

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
import warnings
warnings.filterwarnings('ignore')
```

```
df=pd.read_csv('/content/Churn Modelling.csv')
```

```
df.head()
```



|   | RowNumber | CustomerId | Surname  | CreditScore | Geography | Gender | Age | Tenure |
|---|-----------|------------|----------|-------------|-----------|--------|-----|--------|
| 0 | 1         | 15634602   | Hargrave | 619         | France    | Female | 42  | 2      |
| 1 | 2         | 15647311   | Hill     | 608         | Spain     | Female | 41  | 1      |
| 2 | 3         | 15619304   | Onio     | 502         | France    | Female | 42  | 8      |
| 3 | 4         | 15701354   | Boni     | 699         | France    | Female | 39  | 1      |
| 4 | 5         | 15737888   | Mitchell | 850         | Spain     | Female | 43  | 2      |

```
df.describe()
```

|       | RowNumber   | CustomerId   | CreditScore  | Age          | Tenure       | Balar         |
|-------|-------------|--------------|--------------|--------------|--------------|---------------|
| count | 10000.00000 | 1.000000e+04 | 10000.000000 | 10000.000000 | 10000.000000 | 10000.000000  |
| mean  | 5000.50000  | 1.569094e+07 | 650.528800   | 38.921800    | 5.012800     | 76485.889000  |
| std   | 2886.89568  | 7.193619e+04 | 96.653299    | 10.487806    | 2.892174     | 62397.405000  |
| min   | 1.00000     | 1.556570e+07 | 350.000000   | 18.000000    | 0.000000     | 0.000000      |
| 25%   | 2500.75000  | 1.562853e+07 | 584.000000   | 32.000000    | 3.000000     | 0.000000      |
| 50%   | 5000.50000  | 1.569074e+07 | 652.000000   | 37.000000    | 5.000000     | 97198.540000  |
| 75%   | 7500.25000  | 1.575323e+07 | 718.000000   | 44.000000    | 7.000000     | 127644.240000 |
| max   | 10000.00000 | 1.581569e+07 | 850.000000   | 92.000000    | 10.000000    | 250898.090000 |

```
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

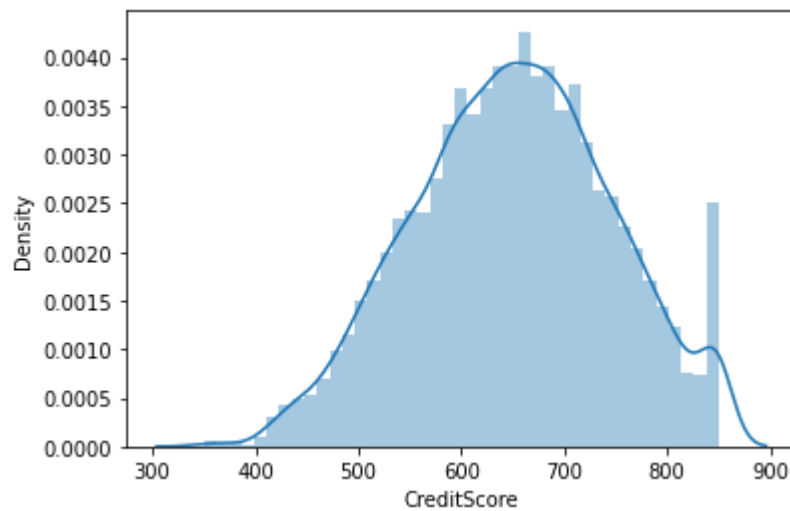
```

```
df.head(2)
```

|   | RowNumber | CustomerId | Surname  | CreditScore | Geography | Gender | Age | Tenure |
|---|-----------|------------|----------|-------------|-----------|--------|-----|--------|
| 0 | 1         | 15634602   | Hargrave | 619         | France    | Female | 42  |        |
| 1 | 2         | 15647311   | Hill     | 608         | Spain     | Female | 41  |        |

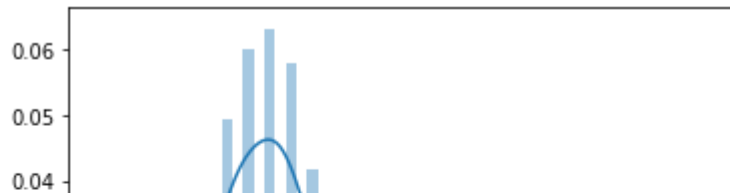
```
sns.distplot(df.CreditScore)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaafd4dd0>
```



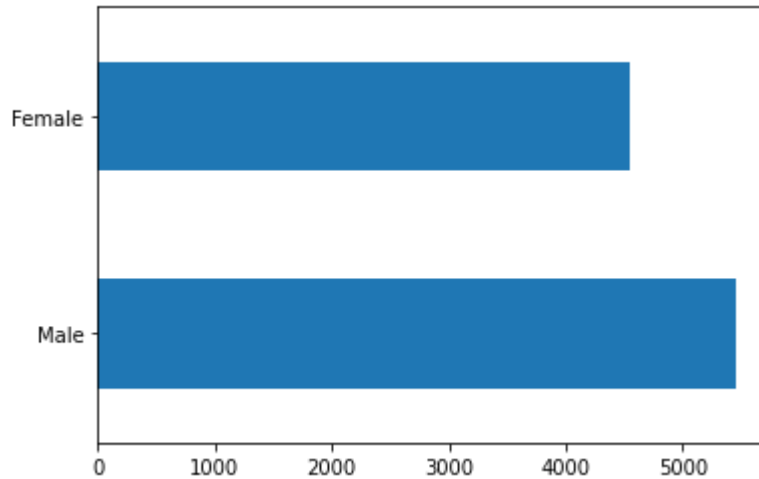
```
sns.distplot(df.Age)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaaed6b50>
```



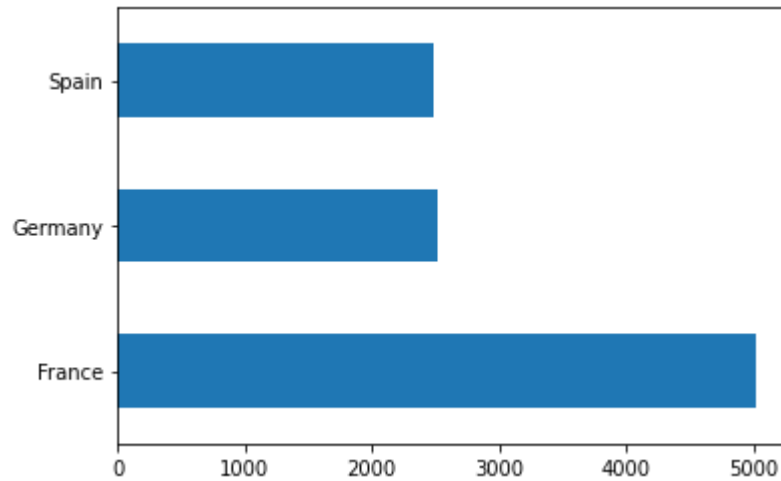
```
df.Gender.value_counts().plot(kind='barh')
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa87e550>
```



```
df.Geography.value_counts().plot(kind='barh')
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa7e3410>
```



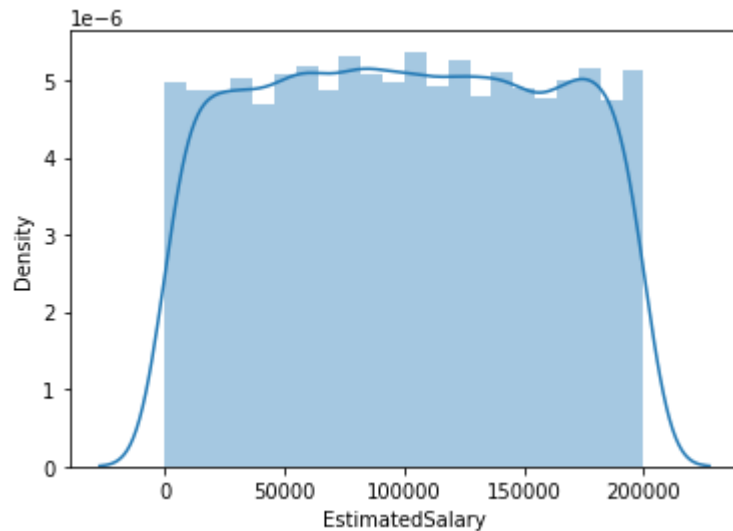
```
df.Tenure.value_counts().plot(kind='barh')
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa7508d0>
```



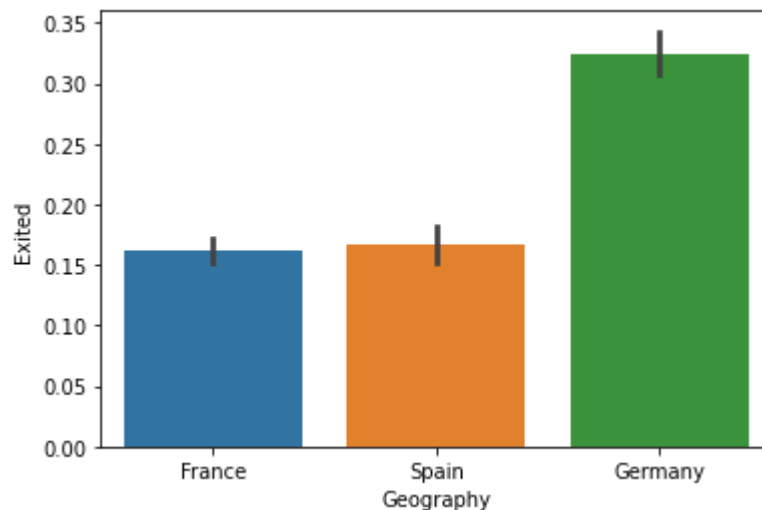
```
sns.distplot(df.EstimatedSalary)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa75f150>
```



```
sns.barplot(df.Geography, df.Exited)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa626550>
```

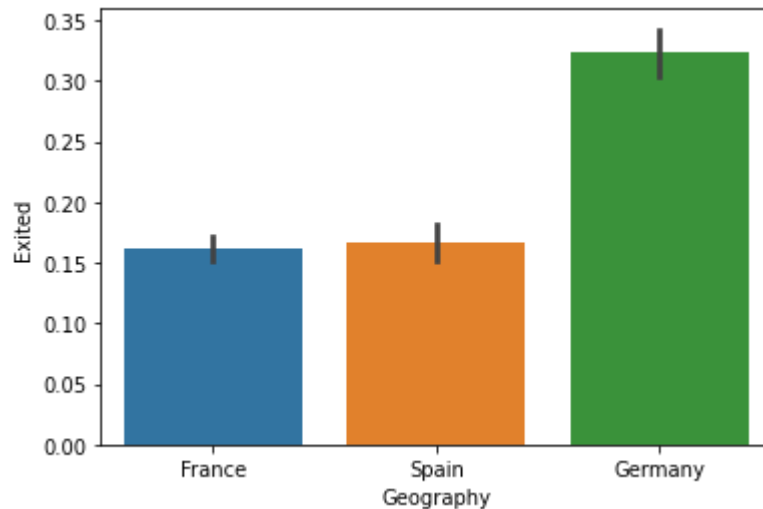


```
df.head(2)
```

|   | RowNumber | CustomerId | Surname  | CreditScore | Geography | Gender | Age | Tenure |
|---|-----------|------------|----------|-------------|-----------|--------|-----|--------|
| 0 | 1         | 15634602   | Hargrave | 619         | France    | Female | 42  |        |
| 1 | 2         | 15647311   | Hill     | 608         | Spain     | Female | 41  |        |

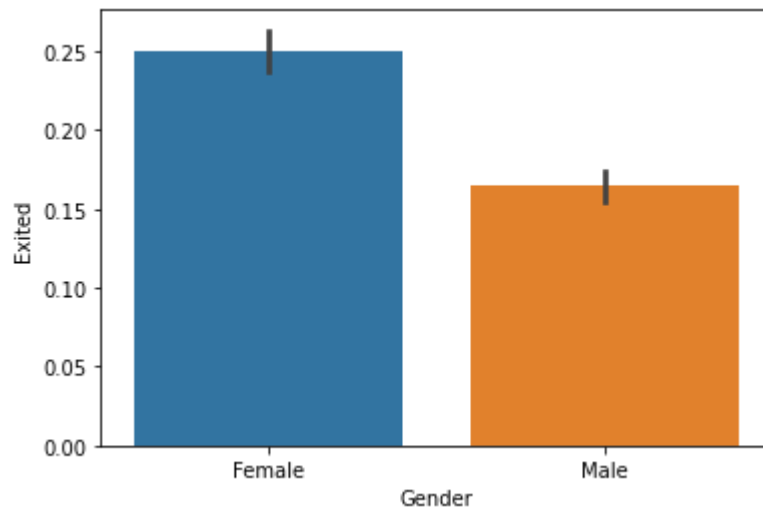
```
sns.barplot(x='Geography', y='Exited', data=df)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa599c50>
```



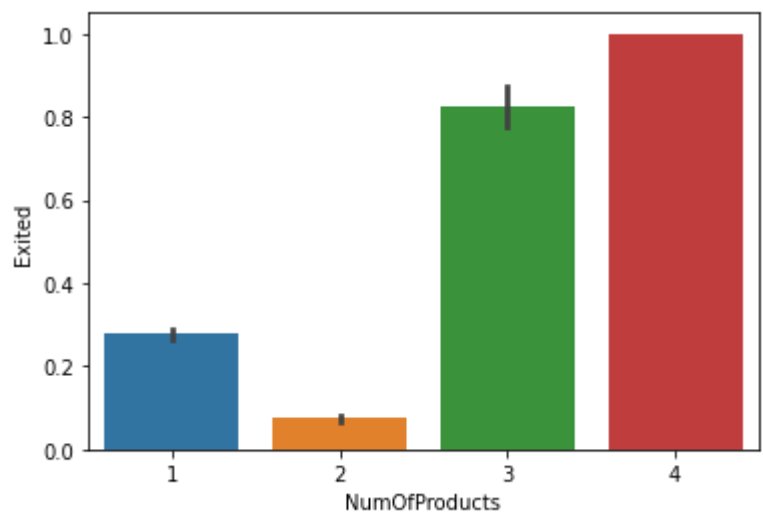
```
sns.barplot(x='Gender',y='Exited',data=df)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa50fdd0>
```



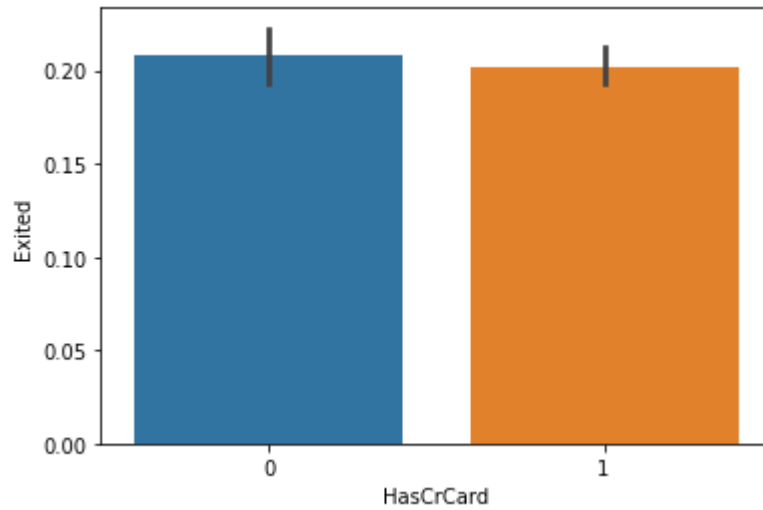
```
sns.barplot(x='NumOfProducts',y='Exited',data=df)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa4ef6d0>
```



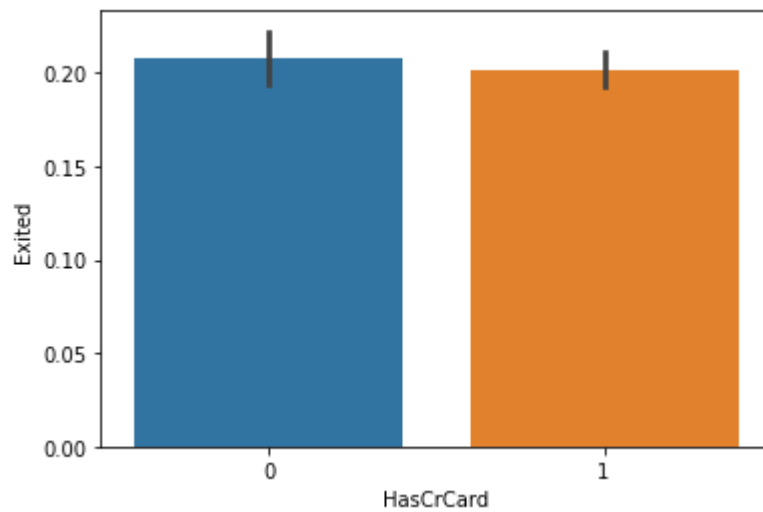
```
sns.barplot(x='HasCrCard',y='Exited',data=df)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa46b990>
```



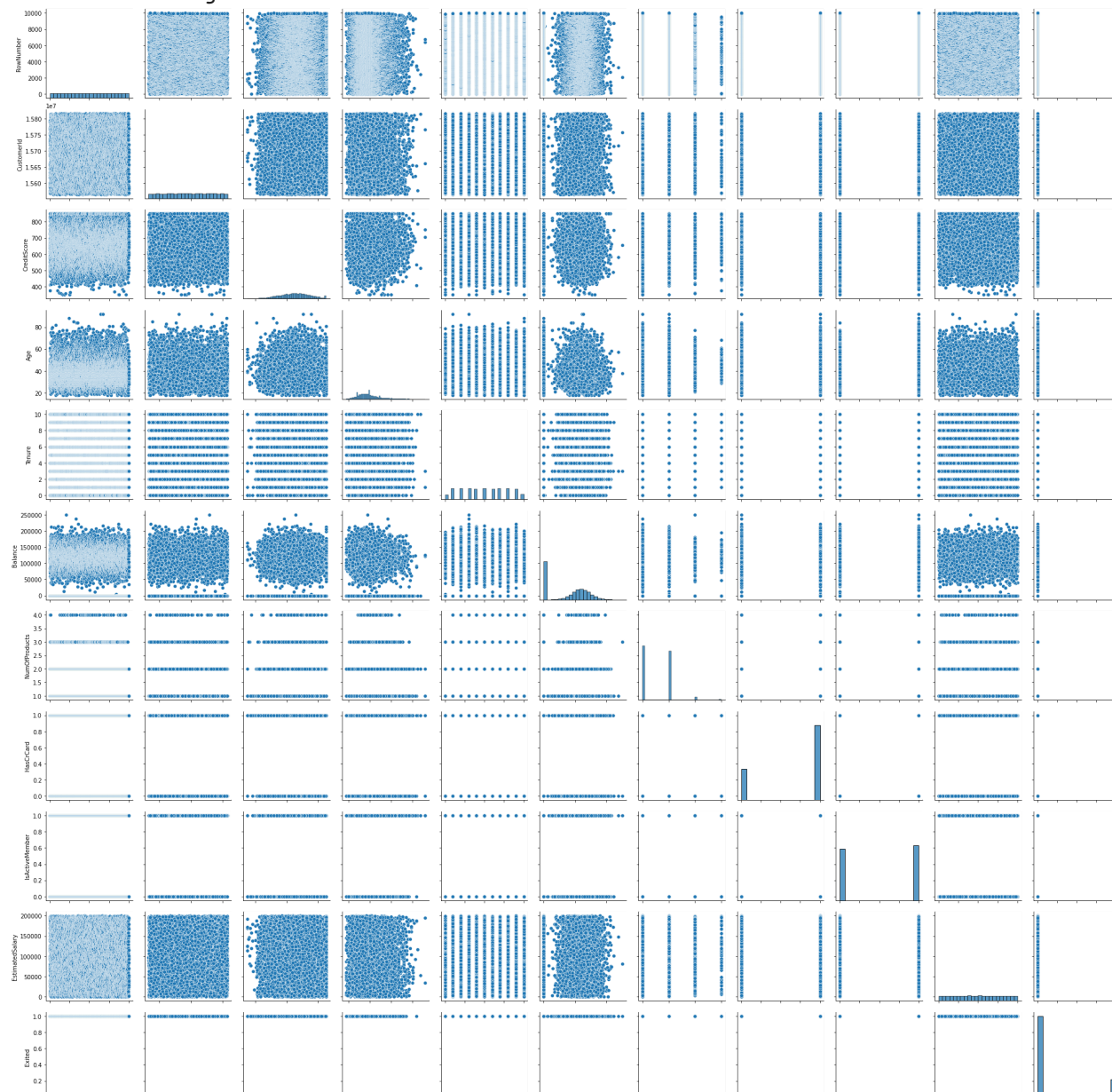
```
sns.barplot(x='HasCrCard',y='Exited',data=df)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaa448690>
```



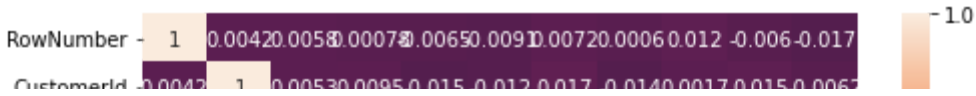
```
sns.pairplot(df)
```

<seaborn.axisgrid.PairGrid at 0x7f4aaa6e4890>



```
plt.figure(figsize=(8,5))
sns.heatmap(df.corr(),annot=True)
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f4aa4ca1d50>



df.Exited.value\_counts()

```
0    7963
1    2037
Name: Exited, dtype: int64
```



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
```

df.head(2)

|   | RowNumber | CustomerId | Surname  | CreditScore | Geography | Gender | Age | Tenure |
|---|-----------|------------|----------|-------------|-----------|--------|-----|--------|
| 0 | 1         | 15634602   | Hargrave | 619         | France    | Female | 42  |        |
| 1 | 2         | 15647311   | Hill     | 608         | Spain     | Female | 41  |        |

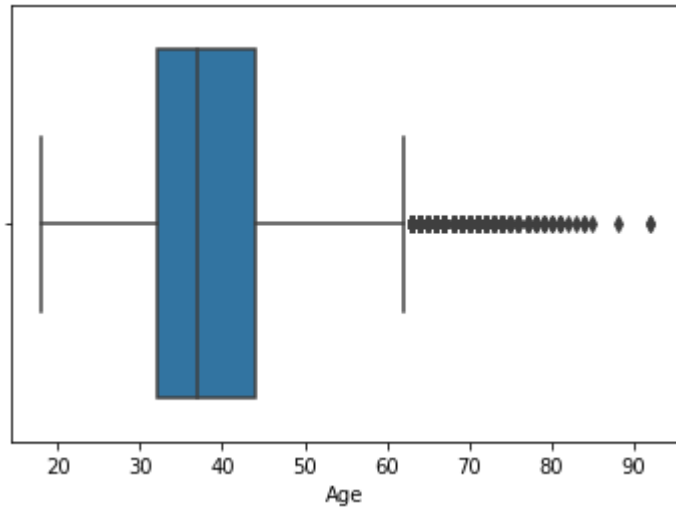
sns.boxplot(df.CreditScore)



```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aa513eb50>
```

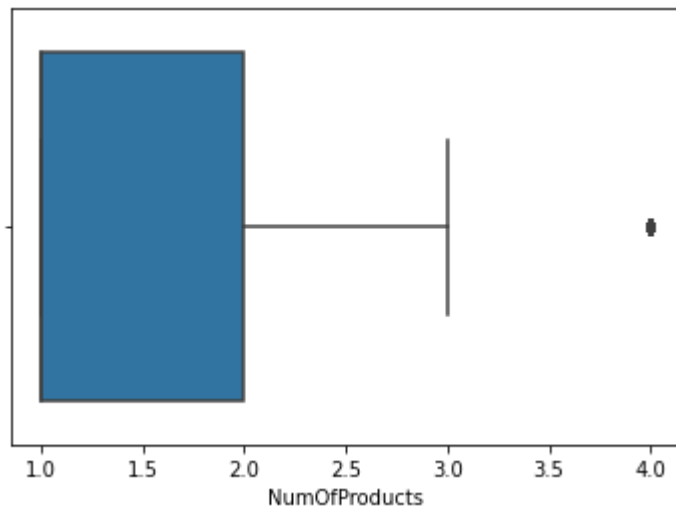
```
sns.boxplot(df.Age)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aa2f7eed0>
```



```
sns.boxplot(df.NumOfProducts)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aa2ee56d0>
```

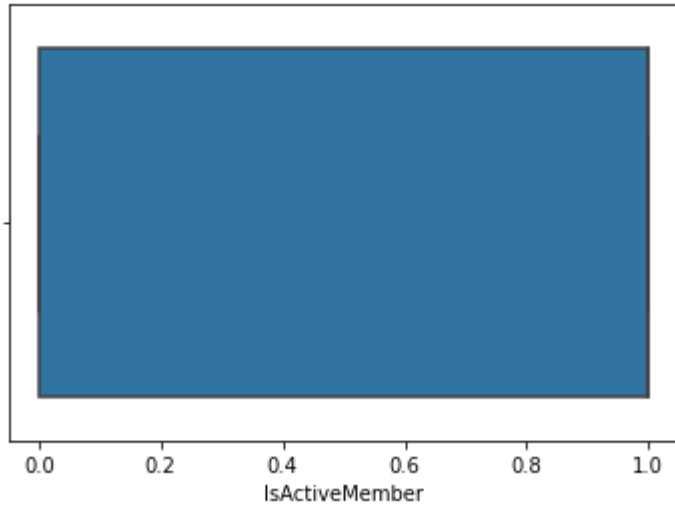


```
sns.boxplot(df.HasCrCard)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aa2e5a790>
```

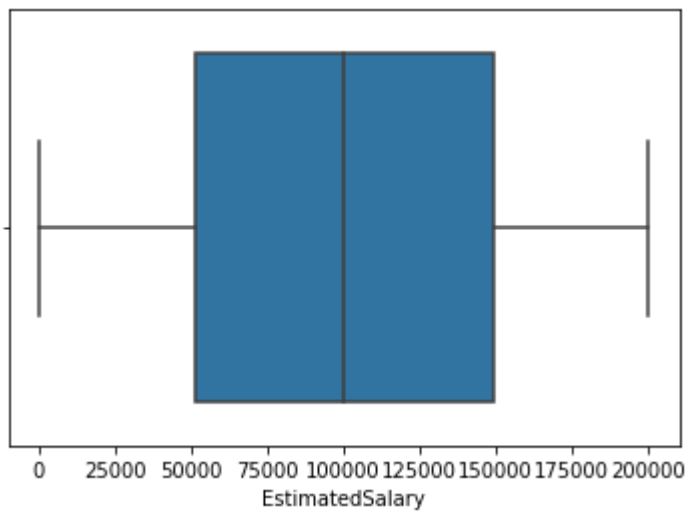
```
sns.boxplot(df.IsActiveMember)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aa2e73d50>
```



```
sns.boxplot(df.EstimatedSalary)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aa2daa650>
```

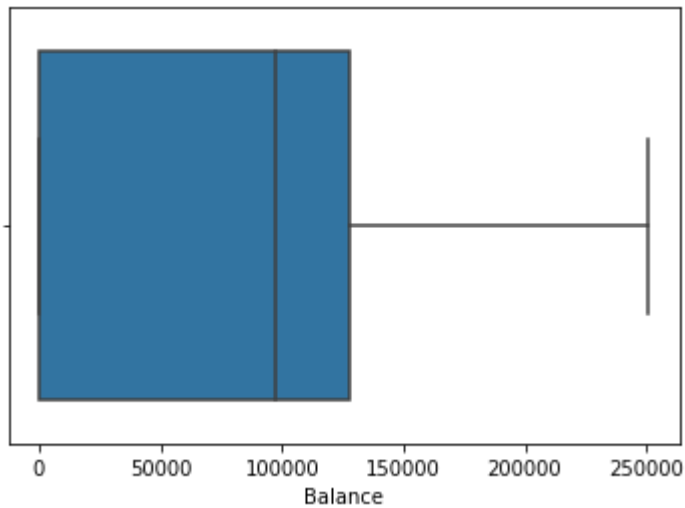


```
sns.boxplot(df.Tenure)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aa2f45390>
```

```
sns.boxplot(df.Balance)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aa2d08550>
```

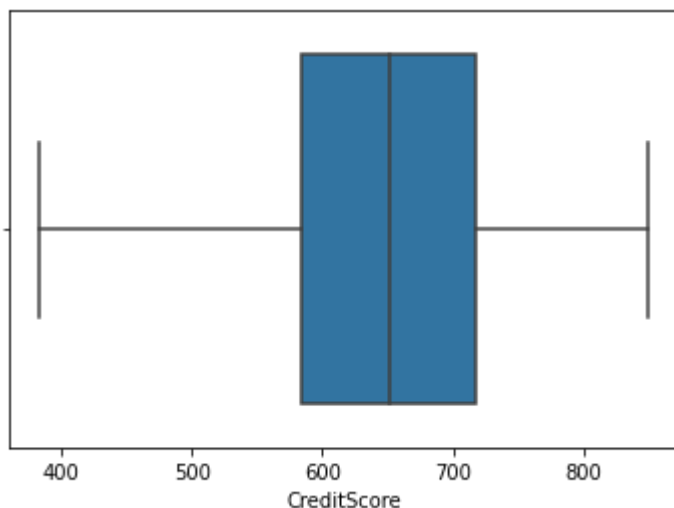


```
def outlier_credit_score(df):  
    IQR = df['CreditScore'].quantile(0.75) - df['CreditScore'].quantile(0.25)  
  
    lower_range = df['CreditScore'].quantile(0.25) - (1.5 * IQR)  
    upper_range = df['CreditScore'].quantile(0.75) + (1.5 * IQR)  
  
    df.loc[df['CreditScore'] <= lower_range, 'CreditScore'] = lower_range  
    df.loc[df['CreditScore'] >= upper_range, 'CreditScore'] = upper_range
```

```
outlier_credit_score(df)
```

```
sns.boxplot(df.CreditScore)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aaafc9390>
```



```
def outlier_NOP(df):  
    IQR = df['NumOfProducts'].quantile(0.75) - df['NumOfProducts'].quantile(0.25)  
  
    lower_range = df['NumOfProducts'].quantile(0.25) - (1.5 * IQR)
```

```
upper_range = df['NumOfProducts'].quantile(0.75) + (1.5 * IQR)
```

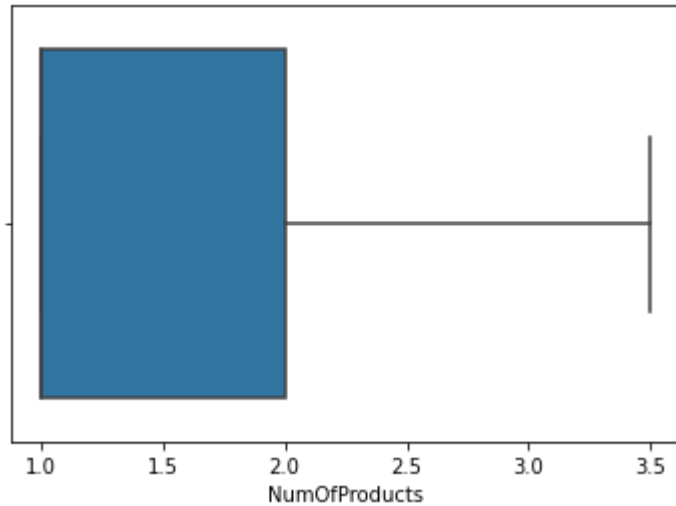
```
df.loc[df['NumOfProducts'] <= lower_range, 'NumOfProducts'] = lower_range
```

```
df.loc[df['NumOfProducts'] >= upper_range, 'NumOfProducts'] = upper_range
```

```
outlier_NOP(df)
```

```
sns.boxplot(df.NumOfProducts)
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f4aa2bcd90>



```
def outlier_age(df):
```

```
    IQR = df['Age'].quantile(0.75) - df['Age'].quantile(0.25)
```

```
    lower_range = df['Age'].quantile(0.25) - (1.5 * IQR)
```

```
    upper_range = df['Age'].quantile(0.75) + (1.5 * IQR)
```

```
    df.loc[df['Age'] <= lower_range, 'Age'] = lower_range
```

```
    df.loc[df['Age'] >= upper_range, 'Age'] = upper_range
```

```
outlier_age(df)
```

```
sns.boxplot(df.Age)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4aa2beaf50>
```

```
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  float64
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(3), int64(8), object(3)
memory usage: 1.1+ MB
```

```
df.head(2)
```

|   | RowNumber | CustomerId | Surname  | CreditScore | Geography | Gender | Age | Tenure |
|---|-----------|------------|----------|-------------|-----------|--------|-----|--------|
| 0 | 1         | 15634602   | Hargrave | 619         | France    | Female | 42  |        |
| 1 | 2         | 15647311   | Hill     | 608         | Spain     | Female | 41  |        |

```
df.drop(['CustomerId', 'RowNumber', 'Surname'], axis=1, inplace=True)
```

```
df.head(2)
```

|   | CreditScore | Geography | Gender | Age | Tenure | Balance  | NumOfProducts | HasCrCard |
|---|-------------|-----------|--------|-----|--------|----------|---------------|-----------|
| 0 | 619         | France    | Female | 42  | 2      | 0.00     | 1.0           |           |
| 1 | 608         | Spain     | Female | 41  | 1      | 83807.86 | 1.0           |           |

```
from sklearn.preprocessing import LabelEncoder
le_geo = LabelEncoder()
```

```
le_gen = LabelEncoder()
df['Sex']=le_gen.fit_transform(df.Gender)
df['Country']=le_geo.fit_transform(df.Geography)
df.drop(['Geography', 'Gender'],axis=1,inplace=True)
```

```
df.head(2)
```

|   | CreditScore | Age | Tenure | Balance  | NumOfProducts | HasCrCard | IsActiveMemb |
|---|-------------|-----|--------|----------|---------------|-----------|--------------|
| 0 | 619         | 42  | 2      | 0.00     | 1.0           | 1         |              |
| 1 | 608         | 41  | 1      | 83807.86 | 1.0           | 0         |              |

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

```
from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
X = sc.fit_transform(X)
```

```
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=42)
```

```
x_train.shape, x_test.shape, y_train.shape, y_test.shape

((8000, 10), (2000, 10), (8000,), (2000,))
```

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