

## ▼ Assignment 3: Logistic Regression

### ▼ Problem statement

You are working as a Data Scientist at a Finance company which proposes home loans. When a customer applies for a home loan, the company studies his demand to decide whether he/she is eligible or not.

Your boss asks you to develop a decision-aid tool to automate the loan eligibility process. The tool is supposed to be given as input the information provided by the customer while filling his/her online loan application form. These information include customer's personal criteria such as Gender, Marital Status, Education, Number of Dependents, Income, Loan Amount, Credit History, etc... To develop the algorithm/model to be applied in the tool, you are given a data set containing historical information about applicants (**features**) as well as their loan status (**output** given as binary variable indicating whether or not a the loan was approved).




#### 1- Load data

```
# imports
import warnings
warnings.filterwarnings('ignore')
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
# show plots in the notebook
%matplotlib inline

from google.colab import drive
drive.mount('/content/drive', force_remount=False)

df = pd.read_csv('/content/drive/MyDrive/loan_prediction.csv')
df.shape

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount(
(614, 13)
```



```
assert df.shape == (614, 13)

df.head()
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome
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



2- Remove the "Loan\_ID" column, and then transform the "Loan\_Status" feature from categorical into numerical values (Y ==> 1, N ==> 0)

```
df.drop("Loan_ID",1, inplace=True)
df["Loan_Status"] = df["Loan_Status"].astype(str).astype("category").cat.codes
```

```
assert df.shape == (614, 12); assert df["Loan_Status"].dtype != 'O'
```

## ▼ Data Exploration

3- Build a dataframe named `stats` displaying information about `df` columns. The index is the columns names and the columns are the following :

- `type` : type of the column
- `# null` : number of null values
- `# unique` : number of unique values
- `unique values` : unique values, concatenated as one text separated with `-` . If the number of unique values is greater than equal 10 ( $\geq 10$ ), print only the the first 20 characters followed by `...`

`stats` would look like this (only the last rows are displayed)

```
stats = pd.DataFrame(columns=["type", "# null", "# unique", "unique values"])
```

```
stats["type"] = df.dtypes
stats["# null"] = df.isnull().sum()
stats["# unique"] = df.nunique()
```

```
for col in df:
    uniqueVal = df[col].unique()
```

```
uniqueVal = "-".join(str(x) for x in uniqueVal)
stats["unique values"][col] = uniqueVal
```

```
stats
```

	type	# null	# unique	unique values
<b>Gender</b>	object	13	2	Male-Female
<b>Married</b>	object	3	2	No-Yes
<b>Dependents</b>	object	15	4	0-1-2-3
<b>Education</b>	object	0	2	Graduate-Not Graduate
<b>Self_Employed</b>	object	32	2	No-Yes
<b>ApplicantIncome</b>	int64	0	505	5849-4583-3000-2583-6000-5417-2333-3036-4000
<b>CoapplicantIncome</b>	float64	0	287	0.0-1508.0-2358.0-4196.0-1516.0-2504.0-1500.0-3000.0
<b>LoanAmount</b>	float64	22	203	nan-128.0-66.0-120.0-141.0-267.0-95.0-158.0-133.0-167.0
<b>Loan_Amount_Term</b>	float64	14	10	360.0-120.0-240.0-nan-180.0-60.0-300.0-48.0-96.0
<b>Credit_History</b>	float64	50	2	1.0-0.0
<b>Property_Area</b>	object	0	3	Urban-Rural-Semiurban
<b>Loan_Status</b>	int8	0	2	0-1

4- Fill the null values in the column **[Married]** by the most frequent one (which is "Yes"). Then, remove all the rows containing at least one NullValue.

```
df = df.fillna(value= {"Married": "Yes"})
df.dropna(inplace=True)
print(df.shape)
```

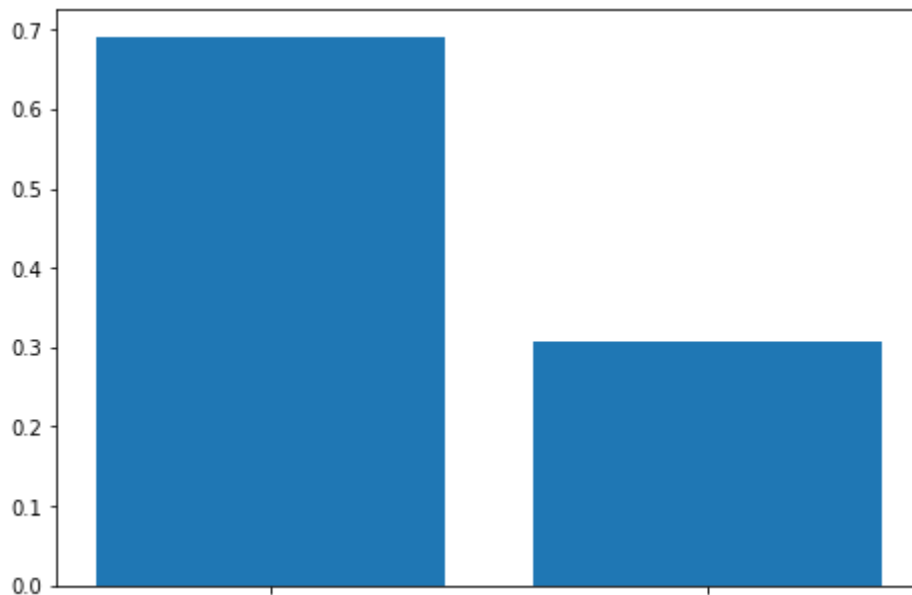
```
(480, 12)
```

```
assert df.isnull().sum().sum() == 0
```

## ▼ Data Visualization

5-1- Plot a **bar plot** of the column **Loan\_Status** (% of each class).

```
ax = plt.figure().add_axes([0,0,1,1])
ax.bar(['Yes', 'No'],df.Loan_Status.value_counts(normalize=True))
plt.show()
```



5-2- How many approved and non approved loans are there ? How much accuracy can you get if prediction is always equal to "1" ?

```
approved = (df['Loan_Status'] == 1).sum()
print("Number of approved loans :")
print(approved)
```

```
nonApproved = (df['Loan_Status'] == 0).sum()
print("Number of non approved loans :")
print(nonApproved)
```

```
Number of approved loans :
332
Number of non approved loans :
148
```

```
accuracy = approved/(approved + nonApproved)
print(accuracy)
```

```
0.6916666666666667
```

6-1- Create a function to be named **make\_bar\_plot**, which takes two parameters as input: a dataframe df (parameter 1) and a column name (parameter 2). The function must output a barplot of the approval rate Loan\_Status by value of the column.

Calling make\_bar\_plot on Property\_Area should look like this

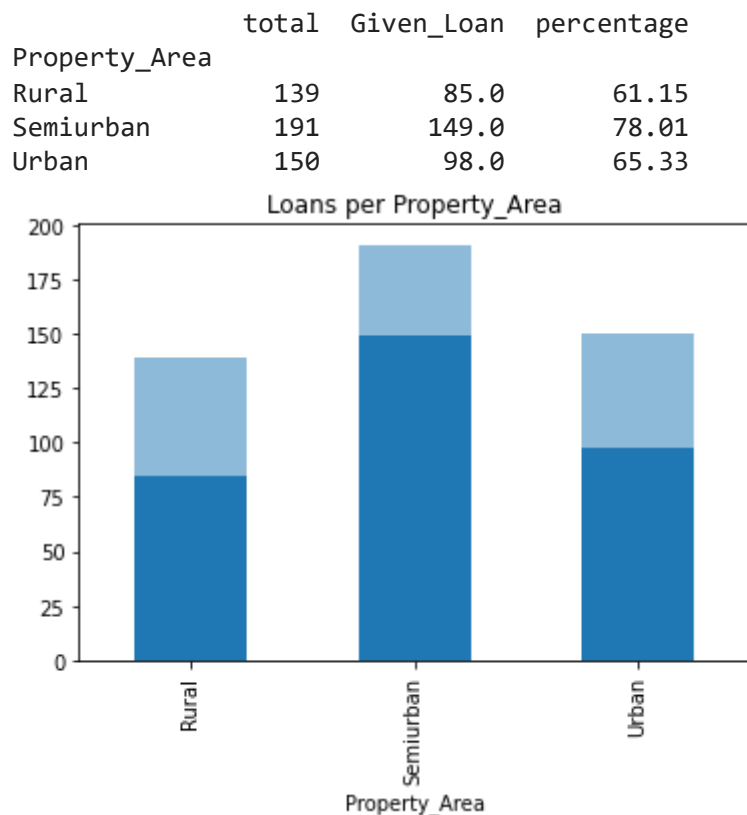
```
def make_bar_plot(df, column):
    ploter = pd.DataFrame()
    ploter['total'] = df.groupby(column).size()
    ploter['Given_Loan'] = df.groupby(column).sum()['Loan_Status']
    ploter['percentage'] = round(ploter['Given_Loan']/ploter['total']*100,2)
    print(ploter)
```

```

ploter['Given_Loan'].plot(kind="bar")
ploter['total'].plot(kind="bar",alpha=0.5,title="Loans per "+str(column))
plt.show()

```

```
make_bar_plot(df, 'Property_Area')
```



6-2- Create a function to be named **make\_box\_plot**, which takes two parameters as input: a dataframe df (parameter 1) and a column name (parameter 2). The function must output a boxplot of the distribution of the column by Loan\_Status .

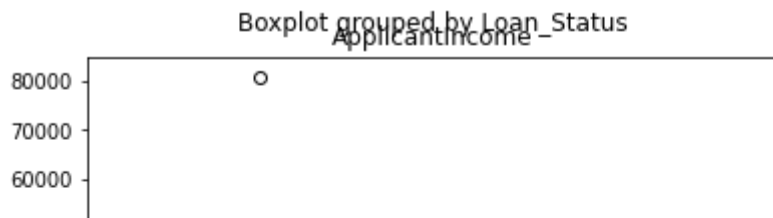
Calling make\_box\_plot on ApplicantIncome should look like this

```

def make_box_plot(df, col):
    df.boxplot(by='Loan_Status',column=col,grid=False)

```

```
make_box_plot(df, 'ApplicantIncome')
```



6-3- Loop through df columns, except for the target and make a plot depending:

- if the column is numeric, call `make_box_plot`
- else, call `make_bar_plot`

```
~~~~~ |          |          |          |
```

```
from pandas.api.types import is_numeric_dtype
```

```
for col in df:
```

```
    if(col!="Loan_Status"):
```

```
        print(f'***** {col} *****')
```

```
        if is_numeric_dtype(df[col]):
```

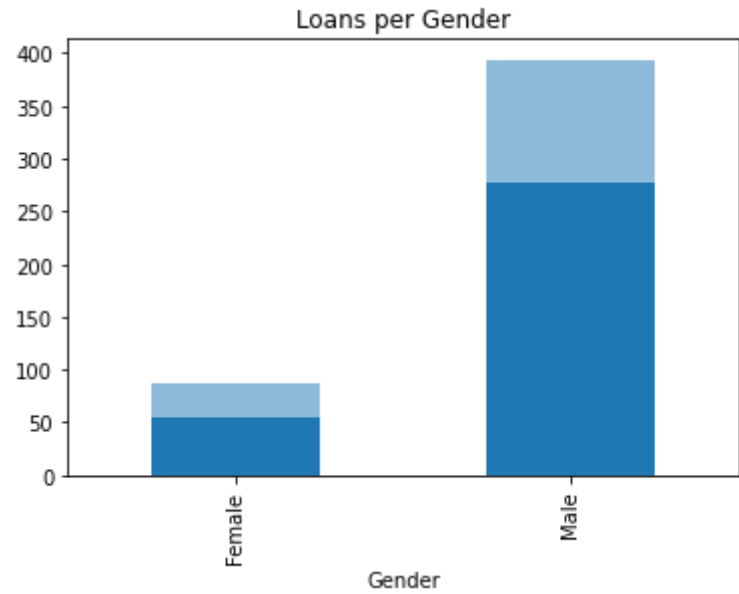
```
            make_box_plot(df,col)
```

```
        else:
```

```
            make_bar_plot(df,col)
```

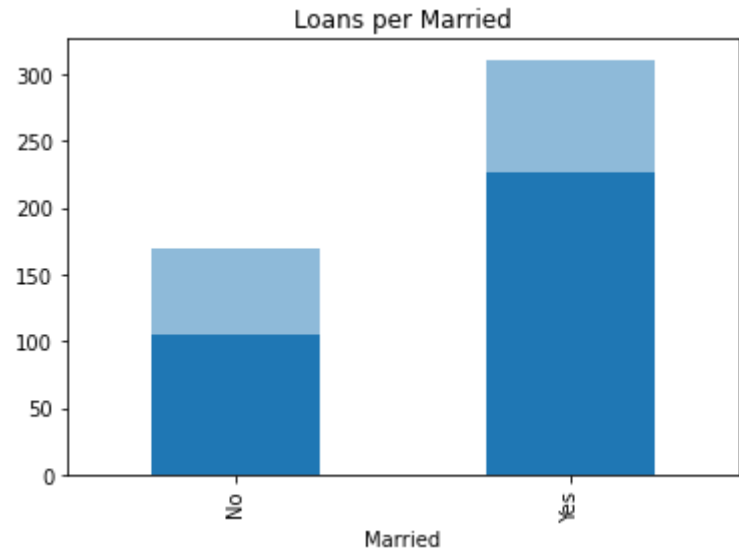
\*\*\*\*\* Gender \*\*\*\*\*

	total	Given_Loan	percentage
Gender			
Female	86	54.0	62.79
Male	394	278.0	70.56



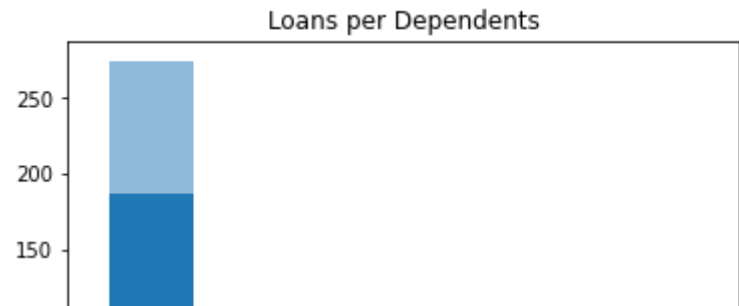
\*\*\*\*\* Married \*\*\*\*\*

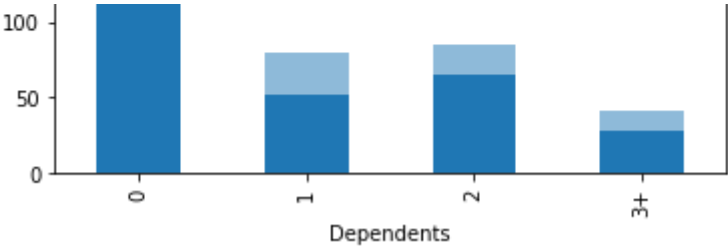
	total	Given_Loan	percentage
Married			
No	169	105.0	62.13
Yes	311	227.0	72.99



\*\*\*\*\* Dependents \*\*\*\*\*

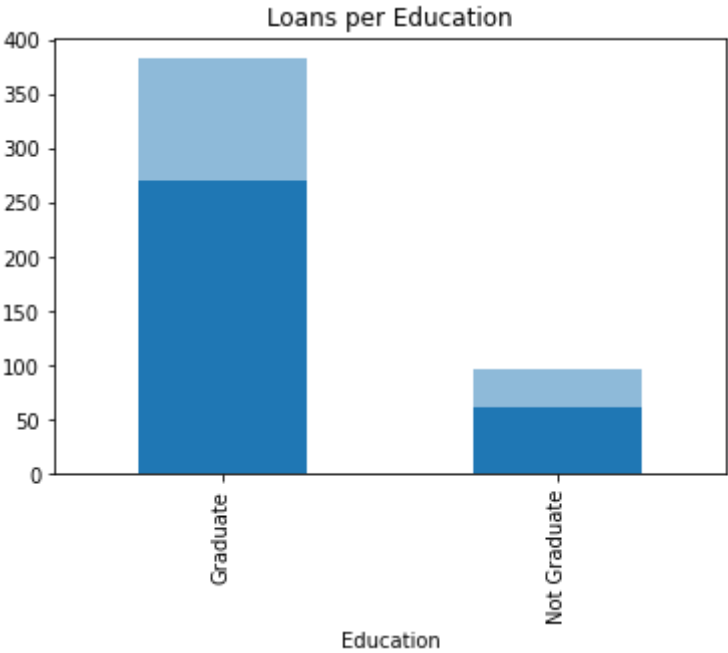
	total	Given_Loan	percentage
Dependents			
0	274	187.0	68.25
1	80	52.0	65.00
2	85	65.0	76.47
3+	41	28.0	68.29





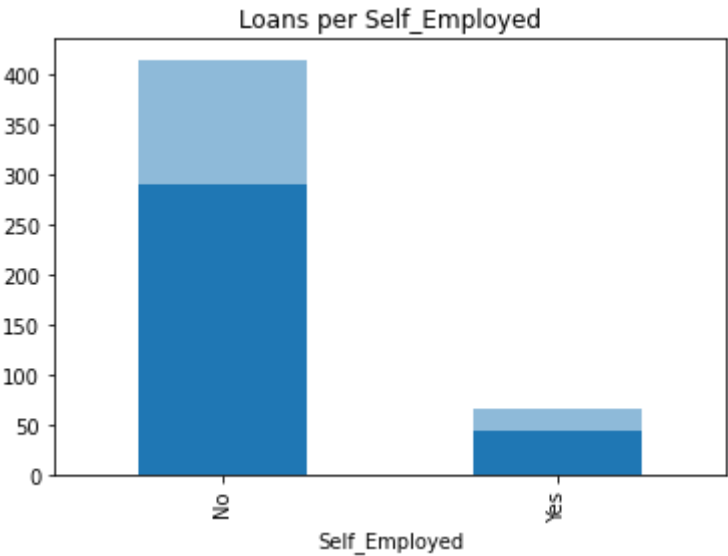
\*\*\*\*\* Education \*\*\*\*\*

	total	Given_Loan	percentage
Education			
Graduate	383	271.0	70.76
Not Graduate	97	61.0	62.89



\*\*\*\*\* Self\_Employed \*\*\*\*\*

	total	Given_Loan	percentage
Self_Employed			
No	414	289.0	69.81
Yes	66	43.0	65.15



\*\*\*\*\* ApplicantIncome \*\*\*\*\*

\*\*\*\*\* CoapplicantIncome \*\*\*\*\*

\*\*\*\*\* LoanAmount \*\*\*\*\*

\*\*\*\*\* Loan\_Amount\_Term \*\*\*\*\*

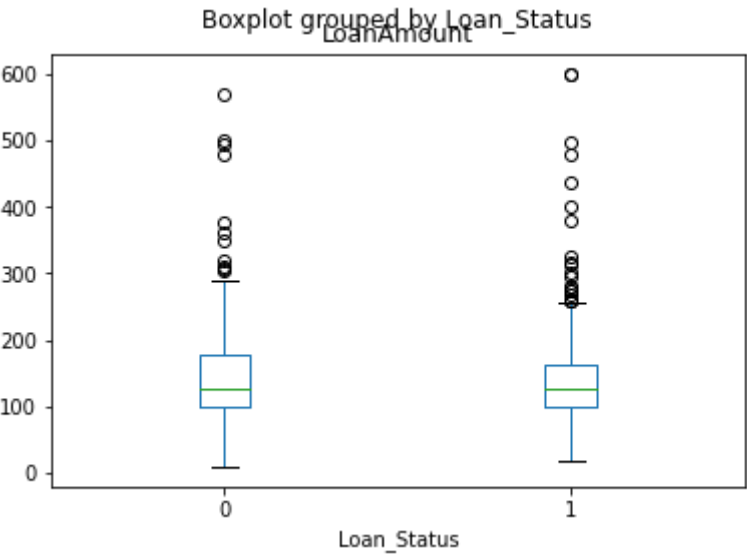
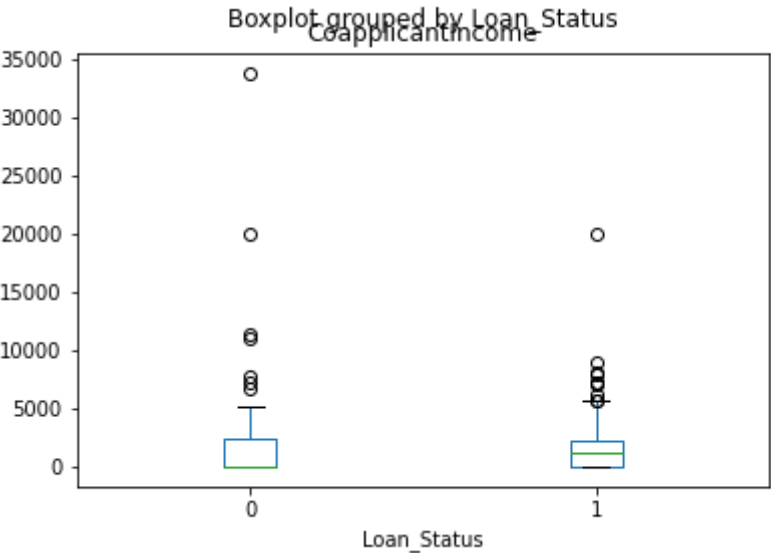
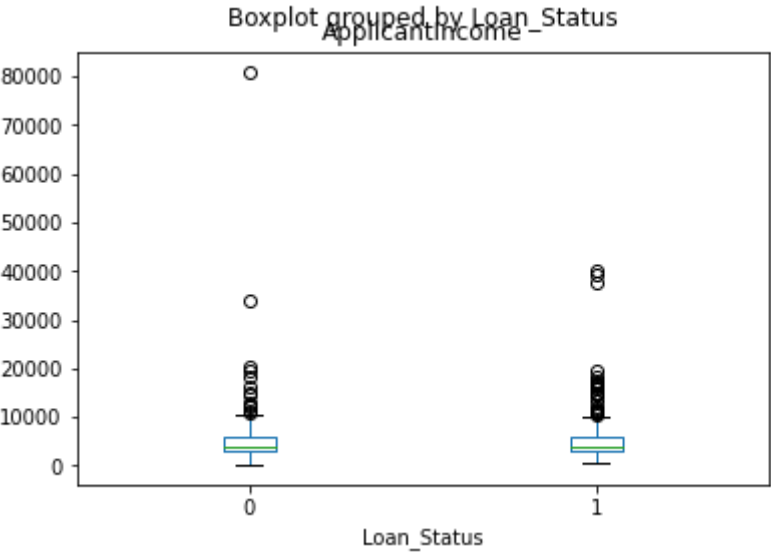
\*\*\*\*\* Credit\_History \*\*\*\*\*

\*\*\*\*\* Property\_Area \*\*\*\*\*

	total	Given_Loan	percentage
--	-------	------------	------------



Property_Area			
Rural	139	85.0	61.15
Semiurban	191	149.0	78.01
Urban	150	98.0	65.33



Boxplot grouped by Loan\_Status

Model Training & Evaluation

7-1- Use [pandas.get\\_dummies](#) to transform Property\_Area and Dependents into dummy variables

```
df = pd.get_dummies(df, columns=["Property_Area", "Dependents"])
```

print the head of df

```
df.head()
```

	Gender	Married	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	Loan_Status
1	Male	Yes	Graduate	No	4583	1508.0	
2	Male	Yes	Graduate	Yes	3000	0.0	
3	Male	Yes	Not Graduate	No	2583	2358.0	
4	Male	No	Graduate	No	6000	0.0	
5	Male	Yes	Graduate	Yes	5417	4196.0	



```
assert df.shape == (480, 17)
```

7-2- Use `pandas.Categorical` (as seen in *Titanic* notebook) to transform the remaining text columns of df into numerical ones

```
for col in df:
    if df[col].dtype == "object":
        df[col] = pd.Categorical(df[col]).codes
```

Print the head of df

```
df.head()
```