

```
In [4]: import pandas as pd
import numpy as np
import warnings
warnings.filterwarnings("ignore")
```

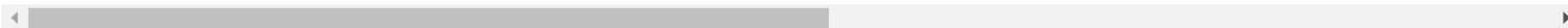
```
In [5]: data=pd.read_csv("/home/placement/Downloads/TelecomCustomerChurn.csv")
```

```
In [6]: data.head()
```

Out[6]:

	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 [7]: data.describe()
```

```
Out[7]:
```

	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 [8]: data.isna().sum()
```

```
Out[8]: 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 [9]: data1=data.fillna(data.median())
```

In [10]: data1

Out[10]:

	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 [11]: 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 [12]: data.dtypes
```

```
Out[12]: customerID      object  
gender      object  
SeniorCitizen  int64  
Partner      object  
Dependents    object  
tenure      int64  
PhoneService  object  
MultipleLines object  
InternetService object  
OnlineSecurity object  
OnlineBackup  object  
DeviceProtection object  
TechSupport   object  
StreamingTV   object  
StreamingMovies object  
Contract      object  
PaperlessBilling object  
PaymentMethod object  
MonthlyCharges float64  
TotalCharges  object  
Churn         object  
dtype: object
```

```
In [13]: data['TotalCharges'] = pd.to_numeric(data['TotalCharges'], errors='coerce')
```

```
In [14]: data.dtypes
```

```
Out[14]: customerID      object  
gender      object  
SeniorCitizen  int64  
Partner      object  
Dependents    object  
tenure      int64  
PhoneService  object  
MultipleLines  object  
InternetService  object  
OnlineSecurity  object  
OnlineBackup    object  
DeviceProtection  object  
TechSupport     object  
StreamingTV     object  
StreamingMovies  object  
Contract        object  
PaperlessBilling  object  
PaymentMethod    object  
MonthlyCharges  float64  
TotalCharges    float64  
Churn           object  
dtype: object
```

```
In [15]: data.isna().sum()
```

```
Out[15]: 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 [16]: databackup=data.copy()
```

```
In [17]: data['TotalCharges']=data['TotalCharges'].fillna(data['TotalCharges'].median())
```

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

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



In [20]: `x.head()`

Out[20]:

	SeniorCitizen	tenure	MonthlyCharges	TotalCharges	gender_Female	gender_Male	Partner_No	Partner_Yes	Dependents_No	Dependents_Yes
0	0	1	29.85	29.85	1	0	0	1	1	0
1	0	34	56.95	1889.50	0	1	1	0	1	0
2	0	2	53.85	108.15	0	1	1	0	1	0
3	0	45	42.30	1840.75	0	1	1	0	1	0
4	0	2	70.70	151.65	1	0	1	0	1	0

5 rows × 11 columns

```
In [ ]: 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 [ ]: 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)
```

```
In [ ]: RFC_cls.best_params_
```

```
In [24]: cls=RandomForestClassifier(n_estimators=75,criterion='entropy',max_depth=10)
```

```
In [25]: cls.fit(x_train,y_train)
```

```
Out[25]: RandomForestClassifier
RandomForestClassifier(criterion='entropy', max_depth=10, n_estimators=75)
```

```
In [26]: rfy_pred=cls.predict(x_test)
```

```
In [27]: rfy_pred
```

```
Out[27]: array(['Yes', 'No', 'No', ..., 'Yes', 'No', 'No'], dtype=object)
```

```
In [28]: from sklearn.metrics import confusion_matrix
confusion_matrix(y_test,rfy_pred)
```

```
Out[28]: array([[1538, 159],
               [ 309, 319]])
```

```
In [29]: from sklearn.metrics import accuracy_score
accuracy_score(y_test,rfy_pred)
```

```
Out[29]: 0.7987096774193548
```

```
In [ ]:
```