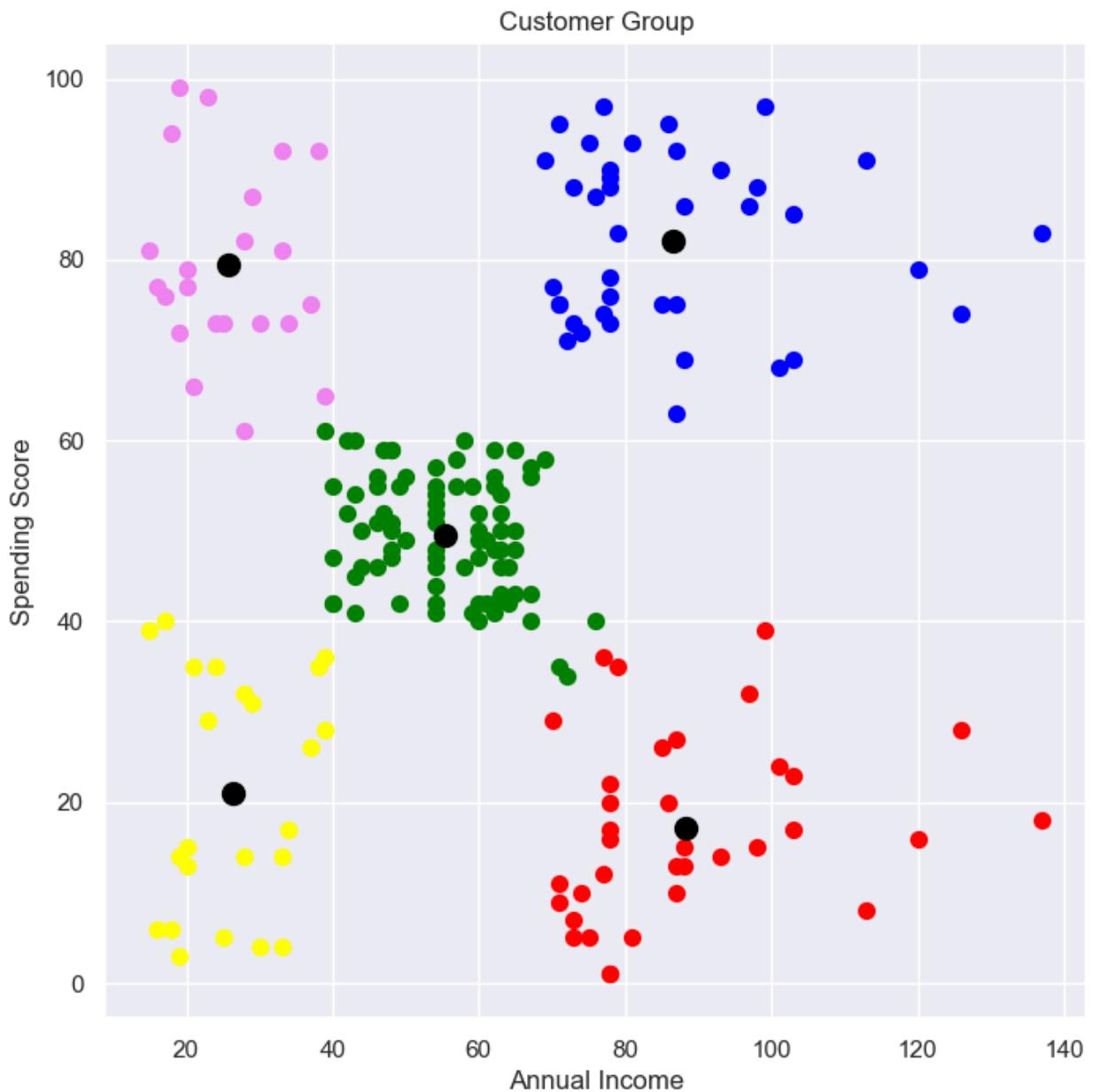
now we Visualizing all 5 clusters

```
plt.figure(figsize=(8,8))
plt.scatter(cust_data_1[y==0,0],cust_data_1[y==0,1],s=50,c='green',label='Cluster 1')
plt.scatter(cust_data_1[y==1,0],cust_data_1[y==1,1],s=50,c='blue',label='Cluster 1')
plt.scatter(cust_data_1[y==2,0],cust_data_1[y==2,1],s=50,c='red',label='Cluster 1')
plt.scatter(cust_data_1[y==3,0],cust_data_1[y==3,1],s=50,c='yellow',label='Cluster 1')
plt.scatter(cust_data_1[y==4,0],cust_data_1[y==4,1],s=50,c='violet',label='Cluster 1')
#in above in 1st y==0,0 this is x coordinate 1st 0 is 0 cluster and
2nd 0 is dataset 1st column and in 2nd y==0,1 is y coordinate

#plot the centroid value
plt.scatter(xkmeans.cluster_centers_[:,0],xkmeans.cluster_centers_[:,1],s=100,c='black',label='Centroid')
#in above 0 is xcentroid point and 1 is y centroid point

plt.title('Customer Group')
plt.xlabel('Annual Income')
plt.ylabel('Spending Score')

plt.show()
```



in above we understand one group of less annual income spending less score and one group is spending more score same for max annual income, now we have to give some offers or some credit score for that group it will help you to improve your spending score

now below we check for age range

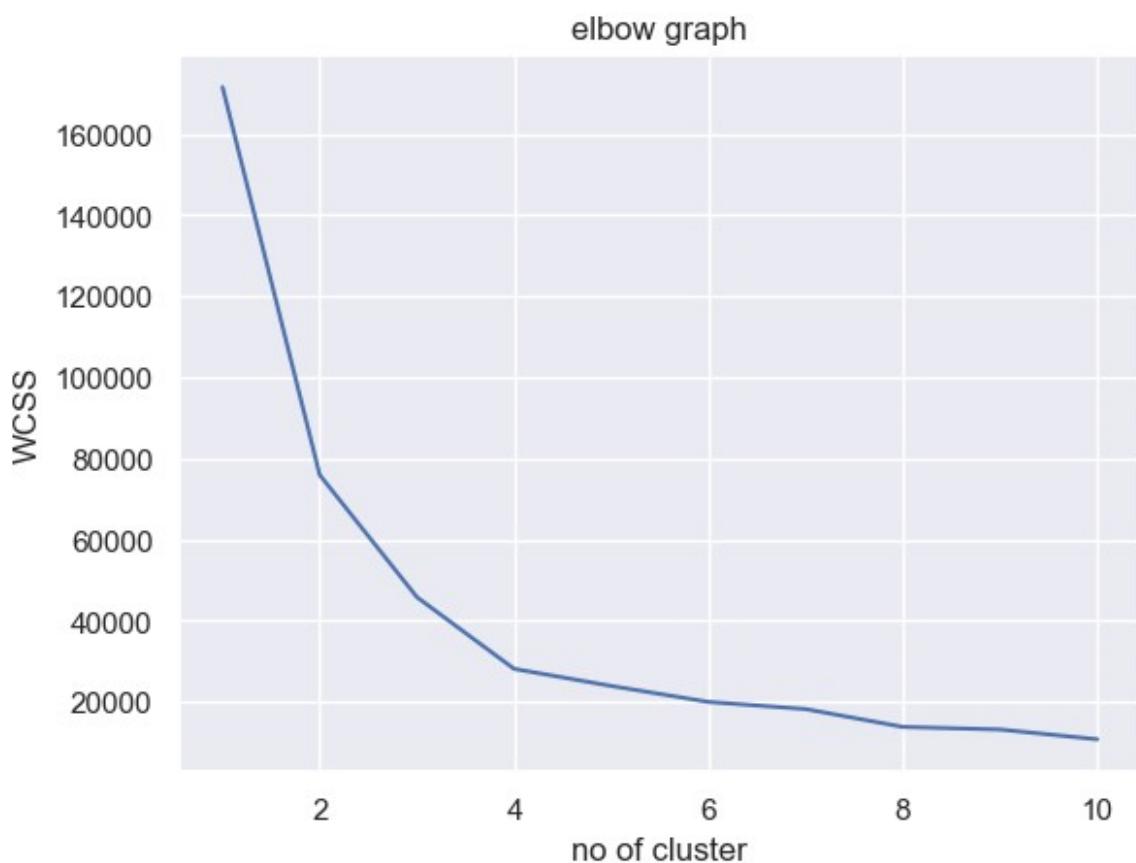
```
#separate columns check with age and score
cust_data_1=cust_data[['Age','Spending Score (1-100)']].values

#finding wcss values for different no of cluster
wcss=[]
```

```
for i in range(1,11):
    xkmeans=KMeans( n_clusters=i,init='k-means++' ,random_state=42)
    xkmeans.fit(cust_data_1)

    wcss.append(xkmeans.inertia_)

#plot elbow graph
sns.set()
plt.plot(range(1,11),wcss)
plt.title('elbow graph')
plt.xlabel('no of cluster')
plt.ylabel('WCSS')
plt.show()
```



```
xkmeans=KMeans(n_clusters=4,init='k-means++',random_state=0)

#label for each data point based on there cluster
y=xkmeans.fit_predict(cust_data_1)
y

array([3, 1, 2, 1, 3, 1, 2, 1, 2, 1, 2, 1, 2, 1, 3, 3, 2, 1, 3,
       1,
       2, 1, 2, 1, 2, 3, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 0, 1, 0,
```

```

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

plt.figure(figsize=(8,8))
plt.scatter(cust_data_1[y==0,0],cust_data_1[y==0,1],s=50,c='green',label='Cluster 1')
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plt.scatter(cust_data_1[y==2,0],cust_data_1[y==2,1],s=50,c='red',label='Cluster 1')
plt.scatter(cust_data_1[y==3,0],cust_data_1[y==3,1],s=50,c='yellow',label='Cluster 1')

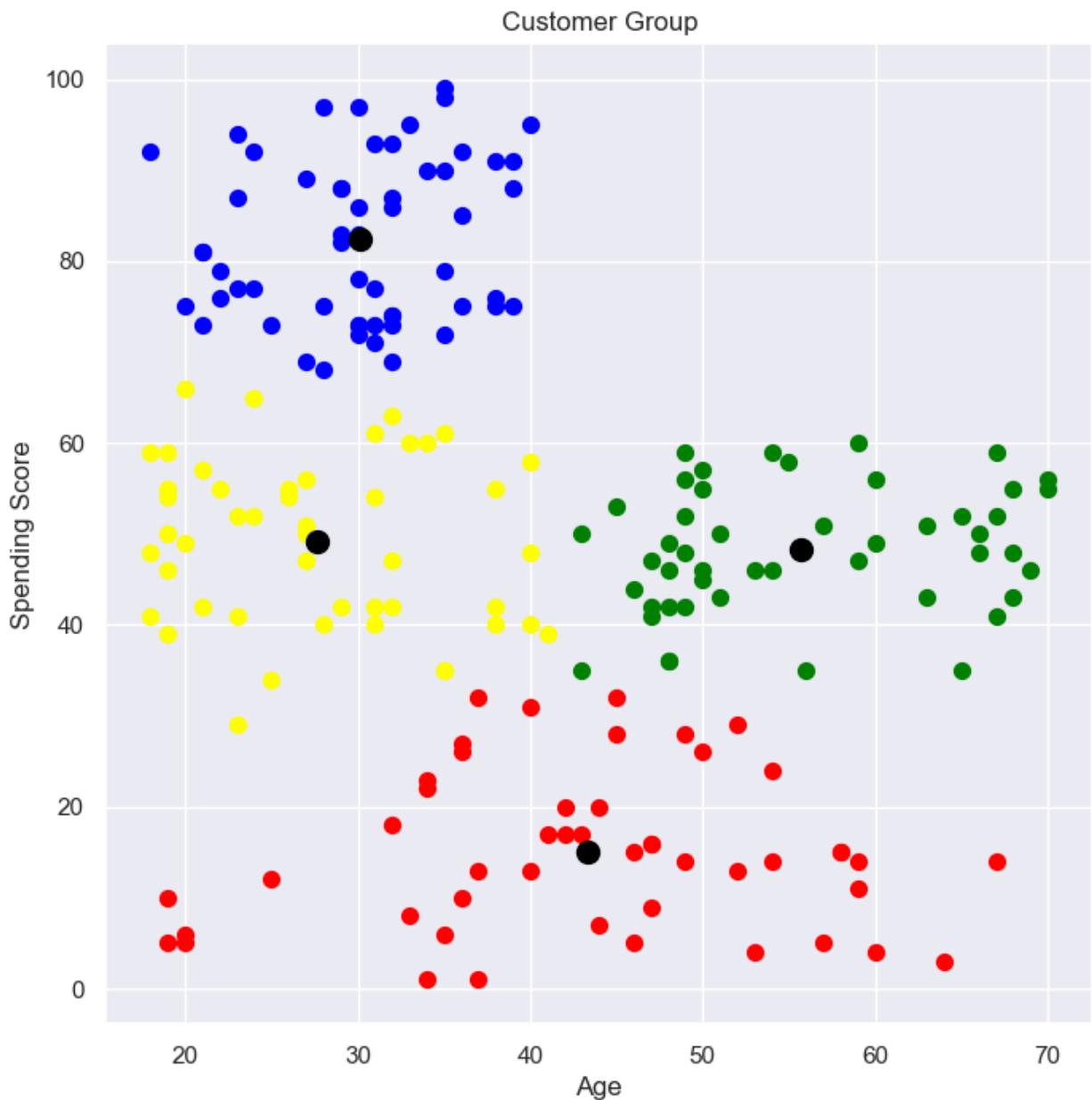
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#2nd 0 is dataset 1st column and in 2nd y==0,1 is y coordinate

#plot the centroid value
plt.scatter(xkmeans.cluster_centers_[:,0],xkmeans.cluster_centers_[:,1],s=100,c='black',label='Centroid')
#in above 0 is xcentroid point and 1 is y centroid point

plt.title('Customer Group')
plt.xlabel('Age')
plt.ylabel('Spending Score')

plt.show()

```



in above we understand more than 40 age people not spending more score for that we have to give some offers for that aged group

```
# Save model & scaler
pickle.dump(xkmeans, open("model.pkl", "wb"))
pickle.dump(scaler, open("scaler.pkl", "wb"))

import os
print(os.getcwd())
C:\Users\Admin
```