

## IMPORTING THE NECESSARY 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
0	41	Yes	Travel_Rarely	1102	Sales	1	2
1	49	No	Travel_Frequently	279	Research & Development	8	1
2	37	Yes	Travel_Rarely	1373	Research & Development	2	2
3	33	No	Travel_Frequently	1392	Research & Development	3	4
4	27	No	Travel_Rarely	591	Research & Development	2	1

5 rows × 35 columns

```
df.shape
```

(1470, 35)

```
df.corr()
```

```
<ipython-input-58-2f6f6606aa2c>:1: FutureWarning: The default value of numeric_only in df.corr()
```

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount
Age	1.000000	0.010661	-0.001686	0.208034	NaN
DailyRate	0.010661	1.000000	-0.004985	-0.016806	NaN
DistanceFromHome	-0.001686	-0.004985	1.000000	0.021042	NaN
Education	0.208034	-0.016806	0.021042	1.000000	NaN
EmployeeCount	NaN	NaN	NaN	NaN	NaN
EmployeeNumber	-0.010145	-0.050990	0.032916	0.042070	NaN
EnvironmentSatisfaction	0.010146	0.018355	-0.016075	-0.027128	NaN

```
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()
```

on	EmployeeCount	EmployeeNu
00	1470.0	1470.00
25	1.0	1024.86
65	0.0	602.02
00	1.0	1.00
00	1.0	491.25
00	1.0	1020.50
00	1.0	1555.75
00	1.0	2068.00

## 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
```

```
df.describe()
```

	Age	DistanceFromHome	EnvironmentSatisfaction	JobInvolvement	JobSatisf
count	1470.000000	1470.000000	1470.000000	1470.000000	1470.
mean	36.923810	9.192517	2.721769	2.729932	2.
std	9.135373	8.106864	1.093082	0.711561	1.
min	18.000000	1.000000	1.000000	1.000000	1.
25%	30.000000	2.000000	2.000000	2.000000	2.
50%	36.000000	7.000000	3.000000	3.000000	3.
75%	43.000000	14.000000	4.000000	3.000000	4.
max	60.000000	29.000000	4.000000	4.000000	4.

dropping all the unnecessary columns

Attrition is the dependent column

```
df.drop(['BusinessTravel', 'DailyRate', 'Department', 'Education', 'EducationField', 'EmployeeNumber', 'StandardHours', 'NumCompaniesWorked', 'JobLev
```

```
df.describe()
```

	Age	DistanceFromHome	EnvironmentSatisfaction	JobInvolvement	JobSatisf
count	1470.000000	1470.000000	1470.000000	1470.000000	1470.
mean	36.923810	9.192517	2.721769	2.729932	2.
std	9.135373	8.106864	1.093082	0.711561	1.
min	18.000000	1.000000	1.000000	1.000000	1.
25%	30.000000	2.000000	2.000000	2.000000	2.
50%	36.000000	7.000000	3.000000	3.000000	3.
75%	43.000000	14.000000	4.000000	3.000000	4.
max	60.000000	29.000000	4.000000	4.000000	4.

```
df.isnull().any()
```

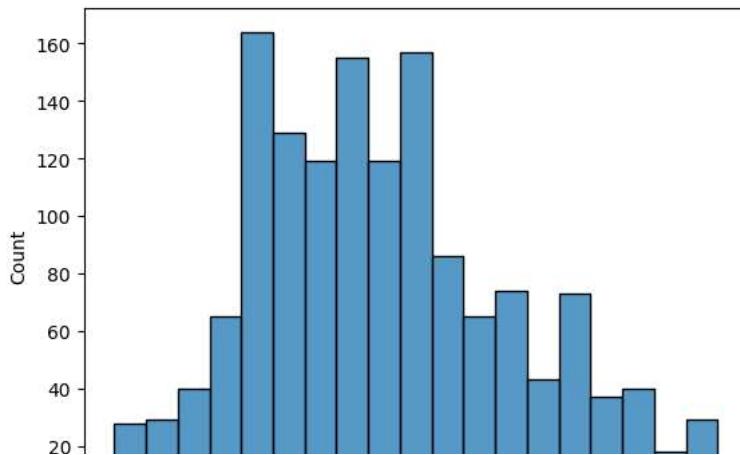
Age	False
Attrition	False
DistanceFromHome	False
EnvironmentSatisfaction	False
Gender	False
JobInvolvement	False
JobSatisfaction	False
MaritalStatus	False
MonthlyIncome	False
OverTime	False
PercentSalaryHike	False
TotalWorkingYears	False
WorkLifeBalance	False
YearsInCurrentRole	False
YearsSinceLastPromotion	False

dtype: bool

## DATA VISUALIZATION

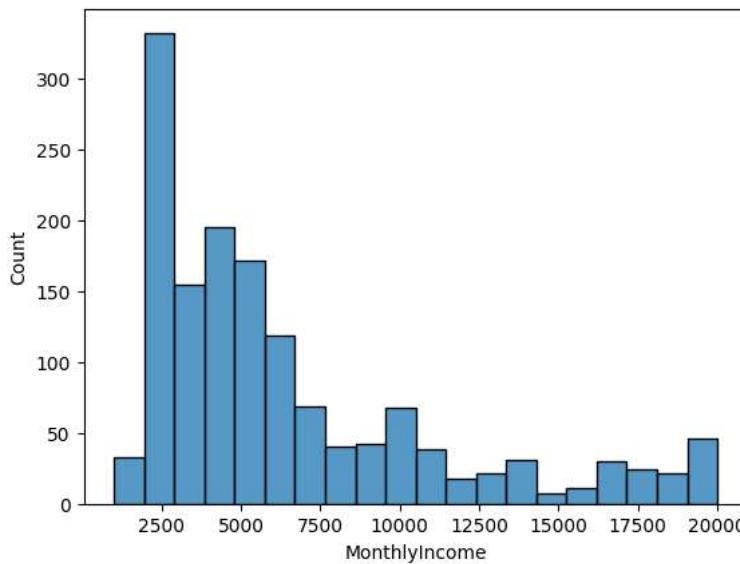
```
# Univariate Analysis
sns.histplot(df["Age"])
```

```
<Axes: xlabel='Age', ylabel='Count'>
```



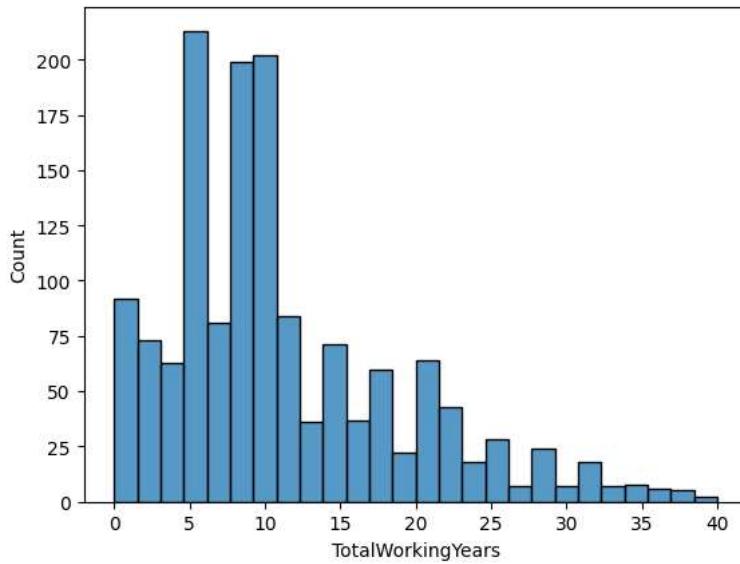
```
sns.histplot(df.MonthlyIncome)
```

```
<Axes: xlabel='MonthlyIncome', ylabel='Count'>
```



```
sns.histplot(df.TotalWorkingYears)
```

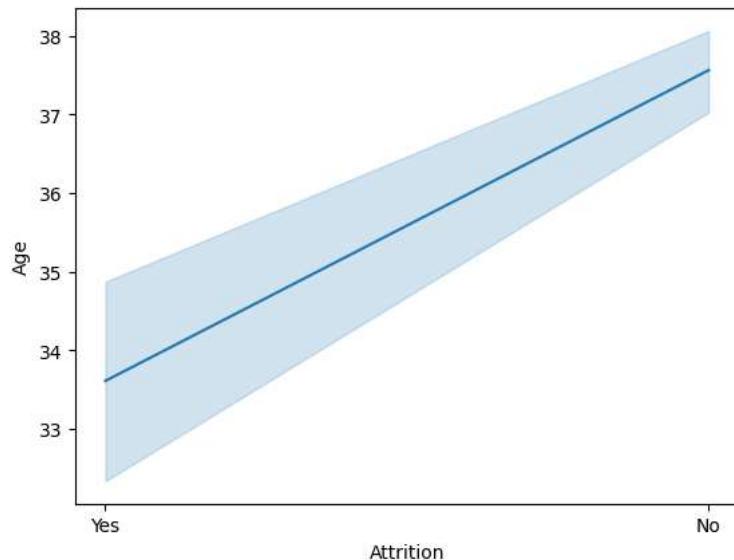
```
<Axes: xlabel='TotalWorkingYears', ylabel='Count'>
```



```
# Bivariate Analysis
```

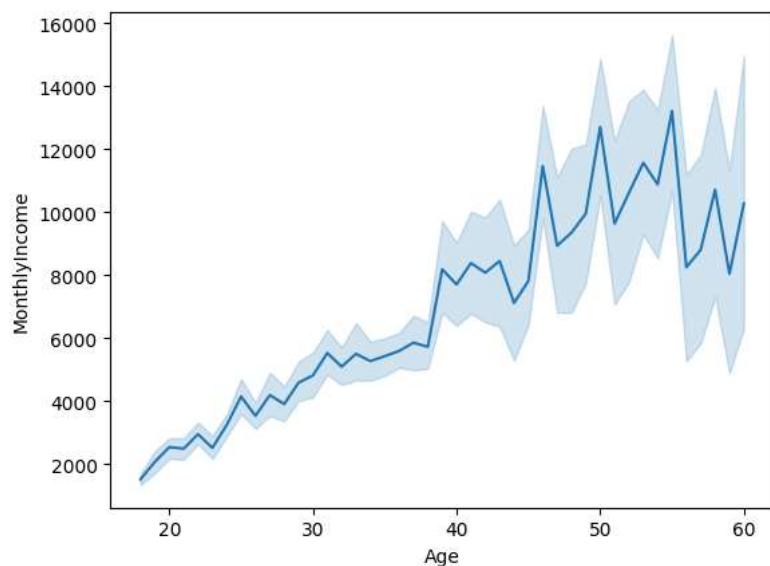
```
sns.lineplot(x = "Attrition", y = "Age", data = df)
```

```
<Axes: xlabel='Attrition', ylabel='Age'>
```



```
sns.lineplot(x = "Age", y = "MonthlyIncome", data = df)
```

```
<Axes: xlabel='Age', ylabel='MonthlyIncome'>
```



```
sns.lineplot(x = "TotalWorkingYears", y = "MonthlyIncome", data = df)
```

```
<Axes: xlabel='TotalWorkingYears', ylabel='MonthlyIncome'>
```

```
    ---
```

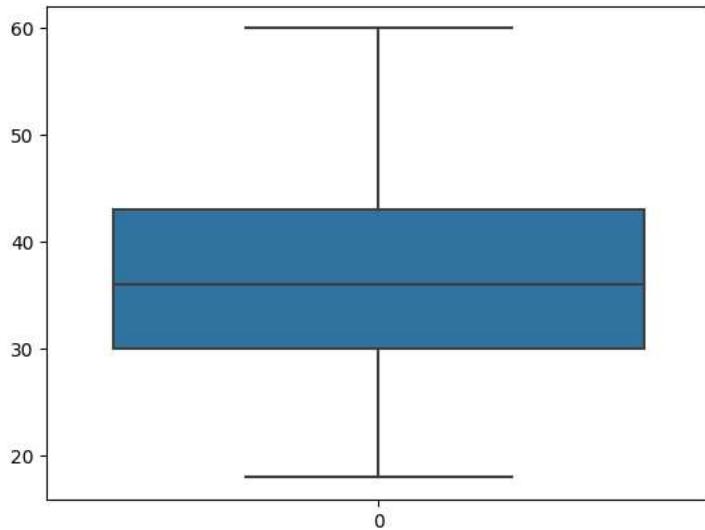
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1470 entries, 0 to 1469
Data columns (total 15 columns):
 #   Column           Non-Null Count  Dtype  
 ---  --  
 0   Age              1470 non-null    int64  
 1   Attrition        1470 non-null    object  
 2   DistanceFromHome 1470 non-null    int64  
 3   EnvironmentSatisfaction 1470 non-null    int64  
 4   Gender            1470 non-null    object  
 5   JobInvolvement   1470 non-null    int64  
 6   JobSatisfaction  1470 non-null    int64  
 7   MaritalStatus    1470 non-null    object  
 8   MonthlyIncome    1470 non-null    int64  
 9   OverTime          1470 non-null    object  
 10  PercentSalaryHike 1470 non-null    int64  
 11  TotalWorkingYears 1470 non-null    int64  
 12  WorkLifeBalance  1470 non-null    int64  
 13  YearsInCurrentRole 1470 non-null    int64  
 14  YearsSinceLastPromotion 1470 non-null    int64  
dtypes: int64(11), object(4)
memory usage: 172.4+ KB
```

## OUTLIER DETECTION

```
sns.boxplot(df.Age)
```

```
<Axes: >
```



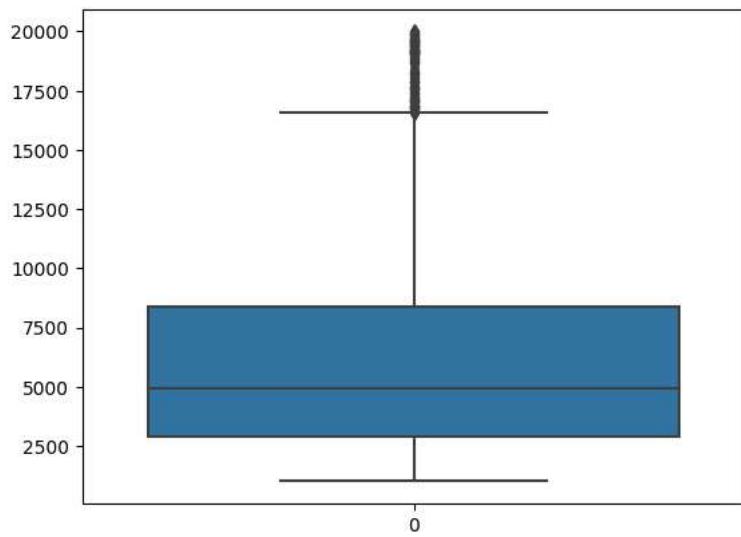
```
sns.boxplot(df.DistanceFromHome)
```

&lt;Axes: &gt;



sns.boxplot(df.MonthlyIncome)

&lt;Axes: &gt;



Q1 = df['MonthlyIncome'].quantile(0.25)

Q3 = df['MonthlyIncome'].quantile(0.75)

IQR = Q3 - Q1

d = 1.5

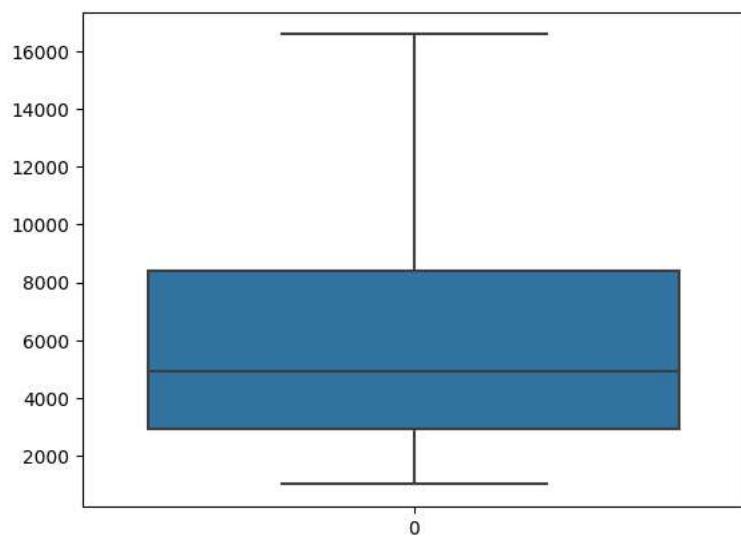
lower\_lim = Q1 -(d\*IQR)

upper\_lim = Q3 + (d\*IQR)

df['MonthlyIncome']=np.where(df['MonthlyIncome']&gt;upper\_lim,upper\_lim,np.where(df['MonthlyIncome']&lt;lower\_lim,lower\_lim,df['MonthlyIncome']))

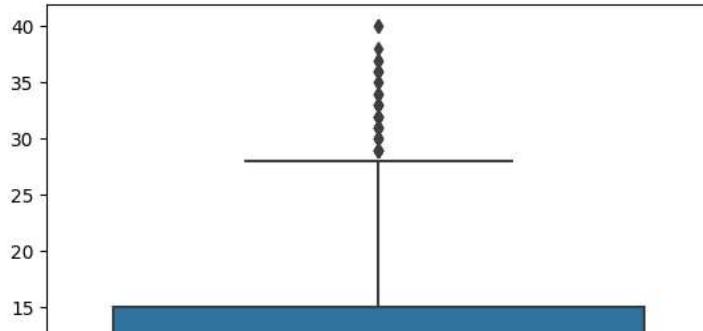
sns.boxplot(df.MonthlyIncome)

&lt;Axes: &gt;



sns.boxplot(df.TotalWorkingYears)

&lt;Axes: &gt;



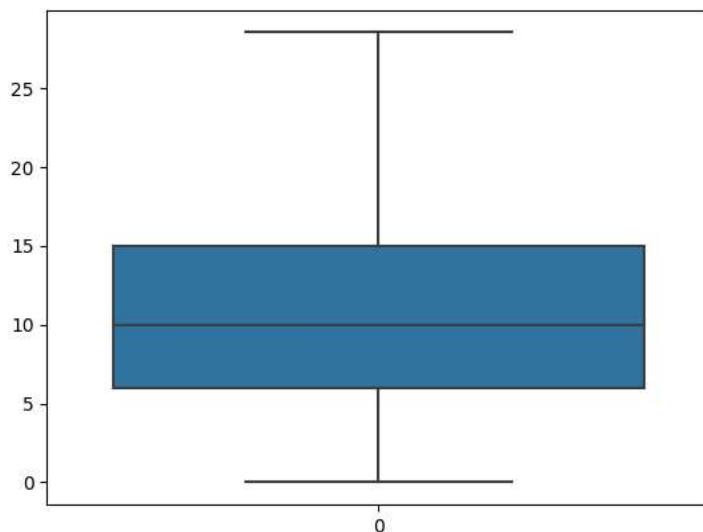
```

Q1 = df['TotalWorkingYears'].quantile(0.25)
Q3 = df['TotalWorkingYears'].quantile(0.75)
IQR = Q3 - Q1
d = 1.5
lower_lim = Q1 - (d*IQR)
upper_lim = Q3 + (d*IQR)
df['TotalWorkingYears']=np.where(df['TotalWorkingYears']>upper_lim,upper_lim,np.where(df['TotalWorkingYears']<lower_lim,lower_lim,df['TotalWorkingYears']))

```

```
sns.boxplot(df.TotalWorkingYears)
```

&lt;Axes: &gt;



## Label Encoding

```

from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
columns = ['Attrition', 'Gender', 'MaritalStatus', 'OverTime']
df[columns] = df[columns].apply(le.fit_transform)

```

## Splitting the dependent and independent variable

```

x = df.drop(columns = "Attrition")
y = df["Attrition"]

x.head()

```

```
Age  DistanceFromHome  EnvironmentSatisfaction  Gender  JobInvolvement  JobSatisfaction
y.head()
0      1
1      0
2      1
3      0
4      0
Name: Attrition, dtype: int64
```

## feature scaling for logistic regression

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

## Splitting into train and test data

```
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)
```

## ▼ Model Building --- Logistic Regression

```
from sklearn.linear_model import LogisticRegression  
model = LogisticRegression()
```

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

```
    ▾ LogisticRegression  
    LogisticRegression()
```

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

```
dfAttrition.value_counts().plot(kind = "pie", autopct = "%1.1f%%")
```

```
<Axes: ylabel='Attrition'>
```

### Evaluation of the classification

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

```
5
```

```
accuracy_score(y_test, pred)
```

```
0.8707482993197279
```

```
confusion_matrix(y_test, pred)
```

```
array([[241,  4],
       [ 34, 15]])
```

```
pd.crosstab(y_test, pred) # Visualized confusion matrix
```

	col_0	0	1	
Attrition				
0	241	4		
1	34	15		

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

	precision	recall	f1-score	support
0	0.88	0.98	0.93	245
1	0.79	0.31	0.44	49
accuracy			0.87	294
macro avg	0.83	0.64	0.68	294
weighted avg	0.86	0.87	0.85	294

```
probability=model.predict_proba(x_test)[:,1] # probability of all the yes values
probability
```

```
0.04599096, 0.61773815, 0.06023599, 0.00959877, 0.27968035,
0.0842562, 0.30259009, 0.02538328, 0.69789422, 0.20782178,
0.0515761, 0.1464348, 0.24578231, 0.02447109, 0.22122421,
0.31186016, 0.02428186, 0.06830356, 0.05988681, 0.42919823,
0.3299274, 0.11381061, 0.0466628, 0.71363351, 0.04911082,
0.02187439, 0.16762865, 0.05888231, 0.13255516, 0.10796323,
0.07749958, 0.07697303, 0.10776168, 0.06121553, 0.06121211,
0.06532578, 0.02067614, 0.01888257, 0.02843847, 0.04773017,
0.42121334, 0.2174884, 0.01128639, 0.7261964, 0.44541614,
0.14534399, 0.57328758, 0.09805092, 0.19111059, 0.5744568,
0.16928395, 0.01950284, 0.16877633, 0.08187363, 0.09909516,
0.0599772, 0.19181103, 0.09649451, 0.04659456, 0.37580431,
0.03868532, 0.18202413, 0.23016278, 0.14492632, 0.15522114,
0.08285658, 0.32017969, 0.04660827, 0.09075653, 0.05204234,
0.11079437, 0.10724618, 0.09165732, 0.12238473, 0.04740504,
0.0132515, 0.03403514, 0.20647038, 0.02810751, 0.03385499,
0.0944743, 0.00750029, 0.0802044, 0.01560663, 0.31123987,
0.45351263, 0.12244881, 0.19410278, 0.24514184, 0.02099799,
0.1780486, 0.32882673, 0.26789763, 0.119946, 0.07207794,
0.20272913, 0.51473141, 0.41198323, 0.02070214, 0.06195954,
0.04716694, 0.06544499, 0.24965305, 0.03763038, 0.26472447,
0.19621147, 0.1002986, 0.02248279, 0.18463981, 0.10535546,
0.04151099, 0.08150019, 0.09715958, 0.01831639, 0.02532645,
0.13620907, 0.07633671, 0.08716973, 0.69723351, 0.02840483,
0.01610639, 0.01600085, 0.09841706, 0.08542302, 0.06429813,
0.02756709, 0.19868174, 0.45278515, 0.18086062, 0.04916228,
0.44738387, 0.49569043, 0.33383642, 0.15580932, 0.25027566,
0.1022055, 0.08073811, 0.11339104, 0.18469605, 0.2565929,
0.02326258, 0.13941762, 0.00760761, 0.12854839, 0.12618728,
0.1393614, 0.34844187, 0.04589769, 0.18069311, 0.03729283,
0.04973812, 0.07705865, 0.05207065, 0.05253495, 0.02416478,
0.46462668, 0.02072513, 0.10699699, 0.77590472, 0.11508102,
```

```

0.03544475, 0.05941833, 0.11161862, 0.05491243, 0.0251996 ,
0.20827328, 0.11700163, 0.13628483, 0.06954993, 0.06988645,
0.02941673, 0.0837539 , 0.0109935 , 0.67691392, 0.03085246,
0.08857201, 0.26988523, 0.03197149, 0.57008189, 0.12436864,
0.43905377, 0.52585167, 0.28197513, 0.05744372, 0.12118214,
0.10946615, 0.03848744, 0.01368207, 0.37489003, 0.08839392,
0.16090726, 0.11710773, 0.41477659, 0.10060868, 0.11785707,
0.03996752, 0.58684823, 0.00428713, 0.06358941, 0.04534226,
0.05625275, 0.2142873 , 0.06301476, 0.11662493, 0.1874911 ,
0.03163527, 0.01841097, 0.11747885, 0.02751903, 0.2651188 ,
0.07446389, 0.13519759, 0.70186102, 0.08672511, 0.38814187,
0.02949061, 0.06394048, 0.15147126, 0.24924537, 0.04506879,
0.03425287, 0.26362348, 0.09956386, 0.01954104, 0.14287163,
0.28912133, 0.06960402, 0.01740465, 0.09153147, 0.00937296,
0.23841396, 0.18150312, 0.01121302, 0.30189644, 0.052956 ,
0.03935999, 0.52911812, 0.40327884, 0.06879774, 0.21014479,
0.17300474, 0.30658594, 0.67374806, 0.05044605, 0.21725974,
0.05178539, 0.02154401, 0.64069311, 0.15948953, 0.16032123,
0.4344616 , 0.04451498, 0.23354616, 0.04093299, 0.05339782,
0.0866061 , 0.01431701, 0.40213906, 0.41223774, 0.1501187 ,
0.11879022, 0.01374278, 0.11379516, 0.02103421, 0.02919475,
0.04201696, 0.10468593, 0.30030105, 0.15293258, 0.23814858,
0.24051204, 0.04890163, 0.14968683, 0.08236902, 0.06229793,
0.32440277, 0.01595591, 0.4437252 , 0.00410957, 0.05043896,
0.28780428, 0.5354455 , 0.03799566, 0.30691028])

```

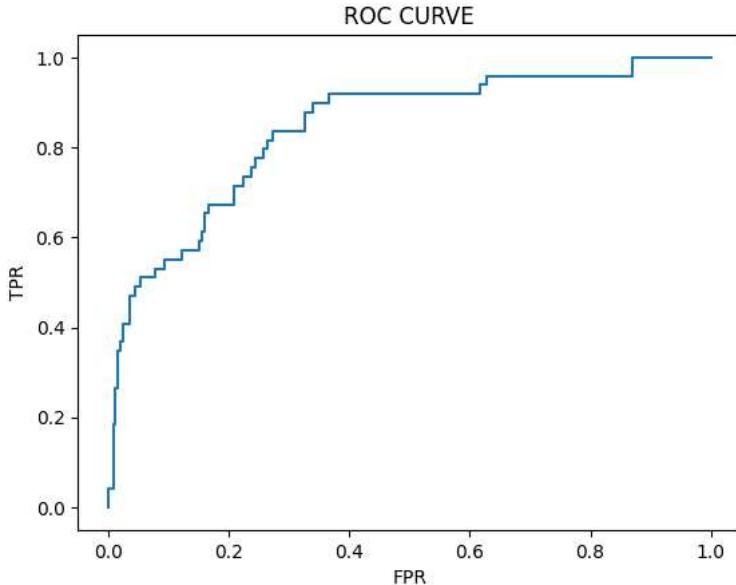
```
roc_curve
```

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

```

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

```



## ▼ Model Building --- Decision Tree

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

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

```

▼ DecisionTreeClassifier
DecisionTreeClassifier()

```

```
y_pred1 = dtc.predict(x_test)
```

```
accuracy_score(y_test, y_pred1)
```

```
0.7857142857142857
```

```
confusion_matrix(y_test, y_pred1)
```

```
array([[213, 32],  
       [ 31, 18]])
```

```
pd.crosstab(y_test, y_pred1)
```

	0	1
Attrition	213	32
0	213	32
1	31	18

```
print(classification_report(y_test, y_pred1))
```

	precision	recall	f1-score	support
0	0.87	0.87	0.87	245
1	0.36	0.37	0.36	49
accuracy			0.79	294
macro avg	0.62	0.62	0.62	294
weighted avg	0.79	0.79	0.79	294

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

```
[1, 0']'),  
Text(0.06566347469220246, 0.6388888888888888, 'gini = 0.0\nsamples = 20\nvalue = [20, 0']'),  
Text(0.06566347469220246, 0.8055555555555555, 'x[9] <= 0.679\nngini = 0.375\nsamples = 8\nvalue = [2, 6']'),  
Text(0.060191518467852256, 0.75, 'gini = 0.0\nsamples = 6\nvalue = [0, 6']'),  
Text(0.07113543091655267, 0.75, 'gini = 0.0\nsamples = 2\nvalue = [2, 0']'),  
Text(0.12038303693570451, 0.8611111111111112, 'x[12] <= 0.028\nngini = 0.426\nsamples = 39\nvalue = [12, 27']'),  
Text(0.11491108071135431, 0.8055555555555556, 'x[0] <= 0.369\nngini = 0.394\nsamples = 37\nvalue = [10, 27']'),  
Text(0.1094391244870041, 0.75, 'x[7] <= 0.119\nngini = 0.353\nsamples = 35\nvalue = [8, 27']'),  
Text(0.09302325581395349, 0.6944444444444444, 'x[10] <= 0.018\nngini = 0.278\nsamples = 30\nvalue = [5, 25']'),  
Text(0.07660738714090287, 0.6388888888888888, 'x[7] <= 0.02\nngini = 0.494\nsamples = 9\nvalue = [4, 5']'),  
Text(0.07113543091655267, 0.5833333333333334, 'gini = 0.0\nsamples = 3\nvalue = [3, 0']'),  
Text(0.08207934336525308, 0.5833333333333334, 'x[1] <= 0.036\nngini = 0.278\nsamples = 6\nvalue = [1, 5']'),  
Text(0.07660738714090287, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [1, 0']'),  
Text(0.08755129958960328, 0.5277777777777778, 'gini = 0.0\nsamples = 5\nvalue = [0, 5']'),  
Text(0.1094391244870041, 0.6388888888888888, 'x[0] <= 0.036\nngini = 0.091\nsamples = 21\nvalue = [1, 20']'),  
Text(0.1039671682626539, 0.5833333333333334, 'x[4] <= 0.5\nngini = 0.32\nsamples = 5\nvalue = [1, 4']'),  
Text(0.09849521203830369, 0.5277777777777778, 'gini = 0.0\nsamples = 4\nvalue = [0, 4']'),  
Text(0.1094391244870041, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [1, 0']'),  
Text(0.11491108071135431, 0.5833333333333334, 'gini = 0.0\nsamples = 16\nvalue = [0, 16']'),  
Text(0.12585499316005472, 0.6944444444444444, 'x[9] <= 0.393\nngini = 0.48\nsamples = 5\nvalue = [3, 2']'),  
Text(0.12038303693570451, 0.6388888888888888, 'gini = 0.0\nsamples = 3\nvalue = [3, 0']'),  
Text(0.13132694938440492, 0.6388888888888888, 'gini = 0.0\nsamples = 2\nvalue = [0, 2']'),  
Text(0.12038303693570451, 0.75, 'gini = 0.0\nsamples = 2\nvalue = [2, 0']'),  
Text(0.12585499316005472, 0.8055555555555556, 'gini = 0.0\nsamples = 2\nvalue = [2, 0']'),  
Text(0.5524164244186046, 0.9166666666666666, 'x[8] <= 0.5\nngini = 0.235\nsamples = 1098\nvalue = [949, 149']'),  
Text(0.2756497948016416, 0.8611111111111112, 'x[11] <= 0.167\nngini = 0.162\nsamples = 798\nvalue = [727, 71']'),  
Text(0.16552667578659372, 0.8055555555555556, 'x[7] <= 0.107\nngini = 0.38\nsamples = 47\nvalue = [35, 12']'),  
Text(0.14227086183310533, 0.75, 'x[2] <= 0.167\nngini = 0.444\nsamples = 6\nvalue = [2, 11']'),
```

```
[2, 4]),  
Text(0.13679890560875513, 0.6944444444444444, 'gini = 0.0\nsamples = 3\nvalue = [0,  
3]),  
Text(0.14774281805745554, 0.6944444444444444, 'x[13] <= 0.333\nngini =  
0.444\nsamples = 3\nvalue = [2, 1]),  
Text(0.14227086183310533, 0.6388888888888888, 'gini = 0.0\nsamples = 2\nvalue = [2,  
0]),  
Text(0.15321477428180574, 0.6388888888888888, 'gini = 0.0\nsamples = 1\nvalue = [0,  
1]),  
Text(0.18878248974008208, 0.75, 'x[13] <= 0.4\nngini = 0.314\nsamples =  
[33, 8]),  
Text(0.17510259917920656, 0.6944444444444444, 'x[1] <= 0.304\nngini = 0.239\nsamples  
= 36\nvalue = [31, 5]),  
Text(0.16415868673050615, 0.6388888888888888, 'x[12] <= 0.056\nngini = 0.08\nsamples  
= 24\nvalue = [23, 1]),  
Text(0.15868673050615595, 0.5833333333333334, 'x[7] <= 0.207\nngini = 0.444\nsamples  
= 3\nvalue = [2, 1]),  
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1]),  
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0]),  
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[21, 0]),  
Text(0.18604651162790697, 0.6388888888888888, 'x[1] <= 0.482\nngini = 0.444\nsamples  
= 12\nvalue = [8, 4]),  
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3]),  
Text(0.19151846785225718, 0.5833333333333334, 'x[5] <= 0.167\nngini = 0.198\nsamples  
= 9\nvalue = [8, 1]),  
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0]),  
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0]),  
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5\nvalue = [2, 3]),  
Text(0.19699042407660738, 0.6388888888888888, 'gini = 0.0\nsamples = 2\nvalue = [2,  
0]),  
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3]),  
Text(0.3857729138166895, 0.8055555555555556, 'x[2] <= 0.167\nngini = 0.145\nsamples  
= 751\nvalue = [692, 59]),  
Text(0.2503419972640219, 0.75, 'x[4] <= 0.167\nngini = 0.237\nsamples = 153\nvalue =  
[132, 21]),  
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10\nvalue = [5, 5]),  
Text(0.2188782489740082, 0.6388888888888888, 'x[7] <= 0.247\nngini = 0.408\nsamples  
= 7\nvalue = [2, 5]),  
Text(0.213406292749658, 0.5833333333333334, 'x[12] <= 0.139\nngini = 0.444\nsamples  
= 3\nvalue = [2, 1]),  
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0]),  
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4]),  
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0]),  
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= 143\nvalue = [127, 16]),  
Text(0.24623803009575923, 0.6388888888888888, 'x[7] <= 0.214\nngini = 0.375\nsamples  
= 24\nvalue = [18, 6]),  
Text(0.23529411764705882, 0.5833333333333334, 'x[0] <= 0.548\nngini = 0.494\nsamples  
= 9\nvalue = [4, 5]),  
Text(0.22982216142270862, 0.5277777777777778, 'x[4] <= 0.5\nngini = 0.444\nsamples =  
6\nvalue = [4, 2]),  
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0]),  
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= 3\nvalue = [1, 2]),  
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2]),  
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0]),  
Text(0.24076607387140903, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue = [0,  
3]),  
Text(0.25718194254445964, 0.5833333333333334, 'x[0] <= 0.393\nngini = 0.124\nsamples  
= 15\nvalue = [14, 1]),  
Text(0.25170998632010944, 0.5277777777777778, 'x[1] <= 0.268\nngini = 0.5\nsamples =  
2\nvalue = [1, 1]),  
Text(0.24623803009575923, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [1,  
0]),
```

```

Text(0.25718194254445964, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [0,
1]'),
Text(0.26265389876880985, 0.5277777777777778, 'gini = 0.0\nsamples = 13\nvalue =
[13, 0]'),
Text(0.3064295485636115, 0.6388888888888888, 'x[7] <= 0.444\nngini = 0.154\nsamples
= 119\nvalue = [109, 10]'),
Text(0.28454172366621067, 0.5833333333333334, 'x[0] <= 0.893\nngini = 0.07\nsamples
= 82\nvalue = [79, 3]'),
Text(0.27359781121751026, 0.5277777777777778, 'x[7] <= 0.228\nngini = 0.05\nsamples
= 78\nvalue = [76, 2]'),
Text(0.26812585499316005, 0.4722222222222222, 'gini = 0.0\nsamples = 48\nvalue =
[48, 0]'),
Text(0.27906976744186046, 0.4722222222222222, 'x[7] <= 0.231\nngini = 0.124\nsamples
= 30\nvalue = [28, 2]'),
Text(0.27359781121751026, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0,
1]'),
Text(0.28454172366621067, 0.4166666666666667, 'x[7] <= 0.248\nngini = 0.067\nsamples
= 29\nvalue = [28, 1]'),
Text(0.27906976744186046, 0.3611111111111111, 'x[3] <= 0.5\nngini = 0.375\nsamples =
4\nvalue = [3, 1]'),
Text(0.27359781121751026, 0.3055555555555556, 'gini = 0.0\nsamples = 3\nvalue = [3,
0]'),
Text(0.28454172366621067, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0,
1]'),
Text(0.29001367989056087, 0.3611111111111111, 'gini = 0.0\nsamples = 25\nvalue =
[25, 0]'),
Text(0.2954856361149111, 0.5277777777777778, 'x[1] <= 0.804\nngini = 0.375\nsamples
= 4\nvalue = [3, 1]'),
Text(0.29001367989056087, 0.4722222222222222, 'gini = 0.0\nsamples = 3\nvalue = [3,
0]'),
Text(0.3009575923392613, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [0,
1]'),
Text(0.3283173734610123, 0.5833333333333334, 'x[7] <= 0.588\nngini = 0.307\nsamples
= 37\nvalue = [30, 7]'),
Text(0.3173734610123119, 0.5277777777777778, 'x[4] <= 0.5\nngini = 0.5\nsamples =
12\nvalue = [6, 6]'),
Text(0.3119015047879617, 0.4722222222222222, 'gini = 0.0\nsamples = 4\nvalue = [4,
0]'),
Text(0.3228454172366621, 0.4722222222222222, 'x[9] <= 0.179\nngini = 0.375\nsamples
= 8\nvalue = [2, 6]'),
Text(0.3173734610123119, 0.4166666666666667, 'gini = 0.0\nsamples = 5\nvalue = [0,
5]'),
Text(0.3283173734610123, 0.4166666666666667, 'x[5] <= 0.5\nngini = 0.444\nsamples =
3\nvalue = [2, 1]'),
Text(0.3228454172366621, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0,
1]'),
Text(0.3337893296853625, 0.3611111111111111, 'gini = 0.0\nsamples = 2\nvalue = [2,
0]'),
Text(0.3392612859097127, 0.5277777777777778, 'x[13] <= 0.7\nngini = 0.077\nsamples =
25\nvalue = [24, 1]'),
Text(0.3337893296853625, 0.4722222222222222, 'gini = 0.0\nsamples = 21\nvalue =
[21, 0]'),
Text(0.3447332421340629, 0.4722222222222222, 'x[12] <= 0.417\nngini = 0.375\nsamples
= 4\nvalue = [3, 1]'),
Text(0.3392612859097127, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0,
1]'),
Text(0.35020519835841313, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue = [3,
0]'),
Text(0.521203830369357, 0.75, 'x[0] <= 0.345\nngini = 0.119\nsamples = 598\nvalue =
[560, 38]'),
Text(0.4100547195622435, 0.6944444444444444, 'x[7] <= 0.067\nngini = 0.205\nsamples
= 198\nvalue = [175, 23]'),
Text(0.4045827633378933, 0.6388888888888888, 'gini = 0.0\nsamples = 2\nvalue = [0,
2]'),
Text(0.4155266757865937, 0.6388888888888888, 'x[9] <= 0.036\nngini = 0.191\nsamples
= 196\nvalue = [175, 21]'),
Text(0.37756497948016415, 0.5833333333333334, 'x[4] <= 0.5\nngini = 0.384\nsamples =
27\nvalue = [20, 7]'),
Text(0.36114911080711354, 0.5277777777777778, 'x[7] <= 0.199\nngini = 0.494\nsamples
= 9\nvalue = [4, 5]'),
Text(0.35567715458276333, 0.4722222222222222, 'gini = 0.0\nsamples = 3\nvalue = [0,
3]'),
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= 6\nvalue = [4, 2]'),
Text(0.36114911080711354, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue = [4,
0]'),
Text(0.37209302325581395, 0.4166666666666667, 'gini = 0.0\nsamples = 2\nvalue = [0,
2]'),
Text(0.39398084815321477, 0.5277777777777778, 'x[1] <= 0.071\nngini = 0.198\nsamples
= 18\nvalue = [16, 2]'),
Text(0.38850889192886456, 0.4722222222222222, 'x[0] <= 0.274\nngini = 0.444\nsamples
= 6\nvalue = [4, 2]'),
Text(0.38303693570451436, 0.4166666666666667, 'x[6] <= 0.25\nngini = 0.32\nsamples =
5\nvalue = [4, 1]'),
Text(0.37756497948016415, 0.3611111111111111, 'x[5] <= 0.167\nngini = 0.5\nsamples =

```

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 2\nvalue = [1, 1']),
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1']),
 Text(0.38303693570451436, 0.30555555555555556, 'gini = 0.0\nsamples = 1\nvalue = [1,
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1']),
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 Text(0.45348837209302323, 0.5833333333333334, 'x[10] <= 0.088\ngini =
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 Text(0.4158686730506156, 0.5277777777777778, 'x[9] <= 0.179\ngini = 0.42\nsamples =
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 Text(0.4158686730506156, 0.41666666666666667, 'gini = 0.0\nsamples = 7\nvalue = [7,
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11\nvalue = [7, 4']),
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3]),
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78\nvalue = [73, 5]),
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39\nvalue = [36, 3]),
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[15, 3']),
 Text(0.5034199726402189, 0.1944444444444445, 'x[9] <= 0.321\ngini = 0.208\nsamples =
17\nvalue = [15, 2]),
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6\nvalue = [4, 2]),
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0.444\nsamples = 3\nvalue = [1, 2]),
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 Text(0.5143638850889193, 0.1944444444444445, 'gini = 0.0\nsamples = 1\nvalue = [0,
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 Text(0.53077975376197, 0.3611111111111111, 'x[0] <= 0.214\ngini = 0.346\nsamples =
9\nvalue = [7, 2]),

```

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Text(0.525307/9/53/6197, 0.305555555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
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Text(0.53077975376197, 0.25, 'gini = 0.0\nsamples = 7\nvalue = [7, 0]'),
Text(0.5417236662106704, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.53077975376197, 0.41666666666666667, 'gini = 0.0\nsamples = 42\nvalue = [42, 0]'),
Text(0.6323529411764706, 0.694444444444444444, 'x[13] <= 0.967\nngini = 0.072\nsamples = 400\nvalue = [385, 15]'),
Text(0.6025991792065664, 0.6388888888888888, 'x[9] <= 0.321\nngini = 0.064\nsamples = 394\nvalue = [381, 13]'),
Text(0.5595075239398085, 0.58333333333333334, 'x[7] <= 0.069\nngini = 0.031\nsamples = 253\nvalue = [249, 4]'),
Text(0.5417236662106704, 0.52777777777777778, 'x[7] <= 0.068\nngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.5362517099863201, 0.4722222222222222, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.5471956224350205, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5772913816689467, 0.52777777777777778, 'x[5] <= 0.167\nngini = 0.024\nsamples = 249\nvalue = [246, 3]'),
Text(0.5581395348837209, 0.4722222222222222, 'x[7] <= 0.128\nngini = 0.089\nsamples = 43\nvalue = [41, 2]'),
Text(0.5471956224350205, 0.41666666666666667, 'x[7] <= 0.112\nngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.5417236662106704, 0.3611111111111111, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.5526675786593708, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5690834473324213, 0.41666666666666667, 'x[1] <= 0.018\nngini = 0.049\nsamples = 40\nvalue = [39, 1]'),
Text(0.5636114911080712, 0.3611111111111111, 'x[10] <= 0.719\nngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
Text(0.5581395348837209, 0.3055555555555556, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.5690834473324213, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5745554035567716, 0.3611111111111111, 'gini = 0.0\nsamples = 34\nvalue = [34, 0]'),
Text(0.5964432284541724, 0.4722222222222222, 'x[9] <= 0.036\nngini = 0.01\nsamples = 206\nvalue = [205, 1]'),
Text(0.5909712722298222, 0.41666666666666667, 'x[10] <= 0.263\nngini = 0.043\nsamples = 46\nvalue = [45, 1]'),
Text(0.585499316005472, 0.3611111111111111, 'x[5] <= 0.5\nngini = 0.198\nsamples = 9\nvalue = [8, 1]'),
Text(0.5800273597811217, 0.3055555555555556, 'x[7] <= 0.173\nngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.5745554035567716, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.585499316005472, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5909712722298222, 0.3055555555555556, 'gini = 0.0\nsamples = 7\nvalue = [7, 0]'),
Text(0.5611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 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']
}
= 103\nvalue = [94, 9]),

grid_search=GridSearchCV(estimator=dtc.param_grid, cv=5, scoring="accuracy")
```

```
Text(0.612859097127223, 0.41666666666666667, 'x[0] <= 0.393\nngini = 0.444\nnsamples =
```

```
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')
```

## GridSearchCV

► estimator: DecisionTreeClassifier

► [DecisionTreeClassifier](#)

Digitized by srujanika@gmail.com

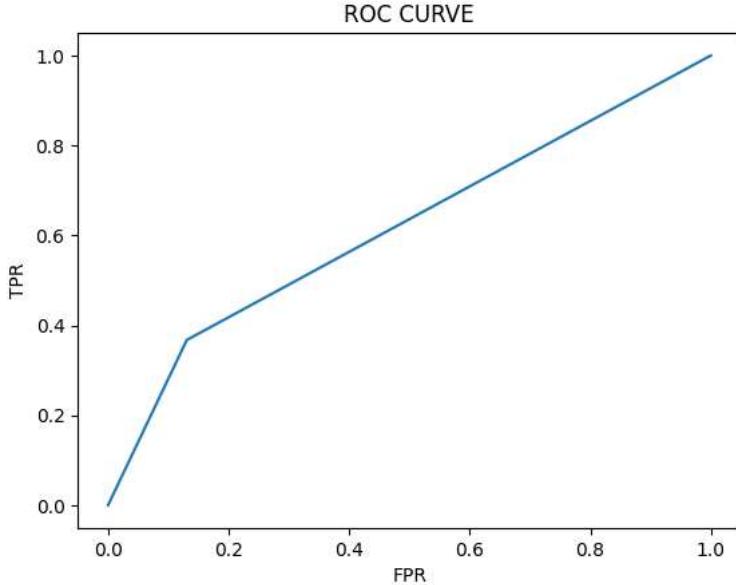
```
grid_search.fit(x_train, y_train)
```

```
> GridSearchCV
> estimator: DecisionTreeClassifier
>     DecisionTreeClassifier
```









## ▼ Model\_Building --- Random Forest Classifier

```
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, 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(
sklearn.utils._param_validation.InvalidParameterError: The 'max_features' parameter of RandomForestClassifier must be an int in the rang

  warnings.warn(some_fits_failed_message, FitFailedWarning)
/usr/local/lib/python3.10/dist-packages/sklearn/model_selection/_search.py:952: UserWarning: One or more of the test scores are non-fini
  0.86136462 0.86392148 0.86051717 0.8622483 0.85287556 0.85370129
  0.85287556 0.85882949      nan 0.85543242 0.85628712 0.85794582
  0.85624366 0.85538896 0.85882225 0.86220484 0.85710561 0.86137911
  0.86053165 0.85627264 0.8656164 0.86222657      nan 0.85033319
  0.85795306 0.85794582 0.85880776 0.86050992 0.85627264 0.85626539
  0.86476894 0.86137187 0.86393597 0.86050268 0.85883674 0.85374475
      nan 0.85373751 0.85966971 0.85880776 0.86049544 0.85882225
  0.86137187 0.85625815 0.86221208 0.86223381 0.85880776 0.8520281
  0.85541793 0.86308127      nan 0.85373751 0.85882225 0.86051717
  0.8621976 0.85966247 0.85882949 0.85371578 0.85796755 0.85798204
  0.85882949 0.85712009 0.85884398 0.85880776]
  warnings.warn(
    GridSearchCV
    estimator: RandomForestClassifier
      RandomForestClassifier
)

```

```
pred3 = rfc_cv.predict(x_test)
```

```
print(classification_report(y_test, pred3))
```

	precision	recall	f1-score	support
0	0.87	0.96	0.91	245
1	0.54	0.27	0.36	49
accuracy			0.84	294
macro avg	0.70	0.61	0.63	294
weighted avg	0.81	0.84	0.82	294

```
probability=rfc_cv.predict_proba(x_test)[:,1] # probability of all the yes values
probability
```

```
array([1.4484457e-02, 7.18997774e-02, 1.63456920e-01, 1.28918317e-01,
 8.40000000e-01, 1.23878736e-01, 6.20384615e-01, 3.18984916e-02,
2.63016545e-03, 1.13679194e-01, 1.99574428e-03, 1.62936581e-01,
9.86635353e-02, 7.03888889e-01, 2.32222222e-01, 3.32416721e-02,
9.70126549e-02, 4.67050000e-01, 1.36930551e-02, 1.41492588e-01,
6.65663781e-01, 1.60486101e-02, 7.16864623e-02, 2.66358004e-02,
3.11074830e-01, 7.38142402e-02, 5.61164804e-02, 3.15762543e-02,
6.94477124e-01, 4.57037414e-02, 6.37677132e-02, 2.25724863e-02,
2.63760358e-01, 1.24089365e-01, 9.56206544e-03, 7.58315547e-02,
6.86789995e-02, 1.24320652e-01, 1.24154089e-01, 7.01970692e-02,
7.51464221e-02, 2.02172371e-02, 1.33426007e-01, 1.09248210e-01,
4.95588777e-02, 5.82911111e-01, 1.48914581e-01, 7.74037177e-02,
7.30000000e-01, 5.40588235e-01, 3.13474026e-01, 6.16821446e-01,
1.24582454e-01, 3.01066611e-02, 4.73750000e-01, 1.24002788e-01,
6.43441018e-03, 2.43915285e-02, 7.20310794e-02, 1.07224391e-01,
5.00692848e-04, 1.74999172e-01, 1.23677449e-01, 1.52577317e-01,
```

```

4.016856100e-01, 7.303141470e-02, 2.00111111e-01, 1.60928686e-01,
1.035427020e-01, 9.09038900e-02, 2.06187491e-01, 6.95000000e-01,
8.01908978e-02, 6.43922113e-03, 2.03245393e-02, 5.59955845e-02,
8.13606929e-02, 1.90508046e-01, 1.18808662e-01, 3.30913689e-03,
3.72477777e-03, 6.39074762e-03, 8.68536899e-02, 1.64147662e-01,
4.28307597e-02, 9.52109758e-02, 5.02057353e-02, 4.86000000e-01,
6.52498355e-02, 2.03708961e-01, 4.92222222e-01, 2.08576354e-03,
1.09746524e-01, 1.79154457e-01, 1.01683388e-01, 2.94850254e-02,
9.22986003e-02, 1.73766759e-01, 6.45540791e-02, 2.92723386e-02,
2.63758307e-01, 3.67567254e-01, 4.80000000e-01, 1.02498772e-02,
2.46758317e-02, 6.185580337e-02, 1.62275488e-02, 5.37935401e-01,
1.54466458e-01, 1.26318587e-01, 2.32792182e-01, 4.71307113e-01,
5.62904338e-02, 5.30755828e-02, 9.54022754e-02, 3.52410686e-02,
1.51221755e-01, 1.19914927e-01, 1.29421728e-02, 8.83275382e-03,
3.71417665e-02, 1.04258879e-01, 1.87029472e-02, 8.92222222e-01,
1.05323030e-01, 1.58076205e-01, 3.75798615e-02, 1.12875845e-01,
5.15371606e-02, 2.17453780e-02, 1.84680314e-01, 4.10270270e-01,
2.54463065e-01, 3.60092427e-01, 2.60420638e-01, 2.70000000e-01,
3.14768803e-01, 5.71115936e-02, 1.61895110e-01, 6.21797486e-02,
9.93960761e-02, 1.30133333e-01, 1.53503224e-01, 5.24528242e-02,
4.56666667e-01, 3.19640356e-02, 1.00270270e-01, 1.32919365e-01,
1.53739935e-02, 1.49171114e-01, 1.29164571e-01, 2.82795788e-01,
1.03522265e-01, 2.70507113e-01, 1.37421215e-01, 6.57593026e-02,
2.72411380e-01, 1.09921230e-02, 1.02431060e-01, 4.72988296e-02,
4.40896154e-01, 3.44616473e-03, 3.33557717e-01, 6.30000000e-01,
1.93333333e-01, 1.66289545e-01, 7.15969326e-02, 9.51442013e-02,
1.74575472e-02, 6.11530259e-02, 2.64650948e-03, 3.89281837e-02,
1.21267745e-01, 4.63602731e-02, 2.64386612e-01, 5.77369697e-01,
2.39916573e-02, 4.80998916e-01, 6.65107862e-02, 8.08120476e-02,
3.71555365e-02, 2.57822808e-02, 6.07937722e-02, 5.68656417e-01,
4.98246304e-02, 7.877965770e-03, 2.76901742e-01, 2.07702066e-01,
3.99414532e-01, 1.14980227e-01, 5.67967427e-01, 9.00000000e-01,
2.51491803e-01, 4.69629287e-03, 1.25038336e-01, 4.20245874e-02,
1.56563046e-02, 2.45855407e-02, 4.25909091e-01, 1.27805374e-01,
1.70535847e-01, 1.12292449e-01, 3.50270112e-01, 4.35831058e-02,
9.43722741e-02, 9.09906510e-02, 5.55357143e-01, 3.20782743e-02,
8.24113071e-02, 9.72946829e-03, 1.41188883e-01, 8.97116215e-02,
4.06638808e-02, 9.76430134e-02, 3.74546339e-02, 7.52108967e-02,
1.20393606e-02, 1.94894141e-01, 4.33815006e-02, 6.20000000e-01,
1.16536183e-01, 2.03037097e-01, 7.30666667e-01, 6.10442260e-02,
1.56135470e-01, 7.63398245e-02, 2.28315651e-02, 2.62212681e-01,
1.202702700e-01 2.1112316160e-02 1.12879260e-02 1.06959776e-01

```

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()

```

