

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	421219104015
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')
```

	RowNumber	CustomerId	Surname		CreditScore		Geography	Gender
Age \								
0	1	15634602	Hargrave		619	France	Female	
42								
1	2	15647311	Hill	608	Spain	Female		
41								
2	3	15619304	Onio	502	France	Female		
42								
3	4	15701354	Boni	699	France	Female		
39								
4	5	15737888	Mitchell		850	Spain	Female	
43								
...
...								
9995	9996	15606229	Obijiaku	771	France	Male		
39								
9996	9997	15569892	Johnstone	516	France	Male		
35								
9997	9998	15584532	Liu	709	France	Female		
36								
9998	9999	15682355	Sabbatini	772	Germany		Male	
42								
9999	10000	15628319	Walker	792	France	Female		
28								

	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	
0	2	0.00	1	1		
1	1	83807.86	1	0	1	
2	8	159660.80	3	1	0	
3	1	0.00	2	0	4	125510.82
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.

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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    int64
EstimatedSalary   float64
Exited            int64
dtype: object

```

```

df_num = df[['RowNumber','Tenure','CustomerId','CreditScore','Age','NumOfProducts',
            'HasCrCard','IsActiveMember','Exited']]

```

```
df_cat = df[['Surname','Geography','Gender']] df_num.head()
```

	RowNumber	Tenure	CustomerId	CreditScore	Age	NumOfProducts	HasCrCard \
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	Exited
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	10000.00000	10000.000000	1.000000e+04	10000.000000
				10000.000000
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
```

```
HANDLEfreq THE MISSING32          VALUES5014 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	country	gender	age	tenure	balance
0	619	France	Female	42	2	0.00
1						
1	608	Spain	Female	41	1	83807.86
1						
2	502	France	Female	42	8	159660.80
3						
3	699	France	Female	39	1	0.00
2						
4	850	Spain	Female	43	2	125510.82
1						

	owns_credit_card	is_active_member	estimated_salary	exited
0	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 VARIABLES

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	country	gender	age	tenure	balance
number_products \						
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
2						

	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

```
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,)
```