	SeniorCitizen	tenure	MonthlyCharges	TotalCharges
count	7043.000000	7043.000000	7043.000000	7032.000000
mean	0.162147	32.371149	64.761692	2283.300441
std	0.368612	24.559481	30.090047	2266.771362
min	0.000000	0.000000	18.250000	18.800000
25%	0.000000	9.000000	35.500000	401.450000
50%	0.000000	29.000000	70.350000	1397.475000
75%	0.000000	55.000000	89.850000	3794.737500
max	1.000000	72.000000	118.750000	8684.800000

```
In [5]: data.isna().sum()
Out[5]: customerID
                              0
        gender
                              0
        SeniorCitizen
        Partner
        Dependents
        tenure
        PhoneService
        MultipleLines
        InternetService
        OnlineSecurity
        OnlineBackup
        DeviceProtection
        TechSupport
        StreamingTV
        StreamingMovies
        Contract
        PaperlessBilling
        PaymentMethod
        MonthlyCharges
        TotalCharges
                            11
        Churn
                              0
        dtype: int64
```

In [6]: data1=data.fillna(data.median())

/tmp/ipykernel 2919/3060338577.py:1: FutureWarning: The default value of numeric only in DataFrame.median i s deprecated. In a future version, it will default to False. In addition, specifying 'numeric only=None' is deprecated. Select only valid columns or specify the value of numeric only to silence this warning. data1=data.fillna(data.median())

In [7]:	data1.isna().sum()	
Out[7]:	customerID	0
	gender	0
	SeniorCitizen	0
	Partner	0
	Dependents	0
	tenure	0
	PhoneService	0
	MultipleLines	0
	InternetService	0
	OnlineSecurity	0
	OnlineBackup	0
	DeviceProtection	0
	TechSupport	0
	StreamingTV	0
	StreamingMovies	0
	Contract	0
	PaperlessBilling	0
	PaymentMethod	0
	MonthlyCharges	0
	TotalCharges	0
	Churn	0
	dtype: int64	

In [8]: data1.info()

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

```
In [9]: list(data1)
     Out[9]: ['customerID',
                                                     'gender',
                                                    'SeniorCitizen',
                                                    'Partner',
                                                    'Dependents',
                                                    'tenure',
                                                    'PhoneService',
                                                    'MultipleLines',
                                                    'InternetService',
                                                    'OnlineSecurity',
                                                    'OnlineBackup',
                                                    'DeviceProtection',
                                                    'TechSupport',
                                                    'StreamingTV',
                                                    'StreamingMovies',
                                                    'Contract',
                                                    'PaperlessBilling',
                                                    'PaymentMethod',
                                                    'MonthlyCharges',
                                                    'TotalCharges',
                                                    'Churn']
In [10]: data1.shape
Out[10]: (7043, 21)
In [12]: SeniorCitizen', 'PhoneService', 'OnlineBackup', 'Partner', 'Dependents', 'OnlineSecurity', 'StreamingTV', 'Strea
```

In [13]: data2

Out[13]:

gender	tenure	MultipleLines	InternetService	TechSupport	Contract	PaymentMethod	MonthlyCharges	TotalCharges	Churn
Female	1	No phone service	DSL	No	Month-to- month	Electronic check	29.85	29.85	No
Male	34	No	DSL	No	One year	Mailed check	56.95	1889.50	No
Male	2	No	DSL	No	Month-to- month	Mailed check	53.85	108.15	Yes
Male	45	No phone service	DSL	Yes	One year	Bank transfer (automatic)	42.30	1840.75	No
Female	2	No	Fiber optic	No	Month-to- month	Electronic check	70.70	151.65	Yes
Male	24	Yes	DSL	Yes	One year	Mailed check	84.80	1990.50	No
Female	72	Yes	Fiber optic	No	One year	Credit card (automatic)	103.20	7362.90	No
Female	11	No phone service	DSL	No	Month-to- month	Electronic check	29.60	346.45	No
Male	4	Yes	Fiber optic	No	Month-to- month	Mailed check	74.40	306.60	Yes
Male	66	No	Fiber optic	Yes	Two year	Bank transfer (automatic)	105.65	6844.50	No
	=emale Male Male Male =emale Male =emale =emale Male	Female 1 Male 34 Male 2 Male 45 Female 2 Male 24 Female 72 Female 11 Male 4	Female 1 No phone service Male 34 No Male 2 No Male 45 No phone service Female 2 No Male 24 Yes Female 72 Yes Female 11 No phone service Male 4 Yes Male 4 Yes	Female 1 No phone service DSL Male 34 No DSL Male 2 No DSL Male 45 No phone service DSL Female 2 No Fiber optic	Female 1 No phone service DSL No Male 34 No DSL No Male 2 No DSL No Male 45 No phone service DSL Yes Female 2 No Fiber optic No	Female 1 No phone service DSL No Month-to-month Male 34 No DSL No One year Male 2 No DSL No Month-to-month Male 45 No phone service DSL Yes One year Female 2 No Fiber optic No Month-to-month Male 24 Yes DSL Yes One year Female 72 Yes Fiber optic No One year Female 72 Yes Fiber optic No One year Female 4 Yes Fiber optic No One year Female 4 Yes Fiber optic No Month-to-month Male 4 Yes Fiber optic No Month-to-month	Female 1 No phone service DSL No Month-to-month Electronic check Male 34 No DSL No One year Mailed check Male 2 No DSL No Month-to-month Mailed check Male 45 No phone service DSL Yes One year Bank transfer (automatic) Female 2 No Fiber optic No Month-to-month Electronic check	Female 1 No phone service DSL No Month-tomonth Electronic check 29.85 Male 34 No DSL No One year Mailed check 56.95 Male 2 No DSL No Month-tomonth Mailed check 53.85 Male 45 No phone service DSL Yes One year Bank transfer (automatic) 42.30 Female 2 No Fiber optic No Month-tomonth Electronic check 70.70	Female 1

7043 rows × 10 columns

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

In [15]: data2

Out[15]:

	gender	tenure	MultipleLines	InternetService	TechSupport	Contract	PaymentMethod	MonthlyCharges	TotalCharges	Churn
0	Female	1	No phone service	DSL	No	Month-to- month	Electronic check	29.85	29.85	0
1	Male	34	No	DSL	No	One year	Mailed check	56.95	1889.50	0
2	Male	2	No	DSL	No	Month-to- month	Mailed check	53.85	108.15	1
3	Male	45	No phone service	DSL	Yes	One year	Bank transfer (automatic)	42.30	1840.75	0
4	Female	2	No	Fiber optic	No	Month-to- month	Electronic check	70.70	151.65	1
7038	Male	24	Yes	DSL	Yes	One year	Mailed check	84.80	1990.50	0
7039	Female	72	Yes	Fiber optic	No	One year	Credit card (automatic)	103.20	7362.90	0
7040	Female	11	No phone service	DSL	No	Month-to- month	Electronic check	29.60	346.45	0
7041	Male	4	Yes	Fiber optic	No	Month-to- month	Mailed check	74.40	306.60	1
7042	Male	66	No	Fiber optic	Yes	Two year	Bank transfer (automatic)	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 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 × 22 columns

```
In [18]: data3.shape
Out[18]: (7043, 22)
In [19]: y=data3['Churn']
    x=data3.drop('Churn',axis=1)
In [20]: 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)
```

localhost:8888/notebooks/telecom.ipynb

```
In [21]: from sklearn.linear model import LogisticRegression
         reg=LogisticRegression()
         reg.fit(x train,y train)
Out[21]:
          ▼ LogisticRegression
          LogisticRegression()
In [22]: y pred=reg.predict(x test)
In [23]: from sklearn.metrics import confusion matrix
         confusion_matrix(y_test,y_pred)
Out[23]: array([[1522, 175],
                [ 270, 358]])
In [24]: from sklearn.metrics import accuracy_score
         accuracy_score(y_test,y_pred)
Out[24]: 0.8086021505376344
In [ ]:
```