## Importing Package

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

## 1.Loading dataset

df =pd.read\_csv("/content/Churn\_Modelling.csv")

df

a i							
Λαο	RowNumb	er Custon	nerId	Surname	CreditScore	Geography	Gender
Age 0 42	\	1 0.27	5616	Hargrave	619	France	Female
1		2 0.32	26454	Hill	608	Spain	Female
41 2		3 0.21	.4421	Onio	502	France	Female
42 3		4 0.54	2636	Boni	699	France	Female
39 4 43		5 0.68	88778	Mitchell	850	Spain	Female
43							
9995 39	99	96 0.16	2119	0bijiaku	771	France	Male
9996 35	99	97 0.01	.6765	Johnstone	516	France	Male
9997	99	98 0.07	5327	Liu	709	France	Female
36 9998 42	99	99 0.46	6637	Sabbatini	772	Germany	Male
9999 28	100	00 0.25	0483	Walker	792	France	Female
0 1 2 3 4	Tenure 2 1 8 1 2	Balance 0.00 83807.80 159660.80 0.00 125510.82	) ; )	nOfProducts 1 1 3 2 1	HasCrCard 1 0 1 0 1	IsActiveMem	ber \ 1
9995 9996 9997 9998	5 10 7 3	0.00 0.00 57369.61 0.00 75075.31	- )	 2 1 1 2	 1 1 0 1		 0 1 1

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

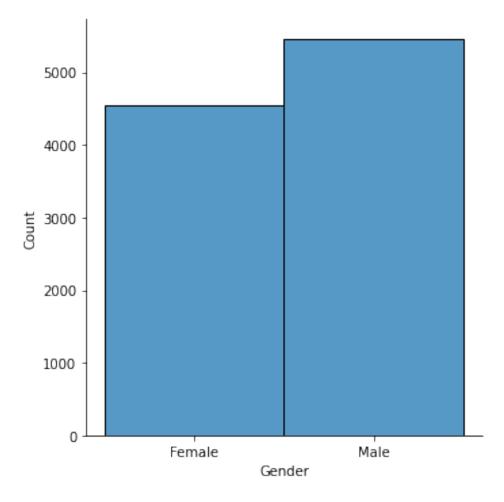
[10000 rows x 14 columns]

Visualization

a) Univariate analysis

sns.displot (df.Gender)

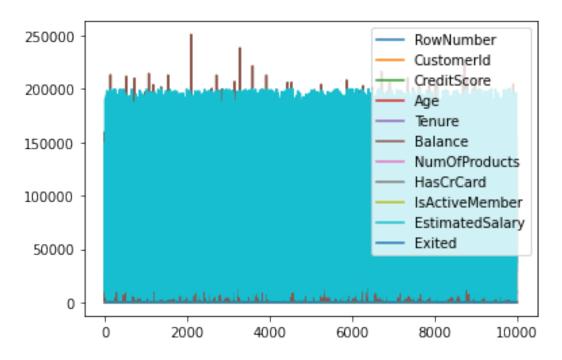
<seaborn.axisgrid.FacetGrid at 0x7fa2127ec990>



b) Bi-Variate Analysis

df.plot.line()

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fa21262e890>

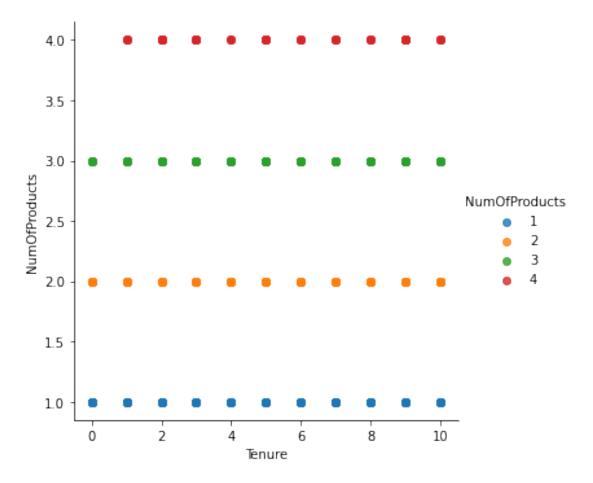


## c) Multi Variate Analysis

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 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning



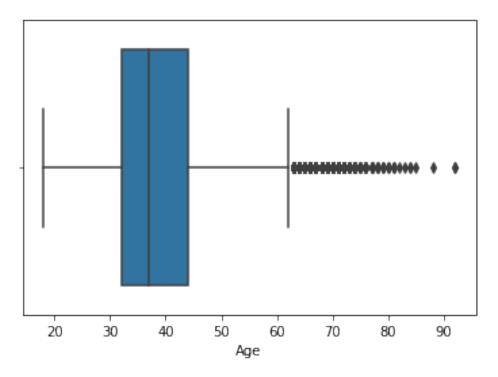
Perform descriptive statistics on the dataset df.describe()

RowNumber	CustomerId	CreditScore	Age			
Tenure \						
count 10000.00000	10000.000000	10000.000000	10000.000000			
10000.000000						
mean 5000.50000	0.500980	650.528800	36.533900			
5.012800						
std 2886.89568	0.287757	96.653299	6.473843			
2.892174						
min 1.00000	0.000000	350.000000	20.000000			
0.000000						
25% 2500.75000	0.251320	584.000000	32.000000			
3.000000						
50% 5000.50000	0.500170	652.000000	37.000000			
5.000000						
75% 7500.25000	0.750164	718.000000	40.000000			
7.000000						
max 10000.00000	1.000000	850.000000	50.000000			
10.000000						

```
NumOfProducts
                                         HasCrCard
             Balance
                                                     IsActiveMember
        10000.000000
                        10000.000000
                                       10000.00000
                                                       10000.000000
count
mean
        76485.889288
                            1.530200
                                           0.70550
                                                           0.515100
        62397.405202
                            0.581654
                                           0.45584
                                                           0.499797
std
min
            0.000000
                            1.000000
                                           0.00000
                                                           0.000000
25%
            0.000000
                            1.000000
                                           0.00000
                                                           0.000000
50%
        97198.540000
                            1.000000
                                           1.00000
                                                           1.000000
75%
       127644.240000
                            2,000000
                                           1.00000
                                                           1.000000
       250898.090000
                            4.000000
                                           1.00000
                                                           1.000000
max
       EstimatedSalary
                               Exited
          10000.000000
                         10000.000000
count
         100090.239881
mean
                             0.203700
std
          57510.492818
                             0.402769
min
              11.580000
                             0.000000
25%
          51002.110000
                             0.000000
50%
         100193.915000
                             0.000000
75%
                             0.000000
         149388.247500
         199992.480000
                             1.000000
max
Handle the missing values
data = pd.read csv("/content/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
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 only valid positional argument will be `data`, and
passing other arguments without an explicit keyword will result in an
error or misinterpretation.
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fa21390b290>

FutureWarning



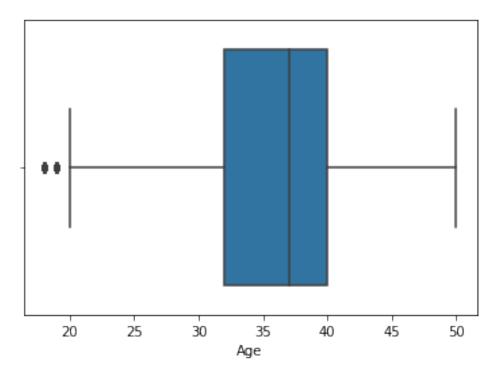
```
df['Age']=np.where(df['Age']>50,40,df['Age'])
df['Age']
0
        42
1
        41
2
        42
3
        39
        43
        39
9995
9996
        35
        36
9997
9998
        42
9999
        28
Name: Age, Length: 10000, dtype: int64
```

/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 only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an

error or misinterpretation. FutureWarning

sns.boxplot(df['Age'])

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fa213879fd0>



```
df['Age']=np.where(df['Age']<20,35,df['Age'])</pre>
df['Age']
0
         42
1
        41
2
        42
3
         39
         43
         . .
39
9995
9996
         35
9997
         36
9998
         42
9999
         28
Name: Age, Length: 10000, dtype: int64
```

Check for categorical Columns and perform encoding

pd.get\_dummies(df,columns=["Gender","Age"],prefix=["Age","Gender"]).he
ad()

RowNumb	er	CustomerId	Surname	CreditScore	Geography	Tenure
Balance \	1	0.275616	Hargrave	619	France	2
0.00	_	0.1270020	9	0_0		_
1	2	0.326454	Hill	608	Spain	1
83807.86 2 159660.80	3	0.214421	Onio	502	France	8
3	4	0.542636	Boni	699	France	1

```
0.00
                  0.688778 Mitchell
                                                 850
                                                          Spain
                                                                        2
4
            5
125510.82
   NumOfProducts HasCrCard
                                                        Gender_41 Gender_42
                                IsActiveMember
                                                 . . .
\
0
                                                                 0
                1
                             1
                                               1
                                                                             1
                                                   . . .
                1
                             0
                                                                             0
1
                                               1
                                                                 1
2
                3
                             1
                                               0
                                                                 0
                                                                             1
3
                2
                             0
                                                                             0
                                                                 0
                1
                             1
                                               1
                                                                             0
4
                                                                 0
                                                  . . .
   Gender_43
               Gender 44
                            Gender_45
                                        Gender_46
                                                    Gender 47
                                                                 Gender 48
                                     0
0
            0
                        0
                                                 0
                                                              0
            0
                        0
                                     0
                                                 0
                                                              0
1
                                                                          0
2
            0
                        0
                                                 0
                                     0
                                                              0
3
            0
                        0
                                     0
                                                 0
                                                              0
                                                                          0
4
            1
                        0
                                     0
                                                              0
                                                                          0
   Gender_49
               Gender_50
0
            0
            0
                        0
1
2
            0
                        0
3
            0
                        0
[5 rows x 45 columns]
Split the data into dependent and independent Variables
a) Split the data into independent Variables
X = df.iloc[:, :-1].values
print(X)
[[1 0.2756161271095934 'Hargrave' ... 1 1 101348.88]
 [2 0.32645436399201344 'Hill' ... 0 1 112542.58]
 [3 0.21442143454311946 'Onio' ... 1 0 113931.57]
 [9998 0.07532731440183227 'Liu' ... 0 1 42085.58]
 [9999 0.4666365320074064 'Sabbatini' ... 1 0 92888.52]
 [10000 0.25048302125293276 '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]

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

print(df)							
۸۵۵	RowNumb	er Custom	erId	Surname	CreditScore	Geography	Gender
Age 0	\	1 0.27	5616	Hargrave	619	France	Female
42 1		2 0.32	6454	Hill	608	Spain	Female
41		3 0.21	4421	Onio	502	? France	Female
42 3		4 0.54	2636	Boni	699	France	Female
39 4		5 0.68	8778	Mitchell	850	) Spain	Female
43 							
9995	999	96 0.16	2119	0bijiaku	771	. France	Male
39 9996	999	97 0.01	6765	Johnstone	516	5 France	Male
35 9997	999	98 0.07	5327	Liu	709	France	Female
36 9998	999	99 0.46	6637	Sabbatini	772	ermany	Male
42 9999	100	00 0.25	9483	Walker	792	? France	Female
28							
0	Tenure 2	Balance 0.00	Num	OfProducts	HasCrCard	IsActiveMem	
0 1	1	83807.86		1 1	1 0		1 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	ī		ĺ
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
                               0
1
             112542.58
2
             113931.57
                               1
3
              93826.63
                               0
4
              79084.10
                               0
                             . . .
              96270.64
9995
                               0
9996
             101699.77
                               0
                               1
9997
              42085.58
                               1
9998
              92888.52
              38190.78
                               0
9999
[10000 \text{ rows } \times 14 \text{ columns}]
Split the data into training and testing
from sklearn.model selection import train test split
train size=0.8
X = d\overline{f}.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)