

```
import numpy as np
```

```
import pandas as pd
```

```
df=pd.read_csv("Churn_Modelling.csv")
```

```
df
```

	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						
...	...	...	...	...	...	...
...						
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.shape
```

```
(10000, 14)
```

```
df.columns
```

```

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

```

```
df["NumOfProducts"].unique()
```

```
array([1, 3, 2, 4])
```

```
df["NumOfProducts"].value_counts()
```

```

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

```

```
df.dtypes
```

```

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

```

```
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.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

	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	

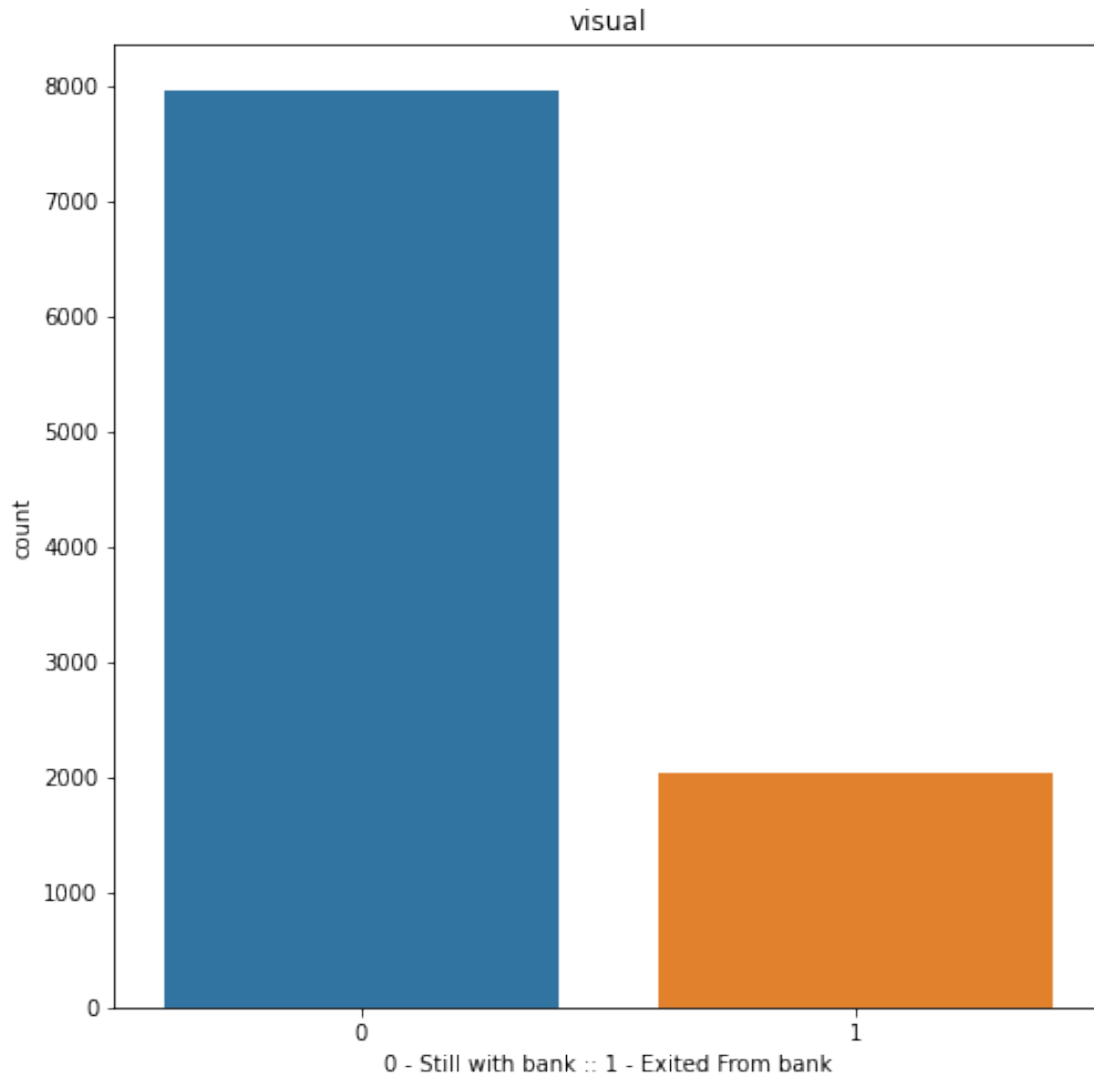
	EstimatedSalary	Exited
count	10000.000000	10000.000000
mean	100090.239881	0.203700
std	57510.492818	0.402769
min	11.580000	0.000000
25%	51002.110000	0.000000
50%	100193.915000	0.000000
75%	149388.247500	0.000000
max	199992.480000	1.000000

```
import matplotlib.pyplot as plt
```

```
import seaborn as sns
```

```
%matplotlib inline
```

```
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()
```



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

```
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
```

```
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
```

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

```
df1.shape
```

```
(10000, 14)
```

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

```
updated_df['Balance']=updated_df['Balance'].fillna(updated_df['Balance']  
'').mean())
```

```
updated_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 10000 entries, 0 to 9999
```

```
Data columns (total 14 columns):
```

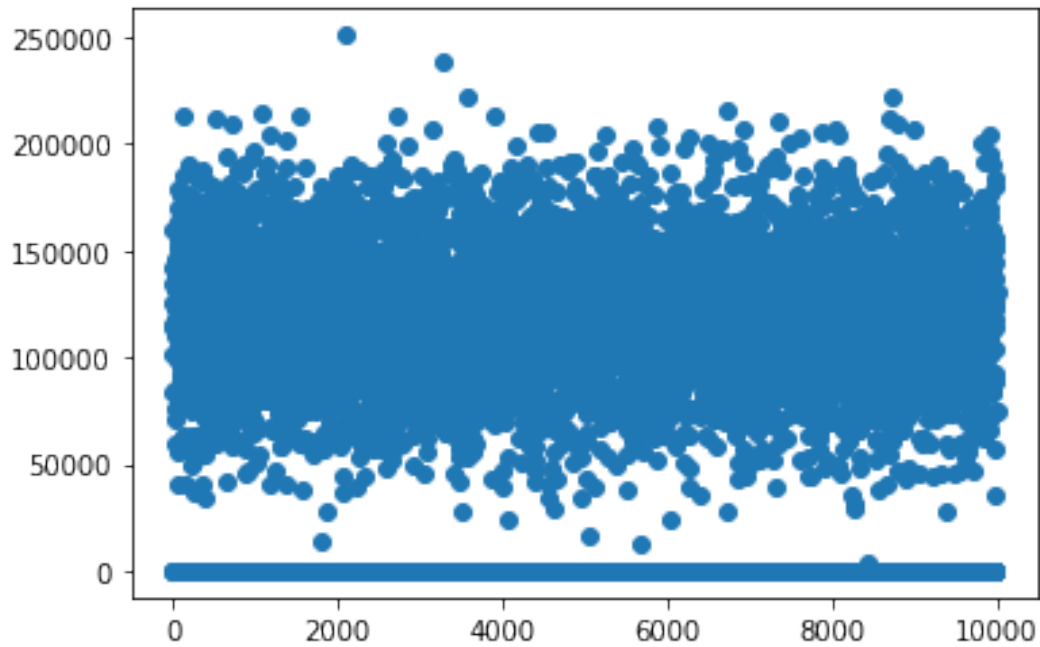
#	Column	Non-Null Count	Dtype
0	RowNumber	10000 non-null	int64
1	CustomerId	10000 non-null	int64
2	Surname	10000 non-null	object
3	CreditScore	10000 non-null	int64
4	Geography	10000 non-null	object
5	Gender	10000 non-null	object
6	Age	10000 non-null	int64
7	Tenure	10000 non-null	int64
8	Balance	10000 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)
```

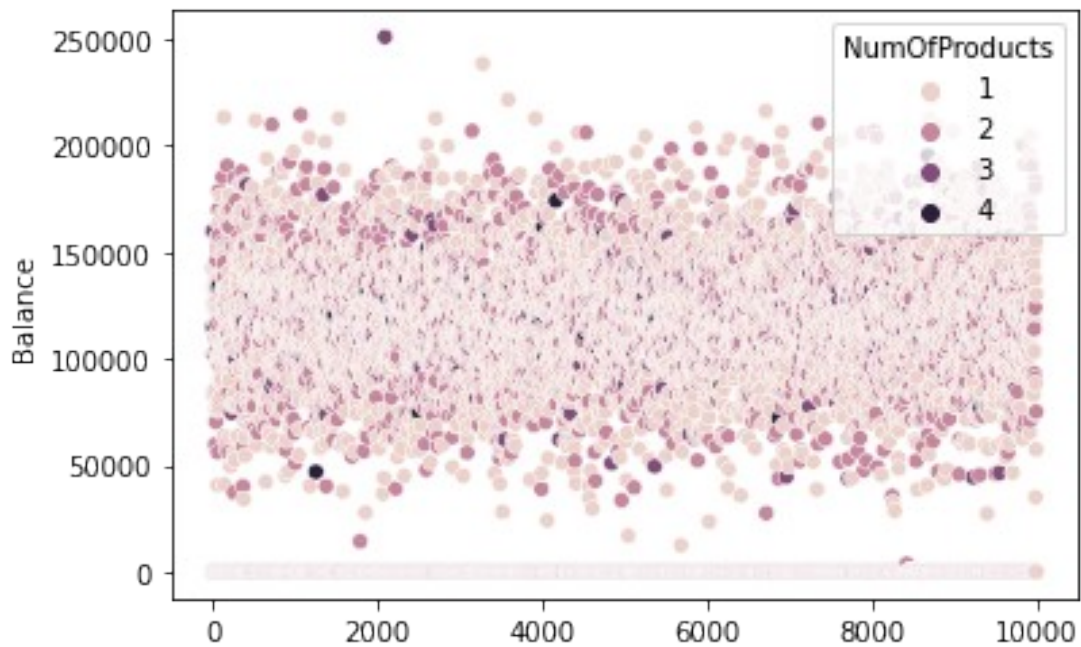
```
memory usage: 1.1+ MB
```

```
plt.scatter(df.index,df['Balance'])
```

```
plt.show()
```

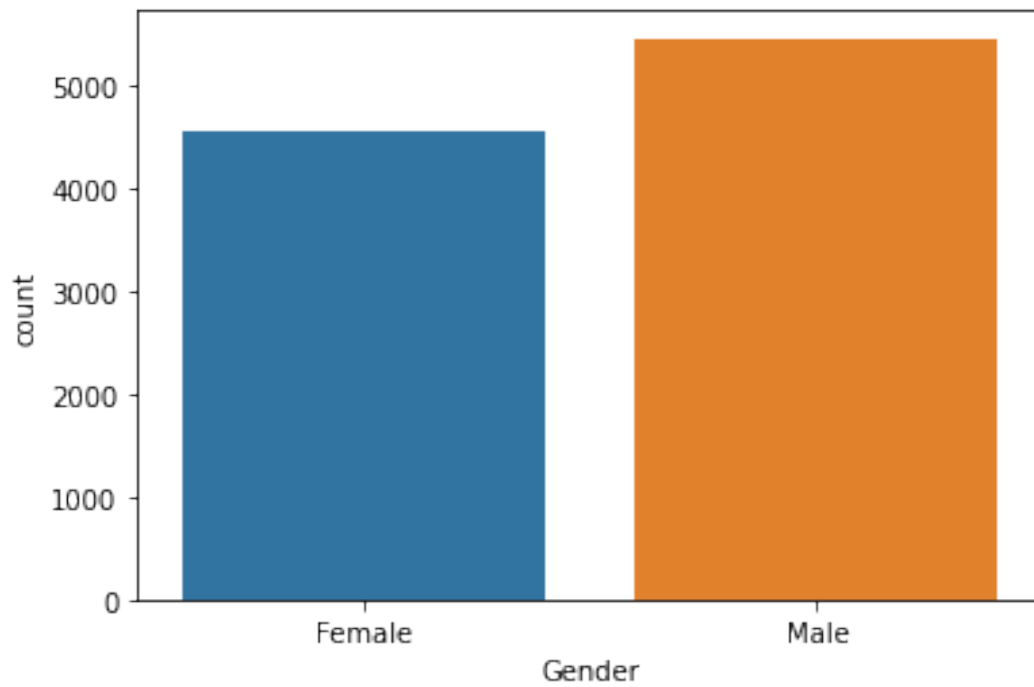


```
sns.scatterplot(x=df.index,y=df['Balance'],hue=df['NumOfProducts'])
<matplotlib.axes._subplots.AxesSubplot at 0x7f2251ff7e50>
```

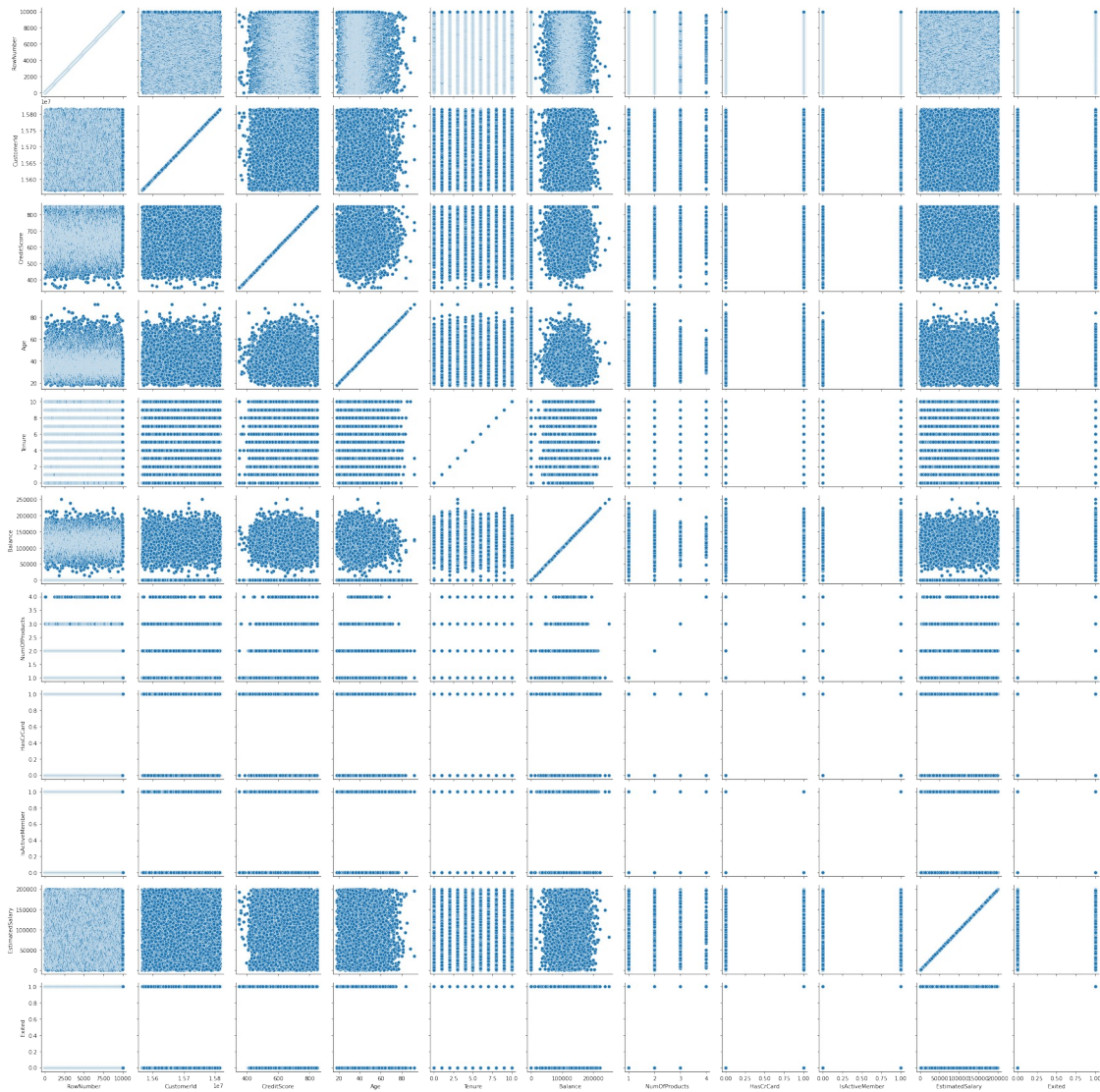


```
sns.barplot(x='Gender',y='Exited',data=df)
sns.countplot(x='Gender',data=df)
<matplotlib.axes._subplots.AxesSubplot at 0x7f2251d44210>
```



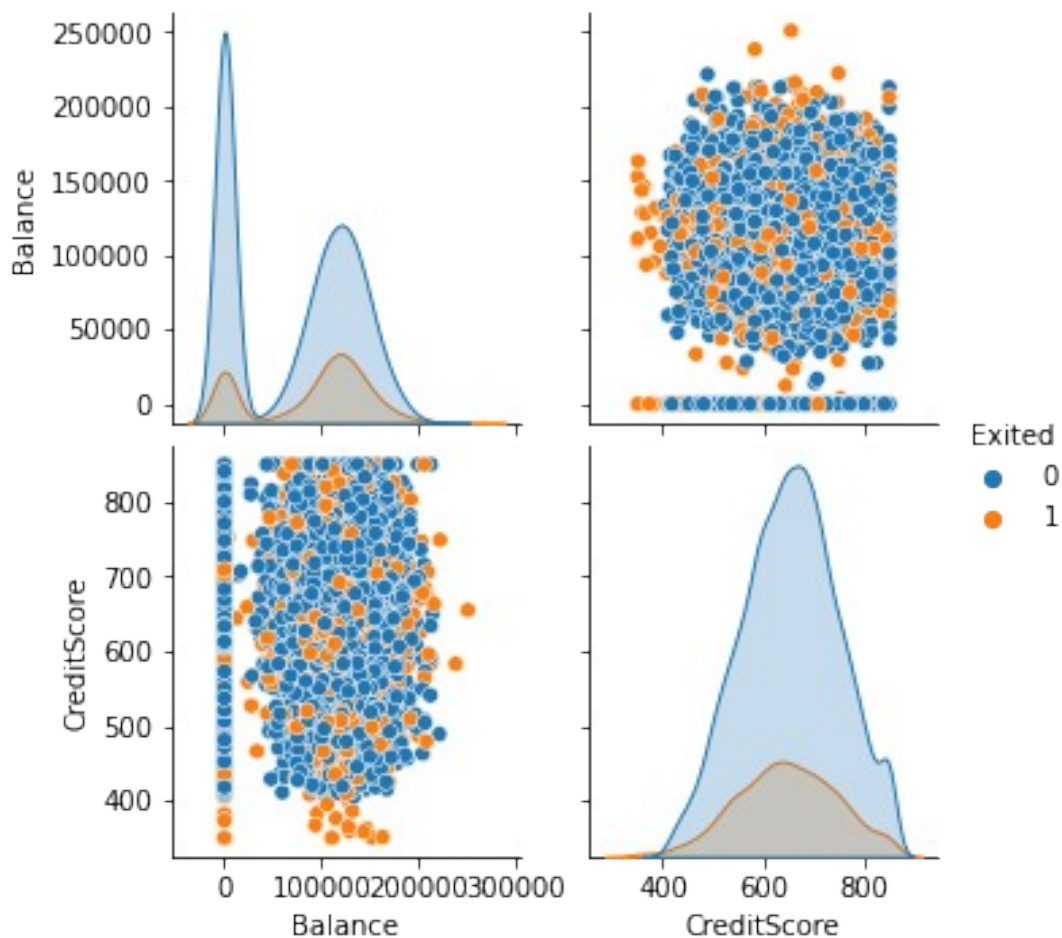


```
g=sns.PairGrid(df)
g.map(sns.scatterplot)
<seaborn.axisgrid.PairGrid at 0x7f2251f6a850>
```



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

```
<seaborn.axisgrid.PairGrid at 0x7f224bd66390>
```



```
df.describe(include='all')
```

	RowNumber	CustomerId	Surname	CreditScore	Geography
Gender	\				
count	10000.00000	1.000000e+04	10000	10000.000000	10000
unique	NaN	NaN	2932	NaN	3
2					
top	NaN	NaN	Smith	NaN	France
Male					
freq	NaN	NaN	32	NaN	5014
5457					
mean	5000.50000	1.569094e+07	NaN	650.528800	NaN
NaN					
std	2886.89568	7.193619e+04	NaN	96.653299	NaN
NaN					
min	1.00000	1.556570e+07	NaN	350.000000	NaN
NaN					
25%	2500.75000	1.562853e+07	NaN	584.000000	NaN
NaN					
50%	5000.50000	1.569074e+07	NaN	652.000000	NaN

NaN					
75%	7500.25000	1.575323e+07	NaN	718.000000	NaN
NaN					
max	10000.00000	1.581569e+07	NaN	850.000000	NaN
NaN					

	Age	Tenure	Balance	NumOfProducts
HasCrCard \				
count	10000.000000	10000.000000	10000.000000	10000.000000
10000.000000				
unique	NaN	NaN	NaN	NaN
NaN				
top	NaN	NaN	NaN	NaN
NaN				
freq	NaN	NaN	NaN	NaN
NaN				
mean	38.921800	5.012800	76485.889288	1.530200
0.70550				
std	10.487806	2.892174	62397.405202	0.581654
0.45584				
min	18.000000	0.000000	0.000000	1.000000
0.00000				
25%	32.000000	3.000000	0.000000	1.000000
0.00000				
50%	37.000000	5.000000	97198.540000	1.000000
1.00000				
75%	44.000000	7.000000	127644.240000	2.000000
1.00000				
max	92.000000	10.000000	250898.090000	4.000000
1.00000				

	IsActiveMember	EstimatedSalary	Exited
count	10000.000000	10000.000000	10000.000000
unique	NaN	NaN	NaN
NaN			
top	NaN	NaN	NaN
NaN			
freq	NaN	NaN	NaN
NaN			
mean	0.515100	100090.239881	0.203700
std	0.499797	57510.492818	0.402769
min	0.000000	11.580000	0.000000
25%	0.000000	51002.110000	0.000000
50%	1.000000	100193.915000	0.000000
75%	1.000000	149388.247500	0.000000
max	1.000000	199992.480000	1.000000

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

	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						
...	...	...	...	...	...	...
...						
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[(df['NumOfProducts']>2)]
```

Gender \	RowNumber	CustomerId	Surname	CreditScore	Geography
2 Female	3	15619304	Onio	502	France
7 Female	8	15656148	Obinna	376	Germany
30 Female	31	15589475	Azikiwe	591	Spain
70 Male	71	15703793	Konovalova	738	Germany
88 Female	89	15622897	Sharpe	646	France
...	...	...	...	...	...
9737 Male	9738	15741197	Calzada	710	Spain
9747 Female	9748	15775761	Iweobiegbonam	610	Germany
9800 Female	9801	15640507	Li	762	Spain
9877 Female	9878	15572182	Onwuamaeze	505	Germany
9895 Female	9896	15796764	Bruno	684	Germany

\	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember
2	42	8	159660.80	3	1	0
7	29	4	115046.74	4	1	0
30	39	3	0.00	3	1	0
70	58	2	133745.44	4	1	0
88	46	4	0.00	3	1	0
...	...	...	...	...	...	...
9737	22	8	0.00	3	1	0
9747	69	5	86038.21	3	0	0
9800	35	3	119349.69	3	1	1
9877	33	3	106506.77	3	1	0
9895	56	3	127585.98	3	1	1

	EstimatedSalary	Exited
2	113931.57	1
7	119346.88	1
30	140469.38	1
70	28373.86	1
88	93251.42	1
...	...	...
9737	107292.91	0
9747	192743.06	1
9800	47114.18	1
9877	45445.78	1
9895	80593.49	1

[326 rows x 14 columns]

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

```

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
```

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

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender
Age \						
0	1	15634602	Hargrave	619	France	Female
42.0						
1	2	15647311	Hill	608	Spain	Female
41.0						
2	3	15619304	Onio	502	France	Female
42.0						
3	4	15701354	Boni	699	France	Female
39.0						
4	5	15737888	Mitchell	850	Spain	Female
43.0						

```
Balance  NumOfProducts  HasCrCard  ...  Tenure_1  Tenure_2
```

Tenure_3 \					
0	0.00	1	1	...	0
0					1
1	83807.86	1	0	...	0
0					
2	159660.80	3	1	...	0
0					0
3	0.00	2	0	...	0
0					1
4	125510.82	1	1	...	0
0					1

	Tenure_4	Tenure_5	Tenure_6	Tenure_7	Tenure_8	Tenure_9
Tenure_10						
0	0	0	0	0	0	0
0						
1	0	0	0	0	0	0
0						
2	0	0	0	0	1	0
0						
3	0	0	0	0	0	0
0						
4	0	0	0	0	0	0
0						

[5 rows x 24 columns]

```
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']
```

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

```
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)
```



```
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]
```

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

```
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
5876	704	France
5877	15585379	Male

Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember
5876	2	111525.02	1	0

	EstimatedSalary	Exited
5876	199484.96	0