

```
In [1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")
```

```
In [2]: loan_dataset = pd.read_csv(r"C:\Users\s323\Desktop\Gatherings\Data Science\Dataset\
```

```
In [3]: loan_dataset.head()
```

```
Out[3]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coapplic
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	

```
In [4]: loan_dataset.shape
```

```
Out[4]: (614, 13)
```

```
In [6]: loan_dataset.describe()
```

```
Out[6]:
```

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_History
count	614.000000	614.000000	592.000000	600.00000	564.000000
mean	5403.459283	1621.245798	146.412162	342.00000	0.842199
std	6109.041673	2926.248369	85.587325	65.12041	0.364878
min	150.000000	0.000000	9.000000	12.00000	0.000000
25%	2877.500000	0.000000	100.000000	360.00000	1.000000
50%	3812.500000	1188.500000	128.000000	360.00000	1.000000
75%	5795.000000	2297.250000	168.000000	360.00000	1.000000
max	81000.000000	41667.000000	700.000000	480.00000	1.000000

```
In [7]: loan_dataset.columns
```

```
Out[7]: Index(['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',
'Self_Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
'Loan_Amount_Term', 'Credit_History', 'Property_Area', 'Loan_Status'],
dtype='object')
```

```
In [9]: loan_dataset.isnull().any()
```

```
Out[9]: Loan_ID      False
        Gender      True
        Married     True
        Dependents  True
        Education   False
        Self_Employed True
        ApplicantIncome False
        CoapplicantIncome False
        LoanAmount  True
        Loan_Amount_Term True
        Credit_History True
        Property_Area False
        Loan_Status False
        dtype: bool
```

```
In [15]: loan_dataset.isnull().sum()
```

```
Out[15]: 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
```

```
In [16]: # dropping the missing values since we have cateogorical value
        loan_dataset = loan_dataset.dropna()
```

```
In [19]: loan_dataset.isnull().sum()
```

```
Out[19]: 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
```

## Label Encoding

- Converting cateogrical data/Textual data to numerical data
- Y- 1, N-0

```
In [29]: loan_dataset.replace({"Loan_Status":{"N":0,"Y":1}},inplace = True)
```

```
In [30]: loan_dataset.head()
```

Out[30]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coapplic
1	LP001003	1	1	1	1	0	4583	
2	LP001005	1	1	0	1	1	3000	
3	LP001006	1	1	0	0	0	2583	
4	LP001008	1	0	0	1	0	6000	
5	LP001011	1	1	2	1	1	5417	

In [31]: *## Dependent col values*  
loan\_dataset["Dependents"].value\_counts()

Out[31]:

```
0    274
2     85
1    80
4     41
Name: Dependents, dtype: int64
```

In [32]: *# Replacing 3+ values with 4 becuae - 3+ is not good for our dataset*  
loan\_dataset=loan\_dataset.replace(to\_replace = "3+", value =4 )

In [33]: loan\_dataset['Dependents'].value\_counts()

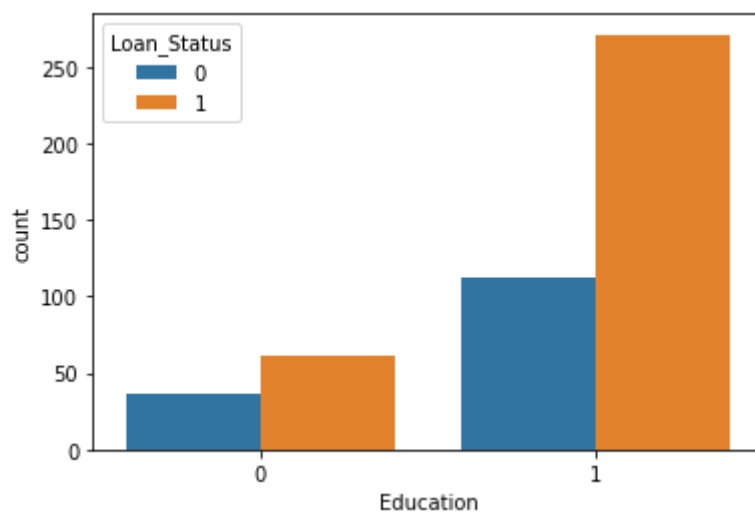
Out[33]:

```
0    274
2     85
1    80
4     41
Name: Dependents, dtype: int64
```

## Data Visulaization

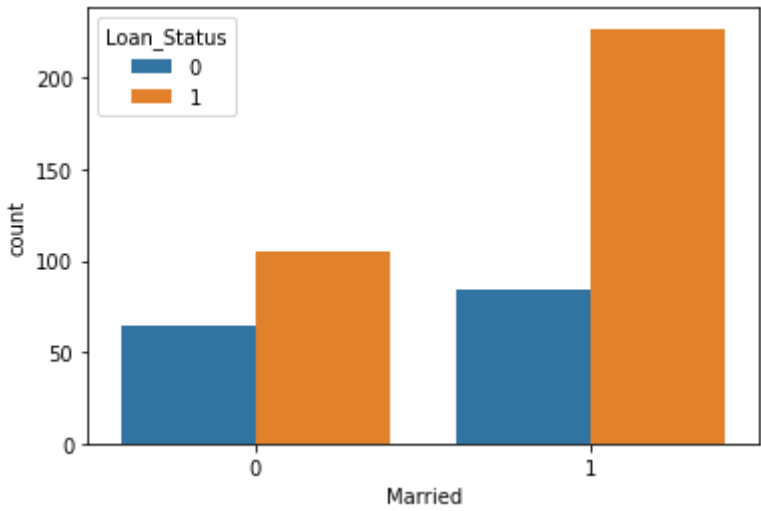
In [34]: *# education & Loan Status*  
sns.countplot(x="Education", hue="Loan\_Status", data =loan\_dataset)

Out[34]: <AxesSubplot:xlabel='Education', ylabel='count'>



In [35]: *# marital status & Loan Status*  
sns.countplot(x="Married", hue="Loan\_Status", data= loan\_dataset)

Out[35]: <AxesSubplot:xlabel='Married', ylabel='count'>



## Convert categorical columns to numerical values

```
In [36]: loan_dataset.replace({"Married":{"No":0,"Yes":1},"Gender":{"Male":1,"Female":0},"Self_Employed":{"Not_Self_Employed":0,"Self_Employed":1},"Property_Area":{"Rural":0,"Semiurban":1,"Urban":2},"Education":{"Graduate":0,"Not_Graduate":1}}
```

```
In [37]: loan_dataset.head()
```

Out[37]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome
1	LP001003	1	1	1	1	0	4583	
2	LP001005	1	1	0	1	1	3000	
3	LP001006	1	1	0	0	0	2583	
4	LP001008	1	0	0	1	0	6000	
5	LP001011	1	1	2	1	1	5417	

## Splitting data into X and Y

- data -x, label-y

```
In [39]: X = loan_dataset.drop(["Loan_ID","Loan_Status"],axis=1)
Y = loan_dataset["Loan_Status"]
```

```
In [42]: 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

## Splitting the data into train and test

```
In [44]: from sklearn.model_selection import train_test_split
```

```
In [45]: x_train,x_test,y_train,y_test = train_test_split(X,Y,test_size=0.1,stratify = Y, r
```

```
In [46]: print(X.shape,x_train.shape,x_test.shape)
```

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

## Training the model: Support Vector Machine

```
In [48]: from sklearn import svm
```

```
In [49]: classifier = svm.SVC(kernel="linear")  
# SVC = Classifier, other type of problem is regression
```

```
In [50]: classifier.fit(x_train,y_train)
```

```
Out[50]: SVC(kernel='linear')
```

## Model Evaluation

```
In [51]: from sklearn.metrics import accuracy_score
```

```
In [53]: trained_predictions = classifier.predict(x_train)  
training_data_accuracy = accuracy_score(trained_predictions,y_train)
```

```
In [54]: training_data_accuracy  
# we want training data score, it is not important though to check overfitting
```

```
Out[54]: 0.7986111111111112
```

```
In [55]: tested_prediction = classifier.predict(x_test)  
testing_data_accuracy = accuracy_score(tested_prediction,y_test)
```

```
In [56]: testing_data_accuracy
```

```
Out[56]: 0.8333333333333334
```