

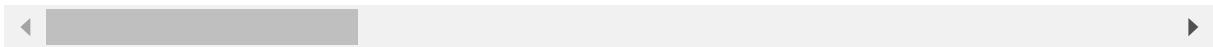
```
In [ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [ ]: data = pd.read_csv('WA_Fn-UseC_-HR-Employee-Attrition.csv')
data.head()
```

Out[3]:

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

5 rows × 35 columns



```
In [ ]: data.tail()
```

Out[4]:

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

5 rows × 35 columns



In [ ]: `data.info()`

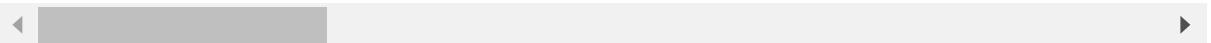
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1470 entries, 0 to 1469
Data columns (total 35 columns):
 #   Column           Non-Null Count Dtype  
 --- 
 0   Age              1470 non-null   int64  
 1   Attrition        1470 non-null   object  
 2   BusinessTravel   1470 non-null   object  
 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
```

In [ ]: `data.describe()`

Out[6]:

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount	EmployeeNu
<b>count</b>	1470.000000	1470.000000	1470.000000	1470.000000	1470.0	1470.00
<b>mean</b>	36.923810	802.485714	9.192517	2.912925	1.0	1024.86
<b>std</b>	9.135373	403.509100	8.106864	1.024165	0.0	602.02
<b>min</b>	18.000000	102.000000	1.000000	1.000000	1.0	1.00
<b>25%</b>	30.000000	465.000000	2.000000	2.000000	1.0	491.25
<b>50%</b>	36.000000	802.000000	7.000000	3.000000	1.0	1020.50
<b>75%</b>	43.000000	1157.000000	14.000000	4.000000	1.0	1555.75
<b>max</b>	60.000000	1499.000000	29.000000	5.000000	1.0	2068.00

8 rows × 26 columns



```
In [ ]: data.isnull().sum()
```

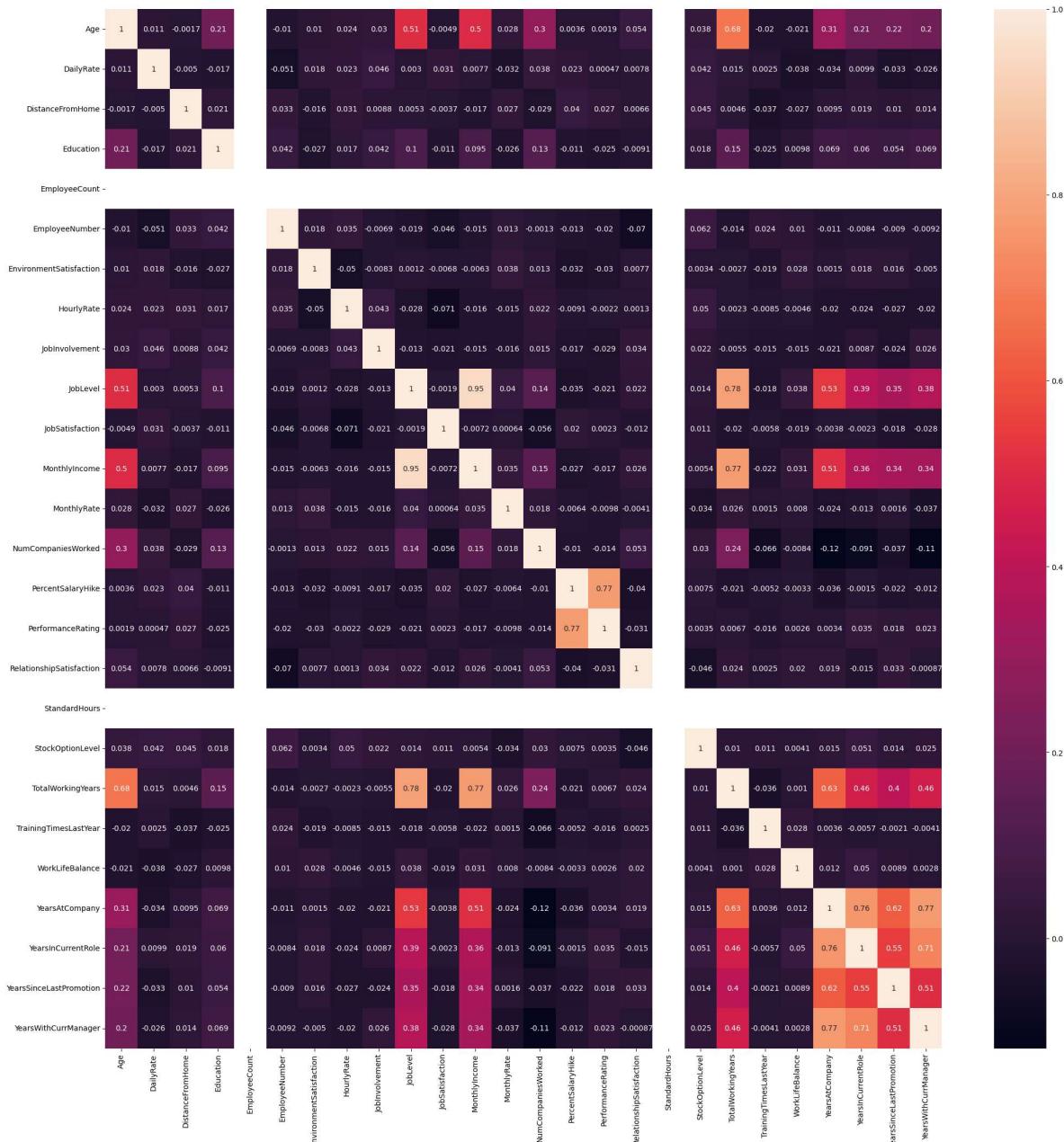
```
Out[7]: Age                  0
Attrition            0
BusinessTravel        0
DailyRate              0
Department            0
DistanceFromHome      0
Education              0
EducationField         0
EmployeeCount          0
EmployeeNumber         0
EnvironmentSatisfaction 0
Gender                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
```

```
In [ ]: cor = data.corr()
```

```
<ipython-input-8-06847dd9a2e1>: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.
cor = data.corr()
```

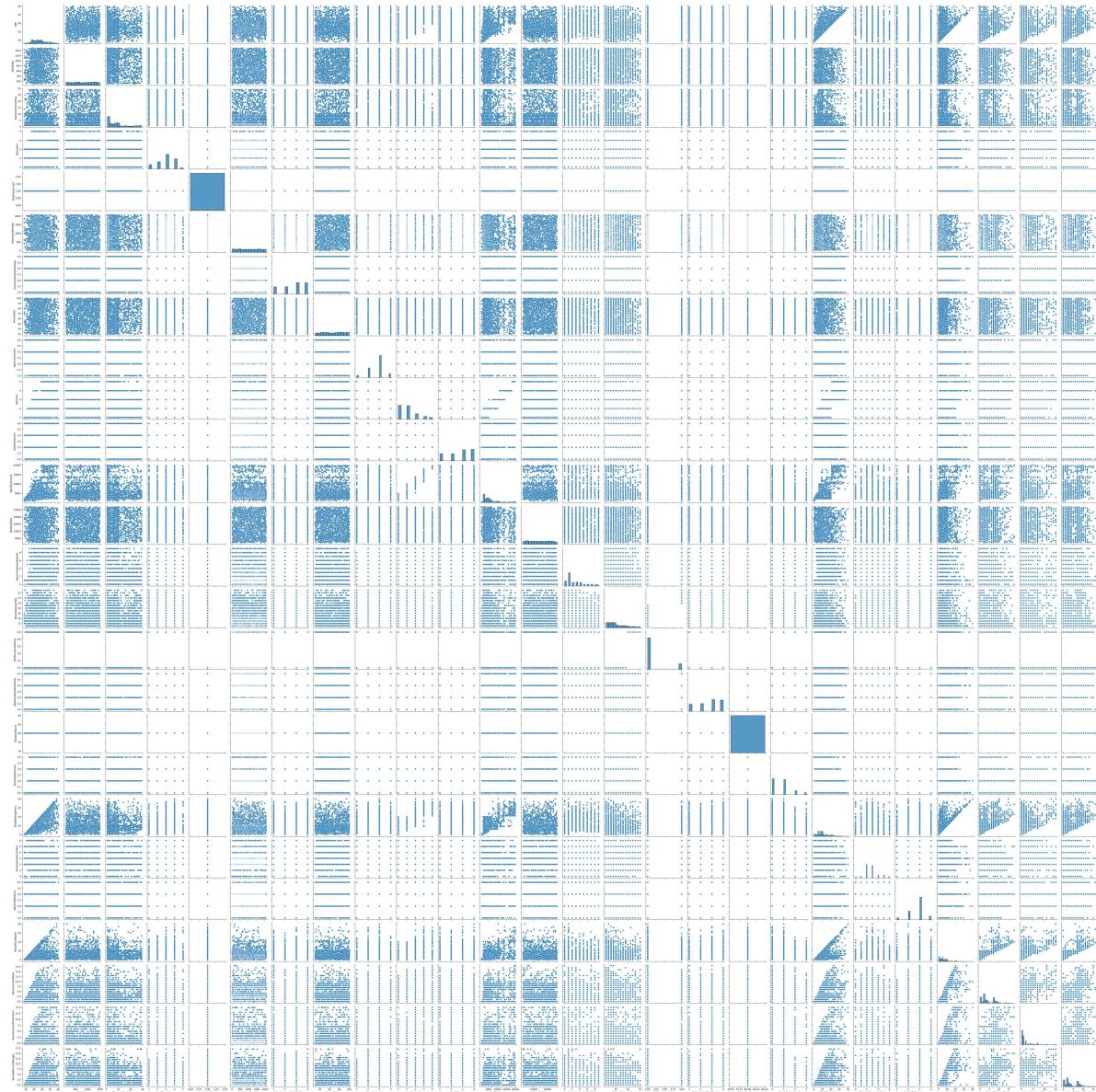
```
In [ ]: fig, ax = plt.subplots(figsize=(25,25))
sns.heatmap(cor, annot=True)
```

Out[9]: <Axes: >



```
In [11]: sns.pairplot(data)
```

```
Out[11]: <seaborn.axisgrid.PairGrid at 0x7bcdd83535e0>
```



```
In [12]: from sklearn.preprocessing import LabelEncoder
```

```
In [13]: le=LabelEncoder()
```

```
In [14]: data["BusinessTravel"]=le.fit_transform(data["BusinessTravel"])
```

```
In [15]: data["Department"]=le.fit_transform(data["Department"])
```

```
In [17]: data["EducationField"]=le.fit_transform(data["EducationField"])
```

```
In [16]: data["Gender"] = le.fit_transform(data["Gender"])
```

```
In [18]: data["JobRole"] = le.fit_transform(data["JobRole"])
```

```
In [19]: data["MaritalStatus"] = le.fit_transform(data["MaritalStatus"])
```

```
In [20]: data["Over18"] = le.fit_transform(data["Over18"])
```

```
In [21]: data["OverTime"] = le.fit_transform(data["OverTime"])
```

```
In [22]: data.head()
```

Out[22]:

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	Educat
0	41	Yes	2	1102	2		1	2
1	49	No	1	279	1		8	1
2	37	Yes	2	1373	1		2	2
3	33	No	1	1392	1		3	4
4	27	No	2	591	1		2	1

5 rows × 35 columns



```
In [23]: data.tail()
```

Out[23]:

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	Edu
1465	36	No	1	884	1		23	2
1466	39	No	2	613	1		6	1
1467	27	No	2	155	1		4	3
1468	49	No	1	1023	2		2	3
1469	34	No	2	628	1		8	3

5 rows × 35 columns



```
In [24]: X = data.drop(columns=["EmployeeNumber", "EmployeeCount", "StandardHours", "Attriti
```

```
In [25]: y = data["Attrition"]
```

```
In [26]: from sklearn.preprocessing import MinMaxScaler  
ms = MinMaxScaler()
```

```
In [27]: X_Scaled=ms.fit_transform(X)
```

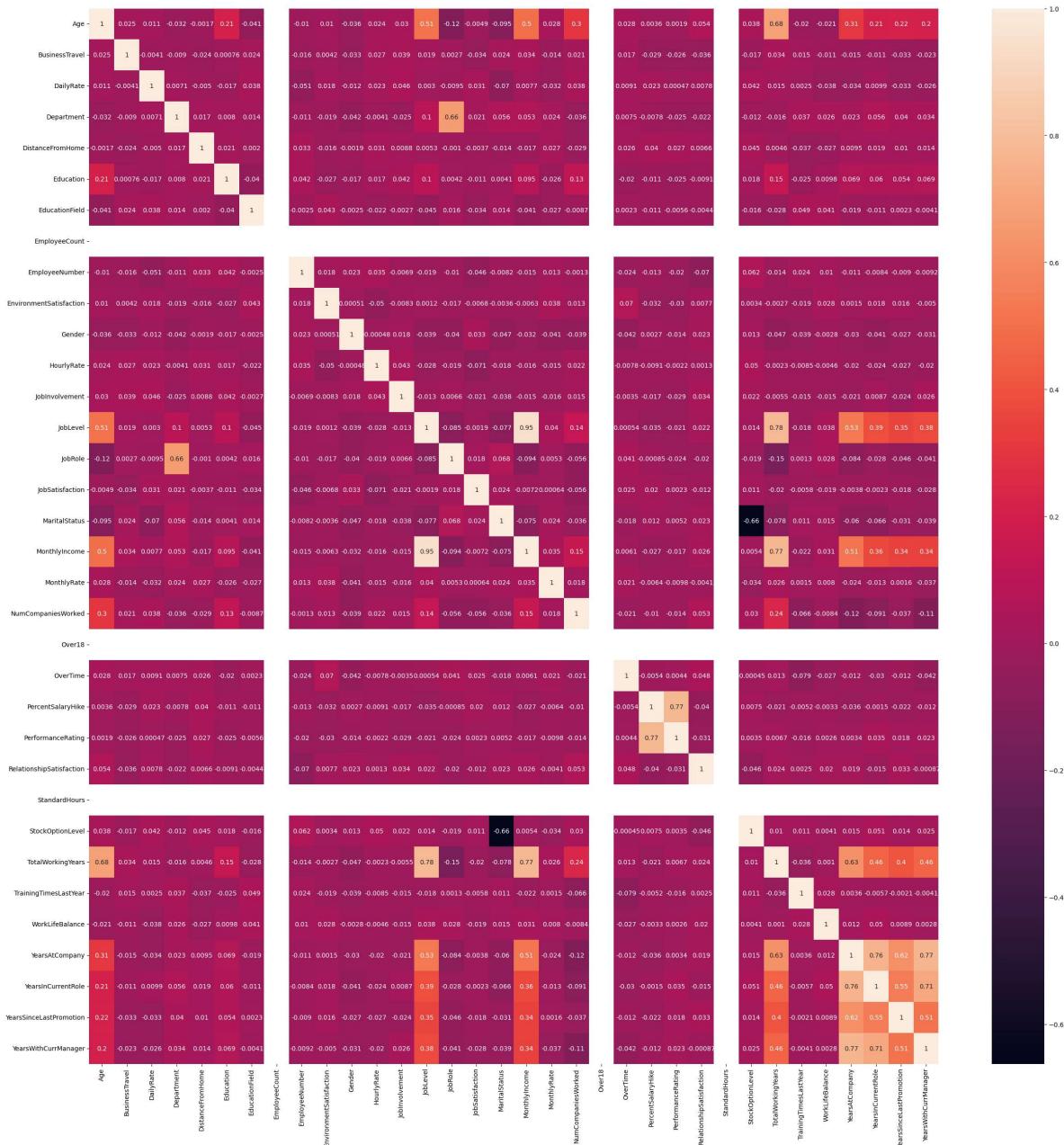
```
In [28]: cor=data.corr()
```

<ipython-input-28-410fe4458127>: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.

```
cor=data.corr()
```

```
In [29]: fig, ax = plt.subplots(figsize=(30,30))
sns.heatmap(cor, annot=True)
```

Out[29]: <Axes: >



```
In [30]: 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)
```

```
In [31]: from sklearn.linear_model import LogisticRegression
classifier = LogisticRegression(random_state=0)
classifier.fit(x_train,y_train)
```

Out[31]: LogisticRegression(random\_state=0)

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```
In [32]: from sklearn.metrics import accuracy_score,confusion_matrix
y_pred = classifier.predict(x_test)
cm = confusion_matrix(y_test, y_pred)
print(cm)
accuracy_score(y_test, y_pred)*100
```

```
[[242  3]
 [ 32 17]]
```

Out[32]: 88.09523809523809

```
In [33]: from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
```

```
In [34]: print(classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
No	0.88	0.99	0.93	245
Yes	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 [35]: from sklearn.tree import DecisionTreeClassifier
dtc=DecisionTreeClassifier()
```

```
In [36]: dtc.fit(x_train,y_train)
```

Out[36]: DecisionTreeClassifier()

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```
In [37]: from sklearn.metrics import accuracy_score,confusion_matrix
y_pred = dtc.predict(x_test)
cm = confusion_matrix(y_test, y_pred)
print(cm)
accuracy_score(y_test, y_pred)*100
```

```
[[205  40]
 [ 34  15]]
```

Out[37]: 74.82993197278913

```
In [38]: from sklearn import tree
plt.figure(figsize=(25,15))
tree.plot_tree(dtc,filled=True)
```

```
Out[38]: [Text(0.319423577724359, 0.9722222222222222, 'x[23] <= 0.038\n gini = 0.269
\nsamples = 1176\nvalue = [988, 188]'),
Text(0.07692307692307693, 0.9166666666666666, 'x[14] <= 0.75\n gini = 0.5
\nsamples = 78\nvalue = [39, 39]'),
Text(0.04807692307692308, 0.8611111111111112, 'x[4] <= 0.554\n gini = 0.42
6\nsamples = 39\nvalue = [27, 12]'),
Text(0.03205128205128205, 0.8055555555555556, 'x[13] <= 0.167\n gini = 0.3
12\nsamples = 31\nvalue = [25, 6]'),
Text(0.019230769230769232, 0.75, 'x[18] <= 0.5\n gini = 0.49\nsamples = 7
\nvalue = [3, 4]'),
Text(0.01282051282051282, 0.6944444444444444, 'x[2] <= 0.298\n gini = 0.37
5\nsamples = 4\nvalue = [3, 1]'),
Text(0.00641025641025641, 0.6388888888888888, 'gini = 0.0\nsamples = 1\nv
alue = [0, 1]'),
Text(0.019230769230769232, 0.6388888888888888, 'gini = 0.0\nsamples = 3\nn
value = [3, 0]'),
Text(0.02564102564102564, 0.6944444444444444, 'gini = 0.0\nsamples = 3\nv
alue = [0, 3]'),
Text(0.04487179487179487, 0.75, 'x[17] <= 0.056\n gini = 0.153\nsamples =
22\nvalue = [22, 22]')]
```

```
In [39]: from sklearn.model_selection import GridSearchCV
parameter={
    'criterion':['gini','entropy'],
    'splitter':['best','random'],
    'max_depth':[1,2,3,4,5,6,7,8,9,10],
    'max_features':['auto', 'sqrt', 'log2']

}
```

```
In [40]: grid_search=GridSearchCV(estimator=dtc,param_grid=parameter,cv=5,scoring="accu
```

In [41]: `grid_search.fit(x_train,y_train)`

```
/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'`.

.
.
```

In [42]: `grid_search.best_params_`

Out[42]: `{'criterion': 'gini',  
 'max_depth': 5,  
 'max_features': 'log2',  
 'splitter': 'best'}`

In [43]: `dtc_cv=DecisionTreeClassifier(criterion= 'entropy',  
 max_depth= 4,  
 max_features= 'sqrt',  
 splitter= 'best')  
 dtc_cv.fit(x_train,y_train)`

Out[43]: `DecisionTreeClassifier(criterion='entropy', max_depth=4, max_features='sqrt')`

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In [44]: `print(classification_report(y_test,y_pred))`

	precision	recall	f1-score	support
No	0.86	0.84	0.85	245
Yes	0.27	0.31	0.29	49
accuracy			0.75	294
macro avg	0.57	0.57	0.57	294
weighted avg	0.76	0.75	0.75	294

```
In [45]: from sklearn.ensemble import RandomForestClassifier  
classifier = RandomForestClassifier(n_estimators = 1000, criterion = 'entropy')  
classifier.fit(x_train, y_train)
```

```
Out[45]: RandomForestClassifier(criterion='entropy', n_estimators=1000, random_state=0)
```

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```
In [46]: from sklearn.metrics import confusion_matrix, accuracy_score  
y_pred = classifier.predict(x_test)  
cm = confusion_matrix(y_test, y_pred)  
print(cm)  
accuracy_score(y_test, y_pred)
```

```
[[243  2]  
 [ 41  8]]
```

```
Out[46]: 0.8537414965986394
```

```
In [47]: from sklearn.ensemble import RandomForestClassifier
```

```
In [48]: rfc=RandomForestClassifier()
```

```
In [49]: forest_params = [{ 'max_depth': list(range(10, 15)), 'max_features': list(range(10, 15)) }]
```

```
In [50]: rfc_cv=GridSearchCV(rfc,param_grid=forest_params,cv=10,scoring="accuracy")
```

```
In [51]: 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 erro  
r_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/_vali  
dation.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, i  
n _validate_params  
    validate_parameter_constraints()  
  File "/usr/local/lib/python3.10/dist-packages/sklearn/utils/_param_validati  
on.py", line 97, in validate_parameter_constraints  
    raise InvalidParameterError(  
sklearn.utils._param_validation.InvalidParameterError: The 'max_features' pa  
rameter of RandomForestClassifier must be an int in the range [1, inf), a flo  
at 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:95  
2: UserWarning: One or more of the test scores are non-finite: [      nan 0.  
84608866 0.85374475 0.85713458 0.85969144 0.86222657  
  0.85881501 0.86053165 0.85880776 0.86562364 0.85712734 0.86050992  
  0.86051717 0.86308851      nan 0.84950746 0.85628712 0.86052441  
  0.85882949 0.86306678 0.86307403 0.85456323 0.85370853 0.86392873  
  0.85627264 0.85885122 0.85882225 0.85625815      nan 0.84864552  
  0.85544691 0.85714182 0.8596842 0.86137911 0.86136462 0.85966971  
  0.85966247 0.85965522 0.85883674 0.86221932 0.85882225 0.85882225  
      nan 0.84695784 0.8613936 0.86138635 0.85712009 0.86224105  
  0.85967695 0.85882949 0.85798204 0.86137187 0.86394321 0.86479067  
  0.85541793 0.86308127      nan 0.85204983 0.85883674 0.85629436  
  0.85967695 0.85798928 0.86053165 0.85964798 0.86222657 0.85966971  
  0.86137911 0.86647834 0.85627264 0.86135014]  
  warnings.warn(
```

```
Out[51]: GridSearchCV(cv=10, estimator=RandomForestClassifier(),
                      param_grid=[{'max_depth': [10, 11, 12, 13, 14],
                                   'max_features': [0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10,
                                   11,
                                   12, 13]}],
                      scoring='accuracy')
```

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```
In [52]: print(classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
No	0.86	0.99	0.92	245
Yes	0.80	0.16	0.27	49
accuracy			0.85	294
macro avg	0.83	0.58	0.59	294
weighted avg	0.85	0.85	0.81	294