In [1]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

In [2]:

```
df = pd.read_csv('LoanData.csv')
df.head()
```

Out[2]:

| | Loan_ID | Gender | Married | Dependents | Education | Self_Employed | ApplicantIncome | Coa |
|---|----------|--------|---------|------------|-----------------|---------------|-----------------|-----|
| 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 | |
| 4 | | | | | | | | • |

In [3]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 614 entries, 0 to 613
Data columns (total 13 columns):
```

| # | Column | Non-Null Count | Dtype |
|----|-------------------|----------------|---------|
| | | | |
| 0 | Loan_ID | 614 non-null | object |
| 1 | Gender | 601 non-null | object |
| 2 | Married | 611 non-null | object |
| 3 | Dependents | 599 non-null | object |
| 4 | Education | 614 non-null | object |
| 5 | Self_Employed | 582 non-null | object |
| 6 | ApplicantIncome | 614 non-null | int64 |
| 7 | CoapplicantIncome | 614 non-null | float64 |
| 8 | LoanAmount | 592 non-null | float64 |
| 9 | Loan_Amount_Term | 600 non-null | float64 |
| 10 | Credit_History | 564 non-null | float64 |
| 11 | Property_Area | 614 non-null | object |
| 12 | Loan_Status | 614 non-null | object |

dtypes: float64(4), int64(1), object(8)

memory usage: 62.5+ KB

```
In [4]:
df.shape
Out[4]:
(614, 13)
In [5]:
df.isnull().sum()
Out[5]:
Loan_ID
                       0
Gender
                      13
Married
                       3
                      15
Dependents
Education
                       0
Self_Employed
                      32
ApplicantIncome
                       0
CoapplicantIncome
                       0
                      22
LoanAmount
Loan_Amount_Term
                      14
Credit_History
                      50
Property_Area
                       0
Loan_Status
                       0
dtype: int64
In [6]:
df['LoanAmount'] = df['LoanAmount'].fillna(df['LoanAmount'].mean())
In [7]:
df['Credit_History'] = df['Credit_History'].fillna(df['Credit_History'].median())
In [8]:
df.isnull().sum()
Out[8]:
Loan_ID
                       0
Gender
                      13
Married
                       3
                      15
Dependents
Education
                       0
Self_Employed
                      32
                       0
ApplicantIncome
CoapplicantIncome
                       0
                       0
LoanAmount
                      14
Loan_Amount_Term
                       0
Credit_History
                       0
Property_Area
                       0
Loan_Status
dtype: int64
```

```
In [9]:
```

```
df.dropna(inplace=True)
```

In [10]:

```
df.isnull().sum()
```

Out[10]:

Loan_ID 0 Gender 0 Married 0 Dependents 0 Education 0 Self_Employed 0 ApplicantIncome 0 CoapplicantIncome 0 LoanAmount 0 Loan_Amount_Term 0 Credit_History 0 0 Property_Area Loan_Status 0 dtype: int64

In [11]:

```
df.shape
```

Out[11]:

(542, 13)

In [12]:

```
plt.figure(figsize = (100, 50))
sns.set(font_scale = 5)
plt.subplot(331)
sns.countplot(df['Gender'],hue=df['Loan_Status'])

plt.subplot(332)
sns.countplot(df['Married'],hue=df['Loan_Status'])

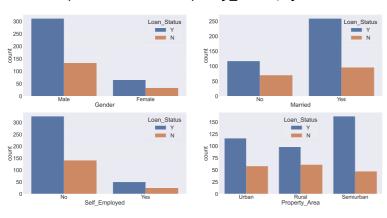
plt.subplot(333)
sns.countplot(df['Education'],hue=df['Loan_Status'])

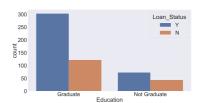
plt.subplot(334)
sns.countplot(df['Self_Employed'],hue=df['Loan_Status'])

plt.subplot(335)
sns.countplot(df['Property_Area'],hue=df['Loan_Status'])
```

Out[12]:

<AxesSubplot:xlabel='Property_Area', ylabel='count'>





In [13]:

```
df['Loan_Status'].replace('Y',1,inplace=True)
df['Loan_Status'].replace('N',0,inplace=True)
```

In [14]:

```
df['Loan_Status'].value_counts()
```

Out[14]:

376
 166

Name: Loan_Status, dtype: int64

In [15]:

```
df.Gender=df.Gender.map({'Male':1, 'Female':0})
df['Gender'].value_counts()
```

Out[15]:

444
 98

Name: Gender, dtype: int64

```
In [16]:
df.Married=df.Married.map({'Yes':1,'No':0})
df['Married'].value_counts()
Out[16]:
1
     355
0
     187
Name: Married, dtype: int64
In [17]:
df.Dependents=df.Dependents.map({'0':0,'1':1,'2':2,'3+':3})
df['Dependents'].value_counts()
Out[17]:
     309
0
1
      94
      94
2
3
      45
Name: Dependents, dtype: int64
In [18]:
df.Education=df.Education.map({'Graduate':1,'Not Graduate':0})
df['Education'].value_counts()
Out[18]:
1
     425
     117
Name: Education, dtype: int64
In [19]:
df.Self_Employed=df.Self_Employed.map({'Yes':1,'No':0})
df['Self_Employed'].value_counts()
Out[19]:
     467
0
1
      75
Name: Self_Employed, dtype: int64
In [20]:
df.Property_Area = df.Property_Area.map({'Urban':2, 'Rural':0, 'Semiurban':1})
df['Property_Area'].value_counts()
Out[20]:
     209
1
     174
2
     159
Name: Property_Area, dtype: int64
```

```
In [21]:
df['LoanAmount'].value_counts()
Out[21]:
146.412162
              19
120.000000
100.000000
              14
110.000000
              13
187.000000
              12
280.000000
               1
240.000000
               1
214.000000
               1
59.000000
               1
               1
253.000000
Name: LoanAmount, Length: 195, dtype: int64
In [22]:
df['Loan_Amount_Term'].value_counts()
Out[22]:
360.0
         464
180.0
          38
480.0
          13
300.0
          12
84.0
           4
           3
120.0
240.0
           3
60.0
           2
           2
36.0
12.0
           1
Name: Loan_Amount_Term, dtype: int64
In [23]:
df['Credit_History'].value_counts()
Out[23]:
       468
1.0
0.0
        74
Name: Credit_History, dtype: int64
```

```
In [24]:
df.head()
Out[24]:
    Loan_ID Gender Married Dependents Education Self_Employed ApplicantIncome Coa
 0 LP001002
                  1
                         0
                                    0
                                              1
                                                                        5849
 1 LP001003
                                    1
                                                            0
                                                                        4583
                 1
                         1
                                              1
 2 LP001005
                  1
                         1
                                    0
                                              1
                                                            1
                                                                        3000
 3 LP001006
                                                            0
                                                                        2583
                  1
                         1
                                    0
                                              0
 4 LP001008
                  1
                         0
                                    0
                                                            0
                                                                        6000
                                              1
In [46]:
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn import metrics
lr=LogisticRegression()
In [58]:
x = df.iloc[1:542,1:12].values
y = df.iloc[1:542,12].values
In [65]:
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2,random_state=40)
In [66]:
x_train.shape,x_test.shape,y_train.shape,y_test.shape
Out[66]:
((432, 11), (109, 11), (432,), (109,))
In [67]:
model=lr.fit(x_test,y_test)
lr.score(x_test,y_test)*100
Out[67]:
80.73394495412845
In [ ]:
```

| In []: | |
|---------|--|
| | |
| In []: | |
| | |
| In []: | |
| | |
| In []: | |
| | |