

```
# Rayidi Abhiram
# 21BCE9261
# VITAP MORNING SLOT
# ASSIGNMENT-4
# Performing Multiple models (Logistic Regression, Decision Tree Regressor, Random Forest Regressor) on Employee Attrition Dataset
```

Import neccessary libraries

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

Import the dataset

```
df = pd.read_csv("/content/drive/MyDrive/Colab Datasets/WA_Fn-UseC_-HR-Employee-Attrition.csv")
```

```
df.head()
```

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	EducationalField
0	41	Yes	Travel_Rarely	1102	Sales	1	Business_University
1	49	No	Travel_Frequently	279	Research & Development	8	Business_University
2	37	Yes	Travel_Rarely	1373	Research & Development	2	Business_University
3	33	No	Travel_Frequently	1392	Research & Development	3	Business_University
4	27	No	Travel_Rarely	591	Research & Development	2	Business_University

5 rows × 35 columns

```
df.shape
```

(1470, 35)

```
df.info()
```

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

```
32 YearsInCurrentRole      1470 non-null   int64
33 YearsSinceLastPromotion  1470 non-null   int64
34 YearsWithCurrManager    1470 non-null   int64
dtypes: int64(26), object(9)
memory usage: 402.1+ KB
```

```
df.describe()
```

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

8 rows × 26 columns

Handling null values

```
df.isnull().any()
# No null values found
```

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()
# Since the dataset does not contain any null values, we can proceed to next step.
```

Age	0
Attrition	0
BusinessTravel	0
DailyRate	0
Department	0
DistanceFromHome	0
Education	0
EducationField	0
EmployeeCount	0
EmployeeNumber	0
EnvironmentSatisfaction	0
Gender	0
HourlyRate	0

```
JobInvolvement      0
JobLevel            0
JobRole             0
JobSatisfaction      0
MaritalStatus       0
MonthlyIncome       0
MonthlyRate         0
NumCompaniesWorked  0
Over18              0
OverTime            0
PercentSalaryHike    0
PerformanceRating    0
RelationshipSatisfaction 0
StandardHours       0
StockOptionLevel    0
TotalWorkingYears   0
TrainingTimesLastYear 0
WorkLifeBalance     0
YearsAtCompany      0
YearsInCurrentRole  0
YearsSinceLastPromotion 0
YearsWithCurrManager 0
dtype: int64
```

Data Visualization

```
df.corr()

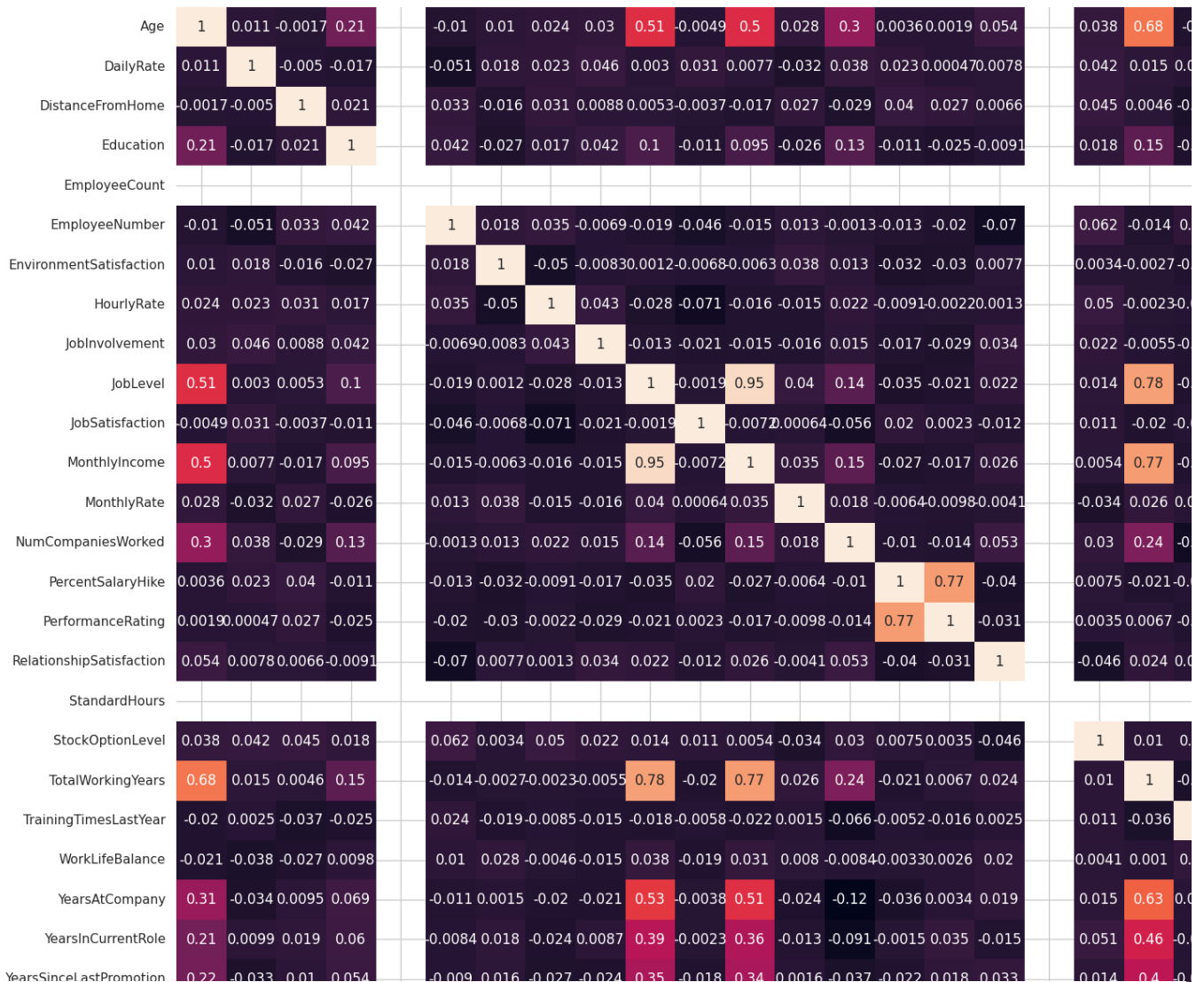
<ipython-input-210-2f6f6606aa2c>:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version of pandas, only DataFrame.corr(numeric=True) will be allowed.
df.corr()
```

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

26 rows × 26 columns

```
plt.figure(figsize=(25,16))
sns.heatmap(df.corr(),annot=True)
```

```
<ipython-input-211-1020bb8eeb59>:2: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version of pandas, this will be set to False. You can avoid this warning by setting numeric_only=True.
sns.heatmap(df.corr(),annot=True)
<Axes: >
```



```
categorical_columns = df.select_dtypes(include=['object']).columns
```

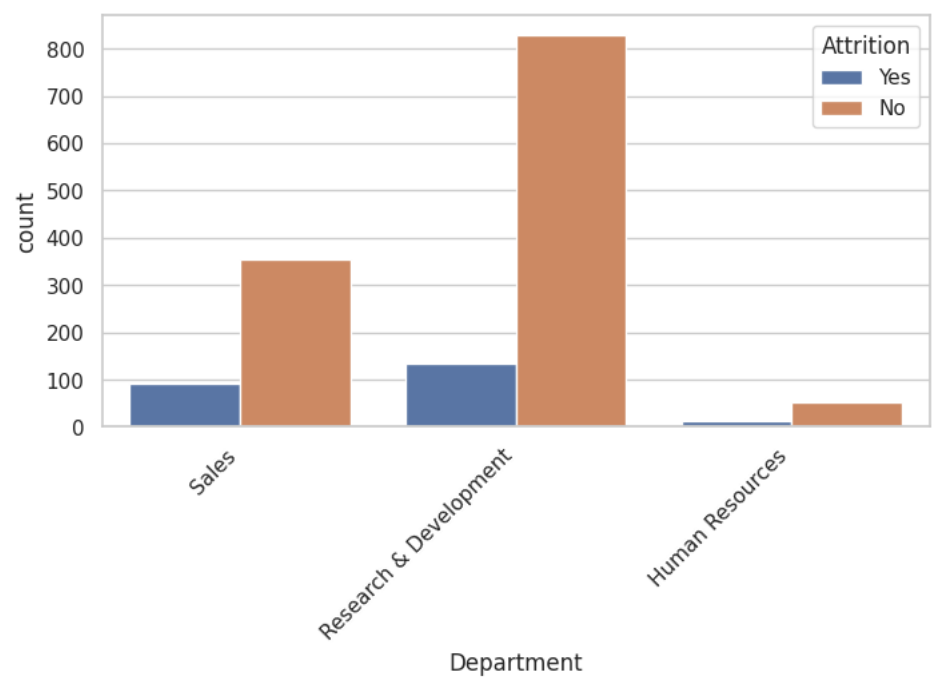
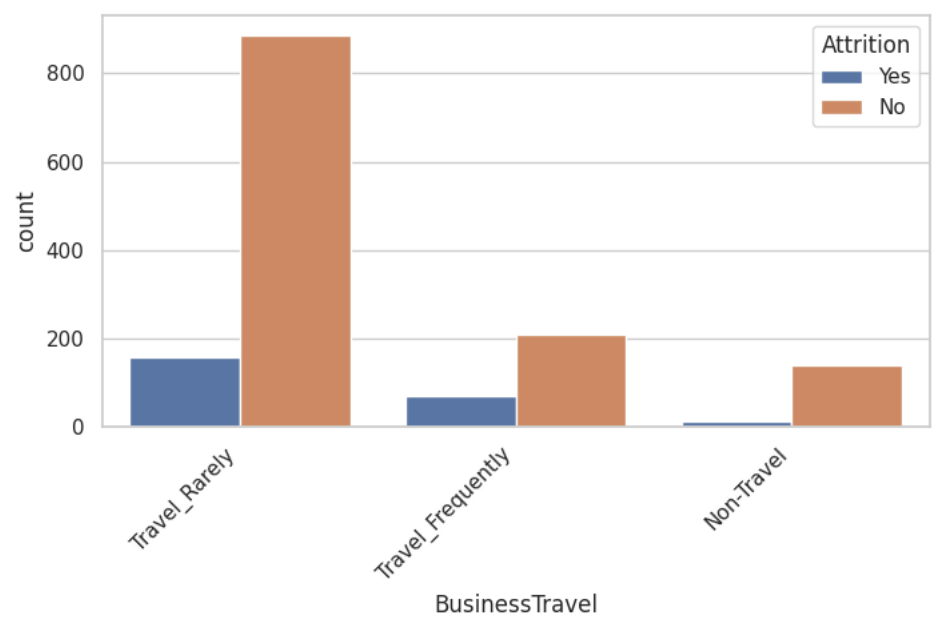
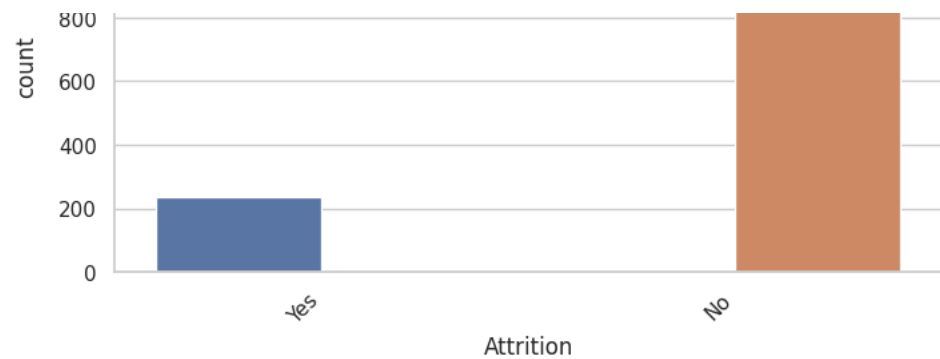
```
for col in categorical_columns:
    plt.subplots(figsize=(8, 4))
    sns.countplot(data=df, x=col, hue="Attrition")
    plt.xticks(rotation=45, ha='right')
```

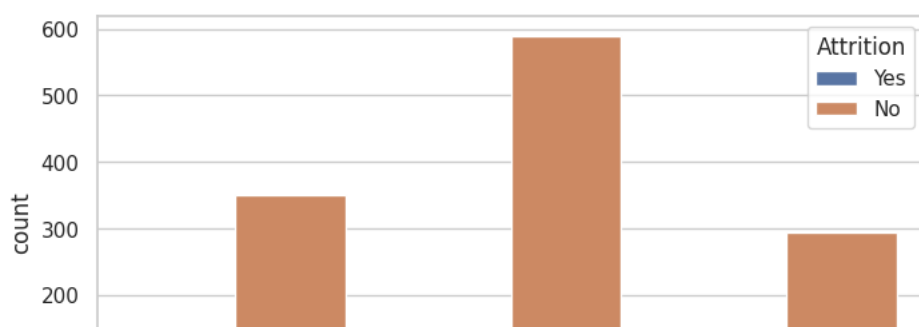
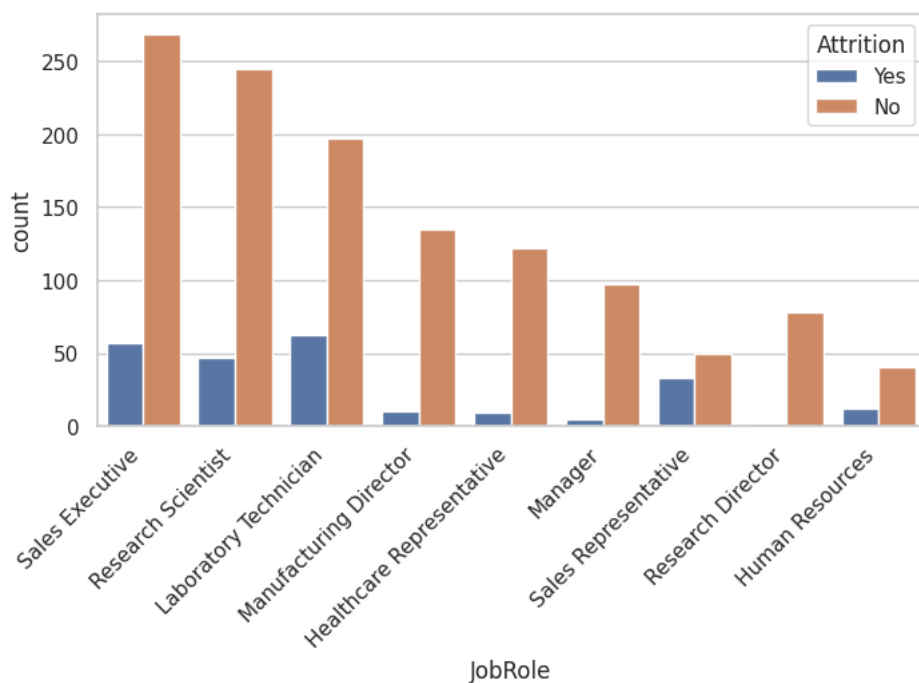
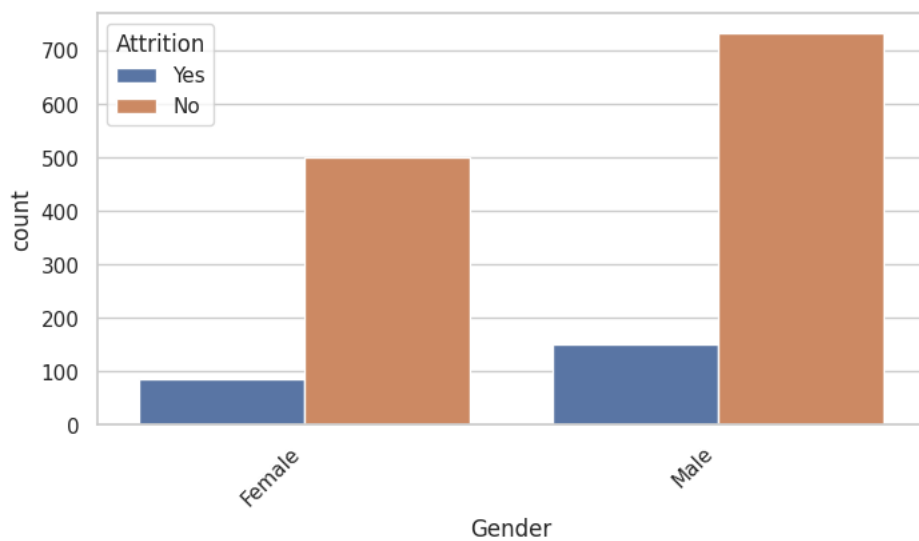
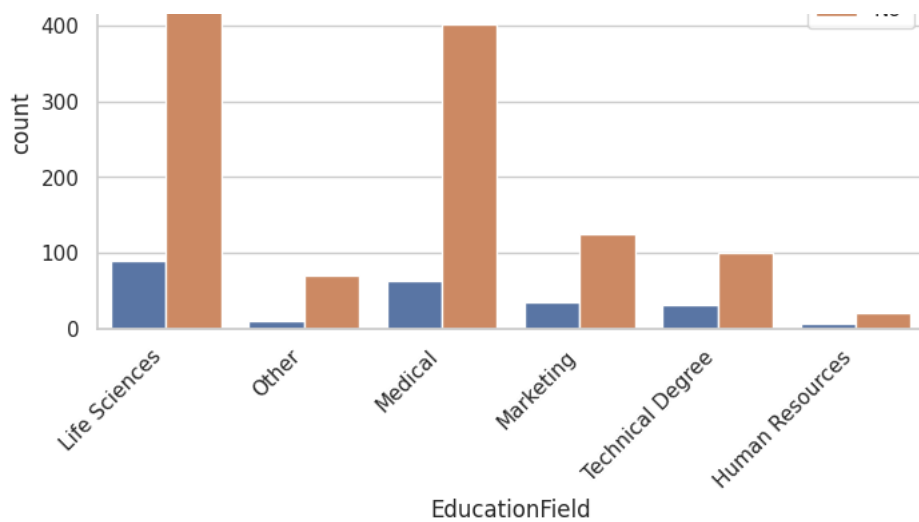
```
"""
```

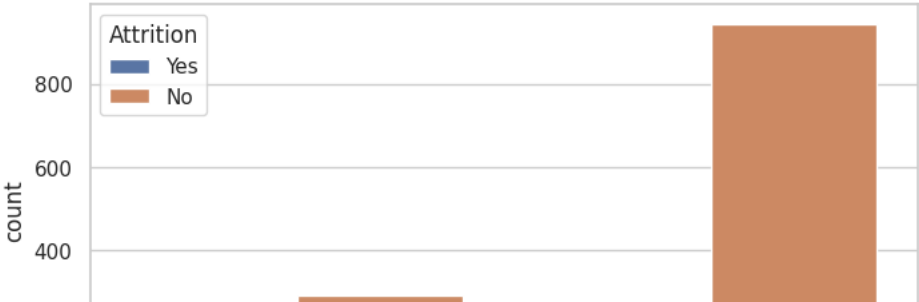
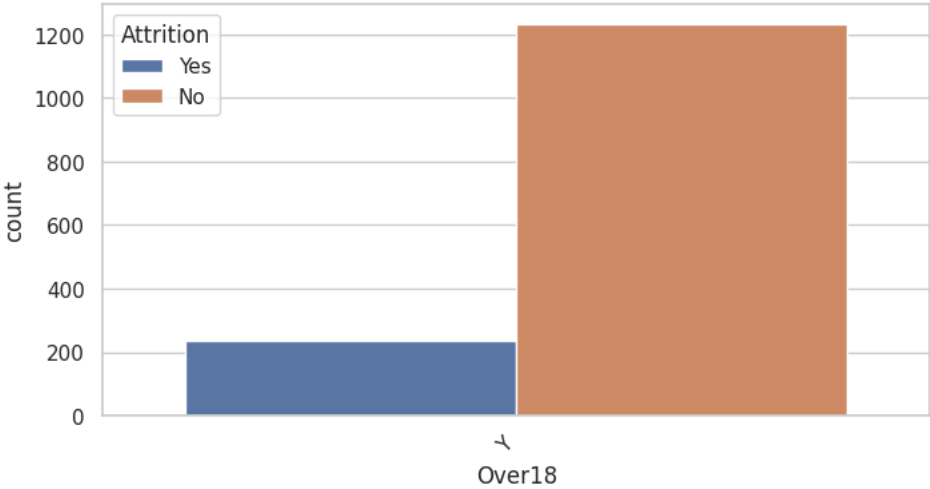
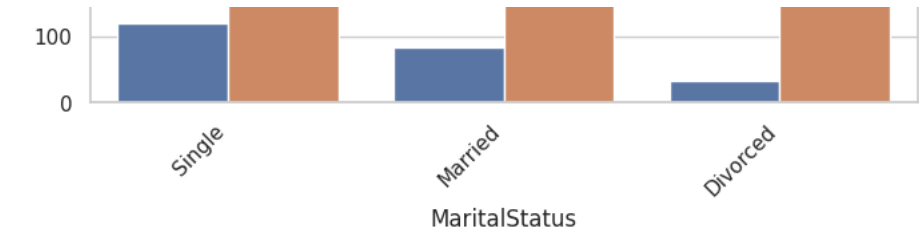
Inference :

- 1) We can say that the employees who travel frequently are more likely to be Attrited followed by travels rarely followed by non travel.  
So people who not travel has less Attrtion.
- 2) Similarly there isn't much linear trend regarding Department and Attrition.
- 3) Intrestingly the count of Attrition is more in Research and Development but the number changes when comes to percentage.
- 4) Also the gender may not affect much Attrition.
- 5) Research Director Designation has less Attrition which is leading to security of role, Also we can find that more than 50% are Attrited in Sales Representative Designation.
- 6) Not so suprised by how the Bachelors has been most Attrited compared to Married and Divorced.
- 7) Not much progressed by the analytics of over18.
- 8) Major point founded is that people who do overtime has more possibility of Attrition.

```
"""
```

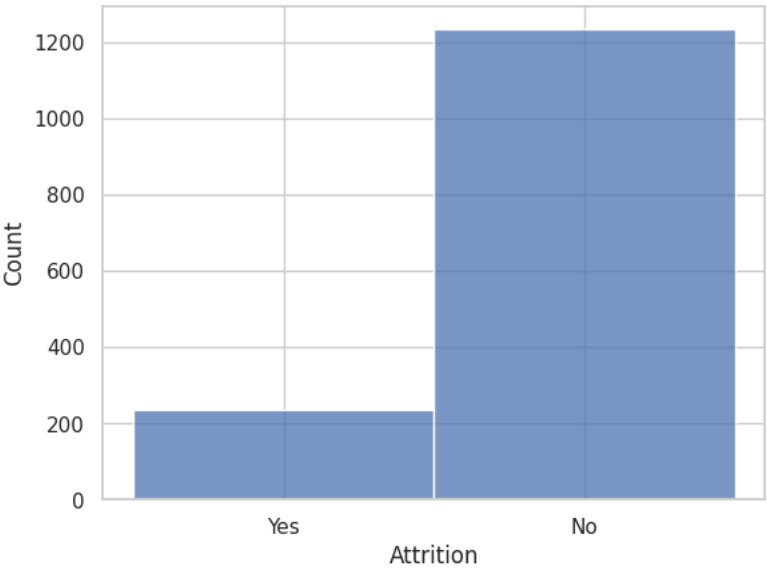






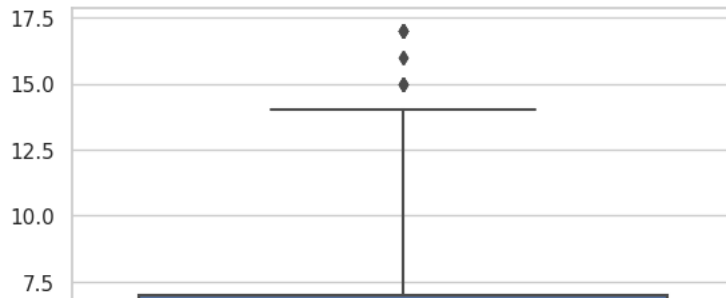
```
sns.histplot(df["Attrition"])
```

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



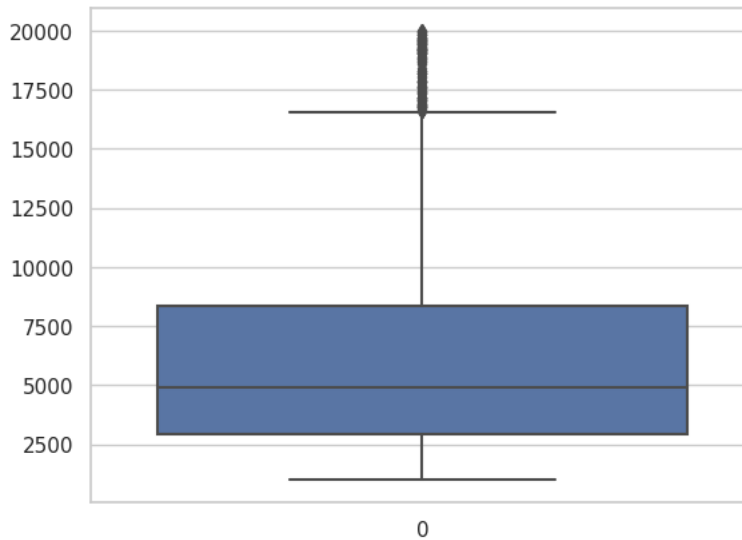
```
sns.boxplot(df.YearsWithCurrManager )
```

&lt;Axes: &gt;



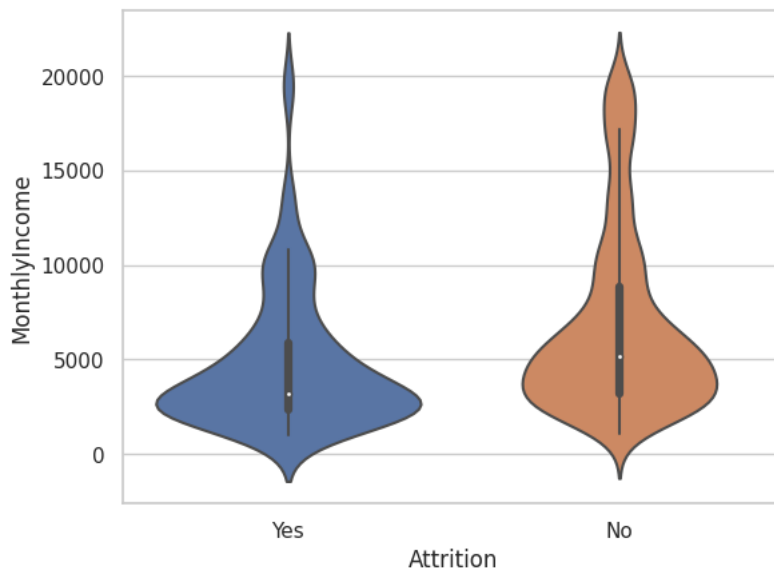
```
sns.boxplot(df.MonthlyIncome )
```

&lt;Axes: &gt;



```
sns.violinplot(x= "Attrition", y= "MonthlyIncome", data=df)
```

&lt;Axes: xlabel='Attrition', ylabel='MonthlyIncome'&gt;



### Splitting Dependent and Independent variables

```
x = df.drop(columns=['Attrition', 'DailyRate', 'StockOptionLevel', 'StandardHours', 'Over18'])
```

```
y = df['Attrition']
```

```
x.head()
```



	Age	BusinessTravel	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	EnvironmentSatisfac
0	41	Travel_Rarely	Sales	1	2	Life Sciences	1	1	
1	49	Travel_Frequently	Research & Development	8	1	Life Sciences	1	2	
2	37	Travel_Rarely	Research & Development	2	2	Other	1	4	
3	33	Travel_Frequently	Research & Development	3	4	Life Sciences	1	5	
4	27	Travel_Rarely	Research & Development	2	1	Medical	1	7	

```

y.head()

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

```

## Encoding

```

from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()

x["Gender"] = le.fit_transform(x["Gender"])
x["BusinessTravel"] = le.fit_transform(x["BusinessTravel"])
x['Department'] = le.fit_transform(x['Department'])
x['EducationField'] = le.fit_transform(x['EducationField'])
x['JobRole'] = le.fit_transform(x['JobRole'])
x['MaritalStatus'] = le.fit_transform(x['MaritalStatus'])
x['OverTime'] = le.fit_transform(x['OverTime'])
y = le.fit_transform(y)

```

```

x.shape

(1470, 30)

```

## Feature Scaling

```

from sklearn.preprocessing import StandardScaler

sc = StandardScaler()

x_scaled = pd.DataFrame(sc.fit_transform(x), columns = x.columns)

x_scaled

```

	Age	BusinessTravel	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	Environments
0	0.446350	0.590048	1.401512	-1.010909	-0.891688	-0.937414	0.0	-1.701283	
1	1.322365	-0.913194	-0.493817	-0.147150	-1.868426	-0.937414	0.0	-1.699621	
2	0.008343	0.590048	-0.493817	-0.887515	-0.891688	1.316673	0.0	-1.696298	
3	-0.429664	-0.913194	-0.493817	-0.764121	1.061787	-0.937414	0.0	-1.694636	
4	-1.086676	0.590048	-0.493817	-0.887515	-1.868426	0.565311	0.0	-1.691313	
...	...	...	...	...	...	...	...	...	...
1465	-0.101159	-0.913194	-0.493817	1.703764	-0.891688	0.565311	0.0	1.721670	
1466	0.227347	0.590048	-0.493817	-0.393938	-1.868426	0.565311	0.0	1.723332	
1467	-1.086676	0.590048	-0.493817	-0.640727	0.085049	-0.937414	0.0	1.726655	
1468	1.322365	-0.913194	1.401512	-0.887515	0.085049	0.565311	0.0	1.728317	
1469	-0.320163	0.590048	-0.493817	-0.147150	0.085049	0.565311	0.0	1.733302	

1470 rows × 30 columns

## Splitting Testing and Training 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)
```

```
x_train.shape, x_test.shape, y_train.shape, y_test.shape

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

## Model Building

### Logistic Regression

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

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

```
▼ LogisticRegression
LogisticRegression()
```

```
y_pred= lo.predict(x_test)
```

### Evaluation Metrics

```
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report, roc_auc_score, roc_curve
accuracy_score(y_test, y_pred)
```

```
0.8843537414965986
```

```
confusion_matrix(y_test, y_pred)
```

```
array([[240,  5],
       [ 29, 20]])
```

```
print(classification_report(y_test, y_pred))
```

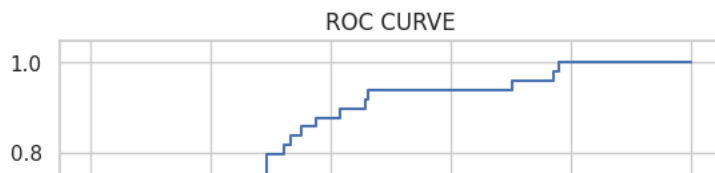
	precision	recall	f1-score	support
0	0.89	0.98	0.93	245
1	0.80	0.41	0.54	49
accuracy			0.88	294
macro avg	0.85	0.69	0.74	294
weighted avg	0.88	0.88	0.87	294

```
# ROC-AUC Curve
```

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

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

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



### Decision Tree Regressor

```
from sklearn.tree import DecisionTreeClassifier
dt_model = DecisionTreeClassifier()
dt_model.fit(x_train, y_train)

dt_pred = dt_model.predict(x_test)
```

### Evaluation Metrics

```
accuracy_score(y_test, dt_pred)
```

```
0.7312925170068028
```

```
confusion_matrix(y_test, dt_pred)
```

```
array([[201, 44],
       [ 35, 14]])
```

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

```
col_0    0    1
row_0
0      201  44
1       35  14
```

```
print(classification_report(y_test,y_pred))
```

```

              precision    recall  f1-score   support

     0       0.89      0.98      0.93        245
     1       0.80      0.41      0.54         49

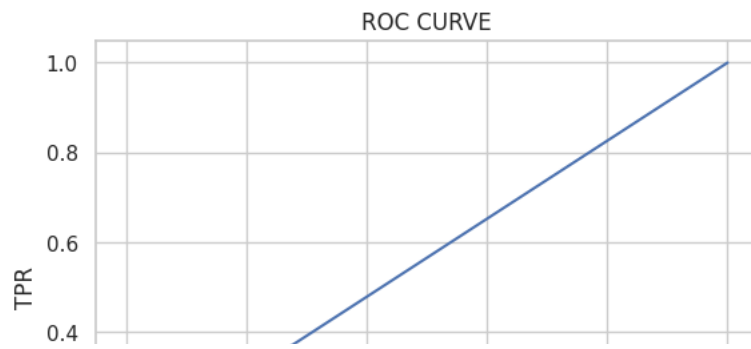
 accuracy      0.88
 macro avg     0.85      0.69      0.74        294
 weighted avg  0.88      0.88      0.87        294
```

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

```
# 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()
```



#### Parameter Tuning

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

```
[Text(0.322849025974026, 0.972222222222222, 'x[23] <= -1.257\ngini = 0.269\nsamples = 1176\nvalue = [988, 188]'),
Text(0.07142857142857142, 0.9166666666666666, 'x[15] <= 0.552\ngini = 0.5\nsamples = 78\nvalue = [39, 39]'),
Text(0.04220779220779221, 0.8611111111111112, 'x[3] <= 0.902\ngini = 0.426\nsamples = 39\nvalue = [27, 12]'),
Text(0.025974025974025976, 0.8055555555555556, 'x[14] <= -1.114\ngini = 0.312\nsamples = 31\nvalue = [25, 6]'),
Text(0.12987012987012988, 0.75, 'x[16] <= -0.937\ngini = 7\nvalue = [3, 4]'),
Text(0.006493506493506494, 0.6944444444444444, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.01948051948051948, 0.6944444444444444, 'x[17] <= -0.838\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.012987012987012988, 0.6388888888888888, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.025974025974025976, 0.6388888888888888, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.03896103896103896, 0.75, 'x[18] <= -0.878\ngini = 0.153\nsamples = 24\nvalue = [22, 2]'),
Text(0.032467532467532464, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.045454545454545456, 0.6944444444444444, 'x[8] <= -1.118\ngini = 0.083\nsamples = 23\nvalue = [22, 1]'),
Text(0.03896103896103896, 0.6388888888888888, 'x[21] <= 0.96\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.032467532467532464, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.045454545454545456, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.05194805194805195, 0.6388888888888888, 'gini = 0.0\nsamples = 21\nvalue = [21, 0]'),
Text(0.05844155844155844, 0.8055555555555556, 'x[7] <= -0.378\ngini = 0.375\nsamples = 8\nvalue = [2, 6]'),
Text(0.05194805194805195, 0.75, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.06493506493506493, 0.75, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.10064935064935066, 0.8611111111111112, 'x[10] <= -0.511\ngini = 0.426\nsamples = 39\nvalue = [12, 27]'),
Text(0.08441558441558442, 0.8055555555555556, 'x[16] <= -0.763\ngini = 0.133\nsamples = 14\nvalue = [1, 13]'),
Text(0.07792207792207792, 0.75, 'gini = 0.0\nsamples = 13\nvalue = [0, 13]'),
Text(0.09090909090909091, 0.75, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.11688311688311688, 0.8055555555555556, 'x[7] <= -1.341\ngini = 0.493\nsamples = 25\nvalue = [11, 14]'),
Text(0.1038961038961039, 0.75, 'x[19] <= 0.482\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
Text(0.09740259740259741, 0.6944444444444444, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.11038961038961038, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.12987012987012986, 0.75, 'x[14] <= -0.207\ngini = 0.432\nsamples = 19\nvalue = [6, 13]'),
Text(0.12337662337662338, 0.6944444444444444, 'gini = 0.0\nsamples = 7\nvalue = [0, 7]'),
Text(0.13636363636363635, 0.6944444444444444, 'x[5] <= -0.186\ngini = 0.5\nsamples = 12\nvalue = [6, 6]'),
Text(0.12337662337662338, 0.6388888888888888, 'x[11] <= -1.729\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
Text(0.11688311688311688, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.12987012987012986, 0.5833333333333334, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.14935064935064934, 0.6388888888888888, 'x[7] <= -0.845\ngini = 0.278\nsamples = 6\nvalue = [1, 5]'),
Text(0.14285714285714285, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.15584415584415584, 0.5833333333333334, 'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
Text(0.5742694805194806, 0.9166666666666666, 'x[19] <= 0.482\ngini = 0.235\nsamples = 1098\nvalue = [949, 149]'),
Text(0.3234070616883117, 0.8611111111111112, 'x[25] <= -1.786\ngini = 0.162\nsamples = 798\nvalue = [727, 71]'),
Text(0.16883116883116883, 0.8055555555555556, 'x[7] <= -0.173\ngini = 0.38\nsamples = 47\nvalue = [35, 12]'),
Text(0.15584415584415584, 0.75, 'x[15] <= 0.552\ngini = 0.1\nsamples = 19\nvalue = [18, 1]'),
Text(0.14935064935064934, 0.6944444444444444, 'gini = 0.0\nsamples = 18\nvalue = [18, 0]'),
Text(0.16233766233766234, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.18181818181818182, 0.75, 'x[16] <= -0.789\ngini = 0.477\nsamples = 28\nvalue = [17, 11]'),
Text(0.17532467532467533, 0.6944444444444444, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),
Text(0.18831168831168832, 0.6944444444444444, 'x[7] <= 0.099\ngini = 0.413\nsamples = 24\nvalue = [17, 7]'),
Text(0.18181818181818182, 0.6388888888888888, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.19480519480519481, 0.6388888888888888, 'x[29] <= 0.386\ngini = 0.351\nsamples = 22\nvalue = [17, 5]'),
Text(0.18181818181818182, 0.5833333333333334, 'x[24] <= 0.544\ngini = 0.133\nsamples = 14\nvalue = [13, 1]'),
Text(0.17532467532467533, 0.5277777777777778, 'gini = 0.0\nsamples = 12\nvalue = [12, 0]'),
Text(0.18831168831168832, 0.5277777777777778, 'x[5] <= 0.565\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.18181818181818182, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.19480519480519481, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.2077922077922078, 0.5833333333333334, 'x[18] <= -0.878\ngini = 0.5\nsamples = 8\nvalue = [4, 4]'),
Text(0.2012987012987013, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.21428571428571427, 0.5277777777777778, 'x[8] <= -0.203\ngini = 0.444\nsamples = 6\nvalue = [2, 4]'),
Text(0.2077922077922078, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.22077922077922077, 0.4722222222222222, 'x[23] <= 1.893\ngini = 0.32\nsamples = 5\nvalue = [1, 4]'),
Text(0.21428571428571427, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),
Text(0.22727272727272727, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.47798295454545453, 0.8055555555555556, 'x[23] <= 3.564\ngini = 0.145\nsamples = 751\nvalue = [692, 59]'),
Text(0.47148944805194803, 0.75, 'x[26] <= -0.41\ngini = 0.143\nsamples = 750\nvalue = [692, 58]'),
Text(0.3449675324675325, 0.6944444444444444, 'x[8] <= -1.118\ngini = 0.218\nsamples = 257\nvalue = [225, 32]'),
Text(0.30357142857142855, 0.6388888888888888, 'x[29] <= -0.455\ngini = 0.355\nsamples = 65\nvalue = [50, 15]'),
Text(0.2824675324675325, 0.5833333333333334, 'x[29] <= -1.016\ngini = 0.303\nsamples = 59\nvalue = [48, 11]'),
Text(0.2597402597402597, 0.5277777777777778, 'x[11] <= -0.323\ngini = 0.463\nsamples = 22\nvalue = [14, 8]'),
Text(0.24675324675324675, 0.4722222222222222, 'x[10] <= -1.151\ngini = 0.198\nsamples = 9\nvalue = [8, 1]'),
Text(0.24025974025974026, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.2532467532467532, 0.4166666666666667, 'gini = 0.0\nsamples = 8\nvalue = [8, 0]'),
Text(0.2727272727272727, 0.4722222222222222, 'x[10] <= -0.388\ngini = 0.497\nsamples = 13\nvalue = [6, 7]'),
Text(0.2662337662337662, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.2792207792207792, 0.4166666666666667, 'x[3] <= -0.024\ngini = 0.346\nsamples = 9\nvalue = [2, 7]'),
Text(0.2727272727272727, 0.3611111111111111, 'x[23] <= -0.679\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.2662337662337662, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.2792207792207792, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.2857142857142857, 0.3611111111111111, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.3051948051948052, 0.5277777777777778, 'x[14] <= -1.114\ngini = 0.149\nsamples = 37\nvalue = [34, 3]'),
Text(0.2987012987012987, 0.4722222222222222, 'x[25] <= -0.37\ngini = 0.5\nsamples = 6\nvalue = [3, 3]'),
Text(0.2922077922077922, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.3051948051948052, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.3116883116883117, 0.4722222222222222, 'gini = 0.0\nsamples = 31\nvalue = [31, 0]'),
Text(0.3246753246753247, 0.5833333333333334, 'x[7] <= -1.479\ngini = 0.444\nsamples = 6\nvalue = [2, 4]'),
Text(0.3181818181818182, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.33116883116883117, 0.5277777777777778, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),
Text(0.38636363636363635, 0.6388888888888888, 'x[0] <= -0.594\ngini = 0.161\nsamples = 192\nvalue = [175, 17]'),
Text(0.35064935064935066, 0.5833333333333334, 'x[5] <= -1.313\ngini = 0.294\nsamples = 67\nvalue = [55, 12]'),
Text(0.34415584415584416, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.35714285714285715, 0.5277777777777778, 'x[25] <= -0.37\ngini = 0.26\nsamples = 65\nvalue = [55, 10]'),
Text(0.32792207792207795, 0.4722222222222222, 'x[5] <= 0.19\ngini = 0.469\nsamples = 16\nvalue = [10, 6]'),
Text(0.32142857142857145, 0.4166666666666667, 'gini = 0.0\nsamples = 7\nvalue = [7, 0]'),
Text(0.3344155844155844, 0.4166666666666667, 'x[8] <= 0.712\ngini = 0.444\nsamples = 9\nvalue = [3, 6]'),
```

```
Text(0.32792207792207795, 0.3611111111111111, 'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
Text(0.3409090909090909, 0.3611111111111111, 'x[11] <= -0.323\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.3344155844155844, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.3474025974025974, 0.3055555555555556, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.3863636363636363, 0.4722222222222222, 'x[8] <= -0.203\ngini = 0.15\nsamples = 49\nvalue = [45, 4]'),
Text(0.3733766233766233, 0.4166666666666667, 'x[16] <= -0.715\ngini = 0.444\nsamples = 6\nvalue = [4, 2]'),
Text(0.3668831168831168, 0.3611111111111111, 'x[14] <= -1.114\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.3638961038961037, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.3733766233766233, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.3798701298701298, 0.3611111111111111, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.3993506493506493, 0.4166666666666667, 'x[4] <= 1.55\ngini = 0.089\nsamples = 43\nvalue = [41, 2]'),
Text(0.3928571428571428, 0.3611111111111111, 'x[20] <= -1.014\ngini = 0.046\nsamples = 42\nvalue = [41, 1]'),
Text(0.3863636363636363, 0.3055555555555556, 'x[3] <= -0.456\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.3798701298701298, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.3928571428571428, 0.25, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.3993506493506493, 0.3055555555555556, 'gini = 0.0\nsamples = 39\nvalue = [39, 0]'),
Text(0.4058441558441558, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.4220779220779220, 0.5833333333333334, 'x[7] <= -1.627\ngini = 0.077\nsamples = 125\nvalue = [120, 5]'),
Text(0.4058441558441558, 0.5277777777777778, 'x[13] <= -0.186\ngini = 0.5\nsamples = 4\nvalue = [2, 2]'),
Text(0.3993506493506493, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.4123376623376623, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.4383116883116883, 0.5277777777777778, 'x[17] <= 1.671\ngini = 0.048\nsamples = 121\nvalue = [118, 3]'),
Text(0.4253246753246753, 0.4722222222222222, 'x[13] <= 0.83\ngini = 0.033\nsamples = 118\nvalue = [116, 2]'),
Text(0.4188311688311688, 0.4166666666666667, 'gini = 0.0\nsamples = 95\nvalue = [95, 0]'),
Text(0.4318181818181818, 0.4166666666666667, 'x[11] <= 1.083\ngini = 0.159\nsamples = 23\nvalue = [21, 2]'),
Text(0.4253246753246753, 0.3611111111111111, 'x[25] <= -0.377\ngini = 0.087\nsamples = 22\nvalue = [21, 1]'),
Text(0.4188311688311688, 0.3055555555555556, 'x[14] <= 0.7\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.4123376623376623, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.4253246753246753, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.4318181818181818, 0.3055555555555556, 'gini = 0.0\nsamples = 20\nvalue = [20, 0]'),
Text(0.4383116883116883, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.4512987012987013, 0.4722222222222222, 'x[27] <= -0.892\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.4448051948051948, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.4577922077922078, 0.4166666666666667, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.5980113636363636, 0.6944444444444444, 'x[26] <= 3.999\ngini = 0.1\nsamples = 493\nvalue = [467, 26]'),
Text(0.5596590909090909, 0.6388888888888888, 'x[14] <= -0.207\ngini = 0.094\nsamples = 486\nvalue = [462, 24]'),
Text(0.502435064935065, 0.5833333333333334, 'x[13] <= 1.236\ngini = 0.154\nsamples = 191\nvalue = [175, 16]'),
Text(0.4959415584415584, 0.5277777777777778, 'x[17] <= -0.035\ngini = 0.145\nsamples = 190\nvalue = [175, 15]'),
Text(0.4788961038961039, 0.4722222222222222, 'x[17] <= -0.073\ngini = 0.221\nsamples = 95\nvalue = [83, 12]'),
Text(0.4724025974025974, 0.4166666666666667, 'x[29] <= 2.629\ngini = 0.207\nsamples = 94\nvalue = [83, 11]'),
Text(0.4659090909090909, 0.3611111111111111, 'x[4] <= -0.403\ngini = 0.192\nsamples = 93\nvalue = [83, 10]'),
Text(0.4448051948051948, 0.3055555555555556, 'x[5] <= 1.692\ngini = 0.363\nsamples = 21\nvalue = [16, 5]'),
Text(0.4383116883116883, 0.25, 'x[16] <= 0.501\ngini = 0.266\nsamples = 19\nvalue = [16, 3]'),
Text(0.4253246753246753, 0.1944444444444445, 'x[16] <= -0.597\ngini = 0.117\nsamples = 16\nvalue = [15, 1]'),
Text(0.4188311688311688, 0.1388888888888889, 'x[3] <= 1.334\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.4123376623376623, 0.0833333333333333, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.4253246753246753, 0.0833333333333333, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.4318181818181818, 0.1388888888888889, 'gini = 0.0\nsamples = 14\nvalue = [14, 0]'),
Text(0.4512987012987013, 0.1944444444444445, 'x[26] <= 2.448\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.4448051948051948, 0.1388888888888889, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.4577922077922078, 0.1388888888888889, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.4512987012987013, 0.25, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.487012987012987, 0.3055555555555556, 'x[27] <= -0.477\ngini = 0.129\nsamples = 72\nvalue = [67, 5]'),
Text(0.4707792207792208, 0.25, 'x[7] <= 0.633\ngini = 0.444\nsamples = 6\nvalue = [4, 2]'),
Text(0.4642857142857143, 0.1944444444444445, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.4772727272727273, 0.1944444444444445, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.5032467532467533, 0.25, 'x[10] <= 1.654\ngini = 0.087\nsamples = 66\nvalue = [63, 3]'),
Text(0.4902597402597403, 0.1944444444444445, 'x[24] <= 0.544\ngini = 0.061\nsamples = 64\nvalue = [62, 2]'),
Text(0.4837662337662338, 0.1388888888888889, 'gini = 0.0\nsamples = 51\nvalue = [51, 0]'),
Text(0.4967532467532468, 0.1388888888888889, 'x[8] <= -0.203\ngini = 0.26\nsamples = 13\nvalue = [11, 2]'),
Text(0.4902597402597403, 0.0833333333333333, 'x[27] <= 0.351\ngini = 0.5\nsamples = 4\nvalue = [2, 2]'),
Text(0.4837662337662338, 0.0277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.4967532467532468, 0.0277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.5032467532467533, 0.0833333333333333, 'gini = 0.0\nsamples = 9\nvalue = [9, 0]'),
Text(0.516237662337663, 0.1944444444444445, 'x[9] <= -0.204\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.5097402597402597, 0.1388888888888889, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5227272727272727, 0.1388888888888889, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.4788961038961039, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.4853896103896103, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.512987012987013, 0.4722222222222222, 'x[18] <= 0.724\ngini = 0.061\nsamples = 95\nvalue = [92, 3]'),
Text(0.5064935064935064, 0.4166666666666667, 'gini = 0.0\nsamples = 76\nvalue = [76, 0]'),
Text(0.5194805194805194, 0.4166666666666667, 'x[7] <= -1.149\ngini = 0.266\nsamples = 19\nvalue = [16, 3]'),
Text(0.5064935064935064, 0.3611111111111111, 'x[5] <= 0.19\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.5, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.512987012987013, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.5324675324675324, 0.3611111111111111, 'x[27] <= 2.007\ngini = 0.117\nsamples = 16\nvalue = [15, 1]'),
Text(0.525974025974026, 0.3055555555555556, 'gini = 0.0\nsamples = 14\nvalue = [14, 0]'),
Text(0.538961038961039, 0.3055555555555556, 'x[29] <= 0.106\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.5324675324675324, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5454545454545454, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.5089285714285714, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6168831168831169, 0.5833333333333334, 'x[20] <= -1.014\ngini = 0.053\nsamples = 295\nvalue = [287, 8]'),
Text(0.5941558441558441, 0.5277777777777778, 'x[28] <= 2.58\ngini = 0.159\nsamples = 46\nvalue = [42, 4]'),
Text(0.5876623376623377, 0.4722222222222222, 'x[23] <= 2.085\ngini = 0.124\nsamples = 45\nvalue = [42, 3]'),
Text(0.5714285714285714, 0.4166666666666667, 'x[11] <= -1.729\ngini = 0.089\nsamples = 43\nvalue = [41, 2]'),
Text(0.5584415584415584, 0.3611111111111111, 'x[26] <= 0.733\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.551948051948052, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.564935064935065, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5844155844155844, 0.3611111111111111, 'x[13] <= -1.609\ngini = 0.048\nsamples = 41\nvalue = [40, 1]'),
Text(0.577922077922078, 0.3055555555555556, 'x[8] <= -1.118\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.5714285714285714, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]')
```

```
Text(0.57142031420314, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5844155844155844, 0.25, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.5909090909090909, 0.3055555555555556, 'gini = 0.0\nsamples = 37\nvalue = [37, 0]'),
Text(0.6038961038961039, 0.4166666666666667, 'x[0] <= 1.541\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.5974025974025974, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6103896103896104, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.6006493506493507, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6396103896103896, 0.5277777777777778, 'x[16] <= -0.943\ngini = 0.032\nsamples = 249\nvalue = [245, 4]'),
Text(0.6233766233766234, 0.4722222222222222, 'x[25] <= -0.37\ngini = 0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.6168831168831169, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6298701298701299, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.6558441558441559, 0.4722222222222222, 'x[17] <= 1.198\ngini = 0.024\nsamples = 244\nvalue = [241, 3]'),
Text(0.6428571428571429, 0.4166666666666667, 'x[0] <= 1.268\ngini = 0.01\nsamples = 208\nvalue = [207, 1]'),
Text(0.6363636363636364, 0.3611111111111111, 'gini = 0.0\nsamples = 185\nvalue = [185, 0]'),
Text(0.6493506493506493, 0.3611111111111111, 'x[23] <= -0.036\ngini = 0.083\nsamples = 23\nvalue = [22, 1]'),
Text(0.6428571428571429, 0.3055555555555556, 'x[4] <= 0.573\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.6363636363636364, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6493506493506493, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.6558441558441559, 0.3055555555555556, 'gini = 0.0\nsamples = 21\nvalue = [21, 0]'),
Text(0.6688311688311688, 0.4166666666666667, 'x[17] <= 1.207\ngini = 0.105\nsamples = 36\nvalue = [34, 2]'),
Text(0.6623376623376623, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6753246753246753, 0.3611111111111111, 'x[10] <= -1.372\ngini = 0.056\nsamples = 35\nvalue = [34, 1]'),
Text(0.6688311688311688, 0.3055555555555556, 'x[26] <= 0.162\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.6623376623376623, 0.25, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.6753246753246753, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6818181818181818, 0.3055555555555556, 'gini = 0.0\nsamples = 32\nvalue = [32, 0]'),
Text(0.6363636363636364, 0.6388888888888888, 'x[9] <= -0.204\ngini = 0.408\nsamples = 7\nvalue = [5, 2]'),
Text(0.6298701298701299, 0.5833333333333334, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.6428571428571429, 0.5833333333333334, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.484476461038961, 0.75, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.825318993506493, 0.8611111111111112, 'x[16] <= -0.533\ngini = 0.385\nsamples = 300\nvalue = [222, 78]'),
Text(0.737114448051948, 0.8055555555555556, 'x[11] <= 1.083\ngini = 0.5\nsamples = 96\nvalue = [49, 47]'),
Text(0.7132711038961039, 0.75, 'x[8] <= -1.118\ngini = 0.498\nsamples = 86\nvalue = [40, 46]'),
Text(0.685064935064935, 0.6944444444444444, 'x[10] <= -1.102\ngini = 0.245\nsamples = 14\nvalue = [2, 12]'),
Text(0.6785714285714286, 0.6388888888888888, 'x[18] <= 0.924\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.672077922077922, 0.5833333333333334, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.685064935064935, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6915584415584416, 0.6388888888888888, 'gini = 0.0\nsamples = 11\nvalue = [0, 11]'),
Text(0.7414772727272727, 0.6944444444444444, 'x[0] <= -1.141\ngini = 0.498\nsamples = 72\nvalue = [38, 34]'),
Text(0.7045454545454546, 0.6388888888888888, 'x[28] <= -0.214\ngini = 0.32\nsamples = 10\nvalue = [2, 8]'),
Text(0.698051948051948, 0.5833333333333334, 'gini = 0.0\nsamples = 7\nvalue = [0, 7]'),
Text(0.711038961038961, 0.5833333333333334, 'x[13] <= -0.186\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.7045454545454546, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.7175324675324676, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.7784090909090909, 0.6388888888888888, 'x[20] <= 1.719\ngini = 0.487\nsamples = 62\nvalue = [36, 26]'),
Text(0.7516233766233766, 0.5833333333333334, 'x[28] <= 1.183\ngini = 0.46\nsamples = 53\nvalue = [34, 19]'),
Text(0.7305194805194806, 0.5277777777777778, 'x[18] <= 0.724\ngini = 0.423\nsamples = 46\nvalue = [32, 14]'),
Text(0.7077922077922078, 0.4722222222222222, 'x[7] <= -1.093\ngini = 0.334\nsamples = 33\nvalue = [26, 7]'),
Text(0.6948051948051948, 0.4166666666666667, 'x[23] <= -0.807\ngini = 0.49\nsamples = 7\nvalue = [3, 4]'),
Text(0.6883116883116883, 0.3611111111111111, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),
Text(0.7012987012987013, 0.3611111111111111, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.7207792207792208, 0.4166666666666667, 'x[20] <= -1.014\ngini = 0.204\nsamples = 26\nvalue = [23, 3]'),
Text(0.7142857142857143, 0.3611111111111111, 'x[3] <= -0.456\ngini = 0.49\nsamples = 7\nvalue = [4, 3]'),
Text(0.7077922077922078, 0.3055555555555556, 'x[7] <= 0.471\ngini = 0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.7012987012987013, 0.25, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.7142857142857143, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.7207792207792208, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.7272727272727273, 0.3611111111111111, 'gini = 0.0\nsamples = 19\nvalue = [19, 0]'),
Text(0.7532467532467533, 0.4722222222222222, 'x[4] <= 0.573\ngini = 0.497\nsamples = 13\nvalue = [6, 7]'),
Text(0.7467532467532467, 0.4166666666666667, 'x[28] <= 0.097\ngini = 0.346\nsamples = 9\nvalue = [2, 7]'),
Text(0.7402597402597403, 0.3611111111111111, 'x[7] <= 1.443\ngini = 0.219\nsamples = 8\nvalue = [1, 7]'),
Text(0.7337662337662337, 0.3055555555555556, 'gini = 0.0\nsamples = 7\nvalue = [0, 7]'),
Text(0.7467532467532467, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.7532467532467533, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.7597402597402597, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.7727272727272727, 0.5277777777777778, 'x[14] <= -1.114\ngini = 0.408\nsamples = 7\nvalue = [2, 5]'),
Text(0.7662337662337663, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.7792077920779208, 0.4722222222222222, 'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
Text(0.8051948051948052, 0.5833333333333334, 'x[2] <= 0.454\ngini = 0.346\nsamples = 9\nvalue = [2, 7]'),
Text(0.7987012987012987, 0.5277777777777778, 'x[16] <= -0.548\ngini = 0.219\nsamples = 8\nvalue = [1, 7]'),
Text(0.7922077922077922, 0.4722222222222222, 'gini = 0.0\nsamples = 7\nvalue = [0, 7]'),
Text(0.8051948051948052, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.8116883116883117, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.7609577922077922, 0.75, 'x[10] <= -0.364\ngini = 0.18\nsamples = 10\nvalue = [9, 1]'),
Text(0.7544642857142857, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.7674512987012987, 0.6944444444444444, 'gini = 0.0\nsamples = 9\nvalue = [9, 0]'),
Text(0.9131493506493507, 0.8055555555555556, 'x[15] <= 0.552\ngini = 0.258\nsamples = 204\nvalue = [173, 31]'),
Text(0.862012987012987, 0.75, 'x[16] <= 2.837\ngini = 0.138\nsamples = 147\nvalue = [136, 11]'),
Text(0.8555194805194806, 0.6944444444444444, 'x[3] <= 0.655\ngini = 0.128\nsamples = 146\nvalue = [136, 10]'),
Text(0.8376623376623377, 0.6388888888888888, 'x[26] <= -0.736\ngini = 0.038\nsamples = 104\nvalue = [102, 2]'),
Text(0.8311688311688312, 0.5833333333333334, 'x[10] <= -1.102\ngini = 0.32\nsamples = 10\nvalue = [8, 2]'),
Text(0.8246753246753247, 0.5277777777777778, 'x[0] <= 0.501\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.8181818181818182, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.8311688311688312, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.8376623376623377, 0.5277777777777778, 'gini = 0.0\nsamples = 7\nvalue = [7, 0]'),
Text(0.8441558441558441, 0.5833333333333334, 'gini = 0.0\nsamples = 94\nvalue = [94, 0]'),
Text(0.8733766233766234, 0.6388888888888888, 'x[8] <= -1.118\ngini = 0.308\nsamples = 42\nvalue = [34, 8]'),
Text(0.8571428571428571, 0.5833333333333334, 'x[17] <= -1.039\ngini = 0.375\nsamples = 4\nvalue = [1, 3]'),
Text(0.8506493506493507, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.8636363636363636, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.8896103896103896, 0.5833333333333334, 'x[0] <= -0.265\ngini = 0.229\nsamples = 38\nvalue = [33, 5]'),
```

```

Text(0.8766233766233766, 0.5277777777777778, 'x[8] <= -0.203\ngini = 0.5\nsamples = 6\nvalue = [3, 3]'),
Text(0.8701298701298701, 0.4722222222222222, 'x[27] <= 1.593\ngini = 0.375\nsamples = 4\nvalue = [1, 3]'),
Text(0.8636363636363636, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.8766233766233766, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]')

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

}

Text(0.9220772207722, 0.5555555555555556, 'x[40] <= 0.555\ngini = 0.444\nsamples = 5\nvalue = [4, 1]'),

dtc_cv=GridSearchCV(estimator=dt_model,param_grid=parameter,cv=5,scoring="accuracy")
Text(0.9330043300433, 0.5555555555555556, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),

dtc_cv.fit(x_train,y_train) # fitting the data

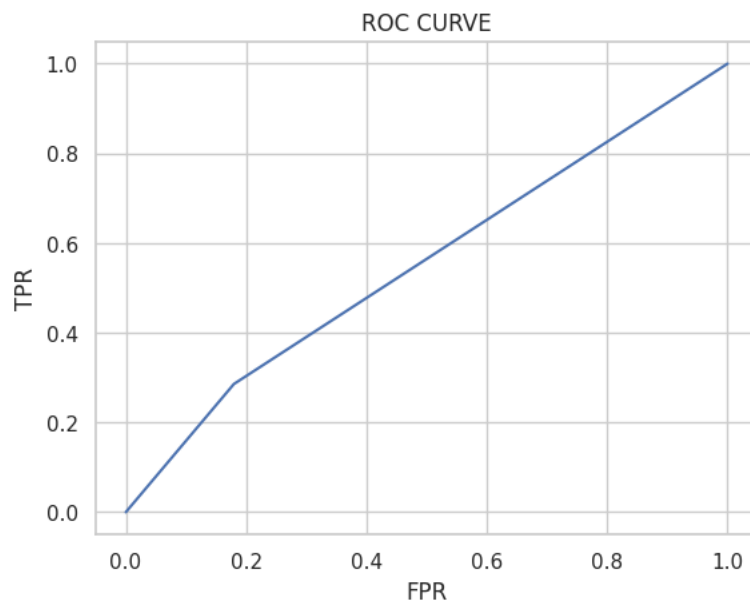
```



17/20

<https://colab.research.google.com/drive/1Vbbpc-9wJcQ0464WcWVi6BZB7XfosiYD#scrollTo=lpCw8kRMihiH&printMode=true> 18/20

```
fpr, tpr, thresholds = roc_curve(y_test, probability)
plt.plot(fpr, tpr)
plt.xlabel('FPR')
plt.ylabel('TPR')
plt.title('ROC CURVE')
plt.show()
```



### Random Forest Regressor

```
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()

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

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

rfc_cv.fit(x_train,y_train)
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/model_selection/_validation.py:378: FitFailedWarning:
50 fits failed out of a total of 700.
```

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

### Evaluation Metrics

```
50 fits failed with the following error:
```

```
accuracy_score(y_test,pred_rf)
```

```
0.8469387755102041
```

```
self. validate params()
```

```
confusion_matrix(y_test,pred_rf)
```

```
array([[240,   5],
       [ 40,   9]])
```

```
print(classification_report(y_test,pred_rf))
```

	precision	recall	f1-score	support
0	0.86	0.98	0.91	245
1	0.64	0.18	0.29	49
accuracy			0.85	294
macro avg	0.75	0.58	0.60	294
weighted avg	0.82	0.85	0.81	294

```
0.85796031 0.85541069 0.85287556 0.857120091
```

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

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

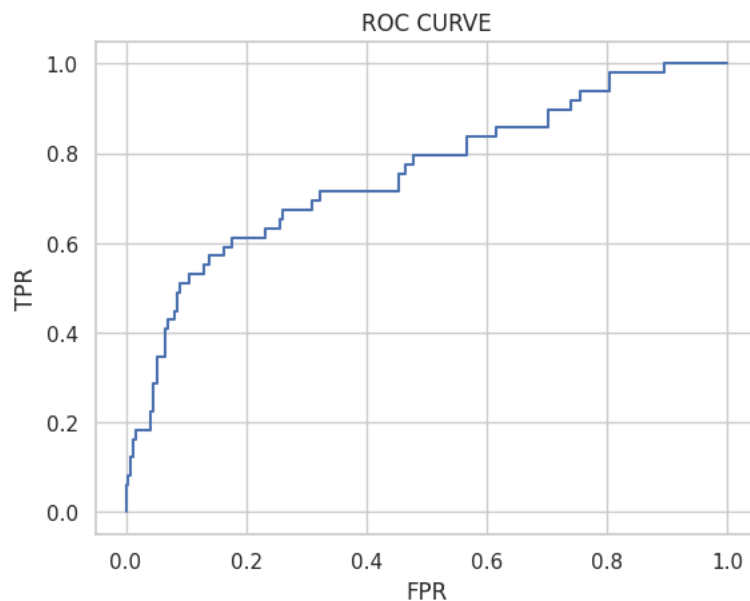
```
plt.plot(fpr, tpr)
```

```
plt.xlabel('FPR')
```

```
plt.ylabel('TPR')
```

```
plt.title('ROC CURVE')
```

```
plt.show()
```



```
''' THE END '''
```

```
' THE END '
```

```
''' Thank You '''
```

```
# Rayidi Abhiram
```

```
# 21bce9261
```

```
' Thank You '
```