Assignment - 2

Python Programming

Assignment Date	21/09/2022
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Student Roll Number	110519106002
Maximum Mark	2 Mark

```
In [1]: import os
    os.chdir('drive/MyDrive/Nalaiya Thiran')
```

1.Download dataset

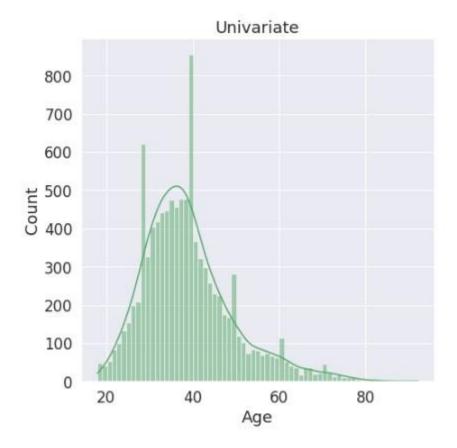
2.Load dataset

	data	= pd.read_0	SV CHAIL	ioucilling.	,										
3		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	11	11	1	101348.88	
	1	2	15647311	Hill	608	Spain	Female	41	1	83807,86	1	0	1	112542.58	(
	2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	
	3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	
	4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	10
				-			***		***		***			***	
	9995	9996	15606229	Obijiaku	771	France	Male	39	5	0.00	2	1	0	96270.64	- 9
	9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	101699.77	(
	9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	42085.58	
	9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	75075.31	2	- 1	0	92888.52	
	9999	10000	15628319	Walker	792	France	Female	28	4	130142.79	1	1	0	38190.78	(

3.Perform Below Visualizations

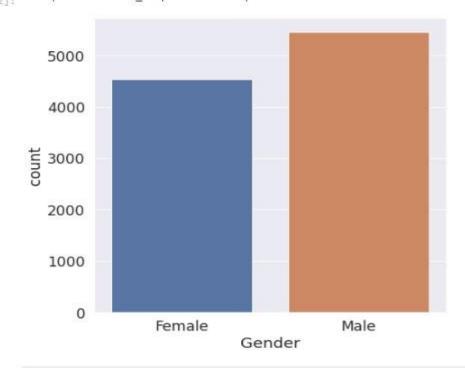
- 1. Univariate Analysis
- 2. Bi-Variate Analysis
- 3. Multi-variate Analysis

```
In [35]: #univariate analysis
import seaborn as sns
import matplotlib.pyplot as plt
#myplt = plt.hist(data["Age"])
sns.histplot(data["Age"],kde=True,color='g')
plt.title("Univariate")
Out[35]: Text(0.5, 1.0, 'Univariate')
```



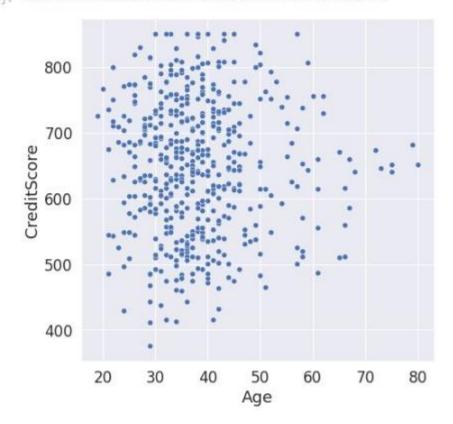
In [42]: #bivariate analysis 1
sns.countplot(x='Gender', data = data)

Out[42]: <matplotlib.axes._subplots.AxesSubplot at 0x7f30022e0610>

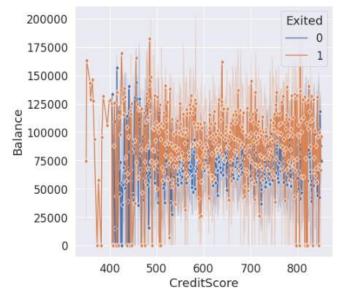


```
In [54]: #bivariate analysis 2
    df = data.head(500)
    sns.scatterplot(x='Age',y='CreditScore', data=df)
```

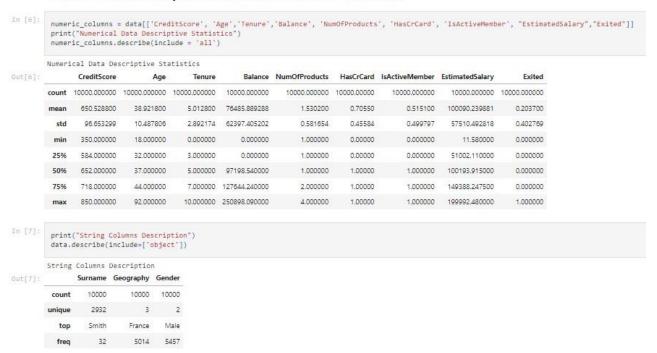
Out[54]: <matplotlib.axes._subplots.AxesSubplot at 0x7f2ffadce150>







4. Perform descriptive statistics on the dataset.



5. Handle the Missing values

```
In [57]: data.isnull().sum().sum()
Out[57]: 0
In [56]: data['Tenure'].isna().sum()
Out[56]: 0
```

6. Find the outliers and replace the outliers

Age



```
In [184_
            df2
Out[184_
                  Age
               0
                   42
           1 41
            3 39
               4 43
            9995 39
            9996 35
            9997
                   36
            9998 42
            9999 28
           10000 rows × 1 columns
            #upper_extreme = q3+1.5*IQR
#Lower_extreme = q1-1.5*IQR
#IQR = q3-q1
            qnt = df2.quantile(q=[0.25,0.75])
             Age
Out[117_
            0.25 32.0
            0.75 44.0
In [186_
            IQR = qnt.loc[0.75] - qnt.loc[0.25]
            upper_extreme = qnt.loc[0.75]+1.5*IQR
lower_extreme = qnt.loc[0.25]-1.5*IQR
            lower_extreme
           Age 14.0
Out[106...
           dtype: float64
 In [84]:
            upper_extreme
            #now we got the upper extreme which is the outlier so replace it now
           Age 62.0
Out[84]:
           dtype: float64
In [191_
            med = df2['Age'].median()
            med
           37.0
In [185_
            upper_extreme
           Age 62.0
Out[185_
           dtype: float64
In [195_
            #treating outliers using capping
            #here 14 is Lower_extreme
            #62 is upper_extreme
            import numpy as np
            print(df2['Age'])
df2["New_Age"] = df2["Age"].map(
              lambda x: med
                if x < 14 else x)
            df2["New_Age"] = df2['Age'].map(lambda x:med
                                              if x > 62 else x)
            df2
```

9

42

```
Ð
     42
      41
2
      42
3
       39
4
       43
      39
9995
9996
9997
      35
       36
      42
9998
9999
      28
Name: Age, Length: 10000, dtype: int64
```

Out[195...

	Age	New_Age
0	42	42.0
-1	41	41.0
2	42	42.0
3	39	39.0
4	43	43.0
	4	-
9995	39	39.0
9996	35	35.0
9997	36	36.0
9998	42	42.0
9999	28	28.0

10000 rows × 2 columns

7. Check for categorical columns and perform encoding

```
In [125...
            #extracting categorical columns
print(data['Geography'].unique())
            print(data['Gender'].unique())
           ['France' 'Spain' 'Germany']
['Female' 'Male']
In [122...
            data2 = data.copy()
            data2['Gender'].replace(['Female', 'Male'],[0,1],inplace=True)
            data2['Geography'].replace(['France', 'Spain', 'Germany'],[0,1,2],inplace=True)
                 RowNumber CreditScore Geography Gender Age Tenure
                                                                           Balance EstimatedSalary
              0
                                                                               0.00
                                     619
                                                              42
                                                                                          101348.88
                                                                       1 83807.86
                                                                       8 159660.80
                                                                                         113931.57
                           3
                           4
                                     699
                                                  0
                                                          0
                                                              39
                                                                             0.00
                                                                                         93826.63
                                     850
                                                          0 43
                                                                       2 125510.82
                                                                                          79084 10
           9995
                        9996
                                                              39
                                                                              0.00
                                                                                          96270.64
           9996
                        9997
                                     516
                                                              35
                                                                      10 57369.61
                                                                                          101699.77
           9997
                        9998
                                     709
                                                  0
                                                          0
                                                             36
                                                                               0.00
                                                                                          42085.58
           9998
                        9999
                                     772
                                                              42
                                                                       3 75075.31
                                                                                          92888.52
                       10000
                                                                       4 130142.79
                                                                                          38190.78
          10000 rows × 8 columns
        d2 = pd.get_dummies(data,columns = ['Geography','Gender'])
            RowNumber Customerld Surname CreditScore Age Tenure Balance NumOffProducts HasCrCard IsActiveMember EstimatedSalary Exited Geography_France Geogra
                 1 15634602 Hargrave
                                                    2 0.00
                2 15647311 Hill
                                       608 41 1 83807.86
                   3 15619304
                                          502 42
                                                                                                     113931.57
                 4 15701354 Boni 699 39 1 0.00
                                                                                             0 93826.63
                    5 15737888 Mitchell
                                          850 43
                                                      2 125510.82
                                                                                                      79084.10
        9995
                 9996 15606229 Obijiaku
                                           771 39
                                                     5 0.00
                                                                          2
                                                                                               0
                                                                                                      96270 64
        9996
                 9997 15569892 Johnstone
                                         516 35 10 57369.61
                                                                                                     101699.77
                                                                                                      42085.58
        9997
                 9998
                      15584532
                                 Liu
                                           709 36
                                                            0.00
                                                                                  0
                 9999 15682355 Sabbatini
                                          772 42 3 75075.31
                                                                                                      92888.52 1
                                                                                                                             0
        9998
                 10000 15628319 Walker
                                           792 28
                                                      4 130142.79
                                                                                                      38190.78
       10000 rows × 17 columns
```

4

8. Split the data into dependent and independent variables.

```
n [138.
                       col= data.iloc[:,:-1].columns
                        print(col)
                        #dependant variable
                        y = data.iloc[:,-1:].columns
                        print(y)
                     dtype='object')
                      Index(['Exited'], dtype='object')
n [139_
                      dep_data_var = data[col]
                       print(dep_data_var)
                       print(data[y])
                                  RowNumber CustomerId Surname CreditScore Geography Gender Age \

        owNumber
        CustomerId
        Surname
        CreditScore Geography
        Gender Age

        1
        15634602
        Hargrave
        619
        France
        Female
        42

        2
        15647311
        Hill
        608
        Spain
        Female
        41

        3
        15619304
        Onio
        502
        France
        Female
        42

        4
        15701354
        Boni
        699
        France
        Female
        39

        5
        15737888
        Mitchell
        850
        Spain
        Female
        43

        ...
        ...
        ...
        ...
        ...
        ...
        ...

        9996
        15606229
        Obijiaku
        771
        France
        Male
        39

        9997
        1550892
        Johnstone
        516
        France
        Male
        35

        9998
        15584532
        Liu
        709
        France
        Female
        36

        9999
        15682355
        Sabbatini
        772
        Germany
        Male
        42

        10000
        15628319
        Walker
        792
        <t
                      4
                      9995
                     9996
                     9997
                      9998
                               Tenure Balance NumOfProducts HasCrCard IsActiveMember \
                                        2 0.00
1 83807.86
8 159660.80
                                                                                                              1
                 1
                                           1 0.00
2 125510.82
                                                    0.00
                 9995
                                         10 57369.61
                 9996
                                      7 0.00
3 75075.31
4 130142.79
                 9998
                 9999
                           EstimatedSalary
                                        101348.88
112542.58
                 7
                                              113931.57
                                              93826.63
79084.10
                                                 96270.64
                                      96270.64
101699.77
42085.58
92888.52
                 9998
                 9999
                                                 38190.78
                 [10000 rows x 13 columns]
                            Exited
                                             0
                 4
                                           0
                 9996
                 9997
                 9998
                 [10000 rows x 1 columns]
```

9. Scale the independent variables

```
import pandas as pd
            import matplotlib.pyplot as plt
            # Import StandardScaler
            from sklearn.preprocessing import StandardScaler
            fig, ax = plt.subplots(figsize=(12, 4))
            cols = ['CreditScore', 'Tenure', 'EstimatedSalary']
            scaler = StandardScaler()
            x_std = scaler.fit_transform(data[cols])
            ax.hist(x_std[:,0])
            ax.hist(x_std[:,1])
Out[157._ (array([ 413., 1035., 1048., 1009., 2001.,
                                                             0., 1995.,
                    1474.]),
            array([-1.73331549, -1.38753759, -1.04175968, -0.69598177, -0.35020386, -0.00442596, 0.34135195, 0.68712986, 1.03290776, 1.37868567,
                     1.72446358]),
            <a list of 10 Patch objects>)
            2000
            1500
            1000
             500
                 0
                                           -2
                                                             -1
                                                                                0
                         -3
```

10.Split the data into training and testing

```
from sklearn.model_selection import train_test_split
    x = data['Age'].values
    print(x)
    y = data.iloc[:,-1].values
    print(y)
    X_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.33, random_state=42)
    print(X_train)
    print("Total Training data ",len(X_train))

[42 41 42 ... 36 42 28]
    [1 0 1 ... 1 1 0]
    [29 37 49 ... 38 43 51]
    Total Training data 6700
```