```
MAHENDRA ENGINEERING COLLEGE FOR WOMEN
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    CLASS:IV YEAR-CSE
    SUB:IBM
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#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 CustomerId Surname CreditScore Geography Gender Age
                                619 France Female 42
        15634602 Hargrave
     2
        15647311
                    Hill
                             608
                                  Spain Female 41
        15619304
                              502 France Female 42
     3
                    Onio
        15701354
                    Boni
                              699
                                  France Female 39
        15737888 Mitchell
                               850
                                     Spain Female 43
        15574012
                     Chu
                              645
                                    Spain Male 44
        15592531 Bartlett
                              822 France Male 50
     7
        15656148 Obinna
                               376 Germany Female 29
        15792365
                      He
                             501
                                  France Male 44
                              684 France Male 27
     10
        15592389
                      H?
        Balance NumOfProducts HasCrCard IsActiveMember \
 Tenure
    2
        0.00
                    1
                          1
                                  1
    1 83807.86
                      1
                            0
                                     1
    8 159660.80
                       3
                                     0
                             1
    1
         0.00
                    2
                          0
                                  0
    2 125510.82
                      1
                             1
                                     1
    8 113755.78
                       2
                             1
                                     0
                    2
                          1
                                  1
    7
         0.00
    4 115046.74
                       4
                             1
                                     0
    4 142051.07
                       2
                             0
                                     1
    2 134603.88
                             1
                                      1
 EstimatedSalary Exited
    101348.88
                 1
                 0
    112542.58
```

1

2

3

4

5

6

7

8

9

0

1

2

3

4

5

6 7

8

0

1

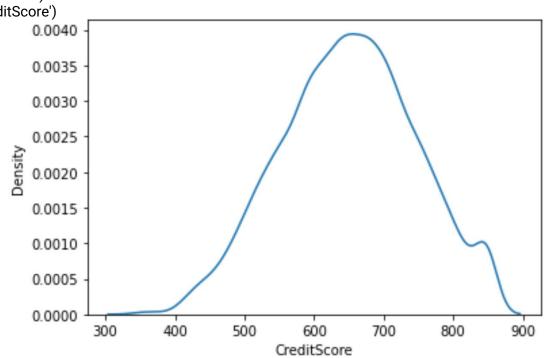
2

113931.57

```
79084.10
                  0
     149756.71
                  1
6
     10062.80
                  0
                  1
     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
0 RowNumber
                   10000 non-null int64
1 CustomerId
                 10000 non-null int64
2 Surname
                10000 non-null object
3 CreditScore
                 10000 non-null int64
4 Geography
                 10000 non-null object
5 Gender
               10000 non-null object
6 Age
              10000 non-null int64
               10000 non-null int64
7 Tenure
               10000 non-null float64
8 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
13 Exited
               10000 non-null int64
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>
```

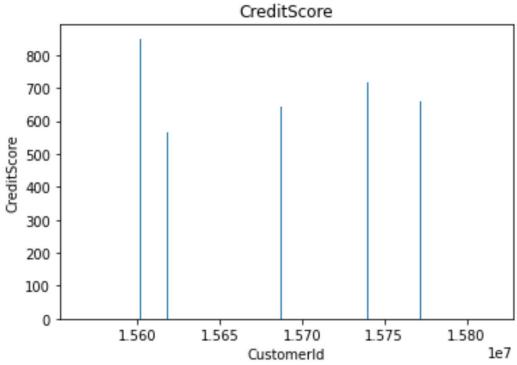
93826.63

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

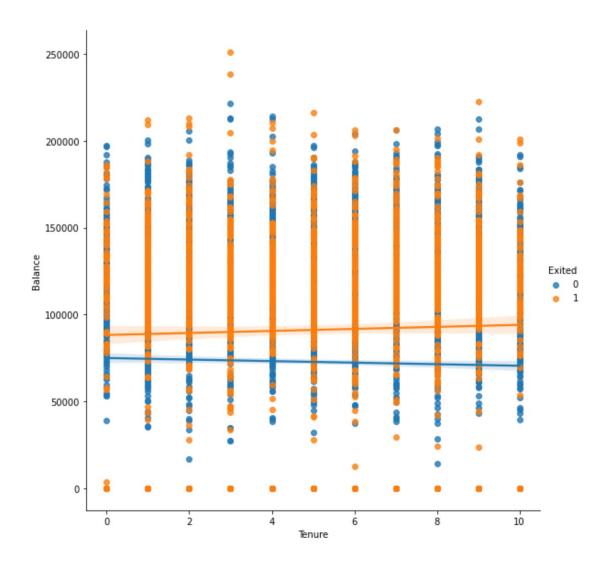


sns.Implot(x='Tenure', y='Balance', data=df,hue='Exited',size=8)
/usr/local/lib/python3.7/dist-packages/seaborn/regression.py:581:
UserWarning: The `size` parameter has been renamed to `height`; please update your code.

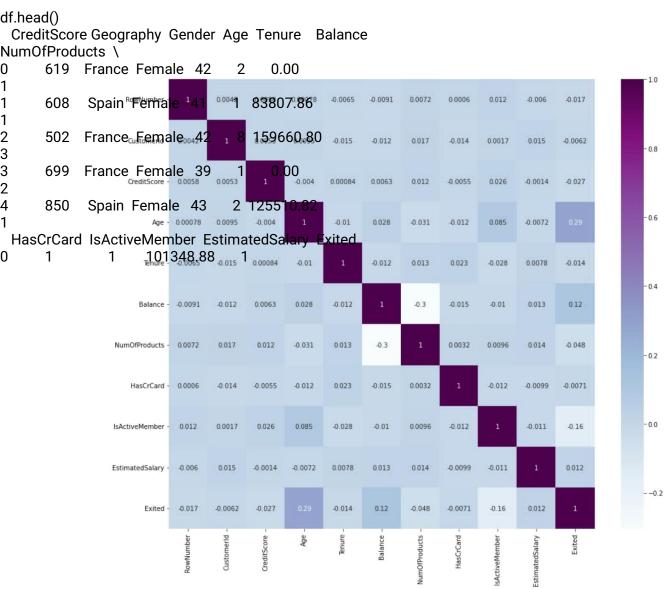
warnings.warn(msg, UserWarning) <seaborn.axisgrid.FacetGrid at 0x7fc4a149e2d0>



#Multi - Variate Analysis ax = df[["CreditScore","Age","Tenure","Balance"]].plot(figsize=(80,40)) ax.legend(loc='center left', bbox_to_anchor=(1, 0.5));



```
df.isnull().sum()
RowNumber
                  0
CustomerId
                 0
Surname
               0
CreditScore
                0
Geography
                0
              0
Gender
             0
Age
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(),annot=True,cmap='BuPu')
plt.show()
```



```
1
               1
                                   0
                     112542.58
2
               0
                     113931.57
                                   1
3
      0
               0
                     93826.63
                                  0
      1
               1
                     79084.10
                                  0
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
                 10000 non-null object
1 Geography
2 Gender
               10000 non-null object
              10000 non-null int64
3 Age
               10000 non-null int64
4 Tenure
               10000 non-null float64
5 Balance
6 NumOfProducts 10000 non-null int64
7 HasCrCard
                 10000 non-null int64
8 IsActiveMember 10000 non-null int64
9 EstimatedSalary 10000 non-null float64
               10000 non-null int64
10 Exited
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
              1
2
    1
          0
              0
3
              0
    1
          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	608	Spain Fe	emale 4	11	1 83	3807.86		
1 2	502	France F	emale	42	8 1	59660.8	0	
3 3	699	France F	emale	39	1	0.00		
2 4 1	850	Spain Fe	emale 4	13	2 12	25510.82	2	
 9995 2	771	France	Male	39	5	0.00		
9996 1	516	France	Male	35	10	57369.	51	
9997 1	709	France	Female	36	7	0.00		
9998 2	772	German	y Male	e 42	3	75075	.31	
9999 1	792	France	Female	28	4	130142	2.79	
		IsActive	Member	Estir	mate	dSalary	Exited	France
Germar 0	1y \ 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 								
 9995 0	1	0	9627	0.64	0	1		
9996 0	1	1	10169	99.77	0	1		
9997 0	0	1	4208	5.58	1	1		
9998 1	1	0	9288	8.52	1	0		
9999 0	1	0	3819	0.78	0	1		
		ale Male 0						

619 France Female 42 2 0.00

```
2
     0
          1
             0
3
     0
          1
             0
     1
          1
             0
9995
       0
           0
               1
9996
               1
       0
           0
9997
       0
            1
               0
9998
       0
            0
               1
9999
       0
            1
               0
[10000 rows x 16 columns]
df.drop(["Geography","Gender"], axis=1, inplace=True)
df.head()
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
     699 39
3
                1
                     0.00
                                2
                                       0
     850 43
                2 125510.82
IsActiveMember EstimatedSalary Exited France Germany Spain
Female \
                                         0
0
        1
             101348.88
                           1
                                1
                                     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
x=df.drop('Exited',axis=1)
Χ
  CreditScore Age Tenure Balance NumOfProducts HasCrCard \
0
       619 42
                  2
                       0.00
                                  1
```

608 41

1 83807.86

```
3
       699 39
                                   2
                                          0
                   1
                       0.00
4
       850 43
                   2 125510.82
                                      1
                                             1
       ... ...
             ...
                    ...
                           ...
9995
         771 39
                    5
                         0.00
                                     2
9996
         516 35
                    10 57369.61
                                        1
                                               1
9997
         709 36
                    7
                         0.00
                                     1
9998
         772 42
                    3 75075.31
                                        2
                                              1
9999
         792 28
                    4 130142.79
                                        1
  IsActiveMember EstimatedSalary France Germany Spain Female
Male
               101348.88
                                       0
0
          1
                             1
                                   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
•••
        •••
                          ... ...
                                  • • •
9995
            0
                  96270.64
                               1
                                    0
                                        0
                                             0
1
                               1
                                    0
                                         0
9996
            1
                 101699.77
                                              0
1
9997
            1
                  42085.58
                                        0
                                             1
                               1
                                    0
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']
0
    1
1
    0
2
    1
```

9995 0 9996 0

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)
```