# MACHINE LEARNING LAB

### **EXERCISE 3**

### Aim:

Use the dataset, perform necessary pre-processing and build a logistic regression model. divide the train data itself into 70-30 ratio and print the performance metrics

## Algorithm:

- 1. Load the dataset.
- 2. Perform necessary pre-processing steps (e.g., handle missing values, encode categorical variables).
- 3. Split the dataset into training and testing sets with a 70-30 ratio.
- 4. Build a logistic regression model using the training data.
- 5. Evaluate the model using performance metrics (e.g., accuracy, precision, recall, F1-score) on the testing set and print the results.

# **Code and Output:**

```
In [1]:
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          import seaborn as sns
In [2]:
          data=pd.read_csv(r"C:\Users\TEJU\Downloads\telecom_customer_churn (1).csv")
In [3]:
          from sklearn import preprocessing
In [4]:
          data.head()
Out[4]:
                                                                                                 Number
                                              Number of
            Customer
                                                                     Zip
                                                                           Latitude
                       Gender Age Married
                                                                                      Longitude
                                                                                                      of
                   ID
                                             Dependents
                                                                    Code
                                                                                                Referrals
                0002-
                                                            Frazier
         0
                       Female
                                                      0
                                                                   93225
                                37
                                                                          34.827662
                                                                                   -118.999073
                                                                                                       2
                                        Yes
               ORFBO
                                                              Park
                0003-
                                                          Glendale 91206 34.162515 -118.203869
                                                                                                       C
                         Male
                                46
                                         No
               MKNFE
                0004-
                                                             Costa
         2
                         Male
                                50
                                         No
                                                      0
                                                                   92627 33.645672 -117.922613
                                                                                                       C
                TLHLJ
                                                             Mesa
```

3

4

0011-

**IGKFF** 

0013-

Male

Female

78

75

Yes

Yes

3

Martinez 94553 38.014457 -122.115432

0 Camarillo 93010 34.227846 -119.079903

Customer Gender Age Married Number of City Zip Latitude Longitude of ID Referrals

**EXCHZ** 

5 rows × 38 columns

In [5]:
 data=data.drop(['Customer ID', 'City', 'Zip Code', 'Latitude', 'Longitude', 'Churn C
 data.head()

Out[5]: Avg Number **Tenure** Monthly **Number of Phone** Multiple Gender Age Married in Offer Long Cc Service **Dependents** Lines Referrals **Distance Months** Charges 0 2 0 Female 37 Yes 9 None Yes 42.39 No Oı 1 Male 46 0 0 None Yes 10.69 No Yes ١ Offer 2 Male 50 0 0 Yes 33.65 No No Offer 3 0 13 27.82 Male 78 Yes 1 Yes No D

0

5 rows × 31 columns

Female

75

Yes



3

3

None

7.38

No

Yes

In [6]: data.shape

Out[6]: (6589, 31)

In [7]:
 data['Avg Monthly Long Distance Charges'] = data['Avg Monthly Long Distance Charges'
 data['Avg Monthly GB Download'] = data['Avg Monthly GB Download'].fillna(0)
 data['Multiple Lines'] = data['Multiple Lines'].fillna('No')

In [8]:
 cols = ['Internet Type', 'Online Security', 'Online Backup', 'Device Protection Plan
 data[cols] = data[cols].fillna('No')
 data.head()

Out[8]:

Out[9]:

	Gender	Age	Married	Number of Dependents	Number of Referrals	Tenure in Months	Offer	Phone Service	Avg Monthly Long Distance Charges	Multiple Lines	•••	Cc
	<b>)</b> Female	37	Yes	0	2	9	None	Yes	42.39	No		Oı
	<b>I</b> Male	46	No	0	0	9	None	Yes	10.69	Yes		1
i	2 Male	50	No	0	0	4	Offer E	Yes	33.65	No		١
:	<b>3</b> Male	78	Yes	0	1	13	Offer D	Yes	27.82	No		1
	<b>1</b> Female	75	Yes	0	3	3	None	Yes	7.38	No		1

5 rows × 31 columns

```
In [9]:
    obj = list(data.select_dtypes(include=['object']).columns)
    data[obj]
    label_encoder = preprocessing.LabelEncoder()
    for col in obj:
        data[col] = label_encoder.fit_transform(data[col])
    data.head()
```

Avg Number Monthly Tenure **Number of** Phone Multiple Gender Age Married in Offer of Long **Dependents** Service Lines Referrals Months **Distance** Charges 0 2 0 0 37 1 9 0 42.39 0 1 46 0 0 0 9 0 1 10.69 2 50 0 0 0 4 5 1 33.65 0 3 78 1 0 1 1 13 4 27.82 0 75 1 0 3 3 0 1 7.38 0

5 rows × 31 columns

```
In [10]: X = data.iloc[:,:-1].values
Y = data.iloc[:,-1].values
```

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.3, random_stat
```

```
In [12]:
          from sklearn.linear_model import LogisticRegression
          model = LogisticRegression(random_state=42, max_iter=5000)
          model.fit(X_train, y_train)
          y_pred = model.predict(X_test)
          print('Y_pred', y_pred)
          print('Y_test', y_test)
         Y_pred [0 1 1 ... 0 1 0]
         Y_test [0 1 1 ... 1 1 0]
         C:\Users\TEJU\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:469: Conv
         ergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
           n_iter_i = _check_optimize_result(
In [13]:
          from sklearn.metrics import accuracy_score,confusion_matrix
          accuracy=accuracy_score(y_test,y_pred)*100
          print("Accuracy of the model is {:.2f}".format(accuracy))
         Accuracy of the model is 84.52
In [14]:
          from sklearn.metrics import f1_score
          f1=f1_score(y_test,y_pred,average="macro")*100
          print("F1 score of the model is {:.2f}".format(f1))
         F1 score of the model is 80.94
In [15]:
          from sklearn.metrics import recall_score
          recall=recall score(y test,y pred,average="macro")*100
          print("Recall score of the model is {:.2f}".format(recall))
         Recall score of the model is 80.62
In [16]:
          from sklearn.metrics import precision_score
          precision=precision_score(y_test,y_pred,average="macro")*100
          print("Precision score of the model is {:.2f}".format(precision))
```

Precision score of the model is 81.29

#### Result:

Therefore, we were successfully able to build a logistic regression model for the given dataset and compute the performance metrics of the same.