

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sea
```

```
In [2]: df = pd.read_csv(r"C:\Users\user\Downloads\C8_loan-train.csv")
df
```

Out[2]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amou
0	LP001002	Male	No	0	Graduate	No	5849	0.0	NaN	
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	
...
609	LP002978	Female	No	0	Graduate	No	2900	0.0	71.0	
610	LP002979	Male	Yes	3+	Graduate	No	4106	0.0	40.0	
611	LP002983	Male	Yes	1	Graduate	No	8072	240.0	253.0	
612	LP002984	Male	Yes	2	Graduate	No	7583	0.0	187.0	
613	LP002990	Female	No	0	Graduate	Yes	4583	0.0	133.0	

614 rows × 13 columns

```
In [3]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 614 entries, 0 to 613
Data columns (total 13 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Loan_ID               614 non-null   object
1   Gender                601 non-null   object
2   Married               611 non-null   object
3   Dependents            599 non-null   object
4   Education             614 non-null   object
5   Self_Employed         582 non-null   object
6   ApplicantIncome       614 non-null   int64
7   CoapplicantIncome     614 non-null   float64
8   LoanAmount            592 non-null   float64
9   Loan_Amount_Term      600 non-null   float64
10  Credit_History         564 non-null   float64
11  Property_Area         614 non-null   object
12  Loan_Status           614 non-null   object
dtypes: float64(4), int64(1), object(8)
memory usage: 62.5+ KB
```

```
In [4]: df2 = df.fillna(0)
```

In [5]: df2

Out[5]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amou
0	LP001002	Male	No	0	Graduate	No	5849	0.0	0.0	
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	
...
609	LP002978	Female	No	0	Graduate	No	2900	0.0	71.0	
610	LP002979	Male	Yes	3+	Graduate	No	4106	0.0	40.0	
611	LP002983	Male	Yes	1	Graduate	No	8072	240.0	253.0	
612	LP002984	Male	Yes	2	Graduate	No	7583	0.0	187.0	
613	LP002990	Female	No	0	Graduate	Yes	4583	0.0	133.0	

614 rows × 13 columns

In [6]: df2.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 614 entries, 0 to 613
Data columns (total 13 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Loan_ID               614 non-null   object
1   Gender                614 non-null   object
2   Married               614 non-null   object
3   Dependents            614 non-null   object
4   Education             614 non-null   object
5   Self_Employed         614 non-null   object
6   ApplicantIncome       614 non-null   int64
7   CoapplicantIncome     614 non-null   float64
8   LoanAmount            614 non-null   float64
9   Loan_Amount_Term      614 non-null   float64
10  Credit_History         614 non-null   float64
11  Property_Area         614 non-null   object
12  Loan_Status           614 non-null   object
dtypes: float64(4), int64(1), object(8)
memory usage: 62.5+ KB
```

In [7]: df3 = df2[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount', 'Loan_Amount_Term', 'Credit_History', 'Loan_Status']]

Out[7]:

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Loan_Status
0	5849	0.0	0.0	360.0	1.0	Y
1	4583	1508.0	128.0	360.0	1.0	N
2	3000	0.0	66.0	360.0	1.0	Y
3	2583	2358.0	120.0	360.0	1.0	Y
4	6000	0.0	141.0	360.0	1.0	Y
...
609	2900	0.0	71.0	360.0	1.0	Y
610	4106	0.0	40.0	180.0	1.0	Y
611	8072	240.0	253.0	360.0	1.0	Y
612	7583	0.0	187.0	360.0	1.0	Y
613	4583	0.0	133.0	360.0	0.0	N

614 rows × 6 columns

In [8]: df3.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 614 entries, 0 to 613
Data columns (total 6 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ApplicantIncome       614 non-null   int64
1   CoapplicantIncome     614 non-null   float64
2   LoanAmount            614 non-null   float64
3   Loan_Amount_Term      614 non-null   float64
4   Credit_History        614 non-null   float64
5   Loan_Status           614 non-null   object
dtypes: float64(4), int64(1), object(1)
memory usage: 28.9+ KB
```

In [10]: df3['Loan_Status'].value_counts()

```
Out[10]: Y    422
         N    192
         Name: Loan_Status, dtype: int64
```

In [11]: ch = {"Loan_Status":{"Y":1,'N':0}}
df3 = df3.replace(ch)
df3

Out[11]:

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Loan_Status
0	5849	0.0	0.0	360.0	1.0	1
1	4583	1508.0	128.0	360.0	1.0	0
2	3000	0.0	66.0	360.0	1.0	1
3	2583	2358.0	120.0	360.0	1.0	1
4	6000	0.0	141.0	360.0	1.0	1
...
609	2900	0.0	71.0	360.0	1.0	1
610	4106	0.0	40.0	180.0	1.0	1
611	8072	240.0	253.0	360.0	1.0	1
612	7583	0.0	187.0	360.0	1.0	1
613	4583	0.0	133.0	360.0	0.0	0

614 rows × 6 columns

In [14]: x = df3[['ApplicantIncome','CoapplicantIncome','LoanAmount','Loan_Amount_Term','Credit_History']]
y = df3['Loan_Status']

In [15]: from sklearn.model_selection import train_test_split

In [16]: x_train,x_test,y_train,y_test = train_test_split(x,y,train_size=0.70)

In [17]: from sklearn.ensemble import RandomForestClassifier

In [18]: rfc = RandomForestClassifier()
rfc.fit(x_train,y_train)

Out[18]: RandomForestClassifier()

In [19]: parameters = {
 'max_depth':[1,2,3,4,5],
 'min_samples_leaf':[5,10,15,20,25],
 'n_estimators':[10,20,30,40,50]
}

```
In [20]: from sklearn.model_selection import GridSearchCV
```

```
In [21]: grid_search = GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring='accuracy')
grid_search.fit(x_train,y_train)
```

```
Out[21]: GridSearchCV(cv=2, estimator=RandomForestClassifier(),
    param_grid={'max_depth': [1, 2, 3, 4, 5],
    'min_samples_leaf': [5, 10, 15, 20, 25],
    'n_estimators': [10, 20, 30, 40, 50]},
    scoring='accuracy')
```

```
In [22]: grid_search.best_score_
```

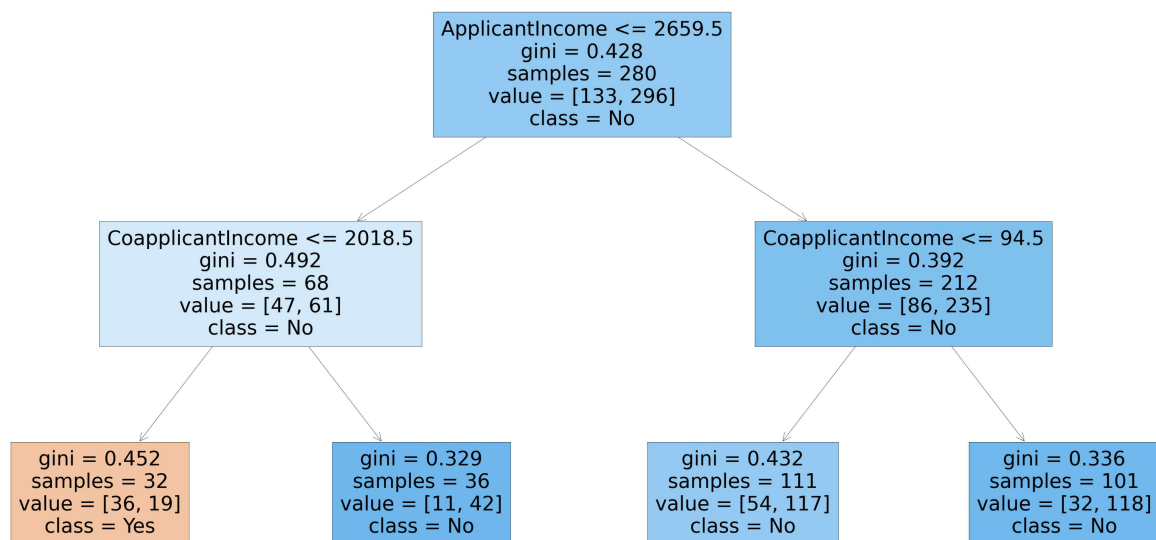
```
Out[22]: 0.7809280591175831
```

```
In [23]: from sklearn.tree import plot_tree
```

```
In [24]: rfc_best= grid_search.best_estimator_
```

```
In [25]: plt.figure(figsize=(80,40))
plot_tree(rfc_best.estimators_[5],feature_names=x.columns,class_names=["Yes","No"],filled=True)
```

```
Out[25]: [Text(2232.0, 1812.0, 'ApplicantIncome <= 2659.5\nngini = 0.428\nsamples = 280\nvalue = [133, 296]\nnclass = N
o'),
Text(1116.0, 1087.2, 'CoapplicantIncome <= 2018.5\nngini = 0.492\nsamples = 68\nvalue = [47, 61]\nnclass = N
o'),
Text(558.0, 362.39999999999986, 'gini = 0.452\nsamples = 32\nvalue = [36, 19]\nnclass = Yes'),
Text(1674.0, 362.39999999999986, 'gini = 0.329\nsamples = 36\nvalue = [11, 42]\nnclass = No'),
Text(3348.0, 1087.2, 'CoapplicantIncome <= 94.5\nngini = 0.392\nsamples = 212\nvalue = [86, 235]\nnclass = N
o'),
Text(2790.0, 362.39999999999986, 'gini = 0.432\nsamples = 111\nvalue = [54, 117]\nnclass = No'),
Text(3906.0, 362.39999999999986, 'gini = 0.336\nsamples = 101\nvalue = [32, 118]\nnclass = No')]
```



```
In [ ]:
```