Assignment - 2
Data Visualization and pre-processing

Assignment submission	26 September 2022
Student Name	Rajaguru G
Student Roll Number	951919CS077
Maximum Marks	2 Marks

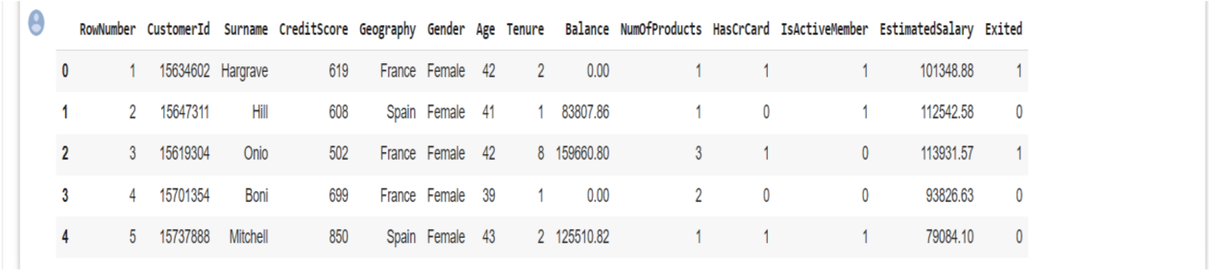
1. Download the Dataset

2. Import required library

```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
sns.set_style('darkgrid')
sns.set(font_scale=1.3)
```

3. Read dataset and do pre-processing

```
df=pd.read_csv("/content/drive/MyDrive/IBM/Assignment - 2 /Churn_Modelling.csv")
df.head()
```

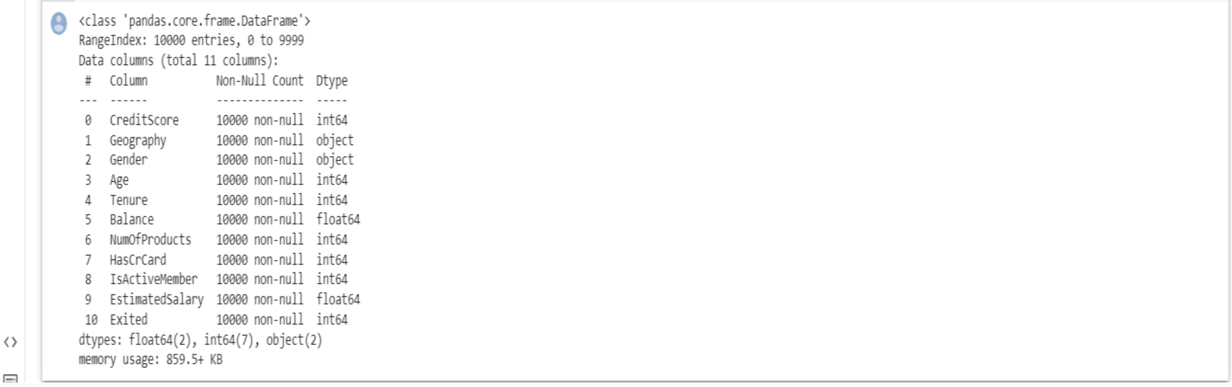


	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	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

Drop the columns that are not required for the neural network.

```
df.drop(["RowNumber","CustomerId","Surname"],axis=1,inplace=True)
```

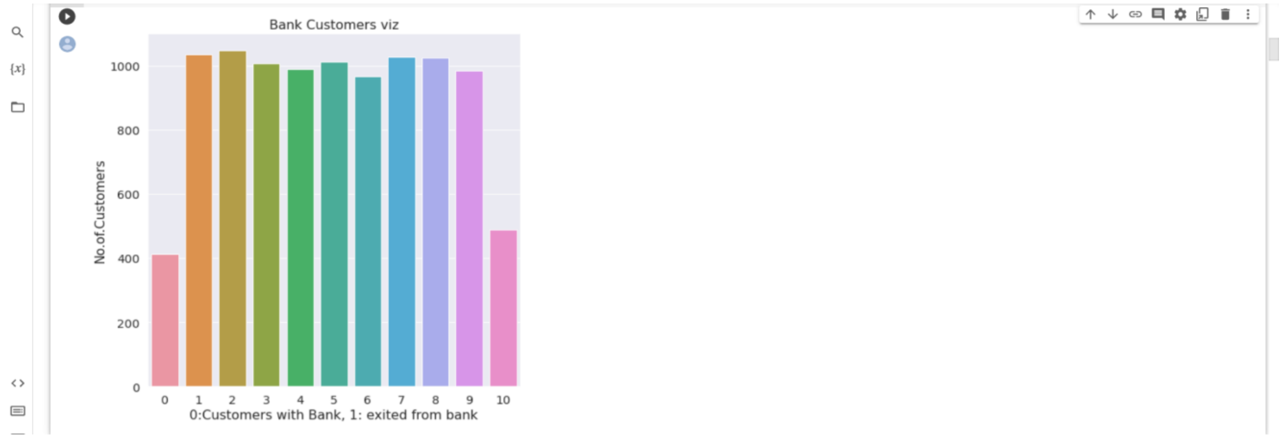
```
df.info()
```



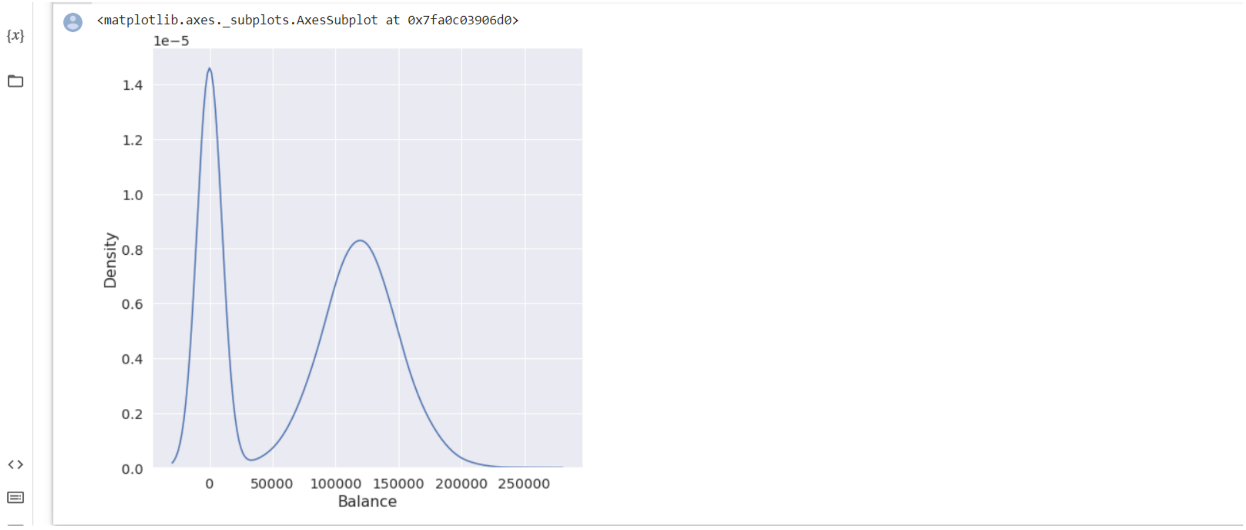
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 11 columns):
 #   Column           Non-Null Count  Dtype  
---  ------           --------------  -----  
 0   CreditScore      10000 non-null  int64  
 1   Geography        10000 non-null  object 
 2   Gender           10000 non-null  object 
 3   Age              10000 non-null  int64  
 4   Tenure           10000 non-null  int64  
 5   Balance          10000 non-null  float64
 6   NumOfProducts    10000 non-null  int64  
 7   HasCrCard        10000 non-null  int64  
 8   IsActiveMember   10000 non-null  int64  
 9   EstimatedSalary  10000 non-null  float64
 10  Exited           10000 non-null  int64  
dtypes: float64(2), int64(7), object(2)
memory usage: 859.5+ KB
```

4.A. Perform Univariate Analysis

```
plt.figure(figsize=(8,8))
sns.countplot(x='Tenure',data=df)
plt.xlabel('0:Customers with Bank, 1: exited from bank')
plt.ylabel('No.of.Customers')
plt.title("Bank Customers viz")
plt.show()
```



```
plt.figure(figsize=(8,8))
sns.kdeplot(x=df['Balance'])
```

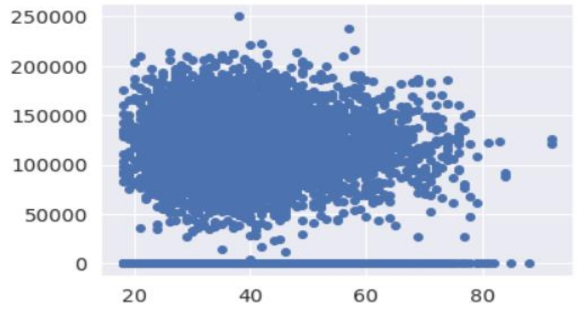


```
<matplotlib.axes._subplots.AxesSubplot at 0x7fa0c03906d0>
```

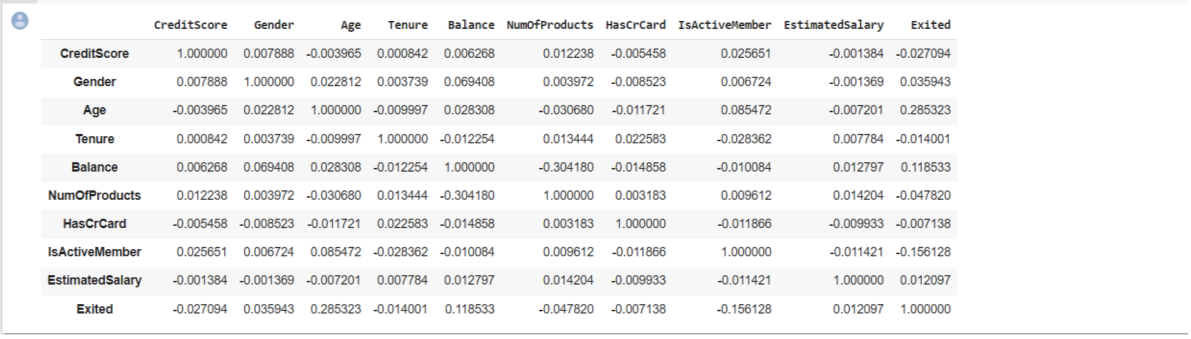
4.B. Perform Bi-variate Analysis

```
plt.scatter(df.Age,df.Balance)
```

```
<matplotlib.collections.PathCollection at 0x7fa0d35a7dd0>
```

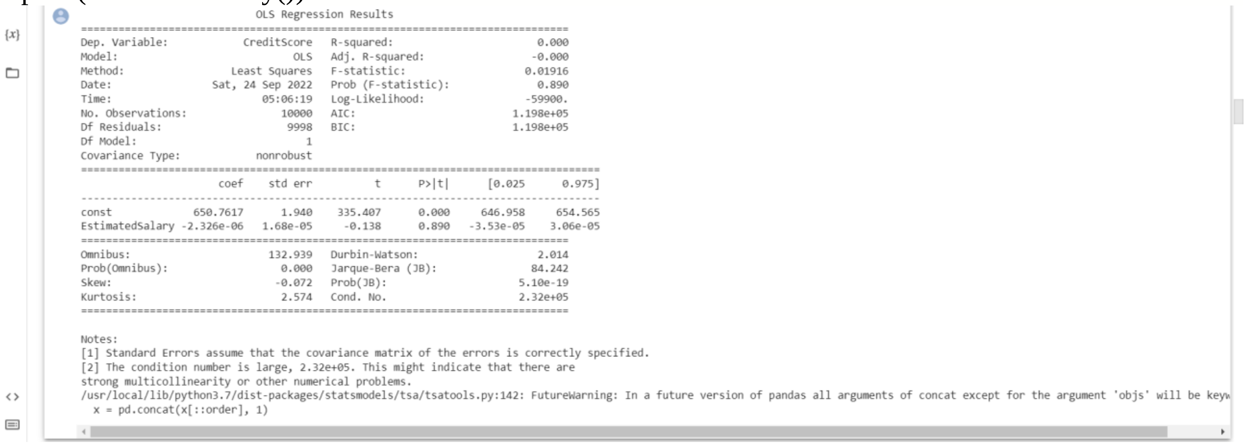


```
df.corr()
```



	CreditScore	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
CreditScore	1.000000	0.007888	-0.003965	0.000842	0.006268	0.012238	-0.005458	0.025651	-0.001384	-0.027094
Gender	0.007888	1.000000	0.022812	0.003739	0.069408	0.003972	-0.008523	0.006724	-0.001369	0.035943
Age	-0.003965	0.022812	1.000000	-0.009997	0.028308	-0.030680	-0.011721	0.085472	-0.007201	0.285323
Tenure	0.000842	0.003739	-0.009997	1.000000	-0.012254	0.013444	0.022583	-0.028362	0.007784	-0.014001
Balance	0.006268	0.069408	0.028308	-0.012254	1.000000	-0.304180	-0.014858	-0.010084	0.012797	0.118533
NumOfProducts	0.012238	0.003972	-0.030680	0.013444	-0.304180	1.000000	0.003183	0.009612	0.014204	-0.047820
HasCrCard	-0.005458	-0.008523	-0.011721	0.022583	-0.014858	0.003183	1.000000	-0.011866	-0.009933	-0.007138
IsActiveMember	0.025651	0.006724	0.085472	-0.028362	-0.010084	0.009612	-0.011866	1.000000	-0.011421	-0.156128
EstimatedSalary	-0.001384	-0.001369	-0.007201	0.007784	0.012797	0.014204	-0.009933	-0.011421	1.000000	0.012097
Exited	-0.027094	0.035943	0.285323	-0.014001	0.118533	-0.047820	-0.007138	-0.156128	0.012097	1.000000

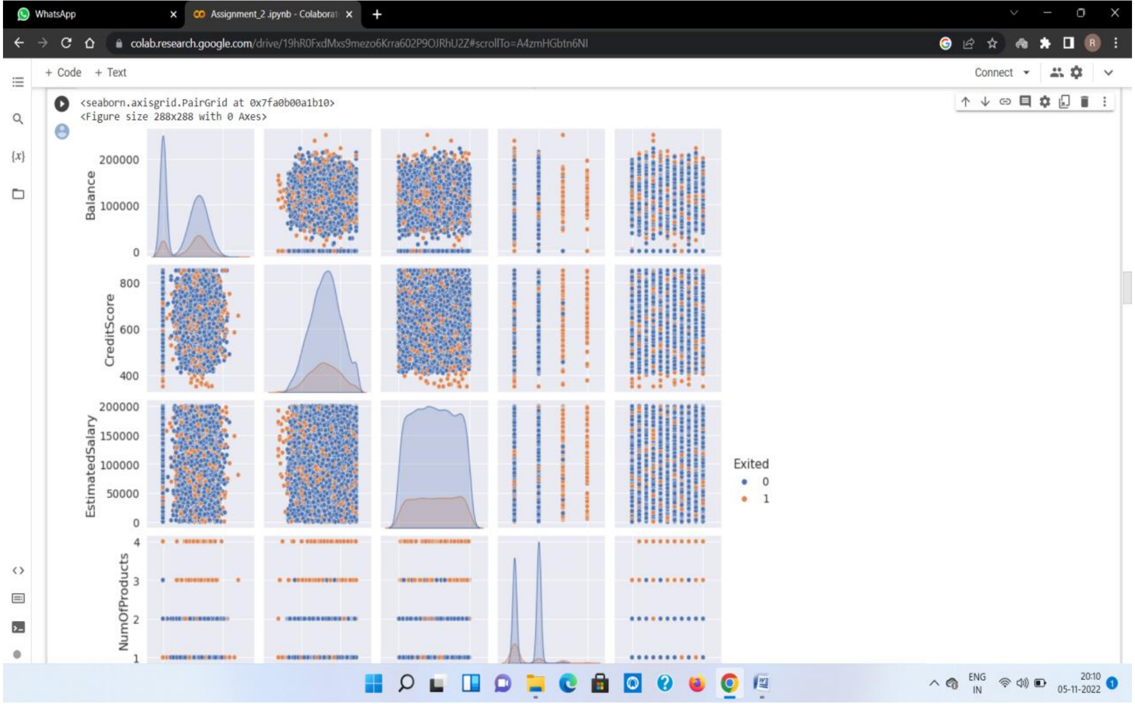
```
#Perform Bivariate Analysis
import statsmodels.api as sm

#define response variable
y = df['CreditScore']

#define explanatory variable
x = df[['EstimatedSalary']]

#add constant to predictor variables
x = sm.add_constant(x)

#fit linear regression model
model = sm.OLS(y, x).fit()

#view model summary
print(model.summary())
```



```
                            OLS Regression Results
==============================================================================
Dep. Variable:            CreditScore   R-squared:                       0.000
Model:                            OLS   Adj. R-squared:                 -0.000
Method:                 Least Squares   F-statistic:                   0.01916
Date:                Sat, 24 Sep 2022   Prob (F-statistic):              0.890
Time:                        05:06:19   Log-Likelihood:                -59900.
No. Observations:               10000   AIC:                         1.198e+05
Df Residuals:                    9998   BIC:                         1.198e+05
Df Model:                           1
Covariance Type:            nonrobust
===================================================================================
                      coef    std err          t      P>|t|      [0.025      0.975]
-----------------------------------------------------------------------------------
const             650.7617      1.940    335.407      0.000     646.958     654.565
EstimatedSalary -2.326e-06   1.68e-05     -0.138      0.890   -3.53e-05    3.06e-05
==============================================================================
Omnibus:                      132.939   Durbin-Watson:                   2.014
Prob(Omnibus):                  0.000   Jarque-Bera (JB):               84.242
Skew:                          -0.072   Prob(JB):                     5.10e-19
Kurtosis:                       2.574   Cond. No.                     2.32e+05
==============================================================================

Notes:
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
[2] The condition number is large, 2.32e+05. This might indicate that there are
strong multicollinearity or other numerical problems.
/usr/local/lib/python3.7/dist-packages/statsmodels/tsa/tsatools.py:142: FutureWarning: In a future version of pandas all arguments of concat except for the argument 'objs' will be keyw
  x = pd.concat(x[::order], 1)
```

4.C. Perform Multi-variate Analysis

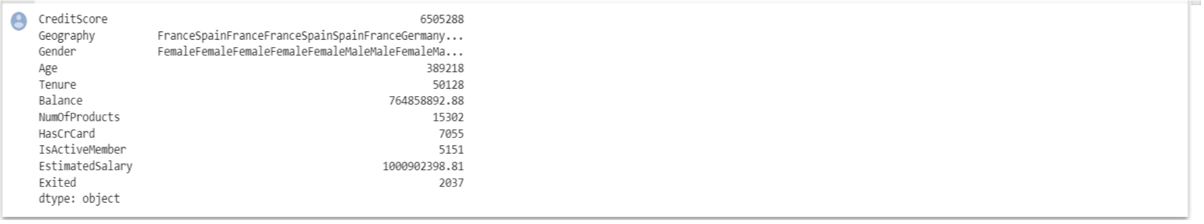
```
#Perform Multivariate Analysis
plt.figure(figsize=(4,4))
sns.pairplot(data=df[["Balance","CreditScore","EstimatedSalary","NumOfProducts","Tenure",
"Exited"]],hue="Exited")
```



```
<seaborn.axisgrid.PairGrid at 0x7fa0b00a1b10>
<Figure size 288x288 with 0 Axes>
```

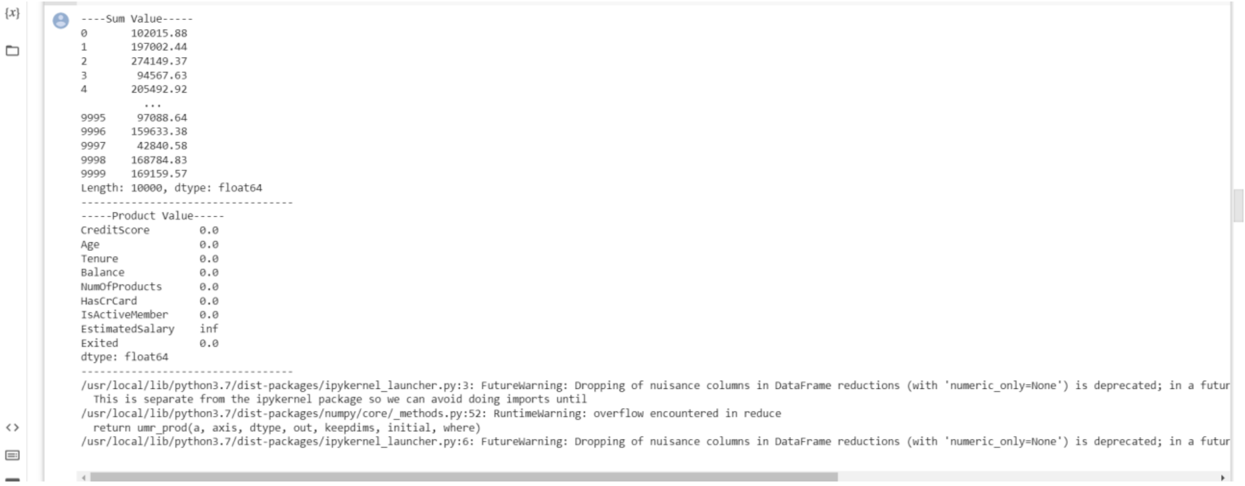
5.Perform descriptive statistics on the datasets:

```
df=pd.DataFrame(df)
print(df.sum())
```



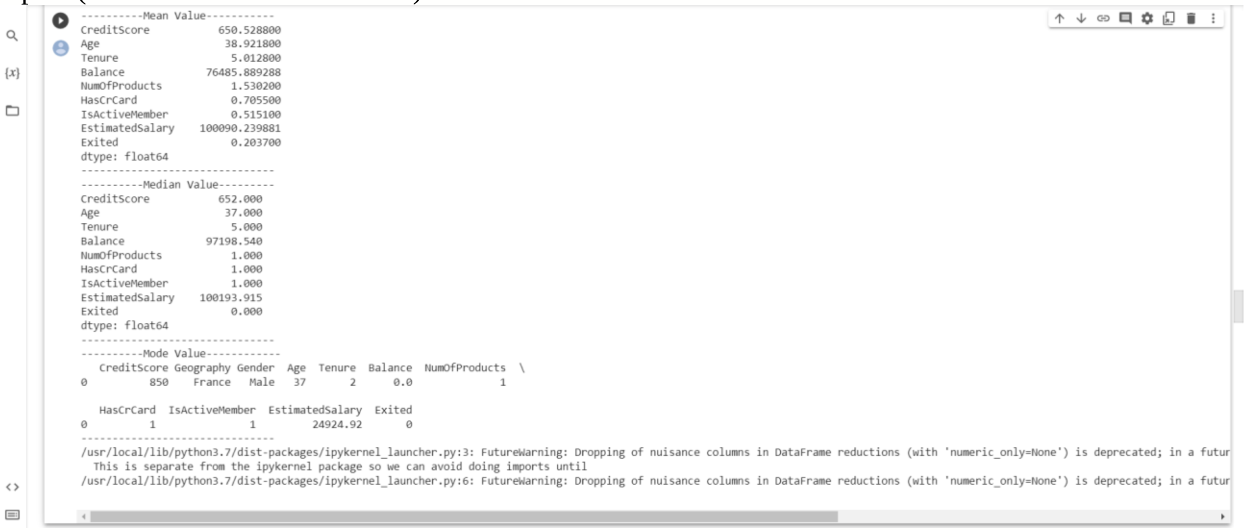
```
CreditScore                                                6505288
Geography          FranceSpainFranceFranceSpainSpainFranceGermany...
Gender             FemaleFemaleFemaleFemaleFemaleMaleMaleFemaleMa...
Age                                                         389218
Tenure                                                       50128
Balance                                               764858892.88
NumOfProducts                                                15302
HasCrCard                                                     7055
IsActiveMember                                                5151
EstimatedSalary                                      1000902398.81
Exited                                                        2037
dtype: object
```

```
#Perform Descriptive Statistics
print("----Sum Value-----")
print(df.sum(1))
print("-----------------------------------")
print("-----Product Value-----")
print(df.prod())
print("-----------------------------------")
```



```
----Sum Value-----
0       102015.88
1       197002.44
2       274149.37
3        94567.63
4       205492.92
          ...
9995     97088.64
9996    159633.38
9997     42840.58
9998    168784.83
9999    169159.57
Length: 10000, dtype: float64
----------------------------------
-----Product Value-----
CreditScore        0.0
Age                0.0
Tenure             0.0
Balance            0.0
NumOfProducts      0.0
HasCrCard          0.0
IsActiveMember     0.0
EstimatedSalary    inf
Exited             0.0
dtype: float64
----------------------------------
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:3: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a futur
  This is separate from the ipykernel package so we can avoid doing imports until
/usr/local/lib/python3.7/dist-packages/numpy/core/_methods.py:52: RuntimeWarning: overflow encountered in reduce
  return umr_prod(a, axis, dtype, out, keepdims, initial, where)
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:6: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a futur
```

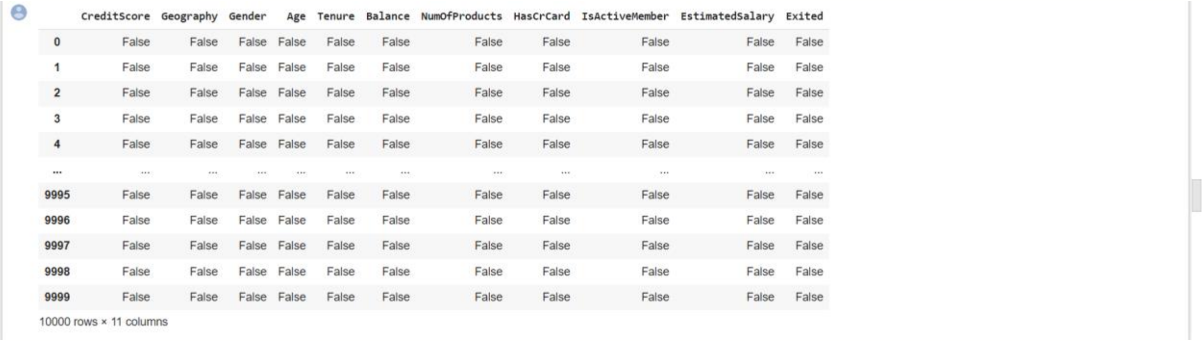
```
#Perform Descriptive Statistics
print("----------Mean Value-----------")
print(df.mean())
print("-------------------------------")
print("----------Median Value---------")
print(df.median())
print("-------------------------------")
print("----------Mode Value------------")
print(df.mode())
print("-------------------------------")
```



```
----------Mean Value-----------
CreditScore           650.528800
Age                    38.921800
Tenure                  5.012800
Balance             76485.889288
NumOfProducts           1.530200
HasCrCard               0.705500
IsActiveMember          0.515100
EstimatedSalary    100090.239881
Exited                  0.203700
dtype: float64
-------------------------------
----------Median Value---------
CreditScore           652.000
Age                    37.000
Tenure                  5.000
Balance             97198.540
NumOfProducts           1.000
HasCrCard               1.000
IsActiveMember          1.000
EstimatedSalary    100193.915
Exited                  0.000
dtype: float64
-------------------------------
----------Mode Value------------
   CreditScore Geography Gender  Age  Tenure  Balance  NumOfProducts  \
0          850    France   Male   37       2      0.0              1

   HasCrCard  IsActiveMember  EstimatedSalary  Exited
0          1               1         24924.92       0
-------------------------------
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:3: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a futur
  This is separate from the ipykernel package so we can avoid doing imports until
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:6: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a futur
```

6.Handle the missing values:

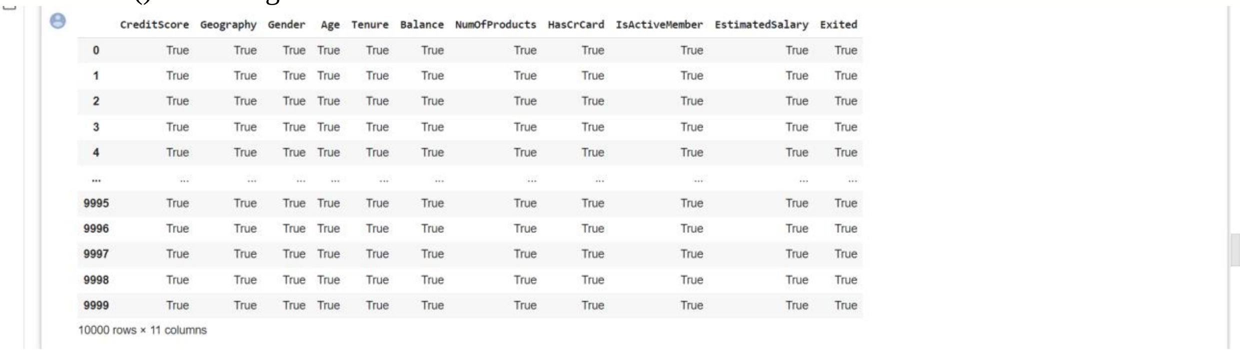
```
df.isnull()#Checking values are null
```



	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	False	False	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	False	False
...	...	...	...	...	...	...	...	...	...	...	...
9995	False	False	False	False	False	False	False	False	False	False	False
9996	False	False	False	False	False	False	False	False	False	False	False
9997	False	False	False	False	False	False	False	False	False	False	False
9998	False	False	False	False	False	False	False	False	False	False	False
9999	False	False	False	False	False	False	False	False	False	False	False

10000 rows × 11 columns

#Handling with missing Values
df.notnull()#Checking values are not null

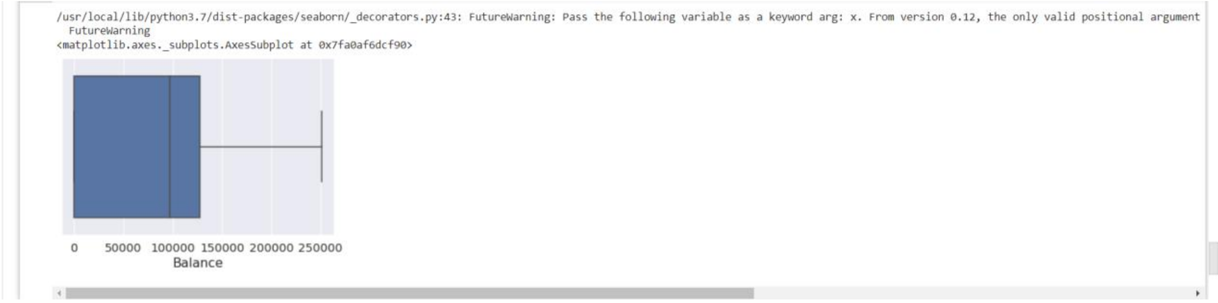


	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	True	True	True	True	True	True	True	True	True	True	True
1	True	True	True	True	True	True	True	True	True	True	True
2	True	True	True	True	True	True	True	True	True	True	True
3	True	True	True	True	True	True	True	True	True	True	True
4	True	True	True	True	True	True	True	True	True	True	True
...	...	...	...	...	...	...	...	...	...	...	...
9995	True	True	True	True	True	True	True	True	True	True	True
9996	True	True	True	True	True	True	True	True	True	True	True
9997	True	True	True	True	True	True	True	True	True	True	True
9998	True	True	True	True	True	True	True	True	True	True	True
9999	True	True	True	True	True	True	True	True	True	True	True

10000 rows × 11 columns

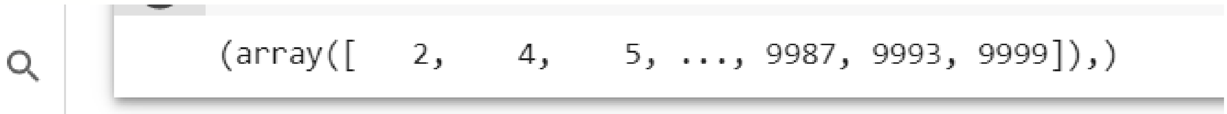
7.Find outlier and replace the outlier:

sns.boxplot(df['Balance'])



```
/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
  FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7fa0af6dcf90>
```

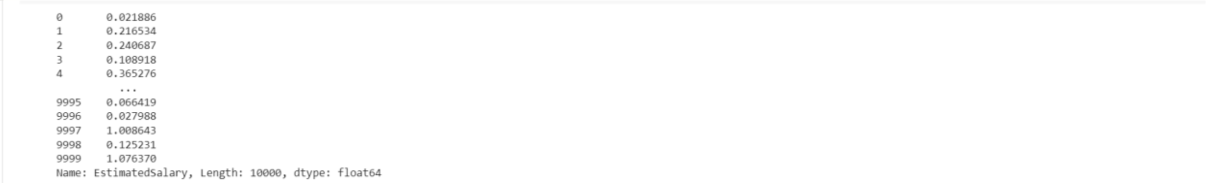
print(np.where(df['Balance']>100000))



```
(array([   2,    4,    5, ..., 9987, 9993, 9999]),)
```

#Find outliers & replace the outliers
from scipy import stats
import numpy as np

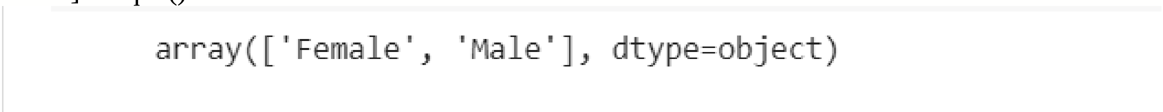
z = np.abs(stats.zscore(df["EstimatedSalary"]))
print(z)



```
0       0.021886
1       0.216534
2       0.240687
3       0.108918
4       0.365276
          ...
9995    0.066419
9996    0.027988
9997    1.008643
9998    0.125231
9999    1.076370
Name: EstimatedSalary, Length: 10000, dtype: float64
```

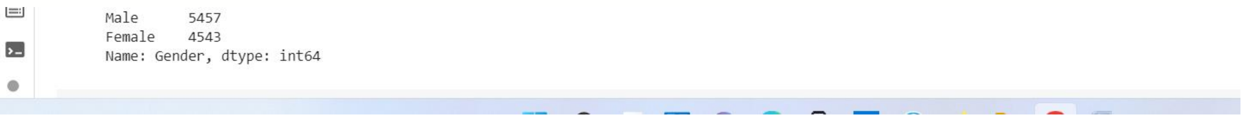
8.Check for categorical columns & performs encoding:

from sklearn.preprocessing import LabelEncoder
df['Gender'].unique()



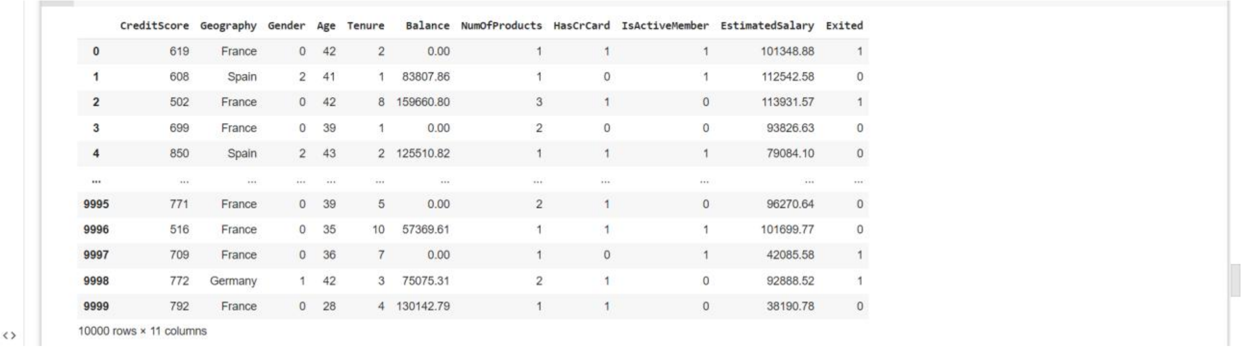
```
array(['Female', 'Male'], dtype=object)
```

df['Gender'].value_counts()



```
Male      5457
Female    4543
Name: Gender, dtype: int64
```

```
#Check for categorical columns & performs encoding
encoding=LabelEncoder()
df["Gender"]=encoding.fit_transform(df.iloc[:,1].values)
df
```

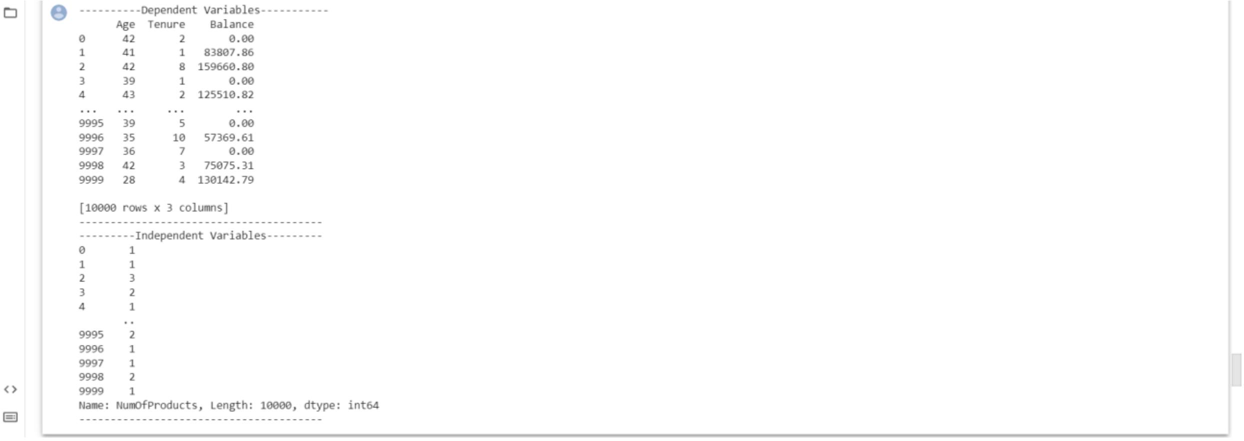


	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	619	France	0	42	2	0.00	1	1	1	101348.88	1
1	608	Spain	2	41	1	83807.86	1	0	1	112542.58	0
2	502	France	0	42	8	159660.80	3	1	0	113931.57	1
3	699	France	0	39	1	0.00	2	0	0	93826.63	0
4	850	Spain	2	43	2	125510.82	1	1	1	79084.10	0
...	...	...	...	...	...	...	...	...	...	...	...
9995	771	France	0	39	5	0.00	2	1	0	96270.64	0
9996	516	France	0	35	10	57369.61	1	1	1	101699.77	0
9997	709	France	0	36	7	0.00	1	0	1	42085.58	1
9998	772	Germany	1	42	3	75075.31	2	1	0	92888.52	1
9999	792	France	0	28	4	130142.79	1	1	0	38190.78	0

10000 rows × 11 columns

9.Split the data into Dependent & Independent Variables:

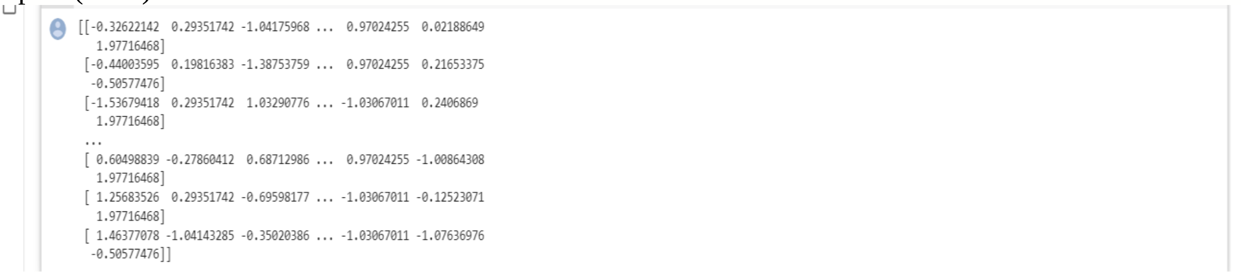
```
print("----------Dependent Variables----------")
X=df.iloc[:,1:4]
print(X)
print("--------------------------------------")
print("---------Independent Variables---------")
Y=df.iloc[:,4]
print(Y)
print("--------------------------------------")
```



```
----------Dependent Variables----------
      Age  Tenure    Balance
0      42       2       0.00
1      41       1   83807.86
2      42       8  159660.80
3      39       1       0.00
4      43       2  125510.82
...   ...     ...        ...
9995   39       5       0.00
9996   35      10   57369.61
9997   36       7       0.00
9998   42       3   75075.31
9999   28       4  130142.79

[10000 rows x 3 columns]
--------------------------------------
---------Independent Variables---------
0       1
1       1
2       3
3       2
4       1
       ..
9995    2
9996    1
9997    1
9998    2
9999    1
Name: NumOfProducts, Length: 10000, dtype: int64
--------------------------------------
```

10.Scale the independent variables:

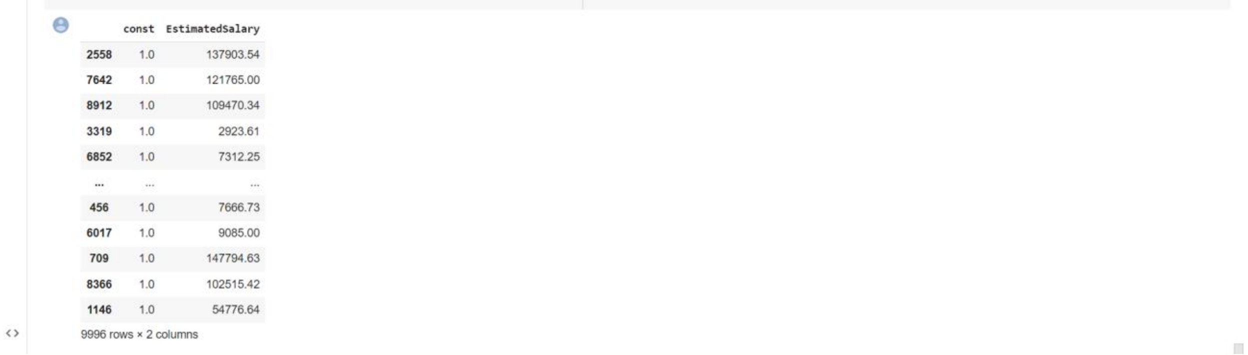
```
from sklearn.preprocessing import StandardScaler
object= StandardScaler()
# standardization
scale = object.fit_transform(df)
print(scale)
```



```
[[-0.32622142  0.29351742 -1.04175968 ...  0.97024255  0.02188649
   1.97716468]
 [-0.44003595  0.19816383 -1.38753759 ...  0.97024255  0.21653375
  -0.50577476]
 [-1.53679418  0.29351742  1.03290776 ... -1.03067011  0.2406869
   1.97716468]
 ...
 [ 0.60498839 -0.27860412  0.68712986 ...  0.97024255 -1.00864308
   1.97716468]
 [ 1.25683526  0.29351742 -0.69598177 ... -1.03067011 -0.12523071
   1.97716468]
 [ 1.46377078 -1.04143285 -0.35020386 ... -1.03067011 -1.07636976
  -0.50577476]]
```

11. Split the data into training & testing:

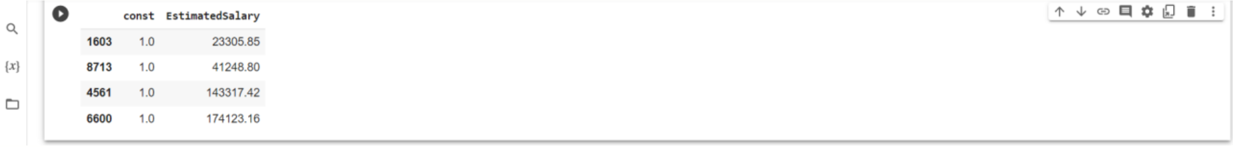
from sklearn.model_selection import train_test_split



	const	EstimatedSalary
2558	1.0	137903.54
7642	1.0	121765.00
8912	1.0	109470.34
3319	1.0	2923.61
6852	1.0	7312.25
...	...	...
456	1.0	7666.73
6017	1.0	9085.00
709	1.0	147794.63
8366	1.0	102515.42
1146	1.0	54776.64

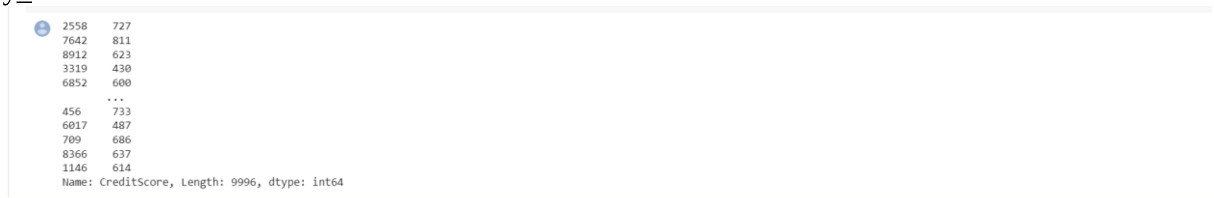
9996 rows × 2 columns

x_test



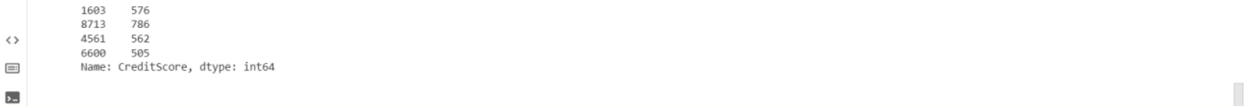
	const	EstimatedSalary
1603	1.0	23305.85
8713	1.0	41248.80
4561	1.0	143317.42
6600	1.0	174123.16

y_train



```
2558    727
7642    811
8912    623
3319    430
6852    600
       ...
456     733
6017    487
709     686
8366    637
1146    614
Name: CreditScore, Length: 9996, dtype: int64
```

y_test



```
1603    576
8713    786
4561    562
6600    505
Name: CreditScore, dtype: int64
```