

# assignment-4

21BAI10418

```
[187]: #Import the Libraries.  
import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
import seaborn as sns
```

```
[188]: df=pd.read_csv("WA_Fn-UseC_-HR-Employee-Attrition.csv")
```

```
[189]: df.shape
```

```
[189]: (1470, 35)
```

```
[190]: df.head()
```

```
[190]:   Age Attrition   BusinessTravel   DailyRate   Department \  
0    41      Yes   Travel_Rarely    1102      Sales  
1    49      No  Travel_Frequently    279  Research & Development  
2    37      Yes   Travel_Rarely    1373  Research & Development  
3    33      No  Travel_Frequently    1392  Research & Development  
4    27      No   Travel_Rarely    591   Research & Development  
  
   DistanceFromHome   Education   EducationField   EmployeeCount   EmployeeNumber \  
0                1           2   Life Sciences           1           1  
1                8           1   Life Sciences           1           2  
2                2           2         Other           1           4  
3                3           4   Life Sciences           1           5  
4                2           1         Medical           1           7  
  
   ... RelationshipSatisfaction   StandardHours   StockOptionLevel \  
0   ...                1           80           0  
1   ...                4           80           1  
2   ...                2           80           0  
3   ...                3           80           0  
4   ...                4           80           1  
  
   TotalWorkingYears   TrainingTimesLastYear   WorkLifeBalance   YearsAtCompany \  
0                8                0                1                6
```

1	10	3	3	10
2	7	3	3	0
3	8	3	3	8
4	6	3	3	2

	YearsInCurrentRole	YearsSinceLastPromotion	YearsWithCurrManager
0	4	0	5
1	7	1	7
2	0	0	0
3	7	3	0
4	2	2	2

[5 rows x 35 columns]

```
[191]: 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

```
[192]: df.describe()
```

```

[192]:
      Age  DailyRate  DistanceFromHome  Education  EmployeeCount  \
count  1470.000000  1470.000000      1470.000000  1470.000000      1470.0
mean    36.923810   802.485714         9.192517    2.912925         1.0
std      9.135373   403.509100         8.106864    1.024165         0.0
min     18.000000   102.000000         1.000000    1.000000         1.0
25%     30.000000   465.000000         2.000000    2.000000         1.0
50%     36.000000   802.000000         7.000000    3.000000         1.0
75%     43.000000  1157.000000        14.000000    4.000000         1.0
max     60.000000  1499.000000        29.000000    5.000000         1.0

```

```

      EmployeeNumber  EnvironmentSatisfaction  HourlyRate  JobInvolvement  \
count    1470.000000          1470.000000  1470.000000  1470.000000
mean    1024.865306           2.721769    65.891156    2.729932
std      602.024335           1.093082   20.329428    0.711561
min         1.000000           1.000000   30.000000    1.000000
25%      491.250000           2.000000   48.000000    2.000000
50%     1020.500000           3.000000   66.000000    3.000000
75%     1555.750000           4.000000   83.750000    3.000000
max     2068.000000           4.000000  100.000000    4.000000

```

```

      JobLevel  ...  RelationshipSatisfaction  StandardHours  \
count  1470.000000  ...          1470.000000      1470.0
mean     2.063946  ...           2.712245        80.0
std      1.106940  ...           1.081209         0.0
min      1.000000  ...           1.000000        80.0
25%      1.000000  ...           2.000000        80.0
50%      2.000000  ...           3.000000        80.0
75%      3.000000  ...           4.000000        80.0
max      5.000000  ...           4.000000        80.0

```

```

      StockOptionLevel  TotalWorkingYears  TrainingTimesLastYear  \
count    1470.000000          1470.000000      1470.000000
mean         0.793878          11.279592         2.799320

```

std	0.852077	7.780782	1.289271
min	0.000000	0.000000	0.000000
25%	0.000000	6.000000	2.000000
50%	1.000000	10.000000	3.000000
75%	1.000000	15.000000	3.000000
max	3.000000	40.000000	6.000000

	WorkLifeBalance	YearsAtCompany	YearsInCurrentRole \
count	1470.000000	1470.000000	1470.000000
mean	2.761224	7.008163	4.229252
std	0.706476	6.126525	3.623137
min	1.000000	0.000000	0.000000
25%	2.000000	3.000000	2.000000
50%	3.000000	5.000000	3.000000
75%	3.000000	9.000000	7.000000
max	4.000000	40.000000	18.000000

	YearsSinceLastPromotion	YearsWithCurrManager
count	1470.000000	1470.000000
mean	2.187755	4.123129
std	3.222430	3.568136
min	0.000000	0.000000
25%	0.000000	2.000000
50%	1.000000	3.000000
75%	3.000000	7.000000
max	15.000000	17.000000

[8 rows x 26 columns]

## 1 Dropping the Unnesscary Colums

```
[193]: df.drop(columns=['EmployeeCount', 'Over18', 'StandardHours', 'EmployeeNumber'],
    ↪inplace=True)
df.shape
```

[193]: (1470, 31)

## 2 Checking and hanldeling null values

```
[194]: df.isnull().any()
```

```
[194]: Age                False
Attrition                False
BusinessTravel           False
DailyRate                False
```

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

```
[195]: print(df.isna().sum())
       '''Therefore There are no null values Present in the dataset'''
```

Age	0
Attrition	0
BusinessTravel	0
DailyRate	0
Department	0
DistanceFromHome	0
Education	0
EducationField	0
EnvironmentSatisfaction	0
Gender	0
HourlyRate	0
JobInvolvement	0
JobLevel	0
JobRole	0
JobSatisfaction	0

```

MaritalStatus      0
MonthlyIncome      0
MonthlyRate        0
NumCompaniesWorked 0
OverTime           0
PercentSalaryHike   0
PerformanceRating   0
RelationshipSatisfaction 0
StockOptionLevel    0
TotalWorkingYears   0
TrainingTimesLastYear 0
WorkLifeBalance     0
YearsAtCompany      0
YearsInCurrentRole   0
YearsSinceLastPromotion 0
YearsWithCurrManager 0
dtype: int64

```

[195]: 'Therefore There are no null values Present in the dataset'

```

[196]: #check for Duplicates
print(df.duplicated().sum())
'''Therefore There are no duplicate values Present in the dataset'''

```

0

[196]: 'Therefore There are no duplicate values Present in the dataset'

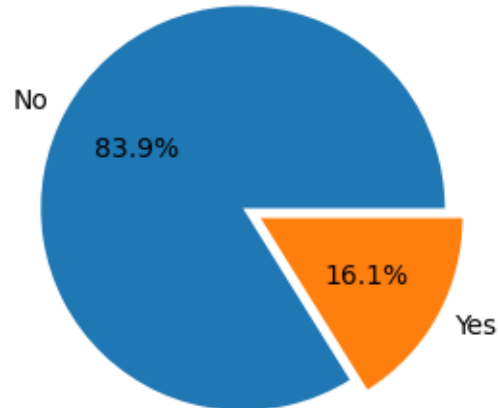
### 3 Data Visualization

```

[197]: department_counts = df['Attrition'].value_counts()
plt.figure(figsize=(4, 3))
Explode=[0.1,0]
plt.pie(department_counts, labels=department_counts.
    ↪index,explode=Explode,autopct='%1.1f%%')
plt.title('Distribution of Employees by Department')
plt.axis('equal')
plt.show()

```

Distribution of Employees by Department



```
[198]: df.columns
```

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

```
[199]: columns=['Attrition', 'BusinessTravel', 'Department', 'Education',
               ↪ 'EducationField', 'EnvironmentSatisfaction',
               ↪ 'Gender', 'JobInvolvement', 'JobLevel', 'JobRole', 'JobSatisfaction',
               ↪ 'MaritalStatus', 'NumCompaniesWorked', 'OverTime', 'PercentSalaryHike',
               ↪ 'PerformanceRating', 'RelationshipSatisfaction', 'StockOptionLevel',
               ↪ 'TrainingTimesLastYear', 'WorkLifeBalance']
```

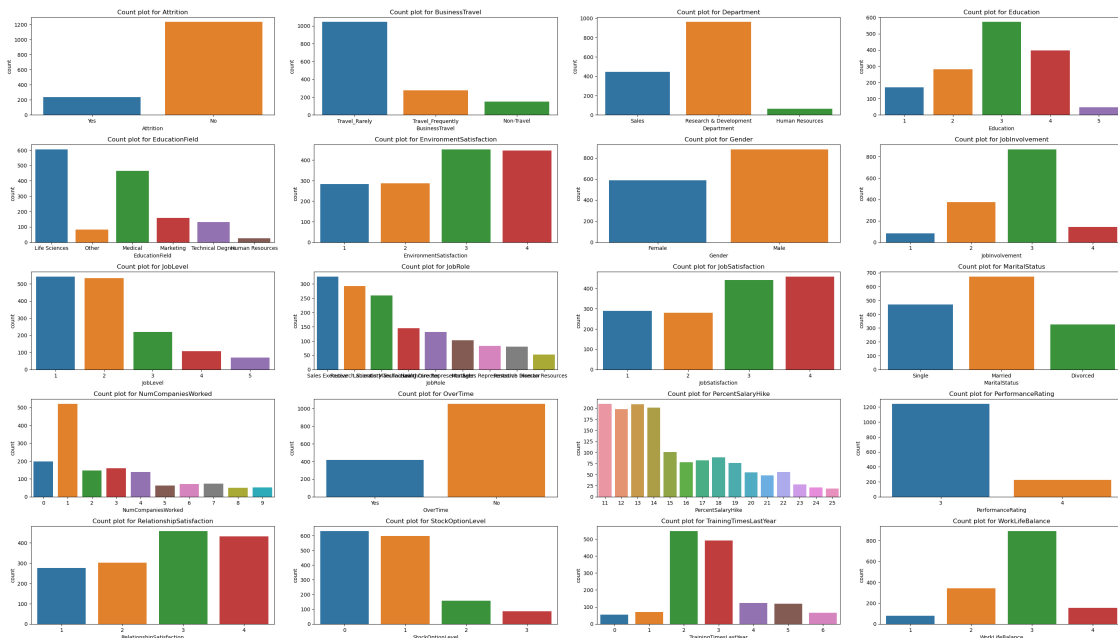
```
[200]: #visualizing the categorical attributes

# Create subplots for count plots
plt.figure(figsize=(30, 20)) # Adjust the figure size

for i, column in enumerate(columns, start=1):
    plt.subplot(6, 4, i) # Adjust the subplot layout (3 rows, 4 columns)
    sns.countplot(data=df, x=column)
    plt.title(f'Count plot for {column}')
```

```
# Adjust layout
plt.tight_layout()
```

```
# Show the plots
plt.show()
```



```
[201]: #visualizing the continous distrubution attributes
dist_columns = ['Age', 'MonthlyIncome', 'DailyRate', 'HourlyRate', '
↳ 'MonthlyRate', 'TotalWorkingYears', 'YearsAtCompany', 'YearsInCurrentRole', '
↳ 'YearsSinceLastPromotion', 'YearsWithCurrManager']
```

```
[202]: #visualizing the distribution of numerical attributes
```

```
plt.figure(figsize=(30, 20))
for i, column in enumerate(dist_columns, start=1):
    plt.subplot(4, 3, i)
    sns.distplot(df[column], kde=True)
    plt.title(f'Distribution of {column}')
plt.tight_layout()
plt.show()
```

C:\Users\adhar\AppData\Local\Temp\ipykernel\_1760\1121139151.py:6: 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[column],kde=True)
```

C:\Users\adhar\AppData\Local\Temp\ipykernel\_1760\1121139151.py:6: 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[column],kde=True)
```

C:\Users\adhar\AppData\Local\Temp\ipykernel\_1760\1121139151.py:6: 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[column],kde=True)
```

C:\Users\adhar\AppData\Local\Temp\ipykernel\_1760\1121139151.py:6: 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[column],kde=True)
```

C:\Users\adhar\AppData\Local\Temp\ipykernel\_1760\1121139151.py:6: 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[column],kde=True)
C:\Users\adhar\AppData\Local\Temp\ipykernel_1760\1121139151.py:6: 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[column],kde=True)
C:\Users\adhar\AppData\Local\Temp\ipykernel_1760\1121139151.py:6: 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[column],kde=True)
C:\Users\adhar\AppData\Local\Temp\ipykernel_1760\1121139151.py:6: 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[column],kde=True)
C:\Users\adhar\AppData\Local\Temp\ipykernel_1760\1121139151.py:6: 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>

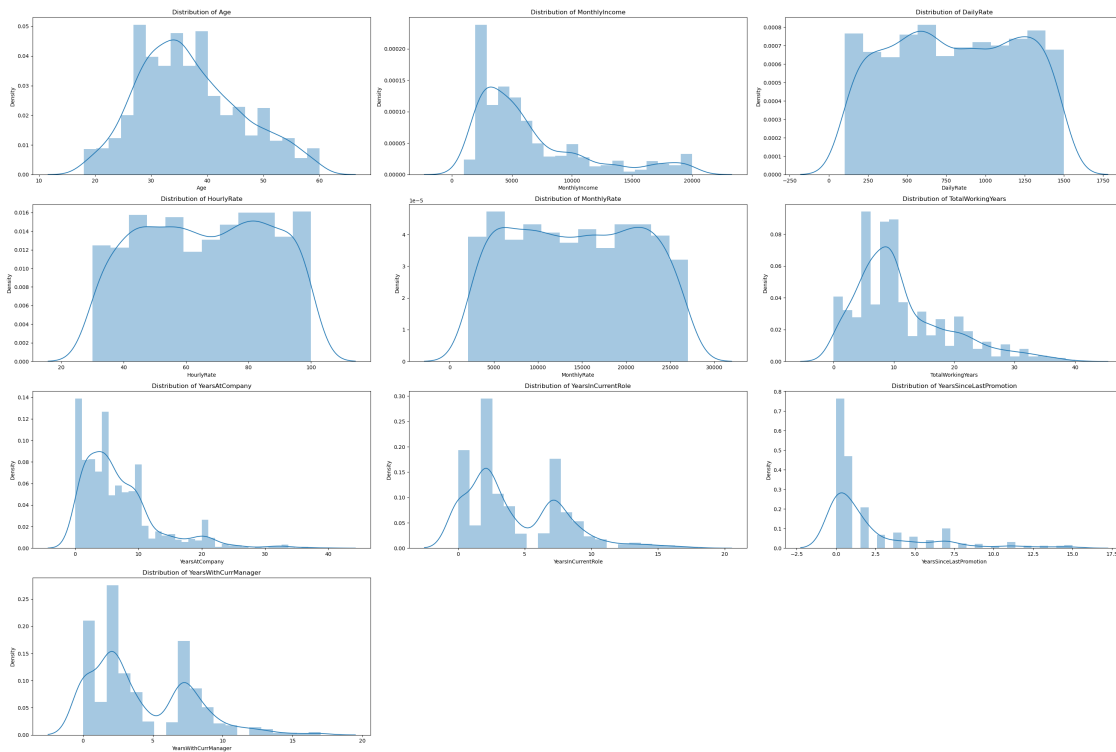
```
sns.distplot(df[column],kde=True)
C:\Users\adhar\AppData\Local\Temp\ipykernel_1760\1121139151.py:6: UserWarning:
```

`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[column],kde=True)
```



```
[203]: df.corr()
```

C:\Users\adhar\AppData\Local\Temp\ipykernel\_1760\1134722465.py:1: FutureWarning: The default value of `numeric_only` in `DataFrame.corr` is deprecated. In a future version, it will default to `False`. Select only valid columns or specify the value of `numeric_only` to silence this warning.

```
df.corr()
```

```
[203]:
```

	Age	DailyRate	DistanceFromHome	Education	\
Age	1.000000	0.010661	-0.001686	0.208034	
DailyRate	0.010661	1.000000	-0.004985	-0.016806	
DistanceFromHome	-0.001686	-0.004985	1.000000	0.021042	
Education	0.208034	-0.016806	0.021042	1.000000	
EnvironmentSatisfaction	0.010146	0.018355	-0.016075	-0.027128	
HourlyRate	0.024287	0.023381	0.031131	0.016775	

JobInvolvement	0.029820	0.046135	0.008783	0.042438
JobLevel	0.509604	0.002966	0.005303	0.101589
JobSatisfaction	-0.004892	0.030571	-0.003669	-0.011296
MonthlyIncome	0.497855	0.007707	-0.017014	0.094961
MonthlyRate	0.028051	-0.032182	0.027473	-0.026084
NumCompaniesWorked	0.299635	0.038153	-0.029251	0.126317
PercentSalaryHike	0.003634	0.022704	0.040235	-0.011111
PerformanceRating	0.001904	0.000473	0.027110	-0.024539
RelationshipSatisfaction	0.053535	0.007846	0.006557	-0.009118
StockOptionLevel	0.037510	0.042143	0.044872	0.018422
TotalWorkingYears	0.680381	0.014515	0.004628	0.148280
TrainingTimesLastYear	-0.019621	0.002453	-0.036942	-0.025100
WorkLifeBalance	-0.021490	-0.037848	-0.026556	0.009819
YearsAtCompany	0.311309	-0.034055	0.009508	0.069114
YearsInCurrentRole	0.212901	0.009932	0.018845	0.060236
YearsSinceLastPromotion	0.216513	-0.033229	0.010029	0.054254
YearsWithCurrManager	0.202089	-0.026363	0.014406	0.069065

	EnvironmentSatisfaction	HourlyRate	JobInvolvement	\
Age	0.010146	0.024287	0.029820	
DailyRate	0.018355	0.023381	0.046135	
DistanceFromHome	-0.016075	0.031131	0.008783	
Education	-0.027128	0.016775	0.042438	
EnvironmentSatisfaction	1.000000	-0.049857	-0.008278	
HourlyRate	-0.049857	1.000000	0.042861	
JobInvolvement	-0.008278	0.042861	1.000000	
JobLevel	0.001212	-0.027853	-0.012630	
JobSatisfaction	-0.006784	-0.071335	-0.021476	
MonthlyIncome	-0.006259	-0.015794	-0.015271	
MonthlyRate	0.037600	-0.015297	-0.016322	
NumCompaniesWorked	0.012594	0.022157	0.015012	
PercentSalaryHike	-0.031701	-0.009062	-0.017205	
PerformanceRating	-0.029548	-0.002172	-0.029071	
RelationshipSatisfaction	0.007665	0.001330	0.034297	
StockOptionLevel	0.003432	0.050263	0.021523	
TotalWorkingYears	-0.002693	-0.002334	-0.005533	
TrainingTimesLastYear	-0.019359	-0.008548	-0.015338	
WorkLifeBalance	0.027627	-0.004607	-0.014617	
YearsAtCompany	0.001458	-0.019582	-0.021355	
YearsInCurrentRole	0.018007	-0.024106	0.008717	
YearsSinceLastPromotion	0.016194	-0.026716	-0.024184	
YearsWithCurrManager	-0.004999	-0.020123	0.025976	

	JobLevel	JobSatisfaction	MonthlyIncome	...	\
Age	0.509604	-0.004892	0.497855	...	
DailyRate	0.002966	0.030571	0.007707	...	
DistanceFromHome	0.005303	-0.003669	-0.017014	...	

Education	0.101589	-0.011296	0.094961	...
EnvironmentSatisfaction	0.001212	-0.006784	-0.006259	...
HourlyRate	-0.027853	-0.071335	-0.015794	...
JobInvolvement	-0.012630	-0.021476	-0.015271	...
JobLevel	1.000000	-0.001944	0.950300	...
JobSatisfaction	-0.001944	1.000000	-0.007157	...
MonthlyIncome	0.950300	-0.007157	1.000000	...
MonthlyRate	0.039563	0.000644	0.034814	...
NumCompaniesWorked	0.142501	-0.055699	0.149515	...
PercentSalaryHike	-0.034730	0.020002	-0.027269	...
PerformanceRating	-0.021222	0.002297	-0.017120	...
RelationshipSatisfaction	0.021642	-0.012454	0.025873	...
StockOptionLevel	0.013984	0.010690	0.005408	...
TotalWorkingYears	0.782208	-0.020185	0.772893	...
TrainingTimesLastYear	-0.018191	-0.005779	-0.021736	...
WorkLifeBalance	0.037818	-0.019459	0.030683	...
YearsAtCompany	0.534739	-0.003803	0.514285	...
YearsInCurrentRole	0.389447	-0.002305	0.363818	...
YearsSinceLastPromotion	0.353885	-0.018214	0.344978	...
YearsWithCurrManager	0.375281	-0.027656	0.344079	...

	PerformanceRating	RelationshipSatisfaction \
Age	0.001904	0.053535
DailyRate	0.000473	0.007846
DistanceFromHome	0.027110	0.006557
Education	-0.024539	-0.009118
EnvironmentSatisfaction	-0.029548	0.007665
HourlyRate	-0.002172	0.001330
JobInvolvement	-0.029071	0.034297
JobLevel	-0.021222	0.021642
JobSatisfaction	0.002297	-0.012454
MonthlyIncome	-0.017120	0.025873
MonthlyRate	-0.009811	-0.004085
NumCompaniesWorked	-0.014095	0.052733
PercentSalaryHike	0.773550	-0.040490
PerformanceRating	1.000000	-0.031351
RelationshipSatisfaction	-0.031351	1.000000
StockOptionLevel	0.003506	-0.045952
TotalWorkingYears	0.006744	0.024054
TrainingTimesLastYear	-0.015579	0.002497
WorkLifeBalance	0.002572	0.019604
YearsAtCompany	0.003435	0.019367
YearsInCurrentRole	0.034986	-0.015123
YearsSinceLastPromotion	0.017896	0.033493
YearsWithCurrManager	0.022827	-0.000867

StockOptionLevel TotalWorkingYears \

Age	0.037510	0.680381
DailyRate	0.042143	0.014515
DistanceFromHome	0.044872	0.004628
Education	0.018422	0.148280
EnvironmentSatisfaction	0.003432	-0.002693
HourlyRate	0.050263	-0.002334
JobInvolvement	0.021523	-0.005533
JobLevel	0.013984	0.782208
JobSatisfaction	0.010690	-0.020185
MonthlyIncome	0.005408	0.772893
MonthlyRate	-0.034323	0.026442
NumCompaniesWorked	0.030075	0.237639
PercentSalaryHike	0.007528	-0.020608
PerformanceRating	0.003506	0.006744
RelationshipSatisfaction	-0.045952	0.024054
StockOptionLevel	1.000000	0.010136
TotalWorkingYears	0.010136	1.000000
TrainingTimesLastYear	0.011274	-0.035662
WorkLifeBalance	0.004129	0.001008
YearsAtCompany	0.015058	0.628133
YearsInCurrentRole	0.050818	0.460365
YearsSinceLastPromotion	0.014352	0.404858
YearsWithCurrManager	0.024698	0.459188

	TrainingTimesLastYear	WorkLifeBalance \
Age	-0.019621	-0.021490
DailyRate	0.002453	-0.037848
DistanceFromHome	-0.036942	-0.026556
Education	-0.025100	0.009819
EnvironmentSatisfaction	-0.019359	0.027627
HourlyRate	-0.008548	-0.004607
JobInvolvement	-0.015338	-0.014617
JobLevel	-0.018191	0.037818
JobSatisfaction	-0.005779	-0.019459
MonthlyIncome	-0.021736	0.030683
MonthlyRate	0.001467	0.007963
NumCompaniesWorked	-0.066054	-0.008366
PercentSalaryHike	-0.005221	-0.003280
PerformanceRating	-0.015579	0.002572
RelationshipSatisfaction	0.002497	0.019604
StockOptionLevel	0.011274	0.004129
TotalWorkingYears	-0.035662	0.001008
TrainingTimesLastYear	1.000000	0.028072
WorkLifeBalance	0.028072	1.000000
YearsAtCompany	0.003569	0.012089
YearsInCurrentRole	-0.005738	0.049856
YearsSinceLastPromotion	-0.002067	0.008941

YearsWithCurrManager	-0.004096	0.002759
----------------------	-----------	----------

	YearsAtCompany	YearsInCurrentRole \
Age	0.311309	0.212901
DailyRate	-0.034055	0.009932
DistanceFromHome	0.009508	0.018845
Education	0.069114	0.060236
EnvironmentSatisfaction	0.001458	0.018007
HourlyRate	-0.019582	-0.024106
JobInvolvement	-0.021355	0.008717
JobLevel	0.534739	0.389447
JobSatisfaction	-0.003803	-0.002305
MonthlyIncome	0.514285	0.363818
MonthlyRate	-0.023655	-0.012815
NumCompaniesWorked	-0.118421	-0.090754
PercentSalaryHike	-0.035991	-0.001520
PerformanceRating	0.003435	0.034986
RelationshipSatisfaction	0.019367	-0.015123
StockOptionLevel	0.015058	0.050818
TotalWorkingYears	0.628133	0.460365
TrainingTimesLastYear	0.003569	-0.005738
WorkLifeBalance	0.012089	0.049856
YearsAtCompany	1.000000	0.758754
YearsInCurrentRole	0.758754	1.000000
YearsSinceLastPromotion	0.618409	0.548056
YearsWithCurrManager	0.769212	0.714365

	YearsSinceLastPromotion	YearsWithCurrManager
Age	0.216513	0.202089
DailyRate	-0.033229	-0.026363
DistanceFromHome	0.010029	0.014406
Education	0.054254	0.069065
EnvironmentSatisfaction	0.016194	-0.004999
HourlyRate	-0.026716	-0.020123
JobInvolvement	-0.024184	0.025976
JobLevel	0.353885	0.375281
JobSatisfaction	-0.018214	-0.027656
MonthlyIncome	0.344978	0.344079
MonthlyRate	0.001567	-0.036746
NumCompaniesWorked	-0.036814	-0.110319
PercentSalaryHike	-0.022154	-0.011985
PerformanceRating	0.017896	0.022827
RelationshipSatisfaction	0.033493	-0.000867
StockOptionLevel	0.014352	0.024698
TotalWorkingYears	0.404858	0.459188
TrainingTimesLastYear	-0.002067	-0.004096
WorkLifeBalance	0.008941	0.002759

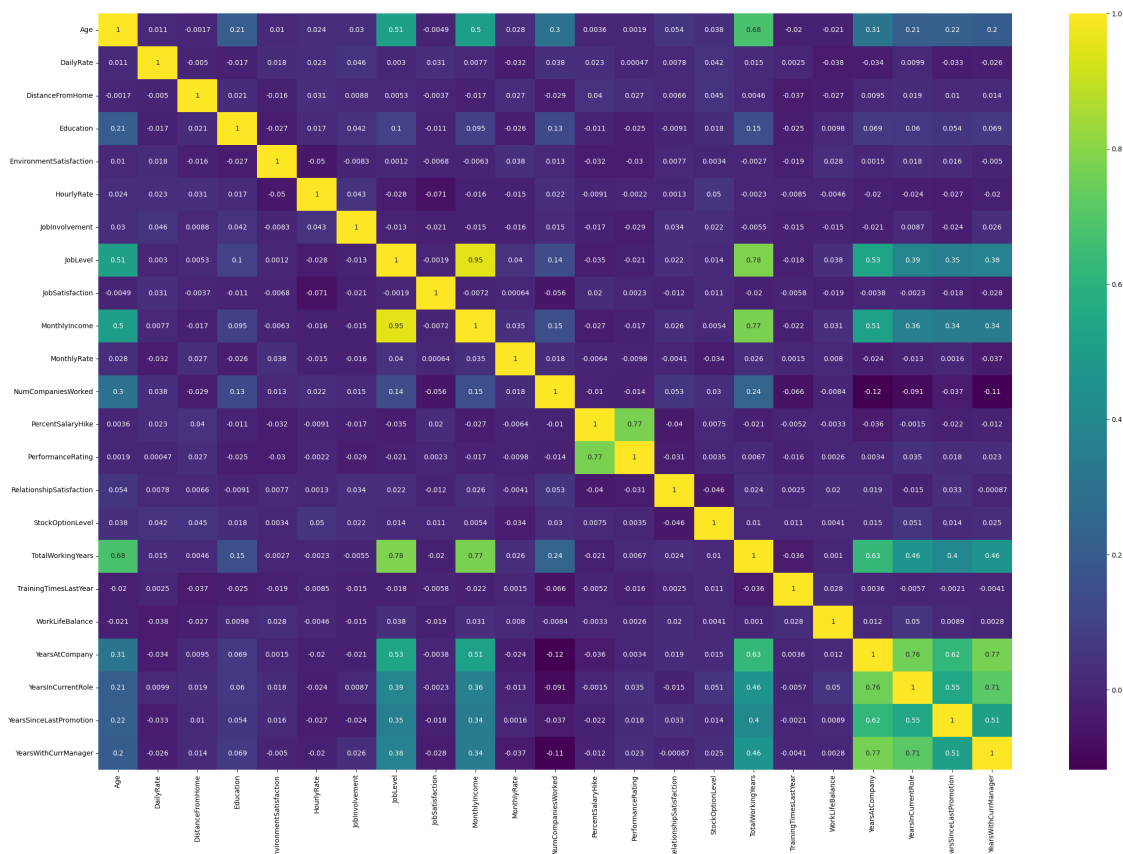
YearsAtCompany	0.618409	0.769212
YearsInCurrentRole	0.548056	0.714365
YearsSinceLastPromotion	1.000000	0.510224
YearsWithCurrManager	0.510224	1.000000

[23 rows x 23 columns]

```
[204]: # Heat maps based on the correlation between all the variables
plt.figure(figsize=(30, 20))
sns.heatmap(df.corr(),annot=True,cmap="viridis")
plt.show()
```

C:\Users\adhar\AppData\Local\Temp\ipykernel\_1760\3278965269.py:3: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning.

```
sns.heatmap(df.corr(),annot=True,cmap="viridis")
```



```
[205]: #encoding Attrition
from sklearn.preprocessing import LabelEncoder
```



```
le=LabelEncoder()
df["Attrition"]=le.fit_transform(df["Attrition"])
print(le.classes_)
```

```
['No' 'Yes']
```

```
[206]: mapping=dict(zip(le.classes_,range(len(le.classes_))))
mapping
```

```
[206]: {'No': 0, 'Yes': 1}
```

## 4 Splitting Dependent and Independent variables

```
[207]: df.head()
```

```
[207]:
```

	Age	Attrition	BusinessTravel	DailyRate	Department	\
0	41	1	Travel_Rarely	1102	Sales	
1	49	0	Travel_Frequently	279	Research & Development	
2	37	1	Travel_Rarely	1373	Research & Development	
3	33	0	Travel_Frequently	1392	Research & Development	
4	27	0	Travel_Rarely	591	Research & Development	

	DistanceFromHome	Education	EducationField	EnvironmentSatisfaction	\
0	1	2	Life Sciences	2	
1	8	1	Life Sciences	3	
2	2	2	Other	4	
3	3	4	Life Sciences	4	
4	2	1	Medical	1	

	Gender	...	PerformanceRating	RelationshipSatisfaction	StockOptionLevel	\
0	Female	...	3	1	0	
1	Male	...	4	4	1	
2	Male	...	3	2	0	
3	Female	...	3	3	0	
4	Male	...	3	4	1	

	TotalWorkingYears	TrainingTimesLastYear	WorkLifeBalance	YearsAtCompany	\
0	8	0	1	6	
1	10	3	3	10	
2	7	3	3	0	
3	8	3	3	8	
4	6	3	3	2	

	YearsInCurrentRole	YearsSinceLastPromotion	YearsWithCurrManager
0	4	0	5
1	7	1	7

2	0	0	0
3	7	3	0
4	2	2	2

[5 rows x 31 columns]

```
[208]: X=df.drop(columns=["Attrition"],axis=1)
X.head()
```

```
[208]:   Age      BusinessTravel  DailyRate      Department \
0   41      Travel_Rarely      1102      Sales
1   49  Travel_Frequently      279  Research & Development
2   37      Travel_Rarely      1373  Research & Development
3   33  Travel_Frequently      1392  Research & Development
4   27      Travel_Rarely      591  Research & Development

      DistanceFromHome  Education  EducationField  EnvironmentSatisfaction \
0                1          2  Life Sciences                2
1                8          1  Life Sciences                3
2                2          2          Other                4
3                3          4  Life Sciences                4
4                2          1          Medical                1

      Gender  HourlyRate  ...  PerformanceRating  RelationshipSatisfaction \
0  Female        94  ...                3                1
1   Male        61  ...                4                4
2   Male        92  ...                3                2
3  Female        56  ...                3                3
4   Male        40  ...                3                4

      StockOptionLevel  TotalWorkingYears  TrainingTimesLastYear  WorkLifeBalance \
0                0                8                0                1
1                1               10                3                3
2                0                7                3                3
3                0                8                3                3
4                1                6                3                3

      YearsAtCompany  YearsInCurrentRole  YearsSinceLastPromotion \
0                6                4                0
1               10                7                1
2                0                0                0
3                8                7                3
4                2                2                2

      YearsWithCurrManager
0                5
1                7
```

```

2          0
3          0
4          2

```

[5 rows x 30 columns]

```

[209]: print(X.shape,X.info())
       print(type(X))

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1470 entries, 0 to 1469
Data columns (total 30 columns):
 #   Column                                Non-Null Count  Dtype
---  -
 0   Age                                  1470 non-null   int64
 1   BusinessTravel                      1470 non-null   object
 2   DailyRate                           1470 non-null   int64
 3   Department                          1470 non-null   object
 4   DistanceFromHome                   1470 non-null   int64
 5   Education                           1470 non-null   int64
 6   EducationField                      1470 non-null   object
 7   EnvironmentSatisfaction             1470 non-null   int64
 8   Gender                              1470 non-null   object
 9   HourlyRate                          1470 non-null   int64
10   JobInvolvement                     1470 non-null   int64
11   JobLevel                           1470 non-null   int64
12   JobRole                             1470 non-null   object
13   JobSatisfaction                    1470 non-null   int64
14   MaritalStatus                      1470 non-null   object
15   MonthlyIncome                      1470 non-null   int64
16   MonthlyRate                        1470 non-null   int64
17   NumCompaniesWorked                 1470 non-null   int64
18   OverTime                           1470 non-null   object
19   PercentSalaryHike                  1470 non-null   int64
20   PerformanceRating                  1470 non-null   int64
21   RelationshipSatisfaction            1470 non-null   int64
22   StockOptionLevel                   1470 non-null   int64
23   TotalWorkingYears                  1470 non-null   int64
24   TrainingTimesLastYear              1470 non-null   int64
25   WorkLifeBalance                    1470 non-null   int64
26   YearsAtCompany                     1470 non-null   int64
27   YearsInCurrentRole                 1470 non-null   int64
28   YearsSinceLastPromotion             1470 non-null   int64
29   YearsWithCurrManager                1470 non-null   int64
dtypes: int64(23), object(7)
memory usage: 344.7+ KB
(1470, 30) None

```

```
<class 'pandas.core.frame.DataFrame'>
```

```
[210]: y=df["Attrition"]  
y.head()
```

```
[210]: 0    1  
      1    0  
      2    1  
      3    0  
      4    0  
      Name: Attrition, dtype: int32
```

## 5 Lable Encoding

```
[211]: from sklearn.preprocessing import LabelEncoder  
le=LabelEncoder()
```

```
[212]: X["BusinessTravel"]=le.fit_transform(X["BusinessTravel"])  
print(le.classes_)  
m1=dict(zip(le.classes_,range(len(le.classes_))))  
print(m1)  
X["Department"]=le.fit_transform(X["Department"])  
print(le.classes_)  
m2=dict(zip(le.classes_,range(len(le.classes_))))  
print(m2)  
X["EducationField"]=le.fit_transform(X["EducationField"])  
print(le.classes_)  
m3=dict(zip(le.classes_,range(len(le.classes_))))  
print(m3)  
X["Gender"]=le.fit_transform(X["Gender"])  
print(le.classes_)  
m4=dict(zip(le.classes_,range(len(le.classes_))))  
print(m4)  
X["JobRole"]=le.fit_transform(X["JobRole"])  
print(le.classes_)  
m5=dict(zip(le.classes_,range(len(le.classes_))))  
print(m5)  
X["MaritalStatus"]=le.fit_transform(X["MaritalStatus"])  
print(le.classes_)  
m6=dict(zip(le.classes_,range(len(le.classes_))))  
print(m6)  
X["OverTime"]=le.fit_transform(X["OverTime"])  
print(le.classes_)  
m6=dict(zip(le.classes_,range(len(le.classes_))))  
print(m6)
```

```
['Non-Travel' 'Travel_Frequently' 'Travel_Rarely']
```

```
{'Non-Travel': 0, 'Travel_Frequently': 1, 'Travel_Rarely': 2}
['Human Resources' 'Research & Development' 'Sales']
{'Human Resources': 0, 'Research & Development': 1, 'Sales': 2}
['Human Resources' 'Life Sciences' 'Marketing' 'Medical' 'Other'
 'Technical Degree']
{'Human Resources': 0, 'Life Sciences': 1, 'Marketing': 2, 'Medical': 3,
 'Other': 4, 'Technical Degree': 5}
['Female' 'Male']
{'Female': 0, 'Male': 1}
['Healthcare Representative' 'Human Resources' 'Laboratory Technician'
 'Manager' 'Manufacturing Director' 'Research Director'
 'Research Scientist' 'Sales Executive' 'Sales Representative']
{'Healthcare Representative': 0, 'Human Resources': 1, 'Laboratory Technician':
 2, 'Manager': 3, 'Manufacturing Director': 4, 'Research Director': 5, 'Research
 Scientist': 6, 'Sales Executive': 7, 'Sales Representative': 8}
['Divorced' 'Married' 'Single']
{'Divorced': 0, 'Married': 1, 'Single': 2}
['No' 'Yes']
{'No': 0, 'Yes': 1}
```

```
[213]: X.head()
```

```
[213]:   Age  BusinessTravel  DailyRate  Department  DistanceFromHome  Education  \
0    41                2      1102           2                1          2
1    49                1       279           1                8          1
2    37                2     1373           1                2          2
3    33                1     1392           1                3          4
4    27                2       591           1                2          1

   EducationField  EnvironmentSatisfaction  Gender  HourlyRate  ...  \
0                1                      2        0         94  ...
1                1                      3        1         61  ...
2                4                      4        1         92  ...
3                1                      4        0         56  ...
4                3                      1        1         40  ...

   PerformanceRating  RelationshipSatisfaction  StockOptionLevel  \
0                   3                        1                  0
1                   4                        4                  1
2                   3                        2                  0
3                   3                        3                  0
4                   3                        4                  1

   TotalWorkingYears  TrainingTimesLastYear  WorkLifeBalance  YearsAtCompany  \
0                   8                      0                  1              6
1                  10                      3                  3             10
2                   7                      3                  3              0
```

3	8	3	3	8
4	6	3	3	2

	YearsInCurrentRole	YearsSinceLastPromotion	YearsWithCurrManager
0	4	0	5
1	7	1	7
2	0	0	0
3	7	3	0
4	2	2	2

[5 rows x 30 columns]

```
[214]: y.head()
#all categories are encoded
```

```
[214]: 0    1
1    0
2    1
3    0
4    0
Name: Attrition, dtype: int32
```

## 6 Feature Scaling

```
[215]: #min max scaling from 0 to 1
from sklearn.preprocessing import MinMaxScaler
ms=MinMaxScaler()
```

```
[216]: X_Scaled=ms.fit_transform(X)
X_Scaled=pd.DataFrame(ms.fit_transform(X),columns=X.columns)
X_Scaled.head()
```

```
[216]:      Age  BusinessTravel  DailyRate  Department  DistanceFromHome  \
0  0.547619         1.0    0.715820          1.0         0.000000
1  0.738095         0.5    0.126700          0.5         0.250000
2  0.452381         1.0    0.909807          0.5         0.035714
3  0.357143         0.5    0.923407          0.5         0.071429
4  0.214286         1.0    0.350036          0.5         0.035714

      Education  EducationField  EnvironmentSatisfaction  Gender  HourlyRate  \
0         0.25             0.2             0.333333      0.0    0.914286
1         0.00             0.2             0.666667      1.0    0.442857
2         0.25             0.8             1.000000      1.0    0.885714
3         0.75             0.2             1.000000      0.0    0.371429
4         0.00             0.6             0.000000      1.0    0.142857
```

	...	PerformanceRating	RelationshipSatisfaction	StockOptionLevel	\
0	...	0.0	0.000000	0.000000	
1	...	1.0	1.000000	0.333333	
2	...	0.0	0.333333	0.000000	
3	...	0.0	0.666667	0.000000	
4	...	0.0	1.000000	0.333333	

	TotalWorkingYears	TrainingTimesLastYear	WorkLifeBalance	YearsAtCompany	\
0	0.200	0.0	0.000000	0.15	
1	0.250	0.5	0.666667	0.25	
2	0.175	0.5	0.666667	0.00	
3	0.200	0.5	0.666667	0.20	
4	0.150	0.5	0.666667	0.05	

	YearsInCurrentRole	YearsSinceLastPromotion	YearsWithCurrManager
0	0.222222	0.000000	0.294118
1	0.388889	0.066667	0.411765
2	0.000000	0.000000	0.000000
3	0.388889	0.200000	0.000000
4	0.111111	0.133333	0.117647

[5 rows x 30 columns]

## 7 Splitting DataSet into Train and Test¶

```
[217]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(X_Scaled,y,test_size =0.
↪2,random_state =0)
```

```
[218]: print(x_train.shape,x_test.shape,y_train.shape,y_test.shape)
```

```
(1176, 30) (294, 30) (1176,) (294,)
```

```
[219]: x_train.head()
```

```
[219]:
```

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	\
1374	0.952381	1.0	0.360057	1.0	0.714286	
1092	0.642857	1.0	0.607015	0.5	0.964286	
768	0.523810	1.0	0.141732	1.0	0.892857	
569	0.428571	0.0	0.953472	1.0	0.250000	
911	0.166667	0.5	0.355762	1.0	0.821429	

	Education	EducationField	EnvironmentSatisfaction	Gender	HourlyRate	\
1374	0.50	0.2	1.000000	0.0	0.600000	
1092	0.50	1.0	1.000000	1.0	0.957143	
768	0.50	0.4	0.666667	1.0	0.628571	

569	0.75	0.2	0.000000	1.0	0.657143
911	0.00	0.2	0.666667	1.0	0.614286

	...	PerformanceRating	RelationshipSatisfaction	StockOptionLevel	\
1374	...	0.0	0.666667	0.333333	
1092	...	1.0	1.000000	0.333333	
768	...	0.0	0.333333	0.333333	
569	...	0.0	0.333333	0.000000	
911	...	0.0	1.000000	0.000000	

	TotalWorkingYears	TrainingTimesLastYear	WorkLifeBalance	\
1374	0.725	0.333333	0.333333	
1092	0.200	0.500000	0.666667	
768	0.200	0.500000	0.333333	
569	0.250	0.166667	0.666667	
911	0.025	0.666667	0.666667	

	YearsAtCompany	YearsInCurrentRole	YearsSinceLastPromotion	\
1374	0.025	0.000000	0.000000	
1092	0.125	0.222222	0.000000	
768	0.175	0.388889	0.466667	
569	0.250	0.388889	0.000000	
911	0.025	0.000000	0.066667	

	YearsWithCurrManager
1374	0.000000
1092	0.176471
768	0.294118
569	0.529412
911	0.000000

[5 rows x 30 columns]

## 8 LogisticRegression Model Building

```
[220]: from sklearn.linear_model import LogisticRegression
model=LogisticRegression()
```

```
[221]: model.fit(x_train,y_train)
```

```
[221]: LogisticRegression()
```

```
[222]: pred=model.predict(x_test)
```

```
[223]: pred
```



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

```
[224]: y_test.head()
```

```
[224]: 442      0
      1091     0
      981      1
      785     0
      1332     1
      Name: Attrition, dtype: int32
```

## 9 Evaluation of classification model

```
[225]: #Accuracy score
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report, roc_auc_score, roc_curve
```

```
[226]: accuracy_score(y_test, pred)
```

```
[226]: 0.8809523809523809
```

```
[227]: confusion_matrix(y_test, pred)
```

```
[227]: array([[242,   3],
            [ 32,  17]], dtype=int64)
```

```
[228]: pd.crosstab(y_test, pred)
```

```
[228]: col_0      0      1
      Attrition
      0      242      3
      1      32     17
```

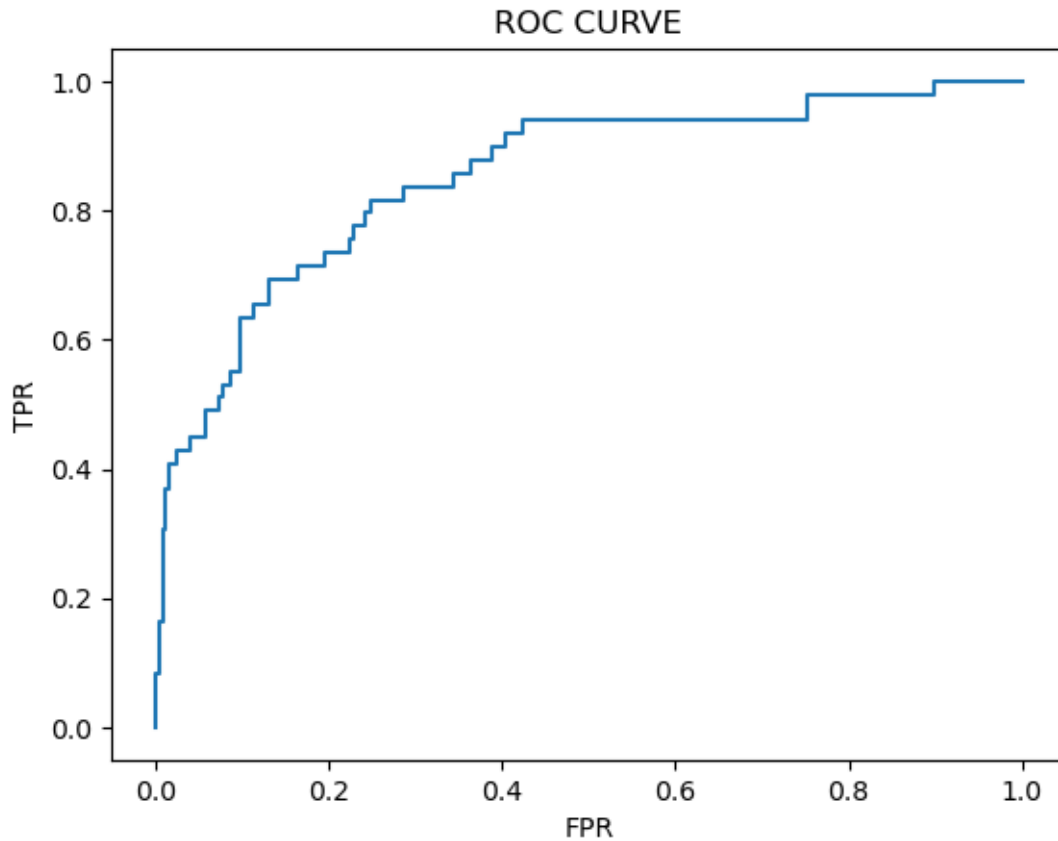
```
[229]: print(classification_report(y_test,pred))
```

	precision	recall	f1-score	support
0	0.88	0.99	0.93	245
1	0.85	0.35	0.49	49
accuracy			0.88	294
macro avg	0.87	0.67	0.71	294
weighted avg	0.88	0.88	0.86	294

## 10 ROC-AUC curve

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

```
[231]: # roc_curve
fpr,tpr,threshholds = roc_curve(y_test,probability)
plt.plot(fpr,tpr)
plt.xlabel('FPR')
plt.ylabel('TPR')
plt.title('ROC CURVE')
plt.show()
```



## 11 Decision Tree Model Building

```
[232]: from sklearn.tree import DecisionTreeClassifier
      dtc=DecisionTreeClassifier()
```

```
[233]: dtc.fit(x_train,y_train)
```

```
[233]: DecisionTreeClassifier()
```

```
[234]: pred=dtc.predict(x_test)
      pred
```

```
[234]: array([0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 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, 0,
            0, 1, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0,
            0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1,
            0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
            1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0,
            0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 0,
```

```

0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0,
0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1,
0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1,
0, 0, 1, 0, 0, 0, 0, 0, 0])

```

```

[235]: #Accuracy score
from sklearn.metrics import
    accuracy_score, confusion_matrix, classification_report, roc_auc_score, roc_curve
accuracy_score(y_test, pred)

```

```

[235]: 0.7517006802721088

```

```

[236]: confusion_matrix(y_test, pred)

```

```

[236]: array([[204,  41],
           [ 32,  17]], dtype=int64)

```

```

[237]: pd.crosstab(y_test, pred)

```

```

[237]: col_0      0      1
Attrition
0         204    41
1          32    17

```

```

[238]: print(classification_report(y_test, pred))

```

	precision	recall	f1-score	support
0	0.86	0.83	0.85	245
1	0.29	0.35	0.32	49
accuracy			0.75	294
macro avg	0.58	0.59	0.58	294
weighted avg	0.77	0.75	0.76	294

```

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

```

```

[239]: array([0., 0., 1., 0., 1., 0., 1., 0., 0., 0., 0., 0., 0., 1., 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., 0., 0., 1., 0., 0., 1., 1., 1.,
          1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 1., 0., 0., 0.,
          1., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0.,

```

```

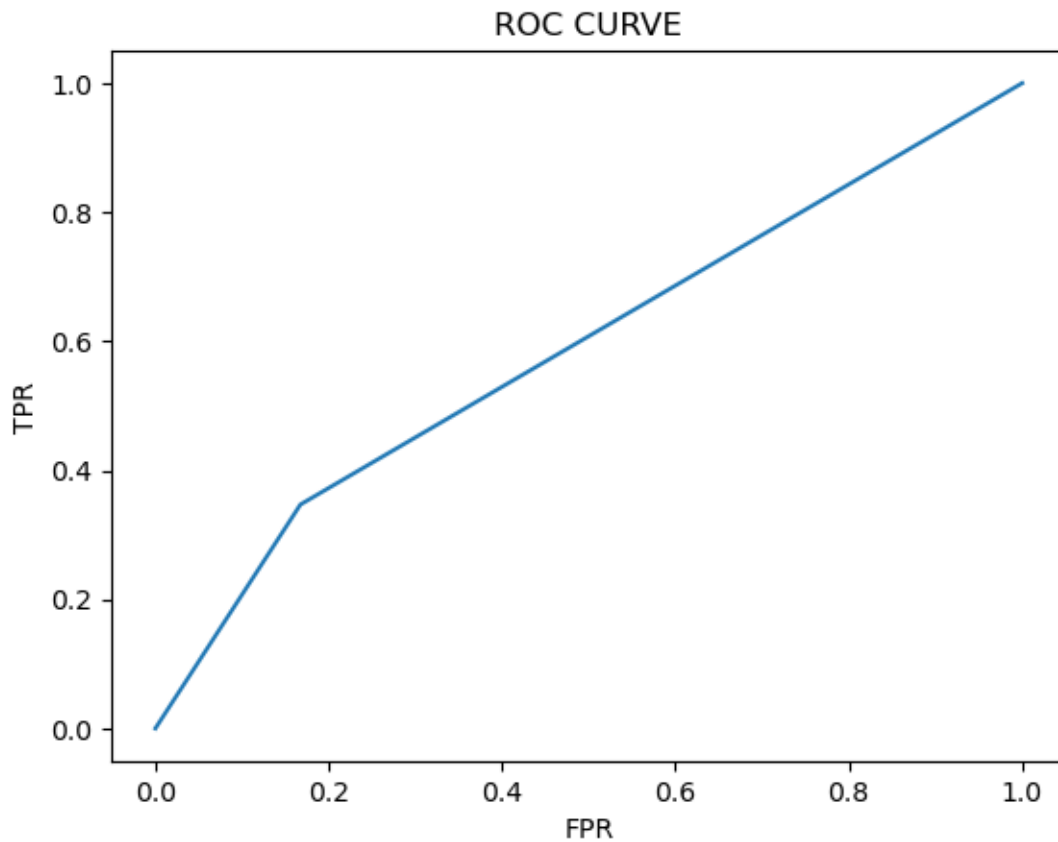
0., 1., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 1., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 1., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 0.,
0., 0., 1., 0., 0., 0., 1., 0., 1., 0., 0., 1., 0., 0., 0., 1., 1.,
0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 1., 1., 0., 0., 0., 0., 1., 0., 0., 1.,
0., 0., 1., 1., 1., 1., 1., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0.,
1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 1., 0., 0., 0., 0., 1., 1., 0., 0., 1., 0., 0., 0., 0., 0., 0.,
0., 1., 0., 0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0., 1., 0., 1.,
0., 0., 1., 0., 0., 0., 0., 0., 1., 0., 1., 0., 1., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 1., 0., 0., 1., 0., 0., 1., 0., 0., 1.,
0., 0., 0., 0., 0.]

```

```

[240]: # roc_curve
fpr,tpr,threshsholds = roc_curve(y_test,probability)
plt.plot(fpr,tpr)
plt.xlabel('FPR')
plt.ylabel('TPR')
plt.title('ROC CURVE')
plt.show()

```



## 12 Hyperparameter tuning

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

```
[241]: [Text(0.32688674428104575, 0.9722222222222222, 'x[23] <= 0.038\ngini =
0.269\nsamples = 1176\nvalue = [988, 188]'),
Text(0.0784313725490196, 0.9166666666666666, 'x[14] <= 0.75\ngini =
0.5\nsamples = 78\nvalue = [39, 39]'),
Text(0.049019607843137254, 0.8611111111111112, 'x[4] <= 0.554\ngini =
0.426\nsamples = 39\nvalue = [27, 12]'),
Text(0.032679738562091505, 0.8055555555555556, 'x[13] <= 0.167\ngini =
0.312\nsamples = 31\nvalue = [25, 6]'),
Text(0.0196078431372549, 0.75, 'x[18] <= 0.5\ngini = 0.49\nsamples = 7\nvalue =
[3, 4]'),
Text(0.013071895424836602, 0.6944444444444444, 'x[14] <= 0.25\ngini =
0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.006535947712418301, 0.6388888888888888, 'gini = 0.0\nsamples = 3\nvalue
= [3, 0]'),
Text(0.0196078431372549, 0.6388888888888888, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.026143790849673203, 0.6944444444444444, 'gini = 0.0\nsamples = 3\nvalue
= [0, 3]'),
Text(0.0457516339869281, 0.75, 'x[17] <= 0.056\ngini = 0.153\nsamples =
24\nvalue = [22, 2]'),
Text(0.0392156862745098, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.05228758169934641, 0.6944444444444444, 'x[7] <= 0.167\ngini =
0.083\nsamples = 23\nvalue = [22, 1]'),
Text(0.0457516339869281, 0.6388888888888888, 'x[13] <= 0.667\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.0392156862745098, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.05228758169934641, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.058823529411764705, 0.6388888888888888, 'gini = 0.0\nsamples = 21\nvalue
= [21, 0]'),
Text(0.06535947712418301, 0.8055555555555556, 'x[19] <= 0.679\ngini =
0.375\nsamples = 8\nvalue = [2, 6]'),
Text(0.058823529411764705, 0.75, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.0718954248366013, 0.75, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.10784313725490197, 0.8611111111111112, 'x[9] <= 0.364\ngini =
0.426\nsamples = 39\nvalue = [12, 27]'),
Text(0.0915032679738562, 0.8055555555555556, 'x[25] <= 0.167\ngini =
```

```

0.133\nsamples = 14\nvalue = [1, 13]'),
Text(0.08496732026143791, 0.75, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.09803921568627451, 0.75, 'gini = 0.0\nsamples = 13\nvalue = [0, 13]'),
Text(0.12418300653594772, 0.8055555555555556, 'x[18] <= 0.5\ngini =
0.493\nsamples = 25\nvalue = [11, 14]'),
Text(0.1111111111111111, 0.75, 'x[6] <= 0.7\ngini = 0.484\nsamples = 17\nvalue
= [10, 7]'),
Text(0.10457516339869281, 0.6944444444444444, 'x[16] <= 0.892\ngini =
0.408\nsamples = 14\nvalue = [10, 4]'),
Text(0.09803921568627451, 0.6388888888888888, 'x[2] <= 0.106\ngini =
0.278\nsamples = 12\nvalue = [10, 2]'),
Text(0.0915032679738562, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.10457516339869281, 0.5833333333333334, 'x[25] <= 0.5\ngini =
0.165\nsamples = 11\nvalue = [10, 1]'),
Text(0.09803921568627451, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.1111111111111111, 0.5277777777777778, 'gini = 0.0\nsamples = 10\nvalue =
[10, 0]'),
Text(0.1111111111111111, 0.6388888888888888, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.11764705882352941, 0.6944444444444444, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.13725490196078433, 0.75, 'x[4] <= 0.839\ngini = 0.219\nsamples =
8\nvalue = [1, 7]'),
Text(0.13071895424836602, 0.6944444444444444, 'gini = 0.0\nsamples = 7\nvalue =
[0, 7]'),
Text(0.1437908496732026, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.575342116013072, 0.9166666666666666, 'x[18] <= 0.5\ngini =
0.235\nsamples = 1098\nvalue = [949, 149]'),
Text(0.32388684640522875, 0.8611111111111112, 'x[25] <= 0.167\ngini =
0.162\nsamples = 798\nvalue = [727, 71]'),
Text(0.18627450980392157, 0.8055555555555556, 'x[2] <= 0.747\ngini =
0.38\nsamples = 47\nvalue = [35, 12]'),
Text(0.17973856209150327, 0.75, 'x[10] <= 0.5\ngini = 0.463\nsamples =
33\nvalue = [21, 12]'),
Text(0.1568627450980392, 0.6944444444444444, 'x[4] <= 0.446\ngini =
0.42\nsamples = 10\nvalue = [3, 7]'),
Text(0.1503267973856209, 0.6388888888888888, 'gini = 0.0\nsamples = 6\nvalue =
[0, 6]'),
Text(0.16339869281045752, 0.6388888888888888, 'x[13] <= 0.167\ngini =
0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.1568627450980392, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.16993464052287582, 0.5833333333333334, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),

```

```

Text(0.20261437908496732, 0.6944444444444444, 'x[24] <= 0.583\ngini =
0.34\nsamples = 23\nvalue = [18, 5]'),
Text(0.1895424836601307, 0.6388888888888888, 'x[9] <= 0.107\ngini =
0.117\nsamples = 16\nvalue = [15, 1]'),
Text(0.1830065359477124, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.19607843137254902, 0.5833333333333334, 'gini = 0.0\nsamples = 15\nvalue
= [15, 0]'),
Text(0.21568627450980393, 0.6388888888888888, 'x[26] <= 0.287\ngini =
0.49\nsamples = 7\nvalue = [3, 4]'),
Text(0.20915032679738563, 0.5833333333333334, 'gini = 0.0\nsamples = 4\nvalue =
[0, 4]'),
Text(0.2222222222222222, 0.5833333333333334, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),
Text(0.19281045751633988, 0.75, 'gini = 0.0\nsamples = 14\nvalue = [14, 0]'),
Text(0.461499183006536, 0.8055555555555556, 'x[23] <= 0.975\ngini =
0.145\nsamples = 751\nvalue = [692, 59]'),
Text(0.45496323529411764, 0.75, 'x[26] <= 0.113\ngini = 0.143\nsamples =
750\nvalue = [692, 58]'),
Text(0.31699346405228757, 0.6944444444444444, 'x[7] <= 0.167\ngini =
0.218\nsamples = 257\nvalue = [225, 32]'),
Text(0.2565359477124183, 0.6388888888888888, 'x[29] <= 0.147\ngini =
0.355\nsamples = 65\nvalue = [50, 15]'),
Text(0.23529411764705882, 0.5833333333333334, 'x[29] <= 0.029\ngini =
0.303\nsamples = 59\nvalue = [48, 11]'),
Text(0.21241830065359477, 0.5277777777777778, 'x[10] <= 0.5\ngini =
0.463\nsamples = 22\nvalue = [14, 8]'),
Text(0.19934640522875818, 0.4722222222222222, 'x[9] <= 0.179\ngini =
0.198\nsamples = 9\nvalue = [8, 1]'),
Text(0.19281045751633988, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.20588235294117646, 0.4166666666666667, 'gini = 0.0\nsamples = 8\nvalue =
[8, 0]'),
Text(0.22549019607843138, 0.4722222222222222, 'x[9] <= 0.4\ngini =
0.497\nsamples = 13\nvalue = [6, 7]'),
Text(0.21895424836601307, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue =
[4, 0]'),
Text(0.23202614379084968, 0.4166666666666667, 'x[4] <= 0.286\ngini =
0.346\nsamples = 9\nvalue = [2, 7]'),
Text(0.22549019607843138, 0.3611111111111111, 'x[5] <= 0.5\ngini =
0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.21895424836601307, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.23202614379084968, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.238562091503268, 0.3611111111111111, 'gini = 0.0\nsamples = 6\nvalue =
[0, 6]'),

```



```

Text(0.2581699346405229, 0.5277777777777778, 'x[13] <= 0.167\ngini =
0.149\nsamples = 37\nvalue = [34, 3]'),
Text(0.25163398692810457, 0.4722222222222222, 'x[26] <= 0.088\ngini =
0.5\nsamples = 6\nvalue = [3, 3]'),
Text(0.24509803921568626, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.2581699346405229, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),
Text(0.2647058823529412, 0.4722222222222222, 'gini = 0.0\nsamples = 31\nvalue =
[31, 0]'),
Text(0.2777777777777778, 0.5833333333333334, 'x[21] <= 0.667\ngini =
0.444\nsamples = 6\nvalue = [2, 4]'),
Text(0.27124183006535946, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.28431372549019607, 0.5277777777777778, 'x[28] <= 0.033\ngini =
0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.2777777777777778, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.2908496732026144, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.37745098039215685, 0.6388888888888888, 'x[0] <= 0.321\ngini =
0.161\nsamples = 192\nvalue = [175, 17]'),
Text(0.3202614379084967, 0.5833333333333334, 'x[6] <= 0.1\ngini =
0.294\nsamples = 67\nvalue = [55, 12]'),
Text(0.3137254901960784, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.32679738562091504, 0.5277777777777778, 'x[25] <= 0.5\ngini =
0.26\nsamples = 65\nvalue = [55, 10]'),
Text(0.30392156862745096, 0.4722222222222222, 'x[6] <= 0.5\ngini =
0.469\nsamples = 16\nvalue = [10, 6]'),
Text(0.2973856209150327, 0.4166666666666667, 'gini = 0.0\nsamples = 7\nvalue =
[7, 0]'),
Text(0.3104575163398693, 0.4166666666666667, 'x[7] <= 0.833\ngini =
0.444\nsamples = 9\nvalue = [3, 6]'),
Text(0.30392156862745096, 0.3611111111111111, 'gini = 0.0\nsamples = 5\nvalue =
[0, 5]'),
Text(0.31699346405228757, 0.3611111111111111, 'x[11] <= 0.125\ngini =
0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.3104575163398693, 0.3055555555555556, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),
Text(0.3235294117647059, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.34967320261437906, 0.4722222222222222, 'x[2] <= 0.037\ngini =
0.15\nsamples = 49\nvalue = [45, 4]'),
Text(0.3431372549019608, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.3562091503267974, 0.4166666666666667, 'x[2] <= 0.938\ngini =

```

```

0.117\nsamples = 48\nvalue = [45, 3]'),
Text(0.34967320261437906, 0.3611111111111111, 'x[5] <= 0.875\ngini =
0.081\nsamples = 47\nvalue = [45, 2]'),
Text(0.3366013071895425, 0.3055555555555556, 'x[10] <= 0.167\ngini =
0.043\nsamples = 45\nvalue = [44, 1]'),
Text(0.3300653594771242, 0.25, 'x[3] <= 0.75\ngini = 0.444\nsamples = 3\nvalue
= [2, 1]'),
Text(0.3235294117647059, 0.19444444444444445, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.3366013071895425, 0.19444444444444445, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.3431372549019608, 0.25, 'gini = 0.0\nsamples = 42\nvalue = [42, 0]'),
Text(0.3627450980392157, 0.3055555555555556, 'x[14] <= 0.75\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.3562091503267974, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.369281045751634, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.3627450980392157, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.434640522875817, 0.5833333333333334, 'x[6] <= 0.9\ngini = 0.077\nsamples
= 125\nvalue = [120, 5]'),
Text(0.4215686274509804, 0.5277777777777778, 'x[0] <= 0.393\ngini =
0.05\nsamples = 118\nvalue = [115, 3]'),
Text(0.4150326797385621, 0.4722222222222222, 'x[2] <= 0.956\ngini =
0.185\nsamples = 29\nvalue = [26, 3]'),
Text(0.4084967320261438, 0.4166666666666667, 'x[29] <= 0.147\ngini =
0.133\nsamples = 28\nvalue = [26, 2]'),
Text(0.3954248366013072, 0.3611111111111111, 'x[10] <= 0.167\ngini =
0.074\nsamples = 26\nvalue = [25, 1]'),
Text(0.3888888888888889, 0.3055555555555556, 'x[19] <= 0.214\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.38235294117647056, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.3954248366013072, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.4019607843137255, 0.3055555555555556, 'gini = 0.0\nsamples = 24\nvalue =
[24, 0]'),
Text(0.4215686274509804, 0.3611111111111111, 'x[17] <= 0.111\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.4150326797385621, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.42810457516339867, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.4215686274509804, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.42810457516339867, 0.4722222222222222, 'gini = 0.0\nsamples = 89\nvalue
= [89, 0]'),
Text(0.4477124183006536, 0.5277777777777778, 'x[2] <= 0.594\ngini =
0.408\nsamples = 7\nvalue = [5, 2]'),
Text(0.4411764705882353, 0.4722222222222222, 'x[27] <= 0.056\ngini =

```

```

0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.434640522875817, 0.416666666666667, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.4477124183006536, 0.416666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.4542483660130719, 0.472222222222222, 'gini = 0.0\nsamples = 4\nvalue =
[4, 0]'),
Text(0.5929330065359477, 0.694444444444444, 'x[26] <= 0.787\ngini =
0.1\nsamples = 493\nvalue = [467, 26]'),
Text(0.5584150326797386, 0.638888888888888, 'x[13] <= 0.5\ngini =
0.094\nsamples = 486\nvalue = [462, 24]'),
Text(0.5089869281045751, 0.583333333333333, 'x[12] <= 0.938\ngini =
0.154\nsamples = 191\nvalue = [175, 16]'),
Text(0.5024509803921569, 0.527777777777778, 'x[16] <= 0.481\ngini =
0.145\nsamples = 190\nvalue = [175, 15]'),
Text(0.4820261437908497, 0.472222222222222, 'x[16] <= 0.47\ngini =
0.221\nsamples = 95\nvalue = [83, 12]'),
Text(0.47549019607843135, 0.416666666666667, 'x[29] <= 0.794\ngini =
0.207\nsamples = 94\nvalue = [83, 11]'),
Text(0.46895424836601307, 0.361111111111111, 'x[5] <= 0.375\ngini =
0.192\nsamples = 93\nvalue = [83, 10]'),
Text(0.4411764705882353, 0.305555555555556, 'x[6] <= 0.9\ngini =
0.363\nsamples = 21\nvalue = [16, 5]'),
Text(0.434640522875817, 0.25, 'x[15] <= 0.413\ngini = 0.266\nsamples =
19\nvalue = [16, 3]'),
Text(0.4215686274509804, 0.194444444444444, 'x[17] <= 0.056\ngini =
0.117\nsamples = 16\nvalue = [15, 1]'),
Text(0.4150326797385621, 0.138888888888889, 'x[6] <= 0.4\ngini = 0.5\nsamples
= 2\nvalue = [1, 1]'),
Text(0.4084967320261438, 0.083333333333333, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.4215686274509804, 0.083333333333333, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.42810457516339867, 0.138888888888889, 'gini = 0.0\nsamples = 14\nvalue
= [14, 0]'),
Text(0.4477124183006536, 0.194444444444444, 'x[6] <= 0.5\ngini =
0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.4411764705882353, 0.138888888888889, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.4542483660130719, 0.138888888888889, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.4477124183006536, 0.25, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.49673202614379086, 0.305555555555556, 'x[27] <= 0.139\ngini =
0.129\nsamples = 72\nvalue = [67, 5]'),
Text(0.4803921568627451, 0.25, 'x[15] <= 0.244\ngini = 0.444\nsamples =
6\nvalue = [4, 2]'),
Text(0.4738562091503268, 0.194444444444444, 'x[9] <= 0.686\ngini =

```

```

0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.4673202614379085, 0.1388888888888889, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.4803921568627451, 0.1388888888888889, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.4869281045751634, 0.19444444444444445, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),
Text(0.5130718954248366, 0.25, 'x[2] <= 0.958\ngini = 0.087\nsamples =
66\nvalue = [63, 3]'),
Text(0.5, 0.19444444444444445, 'x[24] <= 0.583\ngini = 0.061\nsamples =
64\nvalue = [62, 2]'),
Text(0.4934640522875817, 0.1388888888888889, 'gini = 0.0\nsamples = 52\nvalue =
[52, 0]'),
Text(0.5065359477124183, 0.1388888888888889, 'x[12] <= 0.812\ngini =
0.278\nsamples = 12\nvalue = [10, 2]'),
Text(0.5, 0.08333333333333333, 'gini = 0.0\nsamples = 9\nvalue = [9, 0]'),
Text(0.5130718954248366, 0.08333333333333333, 'x[7] <= 0.5\ngini =
0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.5065359477124183, 0.02777777777777776, 'gini = 0.0\nsamples = 2\nvalue
= [0, 2]'),
Text(0.5196078431372549, 0.02777777777777776, 'gini = 0.0\nsamples = 1\nvalue
= [1, 0]'),
Text(0.5261437908496732, 0.19444444444444445, 'x[28] <= 0.567\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.5196078431372549, 0.1388888888888889, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.5326797385620915, 0.1388888888888889, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.4820261437908497, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.48856209150326796, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.5228758169934641, 0.4722222222222222, 'x[17] <= 0.5\ngini =
0.061\nsamples = 95\nvalue = [92, 3]'),
Text(0.5163398692810458, 0.4166666666666667, 'gini = 0.0\nsamples = 76\nvalue =
[76, 0]'),
Text(0.5294117647058824, 0.4166666666666667, 'x[29] <= 0.088\ngini =
0.266\nsamples = 19\nvalue = [16, 3]'),
Text(0.5163398692810458, 0.3611111111111111, 'x[21] <= 0.833\ngini =
0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.5098039215686274, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.5228758169934641, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.5424836601307189, 0.3611111111111111, 'x[15] <= 0.108\ngini =
0.117\nsamples = 16\nvalue = [15, 1]'),
Text(0.5359477124183006, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =

```

```

[0, 1]'),
  Text(0.5490196078431373, 0.3055555555555556, 'gini = 0.0\nsamples = 15\nvalue =
[15, 0]'),
  Text(0.5155228758169934, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.6078431372549019, 0.5833333333333334, 'x[19] <= 0.036\ngini =
0.053\nsamples = 295\nvalue = [287, 8]'),
  Text(0.5849673202614379, 0.5277777777777778, 'x[28] <= 0.7\ngini =
0.159\nsamples = 46\nvalue = [42, 4]'),
  Text(0.5784313725490197, 0.4722222222222222, 'x[10] <= 0.167\ngini =
0.124\nsamples = 45\nvalue = [42, 3]'),
  Text(0.5620915032679739, 0.4166666666666667, 'x[5] <= 0.625\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
  Text(0.5555555555555556, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.5686274509803921, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.5947712418300654, 0.4166666666666667, 'x[23] <= 0.688\ngini =
0.089\nsamples = 43\nvalue = [41, 2]'),
  Text(0.5816993464052288, 0.3611111111111111, 'x[12] <= 0.062\ngini =
0.048\nsamples = 41\nvalue = [40, 1]'),
  Text(0.5751633986928104, 0.3055555555555556, 'x[0] <= 0.595\ngini =
0.375\nsamples = 4\nvalue = [3, 1]'),
  Text(0.5686274509803921, 0.25, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
  Text(0.5816993464052288, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
  Text(0.5882352941176471, 0.3055555555555556, 'gini = 0.0\nsamples = 37\nvalue =
[37, 0]'),
  Text(0.6078431372549019, 0.3611111111111111, 'x[28] <= 0.167\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
  Text(0.6013071895424836, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.6143790849673203, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.5915032679738562, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.630718954248366, 0.5277777777777778, 'x[15] <= 0.056\ngini =
0.032\nsamples = 249\nvalue = [245, 4]'),
  Text(0.6143790849673203, 0.4722222222222222, 'x[15] <= 0.054\ngini =
0.32\nsamples = 5\nvalue = [4, 1]'),
  Text(0.6078431372549019, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue =
[4, 0]'),
  Text(0.6209150326797386, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.6470588235294118, 0.4722222222222222, 'x[2] <= 0.015\ngini =
0.024\nsamples = 244\nvalue = [241, 3]'),
  Text(0.6339869281045751, 0.4166666666666667, 'x[8] <= 0.5\ngini =
0.278\nsamples = 6\nvalue = [5, 1]'),

```

```

Text(0.6274509803921569, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.6405228758169934, 0.3611111111111111, 'gini = 0.0\nsamples = 5\nvalue =
[5, 0]'),
Text(0.6601307189542484, 0.4166666666666667, 'x[21] <= 0.167\ngini =
0.017\nsamples = 238\nvalue = [236, 2]'),
Text(0.6535947712418301, 0.3611111111111111, 'x[25] <= 0.833\ngini =
0.073\nsamples = 53\nvalue = [51, 2]'),
Text(0.6405228758169934, 0.3055555555555556, 'x[29] <= 0.088\ngini =
0.041\nsamples = 48\nvalue = [47, 1]'),
Text(0.6339869281045751, 0.25, 'x[12] <= 0.312\ngini = 0.245\nsamples =
7\nvalue = [6, 1]'),
Text(0.6274509803921569, 0.19444444444444445, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.6405228758169934, 0.19444444444444445, 'gini = 0.0\nsamples = 6\nvalue =
[6, 0]'),
Text(0.6470588235294118, 0.25, 'gini = 0.0\nsamples = 41\nvalue = [41, 0]'),
Text(0.6666666666666666, 0.3055555555555556, 'x[15] <= 0.38\ngini =
0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.6601307189542484, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.673202614379085, 0.25, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.6666666666666666, 0.3611111111111111, 'gini = 0.0\nsamples = 185\nvalue
= [185, 0]'),
Text(0.6274509803921569, 0.6388888888888888, 'x[2] <= 0.366\ngini =
0.408\nsamples = 7\nvalue = [5, 2]'),
Text(0.6209150326797386, 0.5833333333333334, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.6339869281045751, 0.5833333333333334, 'gini = 0.0\nsamples = 5\nvalue =
[5, 0]'),
Text(0.46803513071895425, 0.75, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.826797385620915, 0.8611111111111112, 'x[15] <= 0.157\ngini =
0.385\nsamples = 300\nvalue = [222, 78]'),
Text(0.7410130718954249, 0.8055555555555556, 'x[22] <= 0.167\ngini =
0.5\nsamples = 96\nvalue = [49, 47]'),
Text(0.7058823529411765, 0.75, 'x[4] <= 0.161\ngini = 0.459\nsamples =
42\nvalue = [15, 27]'),
Text(0.6797385620915033, 0.6944444444444444, 'x[16] <= 0.41\ngini =
0.499\nsamples = 23\nvalue = [12, 11]'),
Text(0.6601307189542484, 0.6388888888888888, 'x[15] <= 0.061\ngini =
0.426\nsamples = 13\nvalue = [4, 9]'),
Text(0.6535947712418301, 0.5833333333333334, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.6666666666666666, 0.5833333333333334, 'x[24] <= 0.25\ngini =
0.298\nsamples = 11\nvalue = [2, 9]'),
Text(0.6601307189542484, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.673202614379085, 0.5277777777777778, 'x[20] <= 0.5\ngini = 0.18\nsamples

```

```

= 10\nvalue = [1, 9]'),
  Text(0.6666666666666666, 0.4722222222222222, 'gini = 0.0\nsamples = 8\nvalue =
[0, 8]'),
  Text(0.6797385620915033, 0.4722222222222222, 'x[15] <= 0.083\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
  Text(0.673202614379085, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.6862745098039216, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.6993464052287581, 0.6388888888888888, 'x[7] <= 0.167\ngini =
0.32\nsamples = 10\nvalue = [8, 2]'),
  Text(0.6928104575163399, 0.5833333333333334, 'x[15] <= 0.137\ngini =
0.444\nsamples = 3\nvalue = [1, 2]'),
  Text(0.6862745098039216, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
  Text(0.6993464052287581, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.7058823529411765, 0.5833333333333334, 'gini = 0.0\nsamples = 7\nvalue =
[7, 0]'),
  Text(0.7320261437908496, 0.6944444444444444, 'x[23] <= 0.35\ngini =
0.266\nsamples = 19\nvalue = [3, 16]'),
  Text(0.7254901960784313, 0.6388888888888888, 'x[9] <= 0.2\ngini =
0.198\nsamples = 18\nvalue = [2, 16]'),
  Text(0.7189542483660131, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.7320261437908496, 0.5833333333333334, 'x[27] <= 0.306\ngini =
0.111\nsamples = 17\nvalue = [1, 16]'),
  Text(0.7254901960784313, 0.5277777777777778, 'gini = 0.0\nsamples = 15\nvalue =
[0, 15]'),
  Text(0.738562091503268, 0.5277777777777778, 'x[26] <= 0.188\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
  Text(0.7320261437908496, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.7450980392156863, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.738562091503268, 0.6388888888888888, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.7761437908496732, 0.75, 'x[0] <= 0.202\ngini = 0.466\nsamples =
54\nvalue = [34, 20]'),
  Text(0.7581699346405228, 0.6944444444444444, 'x[0] <= 0.107\ngini =
0.245\nsamples = 7\nvalue = [1, 6]'),
  Text(0.7516339869281046, 0.6388888888888888, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.7647058823529411, 0.6388888888888888, 'gini = 0.0\nsamples = 6\nvalue =
[0, 6]'),
  Text(0.7941176470588235, 0.6944444444444444, 'x[2] <= 0.622\ngini =
0.418\nsamples = 47\nvalue = [33, 14]'),

```

```

Text(0.7777777777777778, 0.6388888888888888, 'x[2] <= 0.145\ngini =
0.482\nsamples = 32\nvalue = [19, 13]'),
Text(0.7647058823529411, 0.5833333333333334, 'x[26] <= 0.237\ngini =
0.18\nsamples = 10\nvalue = [9, 1]'),
Text(0.7581699346405228, 0.5277777777777778, 'gini = 0.0\nsamples = 9\nvalue =
[9, 0]'),
Text(0.7712418300653595, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.7908496732026143, 0.5833333333333334, 'x[16] <= 0.87\ngini =
0.496\nsamples = 22\nvalue = [10, 12]'),
Text(0.7843137254901961, 0.5277777777777778, 'x[25] <= 0.833\ngini =
0.465\nsamples = 19\nvalue = [7, 12]'),
Text(0.7777777777777778, 0.4722222222222222, 'x[17] <= 0.167\ngini =
0.415\nsamples = 17\nvalue = [5, 12]'),
Text(0.7647058823529411, 0.4166666666666667, 'x[19] <= 0.321\ngini =
0.49\nsamples = 7\nvalue = [4, 3]'),
Text(0.7581699346405228, 0.3611111111111111, 'gini = 0.0\nsamples = 4\nvalue =
[4, 0]'),
Text(0.7712418300653595, 0.3611111111111111, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.7908496732026143, 0.4166666666666667, 'x[16] <= 0.235\ngini =
0.18\nsamples = 10\nvalue = [1, 9]'),
Text(0.7843137254901961, 0.3611111111111111, 'x[24] <= 0.417\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.7777777777777778, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.7908496732026143, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.7973856209150327, 0.3611111111111111, 'gini = 0.0\nsamples = 8\nvalue =
[0, 8]'),
Text(0.7908496732026143, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.7973856209150327, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),
Text(0.8104575163398693, 0.6388888888888888, 'x[9] <= 0.064\ngini =
0.124\nsamples = 15\nvalue = [14, 1]'),
Text(0.803921568627451, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.8169934640522876, 0.5833333333333334, 'gini = 0.0\nsamples = 14\nvalue =
[14, 0]'),
Text(0.9125816993464052, 0.8055555555555556, 'x[14] <= 0.75\ngini =
0.258\nsamples = 204\nvalue = [173, 31]'),
Text(0.8611111111111112, 0.75, 'x[15] <= 0.992\ngini = 0.138\nsamples =
147\nvalue = [136, 11]'),
Text(0.8545751633986928, 0.6944444444444444, 'x[4] <= 0.482\ngini =
0.128\nsamples = 146\nvalue = [136, 10]'),
Text(0.8366013071895425, 0.6388888888888888, 'x[26] <= 0.063\ngini =

```



```

0.038\nsamples = 104\nvalue = [102, 2]'),
  Text(0.8300653594771242, 0.5833333333333334, 'x[9] <= 0.193\ngini =
0.32\nsamples = 10\nvalue = [8, 2]'),
  Text(0.8235294117647058, 0.5277777777777778, 'x[11] <= 0.625\ngini =
0.444\nsamples = 3\nvalue = [1, 2]'),
  Text(0.8169934640522876, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
  Text(0.8300653594771242, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.8366013071895425, 0.5277777777777778, 'gini = 0.0\nsamples = 7\nvalue =
[7, 0]'),
  Text(0.8431372549019608, 0.5833333333333334, 'gini = 0.0\nsamples = 94\nvalue =
[94, 0]'),
  Text(0.8725490196078431, 0.6388888888888888, 'x[7] <= 0.167\ngini =
0.308\nsamples = 42\nvalue = [34, 8]'),
  Text(0.8562091503267973, 0.5833333333333334, 'x[25] <= 0.833\ngini =
0.375\nsamples = 4\nvalue = [1, 3]'),
  Text(0.8496732026143791, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
  Text(0.8627450980392157, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.8888888888888888, 0.5833333333333334, 'x[0] <= 0.393\ngini =
0.229\nsamples = 38\nvalue = [33, 5]'),
  Text(0.8758169934640523, 0.5277777777777778, 'x[7] <= 0.5\ngini = 0.5\nsamples
= 6\nvalue = [3, 3]'),
  Text(0.869281045751634, 0.4722222222222222, 'x[4] <= 0.875\ngini =
0.375\nsamples = 4\nvalue = [1, 3]'),
  Text(0.8627450980392157, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
  Text(0.8758169934640523, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
  Text(0.8823529411764706, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
  Text(0.9019607843137255, 0.5277777777777778, 'x[24] <= 0.917\ngini =
0.117\nsamples = 32\nvalue = [30, 2]'),
  Text(0.8954248366013072, 0.4722222222222222, 'x[12] <= 0.812\ngini =
0.062\nsamples = 31\nvalue = [30, 1]'),
  Text(0.8888888888888888, 0.4166666666666667, 'gini = 0.0\nsamples = 28\nvalue =
[28, 0]'),
  Text(0.9019607843137255, 0.4166666666666667, 'x[2] <= 0.561\ngini =
0.444\nsamples = 3\nvalue = [2, 1]'),
  Text(0.8954248366013072, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
  Text(0.9084967320261438, 0.3611111111111111, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
  Text(0.9084967320261438, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),

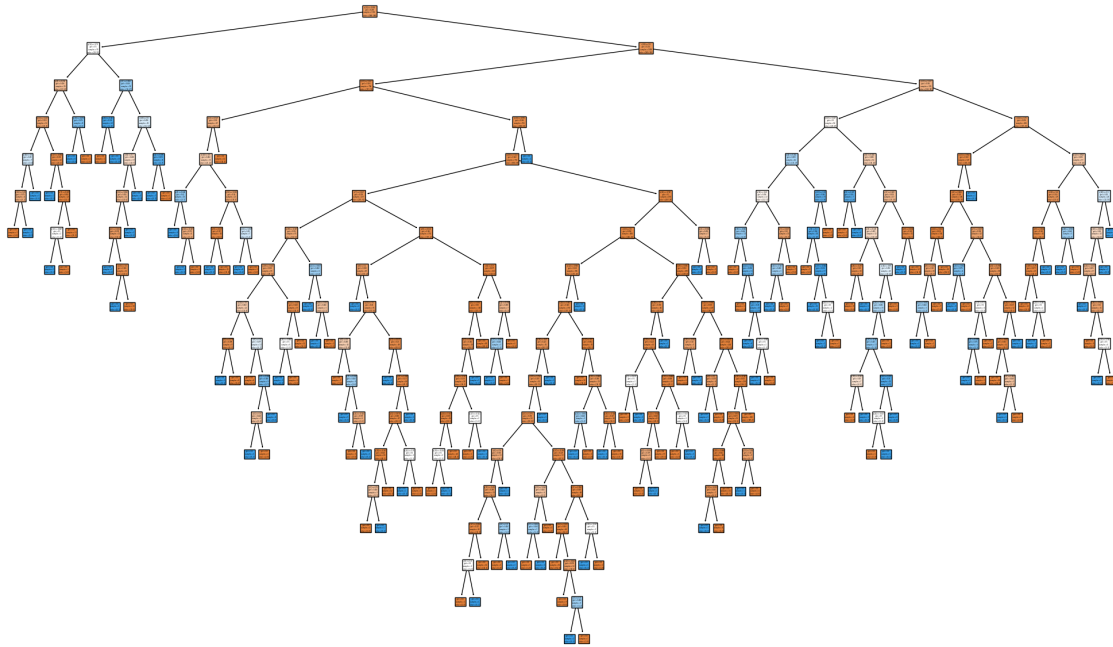
```

```

Text(0.8676470588235294, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.9640522875816994, 0.75, 'x[12] <= 0.812\ngini = 0.456\nsamples =
57\nvalue = [37, 20]'),
Text(0.9411764705882353, 0.6944444444444444, 'x[28] <= 0.4\ngini =
0.238\nsamples = 29\nvalue = [25, 4]'),
Text(0.9281045751633987, 0.6388888888888888, 'x[9] <= 0.964\ngini =
0.142\nsamples = 26\nvalue = [24, 2]'),
Text(0.9215686274509803, 0.5833333333333334, 'x[19] <= 0.75\ngini =
0.077\nsamples = 25\nvalue = [24, 1]'),
Text(0.9150326797385621, 0.5277777777777778, 'gini = 0.0\nsamples = 23\nvalue =
[23, 0]'),
Text(0.9281045751633987, 0.5277777777777778, 'x[23] <= 0.263\ngini =
0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.9215686274509803, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.934640522875817, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.934640522875817, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.954248366013072, 0.6388888888888888, 'x[28] <= 0.933\ngini =
0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.9477124183006536, 0.5833333333333334, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.9607843137254902, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.9869281045751634, 0.6944444444444444, 'x[28] <= 0.1\ngini =
0.49\nsamples = 28\nvalue = [12, 16]'),
Text(0.9803921568627451, 0.6388888888888888, 'x[10] <= 0.833\ngini =
0.48\nsamples = 20\nvalue = [12, 8]'),
Text(0.9738562091503268, 0.5833333333333334, 'x[26] <= 0.013\ngini =
0.415\nsamples = 17\nvalue = [12, 5]'),
Text(0.9673202614379085, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.9803921568627451, 0.5277777777777778, 'x[16] <= 0.505\ngini =
0.32\nsamples = 15\nvalue = [12, 3]'),
Text(0.9738562091503268, 0.4722222222222222, 'gini = 0.0\nsamples = 9\nvalue =
[9, 0]'),
Text(0.9869281045751634, 0.4722222222222222, 'x[16] <= 0.706\ngini =
0.5\nsamples = 6\nvalue = [3, 3]'),
Text(0.9803921568627451, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.9934640522875817, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),
Text(0.9869281045751634, 0.5833333333333334, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.9934640522875817, 0.6388888888888888, 'gini = 0.0\nsamples = 8\nvalue =

```

```
[0, 8]'))]
```



```
[242]: 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']
}
```

```
[243]: grid_search=GridSearchCV(estimator=dtc,param_grid=parameter,cv=5,scoring="accuracy")
```

```
[244]: grid_search.fit(x_train,y_train)
```

```
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be
removed in 1.3. To keep the past behaviour, explicitly set
`max_features='sqrt'`.
  warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be
removed in 1.3. To keep the past behaviour, explicitly set
`max_features='sqrt'`.
  warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
```





[illegible]







```
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
    warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
    warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
    warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
    warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
    warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
    warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
    warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max features='sqrt'`.
    warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set
```



[illegible]





```
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be
removed in 1.3. To keep the past behaviour, explicitly set
`max_features='sqrt'`.
```

```
warnings.warn(
```

```
C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py:269:
```

```
FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be
removed in 1.3. To keep the past behaviour, explicitly set
`max_features='sqrt'`.
```

```
warnings.warn(
```

```
[244]: GridSearchCV(cv=5, estimator=DecisionTreeClassifier(),
                  param_grid={'criterion': ['gini', 'entropy'],
                              'max_depth': [1, 2, 3, 4, 5],
                              'max_features': ['auto', 'sqrt', 'log2'],
                              'splitter': ['best', 'random']},
                  scoring='accuracy')
```

```
[245]: grid_search.best_params_
```

```
[245]: {'criterion': 'entropy',
        'max_depth': 3,
        'max_features': 'log2',
        'splitter': 'best'}
```

```
[246]: dtc_cv=DecisionTreeClassifier(criterion= 'entropy',
        max_depth=3,
        max_features='sqrt',
        splitter='best')
dtc_cv.fit(x_train,y_train)
```

```
[246]: DecisionTreeClassifier(criterion='entropy', max_depth=3, max_features='sqrt')
```

```
[247]: pred=dtc_cv.predict(x_test)
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

13 we can see the significant improvement in the accuracy after hyperparameter tuning

## 14 Random Forest Model Bulding

```
[249]: from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
```

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

```
[252]: rfc_cv= GridSearchCV(rfc,param_grid=forest_params,cv=10,scoring="accuracy")
```

```
[253]: rfc_cv.fit(x_train,y_train)
```

```
C:\ProgramData\anaconda3\lib\site-
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 "C:\ProgramData\anaconda3\lib\site-
packages\sklearn\model_selection\_validation.py", line 686, in _fit_and_score
    estimator.fit(X_train, y_train, **fit_params)
  File "C:\ProgramData\anaconda3\lib\site-packages\sklearn\ensemble\_forest.py",
line 340, in fit
    self._validate_params()
  File "C:\ProgramData\anaconda3\lib\site-packages\sklearn\base.py", line 581,
in _validate_params
    validate_parameter_constraints(
  File "C:\ProgramData\anaconda3\lib\site-
packages\sklearn\utils\_param_validation.py", line 97, in
validate_parameter_constraints
    raise InvalidParameterError(
sklearn.utils._param_validation.InvalidParameterError: The 'max_features'
parameter of RandomForestClassifier must be an int in the range [1, inf), a
float in the range (0.0, 1.0], a str among {'auto' (deprecated), 'log2', 'sqrt'}
or None. Got 0 instead.
```

```
warnings.warn(some_fits_failed_message, FitFailedWarning)
C:\ProgramData\anaconda3\lib\site-
```

```

packages\sklearn\model_selection\_search.py:952: UserWarning: One or more of the
test scores are non-finite: [          nan 0.84950022 0.85631609 0.86137911
0.85967695 0.86140808
0.86224105 0.85966971 0.86221932 0.86053165 0.85882225 0.86135738
0.86053165 0.86306678          nan 0.85118065 0.85715631 0.86224105
0.86224105 0.8596842 0.86393597 0.85628712 0.8622483 0.85882225
0.85626539 0.85542518 0.85457048 0.85967695          nan 0.85035492
0.85630885 0.85545415 0.86308127 0.85796755 0.86138635 0.86223381
0.85373026 0.85797479 0.85883674 0.86307403 0.85540345 0.85626539
          nan 0.84863103 0.85542518 0.85883674 0.86224105 0.85798928
0.86053165 0.86475445 0.86224105 0.85966247 0.86140084 0.86137187
0.85882225 0.85625815          nan 0.84865276 0.85545415 0.85884398
0.85882949 0.85967695 0.86054614 0.86223381 0.86308127 0.85798204
0.86308127 0.85881501 0.85796031 0.86223381]
warnings.warn(

```

```

[253]: GridSearchCV(cv=10, estimator=RandomForestClassifier(),
                param_grid=[{'max_depth': [10, 11, 12, 13, 14],
                              'max_features': [0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11,
                                                12, 13]}],
                scoring='accuracy')

```

```

[254]: pred=rfc_cv.predict(x_test)
print(classification_report(y_test,pred))
print(rfc_cv.best_params_)

```

	precision	recall	f1-score	support
0	0.86	0.98	0.92	245
1	0.69	0.18	0.29	49
accuracy			0.85	294
macro avg	0.77	0.58	0.60	294
weighted avg	0.83	0.85	0.81	294

```
{'max_depth': 13, 'max_features': 7}
```

```

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

```

```

[256]: array([0., 0., 1., 0., 1., 0., 1., 0., 0., 0., 0., 0., 0., 1., 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., 0., 0., 0., 0., 1., 0., 0., 1., 1., 1.,
1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 1., 0., 0., 0.,
1., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0.,
0., 1., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
1., 0., 0., 0., 0., 0., 0., 0., 0., 1., 1., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 1., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 0.,

```



```

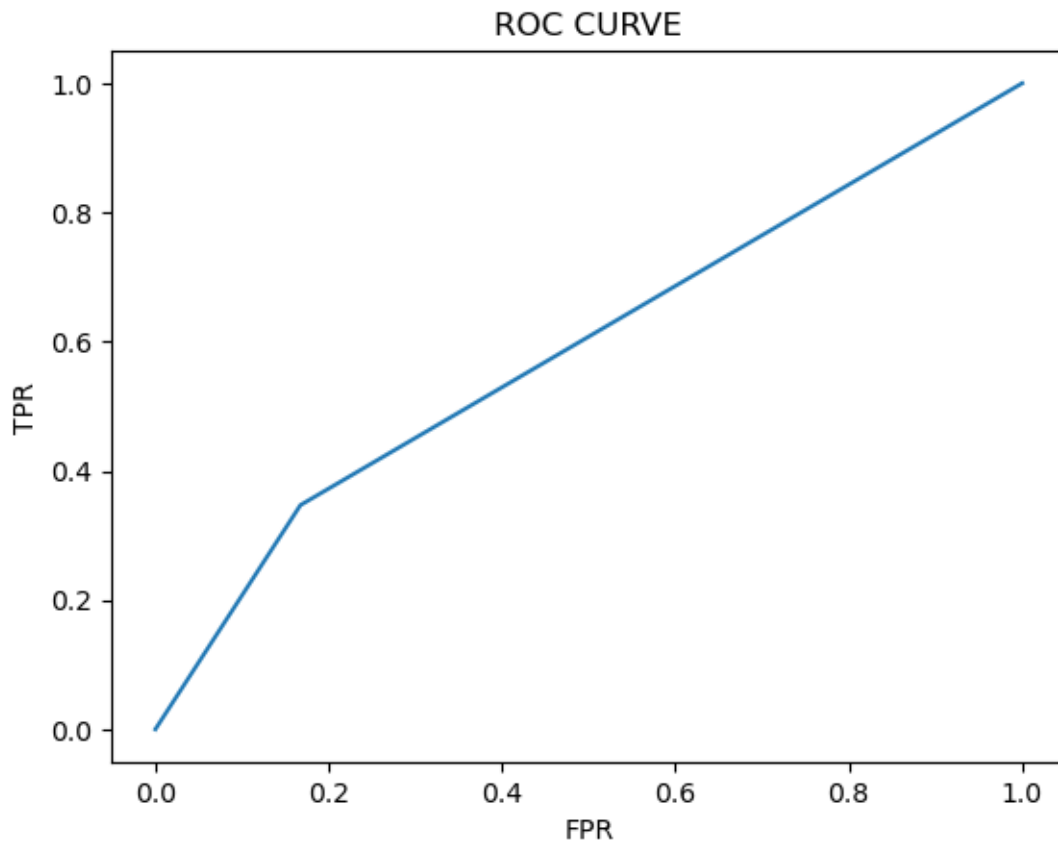
0., 0., 1., 0., 0., 0., 1., 0., 1., 0., 0., 1., 0., 0., 0., 1., 1.,
0., 0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 1., 1., 0., 0., 0., 0., 1., 0., 0., 1.,
0., 0., 1., 1., 1., 1., 1., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0.,
1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 1., 0., 0., 0., 0., 1., 1., 0., 0., 1., 0., 0., 0., 0., 0., 0.,
0., 1., 0., 0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0., 1., 0., 1.,
0., 0., 1., 0., 0., 0., 0., 0., 1., 0., 1., 0., 1., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 1., 0., 0., 1., 0., 0., 1., 0., 0., 1.,
0., 0., 0., 0., 0.]

```

```

[257]: # roc_curve
fpr,tpr,threshholds = roc_curve(y_test,probability)
plt.plot(fpr,tpr)
plt.xlabel('FPR')
plt.ylabel('TPR')
plt.title('ROC CURVE')
plt.show()

```



```
[ ]:
```