

ASSIGNMENT – 4

PROBLEM STATEMENT: CUSTOMER SEGMENTATION ANALYSIS

ASSIGNMENT DATE	17 OCTOBER 2022
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STUDENT ROLL NUMBER	19BEC027
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
```

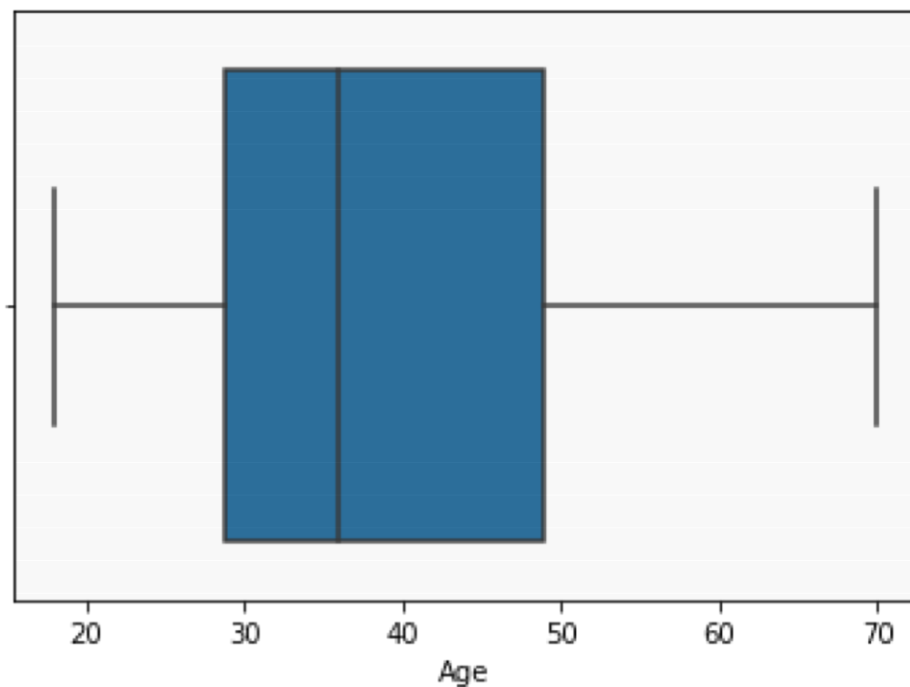
```
Data columns (total 5 columns):
```

```
# Column          Non-Null Count  Dtype
```

```
---  ---
```

```
0 CustomerID      200 non-null  int64 1 Gender      200 nonnull  
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)
```

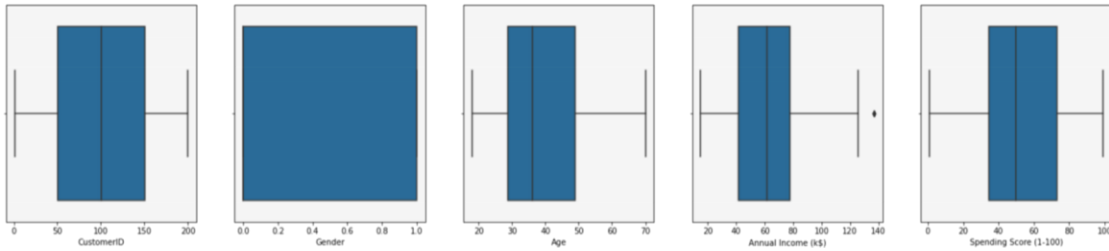
```
sns.boxplot(x=df["CustomerID"]) plt.subplot(1,5,2) sns.boxplot(x=df["Gender"]) plt.subplot(1,5,3)
```

```
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 (1100)

130.375 dtype: float64 df.mean()

CustomerID 100.50

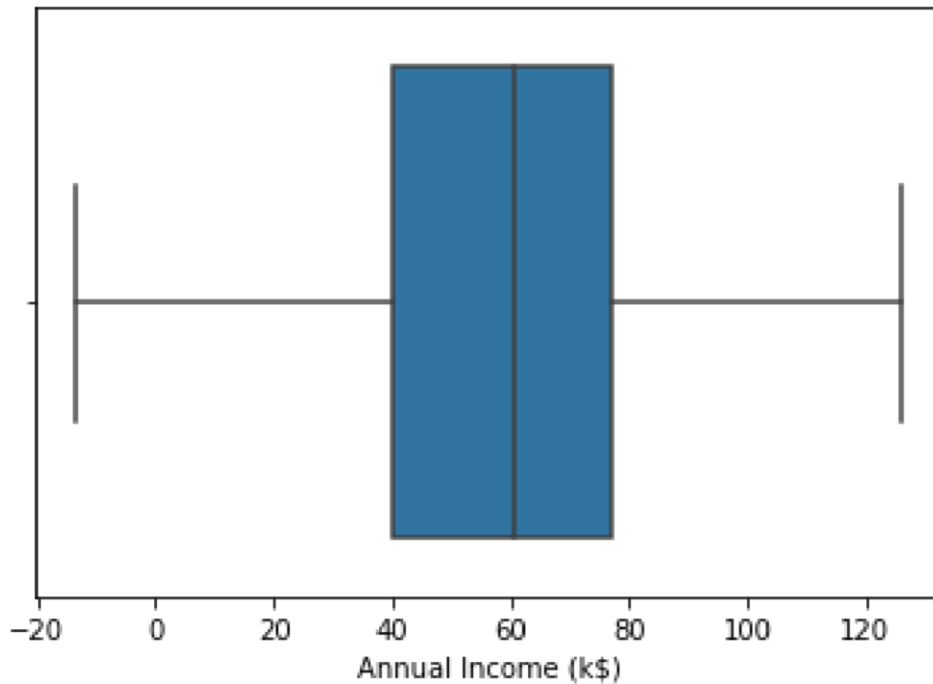
Gender 0.44 Age 38.85

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 ], [1. , 0.05769231,
0.20287253, 0.81632653],
[0. , 0.03846154, 0.21005386, 0.05102041], [0. , 0.09615385,
0.21005386, 0.7755102 ], [0. , 0.25 , 0.21723519, 0.39795918],
[0. , 0.07692308, 0.21723519, 0.76530612],
[0. , 0.32692308, 0.22441652, 0.05102041],
[0. , 0.09615385, 0.22441652, 0.94897959],
```

[1. , 0.88461538, 0.23159785, 0.02040816], [0. , 0.23076923, 0.23159785, 0.7244898], [1. , 0.94230769, 0.23159785, 0.13265306], [0. , 0.32692308, 0.23159785, 1.], [0. , 0.76923077, 0.23877917, 0.14285714], [0. , 0.11538462, 0.23877917, 0.7755102], [1. , 0.36538462, 0.23877917, 0.12244898],

[1. , 0.07692308, 0.23877917, 0.79591837], [0. , 0.32692308, 0.2459605 , 0.34693878],

[1. , 0.03846154, 0.2459605 , 0.66326531], [1. , 0.65384615, 0.26032316, 0.28571429],

[0. , , 0.32692308, 0.26032316, 0.98979592],

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[1. , 0.13461538, 0.26750449, 0.73469388],

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[1. , 0.21153846, 0.2962298 , 0.82653061],

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],

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[0. , 0.25 , 0.38240575, 0.41836735], [0. , 0.59615385, 0.3967684 ,
0.52040816],

[1. , 0.28846154, 0.3967684 , 0.60204082], [0. , 0.25 , 0.40394973,
0.54081633],

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0.45918367],

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, 0.55102041],

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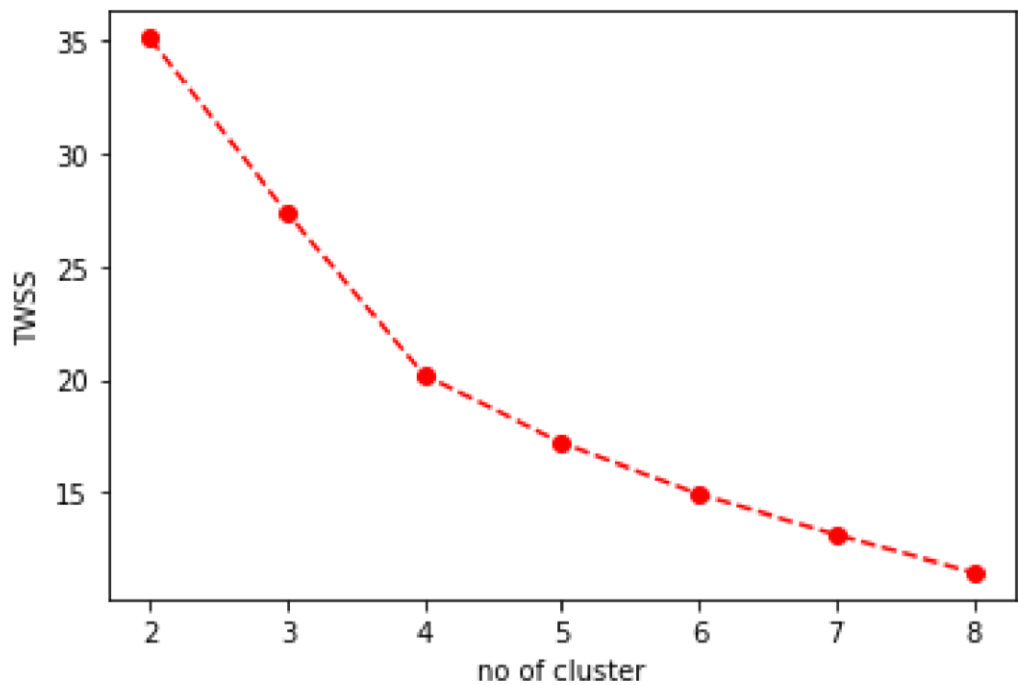
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```

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[1. , 0.28846154, 0.90664273, 0.07142857],
[0. , 0.38461538, 0.90664273, 0.91836735],
[0. , 0.55769231, 0.95691203, 0.15306122], [0. , 0.32692308, 0.95691203,
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0.74489796],
[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]])
```

