### Assignment -2

Project Name	AI Based Discourse for Banking Industry
Student Name	Pushparaj.D
Student Roll Number	73151921038

# **IMPORT LIBRARIES**

import numpy as np import pandas as pd import matplotlib.pyplot as pltimport seaborn as sns

# LOADING THE DATASET

 $df = pd.read\_csv('Churn\_Modelling.csv', encoding='latin-1')df$ 

	RowNuml	ber	CustomerId	Surna	me	CreditSc	ore	Geography	Gender
Age 0	\	1	1563460	2 Hargr	ave	(	619	France	Female
42 1		2	1564731	1	Hill	(	608	Spain	Female
41 2		3	1561930	)4 C	nio	:	502	France	Female
42 3		4	1570135	54 B	oni	(	699	France	Female
39 4		5	1573788	88 Mitc	hell	:	850	Spain	Female
43		•	•						
 9995 39	99	96	1560622	9 Obijiaku		,	771	France	Male
9996 35	99	97	1556989	Johnst	one		516	France	Male
9997 36	99	98	1558453	32	Liu	,	709	France	Female
9998 42	99	99	1568235	Sabba	tini	,	772	Germany	Male
9999 28	100	00	1562831	9 Wal	ker	,	792	France	Female
20	Tenure		Balance	NumOfProducts		HasCrCard	r	sActiveMember	\
0	2		0.00	Trumon roducts	1	1		37 ICH VCIVICIIIOCI	1
1	1	83	3807.86		1	0			1
2	8		9660.80		3	1			0
3	1		0.00		2	0	1		0
4	2	125	5510.82		1	1			1
9995	5		0.00		2	1			0
9996	10	5	7369.61		1	1			1

9997	7	0.00	1	0	1
9998	3	75075.31	2	1	0
9999	4	130142.79	1	1	0

	EstimatedSalary	Exited
0	101348.88	1
1	112542.58	0
2	113931.57	1
3	93826.63	0
4	79084.10	0
9995	96270.64	0
9996	101699.77	0
9997	42085.58	1
9998	92888.52	1
9999	38190.78	0

[10000 rows x 14 columns]

#### **VISUALIZATIONS**

### #visualization of categorical features

```
fig, ax = plt.subplots(3, 2, figsize = (15, 12))plt.title("Visualization")
sns.countplot('Geography', hue = 'Exited', data = df, ax = ax[0][0],palette='spring')
sns.countplot('Gender', hue = 'Exited', data = df, ax = ax[0][1],palette='spring')
sns.countplot('Tenure', hue = 'Exited', data = df, ax = ax[1][0],palette='spring')
sns.countplot('NumOfProducts', hue = 'Exited', data = df, ax = ax[1][1],palette='spring')
sns.countplot('HasCrCard', hue = 'Exited', data = df, ax = ax[2][0],palette='spring')
sns.countplot('IsActiveMember', hue = 'Exited', data = df, ax = ax[2][1],palette='spring')
```

```
ax[0][0].set_title('Count Plot of Geography',color='red',fontsize=15)ax[0][1].set_title('Count Plot of Gender',color='red',fontsize=15) ax[1][0].set_title('Count Plot of Tenure',color='red',fontsize=15) ax[1][1].set_title('Count Plot of NumOfProducts',color='red',fontsize=15) ax[2][0].set_title('Count Plot of HasCrCard',color='red',fontsize=15)ax[2][1].set_title('Count Plot of IsActiveMember',color='red',fontsize=15)
```

#### plt.tight\_layout()plt.show()

/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 anerror or misinterpretation.

#### **FutureWarning**

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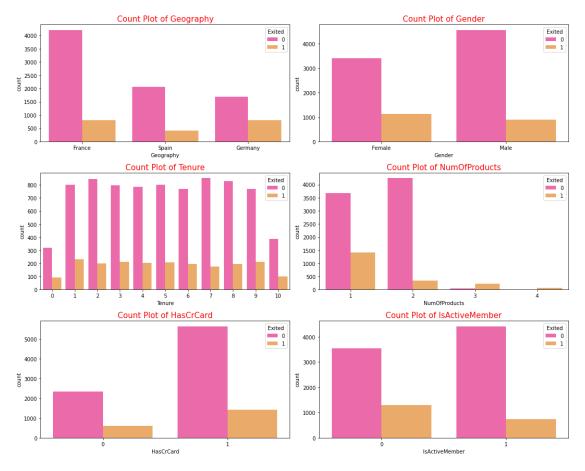
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#### **FutureWarning**



# **DESCRIPTIVE STATISTICS**

# 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 int64l	EstimatedSalary
float64Exited	int64
dtype: object	

$$\label{eq:condition} \begin{split} df\_num &= df[['RowNumber', 'Tenure', 'CustomerId', 'CreditScore', 'Age', 'NumOfProducts', 'HasCrCard', 'IsActiveMember', 'Exited']] \end{split}$$

 $df\_cat = df[['Surname', 'Geography', 'Gender']]df\_num.head()$ 

	RowNumber Tenure CustomerId CreditScore Age NumOfProductsHasCrCard \						
0	1	2	15634602	619	42	1	
1							
1	2	1	15647311	608	41	1	
0							
2	3	8	15619304	502	42	3	
1							
3	4	1	15701354	699	39	2	
0							
4	5	2	15737888	850	43	1	
1							
	IsActiveMember	Exite	d				
0	1		1				
1	1		0				
2	0		1				
3	0		0				
4	1		0				

# df\_cat.head()

# Surname Geography Gender

0	Hargrave	France Female
1	Hill	Spain Female
2	Onio	France Female
3	Boni	France Female
4	Mitchell	Spain Female

# df\_num.describe()

	RowNumber	Tenure	CustomerId	CreditScore						
Age \										
count 100	count 10000.00000 10000.000000 1.000000e+04 10000.000000									
10000.000	000									
mean	5000.50000	5.012800	1.569094e+07	650.528800						
38.921800										
std	2886.89568	2.892174	7.193619e+04	96.653299						
10.487806										
min	1.00000	0.000000	1.556570e+07	350.000000						
18.000000										
25%	2500.75000	3.000000	1.562853e+07	584.000000						
32.000000										
50%	5000.50000	5.000000	1.569074e+07	652.000000						
37.000000										
75%	7500.25000	7.000000	1.575323e+07	718.000000						
44.000000										
max	10000.00000	10.000000	1.581569e+07	850.000000						

#### 92.000000

	NumOfProducts	HasCrCard	IsActiveMember	Exited
count	10000.000000	10000.00000	10000.000000	10000.000000
mean	1.530200	0.70550	0.515100	0.203700
std	0.581654	0.45584	0.499797	0.402769
min	1.000000	0.00000	0.000000	0.000000
25%	1.000000	0.00000	0.000000	0.000000
50%	1.000000	1.00000	1.000000	0.000000
75%	2.000000	1.00000	1.000000	0.000000
max	4.000000	1.00000	1.000000	1.000000

df\_cat.describe(exclude = ['int64','float64'])Surname Geography Gender

 count
 10000
 10000
 10000

 unique
 2932
 3
 2

 top
 Smith
 France
 Male

 FMANDLE THE MASSING VALU
 \$5457

Column	Missing values	
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

print(f"Our target variable is Exited. We can observe that it has onlytwo possible variables: {df['Exited'].unique().tolist()}")

Our target variable is Exited. We can observe that it has only twopossible variables: [1, 0] df.drop(['RowNumber', 'CustomerId', 'Surname'], axis=1, inplace=True)

df.rename(columns=new\_names, inplace=True)df.head()

n	credit_score umber_products		country	gender	age	tenure	balance
0	•	619	France	Female	42	2	0.00
1		608	Spain	Female	41	1	83807.86
1 2		502	France	Female	42	8	159660.80
3		699	France	Female	39	1	0.00
2		850	Spain	Female	43	2	125510.82
1		050	Spain	1 cmaic	73	2	123310.02

	owns_credit_card	is_active_member	estimated_salary ex	ited0	1
		1	101348.88	1	
1		0	1	112542.58	0
2		1	0	113931.57	1
3		0	0	93826.63	0
4		1	1	79084.10	0

### **REPLACE OUTLIERS**

```
def detect_outlier(df):
    outlier = [] threshold =
    3 mean = np.mean(df)std =
    np.std(df) for i in df:
        z_score = (i - mean)/std
    if np.abs(z_score)>threshold:
        outlier.append(i)
    return outlier
    CreditScore_list = df['CreditScore'].tolist()Balance_list =
    df['Balance'].tolist()
```

```
EstimatedSalary_list = df_cat['EstimatedSalary'].tolist()CreditScore_outlier =
 detect outlier(CreditScore list) CreditScore outlier
 Output-[359, 350, 350, 358, 351, 350, 350, 350]
 Balance_outlier = detect_outlier(Balance_list)Balance_outlier
 EstimatedSalary_outlier = detect_outlier(EstimatedSalary_list)
 EstimatedSalary_outlier
print("Shape of Data before removing outliers: {}".format(df.shape))Shape of Data before removing
outliers: (10000, 11)
ENCODING
# Encoding Categorical variables into numerical variables# One Hot Encoding
x = pd.get\_dummies(x)x.head()
x.shape
(10000, 13)
SPLIT THE DATA INTO DEPENDENT AND INDEPENDENT VARIALBLES
# splitting the dataset into x(independent variables) and y(dependent variables)
x = df.iloc[:,0:10]
```

### SCALE THE INDEPENDENT VARIABLES

from sklearn.preprocessing import StandardScalersc = StandardScaler()

### $x_{train} = pd.DataFrame(x_{train})$ $x_{train.head()}$

	credit_score products \	count	ry gender	age tenure		balance
2967 3		579	Germany	Female	39	5 117833.30
700		750	France	Female	32	5 0.00
2 3481		729	Spain	Female	34	9 53299.96
2 1621		689	Spain	Male	38	5 75075.14
1 800		605	France	Male	52	7 0.00
2						
	owns_credit_	_card	is_a	ctive_member		estimated_salary
2967			0		0	5831.00
700			1		0	95611.47
3481			1		1	42855.97
1621			1		1	8651.92
800			1		1	173952.50

### SPLIT THE DATA INTO TRAINING AND TESTING

# splitting the data into training and testing set

```
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)

print(x_train.shape)
print(y_train.shape)
print(x_test.shape)
print(y_test.shape)

(7500, 10)
```

(7500,) (2500, 10) (2500,)