

ASSIGNMENT 2

MAHENDRA ENGINEERING COLLEGE

FOR WOMEN

NAME :R. Sritha

CLASS :4 year ece

SUBJECT:IBM

REGISTER NO:611419106066

#libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

#load dataset

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

df.head(10)

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age
0	1	15634602	Hargrave	619	France	Female	42
1	2	15647311	Hill	608	Spain	Female	41
2	3	15619304	Onio	502	France	Female	42
3	4	15701354	Boni	699	France	Female	39
4	5	15737888	Mitchell	850	Spain	Female	43
5	6	15574012	Chu	645	Spain	Male	44
6	7	15592531	Bartlett	822	France	Male	50
7	8	15656148	Obinna	376	Germany	Female	29
8	9	15792365	He	501	France	Male	44
9	10	15592389	H?	684	France	Male	27

	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	
5	8	113755.78	2	1	0	
6	7	0.00	2	1	1	
7	4	115046.74	4	1	0	

```
3      93826.63      0
4      79084.10      0
5      149756.7      1
6      10062.80      0
7      119346.8      1
8      74940.50      0
9      71725.73      0
```

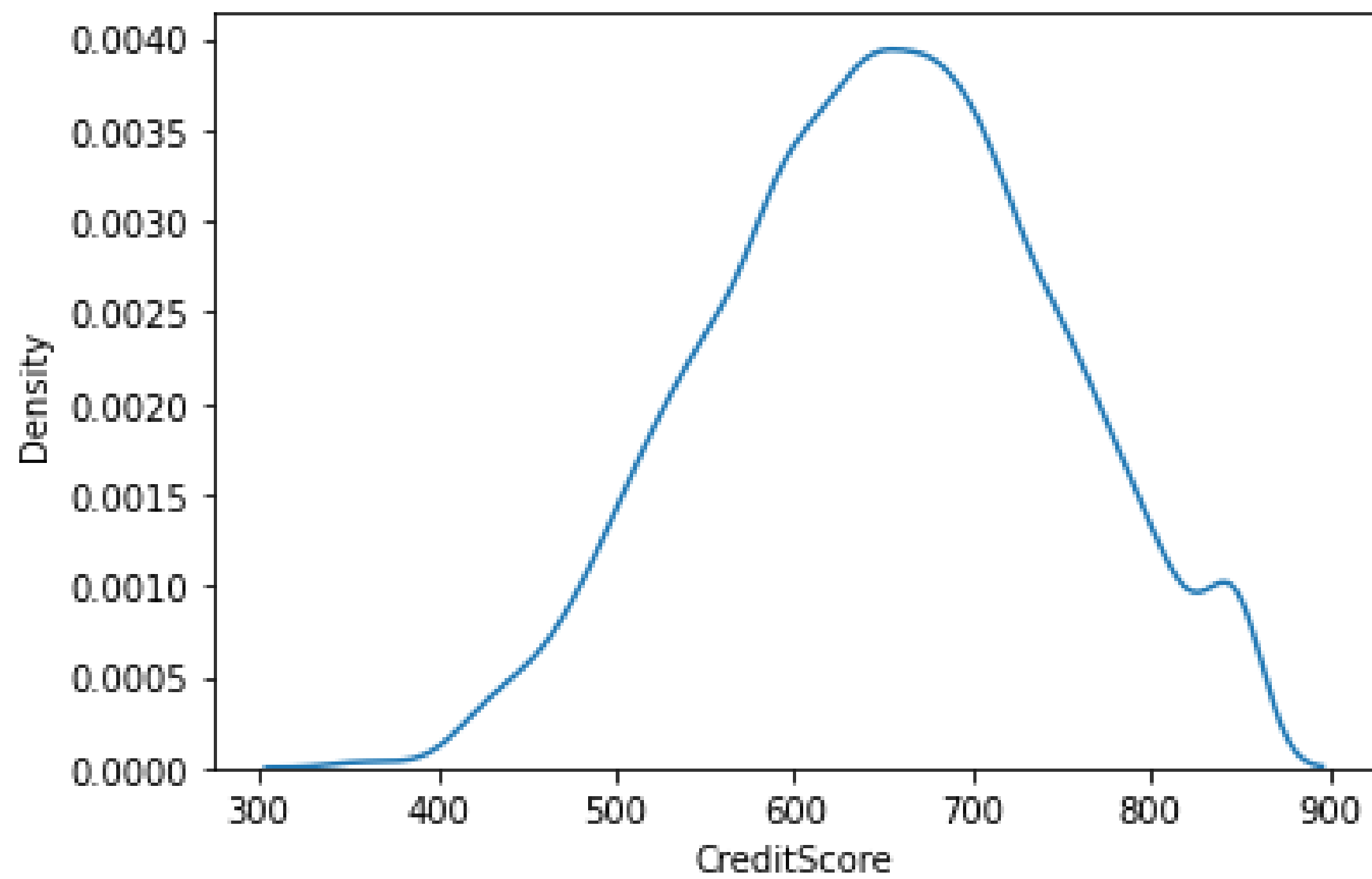
```
df.info(
)
```

```
<class
'pandas.core.frame.DataFrame'>RangeIn
dex:10000entries,0to9999Datacolumns(t
otal 14 columns):
#   Column      Non-NullCountDtype
---  ---
0  RowNumber    10000non-nullint64
1  CustomerId   10000non-nullint64
2  Surname      10000non-null  object
3  CreditScore  10000non-nullint64
4  Geography    10000non-nullobject
5  Gender       10000non-nullobject
6  Age          10000non-nullint64
7  Tenure       10000non-nullint64
8  Balance      10000non-nullfloat64
9  NumOfProducts 10000non-nullint64
10 HasCrCard    10000non-nullint64
11 IsActiveMember 10000non-nullint64
12 EstimatedSalary10000non-null float64
13 Exited      10000
non-nullint64dtypes: float64(2),
int64(9),object(3)memoryusage:1.1+MB
```

```
#Visualizations#Univariate Analysisimport
seabornas sns
```

```
sns.kdeplot(df['CreditScore'])
```

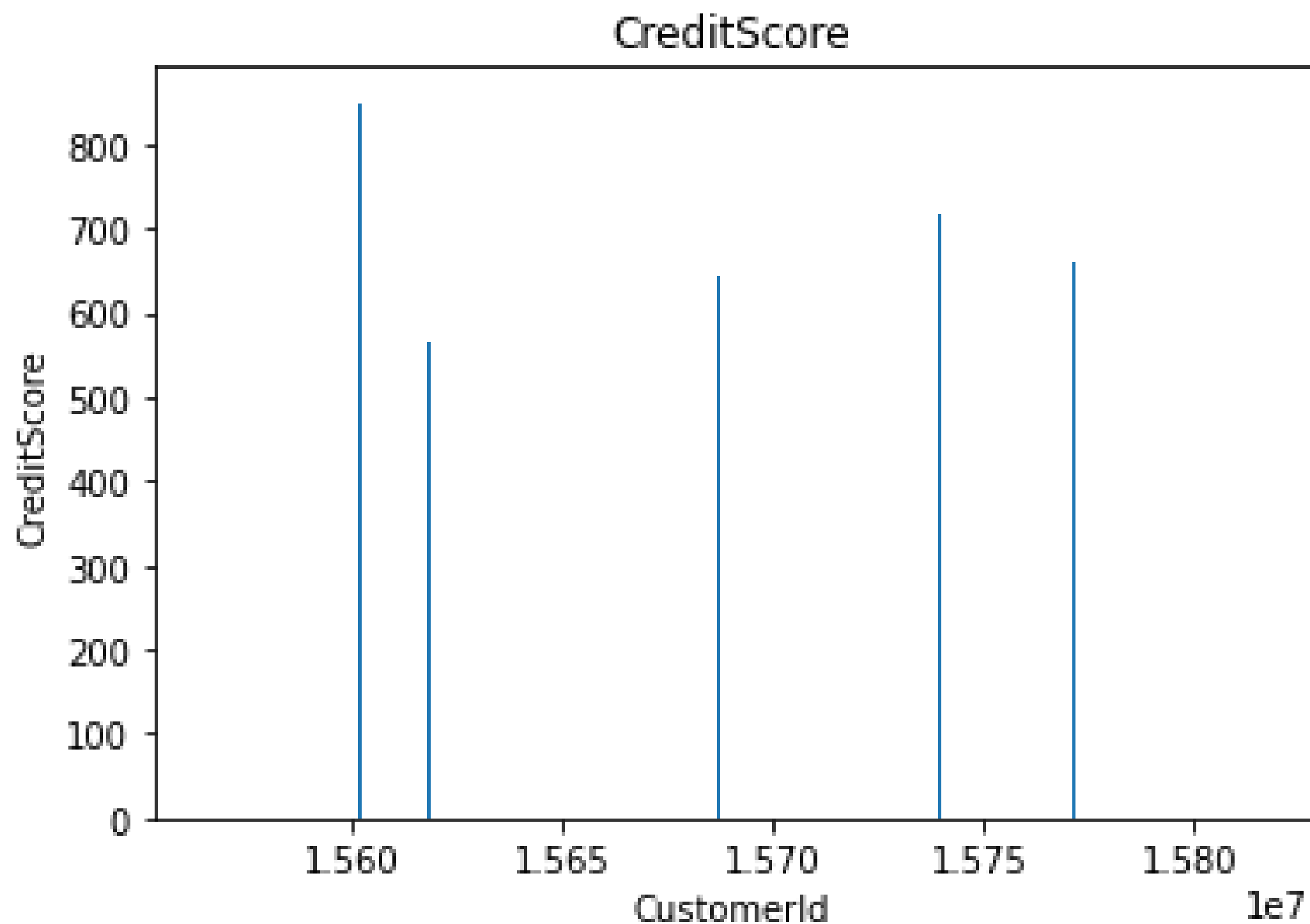
```
<matplotlib.axes._subplots.AxesSubplotat0x7fc4a0cd2790>
```



*#Bi-VariateAnalysis*

```
plt.bar(df.CustomerId,  
df.CreditScore)plt.title('CreditScore'  
)plt.xlabel('CustomerId')plt.ylabel('C  
reditScore')
```

```
Text(0,0.5,'CreditScore')
```

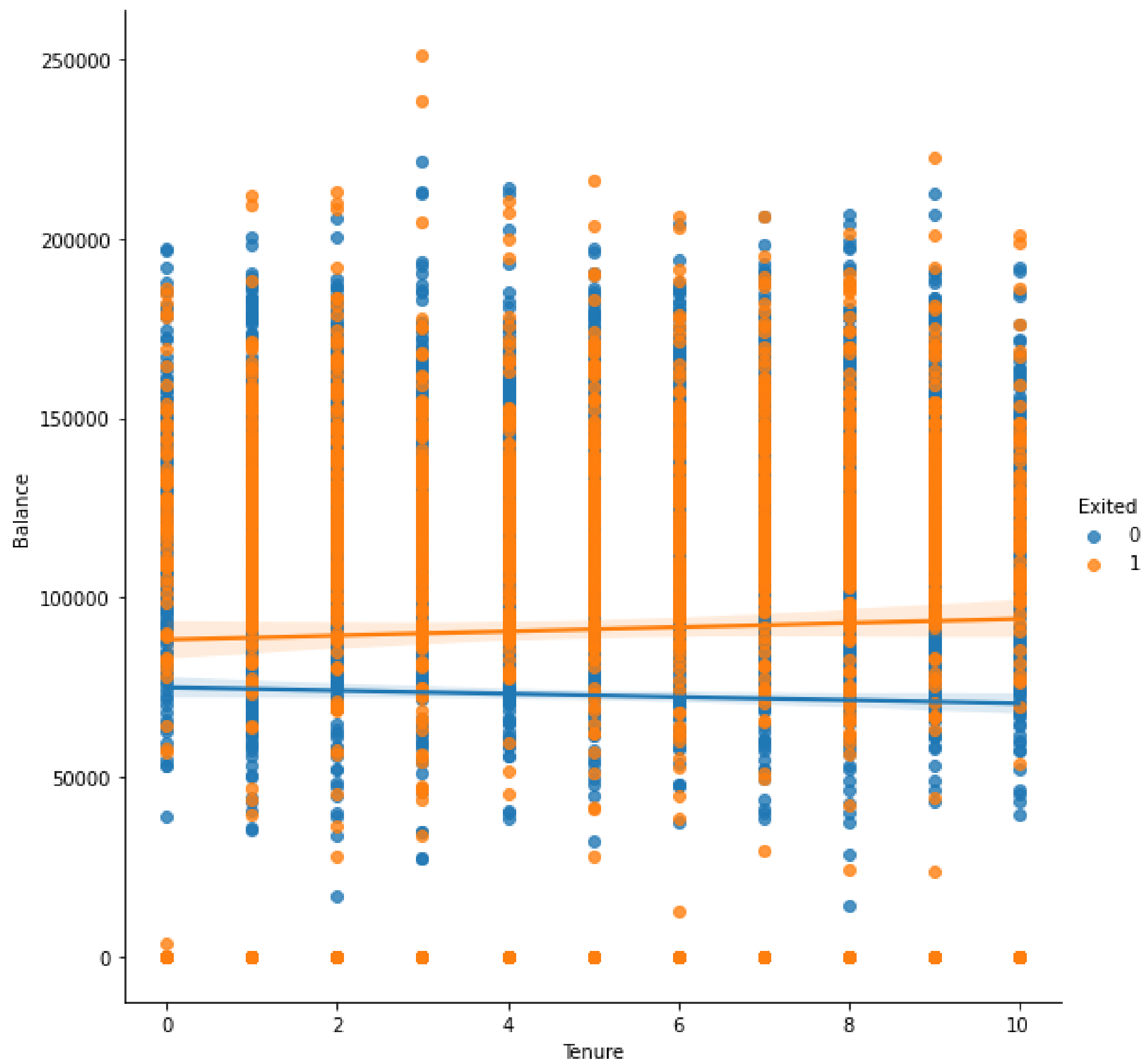


```
sns.lmplot(x='Tenure', y='Balance',data=df,hue='Exited',size=8)
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/regression.py:581:UserWarning: The `size` parameter has been renamed to `height`;  
pleaseupdateyourcode.
```

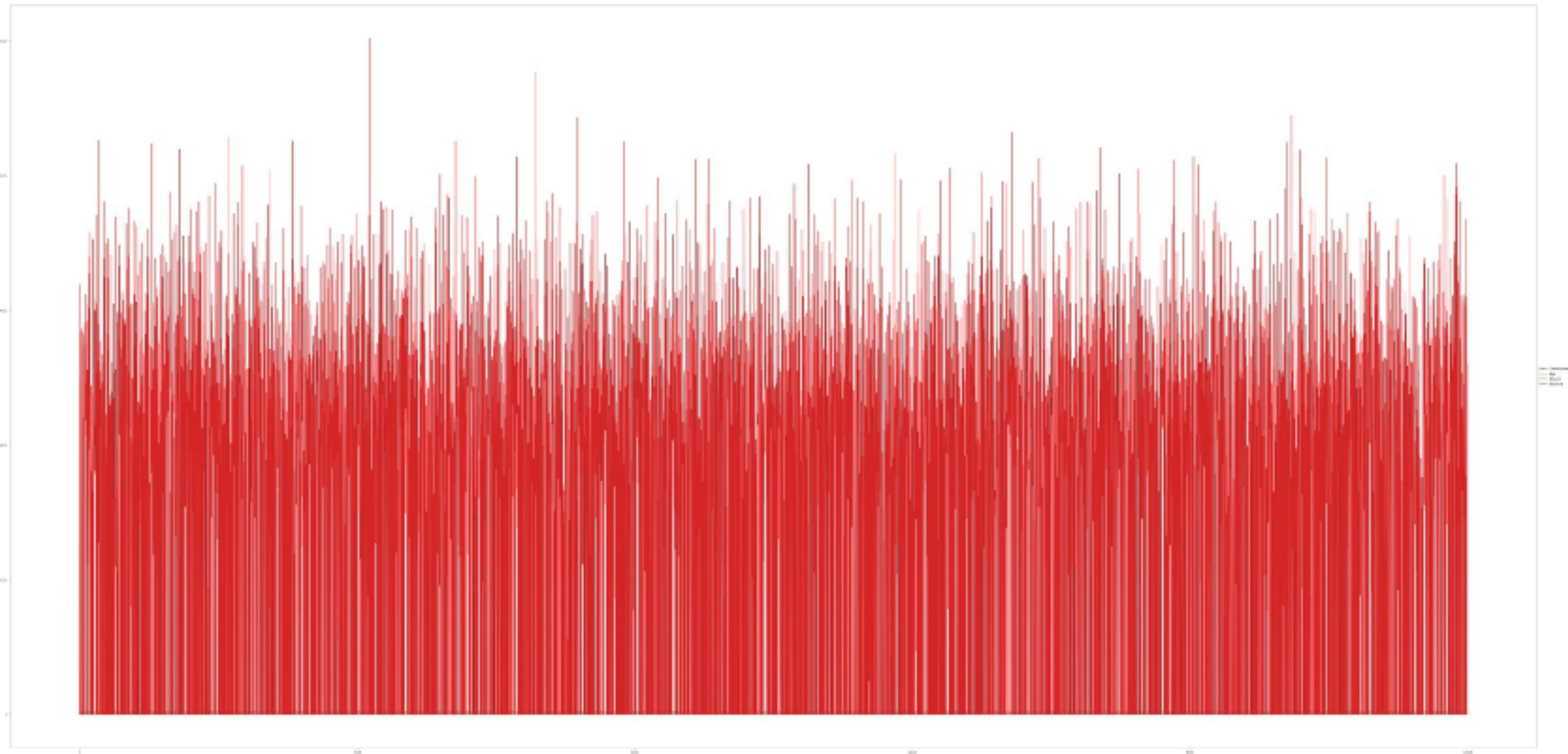
```
warnings.warn(msg,UserWarning)
```

```
<seaborn.axisgrid.FacetGridat0x7fc4a149e2d0>
```



*#Multi -Variate Analysis*

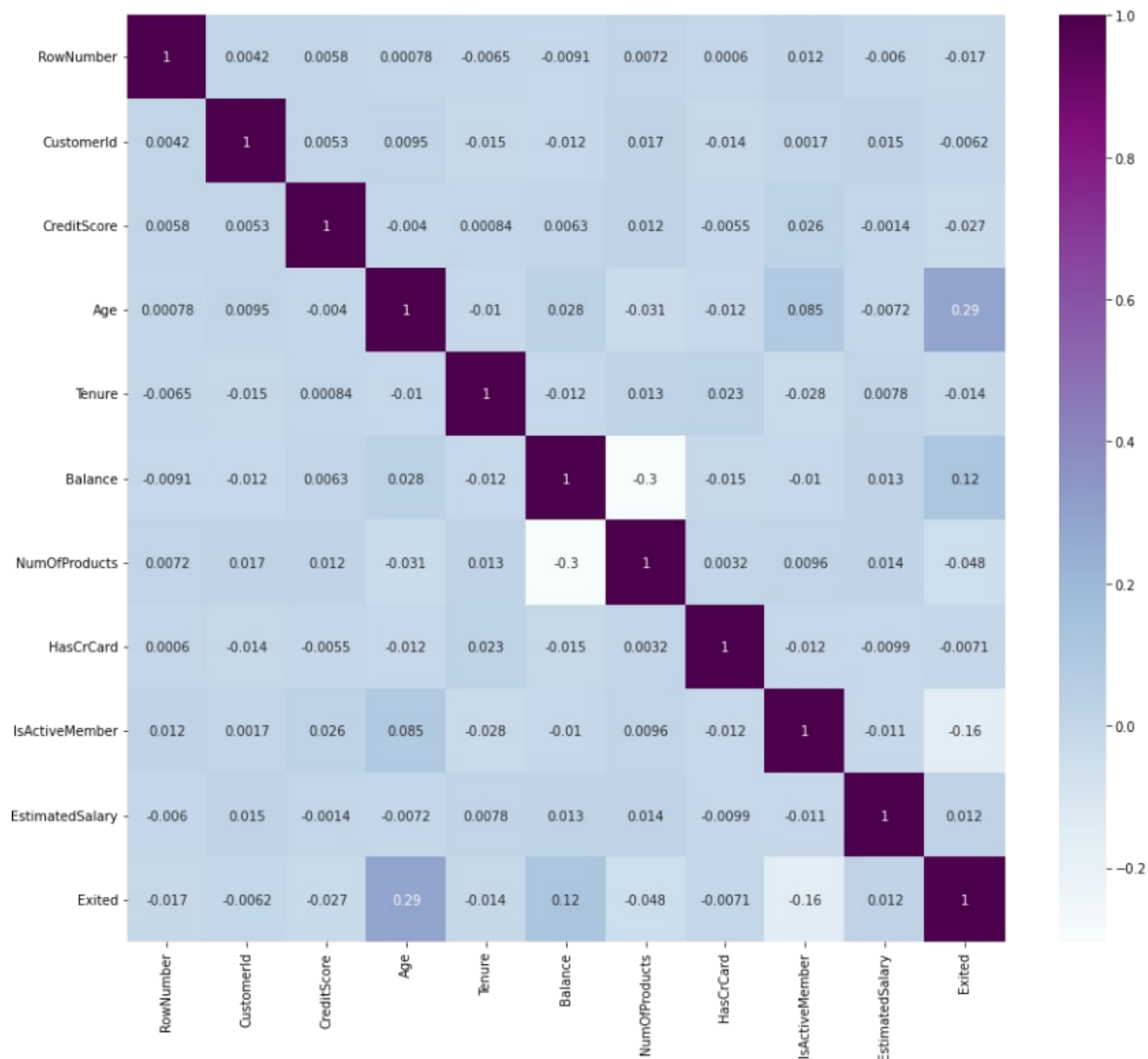
```
ax=df[["CreditScore","Age","Tenure","Balance"]].plot(figsize=(80,40))ax.legend(loc='centerleft',bbox_to_anchor=(1,0.5));
```



```
df.isnull().sum()
```

```
RowNumber      0
CustomerId     0
Surname        0
CreditScore    0
Geography      0
Gender         0
Age           0
Tenure        0
Balance        0
NumOfProducts 0
HasCrCard      0
IsActiveMember 0
EstimatedSalary 0
Exited         0
dtype: int64
```

```
plt.figure(figsize=(15,13))sns.heatmap(df.corr(),anno
t=True,cmap='BuPu')plt.show()
```



```
df.drop(['RowNumber',
'CustomerId', 'Surname'],axis=1,inplace=True)df.head()

  CreditScore Geography  Gender  Age  Tenure  Balance
NumOfProducts \
0          619    France  Female   42     2     0.00
1          608     Spain  Female   41     1  83807.86
1          502    France  Female   42     8 159660.80
2          699    France  Female   39     1     0.00
3          850     Spain  Female   43     2 125510.82
4
1

HasCrCardIsActiveMemberEstimatedSalaryExited0    1
1          101348.881
```







0	619	France	Female	42	2	0.00
1						
1	608	Spain	Female	41	1	83807.86
1						
2	502	France	Female	42	8	159660.80
3						
3	699	France	Female	39	1	0.00
2						
4	850	Spain	Female	43	2	125510.82
1						
...	...	...	...	...	...	...
...						
9995	771	France	Male	39	5	0.00
2						
9996	516	France	Male	35	10	57369.61
1						
9997	709	France	Female	36	7	0.00
1						
9998	772	Germany	Male	42	3	75075.31
2						
9999	792	France	Female	28	4	130142.79
1						

HasCrCardIsActiveMemberEstimatedSalaryExitedFranceGermany\						
0	1	1	101348.88	1	1	
0						
1	0	1	112542.58	0	0	
0						
2	1	0	113931.57	1	1	
0						
3	0	0	93826.63	0	1	
0						
4	1	1	79084.10	0	0	
0						
...	...	...	...	...	...	...
...						
995	1	0	96270.64	0	1	
0						
9996	1	1	101699.77	0	1	
0						
9997	0	1	42085.58	1	1	
0						
9998	1	0	92888.52	1	0	
1						
9999	1	0	38190.78	0	1	
0						
0	Spain	Female	Male			
	0	1	0			

1	1	1	0
2	0	1	0
3	0	1	0
4	1	1	0
...	...	...	...
9995	0	0	1
9996	0	0	1
9997	0	1	0
9998	0	0	1
9999	0	1	0

```
[10000rowsx16columns]df.drop(["Geography", "Gender"],
axis=1,inplace=True)df.head()
```

	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard\0	61942
	2			0.00	1	1	
1	608	41		1	83807.86	1	0
2	502	42		8159660.80		3	1
3	699	39		1	0.00	2	0
4	850	43		2125510.82		1	1

	IsActiveMember	EstimatedSalary	Exited	France	Germany	Spain	Female\
0	1	101348.88	1	1		0	0
1							
1	1	112542.58	0	0		0	1
1							
2	0	113931.57	1	1		0	0
1							
3	0	93826.63	0	1		0	0
1							
4	1	79084.10	0	0		0	1
1							

	Male
0	0
1	0
2	0
3	0
4	0

```
x=df.drop('Exited',axis=1)
```

x

	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	\
0	619	42	2	0.00	1	1	
1	608	41	1	83807.86	1	0	

2	502	42	8	159660.80	3	1
3	699	39	1	0.00	2	0
4	850	43	2	125510.82	1	1
...	...	...	...	...	...	...
9995	771	39	5	0.00	2	1
9996	516	35	10	57369.61	1	1
9997	709	36	7	0.00	1	0
9998	772	42	3	75075.31	2	1
9999	792	28	4	130142.79	1	1

	IsActiveMember	EstimatedSalary	France	Germany	Spain	Female
Male						
0	1	101348.88	1	0	0	1
0						
1	1	112542.58	0	0	1	1
0						
2	0	113931.57	1	0	0	1
0						
3	0	93826.63	1	0	0	1
0						
4	1	79084.10	0	0	1	1
0						
...	...	...	...	...	...	...
...						
995	0	96270.64	1	0	0	0
1						
9996	1	101699.77	1	0	0	0
1						
9997	1	42085.58	1	0	0	1
0						
9998	0	92888.52	0	1	0	0
1						
9999	0	38190.78	1	0	0	1
0						

[10000 rows x 13  
columns]y=df['Exited']

y	
0	1
1	0
2	1
3	0
4	0
...	..
9995	0
9996	0
9997	1
9998	1

```

9999      0
Name: Exited, Length: 10000, dtype:
int64df.shape
(10000, 14)
x.shape(10000,
13)
y.shape(
10000,)

from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test =
train_test_split(x, y, test_size=0.2, random_state=0)
x_train.shape
(8000, 13)
x_test.shape
(2000, 13)
y_test.shape
(2000,)

from sklearn.preprocessing import
StandardScaler
sc = StandardScaler()
x_train =
sc.fit_transform(x_train)
x_train
array([[0.16958176, -0.46460796, 0.00666099, ..., 1.74309049,
        1.09168714, -1.09168714],
       [-2.30455945, 0.30102557, -1.37744033, ..., -0.57369368,
        -0.91601335, 0.91601335],
       [-1.19119591, -0.94312892, -1.031415, ..., -0.57369368,
        1.09168714, -1.09168714],
       ...,
       [0.9015152, -0.36890377, 0.00666099, ..., -0.57369368,
        -0.91601335, 0.91601335],
       [-0.62420521, -0.08179119, 1.39076231, ..., 1.74309049,
        1.09168714, -1.09168714],
       [-0.28401079, 0.87525072, -1.37744033, ..., -0.57369368,
        1.09168714, -1.09168714]])

x_test = sc.transform(x_test)

```

x\_test

```
array([[ -0.55204276, -0.36890377,  1.04473698, ..., -0.57369368,
        1.09168714, -1.09168714],
       [-1.31490297,  0.10961719, -1.031415    , ..., -0.57369368,
        1.09168714, -1.09168714],
       [ 0.57162971,  0.30102557,  1.04473698, ...,  1.74309049,
        1.09168714, -1.09168714],
       ...,
       [-0.74791227, -0.27319958, -1.37744033, ...,  1.74309049,
        -0.91601335,  0.91601335],
       [-0.00566991, -0.46460796, -0.33936434, ..., -0.57369368,
        -0.91601335,  0.91601335],
       [-0.79945688, -0.84742473,  1.04473698, ..., -0.57369368,
        -0.91601335,  0.91601335]])
```