airk8vzg6

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```
[2]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
     from sklearn.linear_model import LogisticRegression
     from sklearn.preprocessing import StandardScaler
[3]: from google.colab import drive
     drive.mount('/content/drive')
    Mounted at /content/drive
[4]: df_train=pd.read_csv("/content/drive/MyDrive/mydatasets/C8_loan-train.csv")
     df_test=pd.read_csv("/content/drive/MyDrive/mydatasets/C8_loan-test.csv")
[5]: df_train.dropna(inplace=True)
     df_test.dropna(inplace=True)
[6]: df_train.info()
    <class 'pandas.core.frame.DataFrame'>
    Int64Index: 480 entries, 1 to 613
    Data columns (total 13 columns):
         Column
                            Non-Null Count Dtype
         Loan_ID
     0
                            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
         CoapplicantIncome 480 non-null
     7
                                            float64
     8
        LoanAmount
                            480 non-null
                                            float64
         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
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dtypes: float64(4), int64(1), object(8)
     memory usage: 52.5+ KB
[7]: df train.columns
[7]: Index(['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',
             'Self_Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
             'Loan_Amount_Term', 'Credit_History', 'Property_Area', 'Loan_Status'],
            dtype='object')
[8]: feature matrix = df_train[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
             'Loan_Amount_Term', 'Credit_History']]
     target_vector = df_train[['Self_Employed']]
[9]: fs = StandardScaler().fit_transform(feature_matrix)
     logr = LogisticRegression()
     logr.fit(fs,target_vector)
     /usr/local/lib/python3.10/dist-packages/sklearn/utils/validation.py:1143:
     DataConversionWarning: A column-vector y was passed when a 1d array was
     expected. Please change the shape of y to (n_samples, ), for example using
     ravel().
       y = column_or_1d(y, warn=True)
[9]: LogisticRegression()
[10]: observation = df test[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
             'Loan_Amount_Term', 'Credit_History']]
     prediction = logr.predict(observation)
     print(prediction)
     ['Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' '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'
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'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes' 'Yes'
      'Yes'l
     /usr/local/lib/python3.10/dist-packages/sklearn/base.py:432: UserWarning: X has
     feature names, but LogisticRegression was fitted without feature names
       warnings.warn(
[11]: logr.classes_
[11]: array(['No', 'Yes'], dtype=object)
[12]: logr.predict_proba(observation)[0][0]
     /usr/local/lib/python3.10/dist-packages/sklearn/base.py:432: UserWarning: X has
     feature names, but LogisticRegression was fitted without feature names
       warnings.warn(
[12]: 0.0
     Random Forest
[22]: df2=pd.read_csv("/content/drive/MyDrive/mydatasets/C8_loan-test.csv")
      df2.dropna(inplace=True)
      df2['Self_Employed'].value_counts()
[22]: No
             257
      Yes
              32
      Name: Self_Employed, dtype: int64
[23]: x=df2[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
             'Loan_Amount_Term', 'Credit_History']]
      y=df2['Self_Employed']
[24]: g1={'Self_Employed':{'No':1, 'Yes':2}}
      df2=df2.replace(g1)
      df2
[24]:
            Loan_ID Gender Married Dependents
                                                   Education
                                                              Self_Employed
      0
           LP001015
                      Male
                               Yes
                                            0
                                                   Graduate
                                                                          1
           LP001022
                      Male
                               Yes
                                            1
                                                   Graduate
                                                                          1
      1
      2
                                                   Graduate
           LP001031
                      Male
                               Yes
                                            2
                                                                          1
      4
           LP001051
                      Male
                               No
                                            0 Not Graduate
                                                                          1
      5
           LP001054
                      Male
                               Yes
                                            0 Not Graduate
                                                                          2
```

```
361 LP002969
                       Male
                                 Yes
                                               1
                                                      Graduate
                                                                              1
      362 LP002971
                       Male
                                 Yes
                                              3+
                                                  Not Graduate
                                                                              2
      363 LP002975
                       Male
                                               0
                                                      Graduate
                                 Yes
                                                                              1
      365
          LP002986
                       Male
                                 Yes
                                               0
                                                      Graduate
                                                                              1
          LP002989
                                                      Graduate
                                                                              2
      366
                       Male
                                  No
                                               0
           ApplicantIncome
                             CoapplicantIncome
                                                 LoanAmount
                                                              Loan_Amount_Term
      0
                       5720
                                                       110.0
                                                                           360.0
      1
                       3076
                                           1500
                                                       126.0
                                                                           360.0
      2
                       5000
                                           1800
                                                       208.0
                                                                           360.0
      4
                       3276
                                               0
                                                        78.0
                                                                           360.0
      5
                       2165
                                           3422
                                                       152.0
                                                                           360.0
                                           2167
                                                        99.0
                                                                           360.0
      361
                       2269
      362
                       4009
                                                                           360.0
                                           1777
                                                       113.0
      363
                                            709
                                                                           360.0
                       4158
                                                       115.0
      365
                       5000
                                           2393
                                                       158.0
                                                                           360.0
      366
                                                        98.0
                                                                           180.0
                       9200
                                               0
           Credit_History Property_Area
      0
                       1.0
                                    Urban
      1
                       1.0
                                    Urban
      2
                       1.0
                                    Urban
      4
                       1.0
                                    Urban
      5
                       1.0
                                    Urban
      . .
      361
                       1.0
                                Semiurban
                                    Urban
      362
                       1.0
      363
                       1.0
                                    Urban
      365
                       1.0
                                    Rural
      366
                       1.0
                                    Rural
      [289 rows x 12 columns]
[25]: from sklearn.model_selection import train_test_split
[26]: x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
[27]: from sklearn.ensemble import RandomForestClassifier
[28]: rfc=RandomForestClassifier()
      rfc.fit(x_train,y_train)
```

[28]: RandomForestClassifier()

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[29]: parameters={'max_depth':[1,2,3,4,5],
                  'min_samples_leaf':[5,10,15,20,25],
                  'n_estimators': [10,20,30,40,50]
      }
[30]: 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)
[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')
[31]: grid_search.best_score_
[31]: 0.8762376237623762
[32]: rfc_best=grid_search.best_estimator_
[33]: from sklearn.tree import plot_tree
      plt.figure(figsize=(80,40))
      plot_tree(rfc_best.estimators_[5],feature_names=x.
       [33]: [Text(0.5, 0.75, 'LoanAmount <= 85.0 \ngini = 0.253 \nsamples = 125 \nvalue = [172, 172]
      30] \nclass = a'),
       Text(0.25, 0.25, 'gini = 0.0 \setminus samples = 16 \setminus ue = [24, 0] \setminus ue = a'),
       Text(0.75, 0.25, 'gini = 0.28 \setminus samples = 109 \setminus value = [148, 30] \setminus class = a')]
```

LoanAmount \leq 85.0 gini = 0.253 samples = 125 value = [172, 30] class = a

gini = 0.0 samples = 16 value = [24, 0] class = a gini = 0.28 samples = 109 value = [148, 30] class = a