Assignment -2 Data Visualization and Pre-Processing

Assignment Date	21 September 2022
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Maximum Marks	2 Marks

1.Download Dataset

importing package

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

2.Loading dataset

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

Out[]: Row Cust Sur Cred Te Bal NumO Has IsActiv Estima Ex Geo Ge A Num omer na itSco grap nd g nu anc **fProdu** CrC eMem tedSal ite ber Id hy ard ber d me re er \mathbf{e} re e cts ary Har Fe 1563 101348 Fran 4 0 1 619 2 0.00 1 1 1 ma gra 4602 2 .88 ce le ve 838 Fe 4 112542 1564 Spai 1 Hill 608 07.8 1 0 ma 7311 1 .58 n le 6 Fe 159 1561 Oni Fran 4 113931 2 3 502 3 660. 1 1 ma 9304 2 o ce .57 le 80 Fe 1570 3 93826. Bon Fran 699 3 0.00 2 0 4 0 ma 1354 9 i 63 ce le Fe 125 Mit 1573 4 79084. Spai chel 850 510. 1 1 0 ma 7888 n 3 10 82 1 le

	Row Num ber	Cust omer Id	Sur na me	Cred itSco re	Geo grap hy	Ge nd er	A g e	Te nu re	Bal anc e	NumO fProdu cts	Has CrC ard	IsActiv eMem ber	Estima tedSal ary	Ex ite d
•••														
9 9 9 5	9996	1560 6229	Obi jiak u	771	Fran ce	Ma le	3 9	5	0.00	2	1	0	96270. 64	0
9 9 9 6	9997	1556 9892	Joh nsto ne	516	Fran ce	Ma le	3 5	10	573 69.6 1	1	1	1	101699 .77	0
9 9 9 7	9998	1558 4532	Liu	709	Fran ce	Fe ma le	3 6	7	0.00	1	0	1	42085. 58	1
9 9 9 8	9999	1568 2355	Sab bati ni	772	Ger man y	Ma le	4 2	3	750 75.3 1	2	1	0	92888. 52	1
9 9 9	1000	1562 8319	Wal ker	792	Fran ce	Fe ma le	2 8	4	130 142. 79	1	1	0	38190. 78	0

1. 10000 rows × 14 columns

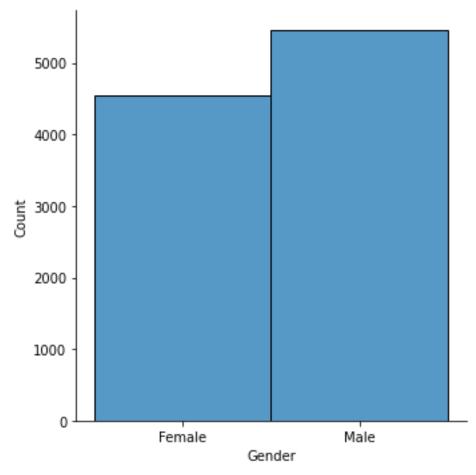
3. Visualizations

a) Univariate Analysis

sns.displot(df.Gender)

Out[]:

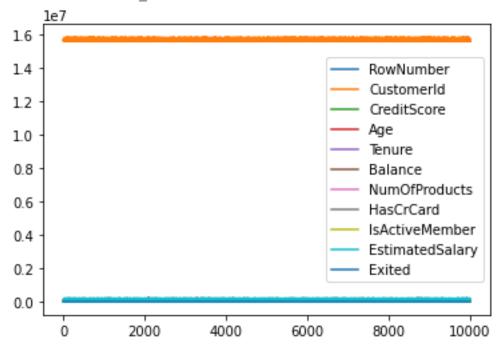
<seaborn.axisgrid.FacetGrid at 0x7f2935aae790>



b) Bi-Variate Analysis

df.plot.line()

<matplotlib.axes._subplots.AxesSubplot at 0x7f2932d2df50>



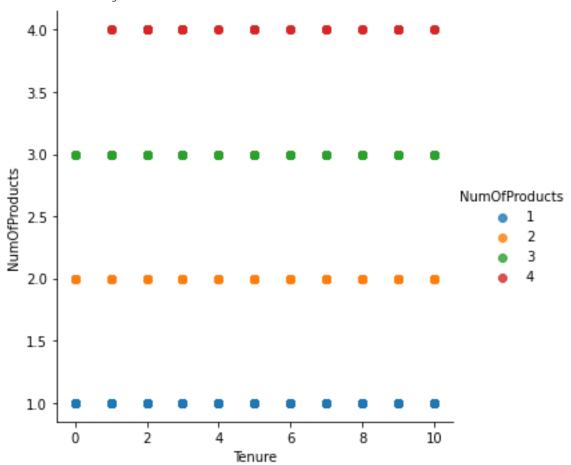
Out[]:

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: FutureWar ning: Pass the following variables as keyword args: x, y, data. From versio n 0.12, the only valid positional argument will be `data`, and passing othe r arguments without an explicit keyword will result in an error or misinter

FutureWarning

pretation.



4.Perform descriptive statistics on the dataset.

df.describe()

											Out[]:
	RowN umbe r	Custo merId	Credit Score	Age	Tenur e	Balanc e	NumOf Product s	HasC rCar d	IsActive Membe r	Estimat edSalar y	Exited
co un t	10000 .0000 0	1.0000 00e+0 4	10000. 00000 0	10000. 00000 0	10000. 00000 0	10000. 000000	10000.0 00000	10000 .0000 0	10000.0 00000	10000.0 00000	10000. 00000 0

	RowN umbe r	Custo merId	Credit Score	Age	Tenur e	Balanc e	NumOf Product s	HasC rCar d	IsActive Membe r	Estimat edSalar y	Exited
m ea n	5000. 50000	1.5690 94e+0 7	650.52 8800	38.921 800	5.0128 00	76485. 889288	1.53020	0.705 50	0.51510 0	100090. 239881	0.2037
st d	2886. 89568	7.1936 19e+0 4	96.653 299	10.487 806	2.8921 74	62397. 405202	0.58165 4	0.455 84	0.49979 7	57510.4 92818	0.4027 69
mi n	1.000	1.5565 70e+0 7	350.00 0000	18.000 000	0.0000	0.0000	1.00000	0.000	0.00000	11.5800 00	0.0000
25 %	2500. 75000	1.5628 53e+0 7	584.00 0000	32.000 000	3.0000	0.0000	1.00000	0.000	0.00000	51002.1 10000	0.0000
50 %	5000. 50000	1.5690 74e+0 7	652.00 0000	37.000 000	5.0000	97198. 540000	1.00000	1.000	1.00000	100193. 915000	0.0000
75 %	7500. 25000	1.5753 23e+0 7	718.00 0000	44.000 000	7.0000	127644 .24000 0	2.00000	1.000	1.00000	149388. 247500	0.0000
m ax	10000 .0000 0	1.5815 69e+0 7	850.00 0000	92.000 000	10.000	250898 .09000 0	4.00000	1.000	1.00000	199992. 480000	1.0000

5.Handle the Missing values.

data = pd.read_csv("Churn_Modelling.csv")
pd.isnull(data["Gender"])

Out[]:

```
0
      False
1
       False
2
       False
3
       False
      False
       . . .
9995
      False
9996
      False
9997
       False
      False
9998
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: FutureWar ning: Pass the following variable as a keyword arg: x. From version 0.12, t he only valid positional argument will be `data`, and passing other argumen ts without an explicit keyword will result in an error or misinterpretation

FutureWarning

<matplotlib.axes._subplots.AxesSubplot at 0x7f2932b8c650>

Out[]:

```
20 30 40 50 60 70 80 90
Age
```

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

```
0
         42
1
         41
2
         42
3
         39
         43
9995
         39
9996
         35
9997
         36
9998
         42
9999
         28
```

Name: Age, Length: 10000, dtype: int64

sns.boxplot(df['Age'])

In []:

Out[]:

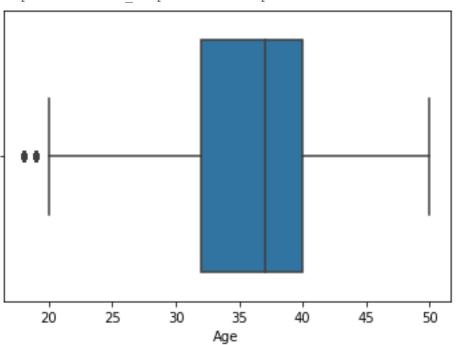
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWar ning: Pass the following variable as a keyword arg: x. From version 0.12, t he only valid positional argument will be `data`, and passing other argumen

ts without an explicit keyword will result in an error or misinterpretation . Future Warning

Out[]:

In []:

<matplotlib.axes. subplots.AxesSubplot at 0x7f2932a9a450>



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

Out[]:
0     42
1     41
2     42
3     39
4     43
...</pre>
```

9995 39 9996 35 9997 36 9998 42

28

9999

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()

								Out[]:	
	RowNumber	CustomerId	Surname	CreditScore	Geography	Tenure	Balance	NumOfProducts	HasCrCard
0	1	15634602	Hargrave	619	France	2	0.00	1	1

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

5 rows × 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"]])
                                                               In []:
print(df)
     RowNumber CustomerId
                          Surname CreditScore Geography Gender
\
0
                0.275616 Hargrave
                                                 France Female 42
            1
                                          619
            2
                                                 Spain Female
1
                0.326454
                           Hill
                                          608
                                                               41
                0.214421
2
            3
                                          502
                             Onio
                                                France Female 42
3
               0.542636
                                          699 France Female 39
            4
                             Boni
            5
                0.688778 Mitchell
                                          850
                                                 Spain Female 43
          . . .
                                                          . . .
                                          . . .
                                                   . . .
         9996 0.162119 Obijiaku
9995
                                          771
                                                               39
                                                 France
                                                         Male
                0.016765 Johnstone
9996
         9997
                                          516
                                                        Male
                                                                35
                                                France
9997
         9998
               0.075327
                            Liu
                                          709 France Female
```

```
9999 0.466637 Sabbatini
10000 0.250483 Walker
9998
                                       772 Germany Male
                                                            42
9999
                                        792
                                              France Female
                                                            28
     Tenure Balance NumOfProducts HasCrCard IsActiveMember \
0
        2
            0.00
                         1
                                        1
                                                     1
        1 83807.86
                                        0
                                                      1
1
                              1
         8 159660.80
                              3
                                        1
                                                      0
                              2
                                                      0
3
         1
               0.00
                                        0
        2 125510.82
                              1
                                       1
                                                      1
            • • • •
                             . . .
       . . .
                                      . . .
                                                    . . .
9995
        5
               0.00
                              2
                                       1
                                                     0
                              1
                                       1
9996
       10 57369.61
                                                      1
                              1
        7
9997
             0.00
                                        0
                                                      1
                              2
                                       1
9998
        3 75075.31
                                                     0
9999
        4 130142.79
                                       1
                                                     0
                              1
     EstimatedSalary Exited
0
         101348.88
                   1
         112542.58
1
2
         113931.57
                       1
3
          93826.63
                      0
          79084.10
                      0
              . . .
          96270.64
9995
                    0
                      0
9996
        101699.77
9997
          42085.58
                      1
9998
          92888.52
                      1
9999
          38190.78
                      0
```

10.Split the data into training and testing

[10000 rows x 14 columns]

```
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)
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,)
                                                                        Out[]:
(None, None)
```