```
from google.colab import drive
drive.mount('/content/drive')
```

## Mounted at /content/drive

import numpy as np
import pandas as pd
import seaborn as sns
from matplotlib import pyplot as pt

%matplotlib inline

df = pd.read\_csv('/content/drive/MyDrive/Colab
Notebooks/downloads/Churn\_Modelling.csv')
df

Age 0 42 1 41 2	RowNumb	er Custome	erId	Surname	CreditScore	e Geography	Gender
	\	1 15634	1602	Hargrave	619	) France	Female
		2 15647	7311	Hill	608	Spain	Female
		3 15619	9304	Onio	502	? France	Female
42 3		4 1570	L354	Boni	699	) France	Female
39 4 43		5 15737	7888	Mitchell	850	) Spain	Female
9995	99	96 15606	5229	0bijiaku	773	France	Male
39 9996 35	99	97 15569	9892	Johnstone	516	6 France	Male
9997 36	99	98 15584	1532	Liu	709	) France	Female
9998 42	99	99 15682	2355	Sabbatini	772	2 Germany	Male
9999 28	100	00 15628	3319	Walker	792	? France	Female
0 1 2 3 4	Tenure 2 1 8 1 2	Balance 0.00 83807.86 159660.80 0.00 125510.82	Num	OfProducts 1 1 3 2 1	HasCrCard 1 0 1 0 1	IsActiveMem	ber \ 1
9995 9996 9997	5 10 7	0.00 57369.61 0.00		2 1 1	1 1 0		0 1 1

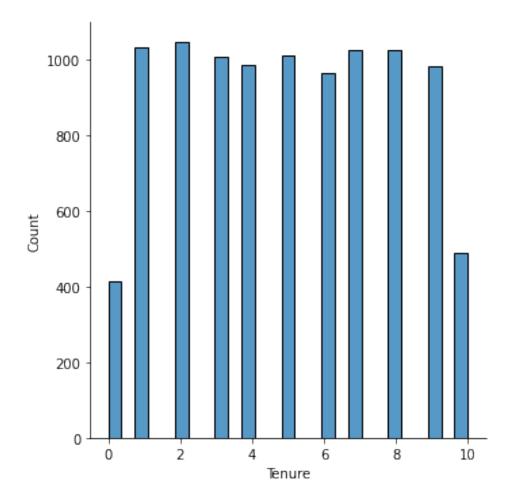
9998	3	75075.31	2	1	0
9999	4	130142.79	1	1	0

	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

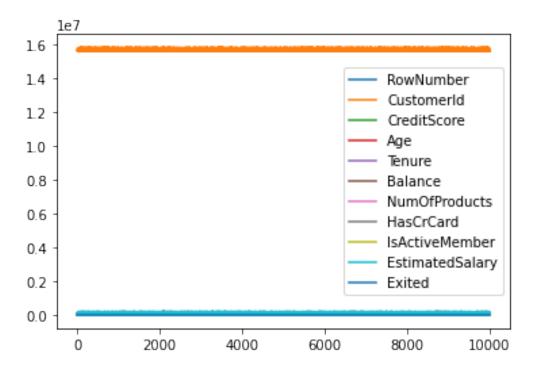
[10000 rows x 14 columns]

sns.displot(df.Tenure)

<seaborn.axisgrid.FacetGrid at 0x7fdebb761250>



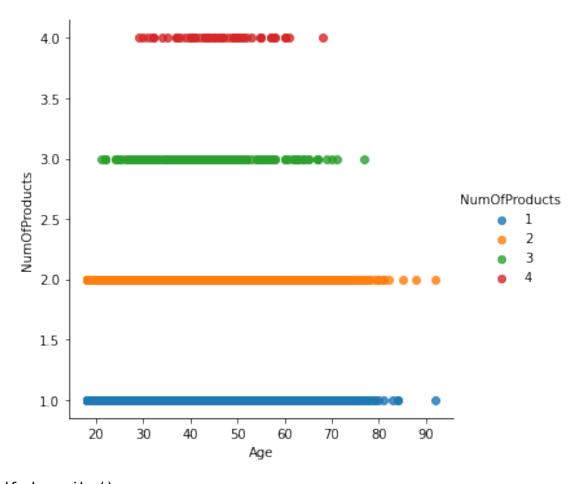
df.plot.line()
<matplotlib.axes. subplots.AxesSubplot at 0x7fdea91aacd0>



sns.lmplot("Age", "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



df.describe()

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

Balance NumOfProducts HasCrCard IsActiveMember \
count 10000.000000 10000.000000 10000.000000

```
76485.889288
                            1.530200
                                           0.70550
                                                           0.515100
mean
std
        62397.405202
                            0.581654
                                           0.45584
                                                           0.499797
                                           0.00000
                                                           0.000000
min
            0.000000
                            1.000000
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
       250898.090000
                            4.000000
                                           1.00000
                                                           1.000000
max
       EstimatedSalary
                                Exited
          10000.000000
                         10000.000000
count
mean
         100090.239881
                             0.203700
std
          57510.492818
                             0.402769
              11.580000
                             0.000000
min
25%
          51002.110000
                             0.000000
50%
         100193.915000
                             0.000000
75%
         149388.247500
                             0.000000
max
         199992.480000
                             1.000000
data = pd.read csv('/content/drive/MyDrive/Colab
Notebooks/downloads/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
df["Tenure"] = np.where(df["Tenure"] >10, np.median,df["Tenure"])
df["Tenure"]
         2
0
1
         1
2
         8
3
         1
         2
         5
9995
9996
        10
9997
         7
         3
9998
9999
         4
Name: Tenure, Length: 10000, dtype: object
```

```
pd.get_dummies(df, columns=["Gender", "Age"], prefix=["Age",
"Gender"]).head()
   RowNumber CustomerId
                               Surname CreditScore Geography Tenure
Balance \
            1
                  15634602
                             Hargrave
                                                  619
                                                          France
                                                                        2
0
0.00
1
            2
                  15647311
                                  Hill
                                                  608
                                                           Spain
                                                                        1
83807.86
            3
                  15619304
                                  Onio
                                                  502
                                                          France
                                                                        8
159660.80
                  15701354
            4
                                                  699
                                                                        1
3
                                  Boni
                                                          France
0.00
            5
                  15737888 Mitchell
                                                  850
                                                           Spain
                                                                        2
125510.82
   NumOfProducts HasCrCard IsActiveMember ...
                                                         Gender 78 Gender 79
0
                 1
                              1
                                                1
                                                                  0
                                                                               0
                                                    . . .
1
                 1
                              0
                                                                  0
                                                                               0
                                                1
                                                    . . .
2
                 3
                              1
                                                                               0
                                                0
                                                                  0
3
                 2
                                                                               0
                              0
                                                0
                                                                  0
4
                 1
                              1
                                                                               0
                                                1
                                                                  0
                                                  . . .
                Gender_81
                                         Gender_83
   Gender_80
                            Gender_82
                                                     Gender_84
                                                                  Gender_85
0
            0
                         0
                                     0
                                                                           0
            0
                         0
                                     0
                                                  0
1
                                                               0
                                                                           0
                                     0
                                                  0
                                                                           0
2
            0
                         0
                                                               0
3
            0
                         0
                                      0
                                                  0
                                                               0
                                                                           0
4
            0
                         0
                                      0
                                                  0
                                                               0
                                                                           0
   Gender_88
                Gender 92
0
                         0
            0
1
            0
                         0
2
            0
                         0
3
            0
                         0
            0
[5 rows x 84 columns]
X = df.iloc[:, :-2].values
print(X)
[[1 15634602 'Hargrave' ... 1 1 1]
[2 15647311 'Hill' ... 1 0 1]
```

```
[3 15619304 'Onio' ... 3 1 0]
 [9998 15584532 'Liu' ... 1 0 1]
 [9999 15682355 'Sabbatini' ... 2 1 0]
 [10000 15628319 'Walker' ... 1 1 0]]
import pandas as pd
df = pd.read csv('/content/drive/MyDrive/Colab
Notebooks/downloads/Churn Modelling.csv')
Y = df.iloc[:, -1].values
print(Y)
[1 \ 0 \ 1 \ \dots \ 1 \ 1 \ 0]
from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler()
df[["RowNumber"]]=scaler.fit transform(df[["RowNumber"]])
print(df)
      RowNumber
                  CustomerId
                                 Surname CreditScore Geography
                                                                    Gender
Age
         0.0000
                    15634602
                                Hargrave
                                                           France
                                                                    Female
0
                                                   619
42
1
         0.0001
                    15647311
                                    Hill
                                                   608
                                                            Spain
                                                                    Female
41
2
         0.0002
                    15619304
                                    Onio
                                                   502
                                                           France
                                                                   Female
42
3
         0.0003
                    15701354
                                    Boni
                                                   699
                                                           France
                                                                   Female
39
         0.0004
                                Mitchell
                                                   850
4
                    15737888
                                                            Spain
                                                                    Female
43
. . .
             . . .
                          . . .
                    15606229
                                Obijiaku
                                                                      Male
9995
         0.9996
                                                   771
                                                           France
39
         0.9997
9996
                    15569892
                               Johnstone
                                                   516
                                                           France
                                                                      Male
35
9997
         0.9998
                    15584532
                                     Liu
                                                   709
                                                           France
                                                                   Female
36
9998
         0.9999
                    15682355
                               Sabbatini
                                                   772
                                                          Germany
                                                                      Male
42
9999
                                                   792
         1.0000
                    15628319
                                  Walker
                                                           France
                                                                   Female
28
                          NumOfProducts
      Tenure
                 Balance
                                          HasCrCard
                                                       IsActiveMember
0
                    0.00
            2
                                        1
                                                   1
                                                                     1
            1
                                                   0
                                                                     1
1
                83807.86
                                        1
2
           8
               159660.80
                                        3
                                                   1
                                                                     0
3
            1
                                        2
                    0.00
                                                   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
           3
                                       2
                                                   1
                                                                    0
9998
                75075.31
9999
           4
               130142.79
                                       1
                                                   1
                                                                    0
      EstimatedSalary
                        Exited
0
             101348.88
                              0
1
            112542.58
2
                              1
             113931.57
3
                              0
             93826.63
4
             79084.10
                              0
9995
             96270.64
                              0
                              0
9996
             101699.77
                              1
9997
             42085.58
9998
                              1
             92888.52
             38190.78
                              0
9999
[10000 rows x 14 columns]
from sklearn.model selection import train test split
train size=0.8
X = d\overline{f}.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)
print(X_valid.shape), print(y_valid.shape)
print(X_test.shape), print(y_test.shape)
(8000, 13)
(8000,)
(1000, 13)
(1000,)
(1000, 13)
(1000,)
(None, None)
```

. . .

. . .

. . .

. . .