

## ▼ 1.Importing the Requiured Package

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

## ▼ 2.Loading the Dataset

```
import pandas as pd

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

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfPro
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	
2	3	15619304	Onio	502	France	Female	42	8	159660.80	

## ▼ 3. Visualization

### 3.1 Univariate Analysis

```
import seaborn as sns
import pandas as pd
```

```
sns.displot(df.Gender)
```

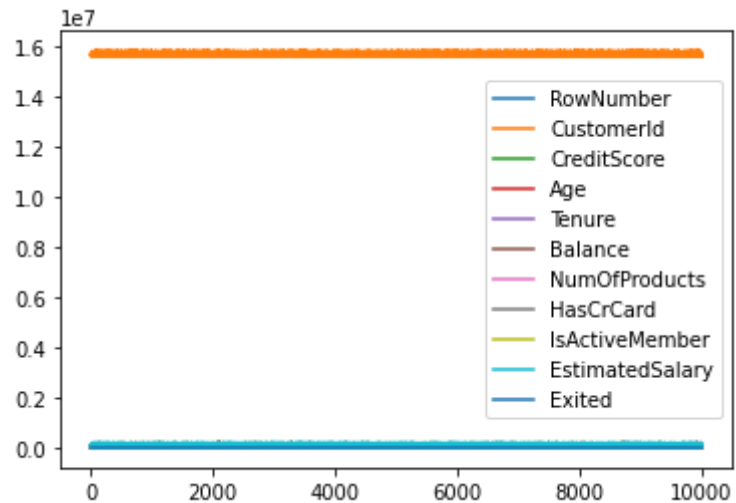
```
<seaborn.axisgrid.FacetGrid at 0x7fbce8c90810>
```

### 3.2 Bi-Variate Analysis

4000 | 

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

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fbce5595210>
```



### 3.3 Multi-Variate Analysis

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





## ▼ 4. Perform description statics on the dataset

```
df.describe()
```

	RowNumber	CustomerId	CreditScore	Age	Tenure	Balance	NumOfProducts
<b>count</b>	10000.00000	1.000000e+04	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000
<b>mean</b>	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	76485.889288	1.530200
<b>std</b>	2886.89568	7.193619e+04	96.653299	10.487806	2.892174	62397.405202	0.581654
<b>min</b>	1.00000	1.556570e+07	350.000000	18.000000	0.000000	0.000000	1.000000
<b>25%</b>	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	0.000000	1.000000
<b>50%</b>	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	97198.540000	1.000000
<b>75%</b>	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	127644.240000	2.000000
<b>max</b>	10000.00000	1.581569e+07	850.000000	92.000000	10.000000	250898.090000	4.000000

## ▼ 5.Handle the Missing values

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

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
```

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

```
import numpy as np

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

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
```

## 7. Check for Categorical columns and perform encoding

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

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

5 rows × 84 columns



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

### 8.1. Split the data into Independent variables

```
#independant
```

```
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]]
```

## ▼ 8.2.Split the data into Dependent variables

```
#dependant
y = df.iloc[:, :-1].values
print(y)
```

```
[[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]]
```

## ▼ 9.Scale the independent variables

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

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

```
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,y,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,)
(5000, 13)
(5000,)
(5000, 13)
(5000,)
(None, None)
```



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