

```
In [1]: import numpy as np
```

```
In [2]: import pandas as pd
```

```
In [3]: df=pd.read_csv("Churn_Modelling.csv")
```

```
In [4]: df
```

Out[4]:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
	0	1	15634602	Hargrave	France	Female	42	2	0.00	1	1	1	101348.88	1
	1	2	15647311	Hill	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
	2	3	15619304	Onio	France	Female	42	8	159660.80	3	1	0	113931.57	1
	3	4	15701354	Boni	France	Female	39	1	0.00	2	0	0	93826.63	0
	4	5	15737888	Mitchell	Spain	Female	43	2	125510.82	1	1	1	79084.10	0

	9995	9996	15606229	Obijaku	France	Male	39	5	0.00	2	1	0	96270.64	0
	9996	9997	15569892	Johnstone	France	Male	35	10	57369.61	1	1	1	101699.77	0
	9997	9998	15584532	Liu	France	Female	36	7	0.00	1	0	1	42085.58	1
	9998	9999	15682355	Sabbatini	Germany	Male	42	3	75075.31	2	1	0	92888.52	1
	9999	10000	15628319	Walker	France	Female	28	4	130142.79	1	1	0	38190.78	0

10000 rows × 14 columns

```
In [5]: df.shape
```

Out[5]: (10000, 14)

```
In [6]: df.columns
```

Out[6]: Index(['RowNumber', 'CustomerId', 'Surname', 'CreditScore', 'Geography', 'Gender', 'Age', 'Tenure', 'Balance', 'NumOfProducts', 'HasCrCard', 'IsActiveMember', 'EstimatedSalary', 'Exited'], dtype='object')

```
In [7]: df["NumOfProducts"].unique()
```

Out[7]: array([1, 3, 2, 4], dtype=int64)

```
In [8]: df["NumOfProducts"].value_counts()
```

Out[8]:

1	5084
2	4590
3	266
4	60

Name: NumOfProducts, dtype: int64

```
In [9]: df.dtypes
```

Out[9]:

RowNumber	int64
CustomerId	int64
Surname	object
CreditScore	int64
Geography	object
Gender	object
Age	int64
Tenure	int64
Balance	float64
NumOfProducts	int64
HasCrCard	int64
IsActiveMember	int64
EstimatedSalary	float64
Exited	int64
dtype:	object

```
In [10]: df.head()
```

```
Out[10]:
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
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	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	1	1	1	79084.10	0

```
In [11]: df.describe()
```

```
Out[11]:
```

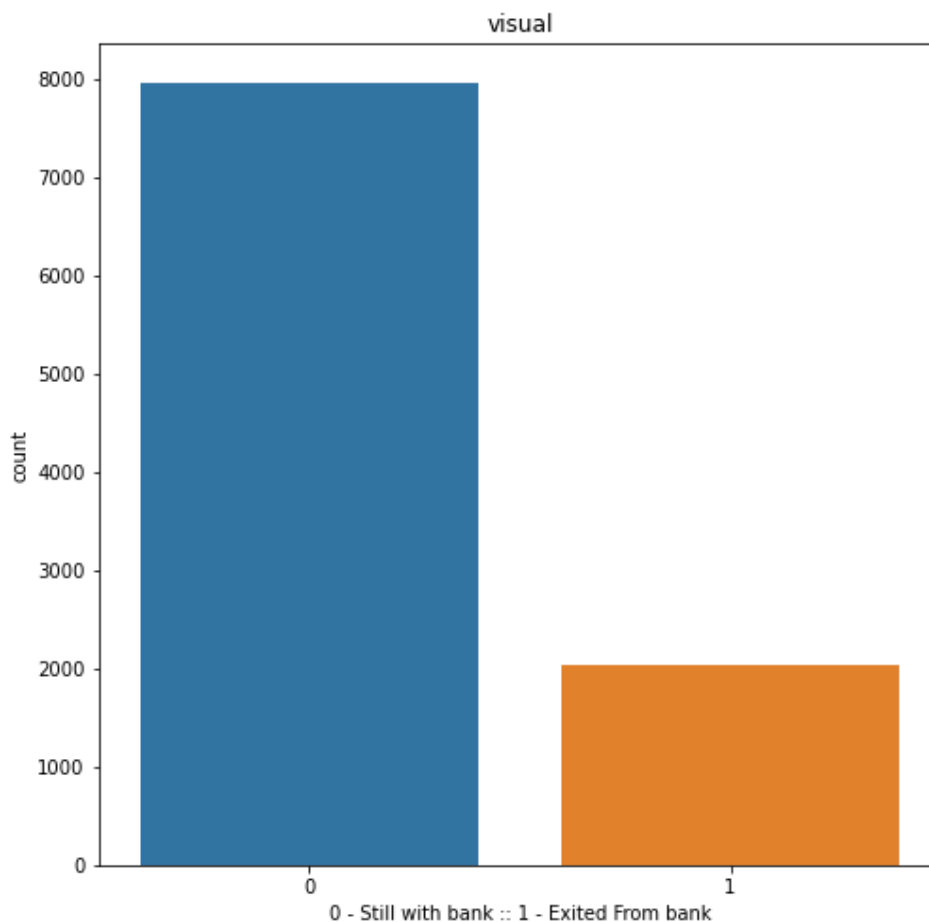
	RowNumber	CustomerId	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
count	10000.00000	1.000000e+04	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000
mean	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	76485.889288	1.530200	0.70550	0.515100	100090.239881	0.203700
std	2886.89568	7.193619e+04	96.653299	10.487806	2.892174	62397.405202	0.581654	0.45584	0.499797	57510.492818	0.402769
min	1.00000	1.556570e+07	350.000000	18.000000	0.000000	0.000000	1.000000	0.00000	0.000000	11.580000	0.000000
25%	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	0.000000	1.000000	0.00000	0.000000	51002.110000	0.000000
50%	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	97198.540000	1.000000	1.00000	1.000000	100193.915000	0.000000
75%	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	127644.240000	2.000000	1.00000	1.000000	149388.247500	0.000000
max	10000.00000	1.581569e+07	850.000000	92.000000	10.000000	250898.090000	4.000000	1.00000	1.000000	199992.480000	1.000000

```
In [12]: import matplotlib.pyplot as plt
```

```
In [13]: import seaborn as sns
```

```
In [14]: %matplotlib inline
```

```
In [15]: plt.figure(figsize=(8,8))
sns.countplot(x='Exited',data=df)
plt.xlabel("0 - Still with bank :: 1 - Exited From bank")
plt.ylabel("count")
plt.title("visual")
plt.show()
```



In [16]:

```
df.info()
```

```
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
```

In [17]:

```
df.isna().any()
```

Out[17]:

```
RowNumber      False
CustomerId     False
Surname         False
CreditScore     False
Geography       False
Gender          False
Age             False
Tenure          False
Balance         False
NumOfProducts  False
HasCrCard       False
IsActiveMember  False
EstimatedSalary False
Exited         False
dtype: bool
```

In [18]:

```
df.isnull().sum()
```

Out[18]:

```
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
```

In [19]:

```
df1=df.copy()
```

In [20]:

```
df1.shape
```

Out[20]:

```
(10000, 14)
```

In [21]:

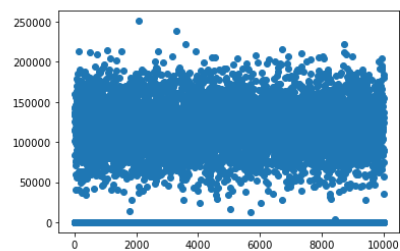
```
updated_df=df.dropna(axis=1)
updated_df.info()
```

```
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
```

```
In [22]: updated_df['Balance']=updated_df['Balance'].fillna(updated_df['Balance'].mean())
updated_df.info()
```

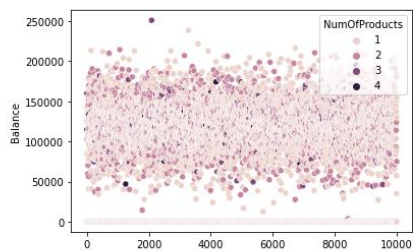
```
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 14 columns):
#   Column      Non-Null Count  Dtype
---  -
0   RowNumber    10000 non-null  int64
1   CustomerId   10000 non-null  int64
2   Surname      10000 non-null  object
3   CreditScore  10000 non-null  int64
4   Geography    10000 non-null  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
```

```
In [23]: plt.scatter(df.index,df['Balance'])
plt.show()
```



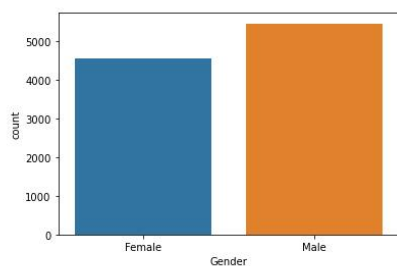
```
In [24]: sns.scatterplot(x=df.index,y=df['Balance'],hue=df['NumOfProducts'])
```

Out[24]:



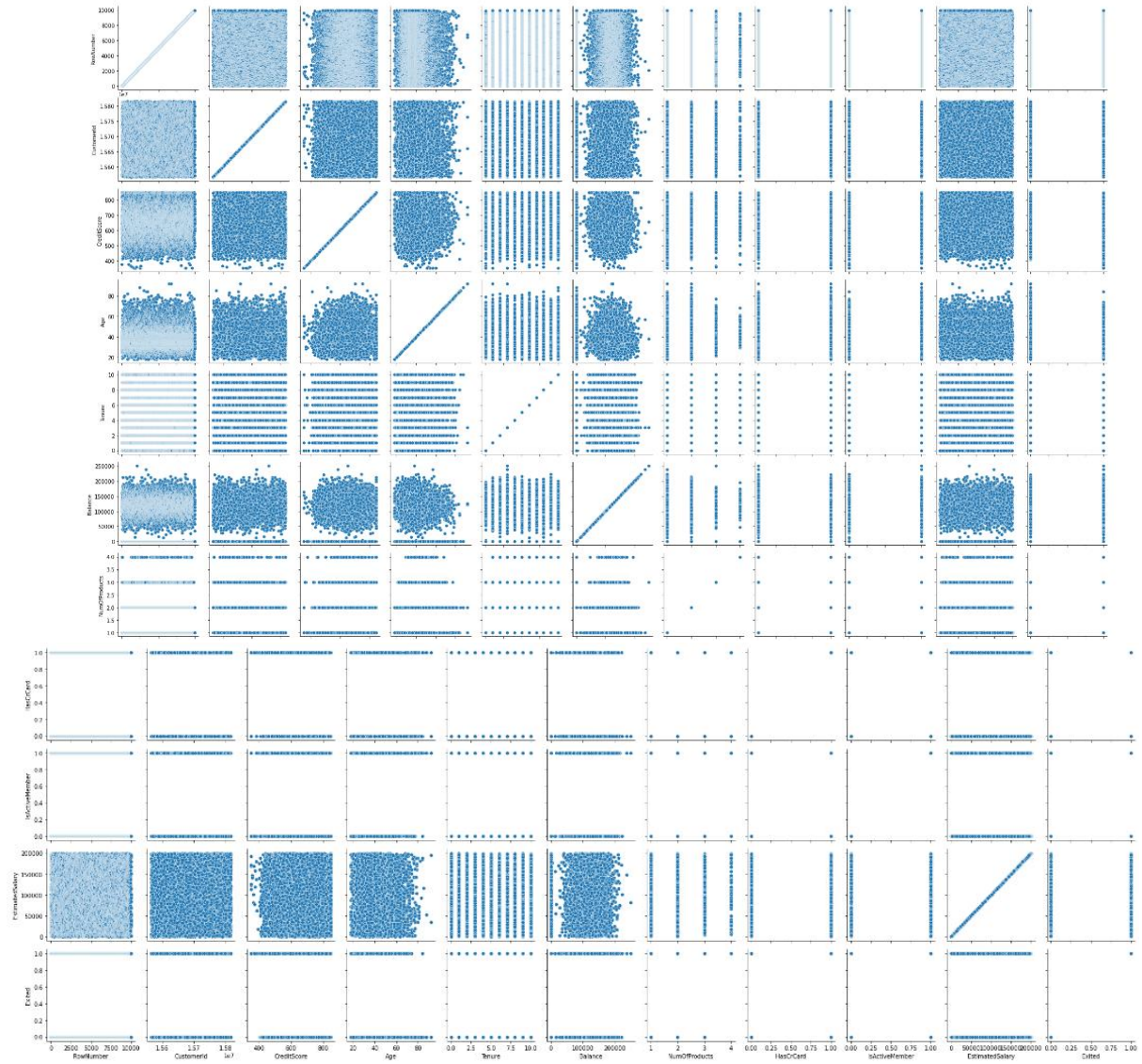
```
In [25]: sns.barplot(x='Gender',y='Exited',data=df)
sns.countplot(x='Gender',data=df)
```

Out[25]:



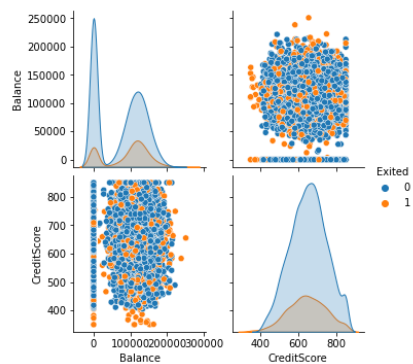
```
In [26]: g=sns.PairGrid(df)
g.map(sns.scatterplot)
```

Out[26]:



```
In [27]: sns.pairplot(data=df[['Balance', 'CreditScore', 'Exited']], hue='Exited')
```

Out[27]:



```
In [28]: df.describe(include='all')
```

Out[28]:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary
count	10000.00000	1.000000e+04	10000	10000.000000	10000	10000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000
unique	NaN	NaN	2932	NaN	3	2	NaN	NaN	NaN	NaN	NaN	NaN	NaN
top	NaN	NaN	Smith	NaN	France	Male	NaN	NaN	NaN	NaN	NaN	NaN	NaN
freq	NaN	NaN	32	NaN	5014	5457	NaN	NaN	NaN	NaN	NaN	NaN	NaN
mean	5000.50000	1.569094e+07	NaN	650.528800	NaN	NaN	38.921800	5.012800	76485.889288	1.530200	0.70550	0.515100	10000.000000
std	2886.89568	7.193619e+04	NaN	96.653299	NaN	NaN	10.487806	2.892174	62397.405202	0.581654	0.45584	0.499797	5716.000000
min	1.00000	1.556570e+07	NaN	350.000000	NaN	NaN	18.000000	0.000000	0.000000	1.000000	0.00000	0.000000	0.000000
25%	2500.75000	1.562853e+07	NaN	584.000000	NaN	NaN	32.000000	3.000000	0.000000	1.000000	0.00000	0.000000	5116.000000
50%	5000.50000	1.569074e+07	NaN	652.000000	NaN	NaN	37.000000	5.000000	97198.540000	1.000000	1.00000	1.000000	10000.000000
75%	7500.25000	1.575323e+07	NaN	718.000000	NaN	NaN	44.000000	7.000000	127644.240000	2.000000	1.00000	1.000000	14900.000000
max	10000.00000	1.581569e+07	NaN	850.000000	NaN	NaN	92.000000	10.000000	250898.090000	4.000000	1.00000	1.000000	19900.000000

In [29]: `df[(df['NumOfProducts']>2) | (df['NumOfProducts']<3)]`

Out[29]:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
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	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	1	1	1	79084.10	0
...
9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	96270.64	0
9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101699.77	0
9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	42085.58	1
9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	75075.31	2	1	0	92888.52	1
9999	10000	15628319	Walker	792	France	Female	28	4	130142.79	1	1	0	38190.78	0

10000 rows × 14 columns

In [30]: `df[(df['NumOfProducts']>2)]`

Out[30]:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
7	8	15656148	Obinna	376	Germany	Female	29	4	115046.74	4	1	0	119346.88	1
30	31	15589475	Azikiwe	591	Spain	Female	39	3	0.00	3	1	0	140469.38	1
70	71	15703793	Kononova	738	Germany	Male	58	2	133745.44	4	1	0	28373.86	1
88	89	15622897	Sharpe	646	France	Female	46	4	0.00	3	1	0	93251.42	1
...
9737	9738	15741197	Calzada	710	Spain	Male	22	8	0.00	3	1	0	107292.91	0
9747	9748	15775761	Iweobiegbonam	610	Germany	Female	69	5	86038.21	3	0	0	192743.06	1
9800	9801	15640507	Li	762	Spain	Female	35	3	119349.69	3	1	1	47114.18	1
9877	9878	15572182	Onwuamaeze	505	Germany	Female	33	3	106506.77	3	1	0	45445.78	1
9895	9896	15796764	Bruno	684	Germany	Female	56	3	127585.98	3	1	1	80593.49	1

326 rows × 14 columns

In [31]: `df['Age']=df['Age'].astype('float')
df.dtypes`

Out[31]:

```

RowNumber      int64
CustomerId      int64
Surname         object
CreditScore     int64
Geography       object
Gender          object
Age             float64
Tenure          int64
Balance         float64
NumOfProducts  int64
HasCrCard       int64
IsActiveMember  int64
EstimatedSalary float64
Exited          int64
dtype: object

```

In [32]: `pd.get_dummies(df,columns=['Tenure']).head()`

```
In [32]: pd.get_dummies(df,columns=['Tenure']).head()
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Balance	NumOfProducts	HasCrCard	...	Tenure_1	Tenure_2	Tenure_3	Tenure_4	Tenure_5	Tenure_6
0	1	15634602	Hargrave	619	France	Female	42.0	0.00	1	1	...	0	1	0	0	0	0
1	2	15647311	Hill	608	Spain	Female	41.0	83807.86	1	0	...	1	0	0	0	0	0
2	3	15619304	Onio	502	France	Female	42.0	159660.80	3	1	...	0	0	0	0	0	0
3	4	15701354	Boni	699	France	Female	39.0	0.00	2	0	...	1	0	0	0	0	0
4	5	15737888	Mitchell	850	Spain	Female	43.0	125510.82	1	1	...	0	1	0	0	0	0

5 rows x 24 columns

```
In [33]: x=df.iloc[:,:-1].values
y=df.iloc[:,4].values
print(x,y)
```

```
[[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]] ['France' 'Spain' 'France' ... 'France' 'Germany' 'France']
```

```
In [34]: x=df.iloc[1:3,:-1].values
x
```

```
Out[34]: array([[2, 15647311, 'Hill', 608, 'Spain', 'Female', 41.0, 1, 83807.86,
          1, 0, 1, 112542.58],
        [3, 15619304, 'Onio', 502, 'France', 'Female', 42.0, 8, 159660.8,
          3, 1, 0, 113931.57]], dtype=object)
```

```
In [35]: x=df[['Gender','Age']]
print(x)
```

	Gender	Age
0	Female	42.0
1	Female	41.0
2	Female	42.0
3	Female	39.0
4	Female	43.0
...
9995	Male	39.0
9996	Male	35.0
9997	Female	36.0
9998	Male	42.0
9999	Female	28.0

[10000 rows x 2 columns]

```
In [36]: from sklearn.model_selection import train_test_split
```

```
In [37]: training_data,testing_data=train_test_split(df,test_size=1,random_state=3)
print(training_data,testing_data)
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	\
6555	6556	15581505	Bales	641	France	Male	
1448	1449	15585367	Diribe	555	Germany	Female	
3351	3352	15792729	Holland	474	Germany	Female	
231	232	15627000	Freeman	610	France	Male	
1204	1205	15650098	Baranova	630	France	Female	
...	
6400	6401	15585907	Collier	676	Spain	Female	
9160	9161	15753679	Mullawirraburka	778	France	Male	
9859	9860	15615430	Adams	678	Germany	Male	
1688	1689	15804610	Valdez	601	France	Female	
5994	5995	15746065	Lo Duca	580	Germany	Male	

	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	\
6555	35.0	5	0.00	2	1	0	
1448	46.0	4	120392.99	1	1	0	
3351	34.0	9	176311.36	1	1	0	
231	40.0	0	0.00	2	1	0	
1204	40.0	7	0.00	2	1	1	
...	
6400	30.0	5	0.00	2	0	0	
9160	24.0	4	0.00	2	1	1	
9859	55.0	4	129646.91	1	1	1	
1688	41.0	1	0.00	2	0	1	
5994	35.0	10	136281.41	2	1	1	

	EstimatedSalary	Exited
6555	93148.93	0
1448	177719.88	1
3351	160213.27	0
231	62232.60	0
1204	34453.17	0
...
6400	179066.58	0
9160	162809.20	0
9859	184125.10	1
1688	160607.06	0
5994	24799.47	0

[9999 rows x 14 columns]

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	\
5876	5877	15585379	Humphries	704	France	Male	39.0	

	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	\
5876	2	111525.02	1	1	0	

	EstimatedSalary	Exited
5876	199484.96	0