ASSIGNMENT 2

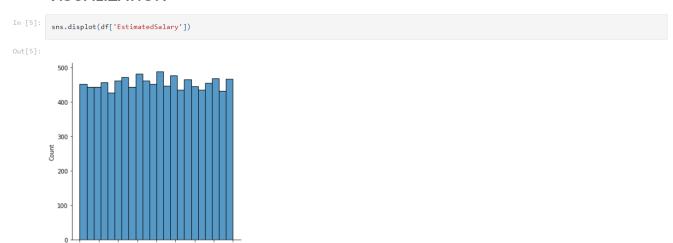
IMPORT THE NECESSARY LIBRARIES

In []:
 import pandas as pd
 import numpy as np
 import matplotlib.pyplot as plt
 import seaborn as sns

IMPORT THE DATA SET INTO DATAFRAME

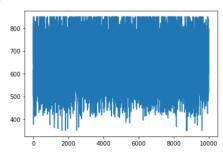
| In [2]: | df=pd.read_csv('Churn_Modelling.csv') | | | | | | | | | | | | | | |
|---------|--|---|----------|----------|-----|--------|--------|----|---|-----------|---|---|--------|-----------|---|
| In [3]: | df.head() | | | | | | | | | | | | | | |
| Out[3]: | RowNumber Customerld Surname CreditScore Geography Gender Age Tenure Balance NumOfProducts HasCrCard IsActiveMember EstimatedSalary Exited | | | | | | | | | | | | 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 |

VISUALIZATION



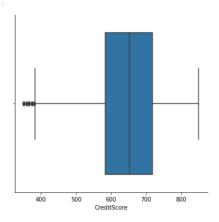
In [15]: #univariate analysis
 df.CreditScore.plot()





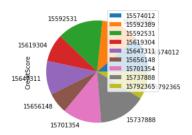
In [6]: sns.catplot(x='CreditScore',kind='box',data=df)

Out[6]



In [16]: df[1:10].groupby(['CustomerId']).sum().plot(kind='pie', y='CreditScore')

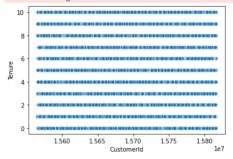
Out[16]:



In [17]:
 sns.scatterplot(df.CustomerId,df.Tenure)
 nlt.show()

C:\Users\darat\AppData\Local\Programs\Python\Python36\lib\site-packages\seaborn_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. 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.plot() 1.6 RowNumber 1.4 Customerld CreditScore 1.2 Age Tenure Balance 1.0 0.8 NumOfProducts 0.6 HasCrCard IsActiveMember 0.4 EstimatedSalary 0.2 0.0 #bivariate analysis df.CreditScore[1:10].plot() df.Balance[1:10].plot() Out[25]: 160000 140000 120000 100000 80000

DESCRIPTIVE ANALYSIS

40000 20000

#multivariate analysis

df.describe() Out[31]: RowNumber CustomerId CreditScore Tenure Balance NumOfProducts HasCrCard IsActiveMember EstimatedSalary Exited count 10000.00000 1.000000e+04 10000.000000 10000.000000 10000.000000 10000.000000 10000.000000 10000.00000 10000.000000 10000.000000 10000.000000 5000.50000 1.569094e+07 650.528800 38.921800 5.012800 76485.889288 1.530200 0.70550 0.515100 100090.239881 0.203700 2886.89568 7.193619e+04 96.653299 2.892174 62397.405202 0.581654 0.45584 0.499797 57510.492818 0.402769 10.487806 std 1.00000 1.556570e+07 min 350.000000 18.000000 0.000000 0.000000 1.000000 0.00000 0.000000 11.580000 0.000000 2500.75000 1.562853e+07 584.000000 32.000000 3.000000 0.000000 1.000000 0.00000 0.000000 51002.110000 0.000000 5000.50000 1.569074e+07 652,000000 1.000000 1.00000 100193,915000 50% 37.000000 5.000000 97198.540000 1.000000 0.000000 7500.25000 1.575323e+07 718.000000 44.000000 7.000000 127644.240000 2.000000 1.00000 1.000000 149388.247500 0.000000 10000.00000 1.581569e+07 850.000000 92.000000 10.000000 250898.090000 4.000000 1.00000 1.000000 199992.480000 1.000000

CHECKING FOR MISSING VALUES

In [32]: df.isnull().any() RowNumber False Out[32]: False CustomerId Surname False CreditScore False Geography False Gender False False Tenure False Balance False NumOfProducts False HasCrCard False IsActiveMember False EstimatedSalary False Exited False dtype: bool

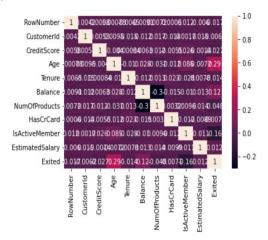
```
df.isnull().sum()
Out[33]: RowNumber
         CustomerId
                            0
         Surname
         CreditScore
                            0
         Geography
         Gender
         Age
         Tenure
         Balance
         NumOfProducts
         HasCrCard
         IsActiveMember
         EstimatedSalary
                            0
         Exited
         dtype: int64
```

HANDLING VALUES

```
In [36]: #No null values to handle

In [37]: sns.heatmap(df.corr(),annot=True)
```

Out[37]:



OUTLIERS

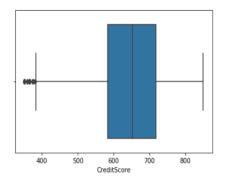
In [38]:

#occurence of outliers
sns.boxplot(df.CreditScore)

C:\Users\darat\AppData\Local\Programs\Python\Python36\lib\site-packages\seaborn_decorators.py:43: FutureWarning: Pass the following variable as a keyw ord arg: x. 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

Out[38]:



```
In [39]: Q1= df.CreditScore.quantile(0.25)
Q3=df.CreditScore.quantile(0.75)
In [40]: IQR=Q3-Q1
In [41]: upper_limit =Q3 + 1.5*IQR
lower_limit =Q1 - 1.5*IQR
```

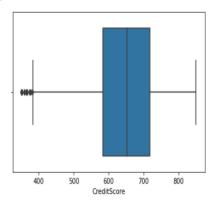
df['CreditScore'] = np.where(df['CreditScore']>upper_limit,30,df['CreditScore'])

In [45]: sns.boxplot(df.CreditScore)

C:\Users\darat\AppData\Local\Programs\Python\Python36\lib\site-packages\seaborn_decorators.py:43: FutureWarning: Pass the following variable as a keyw ord arg: x. 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 missinterpretation.

FutureWarning

Out[45]:



CATEGORICAL COLUMNS _ENCODING

```
In [46]: #Label encoder
from sklearn.preprocessing import LabelEncoder

In [47]: le=LabelEncoder()

In [52]: df.Gender= le.fit_transform(df.Gender)

In [54]: df.head(5)
```

| Out[54]: | | RowNumber | CustomerId | Surname | CreditScore | Geography | Gender | Age | Tenure | Balance | NumOfProducts | HasCrCard | IsActiv | veMember | Estim | natedSala | ry Exited | |
|----------------------|----------|---|---|----------------------------|---------------------|----------------------|----------|--------------|--------|-----------|-------------------------|-------------|---------|----------------|----------------------|------------|-------------|-------|
| | 0 | 1 | 15634602 | Hargrave | 619 | France | 0 | 42 | 2 | 0.00 | 1 | 1 | | 1 | | 101348 | .88 1 | |
| | 1 | 2 | 15647311 | Hill | 608 | Spain | 0 | 41 | 1 | 83807.86 | 1 | 0 | | 1 | | 112542 | .58 0 | |
| | 2 | 3 | 15619304 | Onio | 502 | France | 0 | 42 | 8 | 159660.80 | 3 | 1 | | 0 | | 113931 | .57 1 | |
| | 3 | 4 | 15701354 | Boni | 699 | France | 0 | 39 | 1 | 0.00 | 2 | 0 | | 0 | | 93826 | .63 0 | |
| | 4 | 5 | 15737888 | Mitchell | 850 | Spain | 0 | 43 | 2 | 125510.82 | 1 | 1 | | 1 | | 79084 | .10 0 | |
| In [55]: | #(| one hot enco | odina | | | | | | | | | | | | | | | |
| In [55]: Out[55]: | df | one hot enco f_main=pd.ge f_main.head(RowNumber | et_dummies(| | | | ge Tenur | e Ba | alance | NumOfProd | ucts HasCrCard | IsActiveMen | nber E | EstimatedSa | lary | Exited | Geography_F | ranc |
| | df | f_main=pd.ge f_main.head(| et_dummies(d | | | Gender Ag | • | е В а | alance | NumOfProd | ucts HasCrCard | IsActiveMen | nber E | EstimatedSa | • | Exited | Geography_F | Franc |
| Out[55]: | d† | f_main=pd.ge f_main.head(| et_dummies(d | Surname | CreditScore | Gender Ag | 12 | 2 | | NumOfProd | ucts HasCrCard 1 1 1 0 | IsActiveMen | | | 8.88 | Exited 1 0 | Geography_F | |
| Out[55]: | di di | f_main=pd.ge f_main.head(RowNumber | CustomerId | Surname Hargrave | CreditScore 619 | Gender Ag 0 4 | 12 11 | 2 | 0.00 | NumOfProd | 1 1 | IsActiveMen | 1 | 10134 | 8.88 | 1 | Geography_F | Franc |
| Out[55]: | 0 1 | f_main=pd.ge f_main.head(RowNumber 1 | ct_dummies(a) CustomerId 15634602 15647311 | Surname Hargrave | CreditScore 619 608 | Gender Ag | 12 11 | 2 1 83 | 0.00 | NumOfProd | 1 1 1 | IsActiveMen | 1 | 10134 11254 | 8.88 2.58 1.57 | 1 | Geography_F | |

SEPARATING INDEPENDENT AND DEPENDENT VARIABLES

In [70]: X=df_main.drop(columns=['EstimatedSalary'],axis=1)
 X.head()
 X_scaled=pd.DataFrame(scale(X),columns=X.columns)
 X_scaled.head()

| Out[70]: | | RowNumber | CustomerId | CreditScore | Gender | Age | Tenure | Balance | NumOfProducts | HasCrCard | IsActiveMember | Exited | Geography_France |
|----------|---|-----------|------------|-------------|-----------|----------|-----------|-----------|---------------|-----------|----------------|-----------|------------------|
| | 0 | -1.731878 | -0.783213 | -0.326221 | -1.095988 | 0.293517 | -1.041760 | -1.225848 | -0.911583 | 0.646092 | 0.970243 | 1.977165 | 0.997204 |
| | 1 | -1.731531 | -0.606534 | -0.440036 | -1.095988 | 0.198164 | -1.387538 | 0.117350 | -0.911583 | -1.547768 | 0.970243 | -0.505775 | -1.002804 |
| | 2 | -1.731185 | -0.995885 | -1.536794 | -1.095988 | 0.293517 | 1.032908 | 1.333053 | 2.527057 | 0.646092 | -1.030670 | 1.977165 | 0.997204 |
| | 3 | -1.730838 | 0.144767 | 0.501521 | -1.095988 | 0.007457 | -1.387538 | -1.225848 | 0.807737 | -1.547768 | -1.030670 | -0.505775 | 0.997204 |
| | 4 | -1.730492 | 0.652659 | 2.063884 | -1.095988 | 0.388871 | -1.041760 | 0.785728 | -0.911583 | 0.646092 | 0.970243 | -0.505775 | -1.002804 |

```
In [71]: y=df_main.EstimatedSalary
Out[71]: 0
                101348.88
               112542.58
113931.57
                 93826.63
                79084.10
                 96270.64
         9995
         9996
               101699.77
         9997
                 42085.58
                92888.52
         9999
                 38190.78
         Name: EstimatedSalary, Length: 10000, dtype: float64
         SCALING
          from sklearn.preprocessing import scale
In [73]: X_scaled=pd.DataFrame(scale(X),columns=X.columns)
          X scaled.head()
Out[73]: RowNumber Customerld CreditScore Gender
                                                       Age
                                                                      Balance NumOfProducts HasCrCard IsActiveMember
                                                                                                                      Exited Geography_France
         0
             -1.731878
                        -0.783213 -0.326221 -1.095988 0.293517 -1.041760 -1.225848
                                                                                   -0.911583
                                                                                             0.646092
                                                                                                           0.970243 1.977165
                                                                                                                                    0.997204
            -1.731531 -0.606534 -0.440036 -1.095988 0.198164 -1.387538 0.117350
                                                                                   -0.911583 -1.547768
                                                                                                           0.970243 -0.505775
                                                                                                                                   -1.002804
                                                                                                                   1.977165
             -1.731185
                                  -1.536794 -1.095988 0.293517 1.032908 1.333053
                                                                                             0.646092
                                                                                                           -1.030670
                                                                                                                                    0.997204
                      -0.995885
                                                                                    2.527057
             -1.030670 -0.505775
         3
                                                                                   0.807737 -1.547768
                                                                                                                                    0.997204
             -1.730492 0.652659 2.063884 -1.095988 0.388871 -1.041760 0.785728
                                                                                   -0.911583 0.646092
                                                                                                           0.970243 -0.505775
                                                                                                                                    -1.002804
        TRAIN AND TEST DATA
In [74]:
         from sklearn.model_selection import train_test_split
         X_train,X_test,y_train,y_test =train_test_split(X_scaled,y, test_size=0.3,random_state=0)
In [66]:
         X_train.shape
Out[66]: (7000, 14)
In [67]:
         X_test.shape
Out[67]: (3000, 14)
In [68]:
         y_train.shape
Out[68]: (7000,)
```

In [69]:

Out[69]: (3000,)

y_test.shape