

ASSIGNMENT 2

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```
# Importing required libraries
```

```
import numpy as np import pandas as pd # Reading
```

```
the dataset df =
```

```
pd.read_csv('/content/Churn_Modelling.csv')
```

```
# Visualizing 1st 50 data
```

```
df.head()
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age
0	1	15634602	Hargrave	619	France	Female	42
1	2	15647311	Hill	608	Spain	Female	41
2	3	15619304	Onio	502	France	Female	42
3	4	15701354	Boni	699	France	Female	39
4	5	15737888	Mitchell	850	Spain	Female	43

	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	\
0	2	0.00	1	1	1	
1	1	83807.86	1	0	1	
2	8	159660.80	3	1	0	
3	1	0.00	2	0	0	
4	2	125510.82	1	1	1	

	EstimatedSalary	Exited
0	101348.88	1
1	112542.58	0
2	113931.57	1
3	93826.63	0
4	79084.10	0

```
# Checking for null values
```

```
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
df.dtypes
```

```
RowNumber      int64
CustomerId     int64
Surname        object
CreditScore    int64
Geography      object
Gender         object
Age           int64
Tenure        int64
Balance       float64
NumOfProducts int64
HasCrCard     int64
IsActiveMember int64
EstimatedSalary float64
Exited        int64
dtype: object
```

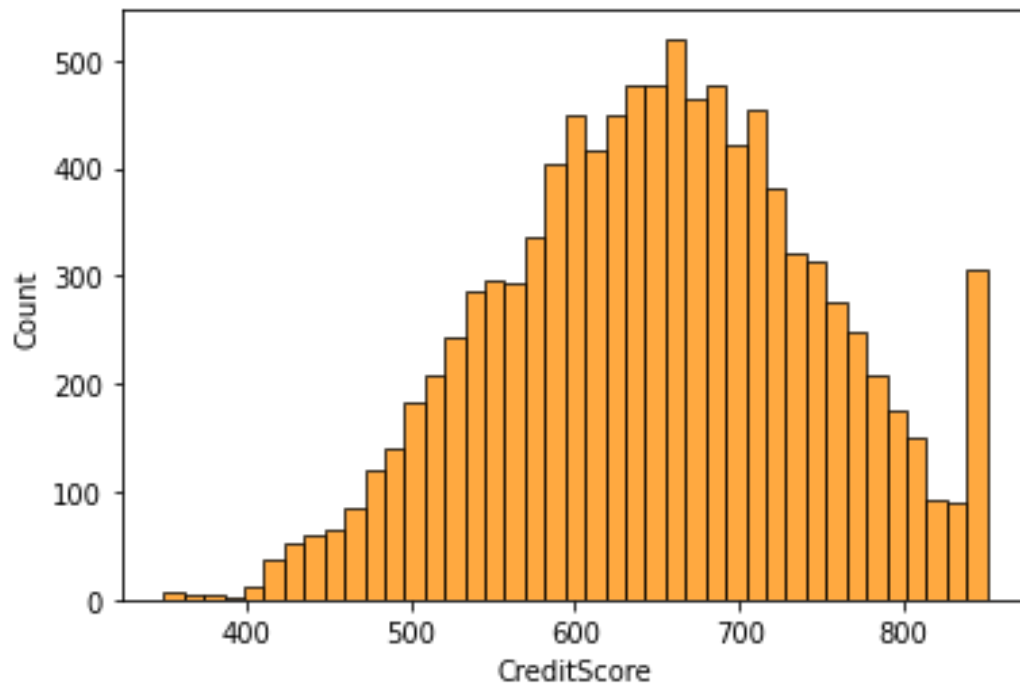
```
import matplotlib.pyplot as plt import seaborn as sns
```

Univariate

Analysis

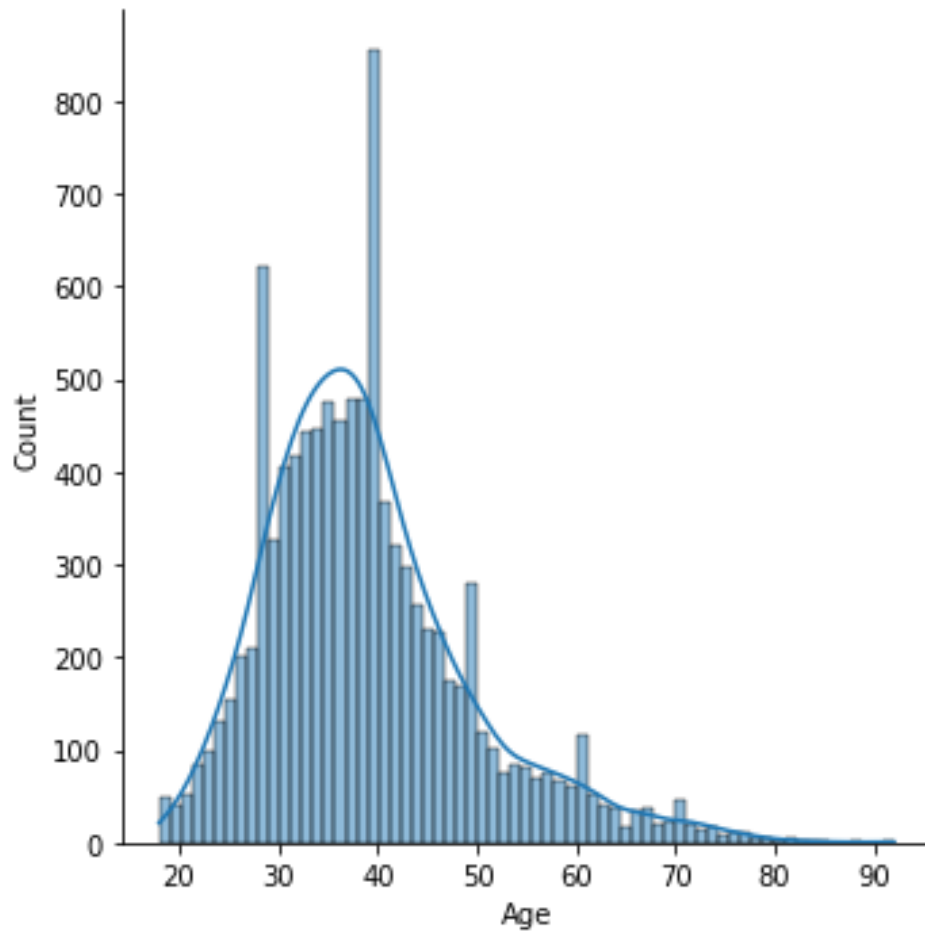
```
sns.histplot(data["CreditScore"],color='darkorange')
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f831677f6d0>
```

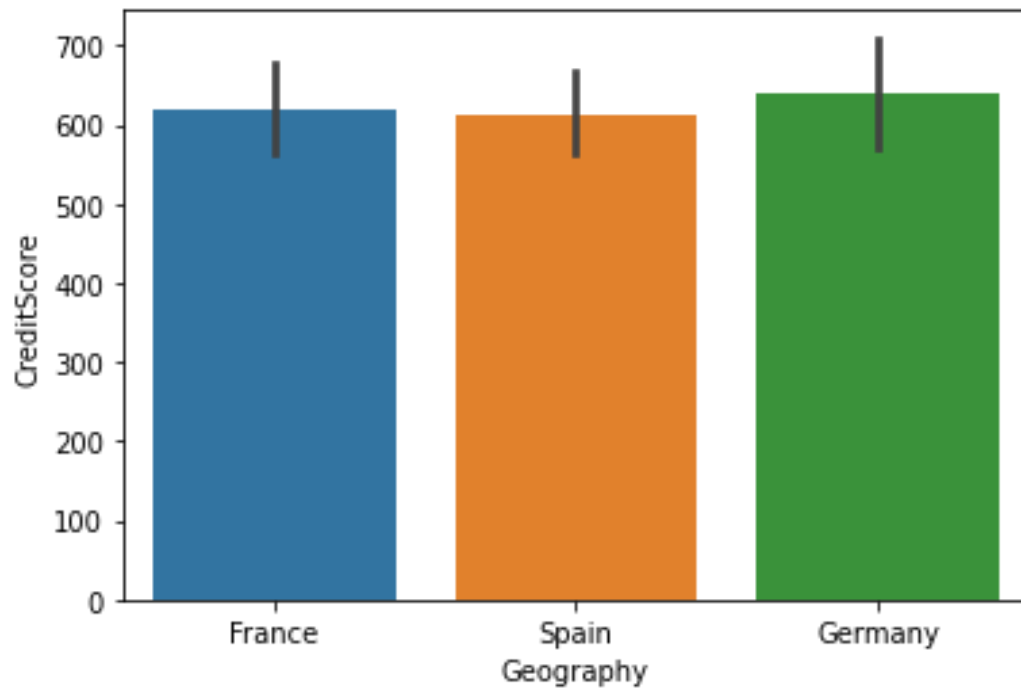


```
sns.displot(data['Age'], kde=True)
```

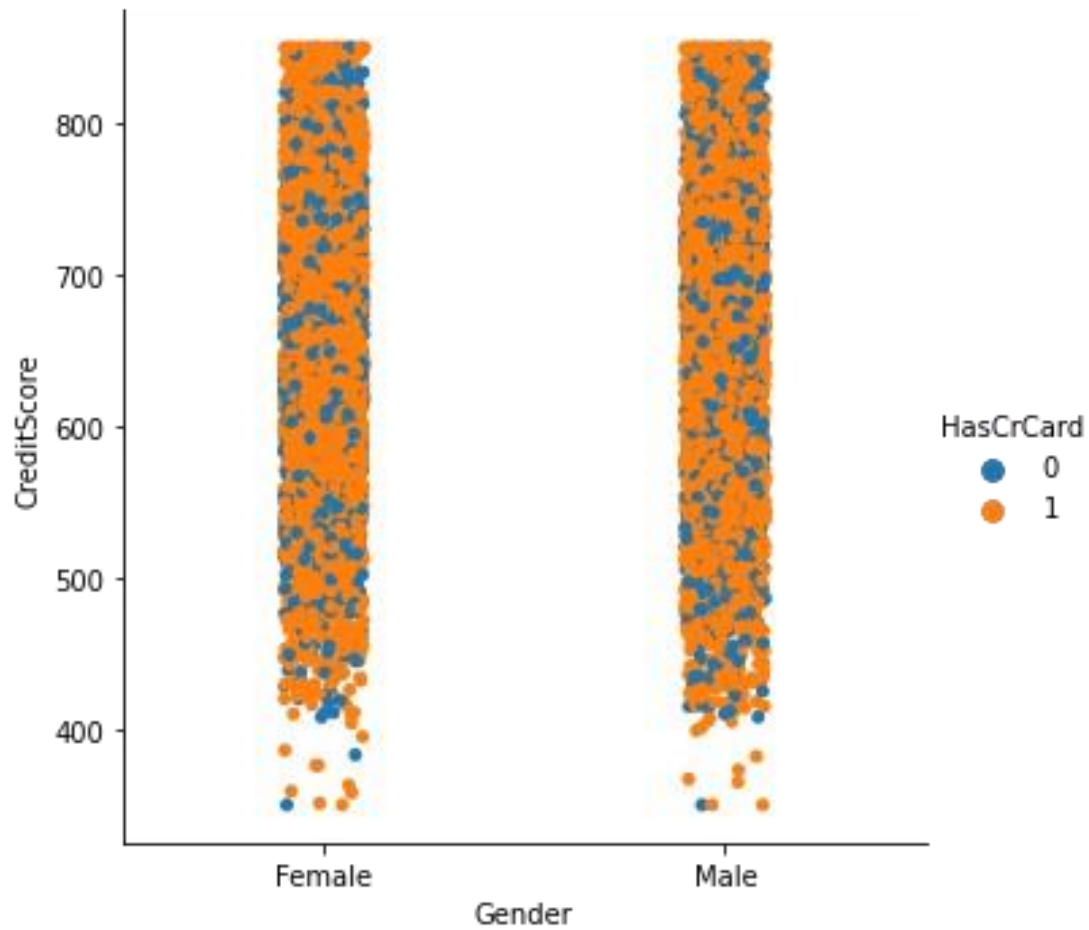
```
<seaborn.axisgrid.FacetGrid at 0x7f831661b210>
```



Bi - Variate Analysis `sns.barplot(data=data.head(50),
x="Geography", y="CreditScore")`
`<matplotlib.axes._subplots.AxesSubplot at 0x7f8313ce63d0>`

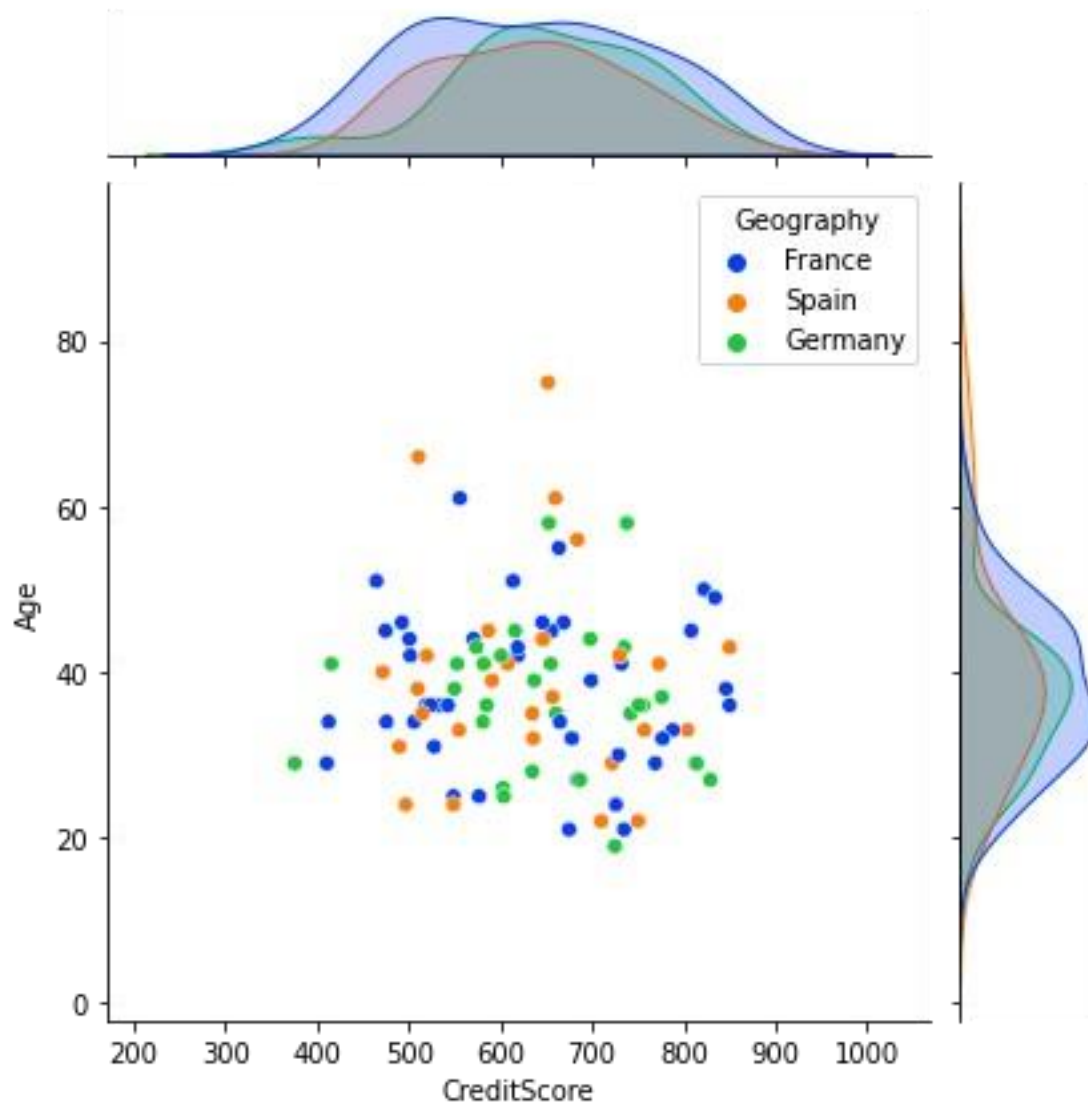


```
sns.catplot(x='Gender', y='CreditScore', hue='HasCrCard', data=data)  
<seaborn.axisgrid.FacetGrid at 0x7f8317198a90>
```



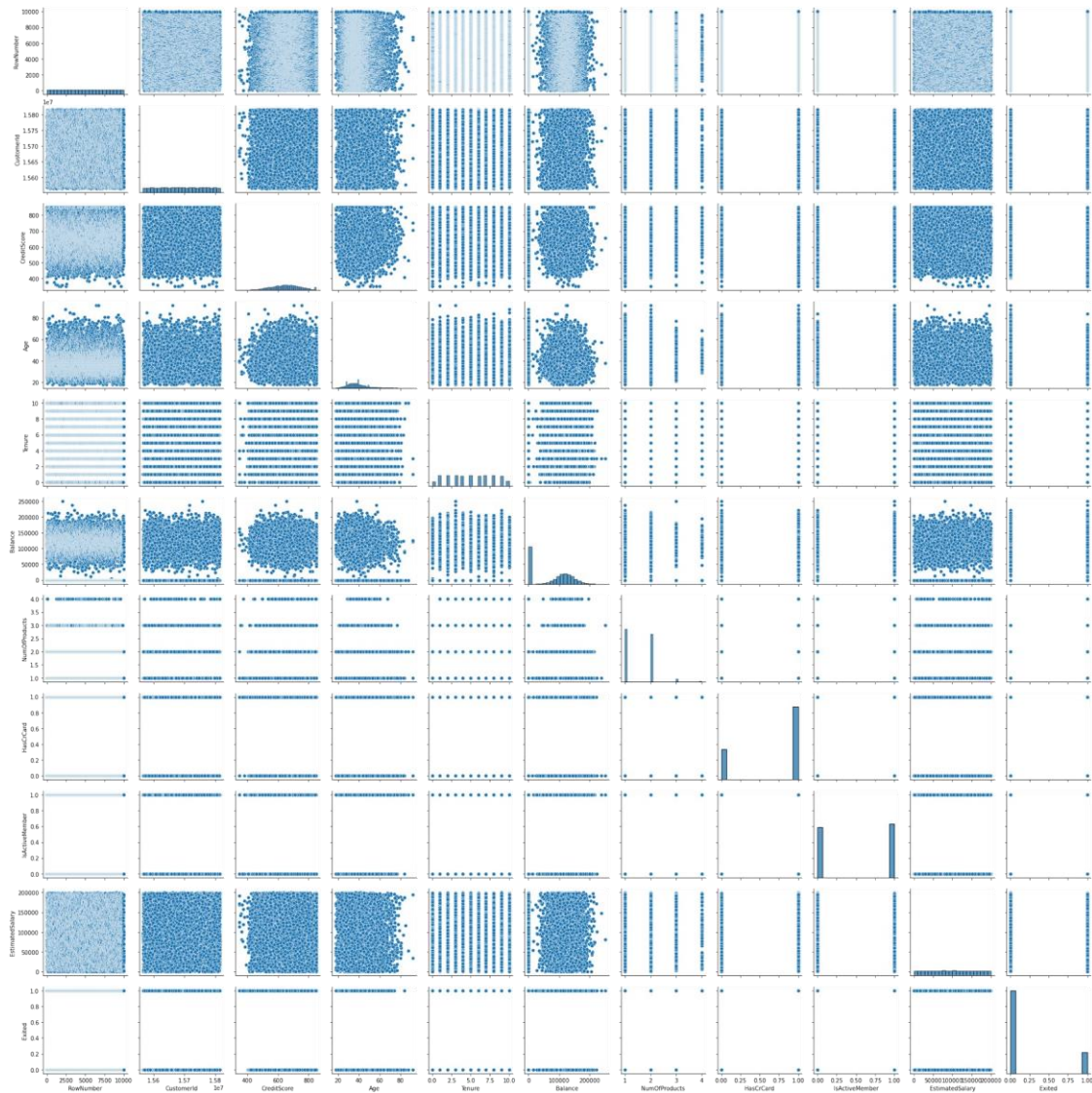
Multi - Variate Analysis

```
sns.jointplot(  
x='CreditScore',    y='Age',  
data=data.head(100),  
palette='bright',  
hue='Geography');
```



```
sns.pairplot(data)
```

```
<seaborn.axisgrid.PairGrid at 0x7f8313a71390>
```



Perform descriptive statistics on the dataset

`data.describe()`

	RowNumber	CustomerId	CreditScore	Age
Tenure \				
count	10000.00000	1.000000e+04	10000.000000	10000.000000
mean	5000.50000	1.569094e+07	650.528800	38.921800
std	2886.89568	7.193619e+04	96.653299	10.487806
min	1.00000	1.556570e+07	350.000000	18.000000

25%	2500.75000	1.562853e+07	584.000000	32.000000
3.000000				
50%	5000.50000	1.569074e+07	652.000000	37.000000
5.000000				
75%	7500.25000	1.575323e+07	718.000000	44.000000
7.000000				
max	10000.00000	1.581569e+07	850.000000	92.000000
10.000000				

	Balance	NumOfProducts	HasCrCard	IsActiveMember	\
count	10000.000000	10000.000000	10000.000000	10000.000000	
mean	76485.889288	1.530200	0.70550	0.515100	std
62397.405202	0.581654	0.45584	0.499797	min	
0.000000	1.000000	0.00000	0.000000	25%	
0.000000	1.000000	0.00000	0.000000		
50%	97198.540000	1.000000	1.00000	1.000000	75%
127644.240000	2.000000	1.00000	1.000000	max	
250898.090000	4.000000	1.00000	1.000000		

	EstimatedSalary	Exited
count	10000.000000	10000.000000
mean	100090.239881	0.203700
57510.492818	0.402769	min
11.580000	0.000000	25%
51002.110000	0.000000	
50%	100193.915000	0.000000
149388.247500	0.000000	max
199992.480000	1.000000	

Handle the Missing values

```
data.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

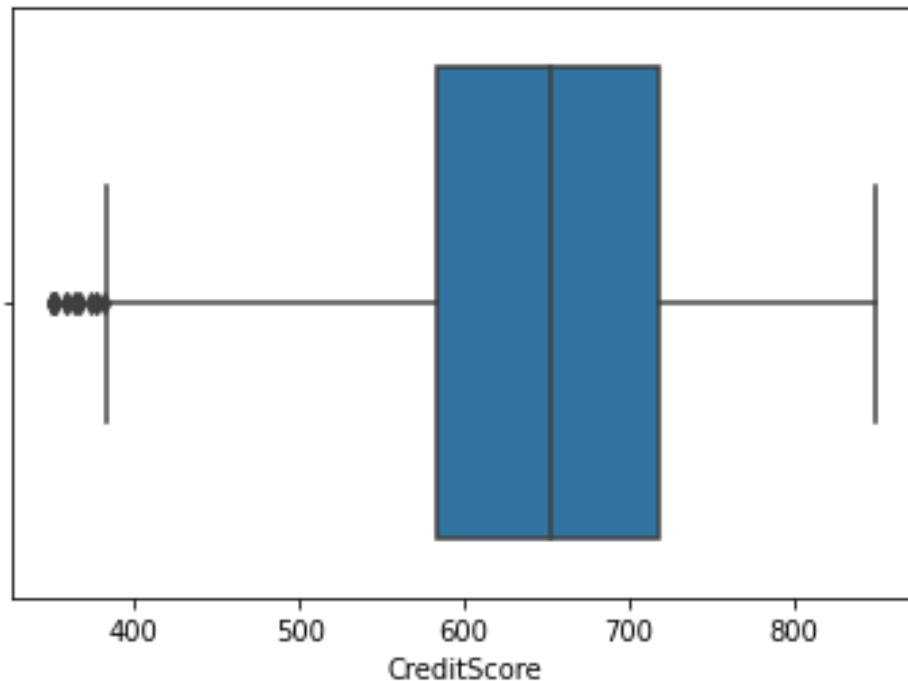
```

Find the outliers and replace the outliers

```
import seaborn as sns
sns.boxplot(data['CreditScore'])
```

```
/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 will be `data`, and
passing other arguments without an explicit keyword will result in an
error or misinterpretation. FutureWarning
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f8310b82990>
```



```
import numpy as np
```

```
Q1 = np.percentile(data['CreditScore'], 25,
interpolation = 'midpoint')
```

```
Q3 = np.percentile(data['CreditScore'], 75,
interpolation = 'midpoint')
```

```
IQR = Q3 - Q1
```

```
#Upper bound
```

```
upper = np.where(data['CreditScore'] >= (Q3+1.5*IQR))
```

```
#Lower bound
```

```
lower = np.where(data['CreditScore'] <= (Q1-1.5*IQR))
```

```
print("Q3: ",Q3)
print("Q1: ",Q1)
print("IQR: ",IQR)
```

```
mean = data["CreditScore"].mean()
```

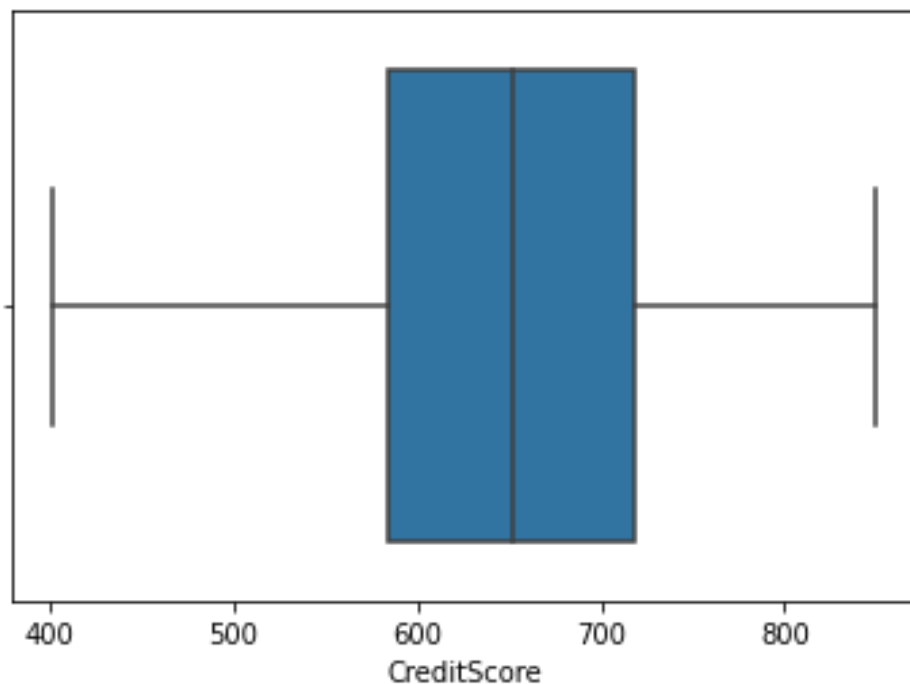
```
data["CreditScore"] = np.where(data["CreditScore"] > 850, mean,
data['CreditScore'])
data["CreditScore"] = np.where(data["CreditScore"] < 400, mean,
data['CreditScore'])
```

```
sns.boxplot(data['CreditScore'])
```

```
Q3: 718.0
Q1: 584.0
IQR: 134.0
```

```
/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 will be `data`, and
passing other arguments without an explicit keyword will result in an
error or misinterpretation. FutureWarning
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f83177a7310>
```



Check for Categorical columns and perform encoding

from sklearn.preprocessing import LabelEncoder

```
le = LabelEncoder()
data['Geography'] = le.fit_transform(data['Geography'])
data['Gender'] = le.fit_transform(data['Gender'])
data.head()
```

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender
Age \						
0	1	15634602	Hargrave	619.0	0	0
42						
1	2	15647311	Hill	608.0	2	0
41						
2	3	15619304	Onio	502.0	0	0
42						
3	4	15701354	Boni	699.0	0	0
39						
4	5	15737888	Mitchell	850.0	2	0
	43					

	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	\
0	2	0.00	1	1	1	
1	1	83807.86	1	0	1	
2	8	159660.80	3	1	0	
3	1	0.00	2	0	0	
4	2	125510.82	1	1	1	

	EstimatedSalary	Exited	0
101348.88		1	
1	112542.58	0	
2	113931.57	1	
3	93826.63	0	
4	79084.10	0	

Split the data into dependent and independent variables

```
y = data['CreditScore'] #dependent
x = data.drop(columns = ['CreditScore'],axis = 1) #independent
x.head()
```

	RowNumber	CustomerId	Surname	Geography	Gender	Age	Tenure
Balance \							
0	1	15634602	Hargrave	0	0	42	2

```

0.00
1      2      15647311      Hill      2      0      41      1
      83807.86
2      3      15619304      Onio      0      0      42      8
      159660.80
3      4      15701354      Boni      0      0      39      1
      0.00
4      5      15737888      Mitchell  2      0      43      2

125510.82

```

	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	1	1	101348.88	1
1	1	0	1	112542.58	0
2	3	1	0	113931.57	1
3	2	0	0	93826.63	0
4	1	1	1	79084.10	0

Scale the independent variables `names`

```
=
```

```
['RowNumber', 'CustomerId', 'Geography', 'Gender', 'Age', 'Tenure', 'Balance',
 'NumOfProducts', 'HasCrCard', 'IsActiveMember', 'EstimatedSalary', 'Exited']
```

```
from sklearn.preprocessing import scale
```

```
x = scale(x[names])
```

```
x
```

```
array([[ -1.73187761,  -0.78321342,  -0.90188624, ...,   0.97024255,
         0.02188649,   1.97716468],
       [ -1.7315312 ,  -0.60653412,   1.51506738, ...,   0.97024255,
         0.21653375,  -0.50577476],
       [ -1.73118479,  -0.99588476,  -0.90188624, ...,  -1.03067011,
         0.2406869 ,   1.97716468],
       ...,
       [  1.73118479,  -1.47928179,  -0.90188624, ...,   0.97024255,
        -1.00864308,   1.97716468],
       [  1.7315312 ,  -0.11935577,   0.30659057, ...,  -1.03067011,
        -0.12523071,   1.97716468],
       [  1.73187761,  -0.87055909,  -0.90188624, ...,  -1.03067011,
        -1.07636976,  -0.50577476]])
```

```
x = pd.DataFrame(x, columns = names)
```

```
x.head()
```

	RowNumber	CustomerId	Geography	Gender	Age	Tenure	Balance \
0	-1.731878	-0.783213	-0.901886	-1.095988	0.293517	-1.041760	-1.225848
1	-1.731531	-0.606534	1.515067	-1.095988	0.198164	-1.387538	0.117350
2	-1.731185	-0.995885	-0.901886	-1.095988	0.293517	1.032908	1.333053
3	-1.730838	0.144767	-0.901886	-1.095988	0.007457	-1.387538	-1.225848
4	-1.730492	0.652659	1.515067	-1.095988	0.388871	-1.041760	0.785728

	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	-0.911583	0.646092	0.970243	0.021886	1.977165
1	-0.911583	-1.547768	0.970243	0.216534	-0.505775
2	2.527057	0.646092	-1.030670	0.240687	1.977165
3	0.807737	-1.547768	-1.030670	-0.108918	-0.505775
4	-0.911583	0.646092	0.970243	-0.365276	-0.505775

Split the data into training and testing from
sklearn.model_selection import train_test_split

Split training and testing data

```
xtrain,xtest,ytrain,ytest =
train_test_split(x,y,test_size=0.20,random_state=0)
# Checking shape of data
```

```
xtrain.shape,xtest.shape
```

```
((8000, 12), (2000, 12))
```