

```
!PIP INSTALL PANDAS
```

```
/bin/bash: line 1: PIP: command not found
```

```
PIP INSTALL MATPLOTLIB
```

```
File "/tmp/ipython-input-14-871634849.py", line 1
  PIP INSTALL MATPLOTLIB
    ^
```

```
SyntaxError: invalid syntax
```

```
PIP INSTALL SEABORN;
```

```
PIP INSTALL NUMPY
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
df = pd.read_csv('Customer Churn.csv')
df
```

```
from google.colab import files
uploaded = files.upload()
```

```
<IPython.core.display.HTML object>
```

```
Saving Customer Churn.csv to Customer Churn (2).csv
```

```
df.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

```

#replacing blank with 0 as tenure is 0 and total charges are recorded

```

df["TotalCharges"] = df["TotalCharges"].replace(" ", "0")
df["TotalCharges"] = df["TotalCharges"].astype(float)
df.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   float64
20  Churn                 7043 non-null   object
dtypes: float64(2), int64(2), object(17)
memory usage: 1.1+ MB

```

#replacing blank with 0 as tenure is 0 and total charges are recorded

```

df.isnull()

{"type": "dataframe"}

df.isnull().sum()

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

df.isnull().sum().sum()

np.int64(0)

```

df.isnull is showing, the null value in dataframes and it makes true, whichever shows false.

df.isnull().sum()- sum check shows 0 in dataframe

df.innull().sum().sum() - it shows overall 0 in dataframe

```

df.describe()

{"summary": "{\n  \"name\": \"df\",\n  \"rows\": 8,\n  \"fields\": [\n    {\n      \"column\": \"SeniorCitizen\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 2489.9992387084,\n        \"min\": 0.0,\n        \"max\": 7043.0,\n        \"num_unique_values\": 5,\n        \"samples\": [\n          0.1621468124378816,\n          1.0,\n          0.36861160561002687\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      }\n    },\n    {\n      \"column\": \"tenure\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 2478.9752758409018,\n        \"min\": 0.0,\n        \"max\": 7043.0,\n        \"num_unique_values\": 5,\n        \"samples\": [\n          0.1621468124378816,\n          1.0,\n          0.36861160561002687\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      }\n    }\n  ]\n}"}

```

```

{"num_unique_values": 8,\n      "samples": [\n        32.37114865824223,\n        29.0,\n        7043.0\n      ],\n      "semantic_type": "\n",\n      "description": "\n"},\n      {\n        "column": "MonthlyCharges",\n        "properties": {\n          "dtype": "number",\n          "std": 2468.7047672837775,\n          "min": 18.25,\n          "max": 7043.0,\n          "num_unique_values": 8,\n          "samples": [\n            64.76169246059918,\n            70.35,\n            7043.0\n          ],\n          "semantic_type": "\n",\n          "description": "\n"},\n          {\n            "column": "TotalCharges",\n            "properties": {\n              "dtype": "number",\n              "std": 3122.5732655623974,\n              "min": 0.0,\n              "max": 8684.8,\n              "num_unique_values": 8,\n              "samples": [\n                2279.7343035638223,\n                1394.55,\n                7043.0\n              ],\n              "semantic_type": "\n",\n              "description": "\n"},\n            }\n          }\n        ],\n        "type": "dataframe"}

df["customerID"].duplicated().sum()

np.int64(0)

df.duplicated().sum()

np.int64(0)

```

#converted 0 and 1 value of Senior Citizen to yes/no to make it easier to understand

```

def conv(value):
    if value == 1:
        return "Yes"
    else:
        return "No"

df['SeniotrCitizen'] = df['SeniorCitizen'].apply(conv)
df.head()

{"type": "dataframe", "variable_name": "df"}

df.head(30)

{"type": "dataframe", "variable_name": "df"}

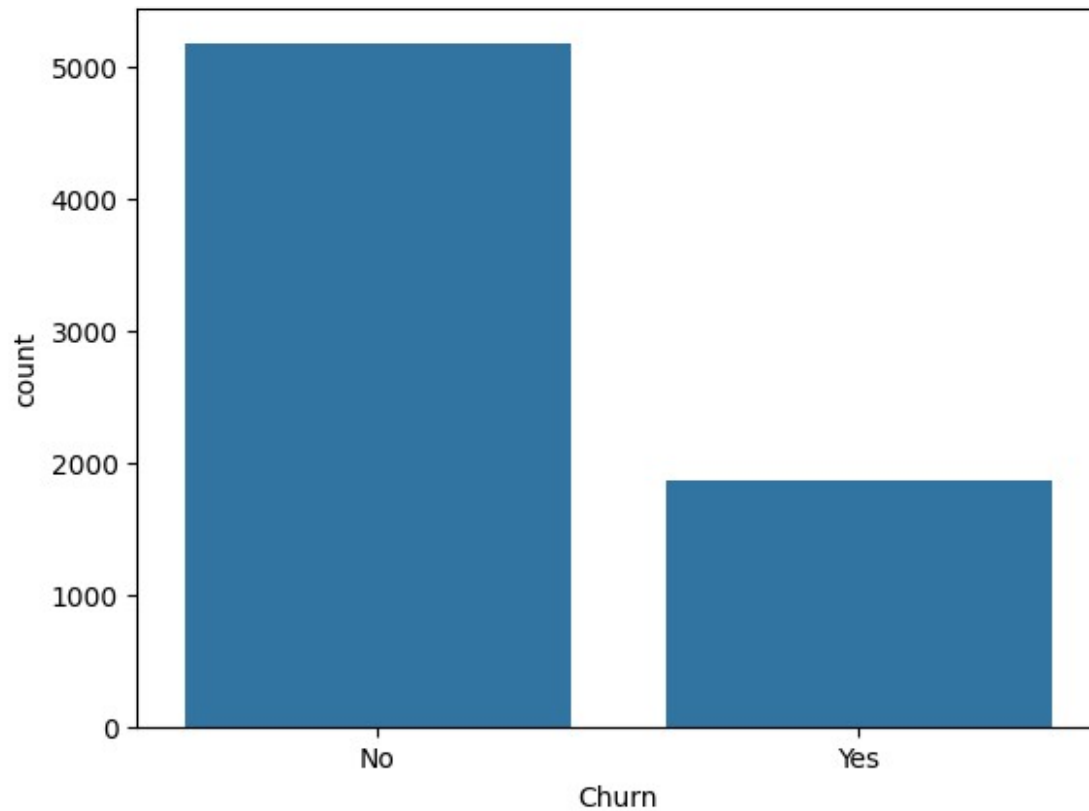
```

#df.head() - it is used for rows 30, 20 whatever, in senior citizen it shows yes in row for checking the rows.

```

sns.countplot(x = df['Churn'], data = df)
plt.show()

```

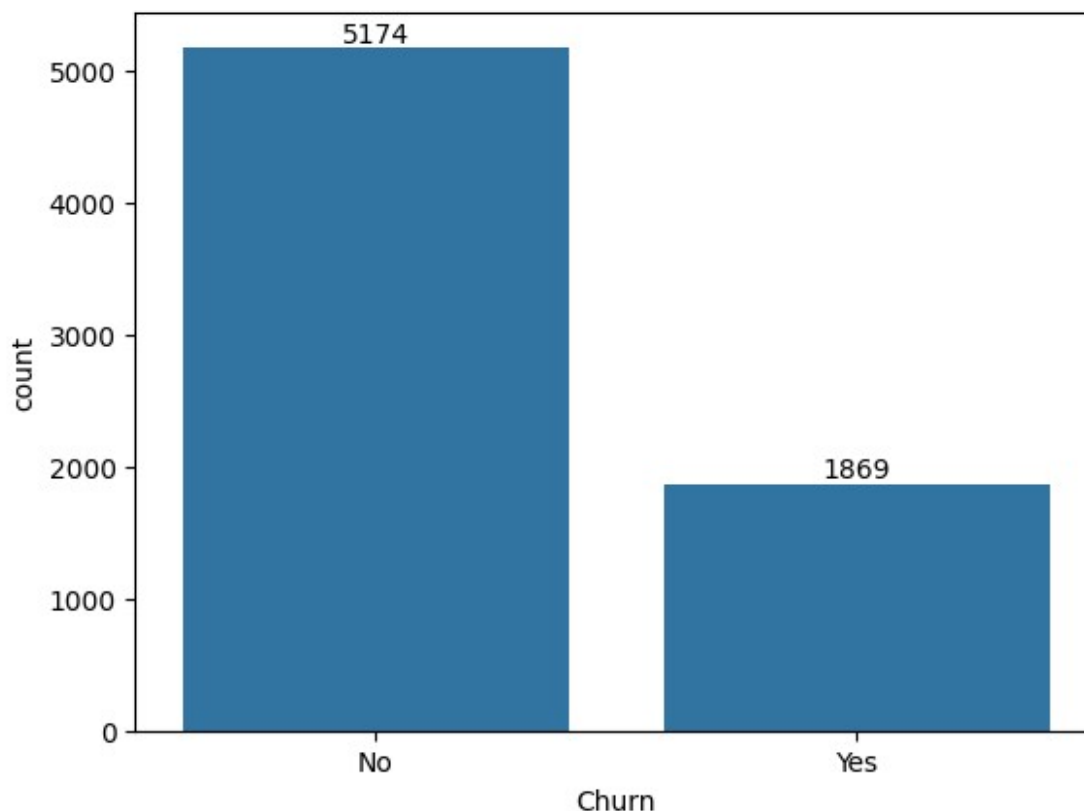


in counplot - it shows yes/no value

```
import seaborn as sns
import matplotlib.pyplot as plt

ax = sns.countplot(x = 'Churn', data = df)

ax.bar_label(ax.containers[0])
plt.show()
```



```
df.describe()
```

```
{"summary": "{\n  \"name\": \"df\",\n  \"rows\": 8,\n  \"fields\": [\n    {\n      \"column\": \"SeniorCitizen\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 2489.9992387084,\n        \"min\": 0.0,\n        \"max\": 7043.0,\n        \"num_unique_values\": 5,\n        \"samples\": [\n          0.1621468124378816,\n          1.0,\n          0.36861160561002687\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      }\n    },\n    {\n      \"column\": \"tenure\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 2478.9752758409018,\n        \"min\": 0.0,\n        \"max\": 7043.0,\n        \"num_unique_values\": 8,\n        \"samples\": [\n          32.37114865824223,\n          29.0,\n          7043.0\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      }\n    },\n    {\n      \"column\": \"MonthlyCharges\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 2468.7047672837775,\n        \"min\": 18.25,\n        \"max\": 7043.0,\n        \"num_unique_values\": 8,\n        \"samples\": [\n          64.76169246059918,\n          70.35,\n          7043.0\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      }\n    },\n    {\n      \"column\": \"TotalCharges\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 3122.5732655623974,\n        \"min\": 0.0,\n        \"max\": 8684.8,\n
```

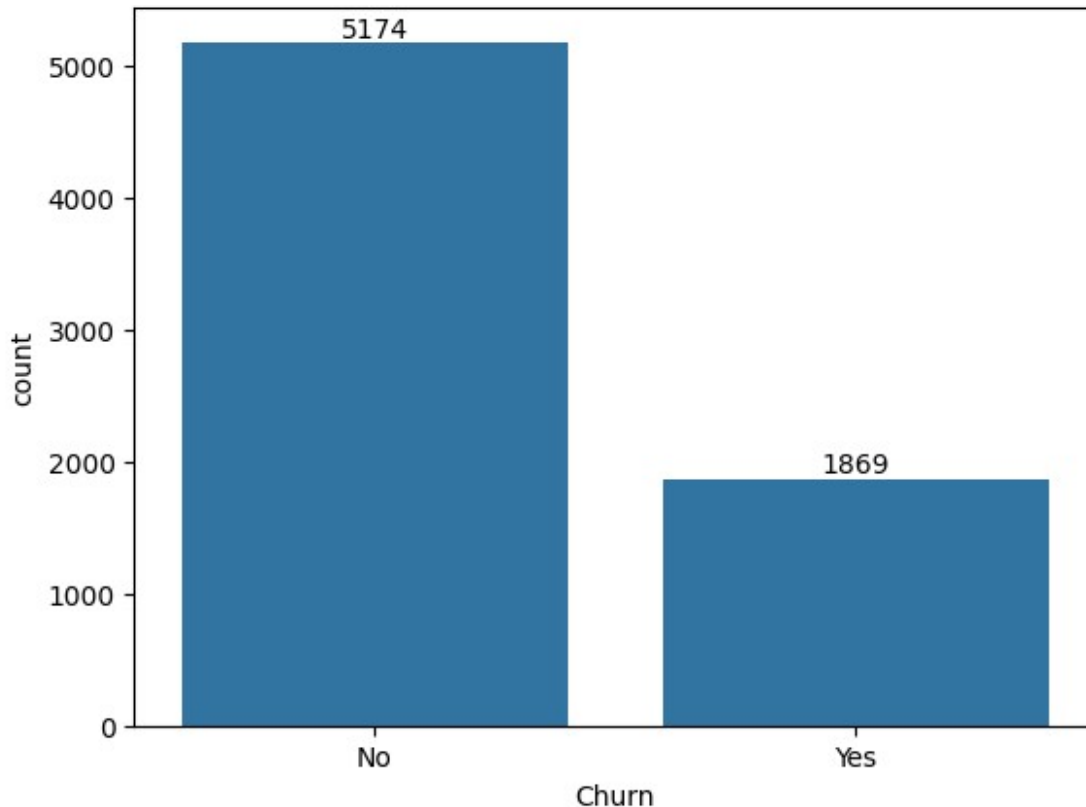
```

\ "num_unique_values\ ": 8,\n          \ "samples\ ": [\n
2279.7343035638223,\n          1394.55,\n          7043.0\n
],\n
\ "semantic_type\ ": \ "\",\n          \ "description\ ": \ "\ "\n
}\n          }\n          ]\n          }", "type": "dataframe"}

ax = sns.countplot(x = 'Churn', data = df)

ax.bar_label(ax.containers[0])
plt.show()

```



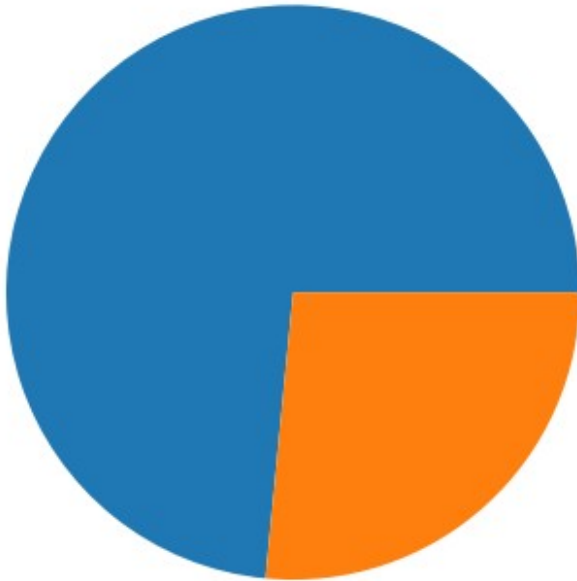
```

gb = df.groupby("Churn").agg({'Churn': "count"})
gb
#plt.pie(df['Churn'])
#plt.show()

{"repr_error": "cannot insert Churn, already
exists", "type": "dataframe", "variable_name": "gb"}

gb = df.groupby("Churn").agg({'Churn': "count"})
plt.pie(gb['Churn'])
plt.show()
gb

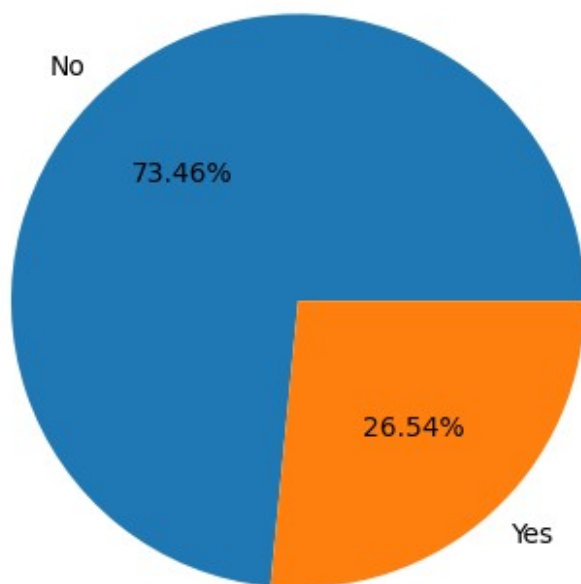
```



```
{"repr_error": "cannot insert Churn, already  
exists", "type": "dataframe", "variable_name": "gb"}  
  
gb = df.groupby("Churn").agg({'Churn': "count"})  
plt.pie(gb['Churn'], labels = gb.index)  
plt.show()
```

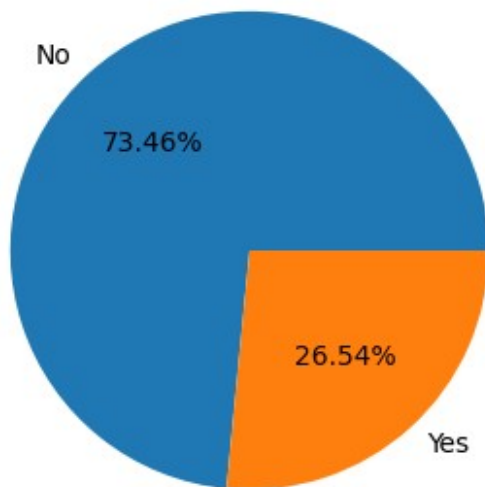



```
gb = df.groupby("Churn").agg({'Churn': "count"})  
plt.pie(gb['Churn'], labels = gb.index, autopct = "%1.2f%%")  
plt.show()
```



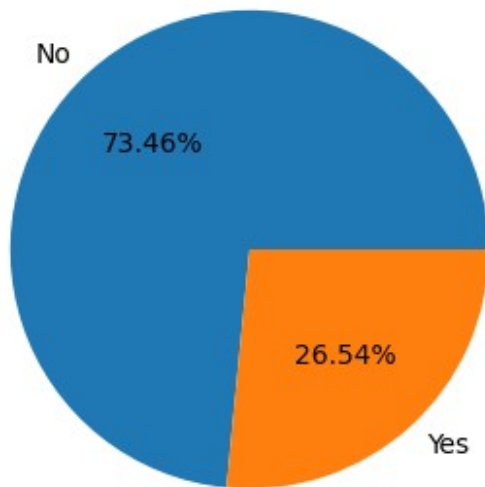
##26.5% customer are churn out from the total customer, i.e.(converted into another sim).

```
plt.figure(figsize = (4,4))
gb = df.groupby("Churn").agg({'Churn':"count"})
plt.pie(gb['Churn'], labels = gb.index, autopct = "%1.2f%%")
plt.show()
```



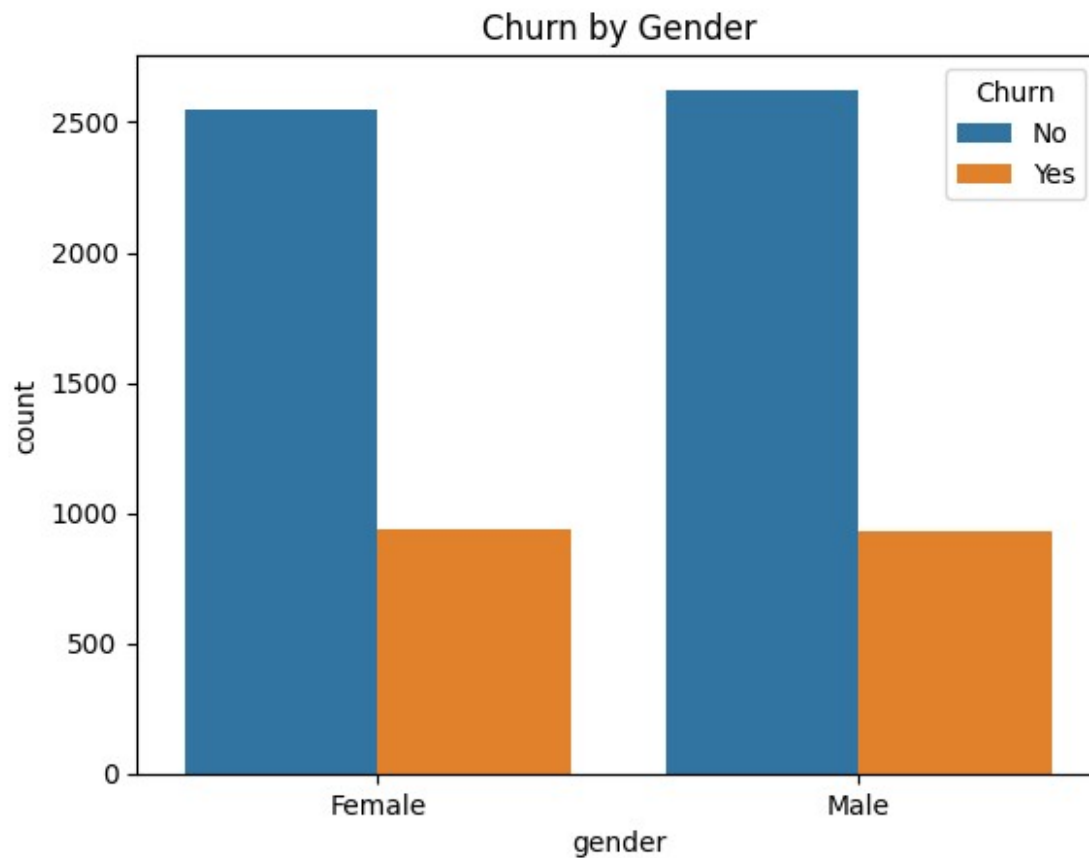
```
plt.figure(figsize = (4,4))
gb = df.groupby("Churn").agg({'Churn':"count"})
plt.pie(gb['Churn'], labels = gb.index, autopct = "%1.2f%%")
plt.title("percentage of Churned Customers",fontsize = 10)
plt.show()
```

percentage of Churned Customers



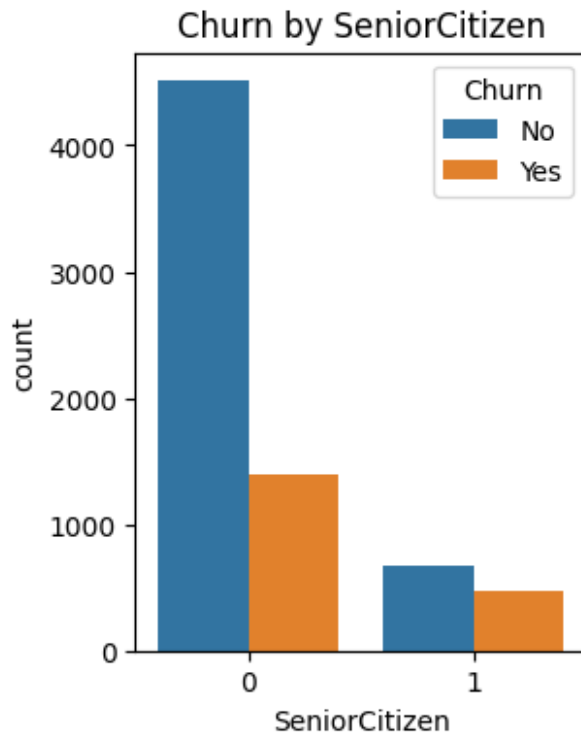
*#from the given pie chart we can conclude that 26.54% of our customers have churned out
#now let's explore the reason behind it.*

```
sns.countplot(x = 'gender', data = df, hue = 'Churn')  
plt.title ("Churn by Gender")  
plt.show()
```



#hue -its denoted, coulumn basis of churn

```
plt.figure(figsize=(3,4))  
sns.countplot(x='SeniorCitizen', data=df, hue='Churn')  
plt.title("Churn by SeniorCitizen")  
plt.show()
```



#copied the code and ask to chgatkpt

```
#plt.figure(figsize=(3,4))
```

```
#sns.countplot(x='SeniorCitizen', data=df, hue='Churn')
```

```
#plt.title("Churn by SeniorCitizen")
```

```
#plt.show()
```

#i want to create a stack bar chart which gives me labels as % to total

#chatgpt coding given below

```
ct = pd.crosstab(df['SeniorCitizen'], df['Churn'], normalize='index')
* 100

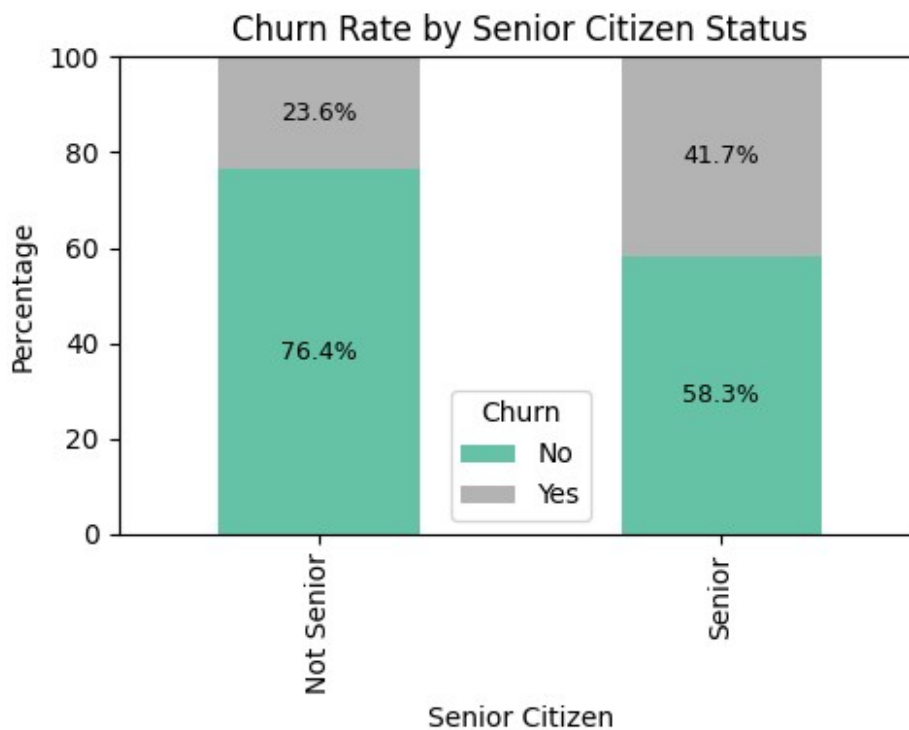
# 2. Plot stacked bar
ax = ct.plot(kind='bar', stacked=True, figsize=(5,4), colormap='Set2')

# 3. Add % labels to each segment
for i, row in ct.iterrows():
    cumulative = 0
    for churn_value in ct.columns:
        percent = row[churn_value]
        if percent > 0:
            ax.text(i, cumulative + percent / 2, f'{percent:.1f}%',
ha='center', va='center', fontsize=9)
```

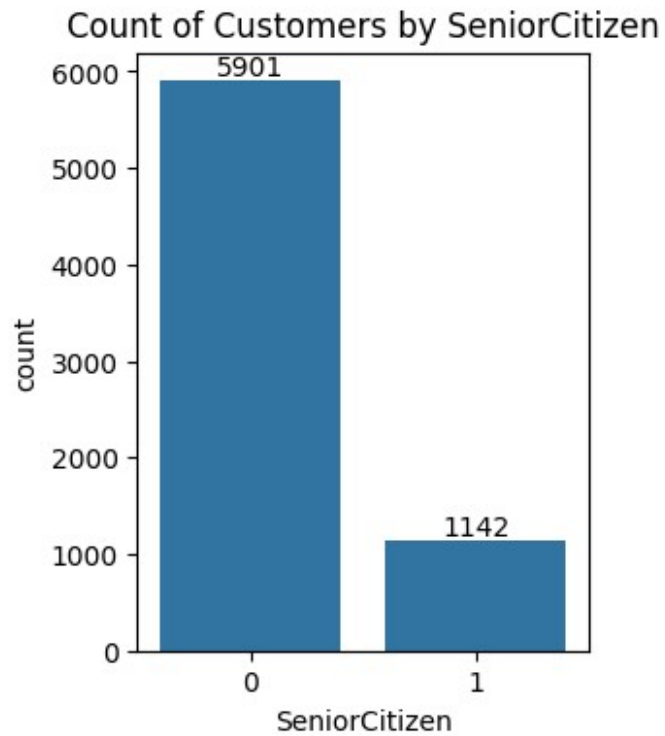
```
cumulative += percent
```

4. Styling

```
plt.title("Churn Rate by Senior Citizen Status")
plt.xlabel("Senior Citizen")
plt.ylabel("Percentage")
plt.xticks([0,1], ["Not Senior", "Senior"]) # Optional: label 0/1
plt.legend(title='Churn')
plt.ylim(0, 100)
plt.tight_layout()
plt.show()
```

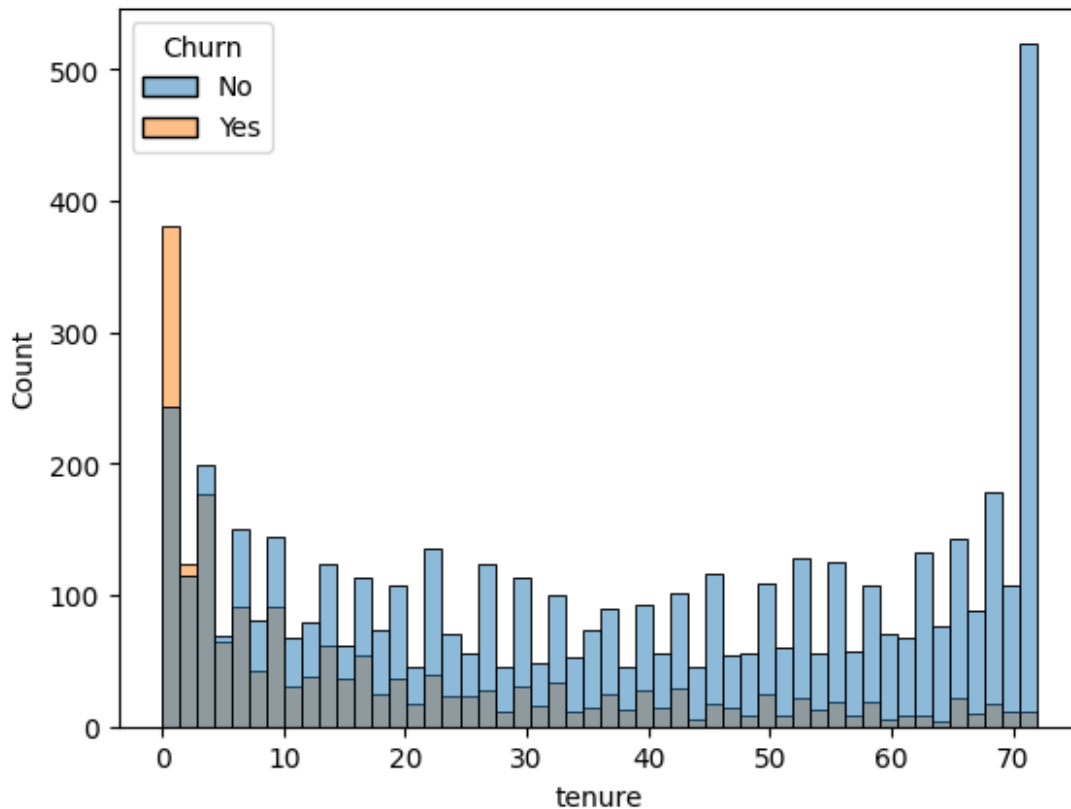


```
plt.figure(figsize=(3,4))
ax = sns.countplot(x='SeniorCitizen', data=df)
ax.bar_label(ax.containers[0])
plt.title("Count of Customers by SeniorCitizen")
plt.show()
```



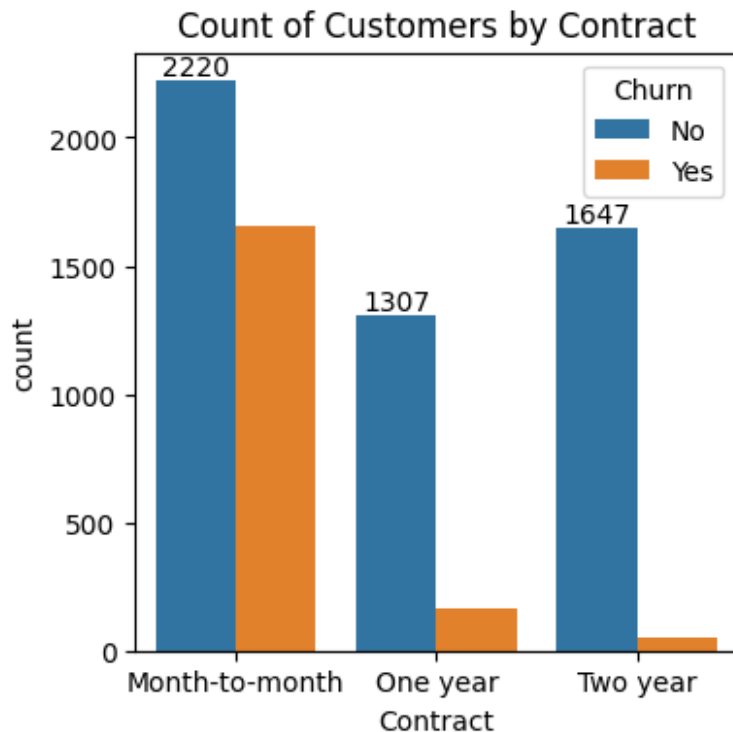
#COMPARATIVELY a greater percentage of people in seniorcitizen category have churched out.

```
sns.histplot (x = 'tenure', data = df, bins = 50, hue = "Churn")  
plt.show()
```



#people who have used our services for a long time have stayed and people who have used our srvcies for 1 or 2 month have churned

```
plt.figure(figsize=(4,4))
ax = sns.countplot(x='Contract', data=df, hue='Churn')
ax.bar_label(ax.containers[0])
plt.title("Count of Customers by Contract")
plt.show()
```

#People who have a contract for month to month likely to churn then to the people who are staying for a year/ 2 year.

```
df.columns.values

array(['customerID', 'gender', 'SeniorCitizen', 'Partner',
      'Dependents',
      'tenure', 'PhoneService', 'MultipleLines', 'InternetService',
      'OnlineSecurity', 'OnlineBackup', 'DeviceProtection',
      'TechSupport', 'StreamingTV', 'StreamingMovies', 'Contract',
      'PaperlessBilling', 'PaymentMethod', 'MonthlyCharges',
      'TotalCharges', 'Churn', 'SeniorCitizen'], dtype=object)

# List of columns to plot
cols = ['PhoneService', 'MultipleLines', 'InternetService',
        'OnlineSecurity', 'OnlineBackup', 'DeviceProtection',
        'TechSupport', 'StreamingTV', 'StreamingMovies']

# Define subplot layout
n_cols = 3
n_rows = (len(cols) + n_cols - 1) // n_cols

# Create figure and axes
fig, axes = plt.subplots(n_rows, n_cols, figsize=(16, n_rows * 4))
axes = axes.flatten() # Flatten in case of multiple rows

# Loop through each column and plot on its subplot
```

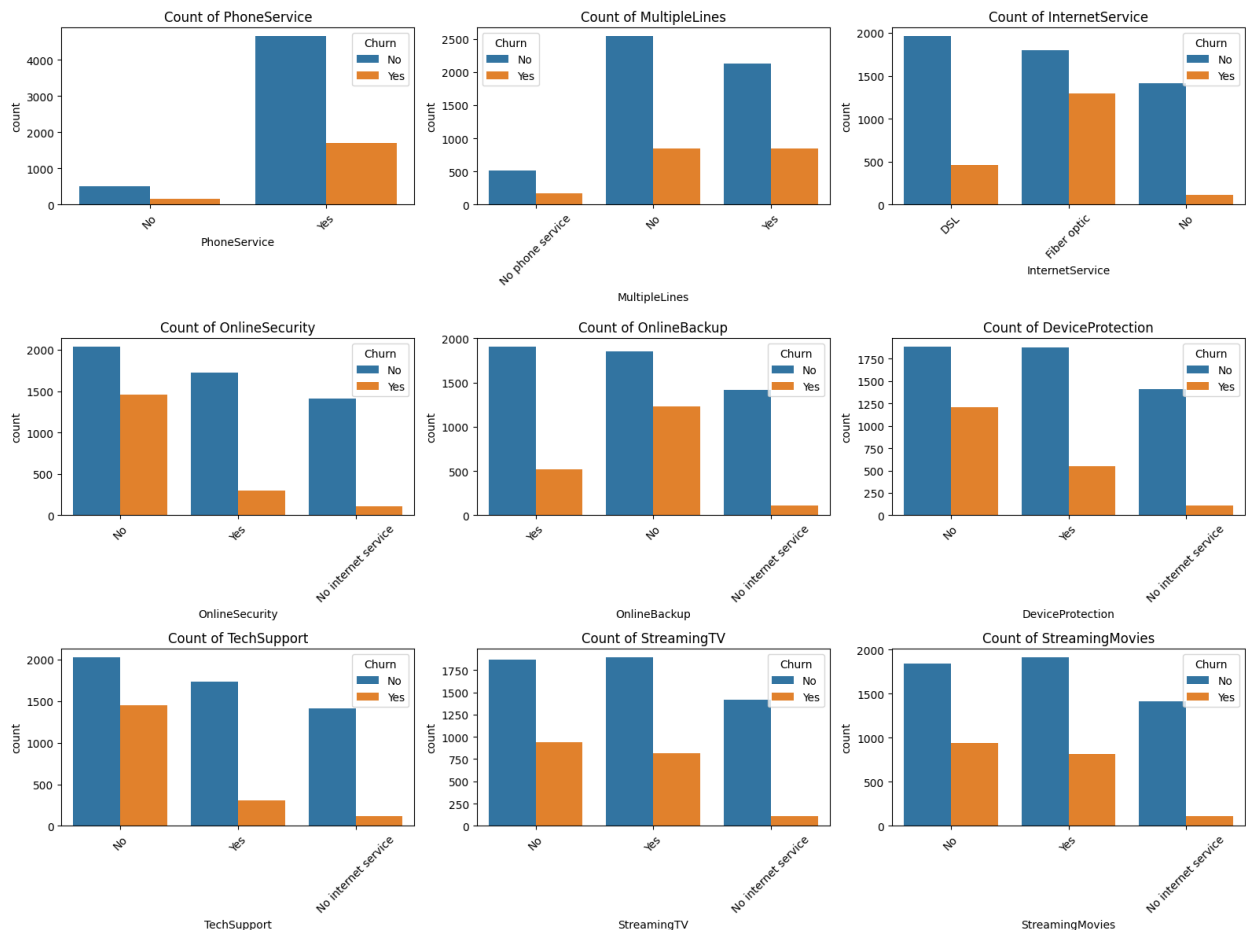
```

for i, col in enumerate(cols):
    sns.countplot(x=col, data=df, ax=axes[i], hue = "Churn")
    axes[i].set_title(f'Count of {col}')
    axes[i].tick_params(axis='x', rotation=45)

# Hide any unused axes
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])

plt.tight_layout()
plt.show()

```



#here in one plot there are many subplots

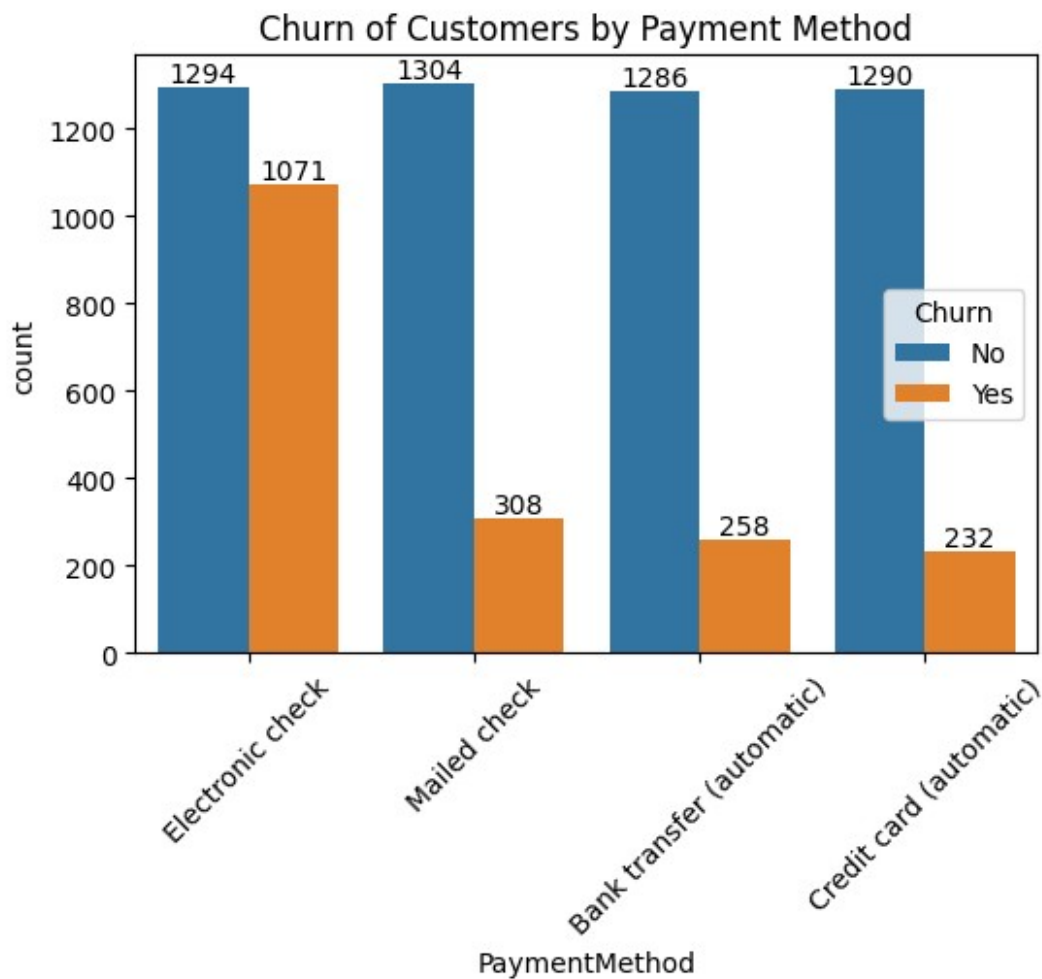
#Customers without internet services or who opted "No" for add-on services like Online Security, Backup, and Tech Support showed higher churn rates.

Those using Fiber Optic Internet also showed a noticeably higher churn compared to DSL or no internet users.

Services such as StreamingTV and StreamingMovies showed mixed churn, but churn was generally higher where services were not active.

This suggests that lack of value-added services **is** strongly associated **with** customer churn.

```
plt.figure(figsize=(6,4))
ax = sns.countplot(x='PaymentMethod', data=df, hue='Churn')
ax.bar_label(ax.containers[0])
ax.bar_label(ax.containers[1])
plt.title("Churn of Customers by Payment Method")
plt.xticks(rotation=45)
plt.show()
```



#customers is likely to churn when he is using electronic check as a payment method