

# UNIVERSITY ADMIT ELIGIBILITY PREDICTOR

## ASSIGNMENT - 4

Date	4th October 2022
Team ID	PNT2022TMID27839
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Domain Name	Education
Project Name	University Admit Eligibility Predictor
Maximum Marks	2 Marks

### 1.)IMPORT THE REQUIRED LIBRARIES

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

### 2.)DOWNLOAD AND UPLOAD THE DATASET

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```
df = pd.read_csv('/Mall_Customers.csv')
df = df.drop(columns=["CustomerID"])
df.head()
```

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

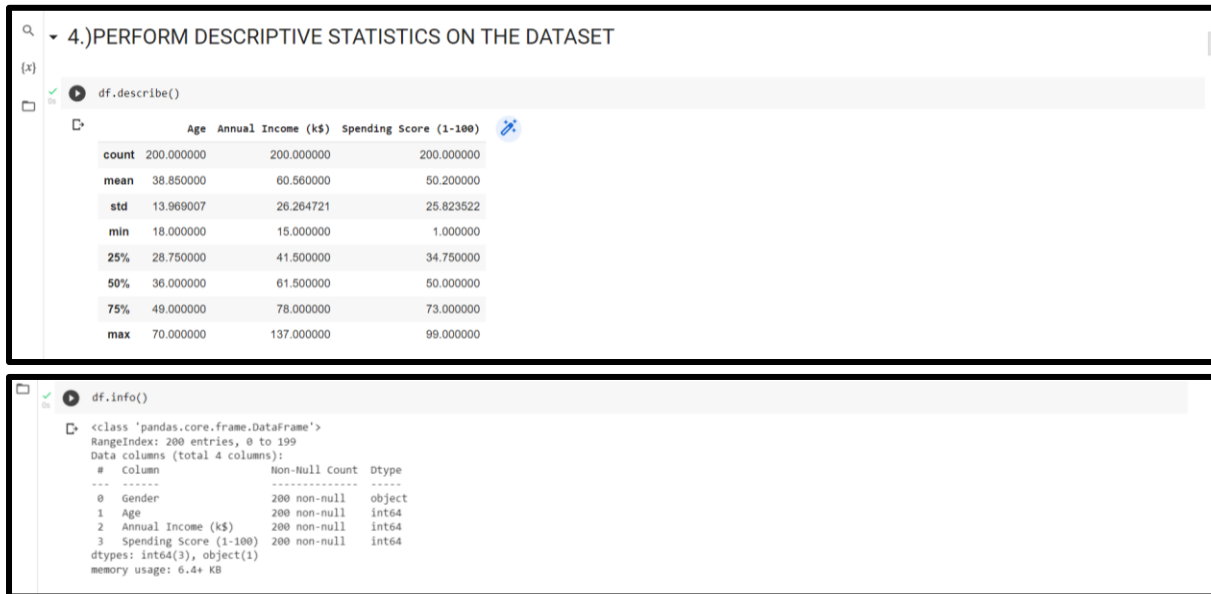
### 3.)HANDLE MISSING VALUES AND DEAL WITH THEM

#### 3.)CHECK FOR MISSING VALUES AND DEAL WITH THEM

```
[3] df.isnull().sum()
```

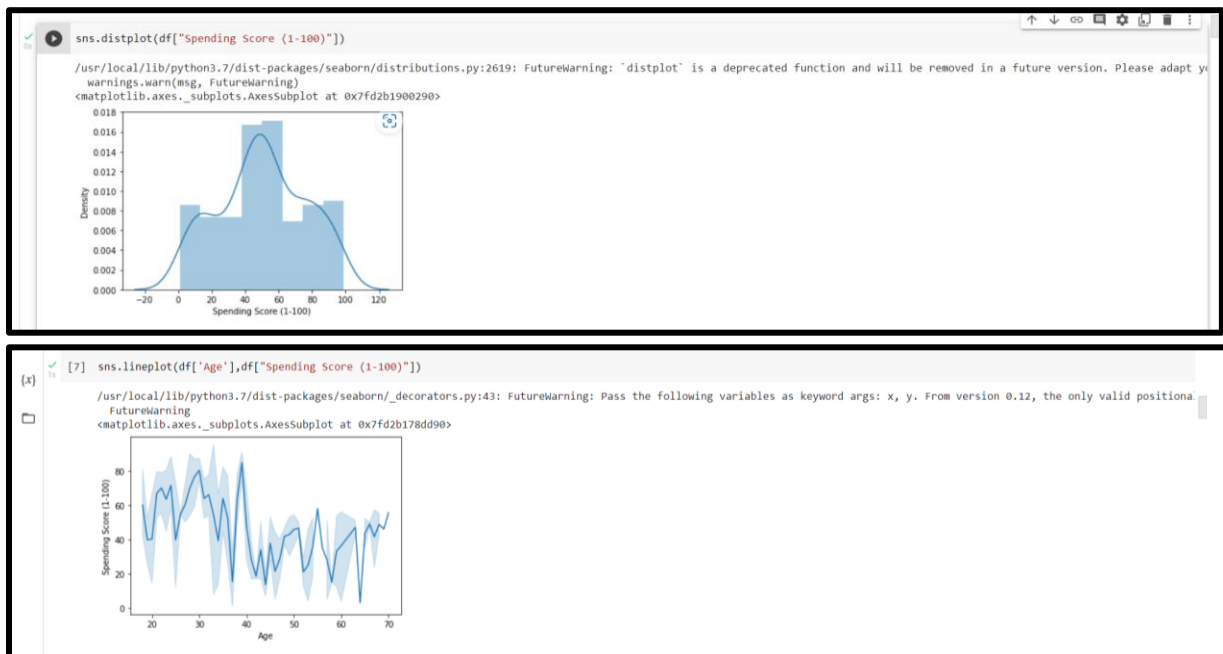
```
Gender      0
Age          0
Annual Income (k$)  0
Spending Score (1-100)  0
dtype: int64
```

## 4.) PERFORM THE DESCRIPTIVE STATISTICS ON THE DATASET

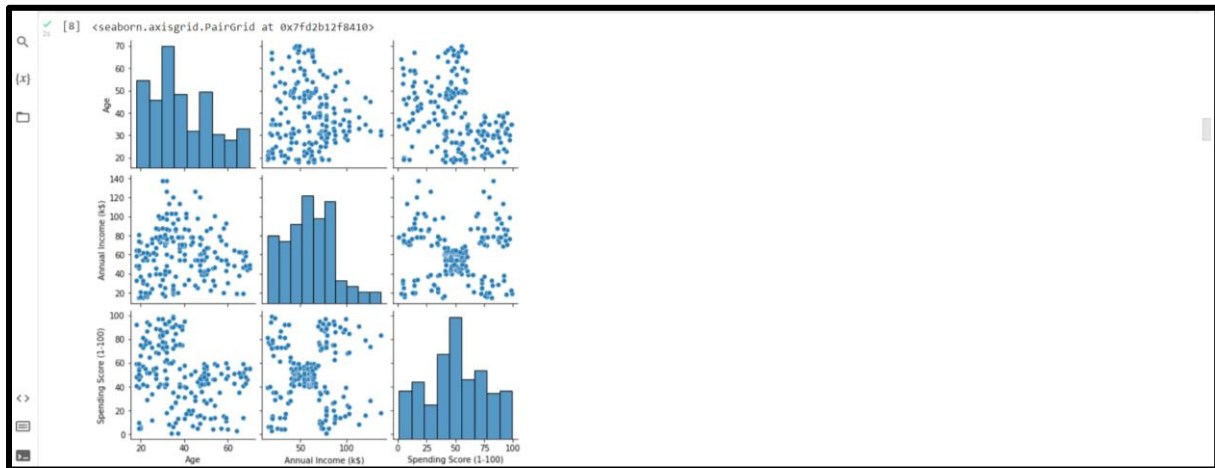


## 5.) PERFORM VARIOUS VISUALISATIONS

### a.) UNIVARIANTE ANALYSIS

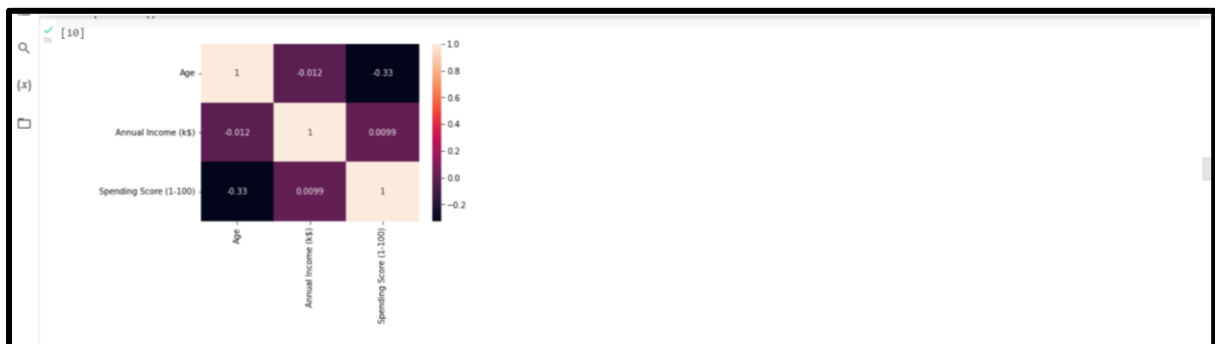


## b.) MULTI - VARIANTE ANALYSIS



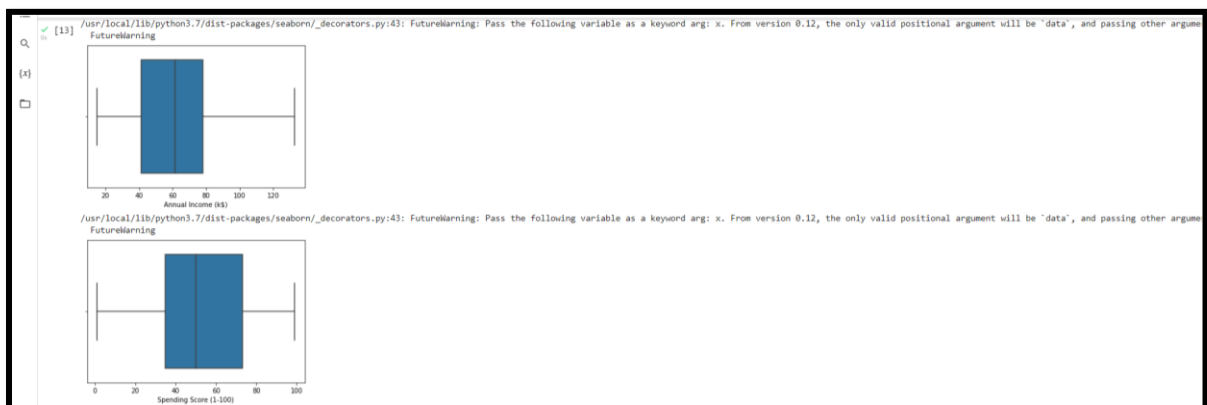
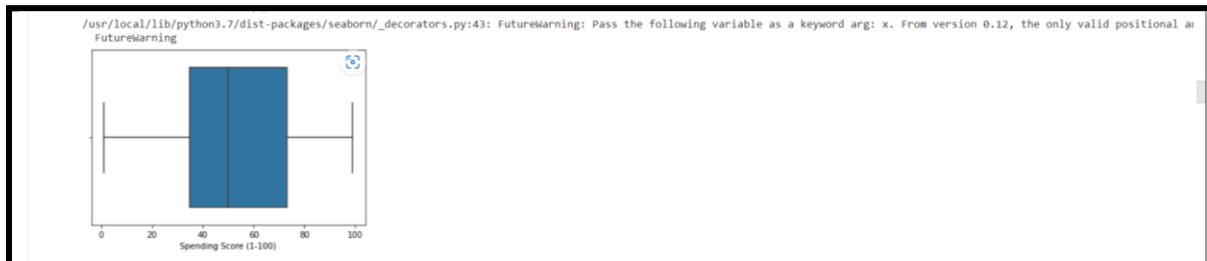
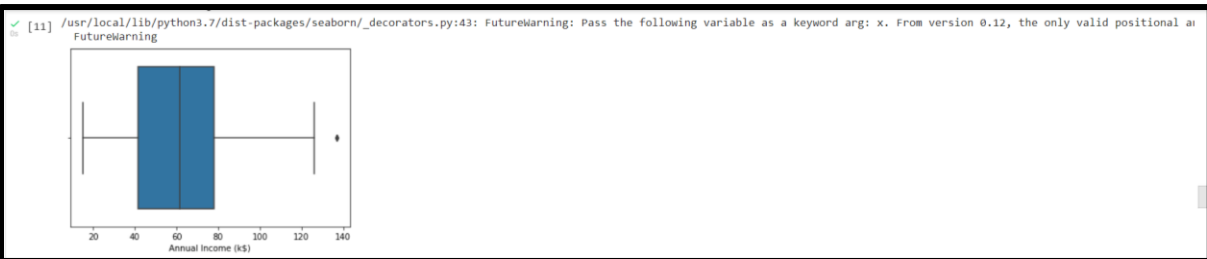
[9]

	Age	Annual Income (k\$)	Spending Score (1-100)
Age	1.000000	-0.012398	-0.327227
Annual Income (k\$)	-0.012398	1.000000	0.009903
Spending Score (1-100)	-0.327227	0.009903	1.000000



## 6.) FIND AND REPLACE THE OUTLIERS





## 7.) CHECK FOR CATEGORICAL COLUMNS AND ENCODE THEM

```
7.)CHECK FOR CATEGORICAL COLUMNS AND PERFORM ENCODING

[14] from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
df.Gender = le.fit_transform(df.Gender)

[15] df.head()
```

	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	19.0	15.0	39.0
1	1	21.0	15.0	81.0
2	0	20.0	16.0	6.0
3	0	23.0	16.0	77.0
4	0	31.0	17.0	40.0

## 8.) SCALING THE DATA

```
8.)SCALING THE DATA

[16] from sklearn.preprocessing import StandardScaler
scale = StandardScaler()
df = pd.DataFrame(scale.fit_transform(df), columns=df.columns)
df.head()
```

	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
0	1.128152	-1.424569	-1.745429	-0.434801
1	1.128152	-1.281035	-1.745429	1.195704
2	-0.886405	-1.352802	-1.707083	-1.715913
3	-0.886405	-1.137502	-1.707083	1.040418
4	-0.886405	-0.563369	-1.668737	-0.395990

## 9.) PERFORMING ANY OF THE CLUSTERING ALGORITHMS

```
[17] from sklearn.cluster import KMeans
error = []
for k in range(1,11):
    kmeans = KMeans(n_clusters=k,init='k-means++')
    kmeans.fit(df)
    error.append(kmeans.inertia_)
plt.plot(range(1,11),error)
plt.title('Elbow Method')
plt.xlabel('no of clusters')
plt.ylabel('error')
plt.grid()
plt.show()
```



The plot titled 'Elbow Method' shows the error (y-axis, ranging from 200 to 800) versus the number of clusters (x-axis, ranging from 1 to 10). The error starts high at 1 cluster (around 750) and decreases rapidly, reaching a minimum around 4 clusters (around 250). After 4 clusters, the error decreases much more slowly, indicating that 4 is a reasonable number of clusters.

```
[18] km = KMeans(n_clusters=8)
Category = km.fit_predict(df)
Category

array([0, 0, 4, 4, 4, 4, 2, 4, 3, 4, 3, 4, 2, 4, 0, 0, 4, 0, 3, 4, 0, 0,
       2, 0, 2, 0, 2, 0, 2, 4, 3, 4, 3, 0, 2, 4, 2, 4, 2, 4, 2, 0, 3, 4,
       2, 4, 2, 4, 4, 4, 2, 0, 4, 3, 2, 3, 2, 3, 4, 3, 3, 0, 2, 2, 3, 0,
       2, 0, 4, 4, 3, 2, 2, 3, 0, 2, 0, 4, 2, 3, 0, 3, 2, 4, 3, 4,
       4, 2, 2, 0, 3, 2, 4, 0, 2, 4, 3, 0, 4, 2, 3, 0, 3, 4, 2, 3, 3, 3,
       3, 4, 1, 0, 4, 4, 2, 2, 2, 2, 2, 0, 1, 5, 6, 1, 5, 7, 6, 3, 6, 7, 6,
       1, 5, 7, 5, 1, 6, 7, 5, 1, 6, 1, 5, 7, 6, 7, 5, 1, 6, 7, 6, 1, 5,
       1, 5, 7, 5, 7, 5, 1, 5, 7, 5, 7, 5, 7, 5, 1, 6, 7, 6, 7, 6, 1, 5,
       7, 6, 7, 6, 1, 5, 7, 5, 1, 6, 1, 6, 1, 5, 1, 5, 7, 5, 1, 5, 1, 6,
       7, 6], dtype=int32)
```

## 10.) ADD THE CLUSTER DATA WITH THE PRIMARY DATASET

```
10.)ADD THE CLUSTER DATA WITH THE PRIMARY DATASET

[19] df[["Category"]] = pd.Series(Category)
df.head()
```

	Gender	Age	Annual Income (k\$)	Spending Score (1-100)	Category	
0	1	128152	-1.424569	-1.745429	-0.434801	0
1	1	128152	-1.281035	-1.745429	1.195704	0
2	-0.886405	-1.352802	-1.707083	-1.715913		4
3	-0.886405	-1.137502	-1.707083	1.040418		4
4	-0.886405	-0.563369	-1.668737	-0.395980		4

## 11. )SPLITTING THE DATA INTO DEPENDENT AND INDEPENDENT VARIABLES

```
11.)SPLITTING THE DATA INTO DEPENDENT AND INDEPENDENT VARIABLES

X = df.drop(columns=["Category"])
Y = df.Category
```

## 12.)SPLIT THE DATA INTO TRAINING AND TESTING DATA

```
11.)SPLITTING THE DATA INTO DEPENDENT AND INDEPENDENT VARIABLES

[20] X = df.drop(columns=["Category"])
Y = df.Category
```

## 13.) BUILD THE MODEL

```
11.)BUILD THE MODEL

[22] from sklearn.ensemble import RandomForestClassifier
model = RandomForestClassifier()
```

## 14.) TRAIN THE MODEL

```
12.)TRAIN THE MODEL

[23] model.fit(x_train,y_train)

RandomForestClassifier()
```

## 15.) TEST THE MODEL

13.)TEST THE MODEL

+ Code

+ Text

```
[ ] y_predict = model.predict(x_test)

[ ] pd.DataFrame({"Actual":y_test,"Predicted":y_predict.round(0)})
```

	Actual	Predicted
18	3	3
170	7	7
107	3	3
98	3	3
177	6	6
182	7	7
5	4	4
148	7	7
12	2	2
152	1	1
61	0	0
125	5	5
180	1	1
154	1	1
80	3	3
7	4	4
33	0	0

## 16.) MEASURE THE PERFORMANCE USING METRICS

14.)MEASURE THE PERFORMANCE USING METRICS

```
[ ] from sklearn import metrics
    metrics.accuracy_score(y_test,y_predict)

0.975
```

```
sns.heatmap(metrics.confusion_matrix(y_test,y_predict),annot=True)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fd2abee7d50>
```

```
[ ] print(metrics.classification_report(y_test,y_predict))
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	2
1	1.00	1.00	1.00	3
2	0.83	1.00	0.91	5
3	1.00	1.00	1.00	7
4	1.00	1.00	1.00	7
5	1.00	0.80	0.89	5
6	1.00	1.00	1.00	6
7	1.00	1.00	1.00	5
accuracy			0.97	40
macro avg	0.98	0.97	0.97	40
weighted avg	0.98	0.97	0.97	40