

```
In [1]: import pandas as pd
import warnings
warnings.filterwarnings("ignore")
```

```
In [2]: data=pd.read_csv("/home/placement/Downloads/TelecomCustomerChurn.csv")
```

```
In [3]: data.describe()
```

Out[3]:

|       | SeniorCitizen | tenure      | MonthlyCharges |
|-------|---------------|-------------|----------------|
| count | 7043.000000   | 7043.000000 | 7043.000000    |
| mean  | 0.162147      | 32.371149   | 64.761692      |
| std   | 0.368612      | 24.559481   | 30.090047      |
| min   | 0.000000      | 0.000000    | 18.250000      |
| 25%   | 0.000000      | 9.000000    | 35.500000      |
| 50%   | 0.000000      | 29.000000   | 70.350000      |
| 75%   | 0.000000      | 55.000000   | 89.850000      |
| max   | 1.000000      | 72.000000   | 118.750000     |

In [4]: data.head()

Out[4]:

|   | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines    | InternetService | OnlineSecurity | ... | DeviceProtec |
|---|------------|--------|---------------|---------|------------|--------|--------------|------------------|-----------------|----------------|-----|--------------|
| 0 | 7590-VHVEG | Female | 0             | Yes     | No         | 1      | No           | No phone service | DSL             | No             | ... |              |
| 1 | 5575-GNVDE | Male   | 0             | No      | No         | 34     | Yes          | No               | DSL             | Yes            | ... |              |
| 2 | 3668-QPYBK | Male   | 0             | No      | No         | 2      | Yes          | No               | DSL             | Yes            | ... |              |
| 3 | 7795-CFOCW | Male   | 0             | No      | No         | 45     | No           | No phone service | DSL             | Yes            | ... |              |
| 4 | 9237-HQITU | Female | 0             | No      | No         | 2      | Yes          | No               | Fiber optic     | No             | ... |              |

5 rows × 21 columns



In [5]: data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7043 entries, 0 to 7042
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype
---  -
0   customerID            7043 non-null   object
1   gender                 7043 non-null   object
2   SeniorCitizen          7043 non-null   int64
3   Partner                7043 non-null   object
4   Dependents             7043 non-null   object
5   tenure                 7043 non-null   int64
6   PhoneService           7043 non-null   object
7   MultipleLines           7043 non-null   object
8   InternetService        7043 non-null   object
9   OnlineSecurity         7043 non-null   object
10  OnlineBackup            7043 non-null   object
11  DeviceProtection       7043 non-null   object
12  TechSupport            7043 non-null   object
13  StreamingTV            7043 non-null   object
14  StreamingMovies        7043 non-null   object
15  Contract               7043 non-null   object
16  PaperlessBilling       7043 non-null   object
17  PaymentMethod          7043 non-null   object
18  MonthlyCharges         7043 non-null   float64
19  TotalCharges           7043 non-null   object
20  Churn                  7043 non-null   object
dtypes: float64(1), int64(2), object(18)
memory usage: 1.1+ MB
```

In [6]: data.shape

Out[6]: (7043, 21)

```
In [7]: list(data)
```

```
Out[7]: ['customerID',  
         'gender',  
         'SeniorCitizen',  
         'Partner',  
         'Dependents',  
         'tenure',  
         'PhoneService',  
         'MultipleLines',  
         'InternetService',  
         'OnlineSecurity',  
         'OnlineBackup',  
         'DeviceProtection',  
         'TechSupport',  
         'StreamingTV',  
         'StreamingMovies',  
         'Contract',  
         'PaperlessBilling',  
         'PaymentMethod',  
         'MonthlyCharges',  
         'TotalCharges',  
         'Churn']
```

```
In [8]: data['TotalCharges']=pd.to_numeric(data['TotalCharges'],errors='coerce')
```

In [9]: data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7043 entries, 0 to 7042
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype
---  -
0   customerID            7043 non-null   object
1   gender                 7043 non-null   object
2   SeniorCitizen          7043 non-null   int64
3   Partner                7043 non-null   object
4   Dependents             7043 non-null   object
5   tenure                 7043 non-null   int64
6   PhoneService           7043 non-null   object
7   MultipleLines           7043 non-null   object
8   InternetService        7043 non-null   object
9   OnlineSecurity         7043 non-null   object
10  OnlineBackup            7043 non-null   object
11  DeviceProtection       7043 non-null   object
12  TechSupport            7043 non-null   object
13  StreamingTV            7043 non-null   object
14  StreamingMovies        7043 non-null   object
15  Contract               7043 non-null   object
16  PaperlessBilling       7043 non-null   object
17  PaymentMethod          7043 non-null   object
18  MonthlyCharges         7043 non-null   float64
19  TotalCharges           7032 non-null   float64
20  Churn                  7043 non-null   object
dtypes: float64(2), int64(2), object(17)
memory usage: 1.1+ MB
```

In [10]: data1=data.drop(['customerID', 'PaymentMethod', 'Dependents', 'Partner', 'PaperlessBilling', 'StreamingTV', 'Strea

In [11]: data1

Out[11]:

|      | gender | tenure | MultipleLines    | InternetService | OnlineBackup | TechSupport | Contract       | MonthlyCharges | TotalCharges | Churn |
|------|--------|--------|------------------|-----------------|--------------|-------------|----------------|----------------|--------------|-------|
| 0    | Female | 1      | No phone service | DSL             | Yes          | No          | Month-to-month | 29.85          | 29.85        | No    |
| 1    | Male   | 34     | No               | DSL             | No           | No          | One year       | 56.95          | 1889.50      | No    |
| 2    | Male   | 2      | No               | DSL             | Yes          | No          | Month-to-month | 53.85          | 108.15       | Yes   |
| 3    | Male   | 45     | No phone service | DSL             | No           | Yes         | One year       | 42.30          | 1840.75      | No    |
| 4    | Female | 2      | No               | Fiber optic     | No           | No          | Month-to-month | 70.70          | 151.65       | Yes   |
| ...  | ...    | ...    | ...              | ...             | ...          | ...         | ...            | ...            | ...          | ...   |
| 7038 | Male   | 24     | Yes              | DSL             | No           | Yes         | One year       | 84.80          | 1990.50      | No    |
| 7039 | Female | 72     | Yes              | Fiber optic     | Yes          | No          | One year       | 103.20         | 7362.90      | No    |
| 7040 | Female | 11     | No phone service | DSL             | No           | No          | Month-to-month | 29.60          | 346.45       | No    |
| 7041 | Male   | 4      | Yes              | Fiber optic     | No           | No          | Month-to-month | 74.40          | 306.60       | Yes   |
| 7042 | Male   | 66     | No               | Fiber optic     | No           | Yes         | Two year       | 105.65         | 6844.50      | No    |

7043 rows × 10 columns

In [12]: data2=data1.fillna(data1.median())

In [13]: data2

Out[13]:

|      | gender | tenure | MultipleLines    | InternetService | OnlineBackup | TechSupport | Contract       | MonthlyCharges | TotalCharges | Churn |
|------|--------|--------|------------------|-----------------|--------------|-------------|----------------|----------------|--------------|-------|
| 0    | Female | 1      | No phone service | DSL             | Yes          | No          | Month-to-month | 29.85          | 29.85        | No    |
| 1    | Male   | 34     | No               | DSL             | No           | No          | One year       | 56.95          | 1889.50      | No    |
| 2    | Male   | 2      | No               | DSL             | Yes          | No          | Month-to-month | 53.85          | 108.15       | Yes   |
| 3    | Male   | 45     | No phone service | DSL             | No           | Yes         | One year       | 42.30          | 1840.75      | No    |
| 4    | Female | 2      | No               | Fiber optic     | No           | No          | Month-to-month | 70.70          | 151.65       | Yes   |
| ...  | ...    | ...    | ...              | ...             | ...          | ...         | ...            | ...            | ...          | ...   |
| 7038 | Male   | 24     | Yes              | DSL             | No           | Yes         | One year       | 84.80          | 1990.50      | No    |
| 7039 | Female | 72     | Yes              | Fiber optic     | Yes          | No          | One year       | 103.20         | 7362.90      | No    |
| 7040 | Female | 11     | No phone service | DSL             | No           | No          | Month-to-month | 29.60          | 346.45       | No    |
| 7041 | Male   | 4      | Yes              | Fiber optic     | No           | No          | Month-to-month | 74.40          | 306.60       | Yes   |
| 7042 | Male   | 66     | No               | Fiber optic     | No           | Yes         | Two year       | 105.65         | 6844.50      | No    |

7043 rows × 10 columns

In [14]: data2['Churn']=data2['Churn'].map({'Yes':1,'No':0})

In [15]: data2

Out[15]:

|      | gender | tenure | MultipleLines    | InternetService | OnlineBackup | TechSupport | Contract       | MonthlyCharges | TotalCharges | Churn |
|------|--------|--------|------------------|-----------------|--------------|-------------|----------------|----------------|--------------|-------|
| 0    | Female | 1      | No phone service | DSL             | Yes          | No          | Month-to-month | 29.85          | 29.85        | 0     |
| 1    | Male   | 34     | No               | DSL             | No           | No          | One year       | 56.95          | 1889.50      | 0     |
| 2    | Male   | 2      | No               | DSL             | Yes          | No          | Month-to-month | 53.85          | 108.15       | 1     |
| 3    | Male   | 45     | No phone service | DSL             | No           | Yes         | One year       | 42.30          | 1840.75      | 0     |
| 4    | Female | 2      | No               | Fiber optic     | No           | No          | Month-to-month | 70.70          | 151.65       | 1     |
| ...  | ...    | ...    | ...              | ...             | ...          | ...         | ...            | ...            | ...          | ...   |
| 7038 | Male   | 24     | Yes              | DSL             | No           | Yes         | One year       | 84.80          | 1990.50      | 0     |
| 7039 | Female | 72     | Yes              | Fiber optic     | Yes          | No          | One year       | 103.20         | 7362.90      | 0     |
| 7040 | Female | 11     | No phone service | DSL             | No           | No          | Month-to-month | 29.60          | 346.45       | 0     |
| 7041 | Male   | 4      | Yes              | Fiber optic     | No           | No          | Month-to-month | 74.40          | 306.60       | 1     |
| 7042 | Male   | 66     | No               | Fiber optic     | No           | Yes         | Two year       | 105.65         | 6844.50      | 0     |

7043 rows × 10 columns

In [16]: data3=pd.get\_dummies(data2)



In [17]: data3

Out[17]:

|      | tenure | MonthlyCharges | TotalCharges | Churn | gender_Female | gender_Male | MultipleLines_No | MultipleLines_No<br>phone service | MultipleLines_Yes | Internet |
|------|--------|----------------|--------------|-------|---------------|-------------|------------------|-----------------------------------|-------------------|----------|
| 0    | 1      | 29.85          | 29.85        | 0     | 1             | 0           | 0                | 1                                 | 0                 |          |
| 1    | 34     | 56.95          | 1889.50      | 0     | 0             | 1           | 1                | 0                                 | 0                 |          |
| 2    | 2      | 53.85          | 108.15       | 1     | 0             | 1           | 1                | 0                                 | 0                 |          |
| 3    | 45     | 42.30          | 1840.75      | 0     | 0             | 1           | 0                | 1                                 | 0                 |          |
| 4    | 2      | 70.70          | 151.65       | 1     | 1             | 0           | 1                | 0                                 | 0                 |          |
| ...  | ...    | ...            | ...          | ...   | ...           | ...         | ...              | ...                               | ...               | ...      |
| 7038 | 24     | 84.80          | 1990.50      | 0     | 0             | 1           | 0                | 0                                 | 1                 |          |
| 7039 | 72     | 103.20         | 7362.90      | 0     | 1             | 0           | 0                | 0                                 | 1                 |          |
| 7040 | 11     | 29.60          | 346.45       | 0     | 1             | 0           | 0                | 1                                 | 0                 |          |
| 7041 | 4      | 74.40          | 306.60       | 1     | 0             | 1           | 0                | 0                                 | 1                 |          |
| 7042 | 66     | 105.65         | 6844.50      | 0     | 0             | 1           | 1                | 0                                 | 0                 |          |

7043 rows × 21 columns

In [18]: `y=data3['Churn']`  
`x=data3.drop('Churn',axis=1)`

In [19]: `from sklearn.model_selection import train_test_split`  
`x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.33,random_state=42)`

```
In [20]: from sklearn.linear_model import LogisticRegression  
classifier= LogisticRegression()  
classifier.fit(x_train,y_train)
```

```
Out[20]: LogisticRegression()
```

```
In [21]: y_pred=classifier.predict(x_test)
```

```
In [22]: y_pred
```

```
Out[22]: array([1, 0, 0, ..., 1, 0, 0])
```

```
In [23]: from sklearn.metrics import confusion_matrix  
confusion_matrix(y_test,y_pred)
```

```
Out[23]: array([[1523, 174],  
                [ 277, 351]])
```

```
In [24]: from sklearn.metrics import accuracy_score  
accuracy_score(y_test,y_pred)
```

```
Out[24]: 0.8060215053763441
```

```
In [ ]:
```