

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

```
#Importing the dataset.
df=pd.read_csv("/content/WA_Fn-UseC_-HR-Employee-Attrition.csv")
```

```
df.head()
```

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

5 rows × 35 columns

```
df.shape
```

(1470, 35)

```
df.Age.value_counts()
```

35	78
34	77
36	69
31	69
29	68
32	61
30	60
33	58
38	58
40	57
37	50
27	48
28	48

```

42  46
39  42
45  41
41  40
26  39
44  33
46  33
43  32
50  30
25  26
24  26
49  24
47  24
55  22
51  19
53  19
48  19
54  18
52  18
22  16
56  14
23  14
58  14
21  13
20  11
59  10
19   9
18   8
60   5
57   4
Name: Age, dtype: int64

```

```
df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1470 entries, 0 to 1469
Data columns (total 35 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   Age                   1470 non-null  int64
 1   Attrition             1470 non-null  object
 2   BusinessTravel        1470 non-null  object
 3   DailyRate             1470 non-null  int64
 4   Department            1470 non-null  object
 5   DistanceFromHome      1470 non-null  int64
 6   Education              1470 non-null  int64
 7   EducationField         1470 non-null  object
 8   EmployeeCount          1470 non-null  int64
 9   EmployeeNumber         1470 non-null  int64
10   EnvironmentSatisfaction 1470 non-null  int64
11   Gender                 1470 non-null  object
12   HourlyRate            1470 non-null  int64
13   JobInvolvement         1470 non-null  int64
14   JobLevel               1470 non-null  int64

```

```

15 JobRole          1470 non-null object
16 JobSatisfaction  1470 non-null int64
17 MaritalStatus    1470 non-null object
18 MonthlyIncome    1470 non-null int64
19 MonthlyRate      1470 non-null int64
20 NumCompaniesWorked 1470 non-null int64
21 Over18           1470 non-null object
22 OverTime         1470 non-null object
23 PercentSalaryHike 1470 non-null int64
24 PerformanceRating 1470 non-null int64
25 RelationshipSatisfaction 1470 non-null int64
26 StandardHours    1470 non-null int64
27 StockOptionLevel  1470 non-null int64
28 TotalWorkingYears 1470 non-null int64
29 TrainingTimesLastYear 1470 non-null int64
30 WorkLifeBalance   1470 non-null int64
31 YearsAtCompany    1470 non-null int64
32 YearsInCurrentRole 1470 non-null int64
33 YearsSinceLastPromotion 1470 non-null int64
34 YearsWithCurrManager 1470 non-null int64

```

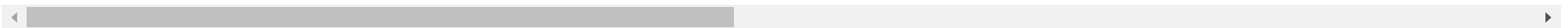
```
dtypes: int64(26), object(9)
```

```
memory usage: 402.1+ KB
```

```
df.describe()
```

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount	EmployeeNumber	EnvironmentSatisfaction	HourlyRate	JobInvolvement	JobLevel	...	Relation:
count	1470.000000	1470.000000	1470.000000	1470.000000	1470.0	1470.000000	1470.000000	1470.000000	1470.000000	1470.000000	...	
mean	36.923810	802.485714	9.192517	2.912925	1.0	1024.865306	2.721769	65.891156	2.729932	2.063946	...	
std	9.135373	403.509100	8.106864	1.024165	0.0	602.024335	1.093082	20.329428	0.711561	1.106940	...	
min	18.000000	102.000000	1.000000	1.000000	1.0	1.000000	1.000000	30.000000	1.000000	1.000000	...	
25%	30.000000	465.000000	2.000000	2.000000	1.0	491.250000	2.000000	48.000000	2.000000	1.000000	...	
50%	36.000000	802.000000	7.000000	3.000000	1.0	1020.500000	3.000000	66.000000	3.000000	2.000000	...	
75%	43.000000	1157.000000	14.000000	4.000000	1.0	1555.750000	4.000000	83.750000	3.000000	3.000000	...	
max	60.000000	1499.000000	29.000000	5.000000	1.0	2068.000000	4.000000	100.000000	4.000000	5.000000	...	

```
8 rows × 26 columns
```



```
#Checking for Null Values.
```

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

```

Age           False
Attrition     False
BusinessTravel False
DailyRate     False

```

```

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

```

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

```

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

```

```
MonthlyRate          0
NumCompaniesWorked    0
Over18                0
OverTime              0
PercentSalaryHike     0
PerformanceRating     0
RelationshipSatisfaction 0
StandardHours         0
StockOptionLevel      0
TotalWorkingYears     0
TrainingTimesLastYear 0
WorkLifeBalance       0
YearsAtCompany        0
YearsInCurrentRole    0
YearsSinceLastPromotion 0
YearsWithCurrManager  0
dtype: int64
```

```
#Data Visualization.
```

```
sns.distplot(df["YearsWithCurrManager"])
```

```
<ipython-input-11-71e8291be26b>:2: UserWarning:
```

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

```
Please adapt your code to use either `displot` (a figure-level function with  
df.corr())
```

```
<ipython-input-12-2f6f6606aa2c>:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only
df.corr()
```

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount	EmployeeNumber	EnvironmentSatisfaction	HourlyRate	JobInvolvement	JobLevel	...
Age	1.000000	0.010661	-0.001686	0.208034	NaN	-0.010145	0.010146	0.024287	0.029820	0.509604	...
DailyRate	0.010661	1.000000	-0.004985	-0.016806	NaN	-0.050990	0.018355	0.023381	0.046135	0.002966	...

```
df.head()
```

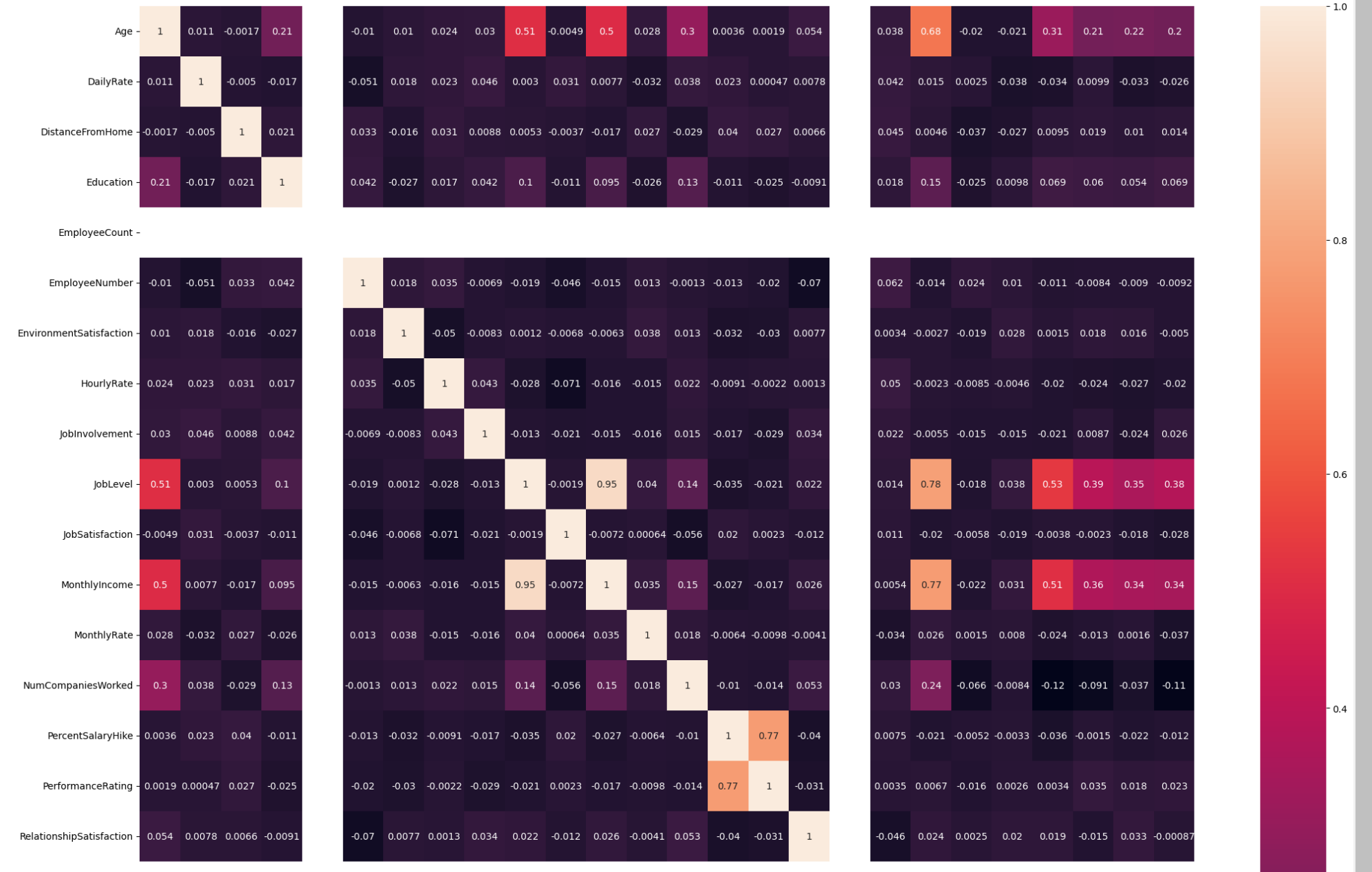
	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	...	RelationshipSatisfaction	StandardHours
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	1	...	1	
1	49	No	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	2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

```
5 rows × 35 columns
```

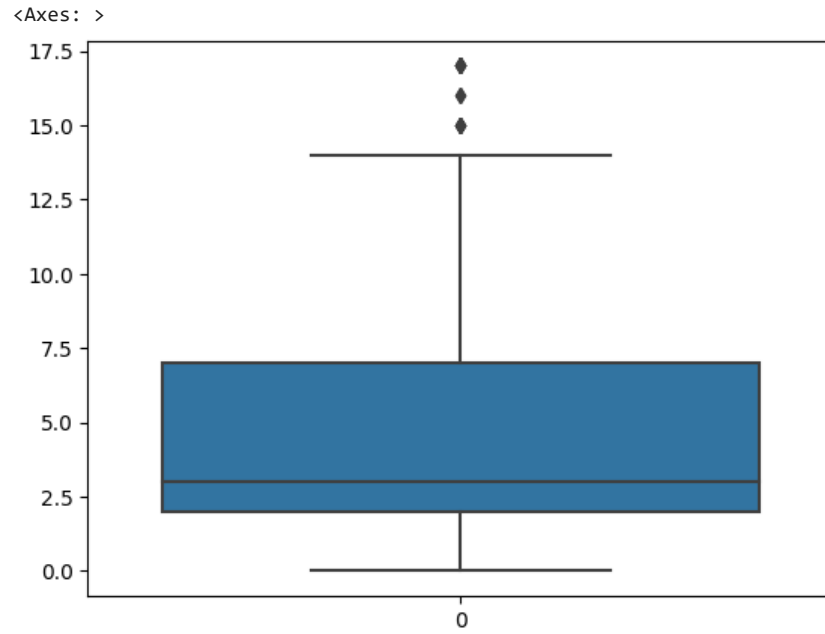
numCompaniesWorked	0.299633	0.036133	-0.029231	0.126317	NaN	-0.001231	0.012394	0.022137	0.013012	0.142301	...
--------------------	----------	----------	-----------	----------	-----	-----------	----------	----------	----------	----------	-----

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

```
<ipython-input-14-9329d5e70af4>:2: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select
sns.heatmap(df.corr(),annot=True)
<Axes: >
```




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



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

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

Outliers detected using Z-Score:

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

1348 16

Name: YearsWithCurrManager, dtype: int64

```

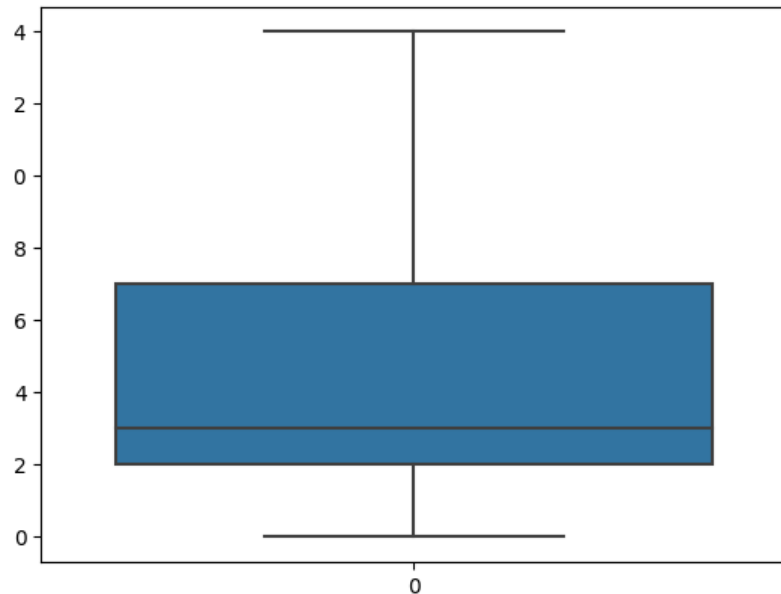
q1 = df.YearsWithCurrManager.quantile(0.25)
q3 = df.YearsWithCurrManager.quantile(0.75)
print(q1)
print(q3)
upperlimit = q3+1.5*(q3-q1)
upperlimit
lowerlimit = q1-1.5*(q3-q1)
lowerlimit
df.median()
df["YearsWithCurrManager"]=np.where(df["YearsWithCurrManager"]>upperlimit,14,df['YearsWithCurrManager'])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

```

```

}
}
python-input-17-3a17581b0650>:9: FutureWarning: The default value of numeric_only in DataFrame.median is deprecated. In a future version, it will default to False. In addition,
df.median()
<es: >

```



```

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

```

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

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

```
df.head()
```

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

5 rows × 35 columns

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

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

5 rows × 34 columns

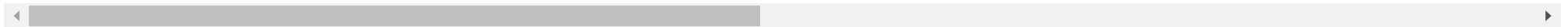
```
y=df.Attrition
y.head()
```

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

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

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

5 rows × 34 columns



```
df.columns
```

```
Index(['Age', 'Attrition', 'BusinessTravel', 'DailyRate', 'Department',
      'DistanceFromHome', 'Education', 'EducationField', 'EmployeeCount',
      'EmployeeNumber', 'EnvironmentSatisfaction', 'Gender', 'HourlyRate',
```

```
'JobInvolvement', 'JobLevel', 'JobRole', 'JobSatisfaction',
'MaritalStatus', 'MonthlyIncome', 'MonthlyRate', 'NumCompaniesWorked',
'Over18', 'OverTime', 'PercentSalaryHike', 'PerformanceRating',
'RelationshipSatisfaction', 'StandardHours', 'StockOptionLevel',
'TotalWorkingYears', 'TrainingTimesLastYear', 'WorkLifeBalance',
'YearsAtCompany', 'YearsInCurrentRole', 'YearsSinceLastPromotion',
'YearsWithCurrManager'],
dtype='object')
```

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

```
x_scaled
```

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	EnvironmentSatisfaction	...	RelationshipS
0	0.547619	1.0	0.715820	1.0	0.000000	0.25	0.2	0.0	0.000000	0.333333	...	
1	0.738095	0.5	0.126700	0.5	0.250000	0.00	0.2	0.0	0.000484	0.666667	...	
2	0.452381	1.0	0.909807	0.5	0.035714	0.25	0.8	0.0	0.001451	1.000000	...	
3	0.357143	0.5	0.923407	0.5	0.071429	0.75	0.2	0.0	0.001935	1.000000	...	
4	0.214286	1.0	0.350036	0.5	0.035714	0.00	0.6	0.0	0.002903	0.000000	...	
...	
1465	0.428571	0.5	0.559771	0.5	0.785714	0.25	0.6	0.0	0.996613	0.666667	...	
1466	0.500000	1.0	0.365784	0.5	0.178571	0.00	0.6	0.0	0.997097	1.000000	...	
1467	0.214286	1.0	0.037938	0.5	0.107143	0.50	0.2	0.0	0.998065	0.333333	...	
1468	0.738095	0.5	0.659270	1.0	0.035714	0.50	0.6	0.0	0.998549	1.000000	...	
1469	0.380952	1.0	0.376521	0.5	0.250000	0.50	0.6	0.0	1.000000	0.333333	...	

1470 rows × 34 columns

```
#Splitting Data into Train and Test.
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, 34), (294, 34), (1176,), (294,))
```

```
x_train.head()
```

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	EnvironmentSatisfaction	...	RelationshipS
1374	0.952381	1.0	0.360057	1.0	0.714286	0.50	0.2	0.0	0.937107	1.000000	...	
1092	0.642857	1.0	0.607015	0.5	0.964286	0.50	1.0	0.0	0.747460	1.000000	...	
768	0.523810	1.0	0.141732	1.0	0.892857	0.50	0.4	0.0	0.515239	0.666667	...	
569	0.428571	0.0	0.953472	1.0	0.250000	0.75	0.2	0.0	0.381229	0.000000	...	
911	0.166667	0.5	0.355762	1.0	0.821429	0.00	0.2	0.0	0.615385	0.666667	...	

5 rows × 34 columns

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

```
model.fit(x_train,y_train)
pred=model.predict(x_test)
pred
```

[illegible]

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

```
y_test
```

```
442      No
1091     No
981      Yes
785      No
1332     Yes
...
1439     No
481      No
124      Yes
198      No
1229     No
Name: Attrition, Length: 294, dtype: object
```

```
df
```

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	...	RelationshipSatisfaction	Stan
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	1	...	1	
1	49	No	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	2	...	4	
2	37	No	Travel_Rarely	1070	Research & Development	2	2	Life Sciences	1	1	...	1	

▼ Evaluation of classification model

#Accuracy score

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

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

```
0.8843537414965986
```

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

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

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

	col_0	No	Yes
Attrition			
No	242	3	
Yes	31	18	

▼ Roc-AUC curve

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

```
0.06205401, 0.61414184, 0.07466397, 0.00797252, 0.39157785,
0.05281564, 0.33160211, 0.02022395, 0.6671328 , 0.19419683,
0.0335299 , 0.10954936, 0.17130578, 0.043804 , 0.2241511 ,
0.23531373, 0.01475346, 0.06562592, 0.05019163, 0.59115162,
0.44667993, 0.07401303, 0.0449937 , 0.67637047, 0.05859033,
0.01545736, 0.03386798, 0.07021403, 0.1707141 , 0.07767295,
0.04154894, 0.08312937, 0.06997437, 0.03567429, 0.05269126,
0.05742727, 0.02144976, 0.01779053, 0.01301572, 0.02825292,
```



```

0.09708281, 0.48942319, 0.07941138, 0.25720931, 0.66861065,
0.26482373, 0.01970983, 0.30281497, 0.02858501, 0.16213966,
0.02040161, 0.2173984, 0.13768821, 0.03568054, 0.37558052,
0.03010741, 0.29718154, 0.15832399, 0.10264349, 0.08700774,
0.0815183, 0.30943969, 0.08708969, 0.07442596, 0.12300414,
0.0618342, 0.04633075, 0.07672219, 0.19834226, 0.03129952,
0.00857215, 0.02394842, 0.13606932, 0.02587787, 0.03217004,
0.0821409, 0.00518749, 0.035308, 0.03813342, 0.14270872,
0.26418695, 0.16461435, 0.27401734, 0.24146954, 0.02119787,
0.17774284, 0.34102562, 0.28338745, 0.06906981, 0.04948532,
0.24465264, 0.74929682, 0.35691434, 0.01878265, 0.08772637,
0.03239915, 0.05413857, 0.15215059, 0.07127406, 0.13828798,
0.09342465, 0.04693869, 0.02494493, 0.15041914, 0.07133392,
0.03025642, 0.05306455, 0.1165452, 0.00872431, 0.01229042,
0.17575238, 0.05005249, 0.09018395, 0.82857166, 0.03066995,
0.0228189, 0.00874605, 0.13496234, 0.16593413, 0.05060052,
0.01520085, 0.29791945, 0.54919611, 0.33581407, 0.0469494,
0.38773566, 0.61348127, 0.14171081, 0.07455884, 0.2409655,
0.09528764, 0.06730943, 0.09797576, 0.20026612, 0.20053142,
0.03046036, 0.14877431, 0.0036571, 0.11146887, 0.15912883,
0.06017571, 0.17964687, 0.06063618, 0.1199213, 0.03284092,
0.02688355, 0.06536903, 0.08335812, 0.01464284, 0.01536292,
0.37701597, 0.01262506, 0.15004068, 0.80530948, 0.11655522,
0.28461049, 0.17042029, 0.15392139, 0.02756879, 0.00599553,
0.04142216, 0.09958411, 0.11567269, 0.10448555, 0.01830036,
0.1444171, 0.1048541, 0.10079777, 0.05099176, 0.09183576,
0.02893646, 0.09754427, 0.00516687, 0.75206394, 0.04227453,
0.04018918, 0.37563319, 0.04457964, 0.72551665, 0.10583031,
0.36656526, 0.38293703, 0.32923777, 0.05248015, 0.08216713,
0.13748888, 0.04309097, 0.01429957, 0.2656631, 0.06297408,
0.16075744, 0.15388494, 0.67190498, 0.05834473, 0.28467369,
0.04694404, 0.46237195, 0.00339026, 0.13927388, 0.02695884,
0.12707414, 0.17395277, 0.0750947, 0.10135673, 0.16496216,
0.02583798, 0.01790826, 0.08850395, 0.02838351, 0.13795992,
0.08655223, 0.22164621, 0.73379009, 0.17294814, 0.40907888,
0.01503347, 0.11411826, 0.21412683, 0.32566668, 0.03366086,
0.04472831, 0.32127248, 0.05442236, 0.0242917, 0.16228044,
0.32858438, 0.22879119, 0.00852736, 0.0798162, 0.01140248,
0.14102568, 0.29116266, 0.01282151, 0.17118076, 0.04051376,
0.04165738, 0.42684273, 0.35009936, 0.0366853, 0.11692325,
0.37940034, 0.31562415, 0.79587005, 0.05488792, 0.21568794,
0.06397987, 0.00569145, 0.66085682, 0.35796045, 0.37592133,
0.3650533, 0.03568965, 0.21192376, 0.05892118, 0.06428028,
0.10143977, 0.00796354, 0.2678938, 0.4288445, 0.0652538,
0.09309022, 0.01226927, 0.14314823, 0.04989664, 0.02304292,
0.02508766, 0.06618985, 0.24272596, 0.26663754, 0.1979951,
0.26504226, 0.01648205, 0.15826843, 0.08519882, 0.02669729,
0.18757572, 0.00768502, 0.27928747, 0.0027473, 0.02506718,
0.22608608, 0.72428674, 0.07739605, 0.26575953])

```

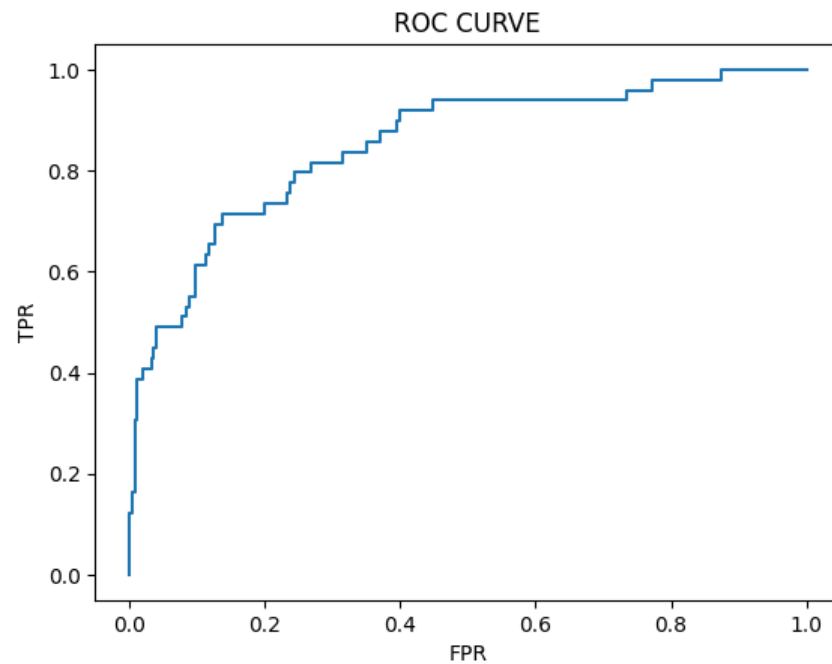
```

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

```

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

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



▼ DecisionTreeClassifier

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

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

```
▼ DecisionTreeClassifier
DecisionTreeClassifier()
```

pred

y_test

df

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	...	RelationshipSatisfaction	Stan
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	1	...	1	
1	49	No	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	2	...	4	
2	37	Yes	Travel_Rarely	1373	Research & Development	2	2	Other	1	4	...	2	
3	33	No	Travel_Frequently	1392	Research & Development	3	4	Life Sciences	1	5	...	3	
4	27	No	Travel_Rarely	591	Research & Development	2	1	Medical	1	7	...	4	
...	
1465	36	No	Travel_Frequently	884	Research & Development	23	2	Medical	1	2061	...	3	
1466	39	No	Travel_Rarely	613	Research & Development	6	1	Medical	1	2062	...	1	
1467	27	No	Travel_Rarely	155	Research & Development	4	3	Life Sciences	1	2064	...	2	
1468	49	No	Travel_Frequently	1023	Sales	2	3	Medical	1	2065	...	4	
1469	34	No	Travel_Rarely	628	Research & Development	8	3	Medical	1	2068	...	1	

1470 rows × 35 columns

▼ Evaluation of classification model

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

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

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

```
0.7619047619047619
```

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

```
array([[208, 37],
       [ 33, 16]])
```

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

col_0	0	1
row_0	208	37
1	33	16

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

	precision	recall	f1-score	support
0	0.86	0.85	0.86	245
1	0.30	0.33	0.31	49
accuracy			0.76	294
macro avg	0.58	0.59	0.58	294
weighted avg	0.77	0.76	0.77	294

▼ Roc-AUC curve

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

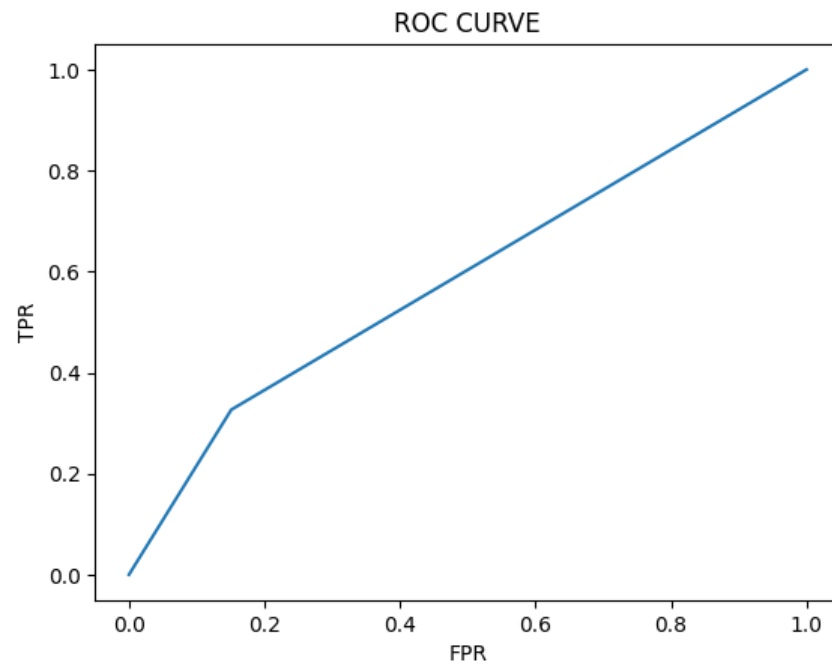
```
probability
```

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

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

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



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

```
[Text(0.321546743697479, 0.972222222222222, 'x[27] <= 0.038\ngini = 0.269\nsamples = 1176\nvalue = [988, 188]'),
Text(0.08067226890756303, 0.916666666666666, 'x[16] <= 0.75\ngini = 0.5\nsamples = 78\nvalue = [39, 39]'),
Text(0.05042016806722689, 0.861111111111111, 'x[4] <= 0.554\ngini = 0.426\nsamples = 39\nvalue = [27, 12]'),
Text(0.03361344537815126, 0.805555555555555, 'x[15] <= 0.167\ngini = 0.312\nsamples = 31\nvalue = [25, 6]'),
Text(0.020168067226890758, 0.75, 'x[16] <= 0.25\ngini = 0.49\nsamples = 7\nvalue = [3, 4]'),
Text(0.013445378151260505, 0.694444444444444, 'x[21] <= 0.5\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.0067226890756302525, 0.638888888888888, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.020168067226890758, 0.638888888888888, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.02689075630252101, 0.694444444444444, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.047058823529411764, 0.75, 'x[19] <= 0.056\ngini = 0.153\nsamples = 24\nvalue = [22, 2]'),
Text(0.040336134453781515, 0.694444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.05378151260504202, 0.694444444444444, 'x[9] <= 0.167\ngini = 0.083\nsamples = 23\nvalue = [22, 1]'),
Text(0.047058823529411764, 0.638888888888888, 'x[15] <= 0.667\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.040336134453781515, 0.583333333333333, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.05378151260504202, 0.583333333333333, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.06050420168067227, 0.638888888888888, 'gini = 0.0\nsamples = 21\nvalue = [21, 0]'),
Text(0.06722689075630252, 0.805555555555555, 'x[8] <= 0.385\ngini = 0.375\nsamples = 8\nvalue = [2, 6]'),
Text(0.06050420168067227, 0.75, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.07394957983193277, 0.75, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.11092436974789915, 0.861111111111111, 'x[11] <= 0.364\ngini = 0.426\nsamples = 39\nvalue = [12, 27]'),
Text(0.09411764705882353, 0.805555555555555, 'x[29] <= 0.167\ngini = 0.133\nsamples = 14\nvalue = [1, 13]'),
Text(0.08739495798319327, 0.75, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.10084033613445378, 0.75, 'gini = 0.0\nsamples = 13\nvalue = [0, 13]'),
Text(0.12773109243697478, 0.805555555555555, 'x[8] <= 0.105\ngini = 0.493\nsamples = 25\nvalue = [11, 14]'),
Text(0.11428571428571428, 0.75, 'x[1] <= 0.75\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
Text(0.10756302521008404, 0.694444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.12100840336134454, 0.694444444444444, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.1411764705882353, 0.75, 'x[15] <= 0.5\ngini = 0.432\nsamples = 19\nvalue = [6, 13]'),
Text(0.13445378151260504, 0.694444444444444, 'gini = 0.0\nsamples = 7\nvalue = [0, 7]'),
Text(0.14789915966386555, 0.694444444444444, 'x[6] <= 0.4\ngini = 0.5\nsamples = 12\nvalue = [6, 6]'),
Text(0.13445378151260504, 0.638888888888888, 'x[3] <= 0.75\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
Text(0.12773109243697478, 0.583333333333333, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.1411764705882353, 0.583333333333333, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.16134453781512606, 0.638888888888888, 'x[8] <= 0.249\ngini = 0.278\nsamples = 6\nvalue = [1, 5]'),
Text(0.1546218487394958, 0.583333333333333, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.16806722689075632, 0.583333333333333, 'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
Text(0.5624212184873949, 0.916666666666666, 'x[21] <= 0.5\ngini = 0.235\nsamples = 1098\nvalue = [949, 149]'),
Text(0.3126575630252101, 0.861111111111111, 'x[29] <= 0.167\ngini = 0.162\nsamples = 798\nvalue = [727, 71]'),
Text(0.1815126050420168, 0.805555555555555, 'x[8] <= 0.445\ngini = 0.38\nsamples = 47\nvalue = [35, 12]'),
Text(0.16806722689075632, 0.75, 'x[16] <= 0.75\ngini = 0.1\nsamples = 19\nvalue = [18, 1]'),
Text(0.16134453781512606, 0.694444444444444, 'gini = 0.0\nsamples = 18\nvalue = [18, 0]'),
Text(0.17478991596638654, 0.694444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.1949579831932773, 0.75, 'x[17] <= 0.094\ngini = 0.477\nsamples = 28\nvalue = [17, 11]'),
Text(0.18823529411764706, 0.694444444444444, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),
Text(0.20168067226890757, 0.694444444444444, 'x[32] <= 0.6\ngini = 0.413\nsamples = 24\nvalue = [17, 7]'),
Text(0.1949579831932773, 0.638888888888888, 'x[11] <= 0.486\ngini = 0.351\nsamples = 22\nvalue = [17, 5]'),
Text(0.18823529411764706, 0.583333333333333, 'x[24] <= 0.5\ngini = 0.496\nsamples = 11\nvalue = [6, 5]'),
Text(0.1815126050420168, 0.527777777777778, 'x[17] <= 0.417\ngini = 0.408\nsamples = 7\nvalue = [2, 5]'),
Text(0.17478991596638654, 0.472222222222222, 'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
Text(0.18823529411764706, 0.472222222222222, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.1949579831932773, 0.527777777777778, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.20168067226890757, 0.583333333333333, 'gini = 0.0\nsamples = 11\nvalue = [11, 0]'),
Text(0.20840336134453782, 0.638888888888888, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.44380252100840334, 0.805555555555555, 'x[27] <= 0.975\ngini = 0.145\nsamples = 751\nvalue = [692, 59]'),
Text(0.4370798319327731, 0.75, 'x[30] <= 0.113\ngini = 0.143\nsamples = 750\nvalue = [692, 58]'),
```

```

Text(0.303781512605042, 0.6944444444444444, 'x[9] <= 0.167\ngini = 0.218\nsamples = 257\nvalue = [225, 32]'),
Text(0.2605042016806723, 0.6388888888888888, 'x[33] <= 0.179\ngini = 0.355\nsamples = 65\nvalue = [50, 15]'),
Text(0.23865546218487396, 0.5833333333333334, 'x[33] <= 0.036\ngini = 0.303\nsamples = 59\nvalue = [48, 11]'),
Text(0.21512605042016808, 0.5277777777777778, 'x[12] <= 0.5\ngini = 0.463\nsamples = 22\nvalue = [14, 8]'),
Text(0.20168067226890757, 0.4722222222222222, 'x[11] <= 0.179\ngini = 0.198\nsamples = 9\nvalue = [8, 1]'),
Text(0.1949579831932773, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.20840336134453782, 0.4166666666666667, 'gini = 0.0\nsamples = 8\nvalue = [8, 0]'),
Text(0.22857142857142856, 0.4722222222222222, 'x[11] <= 0.4\ngini = 0.497\nsamples = 13\nvalue = [6, 7]'),
Text(0.2218487394957983, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.23529411764705882, 0.4166666666666667, 'x[4] <= 0.286\ngini = 0.346\nsamples = 9\nvalue = [2, 7]'),
Text(0.22857142857142856, 0.3611111111111111, 'x[0] <= 0.226\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.2218487394957983, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.23529411764705882, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.24201680672268908, 0.3611111111111111, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.26218487394957984, 0.5277777777777778, 'x[15] <= 0.167\ngini = 0.149\nsamples = 37\nvalue = [34, 3]'),
Text(0.25546218487394956, 0.4722222222222222, 'x[29] <= 0.5\ngini = 0.5\nsamples = 6\nvalue = [3, 3]'),
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Text(0.2689075630252101, 0.4722222222222222, 'gini = 0.0\nsamples = 31\nvalue = [31, 0]'),
Text(0.2823529411764706, 0.5833333333333334, 'x[8] <= 0.065\ngini = 0.444\nsamples = 6\nvalue = [2, 4]'),
Text(0.27563025210084036, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.28907563025210087, 0.5277777777777778, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),
Text(0.34705882352941175, 0.6388888888888888, 'x[0] <= 0.321\ngini = 0.161\nsamples = 192\nvalue = [175, 17]'),
Text(0.3092436974789916, 0.5833333333333334, 'x[6] <= 0.1\ngini = 0.294\nsamples = 67\nvalue = [55, 12]'),
Text(0.3025210084033613, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.31596638655462184, 0.5277777777777778, 'x[29] <= 0.5\ngini = 0.26\nsamples = 65\nvalue = [55, 10]'),
Text(0.292436974789916, 0.4722222222222222, 'x[6] <= 0.5\ngini = 0.469\nsamples = 16\nvalue = [10, 6]'),
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Text(0.2991596638655462, 0.4166666666666667, 'x[9] <= 0.833\ngini = 0.444\nsamples = 9\nvalue = [3, 6]'),
Text(0.292436974789916, 0.3611111111111111, 'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
Text(0.3058823529411765, 0.3611111111111111, 'x[3] <= 0.75\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.2991596638655462, 0.3055555555555556, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.3126050420168067, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.33949579831932775, 0.4722222222222222, 'x[2] <= 0.037\ngini = 0.15\nsamples = 49\nvalue = [45, 4]'),
Text(0.33277310924369746, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.346218487394958, 0.4166666666666667, 'x[2] <= 0.938\ngini = 0.117\nsamples = 48\nvalue = [45, 3]'),
Text(0.33949579831932775, 0.3611111111111111, 'x[5] <= 0.875\ngini = 0.081\nsamples = 47\nvalue = [45, 2]'),
Text(0.32605042016806723, 0.3055555555555556, 'x[12] <= 0.167\ngini = 0.043\nsamples = 45\nvalue = [44, 1]'),
Text(0.31932773109243695, 0.25, 'x[22] <= 0.214\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.3126050420168067, 0.19444444444444445, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.32605042016806723, 0.19444444444444445, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.33277310924369746, 0.25, 'gini = 0.0\nsamples = 42\nvalue = [42, 0]'),
Text(0.35294117647058826, 0.3055555555555556, 'x[8] <= 0.246\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.346218487394958, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.3596638655462185, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.35294117647058826, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.38487394957983195, 0.5833333333333334, 'x[8] <= 0.022\ngini = 0.077\nsamples = 125\nvalue = [120, 5]'),
Text(0.3663865546218487, 0.5277777777777778, 'x[27] <= 0.188\ngini = 0.5\nsamples = 4\nvalue = [2, 2]'),
Text(0.3596638655462185, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.373109243697479, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.40336134453781514, 0.5277777777777778, 'x[18] <= 0.968\ngini = 0.048\nsamples = 121\nvalue = [118, 3]'),
Text(0.3865546218487395, 0.4722222222222222, 'x[2] <= 0.98\ngini = 0.033\nsamples = 118\nvalue = [116, 2]'),
Text(0.373109243697479, 0.4166666666666667, 'x[14] <= 0.938\ngini = 0.017\nsamples = 114\nvalue = [113, 1]'),
Text(0.3663865546218487, 0.3611111111111111, 'gini = 0.0\nsamples = 107\nvalue = [107, 0]'),
Text(0.3798319327731092, 0.3611111111111111, 'x[11] <= 0.193\ngini = 0.245\nsamples = 7\nvalue = [6, 1]'),

```



```

Text(0.373109243697479, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.3865546218487395, 0.3055555555555556, 'gini = 0.0\nsamples = 6\nvalue = [6, 0]'),
Text(0.4, 0.4166666666666667, 'x[19] <= 0.833\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.39327731092436974, 0.3611111111111111, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.40672268907563025, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.42016806722689076, 0.4722222222222222, 'x[23] <= 0.5\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.4134453781512605, 0.4166666666666667, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.426890756302521, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5703781512605042, 0.6944444444444444, 'x[30] <= 0.787\ngini = 0.1\nsamples = 493\nvalue = [467, 26]'),
Text(0.530672268907563, 0.6388888888888888, 'x[15] <= 0.5\ngini = 0.094\nsamples = 486\nvalue = [462, 24]'),
Text(0.4714285714285714, 0.5833333333333334, 'x[14] <= 0.938\ngini = 0.154\nsamples = 191\nvalue = [175, 16]'),
Text(0.4647058823529412, 0.5277777777777778, 'x[18] <= 0.481\ngini = 0.145\nsamples = 190\nvalue = [175, 15]'),
Text(0.4470588235294118, 0.4722222222222222, 'x[18] <= 0.47\ngini = 0.221\nsamples = 95\nvalue = [83, 12]'),
Text(0.4403361344537815, 0.4166666666666667, 'x[33] <= 0.964\ngini = 0.207\nsamples = 94\nvalue = [83, 11]'),
Text(0.4336134453781513, 0.3611111111111111, 'x[5] <= 0.375\ngini = 0.192\nsamples = 93\nvalue = [83, 10]'),
Text(0.4117647058823529, 0.3055555555555556, 'x[6] <= 0.9\ngini = 0.363\nsamples = 21\nvalue = [16, 5]'),
Text(0.4050420168067227, 0.25, 'x[17] <= 0.413\ngini = 0.266\nsamples = 19\nvalue = [16, 3]'),
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Text(0.38487394957983195, 0.1388888888888889, 'x[6] <= 0.4\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
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Text(0.4588235294117647, 0.1944444444444445, 'x[28] <= 0.583\ngini = 0.061\nsamples = 64\nvalue = [62, 2]'),
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Text(0.5663865546218487, 0.5277777777777778, 'x[32] <= 0.7\ngini = 0.159\nsamples = 46\nvalue = [42, 4]'),

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Text(0.5596638655462185, 0.4722222222222222, 'x[27] <= 0.688\ngini = 0.124\nsamples = 45\nvalue = [42, 3]'),
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Text(0.5495798319327732, 0.3055555555555556, 'x[11] <= 0.707\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
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Text(0.5563025210084034, 0.25, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.5630252100840336, 0.3055555555555556, 'gini = 0.0\nsamples = 37\nvalue = [37, 0]'),
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Text(0.6168067226890757, 0.4166666666666667, 'x[26] <= 0.667\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
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Text(0.6235294117647059, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6436974789915967, 0.4166666666666667, 'x[24] <= 0.167\ngini = 0.017\nsamples = 238\nvalue = [236, 2]'),
Text(0.6369747899159663, 0.3611111111111111, 'x[29] <= 0.833\ngini = 0.073\nsamples = 53\nvalue = [51, 2]'),
Text(0.6235294117647059, 0.3055555555555556, 'x[33] <= 0.107\ngini = 0.041\nsamples = 48\nvalue = [47, 1]'),
Text(0.6168067226890757, 0.25, 'x[18] <= 0.824\ngini = 0.245\nsamples = 7\nvalue = [6, 1]'),
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Text(0.6504201680672269, 0.3055555555555556, 'x[17] <= 0.38\ngini = 0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.6436974789915967, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6571428571428571, 0.25, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
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Text(0.8121848739495798, 0.8611111111111112, 'x[17] <= 0.157\ngini = 0.385\nsamples = 300\nvalue = [222, 78]'),
Text(0.7201680672268908, 0.8055555555555556, 'x[26] <= 0.167\ngini = 0.5\nsamples = 96\nvalue = [49, 47]'),
Text(0.6840336134453782, 0.75, 'x[4] <= 0.161\ngini = 0.459\nsamples = 42\nvalue = [15, 27]'),
Text(0.6571428571428571, 0.6944444444444444, 'x[8] <= 0.415\ngini = 0.499\nsamples = 23\nvalue = [12, 11]'),
Text(0.6369747899159663, 0.6388888888888888, 'x[18] <= 0.561\ngini = 0.355\nsamples = 13\nvalue = [3, 10]'),
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Text(0.6773109243697479, 0.6388888888888888, 'x[24] <= 0.167\ngini = 0.18\nsamples = 10\nvalue = [9, 1]'),
Text(0.6705882352941176, 0.5833333333333334, 'x[30] <= 0.088\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
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Text(0.704201680672269, 0.6388888888888888, 'x[11] <= 0.2\ngini = 0.198\nsamples = 18\nvalue = [2, 16]'),
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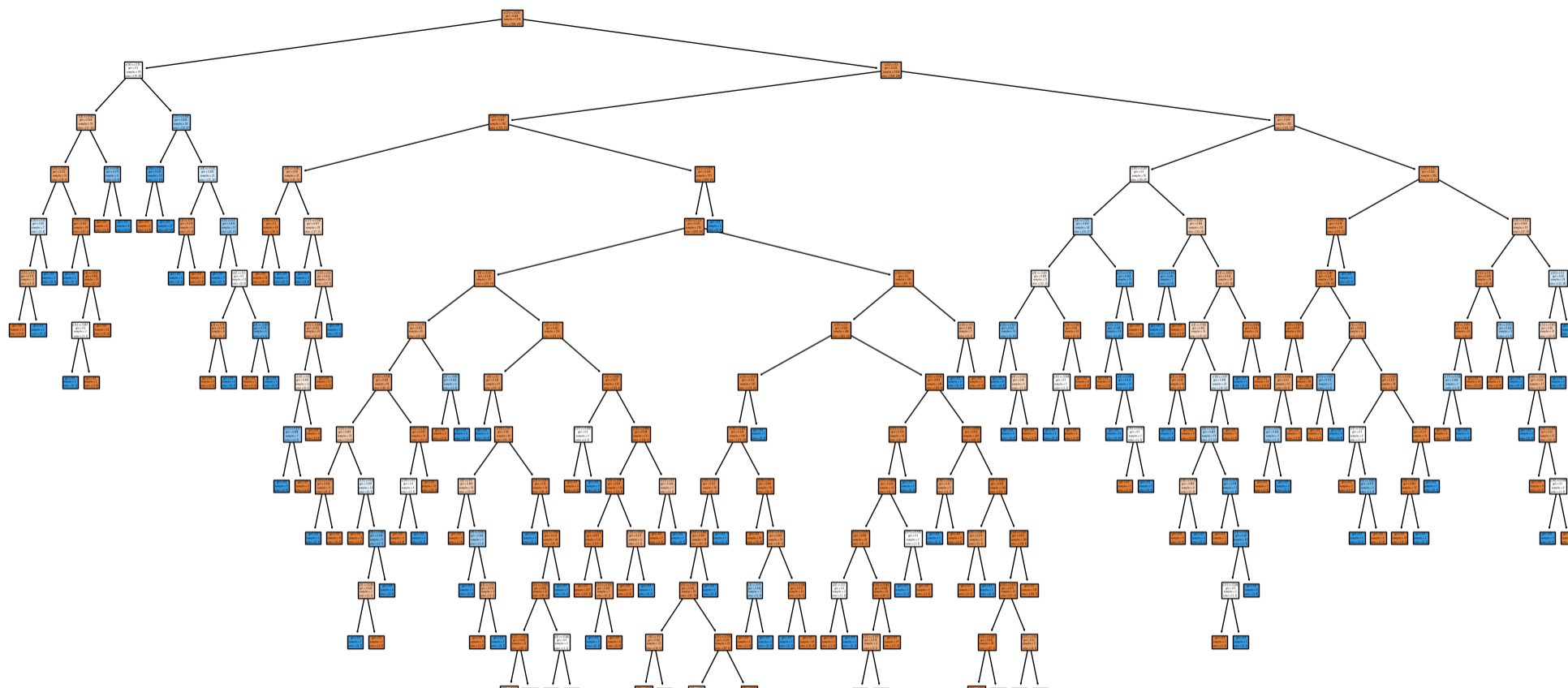
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Text(0.7378151260504202, 0.6944444444444444, 'x[12] <= 0.833\ngini = 0.245\nsamples = 7\nvalue = [1, 6]'),
Text(0.7310924369747899, 0.6388888888888888, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
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Text(0.8521008403361344, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
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Text(0.9394957983193277, 0.6944444444444444, 'x[32] <= 0.4\ngini = 0.238\nsamples = 29\nvalue = [25, 4]'),
Text(0.9260504201680673, 0.6388888888888888, 'x[8] <= 0.071\ngini = 0.142\nsamples = 26\nvalue = [24, 2]'),
Text(0.9193277310924369, 0.5833333333333334, 'x[17] <= 0.199\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.9126050420168067, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.9260504201680673, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.9327731092436975, 0.5833333333333334, 'gini = 0.0\nsamples = 23\nvalue = [23, 0]'),
Text(0.9529411764705882, 0.6388888888888888, 'x[9] <= 0.333\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.946218487394958, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.9596638655462185, 0.5833333333333334, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.9865546218487395, 0.6944444444444444, 'x[32] <= 0.1\ngini = 0.49\nsamples = 28\nvalue = [12, 16]'),
Text(0.9798319327731092, 0.6388888888888888, 'x[12] <= 0.833\ngini = 0.48\nsamples = 20\nvalue = [12, 8]'),
Text(0.973109243697479, 0.5833333333333334, 'x[30] <= 0.013\ngini = 0.415\nsamples = 17\nvalue = [12, 5]'),
Text(0.9663865546218487, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.9798319327731092, 0.5277777777777778, 'x[18] <= 0.505\ngini = 0.32\nsamples = 15\nvalue = [12, 3]'),
Text(0.973109243697479, 0.4722222222222222, 'gini = 0.0\nsamples = 9\nvalue = [9, 0]'),
Text(0.9865546218487395, 0.4722222222222222, 'x[18] <= 0.706\ngini = 0.5\nsamples = 6\nvalue = [3, 3]'),
Text(0.9798319327731092, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.9932773109243698, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.9865546218487395, 0.5833333333333334, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.9932773109243698, 0.6388888888888888, 'gini = 0.0\nsamples = 8\nvalue = [0, 8]')]
```





```
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']
}

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

grid_search.fit(x_train,y_train)
```

<https://colab.research.google.com/drive/1FtFEIbpZYXv07-VfA2mtU7Jwk66c4Oi#scrollTo=tfDa3XCCFqth&printMode=true>

<https://colab.research.google.com/drive/1FtFEIbpZYXv07-VfA2mtU7Jwk66c4Oi> #scrollTo=tfDa3XCCFqth&printMode=true

<https://colab.research.google.com/drive/1FtFEIbpZYXv07-VfA2mtU7Jwk66c4Oi#scrollTo=tfDa3XCCFqth&printMode=true>


```
warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep th
warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep th
warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep th
warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep th
grid_search.best_params_

{'criterion': 'gini',
 'max_depth': 5,
 'max_features': 'auto',
 'splitter': 'random'}

/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep th
dtc_cv=DecisionTreeClassifier(criterion='entropy',
                             max_depth=3,
                             max_features='sqrt',
                             splitter='best')
dtc_cv.fit(x_train,y_train)

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

warnings.warn(

pred=dtc_cv.predict(x_test)

/usr/local/lib/python3.10/dist-packages/sklearn/tree/_classes.py:269: FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep th

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

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

	precision	recall	f1-score	support
0	0.84	0.97	0.90	245
1	0.22	0.04	0.07	49
accuracy			0.82	294
macro avg	0.53	0.51	0.48	294
weighted avg	0.73	0.82	0.76	294

RandomForestClassifier

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

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

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

rfc_cv.fit(x_train,y_train)
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/model_selection/_validation.py:378: FitFailedWarning:
```

```
50 fits failed out of a total of 700
```

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

```
Below are more details about the failures:
```

```
#label encoding
```

```
from sklearn.preprocessing import LabelEncoder
```

```
le=LabelEncoder()
```

```
y=le.fit_transform(y)
```

```
#label encoding
```

```
from sklearn.preprocessing import LabelEncoder
```

```
le=LabelEncoder()
```

```
pred=le.fit_transform(pred)
```

```
File /usr/local/lib/python3.10/dist-packages/sklearn/utils/_param_validation.py, line 97, in validate_parameter_constraints
```

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

	precision	recall	f1-score	support
0	0.85	0.98	0.91	245
1	0.56	0.10	0.17	49
accuracy			0.84	294
macro avg	0.70	0.54	0.54	294
weighted avg	0.80	0.84	0.79	294

```
0.8613/911 0.85/98928 0.85/9/4/9 0.85965522 0.85/96031 0.85/96031
```

```
rfc_cv.best_params_
```

```
{'max_depth': 11, 'max_features': 4}
```

```

GridSearchCV
└─ estimator: RandomForestClassifier
   └─ RandomForestClassifier

```