## Assignment -2

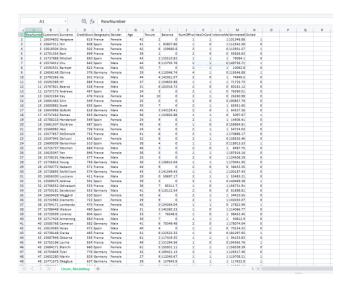
### DATA VISUALISATION AND PRE-PROCESSING

Assignment Date	27 September 2022
Student Name	ABINAYA S
Student Roll Number	612419104002
Maximum Marks	2 Marks

## Question-1:

Download the Dataset

## **SOLUTION:**



## Question-2:

Loading dataset

```
df = pd.read_csv("Churn_Modelling.csv")
df
```

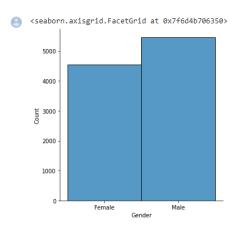
	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58
2	3	15619304	Onio	502	France	Female	42	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
9995	9996	15606229	Obijiaku	771	France	Male	39	5	0.00	2	1	0	96270.64
9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101699.77
9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	42085.58
9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	75075.31	2	1	0	92888.52

## Question-3:

- 1. Visualizations
- a) Univariate Analysis

### **SOLUTION:**

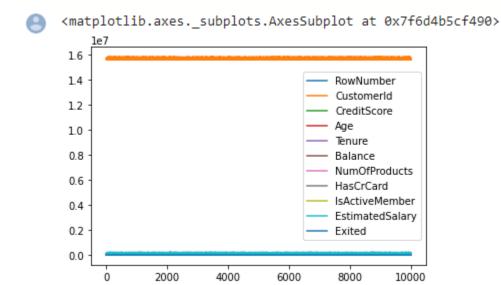
sns.displot(df.Gender)



## b) Bi-Variate Analysis

### **SOLUTION:**

df.plot.line()

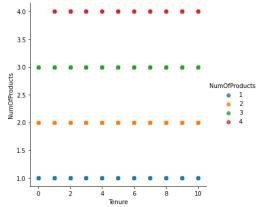


## c) Multi - Variate Analysis

### **SOLUTION:**

```
sns.lmplot("Tenure", "NumOfProducts", df, hue="NumOfProducts", fit reg=False);
```

/usr/local/lib/python3.7/dist-packages/seaborn/\_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y, data. From version FutureWarning



## Question-4:

Perform descriptive statistics on the dataset.

#### **SOLUTION:**

df.describe()

df.describe()													
	RowNumber	CustomerId	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary			
count	10000.00000	1.000000e+04	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.00000	10000.000000	10000.000000	1		
mean	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	76485.889288	1.530200	0.70550	0.515100	100090.239881			
std	2886.89568	7.193619e+04	96.653299	10.487806	2.892174	62397.405202	0.581654	0.45584	0.499797	57510.492818			
min	1.00000	1.556570e+07	350.000000	18.000000	0.000000	0.000000	1.000000	0.00000	0.000000	11.580000			
25%	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	0.000000	1.000000	0.00000	0.000000	51002.110000			
50%	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	97198.540000	1.000000	1.00000	1.000000	100193.915000			
75%	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	127644.240000	2.000000	1.00000	1.000000	149388.247500			
max	10000.00000	1.581569e+07	850.000000	92.000000	10.000000	250898.090000	4.000000	1.00000	1.000000	199992.480000			

## **Question-5:**

Handle the Missing values.

#### **SOLUTION:**

```
data = pd.read_csv("Churn_Modelling.csv")
pd.isnull(data["Gender"])
```

```
False
        False
        False
3
        False
4
        False
9995
        False
9996
        False
9997
        False
9998
        False
        False
9999
Name: Gender, Length: 10000, dtype: bool
```

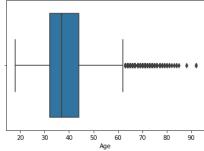
## **Question-6:**

Find the outliers and replace the outliers.

### **SOLUTION:**

```
sns.boxplot(df['Age'])
```

//usr/local/lib/python3.7/dist-packages/seaborn/\_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the FutureWarning 
// FutureWarning 
// Matplotlib.axes.\_subplots.AxesSubplot at 0x7f6d4aaf7110>



#### **SOLUTION:**

```
df['Age']=np.where(df['Age']>50,40,df['Age'])
df['Age']
```

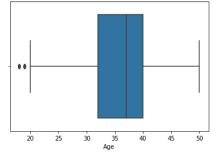
```
42
0
1
        41
2
        42
         39
3
4
        43
         . .
9995
        39
9996
        35
9997
        36
9998
        42
9999
        28
Name: Age, Length: 10000, dtype: int64
```

#### **SOLUTION:**

```
sns.boxplot(df['Age'])
```

 $/usr/local/lib/python 3.7/dist-packages/seaborn/\_decorators.py: 43: Future Warning: Pass the following variable as a keyword arg: x. From version 0.12, the partial part of the partial part of the partial partial$ FutureWarning

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f6d4b7e8990>



```
df['Age']=np.where(df['Age']<20,35,df['Age'])</pre>
df['Age']
```

```
0
        42
        41
1
2
        42
4
        43
9995
        39
9996
        35
9997
        36
9998
        42
9999
        28
Name: Age, Length: 10000, dtype: int64
```

# Question-7:

Check for Categorical columns and perform encoding.

## **SOLUTION:**

```
pd.get_dummies(df, columns=["Gender", "Age"], prefix=["Age", "Gender"]).hea
```

9	nOfProducts	HasCrCard	IsActiveMember	 Gender_41	Gender_42	Gender_43	Gender_44	Gender_45	Gender_46	Gender_47	Gender_48	Gender_49	Gender_50
	1	1	1	 0	1	0	0	0	0	0	0	0	0
	1	0	1	 1	0	0	0	0	0	0	0	0	0
	3	1	0	 0	1	0	0	0	0	0	0	0	0
	2	0	0	 0	0	0	0	0	0	0	0	0	0
	1	1	1	 0	0	1	0	0	0	0	0	0	0

	RowNumber	CustomerId	Surname	CreditScore	Geography	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	• • •	Gender_41	Gender_42	Gende
0	1	15634602	Hargrave	619	France	2	0.00	1	1	1		0	1	
1	2	15647311	Hill	608	Spain	1	83807.86	1	0	1		1	0	
2	3	15619304	Onio	502	France	8	159660.80	3	1	0		0	1	
3	4	15701354	Boni	699	France	1	0.00	2	0	0		0	0	
4	5	15737888	Mitchell	850	Spain	2	125510.82	1	1	1		0	0	

5 rows × 45 columns

## **Question-8:**

- Split the data into dependent and independent variables.
  - a) Split the data into Independent variables.

#### **SOLUTION:**

```
X = df.iloc[:, :-1].values
print(X)
```

```
[[1 15634602 'Hargrave' ... 1 1 101348.88]
[2 15647311 'Hill' ... 0 1 112542.58]
[3 15619304 'Onio' ... 1 0 113931.57]
...
[9998 15584532 'Liu' ... 0 1 42085.58]
[9999 15682355 'Sabbatini' ... 1 0 92888.52]
[10000 15628319 'Walker' ... 1 0 38190.78]]
```

b) Split the data into Dependent variables

```
Y = df.iloc[:, -1].values
print(Y)
[1 0 1 ... 1 1 0]
```

### Question-9:

Scale the independent variables

#### **SOLUTION:**

3

4

93826.63

79084.10

. . .

0

```
import pandas as pd
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
df[["CustomerId"]] = scaler.fit transform(df[["CustomerId"]])
print(df)
        RowNumber CustomerId Surname CreditScore Geography Gender Age \
               1 0.275616 Hargrave
                                               619 France Female 42
                2 0.326454 Hill
  1
                                                    608
                                                             Spain Female
                                                                                 41
                                                    502 France Female 42
699 France Female 39
                                     Onio
  2
                3 0.214421
               4 0.542636 Boni
  3
 3 4 0.542636 Boni 699 France Female 39
4 5 0.688778 Mitchell 850 Spain Female 43
... 9995 9996 0.162119 Obijiaku 771 France Male 39
9996 9997 0.016765 Johnstone 516 France Male 35
9997 9998 0.075327 Liu 709 France Female 36
9998 9999 0.466637 Sabbatini 772 Germany Male 42
9999 10000 0.250483 Walker 792 France Female 28
        Tenure Balance NumOfProducts HasCrCard IsActiveMember \
                     0.00
  0
            2
                                1
                                                      1
                                         1
  1
            1 83807.86
                                                      0
                                                                        1
            8 159660.80
                                         3
                                                     1
  3
            1
                      0.00
                                         2
                                                                        0
  4
            2 125510.82
                                         1
                                                     1
                                                                        1
                                        . . .
                      . . .
                                                   . . .
  . . .
           . . .
  9995
          5
                      0.00
                                         2
                                                     1
  9996
          10 57369.61
                                         1
                                                     1
                                                                        1
                                        1
  9997
            7
                                                     0
                   0.00
                                                                       1
 9998 3 75075.31
9999 4 130142.79
                                         2
                                                     1
                                         1
        EstimatedSalary Exited
  0
              101348.88
  1
              112542.58
  2
             113931.57
                               1
```

## Question-10:

Split the data into training and testing

1. List item

```
from sklearn.model_selection import train_test_split

train_size=0.8

X = df.drop(columns = ['Tenure']).copy()
y = df['Tenure']

X_train, X_rem, y_train, y_rem = train_test_split(X,y, train_size=0.8)
test_size = 0.5

X_valid, X_test, y_valid, y_test = train_test_split(X_rem,y_rem, test_size=0.5)
print(X_train.shape), print(y_train.shape)
print(X_valid.shape), print(y_valid.shape)
print(X_test.shape), print(y_test.shape)
```

```
(8000, 13)
(8000,)
(1000, 13)
(1000,)
(1000, 13)
(1000,)
(None, None)
```