

Customer Churn Prediction: Develop a model to predict customer churn in a subscription-based business.

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn import linear_model as lm
from sklearn.metrics import r2_score, mean_squared_error
import pickle
```

```
df=pd.read_csv('/content/drive/MyDrive/customer_churn_dataset.csv')
df
```

	CustomerID	Gender	Age	Tenure_Months	SubscriptionType	MonthlyCharges	TotalCharges	PaymentMethod	SupportTickets
0	CUST1000	Male	49	54	Standard	99.33	4912.77	Net Banking	1
1	CUST1001	Female	56	8	Standard	12.01	604.59	Debit Card	2
2	CUST1002	Male	66	27	Basic	57.62	2019.39	Cash	4
3	CUST1003	Male	69	27	Premium	97.08	4848.88	Credit Card	7
4	CUST1004	Male	49	34	Standard	54.69	4334.26	Credit Card	9
...
195	CUST1195	Female	47	48	Basic	21.47	1897.88	Debit Card	2
196	CUST1196	Female	55	21	Premium	90.19	1390.19	UPI	3
197	CUST1197	Female	55	39	Basic	12.62	3237.80	Debit Card	6
198	CUST1198	Male	62	36	Standard	54.83	2073.23	Credit Card	0
199	CUST1199	Male	68	33	Basic	43.99	175.66	UPI	5

200 rows × 11 columns

Next steps: [Generate code with df](#) [New interactive sheet](#)

```
y=df['Churn']
x=df[['Tenure_Months','MonthlyCharges','TotalCharges']]
```

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,shuffle=False)
```

```
model=lm.LinearRegression()
model.fit(x_train,y_train)
with open('model_pickle','wb') as f:
    pickle.dump(model,f)
```

```
y_pred=model.predict(x_test)
y_pred
```

```
array([0.30758194, 0.32137416, 0.27305266, 0.29429274, 0.30413932,
        0.27863456, 0.26446263, 0.25805565, 0.30779859, 0.24421493,
        0.29468234, 0.27981053, 0.33530736, 0.27725373, 0.27808355,
        0.25892512, 0.28797862, 0.30300736, 0.31855022, 0.27189399,
        0.29167176, 0.26743392, 0.27112456, 0.29632713, 0.25722897,
        0.30981823, 0.32307379, 0.29537844, 0.29392307, 0.29319153,
        0.25871615, 0.25499324, 0.25641097, 0.28820855, 0.29437996,
        0.2912685 , 0.28934389, 0.2888858 , 0.29407774, 0.27500711])
```

```
r2=r2_score(y_test,y_pred)
r2
```

```
-0.0044649814815447275
```

```
mse=mean_squared_error(y_test,y_pred)
mse
```

```
0.18833718402778965
```

```
plt.scatter(y_test,y_pred)
```

```
plt.xlabel('Actual')
plt.ylabel('Predicted')
```

```
Text(0, 0.5, 'Predicted')
```

