

```
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\loan test.csv")
df
```

Out[2]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIn
0	LP001015	Male	Yes	0	Graduate	No	5720	
1	LP001022	Male	Yes	1	Graduate	No	3076	
2	LP001031	Male	Yes	2	Graduate	No	5000	
3	LP001035	Male	Yes	2	Graduate	No	2340	
4	LP001051	Male	No	0	Not Graduate	No	3276	
...
362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	
363	LP002975	Male	Yes	0	Graduate	No	4158	
364	LP002980	Male	No	0	Graduate	No	3250	
365	LP002986	Male	Yes	0	Graduate	No	5000	
366	LP002989	Male	No	0	Graduate	Yes	9200	

367 rows × 12 columns



```
In [3]: df.columns
```

```
Out[3]: Index(['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',
              'Self_Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
              'Loan_Amount_Term', 'Credit_History', 'Property_Area'],
              dtype='object')
```

```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 367 entries, 0 to 366
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Loan_ID               367 non-null    object
1   Gender                356 non-null    object
2   Married               367 non-null    object
3   Dependents            357 non-null    object
4   Education             367 non-null    object
5   Self_Employed         344 non-null    object
```

```
6 ApplicantIncome 367 non-null int64
7 CoapplicantIncome 367 non-null int64
8 LoanAmount 362 non-null float64
9 Loan_Amount_Term 361 non-null float64
10 Credit_History 338 non-null float64
11 Property_Area 367 non-null object
dtypes: float64(3), int64(2), object(7)
memory usage: 34.5+ KB
```

```
In [5]: df['Property_Area'].value_counts()
```

Out[5]: Urban 140
Semiurban 116
Rural 111
Name: Property_Area, dtype: int64

```
In [6]: x=df[['ApplicantIncome', 'CoapplicantIncome']]
y=df['Property_Area']
```

```
In [7]: g1={"Property_Area":{"Urban":1,'Semiurban':2,'Rural':3,}}
df=df.replace(g1)
print(df)
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	\
0	LP001015	Male	Yes	0	Graduate	No	
1	LP001022	Male	Yes	1	Graduate	No	
2	LP001031	Male	Yes	2	Graduate	No	
3	LP001035	Male	Yes	2	Graduate	No	
4	LP001051	Male	No	0	Not Graduate	No	
..	
362	LP002971	Male	Yes	3+	Not Graduate	Yes	
363	LP002975	Male	Yes	0	Graduate	No	
364	LP002980	Male	No	0	Graduate	No	
365	LP002986	Male	Yes	0	Graduate	No	
366	LP002989	Male	No	0	Graduate	Yes	

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	\
0	5720	0	110.0	360.0	
1	3076	1500	126.0	360.0	
2	5000	1800	208.0	360.0	
3	2340	2546	100.0	360.0	
4	3276	0	78.0	360.0	
..	
362	4009	1777	113.0	360.0	
363	4158	709	115.0	360.0	
364	3250	1993	126.0	360.0	
365	5000	2393	158.0	360.0	
366	9200	0	98.0	180.0	

	Credit_History	Property_Area
0	1.0	1
1	1.0	1
2	1.0	1
3	NaN	1
4	1.0	1
..
362	1.0	1
363	1.0	1
364	NaN	2
365	1.0	3
366	1.0	3

[367 rows x 12 columns]

```
In [8]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.70)
```

```
In [9]: from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(x_train,y_train)
```

Out[9]: RandomForestClassifier()

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

```
In [11]: from sklearn.model_selection import GridSearchCV
grid_search=GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring="accuracy")
grid_search.fit(x_train,y_train)
```

Out[11]: 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 [12]: grid_search.best_score_
```

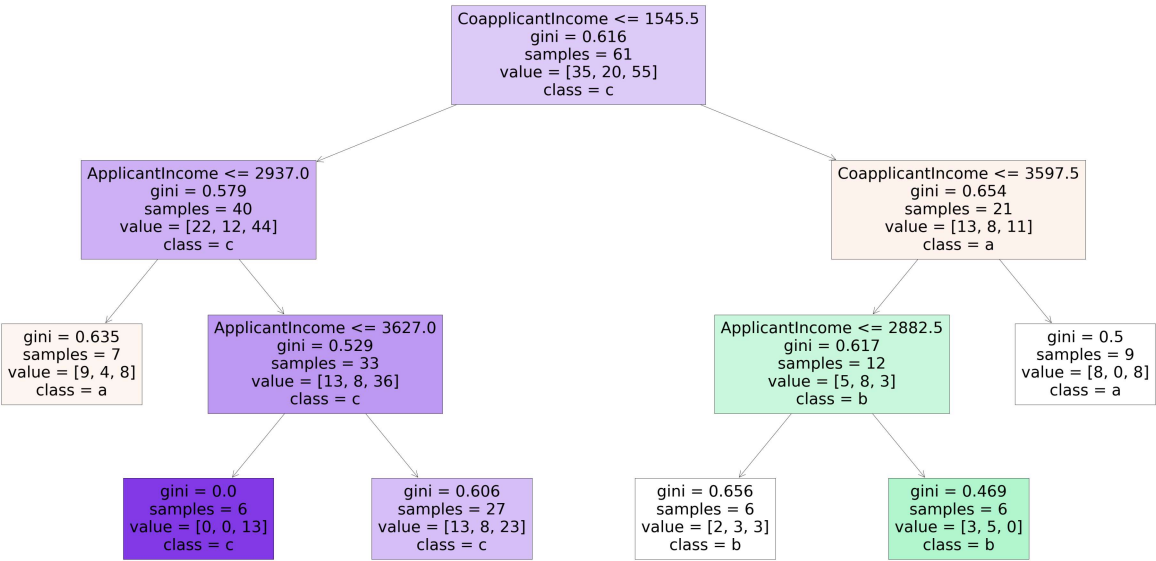
Out[12]: 0.4818181818181818

```
In [13]: rfc_best=grid_search.best_estimator_
```

```
In [14]: from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rfc_best.estimators_[5],feature_names=x.columns,class_names=['a','b','c'],fil
```

Out[14]: [Text(2232.0, 1902.6000000000001, 'CoapplicantIncome <= 1545.5\ngini = 0.616\nsamples = 61\nvalue = [35, 20, 55]\nnclass = c'),
Text(892.8, 1359.0, 'ApplicantIncome <= 2937.0\ngini = 0.579\nsamples = 40\nvalue = [2 2, 12, 44]\nnclass = c'),
Text(446.4, 815.4000000000001, 'gini = 0.635\nsamples = 7\nvalue = [9, 4, 8]\nnclass = a'),
Text(1339.1999999999998, 815.4000000000001, 'ApplicantIncome <= 3627.0\ngini = 0.529\nsamples = 33\nvalue = [13, 8, 36]\nnclass = c'),
Text(892.8, 271.79999999999995, 'gini = 0.0\nsamples = 6\nvalue = [0, 0, 13]\nnclass = c'),
Text(1785.6, 271.79999999999995, 'gini = 0.606\nsamples = 27\nvalue = [13, 8, 23]\nnclass = c'),
Text(3571.2, 1359.0, 'CoapplicantIncome <= 3597.5\ngini = 0.654\nsamples = 21\nvalue = [13, 8, 11]\nnclass = a'),
Text(3124.7999999999997, 815.4000000000001, 'ApplicantIncome <= 2882.5\ngini = 0.617\nns

```
amples = 12\nvalue = [5, 8, 3]\nnclass = b'),  
Text(2678.3999999999996, 271.79999999999995, 'gini = 0.656\nsamples = 6\nvalue = [2, 3,  
3]\nnclass = b'),  
Text(3571.2, 271.79999999999995, 'gini = 0.469\nsamples = 6\nvalue = [3, 5, 0]\nnclass =  
b'),  
Text(4017.6, 815.4000000000001, 'gini = 0.5\nsamples = 9\nvalue = [8, 0, 8]\nnclass =  
a')]]
```



```
In [ ]:
```