

Bivariate Analysis

The dataset is already download in .csv format

IMPORTING THE PACKAGE

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

import warnings
warnings.filterwarnings('ignore')
```

Load the dataset

```
In [2]: df=pd.read_csv("C:\loan_prediction.csv")
```

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

O...	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	Loan
0	LP001002	Male	No	0	Graduate	No	5849	0.0	
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	

```
In [4]: numerical_features = df.select_dtypes(include = [np.number]).columns
categorical_features = df.select_dtypes(include = [np.object]).columns
numerical_features
```

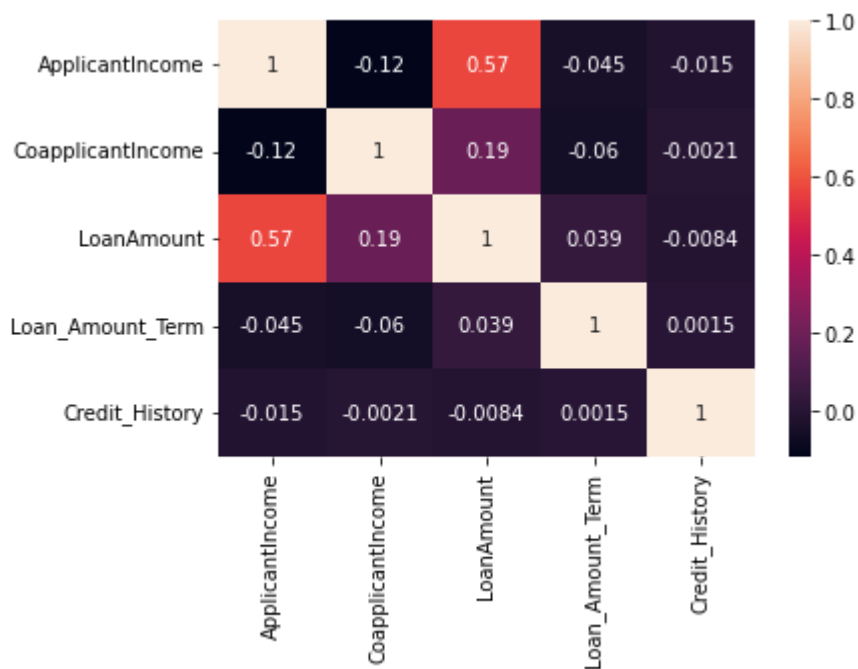
```
Out[4]:Index(['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',
             'Loan_Amount_Term', 'Credit_History'],
            dtype='object')
```

```
In [5]: categorical_features
```

```
Out[5]:Index(['Loan_ID', 'Gender', 'Married', 'Dependents', 'Education',
             'Self_Employed', 'Property_Area', 'Loan_Status'],
            dtype='object')
```

```
In [6]: df.corr()
#plotting the correlation
plt.figure(1)
sns.heatmap(df.corr(), annot = True)
```

```
Out[6]:
```

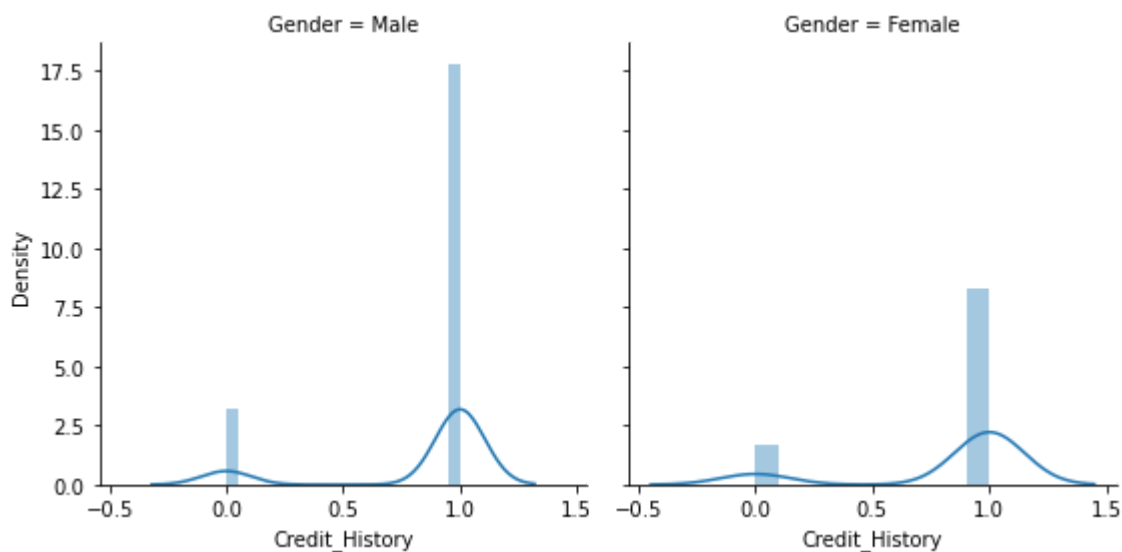


Perform Visualizations

In [7]:

```
#exploring the Gender columns more
plt.figure(3, figsize=(12, 10))
g = sns.FacetGrid(data = df, col= 'Gender', height = 4)
g.map(sns.distplot, 'Credit_History')
```

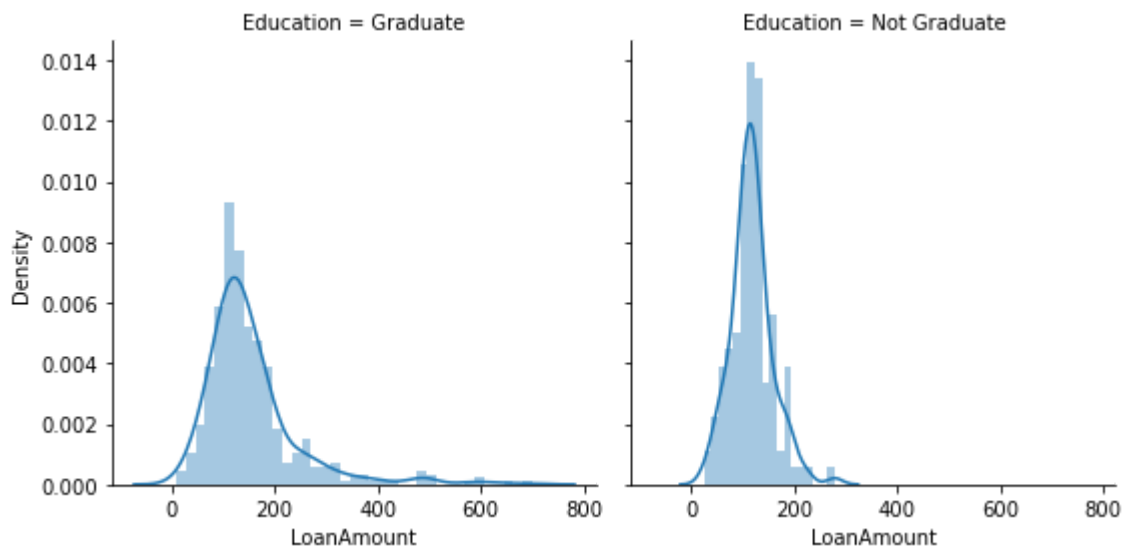
Out[7]:



In [8]:

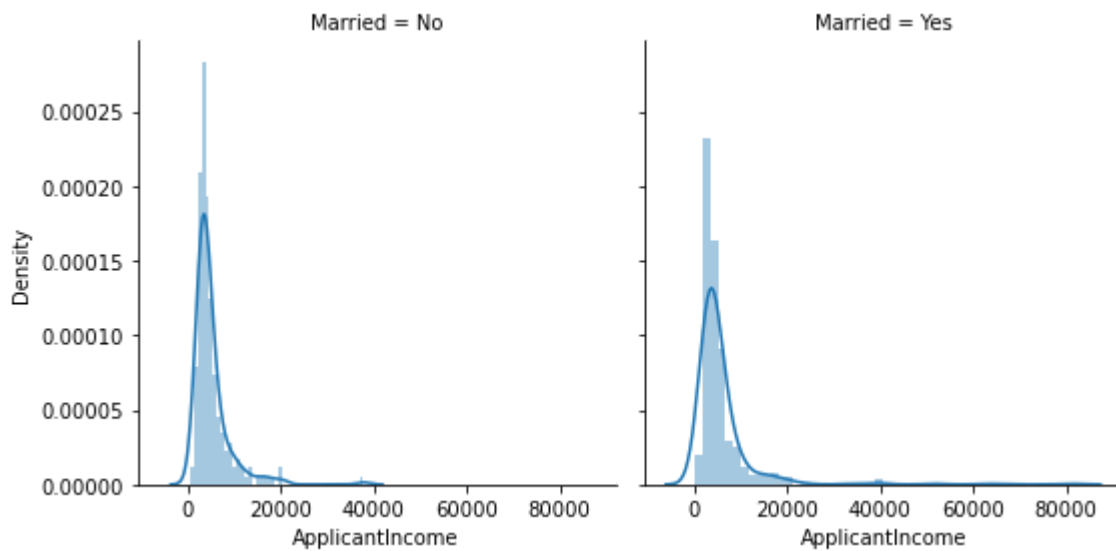
```
#exploring the Education columns more
plt.figure(3, figsize=(12, 10))
g = sns.FacetGrid(data = df, col= 'Education', height = 4)
g.map(sns.distplot, 'LoanAmount')
```

Out[8]:



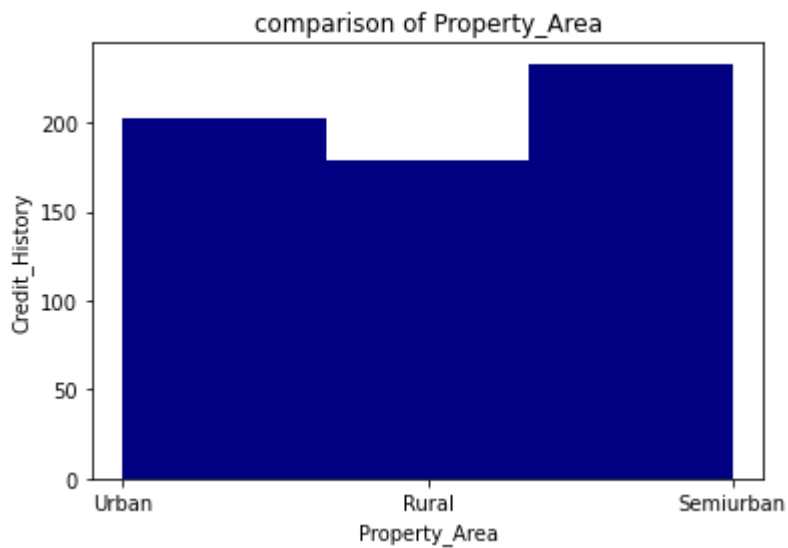
In [9]: *#exploring the Married columns more*
`plt.figure(3, figsize=(12, 10))`
`g = sns.FacetGrid(data = df, col= 'Married', height = 4)`
`g.map(sns.distplot, 'ApplicantIncome')`

Out[9]:



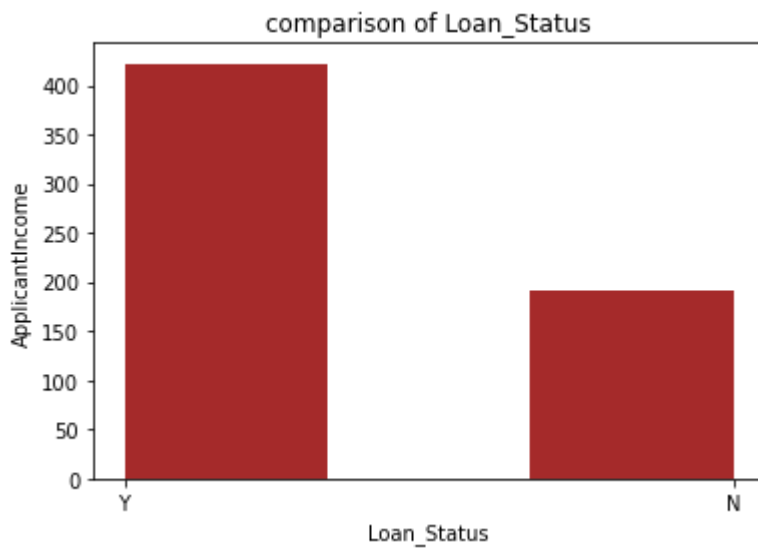
In [10]: *# Plotting the features of the dataset to see the correlation between them*

```
plt.hist(x = df.Property_Area, bins = 3, color="navy")
plt.title('comparison of Property_Area')
plt.xlabel('Property_Area')
plt.ylabel('Credit_History')
plt.show()
```



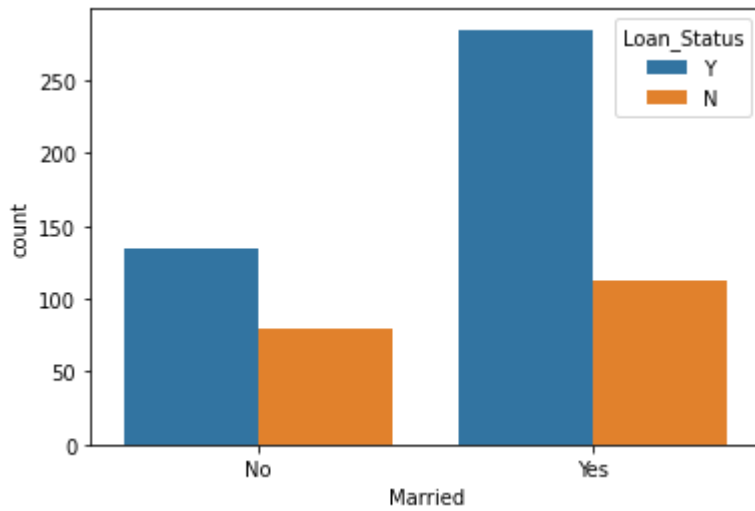
In [11]: *# Plotting the features of the dataset to see the correlation between them*

```
plt.hist(x = df.Loan_Status, bins = 3, color="brown")
plt.title('comparison of Loan_Status')
plt.xlabel('Loan_Status')
plt.ylabel('ApplicantIncome')
plt.show()
```



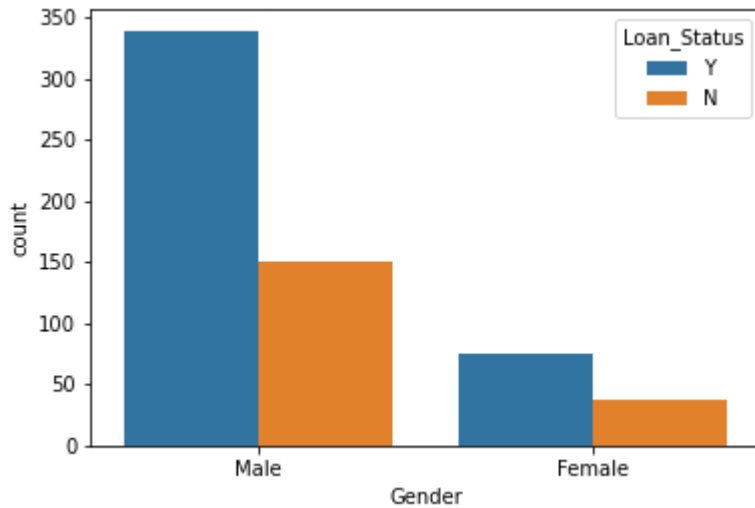
In [12]: *# Now we will find the relationship between Loan_Status*
As there is no relationship between Married and Loan_Status
`sns.countplot(x="Married", hue="Loan_Status", data=df)`

Out[12]:



```
In [13]: # Gender v/s Loan status
sns.countplot(x="Gender", hue="Loan_Status", data=df);

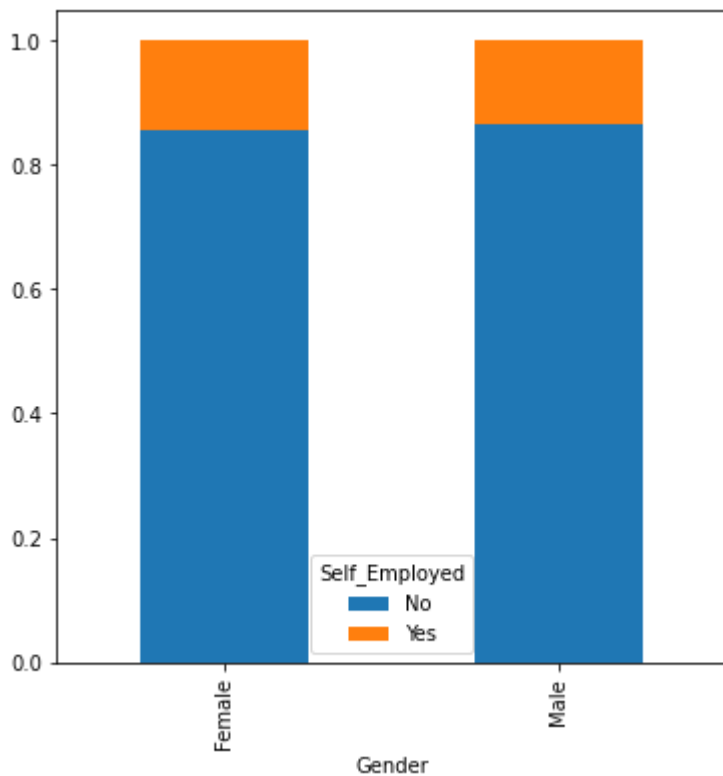
# Most of the males have got there Loans approved.
```



```
In [1... # comparison between Self_Employed and Gender

Gender = pd.crosstab(df['Gender'],df['Self_Employed'])
Gender.div(Gender.sum(1).astype(float), axis=0).plot(kind="bar", stacked=True, figsize=(10, 5))
```

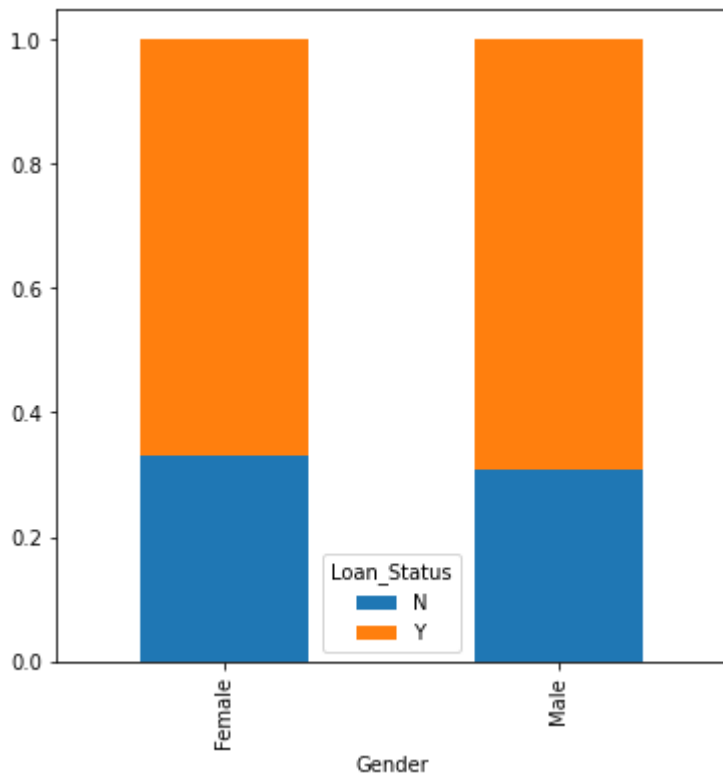
Out[14]:



In [1... *# comparison between Loan_Status and Gender*

```
Gender = pd.crosstab(df['Gender'],df['Loan_Status'])
Gender.div(Gender.sum(1).astype(float), axis=0).plot(kind="bar", stacked=True, figsize=(10, 5))
```

Out[15]:



calculating total Loan_Amount_Term in gender

```
In [16]: total_0 = df.Loan_Amount_Term[df.Dependents == '0'].sum()
total_1 = df.Loan_Amount_Term[df.Dependents == '1'].sum()
```

```

total_2 = df.Loan_Amount_Term[df.Dependents == '2'].sum()
total_3above = df.Loan_Amount_Term[df.Dependents == '3+'].sum()

print("Total Balance in Dependents 0 :",total_0)
print("Total Balance in Dependents 1 :",total_1)
print("Total Balance in Dependents 2 :",total_2)
print("Total Balance in Dependents 3 :",total_3above)

```

```

Total Balance in Dependents 0 : 116268.0
Total Balance in Dependents 1 : 33264.0
Total Balance in Dependents 2 : 34428.0
Total Balance in Dependents 3 : 16260.0

```

In [17]:

```
# plotting a pie chart
```

```

labels = '0', '1','2' ,'3+'
colors = ['brown', 'navy', 'darkgreen' , 'orange']
sizes = [116, 332, 344, 162]
explode = [ 0.01, 0.01, 0.01, 0.01]

```

```
plt.pie(sizes, colors = colors, labels = labels, explode = explode, shadow = True)
```

```

plt.axis('equal')
plt.show()

```

