Assignment -2

Data Visualization and Pre-processing

Assignment Date	25 September 2022
Student Name	Nithiyashree M G
Student Roll Number	1919102098
Maximum Marks	2 Marks

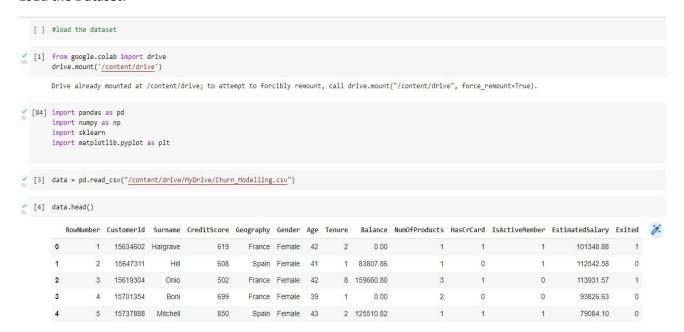
Question-1:

Download the dataset:

The dataset "Churn_Modelling.csv" was downloaded Successfully

Question-2:

Load the Dataset:



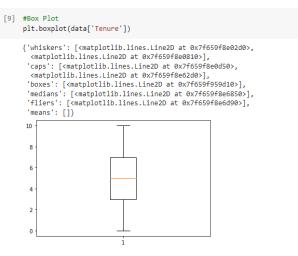
Question-3:

Perform Below Visualization:

Univariate Analysis

[5] #Univariate Analysis for Numerical data

[6] #Histogram data['Age'].plot(kind='hist') <matplotlib.axes._subplots.AxesSubplot at 0x7f65a0462590> 3500 25

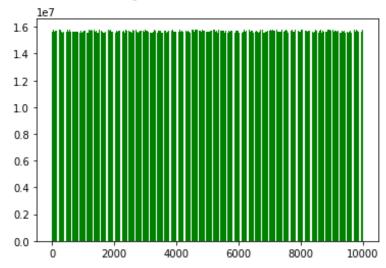


[] #Univariate Analysis for Categorical Data

```
[14] #Bar Chart
    df = pd.DataFrame(data)

X = list(df.iloc[:, 0])
Y = list(df.iloc[:, 1])
plt.bar(X, Y, color='g')
```

<BarContainer object of 10000 artists>



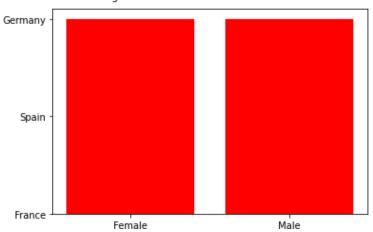
Bivariate Analysis

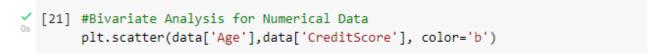
```
[23] #Bivariate Analysis for Categorical Data

#Stacked Bar chart

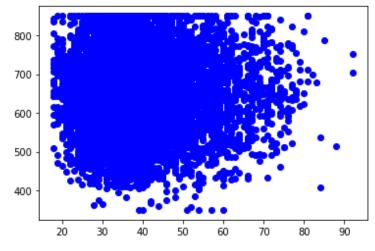
plt.bar(data['Gender'], data['Geography'], color='r')
```

<BarContainer object of 10000 artists>





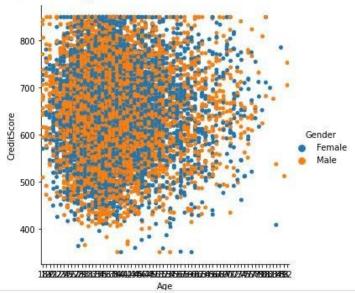
<matplotlib.collections.PathCollection at 0x7f6589f606d0>

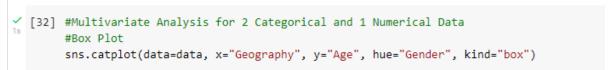


Multivariate Analysis

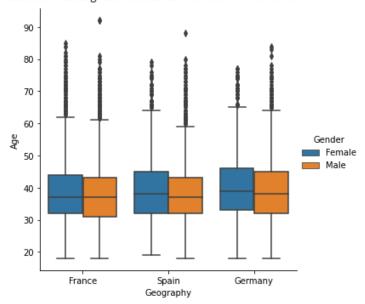
```
#Multivariate Analysis for 2 Numerical and 1 Categorical Data
#Scatter Plot
import seaborn as sns
sns.catplot(data=data, x="Age", y="CreditScore", hue="Gender")
```

<seaborn.axisgrid.FacetGrid at 0x7f657aab5d90>





<seaborn.axisgrid.FacetGrid at 0x7f6575c43490>



Question-4:

(10000, 14)

Perform Descriptive Statistics on the dataset:

```
[ ] #Perform Descriptive Statistics on the Dataset
       data.mean()
   /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1:
           """Entry point for launching an IPython kernel.
         RowNumber
                         1.569094e+07
                               5.0005000+03
        CustomerId
        CreditScore
                              6.505288e+02
        Age
                               3.892180e+01
        Tenure
                                5.012800e+00
        Balance
                                7.648589e+04
         NumOfProducts
                                1.530200e+00
        HasCrCard
                                7.055000e-01
        IsActiveMember
                               5.151000e-01
         EstimatedSalary 1.000902e+05
        Exited
                                2.037000e-01
         dtype: float64
[34] data.median()
         /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1:
           """Entry point for launching an IPython kernel.
         RowNumber
                           5.000500e+03
         CustomerId
                              1.569074e+07
         CreditScore
                               6.520000e+02
         Age
                                3.700000e+01
         Tenure
                                 5.000000e+00
                               9.719854e+04
         Balance
        NumOfProducts
                              1.000000e+00
        HasCrCard
                               1.000000e+00
                               1.000000e+00
         IsActiveMember
         EstimatedSalary
                                1.001939e+05
         Exited
                                0.000000e+00
         dtype: float64
[36] data.describe()
                                          Age
           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 10000.000000 10000.000000 10000.000000 10000.000000 10000.000000 10000.000000 10000.000000
      mean 5000.50000 1.569094e+07 650.528800 38.921800 5.012800 76485.889288 1.530200 0.70550

        std
        2886.89568
        7.193619e+04
        96.653299
        10.487806
        2.892174
        62397.405202
        0.581654
        0.45584
        0.499797
        57510.492818
        0.402769

      min
            1.00000 1.556570e+07 350.000000
                                       18.000000
                                                 0.000000
                                                            0.000000
                                                                      1.000000
                                                                                0.00000
                                                                                          0.000000
                                                                                                     11.580000
                                                                                                               0.000000
                                                           0.000000 1.000000
      25% 2500.75000 1.562853e+07 584.000000 32.000000 3.000000
                                                                               0.00000 0.000000 51002.110000
           5000.50000 1.569074e+07
                             652.000000
                                        37.000000
                                                  5.000000 97198.540000
                                                                       1.000000
                                                                                1.00000
                                                                                          1.000000
                                                                                                  100193.915000
      75% 7500.25000 1.575323e+07 718.00000 44.00000 7.00000 127644.240000 2.00000 1.00000 1.00000 149388.247500 0.000000
      max 10000.00000 1.581569e+07 850.000000 92.000000 10.000000 250898.090000
                                                                      4.000000
                                                                                1.00000
                                                                                          1.000000 199992.480000
                                                                                                               1.000000
[38] data.shape
```

Question-5:

Handle the Missing values:

```
_{
m O_{S}} [39] #Handling the missing values
        data.isnull().sum()
        RowNumber
                            0
        CustomerId
                            0
        Surname
                            0
        CreditScore
                            0
                            0
        Geography
        Gender
                            0
        Age
                            0
        Tenure
        Balance
                            0
        NumOfProducts
                            0
        HasCrCard
                            0
        IsActiveMember
                            0
        EstimatedSalary
        Exited
        dtype: int64
```

Question-6:

Find the outliers and replace the outliers:

```
[ ] #Find the Outliers and replace the outliers

/ [40] sns.boxplot(data['Age'])

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. FutureWarning <a href="mailto:matplotlib.axes._subplots.axesSubplot">matplotlib.axes._subplots.axesSubplot at 0x7f6575aed650></a>
```

(41) qnt=data.quantile(q=[0.25,0.75]) RowNumber CustomerId CreditScore Age Tenure Balance NumOfProducts HasCrCard IsActiveMember EstimatedSalary Exited 🎉 **0.25** 2500.75 15628528.25 584.0 32.0 3.0 0.00 1.0 0.0 0.0 51002.1100 0.75 7500 25 15753233 75 718 0 44 0 7 0 127644 24 20 10 1.0 149388 2475 0.0 [42] IQR = qnt.loc[0.75] - qnt.loc[0.25] IQR 124705.5000 134.0000 12.0000 CustomerId Age Tenure Balance 4.0000 127644.2400 NumOfProducts 1.0000 HasCrCard IsActiveMember 1.0000 EstimatedSalary 98386.1375 Exited dtype: float64 0.0000 [43] upper_extreme = qnt.loc[0.75]+1.5*IQR upper_extreme RowNumber 1.499950e+04 CustomerId 1.594029e+07 CreditScore 9.190000e+02 6.200000e+01 Age Tenure 1.300000e+01 Balance 3.191106e+05 NumOfProducts 3.500000e+00 HasCrCard 2.500000e+00 IsActiveMember 2.500000e+00 EstimatedSalary 2.969675e+05 Exited 0.000000e+00 dtype: float64 [44] lower_extreme = qnt.loc[0.25]-1.5*IQR lower_extreme RowNumber -4.998500e+03 CustomerId 1.544147e+07 CreditScore 3.830000e+02 Age 1.400000e+01 Tenure -3.000000e+00 -1.914664e+05 Balance NumOfProducts -5.000000e-01 HasCrCard -1.500000e+00 IsActiveMember -1.500000e+00 -9.657710e+04 EstimatedSalary Exited 0.000000e+00 dtype: float64

```
[51] df2 = data[(data['Age'] < upper_extreme['Age']) & (data['Age'] > lower_extreme['Age'])]

[50] data.shape
(18000, 14)

[49] df2.shape
(9589, 14)

[52] sns.boxplot(df2['Age'])

//usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. FutureWarning (matplotlib.axes._subplots.AxesSubplot at 0x7f6573caad10>
```

Question-7:

Check for Categorical columns and perform Encoding:

```
[53] #Check for Categorical columns and perform encoding
    #Categorical are Geography and Gender
    from sklearn.preprocessing import LabelEncoder

[75] le=LabelEncoder()
    df2['Geography'] = le.fit_transform(df2['Geography'])
    df2['Gender'] = le.fit_transform(df2['Gender'])
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	0	0	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	2	0	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	0	0	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	0	0	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	850	2	0	43	2	125510.82	1	1	1	79084.10	0

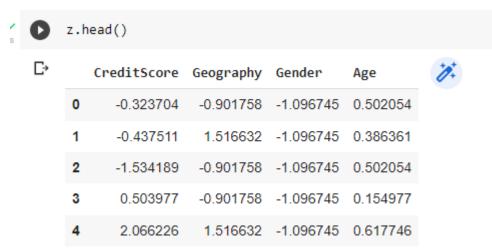
Question-8:

Split the data into dependent and independent variables:

```
[77] #Split the data into dependent and independent variables.
    y=df2['EstimatedSalary']
    x=df2.drop(columns=['EstimatedSalary'],axis=1)
```

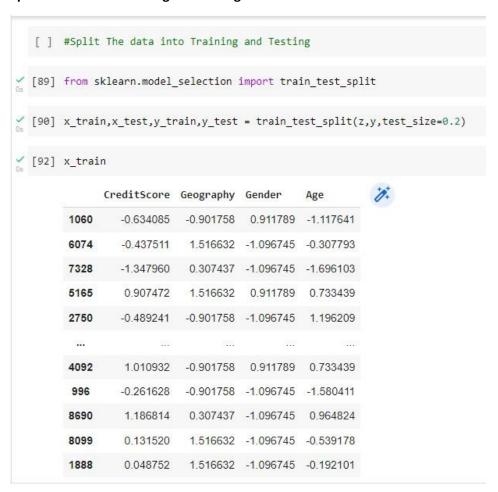
Question-9:

Scale the independent variables:



Question-10:

Split the data into training and testing:



y_train

[→ 1104 151645.96 143463.28 6334 7638 37577.66 5392 43018.82 2851 100478.60 . . . 4269 2048.55 1037 180969.55 9056 166896.01 8440 36864.05 1960 86013.96

Name: EstimatedSalary, Length: 7671, dtype: float64

/ [94] x_test

	CreditScore	Geography	Gender	Age
962	0.772974	0.307437	0.911789	0.154977
5257	1.248890	1.516632	-1.096745	0.386361
7515	-0.841005	0.307437	-1.096745	-0.654871
6844	0.959202	-0.901758	-1.096745	-0.886256
4102	-0.996196	1.516632	-1.096745	0.386361
60	0.379825	0.307437	-1.096745	-1.233333
5555	0.503977	-0.901758	0.911789	-0.076408
5112	1.704115	1.516632	-1.096745	2.237441
138	0.131520	-0.901758	0.911789	-0.423486
4973	0.328095	-0.901758	-1.096745	2.353134

1918 rows × 4 columns

✓ [95] y_test

1002 184023.54 5486 92914.67 7838 132038.65 138780.89 7133 4281 36242.19 61 126494.82 5797 83263.04 5337 38941.44 141 180427.24 706.50 5191

Name: EstimatedSalary, Length: 1918, dtype: float64