

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

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

Out[2]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coappli
0	LP001002	Male	No	0	Graduate	No	5849	
1	LP001003	Male	Yes	1	Graduate	No	4583	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	
4	LP001008	Male	No	0	Graduate	No	6000	
...
609	LP002978	Female	No	0	Graduate	No	2900	
610	LP002979	Male	Yes	3+	Graduate	No	4106	
611	LP002983	Male	Yes	1	Graduate	No	8072	
612	LP002984	Male	Yes	2	Graduate	No	7583	
613	LP002990	Female	No	0	Graduate	Yes	4583	

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 [6]: df.columns
```

```
Out[6]: Index(['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',
              'Self_Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
              'Loan_Amount_Term', 'Credit_History', 'Property_Area', 'Loan_Status'],
              dtype='object')
```

```
In [4]: df['Loan_Status'].value_counts()
```

```
Out[4]: Y    422
        N    192
        Name: Loan_Status, dtype: int64
```

```
In [22]: df1=df[['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',
               'Self_Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
               'Loan_Amount_Term', 'Credit_History', 'Property_Area', 'Loan_Status']]
```

```
In [48]: x=df1[['ApplicantIncome', 'CoapplicantIncome']]
        y=df1['Loan_Status']
```

```
In [49]: g1={'Loan_Status':{'Y':1,'N':2}}
df1=df1.replace(g1)
print(df1)
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	\
0	LP001002	Male	No	0	Graduate	No	
1	LP001003	Male	Yes	1	Graduate	No	
2	LP001005	Male	Yes	0	Graduate	Yes	
3	LP001006	Male	Yes	0	Not Graduate	No	
4	LP001008	Male	No	0	Graduate	No	
..	
609	LP002978	Female	No	0	Graduate	No	
610	LP002979	Male	Yes	3+	Graduate	No	
611	LP002983	Male	Yes	1	Graduate	No	
612	LP002984	Male	Yes	2	Graduate	No	
613	LP002990	Female	No	0	Graduate	Yes	

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

	Credit_History	Property_Area	Loan_Status
0	1.0	Urban	1
1	1.0	Rural	2
2	1.0	Urban	1
3	1.0	Urban	1
4	1.0	Urban	1
..
609	1.0	Rural	1
610	1.0	Rural	1
611	1.0	Urban	1
612	1.0	Urban	1
613	0.0	Semiurban	2

[614 rows x 13 columns]

```
In [50]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=45)
```

```
In [51]: from sklearn.ensemble import RandomForestClassifier
rfc = RandomForestClassifier()
rfc.fit(x_train,y_train)
```

```
Out[51]: RandomForestClassifier()
```

```
In [52]: parameters = {'max_depth':[1,2,3,4,5],  
                      'min_samples_leaf':[5,10,15,20,25],  
                      'n_estimators':[10,20,30,40,50]}
```

```
In [53]: from sklearn.model_selection import GridSearchCV  
  
grid_search = GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring='acc  
grid_search.fit(x_train,y_train)
```

```
Out[53]: 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 [54]: grid_search.best_score_
```

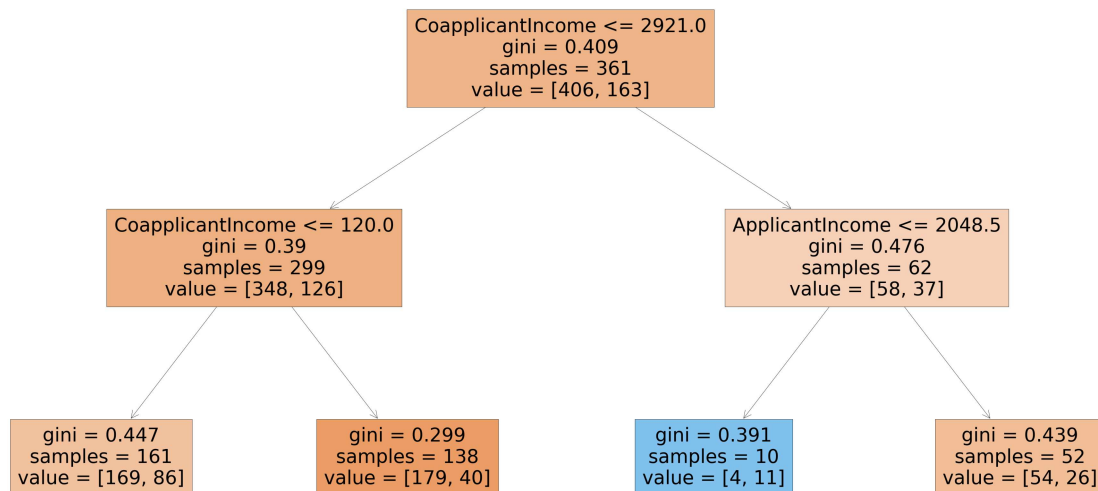
```
Out[54]: 0.6854212997281937
```

```
In [55]: rfc_best = grid_search.best_estimator_
```

In [56]: `from sklearn.tree import plot_tree`

```
plt.figure(figsize=(80,40))
plot_tree(rfc_best.estimators_[4],feature_names=x.columns,filled=True)
```

Out[56]: [Text(2232.0, 1812.0, 'CoapplicantIncome <= 2921.0\ngini = 0.409\nsamples = 361\nvalue = [406, 163]'),
Text(1116.0, 1087.2, 'CoapplicantIncome <= 120.0\ngini = 0.39\nsamples = 299\nvalue = [348, 126]'),
Text(558.0, 362.39999999999986, 'gini = 0.447\nsamples = 161\nvalue = [169, 86]'),
Text(1674.0, 362.39999999999986, 'gini = 0.299\nsamples = 138\nvalue = [179, 40]'),
Text(3348.0, 1087.2, 'ApplicantIncome <= 2048.5\ngini = 0.476\nsamples = 62\nvalue = [58, 37]'),
Text(2790.0, 362.39999999999986, 'gini = 0.391\nsamples = 10\nvalue = [4, 11]'),
Text(3906.0, 362.39999999999986, 'gini = 0.439\nsamples = 52\nvalue = [54, 26]')]



In []:

