

```
import matplotlib.pyplot as plt
%matplotlib inline

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

	RowNumber	CustomerId	CreditSc	Age
\				
0	1	15634602	1	42
1	2	15647311		41
2	3	15619304		42
3	4	15701354		39
4	5	15737888		43
5	6	15574012		44
6	7	15592531		50
7	8	15656148		29
8	9	15792365		44
9	10	15592389		27

	Tenure	Balance	NumOfPro	\
0	2	0.00		
1	1	83807.86		
2	8	159660.80		
3	1	0.00		
4	2	125510.82		
5	8	113755.78		
6	7	0.00		
7	4	115046.74		
8	4	142051.07		
9	2	134603.88		

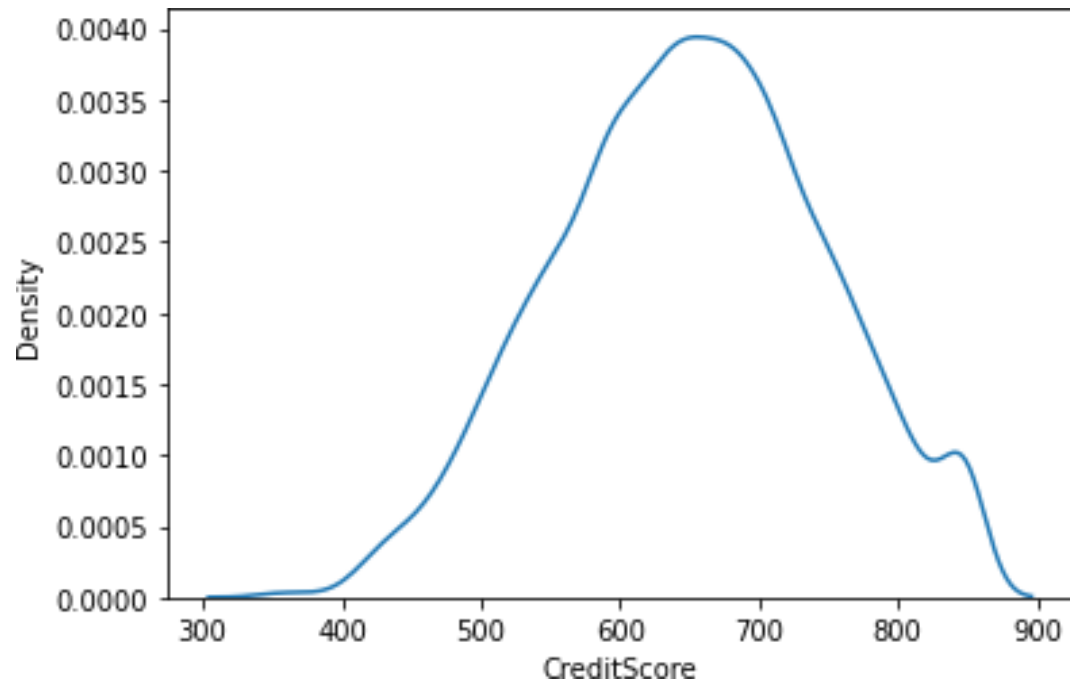
	EstimatedSalary	Exited
0	101348.88	1
1	112542.58	0

df.info()

#	Column	Non-Null Count	Dtype
---	--------	----------------	-------

float64(2), int64(9), object(3) memory usage: 11+ MB

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



#Bi-VariateAnalysis

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

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

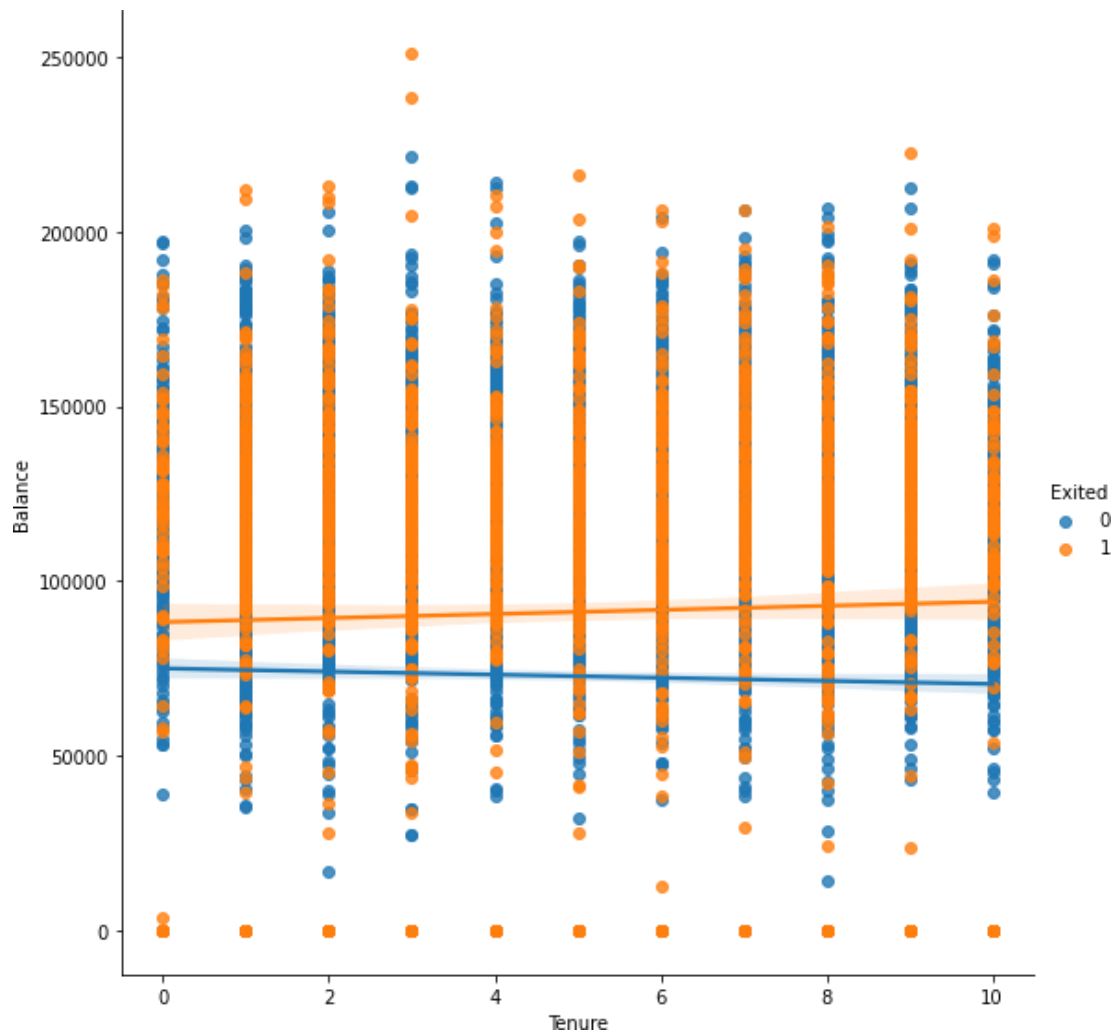


```
sns.Implot(x='Tenure', y='Balance', data=df, hue='Exited', size=8)
```

/usr/local/lib/python3.7/dist-packages/seaborn/regression.py:581: UserWarning: The `size` parameter has been renamed to `height`; please update your code.

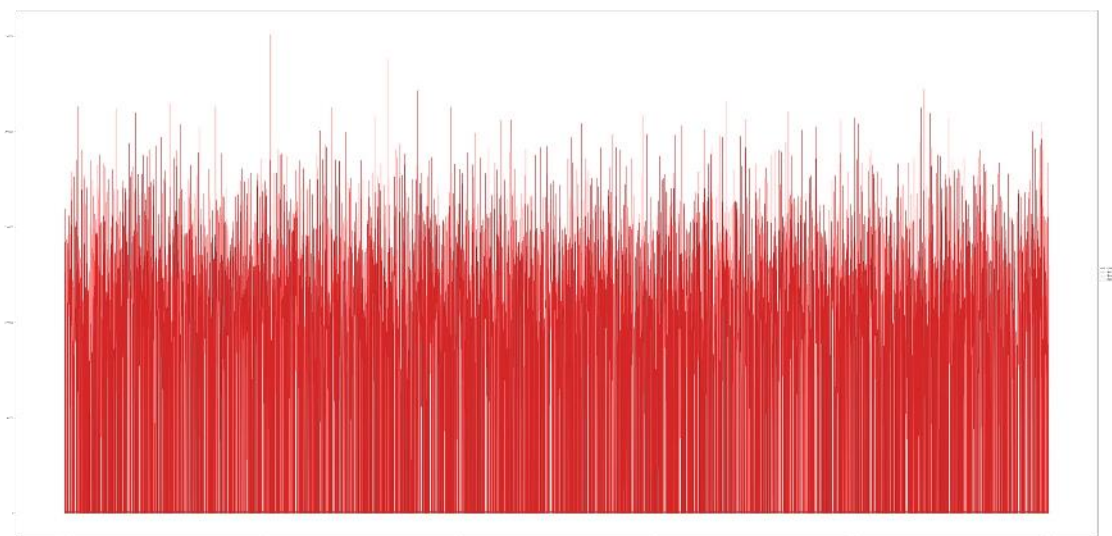
```
warnings.warn(msg, UserWarning)
```

```
<seaborn.axisgrid.FacetGrid at 0x7fc4a149e2d0>
```



#Multi-VariateAnalysis

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

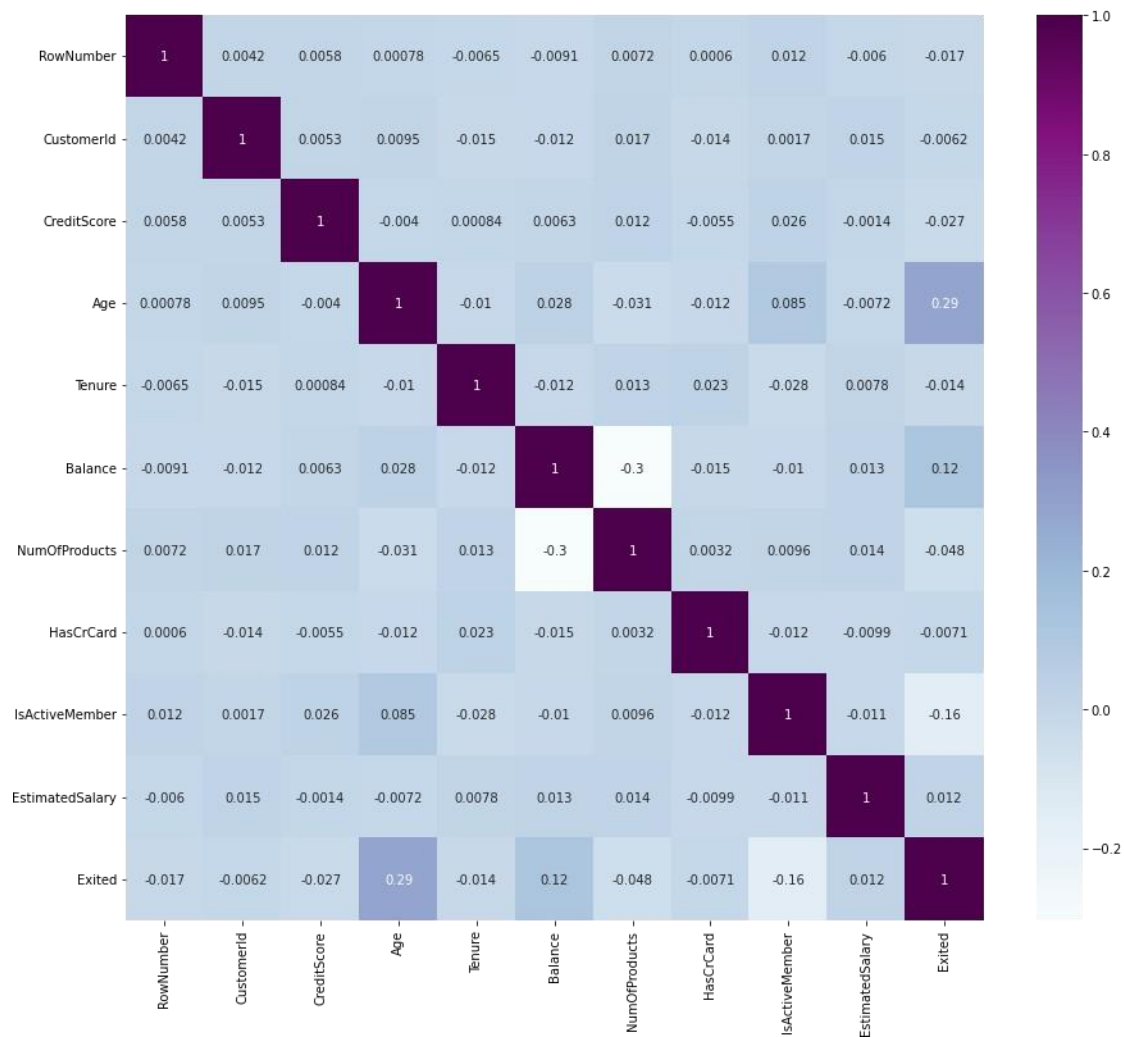


```
df.isnull().sum()
```

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						
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	Exited
0				
1	1		101348.88	1

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	int64

```
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	1
1	0	0	0	1
2	1	0	0	0
3	1	0	0	0
4	0	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						
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						

*HasCrCard IsActiveMember EstimatedSalary Exited FranceGermany *

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					
...
... 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					
0	Spain	Female	Male		
	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	61942
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

	<i>IsActiveMember</i>	<i>EstimatedSalary</i>	<i>France</i>	<i>Germany</i>	<i>Spain</i>	<i>Female</i>
<i>Male</i>						
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						
...
... 9995	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: 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]])
```