

# Naan Mudhalvan Project

## customer churn prediction

### Team Members:

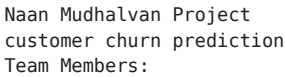
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Double-click (or enter) to edit

```
import matplotlib.pyplot as plt
```

```
df = pd.read_csv('/kaggle/input/telco-customer-churn/WA_Fn-UseC_-Telco-Customer-Churn.csv')
df.head(5)
```

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines
0	7590-VHVEG	Female	0	Yes	No	1	No	No phone service
1	5575-GNVDE	Male	0	No	No	34	Yes	No
2	3668-QPYBK	Male	0	No	No	2	Yes	No
3	7795-CFOCW	Male	0	No	No	45	No	No phone service
4	9237-HQITU	Female	0	No	No	2	Yes	No

5 rows x 21 columns

```
df.drop('customerID',axis='columns',inplace=True)
df.sample(5)
```

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetS
2128	Male	0	Yes	Yes	41	Yes	Yes	

```
df.dtypes
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
```

```
df[pd.to_numeric(df.TotalCharges,errors='coerce').isnull()]
```

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetS
488	Female	0	Yes	Yes	0	No	No phone service	
753	Male	0	No	Yes	0	Yes	No	
936	Female	0	Yes	Yes	0	Yes	No	
1082	Male	0	Yes	Yes	0	Yes	Yes	
1340	Female	0	Yes	Yes	0	No	No phone service	
3331	Male	0	Yes	Yes	0	Yes	No	
3826	Male	0	Yes	Yes	0	Yes	Yes	
4380	Female	0	Yes	Yes	0	Yes	No	
5218	Male	0	Yes	Yes	0	Yes	No	
6670	Female	0	Yes	Yes	0	Yes	Yes	
6754	Male	0	No	Yes	0	Yes	Yes	

```
df1 = df[df.TotalCharges!=' ']

df1['TotalCharges'] = pd.to_numeric(df['TotalCharges'],errors='coerce')

/tmp/ipykernel_20/1694240200.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
df1['TotalCharges'] = pd.to_numeric(df['TotalCharges'],errors='coerce')
```

```
df1.dtypes
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

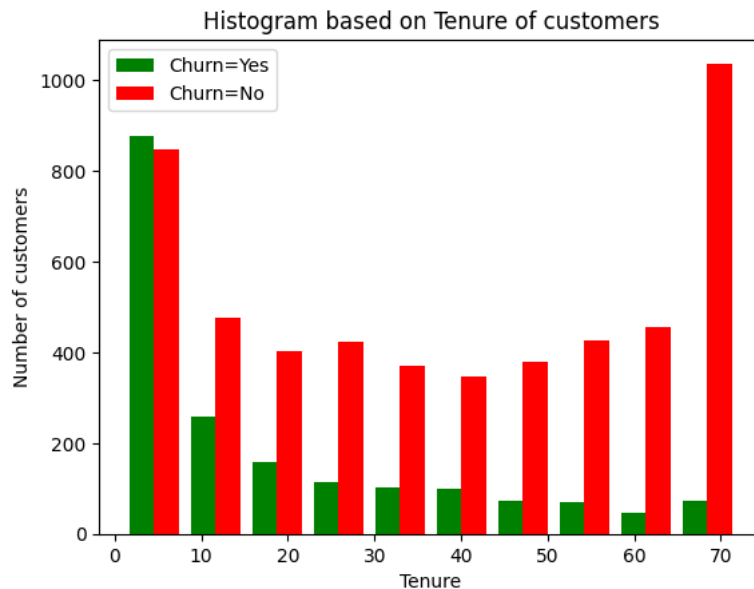
```

```

df_tenure_no = df1[df1.Churn == 'No'].tenure
df_tenure_yes = df1[df1.Churn == 'Yes'].tenure
plt.hist([df_tenure_yes,df_tenure_no],color=['green','red'],label=['Churn=Yes','Churn=No'])
plt.legend()
plt.xlabel('Tenure')
plt.ylabel('Number of customers')
plt.title('Histogram based on Tenure of customers')

```

Text(0.5, 1.0, 'Histogram based on Tenure of customers')

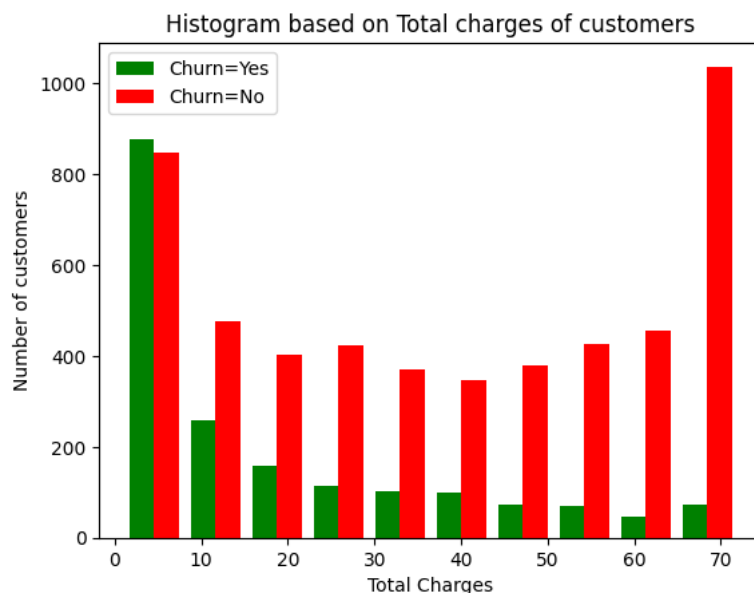


```

df_total_charges_no = df1[df1.Churn == 'No'].TotalCharges
df_total_charges_yes = df1[df1.Churn == 'Yes'].TotalCharges
plt.hist([df_total_charges_yes,df_total_charges_no],color=['green','red'],label=['Churn=Yes','Churn=No'])
plt.legend()
plt.xlabel('Total Charges')
plt.ylabel('Number of customers')
plt.title('Histogram based on Total charges of customers')

```

Text(0.5, 1.0, 'Histogram based on Total charges of customers')



```
def print_unique_values(df):
    for cols in df.columns:
        print(cols,df[cols].unique())
```

```
print_unique_values(df1)
```

```
gender ['Female' 'Male']
SeniorCitizen [0 1]
Partner ['Yes' 'No']
Dependents ['No' 'Yes']
tenure [ 1 34  2 45  8 22 10 28 62 13 16 58 49 25 69 52 71 21 12 30 47 72 17 27
         5 46 11 70 63 43 15 60 18 66  9  3 31 50 64 56  7 42 35 48 29 65 38 68
        32 55 37 36 41  6  4 33 67 23 57 61 14 20 53 40 59 24 44 19 54 51 26 39]
PhoneService ['No' 'Yes']
MultipleLines ['No phone service' 'No' 'Yes']
InternetService ['DSL' 'Fiber optic' 'No']
OnlineSecurity ['No' 'Yes' 'No internet service']
OnlineBackup ['Yes' 'No' 'No internet service']
DeviceProtection ['No' 'Yes' 'No internet service']
TechSupport ['No' 'Yes' 'No internet service']
StreamingTV ['No' 'Yes' 'No internet service']
StreamingMovies ['No' 'Yes' 'No internet service']
Contract ['Month-to-month' 'One year' 'Two year']
PaperlessBilling ['Yes' 'No']
PaymentMethod ['Electronic check' 'Mailed check' 'Bank transfer (automatic)'
               'Credit card (automatic)']
MonthlyCharges [29.85 56.95 53.85 ... 63.1 44.2 78.7 ]
TotalCharges [ 29.85 1889.5 108.15 ... 346.45 306.6 6844.5 ]
Churn ['No' 'Yes']
```

```
df1.replace('No phone service','No',inplace=True)
```

```
/tmp/ipykernel_20/628100714.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
df1.replace('No phone service','No',inplace=True)
```

```
df1.replace('No internet service','No',inplace=True)
```

```
/tmp/ipykernel_20/4127402845.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
df1.replace('No internet service','No',inplace=True)
```

```
print_unique_values(df1)
```

```
gender ['Female' 'Male']
SeniorCitizen [0 1]
Partner ['Yes' 'No']
Dependents ['No' 'Yes']
tenure [ 1 34  2 45  8 22 10 28 62 13 16 58 49 25 69 52 71 21 12 30 47 72 17 27
         5 46 11 70 63 43 15 60 18 66  9  3 31 50 64 56  7 42 35 48 29 65 38 68
        32 55 37 36 41  6  4 33 67 23 57 61 14 20 53 40 59 24 44 19 54 51 26 39]
PhoneService ['No' 'Yes']
MultipleLines ['No' 'Yes']
InternetService ['DSL' 'Fiber optic' 'No']
OnlineSecurity ['No' 'Yes']
OnlineBackup ['Yes' 'No']
DeviceProtection ['No' 'Yes']
TechSupport ['No' 'Yes']
StreamingTV ['No' 'Yes']
StreamingMovies ['No' 'Yes']
Contract ['Month-to-month' 'One year' 'Two year']
PaperlessBilling ['Yes' 'No']
PaymentMethod ['Electronic check' 'Mailed check' 'Bank transfer (automatic)'
               'Credit card (automatic)']
MonthlyCharges [29.85 56.95 53.85 ... 63.1 44.2 78.7 ]
TotalCharges [ 29.85 1889.5 108.15 ... 346.45 306.6 6844.5 ]
Churn ['No' 'Yes']
```

```
df2 = pd.get_dummies(data=df1,columns=['InternetService','Contract','PaymentMethod'])
```

```
df2.shape
```

```
(7032, 27)
```

```

print_unique_values(df2)

gender ['Female' 'Male']
SeniorCitizen [0 1]
Partner ['Yes' 'No']
Dependents ['No' 'Yes']
tenure [ 1 34  2 45  8 22 10 28 62 13 16 58 49 25 69 52 71 21 12 30 47 72 17 27
         5 46 11 70 63 43 15 60 18 66  9  3 31 50 64 56  7 42 35 48 29 65 38 68
        32 55 37 36 41  6  4 33 67 23 57 61 14 20 53 40 59 24 44 19 54 51 26 39]
PhoneService ['No' 'Yes']
MultipleLines ['No' 'Yes']
OnlineSecurity ['No' 'Yes']
OnlineBackup ['Yes' 'No']
DeviceProtection ['No' 'Yes']
TechSupport ['No' 'Yes']
StreamingTV ['No' 'Yes']
StreamingMovies ['No' 'Yes']
PaperlessBilling ['Yes' 'No']
MonthlyCharges [29.85 56.95 53.85 ... 63.1 44.2 78.7 ]
TotalCharges [ 29.85 1889.5 108.15 ... 346.45 306.6 6844.5 ]
Churn ['No' 'Yes']
InternetService_DSL [ True False]
InternetService_Fiber optic [False  True]
InternetService_No [False  True]
Contract_Month-to-month [ True False]
Contract_One year [False  True]
Contract_Two year [False  True]
PaymentMethod_Bank transfer (automatic) [False  True]
PaymentMethod_Credit card (automatic) [False  True]
PaymentMethod_Electronic check [ True False]
PaymentMethod_Mailed check [False  True]

df_true_false_cols = ['InternetService_DSL','InternetService_Fiber optic','InternetService_No','Contract_One year','Contract_

for cols in df_true_false_cols:
    df2[cols].replace({True:1,False:0},inplace=True)

print_unique_values(df2)

gender ['Female' 'Male']
SeniorCitizen [0 1]
Partner ['Yes' 'No']
Dependents ['No' 'Yes']
tenure [ 1 34  2 45  8 22 10 28 62 13 16 58 49 25 69 52 71 21 12 30 47 72 17 27
         5 46 11 70 63 43 15 60 18 66  9  3 31 50 64 56  7 42 35 48 29 65 38 68
        32 55 37 36 41  6  4 33 67 23 57 61 14 20 53 40 59 24 44 19 54 51 26 39]
PhoneService ['No' 'Yes']
MultipleLines ['No' 'Yes']
OnlineSecurity ['No' 'Yes']
OnlineBackup ['Yes' 'No']
DeviceProtection ['No' 'Yes']
TechSupport ['No' 'Yes']
StreamingTV ['No' 'Yes']
StreamingMovies ['No' 'Yes']
PaperlessBilling ['Yes' 'No']
MonthlyCharges [29.85 56.95 53.85 ... 63.1 44.2 78.7 ]
TotalCharges [ 29.85 1889.5 108.15 ... 346.45 306.6 6844.5 ]
Churn ['No' 'Yes']
InternetService_DSL [1 0]
InternetService_Fiber optic [0 1]
InternetService_No [0 1]
Contract_Month-to-month [1 0]
Contract_One year [0 1]
Contract_Two year [0 1]
PaymentMethod_Bank transfer (automatic) [0 1]
PaymentMethod_Credit card (automatic) [0 1]
PaymentMethod_Electronic check [1 0]
PaymentMethod_Mailed check [0 1]

df_yes_no_cols = ['Partner','Dependents','PhoneService','MultipleLines','OnlineSecurity','OnlineBackup','DeviceProtection','T
for cols in df_yes_no_cols:
    df2[cols].replace({'Yes':1,'No':0},inplace=True)
print_unique_values(df2)

gender ['Female' 'Male']
SeniorCitizen [0 1]
Partner [1 0]
Dependents [0 1]
tenure [ 1 34  2 45  8 22 10 28 62 13 16 58 49 25 69 52 71 21 12 30 47 72 17 27
         5 46 11 70 63 43 15 60 18 66  9  3 31 50 64 56  7 42 35 48 29 65 38 68
        32 55 37 36 41  6  4 33 67 23 57 61 14 20 53 40 59 24 44 19 54 51 26 39]
PhoneService [0 1]
MultipleLines [0 1]
OnlineSecurity [0 1]

```

```

OnlineBackup [1 0]
DeviceProtection [0 1]
TechSupport [0 1]
StreamingTV [0 1]
StreamingMovies [0 1]
PaperlessBilling [1 0]
MonthlyCharges [29.85 56.95 53.85 ... 63.1 44.2 78.7 ]
TotalCharges [ 29.85 1889.5 108.15 ... 346.45 306.6 6844.5 ]
Churn [0 1]
InternetService_DSL [1 0]
InternetService_Fiber optic [0 1]
InternetService_No [0 1]
Contract_Month-to-month [1 0]
Contract_One year [0 1]
Contract_Two year [0 1]
PaymentMethod_Bank transfer (automatic) [0 1]
PaymentMethod_Credit card (automatic) [0 1]
PaymentMethod_Electronic check [1 0]
PaymentMethod_Mailed check [0 1]

df2['gender'].replace({'Female':1,'Male':0},inplace=True)
cols_to_scale = ['tenure','MonthlyCharges','TotalCharges']
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
df2[cols_to_scale] = scaler.fit_transform(df2[cols_to_scale])
print_unique_values(df2)

gender [1 0]
SeniorCitizen [0 1]
Partner [1 0]
Dependents [0 1]
tenure [0. 0.46478873 0.01408451 0.61971831 0.09859155 0.29577465
0.12676056 0.38028169 0.85915493 0.16901408 0.21126761 0.8028169
0.67605634 0.33802817 0.95774648 0.71830986 0.98591549 0.28169014
0.15492958 0.4084507 0.64788732 1. 0.22535211 0.36619718
0.05633803 0.63380282 0.14084507 0.97183099 0.87323944 0.5915493
0.1971831 0.83098592 0.23943662 0.91549296 0.11267606 0.02816901
0.42253521 0.69014085 0.88732394 0.77464789 0.08450704 0.57746479
0.47887324 0.66197183 0.3943662 0.90140845 0.52112676 0.94366197
0.43661972 0.76056338 0.50704225 0.49295775 0.56338028 0.07042254
0.04225352 0.45070423 0.92957746 0.30985915 0.78873239 0.84507042
0.18309859 0.26760563 0.73239437 0.54929577 0.81690141 0.32394366
0.6056338 0.25352113 0.74647887 0.70422535 0.35211268 0.53521127]
PhoneService [0 1]
MultipleLines [0 1]
OnlineSecurity [0 1]
OnlineBackup [1 0]
DeviceProtection [0 1]
TechSupport [0 1]
StreamingTV [0 1]
StreamingMovies [0 1]
PaperlessBilling [1 0]
MonthlyCharges [0.11542289 0.38507463 0.35422886 ... 0.44626866 0.25820896 0.60149254]
TotalCharges [0.0012751 0.21586661 0.01031041 ... 0.03780868 0.03321025 0.78764136]
Churn [0 1]
InternetService_DSL [1 0]
InternetService_Fiber optic [0 1]
InternetService_No [0 1]
Contract_Month-to-month [1 0]
Contract_One year [0 1]
Contract_Two year [0 1]
PaymentMethod_Bank transfer (automatic) [0 1]
PaymentMethod_Credit card (automatic) [0 1]
PaymentMethod_Electronic check [1 0]
PaymentMethod_Mailed check [0 1]

```

```
df2.dtypes
```

```

gender                int64
SeniorCitizen         int64
Partner              int64
Dependents           int64
tenure               float64
PhoneService         int64
MultipleLines        int64
OnlineSecurity       int64
OnlineBackup         int64
DeviceProtection     int64
TechSupport         int64
StreamingTV          int64
StreamingMovies      int64
PaperlessBilling     int64
MonthlyCharges       float64
TotalCharges         float64
Churn               int64
InternetService_DSL  int64
InternetService_Fiber optic int64

```

```

InternetService_No          int64
Contract_Month-to-month    int64
Contract_One year           int64
Contract_Two year           int64
PaymentMethod_Bank transfer (automatic) int64
PaymentMethod_Credit card (automatic) int64
PaymentMethod_Electronic check int64
PaymentMethod_Mailed check  int64
dtype: object

X = df2.drop('Churn',axis='columns')
y = df2['Churn']
print('Shape of X:',X.shape)
print('Shape of y:',y.shape)

      Shape of X: (7032, 26)
      Shape of y: (7032,)

from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=5)
print('Shape of X_train:',X_train.shape)
print('Shape of X_test:',X_test.shape)
print('Shape of y_train:',y_train.shape)
print('Shape of y_test:',y_test.shape)

      Shape of X_train: (5625, 26)
      Shape of X_test: (1407, 26)
      Shape of y_train: (5625,)
      Shape of y_test: (1407,)

import tensorflow as tf
from tensorflow import keras

model = keras.Sequential([
    keras.layers.Dense(20,input_shape=(26,),activation='relu'),
    keras.layers.Dense(1,activation='sigmoid')
])
model.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])
model.fit(X_train,y_train,epochs=10)

Epoch 1/10
176/176 [=====] - 1s 1ms/step - loss: 0.5006 - accuracy: 0.7550
Epoch 2/10
176/176 [=====] - 0s 1ms/step - loss: 0.4357 - accuracy: 0.7902
Epoch 3/10
176/176 [=====] - 0s 2ms/step - loss: 0.4244 - accuracy: 0.7977
Epoch 4/10
176/176 [=====] - 0s 1ms/step - loss: 0.4196 - accuracy: 0.8021
Epoch 5/10
176/176 [=====] - 0s 2ms/step - loss: 0.4162 - accuracy: 0.8043
Epoch 6/10
176/176 [=====] - 0s 2ms/step - loss: 0.4148 - accuracy: 0.8048
Epoch 7/10
176/176 [=====] - 0s 1ms/step - loss: 0.4128 - accuracy: 0.8066
Epoch 8/10
176/176 [=====] - 0s 1ms/step - loss: 0.4121 - accuracy: 0.8076
Epoch 9/10
176/176 [=====] - 0s 1ms/step - loss: 0.4110 - accuracy: 0.8071
Epoch 10/10
176/176 [=====] - 0s 1ms/step - loss: 0.4097 - accuracy: 0.8089
<keras.callbacks.History at 0x7e596b5d7df0>

model.evaluate(X_test,y_test)

44/44 [=====] - 0s 1ms/step - loss: 0.4416 - accuracy: 0.7889
[0.44159233570098877, 0.7889125943183899]

from sklearn.metrics import classification_report,confusion_matrix
yp = model.predict(X_test)
yp

44/44 [=====] - 0s 1ms/step
array([[0.19795616],
       [0.3622312 ],
       [0.01491462],
       ...,
       [0.72733843],
       [0.6845046 ],
       [0.55635047]], dtype=float32)

yp.shape

```



```
(1407, 1)

yp[:5]

array([[0.19795616],
       [0.3622312 ],
       [0.01491462],
       [0.7472115 ],
       [0.48185456]], dtype=float32)

y_pred = []
for x in yp:
    if x > 0.5:
        y_pred.append(1)
    else:
        y_pred.append(0)

y_pred[:5]

[0, 0, 0, 1, 0]

y_test[:5]

2660    0
744     0
5579    1
64      1
3287    1
Name: Churn, dtype: int64

print('Classification Report:',classification_report(y_test,y_pred))
```

Classification Report:			precision	recall	f1-score	support
	0	0.84	0.87	0.85		999
	1	0.65	0.58	0.61		408
accuracy			0.79			1407
macro avg		0.74	0.73	0.73		1407
weighted avg		0.78	0.79	0.79		1407

```
import seaborn as sns
cm = confusion_matrix(y_test,y_pred)
sns.heatmap(cm,annot=True,fmt='.2f')
plt.title('Confusion Matrix')
plt.xlabel('Predicted')
plt.ylabel('True')
```

