

# 1.Download the dataset: Dataset

## 2.Load the dataset

```
import numpy as np
import pandas as pd
df = pd.read_csv("Churn_Modelling.csv")
```

## 3.Perform Below Visualizations.

### Univariate Analysis

```
import seaborn as sns
sns.histplot(df.CreditScore,kde=True)
```

### Bi -Variate Analysis

```
import seaborn as sns
import matplotlib.pyplot as plt
sns.scatterplot(df.CreditScore,df.EstimatedSalary)
plt.ylim(0,15000)
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as
keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other
arguments without an explicit keyword will result in an error or misinterpretation.
FutureWarning
(0.0, 15000.0)
```

```
df['Geography'].value_counts()
```

France 5014  
Germany 2509  
Spain 2477  
Name: Geography, dtype: int64

## 5.Handle the Missing values.

```
from ast import increment_lineno
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set(color_codes=True)
df=pd.read_csv("Churn_Modelling.csv")
df.head()
```

Row Number	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	France	Female	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	Spain	Female	43	2	125510.82	1	1	1	79084.10	0

```
df.isnull().sum()
```

```
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
```

No missing values here, so no need to perform further operations

## 6. Find the outliers and replace the outliers

```
import pandas as pd
import matplotlib
from matplotlib import pyplot as pyplot
%matplotlib inline
matplotlib.rcParams['figure.figsize']=(10,4)
df=pd.read_csv("Churn_Modelling.csv")
df.sample(5)
```

Row Number	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCard	IsActiveMember	EstimatedSalary	Exited	
648	649	15633064	Stonebraker	438	France	Female	36	4	0.00	2	1	0	64420.50	0
4872	4873	15645937	Guerin	790	Spain	Male	32	3	0.00	1	1	0	91044.47	0
74	7432	15705379	Upjohn	678	France	Male	38	3	0.00	2	1	0	66561.60	0

```
sns.boxplot(x='CreditScore', data=df)
```

## 7. Check for Categorical columns and perform encoding.

```
df=pd.read_csv("Churn_Modelling.csv")
df.columns
import pandas as pd
import numpy as np
headers=['RowNumber','CustomerId','Surname','CreditScore','Geography',
'Gender','Age','Tenure','Balance','NumofProducts','HasCard',
'IsActiveMember','EstimatedSalary','Exited']
import seaborn as sns
df.head()
```

Row Number	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	112542.58	0

2	3	1561 9304	Onio	502	Fran ce	Fem ale	42	8	1596 60.8 0	3	1	0	1139 31.5 7	1
3	4	1570 1354	Boni	699	Fran ce	Fem ale	39	1	0.00	2	0	0	9382 6.63	0
4	5	1573 7888	Mitc hell	850	Spai n	Fem ale	43	2	1255 10.8 2	1	1	1	7908 4.10	0

*#Splitting the Dataset into the Independent Feature Matrix:*

```
X = df.iloc[:, :-1].values
print(X)
[[1 15634602 'Hargrave' ... 1 1 101348.88]
 [2 15647311 'Hill' ... 0 1 112542.58]
 [3 15619304 'Onio' ... 1 0 113931.57]
 ...
 [9998 15584532 'Liu' ... 0 1 42085.58]
 [9999 15682355 'Sabbatini' ... 1 0 92888.52]
 [10000 15628319 'Walker' ... 1 0 38190.78]]
```

*#Extracting the Dataset to Get the Dependent Vector*

```
Y = df.iloc[:, -1].values
print(Y) [1 0 1 ... 1 1 0]
```

## 9.Scale the independent variables

**from sklearn.preprocessing import StandardScaler**

```
object= StandardScaler()
```

*# standardization*

```
scale=object.fit_transform(x)
```

```
print(scale)
```

```
[[-0.78321342]
```

```
 [-0.60653412]
```

```
 [-0.99588476]
```

```
...
```

```
 [-1.47928179]
```

```
 [-0.11935577]
```

```
 [-0.87055909]]
```

**from sklearn.model\_selection import train\_test\_split**

*# split the dataset*

```
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.05, random_state=0)
```

```
X_train
```

```
array([[800, 15567367, 'Tao', ..., 0, 1, 103315.74],
```

```
 [1070, 15628674, 'Iadanza', ..., 1, 0, 31904.31],
```

```
 [8411, 15609913, 'Clark', ..., 1, 0, 113436.08],
```

```
...,
```

```
 [3265, 15574372, 'Hoolan', ..., 1, 0, 181429.87],
```

```
 [9846, 15664035, 'Parsons', ..., 1, 1, 148750.16],
```

```
 [2733, 15592816, 'Udokamma', ..., 1, 0, 118855.26]], dtype=object)
```

Y\_train

```
array([0, 1, 0, ..., 0, 0, 1])
```

X\_test

```
array([[9395, 15615753, 'Upchurch', ..., 1, 1, 192852.67],  
[899, 15654700, 'Fallaci', ..., 1, 0, 128702.1],  
[2399, 15633877, 'Morrison', ..., 1, 1, 75732.25],  
...,  
[492, 15699005, 'Martin', ..., 1, 1, 9983.88],  
[2022, 15795519, 'Vasiliev', ..., 0, 0, 197322.13],  
[4300, 15711991, 'Chiawuotu', ..., 0, 0, 3183.15]], dtype=object)
```

Y\_test

```
array([0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0,  
0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0,  
0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1,  
0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0,  
1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,  
0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0,  
0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0,  
1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0,  
0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1,  
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1,  
0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,  
0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,  
0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,  
0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,  
1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0,  
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0,  
0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0,  
0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0,  
0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0,  
0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0,  
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,  
0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,  
0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0])
```