

In [4]:

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

In [5]:

```
df = pd.read_csv(r"C:\Users\Om Satyawan Pathak\OneDrive\Desktop\churn analysis\Customer C
hurn.csv")
```

In [6]:

df

Out[6]:

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService | OnlineSe |
|------|------------|--------|---------------|---------|------------|--------|--------------|------------------|-----------------|----------|
| 0 | 7590-VHVEG | Female | 0 | Yes | No | 1 | No | No phone service | DSL | |
| 1 | 5575-GNVDE | Male | 0 | No | No | 34 | Yes | No | DSL | |
| 2 | 3668-QPYBK | Male | 0 | No | No | 2 | Yes | No | DSL | |
| 3 | 7795-CFOCW | Male | 0 | No | No | 45 | No | No phone service | DSL | |
| 4 | 9237-HQITU | Female | 0 | No | No | 2 | Yes | No | Fiber optic | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 7038 | 6840-RESVB | Male | 0 | Yes | Yes | 24 | Yes | Yes | DSL | |
| 7039 | 2234-XADUH | Female | 0 | Yes | Yes | 72 | Yes | Yes | Fiber optic | |
| 7040 | 4801-JZAZL | Female | 0 | Yes | Yes | 11 | No | No phone service | DSL | |
| 7041 | 8361-LTMKD | Male | 1 | Yes | No | 4 | Yes | Yes | Fiber optic | |
| 7042 | 3186-AJIEK | Male | 0 | No | No | 66 | Yes | No | Fiber optic | |

7043 rows x 21 columns



In [7]:

```
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
...
```


| customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService | OnlineSecurity |
|------------|--------|---------------|---------|------------|--------|--------------|---------------|-----------------|----------------|
| 1 | False | False | False | False | False | False | False | False | False |
| 2 | False | False | False | False | False | False | False | False | False |
| 3 | False | False | False | False | False | False | False | False | False |
| 4 | False | False | False | False | False | False | False | False | False |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 7038 | False | False | False | False | False | False | False | False | False |
| 7039 | False | False | False | False | False | False | False | False | False |
| 7040 | False | False | False | False | False | False | False | False | False |
| 7041 | False | False | False | False | False | False | False | False | False |
| 7042 | False | False | False | False | False | False | False | False | False |

7043 rows x 21 columns



In [11]:

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

Out[11]:

```
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 [12]:

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

Out[12]:

0

In [13]:

```
df.describe()
```

Out[13]:

| | 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 |
|---------------|----------|----------------|--------------|-------------|
| SeniorCitizen | tenure | MonthlyCharges | TotalCharges | |
| 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 |

In [14]:

```
df.duplicated()
```

Out[14]:

```
0      False
1      False
2      False
3      False
4      False
...
7038   False
7039   False
7040   False
7041   False
7042   False
Length: 7043, dtype: bool
```

In [15]:

```
df.duplicated().sum()
```

Out[15]:

```
0
```

We will ensure that no duplicate customer-id is there:

In [16]:

```
df['customerID'].duplicated().sum()
```

Out[16]:

```
0
```

In []:

In []:

now we will apply a function to make senior-citizen COLUMN value as yes or no instead of any numerical value:

In [17]:

```
def conv(value):
    if value == 1:
        return "yes"
    else:
        return "no"

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

In [18]:

```
df.head(5) #check for first 5 values
```

Out[18]:

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

5 rows x 21 columns



In []:

In []:

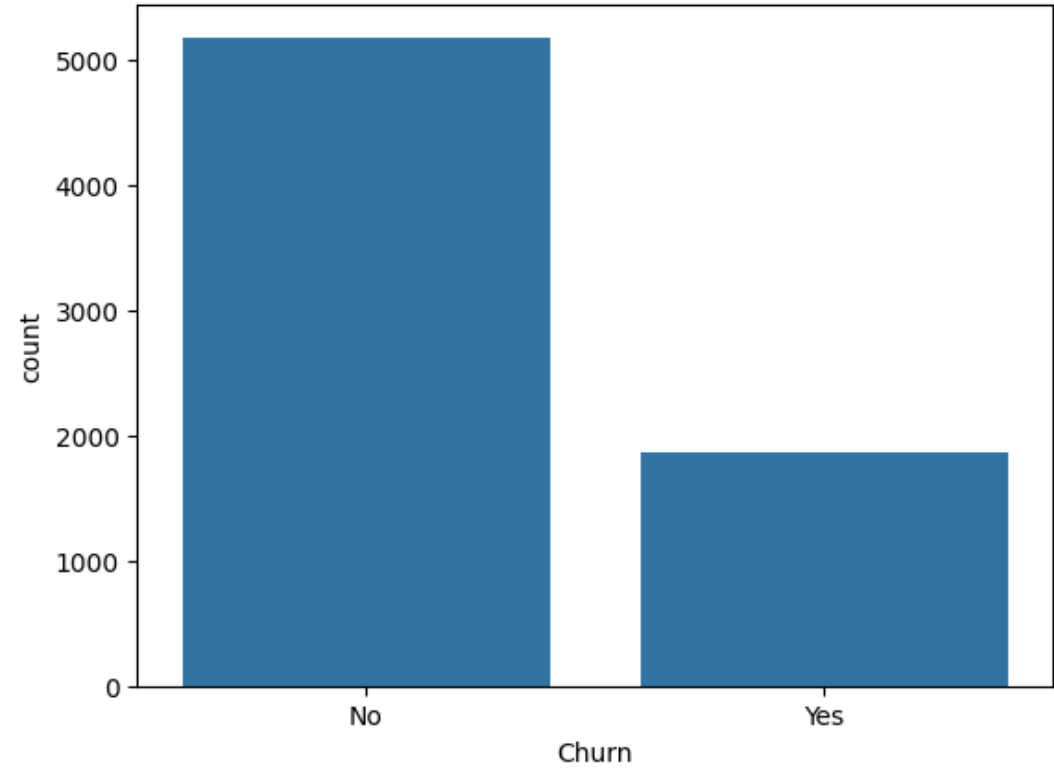
check how many customers have churned-out or not:

In [19]:

```
sns.countplot(x=df["Churn"],data=df)
```

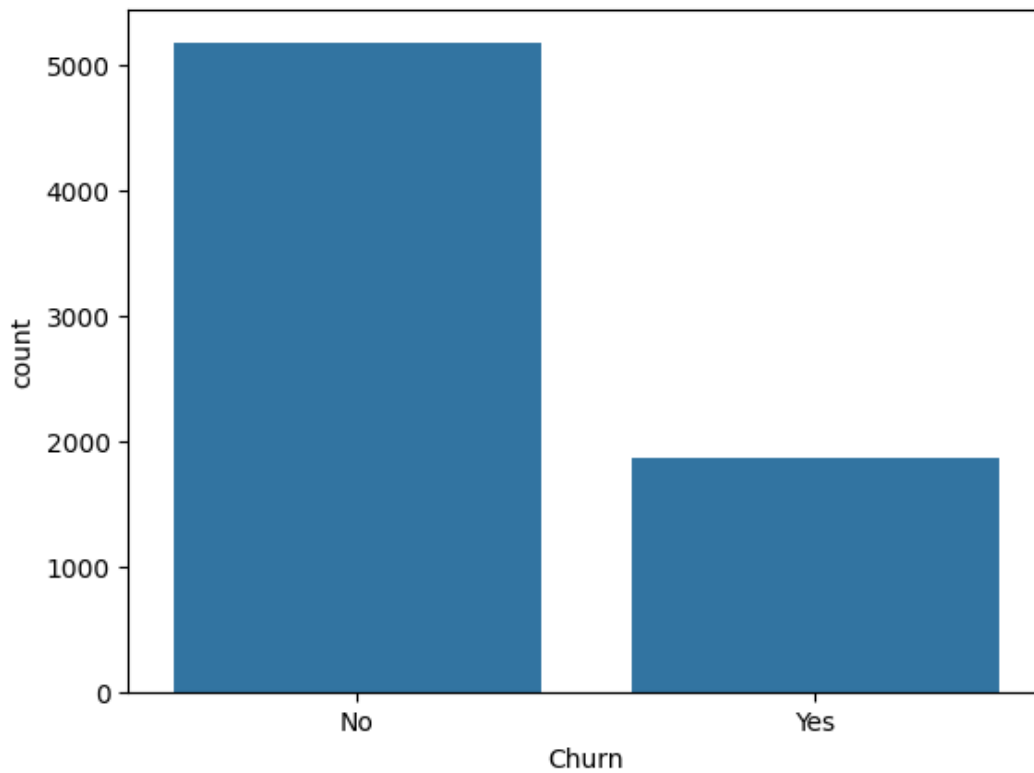
Out[19]:

<Axes: xlabel='Churn', ylabel='count'>



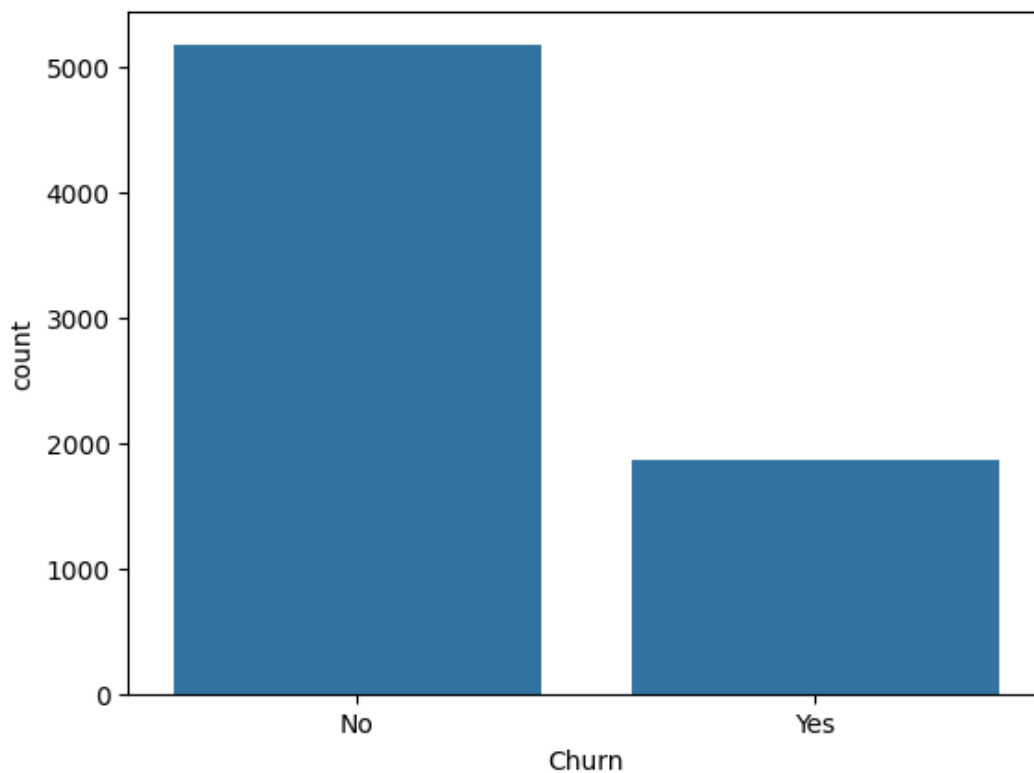
In [20]:

```
sns.countplot(x="Churn", data=df)
m.show()
```



In [21]:

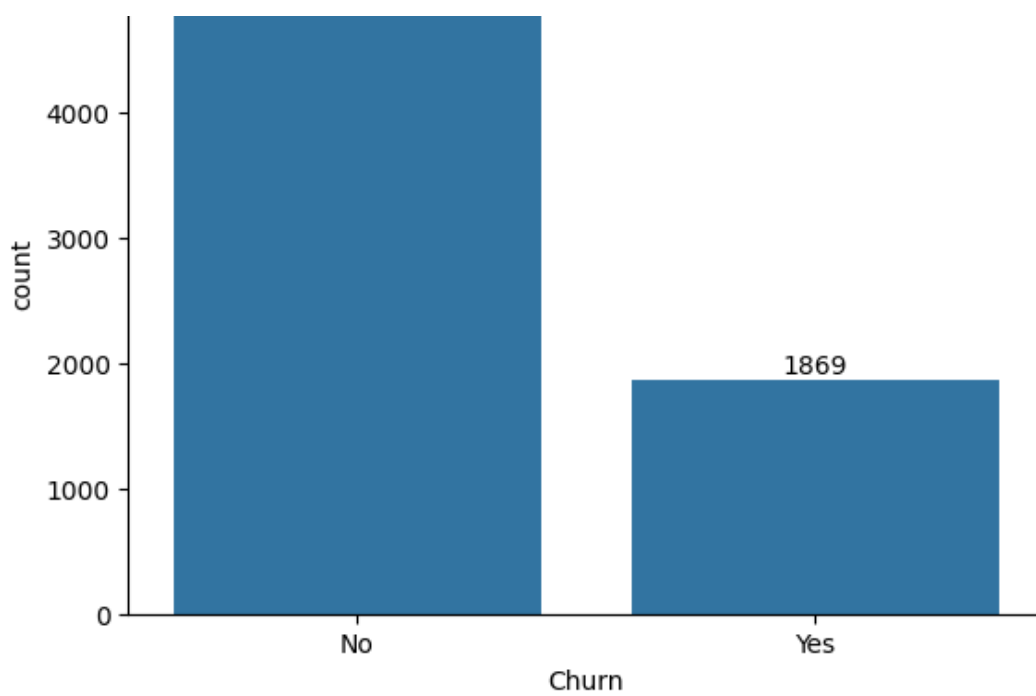
```
ax = sns.countplot(x="Churn", data=df)
m.show()
```



In [22]:

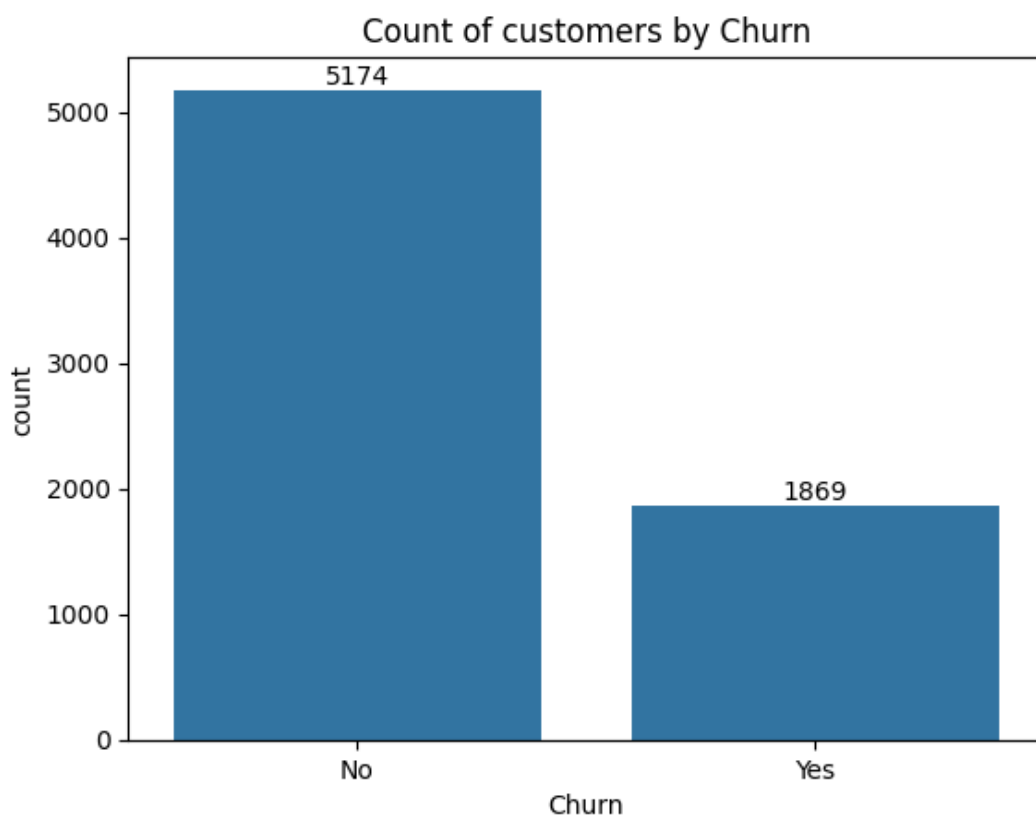
```
ax = sns.countplot(x="Churn", data=df)
ax.bar_label(ax.containers[0]) #check how many customers have churned out or not on top o
f the bars.
m.show()
```





In [23]:

```
ax = sns.countplot(x=df["Churn"], data=df)
ax.bar_label(ax.containers[0]) #check how many customers have churned out or not on top of the bars.
m.title("Count of customers by Churn")
m.show()
```



In [24]:

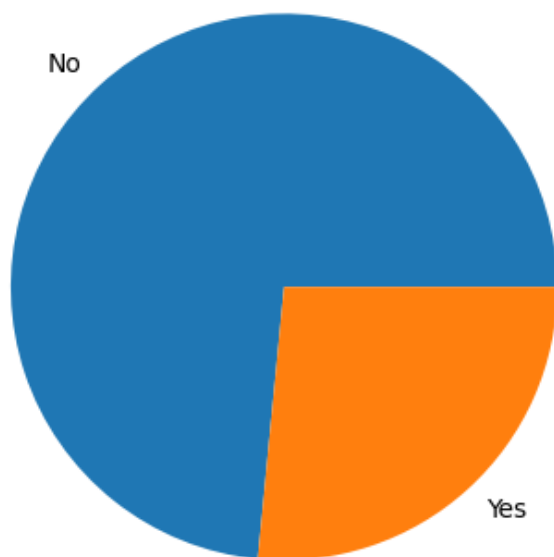
```
gb = df.groupby("Churn").agg({"Churn": "count"})
gb
```

Out[24]:

| Churn | |
|-------|------|
| Churn | |
| No | 5174 |

In [25]:

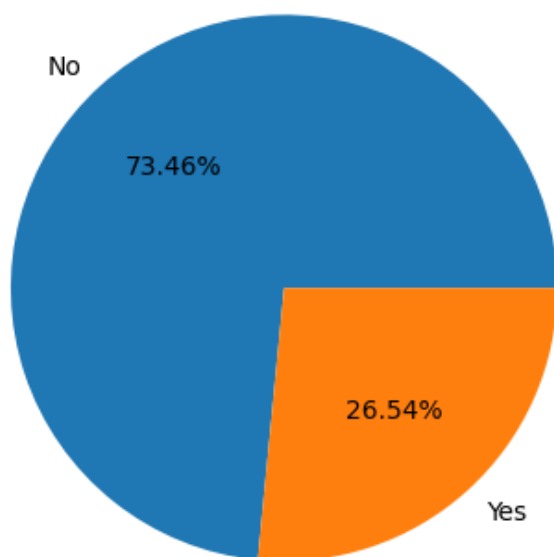
```
gb = df.groupby("Churn").agg({"Churn": "count"})
m.pie(gb["Churn"], labels = gb.index)
m.show()
```



In [26]:

```
gb = df.groupby("Churn").agg({"Churn": "count"})
m.pie(gb["Churn"], labels = gb.index, autopct = "%1.2f%%")
m.title("Percentage of Customers By Churn", fontsize=10)
m.show()
```

Percentage of Customers By Churn



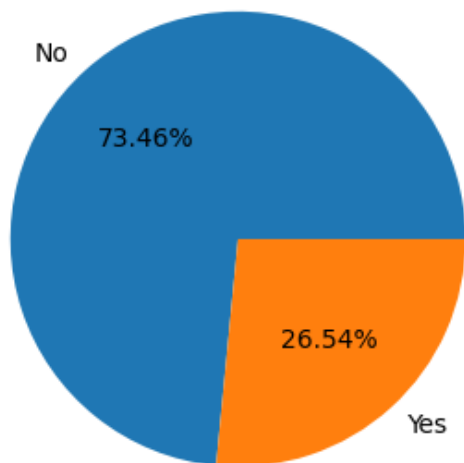
In [27]:

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



```
m.show()
```

Percentage of Customers By Churn

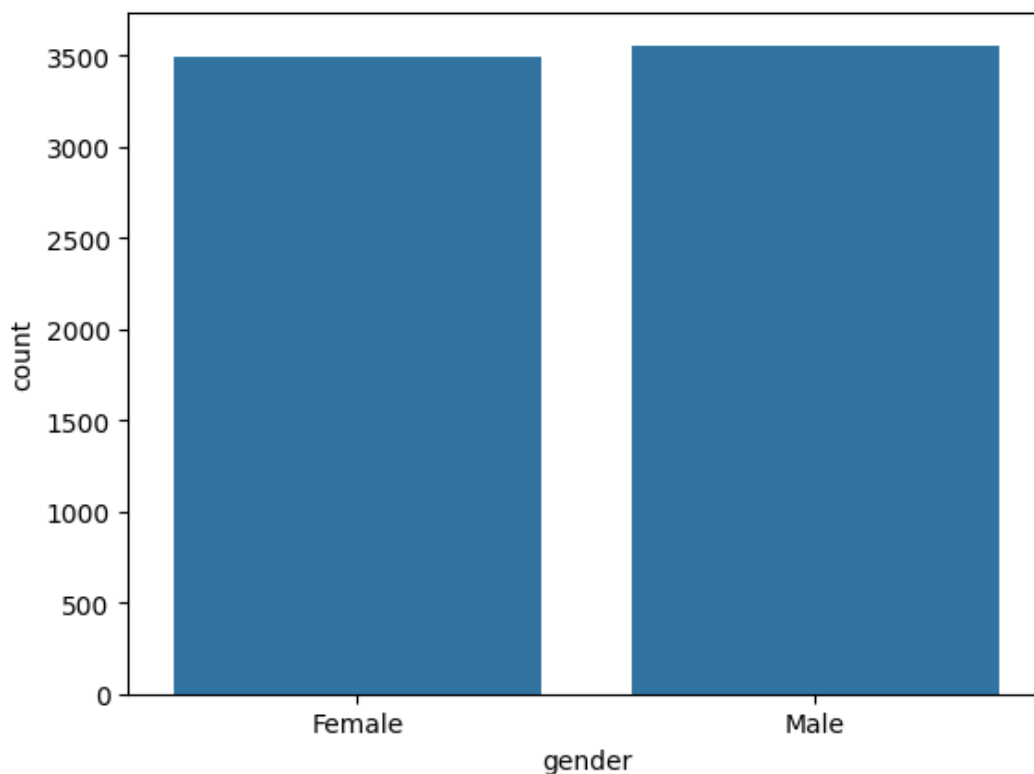


From the given pie chart we can conclude that 26.54% of our customers have churned out.

Now lets explore the reason behind it.

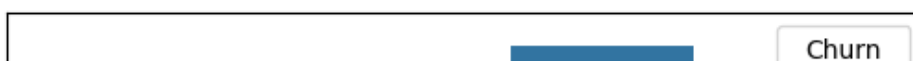
In [28]:

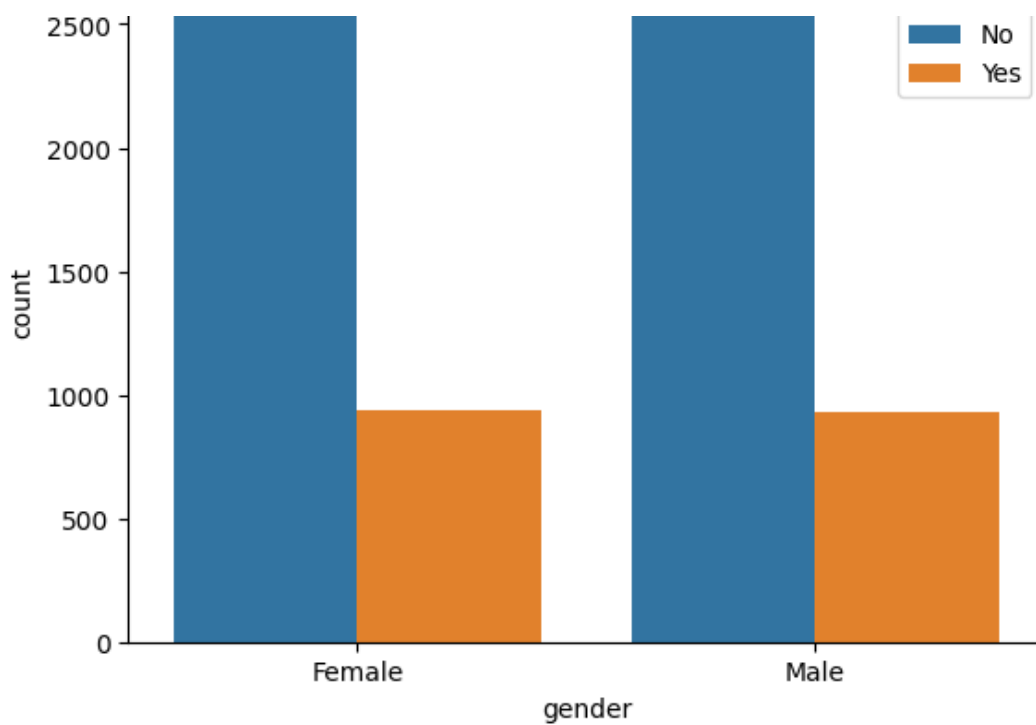
```
sns.countplot(x="gender", data=df)  
m.show()
```



In [29]:

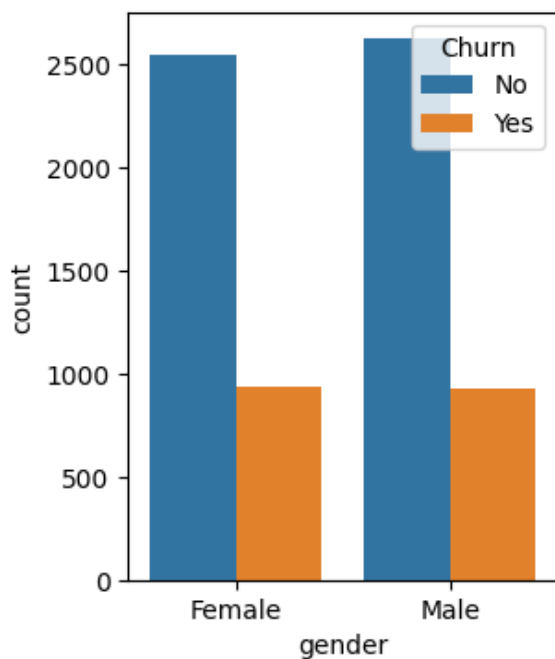
```
sns.countplot(x="gender", data=df, hue="Churn") #check by gender that how many people from  
each gender have churned out  
m.show()
```





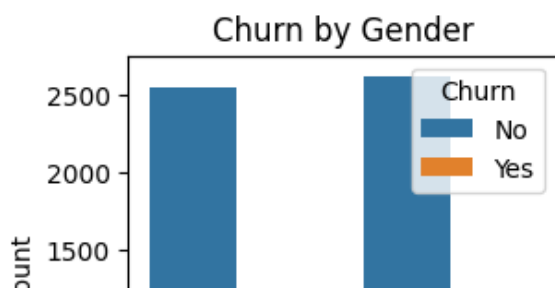
In [30]:

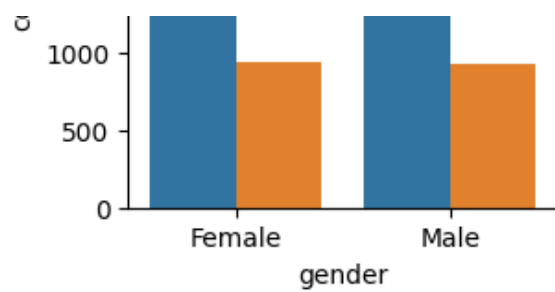
```
m.figure(figsize=(3,4))
sns.countplot(x="gender",data=df,hue="Churn") #check by gender that how many people from
each gender have churned out
m.show()
```



In [31]:

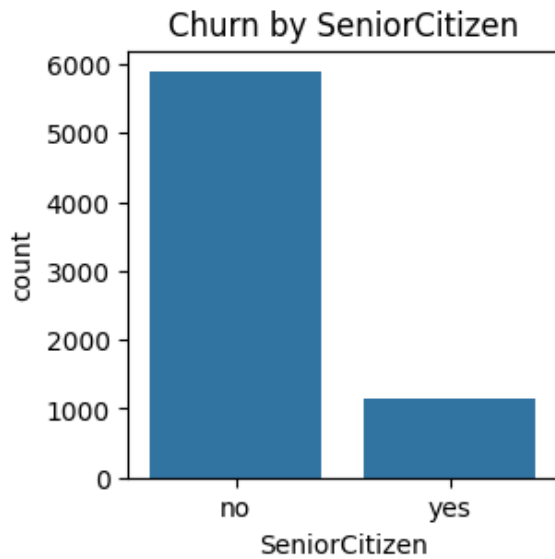
```
m.figure(figsize=(3,3))
sns.countplot(x="gender",data=df,hue="Churn") #check by gender that how many people from
each gender have churned out
m.title("Churn by Gender")
m.show()
```





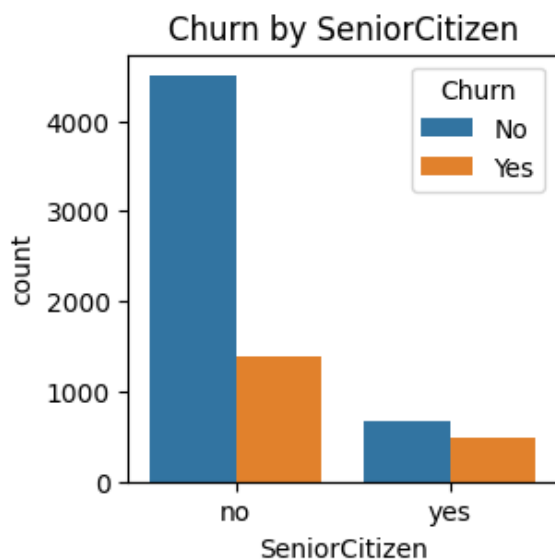
In [32]:

```
m.figure(figsize=(3,3))
sns.countplot(x="SeniorCitizen",data=df) #check by SeniorCitizen that how many people from
m have churned out and how many have not
m.title("Churn by SeniorCitizen")
m.show()
```



In [33]:

```
#how many senior citizens and non-senior citizens have churned (left the service) and how
many stayed:
m.figure(figsize=(3,3))
sns.countplot(x="SeniorCitizen",data=df,hue="Churn")
m.title("Churn by SeniorCitizen")
m.show()
```



In [34]:

```
#THE ABOVE GRAPH IN STACK BAR-CHART FORM:
count_data = df.groupby(['SeniorCitizen', 'Churn']).size().unstack(fill_value=0)
```

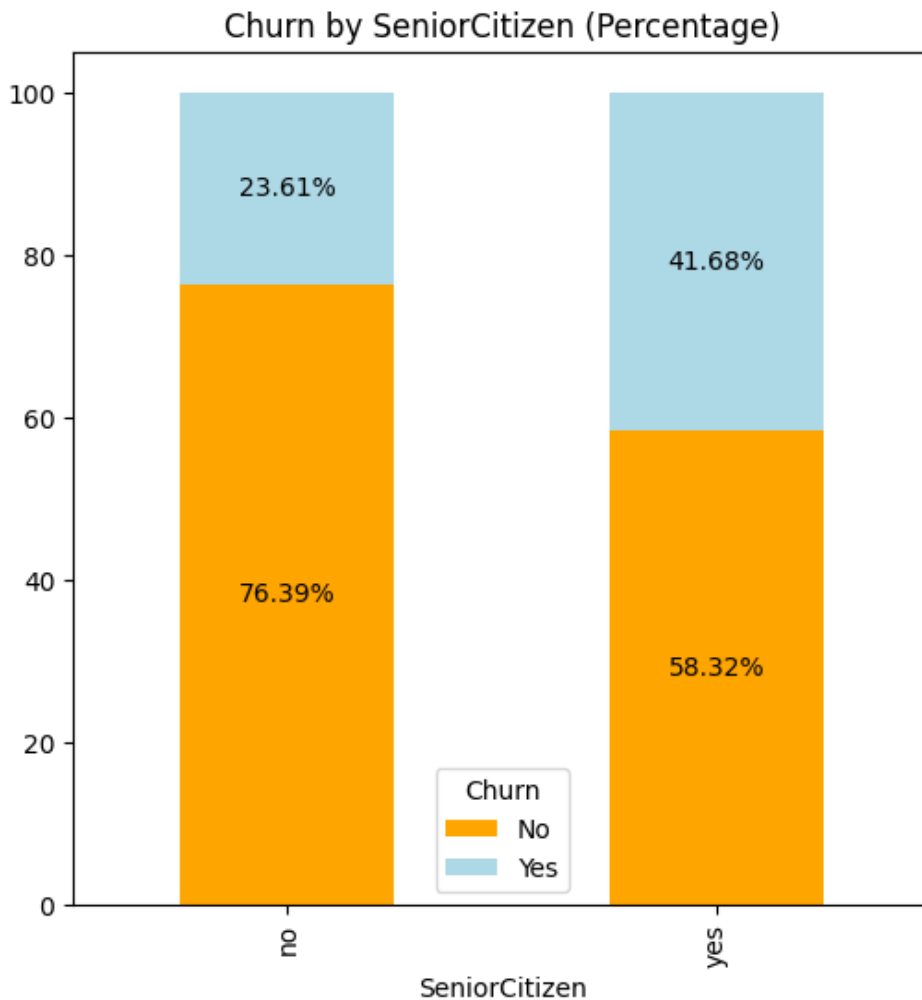
```
percentage_data = count_data.div(count_data.sum(axis=1), axis=0) * 100
```

```
# Plotting the stacked bar chart
```

```
m.figure(figsize=(6,6))
percentage_data.plot(kind='bar', stacked=True, color=['orange', 'lightblue'], ax=m.gca())

# Adding the percentage labels
for p in m.gca().patches:
    height = p.get_height()
    width = p.get_width()
    x, y = p.get_xy() # Get the x and y position of the rectangle
    percentage = round(height, 2) # Get the height (which is the percentage here)
    m.gca().text(x + width / 2, y + height / 2, f'{percentage}%', ha='center', va='center', color='black')

m.title("Churn by SeniorCitizen (Percentage)")
m.show()
```



comparative a gretaed percentage of people in senior-citizen catgeory have churned

In [37]:

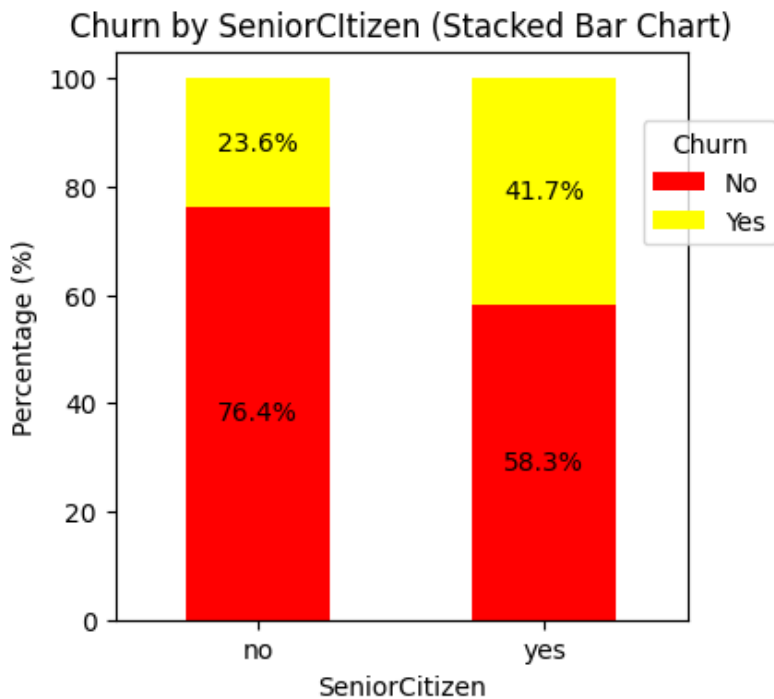
```
total_counts = df.groupby('SeniorCitizen')['Churn'].value_counts(normalize=True).unstack() * 100

fig, ax = m.subplots(figsize=(4, 4))
total_counts.plot(kind='bar', stacked=True, ax=ax, color=['red', 'yellow'])

for p in ax.patches:
    width, height = p.get_width(), p.get_height()
    x, y = p.get_xy()
    ax.text(x + width / 2, y + height / 2, f'{height:.1f}%', ha='center', va='center')
```

```
m.title('Churn by SeniorCitizen (Stacked Bar Chart)')
m.xlabel('SeniorCitizen')
m.ylabel('Percentage (%)')
m.xticks(rotation=0)
m.legend(title='Churn',bbox_to_anchor=(0.9,0.9))

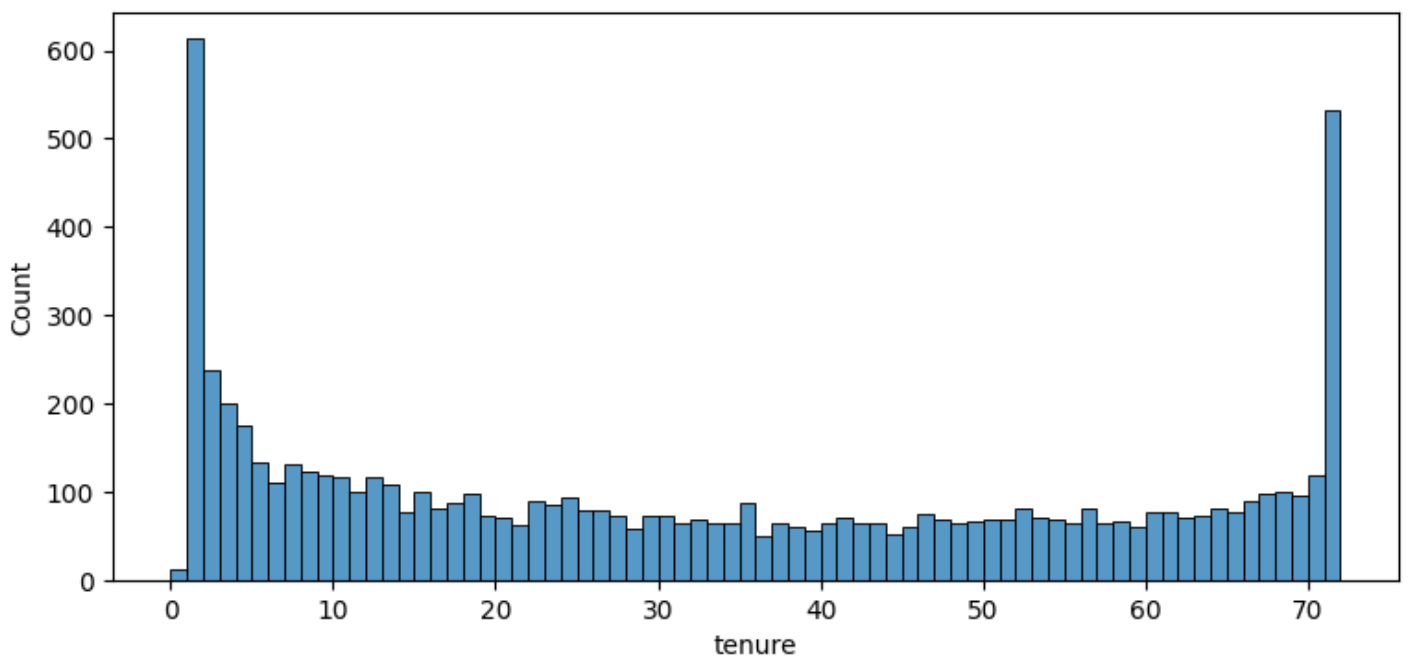
m.show()
```



how many customers have churned out based on tenure:

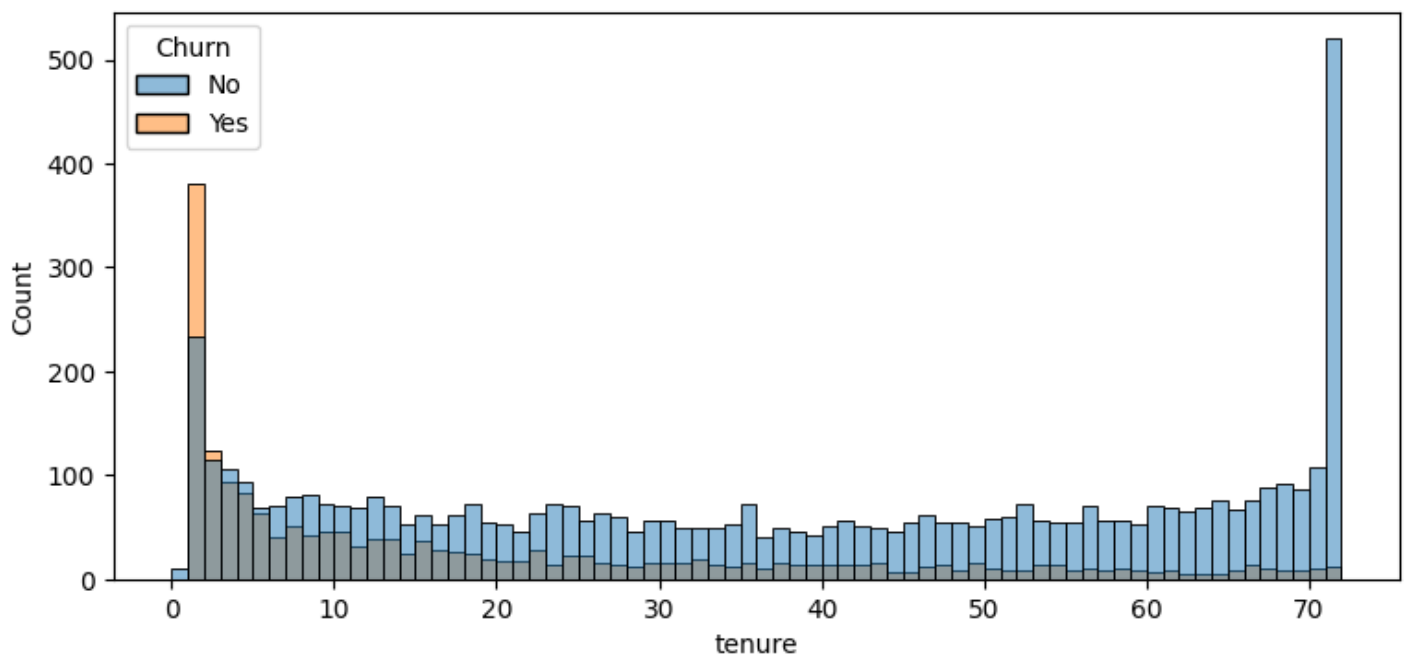
In [47]:

```
m.figure(figsize=(9,4)) #width,height
sns.histplot(x="tenure",data=df,bins=72) #72 bars hence 72 bins
m.show()
```



In [46]:

```
m.figure(figsize=(9,4)) #width,height
sns.histplot(x="tenure",data=df,bins=72, hue="Churn") #72 bars hence 72 bins
m.show()
```

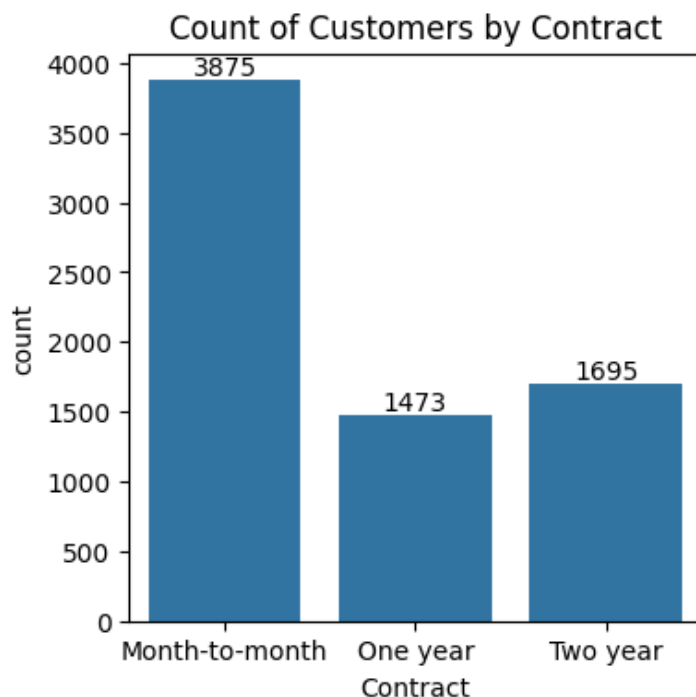


people who have used our services for a longer time have stayed and people who have used our services

for 1 or 2 months have churned

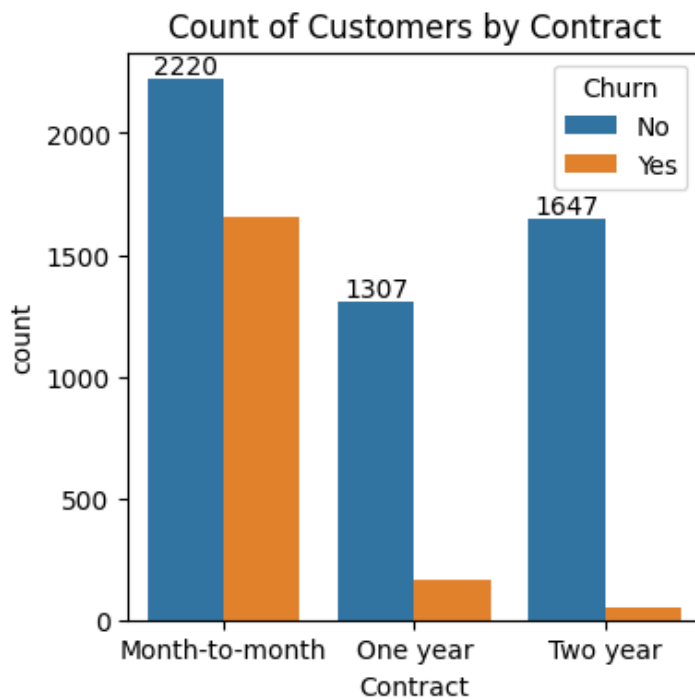
In [51]:

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



In [52]:

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



people who have month-to-month contract are likely to churn out then from those who have 1 or 2 years of contract.

In [53]:

```
df.columns.values
```

Out[53]:

```
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 [58]:

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

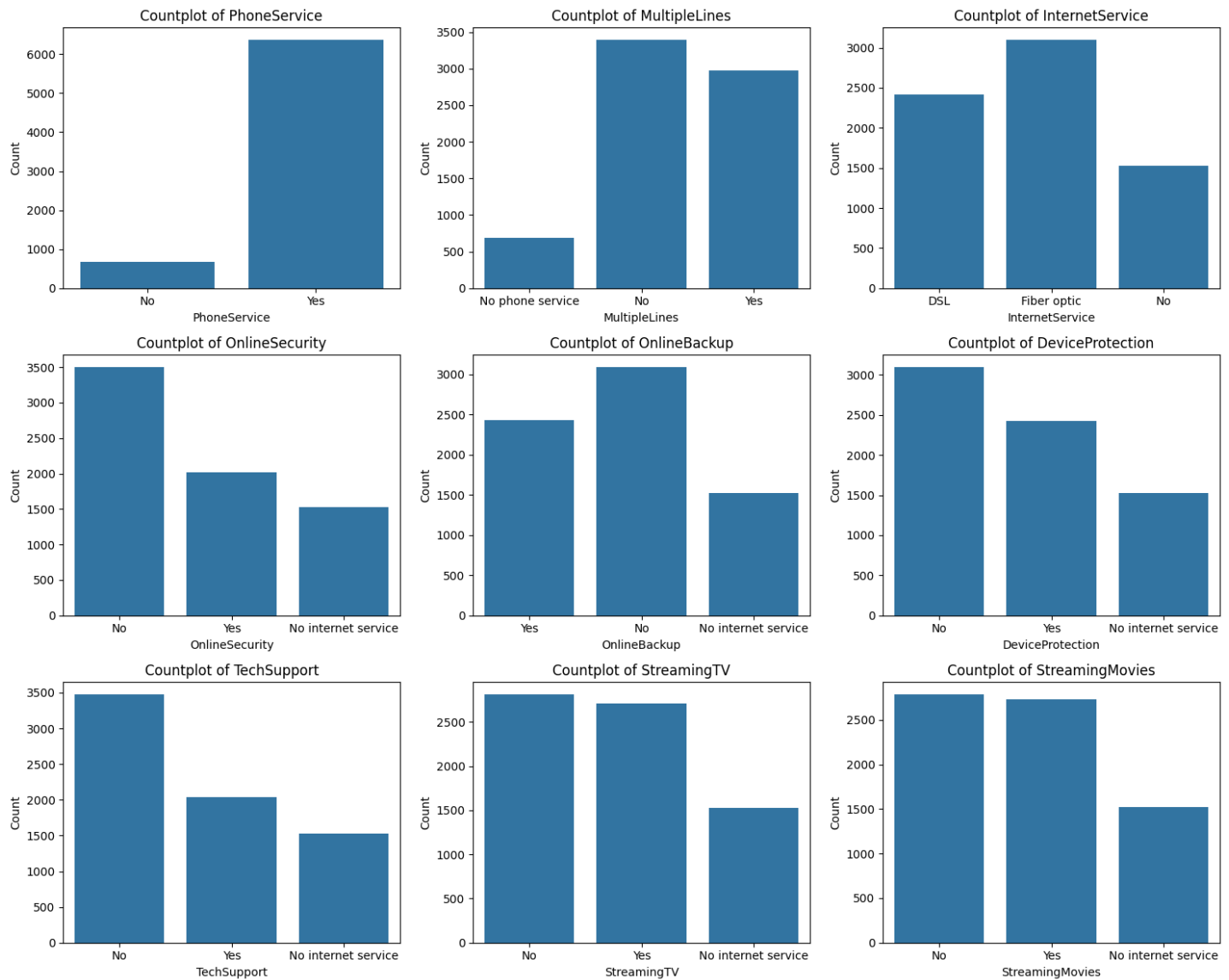
# Define number of rows and columns for subplots
n_cols = 3
n_rows = (len(cols) + n_cols - 1) // n_cols

# Set figure size dynamically
fig, axes = m.subplots(n_rows, n_cols, figsize=(15,n_rows*4))
axes = axes.flatten() # Flatten to make indexing easier

# Create countplots
for i, col in enumerate(cols):
    sns.countplot(x=col, data=df, ax=axes[i])
    axes[i].set_title(f'Countplot of {col}')
    axes[i].set_xlabel(col)
    axes[i].set_ylabel('Count')

# Remove any extra empty subplots (in case of mismatch)
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])
```

```
m.tight_layout()
m.show()
```



In [59]:

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

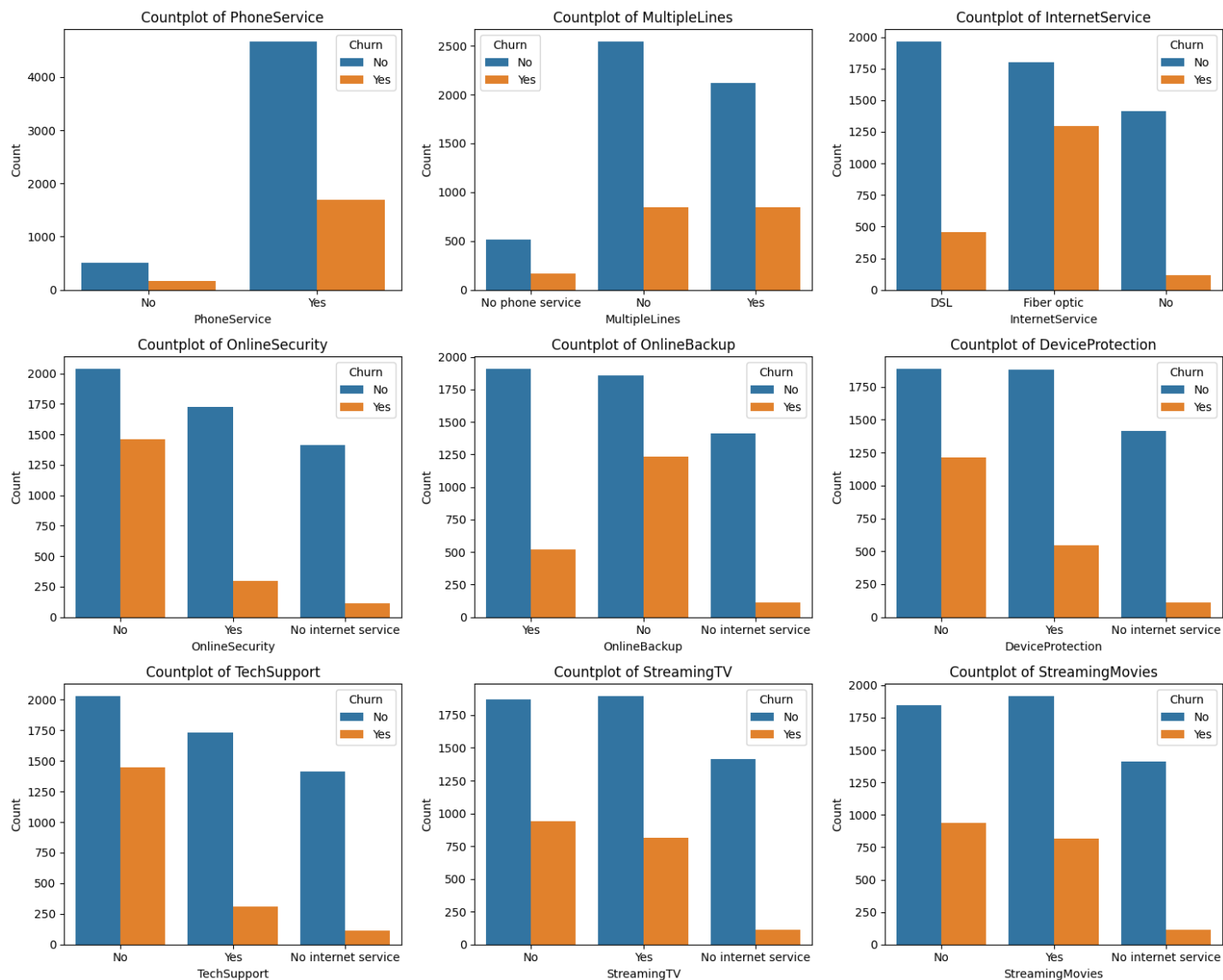
# Define number of rows and columns for subplots
n_cols = 3
n_rows = (len(cols) + n_cols - 1) // n_cols

# Set figure size dynamically
fig, axes = m.subplots(n_rows, n_cols, figsize=(15,n_rows*4))
axes = axes.flatten() # Flatten to make indexing easier

# Create countplots
for i, col in enumerate(cols):
    sns.countplot(x=col, data=df, ax=axes[i], hue=df["Churn"])
    axes[i].set_title(f'Countplot of {col}')
    axes[i].set_xlabel(col)
    axes[i].set_ylabel('Count')

# Remove any extra empty subplots (in case of mismatch)
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])

m.tight_layout()
m.show()
```

In [60]:

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

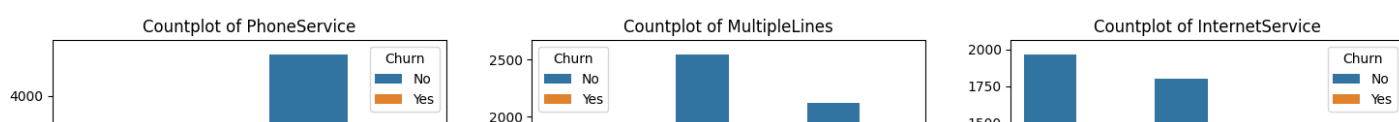
# Define number of rows and columns for subplots
n_cols = 3
n_rows = (len(cols) + n_cols - 1) // n_cols

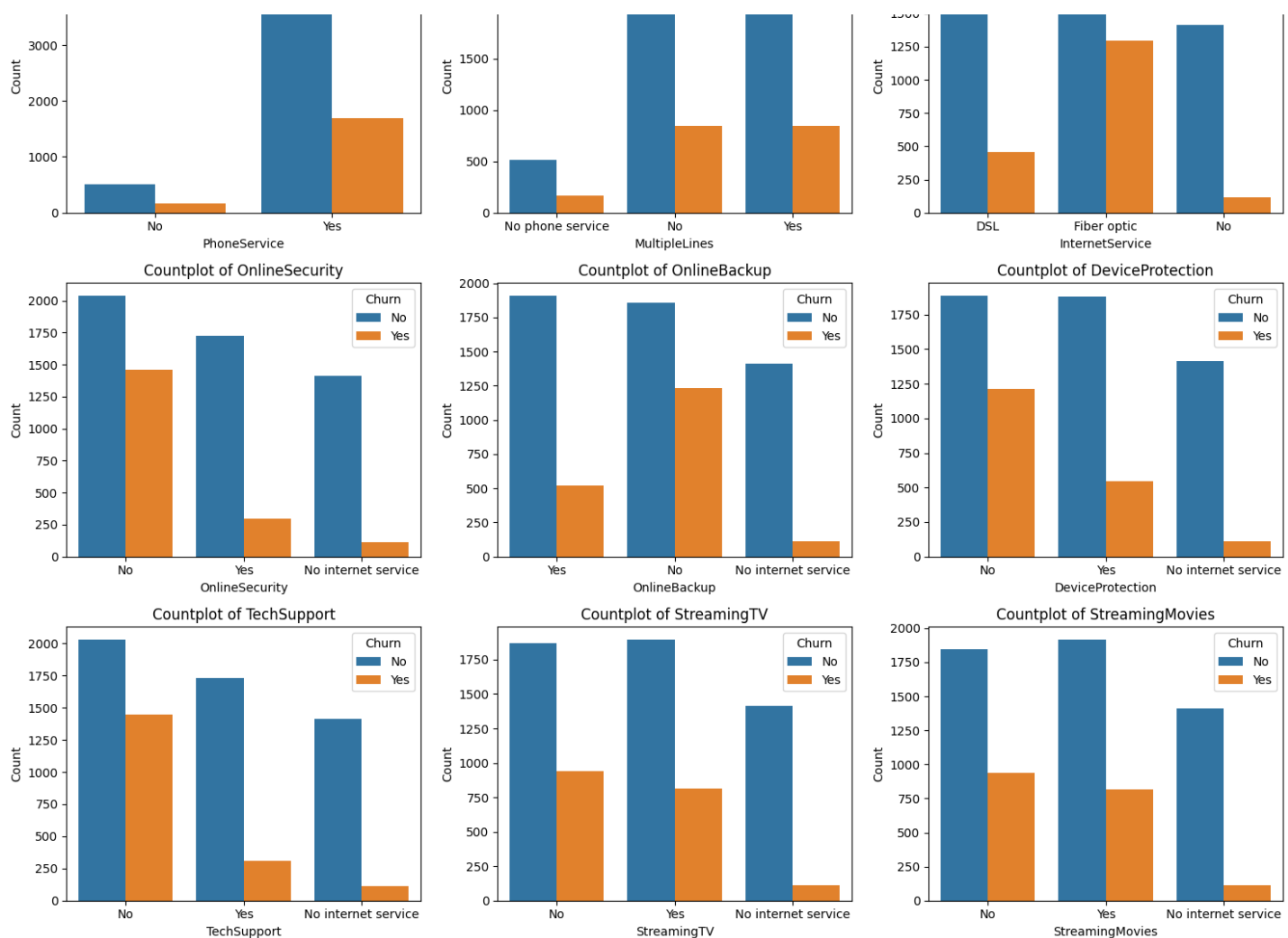
# Set figure size dynamically
fig, axes = m.subplots(n_rows, n_cols, figsize=(15,n_rows*4))
axes = axes.flatten() # Flatten to make indexing easier

# Create countplots
for i, col in enumerate(cols):
    sns.countplot(x=col, data=df, ax=axes[i], hue="Churn")
    axes[i].set_title(f'Countplot of {col}')
    axes[i].set_xlabel(col)
    axes[i].set_ylabel('Count')

# Remove any extra empty subplots (in case of mismatch)
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])

m.tight_layout()
m.show()
```

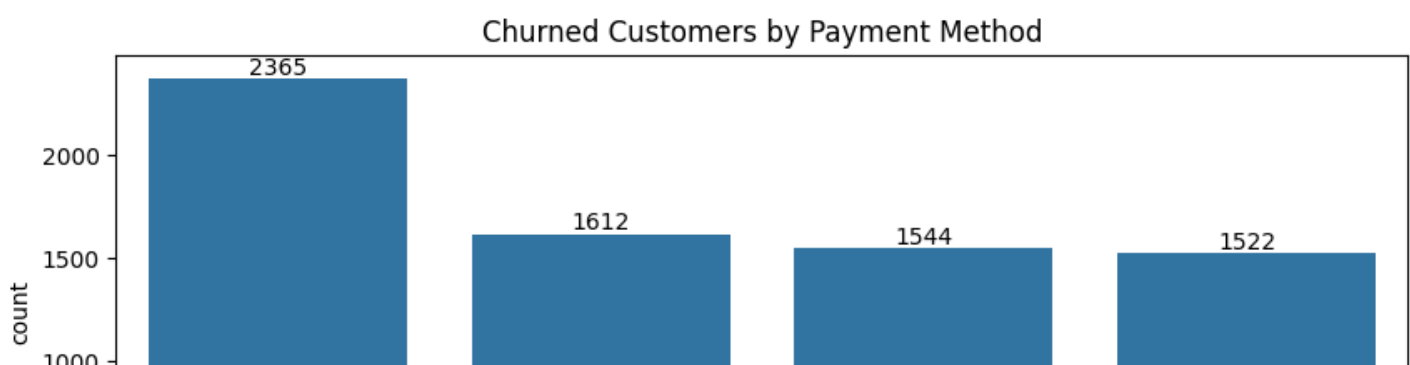


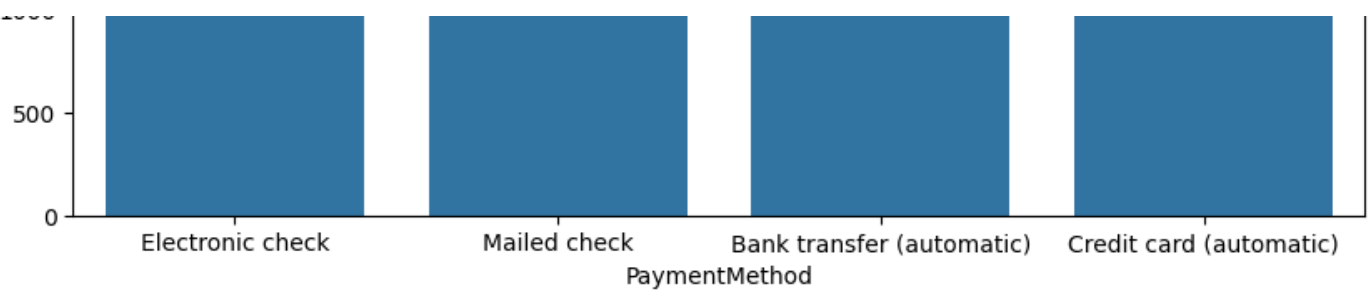


Customers without internet services or related add-ons like online security, backup, and tech support tend to churn less, indicating that minimal service users are more stable. Those using fiber optic internet have significantly higher churn compared to DSL users. Streaming services such as StreamingTV and StreamingMovies show a more balanced churn rate, suggesting they have less impact on customer retention. Overall, customers subscribed to multiple technical services appear more likely to churn, possibly due to higher costs or dissatisfaction.

In [66]:

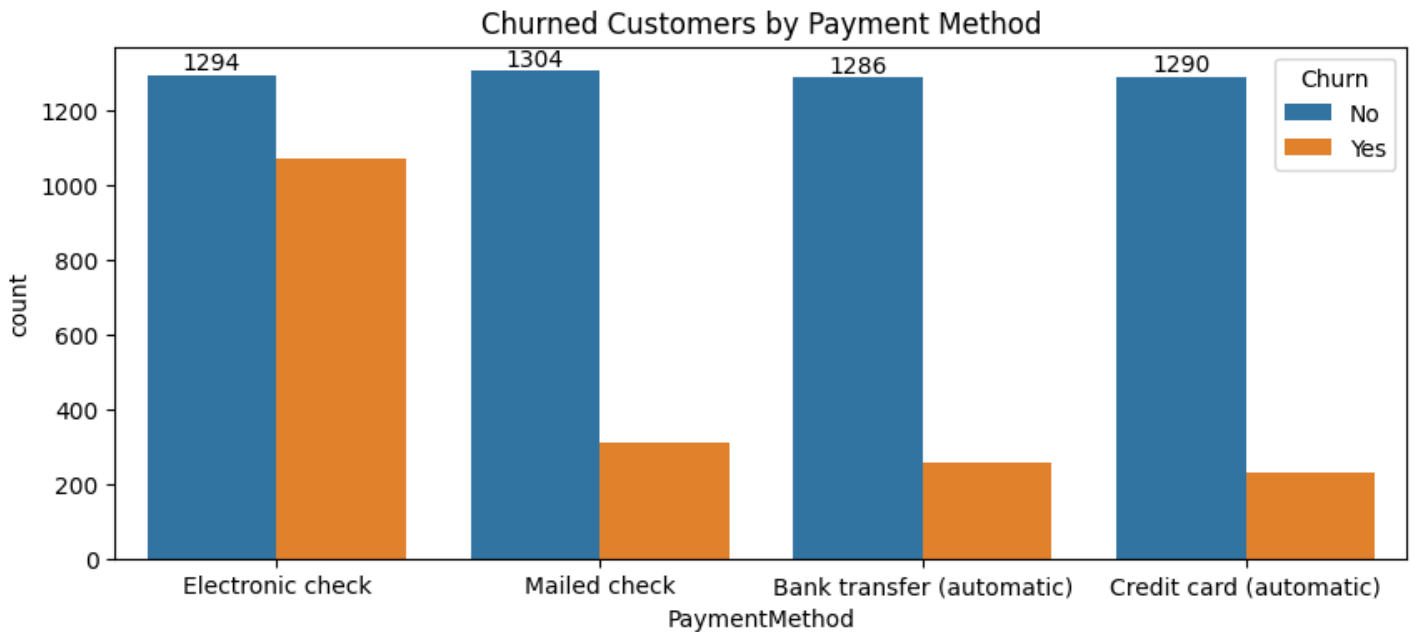
```
m.figure(figsize=(10,4)) #width,height
ax=sns.countplot(x="PaymentMethod",data=df)
ax.bar_label(ax.containers[0])
m.title("Churned Customers by Payment Method")
m.show()
```





In [67]:

```
m.figure(figsize=(10,4)) #width,height
ax=sns.countplot(x="PaymentMethod",data=df,hue="Churn")
ax.bar_label(ax.containers[0])
m.title("Churned Customers by Payment Method")
m.show()
```



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

In []:

In []:

In []:

In []:

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In []: