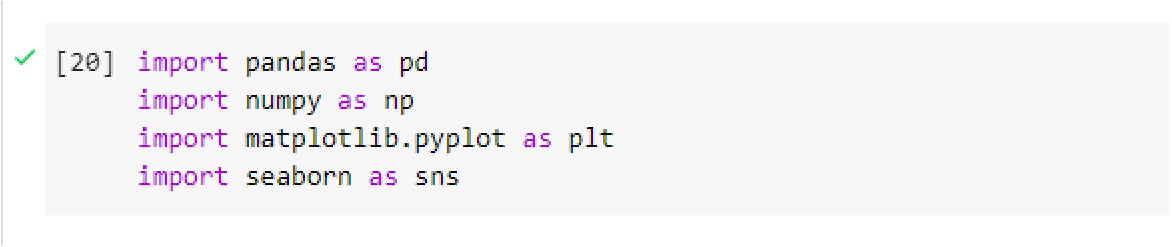
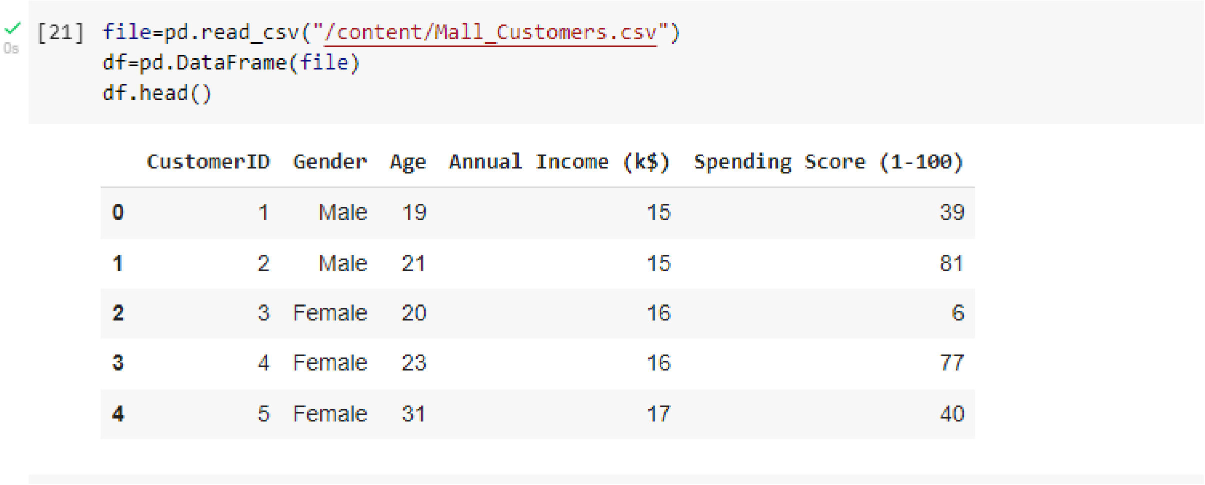
TEAM ID	PNT2022TMID17480
STUDENT NAME	Harini krishnaveni.M
DOMAIN NAME	Artificial intelligence
PROJECT NAME	Early detection of chronic kidney disease using machine learning
MAXIMUM MARKS	2 MARKS



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

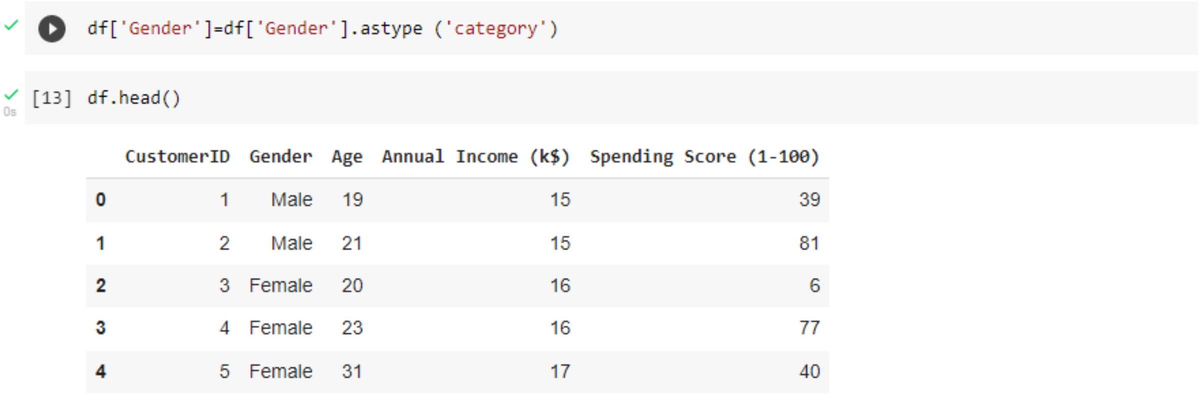
2. LOAD DATASET

2.load dataset



```
[21] file=pd.read_csv("/content/Mall_Customers.csv")
     df=pd.DataFrame(file)
     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



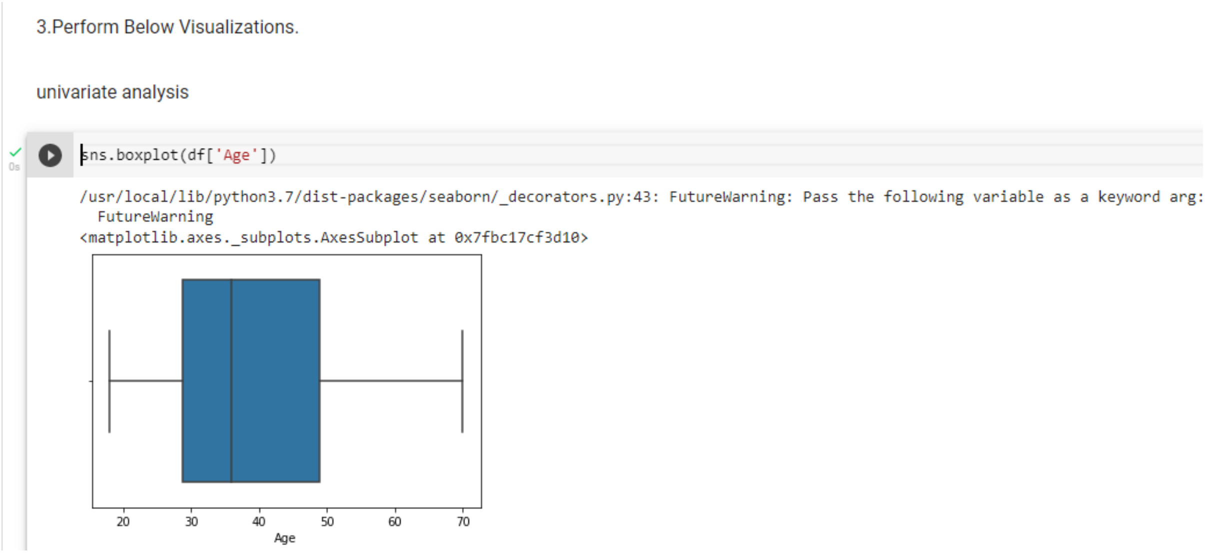
```
df['Gender']=df['Gender'].astype ('category')
```

```
[13] 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

3. PERFORM BELOW VISUALIZATIONS

· UNIVARIATE ANALYSIS

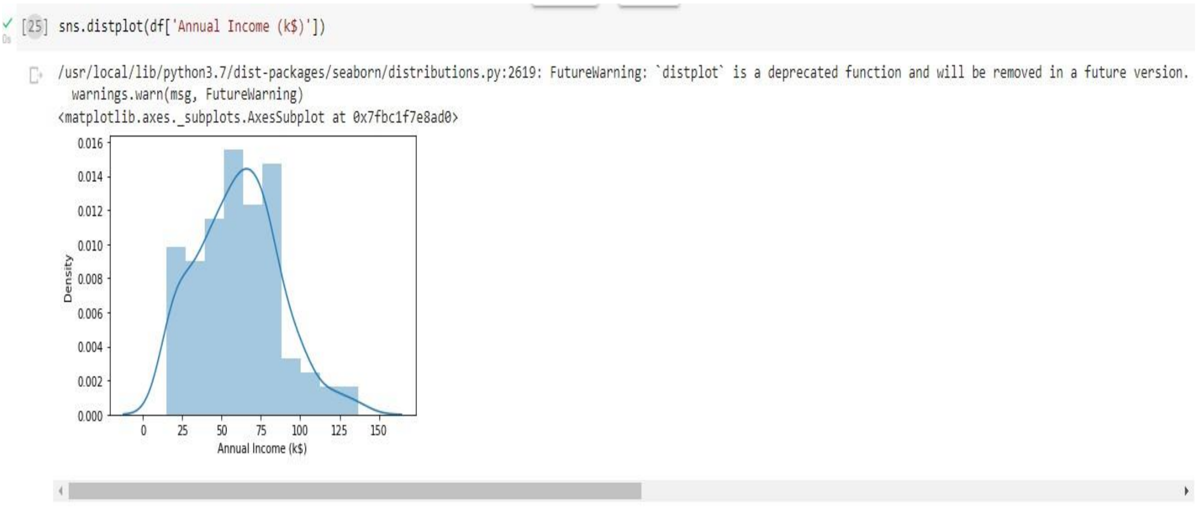


3.Perform Below Visualizations.

univariate analysis

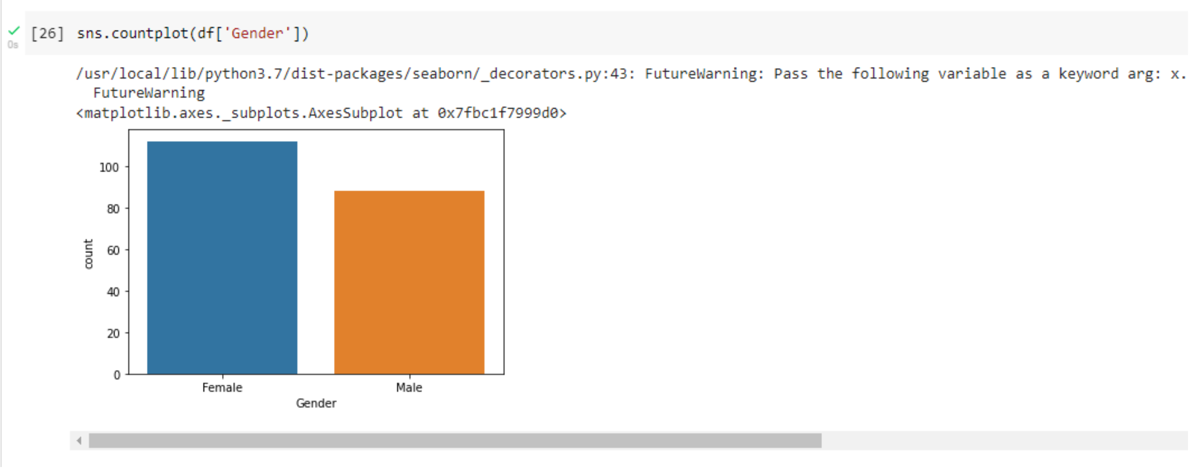
```
sns.boxplot(df['Age'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg:
  FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7fbc17cf3d10>
```



```
[25] sns.distplot(df['Annual Income (k$)'])
```

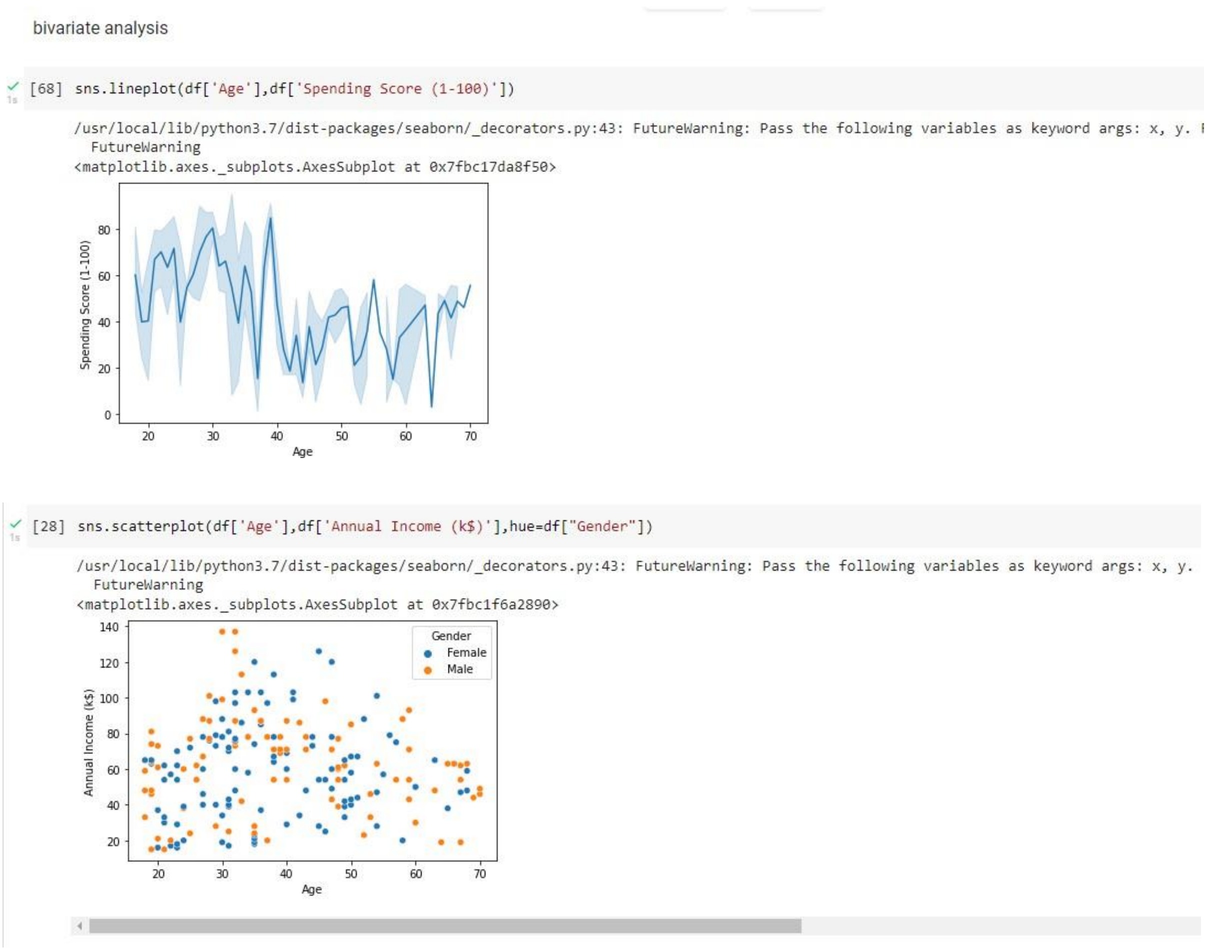
```
/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version.
  warnings.warn(msg, FutureWarning)
<matplotlib.axes._subplots.AxesSubplot at 0x7fbc1f7e8ad0>
```



```
[26] sns.countplot(df['Gender'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x.
  FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7fbc1f7999d0>
```

· BI- VARIATE ANALYSIS



bivariate analysis

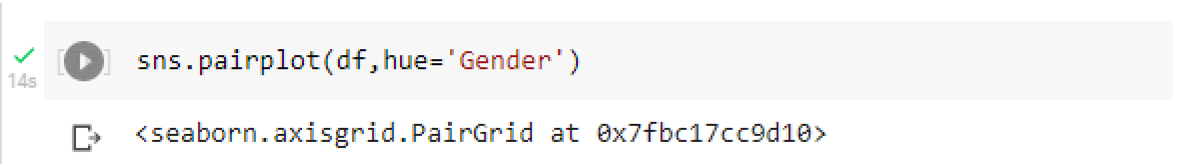
```
[68] sns.lineplot(df['Age'],df['Spending Score (1-100)'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y.
  FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7fbc17da8f50>
```

```
[28] sns.scatterplot(df['Age'],df['Annual Income (k$)'],hue=df["Gender"])
```

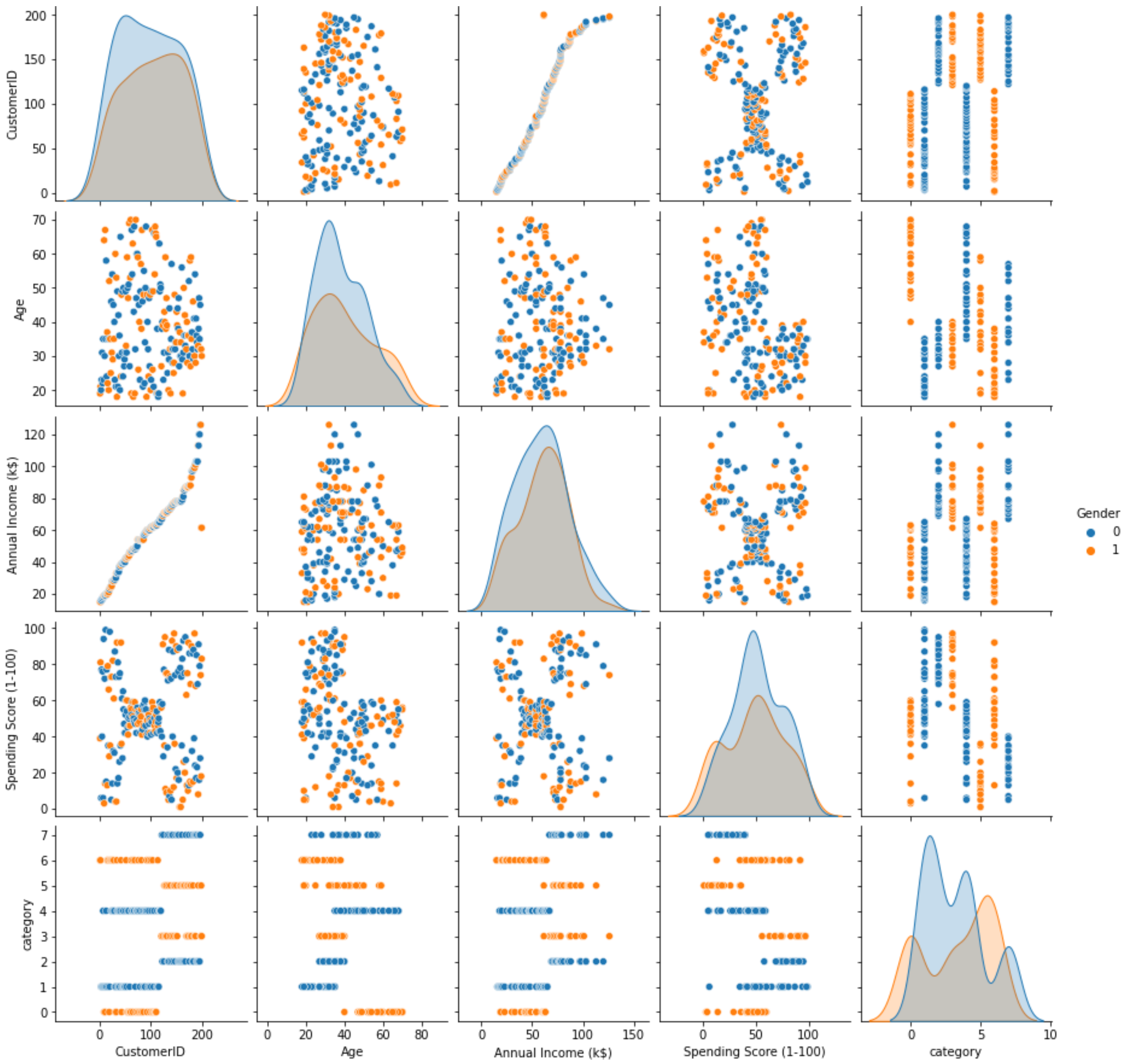
```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y.
  FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7fbc1f6a2890>
```

· MULTI-VARIATE ANALYSIS



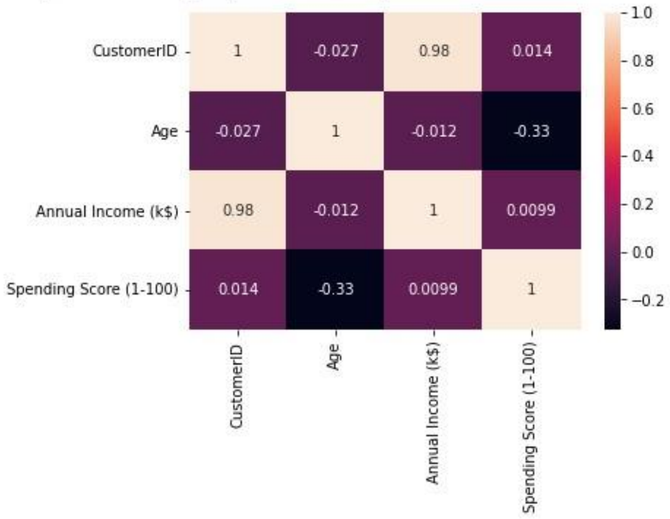
```
sns.pairplot(df,hue='Gender')
```

```
<seaborn.axisgrid.PairGrid at 0x7fbc17cc9d10>
```

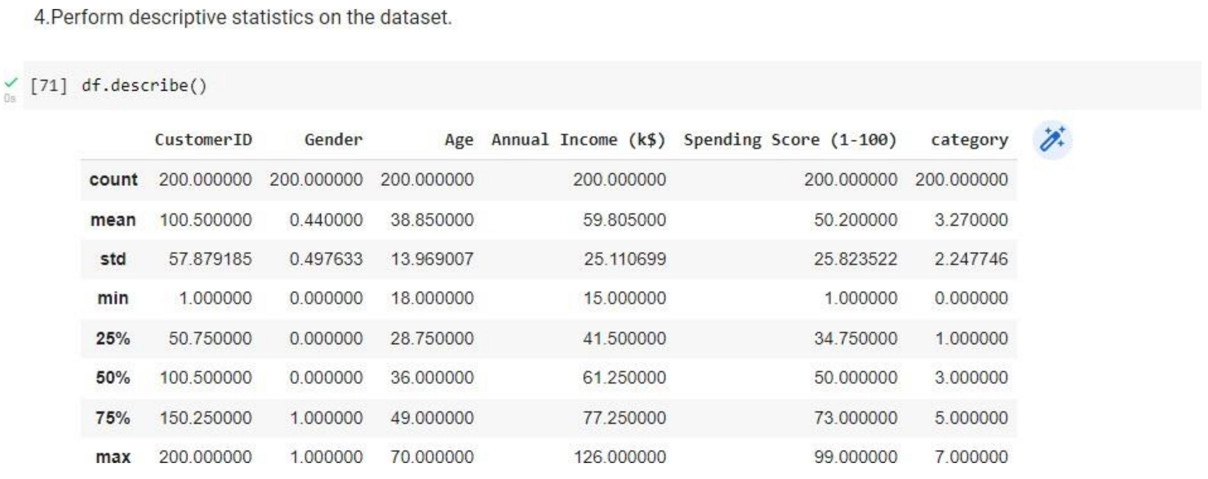


```
[30] sns.heatmap(df.corr(),annot=True)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fbc1f54ab50>
```



4. PERFORM DESCRIPTIVE STATISTICS ON THE DATASET

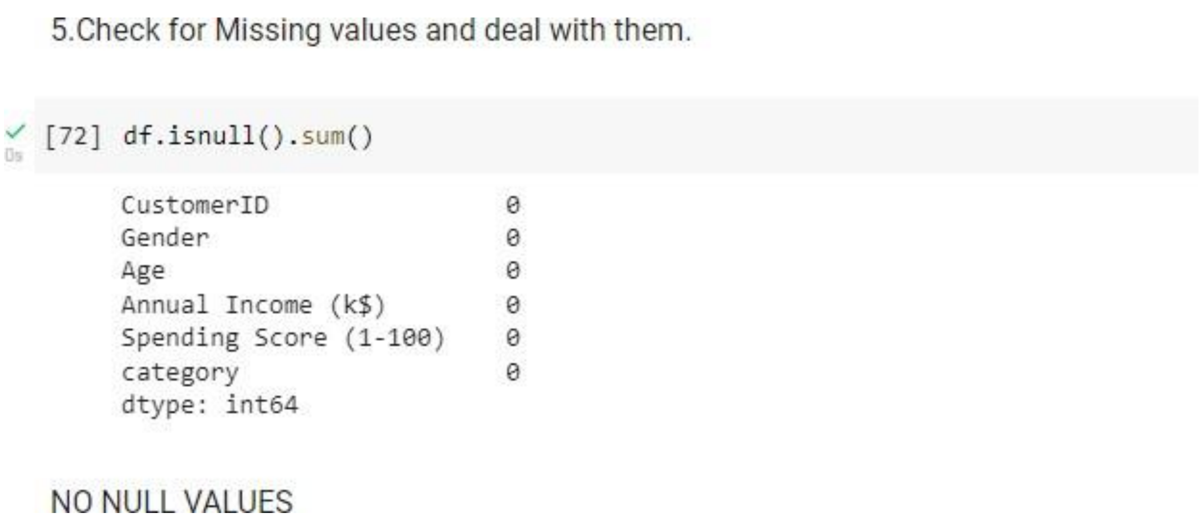


4.Perform descriptive statistics on the dataset.

```
[71] df.describe()
```

	CustomerID	Gender	Age	Annual Income (k$)	Spending Score (1-100)	category
count	200.000000	200.000000	200.000000	200.000000	200.000000	200.000000
mean	100.500000	0.440000	38.850000	59.805000	50.200000	3.270000
std	57.879185	0.497633	13.969007	25.110699	25.823522	2.247746
min	1.000000	0.000000	18.000000	15.000000	1.000000	0.000000
25%	50.750000	0.000000	28.750000	41.500000	34.750000	1.000000
50%	100.500000	0.000000	36.000000	61.250000	50.000000	3.000000
75%	150.250000	1.000000	49.000000	77.250000	73.000000	5.000000
max	200.000000	1.000000	70.000000	126.000000	99.000000	7.000000

5. CHECK FOR MISSING VALUES AND DEAL WITH THEM



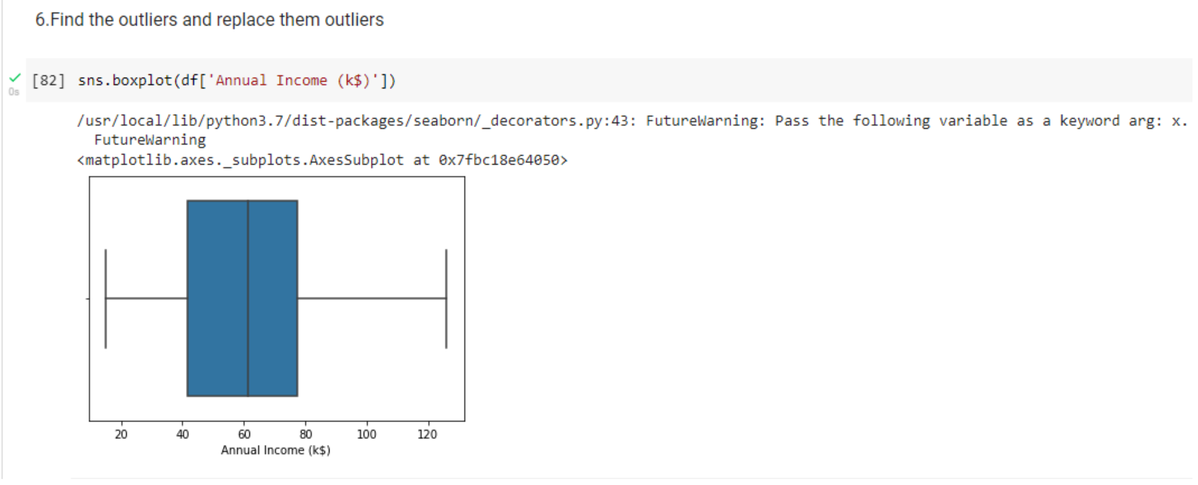
5.Check for Missing values and deal with them.

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

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

NO NULL VALUES

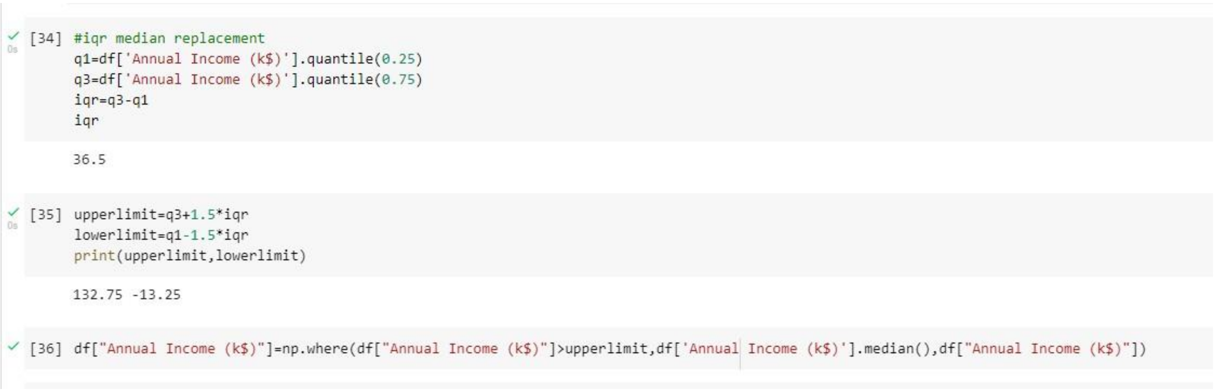
6. FIND THE OUTLIERS AND REPLACE THEM OUTLIERS



6.Find the outliers and replace them outliers

```
[82] sns.boxplot(df['Annual Income (k$)'])
```

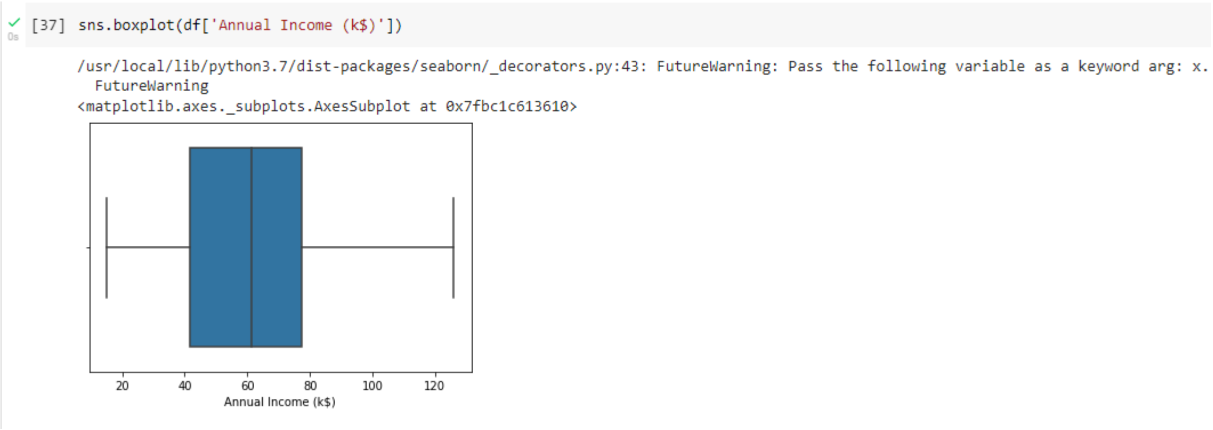
```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x.
  FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7fbc18e64050>
```



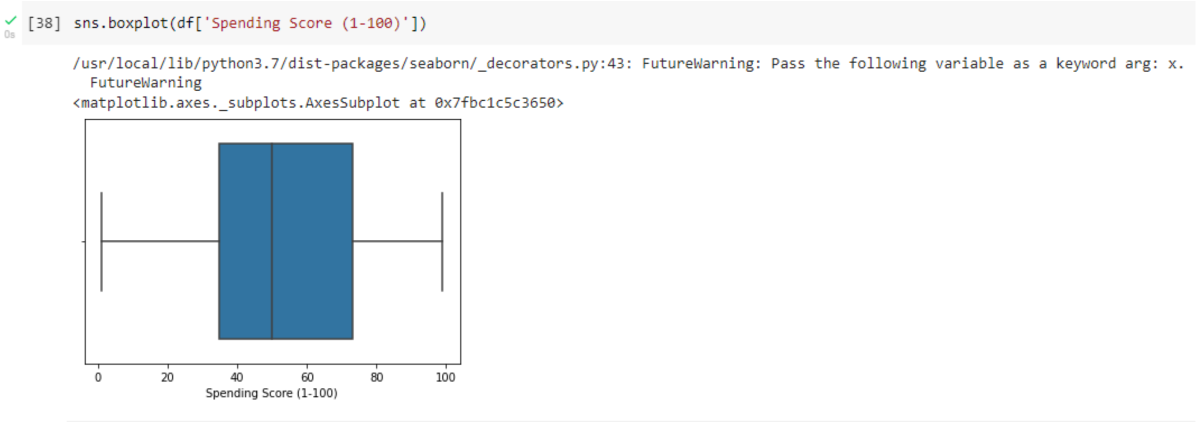
```
[34] #iqr median replacement
     q1=df['Annual Income (k$)'].quantile(0.25)
     q3=df['Annual Income (k$)'].quantile(0.75)
     iqr=q3-q1
     iqr

     36.5
```

```
[35] upperlimit=q3+1.5*iqr
     lowerlimit=q1-1.5*iqr
     print(upperlimit,lowerlimit)

     132.75 -13.25
```

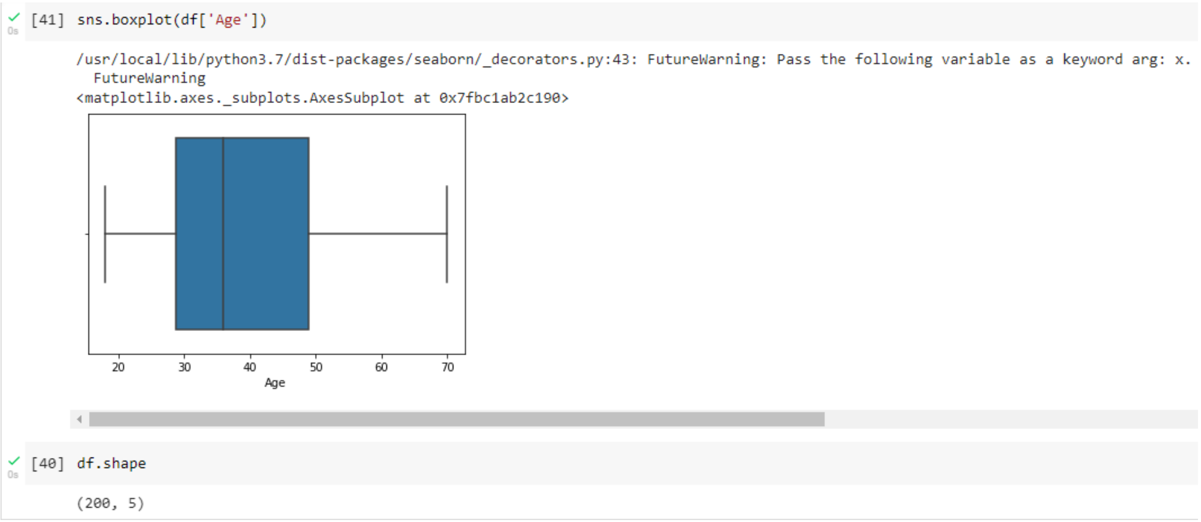
```
[36] df["Annual Income (k$)"]=np.where(df["Annual Income (k$)"]>upperlimit,df['Annual Income (k$)'].median(),df["Annual Income (k$)"])
```



```
[37] sns.boxplot(df['Annual Income (k$)'])

     /usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x.
       FutureWarning
     <matplotlib.axes._subplots.AxesSubplot at 0x7fbc1c613610>
```



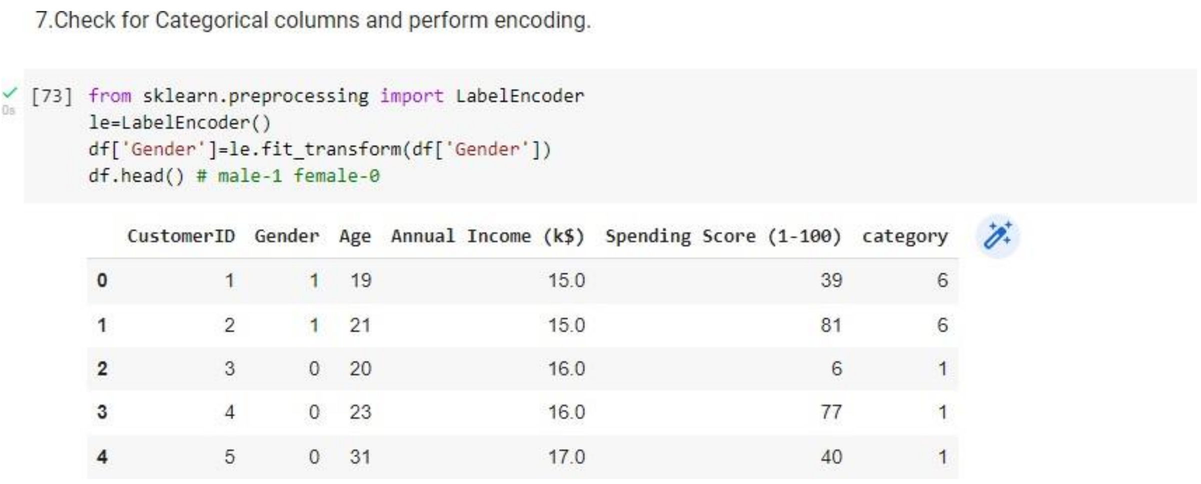
```
[38] sns.boxplot(df['Spending Score (1-100)'])

     /usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x.
       FutureWarning
     <matplotlib.axes._subplots.AxesSubplot at 0x7fbc1c5c3650>
```



```
[41] sns.boxplot(df['Age'])

     /usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x.
       FutureWarning
     <matplotlib.axes._subplots.AxesSubplot at 0x7fbc1ab2c190>
```

```
[40] df.shape

     (200, 5)
```

7. CHECK FOR CATEGORICAL COLUMNS AND PERFORM ENCODING

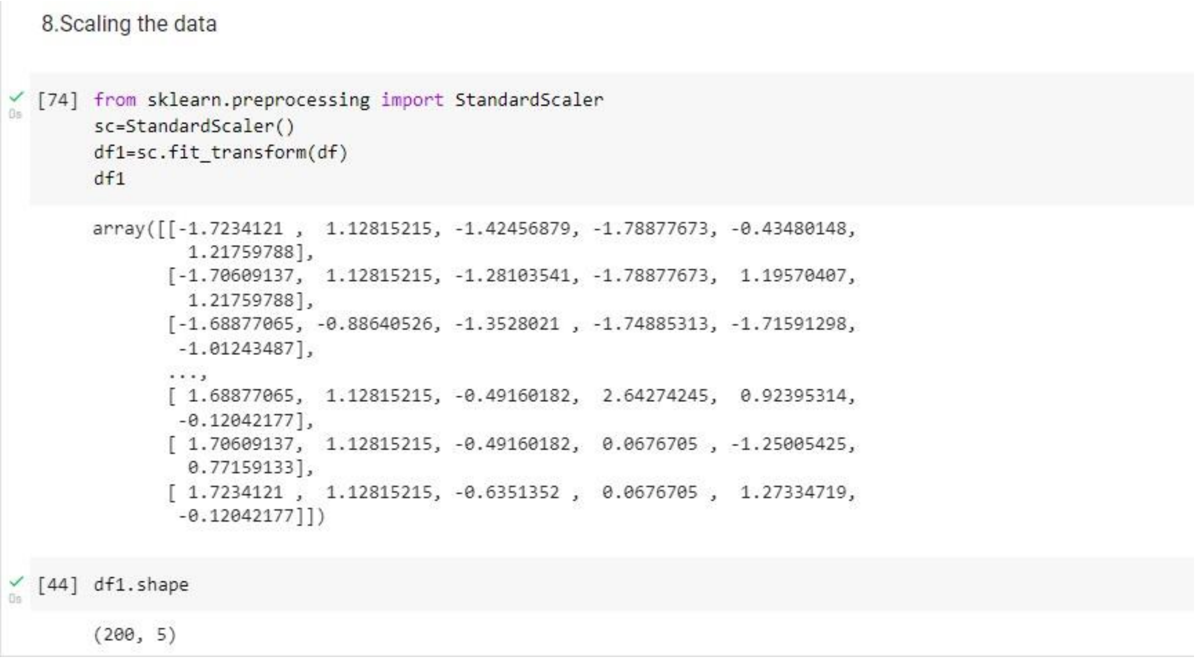


7.Check for Categorical columns and perform encoding.

```
[73] from sklearn.preprocessing import LabelEncoder
     le=LabelEncoder()
     df['Gender']=le.fit_transform(df['Gender'])
     df.head() # male-1 female-0
```

	CustomerID	Gender	Age	Annual Income (k$)	Spending Score (1-100)	category
0	1	1	19	15.0	39	6
1	2	1	21	15.0	81	6
2	3	0	20	16.0	6	1
3	4	0	23	16.0	77	1
4	5	0	31	17.0	40	1

8. SCALING THE DATA



8.Scaling the data

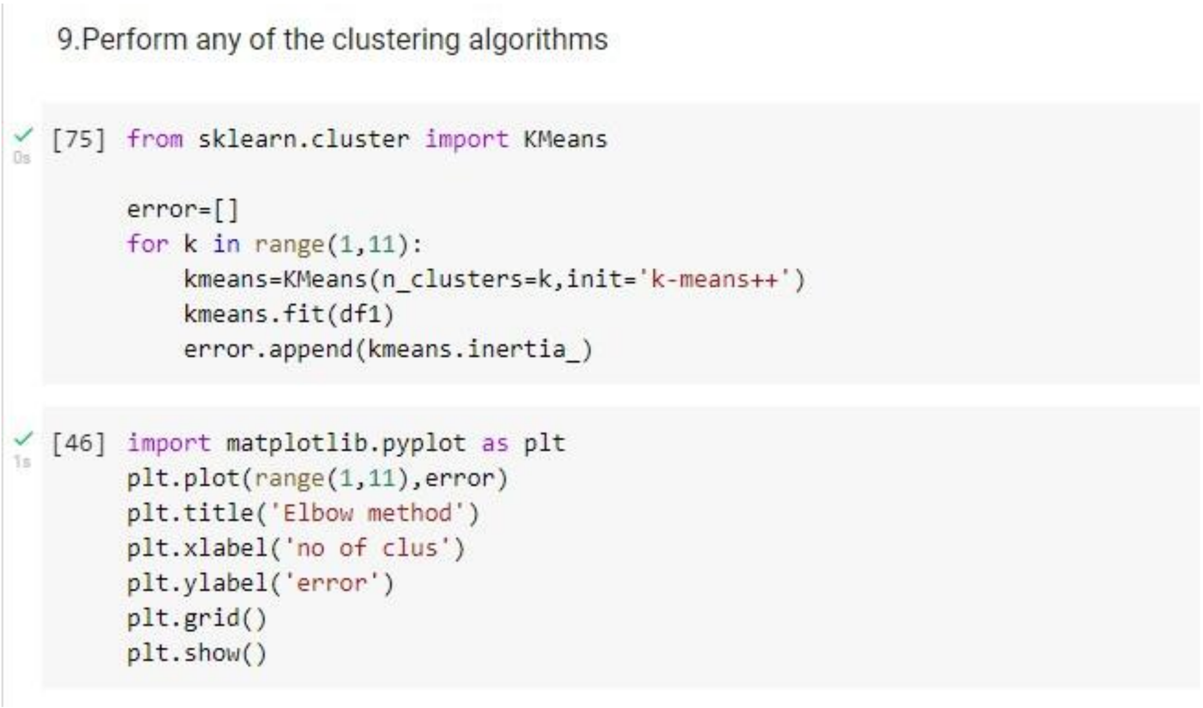
```
[74] from sklearn.preprocessing import StandardScaler
     sc=StandardScaler()
     df1=sc.fit_transform(df)
     df1
```

```
array([[-1.7234121 ,  1.12815215, -1.42456879, -1.78877673, -0.43480148,
         1.21759788],
       [-1.70609137,  1.12815215, -1.28103541, -1.78877673,  1.19570407,
         1.21759788],
       [-1.68877065, -0.88640526, -1.3528021 , -1.74885313, -1.71591298,
        -1.01243487],
       ...,
       [ 1.68877065,  1.12815215, -0.49160182,  2.64274245,  0.92395314,
        -0.12042177],
       [ 1.70609137,  1.12815215, -0.49160182,  0.0676705 , -1.25005425,
         0.77159133],
       [ 1.7234121 ,  1.12815215, -0.6351352 ,  0.0676705 ,  1.27334719,
        -0.12042177]])
```

```
[44] df1.shape
```

```
(200, 5)
```

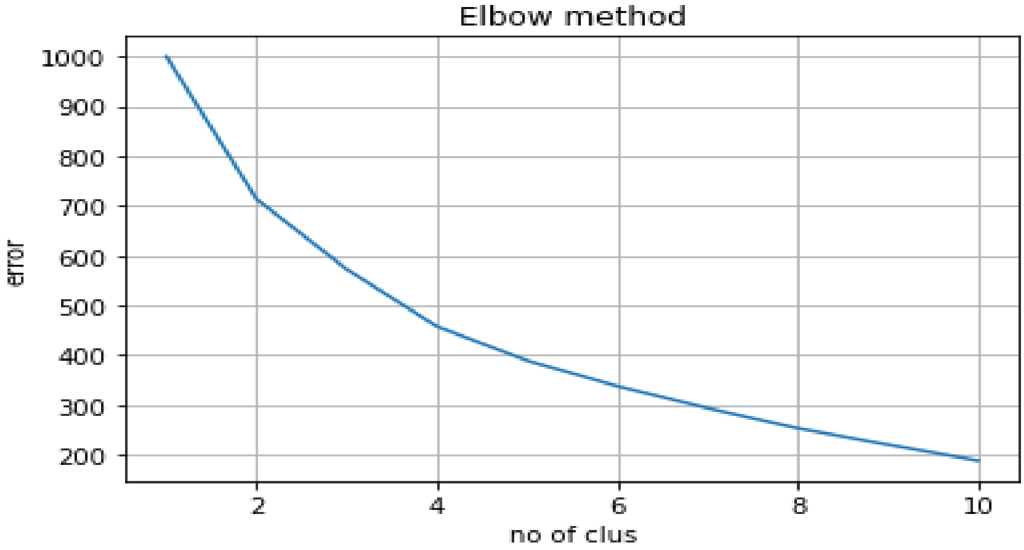
9. PERFORM ANY OF THE CLUSTERING ALGORITHMS

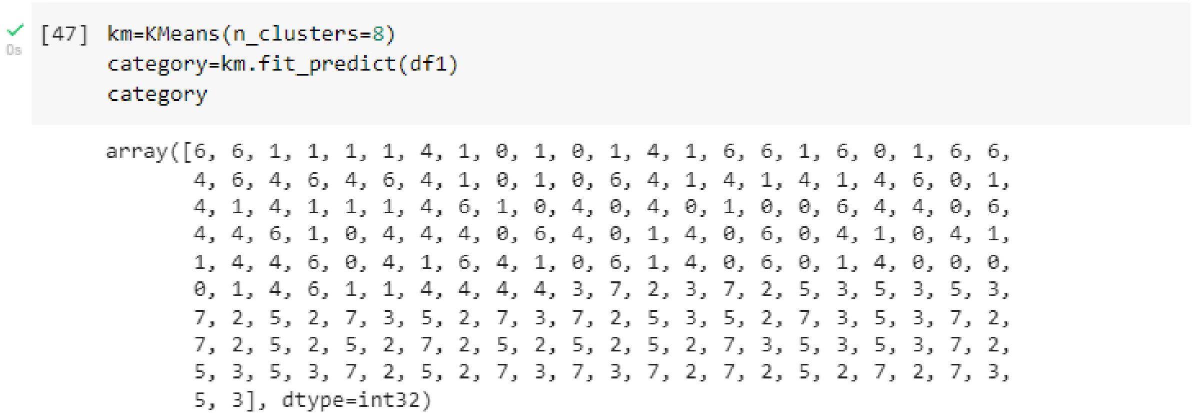


9.Perform any of the clustering algorithms

```
[75] from sklearn.cluster import KMeans

     error=[]
     for k in range(1,11):
         kmeans=KMeans(n_clusters=k,init='k-means++')
         kmeans.fit(df1)
         error.append(kmeans.inertia_)
```

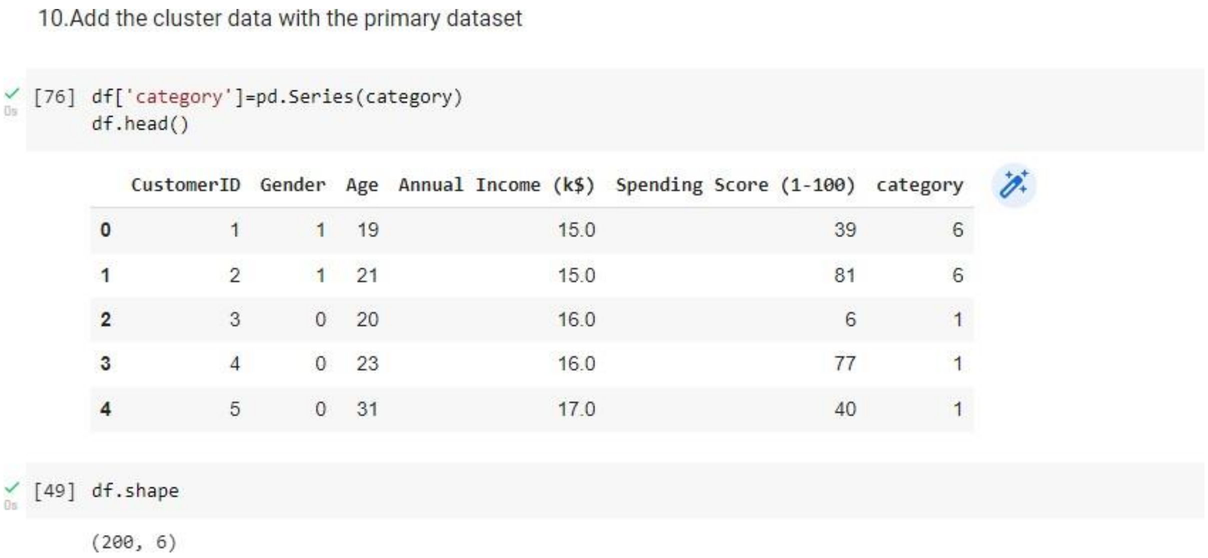
```
[46] import matplotlib.pyplot as plt
     plt.plot(range(1,11),error)
     plt.title('Elbow method')
     plt.xlabel('no of clus')
     plt.ylabel('error')
     plt.grid()
     plt.show()
```





```
[47] km=KMeans(n_clusters=8)
     category=km.fit_predict(df1)
     category

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

10. ADD THE CLUSTER DATA WITH THE PRIMARY DATASET



10.Add the cluster data with the primary dataset

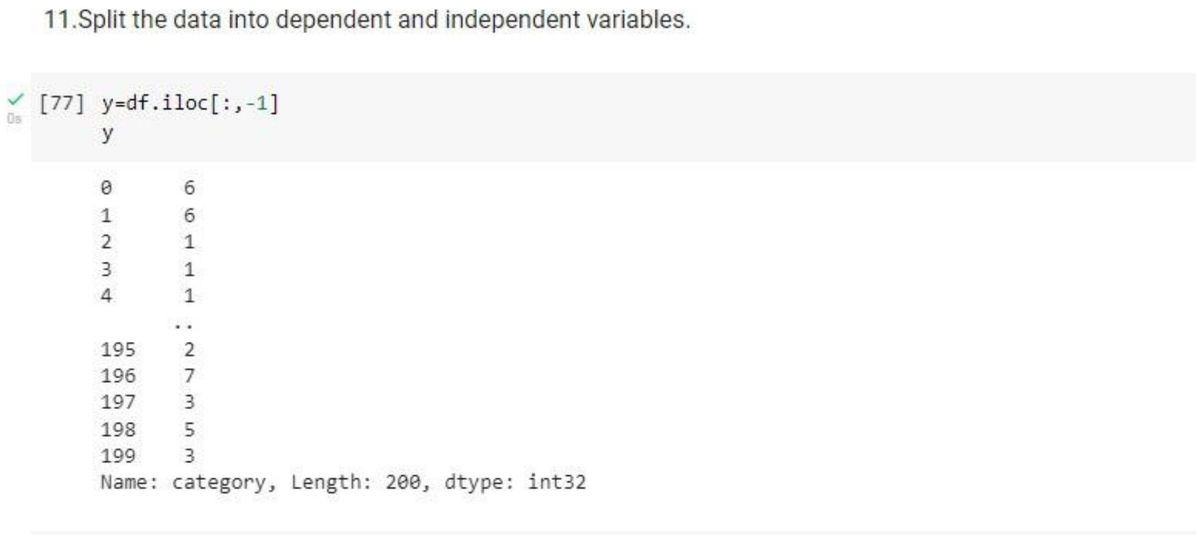
```
[76] df['category']=pd.Series(category)
     df.head()
```

	CustomerID	Gender	Age	Annual Income (k$)	Spending Score (1-100)	category
0	1	1	19	15.0	39	6
1	2	1	21	15.0	81	6
2	3	0	20	16.0	6	1
3	4	0	23	16.0	77	1
4	5	0	31	17.0	40	1

```
[49] df.shape
```

```
(200, 6)
```

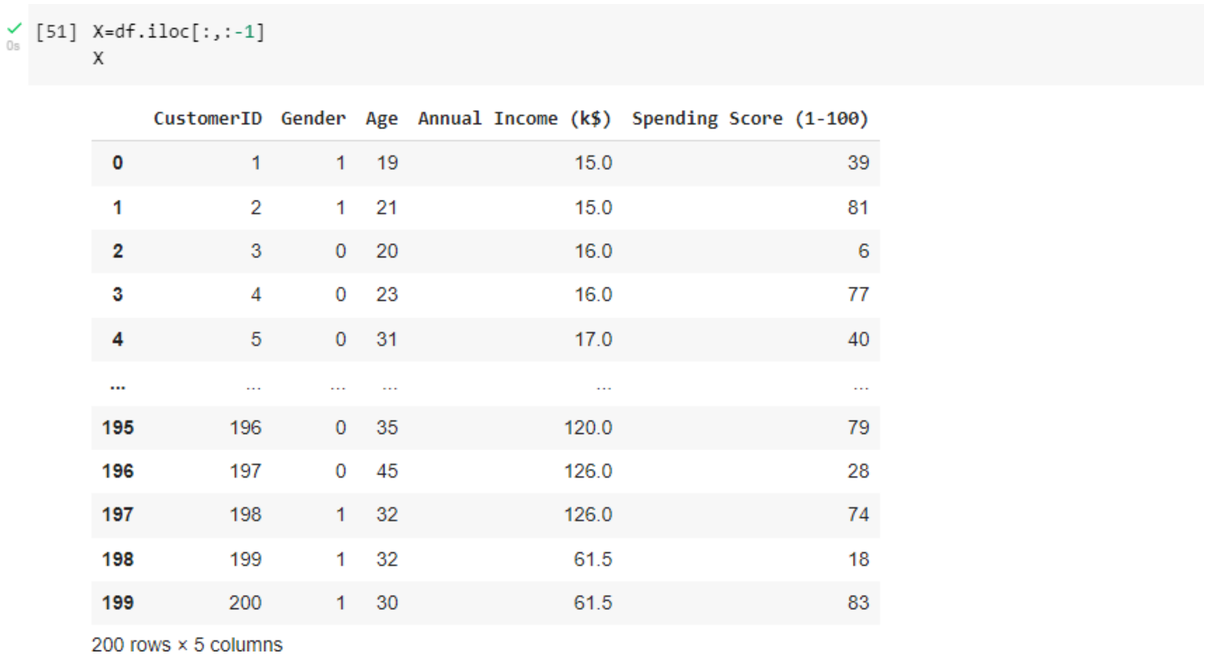
11. SPLIT THE DATA INTO DEPENDENT AND INDEPENDENT VARIABLES.



11.Split the data into dependent and independent variables.

```
[77] y=df.iloc[:,-1]
     y
```

```
0      6
1      6
2      1
3      1
4      1
      ..
195    2
196    7
197    3
198    5
199    3
Name: category, Length: 200, dtype: int32
```

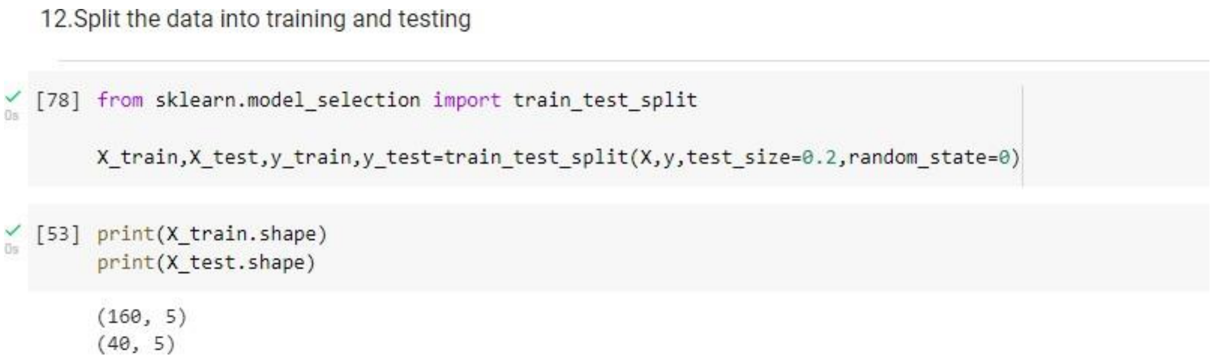


```
[51] X=df.iloc[:,:-1]
     X
```

	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
...	...	...	...	...	...
195	196	0	35	120.0	79
196	197	0	45	126.0	28
197	198	1	32	126.0	74
198	199	1	32	61.5	18
199	200	1	30	61.5	83

200 rows × 5 columns

12. SPLIT THE DATA INTO TRAINING AND TESTING

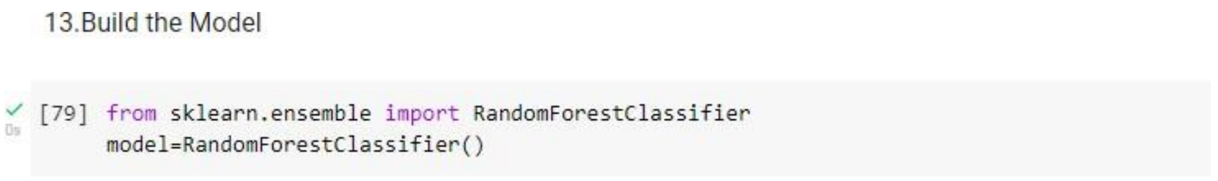


12.Split the data into training and testing

```
[78] 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,random_state=0)
```

```
[53] print(X_train.shape)
     print(X_test.shape)

     (160, 5)
     (40, 5)
```

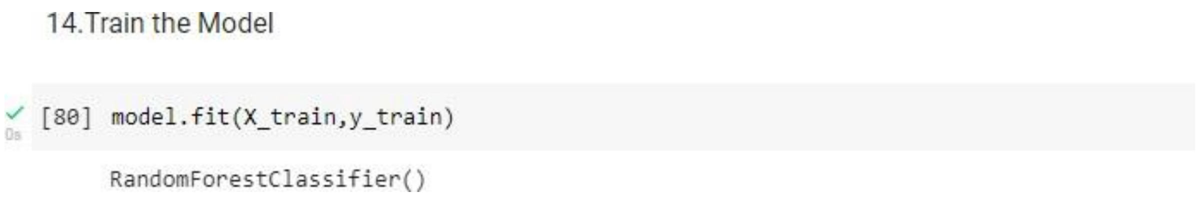
13. BUILD THE MODEL



13.Build the Model

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

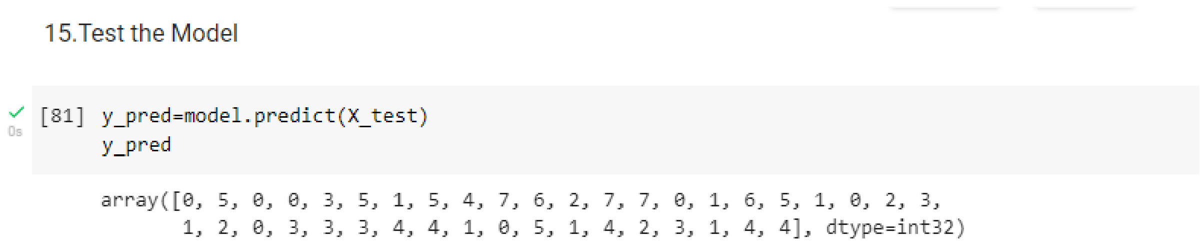
14. TRAIN THE MODEL



14.Train the Model

```
[80] model.fit(X_train,y_train)

     RandomForestClassifier()
```

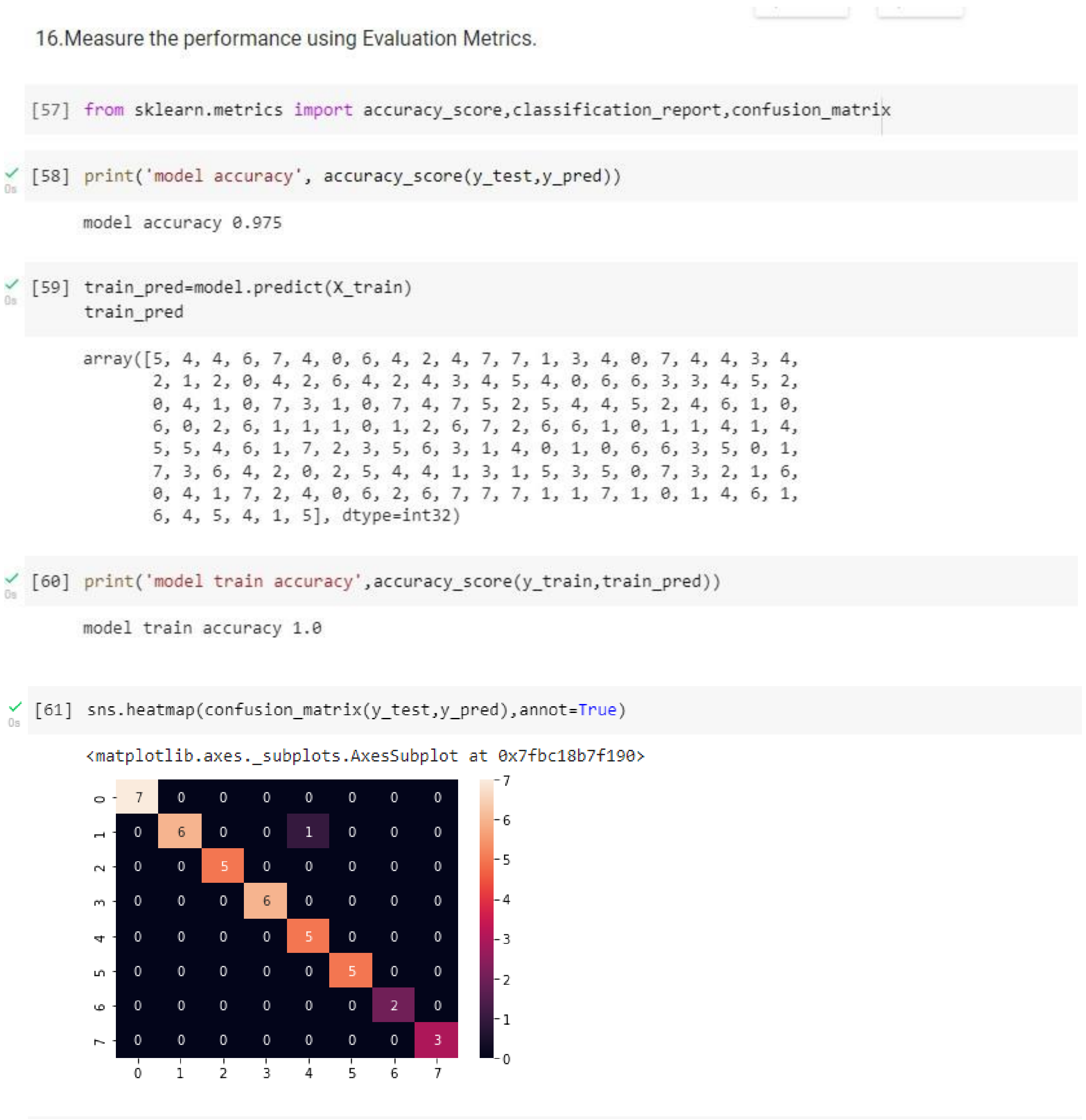
15. TEST THE MODEL



15.Test the Model

```
[81] y_pred=model.predict(X_test)
     y_pred

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

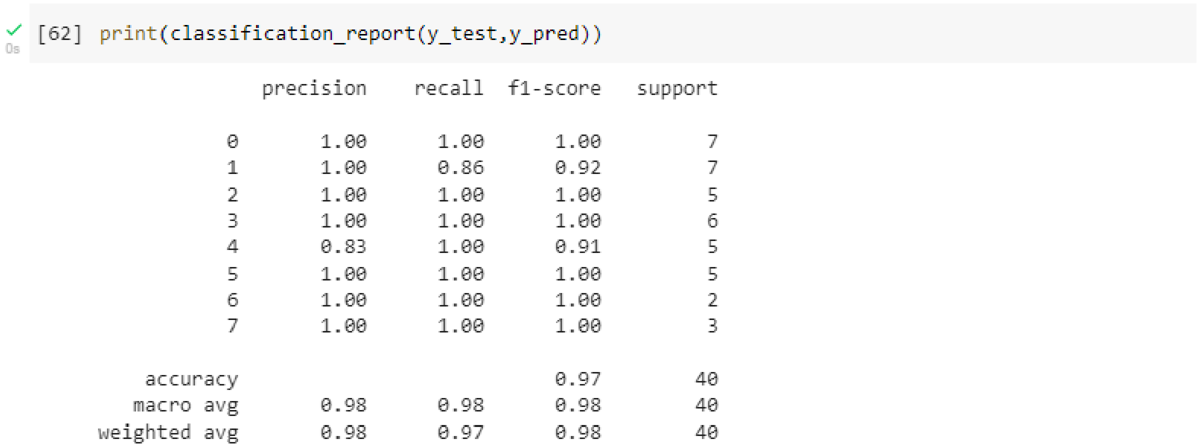
16. MEASURE THE PERFORMANCE USING EVALUATION METRICS.



16.Measure the performance using Evaluation Metrics.

```
[57] from sklearn.metrics import accuracy_score,classification_report,confusion_matrix

[58] print('model accuracy', accuracy_score(y_test,y_pred))

     model accuracy 0.975

[59] train_pred=model.predict(X_train)
     train_pred

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

[60] print('model train accuracy',accuracy_score(y_train,train_pred))

     model train accuracy 1.0

[61] sns.heatmap(confusion_matrix(y_test,y_pred),annot=True)

     <matplotlib.axes._subplots.AxesSubplot at 0x7fbc18b7f190>
```

	0	1	2	3	4	5	6	7
0	7	0	0	0	0	0	0	0
1	0	6	0	0	1	0	0	0
2	0	0	5	0	0	0	0	0
3	0	0	0	6	0	0	0	0
4	0	0	0	0	5	0	0	0
5	0	0	0	0	0	5	0	0
6	0	0	0	0	0	0	2	0
7	0	0	0	0	0	0	0	3



```
[62] print(classification_report(y_test,y_pred))
```

```
              precision    recall  f1-score   support

           0       1.00      1.00      1.00         7
           1       1.00      0.86      0.92         7
           2       1.00      1.00      1.00         5
           3       1.00      1.00      1.00         6
           4       0.83      1.00      0.91         5
           5       1.00      1.00      1.00         5
           6       1.00      1.00      1.00         2
           7       1.00      1.00      1.00         3

    accuracy                           0.97        40
   macro avg       0.98      0.98      0.98        40
weighted avg       0.98      0.97      0.98        40
```