



In [2]: `pip install numpy`

Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: numpy in c:\users\furqu\appdata\roaming\python\python313\site-packages (2.3.2)
Note: you may need to restart the kernel to use updated packages.

In [1]: `pip install numpy`

Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: numpy in c:\users\furqu\appdata\roaming\python\python313\site-packages (2.3.2)
Note: you may need to restart the kernel to use updated packages.

In [2]: `pip install pandas`

Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: pandas in c:\users\furqu\appdata\roaming\python\python313\site-packages (2.3.2)
Requirement already satisfied: numpy>=1.26.0 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from pandas) (2.3.2)
Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from pandas) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from pandas) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from pandas) (2025.2)
Requirement already satisfied: six>=1.5 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from python-dateutil>=2.8.2->pandas) (1.17.0)
Note: you may need to restart the kernel to use updated packages.

In [3]: `pip install numpy`

Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: numpy in c:\users\furqu\appdata\roaming\python\python313\site-packages (2.3.2)
Note: you may need to restart the kernel to use updated packages.

In [5]: `pip install matplotlib`

Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: matplotlib in c:\users\furqu\appdata\roaming\python\python313\site-packages (3.10.7)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from matplotlib) (1.3.3)
Requirement already satisfied: cyclor>=0.10 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from matplotlib) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from matplotlib) (4.60.1)
Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from matplotlib) (1.4.9)
Requirement already satisfied: numpy>=1.23 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from matplotlib) (2.3.2)
Requirement already satisfied: packaging>=20.0 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from matplotlib) (25.0)
Requirement already satisfied: pillow>=8 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from matplotlib) (12.0.0)
Requirement already satisfied: pyparsing>=3 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from matplotlib) (3.2.5)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from matplotlib) (2.9.0.post0)
Requirement already satisfied: six>=1.5 in c:\users\furqu\appdata\roaming\python\python313\site-packages (from python-dateutil>=2.7->matplotlib) (1.17.0)
Note: you may need to restart the kernel to use updated packages.

In []:

```
In [4]: 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.head()
```

Out[4]:

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneServ
--	------------	--------	---------------	---------	------------	--------	-----------

0	7590-VHVEG	Female	0	Yes	No	1	
1	5575-GNVDE	Male	0	No	No	34	
2	3668-QPYBK	Male	0	No	No	2	
3	7795-CFOCW	Male	0	No	No	45	
4	9237-HQITU	Female	0	No	No	2	

5 rows x 21 columns

In [11]: 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

```

Replace blank with 0 as tenure is 0 and no total charges are recorded

Also Converting object (show due to blank rows) convert into a float

```

In [9]: df["TotalCharges"] = df["TotalCharges"].replace(" ", "0")
df["TotalCharges"] = df["TotalCharges"].astype("float")

```

```

In [17]: 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

```

Understading What is isnull()

```

In [10]: #df.isnull() we dont understant what is false so we use sum() here
df.isnull().sum()           #This is show in our data there is null value or not
df.isnull().sum().sum()    # This show the overall data is their any null value

```

```
Out[10]: np.int64(0)
```

Now we use descriptive analysis to understand

```
In [11]: df.describe()
```

```
Out[11]:
```

	SeniorCitizen	tenure	MonthlyCharges	TotalCharges
count	7043.000000	7043.000000	7043.000000	7043.000000
mean	0.162147	32.371149	64.761692	2279.734304
std	0.368612	24.559481	30.090047	2266.794470
min	0.000000	0.000000	18.250000	0.000000
25%	0.000000	9.000000	35.500000	398.550000
50%	0.000000	29.000000	70.350000	1394.550000
75%	0.000000	55.000000	89.850000	3786.600000
max	1.000000	72.000000	118.750000	8684.800000

Check Duplicate Values

```
In [12]: #df.duplicated().sum() #check entire data
df["customerID"].duplicated().sum() # this helps to find a particular column
```

```
Out[12]: np.int64(0)
```

Here we Convert 0/1 value into yes/no to make it easier to understand

```
In [13]: def conv(value):
            if value == 1:
                return "yes"
            else:
                return "no"

df["SeniorCitizen"] = df["SeniorCitizen"].apply(conv)
```

```
In [17]: df.head(10) # here we check 10 starting row is here any seniorcitizen or not
```

Out[17]:

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneSei
0	7590-VHVEG	Female	no	Yes	No	1	
1	5575-GNVDE	Male	no	No	No	34	
2	3668-QPYBK	Male	no	No	No	2	
3	7795-CFOCW	Male	no	No	No	45	
4	9237-HQITU	Female	no	No	No	2	
5	9305-CDSKC	Female	no	No	No	8	
6	1452-KIOVK	Male	no	No	Yes	22	
7	6713-OKOMC	Female	no	No	No	10	
8	7892-POOKP	Female	no	Yes	No	28	
9	6388-TABGU	Male	no	No	Yes	62	
10	9763-GRSKD	Male	no	Yes	Yes	13	
11	7469-LKBCI	Male	no	No	No	16	
12	8091-TTVAX	Male	no	Yes	No	58	
13	0280-XJGEX	Male	no	No	No	49	
14	5129-JLPIS	Male	no	No	No	25	
15	3655-SNQYZ	Female	no	Yes	Yes	69	
16	8191-XWSZG	Female	no	No	No	52	
17	9959-WOFKT	Male	no	No	Yes	71	
18	4190-MFLUW	Female	no	Yes	Yes	10	
19	4183-MYFRB	Female	no	No	No	21	
20	8779-QRDMV	Male	yes	No	No	1	

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneSei
21	1680-VDCWW	Male	no	Yes	No	12	
22	1066-JKSGK	Male	no	No	No	1	
23	3638-WEABW	Female	no	Yes	No	58	
24	6322-HRPFA	Male	no	Yes	Yes	49	
25	6865-JZNKO	Female	no	No	No	30	
26	6467-CHFZW	Male	no	Yes	Yes	47	
27	8665-UTDHZ	Male	no	Yes	Yes	1	
28	5248-YGIJN	Male	no	Yes	No	72	
29	8773-HHUOZ	Female	no	No	Yes	17	

30 rows × 21 columns

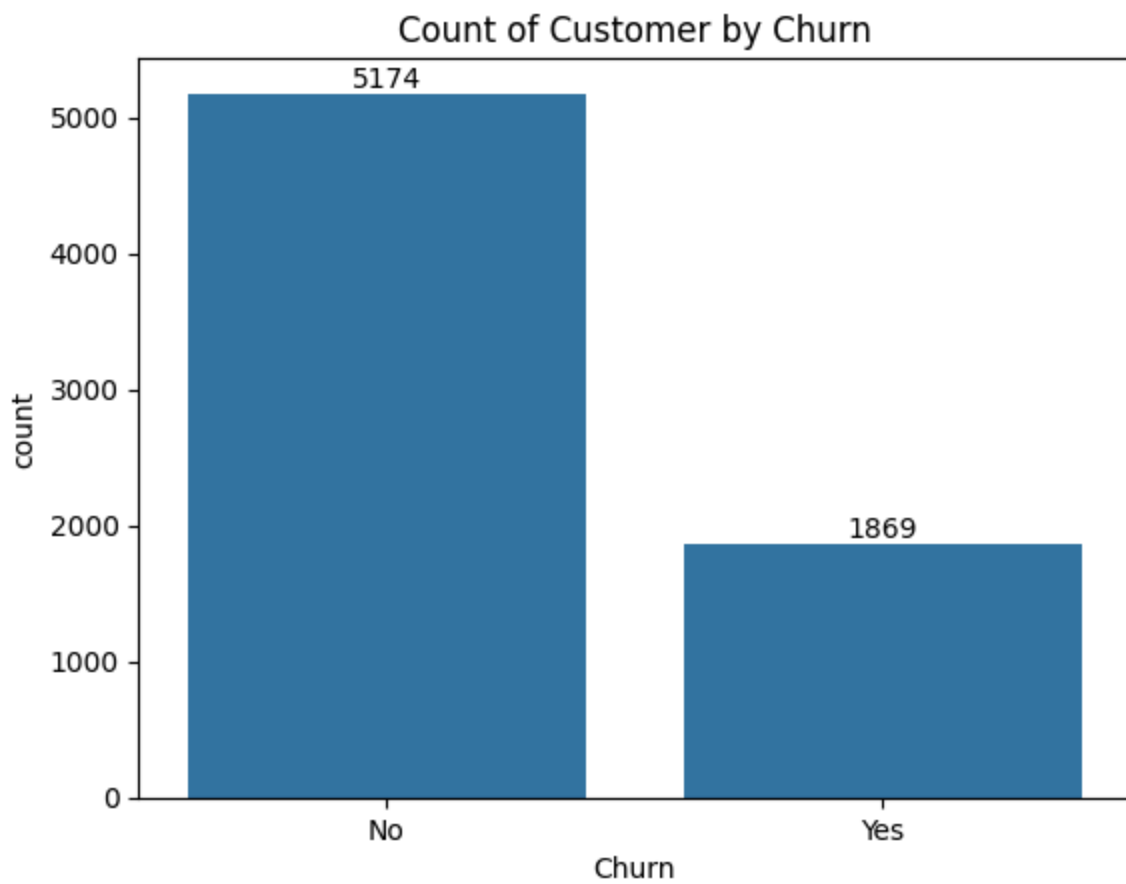
In [19]: `df.tail()` *#is called to retrieve the last rows of the DataFrame. # here easily*

Out[19]:

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneS
7038	6840-RESVB	Male	no	Yes	Yes	24	
7039	2234-XADUH	Female	no	Yes	Yes	72	
7040	4801-JZAZL	Female	no	Yes	Yes	11	
7041	8361-LTMKD	Male	yes	Yes	No	4	
7042	3186-AJIEK	Male	no	No	No	66	

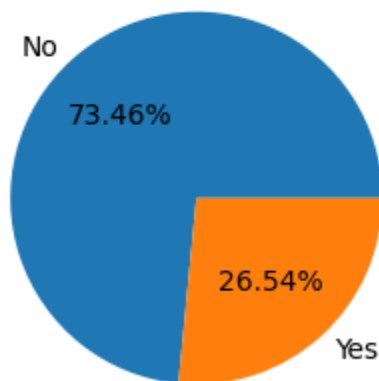
5 rows × 21 columns

In [58]: `ax = sns.countplot(x = 'Churn', data=df)
ax.bar_label(ax.containers[0])
plt.title('Count of Customer by Churn')
plt.show()`



```
In [61]: plt.figure(figsize = (3,4))  
plt.title('Percentage of Churn Customers')  
gb = df.groupby('Churn').agg({'Churn':'count'})  
plt.pie(gb['Churn'], labels = gb.index, autopct = '%1.2f%%')  
plt.show()
```

Percentage of Churn Customers

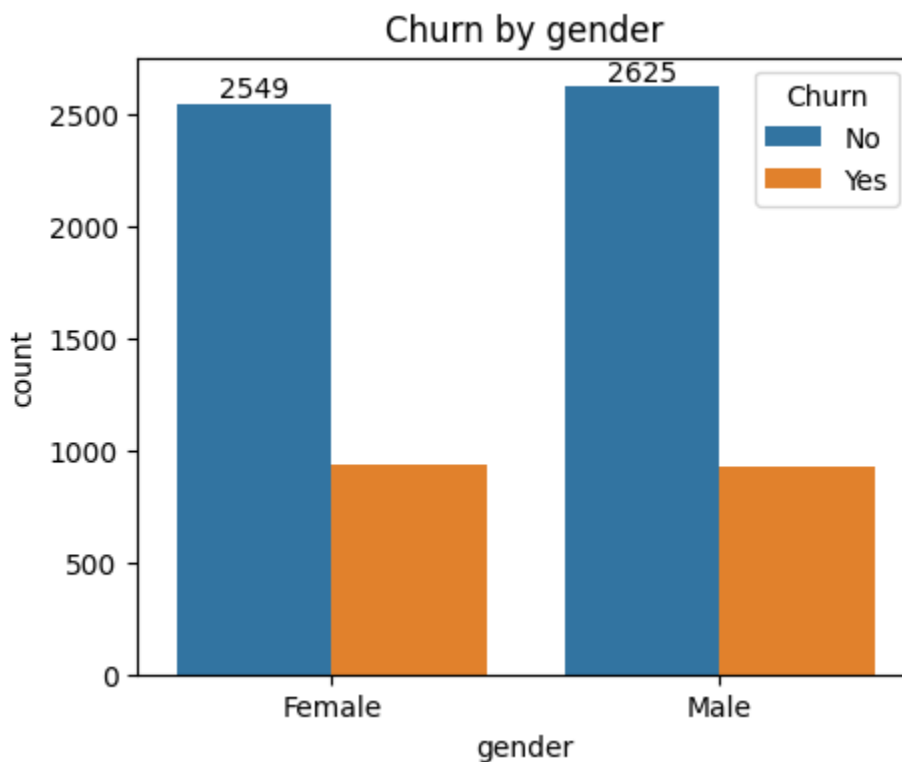


From the given pie chart we can conclude that 26.54% of our customers have churned out

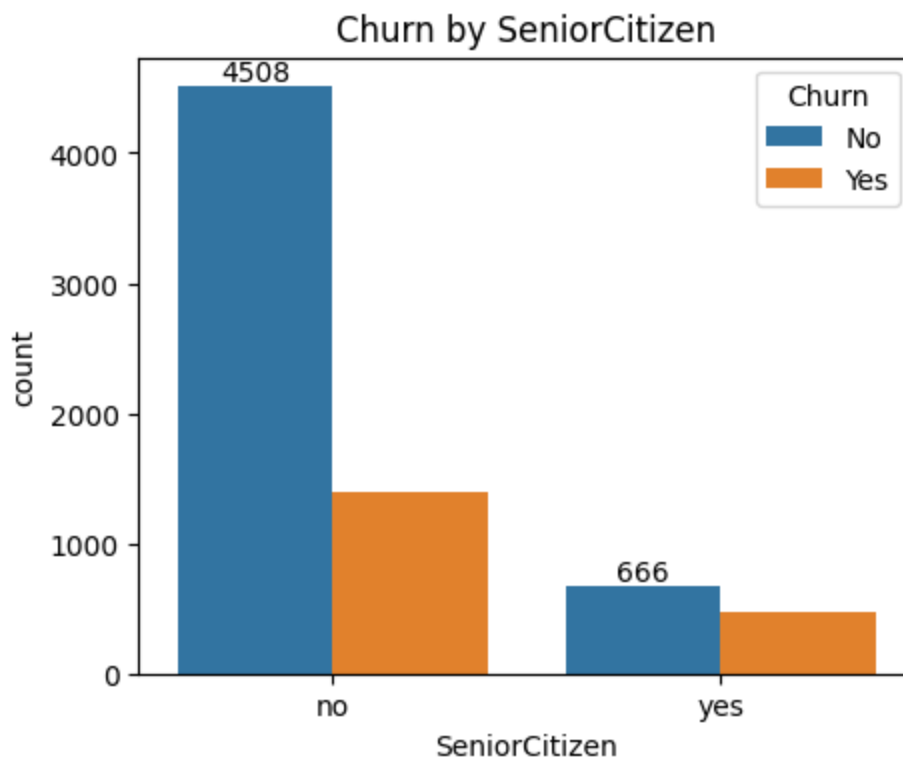
Now let explore reason behind it

Now we explore by gender

```
In [78]: plt.figure(figsize = (5,4))
ax=sns.countplot(x = "gender" ,data = df, hue ="Churn")
plt.title("Churn by gender")
ax.bar_label(ax.containers[0])
plt.show()
```



```
In [81]: plt.figure(figsize = (5,4))
ax=sns.countplot(x = "SeniorCitizen" ,data = df, hue ="Churn")
plt.title("Churn by SeniorCitizen")
ax.bar_label(ax.containers[0])
plt.show()
```



```
In [23]: import pandas as pd
import matplotlib.pyplot as plt

# 1 Group and calculate percentage by SeniorCitizen & Churn
data = (
    df.groupby(['SeniorCitizen', 'Churn'])
      .size()
      .groupby(level=0)
      .apply(lambda x: 100 * x / x.sum())
      .unstack(fill_value=0)
)

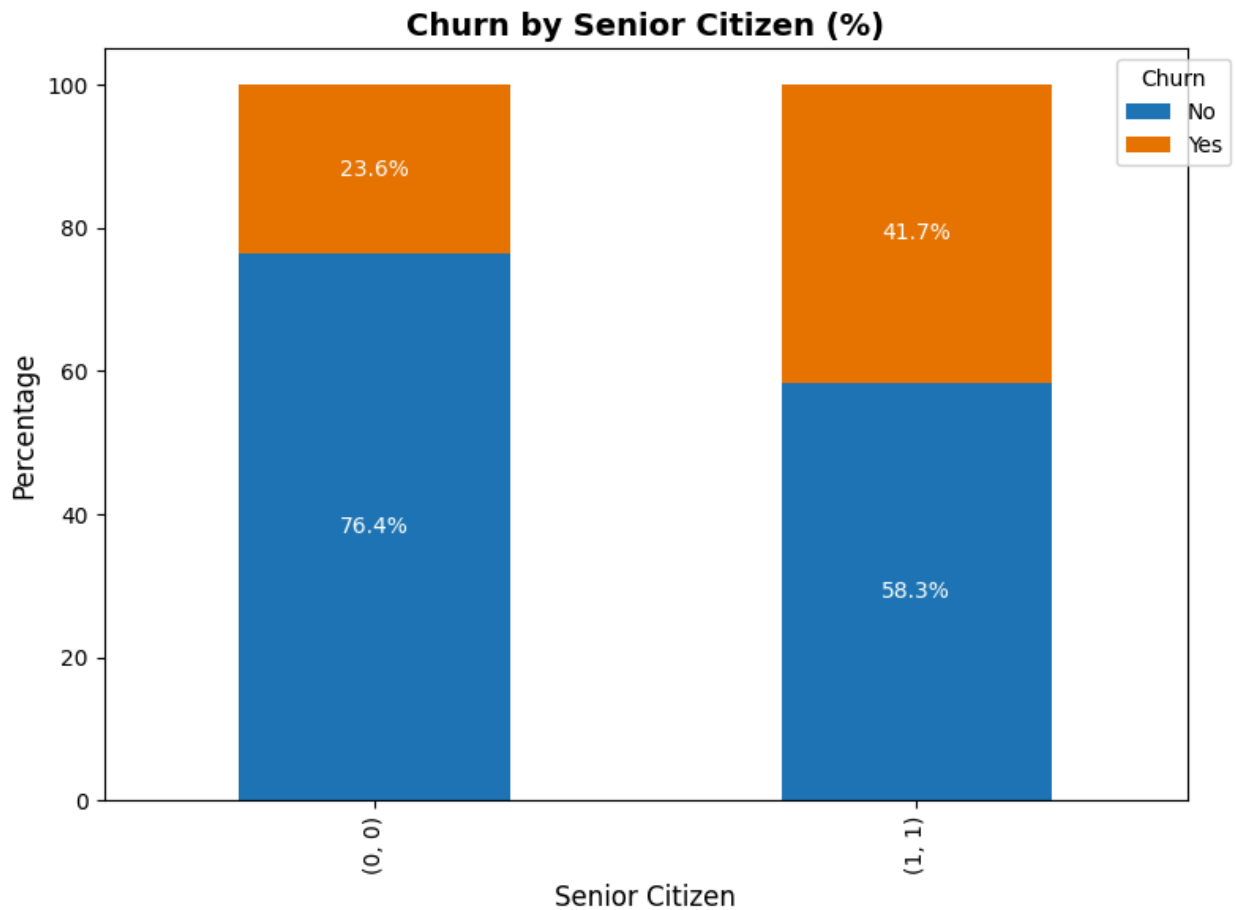
# 2 Create a larger figure
plt.figure(figsize=(6,5)) # You can increase width or height as you like

# 3 Plot stacked bar chart
ax = data.plot(kind='bar', stacked=True, color=['#1f77b4', '#e67300'], figsize=

# 4 Add % labels
for container in ax.containers:
    ax.bar_label(container, fmt='%.1f%%', label_type='center', color='white',

# 5 Add titles and labels
plt.title("Churn by Senior Citizen (%)", fontsize=14, fontweight='bold')
plt.ylabel("Percentage", fontsize=12)
plt.xlabel("Senior Citizen", fontsize=12)
plt.legend(title="Churn", bbox_to_anchor=(1.05, 1))
plt.tight_layout()
plt.show()
```

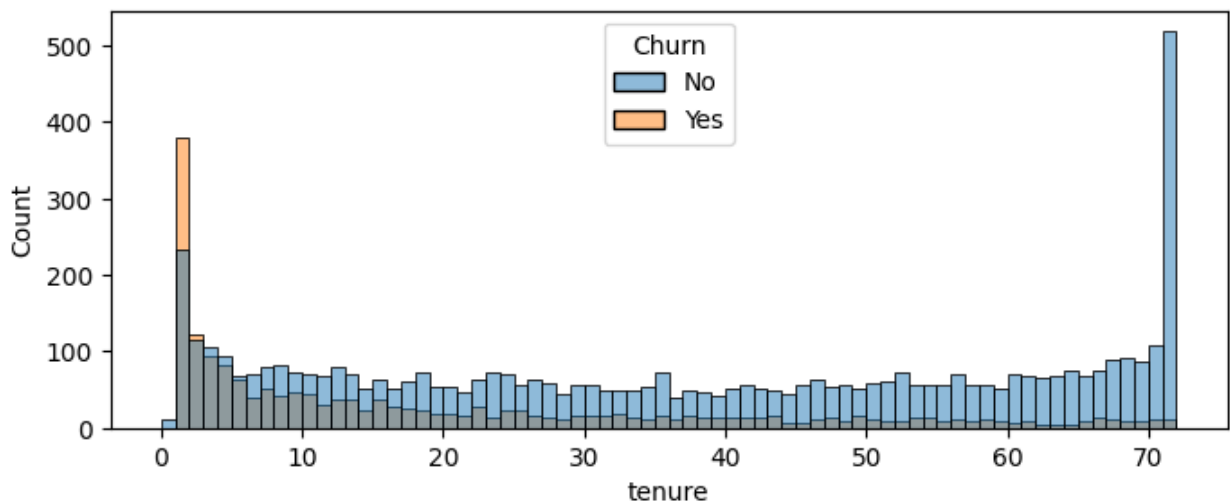
<Figure size 600x500 with 0 Axes>



Comparative a greated percentage of people in senior citizen category have churned

```
In [27]: plt.figure(figsize=(8,3))  
sns.histplot(x="tenure", data=df, bins=72, hue="Churn")  
plt.show
```

```
Out[27]: <function matplotlib.pyplot.show(close=None, block=None)>
```



People who have use our services for a long time have stayed and people who have used our services for a one or two month

In [28]: `df.head()`

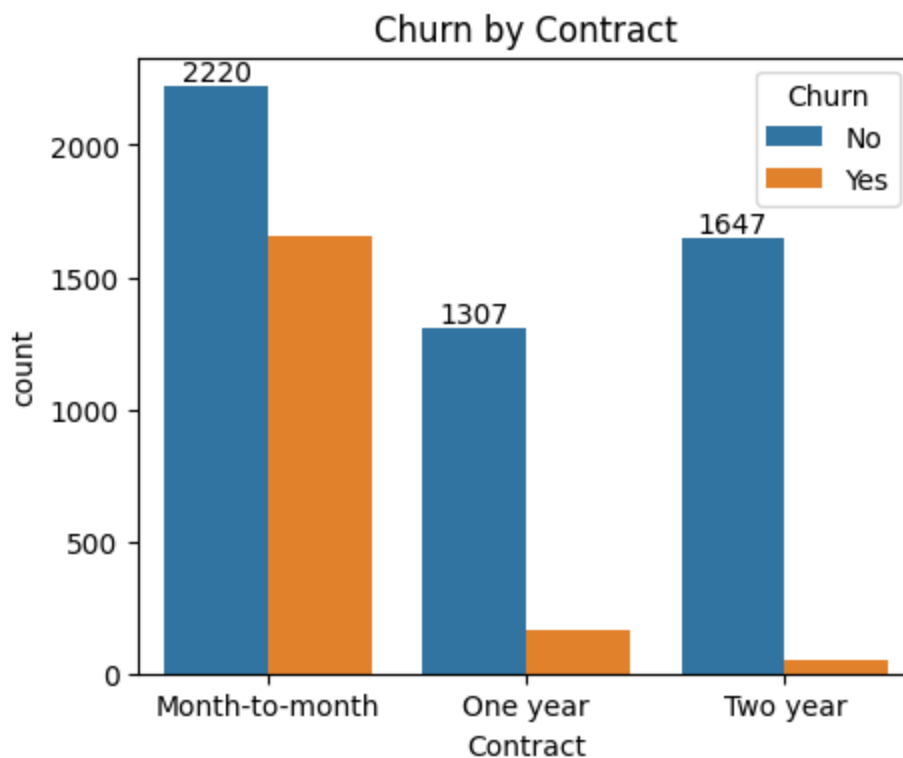
Out[28]:

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

5 rows × 21 columns

In [32]:

```
plt.figure(figsize = (5,4))
ax=sns.countplot(x = "Contract" ,data = df, hue ="Churn")
plt.title("Churn by Contract")
ax.bar_label(ax.containers[0])
plt.show()
```



People who have month to month contract likely to churn then from those who have 1 or 2 year contract

```
In [35]: df.columns.values
```

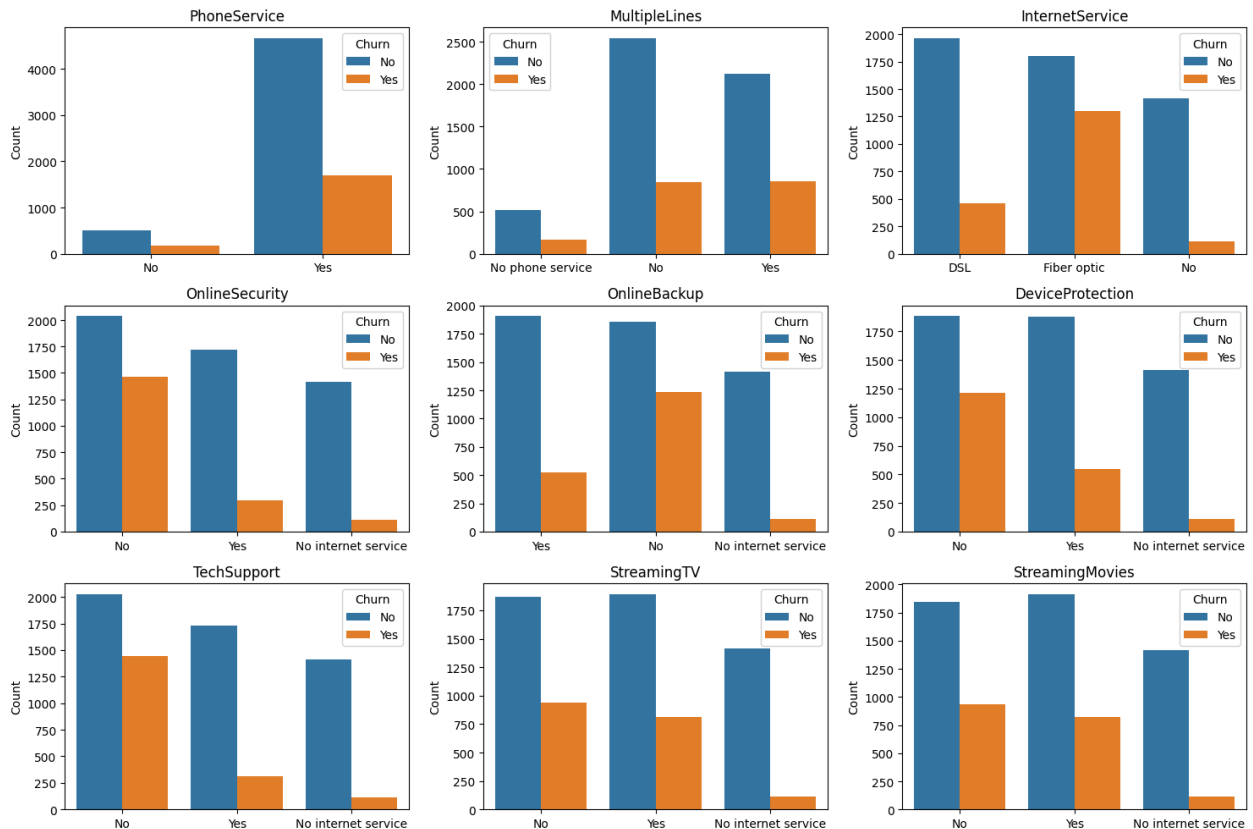
```
Out[35]: array(['customerID', 'gender', 'SeniorCitizen', 'Partner', 'Dependents',  
               'tenure', 'PhoneService', 'MultipleLines', 'InternetService',  
               'OnlineSecurity', 'OnlineBackup', 'DeviceProtection',  
               'TechSupport', 'StreamingTV', 'StreamingMovies', 'Contract',  
               'PaperlessBilling', 'PaymentMethod', 'MonthlyCharges',  
               'TotalCharges', 'Churn'], dtype=object)
```

```
In [41]: import matplotlib.pyplot as plt  
import seaborn as sns  
  
# List of categorical columns  
cols = [  
    'PhoneService', 'MultipleLines', 'InternetService',  
    'OnlineSecurity', 'OnlineBackup', 'DeviceProtection',  
    'TechSupport', 'StreamingTV', 'StreamingMovies'  
]  
  
# Set up the subplot grid (3 rows x 3 columns)  
fig, axes = plt.subplots(3, 3, figsize=(15, 10))  
axes = axes.flatten()
```

```
# Loop through each column and plot countplot
for i, col in enumerate(cols):
    sns.countplot(x=col, data=df, ax=axes[i], hue='Churn')
    axes[i].set_title(col, fontsize=12)
    axes[i].set_xlabel('')
    axes[i].set_ylabel('Count')

# Remove empty subplot if number of plots < total grid cells
for j in range(len(cols), len(axes)):
    fig.delaxes(axes[j])

plt.tight_layout()
plt.show()
```

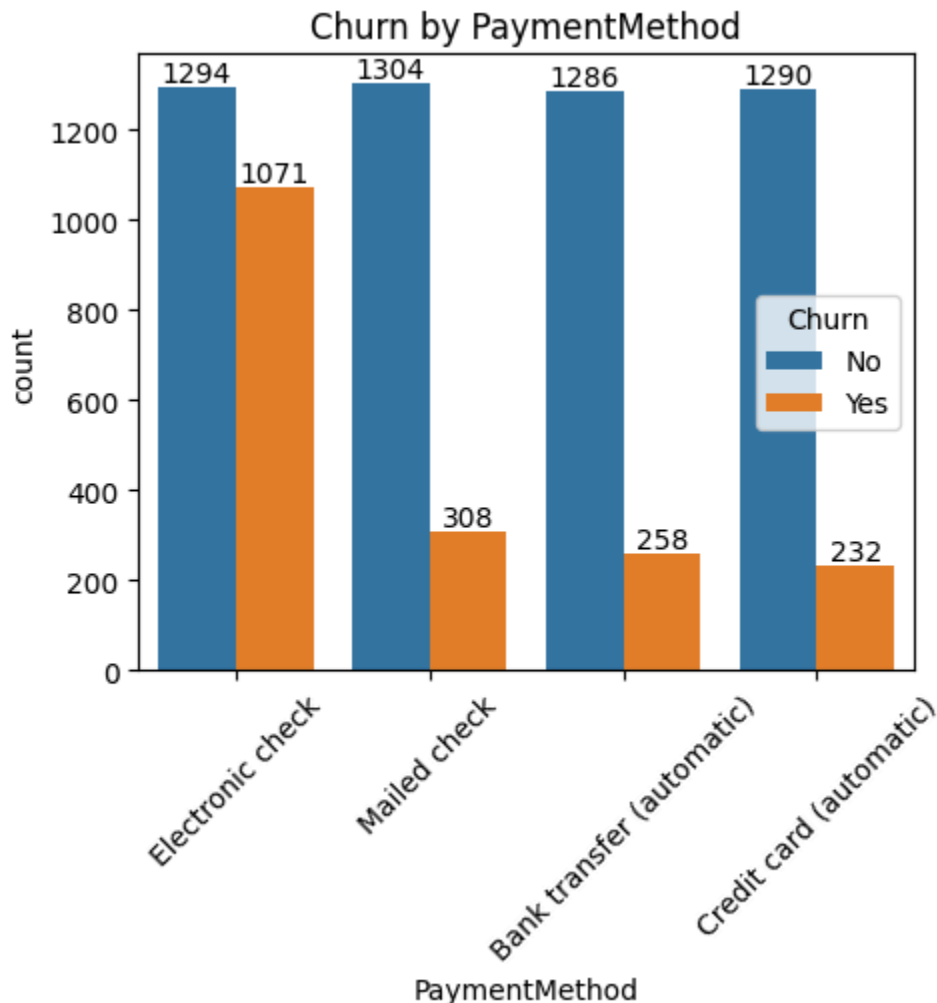


Those using **Fiber optic** internet also have a higher churn rate than DSL users. Overall, customers lacking additional services tend to leave more often, while users with multiple services are generally more loyal.

In [46]: `df.PaymentMethod.values`

```
Out[46]: array(['Electronic check', 'Mailed check', 'Mailed check', ...,  
               'Electronic check', 'Mailed check', 'Bank transfer (automatic)'],  
              shape=(7043,), dtype=object)
```

```
In [66]: plt.figure(figsize = (5,4))  
ax=sns.countplot(x = "PaymentMethod" ,data = df, hue = "Churn")  
plt.title("Churn by PaymentMethod")  
ax.bar_label(ax.containers[0])  
ax.bar_label(ax.containers[1])  
plt.xticks(rotation = 45)  
plt.show()
```



Customer is likely to churn when he is using Electronic check as a payment method

```
In [ ]:
```