Assignment – 2

Assignment Date	05 October 2022
Student Name	Angeline Joy Alex
Student Roll Number	811519104009
Maximum Marks	2 Marks

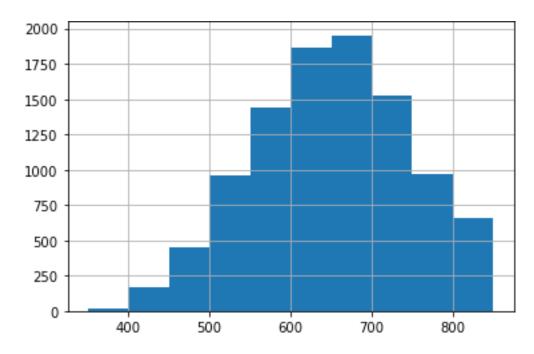
```
import pandas as pd
import numpy as np
from scipy import stats
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels.api as sm
from sklearn.model selection import train test split
churn=pd.read csv("E:/Churn Modelling.csv")
churn.head(10)
   RowNumber CustomerId
                            Surname CreditScore Geography Gender
                                                                       Age
0
                 15634602
                           Hargrave
                                               619
                                                      France
                                                              Female
                                                                        42
1
           2
                 15647311
                               Hill
                                               608
                                                       Spain
                                                               Female
                                                                        41
2
           3
                 15619304
                                Onio
                                               502
                                                      France
                                                              Female
                                                                        42
                 15701354
3
           4
                                               699
                                                                        39
                                Boni
                                                      France Female
4
           5
                 15737888
                           Mitchell
                                               850
                                                       Spain
                                                               Female
                                                                        43
5
                                               645
                 15574012
                                 Chu
                                                       Spain
                                                                 Male
                                                                        44
                 15592531
6
           7
                           Bartlett
                                               822
                                                      France
                                                                Male
                                                                        50
7
                 15656148
                             Obinna
                                               376
                                                                        29
           8
                                                     Germany
                                                              Female
           9
                 15792365
                                               501
8
                                  Не
                                                      France
                                                                Male
                                                                        44
9
          10
                 15592389
                                  H?
                                               684
                                                      France
                                                                Male
                                                                        27
                                                   IsActiveMember
             Balance NumOfProducts HasCrCard
   Tenure
        2
                 0.00
0
                                    1
                                                1
                                                                 1
1
            83807.86
                                                0
                                                                 1
        1
                                    1
2
                                    3
                                                                 0
        8
           159660.80
                                                1
3
        1
                 0.00
                                    2
                                                0
                                                                 0
4
        2
                                    1
                                                1
                                                                 1
           125510.82
5
                                    2
                                                                 0
        8
           113755.78
                                                1
6
        7
                                    2
                 0.00
                                                1
                                                                 1
7
        4
          115046.74
                                    4
                                                                 0
                                                1
8
        4
           142051.07
                                    2
                                                0
                                                                 1
```

134603.88

2	113931.57	1
3	93826.63	0
4	79084.10	0
5	149756.71	1
6	10062.80	0
7	119346.88	1
8	74940.50	0
9	71725.73	0

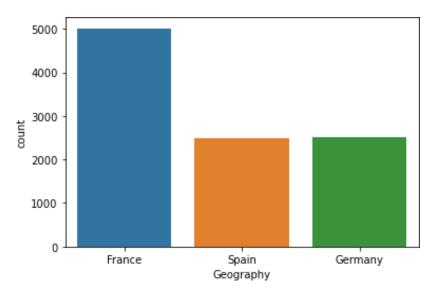
churn.CreditScore.hist()

<AxesSubplot:>



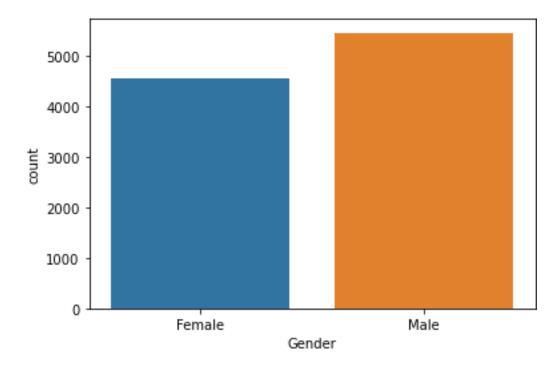
sns.countplot(x="Geography",data=churn)

<AxesSubplot:xlabel='Geography', ylabel='count'>



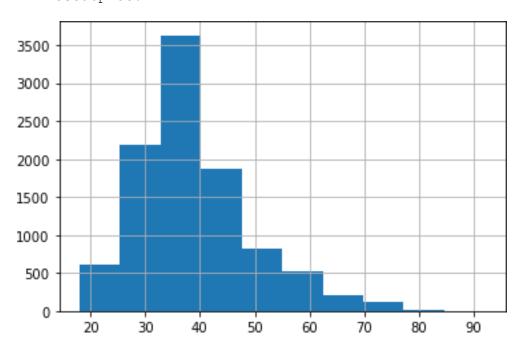
sns.countplot(x="Gender",data=churn)

<AxesSubplot:xlabel='Gender', ylabel='count'>



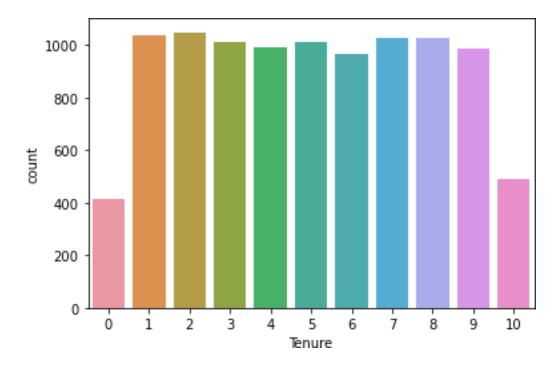
churn.Age.hist()

<AxesSubplot:>



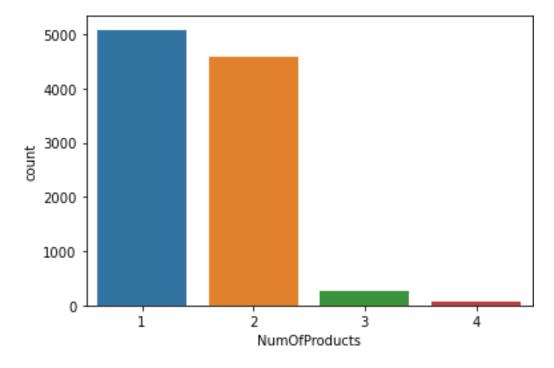
sns.countplot(x="Tenure",data=churn)

<AxesSubplot:xlabel='Tenure', ylabel='count'>



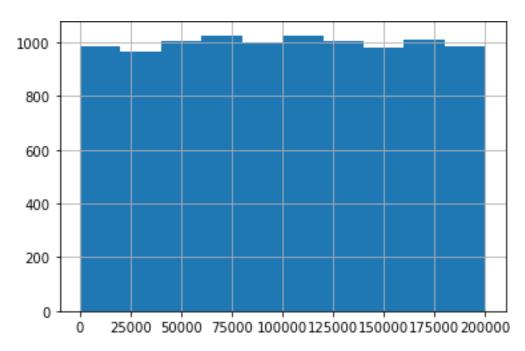
sns.countplot(x="NumOfProducts",data=churn)

<AxesSubplot:xlabel='NumOfProducts', ylabel='count'>

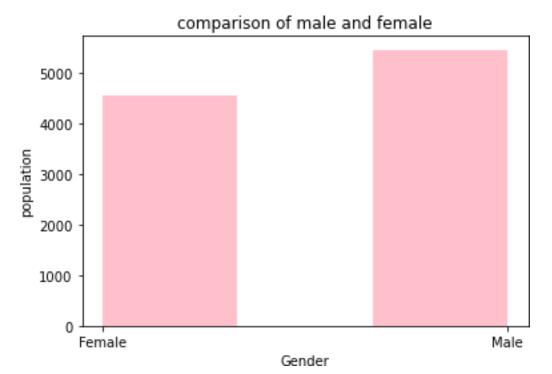


churn.EstimatedSalary.hist()

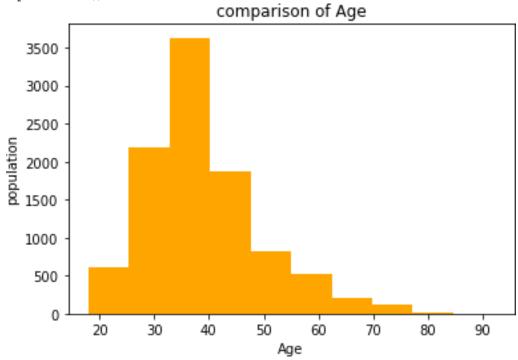
<AxesSubplot:>



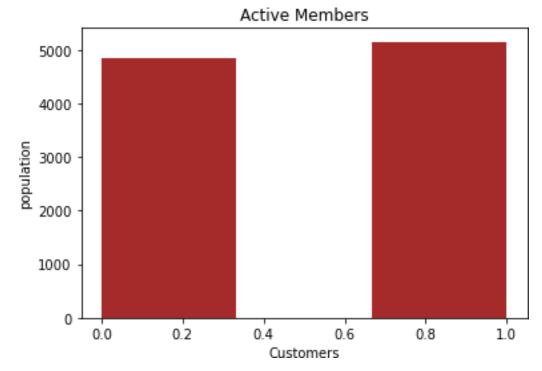
```
plt.hist(x = churn.Gender, bins = 3, color = 'pink')
plt.title('comparison of male and female')
plt.xlabel('Gender')
plt.ylabel('population')
plt.show()
```



```
plt.hist(x = churn.Age, bins = 10, color = 'orange')
plt.title('comparison of Age')
plt.xlabel('Age')
plt.ylabel('population')
plt.show()
```



```
plt.hist(x = churn.IsActiveMember, bins = 3, color = 'brown')
plt.title('Active Members')
plt.xlabel('Customers')
plt.ylabel('population')
plt.show()
```



<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

churn.CreditScore.describe()

count10000.0000000mean650.528800std96.653299min350.00000025%584.00000050%652.00000075%718.000000max850.000000

Name: CreditScore, dtype: float64

churn.Geography.describe()

count 10000 unique 3 top France freq 5014

Name: Geography, dtype: object

churn.Gender.describe()

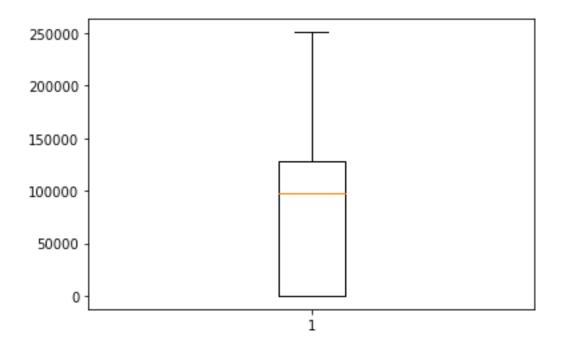
count 10000
unique 2
top Male
freq 5457

Name: Gender, dtype: object

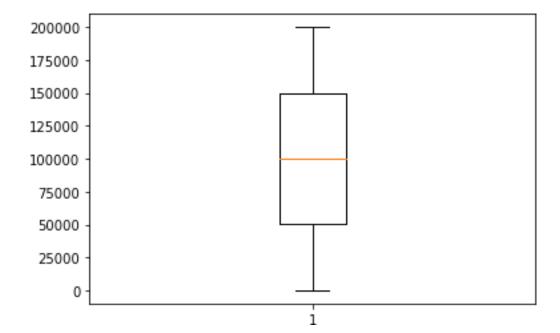
churn.Age.describe()

count 10000.000000

```
mean
            38.921800
std
            10.487806
            18.000000
min
25%
            32.000000
50%
            37.000000
75%
            44.000000
             92.000000
max
Name: Age, dtype: float64
churn.Tenure.describe()
         10000.000000
count
mean
              5.012800
            19.869999
max
Max
Mame: Tenure, odtype oofloat 64
25% 3.000000 ghurn.EstimatedSalary.describe()
          100709888880
75%
coiint
         100090.239881
mean
         57510.492818
std
min
             11.580000
25%
         51002.110000
         100193.915000
50%
75%
         149388.247500
         199992.480000
Name: EstimatedSalary, dtype: float64
churn.isnull().sum()
RowNumber
                    0
                    0
CustomerId
                    0
Surname
CreditScore
                    0
Geography
                    0
Gender
                    0
Aae
                    0
                    0
Tenure
Balance
                    0
NumOfProducts
                    0
HasCrCard
                    \Omega
                    0
IsActiveMember
EstimatedSalary
                    0
                    0
Exited
dtype: int64
plt.boxplot(churn.Balance)
{'whiskers': [<matplotlib.lines.Line2D at 0x1c6e4427340>,
  <matplotlib.lines.Line2D at 0x1c6e4427610>],
 'caps': [<matplotlib.lines.Line2D at 0x1c6e44279a0>,
  <matplotlib.lines.Line2D at 0x1c6e4427bb0>],
 'boxes': [<matplotlib.lines.Line2D at 0x1c6e4427070>],
 'medians': [<matplotlib.lines.Line2D at 0x1c6e4427e80>],
 'fliers': [<matplotlib.lines.Line2D at 0x1c6e4439190>],
 'means': []}
```



plt.boxplot(churn.EstimatedSalary)



```
from sklearn import preprocessing
label encoder = preprocessing.LabelEncoder()
churn['Geography'] = label encoder.fit transform(churn['Geography'])
churn['Gender'] = label encoder.fit transform(churn['Gender'])
churn = churn.drop(['CustomerId', 'Surname', 'RowNumber'], axis = 1)
y=churn.Exited
churn.drop(['Exited'], axis = 1,inplace=True)
x=churn
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler().fit(x)
scaled data=scaler.transform(x)
x= pd.DataFrame(scaled data)
x_train,x_test,y_train,y_test= train_test_split(x,y,train size = 0.8,
test_size = 0.2,random state =42)
x train.head(5)
                   2
                             3
                                           5
9254 0.672
           0.0 1.0 0.189189 0.6
                                     0.000000
                                              0.333333
0.895494
1561 0.564
            0.5 1.0 0.324324 0.4
                                    0.476786 0.333333
                                                        1.0
                                                            1.0
0.979930
1670 0.418
            1.0 1.0 0.081081 0.3
                                    0.457317
                                              0.000000
                                                           0.0
                                                        1.0
0.429438
6087 0.422
            0.0 0.0 0.121622 0.9 0.540606 0.000000 1.0
0.765417
6669 0.334 0.0 1.0 0.513514 0.9 0.566554 0.000000 0.0 0.0
0.197401
y train.head(5)
9254
       0
1561
       0
1670
       1
6087
       1
6669
       1
Name: Exited, dtype: int64
x test.head(5)
                   2
                             3
                                 4
                                           5
                                    0.385452
6252 0.492
            0.5 1.0 0.189189 0.3
                                              0.333333
0.208904
4684 0.546
            0.0 1.0 0.337838 0.1 0.000000
                                              0.333333
                                                        1.0
                                                            1.0
0.731908
1731 0.502 1.0 0.0 0.351351 0.4 0.000000 0.333333 1.0 0.0
```

0.292777 4742 0.312 0.5 1.0 0.554054 0.8 0.474902 0.333333 1.0 1.0 0.853422 4521 0.420 1.0 0.0 0.121622 0.7 0.498194 0.000000 1.0 1.0 0.573346

y_test.head(5)

6252 0 4684 0 1731 0 4742 0 4521 0

Name: Exited, dtype: int64