

Lovely Professional University

Phagwara, Panjab

School of Computer Applications



Project Topic

**Prediction of Loan Approval by using Machine Learning with Detailed Analysis,
Graphs/Figures, and Reports**

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```
import pandas as pd
df2 = pd.read_csv("ProjectPy2.csv")
df2.head(50)
```

```
#no of missing values
df2.isnull().sum()
```

Loan_ID	0
Gender	13
Married	3
Dependents	15
Education	0
Self_Employed	32
ApplicantIncome	0
CoapplicantIncome	0
LoanAmount	22
Loan_Amount_Term	14
Credit_History	50
Property_Area	0
Loan_Status	0
dtype:	int64

```
#type of data
type(df2)
```

```
pandas.core.frame.DataFrame
```

```
# no of rows and columns
df2.shape
```

```
(614, 13)
```

```
# statistical measurment  
df2.describe()
```

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History
count	614.000000	614.000000	592.000000	600.000000	564.000000
mean	5403.459283	1621.245798	146.412162	342.000000	0.842199
std	6109.041673	2926.248369	85.587325	65.12041	0.364878
min	150.000000	0.000000	9.000000	12.000000	0.000000
25%	2877.500000	0.000000	100.000000	360.000000	1.000000
50%	3812.500000	1188.500000	128.000000	360.000000	1.000000
75%	5795.000000	2297.250000	168.000000	360.000000	1.000000
max	81000.000000	41667.000000	700.000000	480.000000	1.000000

```
# drop a missing values  
df2 = df2.dropna()
```

```
df2.isnull().sum()
```

```
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  
Property_Area   0  
Loan_Status     0  
dtype: int64
```

```
In [9]: df2['Dependents'].value_counts()
```

```
Out[9]: 0      274  
        2       85  
        1       80  
        3+      41  
        Name: Dependents, dtype: int64
```

```
In [10]: df2 = df2.replace(to_replace='3+' , value = 4)
```

```
In [11]: df2['Dependents'].value_counts()
```

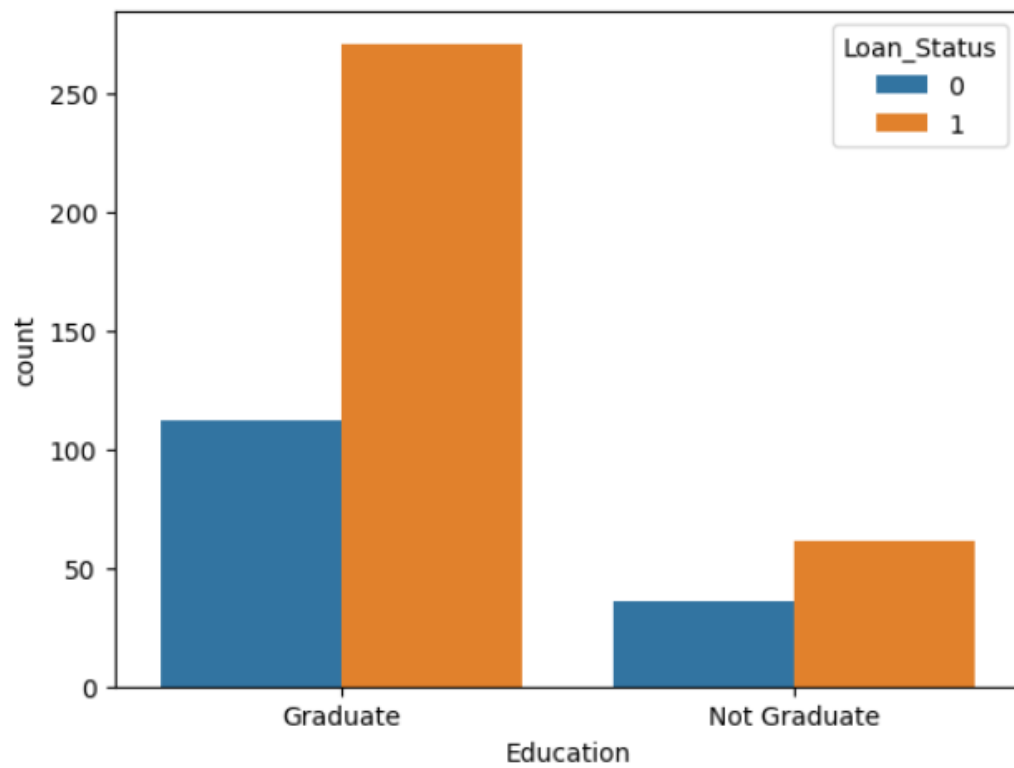
```
Out[11]: 0      274  
         2       85  
         1       80  
         4       41  
         Name: Dependents, dtype: int64
```

```
In [12]: df2.replace({"Loan_Status":{"N":0 , 'Y':1}},inplace=True)
```

```
In [14]: import seaborn as sns  
         sns.countplot(x='Education',hue='Loan_Status',data=df2)
```

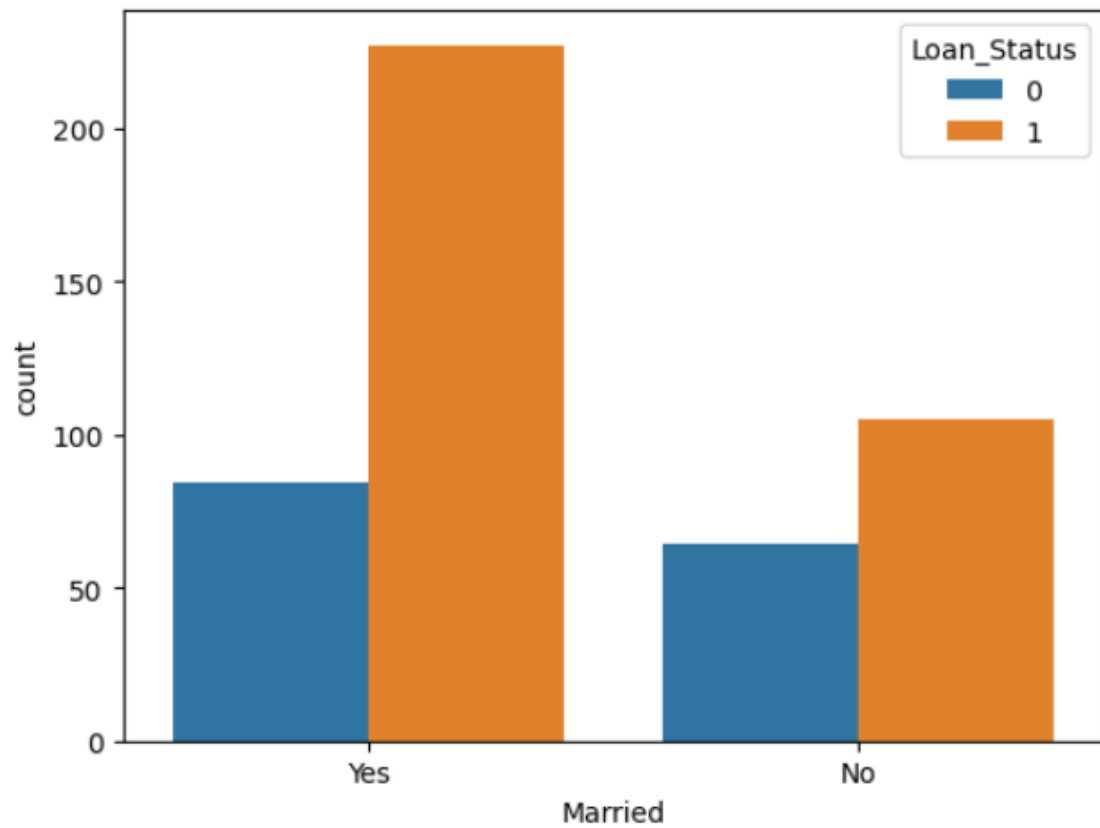
```
Out[14]: <Axes: xlabel='Education', ylabel='count'>
```

```
Out[14]: <Axes: xlabel='Education', ylabel='count'>
```



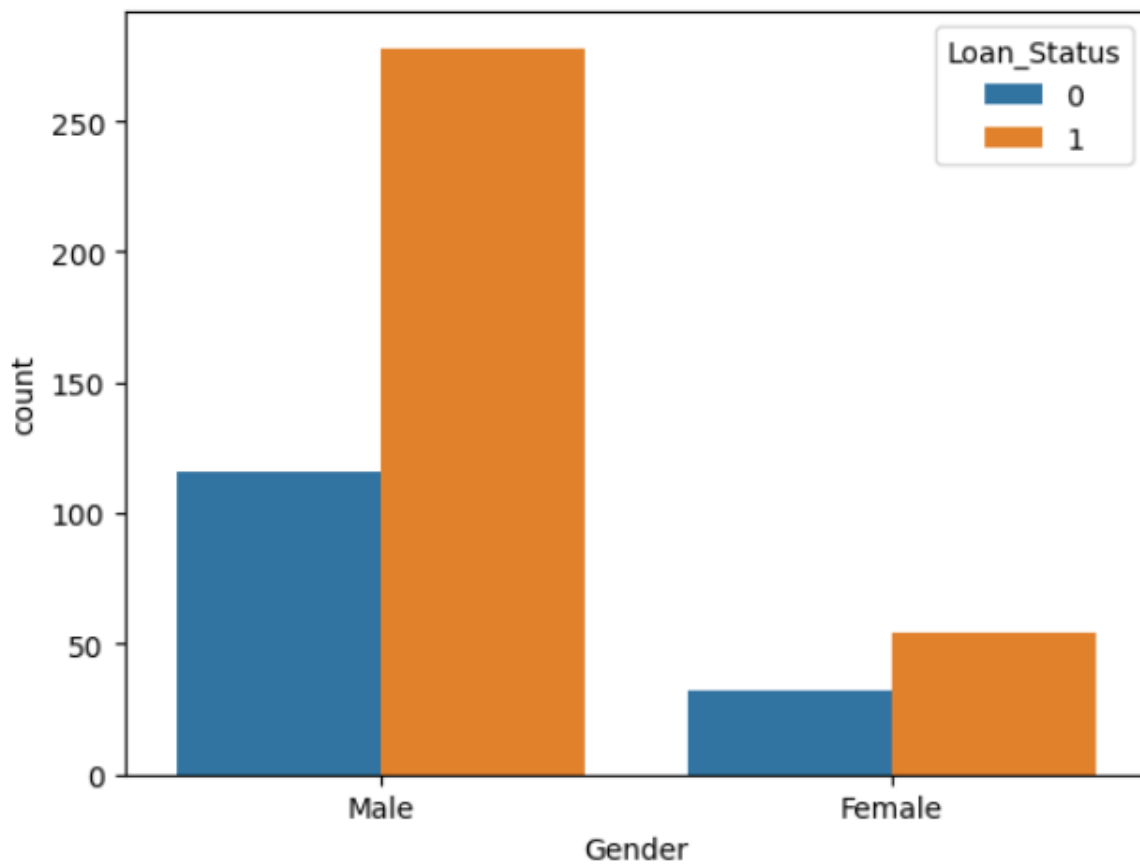
```
In [15]: sns.countplot(x="Married",hue="Loan_Status",data=df2)
```

```
Out[15]: <Axes: xlabel='Married', ylabel='count'>
```



```
[16]: sns.countplot(x="Gender",hue="Loan_Status",data=df2)
```

```
[16]: <Axes: xlabel='Gender', ylabel='count'>
```



```
In [18]: # convert categorical columns to numerical values
df2.replace({"Married":{"Yes":1,"No":0},"Gender":{"Male":1,"Female":0}},inplace=True)
```

```
In [19]: df2.head(10)
```

```
Out[19]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property_Area	Loan_Status
1	LP001003	1	1	1	Graduate	No	4583	1508.0	128.0	360.0	1.0	Rural	0
2	LP001005	1	1	0	Graduate	Yes	3000	0.0	66.0	360.0	1.0	Urban	0
3	LP001006	1	1	0	Not Graduate	No	2583	2358.0	120.0	360.0	1.0	Urban	0
4	LP001008	1	0	0	Graduate	No	6000	0.0	141.0	360.0	1.0	Urban	0

```
In [20]: df2.replace({"Self_Employed":{"No":0,'Yes':1}},inplace=True)
```

```
In [21]: df2.head(20)
```

```
Out[21]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property_Area	Loan_Status
1	LP001003		1	1	1	Graduate	0	4583	1508.0	128.0	360.0	1.0	Rural
2	LP001005		1	1	0	Graduate	1	3000	0.0	66.0	360.0	1.0	Urban
3	LP001006		1	1	0	Not Graduate	0	2583	2358.0	120.0	360.0	1.0	Urban
4	LP001008		1	0	0	Graduate	0	6000	0.0	141.0	360.0	1.0	Urban
5	LP001011		1	1	2	Graduate	1	5417	4196.0	267.0	360.0	1.0	Urban
6	LP001013		1	1	0	Not Graduate	0	2333	1516.0	95.0	360.0	1.0	Urban

```
In [22]: df2.replace({"Property_Area":{"Rural":0,'Urban':2,'Semiurban':1}},inplace=True)
```

```
In [23]: df2.replace({"Education":{"Graduate":1,'Not Graduate':0}},inplace=True)
```

```
In [24]: df2.head(10)
```

```
Out[24]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	Property_Area	Loan_Status
1	LP001003		1	1	1	0	4583	1508.0	128.0	360.0	1.0	0	
2	LP001005		1	1	0	1	3000	0.0	66.0	360.0	1.0	2	
3	LP001006		1	1	0	0	2583	2358.0	120.0	360.0	1.0	2	
4	LP001008		1	0	0	1	6000	0.0	141.0	360.0	1.0	2	
5	LP001011		1	1	2	1	5417	4196.0	267.0	360.0	1.0	2	

```
In [25]: # Separating the data and lable
X = df2.drop(columns=['Loan_ID','Loan_Status'],axis=1)
Y = df2['Loan_Status']
```

```
In [26]: print(X)
print(Y)
```


	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	\
1	1	1	1	1	0	4583	
2	1	1	0	1	1	3000	
3	1	1	0	0	0	2583	
4	1	0	0	1	0	6000	
5	1	1	2	1	1	5417	
..	
609	0	0	0	1	0	2900	
610	1	1	4	1	0	4106	
611	1	1	1	1	0	8072	
612	1	1	2	1	0	7583	
613	0	0	0	1	1	4583	

	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History	\
1	1508.0	128.0	360.0	1.0	
2	0.0	66.0	360.0	1.0	
3	2358.0	120.0	360.0	1.0	
4	0.0	141.0	360.0	1.0	
5	4196.0	267.0	360.0	1.0	
..	
609	0.0	71.0	360.0	1.0	
610	0.0	40.0	180.0	1.0	
611	240.0	253.0	360.0	1.0	
612	0.0	187.0	360.0	1.0	
613	0.0	133.0	360.0	0.0	

```

---
Property_Area
1      0
2      2
3      2
4      2
5      2
..      ...
609     0
610     0
611     2
612     2
613     1

```

```
[480 rows x 11 columns]
```

```

1      0
2      1
3      1
4      1
5      1
..
609    1
610    1
611    1
612    1
613    0

```

```
Name: Loan_Status, Length: 480, dtype: int64
```

```
Name: Loan_Status, Length: 480, dtype: int64
```

```

: # Train test split
  from sklearn.model_selection import train_test_split
  X_train, X_test, Y_train, Y_test = train_test_split(X,Y,test_size=0.1,stratify=Y,random_state=2)

```

```

: print(X.shape, X_train.shape, X_test.shape)

```

```
(480, 11) (432, 11) (48, 11)
```

```

: #Support Vector Machine
  from sklearn import svm
  from sklearn.metrics import accuracy_score
  classifier = svm.SVC(kernel = 'linear')

```

```
classifier.fit(X_train, Y_train)
```

▼ SVC

SVC(kernel='linear')

```
X_train_prediction = classifier.predict(X_train)
training_data_accuracy = accuracy_score(X_train_prediction, Y_train)
```

```
print("Accuracy on data",training_data_accuracy)
```

Accuracy on data 0.7986111111111112

```
X_test_prediction = classifier.predict(X_test)
test_data_accuracy = accuracy_score(X_test_prediction, Y_test)
```

```
print("Accuracy on test data",test_data_accuracy)
```

Accuracy on test data 0.8333333333333334