


```
In [2]: import pandas as pd    #Import libraries
import warnings
warnings.filterwarnings('ignore')
data=pd.read_csv("/home/placement/Downloads/TelecomCustomerChurn.csv") #reading datafile
print(data)
```

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	\
0	7590-VHVEG	Female	0	Yes	No	1	
1	5575-GNVDE	Male	0	No	No	34	
2	3668-QPYBK	Male	0	No	No	2	
3	7795-CF0CW	Male	0	No	No	45	
4	9237-HQITU	Female	0	No	No	2	
...	
7038	6840-RESVB	Male	0	Yes	Yes	24	
7039	2234-XADUH	Female	0	Yes	Yes	72	
7040	4801-JZAZL	Female	0	Yes	Yes	11	
7041	8361-LTMKD	Male	1	Yes	No	4	
7042	3186-AJIEK	Male	0	No	No	66	

	PhoneService	MultipleLines	InternetService	OnlineSecurity	...	\
0	No	No phone service	DSL	No	...	
1	Yes	No	DSL	Yes	...	
2	Yes	No	DSL	Yes	...	
3	No	No phone service	DSL	Yes	...	
4	Yes	No	Fiber optic	No	...	
...	
7038	Yes	Yes	DSL	Yes	...	
7039	Yes	Yes	Fiber optic	No	...	
7040	No	No phone service	DSL	Yes	...	
7041	Yes	Yes	Fiber optic	No	...	
7042	Yes	No	Fiber optic	Yes	...	

	DeviceProtection	TechSupport	StreamingTV	StreamingMovies	Contract	\
0	No	No	No	No	Month-to-month	
1	Yes	No	No	No	One year	
2	No	No	No	No	Month-to-month	
3	Yes	Yes	No	No	One year	
4	No	No	No	No	Month-to-month	
...	
7038	Yes	Yes	Yes	Yes	One year	
7039	Yes	No	Yes	Yes	One year	

```

7040          No          No          No          No  Month-to-month
7041          No          No          No          No  Month-to-month
7042          Yes         Yes         Yes         Yes    Two year

    PaperlessBilling      PaymentMethod  MonthlyCharges  TotalCharges  \
0             Yes      Electronic check      29.85         29.85
1             No      Mailed check      56.95        1889.5
2             Yes      Mailed check      53.85         108.15
3             No  Bank transfer (automatic)      42.30        1840.75
4             Yes      Electronic check      70.70         151.65
...         ...         ...         ...         ...
7038          Yes      Mailed check      84.80        1990.5
7039          Yes  Credit card (automatic)     103.20        7362.9
7040          Yes      Electronic check      29.60         346.45
7041          Yes      Mailed check      74.40          306.6
7042          Yes  Bank transfer (automatic)     105.65        6844.5

    Churn
0      No
1      No
2     Yes
3      No
4     Yes
...     ...
7038    No
7039    No
7040    No
7041   Yes
7042    No

```

[7043 rows x 21 columns]

Exploratory DataAnalysis

```
In [3]: data.columns          #list(data)
```

```
Out[3]: Index(['customerID', 'gender', 'SeniorCitizen', 'Partner', 'Dependents',  
              'tenure', 'PhoneService', 'MultipleLines', 'InternetService',  
              'OnlineSecurity', 'OnlineBackup', 'DeviceProtection', 'TechSupport',  
              'StreamingTV', 'StreamingMovies', 'Contract', 'PaperlessBilling',  
              'PaymentMethod', 'MonthlyCharges', 'TotalCharges', 'Churn'],  
              dtype='object')
```

```
In [4]: data.describe()      #tenure=period of time
```

```
Out[4]:
```

	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 [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['TotalCharges']=pd.to_numeric(data['TotalCharges'],errors='coerce') *#convert the object type into float*

In [7]: 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 [8]: data1=data.drop(columns=['customerID', 'Partner', 'Dependents', 'MultipleLines', 'PaperlessBilling', 'PaymentMeth

In [9]: data1

Out[9]:

	gender	SeniorCitizen	tenure	PhoneService	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection	TechSupport	StreamingTV	S
0	Female	0	1	No	DSL	No	Yes	No	No	No	
1	Male	0	34	Yes	DSL	Yes	No	Yes	No	No	
2	Male	0	2	Yes	DSL	Yes	Yes	No	No	No	
3	Male	0	45	No	DSL	Yes	No	Yes	Yes	No	
4	Female	0	2	Yes	Fiber optic	No	No	No	No	No	
...	
7038	Male	0	24	Yes	DSL	Yes	No	Yes	Yes	Yes	
7039	Female	0	72	Yes	Fiber optic	No	Yes	Yes	No	Yes	
7040	Female	0	11	No	DSL	Yes	No	No	No	No	
7041	Male	1	4	Yes	Fiber optic	No	No	No	No	No	
7042	Male	0	66	Yes	Fiber optic	Yes	No	Yes	Yes	Yes	

7043 rows × 15 columns



```
In [10]: data1.isna().sum()
```

```
Out[10]: gender            0
SeniorCitizen            0
tenure                   0
PhoneService             0
InternetService          0
OnlineSecurity           0
OnlineBackup             0
DeviceProtection        0
TechSupport              0
StreamingTV              0
StreamingMovies          0
Contract                 0
MonthlyCharges           0
TotalCharges            11
Churn                    0
dtype: int64
```

```
In [11]: data2=data1.fillna(data1.median())
```

```
In [12]: data2.isna().sum()
```

```
Out[12]: gender            0
SeniorCitizen            0
tenure                   0
PhoneService             0
InternetService          0
OnlineSecurity           0
OnlineBackup             0
DeviceProtection        0
TechSupport              0
StreamingTV              0
StreamingMovies          0
Contract                 0
MonthlyCharges           0
TotalCharges            0
Churn                    0
dtype: int64
```



```
In [13]: data2['Churn']=data2['Churn'].map({'Yes':1, 'No':0})
```

```
In [14]: data2
```

```
Out[14]:
```

	gender	SeniorCitizen	tenure	PhoneService	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection	TechSupport	StreamingTV	S
0	Female	0	1	No	DSL	No	Yes	No	No	No	
1	Male	0	34	Yes	DSL	Yes	No	Yes	No	No	
2	Male	0	2	Yes	DSL	Yes	Yes	No	No	No	
3	Male	0	45	No	DSL	Yes	No	Yes	Yes	No	
4	Female	0	2	Yes	Fiber optic	No	No	No	No	No	
...	
7038	Male	0	24	Yes	DSL	Yes	No	Yes	Yes	Yes	
7039	Female	0	72	Yes	Fiber optic	No	Yes	Yes	No	Yes	
7040	Female	0	11	No	DSL	Yes	No	No	No	No	
7041	Male	1	4	Yes	Fiber optic	No	No	No	No	No	
7042	Male	0	66	Yes	Fiber optic	Yes	No	Yes	Yes	Yes	

7043 rows × 15 columns



```
In [15]: data3=pd.get_dummies(data2)
```

In [16]: data3

Out[16]:

	SeniorCitizen	tenure	MonthlyCharges	TotalCharges	Churn	gender_Female	gender_Male	PhoneService_No	PhoneService_Yes	InternetService
0	0	1	29.85	29.85	0	1	0	1	0	
1	0	34	56.95	1889.50	0	0	1	0	1	
2	0	2	53.85	108.15	1	0	1	0	1	
3	0	45	42.30	1840.75	0	0	1	1	0	
4	0	2	70.70	151.65	1	1	0	0	1	
...
7038	0	24	84.80	1990.50	0	0	1	0	1	
7039	0	72	103.20	7362.90	0	1	0	0	1	
7040	0	11	29.60	346.45	0	1	0	1	0	
7041	1	4	74.40	306.60	1	0	1	0	1	
7042	0	66	105.65	6844.50	0	0	1	0	1	

7043 rows × 33 columns

In [17]: `y=data3['Churn']`
`x=data3.drop(columns='Churn')`

In [18]: x

Out[18]:

	SeniorCitizen	tenure	MonthlyCharges	TotalCharges	gender_Female	gender_Male	PhoneService_No	PhoneService_Yes	InternetService_DS
0	0	1	29.85	29.85	1	0	1	0	
1	0	34	56.95	1889.50	0	1	0	1	
2	0	2	53.85	108.15	0	1	0	1	
3	0	45	42.30	1840.75	0	1	1	0	
4	0	2	70.70	151.65	1	0	0	1	
...
7038	0	24	84.80	1990.50	0	1	0	1	
7039	0	72	103.20	7362.90	1	0	0	1	
7040	0	11	29.60	346.45	1	0	1	0	
7041	1	4	74.40	306.60	0	1	0	1	
7042	0	66	105.65	6844.50	0	1	0	1	

7043 rows × 32 columns

```
In [103]: 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 [104]: from sklearn.linear_model import LogisticRegression #making model using 64%trainingg data
classifier=LogisticRegression()
classifier.fit(x_train,y_train)
```

Out[104]:

```
▼ LogisticRegression
LogisticRegression()
```

```
In [105]: y_pred=classifier.predict(x_test)    #multiply x_test with classifier
```

```
In [106]: y_pred
```

```
Out[106]: array([1, 0, 0, ..., 1, 1, 0])
```

```
In [107]: from sklearn.metrics import confusion_matrix    #confusion matrix  
          confusion_matrix(y_test,y_pred)
```

```
Out[107]: array([[1517, 180],  
                [ 264, 364]])
```

```
In [108]: from sklearn.metrics import accuracy_score    #accuracy of test data and predicted data  
          accuracy_score(y_test,y_pred)
```

```
Out[108]: 0.8090322580645162
```

```
In [ ]:
```