

## Assignment-4

### **Problem Statement: Customer Segmentation Analysis**

#### **Problem Statement:**

You own the mall and want to understand the customers who can quickly converge [Target Customers] so that the insight can be given to the marketing team and plan the strategy accordingly.

#### **Clustering the data and performing classification algorithms**

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

- 1. Download the dataset: Dataset**
- 2. Load the dataset into the tool.**

[5]

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

Mounted at /content/drive

[6]

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

### 3. Perform Below Visualizations. • Univariate Analysis • Bi- Variate Analysis • Multi-Variate Analysis

[7]

```
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

[8]

```
data.rename(columns={"CustomerID":"customer_id","Gender":"gender","Age":"age",  
                    "Annual Income (k$)":"annual_income",  
                    "Spending Score (1-100)":"spending_scores"},inplace=True)
```

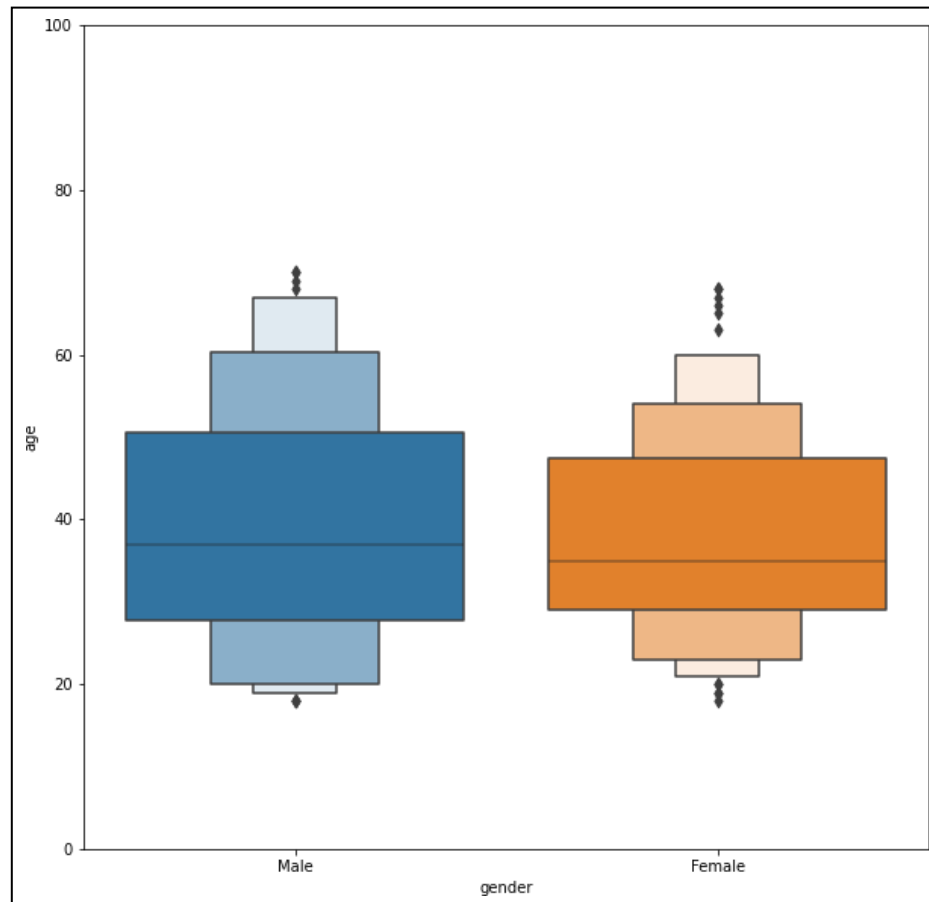
[9]

```
temp = pd.concat([data['age'], data['gender']], axis=1)
```

```
f, ax = plt.subplots(figsize=(10,10))
```

```
fig = sns.boxenplot(x='gender', y="age", data=data)
```

```
fig.axis(ymin=0, ymax=100);
```



## ANALYSIS

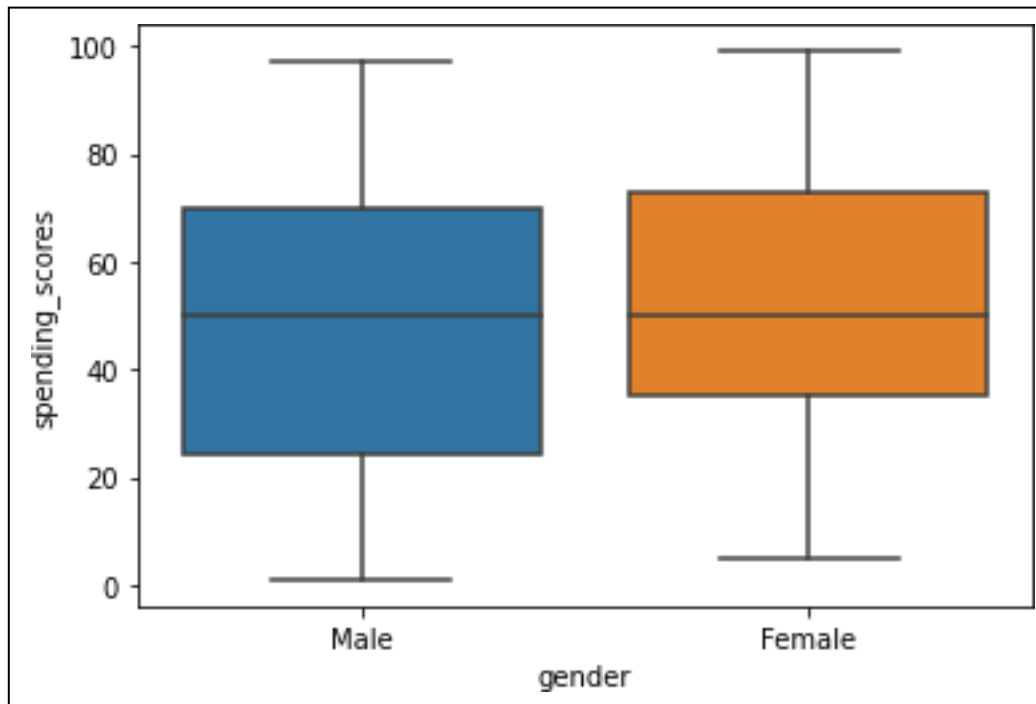
**There is no difference in age of rings for male and female (18-70).**

### Count plot

[10]

```
sns.boxplot(x=data['gender'],y=data['spending_scores'])
```

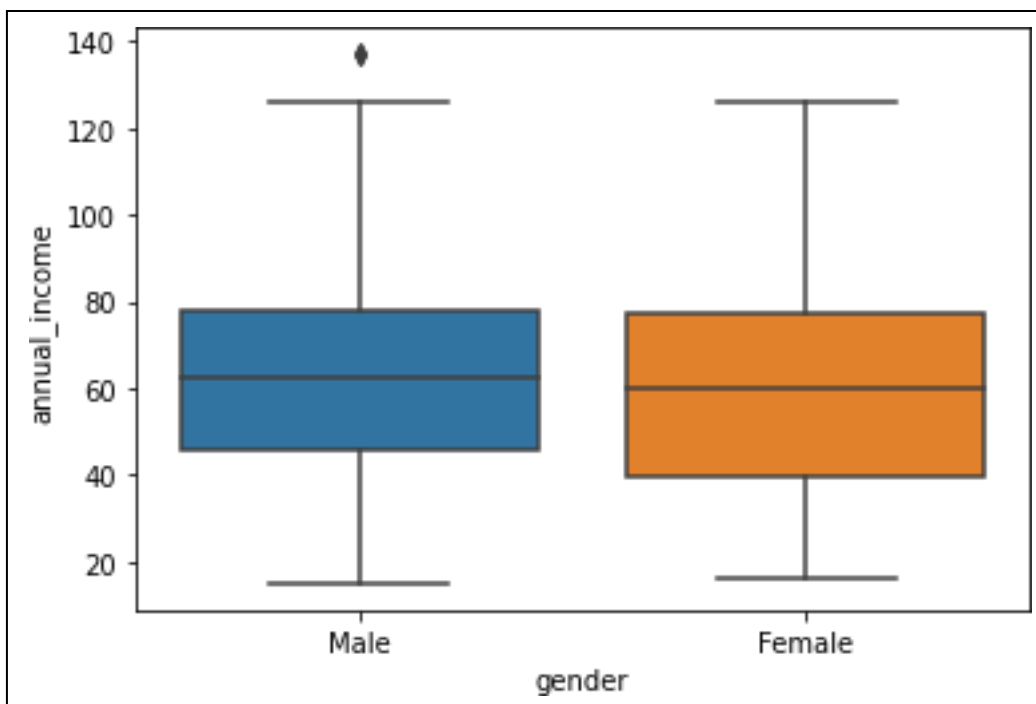
```
<matplotlib.axes._subplots.AxesSubplot at 0x7f801df52310>
```



[11]

```
sns.boxplot(x=data['gender'],y=data['annual_income'])
```

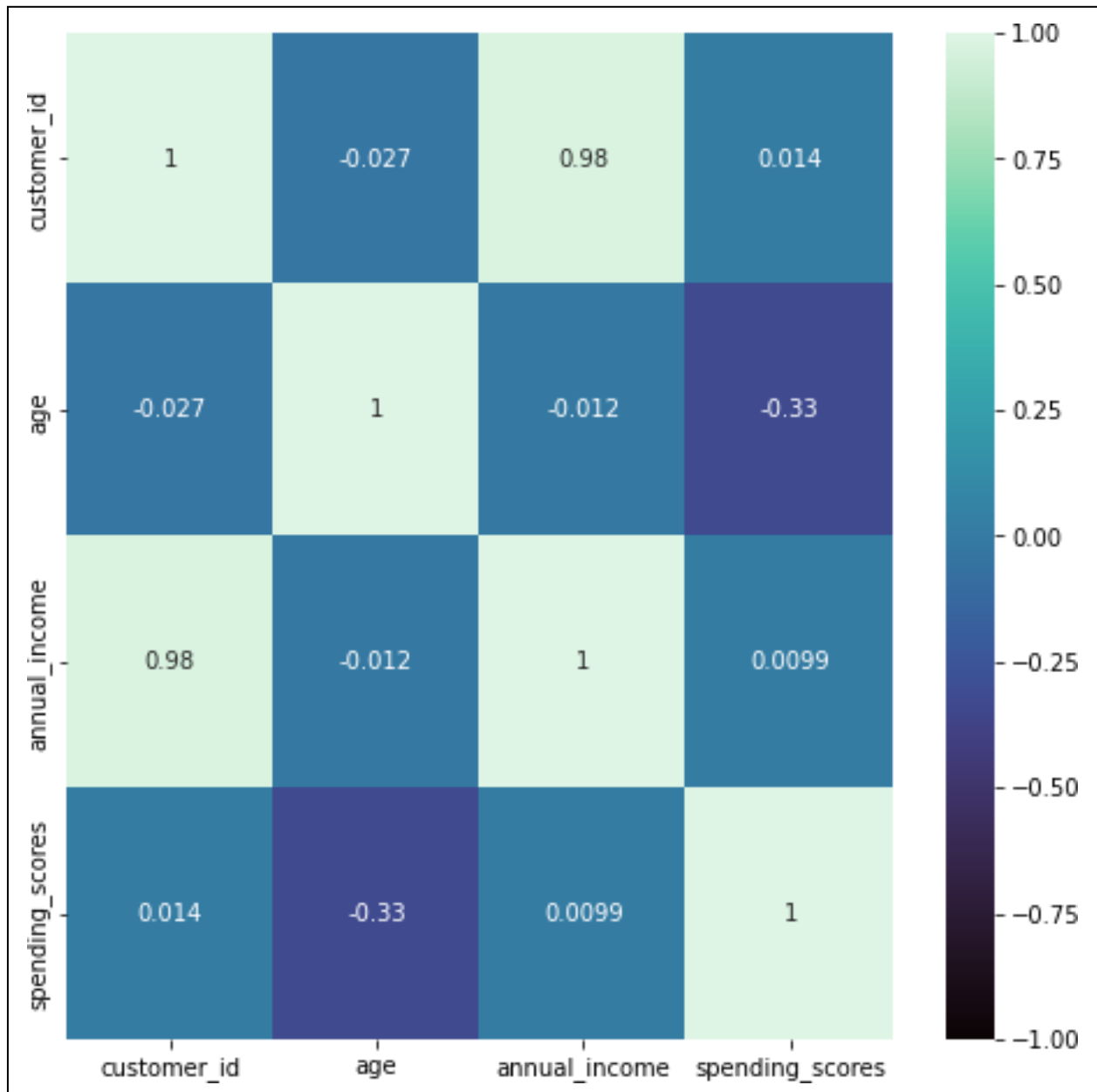
<matplotlib.axes.\_subplots.AxesSubplot at 0x7f801da98610>



Coorelation Plot

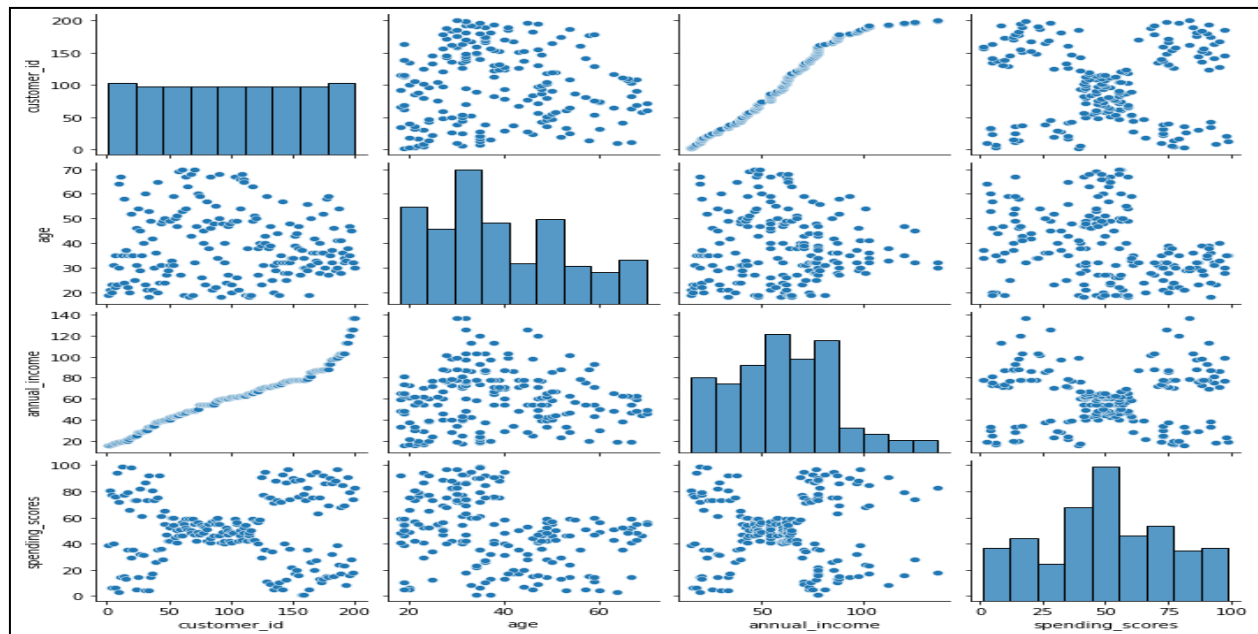
[12]

```
corr=data.corr()  
plt.figure(figsize=(8,8))  
sn=sns.heatmap(corr,vmin=-1,center=0, annot = True, cmap = 'mako')
```



[13]

```
sns.pairplot(data)  
<seaborn.axisgrid.PairGrid at 0x7f801d90d890>
```



#### 4. Perform descriptive statistics on the dataset.

[14]

data.head(10)

	customer_id	gender	age	annual_income	spending_scores
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
5	6	Female	22	17	76
6	7	Female	35	18	6
7	8	Female	23	18	94
8	9	Male	64	19	3
9	10	Female	30	19	72

[15]

data.shape

(200, 5)

[16]

data.describe()

	customer_id	age	annual_income	spending_scores
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

[17]

data.info()

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 200 entries, 0 to 199

Data columns (total 5 columns):

# Column Non-Null Count Dtype

-- --

0 customer\_id 200 non-null int64

1 gender 200 non-null object

2 age 200 non-null int64

3 annual\_income 200 non-null int64

4 spending\_scores 200 non-null int64

dtypes: int64(4), object(1)

memory usage: 7.9+ KB

## 5. Check for Missing values and deal with them.

```
[18]
```

```
data[data.duplicated()]
```

```
[19]
```

```
data.isna().sum()
```

```
customer_id    0
```

```
gender         0
```

```
age            0
```

```
annual_income  0
```

```
spending_scores 0
```

```
dtype: int64
```

**There is no missing values and duplicates in dataframe**

## **6. Find the outliers and replace them outliers**

```
[20]
```

```
for i in data:
```

```
    if data[i].dtype=='int64':
```

```
        q1=data[i].quantile(0.25)
```

```
        q3=data[i].quantile(0.75)
```

```
        iqr=q3-q1
```

```
        upper=q3+1.5*iqr
```

```
        lower=q1-1.5*iqr
```

```
        data[i]=np.where(data[i] >upper, upper, data[i])
```

```
        data[i]=np.where(data[i] <lower, lower, data[i])
```

**After removing outliers, boxplot will be like**

```
[21]
```

```
plt.boxplot(data['age'])
```

```
{'whiskers': [<matplotlib.lines.Line2D at 0x7f801b8fc510>,
```

```
             <matplotlib.lines.Line2D at 0x7f801b8fca50>],
```

```
'caps': [<matplotlib.lines.Line2D at 0x7f801b8fcf90>,
```

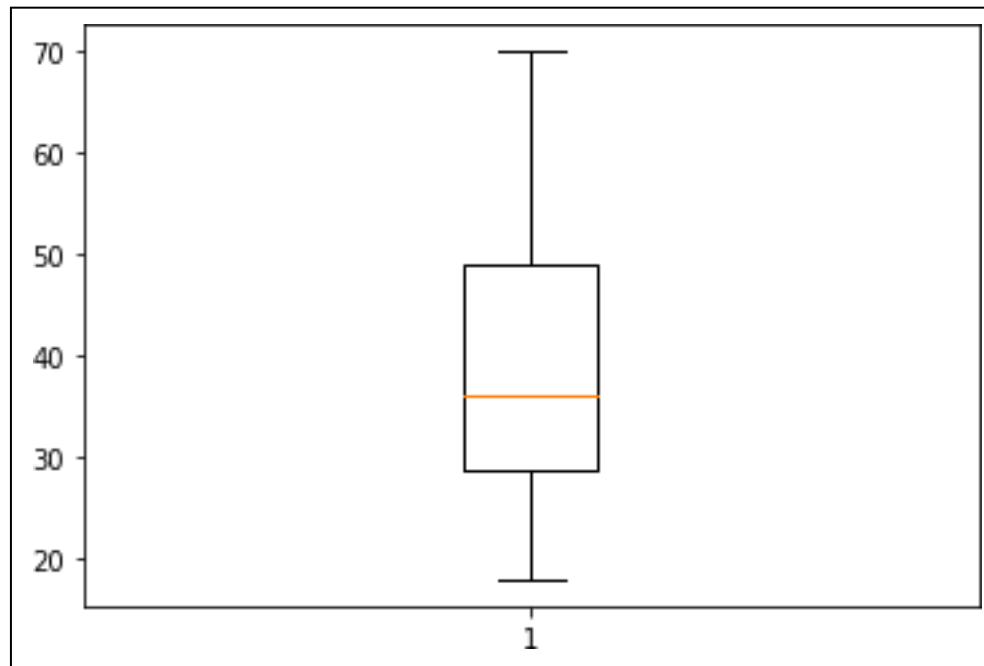
```
         <matplotlib.lines.Line2D at 0x7f801b901510>],
```

```
'boxes': [<matplotlib.lines.Line2D at 0x7f801b8f6ed0>],
```

```
'medians': [<matplotlib.lines.Line2D at 0x7f801b901a90>],
```



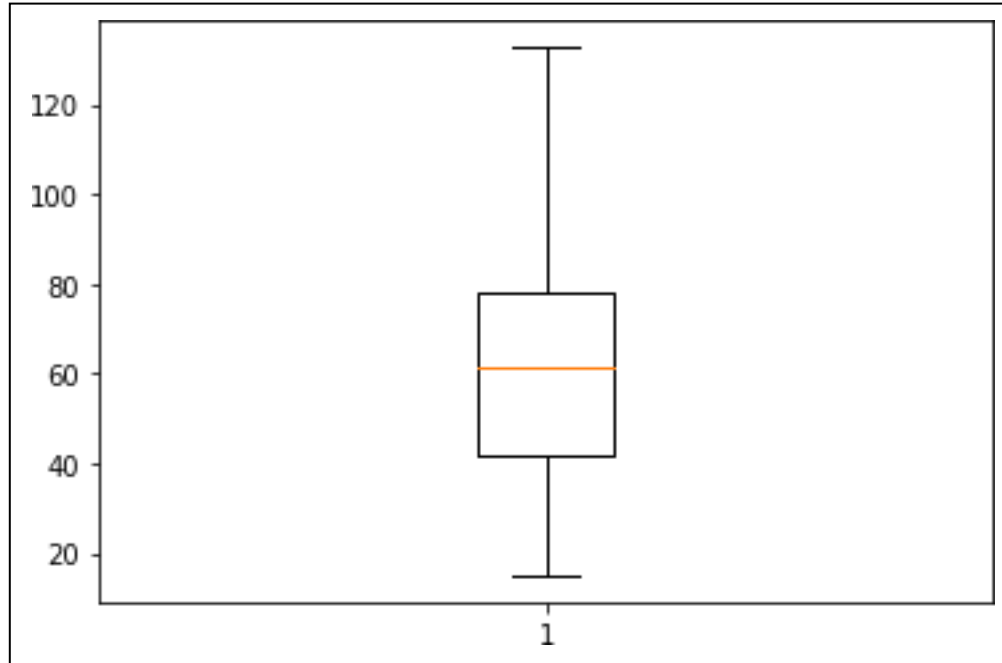
```
'fliers': [<matplotlib.lines.Line2D at 0x7f801b901fd0>],  
'means': []}
```



```
[22]
```

```
plt.boxplot(data['annual_income'])
```

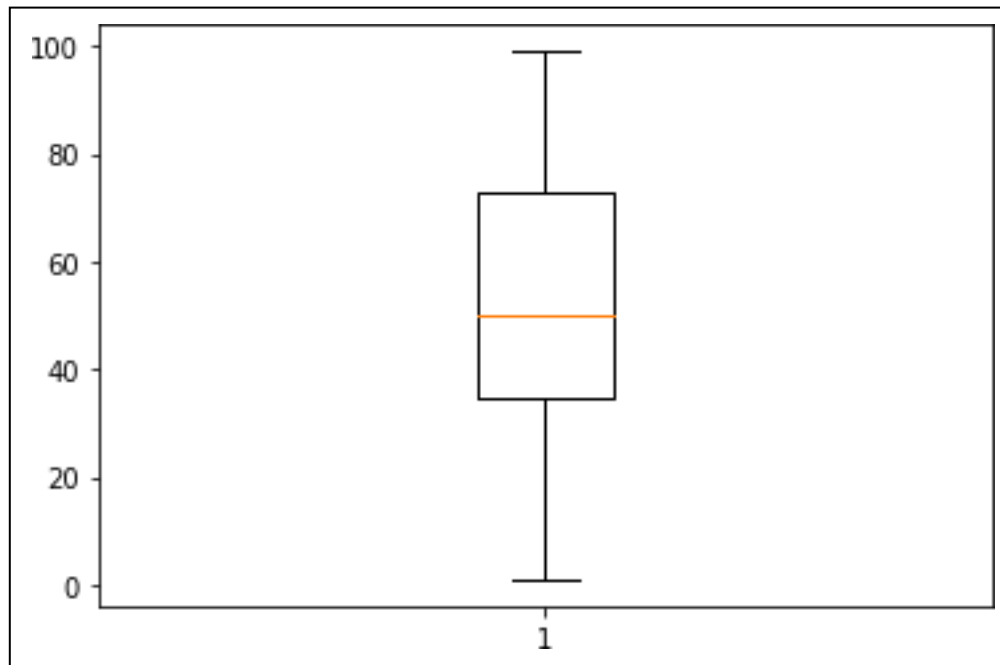
```
{'whiskers': [<matplotlib.lines.Line2D at 0x7f801b8a0510>,  
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'fliers': [<matplotlib.lines.Line2D at 0x7f801b8a8fd0>],  
'means': []}
```



[23]

```
plt.boxplot(data['spending_scores'])
```

```
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 'medians': [<matplotlib.lines.Line2D at 0x7f801b8184d0>],  
 'fliers': [<matplotlib.lines.Line2D at 0x7f801b818a10>],  
 'means': []}
```



## 7. Check for Categorical columns and perform encoding.

[24]

```
from sklearn.preprocessing import LabelEncoder
encoder=LabelEncoder()
data['gender']=encoder.fit_transform(data['gender'])
```

[25]

```
data.head()
```

	customer_id	gender	age	annual_income	spending_scores
<b>0</b>	1.0	1	19.0	15.0	39.0
<b>1</b>	2.0	1	21.0	15.0	81.0
<b>2</b>	3.0	0	20.0	16.0	6.0
<b>3</b>	4.0	0	23.0	16.0	77.0
<b>4</b>	5.0	0	31.0	17.0	40.0

## 8. Scaling the data

[26]

```
from sklearn.preprocessing import StandardScaler  
df=StandardScaler()  
data1=df.fit_transform(data)
```

[27]

data1

```
array([[ -1.7234121,  1.12815215, -1.42456879, -1.74542941, -0.43480148],  
       [ -1.70609137,  1.12815215, -1.28103541, -1.74542941,  1.19570407],  
       [ -1.68877065, -0.88640526, -1.3528021 , -1.70708307, -1.71591298],  
       [ -1.67144992, -0.88640526, -1.13750203, -1.70708307,  1.04041783],  
       [ -1.6541292 , -0.88640526, -0.56336851, -1.66873673, -0.39597992],  
       [ -1.63680847, -0.88640526, -1.20926872, -1.66873673,  1.00159627],  
       [ -1.61948775, -0.88640526, -0.27630176, -1.6303904 , -1.71591298],  
       [ -1.60216702, -0.88640526, -1.13750203, -1.6303904 ,  1.70038436],  
       [ -1.5848463 ,  1.12815215,  1.80493225, -1.59204406, -1.83237767],  
       [ -1.56752558, -0.88640526, -0.6351352 , -1.59204406,  0.84631002],  
       [ -1.55020485,  1.12815215,  2.02023231, -1.59204406, -1.4053405 ],  
       [ -1.53288413, -0.88640526, -0.27630176, -1.59204406,  1.89449216],  
       [ -1.5155634 , -0.88640526,  1.37433211, -1.55369772, -1.36651894],  
       [ -1.49824268, -0.88640526, -1.06573534, -1.55369772,  1.04041783],  
       [ -1.48092195,  1.12815215, -0.13276838, -1.55369772, -1.44416206],  
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```

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```

## 9. Perform any of the clustering algorithms

[28]

```
from sklearn.cluster import KMeans
```

[29]

```
data.drop('customer_id',axis=1,inplace=True)
```

[30]

```
km = KMeans(n_clusters=3, random_state=0)
```

[31]

```
data['Group or Cluster'] = km.fit_predict(data)
```

[32]

```
data.head()
```

	gender	age	annual_income	spending_scores	Group or Cluster
0	1	19.0	15.0	39.0	2
1	1	21.0	15.0	81.0	2
2	0	20.0	16.0	6.0	2
3	0	23.0	16.0	77.0	2
4	0	31.0	17.0	40.0	2

[33]

```
data['Group or Cluster'].value_counts()
```

```
2    123
```

```
1     39
```

```
0     38
```

```
Name: Group or Cluster, dtype: int64
```

[34]

```
import matplotlib.pyplot as plt
```

```
fig,ax = plt.subplots(figsize=(15,8))
```

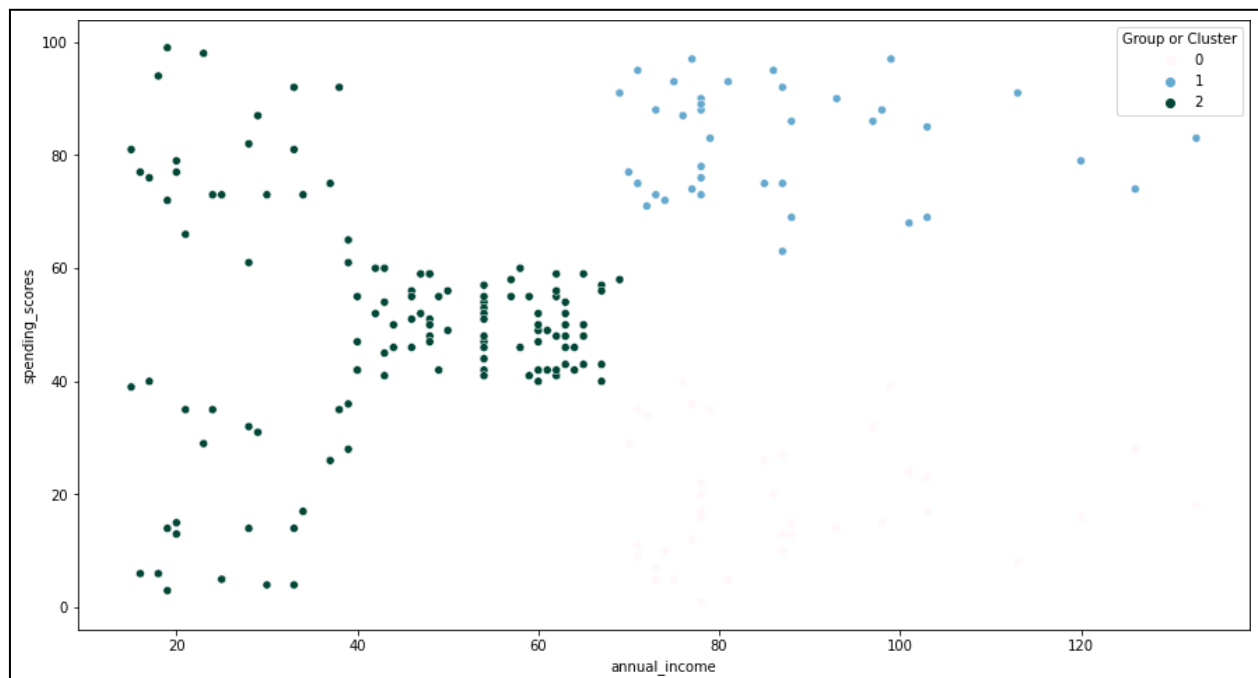
```
sns.scatterplot(x=data['annual_income'],
```

```
                y=data['spending_scores'],
```

```
                hue=data['Group or Cluster'],
```

```
                palette='PuBuGn')
```

```
plt.show()
```



[35]

```
from sklearn.metrics import silhouette_score, silhouette_samples
score = silhouette_score(data,
                        km.labels_,
                        metric='euclidean')

score
```

**0.3842057644019546**

[36]

```
import matplotlib.pyplot as plt
from yellowbrick.cluster import SilhouetteVisualizer

fig, ax = plt.subplots(2, 2, figsize=(20,20))
for i in [2, 3, 4, 5]:
    """
    Create KMeans instance for different number of clusters
    """
    km = KMeans(n_clusters=i,
                init='k-means++',
                n_init=10,
                max_iter=100,
```

```

        random_state=0)
q, mod = divmod(i, 2)
'''

Create SilhouetteVisualizer instance with KMeans instance
Fit the visualizer
'''

visualizer = SilhouetteVisualizer(km,
                                   colors='yellowbrick',
                                   ax=ax[q-1][mod])

visualizer.fit(data)

```

