

#libraries

import pandas as pd

import numpy as npp

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
```

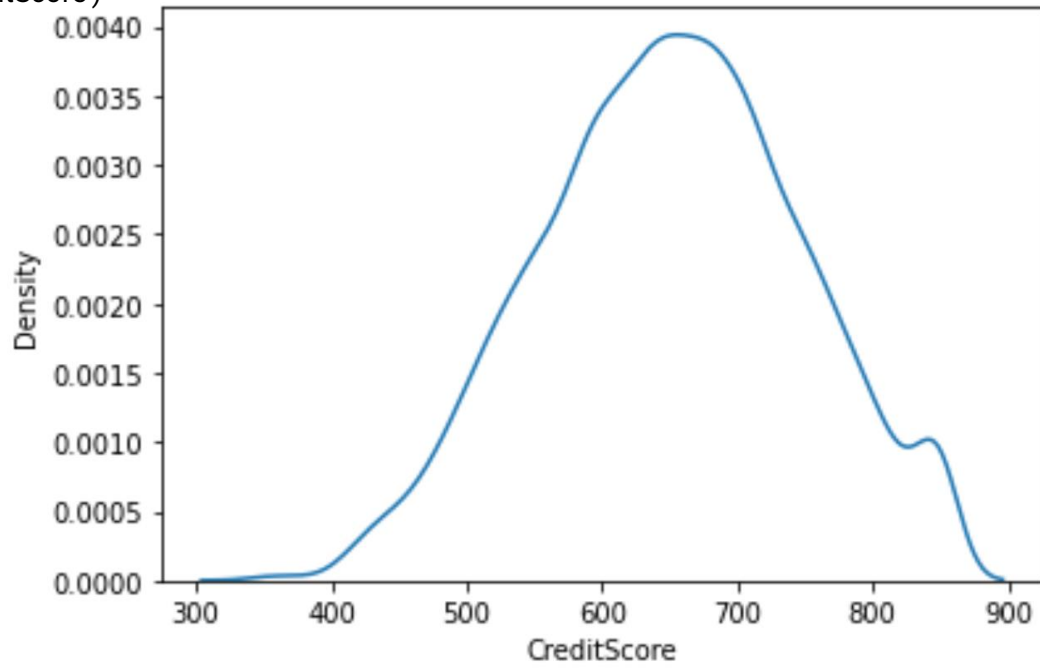
```
plt.bar(df.CustomerId, df.CreditScore)
```

```
plt.title('CreditScore')
```

```
plt.xlabel('CustomerId')
```

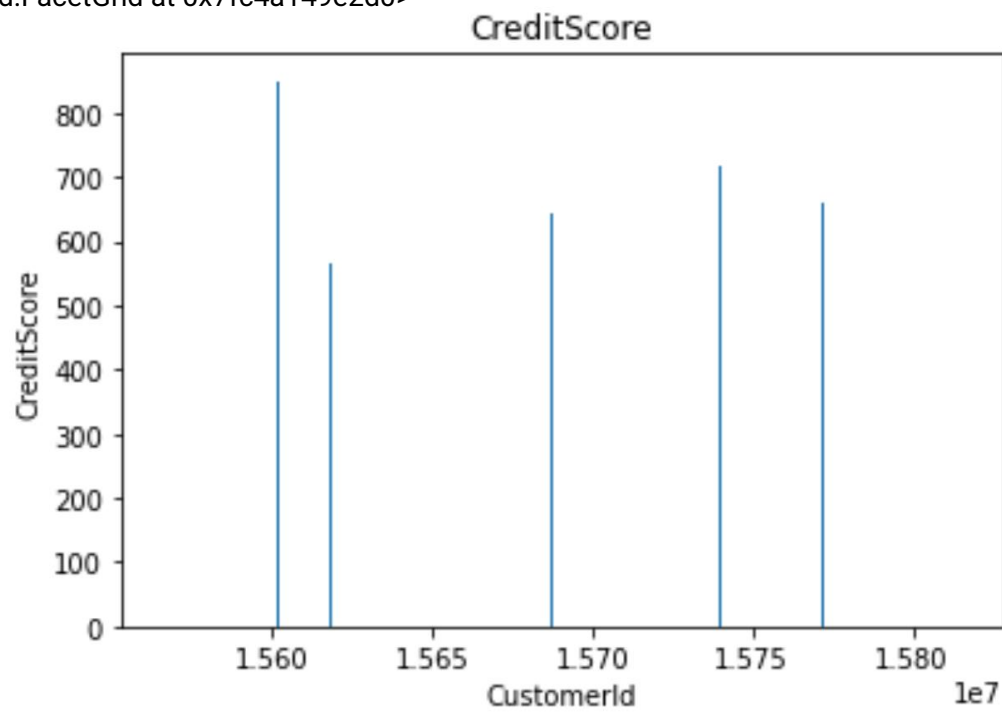
```
plt.ylabel('CreditScore')
```

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



```
sns.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>
```

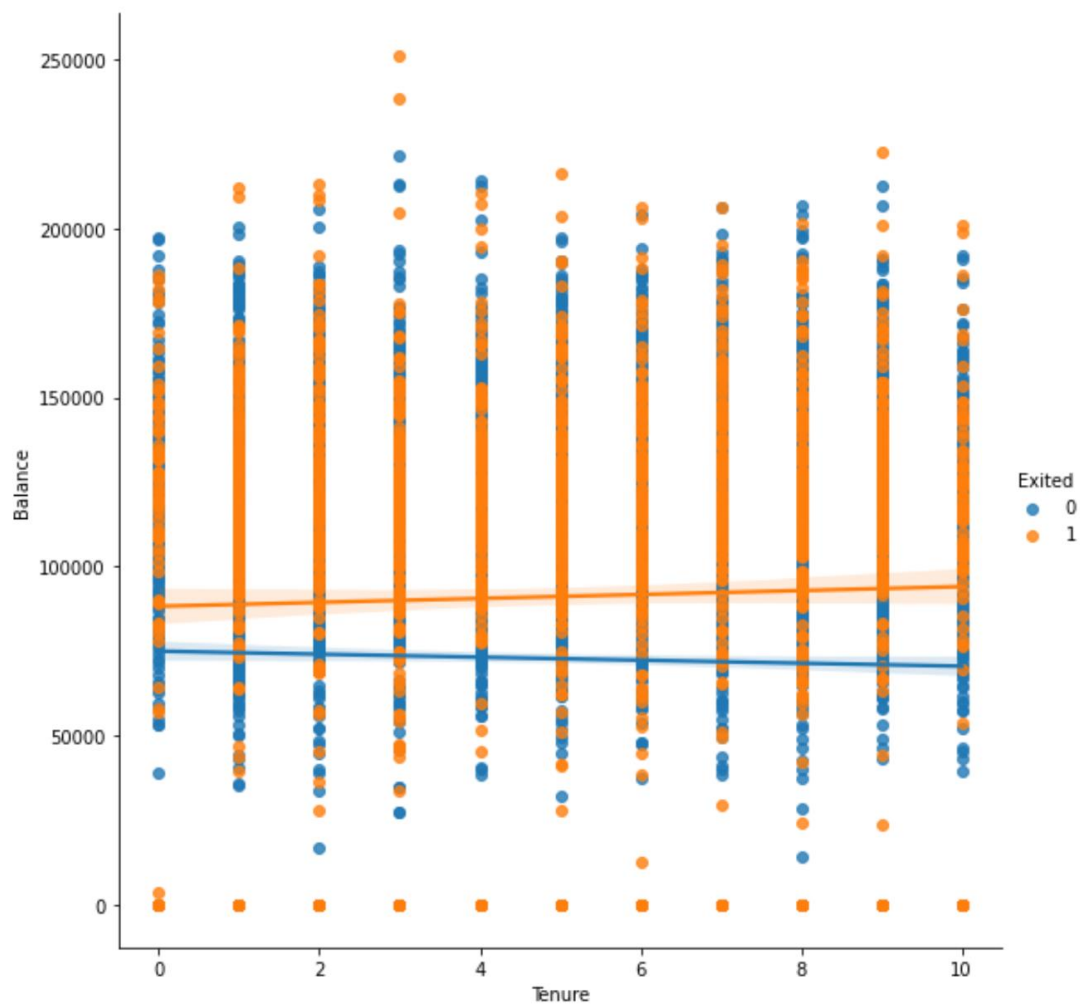


```
#Multi - Variate Analysis
```

```
ax =
```

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

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



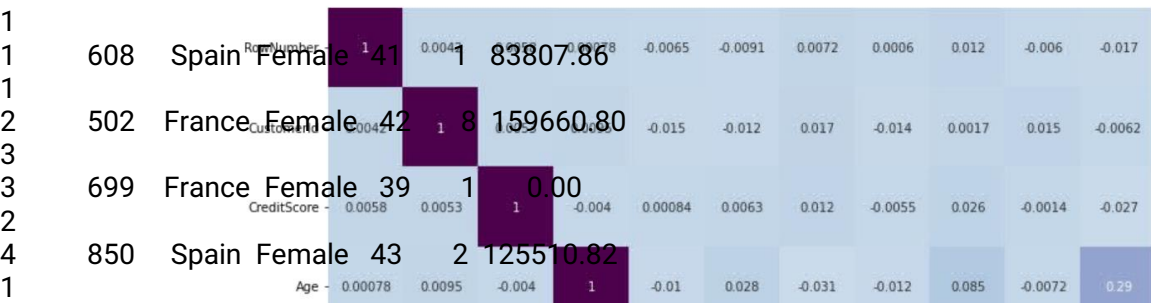
```
df.isnull().sum()
RowNumber      0
CustomerId      0
Surname         0
CreditScore     0
Geography       0
Gender          0
Age            0
Tenure         0
Balance         0
NumOfProducts  0
HasCrCard       0
IsActiveMember  0
EstimatedSalary 0
Exited          0
dtype: int64
plt.figure(figsize=(15,13))
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
```



```
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  |     |         |        |           |     |           |
| 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 France |     |         |        |           |     |           |
| Germany \  |     |         |        |           |     |           |
| 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  |     |         |        |           |     |           |
| 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
```

[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
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
...
...
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]])
```