#### ASSIGNMENT - 2

### MAHENDRA ENGINEERING COLLEGE FOR WOMEN

NAME: KOWSALYA M SUBJECT: IBM CLASS:4th YEAR ECE

REGISTER NO:611419106033

EstimatedSalary

101348.88

112542.58

112021 57

0

1

Exited

1

0

#librariesimport pandas as pdimport as np import matplotlib.pyplot as plt %matplotlib inline

## #load dataset

df = pd.read\_csv(r"/content/Churn\_Modelling.csv")df.head(10)

\	RowNumber	Custome	erld	Surname	CreditScore	Geography	Gender	Age
0	1	15634	602	Hargrave	619	France	Female	42
1	2	15647	311	Hill	608	Spain	Female	41
2	3	15619	304	Onio	502	France	Female	42
3	4	15701	354	Boni	699	France	Female	39
4	5	15737	888	Mitchell	850	Spain	Female	43
5	6	15574	012	Chu	645	Spain	Male	44
6	7	15592	531	Bartlett	822	France	Male	50
7	8	15656	148	Obinna	376	Germany	Female	29
8	9	15792	365	He	501	France	Male	44
9	10	15592	389	H?	684	France	Male	27
0 1 2 3 4 5 6 7 8	8 1 1 2 1 8 1 7 4 1	Balance 0.00 83807.86 59660.80 0.00 25510.82 13755.78 0.00 15046.74 42051.07	Nun	nOfProducts 1 1 3 2 1 2 2 4	HasCrCard  1 0 1 0 1 1 1 1 1	IsActiveMem	ber \ 1	
9		34603.88		1	1		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()

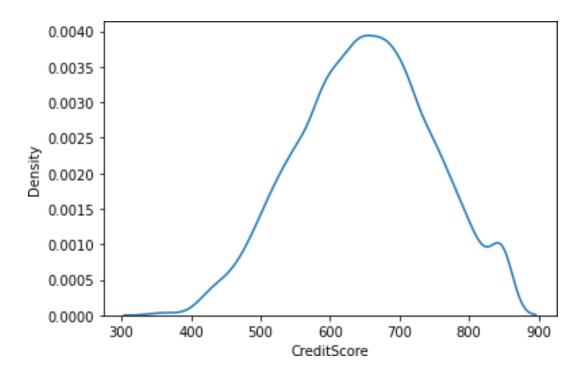
<class 'pandas.core.frame.DataFrame'> RangeIndex: 10000 entries, 0 to 9999 Data columns (total 14 columns):

#	Column	Non-Null Count Dtype
1	RowNumber	10000 non-null int64
2	CustomerId	10000 non-null int64
3	Surname	10000 non-null object
4	CreditScore	10000 non-null int64
5	Geography	10000 non-null object
6	Gender	10000 non-null object
7	Age	10000 non-null int64
8	Tenure	10000 non-null int64
9	Balance	10000 non-null float64
10	NumOfProducts	10000 non-null int64
11	HasCrCard	10000 non-null int64
12	IsActiveMember	10000 non-null int64
13	EstimatedSalary	10000 non-null float64
14	Exited	10000 non-null int64dtypes:
float6	4(2), int64(9), object	ct(3) memory usage: 1.1+ MB

#Visualizations #Univariate Analysis import seaborn as sns

sns.kdeplot(df['CreditScore'])

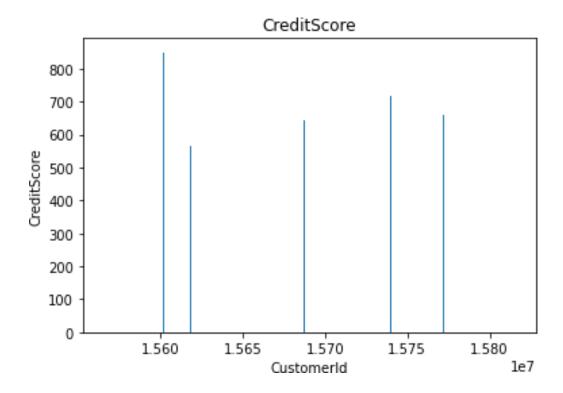
<matplotlib.axes.\_subplots.AxesSubplot at 0x7fc4a0cd2790>



#Bi - Variate Analysis

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

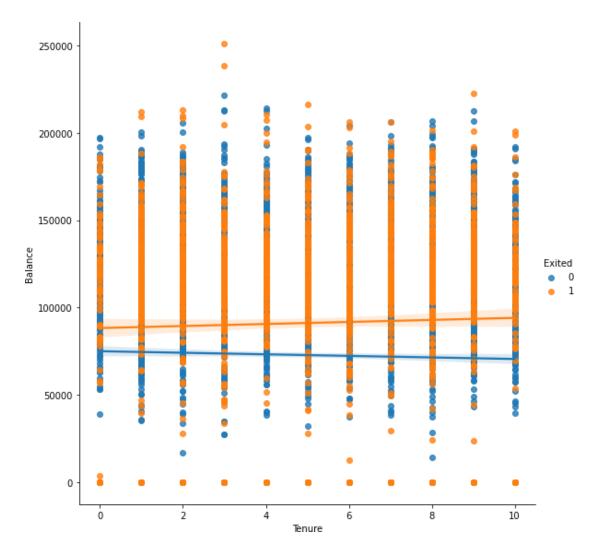
Text(0, 0.5, 'CreditScore')



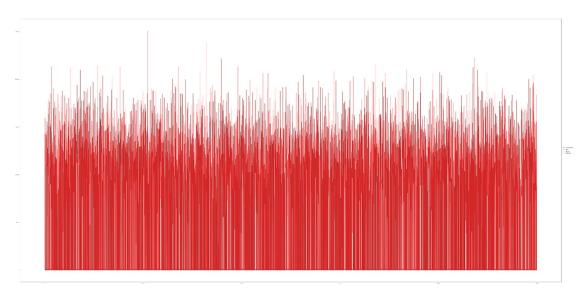
sns.lmplot(x='Tenure', y='Balance', data=df ,hue='Exited',size=8)

/usr/local/lib/python3.7/dist-packages/seaborn/regression.py:581: UserWarning: The `size` parameter has been renamed to `height`; pleaseupdate your code. warnings.warn(msg, UserWarning)

<seaborn.axisgrid.FacetGrid at 0x7fc4a149e2d0>



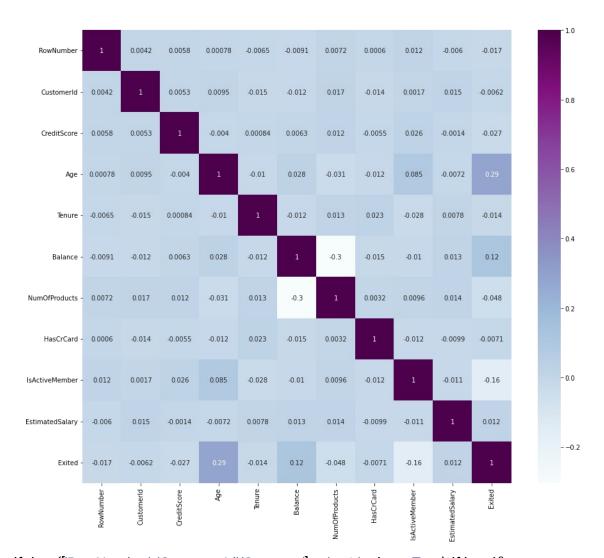
 $\# \textit{Multi - Variate Analysis} \\ ax = df[["CreditScore","Age","Tenure","Balance"]].plot(figsize=(80,40))ax.legend(loc='center left', bbox_to_anchor=(1, 0.5));$ 



# 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 NumOfProducts	Geography \	Gender	Age	Tenure	Balance
0 61	France	Female	42	2	0.00
1 60	Spain	Female	41	1	83807.86
2 50	2 France	Female	42	8	159660.80
3 69	9 France	Female	39	1	0.00
2 4 85 1	) Spain	Female	43	2	125510.82

HasCrCard IsActiveMember EstimatedSalary Exited0 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
1	CreditScore	10000 non-null int64
2	Geography	10000 non-null object
3	Gender	10000 non-null object
4	Age	10000 non-null int64
5	Tenure	10000 non-null int64
6	Balance	10000 non-null float64
7	NumOfProducts	10000 non-null int64
8	HasCrCard	10000 non-null int64
9	IsActiveMember	10000 non-null int64
10	EstimatedSalary	10000 non-null float64
11	Exited	10000 non-null int64dtypes:

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	Spain0	1
	0	0	
1	0	0	1
2	1	0	0
3	1	0	0
4	0	0	1

 $gen=pd.get\_dummies(df["Gender"],drop\_first=False)df=pd.concat([df, dender"]) dender["Gender"],drop\_first=False)df=pd.concat([df, dender"]) dender["Gender"],drop\_first=False]df=pd.concat([df, dender"]) dender["Gender"],drop\_first=False]df=pd.concat([df, dender"]) dender["Gender"],drop\_first=False]df=pd.conca$ 

geo,gen], axis=1)

df

CreditScore Geography Gender Age Tenure NumOfProducts \

Balance

0	619	France	Female	42	2	0.00	
1	608	Spain	Female	41	1	83807.86	
1 2	502	France	Female	42	8	159660.80	
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 1	792	France	Female	28	4	130142.79	
	HasCrCard IsAc	tiveMember Es	stimatedSa	lary Exit	ed France	Germany \	
0	1		1	10	1348.88	1	1
0 1	0		1	11	2542.58	0	0
0 2	1		0	11	3931.57	1	1
0 3	0		0	9	3826.63	0	1
0 4	1		1	7	9084.10	0	0
0 							
 9995	1		0	96	270.64	0	1
0 9996	1		1	10	1699.77	0	1
0 9997	0		1	4	2085.58	1	1
0 9998	1		0	9	2888.52	1	0
1 9999 0	1		0	3	8190.78	0	1

Spain 0

0

Female Male 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()

2 1 2 3 4	CreditScore  60 50 69 85	0.00 8 41 2 42 9 39	Balance Num 1 1 1 83807.8 8 159660.80 1 0.0 2 125510.82	36 ) )0	HasCrCard  1 3 2 1	\0	61942 0 1 0
0 1 1 2 1 3 1 4	IsActiveMem	ber Estimated 1 1 0 0 1	dSalary Exited Fran 101348.88 112542.58 113931.57 93826.63 79084.10	nce Germany : 1 0 1 0 0	SpainFemale 1 0 1 1 1	0 0 0 0 0	0 1 0 0
0 1 2 3 4	Male 0 0 0 0						

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

	CreditScore	Age	Tenure	Balance	<b>NumOfProducts</b>	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

Mala	IsActiveMember	EstimatedSalary	France	Germany	Spain	Female
Male 0	1	101348.88	1	0	0	1
0	1	112542.58	0	0	1	1
2	0	113931.57	1	0	0	1
0 3	0	93826.63	1	0	0	1
0 4 0	1	79084.10	0	0	1	1
					•••	
 9995	0	96270.64	1	0	0	0
9996	1	101699.77	1	0	0	0
9997	1	42085.58	1	0	0	1
0 9998	0	92888.52	0	1	0	0
9999 0	0	38190.78	1	0	0	1

[10000 rows x 13 columns]

y=df['Exited']

```
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 StandardScalersc =
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)
```

9999

## x\_test

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