#### Assignment -2

### **Python Programming**

Assignment Date	30 September2022
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Student Roll Number	111519104100
Maximum Marks	2 Marks

#### Question-1:

1. Download the dataset: Dataset

#### **Solution:**

Downloaded succesfully

#### Question-2:

2. Load the dataset.

#### **Solution:**

```
import pandas as pd
import numpy as np
```

```
file=pd.read_csv("/content/Churn_Modelling (1).csv")
df=pd.DataFrame(file)
df.head()
```

```
        RowNumber
        CustomerId
        Surname
        CreditScore
        Geography
        Gender
        Age
        Tenure
        Balance
        NumOfProducts
        HasCrCard
        IsActiveMember
        EstimatedSalary

        0
        1
        15634602
        Hargrave
        619
        France
        Female
        42
        2
        0.00
        1
        1
        1
        101348.88

        1
        2
        15647311
        Hill
        608
        Spain
        Female
        41
        1
        83807.86
        1
        0
        1
        112542.58

        2
        3
        15619304
        Onio
        502
        France
        Female
        42
        8
        159660.80
        3
        1
        0
        113931.57

        3
        4
        15701354
        Boni
        699
        France
        Female
        39
        1
        0.00
        2
        0
        0
        93826.63

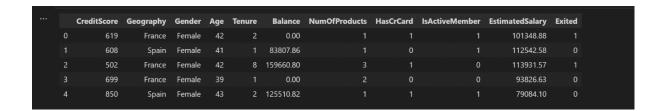
        4
        5
        15737888
        Mitchell
        850
        Spain
        Female
        43
        2
        125510.82
        1
        1
        1
        79084.10
```

```
df['HasCrCard'] = df['HasCrCard'].astype('category')
```

```
df['IsActiveMember'] = df['IsActiveMember'].astype('category')
df['Exited'] = df['Exited'].astype('category')
```

```
df = df.drop(columns=['RowNumber', 'CustomerId', 'Surname'])
```

## df.head()



#### Question 3:

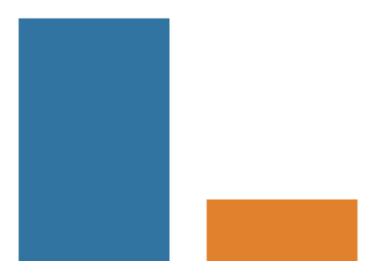
3. Perform Below Visualizations:

Univariate Analysis, Bi - Variate Analysis, Multi - Variate Analysis

#### **Solution:**

```
import seaborn as sns
density = df['Exited'].value_counts(normalize=True).reset_index()
sns.barplot(data=density, x='index', y='Exited', );
density
```

	index	Exited
0	0	0.7963
1	1	0.2037



The data is significantly imbalanced

```
import matplotlib.pyplot as plt
```

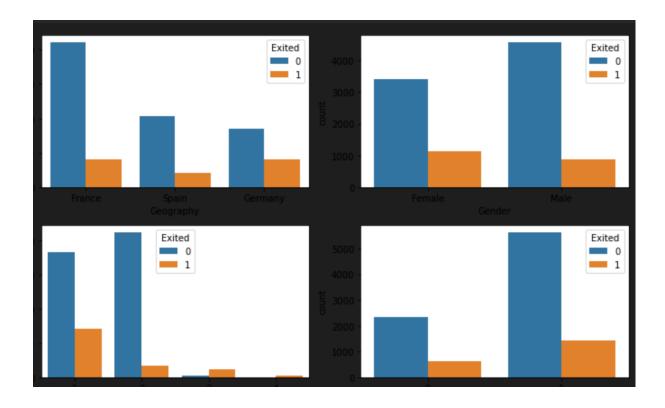
```
categorical = df.drop(columns=['CreditScore', 'Age', 'Tenure', 'Balance',
'EstimatedSalary'])
```

```
rows = int(np.ceil(categorical.shape[1] / 2)) - 1

# create sub-plots anf title them
fig, axes = plt.subplots(nrows=rows, ncols=2, figsize=(10,6))
axes = axes.flatten()

for row in range(rows):
    cols = min(2, categorical.shape[1] - row*2)
    for col in range(cols):
        col_name = categorical.columns[2 * row + col]
        ax = axes[row*2 + col]
        sns.countplot(data=categorical, x=col_name, hue="Exited", ax=ax);

plt.tight_layout()
```



## Question 4:

4. Perform descriptive statistics on the dataset.

#### **Solution:**

```
df.info()
```

# df.describe()

	CreditScore	Age	Tenure	Balance	NumOfProducts	EstimatedSalary
count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000
mean	650.561300	38.660800	5.012800	76485.889288	1.527200	100090.239881
std	96.558702	9.746704	2.892174	62397.405202	0.570081	57510.492818
min	383.000000	18.000000	0.000000	0.000000	1.000000	11.580000
25%	584.000000	32.000000	3.000000	0.000000	1.000000	51002.110000
50%	652.000000	37.000000	5.000000	97198.540000	1.000000	100193.915000
75%	718.000000	44.000000	7.000000	127644.240000	2.000000	149388.247500
max	850.000000	62.000000	10.000000	250898.090000	3.500000	199992.480000

# Question 5:

5. Handle the Missing values.

#### **Solution:**

# df.isna().sum()

```
CreditScore
                0
Geography
Gender
Age
                0
Tenure
                а
Balance
NumOfProducts 0
HasCrCard
               0
IsActiveMember
               0
EstimatedSalary 0
Exited
dtype: int64
```

There is no missing values in dataset

```
for i in df:
    if df[i].dtype=='object' or df[i].dtype=='category':
        print("unique of "+i+" is "+str(len(set(df[i])))+" they are
"+str(set(df[i])))
```

```
unique of Geography is 3 they are {'France','Germany','Spain'} unique of Gender is 2 they are {'Male','Female'} unique of Has CrCard is 2 they are {0,1} unique of Is Active Member is 2 they are {0,1} unique of Exited is 2 they are {0,1}
```

#### Question 6:

6. Find the outliers and replace the outliers.

#### **Solution:**

Checking for outliers

```
def box_scatter(data, x, y):
    fig, (ax1, ax2) = plt.subplots(nrows=2, ncols=1, figsize=(16,6))
    sns.boxplot(data=data, x=x, ax=ax1)
    sns.scatterplot(data=data, x=x,y=y,ax=ax2)
```

```
box_scatter(df,'CreditScore','Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['CreditScore'] < 400])}")</pre>
```

# of bivariate Outliers:19

```
box_scatter(df,'Age','Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['Age'] > 87])}")
```

# # of bivariate Outliers:0

```
box_scatter(df,'Balance','Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['Balance'] > 220000])}")
```

#### # of bivariate Outliers:4

```
box_scatter(df,'EstimatedSalary','Exited');
plt.tight_layout()
```



**Removing Outliers** 

```
for i in df:
    if df[i].dtype=='int64' or df[i].dtypes=='float64':
        q1=df[i].quantile(0.25)
        q3=df[i].quantile(0.75)
        iqr=q3-q1
        upper=q3+1.5*iqr
        lower=q1-1.5*iqr
        df[i]=np.where(df[i] >upper, upper, df[i])
        df[i]=np.where(df[i] <lower, lower, df[i])</pre>
```

After removing outliers, boxplot will be like

```
box_scatter(df,'CreditScore','Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['CreditScore'] < 400])}")</pre>
```

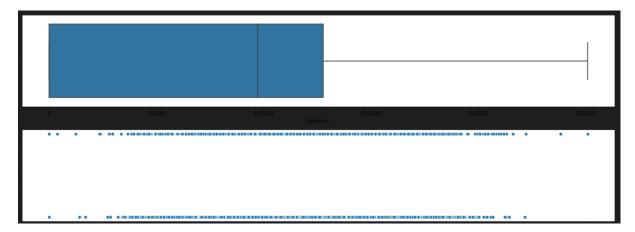
# of bivariate Outliers:19

```
box_scatter(df,'Age','Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['Age'] > 87])}")
```

#### # of bivariate Outliers:0

```
box_scatter(df,'Balance','Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['Balance'] > 220000])}")
```

## # of bivariate Outliers:4



## **Question 7:**

7. Check for Categorical columns and perform encoding.

#### **Solution:**

```
from sklearn.preprocessing import LabelEncoder
encoder=LabelEncoder()
for i in df:
   if df[i].dtype=='object' or df[i].dtype=='category':
        df[i]=encoder.fit_transform(df[i])
```

# **Question 8:**

8. Split the data into dependent and independent variables.

#### **Solution:**

```
x=df.iloc[:,:-1]
x.head()
```

555555		51919191919191919191919191919191	5H5H5H5H5H5H5H5H5H5H5H5	isisisisisisisisi	הומומומומומומומומומומומומומומומומומומומ	notototototototototototo				
	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary
0	619.0			42.0	2.0	0.00	1.0			101348.88
1	608.0	2		41.0	1.0	83807.86	1.0			112542.58
2	502.0			42.0	8.0	159660.80	3.0			113931.57
3	699.0			39.0	1.0	0.00	2.0			93826.63
4	850.0	2		43.0	2.0	125510.82	1.0			79084.10

```
y=df.iloc[:,-1]
y.head()
```

```
0 1
1 0
2 1
3 0
4 0
Name: Exited, dtype: int64
```

# **Question 9:**

9. Scale the independent variables

## **Solution:**

```
from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
x=scaler.fit_transform(x)
```

x

```
array([[-0.32687761, -0.90188624, -1.09598752, ..., 0.64609167, 0.97024255, 0.02188649],
[-0.44080365, 1.51506738, -1.09598752, ..., -1.54776799, 0.97024255, 0.2165375],
[-1.53863634, -0.90188624, -1.09598752, ..., 0.64609167, -1.03067011, 0.2406869],
...,
[ 0.60524449, -0.90188624, -1.09598752, ..., -1.54776799, 0.97024255, -1.00864308],
[ 1.25772996, 0.30659057, 0.91241915, ..., 0.64609167, -1.03067011, -0.12523071],
[ 1.4648682 , -0.90188624, -1.09598752, ..., 0.64609167, -1.03067011, -1.07636976]])
```

# **Question 10:**

10. Split the data into training and testing

#### **Solution:**

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.33)
```