

**Assignment -2 Data
Visualization and Pre-processing**

Assignment submission	28 September 2022
Student Name	B Mydheen Beevi
Student Roll Number	951919CS093
Maximum Marks	2 Marks

1. Download the dataset: Dataset

2. Load the dataset.

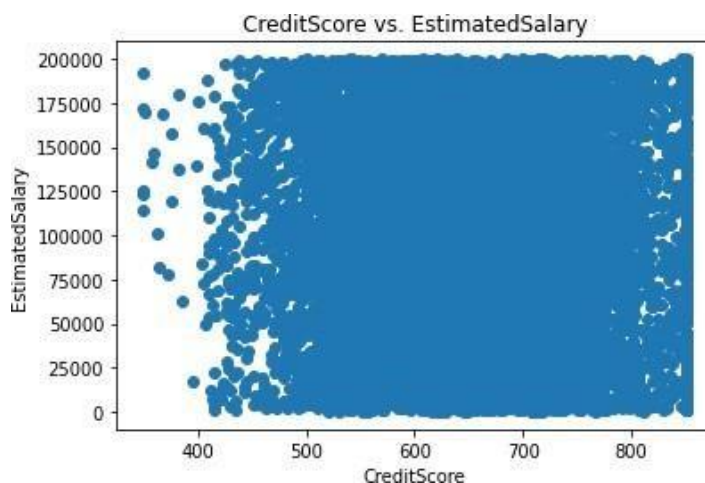
```
import pandas as pd
df=pd.read_csv('Churn_Modelling.csv')
```

3. Perform Below Visualizations

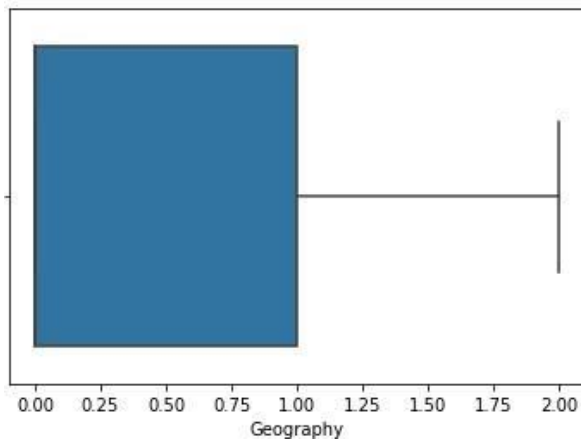
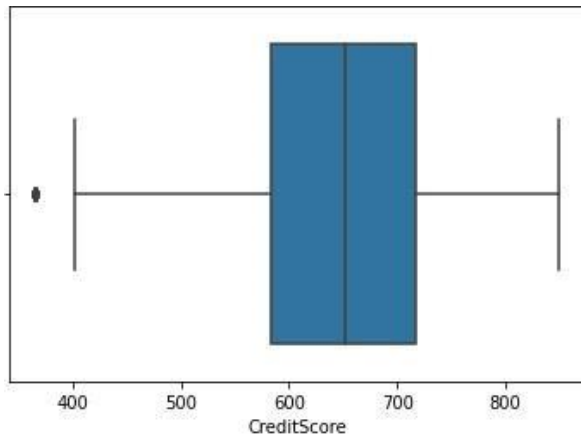
- Univariate Analysis

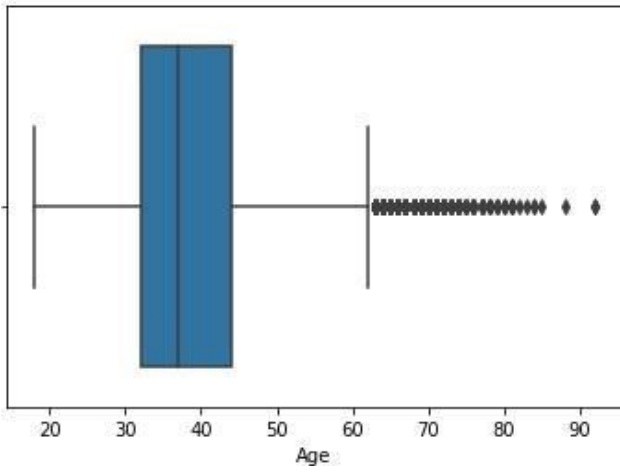
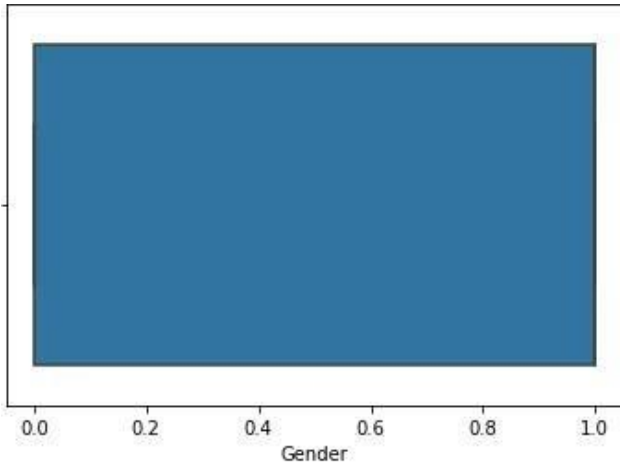
```
import matplotlib.pyplot as plt
```

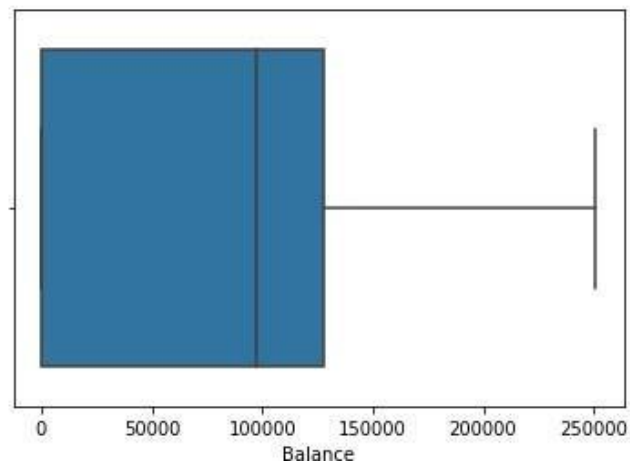
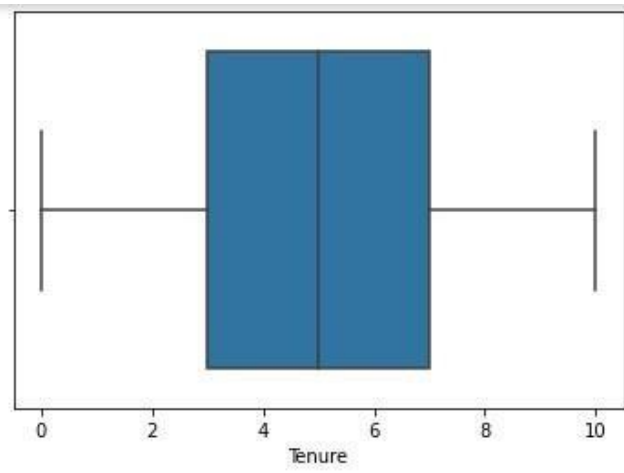
```
%matplotlib inline
plt.scatter(df.CreditScore,df.EstimatedSalary)
plt.title('CreditScore vs. EstimatedSalary')
plt.xlabel('CreditScore') plt.ylabel('EstimatedSalary')
plt.show()
```

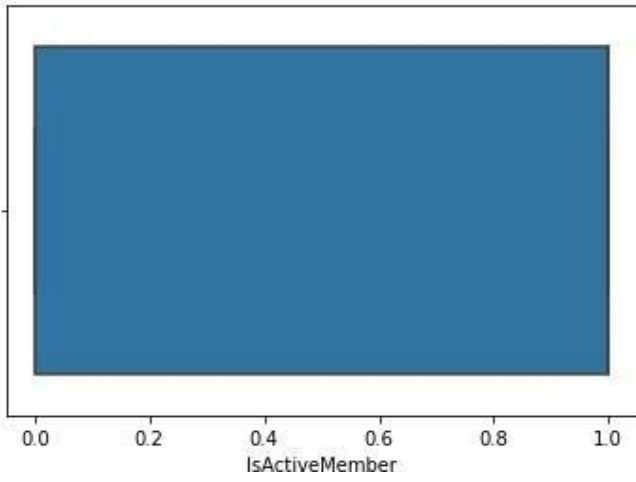
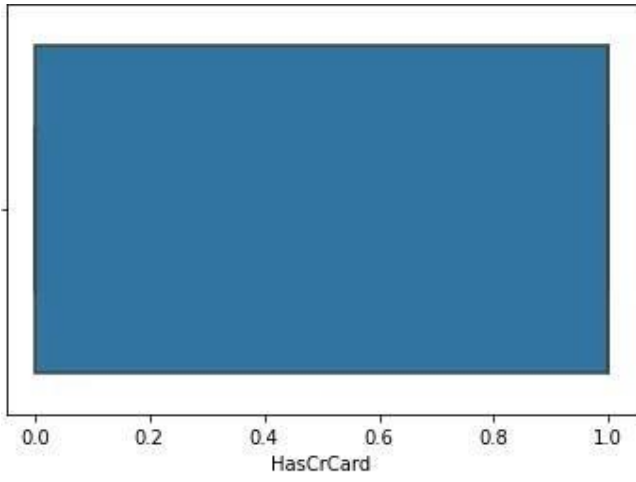


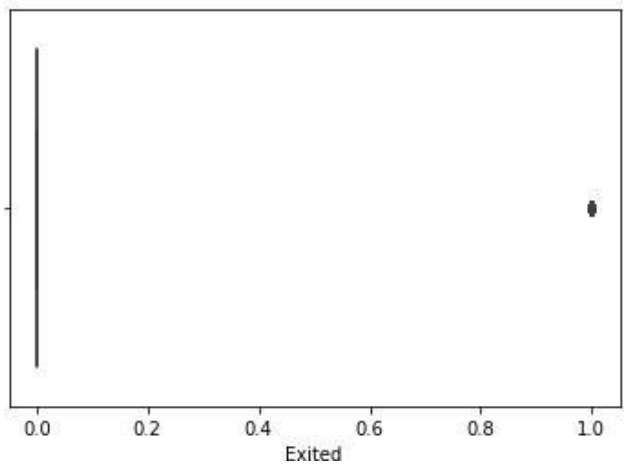
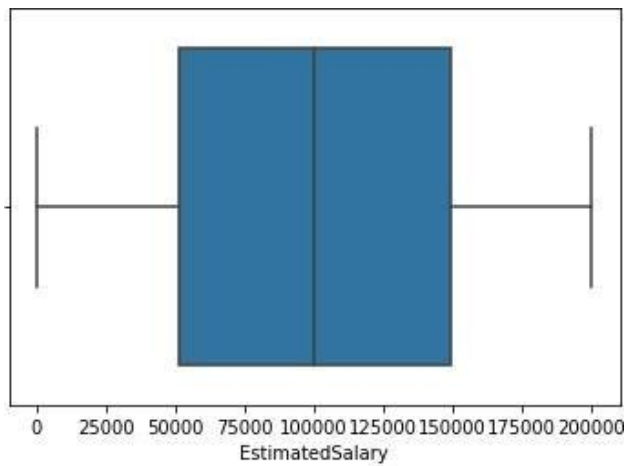
```
for col in df.columns: if(df.dtypes[col]=='int64' or
df.dtypes[col]=='float64' ):
    sns.boxplot(x=df[col]).set( xlabel=col)
plt.show()
```





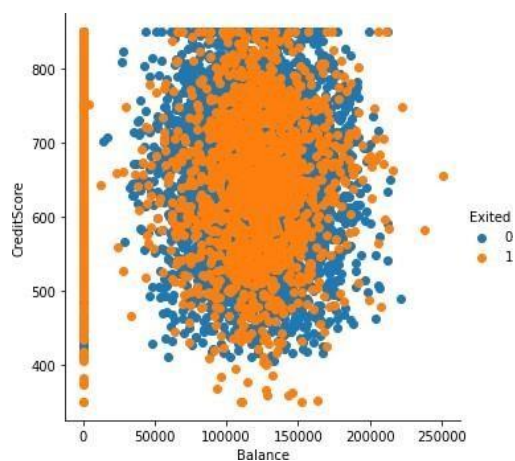




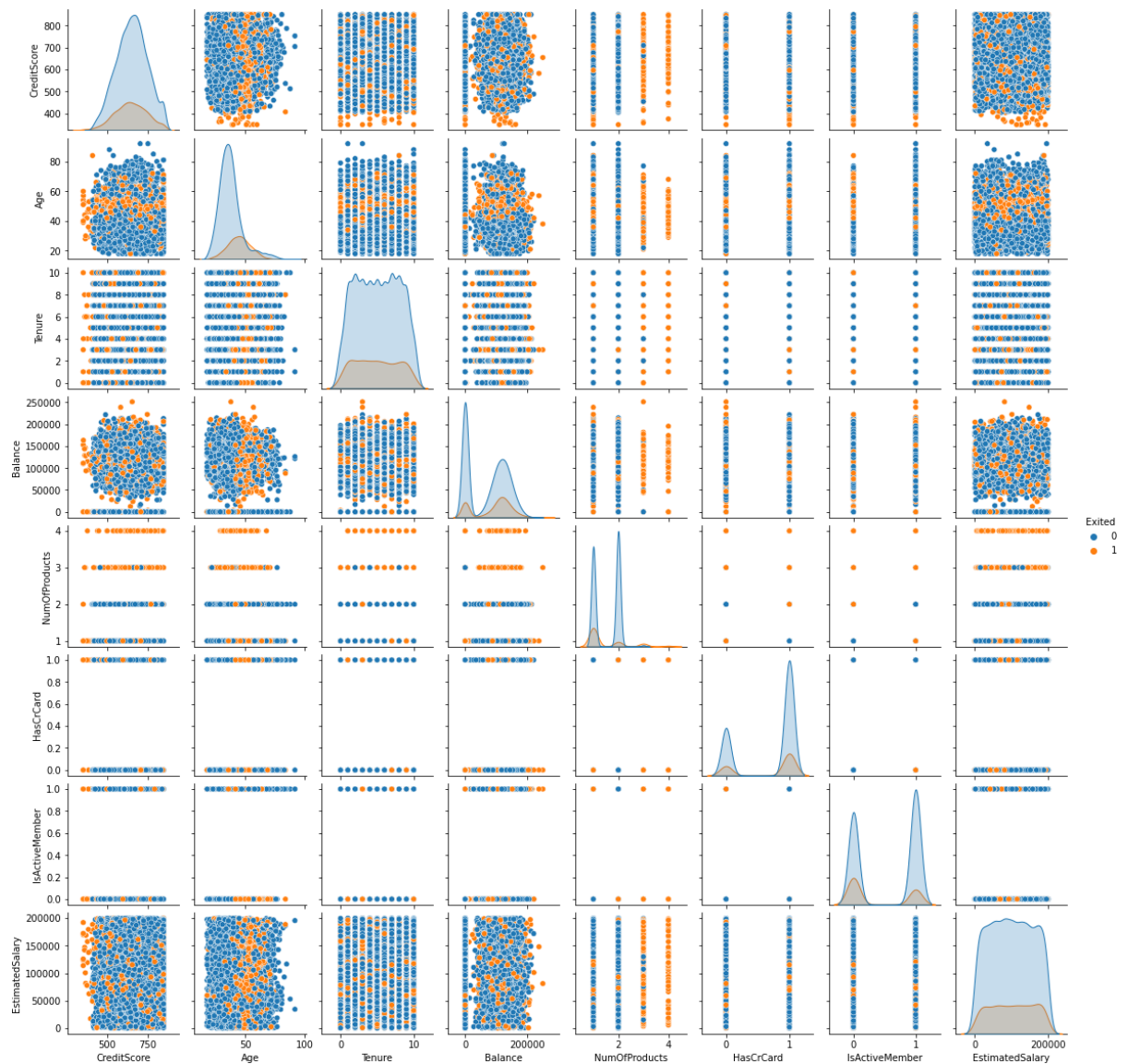


- Bi - Variate Analysis

```
import seaborn as sns
sns.FacetGrid(df,hue='Exited',height=5).map(plt.scatter,"Balance","CreditScore").add_legend()
plt.show()
```



- Multi - Variate Analysis `sns.pairplot(df, hue='Exited', height=2)`



4. Perform descriptive statistics on the dataset.

`df.describe()`

	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
count	10000.00000	10000.00000	10000.00000	10000.00000	10000.00000	10000.00000	9940.00000	10000.00000	10000.00000	10000.00000	10000.00000
mean	650.52400	0.746300	0.545700	38.921800	5.012800	76485.889288	1.515292	0.70550	0.515100	100090.239881	0.203700
std	96.66498	0.827529	0.497932	10.487806	2.892174	62397.405202	0.550743	0.45584	0.499797	57510.492818	0.402769
min	365.00000	0.000000	0.000000	18.000000	0.000000	0.000000	1.000000	0.000000	0.000000	11.580000	0.000000
25%	584.00000	0.000000	0.000000	32.000000	3.000000	0.000000	1.000000	0.000000	0.000000	51002.110000	0.000000
50%	652.00000	0.000000	1.000000	37.000000	5.000000	97198.540000	1.000000	1.000000	1.000000	100193.915000	0.000000
75%	718.00000	1.000000	1.000000	44.000000	7.000000	127644.240000	2.000000	1.000000	1.000000	149388.247500	0.000000
max	850.00000	2.000000	1.000000	92.000000	10.000000	250898.090000	3.000000	1.000000	1.000000	199992.480000	1.000000

5. Handle the Missing values. `df.isnull().sum()`

```
CreditScore      0
Geography        0
Gender           0
Age              0
```

```
Tenure          0
Balance         0
NumOfProducts  0
HasCrCard       0
IsActiveMember  0
EstimatedSalary 0
Exited          0
dtype: int64
```

#there is no missing values

6. Find the outliers and replace the outliers

```
import numpy as np
```

#Outliers are found using the univariate

```
CreditsMedian = df.loc[df['CreditScore']<400, 'CreditScore'].median()
```

```
ProdMedian = df.loc[df['NumOfProducts']>=3.5, 'NumOfProducts'].median()
```

```
df.loc[df.CreditScore < 400, 'CreditScore'] = np.nan df.fillna(CreditsMedian,inplace=True)
```

```
df.loc[df.NumOfProducts > 3, 'NumOfProducts'] = np.nan
```

```
df.fillna(ProdMedian,inplace=True)
```

df

	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	619.0	0	0	42	2	0.00	1.0	1	1	101348.88	1
1	608.0	2	0	41	1	83807.86	1.0	0	1	112542.58	0
2	502.0	0	0	42	8	159660.80	3.0	1	0	113931.57	1
3	699.0	0	0	39	1	0.00	2.0	0	0	93826.63	0
4	850.0	2	0	43	2	125510.82	1.0	1	1	79084.10	0
...
9995	771.0	0	1	39	5	0.00	2.0	1	0	96270.64	0
9996	516.0	0	1	35	10	57369.61	1.0	1	1	101699.77	0
9997	709.0	0	0	36	7	0.00	1.0	0	1	42085.58	1
9998	772.0	1	1	42	3	75075.31	2.0	1	0	92888.52	1
9999	792.0	0	0	28	4	130142.79	1.0	1	0	38190.78	0

10000 rows x 11 columns

7. Check for Categorical columns and perform encoding.

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

```
df.info()
```

#we have 2 categorial information

```
<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

```
from sklearn.preprocessing import
```

```
LabelEncoder,MinMaxScaler
```

```
labelencoder = LabelEncoder() df['Geography']= labelencoder.fit_transform(df['Geography'])
```

```
df['Gender']= labelencoder.fit_transform(df['Gender'])
```

8. Split the data into dependent and independent variables.

```
x= df.iloc[:, :-1]
```

```
y= df.iloc[:, -1:]
```

9. Scale the independent variables

```
from sklearn.preprocessing import MinMaxScaler nm
```

```
=MinMaxScaler()
```

```
X = nm.fit_transform(x)
```

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)
```