Data Visualization and Pre-processing

- 1. Download the dataset: Dataset
- 2. Load the dataset

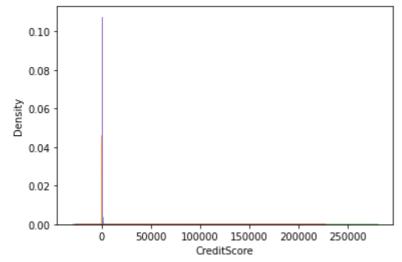
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
data=pd.read_csv('/content/chrun_modelling.csv')
```

3. Perform Below Visualizations

```
sns.kdeplot(data['CreditScore'], shade=True)
sns.kdeplot(data['Age'], shade=True)
sns.kdeplot(data['Balance'], shade=True)
sns.kdeplot(data['EstimatedSalary'], shade=True)
sns.kdeplot(data['Tenure'], shade=True)
```

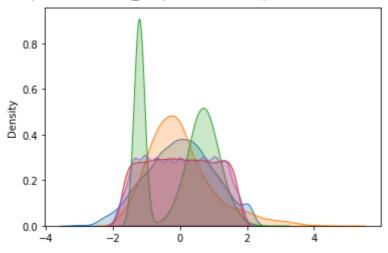


<matplotlib.axes._subplots.AxesSubplot at 0x7fc73ba72d90>



```
from sklearn.preprocessing import StandardScaler
stand= StandardScaler()
for column in ['CreditScore','Age','Balance','EstimatedSalary','Tenure']:
    data[column] = stand.fit_transform(data[column].values.reshape(-1,1))
sns.kdeplot(data['CreditScore'], shade=True)
sns.kdeplot(data['Age'], shade=True)
sns.kdeplot(data['Balance'], shade=True)
sns.kdeplot(data['EstimatedSalary'], shade=True)
sns.kdeplot(data['Tenure'], shade=True)
```

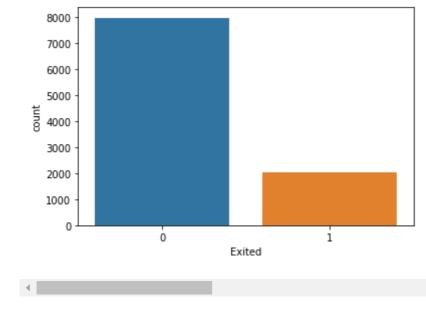
<matplotlib.axes._subplots.AxesSubplot at 0x7fc73b4a6fd0>



sns.countplot(data['Exited'])

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass FutureWarning

<matplotlib.axes._subplots.AxesSubplot at 0x7fc73ac92950>



4. Perform descriptive statistics on the dataset

data.describe()

| 0e+04 1.000000 | 1.000000e+04 |
|-----------------|---------------|
| 9e-16 -6.252776 | -1.078249e-16 |
| 0e+00 1.000050 | 1.000050e+00 |
| 324 | -1.078 |

5 . Handle the Missing values

75% 2500 75000 1 562853a+07 _6 883586a_01 _6 600185a_01 _6 050818a_01 _1 225848 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 | |
| | |

6. Find the outliers and replace the outliers

```
lowerlimit=data['Age'].quantile(0.05)
lowerlimit
data[data['Age']<lowerlimit]
upperlimit=data['Age'].quantile(0.95)
upperlimit
data[data['Age']<upperlimit]
data=data[(data['Age']>lowerlimit)&(data['Age']<upperlimit)]
data</pre>
```

| | RowNumber | CustomerId | Surname | CreditScore | Geography | Gender | Age | Tenure | |
|---|-----------|------------|----------|-------------|-----------|--------|-----|--------|--|
| 0 | 1 | 15634602 | Hargrave | 619 | France | Female | 42 | 2 | |
| 1 | 2 | 15647311 | Hill | 608 | Spain | Female | 41 | 1 | |
| 2 | 3 | 15619304 | Onio | 502 | France | Female | 42 | 8 | |
| 3 | 4 | 15701354 | Boni | 699 | France | Female | 39 | 1 | |

7. Check for Categorical columns and perform encoding

```
x = pd.get_dummies(x)
x.head()
```

| | RowNumber | CustomerId | CreditScore | Age | Tenure | Balance | NumOfProducts | : | | |
|------|-----------------------|------------|-------------|----------|-----------|-----------|---------------|---|--|--|
| 0 | 1 | 15634602 | -0.326221 | 0.293517 | -1.041760 | -1.225848 | 1 | | | |
| 1 | 2 | 15647311 | -0.440036 | 0.198164 | -1.387538 | 0.117350 | 1 | | | |
| 2 | 3 | 15619304 | -1.536794 | 0.293517 | 1.032908 | 1.333053 | 3 | | | |
| 3 | 4 | 15701354 | 0.501521 | 0.007457 | -1.387538 | -1.225848 | 2 | | | |
| 4 | 5 | 15737888 | 2.063884 | 0.388871 | -1.041760 | 0.785728 | 1 | | | |
| 5 rc | 5 rows × 2944 columns | | | | | | | | | |

8. Split the data into dependent and independent variables

```
x = data.iloc[:,0:10]
y = data.iloc[:,10]
print(x.shape)
print(y.shape)
     (10000, 10)
     (10000,)
```

9. Scale the independent variables

```
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test= train_test_split(x,y,test_size=0.25,random_state=0)
sc = StandardScaler()
x_train=sc.fit_transform(x_train)
```

```
x_test = sc.fit_transform(x_test)
x_train = pd.DataFrame(x_train)
x_train.head()
```

| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
|-----------|--|--|--|---|---|--|---|
| -0.702176 | -1.343330 | -0.735507 | 0.015266 | 0.008860 | 0.673160 | 2.535034 | -0.016332 |
| -1.485722 | 1.558330 | 1.024427 | -0.652609 | 0.008860 | -1.207724 | 0.804242 | -0.016332 |
| -0.524522 | -0.655156 | 0.808295 | -0.461788 | 1.393293 | -0.356937 | 0.804242 | -0.016332 |
| -1.167396 | 1.200594 | 0.396614 | -0.080145 | 0.008860 | -0.009356 | -0.926551 | -0.016332 |
| -1.451159 | 0.778798 | -0.467915 | 1.255605 | 0.701077 | -1.207724 | 0.804242 | -0.016332 |
| | -0.702176 -1.485722 -0.524522 -1.167396 | -0.702176 -1.343330 -1.485722 1.558330 -0.524522 -0.655156 -1.167396 1.200594 | -0.702176 -1.343330 -0.735507 -1.485722 1.558330 1.024427 -0.524522 -0.655156 0.808295 -1.167396 1.200594 0.396614 | -0.702176 -1.343330 -0.735507 0.015266 -1.485722 1.558330 1.024427 -0.652609 -0.524522 -0.655156 0.808295 -0.461788 -1.167396 1.200594 0.396614 -0.080145 | -0.702176 -1.343330 -0.735507 0.015266 0.008860 -1.485722 1.558330 1.024427 -0.652609 0.008860 -0.524522 -0.655156 0.808295 -0.461788 1.393293 -1.167396 1.200594 0.396614 -0.080145 0.008860 | -0.702176 -1.343330 -0.735507 0.015266 0.008860 0.673160 -1.485722 1.558330 1.024427 -0.652609 0.008860 -1.207724 -0.524522 -0.655156 0.808295 -0.461788 1.393293 -0.356937 -1.167396 1.200594 0.396614 -0.080145 0.008860 -0.009356 | -0.702176 -1.343330 -0.735507 0.015266 0.008860 0.673160 2.535034 -1.485722 1.558330 1.024427 -0.652609 0.008860 -1.207724 0.804242 -0.524522 -0.655156 0.808295 -0.461788 1.393293 -0.356937 0.804242 -1.167396 1.200594 0.396614 -0.080145 0.008860 -0.009356 -0.926551 |

5 rows × 2944 columns

10. Split the data into training and testing

Colab paid products - Cancel contracts here

