```
In [1]: import pandas as pd
data=pd.read_csv("/home/placement/Downloads/TelecomCustomerChurn.csv")
```

In [2]: data.describe()

Out[2]:

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

## In [4]: data.info()

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

In [5]:	<pre>data.isna().sum()</pre>	
Out[5]:	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	11
	Churn	0
	dtype: int64	

```
In [6]: list(data)
Out[6]: ['customerID',
          'gender',
         'SeniorCitizen',
         'Partner',
         'Dependents',
         'tenure',
         'PhoneService',
         'MultipleLines',
         'InternetService',
         'OnlineSecurity',
         'OnlineBackup',
         'DeviceProtection',
         'TechSupport',
         'StreamingTV',
         'StreamingMovies',
         'Contract',
         'PaperlessBilling',
         'PaymentMethod',
         'MonthlyCharges',
         'TotalCharges',
         'Churn']
```

In [7]: data

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:	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	 DevicePro
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	
7038	6840- RESVB	Male	0	Yes	Yes	24	Yes	Yes	DSL	Yes	
7039	2234- XADUH	Female	0	Yes	Yes	72	Yes	Yes	Fiber optic	No	
7040	4801-JZAZL	Female	0	Yes	Yes	11	No	No phone service	DSL	Yes	
7041	8361- LTMKD	Male	1	Yes	No	4	Yes	Yes	Fiber optic	No	
7042	3186-AJIEK	Male	0	No	No	66	Yes	No	Fiber optic	Yes	

7043 rows × 21 columns

In [8]: data1=data.fillna(data.median())

/tmp/ipykernel\_10166/3060338577.py:1: FutureWarning: The default value of numeric\_only in DataFrame.median is deprecated. In a future version, it will default to False. In addition, specifying 'numeric only=None' i s deprecated. Select only valid columns or specify the value of numeric only to silence this warning. data1=data.fillna(data.median())

```
In [9]: data1.isna().sum()
 Out[9]: customerID
                             0
         gender
                             0
         SeniorCitizen
                             0
         Partner
                             0
         Dependents
                             0
         tenure
         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 [10]: x=data1.drop(['customerID','Churn'],axis=1)
         y=data1['Churn']
```

n [11]:	x.ł	nead()										
ut[11]:		gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection
	0	Female	0	Yes	No	1	No	No phone service	DSL	No	Yes	1
	1	Male	0	No	No	34	Yes	No	DSL	Yes	No	Υ
	2	Male	0	No	No	2	Yes	No	DSL	Yes	Yes	1
	3	Male	0	No	No	45	No	No phone service	DSL	Yes	No	Y
	4	Female	0	No	No	2	Yes	No	Fiber optic	No	No	1
	<b>■</b>											<b>&gt;</b>

In [12]: x=pd.get\_dummies(x)

In [13]: x

Out[13]:

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

7043 rows × 45 columns

In [14]: 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 [15]: from sklearn.model selection import GridSearchCV #GridSearchCV is for parameter tuning

```
from sklearn.ensemble import RandomForestClassifier
         cls=RandomForestClassifier()
         n estimators=[25,50,75,100,125,150,175,200] #number of decision trees in the forest, default = 100
          criterion=['gini','entropy'] #criteria for choosing nodes default = 'gini'
         max depth=[3,5,10] #maximum number of nodes in a tree default = None (it will go till all possible nodes)
         parameters={'n estimators': n estimators, 'criterion':criterion, 'max depth':max depth} #this will undergo 8*2
         RFC cls = GridSearchCV(cls, parameters)
         RFC cls.fit(x train,y train)
Out[15]: GridSearchCV(estimator=RandomForestClassifier(),
                       param grid={'criterion': ['gini', 'entropy'],
                                    'max depth': [3, 5, 10],
                                    'n estimators': [25, 50, 75, 100, 125, 150, 175, 200]})
         In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page with nbyiewer.org.
In [17]: RFC cls.best params
Out[17]: {'criterion': 'entropy', 'max depth': 10, 'n estimators': 75}
In [18]: cls=RandomForestClassifier(n estimators=200,criterion='entropy',max depth=10)
In [19]: cls.fit(x train,y train)
Out[19]: RandomForestClassifier(criterion='entropy', max_depth=10, n_estimators=200)
          In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this page with nbyiewer.org.
In [20]: rfy pred=cls.predict(x test)
In [21]: rfy pred
Out[21]: array(['Yes', 'No', 'No', 'Yes', 'No', 'No'], dtype=object)
```