Assignment -2

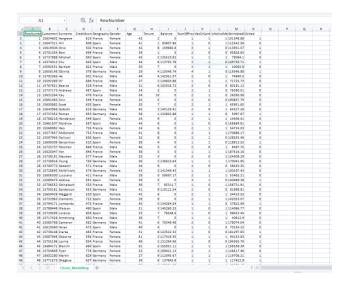
DATA VISUALISATION AND PRE-PROCESSING

Assignment Date	28 September 2022
Student Name	DINESH KUMAR M
Student Roll Number	612419104007
Maximum Marks	2 Marks

Question-1:

Download the Dataset

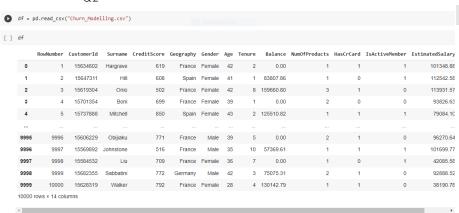
SOLUTION:



Question-2:

Loading dataset

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

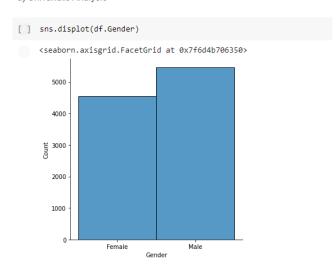


Question-3:

- 1. Visualizations
- a) Univariate Analysis

SOLUTION:

a) Univariate Analysis



b) Bi-Variate Analysis

SOLUTION:

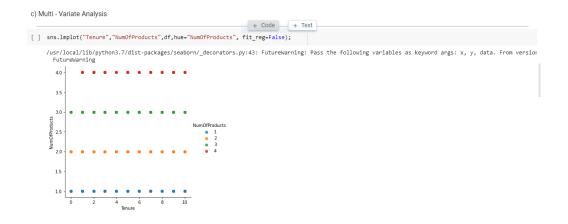
b) Bi-Variate Analysis

[] df.plot.line() <matplotlib.axes._subplots.AxesSubplot at 0x7f6d4b5cf490> 1.4 RowNumber CustomerId 1.2 CreditScore Age 1.0 Tenure Balance 0.8 NumOfProducts 0.6 HasCrCard IsActiveMember 0.4 EstimatedSalary 0.2 Exited 0.0 2000 4000 8000 10000 6000

c) Multi - Variate Analysis

SOLUTION:

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



Question-4:

Perform descriptive statistics on the dataset.

SOLUTION:

df.describe()

count 1000 mean 500 std 288 min 25% 250	RowNumber 0000.00000 5000.50000	CustomerId 1.000000e+04 1.569094e+07 7.193619e+04	10000.000000 650.528800	Age 10000.00000 38.921800	Tenure 10000.000000 5.012800	Balance 10000.000000 76485.889288	NumOfProducts 10000.000000 1.530200	HasCrCard 10000.00000 0.70550	IsActiveMember 10000.000000 0.515100	EstimatedSalary 10000.000000 100090.239881
count 1000 mean 500 std 288 min 25% 250	0000.00000	1.000000e+04 1.569094e+07	10000.000000 650.528800	10000.000000	10000.000000	10000.000000	10000.000000	10000.00000	10000.000000	10000.000000
mean 500 std 288 min 25% 250	5000.50000	1.569094e+07	650.528800							
std 288 min 25% 250				38.921800	5.012800	76485.889288	1.530200	0.70550	0.515100	100090.239881
min 25% 250	2000 00500	7.4020400+04							3.0.10.100	
25 % 250	2886.89568	7.1936196+04	96.653299	10.487806	2.892174	62397.405202	0.581654	0.45584	0.499797	57510.492818
	1.00000	1.556570e+07	350.000000	18.000000	0.000000	0.000000	1.000000	0.00000	0.000000	11.580000
50% 500	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	0.000000	1.000000	0.00000	0.000000	51002.110000
	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	97198.540000	1.000000	1.00000	1.000000	100193.915000
75 % 750	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	127644.240000	2.000000	1.00000	1.000000	149388.247500
max 1000	0000 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"])
```

1. Handle the Missing values.

```
[ ] data = pd.read_csv("Churn_Modelling.csv")
     pd.isnull(data["Gender"])
            False
            False
     1
            False
     3
            False
     4
            False
     9995
            False
     9996
            False
     9997
            False
     9998
            False
     9999
            False
    Name: Gender, Length: 10000, dtype: bool
```

Question-6:

Find the outliers and replace the outliers.

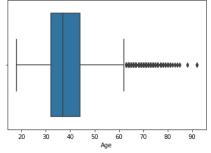
SOLUTION:

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

1. Find the outliers and replace the outliers.

```
[ ] 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 <matplotlib.axes._subplots.AxesSubplot at 0x7f6d4aaf7110>



SOLUTION:

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

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

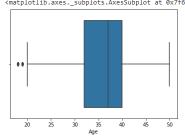
```
0
        42
        41
1
2
        42
3
        39
4
        43
9995
        39
9996
        35
9997
        36
9998
        42
9999
        28
Name: Age, Length: 10000, dtype: int64
```

SOLUTION:

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

```
[ ] 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 0x7f6d4b7e8990>

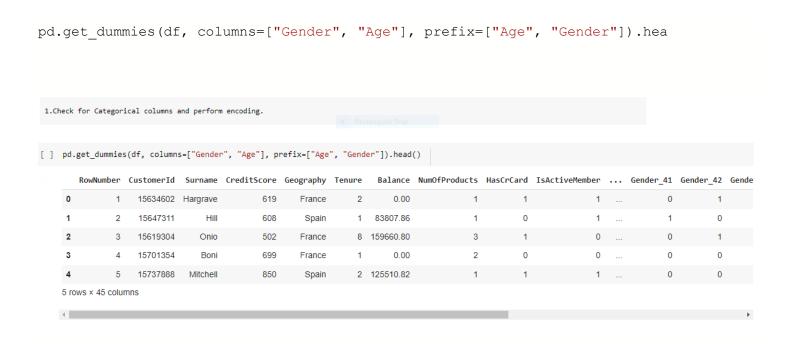


```
df['Age']=np.where(df['Age']<20,35,df['Age'])
df['Age']</pre>
```

```
[] df['Age']=np.where(df['Age']<20,35,df['Age'])
    df['Age']
    0
            42
            41
            42
            39
            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.



1.Ch	eck for Cate	gorical colu	mns and perform en	ncodin	g.									
0	pd.get_dummies(df, columns=["Gender", "Age"], prefix=["Age", "Gender"]).head()													
9 1	OfProducts	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
	4													>

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)
```

a) Split the data into Independent variables.

```
[] 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

SOLUTION:

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

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

```
import pandas as pd
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
df[["CustomerId"]] = scaler.fit_transform(df[["CustomerId"]])
print(df)
```

```
import pandas as pd
   from sklearn.preprocessing import MinMaxScaler
    scaler = MinMaxScaler()
    df[["CustomerId"]] = scaler.fit_transform(df[["CustomerId"]])
                                                                + Code + Text
[ ] print(df)
                                Surname CreditScore Geography
          RowNumber CustomerId
                                                               Gender Age \
                     0.275616 Hargrave
                                                 619
                                                        France
                                                               Female
                                                                        42
                                 Hill
                      0.326454
                                                 608
                                                         Spain
                                                               Female
                                                                        41
    1
                                          502
699
850
                     0.214421
                                    Onio
                                                        France Female
                                                                        42
                     0.542636
                                    Boni
                                                        France Female
                                                                        39
                5 0.688778 Mitchell
    4
                                                        Spain Female 43
                                          771
516
709
772
792
                   0.162119 Obijiaku
0.016765 Johnstone
0.075327 Liu
0.466637 Sabbatini
    9995
              9996
                                                       France
                                                                 Male
                                                                       39
    9996
              9997
                                                        France
                                                                 Male
                                                      France Female
    9997
             9998
    9998
             9999
                                                 772
                                                       Germany
                                                                Male
                                                                        42
            10000 0.250483
    9999
                                Walker
                                                       France Female
          Tenure Balance NumOfProducts HasCrCard IsActiveMember \
    0
                     0.00
              1 83807.86
    1
             8 159660.80
                     0.00
                                     1
                                               1
             2 125510.82
                                                                1
                                              1
1
0
    9995
                     0.00
                                                                0
    9996
             10 57369.61
                                      1
                                                                1
    9997
            7 0.00
3 75075.31
                                                                1
                                                1
    9998
    9999
             4 130142.79
         EstimatedSalary Exited
    0
              101348.88
               112542.58
               113931.57
                93826.63
                79084.10
                             0
                96270.64
                            0
    9995
               101699.77
```

Question-10:

Split the data into training and testing

1. List item

(1000, 13) (1000,) (None, None)

```
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)
```

```
[ ] 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,)
```