## Assignment -2

Assignment Date	17 September 2022
Team ID	PNT2022TMID38845
Project Name	AI Based Discourse for Banking Industry
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Student Roll Number	421219104019
Maximum Marks	2 Marks

## **IMPORT LIBRARIES**

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

## LOADING THE DATASET

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

	RowNumber	CustomerId		Surna	me	Credit	tScore	Geography	Gender
Age	\								
0	1	15634	502	Hargra	ve	619	France	Female	
42									
1	2	15647	311	Hill	608	Spain	Female		
41									
2	3	15619	304	Onio	502	France	Female		
42									
3	4	15701	354	Boni	699	France	Female		
39									
4	5	15737	888	Mitche	11	850	Spain	Female	
43									
•••	•••							•••	•••
	0005	1.50.500	01 1			_			
9995	9996	15606229	Obijiak	tu	771	France	Male		
39	0007	1556000	T. 1		<b>71</b> 6		3.6.1		
9996	9997	15569892	Johnsto	one	516	France	Male		
35	0000	15504522	τ.	700	Г	г 1			
9997 36	9998	15584532	Liu	709	France	Female	;		
	0000	15600055	0.11		770	C		N. 1	
9998 42	9999	15682355	Sabbati	ını	772	German	ny	Male	
9999	10000	15628	210	Walker	702	Emamaa	Female		
28	10000	13028.	519	waiker	192	France	remaie		
20									
	Tenure	Balance	NumOfF	Producte	Ц	asCrCaro	d Ica	ActiveMember	\
0		0.00 1	1	1	110	iscreare	1 157	Activelyiember	\
1		83807.86	1	0	1				
2		159660.80	3	1	0				
3		0 2 0 0 4 2 125	_	_					
9995	5	0.00 2	1	0					

9996	10	57369.61	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 an error or misinterpretation.

### **FutureWarning**

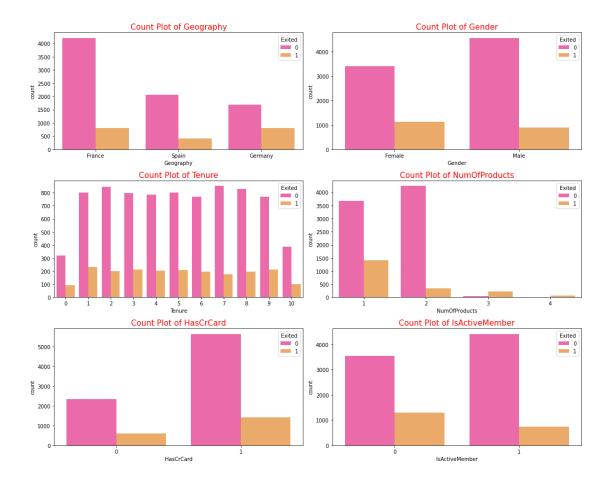
/usr/local/lib/python3.7/dist-packages/seaborn/\_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation. FutureWarning

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/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

/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

/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



### **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 int64 EstimatedSalary

float64 Exited int64

dtype: object

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

18.000000

df_c	$df\_cat = df[['Surname', 'Geography', 'Gender']] df\_num.head()$									
RowNumber Tenure CustomerId CreditScore Age NumOfProducts HasCrCard \										
0 1		1	2	156	34602	619	42	1		
1		2	1	156	47311	608	41	1		
0 2		3	8	156	19304	502	42	3		
1 3		4	1	157	01354	699	39	2		
0		4	1	137	01334	099	39	2		
4 1		5		2	157378	88		850	43	1
1										
	Is	ActiveMe	mber	Exite	d					
0	1	1								
1	1	0								
2	0	1								
3	0	0 4	1	0						
df_ca	t.head(	)								
Surna	ıme Ge	ography	Gender	•						
0	Н	argrave	Fran	ice Fem	ale					
1	1 Hill Spain Female									
2	O	nio Franc	e Fema	ıle						
3	В	oni Franc	e Fema	ıle						
4	M	litchell	Spai	n Fema	le					
df_nu	ım.deso	cribe()								
		RowN	umber		Tenu	ire	Custon	nerId	CreditScore	
Age \										
count 10000.00000 10000.000000 1.000000e+04 10000.000000										
10000	0.0000	00								
mean		5000.50	000		5.0128	00 1.569	094e+07		650.528800	
38.92	1800									
std		2886.89	568		2.8921	74 7.193	619e+04		96.653299	
10.48	7806									
min		1.00	0000		0.0000	00 1.556	570e+07		350.000000	

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
92.00000	0		

	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

HANDLEfreq THE MISSING32 VALUES 5014 5457

print("Column Missing values") print(" ......") df.isnull().sum()

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 only two possible variables:
 {df['Exited'].unique().tolist()}")
 Our target variable is Exited. We can observe that it has only two possible variables: [1, 0]
 df.drop(['RowNumber', 'CustomerId', 'Surname'], axis=1, inplace=True)
 new\_names = {
       'CreditScore': 'credit_score', 'Geography':
       'country', 'Gender': 'gender',
       'Age': 'age',
       'Tenure': 'tenure', 'Balance':
       'balance'.
       'NumOfProducts': 'number_products', 'HasCrCard':
       'owns_credit_card', 'IsActiveMember': 'is_active_member',
       'EstimatedSalary': 'estimated_salary', 'Exited': 'exited'
  }
 df.rename(columns=new_names, inplace=True) df.head()
      credit_score
                                                                           balance
                         country
                                      gender
                                                            tenure
                                                   age
number_products
0
                           France
                                      Female
                                                    42
                                                                 2
                                                                              0.00
                    619
1
 1
                    608
                                Spain Female 41
                                                               83807.86
1
 2
                                France Female 42
                    502
                                                        8
                                                                159660.80
3
 3
                    699
                                France Female 39
                                                               0.00
2
4
                    850
                             Spain
                                                    43
                                                                 2
                                                                        125510.82
                                      Female
1
        owns_credit_card is_active_member estimated_salary exited 0
                                                                                             1
                                                      101348.88
                            1
                                                                     1
 1
                            0
                               1
                                                       0
                                        112542.58
 2
                            1
                               0
                                        113931.57
                                                        1
 3
                               0
                                        93826.63
                                                        0
                                        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]
y = df.iloc[:,10]
print(x.shape) print(y.shape)
print(x.columns) #print(y)
(10000, 10)
(10000,)
Index(['credit_score', 'country', 'gender', 'age', 'tenure', 'balance',
           'number products',
                                   'owns credit card',
                                                            'is active member',
                                                                                      'estimated salary'],
         dtype='object')
```

### SCALE THE INDEPENDENT VARIABLES

from sklearn.preprocessing import StandardScaler sc = StandardScaler()

$x_{train} = pd.DataFrame(x_{train}) x_{train.head}()$							
credit_score						balance	
$number\_products \setminus$							
2967	579	Germany	Female	39	5	117833.30	
3							
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	
000	003	Tallee	Maie	32	/	0.00	

	owns_credit_card	is_active_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

(2500,)

```
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)
```