

## Assignment 2

|                                |                         |
|--------------------------------|-------------------------|
| <b>Assignment Date</b>         | <b>07 November 2022</b> |
| <b>Student Name</b>            | <b>Santhosh S</b>       |
| <b>Student Register Number</b> | <b>620619106027</b>     |
| <b>Maximum Marks</b>           | <b>2</b>                |

### 1.Importing package

```
import pandas as pd
import seaborn as sns
import numpy as np
from matplotlib import pyplot as plt
%matplotlib inline
```

### 2.Loading dataset

```
df = pd.read_csv("/content/drive/MyDrive/Assignment
3/Churn_Modelling.csv")
```

```
df
```

```
   Age  \
0      1  RowNumber  CustomerId  Surname  CreditScore  Geography  Gender
42      1          1    15634602  Hargrave         619      France  Female
41      2          2    15647311    Hill         608        Spain  Female
42      3          3    15619304    Onio         502      France  Female
39      4          4    15701354    Boni         699      France  Female
```

|      |       |          |           |     |         |        |
|------|-------|----------|-----------|-----|---------|--------|
| 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      |

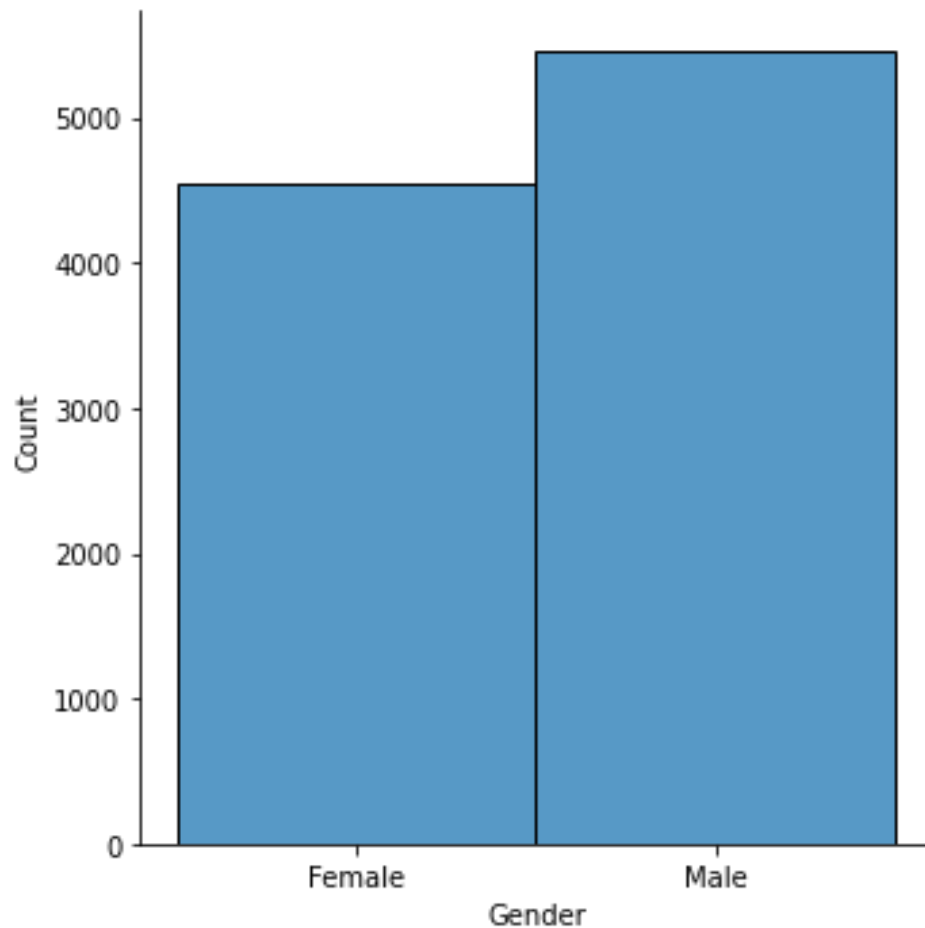
[10000 rows x 14 columns]

### 3. Visualizations

#### a) Univariate Analysis

```
sns.displot(df.Gender)
```

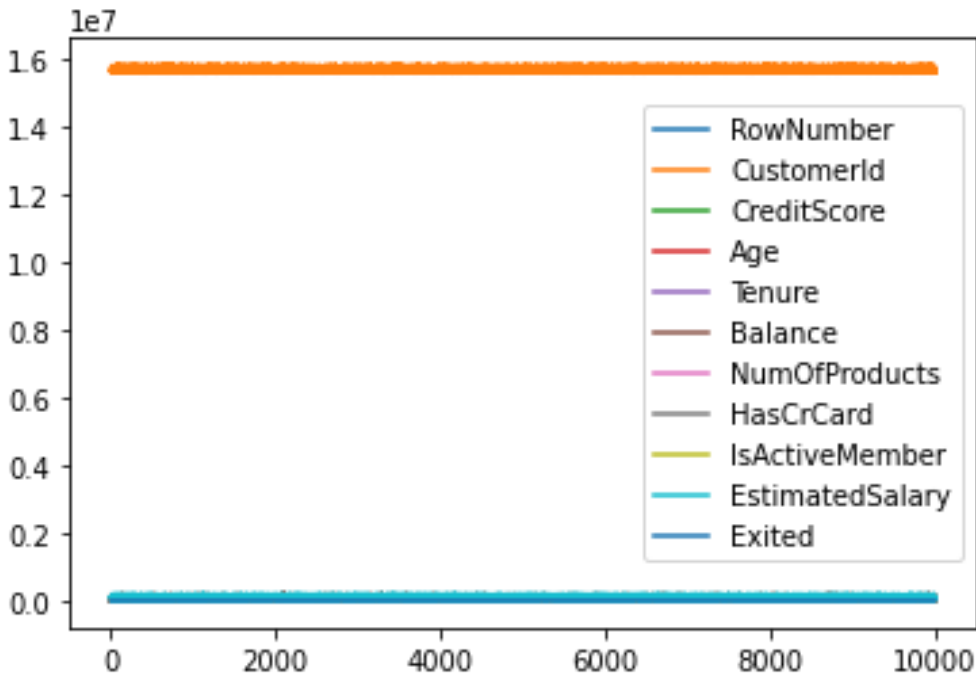
```
<seaborn.axisgrid.FacetGrid at 0x7f3326917d50>
```



#### b) Bi-Variate Analysis

```
df.plot.line()
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f3323aa45d0>
```

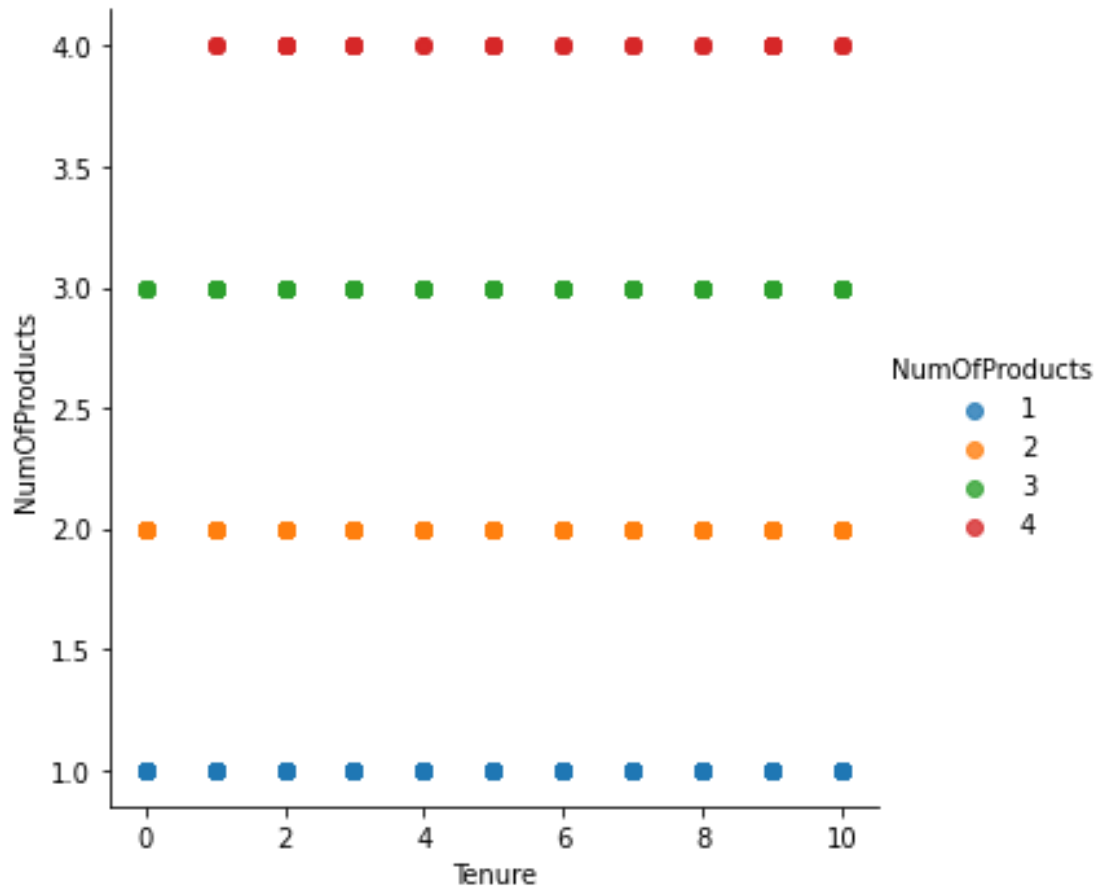


### c) Multi-Variate Analysis

```
sns.lmplot("Tenure", "NumOfProducts", df, hue="NumOfProducts",  
fit_reg=False);
```

/usr/local/lib/python3.7/dist-packages/seaborn/\_decorators.py:43:  
FutureWarning: Pass the following variables as keyword args: x, y,  
data. From version 0.12, the only valid positional argument will be  
`data`, and passing other arguments without an explicit keyword will  
result in an error or misinterpretation.

FutureWarning



#### 4. Perform descriptive statistic on the dataset

df.describe()

|          | RowNumber    | CustomerId   | CreditScore  | Age          |
|----------|--------------|--------------|--------------|--------------|
| Tenure \ |              |              |              |              |
| count    | 10000.000000 | 1.000000e+04 | 10000.000000 | 10000.000000 |
| mean     | 5000.50000   | 1.569094e+07 | 650.528800   | 38.921800    |
| std      | 2886.89568   | 7.193619e+04 | 96.653299    | 10.487806    |
| min      | 1.00000      | 1.556570e+07 | 350.000000   | 18.000000    |
| 25%      | 2500.75000   | 1.562853e+07 | 584.000000   | 32.000000    |
| 50%      | 5000.50000   | 1.569074e+07 | 652.000000   | 37.000000    |
| 75%      | 7500.25000   | 1.575323e+07 | 718.000000   | 44.000000    |
| max      | 10000.00000  | 1.581569e+07 | 850.000000   | 92.000000    |

10.000000

|       | Balance       | NumOfProducts | HasCrCard    | IsActiveMember | \ |
|-------|---------------|---------------|--------------|----------------|---|
| count | 10000.000000  | 10000.000000  | 10000.000000 | 10000.000000   |   |
| mean  | 76485.889288  | 1.530200      | 0.70550      | 0.515100       |   |
| std   | 62397.405202  | 0.581654      | 0.45584      | 0.499797       |   |
| min   | 0.000000      | 1.000000      | 0.00000      | 0.000000       |   |
| 25%   | 0.000000      | 1.000000      | 0.00000      | 0.000000       |   |
| 50%   | 97198.540000  | 1.000000      | 1.00000      | 1.000000       |   |
| 75%   | 127644.240000 | 2.000000      | 1.00000      | 1.000000       |   |
| max   | 250898.090000 | 4.000000      | 1.00000      | 1.000000       |   |

|       | EstimatedSalary | Exited       |
|-------|-----------------|--------------|
| count | 10000.000000    | 10000.000000 |
| mean  | 100090.239881   | 0.203700     |
| std   | 57510.492818    | 0.402769     |
| min   | 11.580000       | 0.000000     |
| 25%   | 51002.110000    | 0.000000     |
| 50%   | 100193.915000   | 0.000000     |
| 75%   | 149388.247500   | 0.000000     |
| max   | 199992.480000   | 1.000000     |

## 5. Handle the Missing values

```
data = pd.read_csv("/content/drive/MyDrive/Assignment  
3/Churn_Modelling.csv")
```

```
pd.isnull(data["Gender"])
```

```
0      False
1      False
2      False
3      False
4      False
...
9995   False
9996   False
9997   False
9998   False
9999   False
```

```
Name: Gender, Length: 10000, dtype: bool
```

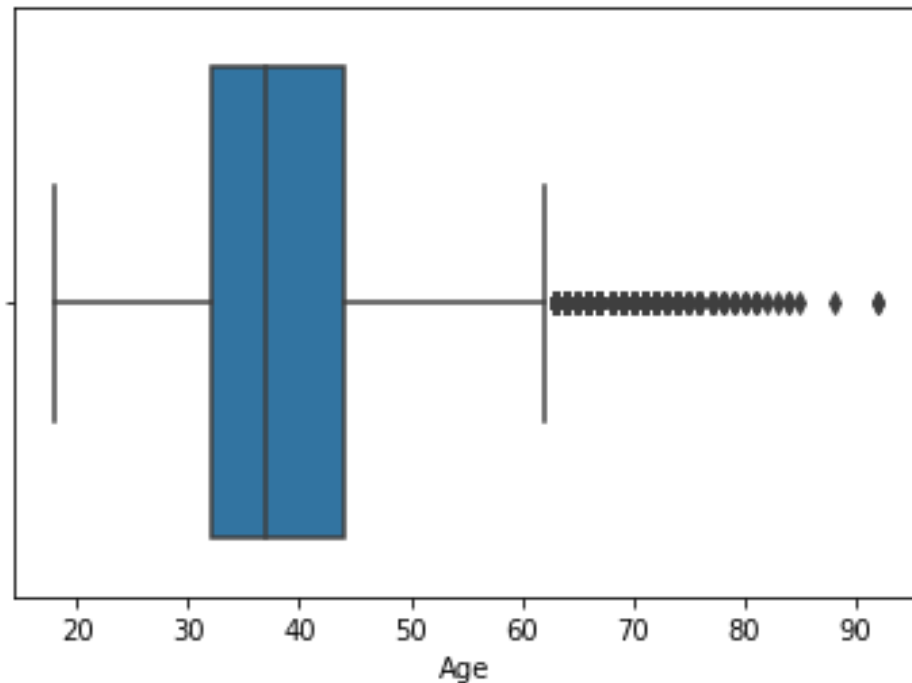
## 6. Find the outliers and replace the outliers

```
sns.boxplot(df['Age'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43:
FutureWarning: Pass the following variable as a keyword arg: x. From
version 0.12, the only valid positional argument will be `data`, and
passing other arguments without an explicit keyword will result in an
error or misinterpretation.
```

```
FutureWarning
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f332390fc50>
```



```
df['Age']=np.where(df['Age']>50,40,df['Age'])
```

```
df['Age']
```

```
0      42
```

```
1      41
```

```
2      42
```

```
3      39
```

```
4      43
```

```
..
```

```
9995   39
```

```
9996   35
```

```
9997   36
```

```
9998   42
```

```
9999   28
```

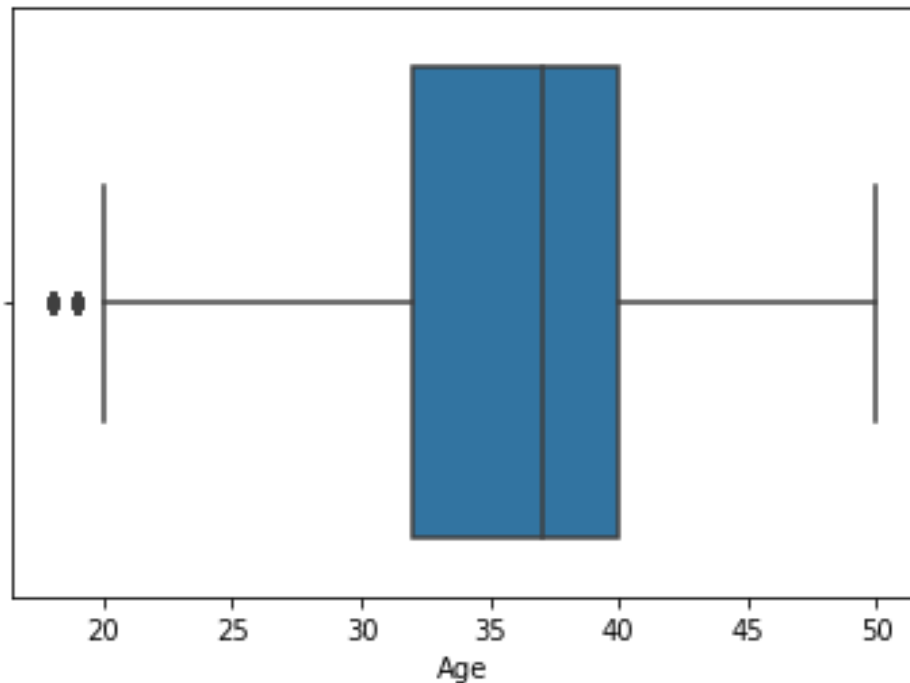
```
Name: Age, Length: 10000, dtype: int64
```

```
sns.boxplot(df['Age'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43:
FutureWarning: Pass the following variable as a keyword arg: x. From
version 0.12, the only valid positional argument will be `data`, and
passing other arguments without an explicit keyword will result in an
error or misinterpretation.
```

FutureWarning

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f332387b390>
```



```
df['Age']=np.where(df['Age']<20,35,df['Age'])
```

```
df['Age']
```

```
0      42
```

```
1      41
```

```
2      42
```

```
3      39
```

```
4      43
```

```
...
```

```
9995    39
```

```
9996    35
```

```
9997    36
```

```
9998    42
```

```
9999    28
```

```
Name: Age, Length: 10000, dtype: int64
```



## 7. Check for Categorical columns and perform encoding

```
pd.get_dummies(df, columns=["Gender", "Age"], prefix=["Age",  
"Gender"]).head()
```

|           | RowNumber | CustomerId | Surname  | CreditScore | Geography | Tenure |
|-----------|-----------|------------|----------|-------------|-----------|--------|
| Balance \ |           |            |          |             |           |        |
| 0         | 1         | 15634602   | Hargrave | 619         | France    | 2      |
| 0.00      |           |            |          |             |           |        |
| 1         | 2         | 15647311   | Hill     | 608         | Spain     | 1      |
| 83807.86  |           |            |          |             |           |        |
| 2         | 3         | 15619304   | Onio     | 502         | France    | 8      |
| 159660.80 |           |            |          |             |           |        |
| 3         | 4         | 15701354   | Boni     | 699         | France    | 1      |
| 0.00      |           |            |          |             |           |        |
| 4         | 5         | 15737888   | Mitchell | 850         | Spain     | 2      |
| 125510.82 |           |            |          |             |           |        |

|   | NumOfProducts | HasCrCard | IsActiveMember | ... | Gender_41 | Gender_42 |
|---|---------------|-----------|----------------|-----|-----------|-----------|
| \ |               |           |                |     |           |           |
| 0 | 1             | 1         | 1              | ... | 0         | 1         |
| 1 | 1             | 0         | 1              | ... | 1         | 0         |
| 2 | 3             | 1         | 0              | ... | 0         | 1         |
| 3 | 2             | 0         | 0              | ... | 0         | 0         |
| 4 | 1             | 1         | 1              | ... | 0         | 0         |

|   | Gender_43 | Gender_44 | Gender_45 | Gender_46 | Gender_47 | Gender_48 | \ |
|---|-----------|-----------|-----------|-----------|-----------|-----------|---|
| 0 | 0         | 0         | 0         | 0         | 0         | 0         |   |
| 1 | 0         | 0         | 0         | 0         | 0         | 0         |   |
| 2 | 0         | 0         | 0         | 0         | 0         | 0         |   |
| 3 | 0         | 0         | 0         | 0         | 0         | 0         |   |
| 4 | 1         | 0         | 0         | 0         | 0         | 0         |   |

|   | Gender_49 | Gender_50 |
|---|-----------|-----------|
| 0 | 0         | 0         |
| 1 | 0         | 0         |
| 2 | 0         | 0         |
| 3 | 0         | 0         |
| 4 | 0         | 0         |

[5 rows x 45 columns]

## 8.Split the data into dependent and independent variables

### a) Split the data into Independent variables.

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

print(X)

[[1 15634602 'Hargrave' ... 1 1 101348.88]
 [2 15647311 'Hill' ... 0 1 112542.58]
 [3 15619304 'Onio' ... 1 0 113931.57]
 ...
 [9998 15584532 'Liu' ... 0 1 42085.58]
 [9999 15682355 'Sabbatini' ... 1 0 92888.52]
 [10000 15628319 'Walker' ... 1 0 38190.78]]
```

### b) Split the data into Dependent variables.

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

print(Y)

[1 0 1 ... 1 1 0]
```

## 9. Scale the independent variables

```
import pandas as pd

from sklearn.preprocessing import MinMaxScaler

scaler = MinMaxScaler()

df[["CustomerId"]] = scaler.fit_transform(df[["CustomerId"]])

print(df)
```

|       | RowNumber | CustomerId | Surname  | CreditScore | Geography | Gender |
|-------|-----------|------------|----------|-------------|-----------|--------|
| Age \ |           |            |          |             |           |        |
| 0     | 1         | 0.275616   | Hargrave | 619         | France    | Female |
| 42    |           |            |          |             |           |        |
| 1     | 2         | 0.326454   | Hill     | 608         | Spain     | Female |
| 41    |           |            |          |             |           |        |
| 2     | 3         | 0.214421   | Onio     | 502         | France    | Female |
| 42    |           |            |          |             |           |        |
| 3     | 4         | 0.542636   | Boni     | 699         | France    | Female |
| 39    |           |            |          |             |           |        |
| 4     | 5         | 0.688778   | Mitchell | 850         | Spain     | Female |
| 43    |           |            |          |             |           |        |
| ...   | ...       | ...        | ...      | ...         | ...       | ...    |
| ...   |           |            |          |             |           |        |

|            |       |          |           |     |         |        |
|------------|-------|----------|-----------|-----|---------|--------|
| 9995<br>39 | 9996  | 0.162119 | Obijiaku  | 771 | France  | Male   |
| 9996<br>35 | 9997  | 0.016765 | Johnstone | 516 | France  | Male   |
| 9997<br>36 | 9998  | 0.075327 | Liu       | 709 | France  | Female |
| 9998<br>42 | 9999  | 0.466637 | Sabbatini | 772 | Germany | Male   |
| 9999<br>28 | 10000 | 0.250483 | Walker    | 792 | France  | Female |

|      | 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]

## 10. Split the data into training and testing

```
from sklearn.model_selection import train_test_split
```

```
train_size=0.8
```

```
X = df.drop(columns = ['Tenure']).copy()
```

```
y = df['Tenure']  
X_train, X_rem, y_train, y_rem = train_test_split(X,y, train_size=0.8)  
test_size = 0.5  
X_valid, X_test, y_valid, y_test = train_test_split(X_rem,y_rem,  
test_size=0.5)  
print(X_train.shape), print(y_train.shape)  
(8000, 13)  
(8000,)  
(None, None)  
print(X_valid.shape), print(y_valid.shape)  
(1000, 13)  
(1000,)  
(None, None)  
print(X_test.shape), print(y_test.shape)  
(1000, 13)  
(1000,)  
(None, None)
```