```
In [1]: import numpy as np
   import pandas as pd
   import seaborn as sns

In [2]:

In [3]: df_train=pd.read_csv("C8_loan-train.csv")
   df_test=pd.read_csv("C8_loan-test.csv")
```

Out[3]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coap
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 [4]:

Out[4]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coap
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	

```
In [5]:
        <class 'pandas.core.frame.DataFrame'>
        Int64Index: 480 entries, 1 to 613
        Data columns (total 13 columns):
            Column
                             Non-Null Count Dtype
            ----
                             -----
         0
            Loan ID
                             480 non-null
                                           object
         1
            Gender
                             480 non-null
                                           object
         2
            Married
                             480 non-null
                                           object
         3
            Dependents
                             480 non-null
                                           object
         4
            Education
                             480 non-null
                                           object
         5
            Self Employed
                             480 non-null
                                           object
         6
            ApplicantIncome
                             480 non-null
                                           int64
         7
            CoapplicantIncome
                             480 non-null
                                           float64
         8
            LoanAmount
                             480 non-null
                                           float64
         9
            Loan_Amount_Term
                             480 non-null
                                           float64
         10 Credit_History
                             480 non-null
                                           float64
         11 Property_Area
                             480 non-null
                                           object
         12 Loan Status
                             480 non-null
                                           object
        dtypes: float64(4), int64(1), object(8)
        memory usage: 52.5+ KB
In [8]:
Out[8]: Index(['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',
               'Self_Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
              'Loan_Amount_Term', 'Credit_History', 'Property_Area', 'Loan_Status'],
             dtype='object')
In [9]: feature_matrix=df1[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
              'Loan_Amount_Term', 'Credit_History']]
In [10]:
Out[10]: (480, 5)
Out[11]: (480, 1)
In [12]:
```

```
logr=LogisticRegression()
In [14]:
         C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py:63: Da
         taConversionWarning: A column-vector y was passed when a 1d array was expecte
         d. Please change the shape of y to (n_samples, ), for example using ravel().
           return f(*args, **kwargs)
Out[14]: LogisticRegression()
In [15]:
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 289 entries, 0 to 366
         Data columns (total 12 columns):
              Column
                                 Non-Null Count Dtype
              ----
                                 -----
                                                 ----
                                 289 non-null
          0
              Loan_ID
                                                 object
          1
              Gender
                                 289 non-null
                                                 object
          2
              Married
                                 289 non-null
                                                 object
          3
              Dependents
                                 289 non-null
                                                 object
          4
              Education
                                 289 non-null
                                                 object
          5
              Self_Employed
                                 289 non-null
                                                 object
          6
              ApplicantIncome
                                 289 non-null
                                                 int64
          7
              CoapplicantIncome
                                 289 non-null
                                                 int64
          8
              LoanAmount
                                 289 non-null
                                                 float64
          9
              Loan Amount Term
                                 289 non-null
                                                 float64
          10 Credit_History
                                 289 non-null
                                                 float64
          11 Property Area
                                 289 non-null
                                                 object
         dtypes: float64(3), int64(2), object(7)
         memory usage: 29.4+ KB
In [16]:
Out[16]: Index(['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',
                'Self_Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
                'Loan_Amount_Term', 'Credit_History', 'Property_Area'],
               dtype='object')
In [17]: observation=df2[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
```

```
In [18]:
                  prediction=logr.predict(observation)
                   ['Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'No' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
                     'Yes']
In [19]: 🖵
Out[19]: array(['No', 'Yes'], dtype=object)
In [20]: ___
Out[20]: array([[0.
                                                      , 1.
                                                                               ],
                                                      , 1.
                                 [0.
                                                                               ],
                                                      , 1.
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                                 [0.
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                                 [0.
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                                 [0.
                                                      , 1.
```

```
In [21]: 0.0
```

RANDOM FOREST

```
Out[22]:
         No
                 257
          Name: Self_Employed, dtype: int64
In [23]: | x=df2[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
                 'Loan_Amount_Term', 'Credit_History']]
In [24]: |g1={'Self_Employed':{'No':1, 'Yes':2}}
         df2=df2.replace(g1)
Out[24]:
                Loan_ID Gender Married Dependents Education Self_Employed ApplicantIncome Coap
            0 LP001015
                          Male
                                   Yes
                                               0
                                                   Graduate
                                                                       1
                                                                                   5720
            1 LP001022
                          Male
                                   Yes
                                                1
                                                   Graduate
                                                                       1
                                                                                   3076
              LP001031
                          Male
                                   Yes
                                                2
                                                   Graduate
                                                                                   5000
                                                       Not
            4 LP001051
                                                0
                                                                                   3276
                          Male
                                   No
                                                   Graduate
                                                        Not
              LP001054
                                                0
                                                                       2
                                                                                   2165
                          Male
                                   Yes
                                                   Graduate
           361 LP002969
                          Male
                                   Yes
                                                   Graduate
                                                                                   2269
                                               1
                                                       Not
           362 LP002971
                          Male
                                   Yes
                                               3+
                                                                                   4009
                                                   Graduate
           363 LP002975
                          Male
                                   Yes
                                               0
                                                   Graduate
                                                                                   4158
                                               0
           365 LP002986
                          Male
                                   Yes
                                                   Graduate
                                                                                   5000
           366 LP002989
                          Male
                                   No
                                                   Graduate
                                                                                   9200
          289 rows × 12 columns
In [25]:
               11 11 12 15 161 161
In [27]:
In [28]: rfc=RandomForestClassifier()
rfc.fit(x_train,y_train)
Out[28]: RandomForestClassifier()
```

```
In [29]: |parameters={'max_depth':[1,2,3,4,5],
                    'min_samples_leaf':[5,10,15,20,25],
                     'n_estimators':[10,20,30,40,50]
In [30]: from sklearn.model_selection import GridSearchCV
         grid_search =GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring="ac
Out[30]: 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 [31]:
Out[31]: 0.9108910891089109
In [32]:
In [33]: from sklearn.tree import plot_tree
         plt.figure(figsize=(80,40))
Out[33]: [Text(2232.0, 1630.8000000000000, 'Credit_History <= 0.5\ngini = 0.085\nsampl</pre>
         es = 125\nvalue = [193, 9]\nclass = a'),
         Text(1116.0, 543.59999999999, 'gini = 0.111\nsamples = 20\nvalue = [32,
         2] \nclass = a'),
          Text(3348.0, 543.59999999999, 'gini = 0.08\nsamples = 105\nvalue = [161,
         7] \cdot (ass = a')
                               Credit_History <= 0.5
                                     gini = 0.085
                                   samples = 125
                                   value = [193, 9]
                                       class = a
                   gini = 0.111
                                                         gini = 0.08
                  samples = 20
                                                      samples = 105
```