

# Assignment-4

Name-Aniket Chattopadhyay

Reg no-21BEC1564

Email- aniket.chattopadhyay2021@vitstudent.ac.in

Campus-VIT Chennai

Slot- Morning

## Data Preprocessing

1. import the necessary libraries
2. import the dataset
3. Handling null values
4. Data visualization
5. outlier detection
6. Seperate Dependent and Independent variables.
7. Encoding
8. Splitting into training and testing set
9. Perform feature scaling

### 1. Import the necessary libraries.

```
In [155... import numpy as np
import seaborn as sns
import pandas as pd
import matplotlib.pyplot as plt
```

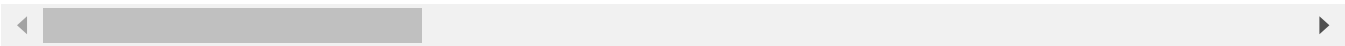
### 2. import the dataset

```
In [156... dataset=pd.read_csv("Employee-Attrition.csv")
dataset.head()
```

Out[156]:

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

5 rows × 35 columns



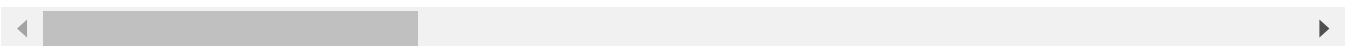
In [157...

dataset.tail()

Out[157]:

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	Ed
1465	36	No	Travel_Frequently	884	Research & Development	23	2	
1466	39	No	Travel_Rarely	613	Research & Development	6	1	
1467	27	No	Travel_Rarely	155	Research & Development	4	3	
1468	49	No	Travel_Frequently	1023	Sales	2	3	
1469	34	No	Travel_Rarely	628	Research & Development	8	3	

5 rows × 35 columns



In [158...

dataset.shape

Out[158]:

(1470, 35)

In [159...

dataset.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

```

Each row in the dataset represents an employee of a fictional company ABC, and contains the following features:

Age: Age of employees

Department: Department of work

Distance from home

Education: 1-Below College; 2-College; 3-Bachelor; 4-Master; 5-Doctor;

Education Field

Environment Satisfaction: 1-Low; 2-Medium; 3-High; 4-Very High

Job Satisfaction: 1-Low; 2-Medium; 3-High; 4-Very High

Marital Status

Monthly Income

Num Companies Worked: Number of companies worked prior to ABC

Work Life Balance: 1-Bad; 2-Good; 3-Better; 4-Best

Years At Company: Current years of service

Attrition: Employee attrition status

In [160...

dataset.describe()

Out[160]:

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount	EmployeeNum
count	1470.000000	1470.000000	1470.000000	1470.000000	1470.0	1470.000000
mean	36.923810	802.485714	9.192517	2.912925	1.0	1024.865000
std	9.135373	403.509100	8.106864	1.024165	0.0	602.024000
min	18.000000	102.000000	1.000000	1.000000	1.0	1.000000
25%	30.000000	465.000000	2.000000	2.000000	1.0	491.250000
50%	36.000000	802.000000	7.000000	3.000000	1.0	1020.500000
75%	43.000000	1157.000000	14.000000	4.000000	1.0	1555.750000
max	60.000000	1499.000000	29.000000	5.000000	1.0	2068.000000

8 rows × 26 columns

In [161...

dataset.corr()

<ipython-input-161-c187c74d1e71>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning.  
dataset.corr()

Out[161]:

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount	En
<b>Age</b>	1.000000	0.010661	-0.001686	0.208034	NaN	
<b>DailyRate</b>	0.010661	1.000000	-0.004985	-0.016806	NaN	
<b>DistanceFromHome</b>	-0.001686	-0.004985	1.000000	0.021042	NaN	
<b>Education</b>	0.208034	-0.016806	0.021042	1.000000	NaN	
<b>EmployeeCount</b>	NaN	NaN	NaN	NaN	NaN	
<b>EmployeeNumber</b>	-0.010145	-0.050990	0.032916	0.042070	NaN	
<b>EnvironmentSatisfaction</b>	0.010146	0.018355	-0.016075	-0.027128	NaN	
<b>HourlyRate</b>	0.024287	0.023381	0.031131	0.016775	NaN	
<b>JobInvolvement</b>	0.029820	0.046135	0.008783	0.042438	NaN	
<b>JobLevel</b>	0.509604	0.002966	0.005303	0.101589	NaN	
<b>JobSatisfaction</b>	-0.004892	0.030571	-0.003669	-0.011296	NaN	
<b>MonthlyIncome</b>	0.497855	0.007707	-0.017014	0.094961	NaN	
<b>MonthlyRate</b>	0.028051	-0.032182	0.027473	-0.026084	NaN	
<b>NumCompaniesWorked</b>	0.299635	0.038153	-0.029251	0.126317	NaN	
<b>PercentSalaryHike</b>	0.003634	0.022704	0.040235	-0.011111	NaN	
<b>PerformanceRating</b>	0.001904	0.000473	0.027110	-0.024539	NaN	
<b>RelationshipSatisfaction</b>	0.053535	0.007846	0.006557	-0.009118	NaN	
<b>StandardHours</b>	NaN	NaN	NaN	NaN	NaN	
<b>StockOptionLevel</b>	0.037510	0.042143	0.044872	0.018422	NaN	
<b>TotalWorkingYears</b>	0.680381	0.014515	0.004628	0.148280	NaN	
<b>TrainingTimesLastYear</b>	-0.019621	0.002453	-0.036942	-0.025100	NaN	
<b>WorkLifeBalance</b>	-0.021490	-0.037848	-0.026556	0.009819	NaN	
<b>YearsAtCompany</b>	0.311309	-0.034055	0.009508	0.069114	NaN	
<b>YearsInCurrentRole</b>	0.212901	0.009932	0.018845	0.060236	NaN	
<b>YearsSinceLastPromotion</b>	0.216513	-0.033229	0.010029	0.054254	NaN	
<b>YearsWithCurrManager</b>	0.202089	-0.026363	0.014406	0.069065	NaN	

26 rows × 26 columns

### 3. Checking for Null Values

In [162... dataset.isnull().any()

```
Out[162]: 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
```

```
In [163... dataset.isnull().sum()
```

```

Out[163]: 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

```

#### 4. Data Visualization

##### Visualizing Distribution of features

```

In [164... plt.figure(figsize = (15, 7))
plt.style.use('seaborn-white')
plt.subplot(331)
sns.distplot(dataset['Age'])
fig = plt.gcf()
fig.set_size_inches(10,10)

plt.subplot(332)
sns.distplot(dataset['DistanceFromHome'])
fig = plt.gcf()
fig.set_size_inches(10,10)

plt.subplot(333)
sns.distplot(dataset['DailyRate'])
fig = plt.gcf()
fig.set_size_inches(10,10)

plt.subplot(334)
sns.distplot(dataset['YearsAtCompany'])
fig = plt.gcf()
fig.set_size_inches(10,10)

```

```
plt.subplot(335)
sns.distplot(dataset['TotalWorkingYears'])
fig = plt.gcf()
fig.set_size_inches(10,10)

plt.subplot(336)
sns.distplot(dataset['NumCompaniesWorked'])
fig = plt.gcf()
fig.set_size_inches(10,10)

plt.subplot(337)
sns.distplot(dataset['MonthlyRate'])
fig = plt.gcf()
fig.set_size_inches(10,10)

plt.subplot(338)
sns.distplot(dataset['MonthlyIncome'])
fig = plt.gcf()
fig.set_size_inches(10,10)

plt.subplot(339)
sns.distplot(dataset['PercentSalaryHike'])
fig = plt.gcf()
fig.set_size_inches(10,10)
```



```
<ipython-input-164-2e0bc05fbe5a>:2: MatplotlibDeprecationWarning: The seaborn styles shipped by Matplotlib are deprecated since 3.6, as they no longer correspond to the styles shipped by seaborn. However, they will remain available as 'seaborn-v0_8-<style>'. Alternatively, directly use the seaborn API instead.  
plt.style.use('seaborn-white')  
<ipython-input-164-2e0bc05fbe5a>:4: UserWarning:  
  
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.  
  
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
  
For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751  
  
sns.distplot(dataset['Age'])  
<ipython-input-164-2e0bc05fbe5a>:9: UserWarning:  
  
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.  
  
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
  
For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751  
  
sns.distplot(dataset['DistanceFromHome'])  
<ipython-input-164-2e0bc05fbe5a>:14: UserWarning:  
  
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.  
  
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
  
For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751  
  
sns.distplot(dataset['DailyRate'])  
<ipython-input-164-2e0bc05fbe5a>:19: UserWarning:  
  
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.  
  
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
  
For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751  
  
sns.distplot(dataset['YearsAtCompany'])  
<ipython-input-164-2e0bc05fbe5a>:24: UserWarning:  
  
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.  
  
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
  
For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751  
  
sns.distplot(dataset['TotalWorkingYears'])  
<ipython-input-164-2e0bc05fbe5a>:29: UserWarning:  
  
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.
```

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

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

```
sns.distplot(dataset['NumCompaniesWorked'])  
<ipython-input-164-2e0bc05fbe5a>:34: UserWarning:
```

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

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

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

```
sns.distplot(dataset['MonthlyRate'])  
<ipython-input-164-2e0bc05fbe5a>:39: UserWarning:
```

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

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

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

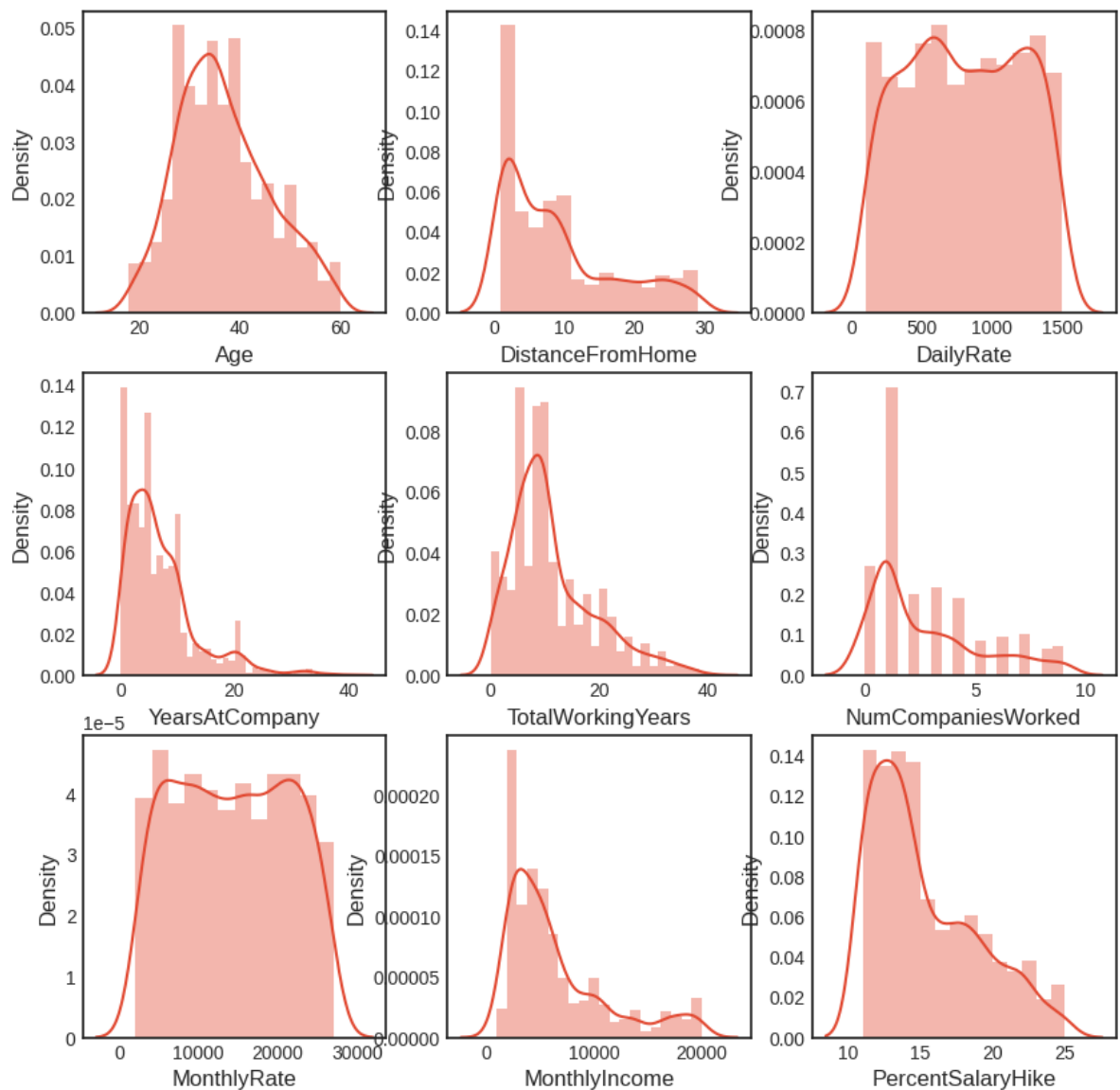
```
sns.distplot(dataset['MonthlyIncome'])  
<ipython-input-164-2e0bc05fbe5a>:44: UserWarning:
```

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

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

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

```
sns.distplot(dataset['PercentSalaryHike'])
```



### Inference:

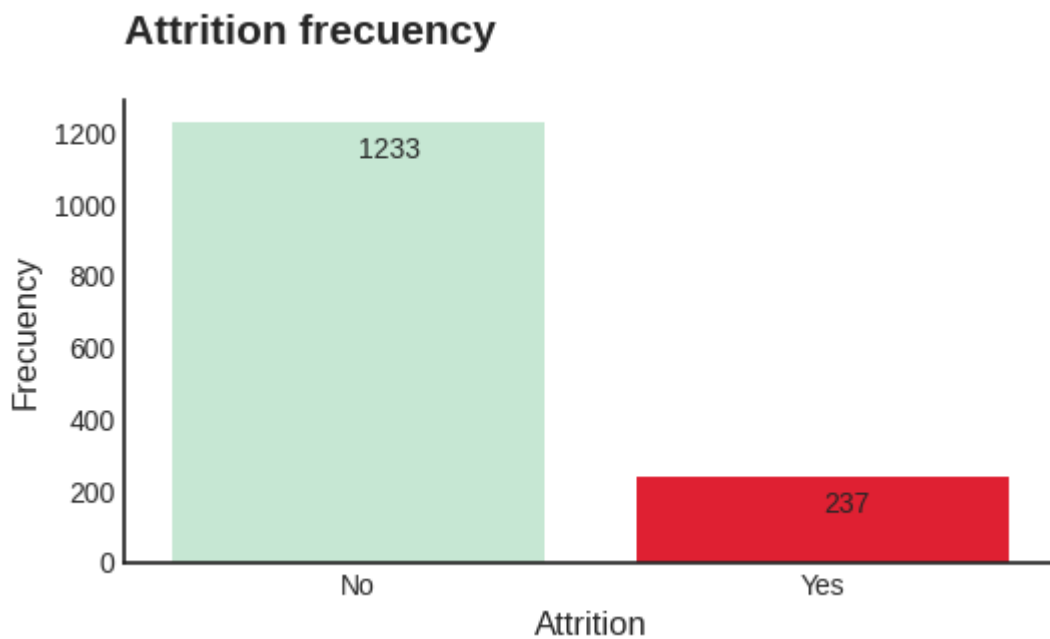
Monthly and Daily Rates has proportional normal distribution

Most of the features except age parameter are right skewed

### Attrition frequency plotting

In [165...

```
colors = ["#C0EDD2", "#FF0018"]
df_class = dataset["Attrition"].value_counts().reset_index()
plt.figure(figsize=(6,3))
ax = sns.barplot(x="index", y="Attrition", data=df_class, palette=colors)
sns.despine()
ax.text(-0.5, 1450, "Attrition frequency",
        fontsize=15,
        fontweight='bold')
for num, text in enumerate(df_class["Attrition"]):
    ax.text(num, text-100, text)
plt.xlabel("Attrition")
plt.ylabel("Frequency")
plt.show(block=False)
```



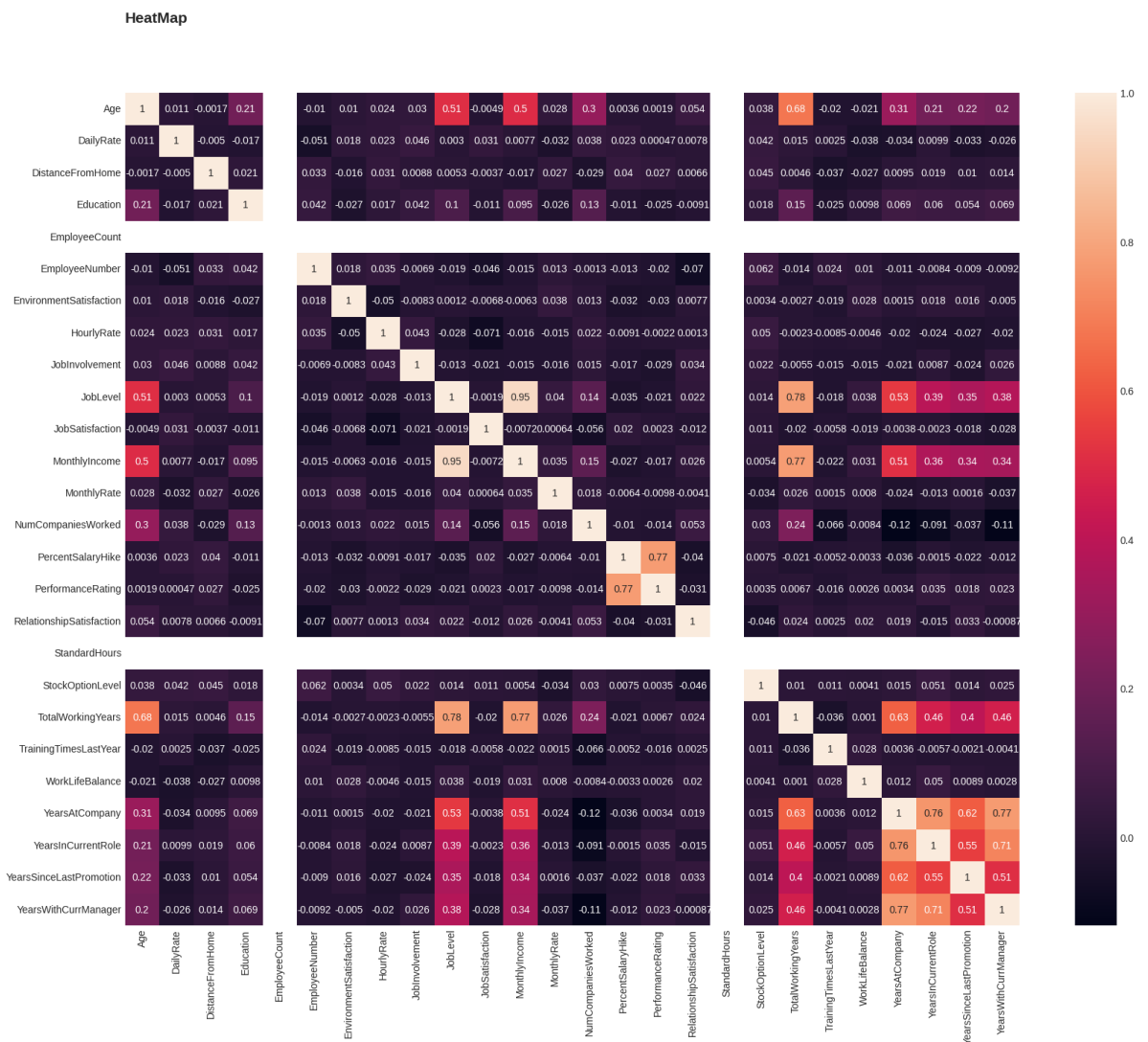
### Inference

Attrition frequency is plotted, which clearly shows no's frequency more than yes'.

### Numeric and Categorical data heatmap plotting

In [166...

```
# numeric and categorical
num_col = dataset.describe().columns.tolist()
cat_col = dataset.describe(include="object").columns.tolist()
corr = dataset[num_col].corr()
plt.subplots(figsize=(20,15))
ax = sns.heatmap(
    corr, annot=True)
ax.text(0, -2.2, "HeatMap",
        fontsize=15,
        fontweight='bold')
plt.show(block=False)
```



## Inference- The band which has lighter colour has high correlation

Boxes in lower right corner have high correlation, but those are relationship between their working period which obviously correlates and has nothing to aid for our objective, so we neglect that

Job level and Monthly income has perfect positive correlation which means higher the job level, better the income

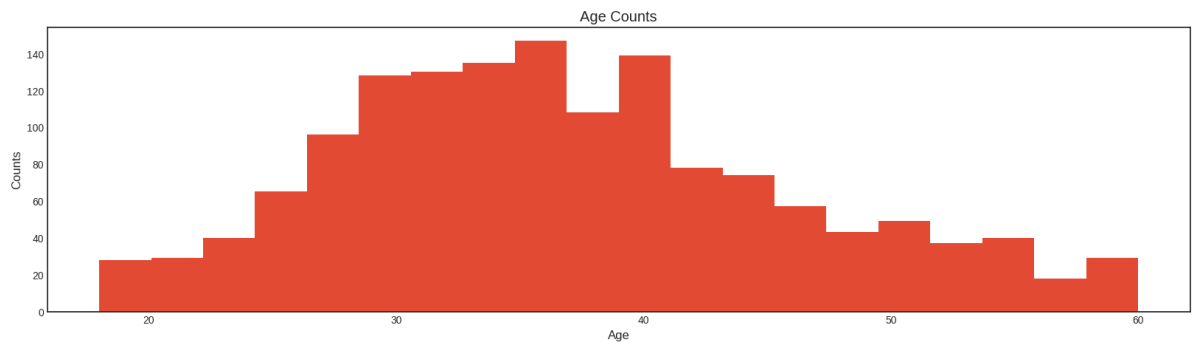
Also Job level has high correlation with Working years, when the employee's year of working increases he/she is being promoted based on seniority

There is also a high correlation between Monthly Income and Total working years which depicts that employees earn higher income owing to their seniority with the firm

## Age Distribution-Histogram

In [167...

```
plt.figure(figsize=(20,5))
plt.hist(dataset.Age,bins=20)
plt.xlabel("Age")
plt.ylabel("Counts")
plt.title("Age Counts")
plt.show()
```



## Inference

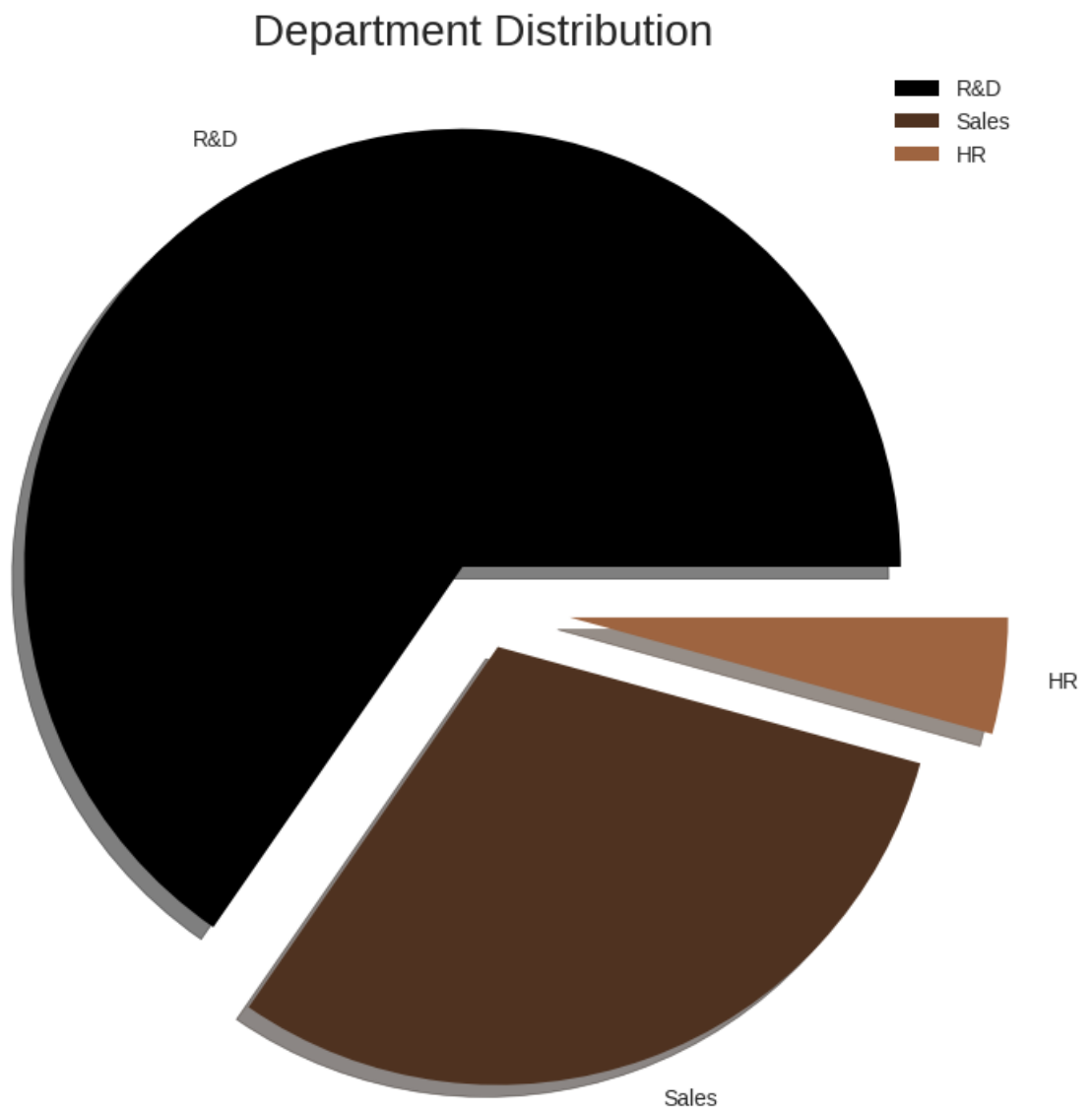
From the age distribution histogram, we can conclude that employees aged between 30-40 are working in IBM

## Department Distribution-Pie Chart

In [168...

```
labels = ['R&D', 'Sales', 'HR']
sizes = dataset['Department'].value_counts()
colors = plt.cm.copper(np.linspace(0, 1, 5))
explode = [0.1, 0.1, 0.2]

plt.rcParams['figure.figsize'] = (9, 9)
plt.pie(sizes, labels = labels, colors = colors, explode = explode, shadow = True)
plt.title('Department Distribution', fontsize = 20)
plt.legend()
plt.show()
```

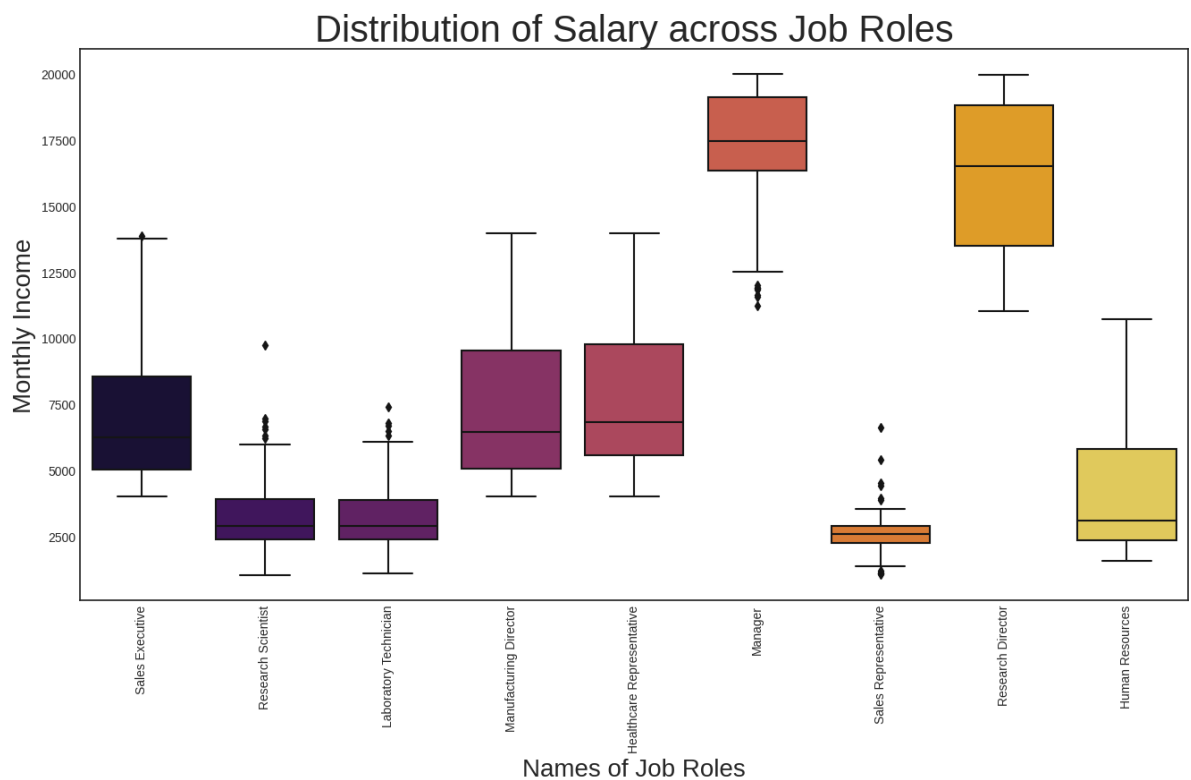


#### Inference:

The given dataset has majority of employees from Research and Development followed by Sales and HR department

#### Distribution of Salary across Job Roles-Boxplot

```
In [169... plt.rcParams['figure.figsize'] = (16, 8)
ax = sns.boxplot(x = dataset['JobRole'], y = dataset['MonthlyIncome'], data = dataset)
ax.set_xlabel(xlabel = 'Names of Job Roles', fontsize = 20)
ax.set_ylabel(ylabel = 'Monthly Income', fontsize = 20)
ax.set_title(label = 'Distribution of Salary across Job Roles', fontsize = 30)
plt.xticks(rotation = 90)
plt.show()
```



### Inference

Highest average monthly income(17500) is given to Managers

Lowest average monthly income(2500) is given to Sales Representative, But there are many outliers in Sales Representative income

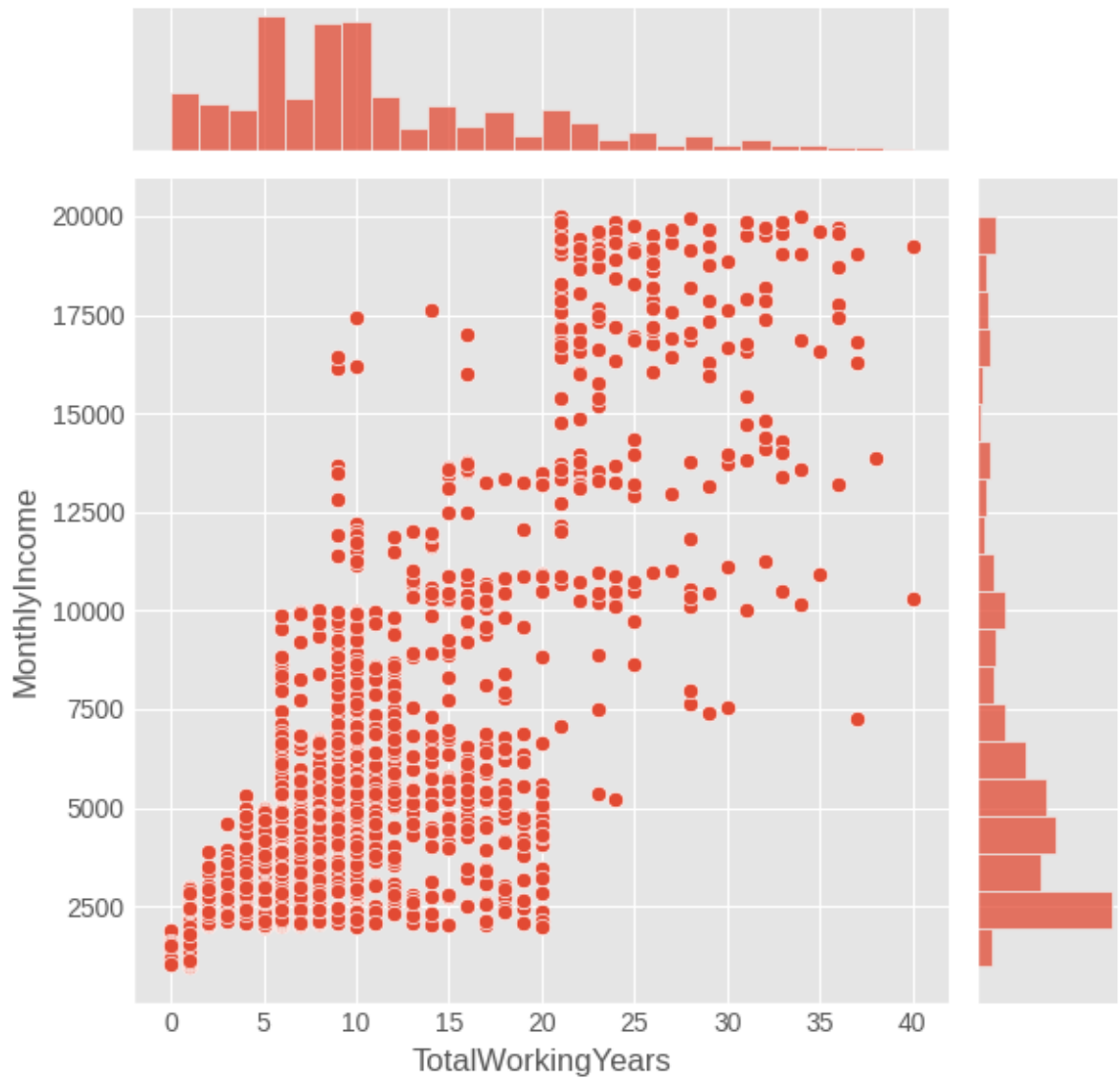
Research Scientist, Laboratory Technician and HR gets almost same average income

### Relationship with working years and income- Joint & Im Plot

```
In [170... plt.figure(figsize=(10,5))
plt.style.use('ggplot')
sns.jointplot(x='TotalWorkingYears', y='MonthlyIncome', data=dataset)
```

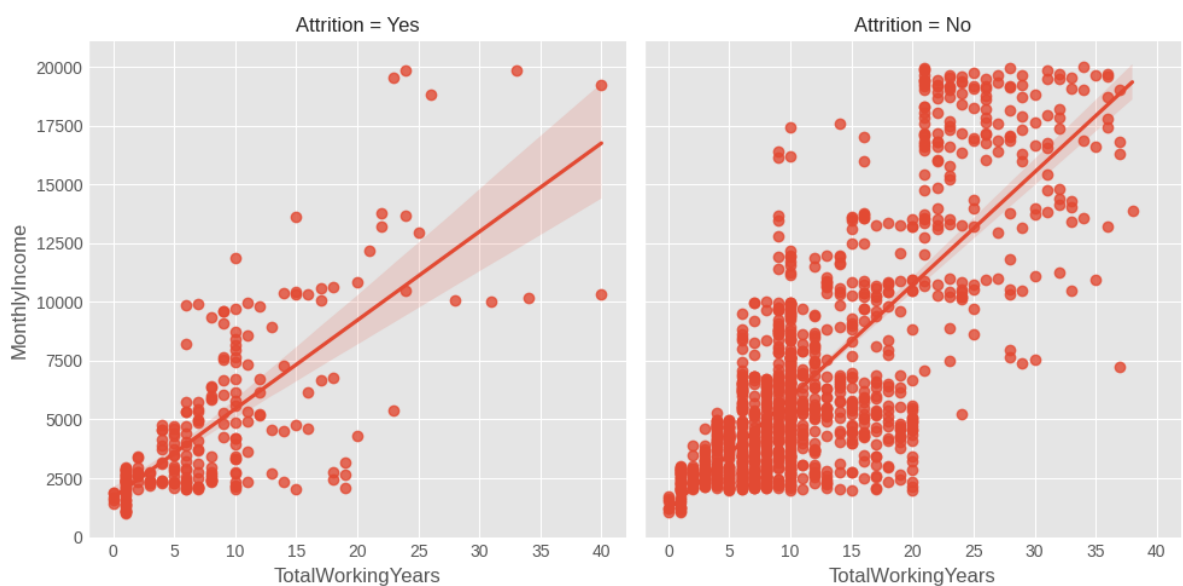
```
Out[170]: <seaborn.axisgrid.JointGrid at 0x7a874f3eada0>
<Figure size 1000x500 with 0 Axes>
```





In [171...

```
sns.lmplot(x = 'TotalWorkingYears', y = 'MonthlyIncome', data = dataset, col = 'Attrition',
plt.show())
```



## Inference

Employees who had less working years were receiving comparatively lower salary

There are many fresher employees working in IBM

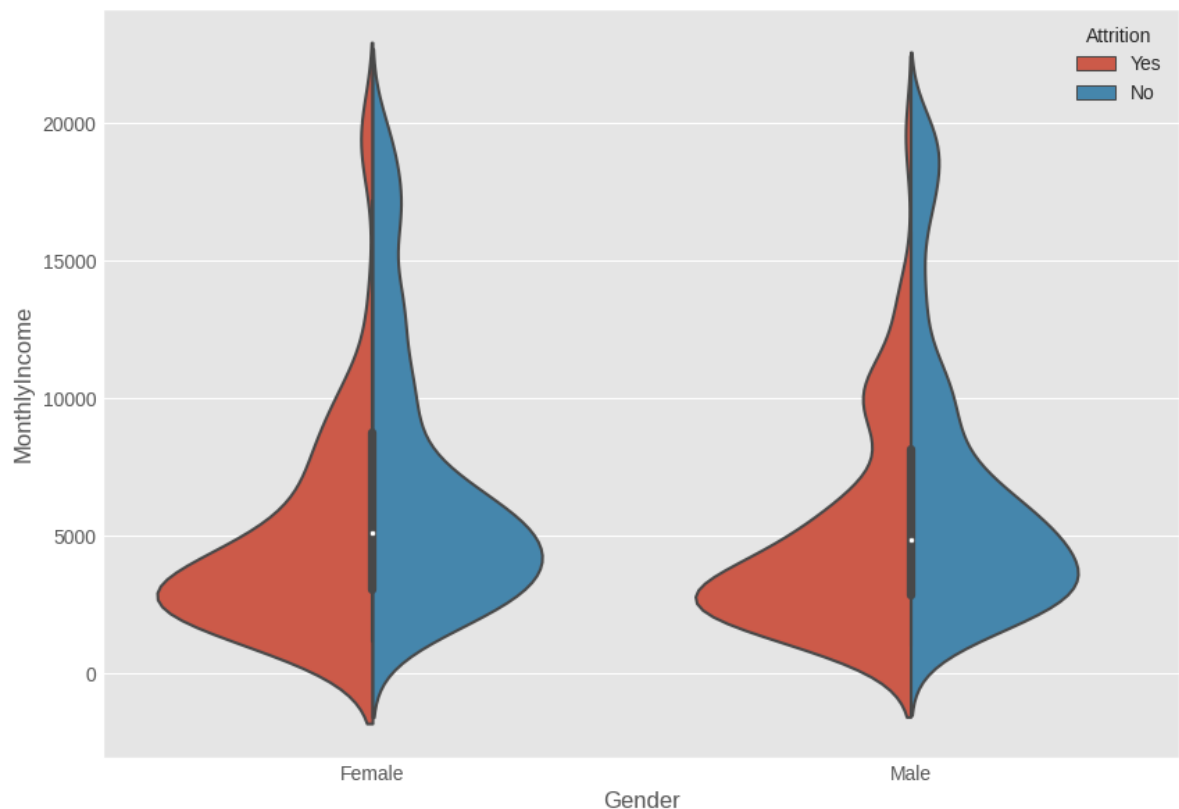
Most of the employees who got laid off had less working years(fresher and junior level) and they were receiving less salary

Very few people from senior category and receiving higher income got laid off

## Gender vs Monthly Income-Violin Plot

```
In [172]: fig,ax = plt.subplots(figsize=(10,7))
sns.violinplot(x='Gender', y='MonthlyIncome',hue='Attrition',split=True,data=dataset)

Out[172]: <Axes: xlabel='Gender', ylabel='MonthlyIncome'>
```



## Inference

Both the genders weren't discriminated in income and the attrition is at the same rate in both the cases

## Relationship with Age and Monthly income-Pair plot

```
In [173]: plt.style.use('ggplot')
g = sns.pairplot(dataset, vars=["MonthlyIncome", "Age"],hue="Attrition",size=5)

/usr/local/lib/python3.10/dist-packages/seaborn/axisgrid.py:2095: UserWarning: The
`size` parameter has been renamed to `height`; please update your code.
  warnings.warn(msg, UserWarning)
```



## Inference

The Income band widens for elder employees between 40-60 years of age and the income band restricts to 10000 for 20-30 year old employees

The employee who were laid off fall under low income(5000-1000) and young age(20-30) category

In [174...

```
dataset.describe(include="object").T #quick summary of categorical variables
```

Out[174]:

	count	unique	top	freq
Attrition	1470	2	No	1233
BusinessTravel	1470	3	Travel_Rarely	1043
Department	1470	3	Research & Development	961
EducationField	1470	6	Life Sciences	606
Gender	1470	2	Male	882
JobRole	1470	9	Sales Executive	326
MaritalStatus	1470	3	Married	673
Over18	1470	1	Y	1470
OverTime	1470	2	No	1054

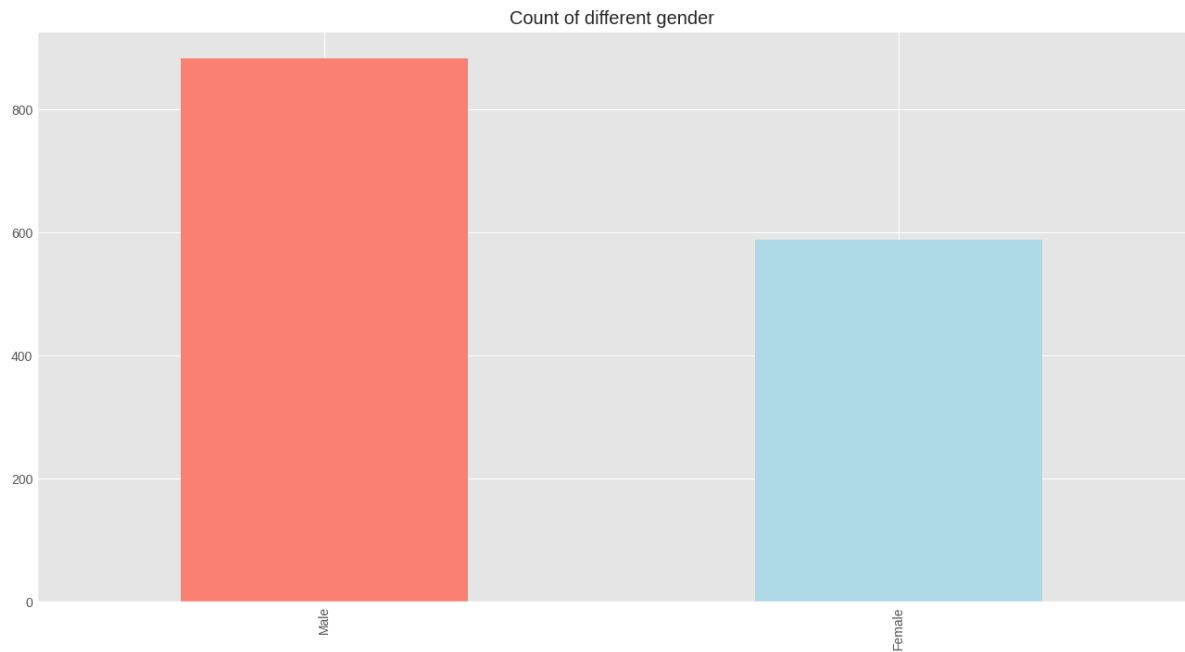
Gender count-Bar Plot

In [175...

```
dataset['Gender'].value_counts().plot(kind='bar',color=['salmon','lightblue'],title='Count of different gender')
```

Out[175]:

<Axes: title={'center': 'Count of different gender'}>



Inference

Shows the gender count.

5. Outlier Detector

In [176...

```
dataset.head()
```

Out[176]:

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

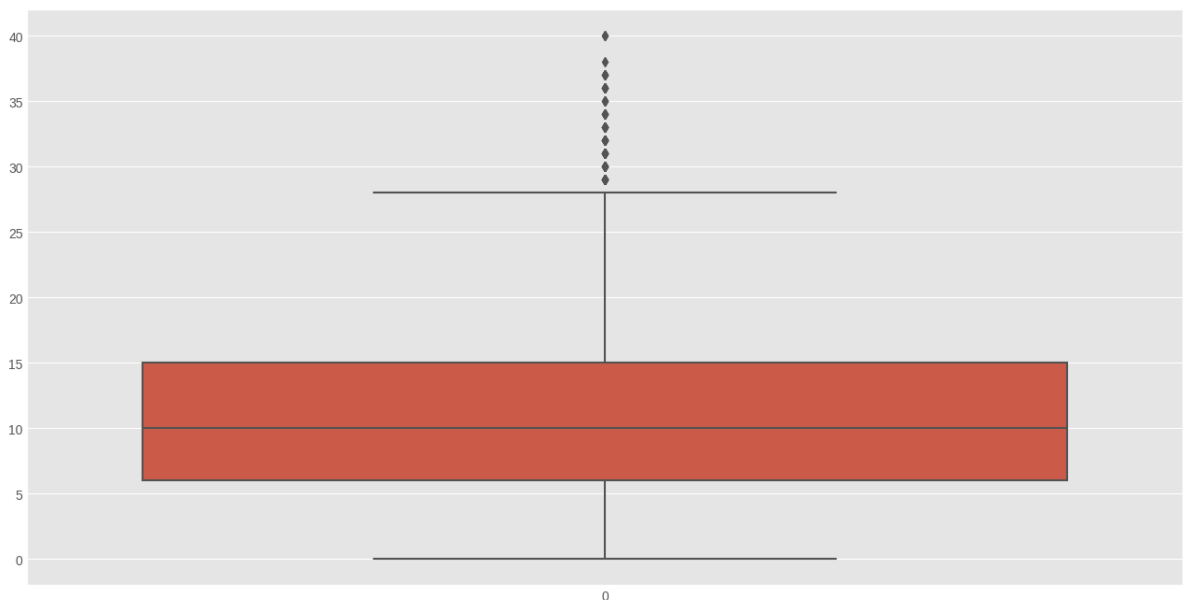
5 rows × 35 columns

In [177...

```
sns.boxplot(dataset["TotalWorkingYears"])
```

Out[177]:

&lt;Axes: &gt;



In [178...

```
q1=dataset.TotalWorkingYears.quantile(0.25)
q3=dataset.TotalWorkingYears.quantile(0.75)
print(q1)
print(q3)
IQR=q3-q1
IQR
upper_limit = q3+1.5*IQR
upper_limit
lower_limit = q1-1.5*IQR
lower_limit
dataset.median()
```

```
6.0
15.0
```

<ipython-input-178-7c0c8ed094ab>:11: FutureWarning: The default value of numeric\_only in DataFrame.median is deprecated. In a future version, it will default to False. In addition, specifying 'numeric\_only=None' is deprecated. Select only valid columns or specify the value of numeric\_only to silence this warning.

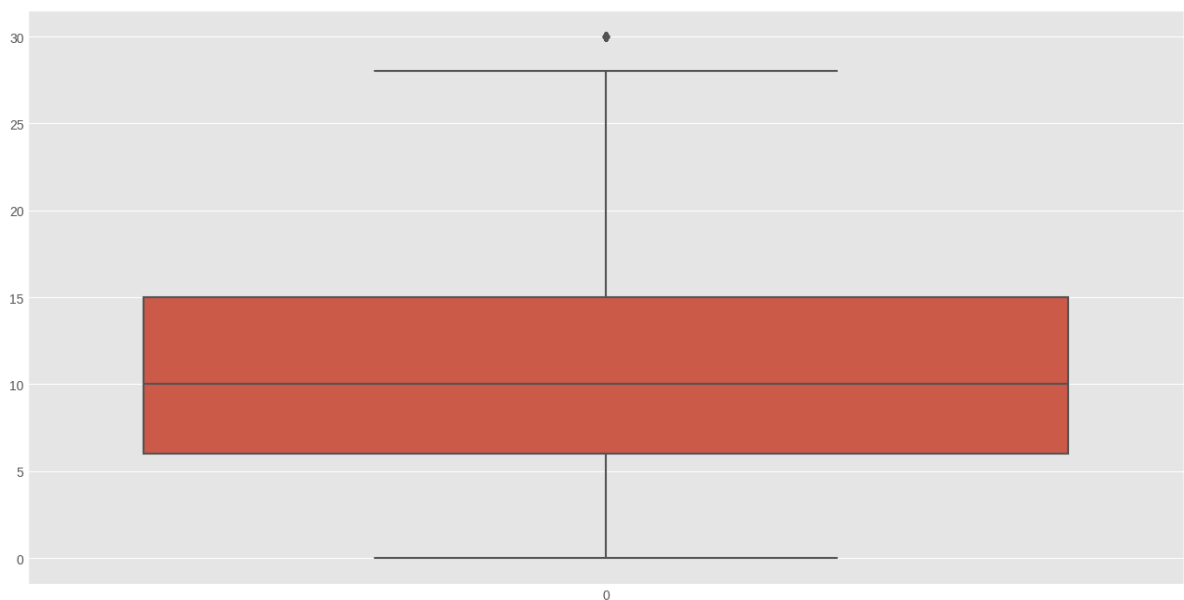
```
dataset.median()
```

```
Out[178]: Age                36.0
DailyRate              802.0
DistanceFromHome        7.0
Education               3.0
EmployeeCount           1.0
EmployeeNumber        1020.5
EnvironmentSatisfaction  3.0
HourlyRate              66.0
JobInvolvement          3.0
JobLevel                2.0
JobSatisfaction          3.0
MonthlyIncome          4919.0
MonthlyRate            14235.5
NumCompaniesWorked       2.0
PercentSalaryHike        14.0
PerformanceRating        3.0
RelationshipSatisfaction  3.0
StandardHours           80.0
StockOptionLevel         1.0
TotalWorkingYears        10.0
TrainingTimesLastYear     3.0
WorkLifeBalance          3.0
YearsAtCompany           5.0
YearsInCurrentRole        3.0
YearsSinceLastPromotion   1.0
YearsWithCurrManager      3.0
dtype: float64
```

```
In [179... dataset['TotalWorkingYears'] = np.where(dataset['TotalWorkingYears'] > upper_limit, 30,
```

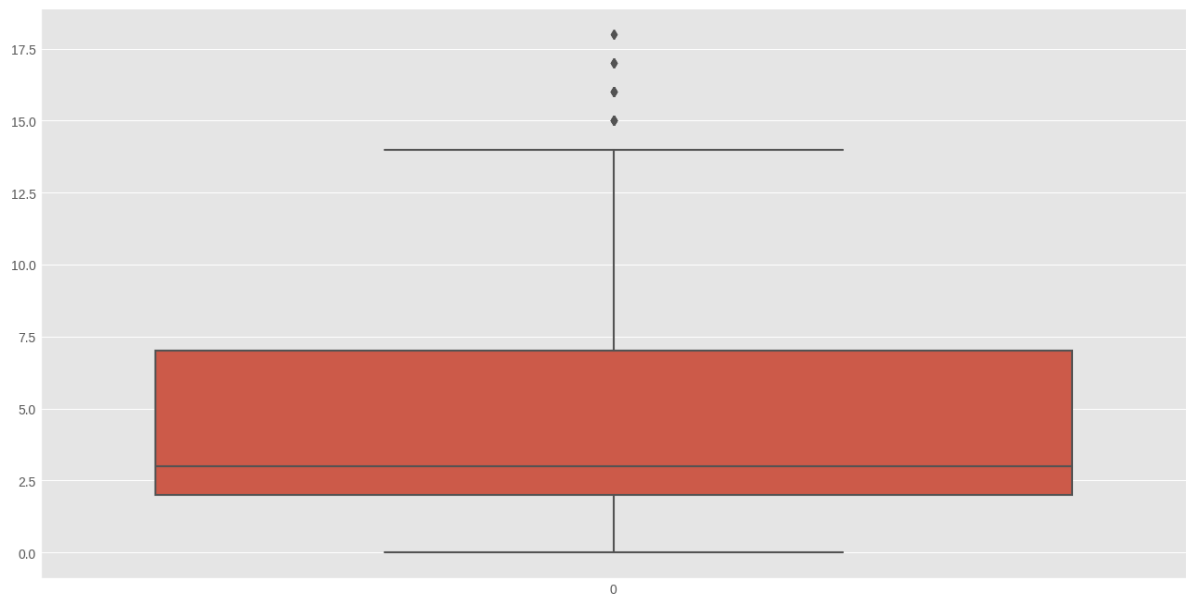
```
In [180... sns.boxplot(dataset["TotalWorkingYears"])
```

```
Out[180]: <Axes: >
```



```
In [181... sns.boxplot(dataset["YearsInCurrentRole"])
```

```
Out[181]: <Axes: >
```



In [182...

```

q1=dataset.YearsInCurrentRole.quantile(0.25)
q3=dataset.YearsInCurrentRole.quantile(0.75)
print(q1)
print(q3)
IQR=q3-q1
IQR
upper_limit = q3+1.5*IQR
upper_limit
lower_limit = q1-1.5*IQR
lower_limit
dataset.median()

```

2.0

7.0

<ipython-input-182-830ef09e96cd>:11: FutureWarning: The default value of numeric\_only in DataFrame.median is deprecated. In a future version, it will default to False. In addition, specifying 'numeric\_only=None' is deprecated. Select only valid columns or specify the value of numeric\_only to silence this warning.

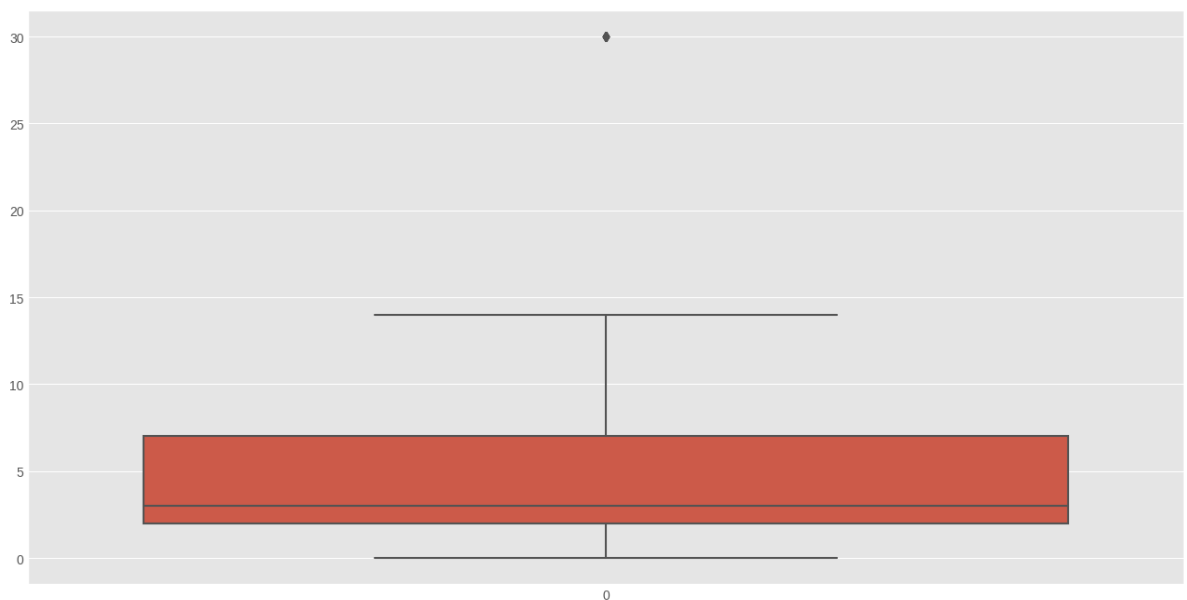
```
dataset.median()
```

```
Out[182]: Age 36.0
DailyRate 802.0
DistanceFromHome 7.0
Education 3.0
EmployeeCount 1.0
EmployeeNumber 1020.5
EnvironmentSatisfaction 3.0
HourlyRate 66.0
JobInvolvement 3.0
JobLevel 2.0
JobSatisfaction 3.0
MonthlyIncome 4919.0
MonthlyRate 14235.5
NumCompaniesWorked 2.0
PercentSalaryHike 14.0
PerformanceRating 3.0
RelationshipSatisfaction 3.0
StandardHours 80.0
StockOptionLevel 1.0
TotalWorkingYears 10.0
TrainingTimesLastYear 3.0
WorkLifeBalance 3.0
YearsAtCompany 5.0
YearsInCurrentRole 3.0
YearsSinceLastPromotion 1.0
YearsWithCurrManager 3.0
dtype: float64
```

```
In [183... dataset['YearsInCurrentRole'] = np.where(dataset['YearsInCurrentRole'] > upper_limit,
```

```
In [184... sns.boxplot(dataset["YearsInCurrentRole"])
```

```
Out[184]: <Axes: >
```



## 6. Seperate Dependent and Independent variables.

```
In [185... dataset.head()
```



Out[185]:

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

5 rows × 35 columns

In [186...]

```
y=dataset['Attrition']  
y
```

Out[186]:

```
0      Yes  
1      No  
2      Yes  
3      No  
4      No  
  
...  
1465   No  
1466   No  
1467   No  
1468   No  
1469   No  
Name: Attrition, Length: 1470, dtype: object
```

In [187...]

```
x=dataset.drop(labels='Attrition',axis=1)
```

In [188...]

```
x.head()
```

Out[188]:

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

5 rows × 34 columns

In [189...]

```
y
```

```
Out[189]: 0      Yes
          1      No
          2      Yes
          3      No
          4      No
          ...
          1465    No
          1466    No
          1467    No
          1468    No
          1469    No
          Name: Attrition, Length: 1470, dtype: object
```

## 6. Encoding

```
In [190... dataset.head()
```

```
Out[190]:
```

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

5 rows × 35 columns

```
In [191... s = (dataset.dtypes == 'object')
object_cols = list(s[s].index)

print("Categorical variables:")
print(object_cols)

Categorical variables:
['Attrition', 'BusinessTravel', 'Department', 'EducationField', 'Gender', 'JobRole', 'MaritalStatus', 'Over18', 'OverTime']
```

```
In [192... from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
```

```
In [193... x['Gender'].value_counts()
```

```
Out[193]: Male      882
          Female   588
          Name: Gender, dtype: int64
```

```
In [194... x["Gender"]=le.fit_transform(x["Gender"])
x["Gender"]
```

```
Out[194]: 0      0
          1      1
          2      1
          3      0
          4      1
          ..
          1465    1
          1466    1
          1467    1
          1468    1
          1469    1
          Name: Gender, Length: 1470, dtype: int64
```

```
In [195... x["Gender"].value_counts()
```

```
Out[195]: 1      882
          0      588
          Name: Gender, dtype: int64

          1- male

          2- female
```

```
In [196... x["Gender"].nunique()
```

```
Out[196]: 2
```

```
In [197... x.BusinessTravel.value_counts()
```

```
Out[197]: Travel_Rarely      1043
          Travel_Frequently    277
          Non-Travel          150
          Name: BusinessTravel, dtype: int64
```

```
In [198... x["BusinessTravel"]=le.fit_transform(x["BusinessTravel"])
          x["BusinessTravel"]
```

```
Out[198]: 0      2
          1      1
          2      2
          3      1
          4      2
          ..
          1465    1
          1466    2
          1467    2
          1468    1
          1469    2
          Name: BusinessTravel, Length: 1470, dtype: int64
```

```
In [199... x.BusinessTravel.value_counts()
```

```
Out[199]: 2      1043
          1      277
          0      150
          Name: BusinessTravel, dtype: int64

          0- Non-Travel

          1- Travel-frequently

          2- Travel-Rarely
```

```
In [200... x.Department.value_counts()
```

```
Out[200]: Research & Development    961  
Sales                        446  
Human Resources             63  
Name: Department, dtype: int64
```

```
In [201... x["Department"]=le.fit_transform(x["Department"])  
x["Department"]
```

```
Out[201]: 0      2  
1      1  
2      1  
3      1  
4      1  
..  
1465   1  
1466   1  
1467   1  
1468   2  
1469   1  
Name: Department, Length: 1470, dtype: int64
```

```
In [202... x.Department.value_counts()
```

```
Out[202]: 1    961  
2    446  
0     63  
Name: Department, dtype: int64
```

0- Human Resources

1- Sales

2- Research & Development

```
In [203... x['EducationField'].value_counts()
```

```
Out[203]: Life Sciences      606  
Medical      464  
Marketing     159  
Technical Degree  132  
Other         82  
Human Resources  27  
Name: EducationField, dtype: int64
```

```
In [204... x["EducationField"]=le.fit_transform(x["EducationField"])  
x["EducationField"]
```

```
Out[204]: 0      1  
1      1  
2      4  
3      1  
4      3  
..  
1465   3  
1466   3  
1467   1  
1468   3  
1469   3  
Name: EducationField, Length: 1470, dtype: int64
```

```
In [205... x['EducationField'].value_counts()
```

```
Out[205]: 1    606
          3    464
          2    159
          5    132
          4     82
          0     27
          Name: EducationField, dtype: int64
```

0-Human Resources

1-Life Sciences

2-Marketing

3-Medical

4-Other

5-Technical Degree

```
In [206... x.JobRole.value_counts()
```

```
Out[206]: Sales Executive      326
          Research Scientist  292
          Laboratory Technician 259
          Manufacturing Director 145
          Healthcare Representative 131
          Manager      102
          Sales Representative    83
          Research Director    80
          Human Resources    52
          Name: JobRole, dtype: int64
```

```
In [207... x["JobRole"]=le.fit_transform(x["JobRole"])
          x["JobRole"]
```

```
Out[207]: 0      7
          1      6
          2      2
          3      6
          4      2
          ..
          1465    2
          1466    0
          1467    4
          1468    7
          1469    2
          Name: JobRole, Length: 1470, dtype: int64
```

```
In [208... x.JobRole.value_counts()
```

```
Out[208]: 7      326
          6      292
          2      259
          4      145
          0      131
          3      102
          8       83
          5       80
          1       52
          Name: JobRole, dtype: int64
```

0-Healthcare Representative

1-Human Resource

2-Laboratory Technician

3-Manager

4-Manufacturing Director

5-Research Director

6-Research Scientist

7-Sales Executive

8-Sales Representative

In [209... `x.MaritalStatus.value_counts()`

Out[209]:

Married	673
Single	470
Divorced	327

Name: MaritalStatus, dtype: int64

In [210... `x["MaritalStatus"]=le.fit_transform(x["MaritalStatus"])`  
`x["MaritalStatus"]`

Out[210]:

0	2
1	1
2	2
3	1
4	1
..	
1465	1
1466	1
1467	1
1468	1
1469	1

Name: MaritalStatus, Length: 1470, dtype: int64

In [211... `x.MaritalStatus.value_counts()`

Out[211]:

1	673
2	470
0	327

Name: MaritalStatus, dtype: int64

0- Divorced

1- Married

2- Single

In [212... `x.Over18.value_counts()`

Out[212]:

Y	1470
---	------

Name: Over18, dtype: int64

In [213... `x["Over18"]=le.fit_transform(x["Over18"])`  
`x["Over18"]`

```
Out[213]: 0      0
          1      0
          2      0
          3      0
          4      0
          ..
          1465    0
          1466    0
          1467    0
          1468    0
          1469    0
          Name: Over18, Length: 1470, dtype: int64
```

```
In [214... x.Over18.value_counts()
```

```
Out[214]: 0      1470
          Name: Over18, dtype: int64

          0- Yes
```

```
In [215... x.OverTime.value_counts()
```

```
Out[215]: No      1054
          Yes      416
          Name: OverTime, dtype: int64
```

```
In [216... x["OverTime"]=le.fit_transform(x["OverTime"])
          x["OverTime"]
```

```
Out[216]: 0      1
          1      0
          2      1
          3      1
          4      0
          ..
          1465    0
          1466    0
          1467    1
          1468    0
          1469    0
          Name: OverTime, Length: 1470, dtype: int64
```

```
In [217... x.OverTime.value_counts()
```

```
Out[217]: 0      1054
          1      416
          Name: OverTime, dtype: int64

          0- No

          1- Yes
```

```
In [218... x.shape
```

```
Out[218]: (1470, 34)
```

```
In [219... x.head()
```

Out[219]:

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	Er
0	41	2	1102	2	1	2		1
1	49	1	279	1	8	1		1
2	37	2	1373	1	2	2		4
3	33	1	1392	1	3	4		1
4	27	2	591	1	2	1		3

5 rows × 34 columns

In [220... `y.head()`

Out[220]:

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

In [222... `y.value_counts()`

Out[222]:

```
No    1233
Yes    237
Name: Attrition, dtype: int64
```

In [224... `y=le.fit_transform(y)`  
`y`

Out[224]: `array([1, 0, 1, ..., 0, 0, 0])`

In [227... `#feature scaling`  
`from sklearn.preprocessing import MinMaxScaler`  
`ms=MinMaxScaler()`  
`x_scaled=pd.DataFrame(ms.fit_transform(x),columns=x.columns)`  
`x_scaled`



Out[227]:

	Age	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationI
<b>0</b>	0.547619	1.0	0.715820	1.0	0.000000	0.25	
<b>1</b>	0.738095	0.5	0.126700	0.5	0.250000	0.00	
<b>2</b>	0.452381	1.0	0.909807	0.5	0.035714	0.25	
<b>3</b>	0.357143	0.5	0.923407	0.5	0.071429	0.75	
<b>4</b>	0.214286	1.0	0.350036	0.5	0.035714	0.00	
...	...	...	...	...	...	...	
<b>1465</b>	0.428571	0.5	0.559771	0.5	0.785714	0.25	
<b>1466</b>	0.500000	1.0	0.365784	0.5	0.178571	0.00	
<b>1467</b>	0.214286	1.0	0.037938	0.5	0.107143	0.50	
<b>1468</b>	0.738095	0.5	0.659270	1.0	0.035714	0.50	
<b>1469</b>	0.380952	1.0	0.376521	0.5	0.250000	0.50	

1470 rows × 34 columns

In [228...]

```
#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_stat
```

In [229...]

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

Out[229]:

```
((1176, 34), (294, 34), (1176,), (294,))
```

## Model Building

- o Import the model building Libraries

- o Initializing the model

- o Training and testing the model

- o Evaluation of Model

- o Save the Model

## Logistic Regression

In [230...]

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

In [231...]

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

Out[231]:

```
▼ LogisticRegression
LogisticRegression()
```

```
In [232... pred=model.predict(x_test)
pred
```

[illegible]

```
In [233... y_test
```

```
Out[233]: array([[0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1,
        1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1,
        0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0,
        0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0,
        1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0,
        0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
        0, 1, 0, 0, 0, 1, 0, 0])
```

```
In [234... #evaluation of classification model
#Accuracy score
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report,
```

```
In [235... accuracy_score(y_test, pred)
```

```
Out[235]: 0.8809523809523809
```

```
In [236... confusion_matrix(y_test, pred)
```

```
Out[236]: array([[242,  3],
                  [ 32, 17]])
```

```
In [237... pd.crosstab(y_test, pred)
```

```
Out[237]:
```

col_0	0	1
row_0		
0	242	3
1	32	17

```
In [238... (242+17)/(242+17+32+3) #accuracy score
```

```
Out[238]: 0.8809523809523809
```

In [239...

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

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

In [240...

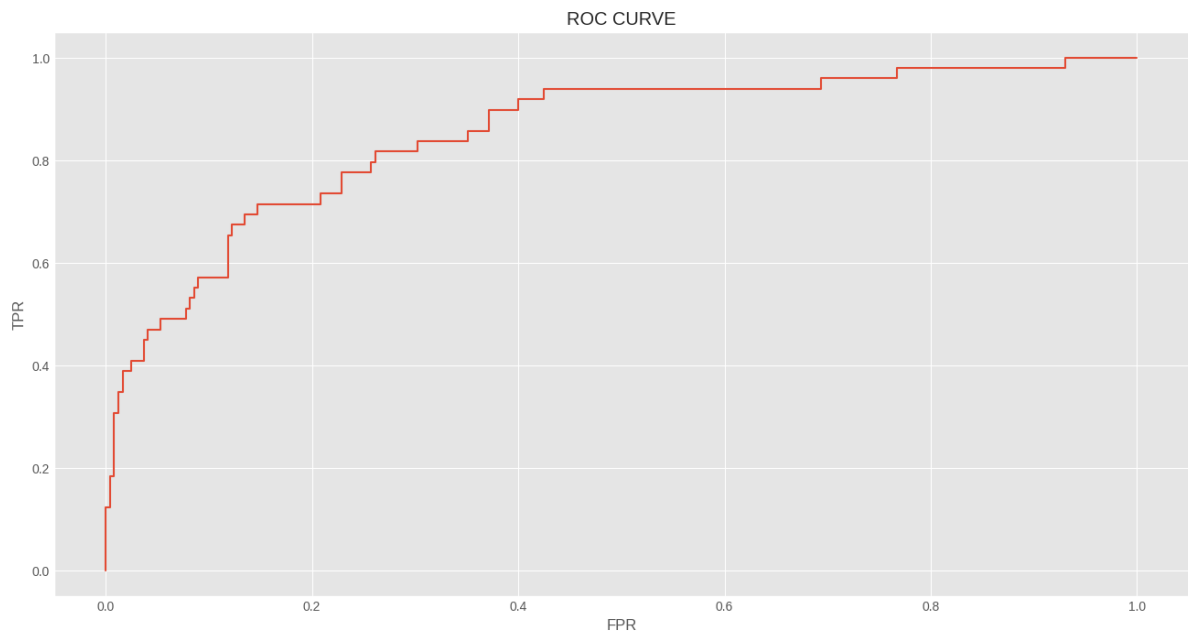
```
#Roc-AUC Curve  
probability=model.predict_proba(x_test)[:,1]  
probability
```

```
Out[240]: array([0.15217402, 0.21632349, 0.31448937, 0.11552231, 0.63195007,
0.06110795, 0.61205171, 0.08073788, 0.00740246, 0.42236585,
0.05133183, 0.32469008, 0.02237374, 0.6920699 , 0.18976212,
0.02894311, 0.11110158, 0.17159697, 0.05136821, 0.21685107,
0.23350228, 0.01976108, 0.07290817, 0.05153163, 0.60410648,
0.49152701, 0.08029735, 0.05034051, 0.67674278, 0.05779457,
0.01808775, 0.03116965, 0.05629576, 0.17137281, 0.07754204,
0.04011103, 0.08277844, 0.066222 , 0.04473409, 0.05074883,
0.0453298 , 0.02518108, 0.01390803, 0.01430808, 0.02992714,
0.47717853, 0.42463578, 0.00342564, 0.73804972, 0.48329632,
0.08750612, 0.50973356, 0.08356101, 0.26756878, 0.63342172,
0.22261944, 0.01664791, 0.27789879, 0.02738251, 0.1516253 ,
0.02255381, 0.21819211, 0.11921312, 0.03932104, 0.32671669,
0.03013093, 0.32903292, 0.15813444, 0.09364017, 0.09358887,
0.0910726 , 0.31432242, 0.09789755, 0.07483434, 0.12863802,
0.05837452, 0.0493793 , 0.06218263, 0.19638662, 0.03325248,
0.00761511, 0.02692652, 0.14003752, 0.02550772, 0.03405416,
0.08306553, 0.00372029, 0.03841149, 0.0385455 , 0.15582131,
0.25915316, 0.16446146, 0.2605143 , 0.2170274 , 0.02308859,
0.17411709, 0.34105565, 0.2836754 , 0.06979948, 0.05412718,
0.23058355, 0.71467057, 0.29623588, 0.01981856, 0.09155609,
0.03062145, 0.06306983, 0.14829133, 0.07471827, 0.131319 ,
0.09039381, 0.04550021, 0.0234031 , 0.14007801, 0.07915213,
0.03325378, 0.05904139, 0.10244746, 0.00769902, 0.01356191,
0.1899012 , 0.04854069, 0.09989016, 0.82471562, 0.02395718,
0.02480647, 0.00740718, 0.13995083, 0.1676497 , 0.06179059,
0.0153163 , 0.27844187, 0.53392829, 0.36999997, 0.04365391,
0.36760733, 0.62429365, 0.1278725 , 0.07357853, 0.20034431,
0.08198194, 0.07412797, 0.11514013, 0.22048825, 0.21971106,
0.03120375, 0.13280402, 0.00281934, 0.11143174, 0.16183701,
0.0632159 , 0.17069567, 0.05761521, 0.12594416, 0.03701899,
0.03134031, 0.05938977, 0.0852415 , 0.01621121, 0.01882977,
0.3645953 , 0.01470022, 0.1478167 , 0.80600097, 0.11850163,
0.28135372, 0.19690363, 0.15121366, 0.02658214, 0.00620571,
0.03477834, 0.11262055, 0.10009029, 0.09904116, 0.01267428,
0.13933536, 0.10267279, 0.13081418, 0.05184247, 0.07418267,
0.02732516, 0.10911227, 0.00416416, 0.75017934, 0.04437636,
0.04324425, 0.3940623 , 0.05143664, 0.71301171, 0.09775088,
0.37841904, 0.37731749, 0.34241499, 0.04783579, 0.08374629,
0.15060733, 0.047484 , 0.01525795, 0.27533347, 0.06190423,
0.15515891, 0.15822481, 0.66396633, 0.06007485, 0.30507983,
0.05444403, 0.45863441, 0.00260234, 0.14983251, 0.03055447,
0.11599359, 0.16925832, 0.08265742, 0.09971406, 0.15622642,
0.02095842, 0.01859019, 0.08868262, 0.02462549, 0.13927623,
0.09226204, 0.21627391, 0.72737143, 0.15794236, 0.41508798,
0.01633093, 0.09689161, 0.21311155, 0.31087529, 0.03492884,
0.04588433, 0.32368325, 0.05863244, 0.02807638, 0.17020541,
0.28403941, 0.24319541, 0.00853641, 0.08857797, 0.01190675,
0.14307901, 0.28821893, 0.01166466, 0.18415389, 0.04799148,
0.04083938, 0.40349113, 0.34270419, 0.03801019, 0.13633343,
0.3579905 , 0.34773119, 0.79705153, 0.04762894, 0.22929036,
0.05777684, 0.00532319, 0.65735633, 0.3691595 , 0.38214768,
0.35711771, 0.03375219, 0.22133642, 0.0614842 , 0.07525809,
0.09793662, 0.00859601, 0.26368491, 0.39075382, 0.06385042,
0.10236432, 0.01395833, 0.14225579, 0.05150353, 0.0246592 ,
0.03247425, 0.07929253, 0.25427181, 0.28548567, 0.19978486,
0.26904127, 0.01644442, 0.15255598, 0.07893817, 0.02843614,
0.19025185, 0.00800355, 0.25237563, 0.00220511, 0.02847355,
0.24199687, 0.73687967, 0.06887902, 0.22136975])
```

```
In [241... #roc_curve
fpr,tpr,thresholds=roc_curve(y_test,probability)
```

In [242...

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



## Decision Tree

In [243...

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

In [244...

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

Out[244]:

```
▼ DecisionTreeClassifier
DecisionTreeClassifier()
```

In [246...

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

Out[246]:

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

In [247...

```
y_test
```

```
Out[247]: array([0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
        0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
        1, 1, 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, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0,
        1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
        0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0,
        1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 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, 1, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0,
        0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
        0, 1, 0, 0, 0, 1, 0, 0])
```

```
In [248... #evaluation of classification model
#Accuracy score
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report,
```

```
In [249... accuracy_score(y_test, pred)
```

```
Out[249]: 0.7619047619047619
```

```
In [250... confusion_matrix(y_test, pred)
```

```
Out[250]: array([[208, 37],
        [ 33, 16]])
```

```
In [251... pd.crosstab(y_test, pred)
```

```
Out[251]: col_0    0    1
row_0
0    208   37
1     33   16
```

```
In [252... #Roc-AUC Curve
probability=dtc.predict_proba(x_test)[: ,1]
probability
```

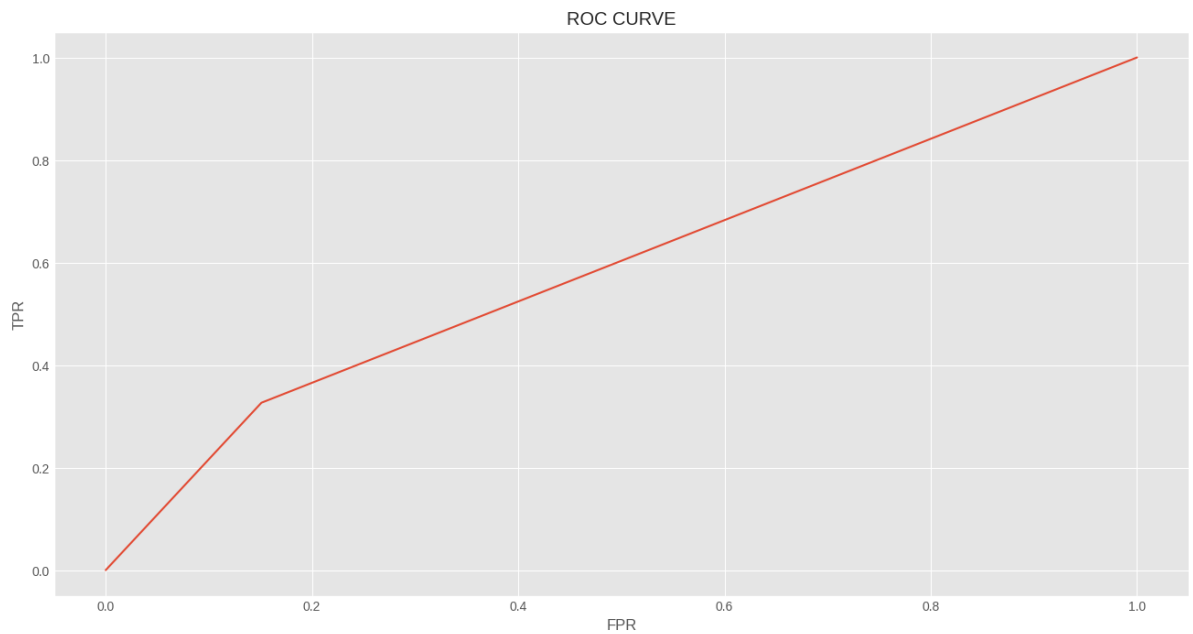
```
Out[252]: array([0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 1., 0., 0., 0.,
        0., 0., 1., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 1., 1.,
        1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 1., 0., 0., 0.,
        0., 0., 0., 0., 0., 1., 0., 0., 1., 0., 0., 0., 0., 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., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 1., 1., 1., 1., 0.,
        0., 0., 1., 0., 0., 0., 0., 0., 1., 0., 0., 1., 0., 0., 0., 0.,
        0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 1., 0., 1., 1., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0.,
        1., 0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 1., 1., 0.,
        0., 1., 0., 0., 0., 0., 1., 1., 0., 0., 1., 0., 0., 0., 0., 0.,
        0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 1., 0., 0., 0., 0., 0., 0., 0., 1., 0., 1., 0., 0., 0.,
        0., 0., 0., 0., 0., 0., 0., 0., 1., 0., 0., 1., 1., 0., 0., 1.,
        0., 0., 0., 0., 0.])
```

```
In [253... #roc_curve
```

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

In [254...

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



### Pre Parameter Tuning

In [256...

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

```

Out[256]: [Text(0.32785292288557216, 0.9722222222222222, 'x[27] <= 0.05\ngini = 0.269\nsamples = 1176\nvalue = [988, 188]'),
Text(0.07960199004975124, 0.9166666666666666, 'x[16] <= 0.75\ngini = 0.5\nsamples = 78\nvalue = [39, 39]'),
Text(0.04975124378109453, 0.8611111111111112, 'x[4] <= 0.554\ngini = 0.426\nsamples = 39\nvalue = [27, 12]'),
Text(0.03316749585406302, 0.8055555555555556, 'x[15] <= 0.167\ngini = 0.312\nsamples = 31\nvalue = [25, 6]'),
Text(0.01990049751243781, 0.75, 'x[21] <= 0.5\ngini = 0.49\nsamples = 7\nvalue = [3, 4]'),
Text(0.013266998341625208, 0.6944444444444444, 'x[10] <= 0.5\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.006633499170812604, 0.6388888888888888, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.01990049751243781, 0.6388888888888888, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.026533996683250415, 0.6944444444444444, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.04643449419568822, 0.75, 'x[19] <= 0.056\ngini = 0.153\nsamples = 24\nvalue = [22, 2]'),
Text(0.03980099502487562, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.05306799336650083, 0.6944444444444444, 'x[9] <= 0.167\ngini = 0.083\nsamples = 23\nvalue = [22, 1]'),
Text(0.04643449419568822, 0.6388888888888888, 'x[23] <= 0.5\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.03980099502487562, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.05306799336650083, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.05970149253731343, 0.6388888888888888, 'gini = 0.0\nsamples = 21\nvalue = [21, 0]'),
Text(0.06633499170812604, 0.8055555555555556, 'x[8] <= 0.385\ngini = 0.375\nsamples = 8\nvalue = [2, 6]'),
Text(0.05970149253731343, 0.75, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.07296849087893864, 0.75, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.10945273631840796, 0.8611111111111112, 'x[11] <= 0.364\ngini = 0.426\nsamples = 39\nvalue = [12, 27]'),
Text(0.09286898839137644, 0.8055555555555556, 'x[29] <= 0.167\ngini = 0.133\nsamples = 14\nvalue = [1, 13]'),
Text(0.08623548922056384, 0.75, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.09950248756218906, 0.75, 'gini = 0.0\nsamples = 13\nvalue = [0, 13]'),
Text(0.12603648424543948, 0.8055555555555556, 'x[8] <= 0.105\ngini = 0.493\nsamples = 25\nvalue = [11, 14]'),
Text(0.11276948590381426, 0.75, 'x[22] <= 0.464\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
Text(0.10613598673300166, 0.6944444444444444, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.11940298507462686, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.13930348258706468, 0.75, 'x[15] <= 0.5\ngini = 0.432\nsamples = 19\nvalue = [6, 13]'),
Text(0.13266998341625208, 0.6944444444444444, 'gini = 0.0\nsamples = 7\nvalue = [0, 7]'),
Text(0.14593698175787728, 0.6944444444444444, 'x[6] <= 0.4\ngini = 0.5\nsamples = 12\nvalue = [6, 6]'),
Text(0.13266998341625208, 0.6388888888888888, 'x[3] <= 0.75\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
Text(0.12603648424543948, 0.5833333333333334, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.13930348258706468, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.15920398009950248, 0.6388888888888888, 'x[8] <= 0.249\ngini = 0.278\nsamples = 6\nvalue = [1, 5]'),

```



```
Text(0.15257048092868988, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.16583747927031509, 0.5833333333333334, 'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
Text(0.576103855721393, 0.9166666666666666, 'x[21] <= 0.5\ngini = 0.235\nsamples = 1098\nvalue = [949, 149]'),
Text(0.3279954394693201, 0.8611111111111112, 'x[29] <= 0.167\ngini = 0.162\nsamples = 798\nvalue = [727, 71]'),
Text(0.1791044776119403, 0.8055555555555556, 'x[8] <= 0.445\ngini = 0.38\nsamples = 47\nvalue = [35, 12]'),
Text(0.16583747927031509, 0.75, 'x[16] <= 0.75\ngini = 0.1\nsamples = 19\nvalue = [18, 1]'),
Text(0.15920398009950248, 0.6944444444444444, 'gini = 0.0\nsamples = 18\nvalue = [18, 0]'),
Text(0.1724709784411277, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.19237147595356552, 0.75, 'x[17] <= 0.094\ngini = 0.477\nsamples = 28\nvalue = [17, 11]'),
Text(0.1857379767827529, 0.6944444444444444, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),
Text(0.19900497512437812, 0.6944444444444444, 'x[8] <= 0.524\ngini = 0.413\nsamples = 24\nvalue = [17, 7]'),
Text(0.19237147595356552, 0.6388888888888888, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.20563847429519072, 0.6388888888888888, 'x[33] <= 0.324\ngini = 0.351\nsamples = 22\nvalue = [17, 5]'),
Text(0.19237147595356552, 0.5833333333333334, 'x[2] <= 0.025\ngini = 0.133\nsamples = 14\nvalue = [13, 1]'),
Text(0.1857379767827529, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.19900497512437812, 0.5277777777777778, 'gini = 0.0\nsamples = 13\nvalue = [13, 0]'),
Text(0.21890547263681592, 0.5833333333333334, 'x[2] <= 0.329\ngini = 0.5\nsamples = 8\nvalue = [4, 4]'),
Text(0.21227197346600332, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.22553897180762852, 0.5277777777777778, 'x[12] <= 0.333\ngini = 0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.21890547263681592, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.23217247097844113, 0.4722222222222222, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.47688640132669985, 0.8055555555555556, 'x[30] <= 0.963\ngini = 0.145\nsamples = 751\nvalue = [692, 59]'),
Text(0.4702529021558872, 0.75, 'x[30] <= 0.113\ngini = 0.143\nsamples = 750\nvalue = [692, 58]'),
Text(0.34618573797678276, 0.6944444444444444, 'x[9] <= 0.167\ngini = 0.218\nsamples = 257\nvalue = [225, 32]'),
Text(0.3034825870646766, 0.6388888888888888, 'x[33] <= 0.147\ngini = 0.355\nsamples = 65\nvalue = [50, 15]'),
Text(0.28192371475953565, 0.5833333333333334, 'x[33] <= 0.029\ngini = 0.303\nsamples = 59\nvalue = [48, 11]'),
Text(0.25870646766169153, 0.5277777777777778, 'x[12] <= 0.5\ngini = 0.463\nsamples = 22\nvalue = [14, 8]'),
Text(0.24543946932006633, 0.4722222222222222, 'x[11] <= 0.179\ngini = 0.198\nsamples = 9\nvalue = [8, 1]'),
Text(0.23880597014925373, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.25207296849087896, 0.4166666666666667, 'gini = 0.0\nsamples = 8\nvalue = [8, 0]'),
Text(0.27197346600331673, 0.4722222222222222, 'x[11] <= 0.4\ngini = 0.497\nsamples = 13\nvalue = [6, 7]'),
Text(0.26533996683250416, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
```

```
Text(0.27860696517412936, 0.4166666666666667, 'x[4] <= 0.286\ngini = 0.346\nsamples = 9\nvalue = [2, 7]'),
Text(0.27197346600331673, 0.3611111111111111, 'x[19] <= 0.944\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.26533996683250416, 0.3055555555555556, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.27860696517412936, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.28524046434494194, 0.3611111111111111, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.30514096185737977, 0.5277777777777778, 'x[15] <= 0.167\ngini = 0.149\nsamples = 37\nvalue = [34, 3]'),
Text(0.29850746268656714, 0.4722222222222222, 'x[29] <= 0.5\ngini = 0.5\nsamples = 6\nvalue = [3, 3]'),
Text(0.29187396351575456, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.30514096185737977, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.3117744610281924, 0.4722222222222222, 'gini = 0.0\nsamples = 31\nvalue = [31, 0]'),
Text(0.3250414593698176, 0.5833333333333334, 'x[8] <= 0.065\ngini = 0.444\nsamples = 6\nvalue = [2, 4]'),
Text(0.31840796019900497, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.33167495854063017, 0.5277777777777778, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),
Text(0.3888888888888889, 0.6388888888888888, 'x[0] <= 0.321\ngini = 0.161\nsamples = 192\nvalue = [175, 17]'),
Text(0.351575456053068, 0.5833333333333334, 'x[6] <= 0.1\ngini = 0.294\nsamples = 67\nvalue = [55, 12]'),
Text(0.3449419568822554, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.3582089552238806, 0.5277777777777778, 'x[29] <= 0.5\ngini = 0.26\nsamples = 65\nvalue = [55, 10]'),
Text(0.33499170812603646, 0.4722222222222222, 'x[6] <= 0.5\ngini = 0.469\nsamples = 16\nvalue = [10, 6]'),
Text(0.3283582089552239, 0.4166666666666667, 'gini = 0.0\nsamples = 7\nvalue = [7, 0]'),
Text(0.3416252072968491, 0.4166666666666667, 'x[9] <= 0.833\ngini = 0.444\nsamples = 9\nvalue = [3, 6]'),
Text(0.33499170812603646, 0.3611111111111111, 'gini = 0.0\nsamples = 5\nvalue = [0, 5]'),
Text(0.3482587064676617, 0.3611111111111111, 'x[18] <= 0.072\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.3416252072968491, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.3548922056384743, 0.3055555555555556, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.3814262023217247, 0.4722222222222222, 'x[2] <= 0.037\ngini = 0.15\nsamples = 49\nvalue = [45, 4]'),
Text(0.3747927031509121, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.3880597014925373, 0.4166666666666667, 'x[2] <= 0.938\ngini = 0.117\nsamples = 48\nvalue = [45, 3]'),
Text(0.3814262023217247, 0.3611111111111111, 'x[5] <= 0.875\ngini = 0.081\nsamples = 47\nvalue = [45, 2]'),
Text(0.3681592039800995, 0.3055555555555556, 'x[12] <= 0.167\ngini = 0.043\nsamples = 45\nvalue = [44, 1]'),
Text(0.3615257048092869, 0.25, 'x[8] <= 0.839\ngini = 0.444\nsamples = 3\nvalue = [2, 1]'),
Text(0.3548922056384743, 0.19444444444444445, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.3681592039800995, 0.19444444444444445, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
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Text(0.3747927031509121, 0.25, 'gini = 0.0\nsamples = 42\nvalue = [42, 0]'),
Text(0.3946932006633499, 0.3055555555555556, 'x[8] <= 0.246\ngini = 0.5\nsamples
= 2\nvalue = [1, 1]'),
Text(0.3880597014925373, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.4013266998341625, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.3946932006633499, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.4262023217247098, 0.5833333333333334, 'x[8] <= 0.022\ngini = 0.077\nsample
s = 125\nvalue = [120, 5]'),
Text(0.4079601990049751, 0.5277777777777778, 'x[27] <= 0.25\ngini = 0.5\nsamples
= 4\nvalue = [2, 2]'),
Text(0.4013266998341625, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.41459369817578773, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.4444444444444444, 0.5277777777777778, 'x[18] <= 0.968\ngini = 0.048\nsampl
es = 121\nvalue = [118, 3]'),
Text(0.42786069651741293, 0.4722222222222222, 'x[2] <= 0.98\ngini = 0.033\nsample
s = 118\nvalue = [116, 2]'),
Text(0.41459369817578773, 0.4166666666666667, 'x[14] <= 0.938\ngini = 0.017\nsamp
les = 114\nvalue = [113, 1]'),
Text(0.4079601990049751, 0.3611111111111111, 'gini = 0.0\nsamples = 107\nvalue =
[107, 0]'),
Text(0.42122719734660036, 0.3611111111111111, 'x[12] <= 0.167\ngini = 0.245\nsamp
les = 7\nvalue = [6, 1]'),
Text(0.41459369817578773, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.42786069651741293, 0.3055555555555556, 'gini = 0.0\nsamples = 6\nvalue =
[6, 0]'),
Text(0.44112769485903813, 0.4166666666666667, 'x[3] <= 0.75\ngini = 0.375\nsample
s = 4\nvalue = [3, 1]'),
Text(0.43449419568822556, 0.3611111111111111, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),
Text(0.44776119402985076, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.46102819237147596, 0.4722222222222222, 'x[19] <= 0.278\ngini = 0.444\nsamp
les = 3\nvalue = [2, 1]'),
Text(0.45439469320066334, 0.4166666666666667, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.46766169154228854, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.5943200663349917, 0.6944444444444444, 'x[30] <= 0.787\ngini = 0.1\nsamples
= 493\nvalue = [467, 26]'),
Text(0.5601160862354893, 0.6388888888888888, 'x[15] <= 0.5\ngini = 0.094\nsamples
= 486\nvalue = [462, 24]'),
Text(0.511608623548922, 0.5833333333333334, 'x[14] <= 0.938\ngini = 0.154\nsample
s = 191\nvalue = [175, 16]'),
Text(0.5049751243781094, 0.5277777777777778, 'x[18] <= 0.481\ngini = 0.145\nsampl
es = 190\nvalue = [175, 15]'),
Text(0.48756218905472637, 0.4722222222222222, 'x[18] <= 0.47\ngini = 0.221\nsampl
es = 95\nvalue = [83, 12]'),
Text(0.48092868988391374, 0.4166666666666667, 'x[33] <= 0.794\ngini = 0.207\nsamp
les = 94\nvalue = [83, 11]'),
Text(0.47429519071310117, 0.3611111111111111, 'x[5] <= 0.375\ngini = 0.192\nsampl
es = 93\nvalue = [83, 10]'),
Text(0.4527363184079602, 0.3055555555555556, 'x[6] <= 0.9\ngini = 0.363\nsamples
= 21\nvalue = [16, 5]'),
Text(0.4461028192371476, 0.25, 'x[17] <= 0.413\ngini = 0.266\nsamples = 19\nvalue
= [16, 3]'),
Text(0.43283582089552236, 0.19444444444444445, 'x[8] <= 0.215\ngini = 0.117\nsamp
les = 16\nvalue = [15, 1]'),
Text(0.4262023217247098, 0.1388888888888889, 'x[0] <= 0.369\ngini = 0.5\nsamples
= 2\nvalue = [1, 1]'),
Text(0.41956882255389716, 0.08333333333333333, 'gini = 0.0\nsamples = 1\nvalue =

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[0, 1]'),
Text(0.43283582089552236, 0.08333333333333333, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.439469320066335, 0.13888888888888889, 'gini = 0.0\nsamples = 14\nvalue = [1
4, 0]'),
Text(0.4593698175787728, 0.19444444444444445, 'x[24] <= 0.833\ngini = 0.444\nsamp
les = 3\nvalue = [1, 2]'),
Text(0.4527363184079602, 0.13888888888888889, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.4660033167495854, 0.13888888888888889, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.4593698175787728, 0.25, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.49585406301824214, 0.30555555555555556, 'x[31] <= 0.083\ngini = 0.129\nsamp
les = 72\nvalue = [67, 5]'),
Text(0.4792703150912106, 0.25, 'x[8] <= 0.68\ngini = 0.444\nsamples = 6\nvalue =
[4, 2]'),
Text(0.472636815920398, 0.19444444444444445, 'gini = 0.0\nsamples = 4\nvalue =
[4, 0]'),
Text(0.4859038142620232, 0.19444444444444445, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.5124378109452736, 0.25, 'x[11] <= 0.993\ngini = 0.087\nsamples = 66\nvalue
= [63, 3]'),
Text(0.49917081260364843, 0.19444444444444445, 'x[28] <= 0.583\ngini = 0.061\nsam
ples = 64\nvalue = [62, 2]'),
Text(0.4925373134328358, 0.13888888888888889, 'gini = 0.0\nsamples = 51\nvalue =
[51, 0]'),
Text(0.5058043117744611, 0.13888888888888889, 'x[9] <= 0.5\ngini = 0.26\nsamples =
13\nvalue = [11, 2]'),
Text(0.49917081260364843, 0.08333333333333333, 'x[17] <= 0.335\ngini = 0.5\nsampl
es = 4\nvalue = [2, 2]'),
Text(0.4925373134328358, 0.027777777777777776, 'gini = 0.0\nsamples = 2\nvalue =
[2, 0]'),
Text(0.5058043117744611, 0.027777777777777776, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.5124378109452736, 0.08333333333333333, 'gini = 0.0\nsamples = 9\nvalue =
[9, 0]'),
Text(0.5257048092868989, 0.19444444444444445, 'x[1] <= 0.25\ngini = 0.5\nsamples
= 2\nvalue = [1, 1]'),
Text(0.5190713101160862, 0.13888888888888889, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.5323383084577115, 0.13888888888888889, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.48756218905472637, 0.36111111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.494195688225539, 0.41666666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0,
1]'),
Text(0.5223880597014925, 0.47222222222222222, 'x[19] <= 0.5\ngini = 0.061\nsamples
= 95\nvalue = [92, 3]'),
Text(0.5157545605306799, 0.41666666666666667, 'gini = 0.0\nsamples = 76\nvalue =
[76, 0]'),
Text(0.5290215588723052, 0.41666666666666667, 'x[33] <= 0.088\ngini = 0.266\nsampl
es = 19\nvalue = [16, 3]'),
Text(0.5157545605306799, 0.36111111111111111, 'x[3] <= 0.75\ngini = 0.444\nsamples
= 3\nvalue = [1, 2]'),
Text(0.5091210613598673, 0.30555555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.5223880597014925, 0.30555555555555556, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.5422885572139303, 0.36111111111111111, 'x[17] <= 0.108\ngini = 0.117\nsampl
es = 16\nvalue = [15, 1]'),
Text(0.5356550580431177, 0.30555555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.548922056384743, 0.30555555555555556, 'gini = 0.0\nsamples = 15\nvalue = [1
5, 0]'),

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Text(0.5182421227197347, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6086235489220564, 0.5833333333333334, 'x[22] <= 0.036\ngini = 0.053\nsamples = 295\nvalue = [287, 8]'),
Text(0.5854063018242123, 0.5277777777777778, 'x[32] <= 0.7\ngini = 0.159\nsamples = 46\nvalue = [42, 4]'),
Text(0.5787728026533997, 0.4722222222222222, 'x[12] <= 0.167\ngini = 0.124\nsamples = 45\nvalue = [42, 3]'),
Text(0.5621890547263682, 0.4166666666666667, 'x[27] <= 0.4\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.5555555555555556, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.5688225538971807, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.5953565505804311, 0.4166666666666667, 'x[27] <= 0.917\ngini = 0.089\nsamples = 43\nvalue = [41, 2]'),
Text(0.582089552238806, 0.3611111111111111, 'x[14] <= 0.062\ngini = 0.048\nsamples = 41\nvalue = [40, 1]'),
Text(0.5754560530679934, 0.3055555555555556, 'x[9] <= 0.167\ngini = 0.375\nsamples = 4\nvalue = [3, 1]'),
Text(0.5688225538971807, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.582089552238806, 0.25, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.5887230514096186, 0.3055555555555556, 'gini = 0.0\nsamples = 37\nvalue = [37, 0]'),
Text(0.6086235489220564, 0.3611111111111111, 'x[30] <= 0.212\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.6019900497512438, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.615257048092869, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.5920398009950248, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6318407960199005, 0.5277777777777778, 'x[17] <= 0.056\ngini = 0.032\nsamples = 249\nvalue = [245, 4]'),
Text(0.615257048092869, 0.4722222222222222, 'x[16] <= 0.75\ngini = 0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.6086235489220564, 0.4166666666666667, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
Text(0.6218905472636815, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.648424543946932, 0.4722222222222222, 'x[2] <= 0.015\ngini = 0.024\nsamples = 244\nvalue = [241, 3]'),
Text(0.6351575456053068, 0.4166666666666667, 'x[11] <= 0.129\ngini = 0.278\nsamples = 6\nvalue = [5, 1]'),
Text(0.6285240464344942, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6417910447761194, 0.3611111111111111, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.6616915422885572, 0.4166666666666667, 'x[24] <= 0.167\ngini = 0.017\nsamples = 238\nvalue = [236, 2]'),
Text(0.6550580431177446, 0.3611111111111111, 'x[29] <= 0.833\ngini = 0.073\nsamples = 53\nvalue = [51, 2]'),
Text(0.6417910447761194, 0.3055555555555556, 'x[33] <= 0.088\ngini = 0.041\nsamples = 48\nvalue = [47, 1]'),
Text(0.6351575456053068, 0.25, 'x[14] <= 0.312\ngini = 0.245\nsamples = 7\nvalue = [6, 1]'),
Text(0.6285240464344942, 0.19444444444444445, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6417910447761194, 0.19444444444444445, 'gini = 0.0\nsamples = 6\nvalue = [6, 0]'),
Text(0.648424543946932, 0.25, 'gini = 0.0\nsamples = 41\nvalue = [41, 0]'),
Text(0.6683250414593698, 0.3055555555555556, 'x[31] <= 0.25\ngini = 0.32\nsamples = 5\nvalue = [4, 1]'),
Text(0.6616915422885572, 0.25, 'gini = 0.0\nsamples = 4\nvalue = [4, 0]'),
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Text(0.6749585406301825, 0.25, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.6683250414593698, 0.3611111111111111, 'gini = 0.0\nsamples = 185\nvalue = [185, 0]'),
Text(0.6285240464344942, 0.6388888888888888, 'x[2] <= 0.366\ngini = 0.408\nsamples = 7\nvalue = [5, 2]'),
Text(0.6218905472636815, 0.5833333333333334, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.6351575456053068, 0.5833333333333334, 'gini = 0.0\nsamples = 5\nvalue = [5, 0]'),
Text(0.4835199004975124, 0.75, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.824212271973466, 0.8611111111111112, 'x[17] <= 0.157\ngini = 0.385\nsamples = 300\nvalue = [222, 78]'),
Text(0.7371475953565506, 0.8055555555555556, 'x[26] <= 0.167\ngini = 0.5\nsamples = 96\nvalue = [49, 47]'),
Text(0.7014925373134329, 0.75, 'x[4] <= 0.161\ngini = 0.459\nsamples = 42\nvalue = [15, 27]'),
Text(0.6749585406301825, 0.6944444444444444, 'x[8] <= 0.415\ngini = 0.499\nsamples = 23\nvalue = [12, 11]'),
Text(0.6550580431177446, 0.6388888888888888, 'x[18] <= 0.561\ngini = 0.355\nsamples = 13\nvalue = [3, 10]'),
Text(0.648424543946932, 0.5833333333333334, 'gini = 0.0\nsamples = 8\nvalue = [0, 8]'),
Text(0.6616915422885572, 0.5833333333333334, 'x[28] <= 0.583\ngini = 0.48\nsamples = 5\nvalue = [3, 2]'),
Text(0.6550580431177446, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue = [3, 0]'),
Text(0.6683250414593698, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.6948590381426202, 0.6388888888888888, 'x[24] <= 0.167\ngini = 0.18\nsamples = 10\nvalue = [9, 1]'),
Text(0.6882255389718076, 0.5833333333333334, 'x[32] <= 0.067\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.681592039800995, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.6948590381426202, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.7014925373134329, 0.5833333333333334, 'gini = 0.0\nsamples = 8\nvalue = [8, 0]'),
Text(0.7280265339966833, 0.6944444444444444, 'x[13] <= 0.125\ngini = 0.266\nsamples = 19\nvalue = [3, 16]'),
Text(0.7213930348258707, 0.6388888888888888, 'x[11] <= 0.2\ngini = 0.198\nsamples = 18\nvalue = [2, 16]'),
Text(0.714759535655058, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.7280265339966833, 0.5833333333333334, 'x[31] <= 0.183\ngini = 0.111\nsamples = 17\nvalue = [1, 16]'),
Text(0.7213930348258707, 0.5277777777777778, 'gini = 0.0\nsamples = 15\nvalue = [0, 15]'),
Text(0.7346600331674958, 0.5277777777777778, 'x[18] <= 0.162\ngini = 0.5\nsamples = 2\nvalue = [1, 1]'),
Text(0.7280265339966833, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.7412935323383084, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.7346600331674958, 0.6388888888888888, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.7728026533996684, 0.75, 'x[0] <= 0.202\ngini = 0.466\nsamples = 54\nvalue = [34, 20]'),
Text(0.7545605306799337, 0.6944444444444444, 'x[12] <= 0.833\ngini = 0.245\nsamples = 7\nvalue = [1, 6]'),
Text(0.7479270315091211, 0.6388888888888888, 'gini = 0.0\nsamples = 6\nvalue = [0, 6]'),
Text(0.7611940298507462, 0.6388888888888888, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
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Text(0.7910447761194029, 0.6944444444444444, 'x[2] <= 0.622\ngini = 0.418\nsample
s = 47\nvalue = [33, 14]'),
Text(0.7744610281923715, 0.6388888888888888, 'x[2] <= 0.145\ngini = 0.482\nsample
s = 32\nvalue = [19, 13]'),
Text(0.7611940298507462, 0.5833333333333334, 'x[2] <= 0.024\ngini = 0.18\nsamples
= 10\nvalue = [9, 1]'),
Text(0.7545605306799337, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.7678275290215588, 0.5277777777777778, 'gini = 0.0\nsamples = 9\nvalue =
[9, 0]'),
Text(0.7877280265339967, 0.5833333333333334, 'x[18] <= 0.87\ngini = 0.496\nsample
s = 22\nvalue = [10, 12]'),
Text(0.7810945273631841, 0.5277777777777778, 'x[8] <= 0.41\ngini = 0.465\nsamples
= 19\nvalue = [7, 12]'),
Text(0.7678275290215588, 0.4722222222222222, 'x[18] <= 0.715\ngini = 0.469\nsampl
es = 8\nvalue = [5, 3]'),
Text(0.7611940298507462, 0.4166666666666667, 'gini = 0.0\nsamples = 5\nvalue =
[5, 0]'),
Text(0.7744610281923715, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.7943615257048093, 0.4722222222222222, 'x[0] <= 0.25\ngini = 0.298\nsamples
= 11\nvalue = [2, 9]'),
Text(0.7877280265339967, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.8009950248756219, 0.4166666666666667, 'x[18] <= 0.202\ngini = 0.18\nsample
s = 10\nvalue = [1, 9]'),
Text(0.7943615257048093, 0.3611111111111111, 'x[15] <= 0.5\ngini = 0.5\nsamples =
2\nvalue = [1, 1]'),
Text(0.7877280265339967, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.8009950248756219, 0.3055555555555556, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.8076285240464345, 0.3611111111111111, 'gini = 0.0\nsamples = 8\nvalue =
[0, 8]'),
Text(0.7943615257048093, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue =
[3, 0]'),
Text(0.8076285240464345, 0.6388888888888888, 'x[11] <= 0.064\ngini = 0.124\nsampl
es = 15\nvalue = [14, 1]'),
Text(0.8009950248756219, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue =
[0, 1]'),
Text(0.814262023217247, 0.5833333333333334, 'gini = 0.0\nsamples = 14\nvalue = [1
4, 0]'),
Text(0.9112769485903814, 0.8055555555555556, 'x[16] <= 0.75\ngini = 0.258\nsample
s = 204\nvalue = [173, 31]'),
Text(0.8590381426202321, 0.75, 'x[17] <= 0.992\ngini = 0.138\nsamples = 147\nvalu
e = [136, 11]'),
Text(0.8524046434494196, 0.6944444444444444, 'x[4] <= 0.482\ngini = 0.128\nsample
s = 146\nvalue = [136, 10]'),
Text(0.8341625207296849, 0.6388888888888888, 'x[30] <= 0.063\ngini = 0.038\nsampl
es = 104\nvalue = [102, 2]'),
Text(0.8275290215588723, 0.5833333333333334, 'x[11] <= 0.193\ngini = 0.32\nsample
s = 10\nvalue = [8, 2]'),
Text(0.8208955223880597, 0.5277777777777778, 'x[28] <= 0.417\ngini = 0.444\nsampl
es = 3\nvalue = [1, 2]'),
Text(0.814262023217247, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [1,
0]'),
Text(0.8275290215588723, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue =
[0, 2]'),
Text(0.8341625207296849, 0.5277777777777778, 'gini = 0.0\nsamples = 7\nvalue =
[7, 0]'),
Text(0.8407960199004975, 0.5833333333333334, 'gini = 0.0\nsamples = 94\nvalue =
[94, 0]'),
Text(0.8706467661691543, 0.6388888888888888, 'x[9] <= 0.167\ngini = 0.308\nsample
s = 42\nvalue = [34, 8]'),

```

```

Text(0.8540630182421227, 0.5833333333333334, 'x[29] <= 0.833\ngini = 0.375\nsamples = 4\nvalue = [1, 3]'),
Text(0.8474295190713101, 0.5277777777777778, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.8606965174129353, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.8872305140961857, 0.5833333333333334, 'x[0] <= 0.393\ngini = 0.229\nsamples = 38\nvalue = [33, 5]'),
Text(0.8739635157545605, 0.5277777777777778, 'x[9] <= 0.5\ngini = 0.5\nsamples = 6\nvalue = [3, 3]'),
Text(0.867330016583748, 0.4722222222222222, 'x[5] <= 0.625\ngini = 0.375\nsamples = 4\nvalue = [1, 3]'),
Text(0.8606965174129353, 0.4166666666666667, 'gini = 0.0\nsamples = 3\nvalue = [0, 3]'),
Text(0.8739635157545605, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.8805970149253731, 0.4722222222222222, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
Text(0.900497512437811, 0.5277777777777778, 'x[8] <= 0.992\ngini = 0.117\nsamples = 32\nvalue = [30, 2]'),
Text(0.8938640132669984, 0.4722222222222222, 'x[28] <= 0.917\ngini = 0.062\nsamples = 31\nvalue = [30, 1]'),
Text(0.8872305140961857, 0.4166666666666667, 'gini = 0.0\nsamples = 30\nvalue = [30, 0]'),
Text(0.900497512437811, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.9071310116086235, 0.4722222222222222, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.8656716417910447, 0.6944444444444444, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]'),
Text(0.9635157545605307, 0.75, 'x[14] <= 0.812\ngini = 0.456\nsamples = 57\nvalue = [37, 20]'),
Text(0.9402985074626866, 0.6944444444444444, 'x[32] <= 0.4\ngini = 0.238\nsamples = 29\nvalue = [25, 4]'),
Text(0.9270315091210614, 0.6388888888888888, 'x[8] <= 0.071\ngini = 0.142\nsamples = 26\nvalue = [24, 2]'),
Text(0.9203980099502488, 0.5833333333333334, 'x[33] <= 0.412\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.9137645107794361, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.9270315091210614, 0.5277777777777778, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.9336650082918739, 0.5833333333333334, 'gini = 0.0\nsamples = 23\nvalue = [23, 0]'),
Text(0.9535655058043118, 0.6388888888888888, 'x[2] <= 0.324\ngini = 0.444\nsamples = 3\nvalue = [1, 2]'),
Text(0.9469320066334992, 0.5833333333333334, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.9601990049751243, 0.5833333333333334, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
Text(0.9867330016583747, 0.6944444444444444, 'x[32] <= 0.1\ngini = 0.49\nsamples = 28\nvalue = [12, 16]'),
Text(0.9800995024875622, 0.6388888888888888, 'x[4] <= 0.804\ngini = 0.48\nsamples = 20\nvalue = [12, 8]'),
Text(0.9734660033167496, 0.5833333333333334, 'x[30] <= 0.013\ngini = 0.415\nsamples = 17\nvalue = [12, 5]'),
Text(0.966832504145937, 0.5277777777777778, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
Text(0.9800995024875622, 0.5277777777777778, 'x[24] <= 0.5\ngini = 0.32\nsamples = 15\nvalue = [12, 3]'),
Text(0.9734660033167496, 0.4722222222222222, 'x[0] <= 0.286\ngini = 0.5\nsamples = 6\nvalue = [3, 3]'),
Text(0.966832504145937, 0.4166666666666667, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),

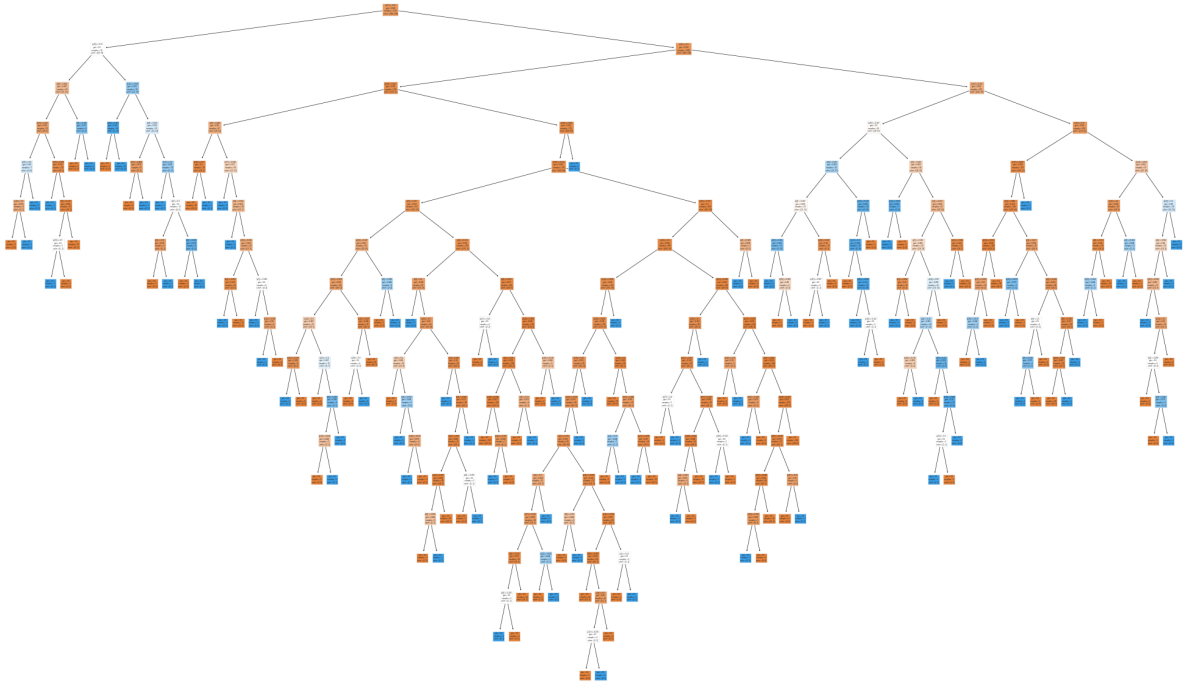
```



```

Text(0.9800995024875622, 0.4166666666666667, 'x[2] <= 0.197\nngini = 0.375\nsample
s = 4\nvalue = [1, 3]'),
Text(0.9734660033167496, 0.3611111111111111, 'gini = 0.0\nsamples = 1\nvalue =
[1, 0]'),
Text(0.9867330016583747, 0.3611111111111111, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.9867330016583747, 0.4722222222222222, 'gini = 0.0\nsamples = 9\nvalue =
[9, 0]'),
Text(0.9867330016583747, 0.5833333333333334, 'gini = 0.0\nsamples = 3\nvalue =
[0, 3]'),
Text(0.9933665008291874, 0.6388888888888888, 'gini = 0.0\nsamples = 8\nvalue =
[0, 8]')]

```



```

In [258...] 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']
}

```

```

In [259...] grid_search=GridSearchCV(estimator=dtc,param_grid=parameter,cv=5,scoring="accuracy")

```

```

In [260...] grid_search.fit(x_train,y_train)

```



[illegible]



[illegible]



[illegible]

[illegible]

```

/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 the past behaviour, explicitly set `max_features='sqrt'`.
  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 the past behaviour, explicitly set `max_features='sqrt'`.
  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 the past behaviour, explicitly set `max_features='sqrt'`.
  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 the past behaviour, explicitly set `max_features='sqrt'`.
  warnings.warn(

```

```

Out[260]:
└─ GridSearchCV
  └─ estimator: DecisionTreeClassifier
    └─ DecisionTreeClassifier

```

```
In [261... grid_search.best_params_
```

```

Out[261]: {'criterion': 'entropy',
           'max_depth': 3,
           'max_features': 'sqrt',
           'splitter': 'best'}

```

```

In [265... dtc_cv=DecisionTreeClassifier(criterion='entropy',
                                     max_depth= 3,
                                     max_features= 'sqrt',
                                     splitter='best')
dtc_cv.fit(x_train,y_train)

```

```

Out[265]:
└─ DecisionTreeClassifier
  DecisionTreeClassifier(criterion='entropy', max_depth=3, max_features='sqrt')

```

```
In [268... pred=dtc_cv.predict(x_test)
```

```
In [270... print(classification_report(y_test,pred))
```

	precision	recall	f1-score	support
0	0.83	1.00	0.91	245
1	0.00	0.00	0.00	49
accuracy			0.83	294
macro avg	0.42	0.50	0.45	294
weighted avg	0.69	0.83	0.76	294

After hyperparameter tuning, accuracy is 83%

## Random Forest

```

In [271... from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()

```



```
In [272...] forest_params=[{'max_depth': list(range(10,15)), 'max_features': list(range(0,14))]}

In [273...] rfc_cv=GridSearchCV(rfc,forest_params,cv=10,scoring="accuracy")

In [274...] rfc_cv.fit(x_train,y_train)
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/model_selection/_validation.py:378: FitFailedWarning:
50 fits failed out of a total of 700.
The score on these train-test partitions for these parameters will be set to nan.
If these failures are not expected, you can try to debug them by setting error_score='raise'.
```

Below are more details about the failures:

-----

50 fits failed with the following error:

Traceback (most recent call last):

File "/usr/local/lib/python3.10/dist-packages/sklearn/model\_selection/\_validation.py", line 686, in \_fit\_and\_score

estimator.fit(X\_train, y\_train, \*\*fit\_params)

File "/usr/local/lib/python3.10/dist-packages/sklearn/ensemble/\_forest.py", line 340, in fit

self.\_validate\_params()

File "/usr/local/lib/python3.10/dist-packages/sklearn/base.py", line 600, in \_validate\_params

validate\_parameter\_constraints(

File "/usr/local/lib/python3.10/dist-packages/sklearn/utils/\_param\_validation.py", line 97, in validate\_parameter\_constraints

raise InvalidParameterError(

sklearn.utils.\_param\_validation.InvalidParameterError: The 'max\_features' parameter of RandomForestClassifier must be an int in the range [1, inf), a float in the range (0.0, 1.0], a str among {'sqrt', 'auto' (deprecated), 'log2'} or None. Got 0 instead.

warnings.warn(some\_fits\_failed\_message, FitFailedWarning)

/usr/local/lib/python3.10/dist-packages/sklearn/model\_selection/\_search.py:952: UserWarning: One or more of the test scores are non-finite: [ nan 0.85120238 0.85630885 0.85459221 0.85291178 0.86053165

0.86137911 0.85798204 0.85882949 0.85966971 0.85798204 0.85966247

0.8605389 0.85797479 nan 0.8443865 0.85204259 0.86224105

0.86052441 0.85967695 0.85969144 0.86138635 0.86477618 0.86137911

0.86393597 0.86137911 0.86052441 0.86307403 nan 0.85034043

0.85373026 0.85714182 0.85712734 0.85712009 0.85714182 0.85627264

0.86221932 0.85457048 0.86222657 0.85627988 0.85542518 0.85795306

nan 0.85204983 0.85969868 0.85712734 0.85457048 0.85966247

0.86137911 0.86051717 0.86478343 0.85287556 0.85628712 0.85798204

0.86052441 0.86647834 nan 0.84950746 0.85546139 0.85971317

0.86138635 0.85627988 0.85967695 0.85882225 0.86223381 0.85628712

0.86221932 0.85458496 0.85880776 0.85882225]

warnings.warn(

Out[274]:

```
GridSearchCV
  estimator: RandomForestClassifier
    RandomForestClassifier
```

```
In [275...] pred=rfc_cv.predict(x_test)
```

```
In [276...] print(classification_report(y_test,pred))
```

	aniket			
	precision	recall	f1-score	support
0	0.86	0.98	0.91	245
1	0.62	0.20	0.31	49
accuracy			0.85	294
macro avg	0.74	0.59	0.61	294
weighted avg	0.82	0.85	0.81	294

Accuracy of this model is 85%

In [ ]: