

MAHENDRA ENGINEERING COLLEGE FOR WOMEN

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CLASS : IBM
ASSIGNMENT- II
#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	
8	4	142051.07	2	0	1	
9	2	134603.88	1	1	1	

	EstimatedSalary	Exited
0	101348.88	1
1	112542.58	0
2	113931.57	1

```
3      93826.63      0
4      79084.10      0
5     149756.71      1
6     10062.80      0
7     119346.88      1
8     74940.50      0
9     71725.73      0

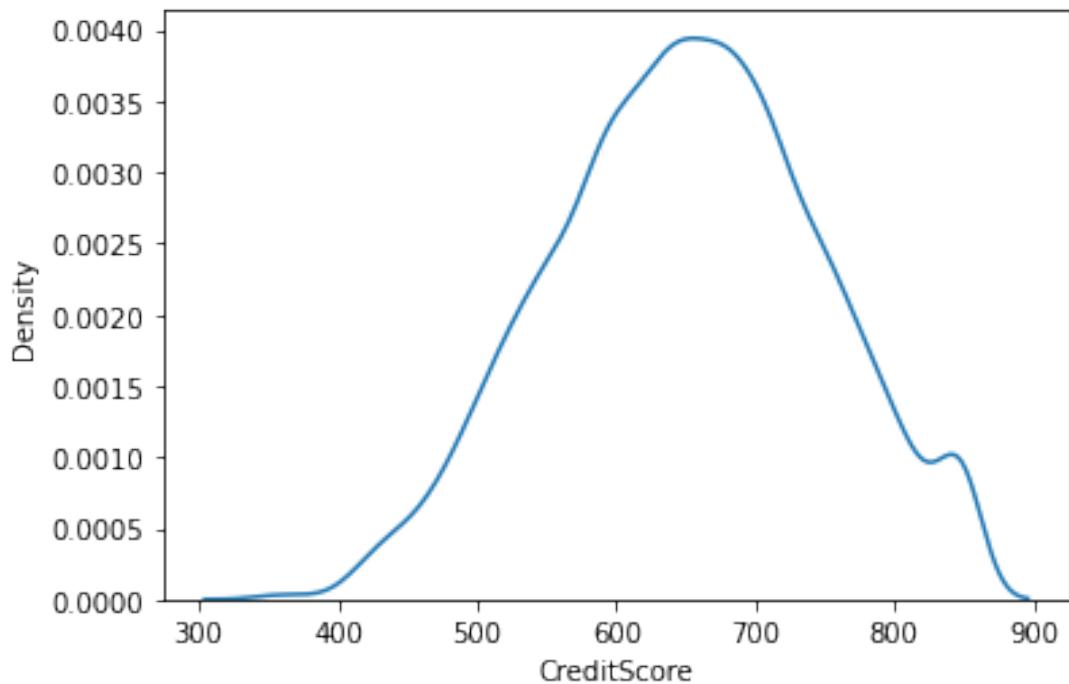
df.info()

<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  
 7   Tenure            10000 non-null   int64  
 8   Balance           10000 non-null   float64 
 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>
```



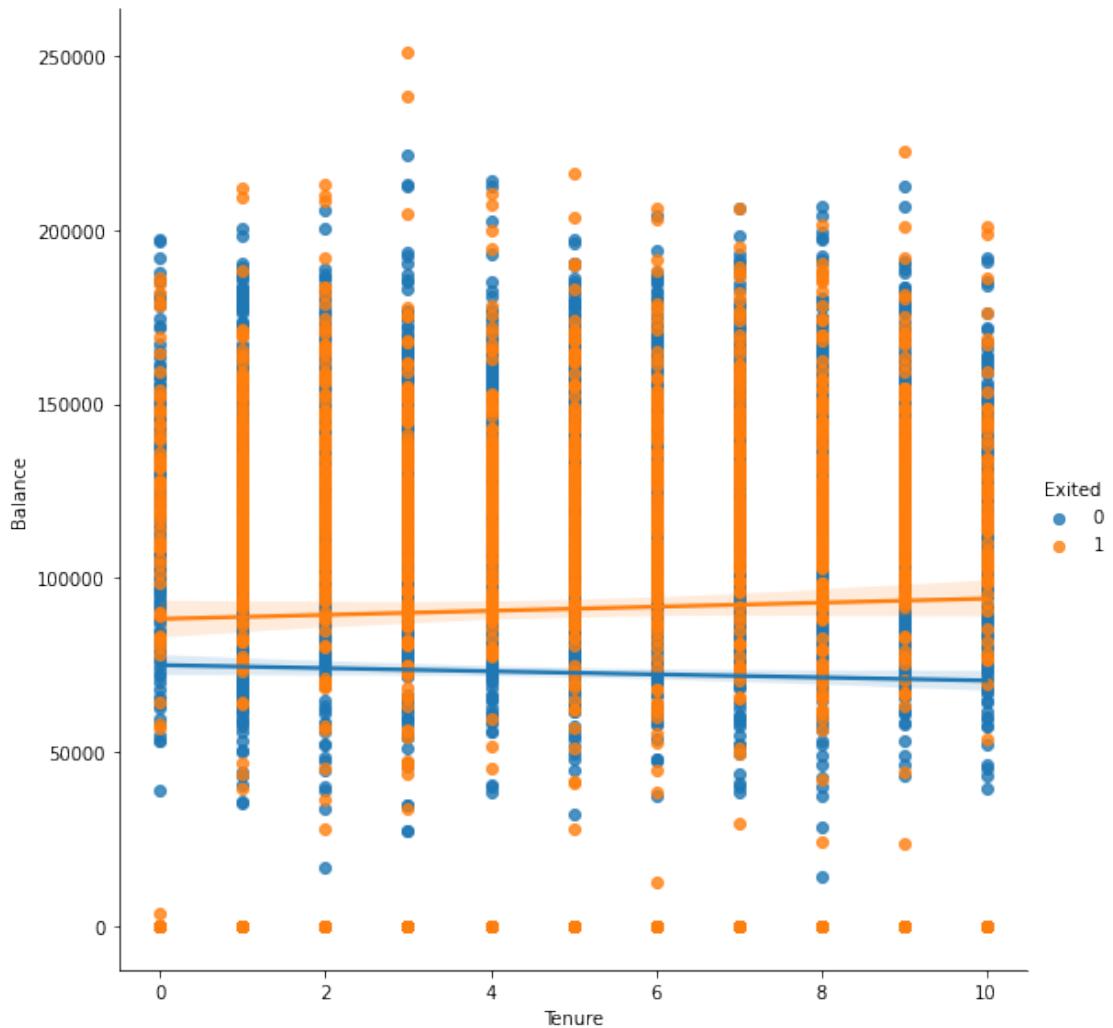
#Bi - Variate Analysis

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

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

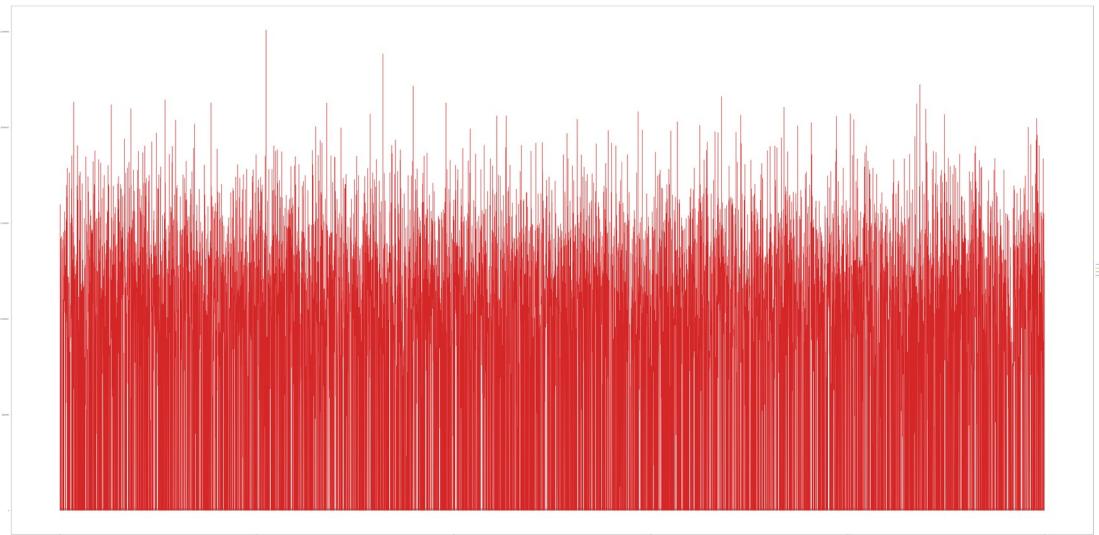


```
sns.lmplot(x='Tenure', y='Balance', data=df, hue='Exited', size=8)  
/usr/local/lib/python3.7/dist-packages/seaborn/regression.py:581:  
UserWarning: The `size` parameter has been renamed to `height`; please  
update your code.  
    warnings.warn(msg, UserWarning)  
<seaborn.axisgrid.FacetGrid at 0x7fc4a149e2d0>
```



```
#Multi - Variate Analysis
```

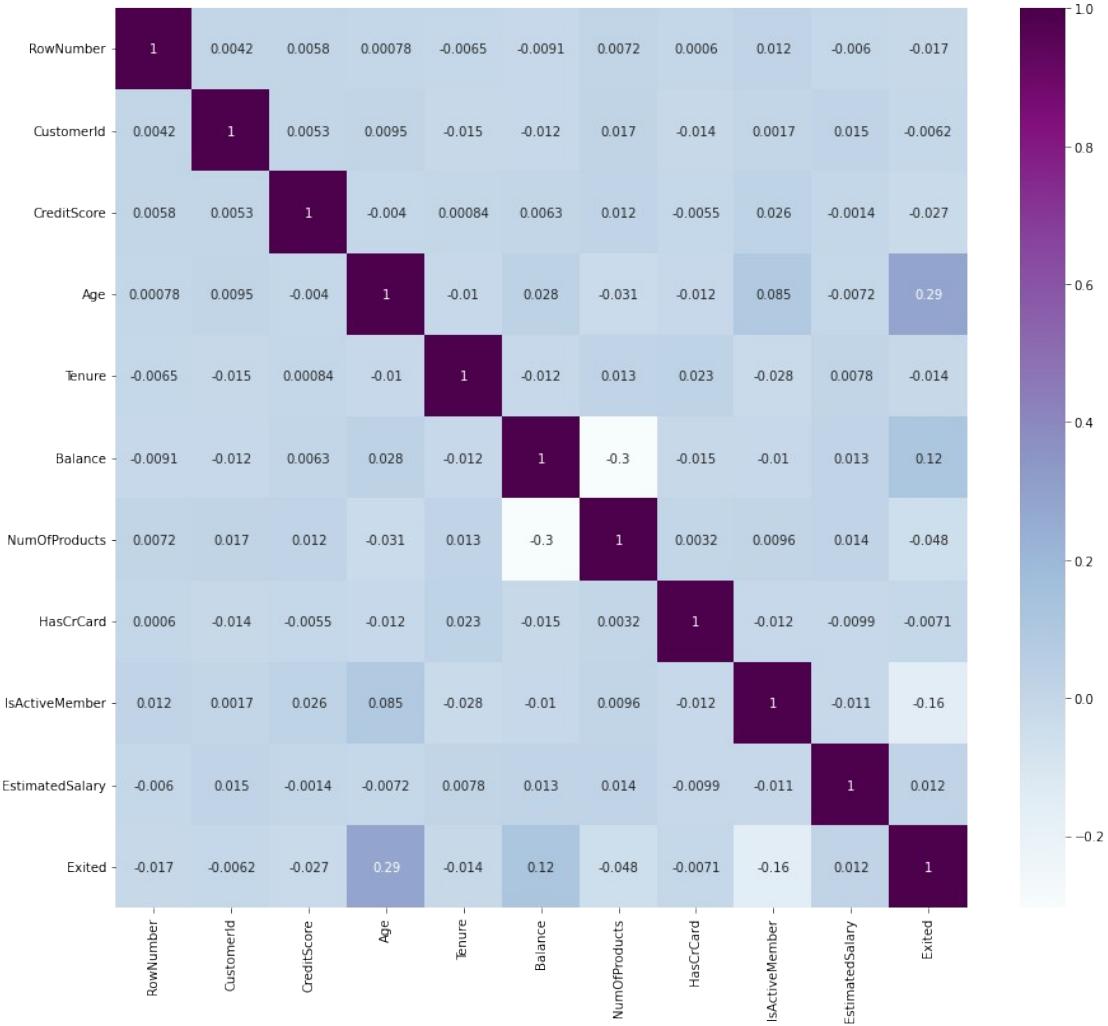
```
ax =
df[["CreditScore", "Age", "Tenure", "Balance"]].plot(figsize=(80,40))
ax.legend(loc='center left', bbox_to_anchor=(1, 0.5));
```



```
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(), annot=True, cmap='BuPu')
plt.show()
```



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

```
df.head()
```

	CreditScore	Geography	Gender	Age	Tenure	Balance
0	619	France	Female	42	2	0.00
1	608	Spain	Female	41	1	83807.86
2	502	France	Female	42	8	159660.80
3	699	France	Female	39	1	0.00
4	850	Spain	Female	43	2	125510.82
1						
0	HasCrCard	IsActiveMember	EstimatedSalary		Exited	
0	1	1	101348.88		1	

```

1      0          1      112542.58      0
2      1          0      113931.57      1
3      0          0      93826.63      0
4      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  
 1   Geography        10000 non-null   object  
 2   Gender            10000 non-null   object  
 3   Age               10000 non-null   int64  
 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
0       1        0       0
1       0        0       1
2       1        0       0
3       1        0       0
4       0        0       1

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 \

```

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

Germany \	HasCrCard	IsActiveMember	EstimatedSalary	Exited	France	
0	1	1	101348.88	1	1	
0	0	1	112542.58	0	0	
1	1	0	113931.57	1	1	
0	3	0	93826.63	0	1	
0	0	1	79084.10	0	0	
4	1	1	
0	
9995	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	

Spain	Female	Male
0	0	1
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

[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
3        699    39       1        0.00           2            0
4        850    43       2   125510.82           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
Male						
0		1		101348.88	1	0
0		1		112542.58	0	1
1		1		113931.57	1	0
0		0		93826.63	1	0
2		0		79084.10	0	1
0		1		96270.64	1	0
3		0		101699.77	1	0
0		1		42085.58	1	0
4		0		92888.52	0	1
0		0		38190.78	1	0
...
9995		0		0	0	0
1		1		0	0	0
9996		1		0	0	0
1		1		0	0	1
9997		0		0	1	0
0		1		0	0	0
9998		0		0	1	0
1		0		0	0	1
9999		0		0	0	1
0		0		0	0	1

[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: 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)
```

```
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]])
```