

Assignment -2

Data Visualization and Pre-processing

Assignment Date	11 October 2022
Student Name	S.HARITHA
Student Roll Number	111619104028
Maximum Marks	2 Marks

Question-1:

Download the dataset:

The dataset "Churn_Modelling.csv" was downloaded Successfully

Question-2:

Load the Dataset:

```
[ ] #Load the dataset

[1] from google.colab import drive
drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

[84] import pandas as pd
import numpy as np
import sklearn
import matplotlib.pyplot as plt

[3] data = pd.read_csv("/content/drive/MyDrive/Churn_Modelling.csv")

[4] data.head()
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101345.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	602	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

Question-3:

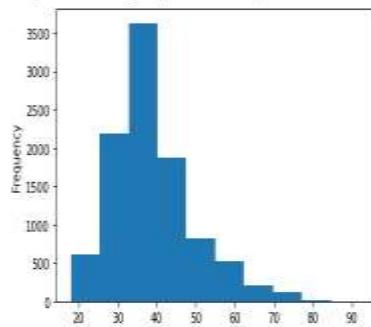
Perform Below Visualization:

Univariate Analysis

[5] #Univariate Analysis for Numerical data

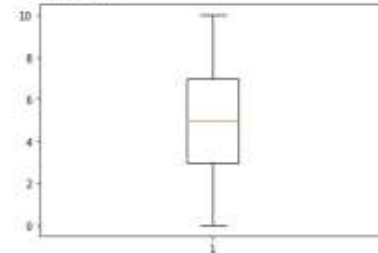
```
[6] #Histogram  
data['Age'].plot(kind='hist')
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f65a0462590>



```
[9] #Box Plot  
plt.boxplot(data['Tenure'])
```

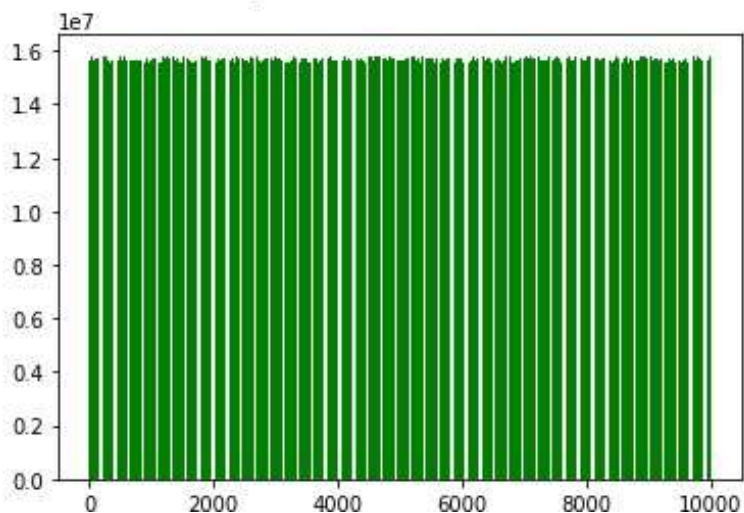
{'whiskers': [
 <matplotlib.lines.Line2D at 0x7f659f8e82d0>,
 <matplotlib.lines.Line2D at 0x7f659f8e8810>],
 'caps': [
 <matplotlib.lines.Line2D at 0x7f659f8e0d50>,
 <matplotlib.lines.Line2D at 0x7f659f8e62d0>],
 'boxes': [
 <matplotlib.lines.Line2D at 0x7f659f059d10>],
 'medians': [
 <matplotlib.lines.Line2D at 0x7f659f8e0850>],
 'fliers': [
 <matplotlib.lines.Line2D at 0x7f659f8e6d90>],
 'means': []}



[] #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>

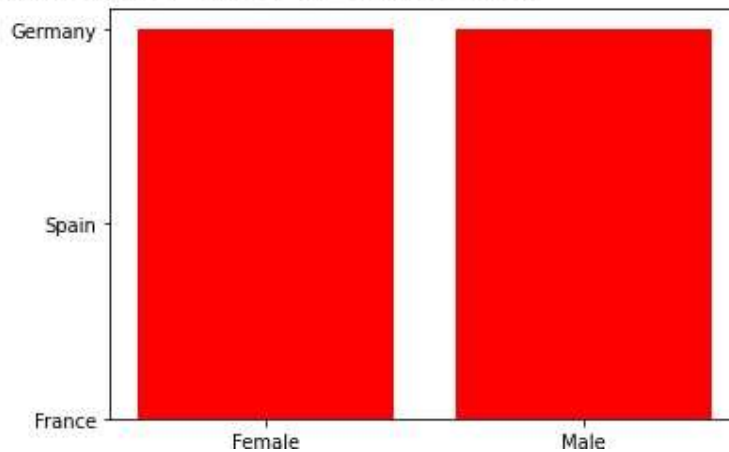


Bivariate Analysis

✓
19s

```
[23] #Bivariate Analysis for Categorical Data  
#Stacked Bar chart  
plt.bar(data['Gender'], data['Geography'], color='r')
```

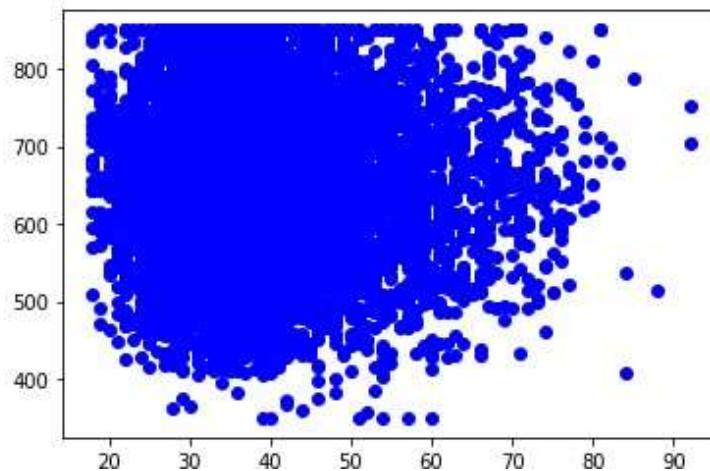
<BarContainer object of 10000 artists>



✓
0s

```
[21] #Bivariate Analysis for Numerical Data  
plt.scatter(data['Age'], data['CreditScore'], color='b')
```

<matplotlib.collections.PathCollection at 0x7f6589f606d0>



Multivariate Analysis

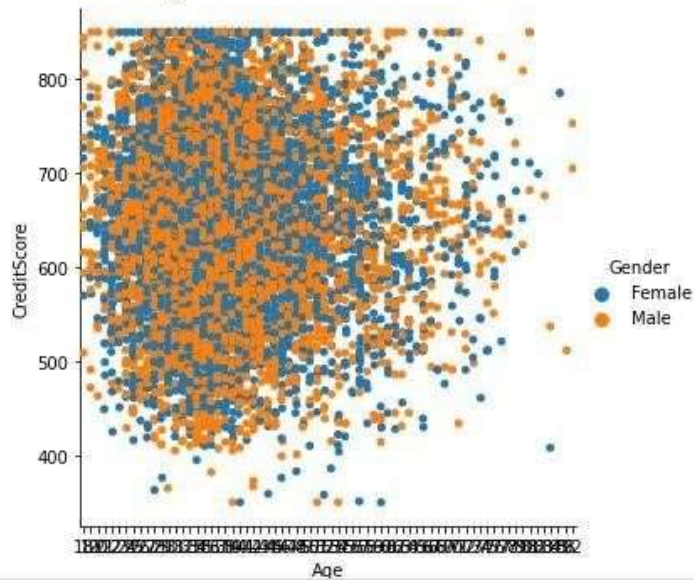
✓
16s



```
#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")
```



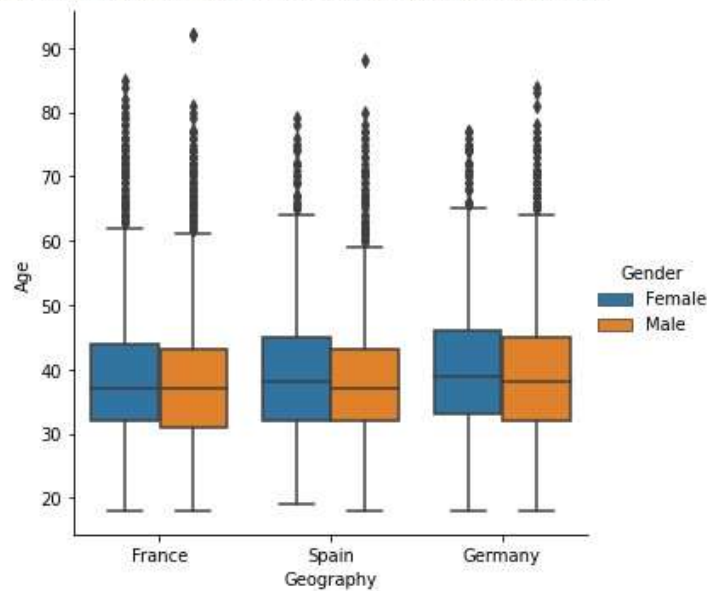
```
<seaborn.axisgrid.FacetGrid at 0x7f657aab5d90>
```



✓
1s

```
[32] #Multivariate Analysis for 2 Categorical and 1 Numerical Data  
#Box Plot  
sns.catplot(data=data, x="Geography", y="Age", hue="Gender", kind="box")
```

```
<seaborn.axisgrid.FacetGrid at 0x7f6575c43490>
```



Question-4:

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      5.000500e+03
CustomerId      1.569094e+07
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
Balance        9.719854e+04
NumOfProducts  1.000000e+00
HasCrCard      1.000000e+00
IsActiveMember 1.000000e+00
EstimatedSalary 1.001939e+05
Exited         0.000000e+00
dtype: float64
```

```
[36] data.describe()
```

	RowNumber	CustomerId	Creditscore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
count	10000.00000	1.002000e+04	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.00000	10000.000000	10000.000000	10000.000000
mean	5000.50000	1.569094e+07	650.520000	38.921800	5.012800	76485.889280	1.530200	0.705500	0.515100	100090.230601	0.203700
std	2086.89588	7.193819e+04	96.653299	10.467006	2.892174	82397.405202	0.581654	0.455040	0.499797	57510.492818	0.402768
min	1.00000	1.554575e+07	350.000000	18.000000	0.000000	0.000000	1.000000	0.000000	0.000000	11.580000	0.000000
25%	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	0.000000	1.000000	0.000000	0.000000	51962.110000	0.000000
50%	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	97196.340000	1.000000	1.000000	1.000000	106193.915000	0.000000
75%	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	127644.240000	2.000000	1.000000	1.000000	149386.247500	0.000000
max	10000.00000	1.581509e+07	850.000000	92.000000	10.000000	256086.090000	4.000000	1.000000	1.000000	199992.480000	1.000000

```
[38] data.shape
```

```
(10000, 14)
```

Question-5:

Handle the Missing values:

```
✓ [39] #Handling the missing values
```

```
data.isnull().sum()
```

```
RowNumber      0
CustomerId      0
Surname         0
CreditScore     0
Geography       0
Gender          0
Age            0
Tenure          0
Balance         0
NumOfProducts  0
HasCrCard       0
IsActiveMember  0
EstimatedSalary 0
Exited          0
dtype: int64
```

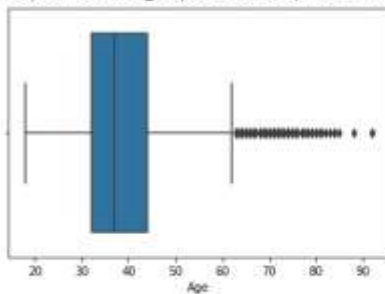
Question-6:

Find the outliers and replace the outliers:

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

```
[48] 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
<matplotlib.axes._subplots.AxesSubplot at 0x7f6575aed650>
```




```
[41] qnt=data.quantile(q=[0.25,0.75])
qnt
```

	RowNumber	CustomerId	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0.25	2800.75	15628528.25	564.0	32.0	3.0	0.00	1.0	0.0	0.0	51002.1100	0.0
0.75	7500.25	15753233.75	718.0	44.0	7.0	127644.24	2.0	1.0	1.0	149388.2475	0.0

```
[42] IQR = qnt.loc[0.75] - qnt.loc[0.25]
IQR
```

RowNumber	4090.5000
CustomerId	124785.5000
CreditScore	134.0000
Age	12.0000
Tenure	4.0000
Balance	127644.2400
NumOfProducts	1.0000
HasCrCard	1.0000
IsActiveMember	1.0000
EstimatedSalary	99386.1375
Exited	0.0000
dtype:	float64

```
[43] upper_extreme = qnt.loc[0.75]+1.5*IQR
upper_extreme
```

RowNumber	1.499950e+04
CustomerId	1.594029e+07
CreditScore	9.190000e+02
Age	6.200000e+01
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
Balance	-1.914664e+05
NumOfProducts	-5.000000e-01
HasCrCard	-1.500000e+00
IsActiveMember	-1.500000e+00
EstimatedSalary	-9.657710e+04
Exited	0.000000e+00
dtype:	float64

```

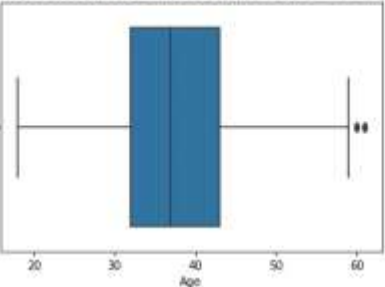
[51] df2 = data[(data['Age'] < upper_extreme['Age']) & (data['Age'] > lower_extreme['Age'])]

[58] 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 0x7f6573caad18>

```



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'])

```

```
[76] df2.head()
```

	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	93626.63	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)
```

✓ [78] x.head()

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	Exited
0	1	15634502	Hargrave	619	0	0	42	2	0.00	1	1	1	1
1	2	15647311	Hill	608	2	0	41	1	83807.86	1	0	1	0
2	3	15619304	Otero	502	0	0	42	8	159660.80	3	1	0	1
3	4	15701354	Boni	699	0	0	39	1	0.00	2	0	0	0
4	5	15737866	Mitchell	850	2	0	43	2	125510.82	1	1	1	0

Question-9:

Scale the independent variables:

✓ [80] #Scale the independent variables
0s names=z.columns
names


```
Index(['CreditScore', 'Geography', 'Gender', 'Age'], dtype='object')
```

✓ [81] from sklearn.preprocessing import scale
0s

✓ [82] z=scale(z)
0s z

```
array([[ -0.32370448, -0.90175758, -1.09674455,  0.50205394],  
       [ -0.43751069,  1.51663241, -1.09674455,  0.38636147],  
       [ -1.53418871, -0.90175758, -1.09674455,  0.50205394],  
       ...,  
       [  0.60743723, -0.90175758, -1.09674455, -0.19210091],  
       [  1.25923643,  0.30743742,  0.91178935,  0.50205394],  
       [  1.46615681, -0.90175758, -1.09674455, -1.1176407 ]])
```

✓ [87] z = pd.DataFrame(z, columns =names)
0s

 `z.head()`

	CreditScore	Geography	Gender	Age
0	-0.323704	-0.901758	-1.096745	0.502054
1	-0.437511	1.516632	-1.096745	0.386361
2	-1.534189	-0.901758	-1.096745	0.502054
3	0.503977	-0.901758	-1.096745	0.154977
4	2.066226	1.516632	-1.096745	0.617746

Question-10:

Split the data into training and testing:

```
[ ] #Split The data into Training and Testing
```

```
[89] from sklearn.model_selection import train_test_split
```

```
[90] x_train,x_test,y_train,y_test = train_test_split(z,y,test_size=0.2)
```

```
[92] x_train
```

	CreditScore	Geography	Gender	Age
1060	-0.634085	-0.901758	0.911789	-1.117641
6074	-0.437511	1.516632	-1.096745	-0.307793
7328	-1.347960	0.307437	-1.096745	-1.696103
5165	0.907472	1.516632	0.911789	0.733439
2750	-0.489241	-0.901758	-1.096745	1.196209
...
4092	1.010932	-0.901758	0.911789	0.733439
996	-0.261628	-0.901758	-1.096745	-1.580411
8690	1.186814	0.307437	-1.096745	0.964824
8099	0.131520	1.516632	-1.096745	-0.539178
1888	0.048752	1.516632	-1.096745	-0.192101

```
✓ y_train
1104    151645.96
6334    143463.28
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 x 4 columns

✓ [95] y_test

```
1002    184023.54
5486     92914.67
7838    132038.65
7133    138780.89
4281     36242.19
```

```
...
61      126494.82
5797     83263.04
5337     38941.44
141     180427.24
5191       706.50
```

```
Name: EstimatedSalary, Length: 1918, dtype: float64
```