Assignment -2

Data Visualization and Pre-processing

| Assignment Date | 30 September 2022 |
|---------------------|-------------------|
| Student Name | A.Ganeshkumar |
| Student Roll Number | CS19011 |
| Maximum Marks | 2 Marks |

```
In []:
# Importing required libraries
import numpy as np import
pandas as pd
In []:
# Reading the dataset
df = pd.read_csv('/content/Churn_Modelling.csv')
In []:
# Visualizing 1st 50 data
df.head()
Out[]:
```

1 15634602 Hargrave RowNumber CustomerId Surname CreditScore Geography Gender Age Tenure Balance NumOfProducts HasCrCar

619

France Female

42

0.00

| _ | | 1 | 2 1564/311 | J | HIII 6 | 08 | Spain | remaie | 41 1 | 83807. | 86 | 1 |
|-----|---|----------|------------|-----|--------|--------|--------|--------|-----------|--------|------|---|
| 3 2 | 3 | 15619304 | Onio | 502 | Franc | e Fema | ile 42 | 8 | 159660.80 | | 3 | |
| | | | | | | | | | | | | |
| 4 | 5 | 15737888 | Mitchell | 850 | Spai | n Fema | ale 43 | 2 | 125510.82 | | 1 | |
| | | 4 | 15701354 |] | Boni 6 | 99 | France | Female | 39 | 1 | 0.00 | 2 |

In []:

Checking for null values

df.isnull().sum() Out[
]:

RowNumber 0 CustomerId 0 0 Surname CreditScore 0 Geography 0 Gender 0 0 Age Tenure 0 Balance 0 NumOfProducts 0 HasCrCard 0 IsActiveMember 0 0 EstimatedSalary Exited dtype: 0

In []:

int64

df.dtypes

Out[]:

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
```

In [47]:

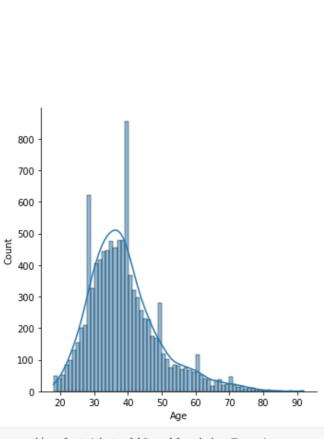
Univariate Analysis

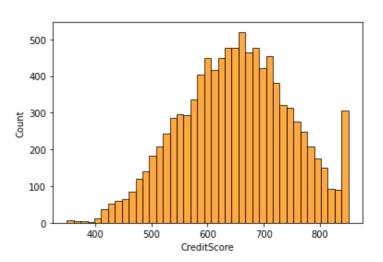
```
sns.histplot(data["CreditScore"],color='darkorange')
```

In [48]:

Out[48]:

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





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

In [49]:

Out[49]:

<seaborn.axisgrid.FacetGrid at 0x7f831661b210>

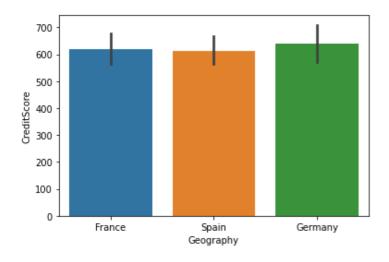
Bi - Variate Analysis

sns.barplot(data=data.head(50), x="Geography", y="CreditScore")

In [50]:

Out[50]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8313ce63d0>

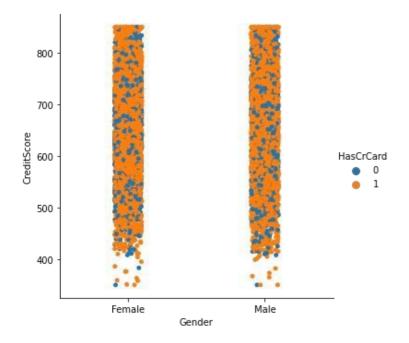


sns.catplot(x='Gender', y='CreditScore', hue='HasCrCard', data=data)

In [51]:

Out[51]:

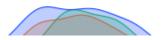
<seaborn.axisgrid.FacetGrid at 0x7f8317198a90>

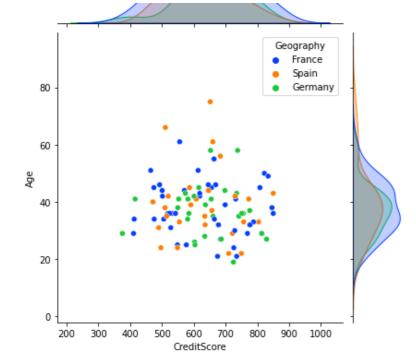


Multi - Variate Analysis

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

In [52]:



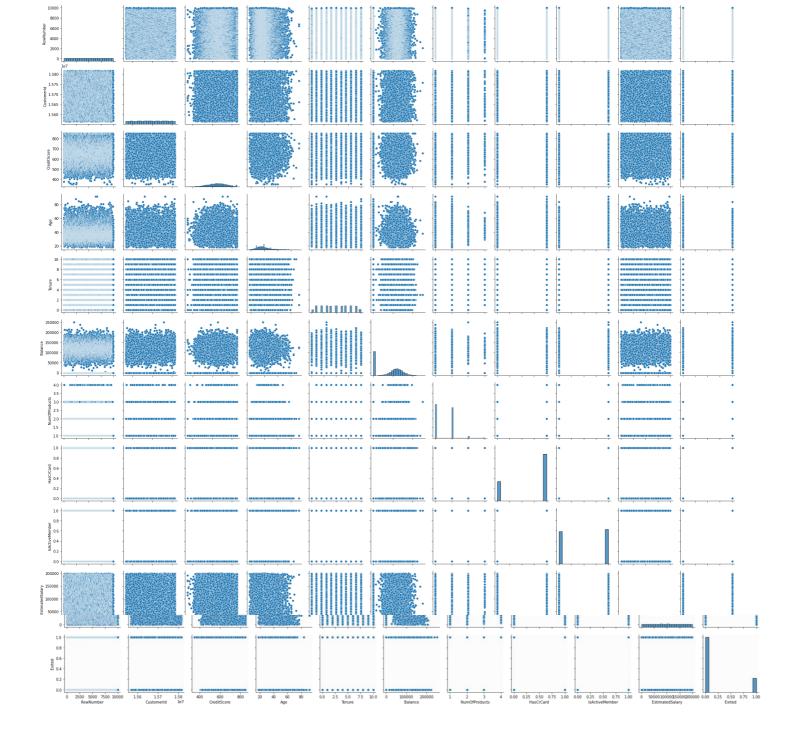


sns.pairplot(data)

In [53]:

Out[53]:

<seaborn.axisgrid.PairGrid at 0x7f8313a71390>



Perform descriptive statistics on the dataset

data.describe()

In [54]:

Out[54]:

RowNum CustomerI CreditScor Age Tenure Balance NumOfProd HasCrCa I ber d e ucts rd s count 10000.000 1.000000e 10000.000 10000.000 10000.000 10000.0000 10000.0000 10000.000 00 +04 000 000 000 00 00

| mean 5 | 5000.5000 | 1.569094e 65 +07 | 0.528800 | 38.921800 | 5.012800 | 76485.8892 88 | | 0.70550 |
|---------------|----------------|----------------------------|----------|--------------|-----------|------------------|----------|---------|
| std 2 | 2886.8956 | 7.193619e 96.653 +04 | 3299 | 10.487806 | 2.892174 | 62397.4052 02 | | 0.45584 |
| min | 1.00000 | 1.556570e 35 +07 | 0.000000 | 18.000000 | 0.000000 | 0.000000 | 1.000000 | 0.00000 |
| 25% 2 | 2500.7500 0 | 1.562853e 58 +07 | 4.000000 | 32.000000 | 3.000000 | 0.000000 | 1.000000 | 0.00000 |
| 50% 5 | 5000.5000 | 1.569074e 65 +07 | 2.000000 | 37.000000 | 5.000000 | 97198.5400 00 | | 1.00000 |
| 75% 7 | 7500.2500 0 | 1.575323e 71 +07 | 8.000000 | 44.000000 | 7.000000 | 127644.240 | 2.000000 | 1.00000 |
| max 10 | 000.000 | 1.581569 +07 00 | | 00 92.000000 | 10.000000 | 250898.090 | 4.000000 | 1.00000 |



Handle the Missing values

data.isnull().sum()

In [55]:

Out[55]:

RowNumber 0 CustomerId 0 Surname 0 CreditScore 0 Geography 0 Gender 0 Age Tenure 0 Balance 0 NumOfProducts HasCrCard 0

IsActiveMember

EstimatedSalary

Exited 0

0

0

0

dtype: int64

Find the outliers and replace the outliers

/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

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

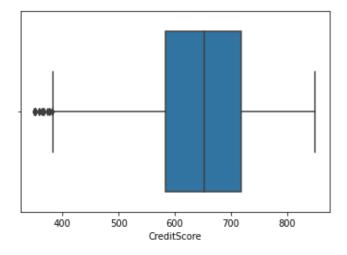
in an error or misinterpretation.

In [56]:

FutureWarning

Out[56]:

<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))</pre>
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'])</pre>
sns.boxplot(data['CreditScore'])
```

In [57]:

Q3: 718.0

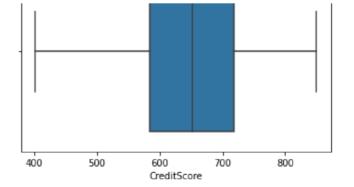
Q1: 584.0

/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

IQR: 134.0 Out[57]:

<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()
```

In [58]:

Out[58]:

| R | owNu | ı <u>mb</u> e | r Custom | erId Suri | name Credi | tScore | Geogr | aphy | Gende | r Age [| Fenure Balance |
|-------|-------|---------------|-------------|-----------|----------------------|--------|-------|------|-------|---------|-----------------------|
| Nu | ımOfl | Produ | ucts1 15634 | 602 HasCr | Car ^{619.0} | 0 | 0 | 42 | 2 | 0.00 | 1 |
| Hargr | ave | | | | | | | | | | |
| | | | | | | | | | | | |
| | | | | | | | | | | | |
| 1 | 2 | 1 | 5647311 | Hill | 608.0 | 2 | 0 | 41 | 1 83 | 807.86 | 1 |
| 2 | | 3 | 15619304 | Onio | 502.0 | 0 | 0 | 42 | 8 159 | 660.80 | 3 |
| | | | | | | | | | | | |
| | | | | | | | | | | | |
| 4 | | 5 | 15737888 | Mitchell | 850.0 | 2 | 0 | 43 | 2 125 | 510.82 | 1 |
| | | | | | | | | | | | |
| | | | | | | | | | | | |
| • | | 4 | 1570125 | 4 D | | 0 | 0 | 20 | 1 | 0.00 | 2 |
| 3 | | 4 | 15701354 | 4 BC | ni 699.0 | 0 | 0 | 39 | 1 | 0.00 | 2 |
| | | | | | | | | | | | |
| 1 | | | | | | | | | | | |
| | | | | | | | | | | | |

Split the data into dependent and independent variables

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

In [59]:

Out[59]:

| | RowNumber Custome | er Surna | a Geograp | Ger | nde Ag | Te | nu Balance | NumOfProd | HasCrC IsActive |
|---|--------------------------|-----------|-----------|-----|--------|----|---------------|-----------|------------------------|
| | Id | me | hy | r | e | re | | ucts | ard Me |
| 0 | 1 15634602 | Hargra ve | 0 | 0 | 42 | 2 | 0.00 | 1 | 1 |
| 1 | 2 15647311 | Hill | 2 | 0 | 41 | 1 | 83807.8 | 1 | 0 |
| 2 | 3 15619304 | Onio | 0 | 0 | 42 | 8 | 159660. 80 | 3 | 1 |
| 3 | 4 15701354 | Boni | 0 | 0 | 39 | 1 | 0.00 | 2 | 0 |
| 4 | 5 15737888 Mitchel | 1 | 2 | 0 | 43 | 2 | 125510. 82 | 1 | 1 |
| | | | | | | | | | |

Scale the independent variables

```
names = ['RowNumber','CustomerId','Geography','Gender','Age','Tenure','Balance','NumOfPro
```

In [60]:

ducts', 'HasCrCard', 'IsActiveMember', 'EstimatedSalary', 'Exited']

In [61]:

from sklearn.preprocessing import scale

```
x = scale(x[names]) x
                    Out[61]:
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]])

In [62]: x = pd.DataFrame(x,columns = names)x

.head()

Out[62]:

| RowNum Cu | stom | er Geogr | ap | Gende A | ge T | | | NumOfProd HasCrCa | |
|---------------------------|-------|------------|----------|-------------|--------|----------|-------|--------------------|-----------|
| ber | Id | hy | r | | | ucts rd | | | mbe |
| | - | | | | | - | | | 0.97024 |
| -1.731878 -0.783213 | - | - 1.0959 0 | 0.29351 | 1.0417 | | 1.2258 | -0 | 0.911583 0.646092 | |
| | 0.90 |)1886 88 | 7 (| 50 | | 48 | | | |
| | - | - | | | | | | | |
| 1 -1.7 | 7315 | 31 -0.606 | 534 1.5 | 15067 1.09 | 59 0.1 | 19816 1. | 3875 | 0.1173 -0.911583 | - 0.97024 |
| | | | 88 | 4 38 | 50 | 1.54 | 7768 | | |
| | | | | - | | | | | |
| 2 -1.731185 -0.995 | 885 | | - 1.0 | 0959 0.293 | 51 1.0 | 329 1 | .3330 | 2.527057 0.646092 | -1.03067 |
| | | 0.90188 | 86 88 | 7 | 08 | 53 | | | |
| | - | | _ | | | | | | |
| 3 -1.7 | 3083 | 8 0.14476 | 67 | - 1.0959 | 0.007 | 45 1.38 | 75 | 1.2258 0.807737 | -1.03067 |
| | | 0.90188 | 86 88 | 7 38 | 48 | 1.54 | 7768 | | |
| | | - | - | | | | | | 0.97024 |
| 4 -1.730492 0.6526 | 559 1 | .515067 1 | 1.0959 (| 0.38887 1.0 |)417 | 0.7857 | 7 | -0.911583 0.646092 | |
| | | | | | 88 | 1 60 | 28 | | |

Split the data into training and testing

In [69]:

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)

```
In [70]:
# Checking shape of data
xtrain.shape,xtest.shape Out[70]:
```

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