ASSIGNMENT – 4

PROBLEM STATEMENT: CUSTOMER SEGMENTATION ANALYSIS

ASSIGNMENT DATE	17 OCTOBER 2022
STUDENT NAME	A.GANESHKUMAR
STUDENT ROLL NUMBER	CS19011
MAXIMUM MARKS	2 MARKS

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

from google.colab import files data_to_load = files.upload()

< IPython.core.display.HTML object >

Saving Mall_Customers.csv to Mall_Customers.csv

import io

df = pd.read_csv(io.BytesIO(data_to_load['Mall_Customers.csv'])) df.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 df.tail()

    CustomerID  Gender  Age  Annual Income (k$)  Spending Score (1100)

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 df.shape

(200 , 5)

df.info()

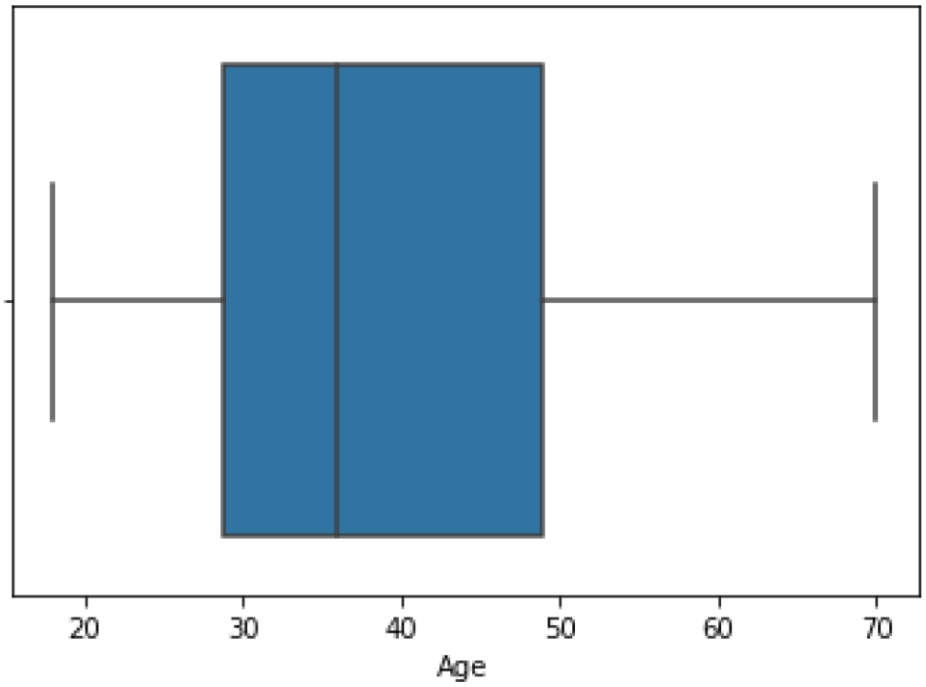
< class 'pandas.core.frame.DataFrame' > RangeIndex: 200 entries, 0 to 199

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

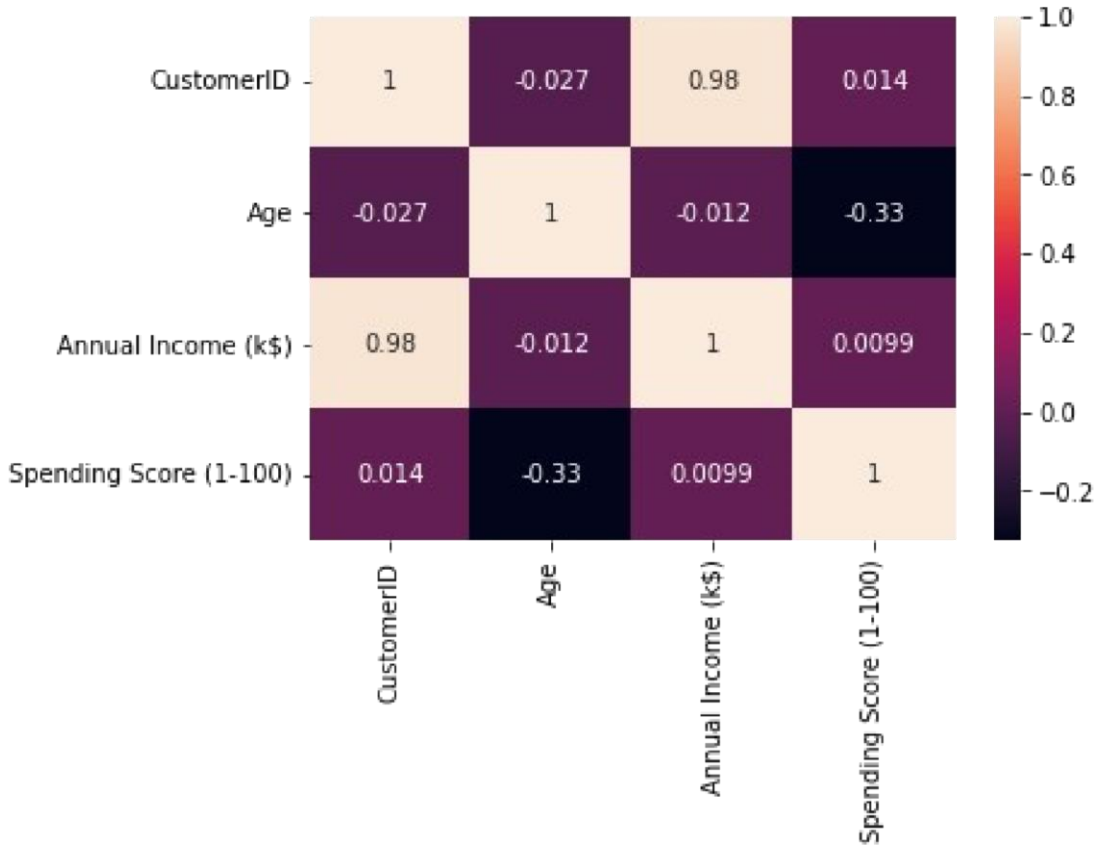
dtypes: int64(4), object(1) memory usage: 7.9+ KB sns.boxplot(x=df['Age'])

< matplotlib.axes._subplots.AxesSubplot at 0x7ff8695244d 0>



sns.heatmap(df.corr(),annot=True)

< matplotlib.axes._subplots.AxesSubplot at 0x7ff869516990>



df.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 df.describe().T

	count	mean	std	min	25%	50%	75 %
CustomerID	200.0	100.50	57.879185	1.0	50.75	100.5	
150.25							
Age	200.0	38.85	13.969007	18.0	28.75	36.0	
49.00							
Annual Income (k$)	200.0	60.56	26.264721	15.0	41.50	61.5	
78.00							
Spending Score (1-100)	200.0	50.20	25.823522	1.0	34.75	50.0	
73.00							

max CustomerID 200.0 Age
70.0 Annual Income (k$) 137.0 Spending Score
(1-100) 99.0 df.isna().sum()

CustomerID 0

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

```
df['Gender'].replace({'Male':1, 'Female':0},inplace=True) df.head()
```

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

df.Gender.unique()
array([1, 0])

```
fig,ax = plt.subplots(figsize=(25,5))

plt.subplot(1,5,1) sns.boxplot(x=df["CustomerID"])
```

```
plt.subplot(1,5,2) sns.boxplot(x=df["Gender"])

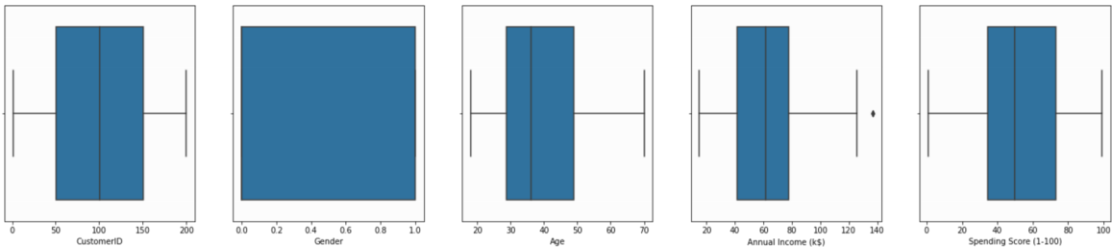
plt.subplot(1,5,3) sns.boxplot(x=df["Age"]) plt.subplot(1,5,4)

sns.boxplot(x=df["Annual Income (k$)"])


plt.subplot(1,5,5)

sns.boxplot(x=df["Spending Score (1-100)"])
```

< matplotlib.axes._subplots.AxesSubplot at 0x 7ff866592c10>




```
qnt=df.quantile(q=[0.25,0.75]) qnt
```

	CustomerID	Gender	Age	Annual Income (k$)	Spending Score (1-100)
0.25	50.75	0.0	28.75	41.5	34.75
0.75	150.25	1.0	49.00	78.0	73.00

```
iqr=qnt.loc[0.75]-qnt.loc[0.25] iqr

CustomerID              99.50

Gender                   1.00 Age                    20.25
Annual Income (k$)       36.50 Spending Score (1-
100)    38.25 dtype: float64

lower=qnt.loc[0.25]-(1.5*iqr) lower

CustomerID             -98.500

Gender                  -1.500 Age                   -1.625
Annual Income (k$)      -13.250 Spending Score (1-
100)   -22.625 dtype: float64

upper=qnt.loc[0.75]+(1.5*iqr) upper

CustomerID             299.500
```

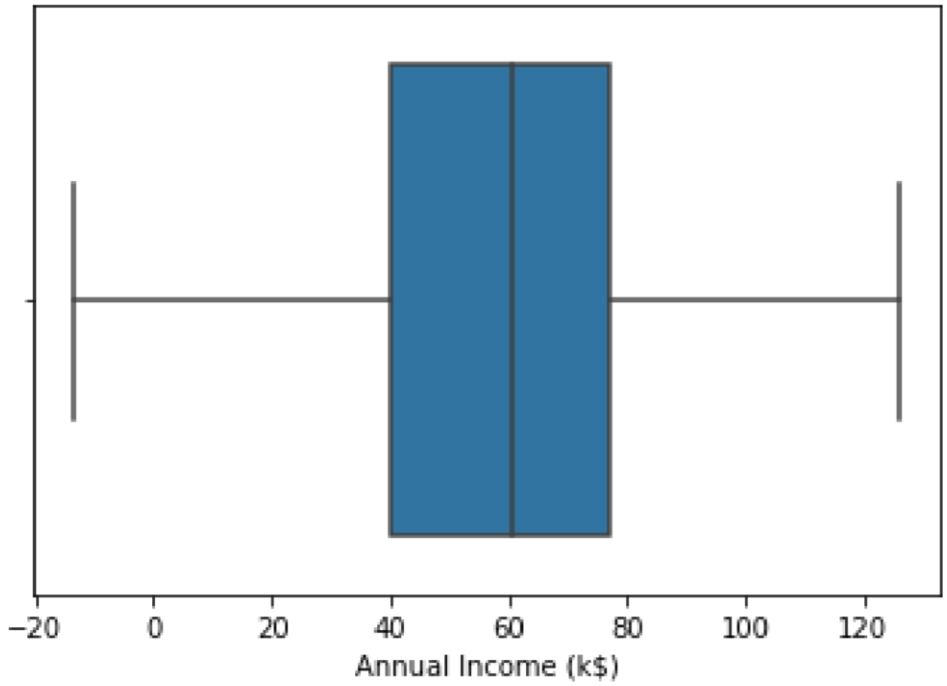
Gender 2.500 Age 79.375
Annual Income (k$) 132.750 Spending Score (1-
100) 130.375 dtype: float64 df.mean()

CustomerID 100.50

Gender 0.44 Age 38.85
Annual Income (k$) 60.56 Spending Score (1-
100) 50.20 dtype: float64

```
df['Annual Income (k$)']=np.where(df['Annual Income (k$)']>132.750,-
13.250,df['Annual Income (k$)']) sns.boxplot(x=df['Annual Income (k$)'])
```

< matplotlib.axes._subplots.AxesSubplot at 0x7ff8664e7ed 0>



df.head()

CustomerID Gender Age Annual Income (k$) Spending Score (1-100) 0 1 1 19
15.0 39

1 2 1 21 15.0 81 2 3 0 20 16.0 6 3 4
0 23 16.0 77 4 5 0 31 17.0 40 df['Gender'] .unique
() array([1, 0])

```
from sklearn.preprocessing import MinMaxScaler sc=MinMaxScaler()

data=sc.fit_transform(df.iloc[:,1:]) data
```

```
array([[1.        , 0.01923077, 0.20287253, 0.3877551 ],
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[1. , 0.26923077, 0. , 0.17346939], [1. , 0.23076923, 0. , 0.83673469]]) from sklearn.cluster import KMeans

TWSS=[] k=list(range(2,9))

```
for i in k:    kmeans=KMeans(n_clusters=i,init='k-means++')
  kmeans.fit(data)
```

```
    TWSS.append(kmeans.inertia_)
```

TWSS

[35.09354046290808 ,

27.37315679730296,

20.211573858371988,

17.210964888908972,

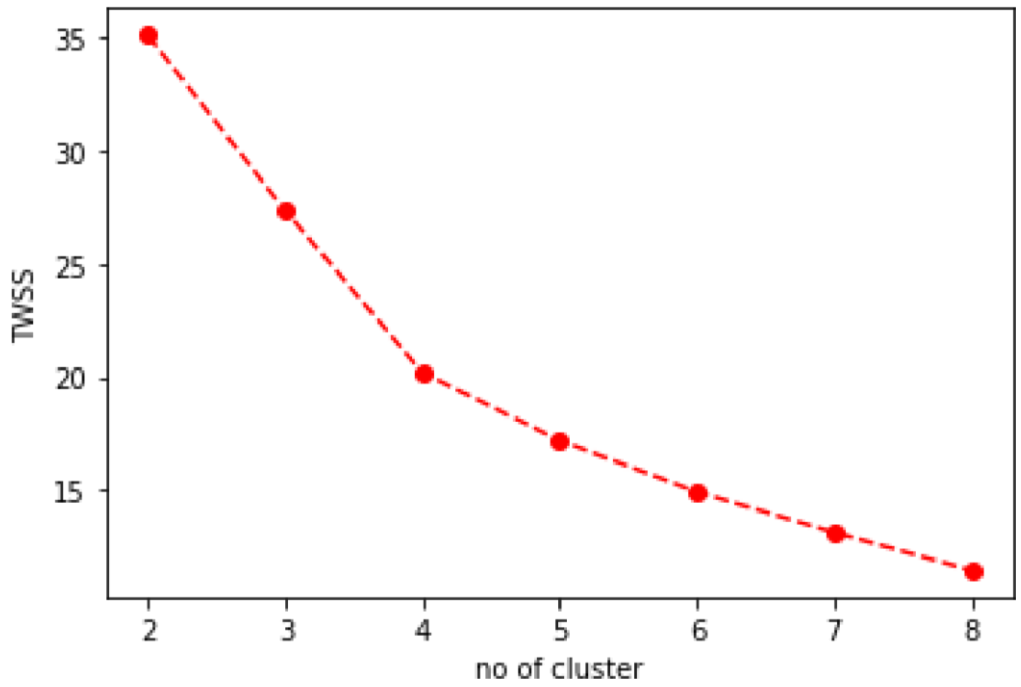
14.941607138943485,

13.153866803186235,

11.485368243450253]

```
plt.plot(k,TWSS,'ro--') plt.xlabel('no of cluster')
plt.ylabel('TWSS')
```

Text(0, 0.5, 'TWSS')




```
model=KMeans(n_clusters=4) model.fit(data)
```

KMeans(n_clusters=4)

```
mb=pd.Series(model.labels_) df.head(3)
```

CustomerID Gender Age Annual Income (k$) Spending Score (1-100) 0 1 1 19 15.0 39

1 2 1 21 15.0 81 2 3 0 20 16.0 6 df.tail()

CustomerID Gender Age Annual Income (k$) Spending Score (1100)

195 196 0 35 120.00

79

196 197 0 45 126.00

28

197 198 1 32 126.00

74

198 199 1 32 -13.25

18

199 200 1 30 -13.25 83

x=df.drop(columns=['CustomerID', 'Gender'],axis=1) x.head()

Age Annual Income (k$) Spending Score (1-100) 0 19 15.0
39

1 21 15.0 81 2 20 16.0 6 3 23
16.0 77 4 31 17.0 40

y=df['Gender'] y

0 1

1 1

2 0

3 0

4 0 ..

195 0

196 0

197 1

198 1

199 1

Name: Gender, Length: 200, dtype: int64 from sklearn.model_selection import train_test_split

x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2) from sklearn.ensemble import

RandomForestClassifier

rf=RandomForestClassifier() rf.fit(x_train,y_train)
RandomForestClassifier()

pred=rf.predict(x_test)

from sklearn.metrics import accuracy_score accuracy_score(y_test,pred)

0.575

from sklearn import metrics metrics.confusion_matrix(y_test,pred)

array([[17, 8],   [9,
6]])