

```
In [43]: import pandas as pd
```

```
In [44]: data=pd.read_csv("/home/placement/Desktop/venkatesh/TelecomCustomerChurn.csv")
```

```
In [45]: data.describe()
```

```
Out[45]:
```

| | 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 [19]: 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 [46]: data.isna().sum()
```

```
Out[46]: 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 [47]: data.dtypes
```

```
Out[47]: 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 [22]: data['TotalCharges']=pd.to_numeric(data['TotalCharges'],errors='coerce')
# or_use_this==> data['TotalCharges']=data['TotalCharges'].replace('',np.nan).astype(float).values
```

```
In [32]: data.isna().sum()
```

```
Out[32]: 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 [23]: data.dtypes
```

```
Out[23]: 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 [33]: data['TotalCharges']=data['TotalCharges'].fillna(data['TotalCharges'].median())
```

```
In [ ]: #for backup
#databackup=data.copy()
```

```
In [35]: X=data.drop(['customerID','Churn'],axis=1)
y=data['Churn']
```

```
In [36]: #data['SeniorCitizen']=data['SeniorCitizen'].map{0:'No',1:'Yes'}
#data['TotalCharges']=data['TotalCharges'].fillna(data['TotalCharges'].median())
```

```
In [37]: X=pd.get_dummies(X)
```

```
In [38]: X.head(10)
```

```
Out[38]:
```

| | 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 | 0 | 8 | 99.65 | 820.50 | 1 | 0 | 1 | 0 | 1 | 0 |
| 6 | 0 | 22 | 89.10 | 1949.40 | 0 | 1 | 1 | 0 | 0 | 1 |
| 7 | 0 | 10 | 29.75 | 301.90 | 1 | 0 | 1 | 0 | 1 | 0 |
| 8 | 0 | 28 | 104.80 | 3046.05 | 1 | 0 | 0 | 1 | 1 | 0 |
| 9 | 0 | 62 | 56.15 | 3487.95 | 0 | 1 | 1 | 0 | 0 | 1 |

10 rows × 45 columns



```
In [49]: list(X)
```

```
Out[49]: ['SeniorCitizen',  
          'tenure',  
          'MonthlyCharges',  
          'TotalCharges',  
          'gender_Female',  
          'gender_Male',  
          'Partner_No',  
          'Partner_Yes',  
          'Dependents_No',  
          'Dependents_Yes',  
          'PhoneService_No',  
          'PhoneService_Yes',  
          'MultipleLines_No',  
          'MultipleLines_No phone service',  
          'MultipleLines_Yes',  
          'InternetService_DSL',  
          'InternetService_Fiber optic',  
          'InternetService_No',  
          'OnlineSecurity_No',  
          'OnlineSecurity_No internet service',  
          'OnlineSecurity_Yes',  
          'OnlineBackup_No',  
          'OnlineBackup_No internet service',  
          'OnlineBackup_Yes',  
          'DeviceProtection_No',  
          'DeviceProtection_No internet service',  
          'DeviceProtection_Yes',  
          'TechSupport_No',  
          'TechSupport_No internet service',  
          'TechSupport_Yes',  
          'StreamingTV_No',  
          'StreamingTV_No internet service',  
          'StreamingTV_Yes',  
          'StreamingMovies_No',  
          'StreamingMovies_No internet service',  
          'StreamingMovies_Yes',  
          'Contract_Month-to-month',  
          'Contract_One year',  
          'Contract_Two year',
```



```
'PaperlessBilling_No',  
'PaperlessBilling_Yes',  
'PaymentMethod_Bank transfer (automatic)',  
'PaymentMethod_Credit card (automatic)',  
'PaymentMethod_Electronic check',  
'PaymentMethod_Mailed check']
```

```
In [39]: 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 [40]: 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[40]: 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 nbviewer.org.

```
In [50]: RFC_cls.best_params_
```

```
Out[50]: {'criterion': 'entropy', 'max_depth': 10, 'n_estimators': 125}
```

```
In [51]: cls=RandomForestClassifier(n_estimators=125,criterion='entropy',max_depth=10)
```

```
In [52]: cls.fit(X_train,y_train)
```

```
Out[52]: RandomForestClassifier(criterion='entropy', max_depth=10, n_estimators=125)
```

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```
In [53]: rfy_pred=cls.predict(X_test)
```

```
In [54]: rfy_pred
```

```
Out[54]: array(['Yes', 'No', 'No', ..., 'Yes', 'No', 'No'], dtype=object)
```

```
In [55]: from sklearn.metrics import confusion_matrix  
confusion_matrix(y_test,rfy_pred)
```

```
Out[55]: array([[1547,  150],  
               [ 299,  329]])
```

```
In [56]: from sklearn.metrics import accuracy_score  
accuracy_score(y_test,rfy_pred)
```

```
Out[56]: 0.8068817204301075
```

```
In [ ]:
```