

## ASSIGNMENT – 2

### Python Programming

Assignment Date	26-09-2022
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Question-1 :

#### 1 . Importing Required Package

**Solution :**

```
import pandas as pd
import seaborn as sns
import numpy as np
from matplotlib import pyplot as plt
%matplotlib inline
```

Question-2 :

#### 2. Loading the Dataset

**Solution :**

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

df
```

**Output:**



	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.64	0
9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101699.77	0
9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	42085.58	1
9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	75075.31	2	1	0	92888.52	1
9999	10000	15628319	Walker	792	France	Female	28	4	130142.79	1	1	0	38190.78	0

10000 rows x 14 columns

### 3. Visualizations

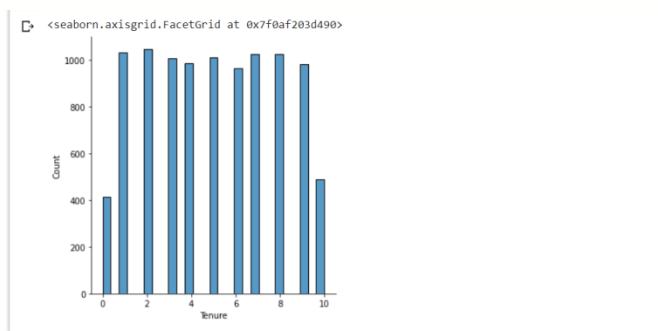
Question-3 :

#### 3.1 Univariate Analysis

**Solution:**

```
sns.displot(df.Tenure)
```

**Output:**

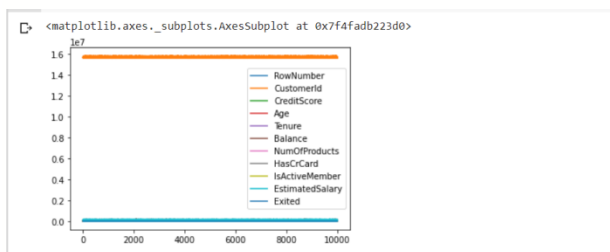


#### 3.2 Bi-Variate Analysis

**Solution:**

```
df.plot.line()
```

**Output:**

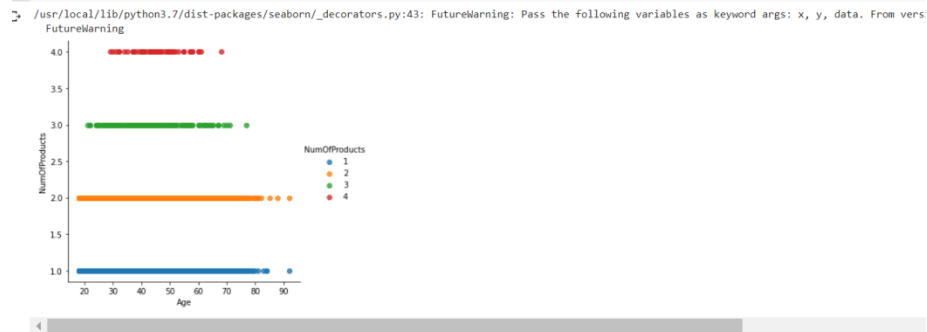


### 3.3 Multi - Variate Analysis

#### Solution:

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

#### Output:



### 4. Perform descriptive statistics on the dataset.

#### Question-4 :

#### Solution:

```
df.describe()
```

#### Output:

	RowNumber	CustomerId	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
count	10000.00000	1.000000e+04	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.00000	10000.000000	10000.000000	10000.000000
mean	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	76485.889288	1.530200	0.70550	0.515100	100090.239881	0.203700
std	2886.89568	7.193619e+04	96.653299	10.487806	2.892174	62397.405202	0.581654	0.45584	0.499797	57510.492818	0.402769
min	1.00000	1.556570e+07	350.000000	18.000000	0.000000	0.000000	1.000000	0.00000	0.000000	11.580000	0.000000
25%	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	0.000000	1.000000	0.00000	0.000000	51002.110000	0.000000
50%	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	97198.540000	1.000000	1.00000	1.000000	100193.915000	0.000000
75%	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	127644.240000	2.000000	1.00000	1.000000	149388.247500	0.000000
max	10000.00000	1.581569e+07	850.000000	92.000000	10.000000	250898.090000	4.000000	1.00000	1.000000	199992.480000	1.000000

## 5. Handle the Missing values.

Question-5 :

**Solution:**

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

**Output:**

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

Question-6:

## 6. Find the outliers and replace the outliers.

**Solution:**

```
df["Tenure"] = np.where(df["Tenure"] > 10, np.median(df["Tenure"]))
df["Tenure"]
```

**Output:**

```
0      2
1      1
2      8
3      1
4      2
...
9995    5
9996   10
9997    7
9998    3
9999    4
Name: Tenure, Length: 10000, dtype: object
```

## Question-7 :

### 7. Check for Categorical columns and perform encoding.

#### Solution:

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

#### Output:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	...	Gender_78
0	1	15634602	Hargrave	619	France	2	0.00	1	1	1	...	0
1	2	15647311	Hill	608	Spain	1	83807.86	1	0	1	...	0
2	3	15619304	Onio	502	France	8	159660.80	3	1	0	...	0
3	4	15701354	Boni	699	France	1	0.00	2	0	0	...	0
4	5	15737888	Mitchell	850	Spain	2	125510.82	1	1	1	...	0

5 rows × 84 columns

#### Output:

	HasCrCard	IsActiveMember	...	Gender_78	Gender_79	Gender_80	Gender_81	Gender_82	Gender_83	Gender_84	Gender_85	Gender_88	Gender_92
0	1	1	...	0	0	0	0	0	0	0	0	0	0
1	0	1	...	0	0	0	0	0	0	0	0	0	0
2	1	0	...	0	0	0	0	0	0	0	0	0	0
3	0	0	...	0	0	0	0	0	0	0	0	0	0
4	1	1	...	0	0	0	0	0	0	0	0	0	0

Question-8:

## 8. Split the data into dependent and independent variables

### 8.1 Split the data into Independent variables.

**Solution:**

```
X = df.iloc[:, :-2].values  
print(X)
```

**Output:**

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

### 8.2 Split the data into Dependent variables.

**Solution:**

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

**Output:**

```
[1 0 1 ... 1 1 0]
```

## Question-9 :

### 9. Scale the independent variables

#### Solution:

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

#### Output:

```
RowNumber  CustomerId  Surname  CreditScore  Geography  Gender  Age \
0          0.00000  15634602  Hargrave        619      France  Female  42
1          0.00001  15647311    Hill         688      Spain  Female  41
2          0.00002  15619304    Onio         582      France  Female  42
3          0.00003  15701354    Boni         699      France  Female  39
4          0.00004  15737888  Mitchell        850      Spain  Female  43
...
9995      0.99996  15686229  Obijaku         771      France   Male   39
9996      0.99997  15569892  Johnstone    516      France   Male   35
9997      0.99998  15584532    Liu         709      France  Female  36
9998      0.99999  15682355  Sabbatini    772      Germany  Male   42
9999      1.00000  15628319    Walker        792      France  Female  28

Tenure      Balance  NumOfProducts  HasCrCard  IsActiveMember \
0           2         0.00             1           1             1
1           1      83807.86             1           0             1
2           8     159660.80             3           1             0
3           1         0.00             2           0             0
4           2     125510.82             1           1             1
...
9995         5         0.00             2           1             0
9996        10     57369.61             1           1             1
9997         7         0.00             1           0             1
9998         3      75075.31             2           1             0
9999         4     130142.79             1           1             0

EstimatedSalary  Exited
0          101348.88         1
1          112542.58         0
2          113931.57         1
3           93826.63         0
4           79084.10         0
...
9995          96270.64         0
9996         101699.77         0
9997          42085.58         1
9998          92888.52         1
9999          38190.78         0

[10000 rows x 14 columns]
```

## Question-10 :

### 10. Split the data into training and testing

#### Solution:

```
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)
```

#### Output:

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