

# ASSIGNMENT 1

MAHENDRA ENGINEERING COLLEGE FOR WOMEN

NAME:SATHIYA.N

CLASS: 4 YEAR ECE

SUBJECT:IBM

REGISTER NO:611419106056

#libraries

import pandas as

pdimport numpy as

np

import matplotlib.pyplot as plt

%matplotlib inline

#load dataset

\	R o w N u m b e r	C u s t o m e r I d	Sur nam e	CreditSco re	Ge og ra ph y	G e n d e r	A g e
0	1	1 5 6 3 4 6 0 2	Har gra ve	619	Fr an ce	F e m a l e	4 2
1	2	1 5 6 4 7 3 1 1	Hil l	608	Sp ai n	F e m a l e	4 1
2	3	1 5 6 1 9 3	Oni o	502	Fr an ce	F e m a l e	4 2

		0 4					
3	4	1 5 7 0 1 3 5 4	Bon i	699	Fr an ce	F e m a l e	3 9
4	5	1 5 7 3 7 8 8 8	Mit che ll	850	Sp ai n	F e m a l e	4 3
5	6	1 5 5 7 4 0 1 2	Chu	645	Sp ai n	M a l e	4 4
6	7	1 5 5 9 2 5 3 1	Bar tle tt	822	Fr an ce	M a l e	5 0
7	8	1 5 6 5 6 1 4 8	Obi nna	376	Ge rm an y	F e m a l e	2 9
8	9	1 5 7 9 2 3	He	501	Fr an ce	M a l e	4 4

		65					
9	10	15592389	H?	684	France	Male	27
	Tenure	Balance	Num Of Pro ducts	Has C r C a r d	I s A c t i v e M e m b e r	\	
0	2	0 . 0 0	1	1	1		
1	1	83807 .86	1	0	1		
2	8	159660 .80	3	1	0		
3	1	0 . 0 0	2	0	0		
4	2	125510 .82	1	1	1		
5	8	113755 .78	2	1	0		
6	7	0 . 0 0	2	1	1		
7	4	115046 .74	4	1	0		
8	4	142051 .07	2	0	1		

9	2	134603 .88	1	1	1	
	E s t i m a t e d S a l a r y		Exited			
0	1 0 1 3 4 8 . 8 8		1			
1	1 1 2 5 4 2 . 5 8		0			
2	1 1 3 9 3 1 . 5 7		1			

```
df = pd.read_csv(r"/content/Churn_Modelling.csv")
df.head(10)
```

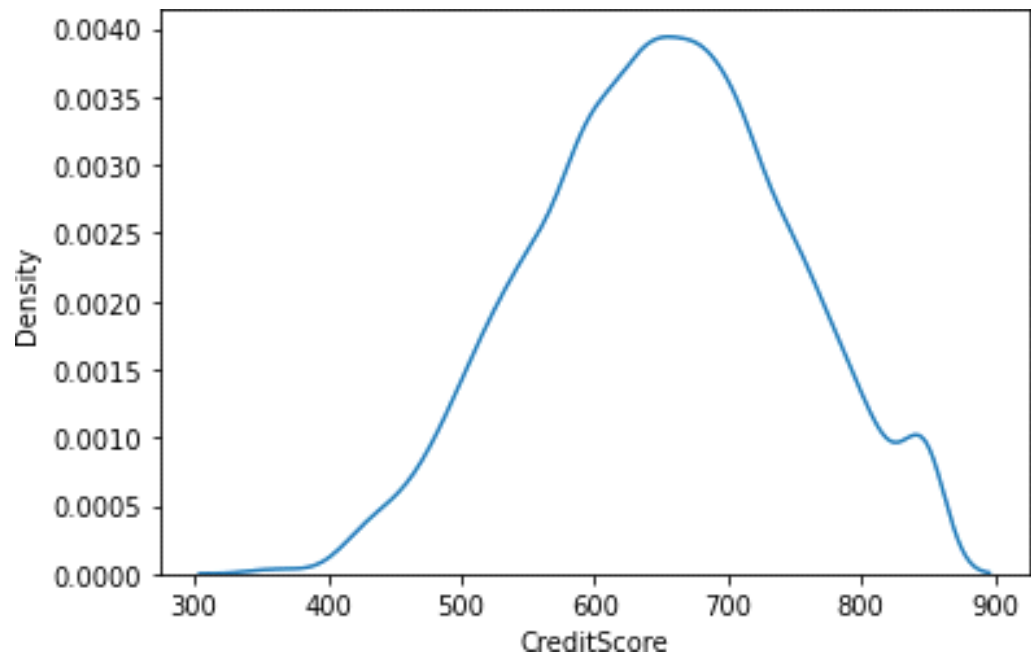
3	93826.63	0
4	79084.10	0
5	149756.71	1
6	10062.80	0
7	119346.88	1
8	74940.50	0
9	71725.73	0
df.info()		

```
<class
'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to
9999 Data columns (total 14
columns):
#      Column                Non-Null Count  Dtype
•      RowNumber            10000 non-null  int64
•      CustomerId           10000 non-null  int64
•      Surname               10000 non-null  object
•      CreditScore           10000 non-null  int64
•      Geography            10000 non-null  object
•      Gender               10000 non-null  object
•      Age                 10000 non-null  int64
•      Tenure              10000 non-null  int64
•      Balance             10000 non-null  float64
•      NumOfProducts       10000 non-null  int64
•      HasCrCard           10000 non-null  int64
•      IsActiveMember     10000 non-null  int64
•      EstimatedSalary     10000 non-null  float64
•      Exited              10000 non-null
int64dtypes: float64(2), int64(9),
object(3) memory usage: 1.1+ MB
```

```
#Visualizations
#Univariate
Analysis import
seaborn as sns
```

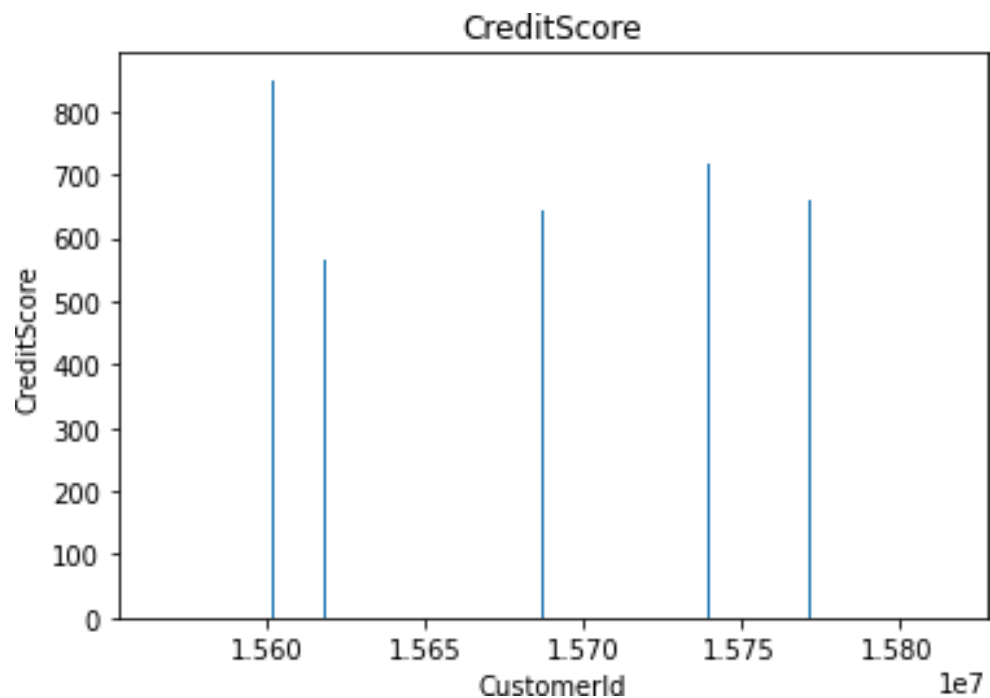
```
sns.kdeplot(df['CreditScore'])
```

```
<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')
```

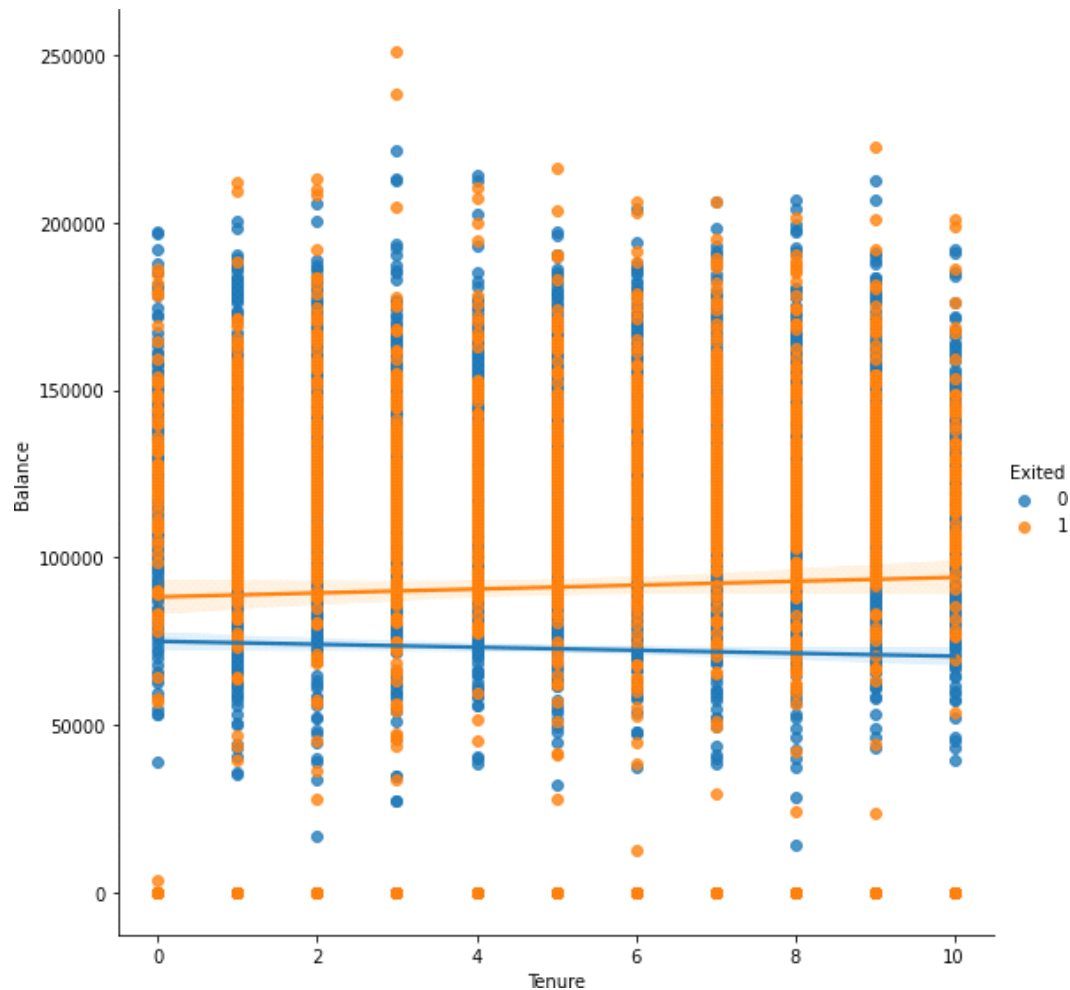


```
sns.lmplot(x='Tenure', y='Balance', data=df ,hue='Exited',size=8)
```

```
/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>
```

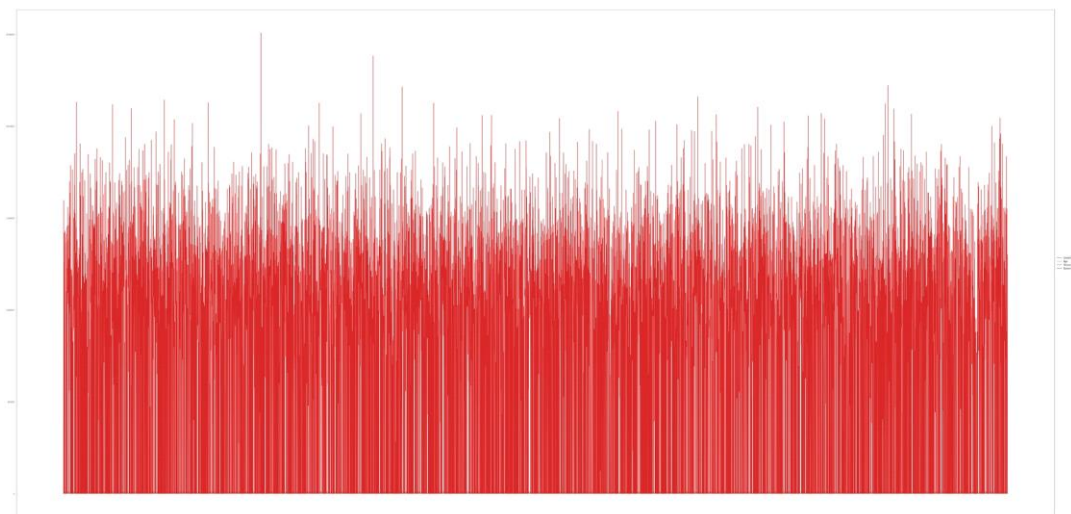


```
#Multi - Variate Analysis
```

```
ax =
```

```
df[["CreditScore", "Age", "Tenure", "Balance"]].plot(figsize=(80,40)
```

```
)ax.legend(loc='center left', bbox_to_anchor=(1, 0.5));
```

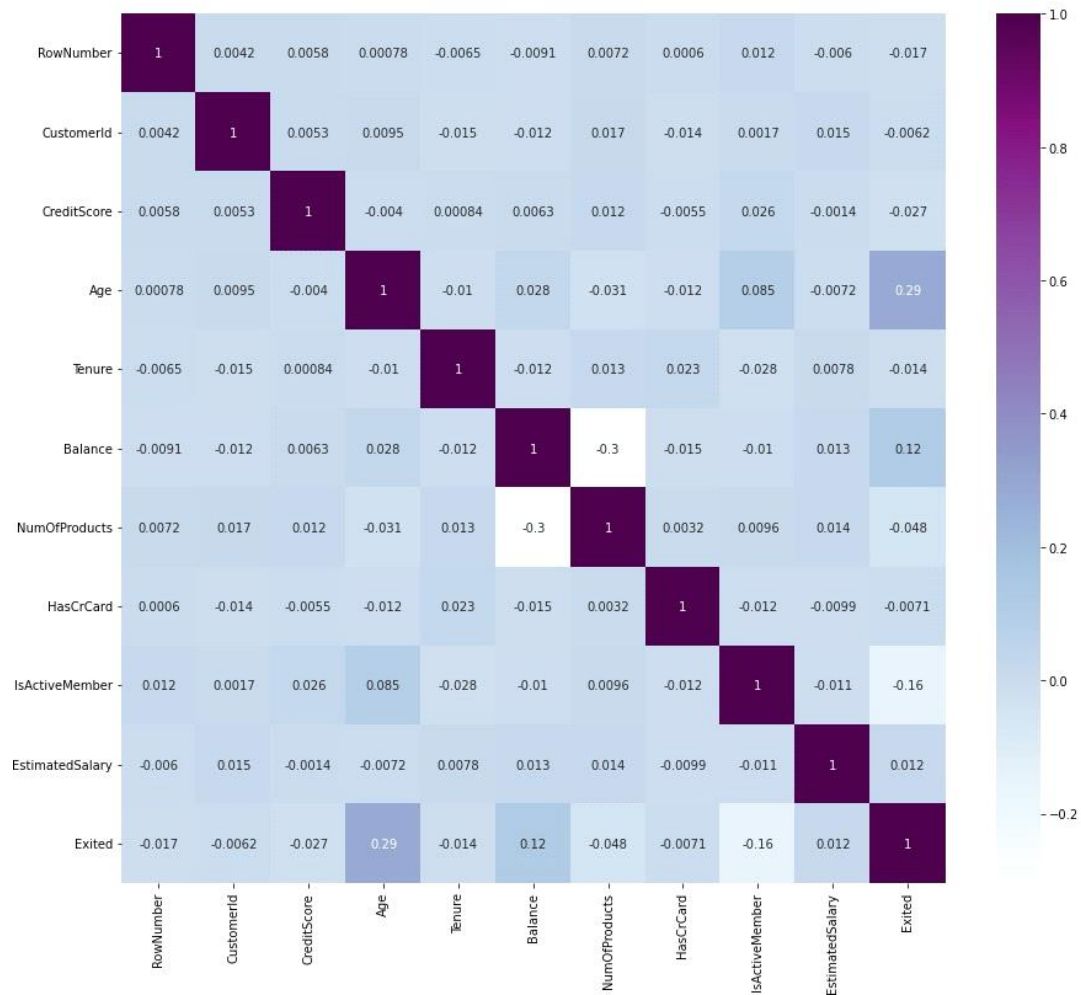


```
df.isnull().sum()
```

```
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
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						

```
HasCrCard    IsActiveMember    EstimatedSalary
Exited0      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
•   CreditScore      10000 non-null  int64
•   Geography        10000 non-null  object
•   Gender            10000 non-null  object
•   Age              10000 non-null  int64
•   Tenure           10000 non-null  int64
•   Balance          10000 non-null  float64
•   NumOfProducts    10000 non-null  int64
•   HasCrCard        10000 non-null  int64
•   IsActiveMember   10000 non-null  int64
•   EstimatedSalary  10000 non-null  float64
•   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
1         0         0
2         1         0
3         1         0
4         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  
 ceNumOfProducts \ Balan

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						

0	1		1013 48.88	1	1
0					
1	0		1125 42.58	0	0
0					
2	1		1139 31.57	1	1
0					
3	0		9382 6.63	0	1
0					
4	1		7908 4.10	0	0
0			..		
..	..	..	.	..	..
..	..	..	..	..	..
.. 9995	1		96270.6 4		1

0					
996	9	1		1016 99.77	0 1
0					
997	9	0		4208 5.58	1 1
0					
998	9	1		9288 8.52	1 0
1					
999	9	1		3819 0.78	0 1
0					
0	pain	emale	M ale		

HasCrCard IsActiveMember EstimatedSalary Exited France  
Germany \

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	101 348	1	1	0	0
---	--	---	------------	---	---	---	---

			.88				
1							
1		1	112 542 .58	0	0	0	1
1							
2		0	113 931 .57	1	1	0	0
1							
3		0	938 26. 63	0	1	0	0
1							
4		1	790 84. 10	0	0	0	1
1							
	Male						
0	0						
1	0						
2	0						
3	0						
4	0						

	Credit Score	Age	Tenure	Balance	NumOfProducts	HasCreditCard	\
0	619	42	2	0.00	1	1	
1	608	41	1	83807.86	1	0	

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

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	IsActiveMember		EstimatedSalary		France	Germany	Spain	Female	
	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				...				
	...		...			...	...	...	...
	...				96270.64				
	9995		0			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.shap
e
(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.sha
pe(8000,
13)
x_test.sha
pe(2000,
13)
y_test.sha
pe(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,