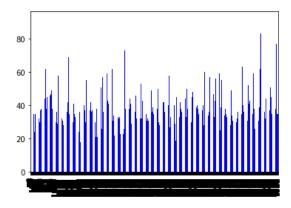
Library Initialization

```
#Required Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import MinMaxScaler
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
#Dataset path initialization
df=pd.read_csv('/content/Churn_Modelling.csv')
```

Dataset Summary

| ат.h | ead() | | | | | | | | | | | | | | | |
|----------------------|----------------|----------------------------------|-------------------------|-------------------|--------------------|---------------------------|----------------------|----------------|-------------|-----------------------|------------|------------|-----------------|------------|--------------------------------|-------------|
| Ro | wNumber | CustomerId | Surname | CreditScore | Geography | Gender | Age | Tenure | Balance | Num | OfProducts | HasCrCard | IsActiveMembe | er Est | imatedSalary E | xited |
| 0 | 1 | 15634602 | Hargrave | 619 | France | Female | 42 | 2 | 0.00 |) | 1 | 1 | | 1 | 101348.88 | 1 |
| 1 | 2 | 15647311 | Hill | 608 | Spain | Female | 41 | 1 | 83807.86 | 5 | 1 | 0 | | 1 | 112542.58 | 0 |
| 2 | 3 | 15619304 | Onio | 502 | France | Female | 42 | 8 | 159660.80 |) | 3 | 1 | | 0 | 113931.57 | 1 |
| 3 | 4 | 15701354 | Boni | 699 | France | Female | 39 | 1 | 0.00 |) | 2 | 0 | | 0 | 93826.63 | 0 |
| 4 | 5 | 15737888 | Mitchell | 850 | Spain | Female | 43 | 2 | 125510.82 | 2 | 1 | 1 | | 1 | 79084.10 | 0 |
| df.t | ail() | | | | | | | | | | | | | | | |
| | RowNumb | er Custome | erId Surna | ıme CreditSc | ore Geogra | phy Gen | der | Age Ter | nure Ba | ance | NumOfProdu | cts HasCrC | Card IsActiveMe | ember | EstimatedSalar | y Exit |
| 9995 | | oer Custome | | | | - | i der Male | Age Ter | nure Ba | lance | NumOfProdu | cts HasCrC | Card IsActiveMe | ember 0 | | _ |
| | 99 | 96 15606 | | aku | 771 Fra | nce N | | _ | 5 | | NumOfProdu | | | | 96270.6 | 4 |
| 9995 9996 9997 | 99 | 96 15606 | 229 Obiji 892 Johnst | aku one | 771 Fra 516 Fra | nce N | //ale //ale | 39 | 5 | 0.00 | NumOfProdu | 2 | 1 | 0 | 96270.6 101699.7 | 4 7 |
| 9996 | 99 99 99 | 96 15606 97 15569 98 15584 | 229 Obiji 892 Johnst | aku one Liu | 771 Fra 516 Fra | nce N nce N nce Fen | //ale //ale | 39 35 | 5 10 573 | 0.00 69.61 0.00 | NumOfProdu | 2 | 1 | 0 | 96270.6 101699.7 42085.5 | 4 7 8 |

```
df.shape
(10000, 14)
 df.isnull().sum()
RowNumber
                   0
CustomerId
Surname
                   0
0
Surname
CreditScore 0
Gender
Age
Tenure 0
Balance 0
NumOfProducts 0
HasCrCard 0
IsActiveMember 0
EstimatedSalary 0
Exited
dtype: int64
 df.drop(['RowNumber','CustomerId','Surname'],axis=1, inplace=True)
 #Data visualization
 df.Geography.value_counts().plot(kind='hist',color="Purple")
 df.Geography.value_counts()
France
          5014
Germany
          2509
Spain
          2477
Name: Geography, dtype: int64
  2.00
  1.75
  1.50
 <sub>글</sub> 1.25
 Lednen
1.00
  0.75
  0.50
  0.25
  0.00
                      3500
                                      4500
               3000
                              4000
 df.Age.describe()
count 10000.000000
           38.921800
mean
           10.487806
std
           18.000000
           32.000000
37.000000
25%
50%
75%
           44.000000
           92.000000
max
Name: Age, dtype: float64
 df.Age.plot(kind='bar',color="blue")
```

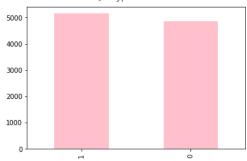


```
df.IsActiveMember.value_counts().plot(kind='bar',color="pink")
df.IsActiveMember.value_counts()
```

```
df.IsActiveMember.value_counts().plot(kind='bar',color="pink")
df.IsActiveMember.value_counts()
```

1 5151 0 4849

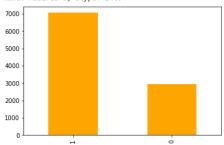
Name: IsActiveMember, dtype: int64



df.HasCrCard.value_counts().plot(kind='bar',color="Orange")
df.HasCrCard.value_counts()

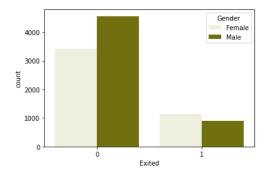
1 7055

Name: HasCrCard, dtype: int64

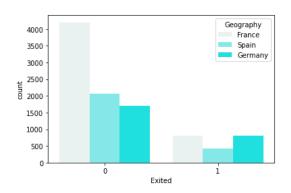


```
df.Tenure.value_counts().plot(kind='bar',color="SkyBlue");
df.Tenure.value_counts()
2
        1048
        1035
        1028
8
        1025
5
        1012
3
        1009
4
9
         989
         984
6
10
         967
         490
         413
0
Name: Tenure, dtype: int64
1000
  800
  600
  400
  200
  \begin{tabular}{ll} $\sf df.NumOfProducts.value\_counts().plot(kind='bar',color="black"); \\ $\sf df.NumOfProducts.value\_counts() \end{tabular} 
      5084
4590
       266
        60
Name: NumOfProducts, dtype: int64
5000
 4000
 3000
 2000
 1000
 df.Exited.value_counts().plot(kind='bar',color="darkgreen");
df.Exited.value_counts()
0
Name: Exited, dtype: int64
 8000
 7000
 6000
 5000
 4000
 3000
 2000
 1000
                      0
```

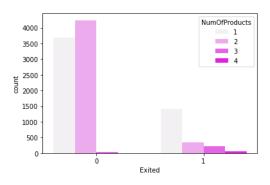
$\verb|sns.countplot(x=df.Exited, hue=df.Gender, color="Olive")|\\$



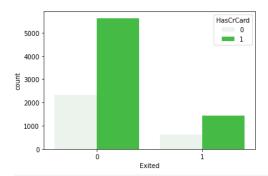
 $\verb|sns.countplot(x=df.Exited, hue=df.Geography, color="cyan")| \\$



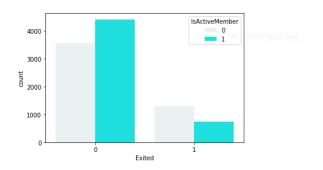
sns.countplot(x=df.Exited,hue=df.NumOfProducts,color="fuchsia")



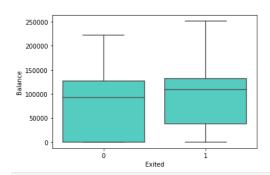
sns.countplot(x=df.Exited,hue=df.HasCrCard,color="limegreen")



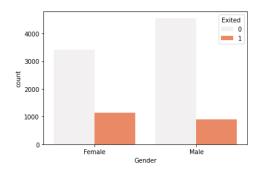
 $\verb|sns.countplot(x=df.Exited, hue=df.IsActiveMember, color="aqua")|\\$



sns.boxplot(x=df.Exited,y=df.Balance,color="turquoise")



 $\verb|sns.countplot(x="Gender", hue="Exited", data=df, color="coral")|\\$



```
df['Geography']=df['Geography'].map({'France':0,'Spain':1,'Germany':2})

X=df.iloc[:,:-1].values

X.shape

(10000, 10)

#Feature Scaling of Data Set
le-LabelEncoder()

X[:,2]=le.fit_transform(X[:,2])

print(X)

[[619 0 0 ... 1 1 101348.88]
[608 1 0 ... 0 1 112542.58]
[502 0 0 ... 1 0 113943.58]
[702 0 0 ... 1 0 113943.58]
[709 0 0 ... 1 0 13943.52]

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```