```
In [1]:
          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
In [2]:
          df_train=pd.read_csv("loan-test.csv").dropna()
          df_test=pd.read_csv("loan-train.csv").dropna()
          df train
Out[2]:
               Loan_ID Gender Married Dependents Education Self_Employed ApplicantIncome CoapplicantIn
           0 LP001015
                          Male
                                    Yes
                                                  0
                                                      Graduate
                                                                          No
                                                                                         5720
           1 LP001022
                          Male
                                    Yes
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                                                      Graduate
                                                                          No
                                                                                         3076
           2 LP001031
                                    Yes
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                                                      Graduate
                                                                          No
                                                                                         5000
                          Male
                                                           Not
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                                                                                         3276
           4 LP001051
                          Male
                                    No
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                                                      Graduate
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           5 LP001054
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                          Male
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                                                                                         2165
                                                      Graduate
         361 LP002969
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                          Male
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         362 LP002971
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                                                                                         4009
                          Male
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                                                      Graduate
         363 LP002975
                          Male
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                                                                                         4158
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                                                      Graduate
                                                                                         5000
         365 LP002986
                          Male
                                    Yes
                                                                          No
                                                                                         9200
         366 LP002989
                          Male
                                    No
                                                      Graduate
                                                                          Yes
        289 rows × 12 columns
In [3]:
          df test
Out[3]:
               Loan ID Gender Married Dependents Education Self Employed ApplicantIncome CoapplicantIII
           1 LP001003
                          Male
                                    Yes
                                                      Graduate
                                                                          No
                                                                                         4583
                                                      Graduate
                                                                                         3000
           2 LP001005
                          Male
                                    Yes
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           3 LP001006
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                                                      Graduate
              LP001008
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                                                      Graduate
                                                                                         6000
                          Male
                                    No
                                                                          No
             LP001011
                          Male
                                    Yes
                                                  2
                                                      Graduate
                                                                          Yes
                                                                                         5417
```

		Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coapplicantl
	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	
	480 r	ows × 13 c	columns						
In [4]:	<pre>df_train.dropna(inplace=True) df_test.dropna(inplace=True)</pre>								
In [5]:	df_	train.inf	0()						
	<pre><class 'pandas.core.frame.dataframe'=""> Int64Index: 289 entries, 0 to 366 Data columns (total 12 columns): # Column Non-Null Count Dtype</class></pre>								
	0	Loan_ID		289	non-null	object			
	1	Gender			non-null	object			
	2 3	Married Dependen	its		non-null non-null	object object			
	4	Education			non-null	object			
	5	Self_Emp			non-null	object			
	6 7	Applicar Coapplic			non-null non-null	int64 int64			
	8	LoanAmou			non-null	float64			
	9	Loan_Amo			non-null	float64			
	10	Credit_H			non-null	float64			
	dtyp	Property es: float ry usage:	64(3),	int64(2)	non-null , object(7)	object			
In [6]:	<pre>feature_matrix = df_train[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount', 'Loan_ target_vector = df_train[['Gender']]</pre>								
In [7]:	<pre>fs = StandardScaler().fit_transform(feature_matrix) logr = LogisticRegression() logr.fit(fs,target_vector) C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py:63: 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(). return f(*args, **kwargs)</pre>								
Out[7]:	Logi	sticRegre	ssion()						
In [8]:	df_	test.info	()						

```
<class 'pandas.core.frame.DataFrame'>
        Int64Index: 480 entries, 1 to 613
        Data columns (total 13 columns):
         #
                                 Non-Null Count
              Column
                                                  Dtype
              Loan ID
         0
                                 480 non-null
                                                  object
              Gender
                                 480 non-null
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         2
             Married
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              Dependents
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              Education
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              ApplicantIncome
                                 480 non-null
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              CoapplicantIncome
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                                                  float64
         8
                                 480 non-null
                                                  float64
              LoanAmount
         9
              Loan Amount Term
                                 480 non-null
                                                  float64
         10
             Credit_History
                                 480 non-null
                                                  float64
             Property_Area
                                 480 non-null
                                                  object
         11
             Loan Status
         12
                                 480 non-null
                                                  object
        dtypes: float64(4), int64(1), object(8)
        memory usage: 52.5+ KB
In [9]:
         observation = df test[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount', 'Loan Amo
         prediction = logr.predict(observation)
         print(prediction)
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Out[11]: 0.0

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In [10]:
         logr.classes_
Out[10]: array(['Female', 'Male'], dtype=object)
In [11]:
         logr.predict_proba(observation)[0][0]
```