Assignment -2

| Assignment Date | 26 September 2022 | | | | |
|---------------------|-------------------|--|--|--|--|
| Student Name | Boomika V.G | | | | |
| Student Roll Number | 621319106009 | | | | |
| Maximum Marks | 2 Marks | | | | |

Question-1 Download the dataset:

```
Importing Libraries

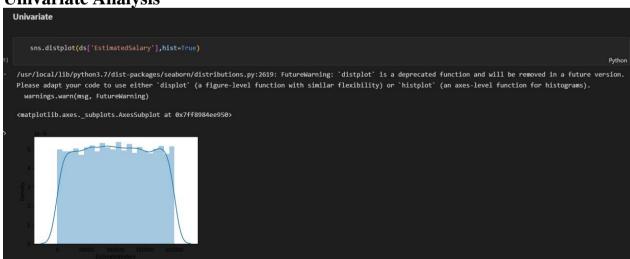
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

Question-2 Load the dataset

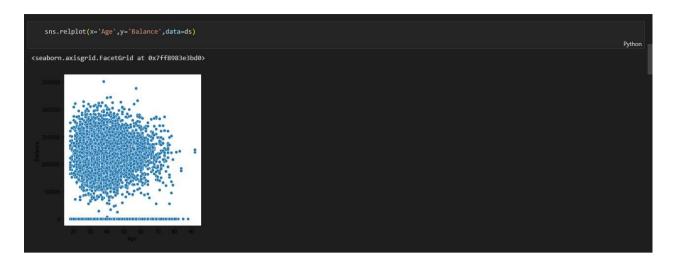


Question-3 perform below visualization

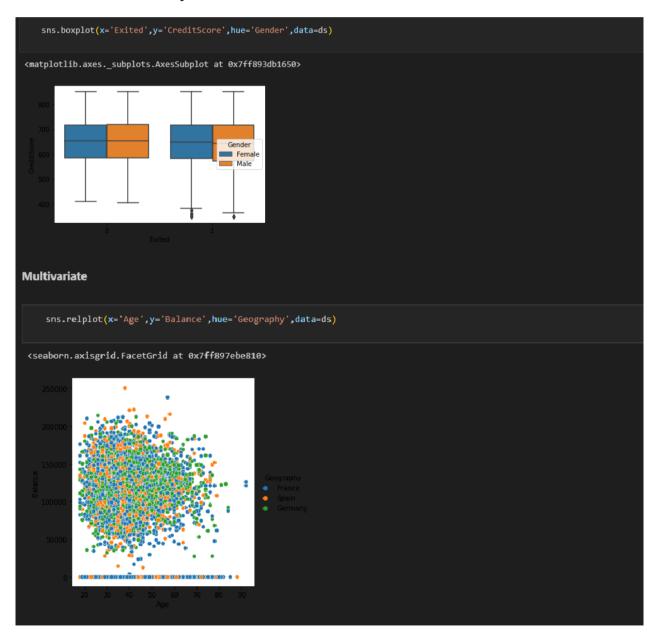
• Univariate Analysis



• Bi-variate Analysis



• Multi-variate Analysis

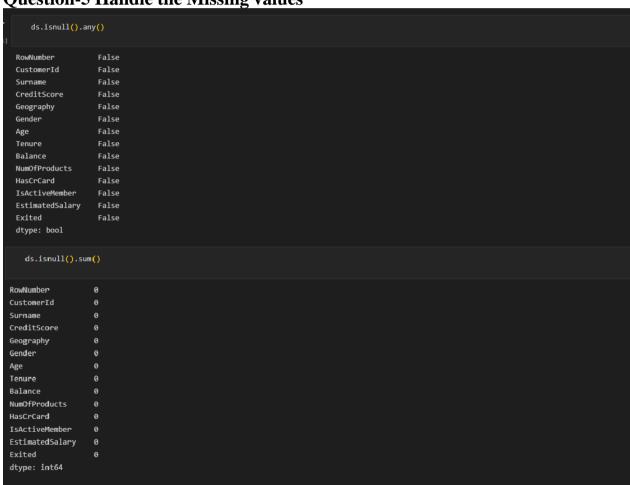


Question-4 perform descriptive statistics on the dataset.



| mean 5000.50000 1.569094e+07 650.528800 38.921800 5.012800 76485.889288 1.530200 0.70550 0.515100 100090.239881 0.2037 std 2886.89568 7.193619e+04 96.653299 10.487806 2.892174 62397.405202 0.581654 0.45584 0.499797 57510.492818 0.40276 min 1.00000 1.556570e+07 350.00000 18.00000 0.000000 1.000000 0.00000 0.00000 1.500000 0.00000 11.580000 0.00000 25% 2500.75000 1.562853e+07 584.000000 37.00000 5.00000 97198.540000 1.00000 | | RowNumber | CustomerId | CreditScore | Age | Tenure | Balance | NumOfProducts | HasCrCard | IsActiveMember | EstimatedSalary | Exited |
|--|-------|-------------|--------------|--------------|--------------|--------------|---------------|---------------|-------------|----------------|-----------------|--------------|
| std 2886.89568 7.193619e+04 96.653299 10.487806 2.892174 62397.405202 0.581654 0.45584 0.499797 57510.492818 0.40276 min 1.00000 1.556570e+07 350.000000 18.000000 0.000000 1.000000 0.00000 0.000000 11.580000 0.00000 25% 2500.75000 1.562853e+07 584.000000 32.000000 3.000000 0.000000 1.000000 0.00000 51002.110000 0.00000 50% 5000.50000 1.569074e+07 652.000000 37.000000 5.000000 97198.540000 1.000000 1.00000 1.00000 1.000000 1.00000 1.000000 1.4938.247500 0.00000 | count | 10000.00000 | 1.000000e+04 | 10000.000000 | 10000.000000 | 10000.000000 | 10000.000000 | 10000.000000 | 10000.00000 | 10000.000000 | 10000.000000 | 10000.000000 |
| min 1.00000 1.556570e+07 350.000000 18.00000 0.000000 0.000000 1.000000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 0.00000 51002.110000 0.00000 50% 5000.50000 1.569074e+07 652.000000 37.000000 5.000000 97198.540000 1.000000 1.00000 1.00000 1.00000 1.00000 1.00000 1.00000 1.00000 1.000000 1.00000 1.00000 1.00000 1.00000 1.000000 1.000000 1.00000 1.00000 1.00000 < | mean | 5000.50000 | 1.569094e+07 | 650.528800 | 38.921800 | 5.012800 | 76485.889288 | 1.530200 | 0.70550 | 0.515100 | 100090.239881 | 0.203700 |
| 25% 2500.75000 1.562853e+07 584.000000 32.000000 3.000000 0.000000 1.000000 0.00000 0.00000 51002.110000 0.00000 50% 5000.50000 1.569074e+07 652.00000 37.000000 5.00000 97198.540000 1.000000 1.000000 1.00000 </td <td>std</td> <td>2886.89568</td> <td>7.193619e+04</td> <td>96.653299</td> <td>10.487806</td> <td>2.892174</td> <td>62397.405202</td> <td>0.581654</td> <td>0.45584</td> <td>0.499797</td> <td>57510.492818</td> <td>0.402769</td> | std | 2886.89568 | 7.193619e+04 | 96.653299 | 10.487806 | 2.892174 | 62397.405202 | 0.581654 | 0.45584 | 0.499797 | 57510.492818 | 0.402769 |
| 50% 5000.50000 1.569074e+07 652.000000 37.000000 5.000000 97198.540000 1.000000 1.00000 1.00000 1.00000 1.00000 0.00000 75% 7500.25000 1.575323e+07 718.000000 44.000000 7.000000 127644.240000 2.000000 1.00000 1.00000 149388.247500 0.00000 | min | 1.00000 | 1.556570e+07 | 350.000000 | 18.000000 | 0.000000 | 0.000000 | 1.000000 | 0.00000 | 0.000000 | 11.580000 | 0.000000 |
| 75% 7500.25000 1.575323e+07 718.000000 44.000000 7.000000 127644.240000 2.000000 1.00000 1.00000 149388.247500 0.00000 | 25% | 2500.75000 | 1.562853e+07 | 584.000000 | 32.000000 | 3.000000 | 0.000000 | 1.000000 | 0.00000 | 0.000000 | 51002.110000 | 0.000000 |
| | 50% | 5000.50000 | 1.569074e+07 | 652.000000 | 37.000000 | 5.000000 | 97198.540000 | 1.000000 | 1.00000 | 1.000000 | 100193.915000 | 0.000000 |
| may 10000 00000 1 581569a+07 850 000000 92 000000 10 000000 250898 090000 4 000000 1 00000 1 000000 1 000000 1 000000 | 75% | 7500.25000 | 1.575323e+07 | 718.000000 | 44.000000 | 7.000000 | 127644.240000 | 2.000000 | 1.00000 | 1.000000 | 149388.247500 | 0.000000 |
| 1100000000 1.301303C101 030.000000 3E.000000 10.000000 E.0000000 1.00000 1.00000 1.300000 1.300000 1.300000 1.300000 | max | 10000.00000 | 1.581569e+07 | 850.000000 | 92.000000 | 10.000000 | 250898.090000 | 4.000000 | 1.00000 | 1.000000 | 199992.480000 | 1.000000 |

Question-5 Handle the Missing values



Question-6 Find the outliers and replace the outliers

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is
deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.
""Entry point for launching an IPython kernel.
RowNumber
                    0.000000
CustomerId
                   0.001149
CreditScore
                   -0.071607
                    1.011320
Age
Tenure
                   0.010991
NumOfProducts
                   -0.901812
IsActiveMember
                   -0.060437
EstimatedSalary
                  0.002085
Exited
                    1.471611
dtype: float64
```

```
sns.boxplot(ds["Age"])
 /usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword 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
 <matplotlib.axes._subplots.AxesSubplot at 0x7ff893caec90>
        q0 = ds["Age"].describe()["25%"]
q1 = ds["Age"].describe()["75%"]
        iqr=q1-q0

lb = q0 -(1.5*iqr)

ub = q1 + (1.5*iqr)
        ds[ds["Age"]<lb]
ds[ds["Age"]>ub]
        #Replacing the outlier
outlier_list = list(ds[ds["Age"] > ub]["Age"])
        print(outlier list)
 [66, 75, 65, 73, 65, 72, 67, 67, 79, 80, 68, 75, 66, 66, 70, 63, 72, 64, 64, 70, 67, 82, 63, 69, 65, 69, 64, 65, 74, 67, 66, 67, 63, 70, 71, 72, 67, 74, 76, 66,
 64, 67, 70, 68, 72, 71, 66, 75, 67, 73, 69, 76, 63, 85, 67, 74, 76, 66, 69, 66, 72, 63, 71, 63, 74, 67, 72, 72, 66, 84, 71, 66, 63, 74, 69, 84, 67, 64, 68, 66,
 77, 70, 67, 79, 67, 76, 73, 66, 67, 64, 73, 76, 72, 64, 71, 63, 70, 65, 66, 65, 80, 66, 63, 63, 63, 63, 63, 64, 69, 63, 64, 76, 75, 68, 69, 77, 64, 66, 74, 71,
 67, 68, 64, 68, 78, 64, 75, 66, 64, 78, 65, 74, 64, 64, 71, 77, 79, 70, 81, 64, 68, 68, 63, 79, 66, 64, 70, 69, 71, 72, 66, 68, 63, 71, 72, 72, 66, 68, 63, 71, 72, 72, 66, 68, 63, 71, 72, 72, 66, 68, 73, 75, 65,
 65, 67, 63, 68, 71, 73, 64, 66, 71, 69, 71, 66, 76, 69, 73, 64, 64, 75, 73, 71, 72, 63, 67, 68, 73, 67, 64, 63, 92, 65, 75, 67, 71, 64, 66, 64, 66, 67, 77, 92,
 67, 64, 81, 73, 63, 67, 74, 83, 69, 71, 78, 63, 70, 69, 72, 70, 63, 74, 80, 69, 72, 67, 76, 71, 67, 71, 78, 63, 63, 68, 64, 70, 78, 69, 68, 64, 64, 77, 77]
       outlier_dict = {}.fromkeys(outlier_list,ub)
print(outlier_dict)
 {66: 62.0, 75: 62.0, 65: 62.0, 73: 62.0, 72: 62.0, 67: 62.0, 79: 62.0, 80: 62.0, 68: 62.0, 79: 62.0, 63: 62.0, 64: 62.0, 82: 62.0, 82: 62.0, 69: 62.0, 71: 62.0, 71: 62.0, 72: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 62.0, 73: 6
 76: 62.0, 77: 62.0, 88: 62.0, 85: 62.0, 84: 62.0, 78: 62.0, 81: 62.0, 92: 62.0, 83: 62.0}
      ds["Age"] = ds["Age"].replace(outlier_dict)
sns.boxplot(ds["Age"])
 /usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword 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.
 <matplotlib.axes. subplots.AxesSubplot at 0x7ff8987b26d0>
```

Question -7 Check for Categorical columns and perform encoding.

```
Check for Categorical columns and perform encoding

from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
ct=ColumnTransformer([('oh',OneHotEncoder(),[1,2])],remainder='passthrough')
x=ct.fit_transform(x)
print(x.shape)

(10000, 13)

# saving the data
import joblib
joblib.dump(ct,"churnct.pkl")
```

Question-8 Split the data into dependent and independent variables.

```
x=ds.iloc[:,3:13].values
print(x.shape)
y=ds.iloc[:,13:14].values
print(y.shape)

(10000, 10)
(10000, 1)
```

Question-10 Split the data into training and testing

```
from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_state=0)
    print(x_train.shape)
    print(x_test.shape)

(8000, 13)

(2000, 13)

from sklearn.preprocessing import StandardScaler
    sc-StandardScaler()
    x_train=sc.fit_transform(x_train)
    x_test=sc.transform(x_test)
    joblib.dump(sc,"churnsc.pkl")

['churnsc.pkl']
```