

## ClassificationRegaression

```
In [24]:
        import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn.model selection import train test split
         from sklearn.preprocessing import StandardScaler,MinMaxScaler
         from sklearn.preprocessing import OneHotEncoder,LabelEncoder
         from sklearn.linear model import LogisticRegression
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.naive bayes import MultinomialNB
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.metrics import accuracy score,classification report
        df=pd.read csv("C:\\Users\\hp\\Downloads\\bank loan.csv")
In [5]:
         df.head()
             Loan ID Gender Married Dependents Education Self Employed Applicant
Out[5]:
         0 LP001002
                         Male
                                                  0
                                                       Graduate
                                    Nο
                                                                            Nο
         1 LP001003
                         Male
                                    Yes
                                                       Graduate
                                                                            No
         2 LP001005
                         Male
                                    Yes
                                                       Graduate
                                                                            Yes
         3 LP001006
                         Male
                                    Yes
                                                                            No
                                                       Graduate
         4 LP001008
                         Male
                                                       Graduate
                                                                            No
                                    No
In [7]:
        df.isnull().sum()
                               0
Out[7]: Loan ID
         Gender
                              13
                               3
         Married
         Dependents
                              15
         Education
                               0
                              32
         Self Employed
         ApplicantIncome
                               0
         CoapplicantIncome
                               0
         LoanAmount
                              22
         Loan Amount Term
                               14
         Credit History
                              50
                               0
         Property Area
         Loan Status
         dtype: int64
In [9]:
        df.shape
Out[9]: (614, 13)
```

```
In [51]: df.columns
Out[51]: Index(['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',
                'Self Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
                 'Loan_Amount_Term', 'Credit_History', 'Property_Area', 'Loan_Status'],
               dtype='object')
         df.fillna(df.median(numeric only=True), inplace=True)
In [11]:
In [13]: df.isnull().sum()
Out[13]: Loan ID
                                0
         Gender
                               13
         Married
                                3
                               15
         Dependents
         Education
                                0
         Self Employed
                               32
         ApplicantIncome
                                0
         CoapplicantIncome
                                0
         LoanAmount
                                0
         Loan Amount_Term
                                0
         Credit History
                                0
         Property Area
                                0
         Loan Status
                                0
         dtype: int64
In [15]: df.fillna(df['Gender'].mode()[0],inplace=True)
In [17]: df.isnull().sum()
Out[17]: Loan ID
                               0
                               0
         Gender
         Married
                               0
         Dependents
                               0
                               0
         Education
         Self Employed
                               0
         ApplicantIncome
                               0
         CoapplicantIncome
                               0
         LoanAmount
                               0
                               0
         Loan Amount Term
         Credit History
                               0
                               0
         Property Area
         Loan Status
                               0
         dtype: int64
In [19]: for col in df.select dtypes(include=[object]).columns:
             df[col]=LabelEncoder().fit transform(df[col])
In [21]: X=df.drop('Loan Status',axis=1)
         y=df[['Loan_Status']]
In [25]: X train,X test,y train,y test=train test split(X,y, test size=0.2,random state
```

```
In [54]: model2 = LogisticRegression(max_iter=1000)
model2.fit(X_train,y_train)
```

C:\ProgramData\anaconda3\Lib\site-packages\sklearn\linear\_model\\_logistic.py:47
3: ConvergenceWarning: lbfgs failed to converge after 1000 iteration(s) (statu s=1):

STOP: TOTAL NO. of ITERATIONS REACHED LIMIT

Increase the number of iterations to improve the convergence (max\_iter=1000). You might also want to scale the data as shown in:

https://scikit-learn.org/stable/modules/preprocessing.html

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear\_model.html#logistic-regressi
on

n\_iter\_i = \_check\_optimize\_result(

#### Out[54]:

•	LogisticRegress	ion 🕕 🕜			
Pa	Parameters				
<u>.</u>	penalty	'12'			
<u>.</u>	dual	False			
<u>.</u>	tol	0.0001			
٠	С	1.0			
٠	fit_intercept	True			
<b>.</b>	intercept_scaling	1			
<u>.</u>	class_weight	None			
<u>.</u>	random_state	None			
٩	solver	'lbfgs'			
٠	max_iter	1000			
٠	multi_class	'deprecated'			
٠	verbose	0			
٠	warm_start	False			
٠	n_jobs	None			
<u>.</u>	l1_ratio	None			

# In [55]: # 5. Evaluate performance y pred=model2.predict(X test)

# In [56]: accuracy=accuracy\_score(y\_test,y\_pred) accuracy

### DecisionTreeClassifier

In [57]: #
 model3=DecisionTreeClassifier(max\_depth=4, criterion='entropy',random\_state=42
 model3.fit(X\_train,y\_train)

Out[57]:

•	DecisionTreeClassifier 🕒 🕐				
Parameters					
<b>.</b>	criterion	'entropy'			
٠	splitter	'best'			
<b>.</b>	max_depth	4			
٠	min_samples_split	2			
٠	min_samples_leaf	1			
٠	min_weight_fraction_leaf	0.0			
٠	max_features	None			
٠	random_state	42			
٠	max_leaf_nodes	None			
٠	min_impurity_decrease	0.0			
٠	class_weight	None			
٠	ccp_alpha	0.0			
٠	monotonic_cst	None			

In [46]: print(classification\_report(y\_test,y\_pred))

```
In [58]: y_pred=model3.predict(X_test)
In [59]: accuracy=accuracy_score(y_test,y_pred)
accuracy
Out[59]: 0.7886178861788617
```

	precision	recall	f1-score	support
0 1	0.95 0.76	0.42 0.99	0.58 0.86	43 80
accuracy macro avg weighted avg	0.85 0.83	0.70 0.79	0.79 0.72 0.76	123 123 123

## Random Forest Classifier

```
In [60]: model4=RandomForestClassifier(n_estimators=100,random_state=42)
model4.fit(X_train,y_train)
```

#### Out[60]:

•	RandomForestClassifier			
Parameters				
<b>.</b>	n_estimators	100		
<b>.</b>	criterion	'gini'		
<b>.</b>	max_depth	None		
<b>.</b>	min_samples_split	2		
<b>.</b>	min_samples_leaf	1		
<u>.</u>	min_weight_fraction_leaf	0.0		
<b>.</b>	max_features	'sqrt'		
<b>.</b>	max_leaf_nodes	None		
<b>.</b>	min_impurity_decrease	0.0		
<b>.</b>	bootstrap	True		
<b>.</b>	oob_score	False		
<u>.</u>	n_jobs	None		
<b>.</b>	random_state	42		
<b>.</b>	verbose	0		
<b>.</b>	warm_start	False		
<b>.</b>	class_weight	None		
<b>.</b>	ccp_alpha	0.0		
	max_samples	None		
<b>.</b>	monotonic_cst	None		

In [50]: print(classification\_report(y\_test,y\_pred))

```
In [61]: y_pred=model4.predict(X_test)
In [62]: accuracy=accuracy_score(y_test,y_pred)
accuracy
Out[62]: 0.7804878048780488
```

```
precision
                       recall f1-score support
                  0.95
                            0.42
                                     0.58
                                                 43
          0
                  0.76
                            0.99
                                                 80
          1
                                     0.86
   accuracy
                                     0.79
                                                123
                  0.85
                            0.70
                                     0.72
                                                123
  macro avg
weighted avg
                  0.83
                            0.79
                                     0.76
                                                123
```

```
In [63]: # 7. Save model
  joblib.dump(model2, "bank_loan_model.pkl")
  print(" Model saved as bank_loan_model.pkl")
```

✓ Model saved as bank\_loan\_model.pkl