

NAME: K S H V SAI HARI KRISHNA
REG NO: 21BCE8069

- 1.Download the Employee Attrition Dataset
- <https://www.kaggle.com/datasets/patelprashant/employee-attrition>
- 2.Perfrom Data Preprocessing
- 3.Model Building using Logistic Regression and Decision Tree and Random Forest
- 4.Calculate Performance metrics

```
#Import the Libraries.  
import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
import seaborn as sns  
  
#Importing the dataset.  
df=pd.read_csv("Employee-Attrition.csv")  
  
df.head()
```

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	1
1	49	No	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	2
2	37	Yes	Travel_Rarely	1373	Research & Development	2	2	Other	1	3
3	33	No	Travel_Frequently	1392	Research & Development	3	4	Life Sciences	1	4
4	27	No	Travel_Rarely	591	Research & Development	2	1	Medical	1	5

5 rows × 35 columns

```
df.shape  
  
(1470, 35)
```

```
df.Age.value_counts()  
  
35    78  
34    77  
36    69  
31    69  
29    68  
32    61  
30    60  
33    58  
38    58  
40    57  
37    50  
27    48  
28    48  
42    46  
39    42  
45    41  
41    40  
26    39  
44    33  
46    33  
43    32  
50    30  
25    26  
24    26  
49    24  
47    24  
55    22  
51    19  
53    19  
48    19  
54    18  
52    18
```

```
22 16
56 14
23 14
58 14
21 13
20 11
59 10
19 9
18 8
60 5
57 4
Name: Age, dtype: int64
```

```
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1470 entries, 0 to 1469
Data columns (total 35 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Age                                    1470 non-null   int64
1   Attrition                            1470 non-null   object
2   BusinessTravel                       1470 non-null   object
3   DailyRate                            1470 non-null   int64
4   Department                           1470 non-null   object
5   DistanceFromHome                    1470 non-null   int64
6   Education                            1470 non-null   int64
7   EducationField                       1470 non-null   object
8   EmployeeCount                        1470 non-null   int64
9   EmployeeNumber                       1470 non-null   int64
10  EnvironmentSatisfaction               1470 non-null   int64
11  Gender                               1470 non-null   object
12  HourlyRate                           1470 non-null   int64
13  JobInvolvement                       1470 non-null   int64
14  JobLevel                             1470 non-null   int64
15  JobRole                               1470 non-null   object
16  JobSatisfaction                       1470 non-null   int64
17  MaritalStatus                        1470 non-null   object
18  MonthlyIncome                        1470 non-null   int64
19  MonthlyRate                           1470 non-null   int64
20  NumCompaniesWorked                   1470 non-null   int64
21  Over18                               1470 non-null   object
22  OverTime                             1470 non-null   object
23  PercentSalaryHike                    1470 non-null   int64
24  PerformanceRating                    1470 non-null   int64
25  RelationshipSatisfaction              1470 non-null   int64
26  StandardHours                        1470 non-null   int64
27  StockOptionLevel                     1470 non-null   int64
28  TotalWorkingYears                    1470 non-null   int64
29  TrainingTimesLastYear                1470 non-null   int64
30  WorkLifeBalance                      1470 non-null   int64
31  YearsAtCompany                       1470 non-null   int64
32  YearsInCurrentRole                   1470 non-null   int64
33  YearsSinceLastPromotion               1470 non-null   int64
34  YearsWithCurrManager                 1470 non-null   int64
dtypes: int64(26), object(9)
memory usage: 402.1+ KB
```

```
df.describe()

      Age  DailyRate  DistanceFromHome  Education  EmployeeCount  EmployeeNumber  EnvironmentSatisfaction  HourlyRate  J
count 1470.000000  1470.000000      1470.000000  1470.000000          1470.0      1470.000000          1470.000000  1470.000000
mean   36.923810   802.485714        9.192517    2.912925            1.0      1024.865306          2.721769    65.891156
std     9.135373   403.509100        8.106864    1.024165            0.0        602.024335          1.093082    20.329428
min    18.000000   102.000000        1.000000    1.000000            1.0         1.000000          1.000000    30.000000
25%    30.000000   465.000000        2.000000    2.000000            1.0        491.250000          2.000000    48.000000
50%    36.000000   802.000000        7.000000    3.000000            1.0       1020.500000          3.000000    66.000000
75%    43.000000  1157.000000       14.000000    4.000000            1.0       1555.750000          4.000000    83.750000
max    60.000000  1499.000000       29.000000    5.000000            1.0       2068.000000          4.000000   100.000000

8 rows × 26 columns
```

```
#Checking for Null Values.
df.isnull().any()

Age                False
Attrition          False
BusinessTravel     False
DailyRate          False
```

```

Department      False
DistanceFromHome False
Education        False
EducationField   False
EmployeeCount    False
EmployeeNumber   False
EnvironmentSatisfaction False
Gender           False
HourlyRate       False
JobInvolvement   False
JobLevel         False
JobRole          False
JobSatisfaction  False
MaritalStatus    False
MonthlyIncome    False
MonthlyRate      False
NumCompaniesWorked False
Over18           False
OverTime         False
PercentSalaryHike False
PerformanceRating False
RelationshipSatisfaction False
StandardHours    False
StockOptionLevel False
TotalWorkingYears False
TrainingTimesLastYear False
WorkLifeBalance  False
YearsAtCompany   False
YearsInCurrentRole False
YearsSinceLastPromotion False
YearsWithCurrManager False
dtype: bool

```

```
df.isnull().sum()
```

```

Age              0
Attrition        0
BusinessTravel   0
DailyRate        0
Department       0
DistanceFromHome 0
Education        0
EducationField   0
EmployeeCount    0
EmployeeNumber   0
EnvironmentSatisfaction 0
Gender           0
HourlyRate       0
JobInvolvement   0
JobLevel         0
JobRole          0
JobSatisfaction  0
MaritalStatus    0
MonthlyIncome    0
MonthlyRate      0
NumCompaniesWorked 0
Over18           0
OverTime         0
PercentSalaryHike 0
PerformanceRating 0
RelationshipSatisfaction 0
StandardHours    0
StockOptionLevel 0
TotalWorkingYears 0
TrainingTimesLastYear 0
WorkLifeBalance  0
YearsAtCompany   0
YearsInCurrentRole 0
YearsSinceLastPromotion 0
YearsWithCurrManager 0
dtype: int64

```

```

#Data Visualization.
sns.distplot(df["YearsWithCurrManager"])

```

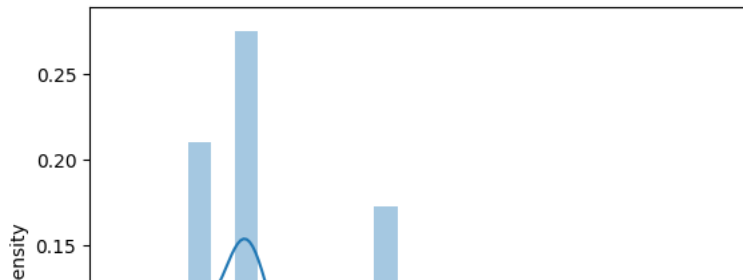
```
<ipython-input-12-71e8291be26b>:2: UserWarning:
```

```
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.
```

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(df["YearsWithCurrManager"])
<Axes: xlabel='YearsWithCurrManager', ylabel='Density'>
```



```
df.corr()
```

```
<ipython-input-13-2f6f6606aa2c>:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future ver
df.corr()
```

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount	EmployeeNumber	EnvironmentSatisfaction	Hc
Age	1.000000	0.010661	-0.001686	0.208034	NaN	-0.010145	0.010146	
DailyRate	0.010661	1.000000	-0.004985	-0.016806	NaN	-0.050990	0.018355	
DistanceFromHome	-0.001686	-0.004985	1.000000	0.021042	NaN	0.032916	-0.016075	
Education	0.208034	-0.016806	0.021042	1.000000	NaN	0.042070	-0.027128	
EmployeeCount	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
EmployeeNumber	-0.010145	-0.050990	0.032916	0.042070	NaN	1.000000	0.017621	
EnvironmentSatisfaction	0.010146	0.018355	-0.016075	-0.027128	NaN	0.017621	1.000000	
HourlyRate	0.024287	0.023381	0.031131	0.016775	NaN	0.035179	-0.049857	
JobInvolvement	0.029820	0.046135	0.008783	0.042438	NaN	-0.006888	-0.008278	
JobLevel	0.509604	0.002966	0.005303	0.101589	NaN	-0.018519	0.001212	
JobSatisfaction	-0.004892	0.030571	-0.003669	-0.011296	NaN	-0.046247	-0.006784	
MonthlyIncome	0.497855	0.007707	-0.017014	0.094961	NaN	-0.014829	-0.006259	
MonthlyRate	0.028051	-0.032182	0.027473	-0.026084	NaN	0.012648	0.037600	
NumCompaniesWorked	0.299635	0.038153	-0.029251	0.126317	NaN	-0.001251	0.012594	
PercentSalaryHike	0.003634	0.022704	0.040235	-0.011111	NaN	-0.012944	-0.031701	
PerformanceRating	0.001904	0.000473	0.027110	-0.024539	NaN	-0.020359	-0.029548	
RelationshipSatisfaction	0.053535	0.007846	0.006557	-0.009118	NaN	-0.069861	0.007665	
StandardHours	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
StockOptionLevel	0.037510	0.042143	0.044872	0.018422	NaN	0.062227	0.003432	
TotalWorkingYears	0.680381	0.014515	0.004628	0.148280	NaN	-0.014365	-0.002693	
TrainingTimesLastYear	-0.019621	0.002453	-0.036942	-0.025100	NaN	0.023603	-0.019359	
WorkLifeBalance	-0.021490	-0.037848	-0.026556	0.009819	NaN	0.010309	0.027627	
YearsAtCompany	0.311309	-0.034055	0.009508	0.069114	NaN	-0.011240	0.001458	
YearsInCurrentRole	0.212901	0.009932	0.018845	0.060236	NaN	-0.008416	0.018007	
YearsSinceLastPromotion	0.216513	-0.033229	0.010029	0.054254	NaN	-0.009019	0.016194	
YearsWithCurrManager	0.202089	-0.026363	0.014406	0.069065	NaN	-0.009197	-0.004999	

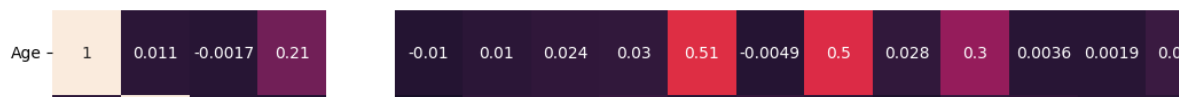
26 rows × 26 columns

```
df.head()
```

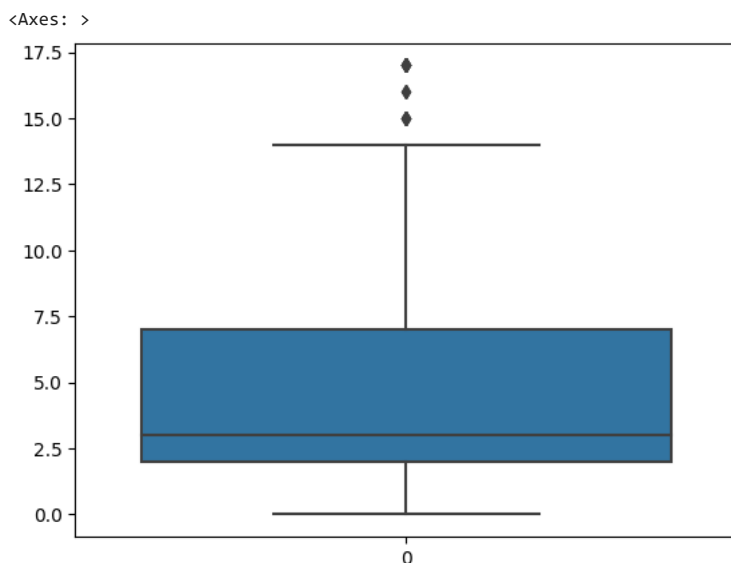
	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber
0	41	Yes	Travel_Rarely	1102	Sales		1	2	Life Sciences	1
1	49	No	Travel_Frequently	279	Research & Development		8	1	Life Sciences	1
2	37	Yes	Travel_Rarely	1373	Research & Development		2	2	Other	1
3	33	No	Travel_Frequently	1392	Research & Development		3	4	Life Sciences	1
4	39	No	Travel_Frequently	1196	Research & Development		2	1	Life Sciences	1

```
plt.subplots(figsize = (25,25))
sns.heatmap(df.corr(),annot=True)
```

```
<ipython-input-15-9329d5e70af4>:2: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future ver
sns.heatmap(df.corr(),annot=True)
<Axes: >
```



```
sns.boxplot(df.YearsWithCurrManager)
```



```
from scipy import stats
z_scores = np.abs(stats.zscore(df['YearsWithCurrManager']))
max_threshold=3
outliers = df['YearsWithCurrManager'][z_scores > max_threshold]
```

```
# Print and visualize the outliers
print("Outliers detected using Z-Score:")
print(outliers)
```

```
Outliers detected using Z-Score:
```

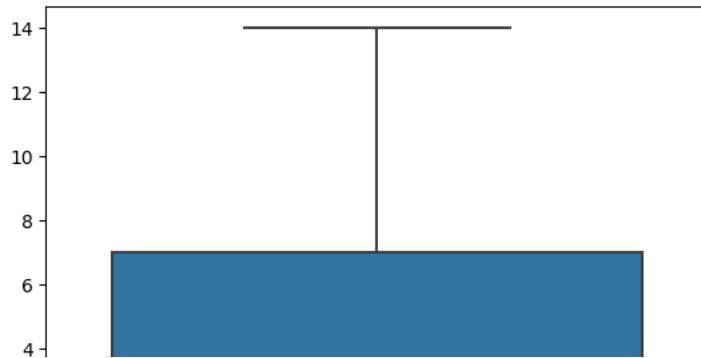
```
28      17
123     15
153     15
187     15
231     15
386     17
561     16
616     17
635     15
686     17
875     17
926     17
1078    17
1348    16
```

```
Name: YearsWithCurrManager, dtype: int64
```

```
q1 = df.YearsWithCurrManager.quantile(0.25)
q3 = df.YearsWithCurrManager.quantile(0.75)
print(q1)
print(q3)
upperlimit = q3+1.5*(q3-q1)
upperlimit
lowerlimit = q1-1.5*(q3-q1)
lowerlimit
df.median()
df["YearsWithCurrManager"]=np.where(df["YearsWithCurrManager"]>upperlimit,14,df['YearsWithCurrManager'])
sns.boxplot(df.YearsWithCurrManager)
```

2.0
7.0

```
<ipython-input-18-3a17581b0650>:9: FutureWarning: The default value of numeric_only in DataFrame.median is deprecated. In a future v
df.median()
<Axes: >
```



```
from scipy import stats
z_scores = np.abs(stats.zscore(df['YearsWithCurrManager']))
max_threshold=3
outliers = df['YearsWithCurrManager'][z_scores > max_threshold]
```

```
# Print and visualize the outliers
print("Outliers detected using Z-Score:")
print(outliers)
```

```
Outliers detected using Z-Score:
Series([], Name: YearsWithCurrManager, dtype: int64)
```

```
df.head()
```

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	1
1	49	No	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	2
2	37	Yes	Travel_Rarely	1373	Research & Development	2	2	Other	1	4
3	33	No	Travel_Frequently	1392	Research & Development	3	4	Life Sciences	1	5
4	27	No	Travel_Rarely	591	Research & Development	2	1	Medical	1	7

5 rows × 35 columns

```
x=df.drop('Attrition',axis=1)
x.head()
```

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	Environment
0	41	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	1	
1	49	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	2	
2	37	Travel_Rarely	1373	Research & Development	2	2	Other	1	4	
3	33	Travel_Frequently	1392	Research & Development	3	4	Life Sciences	1	5	
4	27	Travel_Rarely	591	Research & Development	2	1	Medical	1	7	

5 rows × 34 columns

```
y=df.Attrition
y.head()
```

```
0    Yes
1    No
2    Yes
3    No
4    No
Name: Attrition, dtype: object
```

```
#label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
x.BusinessTravel =le.fit_transform(x.BusinessTravel )
x.head()
x.Department =le.fit_transform(x.Department )
x.head()
x.EducationField =le.fit_transform(x.EducationField )
x.head()
x.Gender=le.fit_transform(x.Gender)
x.head()
x.JobRole =le.fit_transform(x.JobRole )
x.head()
x.MaritalStatus =le.fit_transform(x.MaritalStatus )
x.head()
x.Over18 =le.fit_transform(x.Over18 )
x.head()
x.OverTime =le.fit_transform(x.OverTime )
x.head()
```

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	Environmen
0	41	2	1102	2	1	2	1	1	1	
1	49	1	279	1	8	1	1	1	1	2
2	37	2	1373	1	2	2	4	1	1	4
3	33	1	1392	1	3	4	1	1	1	5
4	27	2	591	1	2	1	3	1	1	7

5 rows × 34 columns

df.columns

```
Index(['Age', 'Attrition', 'BusinessTravel', 'DailyRate', 'Department',
      'DistanceFromHome', 'Education', 'EducationField', 'EmployeeCount',
      'EmployeeNumber', 'EnvironmentSatisfaction', 'Gender', 'HourlyRate',
      'JobInvolvement', 'JobLevel', 'JobRole', 'JobSatisfaction',
      'MaritalStatus', 'MonthlyIncome', 'MonthlyRate', 'NumCompaniesWorked',
      'Over18', 'OverTime', 'PercentSalaryHike', 'PerformanceRating',
      'RelationshipSatisfaction', 'StandardHours', 'StockOptionLevel',
      'TotalWorkingYears', 'TrainingTimesLastYear', 'WorkLifeBalance',
      'YearsAtCompany', 'YearsInCurrentRole', 'YearsSinceLastPromotion',
      'YearsWithCurrManager'],
      dtype='object')
```

```
#feature scaling
from sklearn.preprocessing import MinMaxScaler
ms=MinMaxScaler()
x_scaled=pd.DataFrame(ms.fit_transform(x),columns=x.columns)
```

x_scaled

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	En
0	0.547619	1.0	0.715820	1.0	0.000000	0.25	0.2	0.0	0.000000	
1	0.738095	0.5	0.126700	0.5	0.250000	0.00	0.2	0.0	0.000484	
2	0.452381	1.0	0.909807	0.5	0.035714	0.25	0.8	0.0	0.001451	
3	0.357143	0.5	0.923407	0.5	0.071429	0.75	0.2	0.0	0.001935	
4	0.214286	1.0	0.350036	0.5	0.035714	0.00	0.6	0.0	0.002903	
...
1465	0.428571	0.5	0.559771	0.5	0.785714	0.25	0.6	0.0	0.996613	
1466	0.500000	1.0	0.365784	0.5	0.178571	0.00	0.6	0.0	0.997097	
1467	0.214286	1.0	0.037938	0.5	0.107143	0.50	0.2	0.0	0.998065	
1468	0.738095	0.5	0.659270	1.0	0.035714	0.50	0.6	0.0	0.998549	
1469	0.380952	1.0	0.376521	0.5	0.250000	0.50	0.6	0.0	1.000000	

1470 rows × 34 columns


```
x_train.head()
```

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	En
1374	0.952381	1.0	0.360057	1.0	0.714286	0.50	0.2	0.0	0.937107	
1092	0.642857	1.0	0.607015	0.5	0.964286	0.50	1.0	0.0	0.747460	
768	0.523810	1.0	0.141732	1.0	0.892857	0.50	0.4	0.0	0.515239	
569	0.428571	0.0	0.953472	1.0	0.250000	0.75	0.2	0.0	0.381229	
911	0.166667	0.5	0.355762	1.0	0.821429	0.00	0.2	0.0	0.615385	

y_test

9/22

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNur
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	
1	49	No	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	
2	37	Yes	Travel_Rarely	1373	Research & Development	2	2	Other	1	
3	33	No	Travel_Frequently	1392	Research & Development	3	4	Life Sciences	1	
4	27	No	Travel_Rarely	591	Research & Development	2	1	Medical	1	
...	
1465	36	No	Travel_Frequently	884	Research & Development	23	2	Medical	1	:
1466	39	No	Travel_Rarely	613	Research & Development	6	1	Medical	1	:
1467	27	No	Travel_Rarely	155	Research & Development	4	3	Life Sciences	1	:
1468	49	No	Travel_Frequently	1023	Sales	2	3	Medical	1	:
1469	34	No	Travel_Rarely	628	Research & Development	8	3	Medical	1	:

1470 rows × 35 columns

▼ Evaluation of classification model

```
#Accuracy score
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report,roc_auc_score,roc_curve
```

```
accuracy_score(y_test,pred)

0.8843537414965986
```

```
confusion_matrix(y_test,pred)

array([[242,  3],
       [ 31, 18]])
```

```
pd.crosstab(y_test,pred)
```

col_0	No	Yes
Attrition		
No	242	3
Yes	31	18

▼ Roc-AUC curve

```
probability=model.predict_proba(x_test)[: ,1]
probability

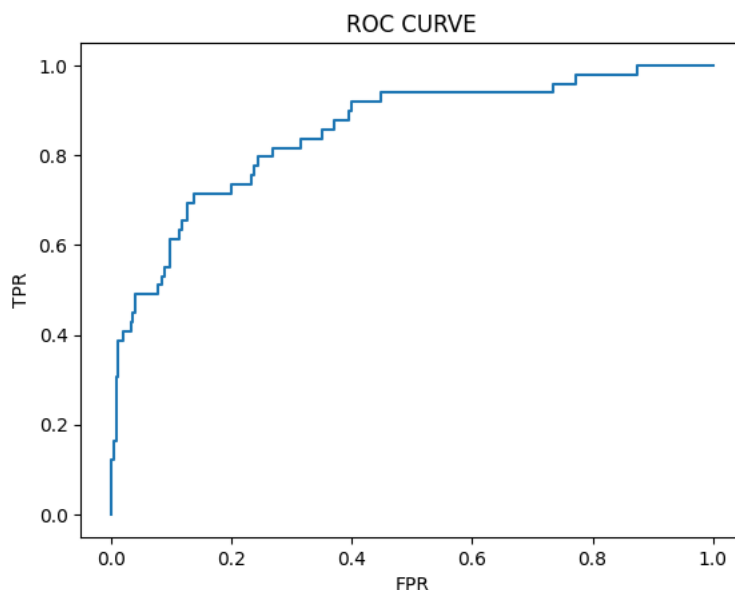
0.06205401, 0.61414184, 0.07466397, 0.00797252, 0.39157785,
0.05281564, 0.33160211, 0.02022395, 0.6671328 , 0.19419683,
0.0335299 , 0.10954936, 0.17130578, 0.043804 , 0.2241511 ,
0.23531373, 0.01475346, 0.06562592, 0.05019163, 0.59115162,
0.44667993, 0.07401303, 0.0449937 , 0.67637047, 0.05859033,
0.01545736, 0.03386798, 0.07021403, 0.1707141 , 0.07767295,
0.04154894, 0.08312937, 0.06997437, 0.03567429, 0.05269126,
0.05742727, 0.02144976, 0.01779053, 0.01301572, 0.02825292,
0.50162054, 0.41541766, 0.00299378, 0.74315718, 0.51799699,
0.09708281, 0.48942319, 0.07941138, 0.25720931, 0.66861063,
0.26482373, 0.01970983, 0.30281497, 0.02858501, 0.16213966,
0.02040161, 0.2173984 , 0.13768821, 0.03568054, 0.37558052,
0.03010741, 0.29718154, 0.15832399, 0.10264349, 0.08700774,
0.0815183 , 0.30943969, 0.08708969, 0.07442596, 0.12300414,
```

```
0.00857413, 0.02334842, 0.13000332, 0.02381787, 0.03217004,
0.0821409 , 0.00518749, 0.035308 , 0.03813342, 0.14270872,
0.26418695, 0.16461435, 0.27401734, 0.24146954, 0.02119787,
0.17774284, 0.34102562, 0.28338745, 0.06906981, 0.04948532,
0.24465264, 0.74929682, 0.35691434, 0.01878265, 0.08772637,
0.03239915, 0.05413857, 0.15215059, 0.07127406, 0.13828798,
0.09342465, 0.04693869, 0.02494493, 0.15041914, 0.07133392,
0.03025642, 0.05306455, 0.1165452 , 0.00872431, 0.01229042,
0.17575238, 0.05005249, 0.09018395, 0.82857166, 0.03066995,
0.0228189 , 0.00874605, 0.13496234, 0.16593413, 0.05060052,
0.01520085, 0.29791945, 0.54919611, 0.33581407, 0.0469494 ,
0.38773566, 0.61348127, 0.14171081, 0.07455884, 0.2409655 ,
0.09528764, 0.06730943, 0.09797576, 0.20026612, 0.20053142,
0.03046036, 0.14877431, 0.0036571 , 0.11146887, 0.15912883,
0.06017571, 0.17964687, 0.06063618, 0.1199213 , 0.03284092,
0.02688355, 0.06536903, 0.08335812, 0.01464284, 0.01536292,
0.37701597, 0.01262506, 0.15004068, 0.80530948, 0.11655522,
0.28461049, 0.17042029, 0.15392139, 0.02756879, 0.00599553,
0.04142216, 0.09958411, 0.11567269, 0.10448555, 0.01830036,
0.1444171 , 0.1048541 , 0.10079777, 0.05099176, 0.09183576,
0.02893646, 0.09754427, 0.00516687, 0.75206394, 0.04227453,
0.04018918, 0.37563319, 0.04457964, 0.72551665, 0.10583031,
0.36656526, 0.38293703, 0.32923777, 0.05248015, 0.08216713,
0.13748888, 0.04309097, 0.01429957, 0.2656631 , 0.06297408,
0.16075744, 0.15388494, 0.67190498, 0.05834473, 0.28467369,
0.04694404, 0.46237195, 0.00339026, 0.13927388, 0.02695884,
0.12707414, 0.17395277, 0.0750947 , 0.10135673, 0.16496216,
0.02583798, 0.01790826, 0.08850395, 0.02838351, 0.13795992,
0.08655223, 0.22164621, 0.73379009, 0.17294814, 0.40907888,
0.01503347, 0.11411826, 0.21412683, 0.32566668, 0.03366086,
0.04472831, 0.32127248, 0.05442236, 0.0242917 , 0.16228044,
0.32858438, 0.22879119, 0.00852736, 0.0798162 , 0.01140248,
0.14102568, 0.29116266, 0.01282151, 0.17118076, 0.04051376,
0.04165738, 0.42684273, 0.35009936, 0.0366853 , 0.11692325,
0.37940034, 0.31562415, 0.79587005, 0.05488792, 0.21568794,
0.06397987, 0.00569145, 0.66085682, 0.35796045, 0.37592133,
0.3650533 , 0.03568965, 0.21192376, 0.05892118, 0.06428028,
0.10143977, 0.00796354, 0.2678938 , 0.4288445 , 0.0652538 ,
0.09309022, 0.01226927, 0.14314823, 0.04989664, 0.02304292,
0.02508766, 0.06618985, 0.24272596, 0.26663754, 0.1979951 ,
0.26504226, 0.01648205, 0.15826843, 0.08519882, 0.02669729,
0.18757572, 0.00768502, 0.27928747, 0.0027473 , 0.02506718,
0.22608608, 0.72428674, 0.07739605, 0.26575953]])
```

```
#label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
y_test=le.fit_transform(y_test)

# roc_curve
fpr, tpr, thresholds = roc_curve(y_test, probability)

plt.plot(fpr, tpr)
plt.xlabel('FPR')
plt.ylabel('TPR')
plt.title('ROC CURVE')
plt.show()
```



```
from sklearn.tree import DecisionTreeClassifier
dtc=DecisionTreeClassifier()
```

```
dtc.fit(x_train,y_train)
```

```
DecisionTreeClassifier
DecisionTreeClassifier()
```

```
pred=dtc.predict(x_test)
```

```
pred
```

```
array(['No', 'No', 'No', 'No', 'No', 'No', 'Yes', 'No', 'No', 'No', 'No',
       'No', 'No', 'Yes', 'No', 'No', 'No', 'No', 'No', 'Yes', 'No',
       'No', 'No', 'No', 'No', 'No', 'No', 'Yes', 'No', 'No', 'No',
       'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'Yes', 'No', 'Yes',
       'No', 'No', 'No', 'No', 'No', 'No', 'Yes', 'No', 'Yes', 'No',
       'Yes', 'Yes', 'No', 'No', 'No', 'Yes', 'No', 'No', 'No', 'No',
       'No', 'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'Yes', 'No', 'No',
       'No', 'No', 'No', 'No', 'No', 'No', 'Yes', 'No', 'No', 'No', 'No',
       'Yes', 'No', 'No', 'Yes', 'Yes', 'No', 'No', 'No', 'No', 'No',
       'No', 'No', 'No', 'Yes', 'No', 'No', 'Yes', 'No', 'No', 'No',
       'Yes', 'No', 'No', 'No', 'Yes', 'No', 'No', 'No', 'No', 'Yes',
       'No', 'No', 'No', 'Yes', 'No', 'No', 'No', 'No', 'No', 'No',
       'No', 'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No',
       'No', 'No', 'No', 'No', 'No', 'Yes', 'No', 'No', 'Yes', 'No', 'No',
       'No', 'No', 'Yes', 'No', 'No', 'Yes', 'No', 'No', 'Yes', 'No',
       'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No',
       'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No',
       'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No',
       'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No',
       'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No',
       'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No', 'No',
       'No', 'No', 'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'No',
       'Yes', 'No', 'Yes', 'No', 'No', 'No', 'No', 'No', 'No', 'No',
       'No', 'Yes', 'No', 'Yes', 'No', 'No', 'Yes', 'Yes', 'No', 'Yes',
       'No', 'No', 'Yes', 'No', 'No', 'No', 'No', 'No', 'No'], dtype=object)
```

```
y_test
```

```
array([0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
       0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0,
       1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
       1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
       0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0,
       1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0,
       0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
       0, 1, 0, 0, 0, 1, 0, 0])
```

```
df
```

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNur
0	41	Yes	Travel_Rarely	1102	Sales		1	2	Life Sciences	1
1	49	No	Travel_Frequently	279	Research & Development		8	1	Life Sciences	1
2	37	Yes	Travel_Rarely	1373	Research & Development		2	2	Other	1
3	33	No	Travel_Frequently	1392	Research & Development		3	4	Life Sciences	1

▼ Evaluation of classification model

```
#Accuracy score
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report,roc_auc_score,roc_curve

#label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
y=le.fit_transform(y)
#label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
pred=le.fit_transform(pred)

y_test

array([[0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
        0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
        1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1,
        1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1,
        0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0,
        0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0,
        1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0,
        0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
        0, 1, 0, 0, 0, 1, 0, 0, 0]])

accuracy_score(y_test,pred)

0.7482993197278912

confusion_matrix(y_test,pred)

array([[203, 42],
       [ 32, 17]])

pd.crosstab(y_test,pred)

col_0    0    1
row_0
0      203  42
1       32  17

print(classification_report(y_test,pred))

              precision    recall  f1-score   support

0               0.86         0.83         0.85         245
1               0.29         0.35         0.31          49

 accuracy
macro avg         0.58         0.59         0.58         294
weighted avg         0.77         0.75         0.76         294
```

▼ Roc-AUC curve

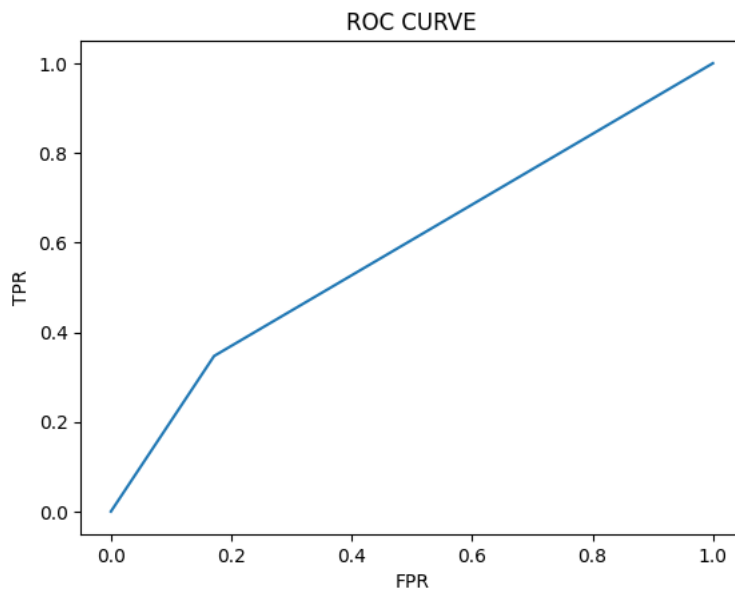
```
probability=dtc.predict_proba(x_test)[: ,1]
```

```
probability
```

```
array([[0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 1., 0., 0., 0.,
        0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0.,
        0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 1., 1.,
        1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 1., 0., 0., 0.,
        0., 0., 0., 1., 0., 1., 0., 1., 1., 0., 0., 0., 1., 0., 0., 0., 0.,
        0., 1., 0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0.,
        1., 0., 0., 0., 0., 1., 0., 0., 1., 1., 0., 0., 0., 0., 0., 0.,
        0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 1., 1., 1., 1., 0.,
        0., 0., 1., 0., 0., 0., 1., 0., 0., 0., 0., 1., 0., 0., 0., 1., 0.,
        0., 0., 0., 0., 0., 0., 1., 0., 0., 1., 0., 0., 0., 0., 0., 0.,
        0., 0., 0., 0., 0., 0., 1., 0., 0., 1., 0., 0., 0., 0., 0., 1., 0.,
        1., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 1., 1., 0.,
        0., 1., 0., 0., 0., 0., 1., 1., 0., 0., 1., 0., 0., 0., 0., 0.,
        0., 1., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 0.,
        0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 1., 0., 1., 0., 0., 0.,
        0., 0., 0., 0., 0., 1., 0., 1., 0., 0., 1., 0., 0., 0., 0.,
        0., 0., 0., 0., 0., 1., 0., 1., 0., 1., 1., 0., 1., 0., 0., 1.,
        0., 0., 0., 0., 0.]])
```

```
fpr,tpr,thresholds = roc_curve(y_test,probability)
```

```
plt.plot(fpr,tpr)
plt.xlabel('FPR')
plt.ylabel('TPR')
plt.title('ROC CURVE')
plt.show()
```



```
from sklearn import tree
plt.figure(figsize=(25,15))
tree.plot_tree(dtc,filled=True)
```

```
[Text(0.3291164340101523, 0.9722222222222222, 'x[27] <= 0.038\ngini = 0.269\nnsamples = 1176\nnvalue = [988, 188]'),
Text(0.08121827411167512, 0.9166666666666666, 'x[16] <= 0.75\ngini = 0.5\nnsamples = 78\nnvalue = [39, 39]'),
Text(0.050761421319796954, 0.8611111111111112, 'x[4] <= 0.554\ngini = 0.426\nnsamples = 39\nnvalue = [27, 12]'),
Text(0.0338409475465313, 0.8055555555555556, 'x[15] <= 0.167\ngini = 0.312\nnsamples = 31\nnvalue = [25, 6]'),
Text(0.02030456852791878, 0.75, 'x[21] <= 0.5\ngini = 0.49\nnsamples = 7\nnvalue = [3, 4]'),
Text(0.01353637901861252, 0.6944444444444444, 'x[22] <= 0.321\ngini = 0.375\nnsamples = 4\nnvalue = [3, 1]'),
Text(0.00676818950930626, 0.6388888888888888, 'gini = 0.0\nnsamples = 3\nnvalue = [3, 0]'),
Text(0.02030456852791878, 0.6388888888888888, 'gini = 0.0\nnsamples = 1\nnvalue = [0, 1]'),
Text(0.02707275803722504, 0.6944444444444444, 'gini = 0.0\nnsamples = 3\nnvalue = [0, 3]'),
Text(0.047377326565143825, 0.75, 'x[19] <= 0.056\ngini = 0.153\nnsamples = 24\nnvalue = [22, 2]'),
Text(0.04060913705583756, 0.6944444444444444, 'gini = 0.0\nnsamples = 1\nnvalue = [0, 1]'),
Text(0.05414551607445008, 0.6944444444444444, 'x[9] <= 0.167\ngini = 0.083\nnsamples = 23\nnvalue = [22, 1]'),
Text(0.047377326565143825, 0.6388888888888888, 'x[0] <= 0.214\ngini = 0.5\nnsamples = 2\nnvalue = [1, 1]'),
Text(0.04060913705583756, 0.5833333333333334, 'gini = 0.0\nnsamples = 1\nnvalue = [1, 0]'),
Text(0.05414551607445008, 0.5833333333333334, 'gini = 0.0\nnsamples = 1\nnvalue = [0, 1]'),
Text(0.06091370558375635, 0.6388888888888888, 'gini = 0.0\nnsamples = 21\nnvalue = [21, 0]'),
Text(0.0676818950930626, 0.8055555555555556, 'x[8] <= 0.385\ngini = 0.375\nnsamples = 8\nnvalue = [2, 6]'),
Text(0.06091370558375635, 0.75, 'gini = 0.0\nnsamples = 2\nnvalue = [2, 0]'),
Text(0.07445008460236886, 0.75, 'gini = 0.0\nnsamples = 6\nnvalue = [0, 6]'),
Text(0.1116751269035533, 0.8611111111111112, 'x[11] <= 0.364\ngini = 0.426\nnsamples = 39\nnvalue = [12, 27]'),
Text(0.09475465313028765, 0.8055555555555556, 'x[29] <= 0.167\ngini = 0.133\nnsamples = 14\nnvalue = [1, 13]'),
Text(0.08798646362098139, 0.75, 'gini = 0.0\nnsamples = 1\nnvalue = [1, 0]'),
Text(0.10152284263959391, 0.75, 'gini = 0.0\nnsamples = 13\nnvalue = [0, 13]'),
Text(0.12859560067681894, 0.8055555555555556, 'x[8] <= 0.105\ngini = 0.493\nnsamples = 25\nnvalue = [11, 14]'),
Text(0.11505922165820642, 0.75, 'x[22] <= 0.464\ngini = 0.278\nnsamples = 6\nnvalue = [5, 1]'),
Text(0.10829103214890017, 0.6944444444444444, 'gini = 0.0\nnsamples = 5\nnvalue = [5, 0]'),
Text(0.1218274111675127, 0.6944444444444444, 'gini = 0.0\nnsamples = 1\nnvalue = [0, 1]'),
Text(0.14213197969543148, 0.75, 'x[15] <= 0.5\ngini = 0.432\nnsamples = 19\nnvalue = [6, 13]'),
Text(0.1353637901861252, 0.6944444444444444, 'gini = 0.0\nnsamples = 7\nnvalue = [0, 7]'),
Text(0.14890016920473773, 0.6944444444444444, 'x[6] <= 0.4\ngini = 0.5\nnsamples = 12\nnvalue = [6, 6]'),
Text(0.1353637901861252, 0.6388888888888888, 'x[3] <= 0.75\ngini = 0.278\nnsamples = 6\nnvalue = [5, 1]'),
Text(0.12859560067681894, 0.5833333333333334, 'gini = 0.0\nnsamples = 5\nnvalue = [5, 0]'),
Text(0.14213197969543148, 0.5833333333333334, 'gini = 0.0\nnsamples = 1\nnvalue = [0, 1]'),
Text(0.16243654822335024, 0.6388888888888888, 'x[8] <= 0.249\ngini = 0.278\nnsamples = 6\nnvalue = [1, 5]'),
Text(0.155668358714044, 0.5833333333333334, 'gini = 0.0\nnsamples = 1\nnvalue = [1, 0]'),
Text(0.1692047377326565, 0.5833333333333334, 'gini = 0.0\nnsamples = 5\nnvalue = [0, 5]'),
Text(0.5770145939086294, 0.9166666666666666, 'x[21] <= 0.5\ngini = 0.235\nnsamples = 1098\nnvalue = [949, 149]'),
Text(0.3325401861252115, 0.8611111111111112, 'x[29] <= 0.167\ngini = 0.162\nnsamples = 798\nnvalue = [727, 71]'),
Text(0.18274111675126903, 0.8055555555555556, 'x[8] <= 0.445\ngini = 0.38\nnsamples = 47\nnvalue = [35, 12]'),
Text(0.1692047377326565, 0.75, 'x[16] <= 0.75\ngini = 0.1\nnsamples = 19\nnvalue = [18, 1]'),
Text(0.16243654822335024, 0.6944444444444444, 'gini = 0.0\nnsamples = 18\nnvalue = [18, 0]'),
Text(0.17597292724196278, 0.6944444444444444, 'gini = 0.0\nnsamples = 1\nnvalue = [0, 1]'),
Text(0.19627749576988154, 0.75, 'x[17] <= 0.094\ngini = 0.477\nnsamples = 28\nnvalue = [17, 11]'),
Text(0.1895093062605753, 0.6944444444444444, 'gini = 0.0\nnsamples = 4\nnvalue = [0, 4]'),
Text(0.20304568527918782, 0.6944444444444444, 'x[8] <= 0.524\ngini = 0.413\nnsamples = 24\nnvalue = [17, 7]'),
Text(0.19627749576988154, 0.6388888888888888, 'gini = 0.0\nnsamples = 2\nnvalue = [0, 2]'),
Text(0.2098138747884941, 0.6388888888888888, 'x[33] <= 0.393\ngini = 0.351\nnsamples = 22\nnvalue = [17, 5]'),
```

```

Text(0.19627749576988154, 0.5833333333333334, 'x[2] <= 0.025\ngini =
0.133\nsamples = 14\nvalue = [13, 1]'),
Text(0.1895093062605753, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.20304568527918782, 0.5277777777777778, 'gini = 0.0\nsamples = 13\nvalue =
[13, 0]'),
Text(0.2233502538071066, 0.5833333333333334, 'x[2] <= 0.329\ngini = 0.5\nsamples
= 8\nvalue = [4, 4]'),
Text(0.21658206429780033, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.23011844331641285, 0.5277777777777778, 'x[12] <= 0.333\ngini =
0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.2233502538071066, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.23688663282571912, 0.4722222222222222, 'gini = 0.0\nsamples = 4\nvalue =
[4, 0]'),
Text(0.48233925549915396, 0.8055555555555556, 'x[30] <= 0.963\ngini =
0.145\nsamples = 751\nvalue = [692, 59]'),
Text(0.4755710659898477, 0.75, 'x[30] <= 0.113\ngini = 0.143\nsamples =
750\nvalue = [692, 58]'),
Text(0.35152284263959394, 0.6944444444444444, 'x[9] <= 0.167\ngini =
0.218\nsamples = 257\nvalue = [225, 32]'),
Text(0.3096446700507614, 0.6388888888888888, 'x[33] <= 0.179\ngini =
0.355\nsamples = 65\nvalue = [50, 15]'),
Text(0.2876480541455161, 0.5833333333333334, 'x[33] <= 0.036\ngini =
0.303\nsamples = 59\nvalue = [48, 11]'),
Text(0.2639593908629442, 0.5277777777777778, 'x[12] <= 0.5\ngini =
0.463\nsamples = 22\nvalue = [14, 8]'),
Text(0.25042301184433163, 0.4722222222222222, 'x[11] <= 0.179\ngini =
0.198\nsamples = 9\nvalue = [8, 1]'),
Text(0.2436548223350254, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.2571912013536379, 0.4166666666666667, 'gini = 0.0\nsamples = 8\nvalue =
[8, 0]'),
Text(0.27749576988155666, 0.4722222222222222, 'x[11] <= 0.4\ngini =
0.497\nsamples = 13\nvalue = [6, 7]'),
Text(0.2707275803722504, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue =
[4, 0]'),
Text(0.28426395939086296, 0.4166666666666667, 'x[4] <= 0.286\ngini =
0.346\nsamples = 9\nvalue = [2, 7]'),
Text(0.27749576988155666, 0.3611111111111111, 'x[0] <= 0.226\ngini =
0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.2707275803722504, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.28426395939086296, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.2910321489001692, 0.3611111111111111, 'gini = 0.0\nsamples = 6\nvalue =
[0, 6]'),
Text(0.311336717428088, 0.5277777777777778, 'x[15] <= 0.167\ngini =
0.149\nsamples = 37\nvalue = [34, 3]'),
Text(0.30456852791878175, 0.4722222222222222, 'x[29] <= 0.5\ngini = 0.5\nsamples
= 6\nvalue = [3, 3]'),
Text(0.29780033840947545, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),
Text(0.311336717428088, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.31810490693739424, 0.4722222222222222, 'gini = 0.0\nsamples = 31\nvalue =
[31, 0]'),
Text(0.3316412859560068, 0.5833333333333334, 'x[8] <= 0.065\ngini =
0.444\nsamples = 6\nvalue = [2, 4]'),
Text(0.3248730964467005, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.338409475465313, 0.5277777777777778, 'gini = 0.0\nsamples = 4\nvalue =
[0, 4]'),
Text(0.3934010152284264, 0.6388888888888888, 'x[0] <= 0.321\ngini =
0.161\nsamples = 192\nvalue = [175, 17]'),
Text(0.3587140439932318, 0.5833333333333334, 'x[6] <= 0.1\ngini = 0.294\nsamples
= 67\nvalue = [55, 12]'),
Text(0.35194585448392557, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.36548223350253806, 0.5277777777777778, 'x[29] <= 0.5\ngini =
0.26\nsamples = 65\nvalue = [55, 10]'),
Text(0.34856175972927245, 0.4722222222222222, 'x[11] <= 0.679\ngini =
0.469\nsamples = 16\nvalue = [10, 6]'),
Text(0.34179357021996615, 0.4166666666666667, 'x[6] <= 0.4\ngini =
0.444\nsamples = 9\nvalue = [3, 6]'),
Text(0.3350253807106599, 0.3611111111111111, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.34856175972927245, 0.3611111111111111, 'x[2] <= 0.126\ngini =
0.245\nsamples = 7\nvalue = [1, 6]'),
Text(0.34179357021996615, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.3553299492385787, 0.3055555555555556, 'gini = 0.0\nsamples = 6\nvalue =
[0, 6]'),
Text(0.3553299492385787, 0.4166666666666667, 'gini = 0.0\nsamples = 7\nvalue =
[7, 0]'),
Text(0.3824027072758037, 0.4722222222222222, 'x[2] <= 0.037\ngini =
0.15\nsamples = 49\nvalue = [45, 4]'),
Text(0.3756345177664975, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.38017000670510007, 0.4166666666666667, 'x[2] <= 0.030\ngini =

```



```
Text(0.3824027072758037, 0.3611111111111111, 'x[5] <= 0.875\ngini = 0.081\nsamples = 47\nvalue = [45, 2]'),  
Text(0.3688663282571912, 0.3055555555555556, 'x[12] <= 0.167\ngini = 0.043\nsamples = 45\nvalue = [44, 1]'),  
Text(0.36209813874788493, 0.25, 'x[3] <= 0.75\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),  
Text(0.3553299492385787, 0.19444444444444445, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),  
Text(0.3688663282571912, 0.19444444444444445, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),  
Text(0.3756345177664975, 0.25, 'gini = 0.0\nsamples = 42\nvalue = [42, 0]'),  
Text(0.39593908629441626, 0.3055555555555556, 'x[32] <= 0.1\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),  
Text(0.38917089678510997, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),  
Text(0.4027072758037225, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),  
Text(0.39593908629441626, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),  
Text(0.428087986463621, 0.5833333333333334, 'x[8] <= 0.022\ngini = 0.077\nsamples = 125\nvalue = [120, 5]'),  
Text(0.40947546531302875, 0.5277777777777778, 'x[2] <= 0.578\ngini = 0.5\nsamples = 4\nvalue = [2, 2]'),  
Text(0.4027072758037225, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),  
Text(0.41624365482233505, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),  
Text(0.4467005076142132, 0.5277777777777778, 'x[18] <= 0.968\ngini = 0.048\nsamples = 121\nvalue = [118, 3]'),  
Text(0.42978003384094754, 0.4722222222222222, 'x[2] <= 0.98\ngini = 0.033\nsamples = 118\nvalue = [116, 2]'),  
Text(0.41624365482233505, 0.4166666666666667, 'x[14] <= 0.938\ngini = 0.017\nsamples = 114\nvalue = [113, 1]'),  
Text(0.40947546531302875, 0.3611111111111111, 'gini = 0.0\nsamples = 107\nvalue = [107, 0]'),  
Text(0.4230118443316413, 0.3611111111111111, 'x[16] <= 0.25\ngini = 0.245\nsamples = 7\nvalue = [6, 1]'),  
Text(0.41624365482233505, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),  
Text(0.42978003384094754, 0.3055555555555556, 'gini = 0.0\nsamples = 6\nvalue = [6, 0]'),  
Text(0.4433164128595601, 0.4166666666666667, 'x[1] <= 0.25\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),  
Text(0.4365482233502538, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),  
Text(0.4500846023688663, 0.3611111111111111, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),  
Text(0.46362098138747887, 0.4722222222222222, 'x[19] <= 0.278\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),  
Text(0.45685279187817257, 0.4166666666666667, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),  
Text(0.4703891708967851, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),  
Text(0.5996192893401016, 0.6944444444444444, 'x[30] <= 0.787\ngini = 0.1\nsamples = 493\nvalue = [467, 26]'),  
Text(0.5647208121827412, 0.6388888888888889, 'x[15] <= 0.5\ngini = 0.094\nsamples = 486\nvalue = [462, 24]'),  
Text(0.5152284263959391, 0.5833333333333334, 'x[14] <= 0.938\ngini = 0.154\nsamples = 191\nvalue = [175, 16]'),  
Text(0.5084602368866328, 0.5277777777777778, 'x[18] <= 0.481\ngini = 0.145\nsamples = 190\nvalue = [175, 15]'),  
Text(0.4906937394247039, 0.4722222222222222, 'x[33] <= 0.964\ngini = 0.221\nsamples = 95\nvalue = [83, 12]'),  
Text(0.48392554991539766, 0.4166666666666667, 'x[18] <= 0.47\ngini = 0.207\nsamples = 94\nvalue = [83, 11]'),  
Text(0.47715736040609136, 0.3611111111111111, 'x[5] <= 0.375\ngini = 0.192\nsamples = 93\nvalue = [83, 10]'),  
Text(0.45516074450084604, 0.3055555555555556, 'x[6] <= 0.9\ngini = 0.363\nsamples = 21\nvalue = [16, 5]'),  
Text(0.44839255499153974, 0.25, 'x[17] <= 0.413\ngini = 0.266\nsamples = 19\nvalue = [16, 3]'),  
Text(0.43485617597292725, 0.19444444444444445, 'x[8] <= 0.215\ngini = 0.117\nsamples = 16\nvalue = [15, 1]'),  
Text(0.428087986463621, 0.1388888888888889, 'x[31] <= 0.417\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),  
Text(0.4213197969543147, 0.08333333333333333, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),  
Text(0.43485617597292725, 0.08333333333333333, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),  
Text(0.4416243654822335, 0.1388888888888889, 'gini = 0.0\nsamples = 14\nvalue = [14, 0]'),  
Text(0.4619289340101523, 0.19444444444444445, 'x[31] <= 0.556\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),  
Text(0.45516074450084604, 0.1388888888888889, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),  
Text(0.4686971235194585, 0.1388888888888889, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),  
Text(0.4619289340101523, 0.25, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),  
Text(0.49915397631133673, 0.3055555555555556, 'x[31] <= 0.139\ngini = 0.129\nsamples = 72\nvalue = [67, 5]'),  
Text(0.48223350253807107, 0.25, 'x[8] <= 0.68\ngini = 0.444\nsamples = 6\nvalue
```

```

= [4, 2]'),
  Text(0.4754653130287648, 0.19444444444444445, 'gini = 0.0\nsamples = 4\nvalue =
[4, 0]'),
  Text(0.4890016920473773, 0.19444444444444445, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
  Text(0.5160744500846024, 0.25, 'x[11] <= 0.993\ngini = 0.087\nsamples =
66\nvalue = [63, 3]'),
  Text(0.5025380710659898, 0.19444444444444445, 'x[28] <= 0.583\ngini =
0.061\nsamples = 64\nvalue = [62, 2]'),
  Text(0.4957698815566836, 0.13888888888888889, 'gini = 0.0\nsamples = 51\nvalue =
[51, 0]'),
  Text(0.5093062605752962, 0.13888888888888889, 'x[9] <= 0.5\ngini = 0.26\nsamples
= 13\nvalue = [11, 2]'),
  Text(0.5025380710659898, 0.08333333333333333, 'x[17] <= 0.335\ngini =
0.5\nsamples = 4\nvalue = [2, 2]'),
  Text(0.4957698815566836, 0.027777777777777776, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
  Text(0.5093062605752962, 0.027777777777777776, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
  Text(0.5160744500846024, 0.08333333333333333, 'gini = 0.0\nsamples = 9\nvalue =
[9, 0]'),
  Text(0.5296108291032149, 0.19444444444444445, 'x[24] <= 0.5\ngini = 0.5\nsamples
= 2\nvalue = [1, 1]'),
  Text(0.5228426395939086, 0.13888888888888889, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.5363790186125211, 0.13888888888888889, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.4906937394247039, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.49746192893401014, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.5262267343485617, 0.4722222222222222, 'x[19] <= 0.5\ngini =
0.061\nsamples = 95\nvalue = [92, 3]'),
  Text(0.5194585448392555, 0.4166666666666667, 'gini = 0.0\nsamples = 76\nvalue =
[76, 0]'),
  Text(0.5329949238578681, 0.4166666666666667, 'x[8] <= 0.161\ngini =
0.266\nsamples = 19\nvalue = [16, 3]'),
  Text(0.5194585448392555, 0.3611111111111111, 'x[24] <= 0.833\ngini =
0.444\nsamples = 3\nvalue = [1, 2]'),
  Text(0.5126903553299492, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
  Text(0.5262267343485617, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.5465313028764806, 0.3611111111111111, 'x[31] <= 0.639\ngini =
0.117\nsamples = 16\nvalue = [15, 1]'),
  Text(0.5397631133671743, 0.3055555555555556, 'gini = 0.0\nsamples = 14\nvalue =
[14, 0]'),
  Text(0.5532994923857868, 0.3055555555555556, 'x[19] <= 0.944\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),

from sklearn.model_selection import GridSearchCV
parameter={
  'criterion':['gini','entropy'],
  'splitter':['best','random'],
  'max_depth':[1,2,3,4,5],
  'max_features':['auto', 'sqrt', 'log2']
}

  Text(0.5668358714043993, 0.4166666666666667, 'x[26] <= 0.5\ngini = 0.5\nsamples
grid_search=GridSearchCV(estimator=dtc,param_grid=parameter,cv=5,scoring="accuracy")
  [0, 1]')
grid_search.fit(x_train,y_train)

```

[illegible]

lab research google.com/drive/1CiN97ilKD7TYi.ISQNhKTiBms4lW7zH3W#scrollTo=34DI_vfmC9Po8&printMode=true 20/22

```

/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.
warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.
warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.
grid_search.best_params_

{'criterion': 'gini',
 'max_depth': 3,
 'max_features': 'auto',
 'splitter': 'random'}

/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.
dtc_cv=DecisionTreeClassifier(criterion='entropy',
                             max_depth=3,
                             max_features='sqrt',
                             splitter='best')
dtc_cv.fit(x_train,y_train)

```

```

DecisionTreeClassifier
DecisionTreeClassifier(criterion='entropy', max_depth=3, max_features='sqrt')

```

```

estimator: DecisionTreeClassifier
pred=dtc_cv.predict(x_test)

```

```

#label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
y=le.fit_transform(y)
#label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
pred=le.fit_transform(pred)

```

```
print(classification_report(y_test,pred))
```

	precision	recall	f1-score	support
0	0.85	0.97	0.90	245
1	0.43	0.12	0.19	49
accuracy			0.83	294
macro avg	0.64	0.54	0.55	294
weighted avg	0.78	0.83	0.78	294

RandomForestClassifier

```

from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()

forest_params = [{'max_depth': list(range(10, 15)), 'max_features': list(range(0,14))}]

rfc_cv= GridSearchCV(rfc,param_grid=forest_params,cv=10,scoring="accuracy")

rfc_cv.fit(x_train,y_train)

```

```
/usr/local/lib/python3.10/dist-packages/sklearn/model_selection/_validation.py:378: FitFailedWarning:
50 fits failed out of a total of 700.
The score on these train-test partitions for these parameters will be set to nan.
If these failures are not expected, you can try to debug them by setting error_score='raise'.
```

Below are more details about the failures:

50 fits failed with the following error:

Traceback (most recent call last):

```
File "/usr/local/lib/python3.10/dist-packages/sklearn/model_selection/_validation.py", line 686, in _fit_and_score
    estimator.fit(X_train, y_train, **fit_params)
File "/usr/local/lib/python3.10/dist-packages/sklearn/ensemble/_forest.py", line 340, in fit
    self._validate_params()
File "/usr/local/lib/python3.10/dist-packages/sklearn/base.py", line 600, in _validate_params
    validate_parameter_constraints(
File "/usr/local/lib/python3.10/dist-packages/sklearn/utils/_param_validation.py", line 97, in validate_parameter_constraints
    raise InvalidParameterError(
```

```
pred=rfc_cv.predict(x_test)
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/model_selection/_search.py:853: UserWarning: One or more of the test scores are nan.
```

```
#label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
y=le.fit_transform(y)
#label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
pred=le.fit_transform(pred)
0.85458496 0.85112009 0.80301403 0.85884398 0.85910592 0.85198928

print(classification_report(y_test,pred))
```

	precision	recall	f1-score	support
0	0.85	0.98	0.91	245
1	0.67	0.16	0.26	49
accuracy			0.85	294
macro avg	0.76	0.57	0.59	294
weighted avg	0.82	0.85	0.81	294

```
rfc_cv.best_params_
```

```
{'max_depth': 12, 'max_features': 9}
```