

Load dataset and importing required library

```
import numpy as np
import pandas as pd
```

```
df = pd.read_csv('/content/sample_data/Churn_Modelling.csv')
```

```
df.head()
```

| | 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 |

| | 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 | |

| | EstimatedSalary | Exited |
|---|-----------------|--------|
| 0 | 101348.88 | 1 |
| 1 | 112542.58 | 0 |
| 2 | 113931.57 | 1 |
| 3 | 93826.63 | 0 |
| 4 | 79084.10 | 0 |

```
df.shape
```

```
(10000, 14)
```

Statistical analysis

```
df.info
```

```
<bound method DataFrame.info of
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
...      ...      ...      ...      ...      ...      ...
...
9995      9996      15606229      Obijiaku      771      France      Male
39
9996      9997      15569892      Johnstone      516      France      Male
35
9997      9998      15584532      Liu          709      France      Female
36
9998      9999      15682355      Sabbatini      772      Germany      Male
42
9999      10000      15628319      Walker      792      France      Female
28

```

```

      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
...      ...      ...      ...      ...      ...
9995      5          0.00          2          1          0
9996     10      57369.61          1          1          1
9997      7          0.00          1          0          1
9998      3      75075.31          2          1          0
9999      4     130142.79          1          1          0

```

```

      EstimatedSalary      Exited
0          101348.88          1
1          112542.58          0
2          113931.57          1
3           93826.63          0
4           79084.10          0
...      ...      ...
9995          96270.64          0
9996         101699.77          0
9997          42085.58          1
9998          92888.52          1
9999          38190.78          0

```

```
[10000 rows x 14 columns]>
```

```
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
```

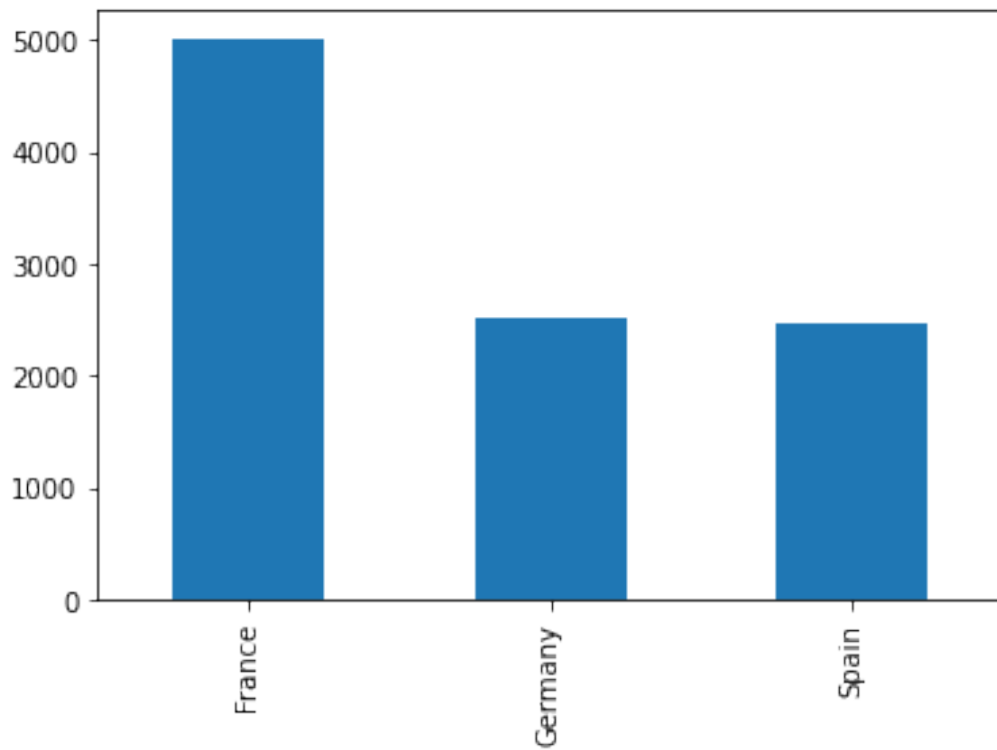
```
df.drop(['RowNumber', 'CustomerId', 'Surname'], axis=1, inplace=True)
```

Data Visualization

```
import matplotlib.pyplot as plt
import seaborn as sns
```

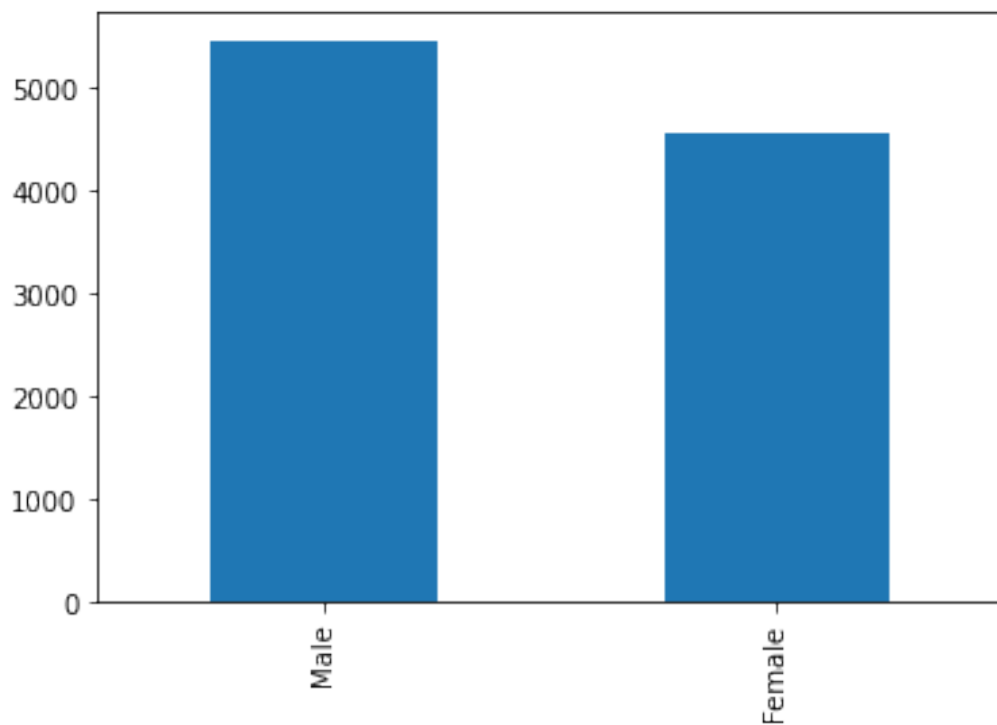
```
df.Geography.value_counts().plot(kind='bar')
df.Geography.value_counts()
```

```
France      5014
Germany     2509
Spain       2477
Name: Geography, dtype: int64
```



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

```
Male      5457  
Female    4543  
Name: Gender, dtype: int64
```

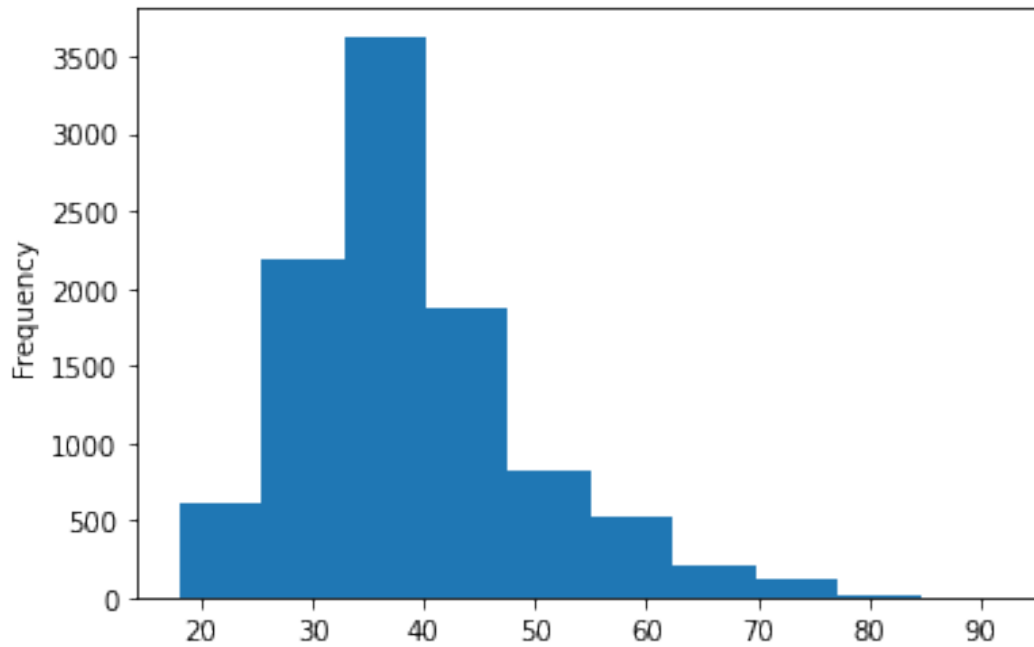


```
df.Age.describe()
```

```
count    10000.000000
mean      38.921800
std       10.487806
min       18.000000
25%       32.000000
50%       37.000000
75%       44.000000
max       92.000000
Name: Age, dtype: float64
```

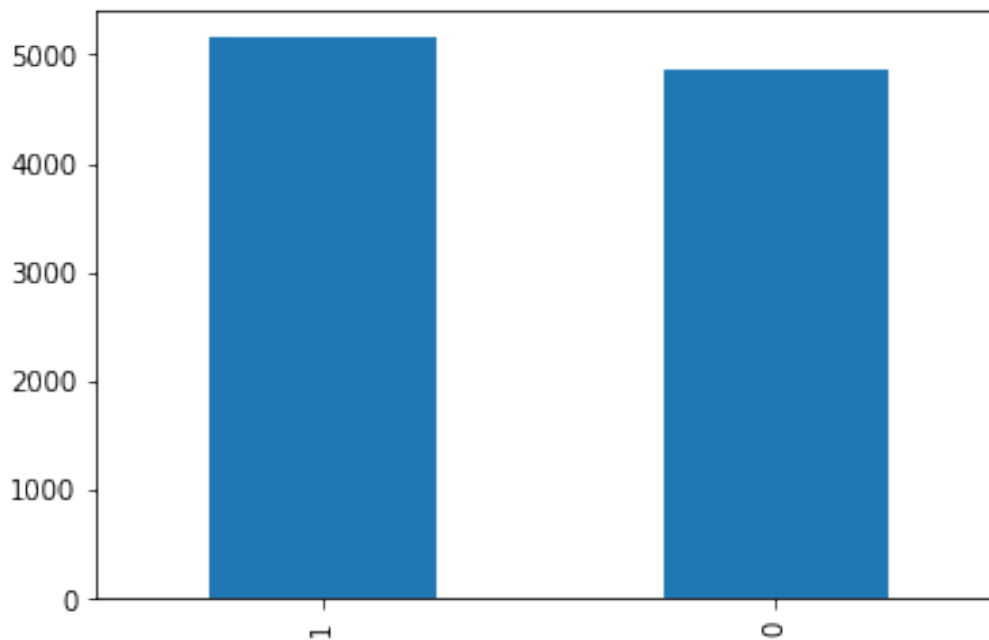
```
df.Age.plot(kind='hist')
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f8482367e90>
```



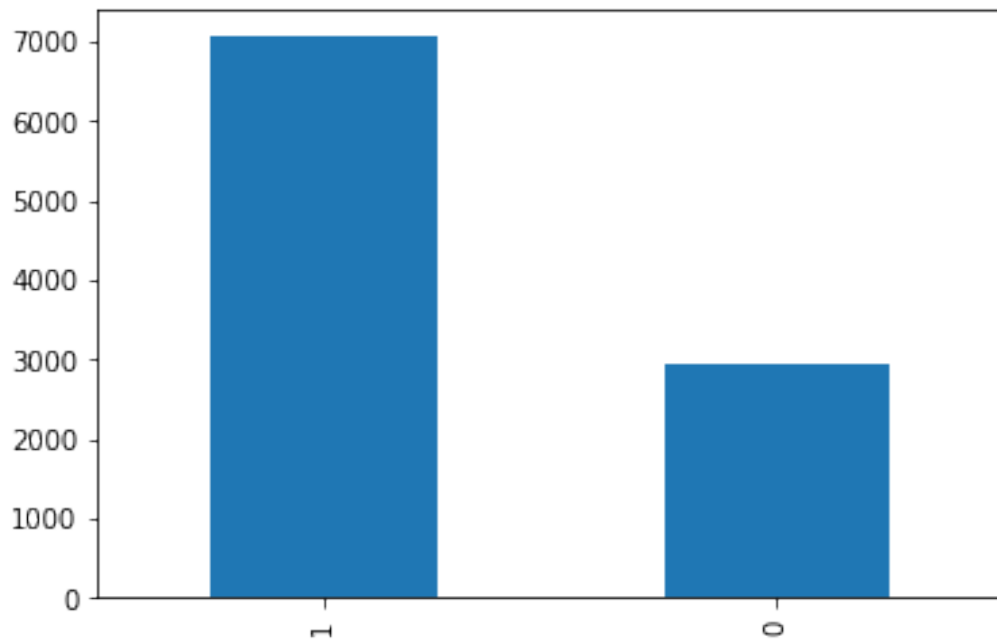
```
df.IsActiveMember.value_counts().plot(kind='bar')
df.IsActiveMember.value_counts()
```

```
1    5151
0    4849
Name: IsActiveMember, dtype: int64
```



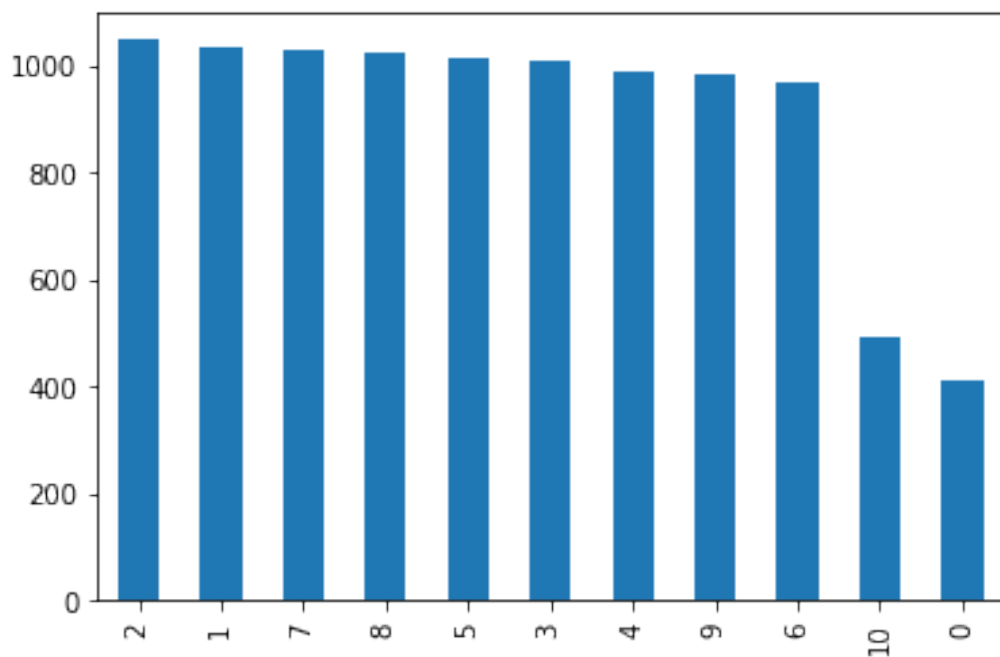
```
df.HasCrCard.value_counts().plot(kind='bar')
df.HasCrCard.value_counts()
```

```
1    7055
0    2945
Name: HasCrCard, dtype: int64
```



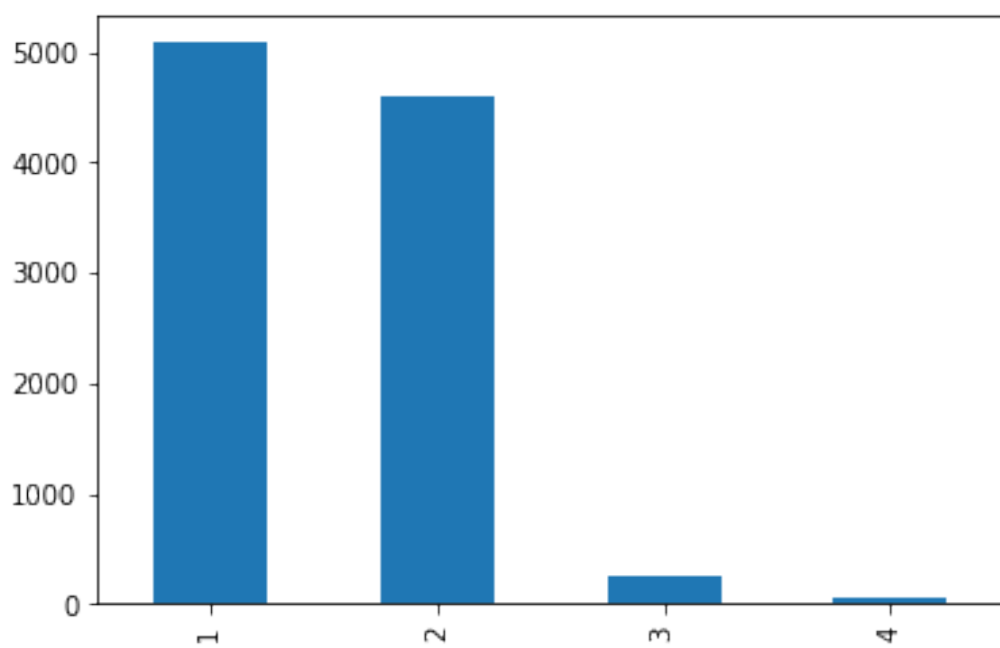
```
df.Tenure.value_counts().plot(kind='bar');
df.Tenure.value_counts()
```

```
2    1048
1    1035
7    1028
8    1025
5    1012
3    1009
4     989
9     984
6     967
10    490
0     413
Name: Tenure, dtype: int64
```



```
df.NumOfProducts.value_counts().plot(kind='bar');
df.NumOfProducts.value_counts()
```

```
1    5084
2    4590
3     266
4      60
Name: NumOfProducts, dtype: int64
```

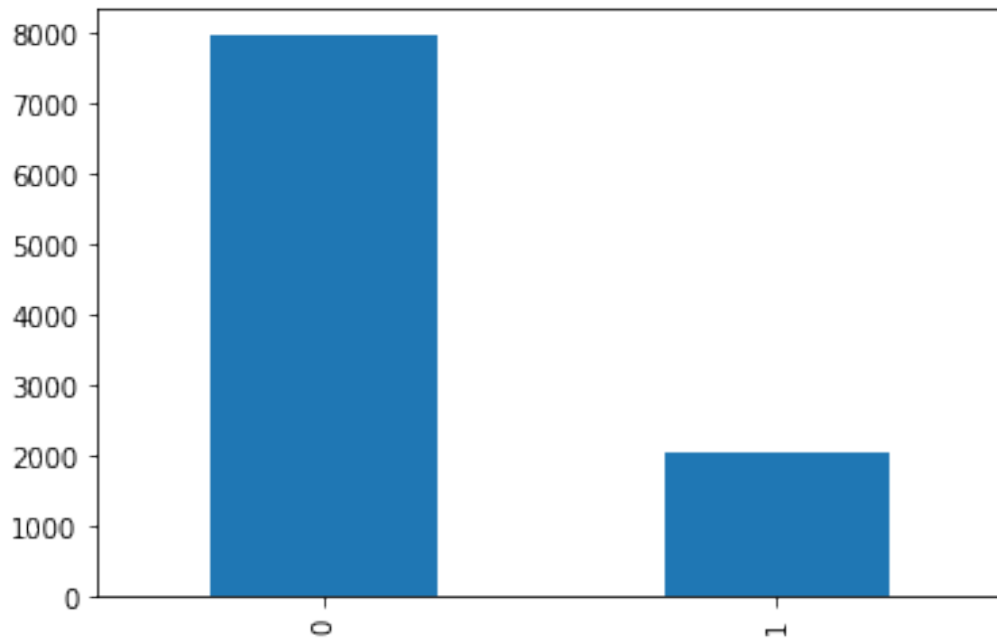



```
df.Exited.value_counts().plot(kind='bar');  
df.Exited.value_counts()
```

```
0    7963
```

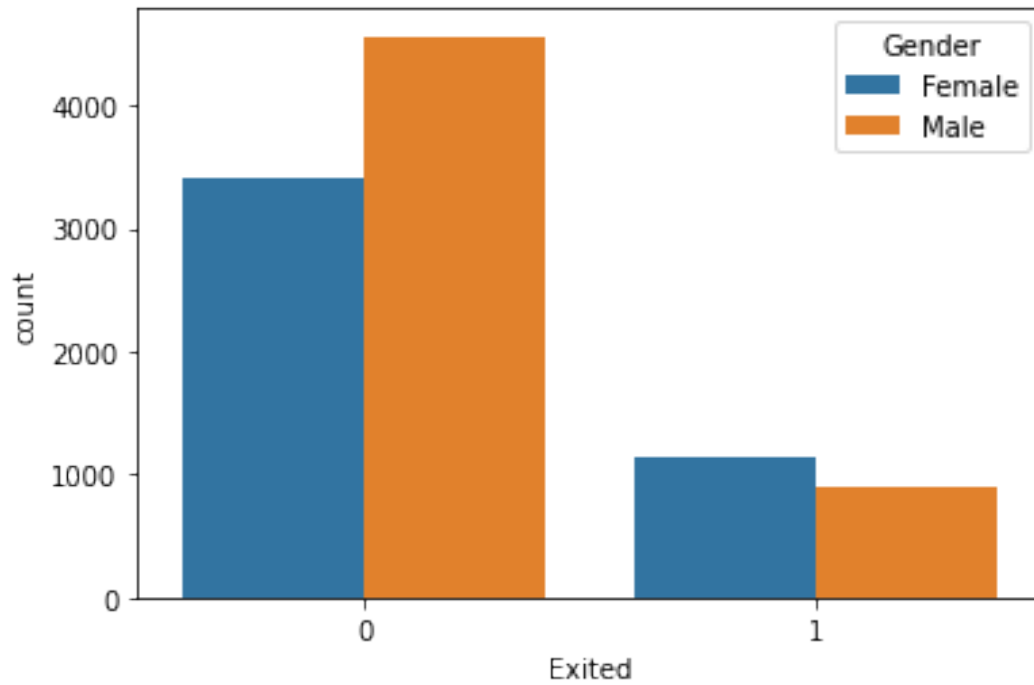
```
1    2037
```

```
Name: Exited, dtype: int64
```



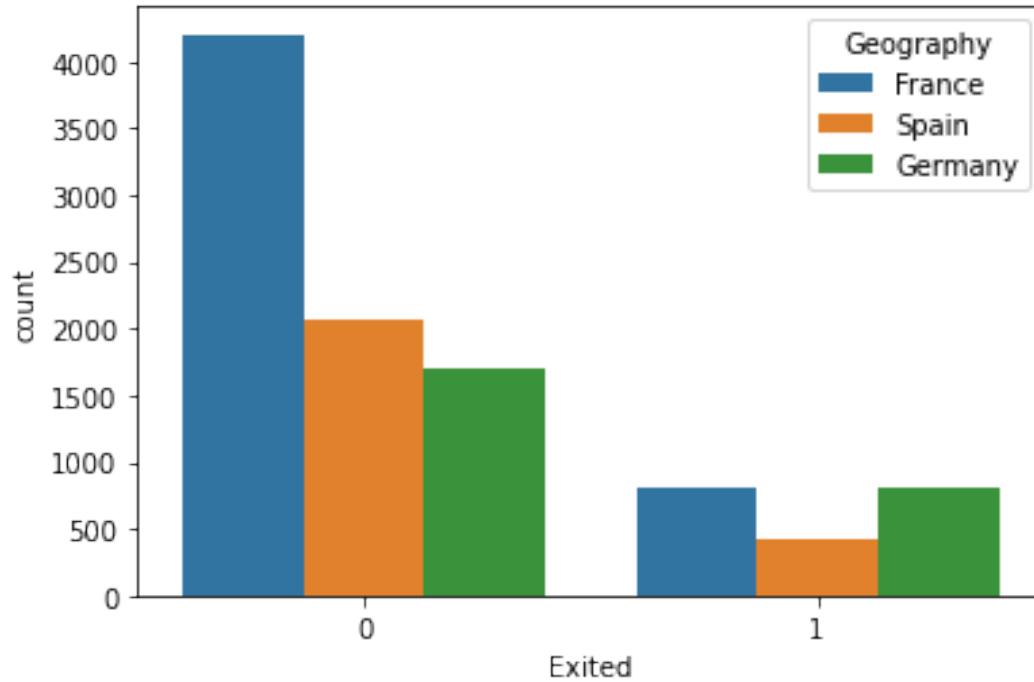
```
sns.countplot(x=df.Exited,hue=df.Gender)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f8481bb1410>
```



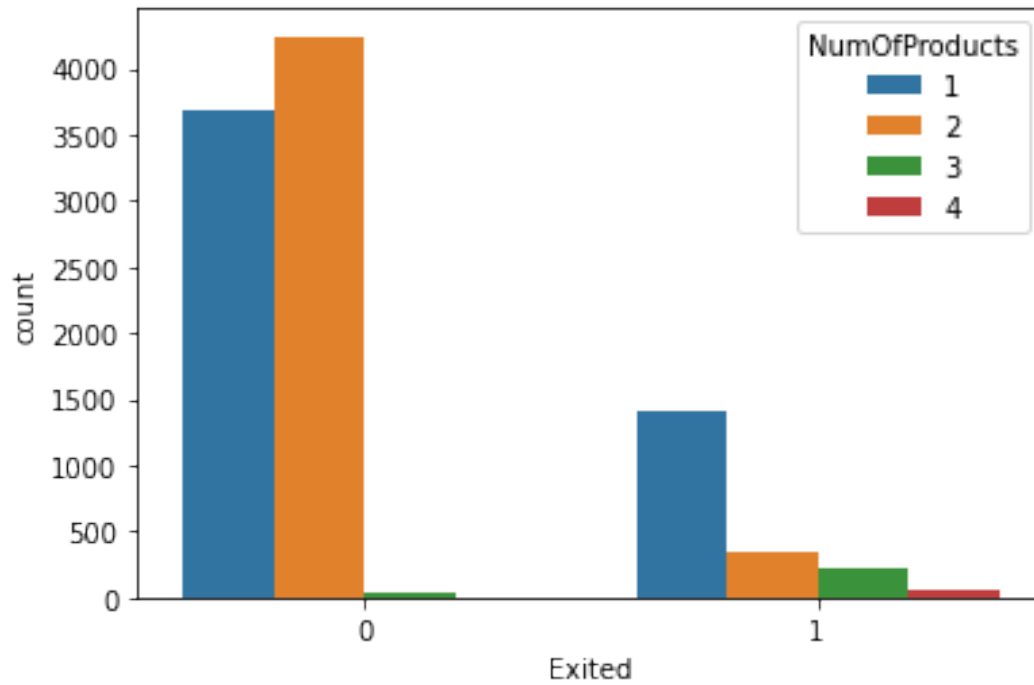
```
sns.countplot(x=df.Exited,hue=df.Geography)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f8481b387d0>
```



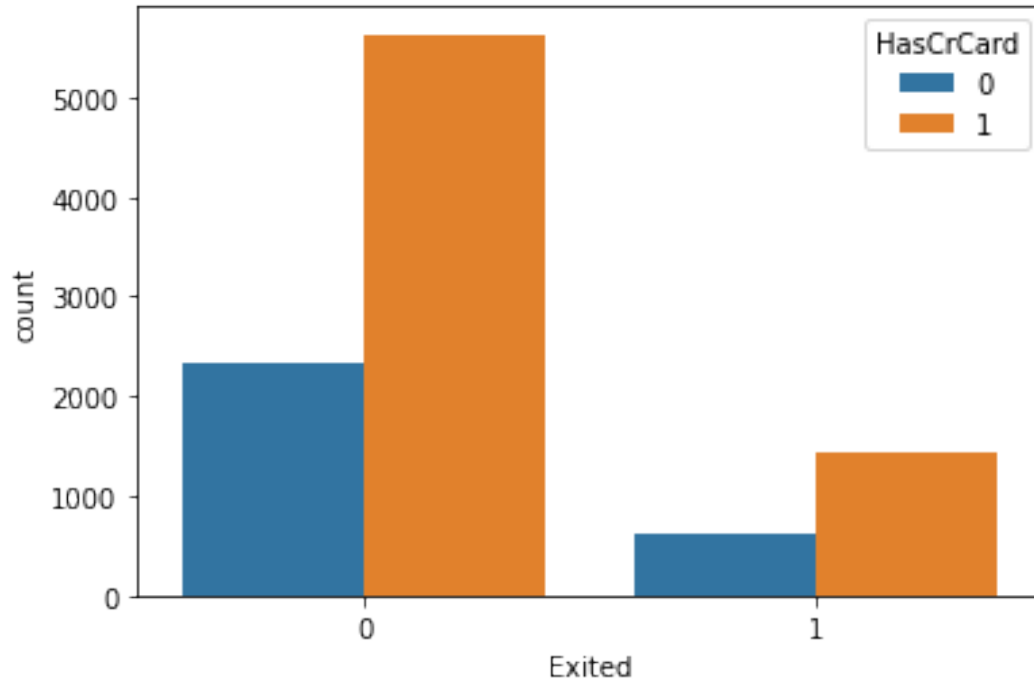
```
sns.countplot(x=df.Exited,hue=df.NumOfProducts)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f8481ab1e50>
```



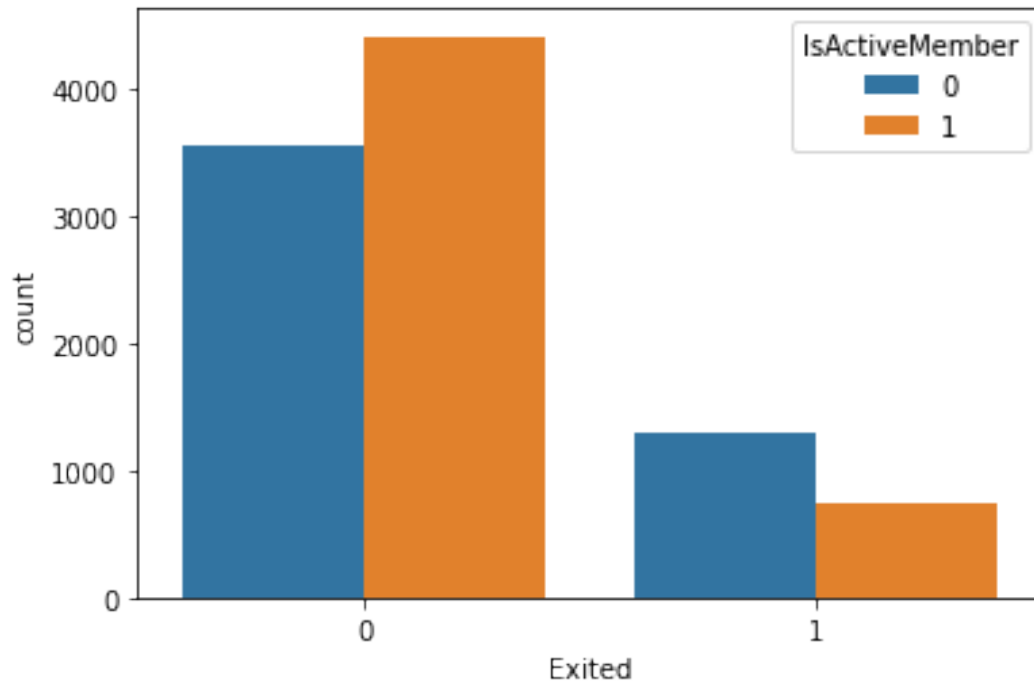
```
sns.countplot(x=df.Exited,hue=df.HasCrCard)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f8481a43a50>
```



```
sns.countplot(x=df.Exited,hue=df.IsActiveMember)
```

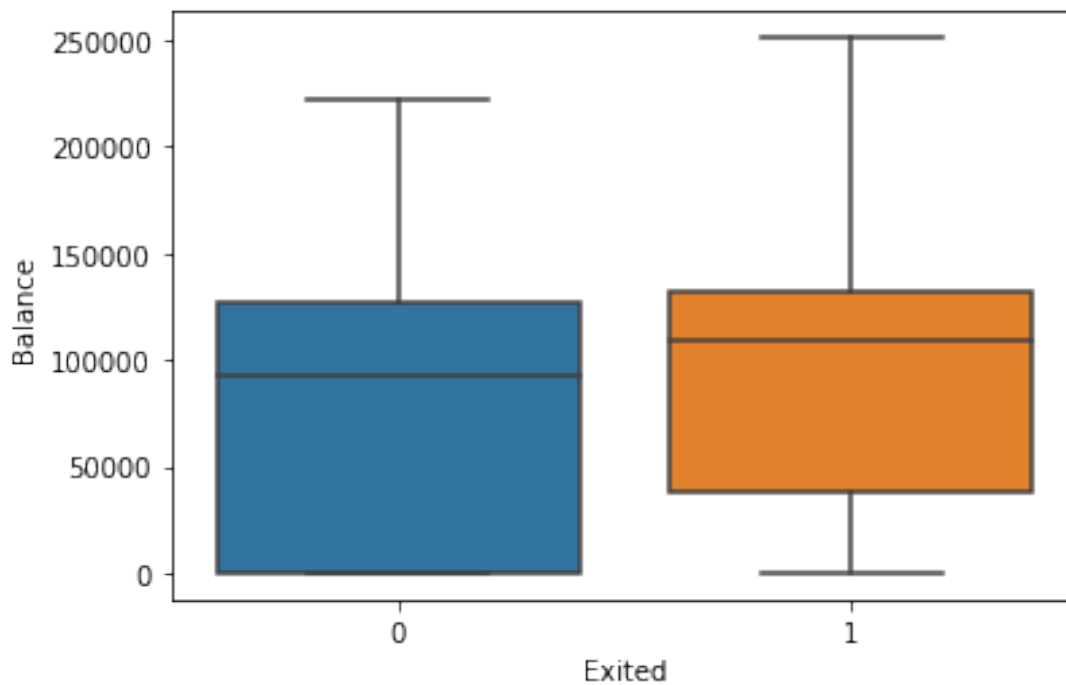
```
<matplotlib.axes._subplots.AxesSubplot at 0x7f84819ade90>
```



Finding Outliers

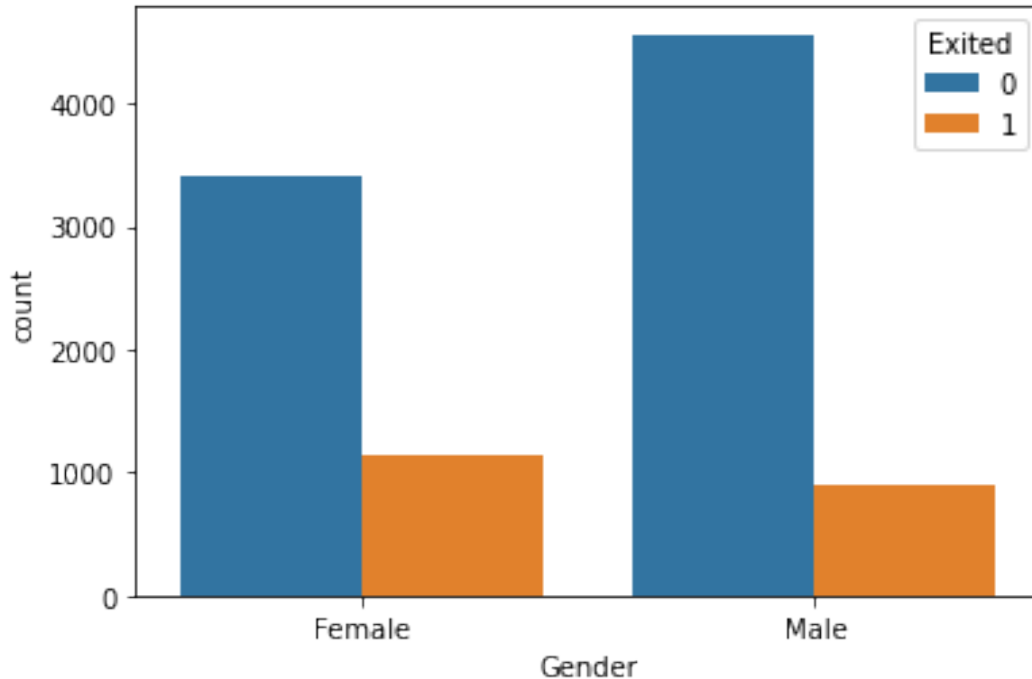
```
sns.boxplot(x=df.Exited,y=df.Balance)
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f8481a2fd90>



```
sns.countplot(x="Gender",hue="Exited",data=df)
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f8481c00050>



categorizing with LabelEncoding

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

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

```
X.shape
```

```
(10000, 10)
```

Feature Scaling

```
from sklearn.preprocessing import LabelEncoder  
from sklearn.preprocessing import MinMaxScaler
```

```
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 113931.57]  
 ...  
 [709 0 0 ... 0 1 42085.58]
```

```
[772 2 1 ... 1 0 92888.52]  
[792 0 0 ... 1 0 38190.78]]
```

```
MnScaler = MinMaxScaler()
```

```
X = MnScaler.fit_transform(X)
```

Train Test Split

```
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)
```

```
from sklearn.preprocessing import StandardScaler
```

```
stdscaler = StandardScaler()
```

```
X_train = stdscaler.fit_transform(X_train)
```

```
X_test = stdscaler.transform(X_test)
```