

Assignment -2

Data Visualization and Data Pre-processing

Assignment Date	17 September 2022
Student Name	Logeshkumar R
Student Roll Number	727719EUCS074
Maximum Marks	2 Marks

Question-1:

Download the dataset: Dataset

Churn_Modelling.csv · Google | X

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Churn_Modelling.csv

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	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z
1	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Bal	EstimatedSalary	Exited	Churn	Churn	Churn	Churn	Churn	Churn	Churn	Churn	Churn	Churn	Churn	Churn	Churn	Churn	
2	1	15634602	Hargrave	619	France	Female	42	2	0	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
3	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	0	1	1	1	1	1	1	1	1	1	1	1	1	
4	3	15619304	Onio	502	France	Female	42	8	159660.8	3	1	0	1	0	1	1	1	1	1	1	1	1	1	1	1	
5	4	15701354	Boni	699	France	Female	39	1	0	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
6	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
7	6	15574012	Chu	645	Spain	Male	44	8	113755.78	2	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	
8	7	15592531	Bartlett	922	France	Male	50	7	0	2	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
9	8	15656148	Obinna	376	Germany	Female	29	4	115046.74	4	1	0	1	0	1	1	1	1	1	1	1	1	1	1	1	
10	9	15792385	He	501	France	Male	44	4	142051.07	2	0	1	0	1	1	1	1	1	1	1	1	1	1	1	1	
11	10	15592389	H7	684	France	Male	27	2	134603.88	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
12	11	15767821	Bearece	528	France	Male	31	6	102016.72	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
13	12	15737173	Andrews	497	Spain	Male	24	3	0	2	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
14	13	15632264	Kay	476	France	Female	34	10	0	2	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
15	14	15691483	Chin	549	France	Female	25	5	0	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
16	15	15600882	Scott	635	Spain	Female	35	7	0	2	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
17	16	15643966	Goforth	616	Germany	Male	45	3	143129.41	2	0	1	0	1	1	1	1	1	1	1	1	1	1	1	1	
18	17	15737452	Romeo	653	Germany	Male	58	1	132602.88	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
19	18	15788218	Henderson	549	Spain	Female	24	9	0	2	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
20	19	15661507	Muldrow	587	Spain	Male	45	6	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
21	20	15568982	Hao	726	France	Female	24	6	0	2	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
22	21	15577657	McDonald	732	France	Male	41	8	0	2	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
23	22	15597945	Dellucci	636	Spain	Female	32	8	0	2	1	1	0	1	1	1	1	1	1	1	1	1	1	1	1	
24	23	15699309	Gerasimov	510	Spain	Female	38	4	0	1	1	1	0	1	1	1	1	1	1	1	1	1	1	1	1	
25	24	15725737	Mosman	669	France	Male	3	0	2	0	1	0	1	0	1	1	1	1	1	1	1	1	1	1	1	
26	25	15625047	Yen	846	France	Female	5	0	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
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Question-2:

Load the dataset.

Solution:

```
import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

data = pd.read_csv("E://Churn_Modelling.csv")
data.head()
```

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
data = pd.read_csv("E://Churn_Modelling.csv")
data.head()
```

Out[1]:

	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

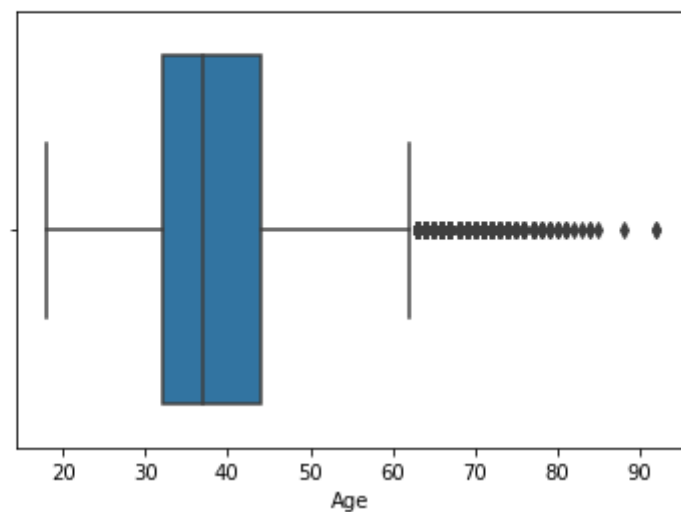
Question-3:

Perform Below Visualizations.

• Univariate Analysis

```
In [3]: sns.boxplot(data.Age)
```

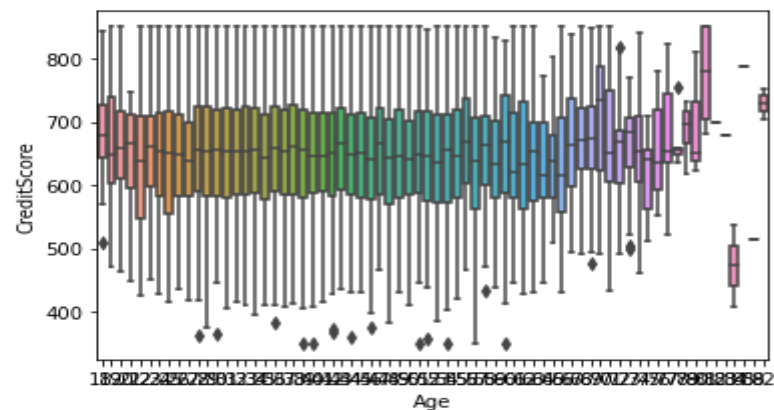
Out[3]: <AxesSubplot:xlabel='Age'>



• Bi - Variate Analysis

```
In [7]: sns.boxplot(y=data.CreditScore,x=data.Age)
```

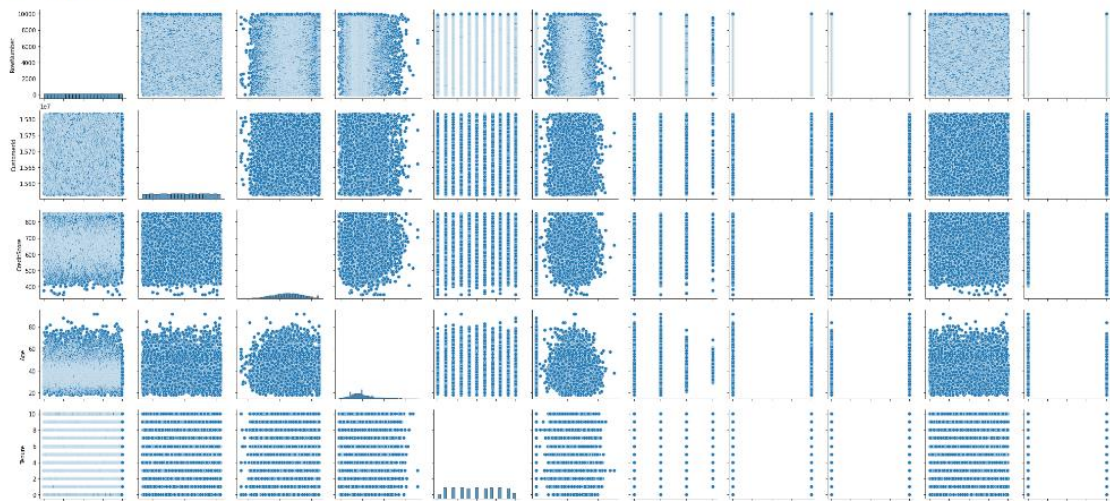
Out[7]: <AxesSubplot:xlabel='Age', ylabel='CreditScore'>



- Multi - Variate Analysis

```
In [8]: sns.pairplot(data)
```

```
Out[8]: <seaborn.axisgrid.PairGrid at 0x22213a81250>
```



Question-4:

Perform descriptive statistics on the dataset.

```
In [9]: data['NumOfProducts'].mean()
```

```
Out[9]: 1.5302
```

```
In [10]: data['EstimatedSalary'].median()
```

```
Out[10]: 100193.915
```

```
In [11]: data['Tenure'].mode()
```

```
Out[11]: 0    2
dtype: int64
```

```
In [13]: data.kurt()
```

```
Out[13]: 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 [16]: data.var()
```

```
Out[16]: 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 [17]: data.std()
```

```
Out[17]: 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 [18]: data.isna().any()
```

```
Out[18]: 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 [19]: data.isna().sum()
```

```
Out[19]: 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 [22]: data['EstimatedSalary'].fillna(data['EstimatedSalary'].mean(),inplace=True)
         data
```

```
Out[22]:
```

	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

10000 rows x 14 columns

Question-6:

Find the outliers and replace the outliers

```
In [23]: Q1=data.Age.quantile(0.25)
         Q2=data.Age.quantile(0.75)
         IQR=Q2-Q1
         print(IQR)
```

```
12.0
```

```
In [24]: data=data[~((data.Age<(Q1-1.5*IQR))|(data.Age>(Q2+1.5*IQR)))]
         data
```

```
Out[24]:
```

	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 [28]: data['Exited'].replace({1:'Yes',0:'No'},inplace=True)
data.head()
```

```
Out[28]:
```

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

Question-8:

Split the data into dependent and independent variables.

```
In [34]: dmain= pd.get_dummies(data,columns=['Gender'])
dmain
```

```
Out[34]:
```

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

```
In [35]: y = dmain['HasCrCard']
y
```

```
Out[35]:
```

0	1
1	0
2	1
3	0
4	1
...	...
9995	1
9996	1
9997	0
9998	1
9999	1

Name: HasCrCard, Length: 10000, dtype: int64

```
In [36]: x = dmain.drop(columns='HasCrCard',axis=1)
x.head()
```

```
Out[36]:
```

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

Question-9:

Scale the independent variables

```
In [37]: x=data.iloc[:,6:7].values
         from sklearn.preprocessing import StandardScaler
         std=StandardScaler()
         x=std.fit_transform(x)
         x
```

```
Out[37]: array([[ 0.29351742],
                [ 0.19816383],
                [ 0.29351742],
                ...,
                [-0.27860412],
                [ 0.29351742],
                [-1.04143285]])
```

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 [40]: y_train
```

```
Out[40]: 7389    1
          9275    1
          2995    1
          5316    1
          356     1
          ..
          9225    1
          4859    1
          3264    1
          9845    1
          2732    1
          Name: HasCrCard, Length: 8000, dtype: int64
```

```
In [41]: y_test
```

```
Out[41]: 9394    1
          898     1
          2398    1
          5906    0
          2343    1
          ..
          1037    1
          2899    1
          9549    1
          2740    1
          6690    1
          Name: HasCrCard, Length: 2000, dtype: int64
```