

```
In [23]: import pandas as pd #Import libraries
import warnings
warnings.filterwarnings('ignore')
accuracy_score(y_test,y_pred)
data=pd.read_csv("/home/placement/kambala/TelecomCustomerChurn.csv")
```

```
In [3]: 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]

In [4]: data.columns

Out[4]: 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 [5]: data.describe()
```

```
Out[5]:
```

	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 [6]: 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 [7]: data['TotalCharges']=pd.to_numeric(data['TotalCharges'],errors='coerce')

```
In [8]: 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 [11]: data1=data.drop(columns=['customerID', 'Partner', 'Dependents', 'MultipleLines', 'PaperlessBilling', 'PaymentMeth
```

```
In [12]: datafrom sklearn.metrics import accuracy_score #accuracy of test data and predicted data
accuracy_score(y_test,y_pred)
```

Out[12]:

	gender	SeniorCitizen	tenure	PhoneService	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection	TechSupport	StreamingTV	St
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 [14]: data1.isna().sum()
```

```
Out[14]: 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 [15]: data2=data1.fillna(data1.median())
```

```
In [16]: data2.isna().sum()
```

```
Out[16]: 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 [17]: data2['Churn']=data2['Churn'].map({'Yes':1,'No':0})
```

```
In [18]: data2
```

```
Out[18]:
```

	gender	SeniorCitizen	tenure	PhoneService	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection	TechSupport	StreamingTV	St
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



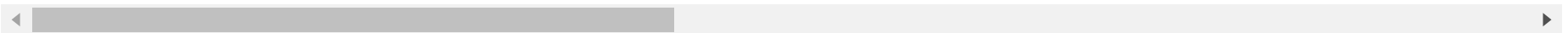
```
In [19]: data3=pd.get_dummies(data2)
```

In [20]: data3

Out[20]:

	SeniorCitizen	tenure	MonthlyCharges	TotalCharges	Churn	gender_Female	gender_Male	PhoneService_No	PhoneService_Yes	InternetServ
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



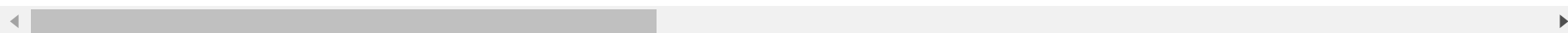
```
In [24]: y=data3['Churn']
x=data3.drop(columns='Churn')
from sklearn.metrics import accuracy_score #accuracy of test data and predicted
accuracy_score(y_test,y_pred)
```

In [25]: x

Out[25]:

	SeniorCitizen	tenure	MonthlyCharges	TotalCharges	gender_Female	gender_Male	PhoneService_No	PhoneService_Yes	InternetService_DSL
0	0	1	29.85	29.85	1	0	1	0	1
1	0	34	56.95	1889.50	0	1	0	1	1
2	0	2	53.85	108.15	0	1	0	1	1
3	0	45	42.30	1840.75	0	1	1	0	1
4	0	2	70.70	151.65	1	0	0	1	(
...
7038	0	24	84.80	1990.50	0	1	0	1	1
7039	0	72	103.20	7362.90	1	0	0	1	(
7040	0	11	29.60	346.45	1	0	1	0	1
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 [27]: 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 [28]: from sklearn.linear_model import LogisticRegression #making model using 64%trainingg data
classifier=LogisticRegression()
classifier.fit(x_train,y_train)
```

Out[28]: LogisticRegression()

```
In [29]: y_pred=classifier.predict(x_test)
```

In [30]: `y_pred`

Out[30]: `array([1, 0, 0, ..., 1, 1, 0])`

In [31]: `from sklearn.metrics import confusion_matrix #confusion matrix`
`confusion_matrix(y_test,y_pred)`

Out[31]: `array([[1518, 179],
 [264, 364]])`

In [32]: `from sklearn.metrics import accuracy_score`
`accuracy_score(y_test,y_pred)`

Out[32]: `0.8094623655913978`

In []: