

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: from sklearn.linear_model import LogisticRegression
```

```
In [3]: df=pd.read_csv("C8 loan csv").dropna()

df
```

Out[3]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Co
0	LP001015	Male	Yes	0	Graduate	No	5720	
1	LP001022	Male	Yes	1	Graduate	No	3076	
2	LP001031	Male	Yes	2	Graduate	No	5000	
4	LP001051	Male	No	0	Not Graduate	No	3276	
5	LP001054	Male	Yes	0	Not Graduate	Yes	2165	
...	...	...	...	...	...	...	...	
361	LP002969	Male	Yes	1	Graduate	No	2269	
362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	
363	LP002975	Male	Yes	0	Graduate	No	4158	

```
In [4]: df.dropna(inplace=True)
```

```
In [5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 289 entries, 0 to 366
Data columns (total 12 columns):
 #   Column                Non-Null Count  Dtype  
---  -
 0   Loan_ID               289 non-null   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 [6]: feature_matrix = df[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount', 'Loan_Amount_Term']]
target_vector = df['Property_Area']
```

```
In [7]: feature_matrix.shape
```

```
Out[7]: (289, 5)
```

```
In [8]: target_vector.shape
```

```
Out[8]: (289,)
```

```
In [9]: from sklearn.preprocessing import StandardScaler
```

```
In [10]: fs = StandardScaler().fit_transform(feature_matrix)
```

```
In [11]: logr = LogisticRegression()
logr.fit(fs, target_vector)
```

```
Out[11]: LogisticRegression()
```

```
In [12]: feature_matrix.shape
```

```
Out[12]: (289, 5)
```

```
In [13]: target_vector.shape
```

```
Out[13]: (289,)
```

```
In [14]: from sklearn.preprocessing import StandardScaler
```

```
In [15]: fs = StandardScaler().fit_transform(feature_matrix)
```

```
In [16]: logr = LogisticRegression()  
logr.fit(fs, target_vector)
```

```
Out[16]: LogisticRegression()
```

```
In [17]: observation=df[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount', 'Loan_Amount_Repaid']]
```

```
In [18]: prediction = logr.predict(observation)  
prediction
```

```
Out[18]: array(['Urban', 'Urban', 'Urban', 'Urban', 'Rural', 'Urban', 'Urban',  
                'Rural', 'Urban', 'Urban', 'Urban', 'Rural', 'Urban', 'Rural',  
                'Rural', 'Rural', 'Urban', 'Urban', 'Rural', 'Rural', 'Urban',  
                'Rural', 'Urban', 'Urban', 'Urban', 'Rural', 'Urban', 'Urban',  
                'Rural', 'Urban', 'Urban', 'Urban', 'Urban', 'Rural', 'Urban',  
                'Urban', 'Rural', 'Urban', 'Rural', 'Rural', 'Urban', 'Urban',  
                'Urban', 'Urban', 'Rural', 'Rural', 'Urban', 'Rural', 'Urban',  
                'Urban', 'Rural', 'Urban', 'Rural', 'Urban', 'Urban', 'Urban',  
                'Rural', 'Urban', 'Urban', 'Rural', 'Rural', 'Rural', 'Urban',  
                'Urban', 'Urban', 'Urban', 'Urban', 'Urban', 'Urban', 'Urban',  
                'Urban', 'Urban', 'Urban', 'Urban', 'Urban', 'Urban', 'Rural',  
                'Urban', 'Urban', 'Rural', 'Urban', 'Rural', 'Urban', 'Urban',  
                'Urban', 'Rural', 'Rural', 'Urban', 'Urban', 'Urban', 'Urban',  
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                'Urban', 'Urban', 'Urban', 'Urban', 'Urban', 'Urban', 'Rural',  
                'Urban', 'Rural', 'Urban', 'Urban', 'Urban', 'Urban', 'Urban',  
                'Urban', 'Rural', 'Urban', 'Urban', 'Rural', 'Urban', 'Rural',  
                'Rural', 'Urban', 'Urban', 'Urban', 'Rural', 'Urban', 'Urban',  
                ...])
```

```
In [19]: logr.classes_
```

```
Out[19]: array(['Rural', 'Semiurban', 'Urban'], dtype=object)
```

```
In [20]: logr.predict_proba(observation)[0][1]
```

```
Out[20]: 0.0
```