## → ASSIGNMENT 4:

#### HARSH KUMAR

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#### ASSIGNMENT 4:

1.Download the Employee Attrition Dataset

https://www.kaggle.com/datasets/patelprashant/employee-attrition

- 2.Perfrom Data Preprocessing
- 3.Model Building using Logistic Regression and Decision Tree and Random Forest
- 4.Calculate Performance metrics

#### DATA COLLECTION

-> Collect the dataset or create the dataset

#### DATA PREPROCESSING

- -> Import the libraries.
- -> Import the dataset.
- -> Check for null values.
- -> Data Visualization.
- -> Outlier Detection.
- -> Splitting Dependent and Independent variables.
- -> Encoding.
- -> Feature Scaling.
- $\mbox{-}\mbox{>}$  Splitting Dataset into Train and Test.

## MODEL BUILDING

- -> Import the model building libraries.
- $\mbox{-}\mbox{>}$  Initializing the model.
- -> Training and testing the model.
- -> Evaluation of Model.
- -> Save the model.

## APPLICATION BUILDING

- -> Create an HTML file.
- -> Build a Python code.

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

df = pd.read_csv("HR-Employee-Attrition.csv")
```

df.head()

Age	Attrition	BusinessTrav	el DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNur
<b>0</b> 41	Yes	Travel_Rare	ely 1102	Sales	1	2	Life Sciences	1	
4 40	Ma	Traval Francisco	.l., 070	Research &	n	1	l ifa Caianasa	1	
ape									
(1470, 3	5)								
trition.	value_count	s()							
	233								
	237 trition, dt	ype: int64							
fo()									
fo()									
	•	e.frame.DataFr stries, 0 to 1							
_		. 35 columns):	.02						
# Col			on-Null Count						
0 4			470 11						
0 Age 1 Att	rition		470 non-null 470 non-null	int64 object					
	inessTravel		470 non-null	object					
	lyRate		470 non-null	int64					
	artment		470 non-null	object					
	tanceFromHo		470 non-null	int64					
6 Edu	cation	1	470 non-null	int64					
7 Edu	cationField	1	470 non-null	object					
8 Emp	loyeeCount	1	470 non-null	int64					
	loyeeNumber		470 non-null	int64					
	ironmentSat		470 non-null	int64					
11 Gen			470 non-null	object					
	rlyRate		470 non-null	int64					
	Involvement		470 non-null	int64					
	Level Role		470 non-null	int64					
	Satisfactio		470 non-null 470 non-null	object int64					
	italStatus		470 non-null	object					
	thlyIncome		470 non-null	int64					
	thlyRate		470 non-null	int64					
20 Num	CompaniesWo	rked 1	470 non-null	int64					
21 Ove	r18	1	470 non-null	object					
22 Ove	rTime	1	470 non-null	object					
23 Per	centSalaryH	like 1	470 non-null	int64					
	formanceRat	_	470 non-null	int64					
		tisfaction 1		int64					
	ndardHours		470 non-null	int64					
	ckOptionLev		470 non-null	int64					
	alWorkingYe		470 non-null	int64					
	<pre>iningTimesL kLifeBalanc</pre>		470 non-null 470 non-null	int64 int64					
	rsAtCompany		470 non-null	int64					
	rsInCurrent		470 non-null	int64					
	rsSinceLast		470 non-null	int64					
	rsWithCurrM		470 non-null	int64					
	. Jeen Circui I I'	upc: 1	o mon mull	±11.00T					

df.describe()

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

8 rows × 26 columns

<sup>#</sup> Detection of null values
df.isnull().any()

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  ${\tt PercentSalaryHike}$ False PerformanceRating False RelationshipSatisfaction StandardHours StockOptionLevel False TotalWorkingYears False TrainingTimesLastYear False False WorkLifeBalance YearsAtCompany False YearsInCurrentRole False YearsSinceLastPromotion False YearsWithCurrManager False dtype: bool

#### df.isnull().sum()

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

## **▼** DATA VISUALIZATION

# @title DATA VISUALIZATION

sns.distplot(df["Age"])

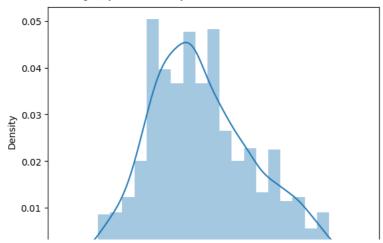
<ipython-input-779-eb5d36fb3f65>:3: UserWarning:

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

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

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

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



df.corr()

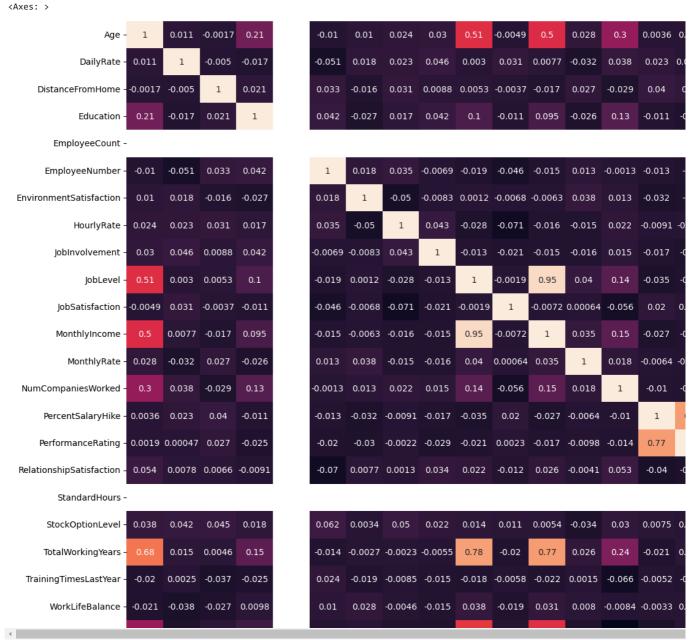
<ipython-input-780-2f6f6606aa2c>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future ve
 df.corr()

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount	EmployeeNumber	EnvironmentSatisfaction	Hou
Age	1.000000	0.010661	-0.001686	0.208034	NaN	-0.010145	0.010146	(
DailyRate	0.010661	1.000000	-0.004985	-0.016806	NaN	-0.050990	0.018355	(
DistanceFromHome	-0.001686	-0.004985	1.000000	0.021042	NaN	0.032916	-0.016075	(
Education	0.208034	-0.016806	0.021042	1.000000	NaN	0.042070	-0.027128	(
EmployeeCount	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
EmployeeNumber	-0.010145	-0.050990	0.032916	0.042070	NaN	1.000000	0.017621	(
EnvironmentSatisfaction	0.010146	0.018355	-0.016075	-0.027128	NaN	0.017621	1.000000	-(
HourlyRate	0.024287	0.023381	0.031131	0.016775	NaN	0.035179	-0.049857	1
Jobinvolvement	0.029820	0.046135	0.008783	0.042438	NaN	-0.006888	-0.008278	(
JobLevel	0.509604	0.002966	0.005303	0.101589	NaN	-0.018519	0.001212	-(
JobSatisfaction	-0.004892	0.030571	-0.003669	-0.011296	NaN	-0.046247	-0.006784	-(
MonthlyIncome	0.497855	0.007707	-0.017014	0.094961	NaN	-0.014829	-0.006259	-(
MonthlyRate	0.028051	-0.032182	0.027473	-0.026084	NaN	0.012648	0.037600	-(
NumCompaniesWorked	0.299635	0.038153	-0.029251	0.126317	NaN	-0.001251	0.012594	(
PercentSalaryHike	0.003634	0.022704	0.040235	-0.011111	NaN	-0.012944	-0.031701	-(
PerformanceRating	0.001904	0.000473	0.027110	-0.024539	NaN	-0.020359	-0.029548	-(
RelationshipSatisfaction	0.053535	0.007846	0.006557	-0.009118	NaN	-0.069861	0.007665	(
StandardHours	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
StockOptionLevel	0.037510	0.042143	0.044872	0.018422	NaN	0.062227	0.003432	(
TotalWorkingYears	0.680381	0.014515	0.004628	0.148280	NaN	-0.014365	-0.002693	-(
TrainingTimesLastYear	-0.019621	0.002453	-0.036942	-0.025100	NaN	0.023603	-0.019359	-(
WorkLifeBalance	-0.021490	-0.037848	-0.026556	0.009819	NaN	0.010309	0.027627	-(
YearsAtCompany	0.311309	-0.034055	0.009508	0.069114	NaN	-0.011240	0.001458	-(
YearsInCurrentRole	0.212901	0.009932	0.018845	0.060236	NaN	-0.008416	0.018007	-(
YearsSinceLastPromotion	0.216513	-0.033229	0.010029	0.054254	NaN	-0.009019	0.016194	-(
YearsWithCurrManager	0.202089	-0.026363	0.014406	0.069065	NaN	-0.009197	-0.004999	-(

26 rows × 26 columns

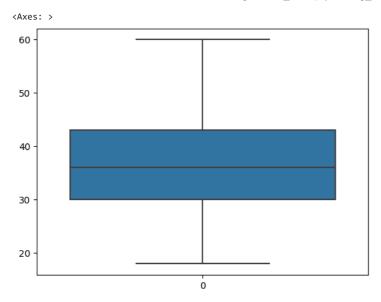
```
plt.subplots(figsize=(25,15))
sns.heatmap(df.corr(), annot = True)
```

<ipython-input-781-dec6949f63e9>:2: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future ve sns.heatmap(df.corr(), annot = True)

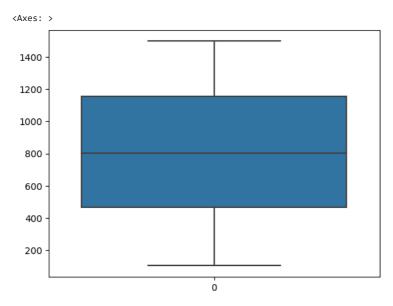


## → OUTLIER TREATMENT

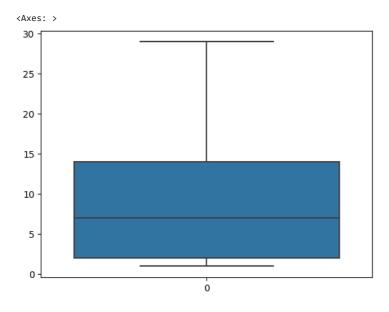
sns.boxplot(df["Age"])



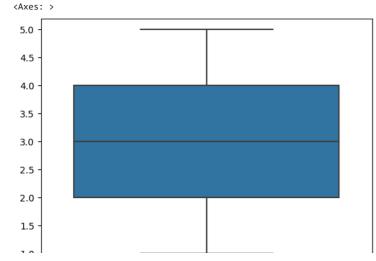
sns.boxplot(df["DailyRate"])



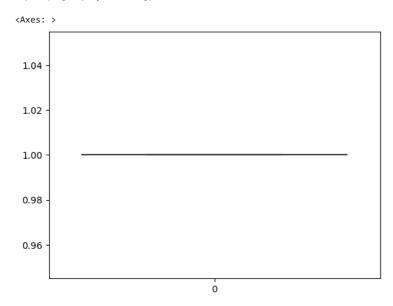
sns.boxplot(df["DistanceFromHome"])



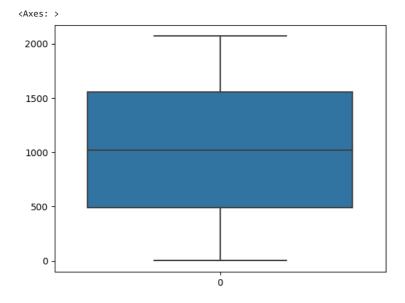
sns.boxplot(df["Education"])



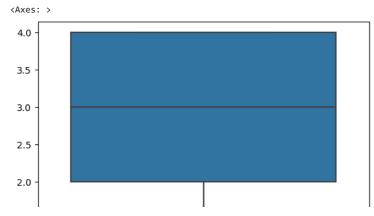
sns.boxplot(df["EmployeeCount"])



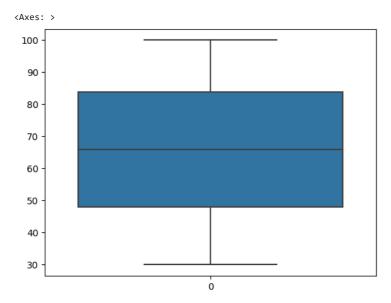
sns.boxplot(df["EmployeeNumber"])



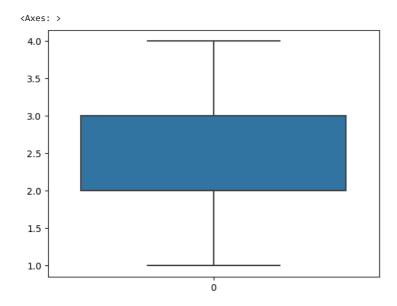
sns.boxplot(df["EnvironmentSatisfaction"])



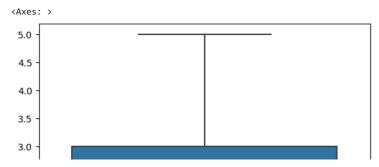
sns.boxplot(df["HourlyRate"])



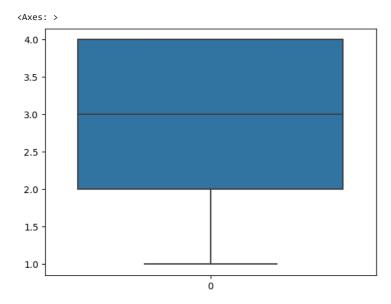
sns.boxplot(df["JobInvolvement"])



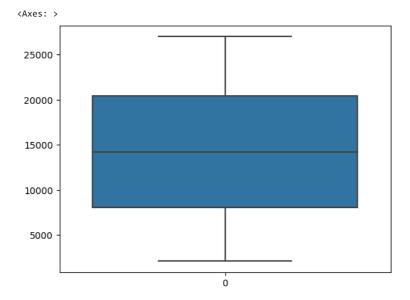
sns.boxplot(df["JobLevel"])



sns.boxplot(df["JobSatisfaction"])



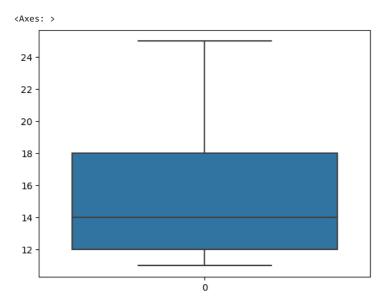
sns.boxplot(df["MonthlyRate"])



sns.boxplot(df["NumCompaniesWorked"])

```
<Axes: >
      8
Q1 = df.NumCompaniesWorked.quantile(0.25)
Q3 = df.NumCompaniesWorked.quantile(.75)
IQR = Q3-Q1
upper_lt = Q3 + 1.5*IQR
lower_lt = Q1 - 1.5*IQR
median = df["NumCompaniesWorked"].median()
\label{eq:df_NumCompaniesWorked'} df['NumCompaniesWorked'] + pper_lt, median, df['NumCompaniesWorked'])
\label{eq:df_NumCompaniesWorked'} df['NumCompaniesWorked'] < lower_lt, median, df['NumCompaniesWorked']) \\
sns.boxplot(df["NumCompaniesWorked"])
     <Axes: >
      8
       7
       6
      5
       4
      3
       2
       1
       0
```

sns.boxplot(df["PercentSalaryHike"])



sns.boxplot(df["PerformanceRating"])

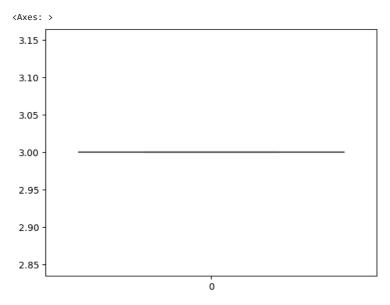
```
<4xes: >
4.0 -
3.8 -
3.6 -
3.4 -
```

```
Q1 = df.PerformanceRating.quantile(0.25)
Q3 = df.PerformanceRating.quantile(.75)
IQR = Q3-Q1
upper_lt = Q3 + 1.5*IQR
lower_lt = Q1 - 1.5*IQR
```

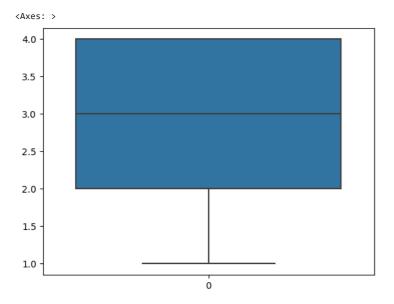
median = df["PerformanceRating"].median()

```
df['PerformanceRating'] = np.where(df['PerformanceRating']>upper_lt, median, df['PerformanceRating'])
df['PerformanceRating'] = np.where(df['PerformanceRating']<lower_lt, median, df['PerformanceRating'])</pre>
```

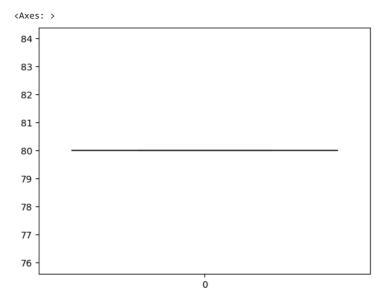
sns.boxplot(df["PerformanceRating"], color = 'red')



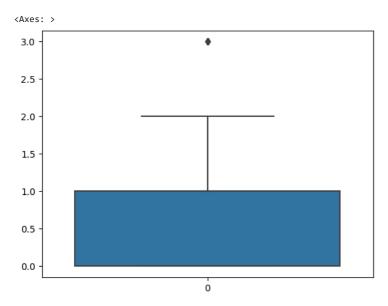
sns.boxplot(df["RelationshipSatisfaction"])



sns.boxplot(df["StandardHours"])



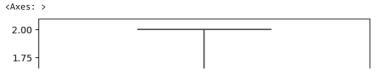
sns.boxplot(df["StockOptionLevel"])



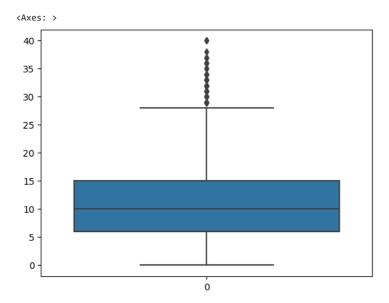
```
Q1 = df.StockOptionLevel.quantile(0.25)
Q3 = df.StockOptionLevel.quantile(.75)
IQR = Q3-Q1
upper_lt = Q3 + 1.5*IQR
lower_lt = Q1 - 1.5*IQR

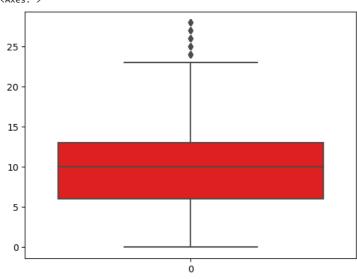
median = df["StockOptionLevel"].median()

df['StockOptionLevel'] = np.where(df['StockOptionLevel']>upper_lt, median, df['StockOptionLevel'])
df['StockOptionLevel'] = np.where(df['StockOptionLevel']<lower_lt, median, df['StockOptionLevel'])
sns.boxplot(df["StockOptionLevel"], color = 'red')</pre>
```



sns.boxplot(df["TotalWorkingYears"])



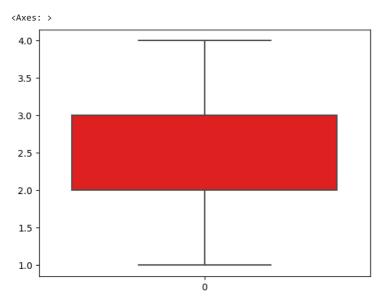


sns.boxplot(df["TrainingTimesLastYear"])

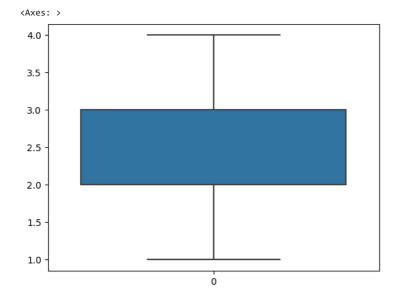
```
Q1 = df.TrainingTimesLastYear.quantile(0.25)
Q3 = df.TrainingTimesLastYear.quantile(.75)
IQR = Q3-Q1
upper_lt = Q3 + 1.5*IQR
lower_lt = Q1 - 1.5*IQR
median = df["TrainingTimesLastYear"].median()
```

 $df['TrainingTimesLastYear'] = np.where(df['TrainingTimesLastYear'] \\ vupper_lt, median, df['TrainingTimesLastYear']) \\ df['TrainingTimesLastYear'] = np.where(df['TrainingTimesLastYear'] \\ vupper_lt, median, df['TrainingTimesLastYear']) \\ vuppe$ 

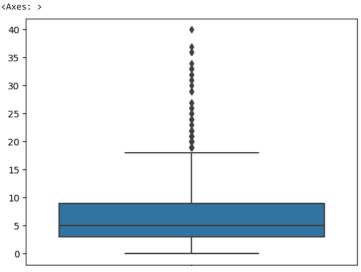
# # after outlier treatment sns.boxplot(df["TrainingTimesLastYear"], color = 'red')



#### sns.boxplot(df["WorkLifeBalance"])



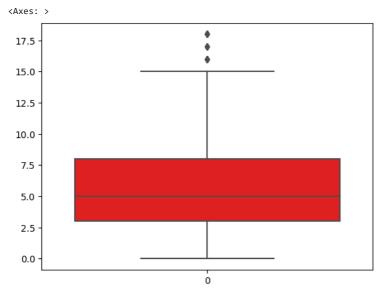
sns.boxplot(df["YearsAtCompany"])



```
Q1 = df.YearsAtCompany.quantile(0.25)
Q3 = df.YearsAtCompany.quantile(.75)
IQR = Q3-Q1
upper_lt = Q3 + 1.5*IQR
lower_lt = Q1 - 1.5*IQR
median = df["YearsAtCompany"].median()

df['YearsAtCompany'] = np.where(df['YearsAtCompany']>upper_lt, median, df['YearsAtCompany'])
df['YearsAtCompany'] = np.where(df['YearsAtCompany']<lower_lt, median, df['YearsAtCompany'])

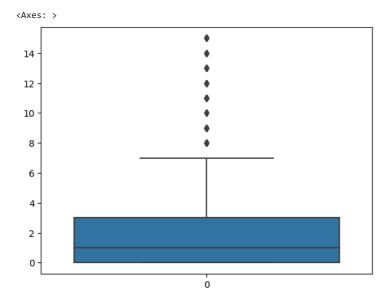
# after outlier treatment
sns.boxplot(df["YearsAtCompany"], color = 'red')</pre>
```



sns.boxplot(df["YearsInCurrentRole"])

```
<Axes: >
Q1 = df.YearsInCurrentRole.quantile(0.25)
Q3 = df.YearsInCurrentRole.quantile(.75)
IQR = Q3-Q1
upper_lt = Q3 + 1.5*IQR
lower_lt = Q1 - 1.5*IQR
median = df["YearsInCurrentRole"].median()
\label{eq:df_'YearsInCurrentRole'} df['YearsInCurrentRole'] = np.where(df['YearsInCurrentRole']) upper_lt, median, df['YearsInCurrentRole'])
\label{eq:df_'YearsInCurrentRole'} $$ df['YearsInCurrentRole'] < lower_lt, median, df['YearsInCurrentRole']) $$
# after outlier treatment
sns.boxplot(df["YearsInCurrentRole"], color = 'red')
     <Axes: >
       14
       12
       10
        8
        6
        2
```

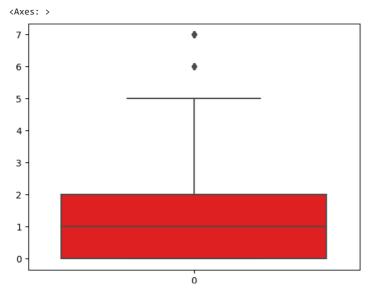
sns.boxplot(df["YearsSinceLastPromotion"])



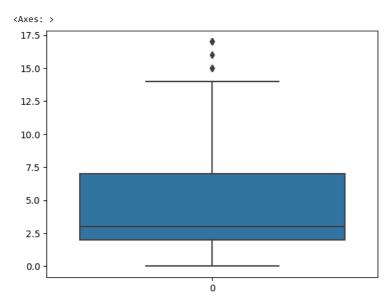
```
Q1 = df.YearsSinceLastPromotion.quantile(0.25)
Q3 = df.YearsSinceLastPromotion.quantile(.75)
IQR = Q3-Q1
upper_lt = Q3 + 1.5*IQR
lower_lt = Q1 - 1.5*IQR
median = df["YearsSinceLastPromotion"].median()

df['YearsSinceLastPromotion'] = np.where(df['YearsSinceLastPromotion']>upper_lt, median, df['YearsSinceLastPromotion'])
df['YearsSinceLastPromotion'] = np.where(df['YearsSinceLastPromotion']<lower_lt, median, df['YearsSinceLastPromotion'])

# after outlier treatment
sns.boxplot(df["YearsSinceLastPromotion"], color = 'red')
```



sns.boxplot(df["YearsWithCurrManager"])



```
Q1 = df.YearsWithCurrManager.quantile(0.25)
Q3 = df.YearsWithCurrManager.quantile(.75)
IQR = Q3-Q1
upper_lt = Q3 + 1.5*IQR
lower_lt = Q1 - 1.5*IQR
median = df["YearsWithCurrManager"].median()

df['YearsWithCurrManager'] = np.where(df['YearsWithCurrManager']>upper_lt, median, df['YearsWithCurrManager'])
df['YearsWithCurrManager'] = np.where(df['YearsWithCurrManager']
# after outlier treatment
sns.boxplot(df["YearsWithCurrManager"], color = 'red')
```

## ▼ Splitting Dependent and Independent Variable

```
# @title Splitting Dependent and Independent Variable
#X = df.iloc[:, 2:]
X = df.drop(['Attrition'], axis=1)
x
```

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	Envir
0	41	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	1	
1	49	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	2	
2	37	Travel_Rarely	1373	Research & Development	2	2	Other	1	4	
3	33	Travel_Frequently	1392	Research & Development	3	4	Life Sciences	1	5	
4	27	Travel_Rarely	591	Research & Development	2	1	Medical	1	7	
1465	36	Travel_Frequently	884	Research & Development	23	2	Medical	1	2061	
1466	39	Travel_Rarely	613	Research & Development	6	1	Medical	1	2062	
1467	27	Travel_Rarely	155	Research & Development	4	3	Life Sciences	1	2064	
1468	49	Travel_Frequently	1023	Sales	2	3	Medical	1	2065	
1469	34	Travel_Rarely	628	Research & Development	8	3	Medical	1	2068	

1470 rows × 34 columns

type(X)

```
pandas.core.frame.DataFrame
Y = df.iloc[:, 1]
     0
             Yes
     1
              Yes
              No
     1465
              No
     1466
1467
               No
               No
     1468
     1469
              No
     Name: Attrition, Length: 1470, dtype: object
type(Y)
     pandas.core.series.Series
```

## Label Encoding

```
# @title Label Encoding
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()

Y = le.fit_transform(Y)

Y
```

```
10/20/23, 1:08 AM
```

```
array([1, 0, 1, ..., 0, 0, 0])
type(Y)
     numpy.ndarray
Y = pd.Series(Y, name = 'Attrition')
     0
     1
             0
     2
     3
     1465
     1466
     1467
     1468
     1469
     Name: Attrition, Length: 1470, dtype: int64
type(Y)
     pandas.core.series.Series
X.BusinessTravel = le.fit_transform(X.BusinessTravel)
X.Department = le.fit_transform(X.Department)
X.EducationField = le.fit_transform(X.EducationField)
X.Gender = le.fit_transform(X.Gender)
X.JobRole = le.fit_transform(X.JobRole)
X.MaritalStatus = le.fit_transform(X.MaritalStatus)
X.OverTime = le.fit_transform(X.OverTime)
X.EducationField = le.fit_transform(X.EducationField)
X.Over18 = le.fit_transform(X.Over18)
Χ
```

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	Enviro
0	41	2	1102	2	1	2	1	1	1	
1	49	1	279	1	8	1	1	1	2	
2	37	2	1373	1	2	2	4	1	4	
3	33	1	1392	1	3	4	1	1	5	
4	27	2	591	1	2	1	3	1	7	
1465	36	1	884	1	23	2	3	1	2061	
1466	39	2	613	1	6	1	3	1	2062	
1467	27	2	155	1	4	3	1	1	2064	
1468	49	1	1023	2	2	3	3	1	2065	
1469	34	2	628	1	8	3	3	1	2068	

1470 rows × 34 columns

```
# Feature Scaling
from sklearn.preprocessing import MinMaxScaler
ms = MinMaxScaler()

# applying fit transform on all columns with 1 command
X_Scaled = pd.DataFrame(ms.fit_transform(X), columns = X.columns)
```

X\_Scaled

Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	Eı
0.547619	1.0	0.715820	1.0	0.000000	0.25	0.2	0.0	0.000000	
0.738095	0.5	0.126700	0.5	0.250000	0.00	0.2	0.0	0.000484	
0.452381	1.0	0.909807	0.5	0.035714	0.25	0.8	0.0	0.001451	
0.357143	0.5	0.923407	0.5	0.071429	0.75	0.2	0.0	0.001935	
0.214286	1.0	0.350036	0.5	0.035714	0.00	0.6	0.0	0.002903	
0.428571	0.5	0.559771	0.5	0.785714	0.25	0.6	0.0	0.996613	
0.500000	1.0	0.365784	0.5	0.178571	0.00	0.6	0.0	0.997097	
0.214286	1.0	0.037938	0.5	0.107143	0.50	0.2	0.0	0.998065	
0.738095	0.5	0.659270	1.0	0.035714	0.50	0.6	0.0	0.998549	
0.380952	1.0	0.376521	0.5	0.250000	0.50	0.6	0.0	1.000000	
	0.547619 0.738095 0.452381 0.357143 0.214286 0.428571 0.500000 0.214286 0.738095	0.547619 1.0 0.738095 0.5 0.452381 1.0 0.357143 0.5 0.214286 1.0  0.428571 0.5 0.500000 1.0 0.214286 1.0 0.738095 0.5	0.547619       1.0       0.715820         0.738095       0.5       0.126700         0.452381       1.0       0.909807         0.357143       0.5       0.923407         0.214286       1.0       0.350036              0.428571       0.5       0.559771         0.500000       1.0       0.365784         0.214286       1.0       0.037938         0.738095       0.5       0.659270	0.547619       1.0       0.715820       1.0         0.738095       0.5       0.126700       0.5         0.452381       1.0       0.909807       0.5         0.357143       0.5       0.923407       0.5         0.214286       1.0       0.350036       0.5               0.428571       0.5       0.559771       0.5         0.500000       1.0       0.365784       0.5         0.214286       1.0       0.037938       0.5         0.738095       0.5       0.659270       1.0	0.547619       1.0       0.715820       1.0       0.000000         0.738095       0.5       0.126700       0.5       0.250000         0.452381       1.0       0.909807       0.5       0.035714         0.357143       0.5       0.923407       0.5       0.071429         0.214286       1.0       0.350036       0.5       0.035714                0.428571       0.5       0.559771       0.5       0.785714         0.500000       1.0       0.365784       0.5       0.178571         0.214286       1.0       0.037938       0.5       0.107143         0.738095       0.5       0.659270       1.0       0.035714	0.547619       1.0       0.715820       1.0       0.000000       0.25         0.738095       0.5       0.126700       0.5       0.250000       0.00         0.452381       1.0       0.909807       0.5       0.035714       0.25         0.357143       0.5       0.923407       0.5       0.071429       0.75         0.214286       1.0       0.350036       0.5       0.035714       0.00                 0.428571       0.5       0.559771       0.5       0.785714       0.25         0.500000       1.0       0.365784       0.5       0.178571       0.00         0.214286       1.0       0.037938       0.5       0.107143       0.50         0.738095       0.5       0.659270       1.0       0.035714       0.50	0.547619       1.0       0.715820       1.0       0.000000       0.25       0.2         0.738095       0.5       0.126700       0.5       0.250000       0.00       0.2         0.452381       1.0       0.909807       0.5       0.035714       0.25       0.8         0.357143       0.5       0.923407       0.5       0.071429       0.75       0.2         0.214286       1.0       0.350036       0.5       0.035714       0.00       0.6                   0.428571       0.5       0.559771       0.5       0.785714       0.25       0.6         0.500000       1.0       0.365784       0.5       0.178571       0.00       0.6         0.214286       1.0       0.037938       0.5       0.107143       0.50       0.2         0.738095       0.5       0.659270       1.0       0.035714       0.50       0.5	0.547619       1.0       0.715820       1.0       0.000000       0.25       0.2       0.0         0.738095       0.5       0.126700       0.5       0.250000       0.00       0.2       0.0         0.452381       1.0       0.909807       0.5       0.035714       0.25       0.8       0.0         0.357143       0.5       0.923407       0.5       0.071429       0.75       0.2       0.0         0.214286       1.0       0.350036       0.5       0.035714       0.00       0.6       0.0         0.428571       0.5       0.559771       0.5       0.785714       0.25       0.6       0.0         0.500000       1.0       0.365784       0.5       0.178571       0.00       0.6       0.0         0.214286       1.0       0.037938       0.5       0.107143       0.50       0.2       0.0         0.738095       0.5       0.659270       1.0       0.035714       0.50       0.6       0.0	0.547619       1.0       0.715820       1.0       0.000000       0.25       0.2       0.0       0.000000         0.738095       0.5       0.126700       0.5       0.250000       0.00       0.2       0.0       0.000484         0.452381       1.0       0.909807       0.5       0.035714       0.25       0.8       0.0       0.001451         0.357143       0.5       0.923407       0.5       0.071429       0.75       0.2       0.0       0.001935         0.214286       1.0       0.350036       0.5       0.035714       0.00       0.6       0.0       0.002903                     0.428571       0.5       0.559771       0.5       0.785714       0.25       0.6       0.0       0.999613         0.500000       1.0       0.365784       0.5       0.178571       0.00       0.6       0.0       0.9997097         0.214286       1.0       0.037938       0.5       0.107143       0.50       0.2       0.0       0.998549         0.738095       0.5       0.659270       1.0       0.035714       0.50

1470 rows × 34 columns

```
#Splitting dataset into train and test
```

## MODEL BUILDING

## ▼ 1) Logistic Regression

```
from sklearn.linear_model import LogisticRegression
lr = LogisticRegression()
lr.fit(X_train, Y_train)
   LogisticRegression
   LogisticRegression()
pred = lr.predict(X_test)
pred
   array([0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       0,\ 0,\ 0,\ 0,\ 0,\ 0,\ 1,\ 0,\ 0,\ 0,\ 1,\ 0,\ 0,\ 1,\ 0,\ 0,\ 0,\ 0,\ 0,
         0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 1, 0, 0])
```

```
Y test
     442
     1091
     981
             1
     785
             0
     1332
            1
     1439
             0
     481
             0
     124
     198
             0
     Name: Attrition, Length: 294, dtype: int64
```

#### ▼ EVALUATION OF LOGISTIC REGRESSION MODEL

print(classification\_report(Y\_test, pred))

	precision	recall	f1-score	support
0	0.89	0.98	0.93	245
1	0.78	0.37	0.50	49
accuracy			0.88	294
macro avg	0.83	0.67	0.72	294
weighted avg	0.87	0.88	0.86	294

```
prob = lr.predict_proba(X_test)[:,1]
```

prob

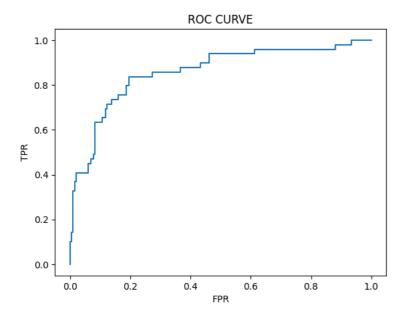
```
0.05676869, 0.30527903, 0.06380828, 0.00595047, 0.31380025,
 0.10429547,\ 0.33245617,\ 0.01998986,\ 0.70947895,\ 0.24794078,
0.0360448 , 0.11540785, 0.19099221, 0.06516867, 0.1790507 ,
0.2589084 , 0.02432868, 0.05030312, 0.07689715, 0.56886403,
0.23296053, 0.06290349, 0.05389915, 0.74340672, 0.07925537,
0.01583639, 0.02864397, 0.07861393, 0.35128791, 0.13789068,
0.05708495, 0.03507929, 0.09451264, 0.05595278, 0.03671325,
0.035616 , 0.02273025, 0.02963269, 0.01479639, 0.04009473,
0.48767355, 0.19876799, 0.00436051, 0.74220658, 0.4680317
0.17045322, 0.52988098, 0.13832449, 0.26216984, 0.62155628,
 0.21414729, \ 0.02622294, \ 0.30058166, \ 0.04111191, \ 0.16207279, 
0.03370787, 0.27528827, 0.07107923, 0.05182189, 0.54261956,
0.04264961, 0.23487739, 0.14701782, 0.07479887, 0.11537307,
0.02300648,\ 0.21190048,\ 0.06772688,\ 0.10704887,\ 0.06176002,
 0.1003125 \ , \ 0.04184002, \ 0.18974364, \ 0.20453703, \ 0.03407722, 
0.00715255, 0.01622941, 0.16602592, 0.02548361, 0.04608612,
0.08087786, 0.06041798, 0.02650985, 0.03486111, 0.10643701,
0.35496811, 0.11770772, 0.19535174, 0.24275894, 0.01508851,
0.22523028,\ 0.36675156,\ 0.29655802,\ 0.06393108,\ 0.05870711,
 0.35811197, \ 0.7698731 \ , \ 0.29150179, \ 0.01667642, \ 0.10890206, 
0.03754651,\ 0.03470175,\ 0.21009494,\ 0.03998842,\ 0.10919276,
0.10808936, 0.0493002 , 0.03383773, 0.1677363 , 0.06762935, 0.05743164, 0.05088314, 0.14647426, 0.00688177, 0.01362541,
 0.17135655, \ 0.04057648, \ 0.06784162, \ 0.85751252, \ 0.02736216, 
0.01298357, 0.00962036, 0.11610702, 0.1516965 , 0.08454238,
 0.02209819, \ 0.25728446, \ 0.58330047, \ 0.3984843 \ , \ 0.09482722, 
0.4400553 , 0.56922942, 0.2215613 , 0.05595997, 0.14503804,
0.12458319, 0.07372571, 0.10975582, 0.07534216, 0.17660223,
```

```
U.U0435814, U.11181529, U.U4U/5243, U.2U22809/, U.U/98UU/4,
0.02351601, 0.01621434, 0.07857288, 0.03167707, 0.01177251,
 0.26684397, \ 0.00708518, \ 0.16607682, \ 0.83355576, \ 0.13668737, \\
0.27016504, 0.13154952, 0.11162142, 0.05424089, 0.00684904,
0.03639676, 0.09558948, 0.13291639, 0.09346684, 0.01258016,
0.1199055 , 0.10625927, 0.05831006, 0.09153553, 0.08750728
 0.03025245, \ 0.146812 \quad , \ 0.01068085, \ 0.79556632, \ 0.03514599, 
0.03962226, 0.38848045, 0.06037585, 0.77262447, 0.13738206,
0.43953659, 0.53967335, 0.17529952, 0.05286558, 0.04592524,
0.14052106, 0.05549132, 0.01089639, 0.36059135, 0.05715995,
0.17629379, 0.1424216 , 0.69568084, 0.05367088, 0.20697085,
0.02818173, 0.46066677, 0.00915616, 0.14115345, 0.03003976,
0.07539821, 0.11656754, 0.05036671, 0.09009807, 0.2495464
 0.02533313, \ 0.02676806, \ 0.10527759, \ 0.0255894 \ , \ 0.15653797 
0.16187907, 0.22097983, 0.79363802, 0.07603148, 0.50428861,
0.01324733,\ 0.10075963,\ 0.24069048,\ 0.34126423,\ 0.04303388,
 \texttt{0.0123262} \text{ , } \texttt{0.3484587} \text{ , } \texttt{0.04632451}, \texttt{0.02116938}, \texttt{0.1376797} \text{ , } 
 \hbox{0.4586222 , 0.20989952, 0.01129264, 0.07275431, 0.0242586} 
0.21843379, 0.29278726, 0.04439659, 0.17762422, 0.09326981,
0.04458596, 0.44716043, 0.35302322, 0.0421056, 0.15808294,
0.34792218, 0.44110432, 0.8631254 , 0.0376125 , 0.22628 , 0.10334198, 0.02781656, 0.68815644, 0.17545869, 0.31002285,
0.45018771,\ 0.02486529,\ 0.27695696,\ 0.06369293,\ 0.06756454,
0.17135499, 0.00604792, 0.24771961, 0.57153276, 0.0912968
0.11724733, 0.01644254, 0.23538889, 0.05308161, 0.02923285,
 \hbox{0.0311262 , 0.07946052, 0.33957449, 0.13195021, 0.18304628, } 
 0.29781204, \ 0.01860209, \ 0.14674178, \ 0.09711585, \ 0.11045661, 
0.26327301,\ 0.01357499,\ 0.19532227,\ 0.01099183,\ 0.01629828,
0.20759601, 0.83454478, 0.04483892, 0.23713331])
```

## ▼ ROC - AUC CURVE FOR LOGISTIC REGRESSION MODEL

```
# @title ROC - AUC CURVE FOR LOGISTIC REGRESSION MODEL
fpr, tpr, thresholds = roc_curve(Y_test, prob)

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



## 2) Decision Tree Classifier

```
pred
```

```
Y_test
```

```
442
        0
1091
        0
981
        1
785
        0
1332
1439
        0
481
        0
124
        1
198
        0
1229
Name: Attrition, Length: 294, dtype: int64
```

#### Evaluation of Decison Tree Classifier model

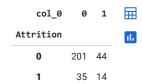
```
# @title Evaluation of Decison Tree Classifier model from sklearn.metrics import accuracy_score, confusion_matrix, classification_report, roc_auc_score, roc_curve
```

```
# Accuracy Score
accuracy = accuracy_score(Y_test, pred)
accuracy
```

0.7312925170068028

 ${\tt confusion\_matrix}({\tt Y\_test},\ {\tt pred})$ 

pd.crosstab(Y\_test, pred)



report = classification\_report(Y\_test, pred)
print(report)

	precision	recall	f1-score	support
0 1	0.85 0.24	0.82 0.29	0.84 0.26	245 49
accuracy macro avg	0.55	0.55	0.73 0.55	294 294
weighted avg	0.75	0.73	0.74	294

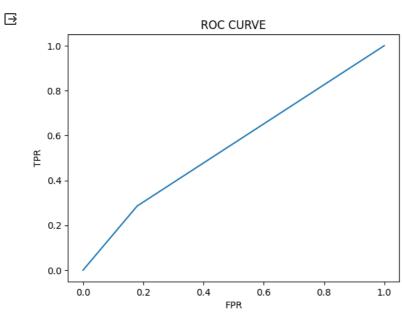
#### ▼ ROC - AUC CURVE FOR DECISION TREE CLASSIFIER MODEL

```
# @title ROC - AUC CURVE FOR DECISION TREE CLASSIFIER MODEL
prob = dtc.predict_proba(X_test)[:,1]
prob
```

```
\mathsf{array}([0.,\ 1.,\ 0.,\ 1.,\ 0.,\ 0.,\ 1.,\ 0.,\ 0.,\ 0.,\ 0.,\ 0.,\ 0.,\ 1.,\ 0.,\ 0.,\ 0.,
     0.,\; 1.,\; 0.,\; 1.,\; 0.,\; 0.,\; 0.,\; 1.,\; 0.,\; 0.,\; 0.,\; 1.,\; 0.,\; 1.,\; 0.,\; 0.,\; 0.,
     0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 1.,
     0., 0., 0., 1., 0., 0., 0., 1., 1., 0., 0., 0., 0., 1., 0., 1., 0.,
     1., 0., 1., 1., 1., 0., 0., 0., 1., 0., 0., 0., 0., 0., 1.,
     0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 0., 1., 1., 0., 0.,
     0.,\;0.,\;1.,\;0.,\;0.,\;0.,\;0.,\;1.,\;0.,\;0.,\;0.,\;0.,\;0.,\;0.,\;0.,\;0.,\;0.,
     0.,\; 0.,\; 1.,\; 0.,\; 0.,\; 0.,\; 0.,\; 1.,\; 0.,\; 0.,\; 0.,\; 0.,\; 1.,\; 0.,\; 0.,\; 0.,
     0.,\;0.,\;0.,\;0.,\;0.,\;1.,\;0.,\;1.,\;1.,\;0.,\;0.,\;0.,\;0.,\;1.,\;0.,\;0.,\;1.,
     0., \ 1., \ 0., \ 0., \ 0., \ 1., \ 1., \ 0., \ 0., \ 1., \ 0., \ 0., \ 0., \ 0., \ 0., \ 0.,
     0.,\ 0.,\ 1.,\ 0.,\ 0.,\ 0.,\ 0.,\ 0.,\ 0.,\ 1.,\ 0.,\ 1.,\ 0.,\ 0.,\ 0.,\ 0.,
```

```
# roc_curve
fpr, tpr, thresholds = roc_curve(Y_test, prob)

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



#### HYPER PARAMETER TUNING

```
from sklearn import tree
```

plt.figure(figsize=(25,15))
tree.plot\_tree(dtc, filled=True)

```
Text(0.006535947712418301, 0.65625, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.0196078431372549, 0.65625, 'x[22] <= 0.036\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
  Text(0.013071895424836602, 0.59375, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.026143790849673203, 0.59375, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.0392156862745098, 0.71875, 'x[19] <= 0.062\ngini = 0.153\nsamples = 24\nvalue = [22, 2]'),
  Text(0.032679738562091505, 0.65625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.0457516339869281, 0.65625, 'x[9] <= 0.167\ngini = 0.083\nsamples = 23\nvalue = [22, 1]'),
Text(0.0392156862745098, 0.59375, 'x[1] <= 0.75\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
  Text(0.032679738562091505, 0.53125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
   Text(0.0457516339869281, 0.53125, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
 Text(0.11764705882352941, 0.78125, 'x[8] <= 0.105\ngini = 0.493\nsamples = 25\nvalue = [11, 14]'),
Text(0.10457516339869281, 0.71875, 'x[22] <= 0.464\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
Text(0.09803921568627451, 0.65625, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.1111111111111111, 0.65625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
 Text(0.1111111111111111, 0.65625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),

Text(0.13071895424836602, 0.71875, 'x[15] <= 0.5\ngini = 0.432\nsamples = 19\nvalue = [6, 13]'),

Text(0.12418300653594772, 0.65625, 'gini = 0.0\nsamples = 7\nvalue = [0, 7]'),

Text(0.13725490196078433, 0.65625, 'x[6] <= 0.4\ngini = 0.5\nsamples = 12\nvalue = [6, 6]'),

Text(0.12418300653594772, 0.59375, 'x[28] <= 0.833\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),

Text(0.11764705882352941, 0.53125, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),

Text(0.13071895424836602, 0.53125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),

Text(0.1503267973856209, 0.59375, 'x[8] <= 0.249\ngini = 0.278\nsamples = 6\nvalue = [1, 5]'),

Text(0.1437908496732026, 0.53125, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),

Text(0.15686745998992, 0.53125, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
  Text(0.1568627450980392, 0.53125, 'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
Text(0.5730698529411765, 0.90625, 'x[21] <= 0.5\ngini = 0.235\nsamples = 1098\nvalue = [949, 149]'),
Text(0.319750816993464, 0.84375, 'x[29] <= 0.167\ngini = 0.162\nsamples = 798\nvalue = [727, 71]'),
 Text(0.1979816993464, 0.64575, 'x[28] <= 0.16/\light(\frac{1}{1}\) = 0.162\light(\frac{1}{1}\) = 0.162\light(\frac{1}\) = 0.162\light(\frac{1}
  Text(0.1895424836601307, 0.59375, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
  Text(0.20915032679738563, 0.65625, 'x[0] <= 0.333\ngini = 0.128\nsamples = 29\nvalue = [27, 2]'),
Text(0.20261437908496732, 0.59375, 'x[17] <= 0.125\ngini = 0.408\nsamples = 7\nvalue = [5, 2]'),
 Text(0.20261437908496732, 0.59375, 'x[17] <= 0.125\ngin1 = 0.408\nsamples = /\nvalue = [5, 2]'),
Text(0.19607843137254902, 0.53125, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.20915032679738563, 0.53125, 'x[13] <= 0.125\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.20261437908496732, 0.46875, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.21568627450980393, 0.46875, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.21568627450980393, 0.59375, 'gini = 0.0\nsamples = 22\nvalue = [22, 0]'),
Text(0.20915032679738563, 0.71875, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
  Text(0.4368872549019608, 0.78125, 'x[30] <= 0.25\ngini = 0.145\nsamples = 751\nvalue = [692, 59]'),
Text(0.32638888888889, 0.71875, 'x[9] <= 0.167\ngini = 0.218\nsamples = 257\nvalue = [225, 32]'),
  Text(0.28594771241830064, 0.65625, 'x[33] <= 0.179\ngini = 0.355\nsamples = 65\nvalue = [50, 15]'),
Text(0.2647058823529412, 0.59375, 'x[33] <= 0.036\ngini = 0.303\nsamples = 59\nvalue = [48, 11]'),
  Text(0.24183006535947713, 0.53125, 'x[12] <= 0.5\ngini = 0.463\nsamples = 22\nvalue = [14, 8]'),
Text(0.22875816993464052, 0.46875, 'x[11] <= 0.179\ngini = 0.198\nsamples = 9\nvalue = [8, 1]'),
Text(0.22222222222222, 0.40625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
  Text(0.24836601307189543, 0.40625, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.26143790849673204, 0.40625, 'x[4] <= 0.286\ngini = 0.346\nsamples = 9\nvalue = [2, 7]'),
  Text(0.2549019607843137, 0.34375, 'x[19] <= 0.312\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
  Text(0.24836601307189543, 0.28125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.26143790849673204, 0.28125, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.2679738562091503, 0.34375, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.2875816993464052, 0.53125, 'x[15] <= 0.167\ngini = 0.149\nsamples = 37\nvalue = [34, 3]'),
  Text(0.28104575163398693, 0.46875, 'x[30] <= 0.10\\ngini = 0.149\\nsamples = 5\\nvalue = [34, 5] 

Text(0.28104575163398693, 0.46875, 'x[30] <= 0.194\\ngini = 0.5\\nsamples = 6\\nvalue = [3, 3]'), 

Text(0.27450980392156865, 0.40625, 'gini = 0.0\\nsamples = 3\\nvalue = [0, 3]'), 

Text(0.2875816993464052, 0.40625, 'gini = 0.0\\nsamples = 3\\nvalue = [3, 0]'),
  Text(0.32679738562091504, 0.53125, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.33986928104575165, 0.53125, 'x[29] <= 0.5\ngini = 0.26\nsamples = 65\nvalue = [55, 10]'),
Text(0.3235294117647059, 0.46875, 'x[11] <= 0.679\ngini = 0.469\nsamples = 16\nvalue = [10, 6]'),
  Text(0.31699346405228757, 0.40625, 'x[6] <= 0.4\ngini = 0.444\nsamples = 9\nvalue = [3, 6]'), Text(0.3104575163398693, 0.34375, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'), Text(0.3235294117647059, 0.34375, 'x[22] <= 0.464\ngini = 0.245\nsamples = 7\nvalue = [1, 6]'),
  Text(0.31699346405228757, 0.28125, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.3300653594771242, 0.28125, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
  Text(0.3300653594771242, 0.49625, 'gini = 0.0\nsamples = 7\nvalue = [7, 0]'),
Text(0.3562091503267974, 0.46875, 'x[2] <= 0.037\ngini = 0.15\nsamples = 49\nvalue = [45, 4]'),
```

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Text(0.34967320261437906, 0.40625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.34967320261437906, 0.21875, 'gini = 0.0\nsamples = 42\nvalue = [42, 0]'),
Text(0.369281045751634, 0.28125, 'x[2] <= 0.337\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
 Text(0.3627450980392157, 0.21875, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.3758169934640523, 0.21875, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.369281045751634, 0.34375, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
 Text(0.40032679738562094, 0.59375, 'x[8] <= 0.022\ngini = 0.077\nsamples = 125\nvalue = [120, 5]'),
Text(0.38235294117647056, 0.53125, 'x[2] <= 0.578\ngini = 0.5\nsamples = 4\nvalue = [2, 2]'),
Text(0.3758169934640523, 0.46875, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.3888888888888, 0.46875, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
 Text(0.38235294117647056, 0.34375, 'gini = 0.0\nsamples = 107\nvalue = [107, 0]'),
Text(0.38235294117647056, 0.34375, 'gini = 0.0\nsamples = 107\nvalue = [107, 0]'),

Text(0.3954248366013072, 0.34375, 'x[11] <= 0.193\ngini = 0.245\nsamples = 7\nvalue = [6, 1]'),

Text(0.38888888888889, 0.28125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),

Text(0.4019607843137255, 0.28125, 'gini = 0.0\nsamples = 6\nvalue = [6, 0]'),

Text(0.4150326797385621, 0.40625, 'x[0] <= 0.405\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),

Text(0.4084967320261438, 0.34375, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),

Text(0.4215686274509804, 0.34375, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),

Text(0.438404522875817, 0.46875, 'x[30] <= 0.083\ngini = 0.4444\nsamples = 3\nvalue = [2, 1]'),

Text(0.43810457516339867, 0.46875, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]')
 Text(0.42810457516339867, 0.40625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'), Text(0.4411764705882353, 0.40625, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
  \label{eq:text}  \text{Text}(0.5473856209150327, \ 0.71875, \ 'x[17] <= 0.299 \\ \text{ngini} = 0.103 \\ \text{nsamples} = 494 \\ \text{nvalue} = [467, 27]'), \\ \text{Text}(0.4812091503267974, \ 0.65625, \ 'x[33] <= 0.964 \\ \text{ngini} = 0.06 \\ \text{nsamples} = 291 \\ \text{nvalue} = [282, 9]'), 
 Text(0.47467320261437906, 0.59375, 'x[8] <= 0.016\ngini = 0.054\nsamples = 290\nvalue = [282, 8]'),
Text(0.4542483660130719, 0.53125, 'x[4] <= 0.411\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.4477124183006536, 0.46875, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.447/124183000530, 0.40675, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.46078431372549017, 0.46875, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.4950980392156863, 0.53125, 'x[17] <= 0.056\ngini = 0.048\nsamples = 287\nvalue = [280, 7]'),
Text(0.4738562091503268, 0.46875, 'x[17] <= 0.054\ngini = 0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.4673202614379085, 0.40625, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
 Text(0.4803921568627451, 0.40625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5163398692810458, 0.46875, 'x[15] <= 0.5\ngini = 0.042\nsamples = 282\nvalue = [276, 6]'),
 Text(0.4738562091503268, 0.28125, 'x[19] <= 0.812\ngini = 0.175\nsamples = 31\nvalue = [28, 3]'),
Text(0.4673202614379085, 0.21875, 'x[31] <= 0.107\ngini = 0.124\nsamples = 30\nvalue = [28, 2]'),
Text(0.454248360130719, 0.15625, 'x[2] <= 0.192\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.4477124183006536, 0.09375, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.46078431372549017, 0.09375, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.4803921568627451, 0.15625, 'x[4] <= 0.946\ngini = 0.069\nsamples = 28\nvalue = [27, 1]'),
Text(0.4803921568627451, 0.09375, 'gini = 0.0\nsamples = 25\nvalue = [25, 0]'),
Text(0.4803921568627451, 0.09375, 'x[1] <= 0.75\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.4803921568627451, 0.03125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
 Text(0.4934640522875817, 0.03125, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.4934640522875817, 0.21875, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.4869281045751634, 0.28125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5065359477124183, 0.34375, 'x[26] <= 0.75\ngini = 0.027\nsamples = 72\nvalue = [71, 1]'),
 Text(0.5, 0.28125, 'gini = 0.0\nsamples = 64\nvalue = [64, 0]'),
 Text(0.5)30718954248366, 0.28125, 'x[2] <= 0.528\ngini = 0.219\nsamples = 8\nvalue = [7, 1]'),
Text(0.5065359477124183, 0.21875, 'gini = 0.0\nsamples = 6\nvalue = [6, 0]'),
Text(0.5196078431372549, 0.21875, 'x[14] <= 0.625\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.5130718954248366, 0.15625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
 Text(0.5261437908496732, 0.15625, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.5392156862745098, 0.40625, 'x[33] <= 0.107\ngini = 0.011\nsamples = 178\nvalue = [177, 1]'),
 Text(0.5326797385620915, 0.34375, 'x[30] <= 0.528\ngini = 0.133\\nsamples = 14\\nvalue = [13, 1]'),
Text(0.5261437908496732, 0.28125, 'gini = 0.0\\nsamples = 12\\nvalue = [12, 0]'),
Text(0.5392156862745098, 0.28125, 'x[10] <= 0.5\\ngini = 0.5\\nsamples = 2\\nvalue = [1, 1]'),
 Text(0.5392158862745098, 0.28125, 'x[10] <= 0.5\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),

Text(0.5326797385620915, 0.21875, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),

Text(0.545751633986928, 0.21875, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),

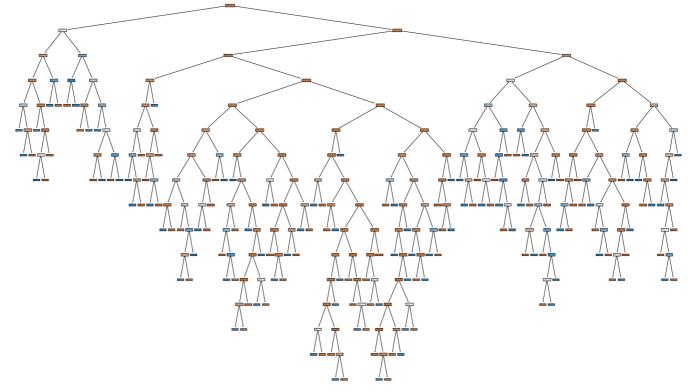
Text(0.545751633986928, 0.34375, 'gini = 0.0\nsamples = 164\nvalue = [164, 0]'),

Text(0.4877450980392157, 0.59375, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),

Text(0.613562091503268, 0.65625, 'x[17] <= 0.5\ngini = 0.162\nsamples = 203\nvalue = [185, 18]'),
 Text(0.5800653594771242, 0.59375, 'x[2] <= 0.033\ngini = 0.275\nsamples = 97\nvalue = [81, 16]'), Text(0.5620915032679739, 0.53125, 'x[17] <= 0.346\ngini = 0.48\nsamples = 5\nvalue = [2, 3]'), Text(0.55555555555556, 0.46875, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
 Text(0.5816993464052288, 0.46875, 'x[18] <= 0.95\ngini = 0.15\nsamples = 70\nValue = [64, 6]'), Text(0.5751633986928104, 0.40625, 'x[17] <= 0.3\ngini = 0.134\nsamples = 69\nvalue = [64, 5]'), Text(0.5686274509803921, 0.34375, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'), Text(0.5816993464052288, 0.34375, 'x[11] <= 0.971\ngini = 0.111\nsamples = 68\nvalue = [64, 4]'), Text(0.5751633986928104, 0.28125, 'x[31] <= 0.786\ngini = 0.086\nsamples = 67\nvalue = [64, 3]'), Text(0.5588235294117647, 0.21875, 'x[6] <= 0.9\ngini = 0.06\nsamples = 65\nvalue = [63, 2]'), Text(0.545751633986928, 0.15625, 'x[28] <= 0.833\ngini = 0.032\nsamples = 61\nvalue = [60, 1]'),
 Text(0.5522875816993464, 0.09375, 'X[22] <= 0.071\ng1n1 = 0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.5545751633986928, 0.03125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5588235294117647, 0.03125, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.5718954248366013, 0.15625, 'x[8] <= 0.412\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.565359477124183, 0.09375, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.5784313725490197, 0.09375, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5915032679738562, 0.21875, 'x[31] <= 0.929\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.5849673202614370, 0.15625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]')
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Text(0.5980392156862745, 0.15625, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.5882352941176471, 0.28125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5882352941176471, 0.40625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6143790849673203, 0.46875, 'x[4] <= 0.268\ngini = 0.434\nsamples = 22\nvalue = [15, 7]'),
Text(0.5947712418300654, 0.34375, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
                                                                               'x[17] <= 0.498\ngini = 0.133\nsamples = 14\nvalue = [13, 1]'),
Text(0.6078431372549019, 0.34375,
Text(0.6013071895424836, 0.28125, 'gini = 0.0\nsamples = 13\nvalue = [13, 0]'),
Text(0.6143790849673203, 0.28125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6274509803921569, 0.40625,
                                                                               'x[33] \leftarrow 0.571 \text{ ngini} = 0.408 \text{ nsamples} = 7 \text{ nvalue} = [2, 5]'),
                                                                               'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.6209150326797386, 0.34375,
Text(0.6339869281045751, 0.34375,
Text(0.6470588235294118, 0.59375,
                                                                               'x[31] \leftarrow 0.964 \text{ ngini} = 0.037 \text{ nsamples} = 106 \text{ nvalue} = [104, 2]'),
Text(0.6405228758169934, 0.53125,
                                                                               'x[30] \leftarrow 0.972 = 0.019 = 105 = 105 = 104, 1'
Text(0.6339869281045751, 0.46875,
                                                                               'gini = 0.0\nsamples = 101\nvalue = [101, 0]'),
                                                                               'x[24] \le 0.833 \text{ ngini} = 0.375 \text{ nsamples} = 4 \text{ nvalue} = [3, 1]'),
Text(0.6470588235294118, 0.46875,
Text(0.6405228758169934, 0.40625,
                                                                               'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.6535947712418301, 0.40625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6535947712418301, 0.53125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.8263888888888, 0.84375, 'x[17] <= 0.157\ngini = 0.385\nsamples = 300\nvalue = [222, 78]'),
Text(0.7426470588235294, 0.78125, 'x[26] <= 0.25 \\ ngini = 0.5 \\ nsamples = 96 \\ nvalue = [49, 47]'), \\ ngini = 0.5 \\ ngi = 0.5 \\ ngini = 0.5 \\ ngi = 0.5 \\ ngini = 0.5 \\ ngini = 0.5 
Text(0.66666666666666, 0.53125, 'gini = 0.0\nsamples = 8\nvalue = [0, 8]'),
Text(0.6797385620915033, 0.53125, 'x[9] <= 0.333\ngini = 0.48\nsamples = 5\nvalue = [3, 2]'),
Text(0.673202614379085, 0.46875, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.6862745098039216, 0.46875,
                                                                                'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.6993464052287581, 0.59375, 'x[14] <= 0.875\ngini = 0.18\nsamples = 10\nvalue = [9, 1]'),
Text(0.6928104575163399, 0.53125, 'gini = 0.0\nsamples = 8\nvalue = [8, 0]'),
 \label{eq:text}      \text{Text}(0.7058823529411765, \ 0.53125, \ 'x[15] <= 0.667 \\      \text{ngini} = 0.5 \\      \text{nsamples} = 2 \\      \text{nvalue} = [1, \ 1]'), \\      \text{Text}(0.6993464052287581, \ 0.46875, \ 'gini = 0.0 \\      \text{nsamples} = 1 \\      \text{nvalue} = [0, \ 1]'), 
Text(0.7124183006535948, 0.46875, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
\label{eq:text} \texttt{Text}(0.7320261437908496,\ 0.65625,\ 'x[27] <= 0.5 \\ \texttt{ngini} = 0.266 \\ \texttt{nsamples} = 19 \\ \texttt{nvalue} = [3,\ 16]'),
Text(0.7254901960784313, 0.59375, 'x[11] <= 0.2\ngini = 0.198\nsamples = 18\nvalue = [2, 16]'),
Text(0.7189542483660131, 0.53125, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.7320261437908496, 0.53125, 'x[32] <= 0.929\ngini = 0.111\nsamples = 17\nvalue = [1, 16]'),
Text(0.7254901960784313, 0.46875, 'gini = 0.0\nsamples = 15\nvalue = [0, 15]'),
Text(0.738562091503268, 0.46875, 'x[5] <= 0.5\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.7320261437908496, 0.40625, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.7450980392156863, 0.40625, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.738562091503268, 0.59375, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.7516339869281046, 0.59375, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'), Text(0.7647058823529411, 0.59375, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.7941176470588235, 0.65625, 'x[2] <= 0.622\ngini = 0.418\nsamples = 47\nvalue = [33, 14]'),
Text(0.7777777777777, 0.59375, 'x[2] <= 0.145\ngini = 0.482\nsamples = 32\nvalue = [19, 13]'), Text(0.7647058823529411, 0.53125, 'x[2] <= 0.024\ngini = 0.18\nsamples = 10\nvalue = [9, 1]'),
Text(0.7647058023323411, 0.53125, 'k[2] <- 0.024\light[in = 0.18\light[in = 0.
Text(0.7647058823529411, 0.34375, 'gini = 0.0 \nsamples = 5 \nvalue = [5, 0]'),
Text(0.777777777778, 0.34375, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.7973856209150327, 0.40625, 'x[0] <= 0.25\ngini = 0.298\nsamples = 11\nvalue = [2, 9]'),
Text(0.7908496732026143, 0.34375, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.803921568627451, 0.34375, 'x[4] <= 0.018\ngini = 0.18\nsamples = 10\nvalue = [1, 9]'),
Text(0.7973856209150327, 0.28125, 'x[11] <= 0.229\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.7908496732026143, 0.21875, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.803921568627451, 0.21875, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
\label{eq:text} \begin{split} \text{Text}(0.8104575163398693, \ 0.28125, \ 'gini = 0.0 \land samples = 8 \land value = [0, \ 8]'), \\ \text{Text}(0.7973856209150327, \ 0.46875, \ 'gini = 0.0 \land samples = 3 \land value = [3, \ 0]'), \end{split}
Text(0.8169934640522876, 0.53125, 'gini = 0.0\nsamples = 14\nvalue = [14, 0]'),
Text(0.9101307189542484, 0.78125, 'x[16] <= 0.75\ngini = 0.258\nsamples = 204\nvalue = [173, 31]'),
Text(0.8627450980392157, 0.71875, 'x[17] <= 0.992\ngini = 0.138\nsamples = 147\nvalue = [136, 11]'),
Text(0.8169934640522876, 0.40625, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.8300653594771242, 0.40625, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.8366013071895425, 0.46875, 'gini = 0.0\nsamples = 7\nvalue = [7, 0]'),
Text(0.8431372549019608, 0.53125, 'gini = 0.0\nsamples = 94\nvalue = [94, 0]'),
Text(0.8758169934649523, 0.59375, 'x[9] <= 0.167\ngini = 0.308\nsamples = 42\nvalue = [34, 8]'), Text(0.8562091503267973, 0.53125, 'x[18] <= 0.194\ngini = 0.375\nsamples = 4\nvalue = [1, 3]'), Text(0.8496732026143791, 0.46875, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'), Text(0.8627450980392157, 0.46875, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'), Text(0.8954248366013072, 0.53125, 'x[0] <= 0.393\ngini = 0.229\nsamples = 38\nvalue = [33, 5]'),
Text(0.8758169934640523, 0.46875, 'x[1] <= 0.25\ngini = 0.5\nsamples = 6\nvalue = [3, 3]'),
Text(0.869281045751634, 0.40625, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.869281045751634, 0.40625, 'y[11] <= 0.643\ngini = 0.375\nsamples = 4\nyalue = [1, 0]'),
Text(0.8823529411764706, 0.40625, 'x[11] <= 0.643\ngini = 0.375\nsamples = 4\nvalue = [1, 3]'),
Text(0.8758169934640523, 0.34375, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.8888888888888, 0.34375, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.9150326797385621, 0.46875, 'x[8] <= 0.992\ngini = 0.117\nsamples = 32\nvalue = [30, 2]'),
Text(0.9084967320261438, 0.40625, 'x[18] <= 0.174\ngini = 0.062\nsamples = 31\nvalue = [30, 1]'),
Text(0.9019607843137255, 0.34375, 'x[14] <= 0.688\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
```

```
Text(0.8954248366013072, 0.28125, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.9084967320261438, 0.28125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
                                                                  'gini = 0.0\nsamples = 29\nvalue = [29, 0]'),
Text(0.9150326797385621, 0.34375,
'x[14] \leftarrow 0.812  | o.456 \ nsamples = 57 \ nvalue = [37, 20]'),
Text(0.9575163398692811, 0.71875,
Text(0.9281045751633987, 0.65625, 'x[8] <= 0.071 \\ lngini = 0.238 \\ lnsamples = 29 \\ lnsa
Text(0.9150326797385621, 0.59375,
                                                                  'x[0] \leftarrow 0.321 \text{ ngini} = 0.444 \text{ nsamples} = 3 \text{ nvalue} = [1, 2]'),
Text(0.9084967320261438, 0.53125,
                                                                  'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.9215686274509803, 0.53125, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.9411764705882353, 0.59375,
                                                                  'x[28] \leftarrow 0.167 \text{ ngini} = 0.142 \text{ nsamples} = 26 \text{ nvalue} = [24, 2]'),
Text(0.934640522875817, 0.53125, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.9477124183006536, 0.53125,
                                                                   'x[17] <= 0.956\ngini = 0.077\nsamples = 25\nvalue = [24, 1]'),
Text(0.9411764705882353, 0.46875, 'gini = 0.0\nsamples = 24\nvalue = [24, 0]'),
                                                                'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.954248366013072, 0.46875,
Text(0.9869281045751634, 0.65625,
                                                                  'x[32] \leftarrow 0.214 \text{ ngini} = 0.49 \text{ nsamples} = 28 \text{ nvalue} = [12, 16]'),
Text(0.9803921568627451, 0.59375,
                                                                  'x[4] \leftarrow 0.804 \text{ ngini} = 0.48 \text{ nsamples} = 20 \text{ nvalue} = [12, 8]'),
Text(0.9738562091503268, 0.53125,
                                                                  'x[30] \leftarrow 0.028 \text{ ngini} = 0.415 \text{ nsamples} = 17 \text{ nvalue} = [12, 5]'),
Text(0.9673202614379085, 0.46875,
                                                                  'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
                                                                  'x[24] \leftarrow 0.5 \neq 0.32 = 15 = 15
Text(0.9803921568627451, 0.46875,
                                                                  'x[27] \leftarrow 0.196 \setminus initial = 0.5 \setminus insamples = 6 \setminus invalue = [3, 3]'),
Text(0.9738562091503268, 0.40625,
Text(0.9673202614379085, 0.34375, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.9803921568627451, 0.34375,
                                                                  'x[6] \leftarrow 0.7 \text{ ngini} = 0.375 \text{ nsamples} = 4 \text{ nvalue} = [1, 3]'),
Text(0.9738562091503268, 0.28125, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.9869281045751634, 0.28125,
                                                                  'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.9869281045751634, 0.40625, 'gini = 0.0 \nsamples = 9 \nvalue = [9, 0]'),
Text(0.9934640522875817, 0.59375, 'gini = 0.0\nsamples = 8\nvalue = [0, 8]')]
```



```
from sklearn.model_selection import GridSearchCV

# search documentation of parameters for Decision Tree Classifier
parameters = {
    'criterion' : ['gini', 'entropy'],
    'splitter' : ['best', 'random'],
    'max_depth' : [1, 2, 3, 4, 5],
    'max_features' : ['auto', 'sqrt', 'log2']
}

# using GridSearchCV to figure out the best hyperparameters
grid_search = GridSearchCV(estimator = dtc, param_grid = parameters, cv=5, scoring ="accuracy")
grid_search.fit(X_train, Y_train)
```

```
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