

# churn-analysis

November 25, 2024

```
[6]: 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()
```

```
[6]: customerID  gender  SeniorCitizen  Partner  Dependents  tenure  PhoneService  \
0  7590-VHVEG  Female           0      Yes           No         1           No
1  5575-GNVDE   Male           0      No            No        34           Yes
2  3668-QPYBK   Male           0      No            No         2           Yes
3  7795-CFOCW   Male           0      No            No        45           No
4  9237-HQITU   Female          0      No            No         2           Yes
```

```
MultipleLines  InternetService  OnlineSecurity  ...  DeviceProtection  \
0  No phone service           DSL              No  ...           No
1              No           DSL              Yes  ...           Yes
2              No           DSL              Yes  ...           No
3  No phone service           DSL              Yes  ...           Yes
4              No  Fiber optic              No  ...           No
```

```
TechSupport  StreamingTV  StreamingMovies  Contract  PaperlessBilling  \
0          No           No              No  Month-to-month           Yes
1          No           No              No    One year           No
2          No           No              No  Month-to-month           Yes
3          Yes          No              No    One year           No
4          No           No              No  Month-to-month           Yes
```

```
PaymentMethod  MonthlyCharges  TotalCharges  Churn
0  Electronic check           29.85           29.85   No
1    Mailed check           56.95          1889.5   No
2    Mailed check           53.85           108.15  Yes
3  Bank transfer (automatic)    42.30          1840.75  No
4    Electronic check           70.70           151.65  Yes
```

[5 rows x 21 columns]

```
[12]: df["TotalCharges"] = df["TotalCharges"].replace(" ", "0 ")
df["TotalCharges"] = df["TotalCharges"].astype("float")
```

## 1 Replacing the blanks with 0 as tenure is 0 no total charges are recorded

```
[13]: 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
```

```
[16]: df.isnull().sum().sum()
```

```
[16]: np.int64(0)
```

```
[17]: df.describe()
```

```
[17]:
```

	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

```
[21]: df["customerID"].duplicated().sum()
```

```
[21]: np.int64(0)
```

2 we have converted 0 and 1 value of senior citizen into yes/no

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

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

```
[24]: df.head(10)
```

```
[24]:  customerID  gender SeniorCitizen Partner Dependents  tenure PhoneService \
0  7590-VHVEG  Female           no      Yes           No         1           No
1  5575-GNVDE   Male           no      No            No        34           Yes
2  3668-QPYBK   Male           no      No            No         2           Yes
3  7795-CFOCW   Male           no      No            No        45           No
4  9237-HQITU  Female           no      No            No         2           Yes
5  9305-CDSKC  Female           no      No            No         8           Yes
6  1452-KIOVK   Male           no      No            Yes        22           Yes
7  6713-OKOMC  Female           no      No            No        10           No
8  7892-POOKP  Female           no      Yes            No        28           Yes
9  6388-TABGU   Male           no      No            Yes        62           Yes
```

	MultipleLines	InternetService	OnlineSecurity	...	DeviceProtection	\
0	No phone service	DSL	No	...	No	
1	No	DSL	Yes	...	Yes	
2	No	DSL	Yes	...	No	
3	No phone service	DSL	Yes	...	Yes	
4	No	Fiber optic	No	...	No	
5	Yes	Fiber optic	No	...	Yes	
6	Yes	Fiber optic	No	...	No	
7	No phone service	DSL	Yes	...	No	
8	Yes	Fiber optic	No	...	Yes	
9	No	DSL	Yes	...	No	

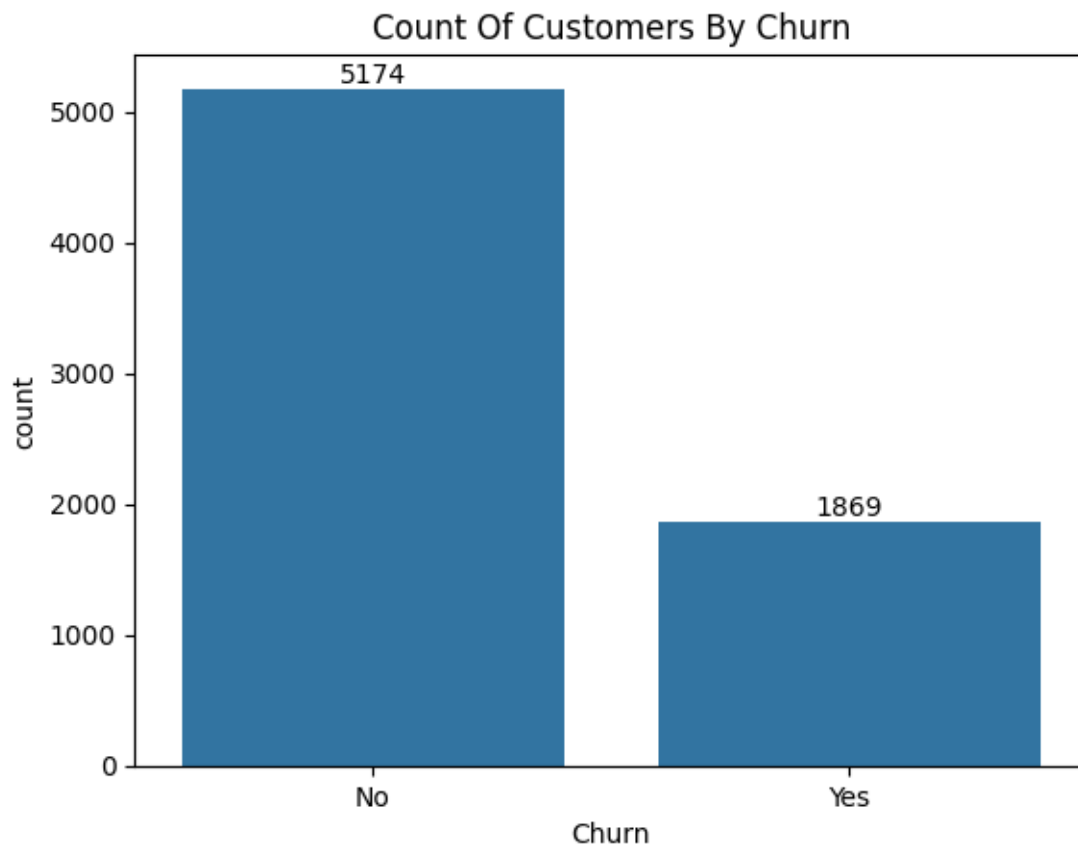
	TechSupport	StreamingTV	StreamingMovies	Contract	PaperlessBilling	\
0	No	No	No	Month-to-month	Yes	
1	No	No	No	One year	No	
2	No	No	No	Month-to-month	Yes	
3	Yes	No	No	One year	No	
4	No	No	No	Month-to-month	Yes	
5	No	Yes	Yes	Month-to-month	Yes	
6	No	Yes	No	Month-to-month	Yes	
7	No	No	No	Month-to-month	No	
8	Yes	Yes	Yes	Month-to-month	Yes	
9	No	No	No	One year	No	

	PaymentMethod	MonthlyCharges	TotalCharges	Churn
0	Electronic check	29.85	29.85	No
1	Mailed check	56.95	1889.50	No
2	Mailed check	53.85	108.15	Yes
3	Bank transfer (automatic)	42.30	1840.75	No
4	Electronic check	70.70	151.65	Yes
5	Electronic check	99.65	820.50	Yes
6	Credit card (automatic)	89.10	1949.40	No
7	Mailed check	29.75	301.90	No
8	Electronic check	104.80	3046.05	Yes
9	Bank transfer (automatic)	56.15	3487.95	No

[10 rows x 21 columns]

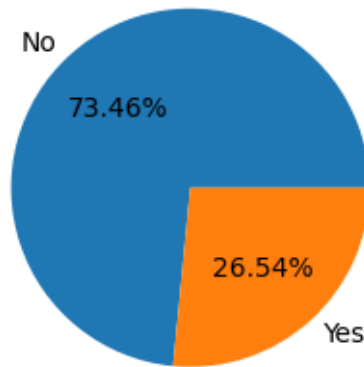
```
[28]: ax = sns.countplot(x= 'Churn' , data = df )

ax.bar_label(ax.containers[0])
plt.title("Count Of Customers By Churn")
plt.show()
```



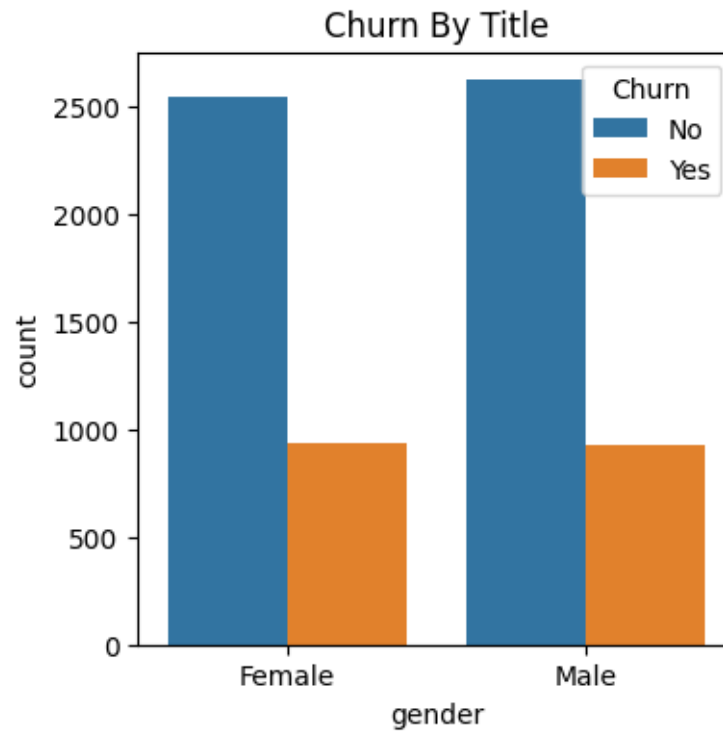
```
[30]: plt.figure(figsize = (3,4))
gb = df.groupby("Churn").agg({'Churn': "count"})
plt.pie(gb["Churn"] , labels = gb.index , autopct = "%1.2f%%")
plt.title("Percentage Of Churned Customer")
plt.show()
```

Percentage Of Churned Customer

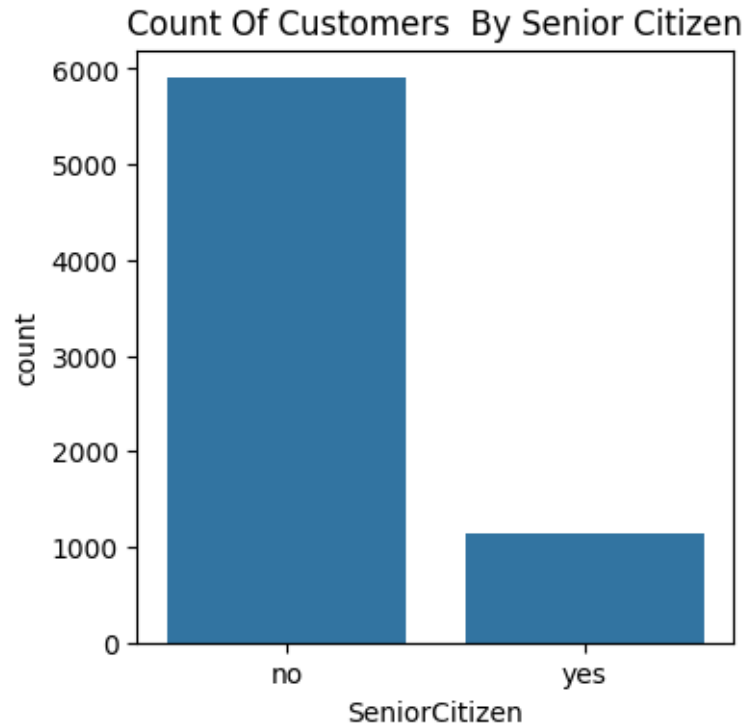


3 from the given pie chart we can conclude that 26.54 % customers have churned out . Now lets figure out the reason behind it

```
[36]: plt.figure(figsize = (4,4))
sns.countplot(x = "gender" , data = df , hue = "Churn")
plt.title("Churn By Title")
plt.show()
```



```
[52]: plt.figure(figsize = (4,4))
sns.countplot(x = "SeniorCitizen" , data = df )
ax.bar_label(ax.containers[0])
plt.title("Count Of Customers By Senior Citizen")
plt.show()
```



```
[43]: grouped = df.groupby(['SeniorCitizen', 'Churn']).size().
      ↪reset_index(name='Count')

# Calculate the percentage contribution within each SeniorCitizen group
grouped['Percentage'] = grouped.groupby('SeniorCitizen')['Count'].
      ↪transform(lambda x: x / x.sum() * 100)

# Pivot the data to prepare for stacked bar plotting
pivot_data = grouped.pivot(index='SeniorCitizen', columns='Churn',
      ↪values='Percentage').fillna(0)

# Step 2: Plot the stacked bar chart
fig, ax = plt.subplots(figsize=(4, 4))
bottoms = [0] * len(pivot_data)

for churn_value in pivot_data.columns:
    ax.bar(
        pivot_data.index, # x-axis labels (SeniorCitizen categories)
        pivot_data[churn_value], # bar heights (percentages)
        bottom=bottoms, # bottom of the bar for stacking
        label=churn_value
    )
    # Update the bottom positions for stacking
```



```

    bottoms = [i + j for i, j in zip(bottoms, pivot_data[churn_value])]

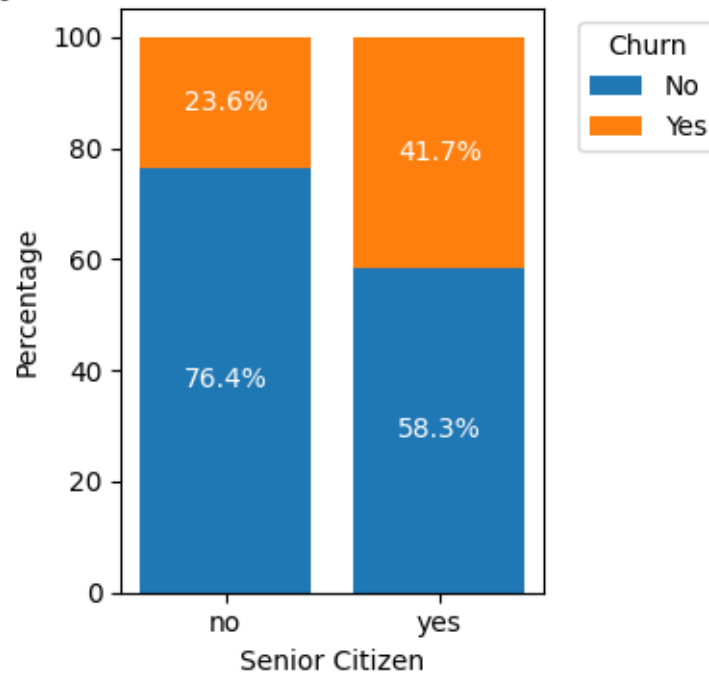
# Step 3: Add percentage labels
for i, senior in enumerate(pivot_data.index):
    cumulative_bottom = 0
    for churn_value in pivot_data.columns:
        value = pivot_data.loc[senior, churn_value]
        if value > 0: # Only annotate non-zero segments
            ax.text(
                i,
                cumulative_bottom + value / 2,
                f'{value:.1f}%',
                ha='center',
                va='center',
                color='white' if value > 10 else 'black', # Adjust text color
                ↪for readability
                fontsize=10
            )
            cumulative_bottom += value

# Step 4: Customize the plot
ax.set_title("Churn by Senior Citizen (Stacked Bar Chart with %)", fontsize=14)
ax.set_xlabel("Senior Citizen")
ax.set_ylabel("Percentage")
ax.legend(title='Churn', bbox_to_anchor=(1.05, 1), loc='upper left')

plt.tight_layout()
plt.show()

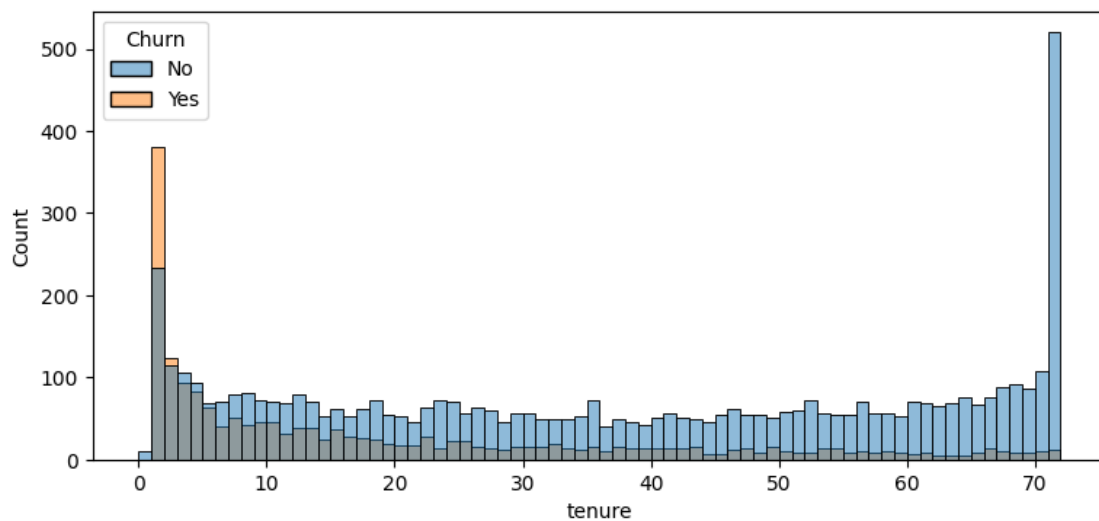
```

Churn by Senior Citizen (Stacked Bar Chart with %)



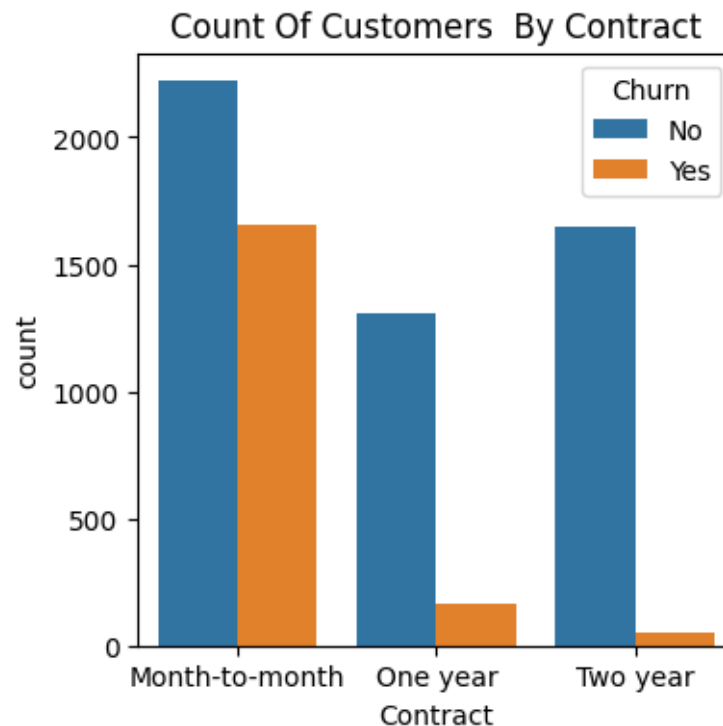
#Comparative a greater percentage of people in senior citizen category have churned

```
[48]: plt.figure(figsize = (9 , 4))
sns.histplot(x = "tenure" , data = df, bins = 72, hue = "Churn")
plt.show()
```



#People who have used our services for a long time have stayed and people who have used our services for 1 -2 months have churned out

```
[53]: plt.figure(figsize = (4,4))
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 month to month contract are likrly to churn then from those who have 1-2 years of contract

```
[54]: df.columns.values
```

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

```
[63]: # Columns to plot (categorical columns)
columns_to_plot = ['PhoneService',
        'MultipleLines', 'InternetService', 'OnlineSecurity', 'OnlineBackup',
```

```

    'DeviceProtection', 'TechSupport', 'StreamingTV', 'StreamingMovies']

# Define the number of rows and columns for subplots
n_cols = 3 # Number of columns in the subplot grid
n_rows = -(-len(columns_to_plot) // n_cols) # Calculate rows needed (ceil_
↳division)

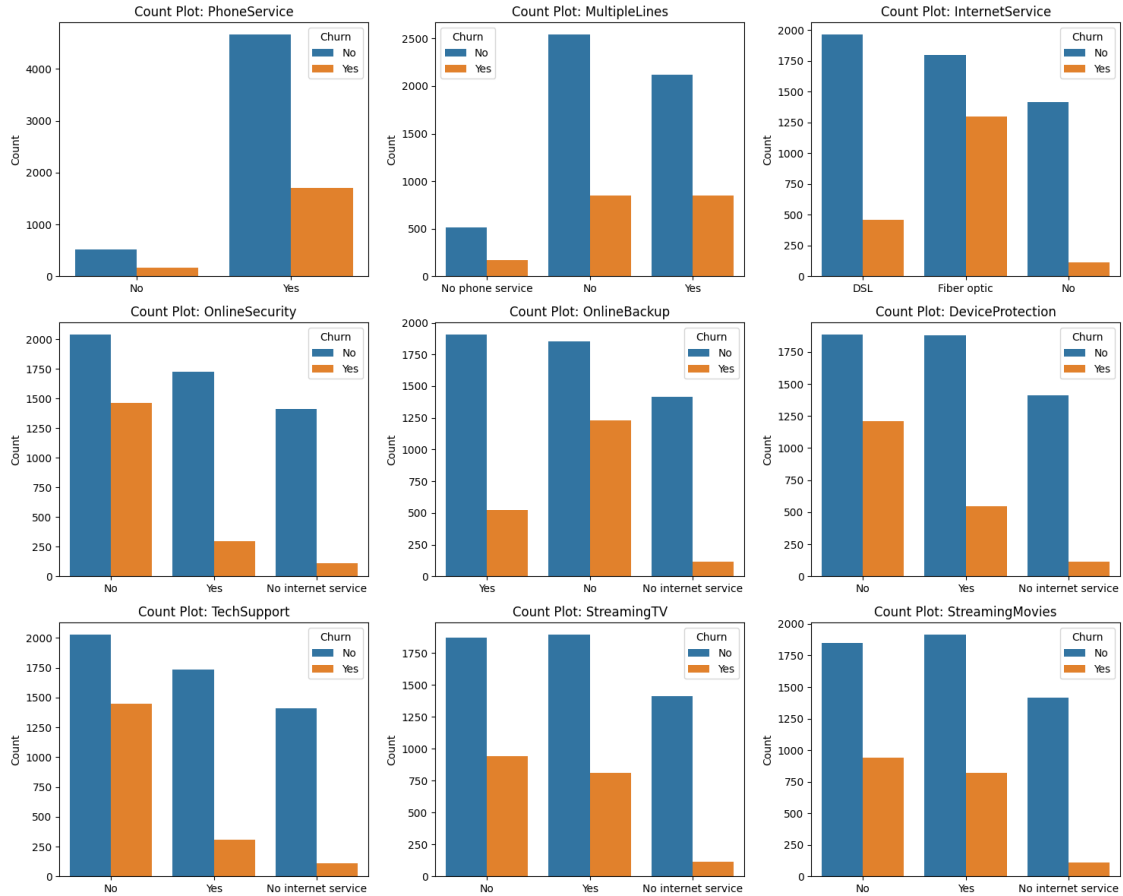
# Create subplots
fig, axes = plt.subplots(n_rows, n_cols, figsize=(15, n_rows * 4))
axes = axes.flatten() # Flatten to iterate easily

# Loop through columns and create count plots
for i, col in enumerate(columns_to_plot):
    ax = axes[i]
    sns.countplot(data=df, x=col, ax=ax, hue = df["Churn"])
    ax.set_title(f"Count Plot: {col}", fontsize=12)
    ax.set_xlabel("")
    ax.set_ylabel("Count")
    ax.tick_params(axis='x', rotation=0)

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

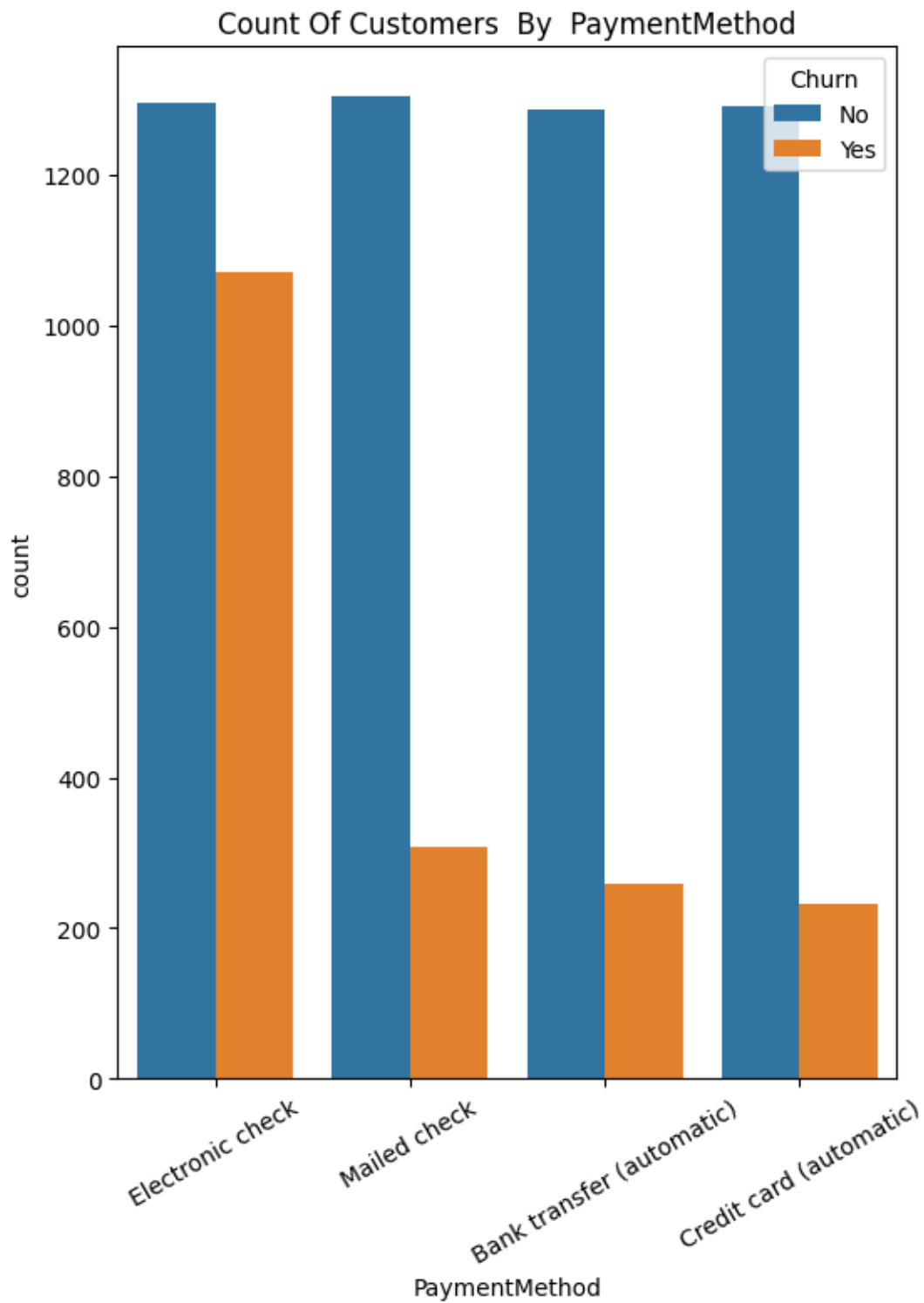
# Adjust layout
plt.tight_layout()
plt.show()

```



#Services such as “OnlineSecurity,” “OnlineBackup,” and “TechSupport” show higher churn rates when customers do not have these features. “Fiber optic” internet service appears to have a higher churn rate compared to “DSL” or no internet service. The availability of features like “StreamingTV” and “StreamingMovies” does not clearly favor retention, as churn is relatively similar between “Yes” and “No” groups.

```
[68]: plt.figure(figsize = (6,8))
sns.countplot(x = "PaymentMethod" , data = df, hue = "Churn")
ax.bar_label(ax.containers[0])
ax.bar_label(ax.containers[1])
plt.xticks(rotation = 30 )
plt.title("Count Of Customers By PaymentMethod")
plt.show()
```



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

[ ]: