

ASSIGNMENT 2

MAHENDRA ENGINEERING COLLEGE

FOR WOMEN

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CLASS :4 year ece

SUBJECT:IBM

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```
#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	

```

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'>RangeIn
dex:10000entries,0to9999Datacolumns(t
otal 14 columns):
#    Column                Non-NullCountDtype
----
0    RowNumber            10000non-nullint64
1    CustomerId           10000non-nullint64
2    Surname              10000non-null  object
3    CreditScore           10000non-nullint64
4    Geography            10000non-nullobject
5    Gender               10000non-nullobject
6    Age                  10000non-nullint64
7    Tenure               10000non-nullint64
8    Balance              10000non-nullfloat64
9    NumOfProducts        10000non-nullint64
10   HasCrCard            10000non-nullint64
11   IsActiveMember       10000non-nullint64
12   EstimatedSalary10000non-null  float64
13   Exited               10000 non-
nullint64dtypes: float64(2),
int64(9),object(3)memoryusage:1.1+MB

```

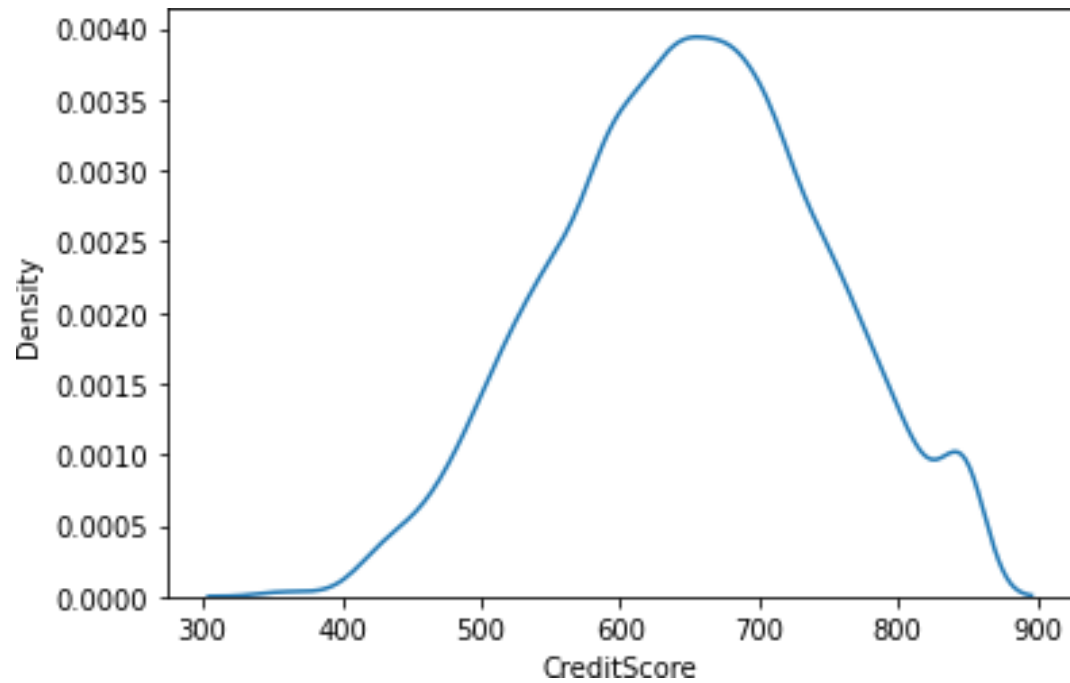
```

#Visualizations#Univa
riate Analysisimport
seabornassns

```

```
sns.kdeplot(df['CreditScore'])
```

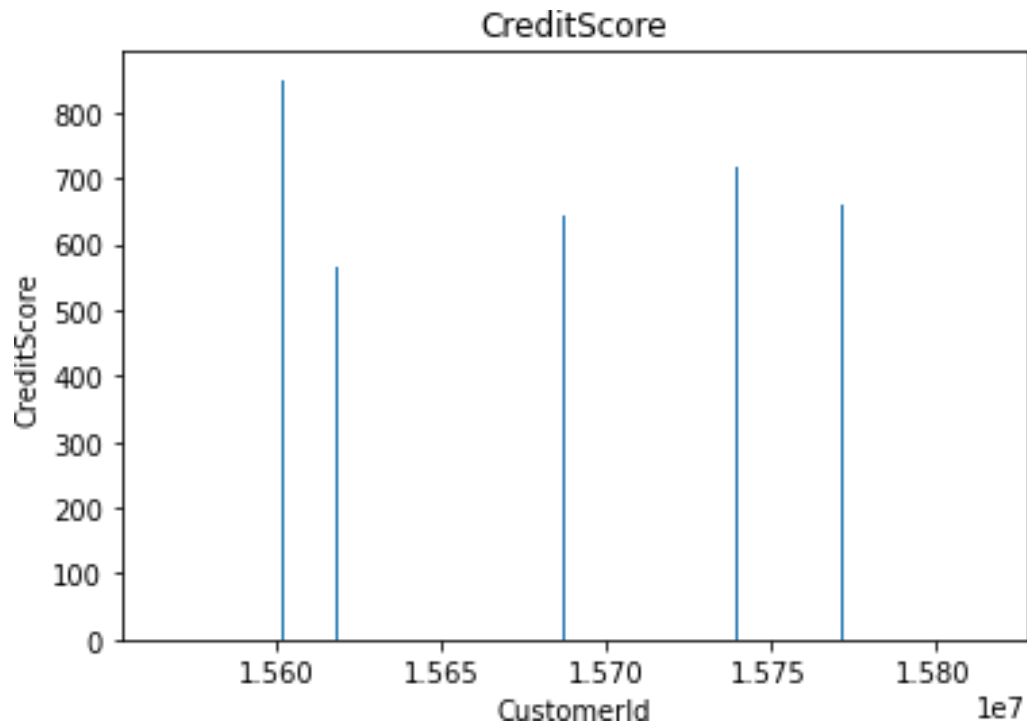
```
<matplotlib.axes._subplots.AxesSubplotat0x7fc4a0cd2790>
```



```
#Bi-VariateAnalysis
```

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

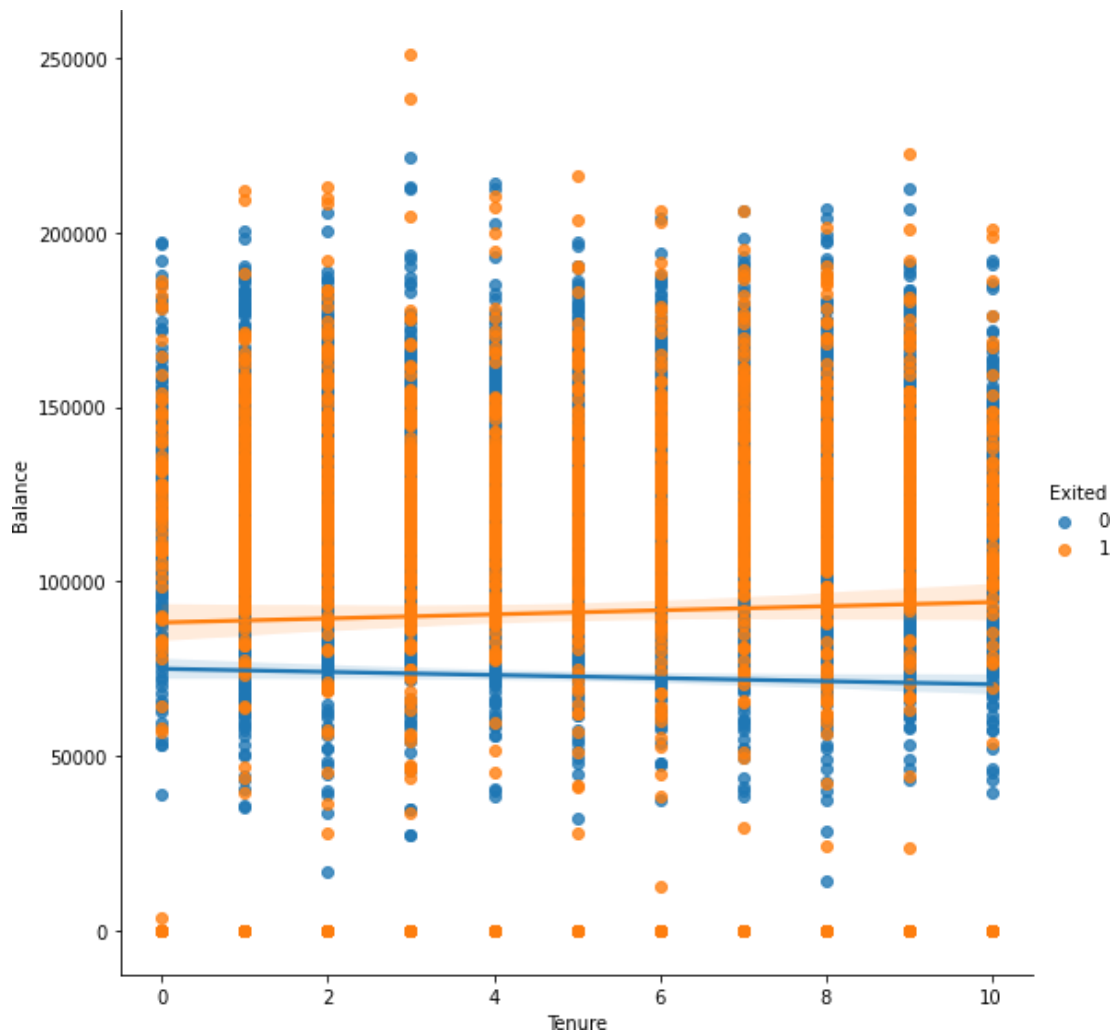
```
Text(0,0.5,'CreditScore')
```



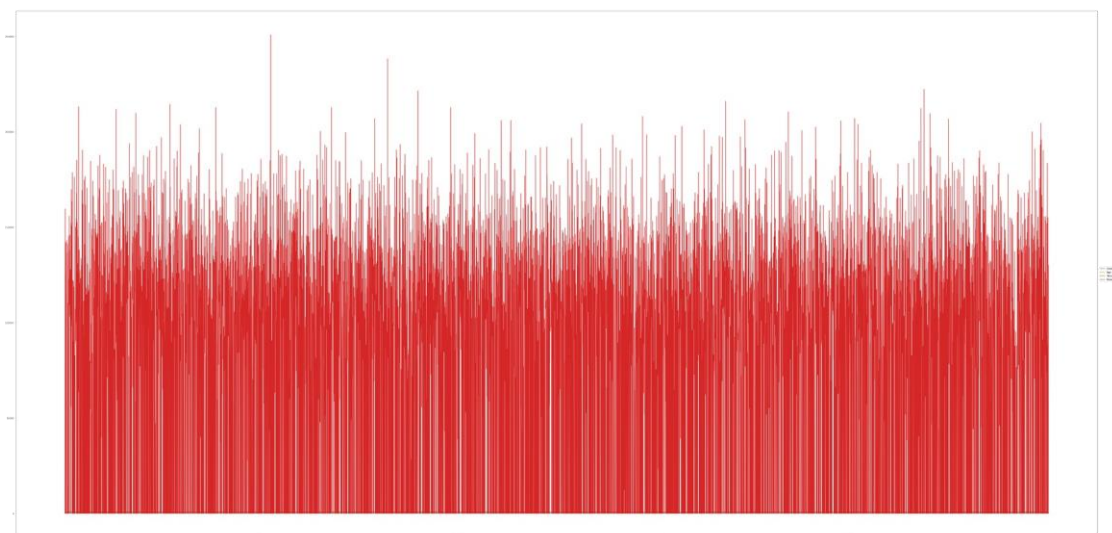
```
sns.lmplot(x='Tenure', y='Balance', data=df, hue='Exited', size=8)

/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>
```



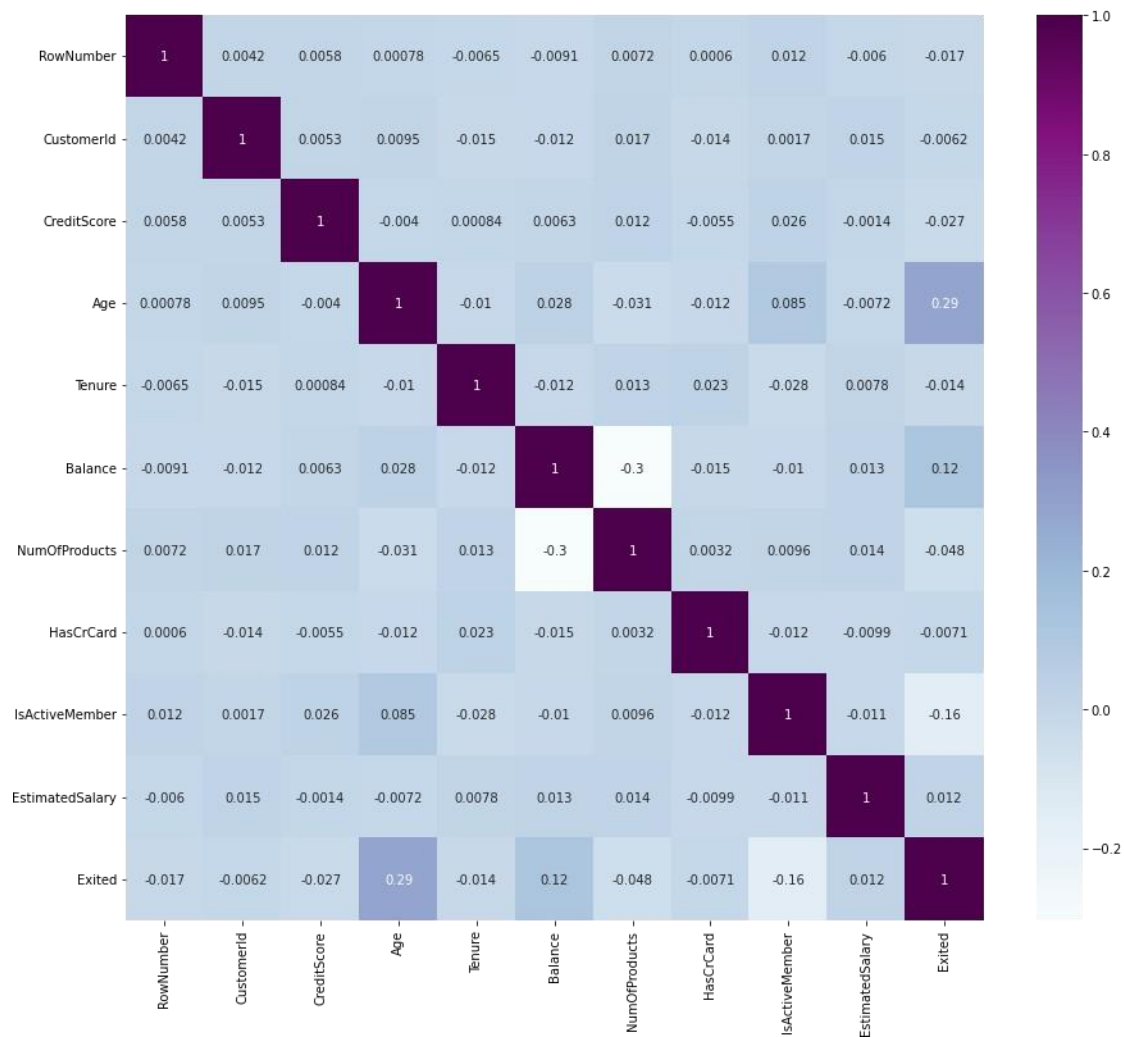
```
#Multi -Variate Analysis
ax=df[["CreditScore","Age","Tenure","Balance"]].plot(figsize=(80,4
0))ax.legend(loc='centerleft',bbox_to_anchor=(1,0.5));
```



```
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(),anno
t=True,cmap='BuPu')plt.show()
```



```
df.drop(['RowNumber',
```

```
'CustomerId', 'Surname'], axis=1, inplace=True) df.head()
```

```

    CreditScore Geography  Gender  Age  Tenure  Balance
NumOfProducts \
0          619    France  Female   42     2     0.00
1
1          608    Spain  Female   41     1  83807.86
1
2          502    France  Female   42     8 159660.80
3
3          699    France  Female   39     1     0.00
2
4          850    Spain  Female   43     2 125510.82
1
```

```

HasCrCardIsActiveMemberEstimatedSalaryExited0    1
1          101348.881
```

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'>RangeIn
dex:10000entries,0to9999Datacolumns(t
otal 11 columns):
#    Column                Non-NullCountDtype
----
0    CreditScore          10000non-nullint64
1    Geography            10000non-nullobject
2    Gender               10000non-nullobject
3    Age                  10000non-nullint64
4    Tenure               10000non-nullint64
5    Balance              10000non-nullfloat64
6    NumOfProducts        10000non-nullint64
7    HasCrCard            10000non-nullint64
8    IsActiveMember       10000non-nullint64
9    EstimatedSalary       10000non-null float64
10   Exited               10000 non-
nullint64dtypes: float64(2),
int64(7),object(2)memoryusage:859.5+KB
```

```
df["Geography"].unique()
```

```
array(['France', 'Spain', 'Germany'],
```

```
dtype=object)df["Gender"].unique()
```

```
array(['Female', 'Male'],dtype=object)geo=pd.get_dummi
```

```
es(df["Geography"],drop_first=False)geo.head()
```

	France	Germany	Spain	0	1
	0	0	0	0	0
1	0	0	0	1	0
2	1	0	0	0	0
3	1	0	0	0	0
4	0	0	0	1	0

```
gen=pd.get_dummies(df["Gender"],drop_first=False)df=p
```

```
d.concat([df,geo,gen],axis=1)
```

```
df
```

	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts\
--	-------------	-----------	--------	-----	--------	---------	----------------

0	619	France	Female	42	2	0.00
1						
1	608	Spain	Female	41	1	83807.86
1						
2	502	France	Female	42	8	159660.80
3						
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	792	France	Female	28	4	130142.79
1						

HasCrCardIsActiveMemberEstimatedSalaryExitedFranceGermany\

0	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						
...
...9						
995	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      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

```

```

[10000rowsx16columns]df.drop(["Geography", "Gender"],
axis=1,inplace=True)df.head()

```

```

      CreditScoreAgeTenure      BalanceNumOfProductsHasCrCard\0      61942
2      0.00  1  1
1      608    41      1    83807.86      1      0
2      502    42      8159660.80      3      1
3      699    39      1      0.00      2      0
4      850    43      2125510.82      1      1

```

```

      IsActiveMemberEstimatedSalaryExitedFranceGermanySpainFemale\
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

	IsActiveMember	EstimatedSalary	France	Germany	Spain	Female
Male						
0	1	101348.88	1	0	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						
...
...						
995	0	96270.64	1	0	0	0
1						
9996	1	101699.77	1	0	0	0
1						
9997	1	42085.58	1	0	0	1
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']

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:

int64df.shape

(10000,14)

x.shape(10000

,13)

y.shape(

10000,)

fromsklearn.model_selectionimporttrain_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]])
```