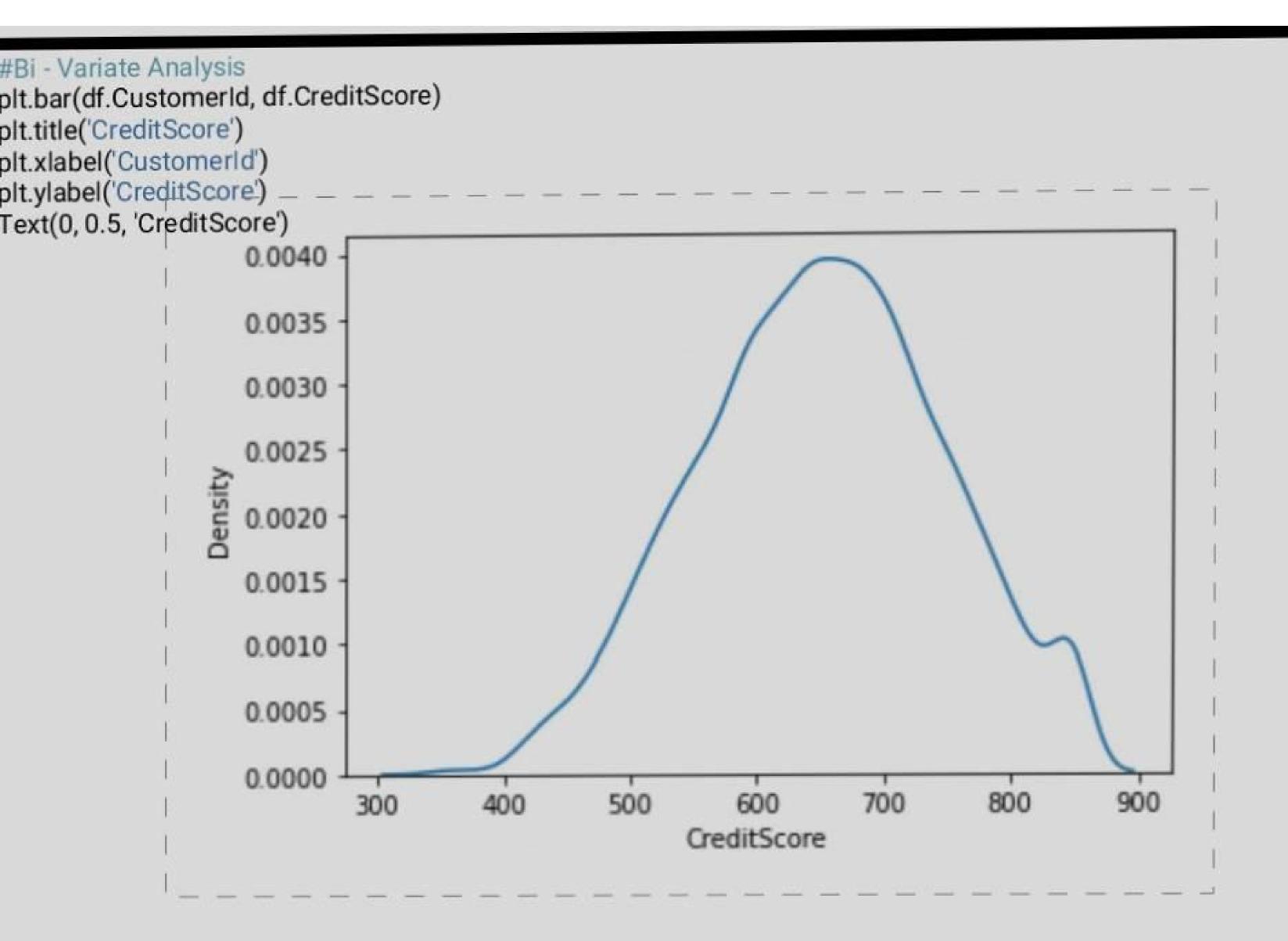
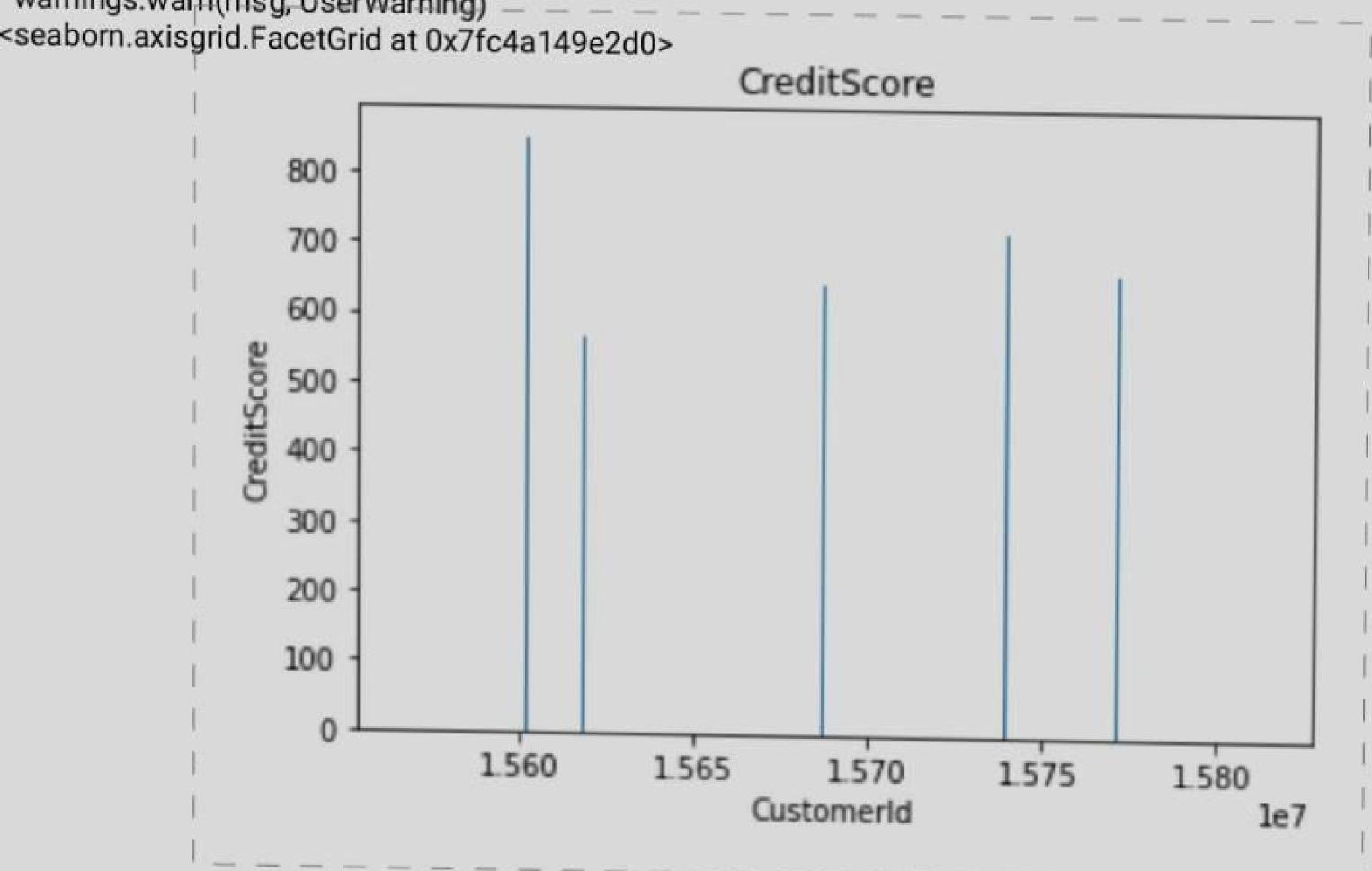
```
NAME: KAVIYA.S
   CLASS: IV YEAR-CSE
   SUB:IBM(AI)
   REG NO:6114191104033
libraries
nport pandas as pd
nport numpy as npp
nport matplotlib.pyplot as plt
matplotlib inline
load dataset
f = pd.read_csv(r"/content/Churn_Modelling.csv")
f.head(10)
RowNumber Customerld Surname CreditScore Geography Gender Age
      15634602 Hargrave
                             619 France Female 42
    2 15647311
                          608 Spain Female 41
                   Hill
       15619304
                   Onio
                            502 France Female 42
       15701354
                   Boni
                           699 France Female 39
       15737888 Mitchell
                            850
                                  Spain Female 43
       15574012
                   Chu
                           645
                                Spain Male 44
       15592531 Bartlett
                            822 France Male 50
       15656148 Obinna
                             376 Germany Female 29
    9 15792365
                           501 France Male 44
                    He
       15592389
                    H?
    10
                           684 France Male 27
Tenure Balance NumOfProducts HasCrCard IsActiveMember \
       0.00
     83807.86
   8 159660.80
                  2
       0.00
                                0
                        0
   2 125510.82
   8 113755.78
       0.00
                  2
   4 115046.74
   4 142051.07
   2 134603.88
EstimatedSalary Exited
    101348.88
    112542.58
```

113931.57

MAHENDRA ENGINEERING COLLEGE FOR WOMEN

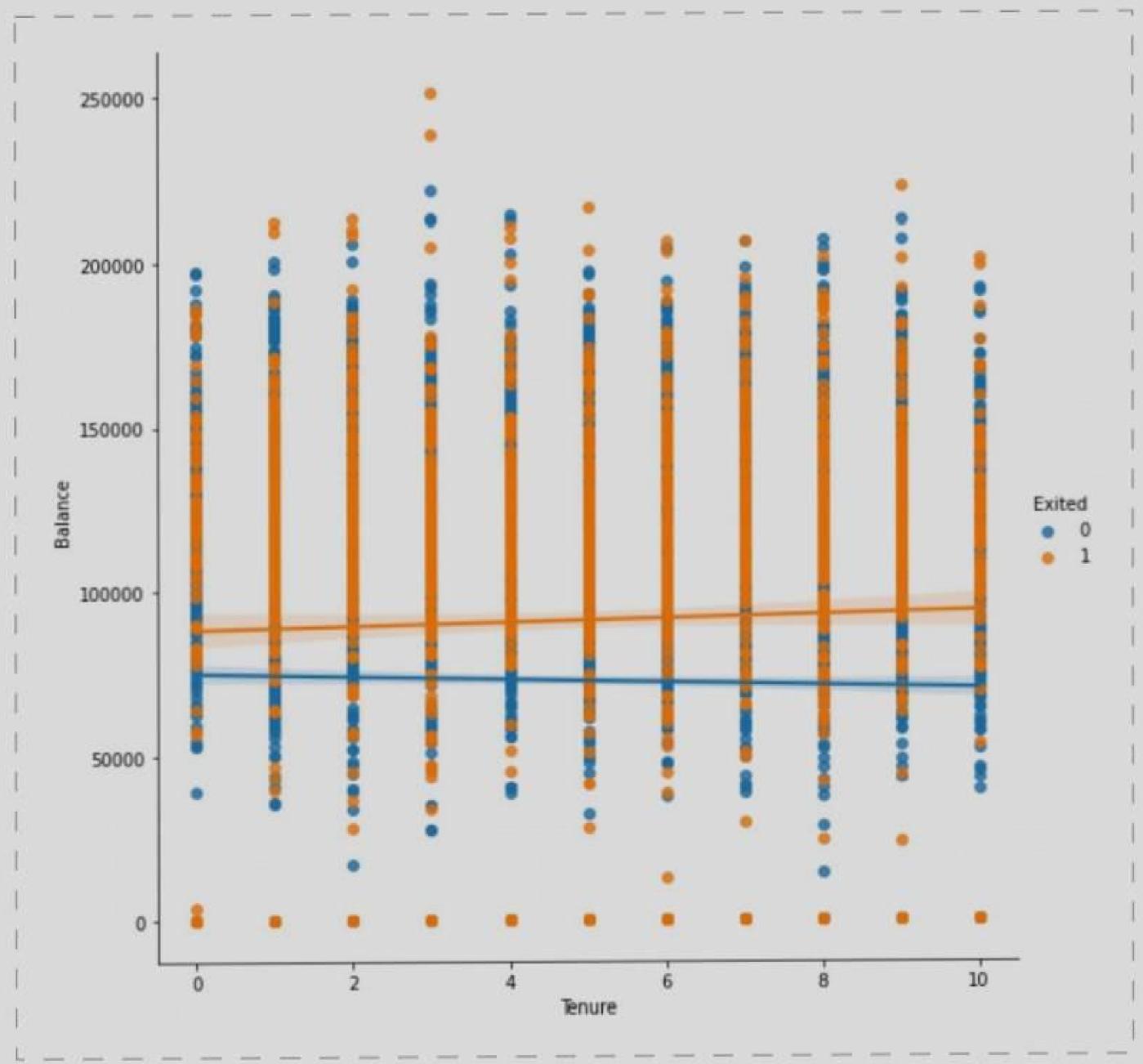
```
93826.63
     79084.10
5
     149756.71
6
     10062.80
     119346.88
     74940.50
     71725.73
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 14 columns):
# Column
                Non-Null Count Dtype
0 RowNumber
                  10000 non-null int64
  Customerld
               10000 non-null int64
                10000 non-null object
2 Surname
  CreditScore
                10000 non-null int64
                 10000 non-null object
  Geography
               10000 non-null object
  Gender
  Age
              10000 non-null int64
               10000 non-null int64
  Tenure
               10000 non-null float64
  Balance
9 NumOfProducts 10000 non-null int64
10 HasCrCard
                 10000 non-null int64
11 IsActiveMember 10000 non-null int64
12 EstimatedSalary 10000 non-null float64
               10000 non-null int64
13 Exited
dtypes: float64(2), int64(9), object(3)
memory usage: 1.1+ MB
#Visualizations
#Univariate Analysis
import seaborn as sns
sns.kdeplot(df['CreditScore'])
<matplotlib.axes._subplots.AxesSubplot at 0x7fc4a0cd2790>
```



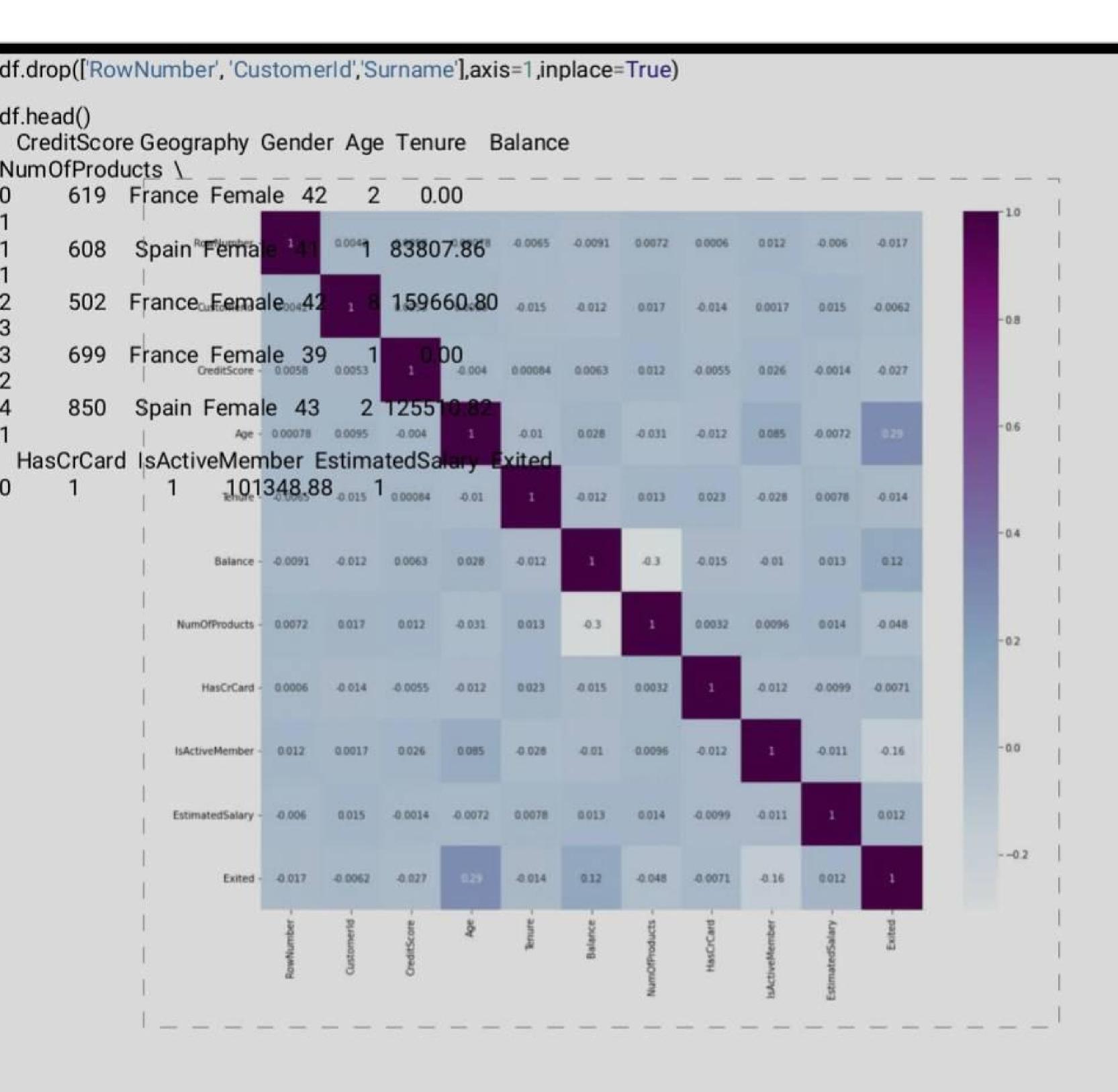


#Multi - Variate Analysis

ax =
df[["CreditScore","Age","Tenure","Balance"]].plot(figsize=(80,40))
ax.legend(loc='center left', bbox\_to\_anchor=(1, 0.5));







```
112542.58
                                 0
                    113931.57
                    93826.63
                    79084.10
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 11 columns):
# Column
               Non-Null Count Dtype
0 CreditScore 10000 non-null int64
1 Geography 10000 non-null object
2 Gender 10000 non-null object
        10000 non-null int64
3 Age
4 Tenure 10000 non-null int64
5 Balance
              10000 non-null float64
6 NumOfProducts 10000 non-null int64
7 HasCrCard
                10000 non-null int64
8 IsActiveMember 10000 non-null int64
9 EstimatedSalary 10000 non-null float64
10 Exited
              10000 non-null int64
dtypes: float64(2), int64(7), object(2)
memory usage: 859.5+ KB
df["Geography"].unique()
array(['France', 'Spain', 'Germany'], dtype=object)
df["Gender"].unique()
array(['Female', 'Male'], dtype=object)
geo=pd.get_dummies(df["Geography"],drop_first=False)
geo.head()
France Germany Spain
gen=pd.get_dummies(df["Gender"],drop_first=False)
df=pd.concat([df, geo,gen], axis=1)
df
  CreditScore Geography Gender Age Tenure Balance
NumOfProducts \
```

```
1 83807.86
      608 Spain Female 41
      502 France Female 42
                              8 159660.80
      699 France Female 39
                                  0.00
           Spain Female 43
                             2 125510.82
      850
                                   0.00
        771 France Male 39
9995
                              10 57369.61
        516 France Male 35
9996
        709 France Female 36
                                    0.00
9997
                                3 75075.31
        772 Germany Male 42
9998
                                4 130142.79
        792 France Female 28
9999
  HasCrCard IsActiveMember EstimatedSalary Exited France
Germany \
                   101348.88
              1
                               0
                                   0
                   112542.58
      0
                   113931.57 1
                   93826.63
      0
              0
                                  0
                   79084.10
0
                     96270.64
                                0
                0
9995
                     101699.77
9996
                     42085.58
9997
                                    0
                     92888.52
                0
9998
                                0
                     38190.78
                0
9999
  Spain Female Male
0
```

France Female 42

619

0.00

```
0
9995
9996
9997
9998
9999
[10000 rows x 16 columns]
df.drop(["Geography", "Gender"], axis=1, inplace=True)
df.head()
CreditScore Age Tenure Balance NumOfProducts HasCrCard \
     619 42
                   0.00
     608 41
             1 83807.86
     502 42
             8 159660.80
     699 39
                   0.00
                              2
     850 43
               2 125510.82
IsActiveMember EstimatedSalary Exited France Germany Spain
Female \
            101348.88
            112542.58
                             0
                        0
       0
            113931.57
            93826.63 0 1
             79084.10
                        0 0
                                 0 1
Male
x=df.drop('Exited',axis=1)
  CreditScore Age Tenure Balance NumOfProducts HasCrCard \
                     0.00
      619 42
                1 83807.86
      608 41
```

```
699 39
                      0.00
                                      0
      850 43
                 2 125510.82
9995
        771 39
                       0.00
        516 35
9996
                  10 57369.61
        709 36
                       0.00
9997
                                        0
        772 42
                   3 75075.31
9998
        792 28
                   4 130142.79
9999
  IsActiveMember EstimatedSalary France Germany Spain Female
Male
              101348.88
                                    0
              112542.58
                           0
                                0
              113931.57
                                0
                                    0
         0
               93826.63
                               0
                                   0
         0
               79084.10
                          0
                               0
9995
                96270.64
           0
                                     0
                                 0
9996
                101699.77
                                  0
                                      0
                                          0
9997
                42085.58
                                 0
                                     0
9998
           0
                92888.52
                                     0
                            0
                                          0
9999
                                 0
                                     0
           0
                38190.78
[10000 rows x 13 columns]
y=df['Exited']
9995
9996
9997
9998
```

502 42

8 159660.80

3

```
9999 0
Name: Exited, Length: 10000, dtype: int64
df.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})
```

```
Lest 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]])
```