Assignment -2

Project Name	AI Based Discourse for Banking Industry
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IMPORT LIBRARIES

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

LOADING THE DATASET

 $df = pd.read_csv('Churn_Modelling.csv', encoding='latin-1')df$

	RowNumb	er Custon	nerId	Surname	CreditSo	core	Geography	Gender
Age 0	\	1 156	34602	Hargrave		619	France	Female
42		1 100		11u1gru (01)	1 1 1 1 1 1 1	
1		2 156	47311	Hill		608	Spain	Female
41 2		3 156	19304	Onio		502	France	Female
42								
3		4 157	01354	Boni		699	France	Female
39 4		5 157	37888	Mitchell		850	Spain	Female
43		5 157.	37000	1/11/01/01		020	Spain	Tomare
	•••		•••	•••		•••	•••	•••
 9995	999	96 156	06229	Obijiaku		771	France	Male
39				· ·				
9996 35	999	97 155	69892	Johnstone		516	France	Male
33 9997	999	98 155	84532	Liu		709	France	Female
36								
9998	999	99 156	82355	Sabbatini		772	Germany	Male
42 9999	1000	00 156	28319	Walker		792	France	Female
28	1000	, 0	_0017	, , , , , , , , , , , , , , , , , , , 			1 1001100	1 01110110
	Tenure	Balance	Ni	ımOfProducts	HasCrCard	I o	ActiveMember	\
0	2	0.00		1		1	Activelylember	1
1	1	83807.86	•	1)		1
2	8	159660.80		3		1		0
3	1	0.00)	2	(0		0
4	2	125510.82		1		1		1
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 anerror 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 anerror or misinterpretation.

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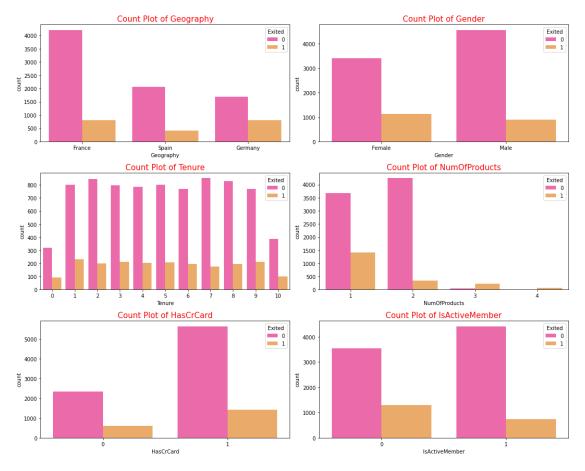
FutureWarning

/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 anerror or misinterpretation.

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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 10	0000.00000 10000.0	00000 1.000000e-	+04 10000.000000	
10000.00	00000			
mean	5000.50000	5.012800 1.	569094e+07	650.528800
38.92180	00			
std	2886.89568	2.892174 7.	193619e+04	96.653299
10.48780				
min	1.00000	0.000000 1.	556570e+07	350.000000
18.00000	00			
25%	2500.75000	3.000000 1	562853e+07	584.000000
32.00000				
50%	5000.50000	5.000000 1	569074e+07	652.000000
37.00000	00			
75%	7500.25000	7.000000 1	575323e+07	718.000000
44.00000	00			
max	10000.00000	10.000000 1.5	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]
```

```
y = df.iloc[:,10]

print(x.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'],

SCALE THE INDEPENDENT VARIABLES

dtype='object')

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

(2500, 10) (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,)
```