Data Visualization and Pre-processing

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

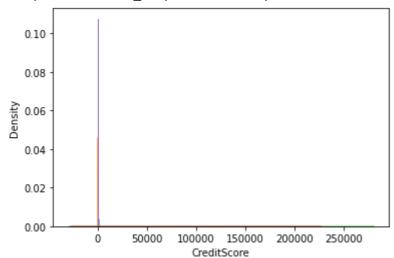
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
data=pd.read_csv('/content/chrun_modelling.csv')
```

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)
```

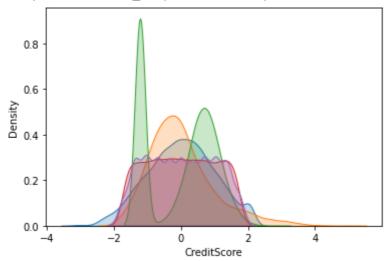


<matplotlib.axes._subplots.AxesSubplot at 0x7fc73ba72d90>



```
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)
```

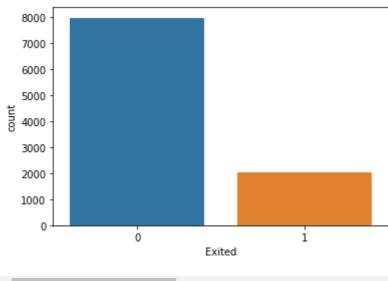
<matplotlib.axes._subplots.AxesSubplot at 0x7fc73b4a6fd0>



sns.countplot(data['Exited'])

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

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



4. Perform descriptive statistics on the dataset

data.describe()

	RowNumber	CustomerId	CreditScore	Age	Tenure	
count	10000.00000	1.000000e+04	1.000000e+04	1.000000e+04	1.000000e+04	1.0000

5. Handle the Missing values

data.isnull().sum()

RowNumber CustomerId 0 Surname CreditScore Geography Gender Age Tenure 0 Balance NumOfProducts HasCrCard 0 IsActiveMember 0 0 EstimatedSalary Exited 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	Tenur
0	1	15634602	Hargrave	619	France	Female	42	:

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

E roug v 2011 columns

10. Split the data into training and testing

Colab paid products - Cancel contracts here