MAHENDRA ENGINEERING COLLEGE FOR WOMEN **ASSIGNMENT-2 SOLUTION** NAME OF THE STUDENT: S. KAVIYA REGISTER NUMBER:611419104033 YEAR/DEPARTMENT:IV-CSE #libraries import pandas as pd import numpy as npp import matplotlib.pyplot as plt %matplotlib inline #load dataset df = pd.read_csv(r"/content/Churn_Modelling.csv") df.head(10)RowNumber Customerld Surname CreditScore Geography Gender Age 619 France Female 42 15634602 Hargrave 15647311 608 Spain Female 41 2 Hill 15619304 502 France Female 42 Onio 4 15701354 Boni 699 France Female 39 15737888 Mitchell 850 Spain Female 43 15574012 Chu 645 Spain Male 44 7 15592531 Bartlett 822 France Male 50 376 Germany Female 29 15656148 Obinna 9 15792365 France Male 44 He 501 15592389 H? 684 France Male 27 10 Balance NumOfProducts HasCrCard IsActiveMember \ Tenure 0.00 83807.86 0 159660.80 0 0.00 2 0 0 2 125510.82 0 8 113755.78 0.00 2 4 115046.74 0 4 142051.07 0 2 134603.88 EstimatedSalary Exited 101348.88

0

2

3

4

5

6

8

9

0

3

4

5

6

7

8

9

0

2

112542.58

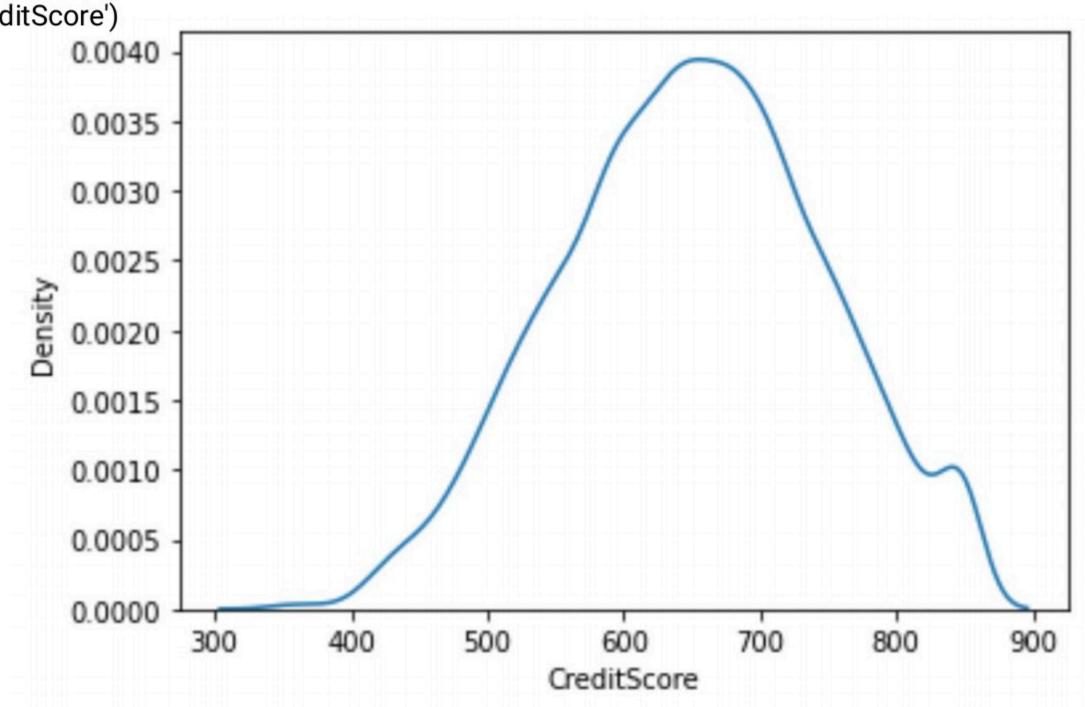
113931.57

0

```
79084.10
                 0
     149756.71
6
                 0
     10062.80
     119346.88
8
     74940.50
                 0
9
     71725.73
                 0
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 14 columns):
# Column
                Non-Null Count Dtype
                  10000 non-null int64
0 RowNumber
  CustomerId
                 10000 non-null int64
                10000 non-null object
2 Surname
  CreditScore
                10000 non-null int64
                10000 non-null object
  Geography
  Gender
               10000 non-null object
  Age
             10000 non-null int64
6
7 Tenure
               10000 non-null int64
               10000 non-null float64
  Balance
9 NumOfProducts 10000 non-null int64
                 10000 non-null int64
10 HasCrCard
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>
```

93826.63

#Bi - Variate Analysis
plt.bar(df.Customerld, df.CreditScore)
plt.title('CreditScore')
plt.xlabel('Customerld')
plt.ylabel('CreditScore')
Text(0, 0.5, 'CreditScore')



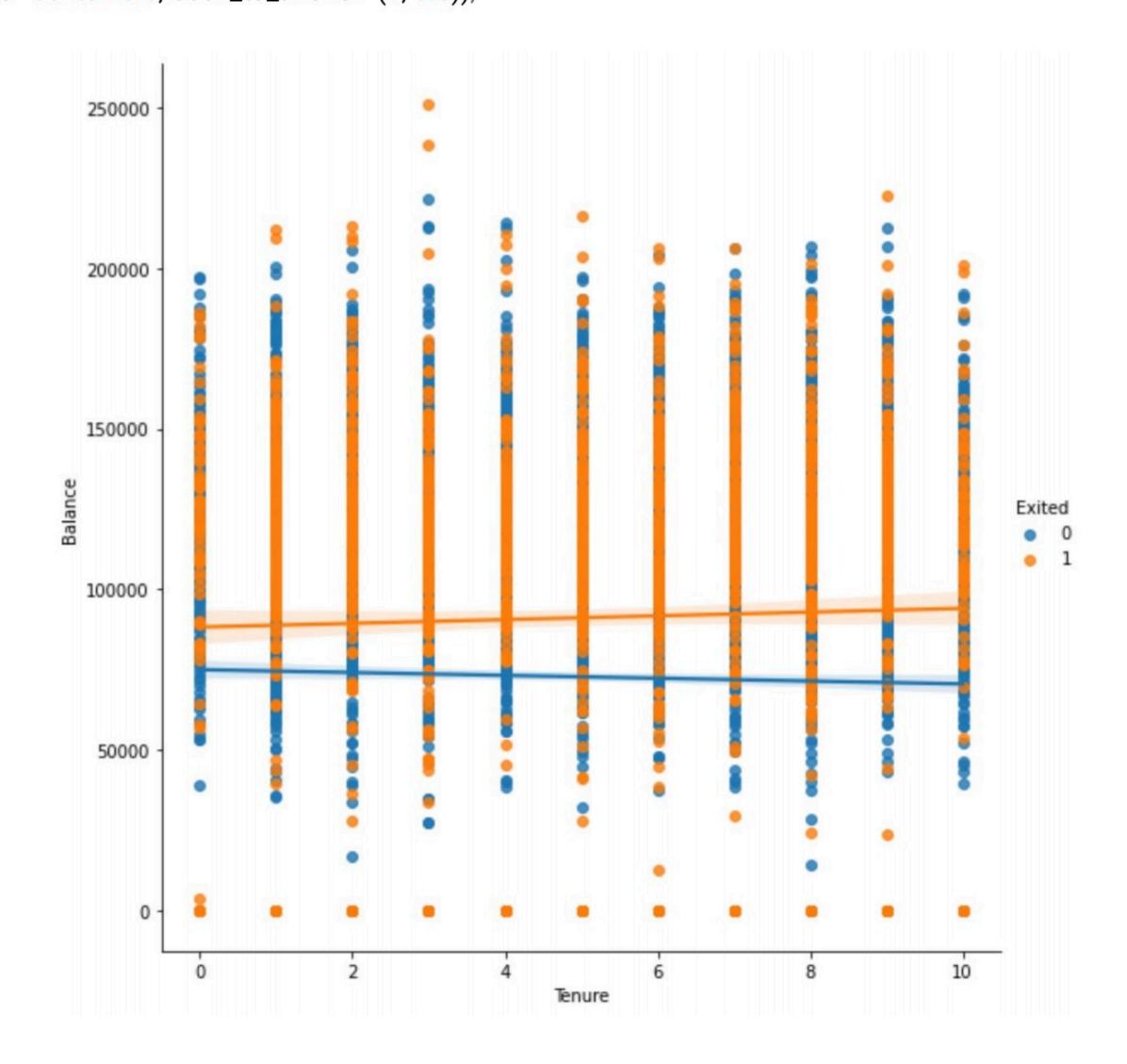
sns.lmplot(x='Tenure', y='Balance', data=df ,hue='Exited',size=8)
/usr/local/lib/python3.7/dist-packages/seaborn/regression.py:581:
UserWarning: The `size` parameter has been renamed to `height`; please update your code.

warnings.warn(msg, UserWarning)

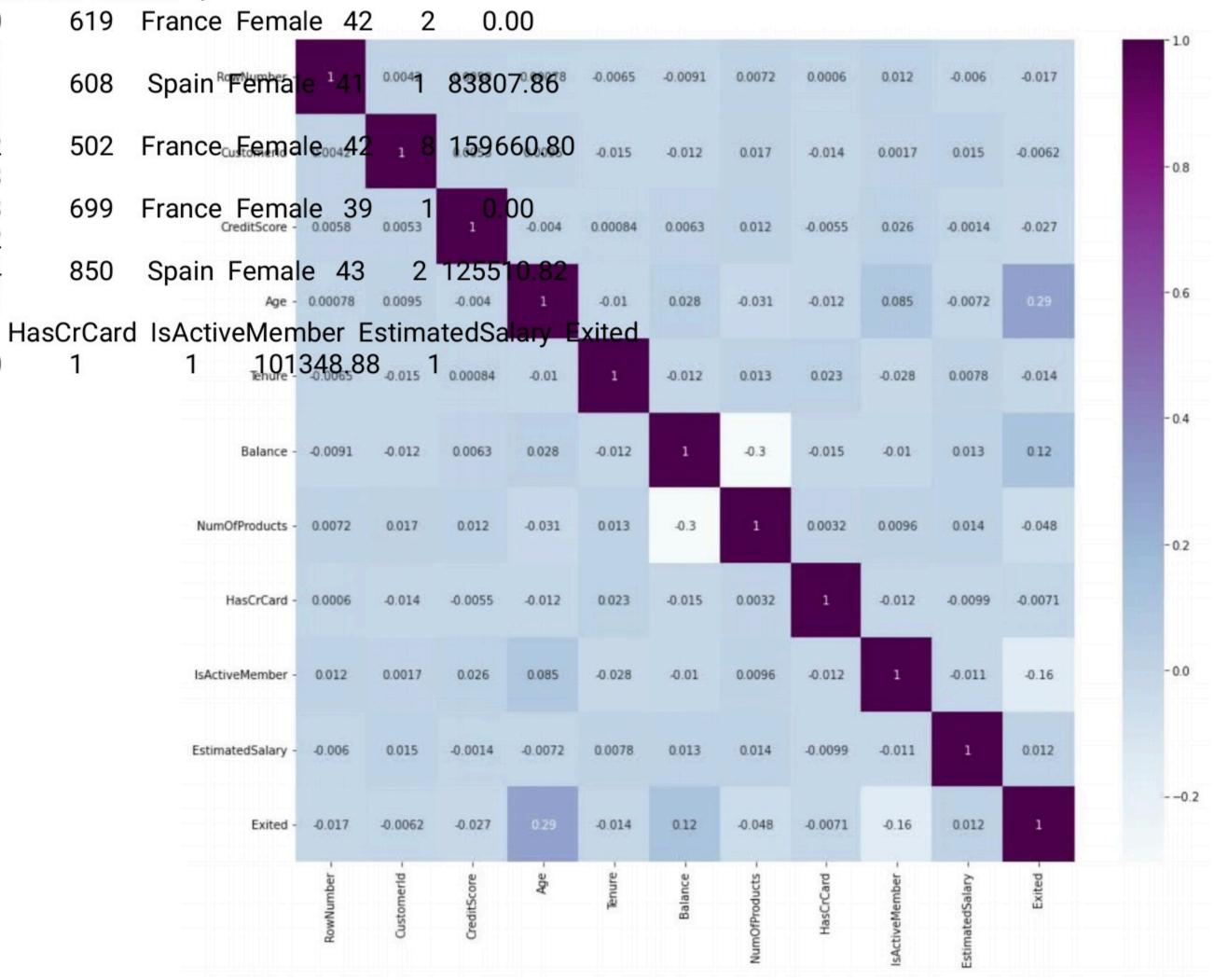
<seaborn.axisgrid.FacetGrid at 0x7fc4a149e2d0>



#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
                    113931.57
     0
                    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
 Geography 10000 non-null object
2 Gender 10000 non-null object
       10000 non-null int64
  Age
  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
0
         0
            0
    0
         0
             0
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								
1 1	608	Spain Fe	emale	41	1 83	3807.86		
2	502	France F	emale	42	8 1	59660.8	0	
3	699	France F	emale	39	1	0.00		
1 1 2 3 3 2 4 1	850	Spain Fe	emale	43	2 12	5510.82	2	
 9995	771	France	Male	39	5	0.00		
2 9996 1	516	France	Male	35	10	57369.6	51	
3610	709	France	Femal	e 36	7	0.00		
9998 2	772	2 Germar	ny Mal	le 42	3	75075	.31	
	792	2 France	Femal	e 28	4	130142	79	
		I IsActive	Membe	r Estir	mate	dSalary	Exited	France
German	y \ 1	1	101348	0 0 0	1	1		
0 0		Į.	101340	0.00		1		
1	0	1	112542	2.58	0	0		
0 2	1	0	11393	1.57	1	1		
2 0					•			
3 0	0	0	93826	.63	0	1		
0 4 0	1	1	79084	.10	0	0		
	•							
 9995 0	1	0	962	70.64	0	1		
9996 0	1	1	1016	599.77	0	1		
9997 0	0	1	4208	85.58	1	1		
9998 1	1	0	928	88.52	1	0		
9999 0	1	0	3819	90.78	0	1		
the second was	n Fem 1	nale Male 0						

619 France Female 42

0.00

```
0
4
             0
9995
           0
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 \
0
     619 42
                    0.00
     608
          41
               1 83807.86
     502 42
               8 159660.80
     699 39
                     0.00
                                      0
                2 125510.82
     850 43
IsActiveMember EstimatedSalary Exited France Germany Spain
Female \
0
             101348.88
                                   0
                                        0
             112542.58
                          0
                              0
                                   0
             113931.57
                                   0
        0
                          1
                              1
                                        0
             93826.63
                                   0
        0
                         0
                                       0
             79084.10
                         0
                              0
                                   0
                                       1
 Male
0
  0
3 0
x=df.drop('Exited',axis=1)
X
  CreditScore Age Tenure Balance NumOfProducts HasCrCard \
0
       619 42 2 0.00
```

608 41 1 83807.86

0

```
3
       699
            39
                       0.00
                                  2
                                        0
       850 43
                  2 125510.82
4
9995
              39
                        0.00
                                    2
         771
9996
         516
              35
                   10 57369.61
         709 36
9997
                        0.00
                                          0
         772 42
9998
                    3 75075.31
         792 28
9999
                    4 130142.79
  IsActiveMember EstimatedSalary France Germany Spain Female
Male
0
               101348.88
                                 0
                                     0
          1
                            1
                                          1
0
               112542.58
                            0
                                 0
                                      1
0
               113931.57
                                     0
          0
                            1
                                 0
0
               93826.63
3
                                 0
                                     0
          0
                                          1
0
                                 0
               79084.10
                           0
4
0
                 96270.64
9995
                                       0
           0
                              1
                                   0
                                            0
                 101699.77
9996
                              1
                                   0
                                       0
                                            0
9997
                 42085.58
                              1
                                   0
                                       0
            1
0
                 92888.52
9998
           0
                             0
                                       0
                                            0
           0
                                   0
                                       0
9999
                 38190.78
                             1
                                            1
[10000 rows x 13 columns]
y=df['Exited']
```

502 42

8 159660.80

```
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)
```