

## ASSIGNMENT -2

### DATA VISUALIZATION AND DATA PRE-PROCESSING

Assignment Date	17 September 2022
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Student Roll Number	727719EUCS091
Maximum Marks	2 Marks

#### Question-1:

Download the dataset: Dataset

The screenshot shows a web browser window with multiple tabs. The active tab is 'Churn\_Modelling.csv - Google Drive'. The address bar shows the Google Drive link: [drive.google.com/file/d/1\\_HcM0K8wt4b7FMUkc1V1dv0y6L9ULzy/view](https://drive.google.com/file/d/1_HcM0K8wt4b7FMUkc1V1dv0y6L9ULzy/view). Below the address bar, there are several icons for Google Classroom, SKCET Online Exam, Classroom - GDB o..., Job Search Welcome, Learning Path, IBM Nalaiya Thiran, and Wipro FUTURESILLS. The main content area shows a preview of the 'Churn\_Modelling.csv' file. The file is a CSV with columns: RowNumber, CustomerId, Surname, CreditScore, Geography, Gender, Age, Tenure, Balance, NumOfProducts, HasCRCard, IsActiveMember, and EstimatedSalary. The preview shows the first 23 rows of data. The bottom of the screenshot shows a Windows taskbar with a search bar and several application icons.

RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCRCard	IsActiveMember	EstimatedSalary
1	15634602	Hargrave	619	France	Female	42	2	0	1	1	1	101348.88
2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.56
3	15619304	Onio	502	France	Female	42	8	159660.8	3	1	0	113931.57
4	15701354	Boni	699	France	Female	39	1	0	2	0	0	93826.63
5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.3
6	15574012	Chu	645	Spain	Male	44	8	113755.78	2	1	0	149756.71
7	15592531	Bartlett	822	France	Male	50	7	0	2	1	1	10062.8
8	15656148	Oblinna	376	Germany	Female	29	4	115046.74	4	1	0	119346.88
9	15792365	He	501	France	Male	44	4	142051.07	2	0	1	74940.5
10	15592389	H7	684	France	Male	27	2	134603.88	1	1	1	71725.73
11	15767821	Bearce	528	France	Male	31	6	102016.72	2	0	0	80181.12
12	15737173	Andrews	497	Spain	Male	24	3	0	2	1	0	76390.01
13	15632264	Kay	476	France	Female	34	10	0	2	1	0	26260.96
14	15691483	Chin	549	France	Female	25	5	0	2	0	0	190857.79
15	15600882	Scott	635	Spain	Female	35	7	0	2	1	1	65951.65
16	15643966	Goforth	616	Germany	Male	45	3	143129.41	2	0	1	64327.26
17	15737452	Romeo	653	Germany	Male	58	1	132602.88	1	1	0	5097.67
18	15788218	Henderson	549	Spain	Female	24	9	0	2	1	1	14406.41
19	15661507	Mulrow	587	Spain	Male	45	6	0	1	0	0	158684.81
20	15568982	Hao	726	France	Female	24	6	0	2	1	1	54724.03
21	15577657	McDonald	732	France	Male	41	8	0	2	1	1	170886.17
22	15687945	Dallucci	636	Spain	Female	32	8	0	2	1	0	138555.46

#### Question-2:

Load the dataset.

**Solution:**

```
import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

df = pd.read_csv("C://Users\Mohana Sowdesh\Desktop//IBM Nalaiya Thiran// Dataset//
Churn_Modelling.csv")
df.head()
```

```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [4]: df = pd.read_csv("C://Users\Mohana Sowdesh//Desktop//IBM Nalaiya Thiran//Dataset//Churn_Modelling.csv")
df.head()
```

```
Out[4]:
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	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

### Question-3:

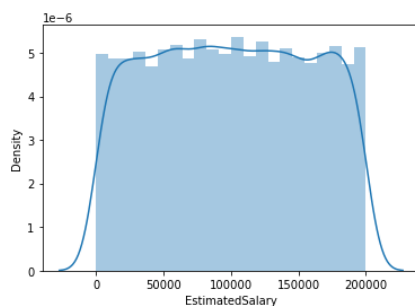
Perform Below Visualizations.

#### • Univariate Analysis

```
In [5]: sns.distplot(df.EstimatedSalary)
```

C:\Users\Mohana Sowdesh\anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
warnings.warn(msg, FutureWarning)

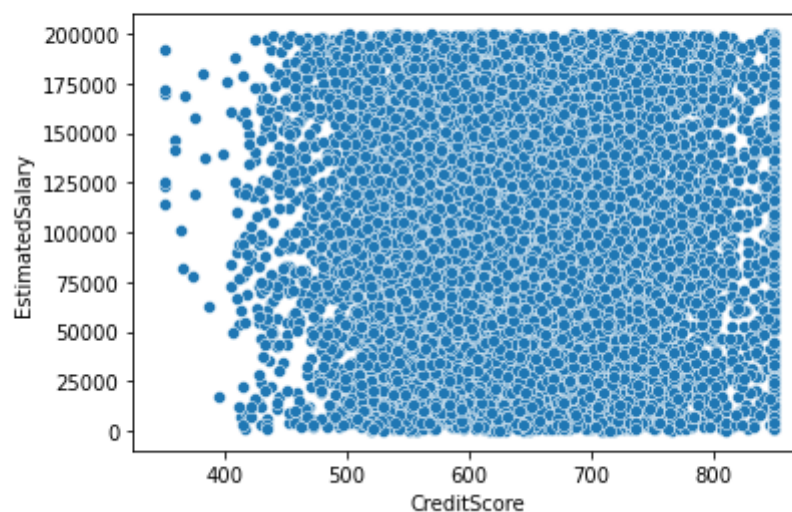
```
Out[5]: <AxesSubplot:xlabel='EstimatedSalary', ylabel='Density'>
```



#### • Bi - Variate Analysis

```
In [8]: sns.scatterplot(df.CreditScore,df.EstimatedSalary)
```

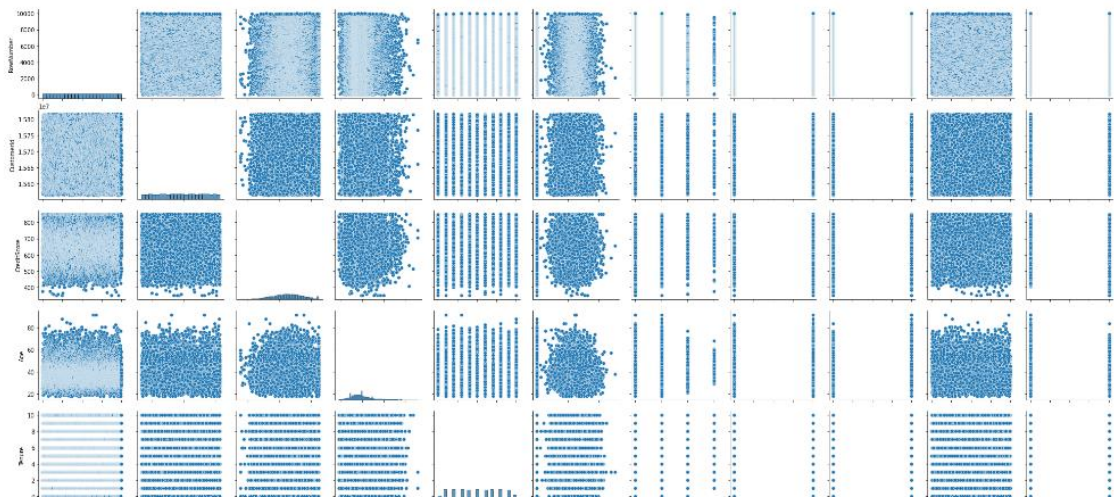
```
Out[8]: <AxesSubplot:xlabel='CreditScore', ylabel='EstimatedSalary'>
```



- Multi - Variate Analysis

```
In [9]: sns.pairplot(df)
```

```
Out[9]: <seaborn.axisgrid.PairGrid at 0x13f5bf2ab20>
```



#### Question-4:

Perform descriptive statistics on the dataset.

```
In [9]: df['Age'].mean()
```

```
Out[9]: 38.9218
```

```
In [11]: df['NumOfProducts'].median()
```

```
Out[11]: 1.0
```

```
In [12]: df['HasCrCard'].mode()
```

```
Out[12]: 0    1
dtype: int64
```

```
In [13]: df.skew()
```

```
Out[13]: RowNumber      0.000000
CustomerId    0.001149
CreditScore  -0.071607
Age           1.011320
Tenure        0.010991
Balance      -0.141109
NumOfProducts 0.745568
HasCrCard    -0.901812
IsActiveMember -0.060437
EstimatedSalary 0.002085
Exited       1.471611
dtype: float64
```

---

```
In [14]: df.kurt()
```

```
Out[14]: RowNumber      -1.200000  
         CustomerId     -1.196113  
         CreditScore    -0.425726  
         Age            1.395347  
         Tenure         -1.165225  
         Balance        -1.489412  
         NumOfProducts   0.582981  
         HasCrCard       -1.186973  
         IsActiveMember -1.996747  
         EstimatedSalary -1.181518  
         Exited          0.165671  
         dtype: float64
```

```
In [19]: df.var()
```

```
Out[19]: RowNumber      8.334167e+06  
         CustomerId      5.174815e+09  
         CreditScore      9.341860e+03  
         Age             1.099941e+02  
         Tenure           8.364673e+00  
         Balance          3.893436e+09  
         NumOfProducts     3.383218e-01  
         HasCrCard         2.077905e-01  
         IsActiveMember     2.497970e-01  
         EstimatedSalary    3.307457e+09  
         Exited            1.622225e-01  
         dtype: float64
```

```
In [20]: df.std()
```

```
Out[20]: RowNumber      2886.895680  
         CustomerId     71936.186123  
         CreditScore      96.653299  
         Age             10.487806  
         Tenure           2.892174  
         Balance          62397.405202  
         NumOfProducts     0.581654  
         HasCrCard         0.455840  
         IsActiveMember     0.499797  
         EstimatedSalary   57510.492818  
         Exited            0.402769  
         dtype: float64
```

### Question-5:

Handle the Missing values.

```
In [21]: df.isna().any()
```

```
Out[21]: RowNumber      False
CustomerId      False
Surname          False
CreditScore      False
Geography        False
Gender           False
Age             False
Tenure           False
Balance          False
NumOfProducts    False
HasCrCard        False
IsActiveMember   False
EstimatedSalary  False
Exited           False
dtype: bool
```

```
In [22]: df.isna().sum()
```

```
Out[22]: 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
```

```
In [24]: df['Age'].fillna(df['Age'].mean(),inplace=True)
df
```

```
Out[24]:
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSa
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	10134
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	11254
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	11393
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	9382
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	7908
...	...	...	...	...	...	...	...	...	...	...	...	...	...
9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	9627
9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	10169
9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	4208
9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	75075.31	2	1	0	9288
9999	10000	15628319	Walker	792	France	Female	28	4	130142.79	1	1	0	3819

10000 rows x 14 columns

## Question-6:

Find the outliers and replace the outliers

```
In [43]: Q1=df.Age.quantile(0.25)
Q2=df.Age.quantile(0.75)
IQR=Q2-Q1
print(IQR)
```

12.0

```
In [44]: df=df[~((df.Age<(Q1-1.5*IQR))|(df.Age>(Q2+1.5*IQR)))]
df
```

Out[44]:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	10134
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	11254
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	11393
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	9382
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	7908
...	...	...	...	...	...	...	...	...	...	...	...	...	...
9995	9996	15606229	Obijaku	771	France	Male	39	5	0.00	2	1	0	9627
9996	9997	15569892	Johnstone	516	France	Male	35	10	57369.61	1	1	1	10169
9997	9998	15584532	Liu	709	France	Female	36	7	0.00	1	0	1	4208
9998	9999	15682355	Sabbatini	772	Germany	Male	42	3	75075.31	2	1	0	9288
9999	10000	15628319	Walker	792	France	Female	28	4	130142.79	1	1	0	3819

9641 rows x 14 columns

## Question-7:

Check for Categorical columns and perform encoding.

```
In [46]: df['HasCrCard'].replace({1:'Yes',0:'No'},inplace=True)
df.head()
```

Out[46]:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	Yes	1	101348.88
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	No	1	112542.58
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	Yes	0	113931.57
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	No	0	93826.63
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	Yes	1	79084.10

## Question-8:

Split the data into dependent and independent variables.

```
In [50]: data_main= pd.get_dummies(df,columns=['Gender'])
data_main
```

Out[50]:

Surname	CreditScore	Geography	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	Gender_Female	Gender_Male	
Hargrave	619	France	42	2	0.00		1	Yes	1	101348.88	1	1	0
Hill	608	Spain	41	1	83807.86		1	No	1	112542.58	0	1	0
Onio	502	France	42	8	159660.80		3	Yes	0	113931.57	1	1	0
Boni	699	France	39	1	0.00		2	No	0	93826.63	0	1	0
Mitchell	850	Spain	43	2	125510.82		1	Yes	1	79084.10	0	1	0
...	...	...	...	...	...	...	...	...	...	...	...	...	...
Obijaku	771	France	39	5	0.00		2	Yes	0	96270.64	0	0	1
Johnstone	516	France	35	10	57369.61		1	Yes	1	101699.77	0	0	1
Liu	709	France	36	7	0.00		1	No	1	42085.58	1	1	0
Sabbatini	772	Germany	42	3	75075.31		2	Yes	0	92888.52	1	0	1
Walker	792	France	28	4	130142.79		1	Yes	0	38190.78	0	1	0

```
In [51]: y = data_main['Tenure']
y
```

```
Out[51]: 0      2
         1      1
         2      8
         3      1
         4      2
         ..
        9995     5
        9996    10
        9997     7
        9998     3
        9999     4
        Name: Tenure, Length: 9641, dtype: int64
```

```
In [52]: x = data_main.drop(columns='Tenure',axis=1)
x.head()
```

```
Out[52]:
```

CustomerId	Surname	CreditScore	Geography	Age	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited	Gender_Female	Gender_Male
i634602	Hargrave	619	France	42	0.00	1	Yes	1	101348.88	1	1	0
i647311	Hill	608	Spain	41	83807.86	1	No	1	112542.58	0	1	0
i619304	Onio	502	France	42	159660.80	3	Yes	0	113931.57	1	1	0
i701354	Boni	699	France	39	0.00	2	No	0	93826.63	0	1	0
i737888	Mitchell	850	Spain	43	125510.82	1	Yes	1	79084.10	0	1	0

### Question-9:

Scale the independent variables

```
In [55]: x=df.iloc[:,6:7].values
from sklearn.preprocessing import StandardScaler
std=StandardScaler()
x=std.fit_transform(x)
x
```

```
Out[55]: array([[ 0.47806838],
                 [ 0.36446646],
                 [ 0.47806838],
                 ...,
                 [-0.20354316],
                 [ 0.47806838],
                 [-1.11235856]])
```

### Question-10:

Split the data into training and testing

```
In [56]: 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)
```

```
In [57]: x_train
```

```
Out[57]: array([[ -2.24837781],
 [  0.59167031],
 [  1.04607801],
 ...,
 [-0.54434894],
 [  1.04607801],
 [-0.43074701]])
```

```
In [58]: x_test
```

```
Out[58]: array([[ 1.50048571],
 [-0.20354316],
 [  0.36446646],
 ...,
 [  0.81887416],
 [-0.88515471],
 [  0.13726261]])
```

```
In [59]: y_train
```

```
Out[59]: 746      2
1788      8
1057      1
7559      2
1141      5
..
8184      3
9567      4
5042      3
3370      6
2819      5
Name: Tenure, Length: 7712, dtype: int64
```

```
In [60]: y_test
```

```
Out[60]: 2454      1
944      8
3938      1
4109      1
8573      8
..
1465      3
8409      9
5624      1
2817      8
6851      9
Name: Tenure, Length: 1929, dtype: int64
```