

In [2]:

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

# Load the dataset
df = pd.read_csv("C:\\\\Users\\\\prems\\\\Videos\\\\power bi\\\\Telco_Cusomer_Churn.csv")

# Basic information about the dataset
print("Dataset Shape:", df.shape)
print("\n" + "*60)
print("\nFirst 5 rows:")
print(df.head())
print("\n" + "*60)
print("\nDataset Information:")
print(df.info())
print("\n" + "*60)
print("\nMissing Values:")
print(df.isnull().sum())

# Convert TotalCharges to numeric (handling empty strings)
df['TotalCharges'] = pd.to_numeric(df['TotalCharges'], errors='coerce')
```

Dataset Shape: (7043, 21)

=====

First 5 rows:

| | 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]

```
=====
Dataset Information:  
<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  
None
```

```
=====  
Missing Values:  
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 [18]:

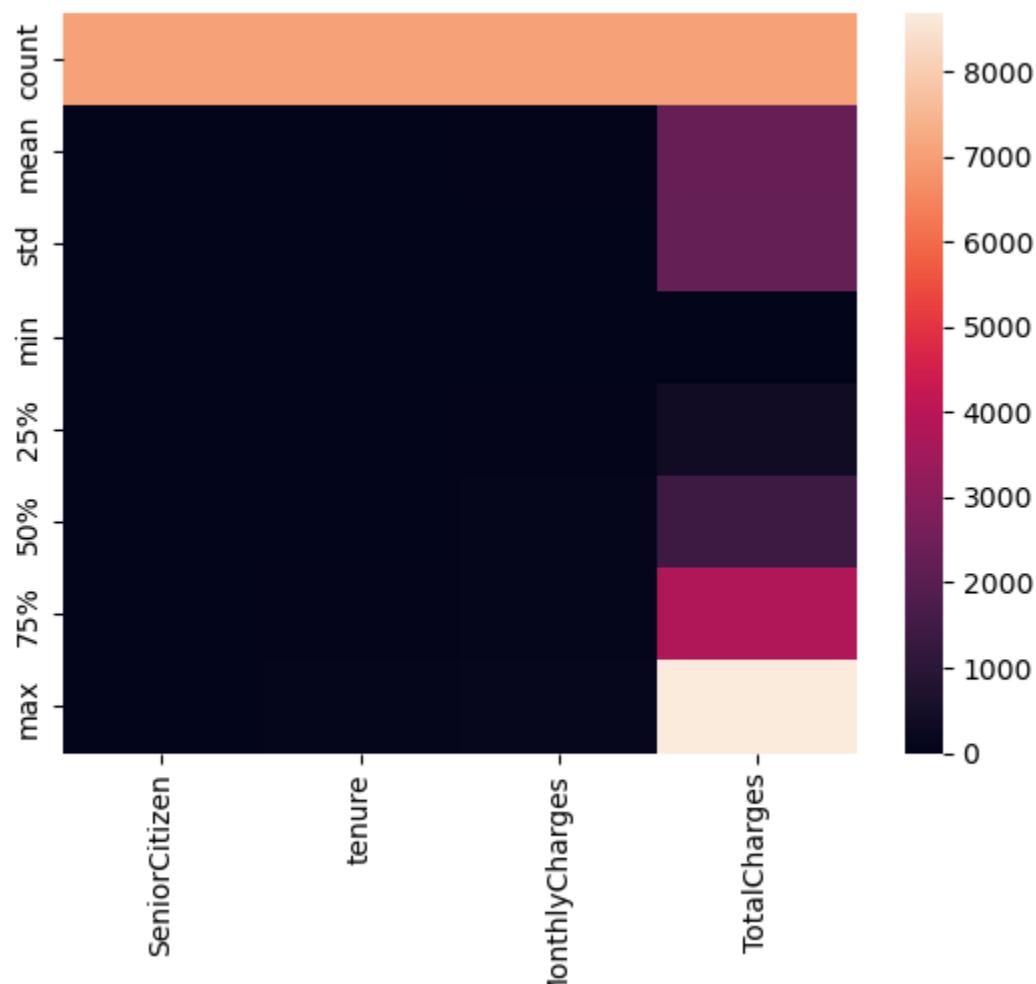
```
# Check basic statistics
print("\nBasic Statistics:")
print(df.describe())
sns.heatmap(df.describe())
```

Basic Statistics:

| | SeniorCitizen | tenure | MonthlyCharges | TotalCharges |
|-------|---------------|-------------|----------------|--------------|
| count | 7043.000000 | 7043.000000 | 7043.000000 | 7032.000000 |
| mean | 0.162147 | 32.371149 | 64.761692 | 2283.300441 |
| std | 0.368612 | 24.559481 | 30.090047 | 2266.771362 |
| min | 0.000000 | 0.000000 | 18.250000 | 18.800000 |
| 25% | 0.000000 | 9.000000 | 35.500000 | 401.450000 |
| 50% | 0.000000 | 29.000000 | 70.350000 | 1397.475000 |
| 75% | 0.000000 | 55.000000 | 89.850000 | 3794.737500 |
| max | 1.000000 | 72.000000 | 118.750000 | 8684.800000 |

Out[18]:

<Axes: >



In []:

```
df.style.set_caption('describe the data')
```

In [30]:

```
print("\nisnull:")
print(df.isnull())
print("\n" + "="*60)
print("\nisnullsum:")
print(df.isnull().sum())
```

```
df.add_prefix('lefttone_')
df.add_suffix('_righttone')
```

| isnull: | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | \ |
|---------|------------------|---------------|-----------------|-----------------|------------|--------|---|
| 0 | False | False | False | False | False | False | |
| 1 | False | False | False | False | False | False | |
| 2 | False | False | False | False | False | False | |
| 3 | False | False | False | False | False | False | |
| 4 | False | False | False | False | False | False | |
| ... | ... | ... | ... | ... | ... | ... | |
| 7038 | False | False | False | False | False | False | |
| 7039 | False | False | False | False | False | False | |
| 7040 | False | False | False | False | False | False | |
| 7041 | False | False | False | False | False | False | |
| 7042 | False | False | False | False | False | False | |
| | PhoneService | MultipleLines | InternetService | OnlineSecurity | ... | \ | |
| 0 | False | False | False | False | False | ... | |
| 1 | False | False | False | False | False | ... | |
| 2 | False | False | False | False | False | ... | |
| 3 | False | False | False | False | False | ... | |
| 4 | False | False | False | False | False | ... | |
| ... | ... | ... | ... | ... | ... | ... | |
| 7038 | False | False | False | False | False | ... | |
| 7039 | False | False | False | False | False | ... | |
| 7040 | False | False | False | False | False | ... | |
| 7041 | False | False | False | False | False | ... | |
| 7042 | False | False | False | False | False | ... | |
| | DeviceProtection | TechSupport | StreamingTV | StreamingMovies | Contract | \ | |
| 0 | False | False | False | False | False | False | |
| 1 | False | False | False | False | False | False | |
| 2 | False | False | False | False | False | False | |
| 3 | False | False | False | False | False | False | |
| 4 | False | False | False | False | False | False | |
| ... | ... | ... | ... | ... | ... | ... | |
| 7038 | False | False | False | False | False | False | |
| 7039 | False | False | False | False | False | False | |
| 7040 | False | False | False | False | False | False | |
| 7041 | False | False | False | False | False | False | |
| 7042 | False | False | False | False | False | False | |
| | PaperlessBilling | PaymentMethod | MonthlyCharges | TotalCharges | Churn | | |
| 0 | False | False | False | False | False | False | |
| 1 | False | False | False | False | False | False | |
| 2 | False | False | False | False | False | False | |
| 3 | False | False | False | False | False | False | |
| 4 | False | False | False | False | False | False | |
| ... | ... | ... | ... | ... | ... | ... | |
| 7038 | False | False | False | False | False | False | |
| 7039 | False | False | False | False | False | False | |
| 7040 | False | False | False | False | False | False | |
| 7041 | False | False | False | False | False | False | |
| 7042 | False | False | False | False | False | False | |

[7043 rows x 21 columns]

```
isnullsum:  
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    11  
Churn           0  
dtype: int64
```

```
Out[30]:
```

| | customerID_righttöne | gender_righttöne | SeniorCitizen_righttöne | Partner_righttöne | Dependents_righttöne |
|------|----------------------|------------------|-------------------------|-------------------|----------------------|
| 0 | 7590-VHVEG | Female | 0 | Yes | |
| 1 | 5575-GNVDE | Male | 0 | No | |
| 2 | 3668-QPYBK | Male | 0 | No | |
| 3 | 7795-CFOCW | Male | 0 | No | |
| 4 | 9237-HQITU | Female | 0 | No | |
| ... | ... | ... | ... | ... | ... |
| 7038 | 6840-RESVB | Male | 0 | Yes | |
| 7039 | 2234-XADUH | Female | 0 | Yes | |
| 7040 | 4801-JZAZL | Female | 0 | Yes | |
| 7041 | 8361-LTMKD | Male | 1 | Yes | |
| 7042 | 3186-AJIEK | Male | 0 | No | |

7043 rows × 21 columns

```
In [13]:
```

```
print("\nremoving the missing values:")
print(df.dropna())
```

```
removing the missing values:
customerID  gender  SeniorCitizen  Partner  Dependents  tenure \
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
```

| | | | | | | |
|------|------------|--------|---|-----|-----|----|
| 7038 | 6840-RESVB | Male | 0 | Yes | Yes | 24 |
| 7039 | 2234-XADUH | Female | 0 | Yes | Yes | 72 |
| 7040 | 4801-JZAZL | Female | 0 | Yes | Yes | 11 |
| 7041 | 8361-LTMKD | Male | 1 | Yes | No | 4 |
| 7042 | 3186-AJIEK | Male | 0 | No | No | 66 |

| PhoneService | | MultipleLines | InternetService | OnlineSecurity | ... | \ |
|--------------|-----|------------------|-----------------|----------------|-----|-----|
| 0 | No | No phone service | | DSL | No | ... |
| 1 | Yes | | No | DSL | Yes | ... |
| 2 | Yes | | No | DSL | Yes | ... |
| 3 | No | No phone service | | DSL | Yes | ... |
| 4 | Yes | | No | Fiber optic | No | ... |
| ... | ... | ... | ... | ... | ... | ... |
| 7038 | Yes | | Yes | DSL | Yes | ... |
| 7039 | Yes | | Yes | Fiber optic | No | ... |
| 7040 | No | No phone service | | DSL | Yes | ... |
| 7041 | Yes | | Yes | Fiber optic | No | ... |
| 7042 | Yes | | No | Fiber optic | Yes | ... |

| DeviceProtection | | TechSupport | StreamingTV | StreamingMovies | Contract | \ |
|------------------|-----|-------------|-------------|-----------------|----------------|-----|
| 0 | No | No | No | No | Month-to-month | |
| 1 | Yes | No | No | No | One year | |
| 2 | No | No | No | No | Month-to-month | |
| 3 | Yes | Yes | No | No | One year | |
| 4 | No | No | No | No | Month-to-month | |
| ... | ... | ... | ... | ... | ... | ... |
| 7038 | Yes | Yes | Yes | Yes | One year | |
| 7039 | Yes | No | Yes | Yes | One year | |
| 7040 | No | No | No | No | Month-to-month | |
| 7041 | No | No | No | No | Month-to-month | |
| 7042 | Yes | Yes | Yes | Yes | Two year | |

| PaperlessBilling | | PaymentMethod | MonthlyCharges | TotalCharges | \ |
|------------------|-----|---------------------------|----------------|--------------|-----|
| 0 | Yes | Electronic check | 29.85 | 29.85 | |
| 1 | No | Mailed check | 56.95 | 1889.50 | |
| 2 | Yes | Mailed check | 53.85 | 108.15 | |
| 3 | No | Bank transfer (automatic) | 42.30 | 1840.75 | |
| 4 | Yes | Electronic check | 70.70 | 151.65 | |
| ... | ... | ... | ... | ... | ... |
| 7038 | Yes | Mailed check | 84.80 | 1990.50 | |
| 7039 | Yes | Credit card (automatic) | 103.20 | 7362.90 | |
| 7040 | Yes | Electronic check | 29.60 | 346.45 | |
| 7041 | Yes | Mailed check | 74.40 | 306.60 | |
| 7042 | Yes | Bank transfer (automatic) | 105.65 | 6844.50 | |

| Churn | |
|-------|-----|
| 0 | No |
| 1 | No |
| 2 | Yes |
| 3 | No |
| 4 | Yes |
| ... | ... |
| 7038 | No |
| 7039 | No |
| 7040 | No |
| 7041 | Yes |
| 7042 | No |

```
[7032 rows x 21 columns]
```

```
In [4]:
```

```
print(df.columns)
```

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

gender is male

```
In [34]:
```

```
df[df.gender=='Male']
```

```
Out[34]:
```

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService | OnlineSecurity | OnlineBackup | DeviceProtection | TechSupport | StreamingTV | StreamingMovies | Contract | PaperlessBilling | PaymentMethod | MonthlyCharges | TotalCharges | Churn |
|------|------------|--------|---------------|---------|------------|--------|--------------|---------------|-----------------|----------------|--------------|------------------|-------------|-------------|-----------------|----------|------------------|---------------|----------------|--------------|-------|
| 1 | 5575-GNVDE | Male | 0 | No | No | 34 | Yes | No | No | No | No | No | No | No | No | No | No | No | No | No | No |
| 2 | 3668-QPYBK | Male | 0 | No | No | 2 | Yes | No | No | No | No | No | No | No | No | No | No | No | No | No | No |
| 3 | 7795-CFOCW | Male | 0 | No | No | 45 | No | No | No | No | No | No | No | No | No | No | No | No | No | No | No |
| 6 | 1452-KIOVK | Male | 0 | No | Yes | 22 | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes |
| 9 | 6388-TABGU | Male | 0 | No | Yes | 62 | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 7033 | 9767-FFLEM | Male | 0 | No | No | 38 | Yes | No | No | No | No | No | No | No | No | No | No | No | No | No | No |
| 7035 | 8456-QDAVC | Male | 0 | No | No | 19 | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes |
| 7038 | 6840-RESVB | Male | 0 | Yes | Yes | 24 | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes |
| 7041 | 8361-LTMKD | Male | 1 | Yes | No | 4 | No | No | No | No | No | No | No | No | No | No | No | No | No | No | No |
| 7042 | 3186-AJIEK | Male | 0 | No | No | 66 | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes |

3555 rows x 21 columns

```
In [ ]:
```

```
sns.displot(df.gender=='male')
```

rename columns like customerid to id

```
In [13]:
```

```
df.rename(columns={'customerID':'id'}) #inplace=true (permanent )
```

Out[13]:

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

7043 rows × 21 columns

In [16]:

```
df[df['InternetService'].str.contains('DSL')].head(10)
```

Out[16]:

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService |
|----|------------|--------|---------------|---------|------------|--------|--------------|------------------|-----------------|
| 0 | 7590-VHVEG | Female | 0 | Yes | No | 1 | No | No phone service | |
| 1 | 5575-GNVDE | Male | 0 | No | No | 34 | Yes | No | |
| 2 | 3668-QPYBK | Male | 0 | No | No | 2 | Yes | No | |
| 3 | 7795-CFOCW | Male | 0 | No | No | 45 | No | No phone service | |
| 7 | 6713-OKOMC | Female | 0 | No | No | 10 | No | No phone service | |
| 9 | 6388-TABGU | Male | 0 | No | Yes | 62 | Yes | No | |
| 10 | 9763-GRSKD | Male | 0 | Yes | Yes | 13 | Yes | No | |

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService |
|----|------------|--------|---------------|---------|------------|--------|--------------|---------------|------------------|
| 18 | 4190-MFLUW | Female | 0 | Yes | Yes | 10 | Yes | No | |
| 20 | 8779-QRDMV | Male | 1 | No | No | 1 | No | No | No phone service |
| 23 | 3638-WEABW | Female | 0 | Yes | No | 58 | Yes | Yes | |

10 rows × 21 columns

find all instance when 'gender is male' and InternetService is DSL'

In [17]:

```
df[(df['gender']=='Male')&(df['InternetService']=='DSL')]
```

Out[17]:

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService |
|------|------------|--------|---------------|---------|------------|--------|--------------|---------------|------------------|
| 1 | 5575-GNVDE | Male | 0 | No | No | 34 | Yes | No | |
| 2 | 3668-QPYBK | Male | 0 | No | No | 2 | Yes | No | |
| 3 | 7795-CFOCW | Male | 0 | No | No | 45 | No | No | No phone service |
| 9 | 6388-TABGU | Male | 0 | No | Yes | 62 | Yes | No | |
| 10 | 9763-GRSKD | Male | 0 | Yes | Yes | 13 | Yes | No | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 7007 | 2274-XUATA | Male | 1 | Yes | No | 72 | No | No | No phone service |
| 7021 | 1699-HPSBG | Male | 0 | No | No | 12 | Yes | No | |
| 7027 | 0550-DCXLH | Male | 0 | No | No | 13 | Yes | No | |
| 7031 | 3605-JISKB | Male | 1 | Yes | No | 55 | Yes | Yes | |
| 7038 | 6840-RESVB | Male | 0 | Yes | Yes | 24 | Yes | Yes | |

1233 rows × 21 columns

In [20]:

```
df[(df['gender']=='Male')&(df['InternetService']=='DSL')|(df['OnlineSecurity']=='Yes')]
```

Out[20]:

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | Intl |
|------|------------|--------|---------------|---------|------------|--------|--------------|------------------|------|
| 1 | 5575-GNVDE | Male | 0 | No | No | 34 | Yes | No | |
| 2 | 3668-QPYBK | Male | 0 | No | No | 2 | Yes | No | |
| 3 | 7795-CFOCW | Male | 0 | No | No | 45 | No | No phone service | |
| 7 | 6713-OKOMC | Female | 0 | No | No | 10 | No | No phone service | |
| 9 | 6388-TABGU | Male | 0 | No | Yes | 62 | Yes | No | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 7031 | 3605-JISKB | Male | 1 | Yes | No | 55 | Yes | Yes | |
| 7034 | 0639-TSIQW | Female | 0 | No | No | 67 | Yes | Yes | |
| 7038 | 6840-RESVB | Male | 0 | Yes | Yes | 24 | Yes | Yes | |
| 7040 | 4801-JZAZL | Female | 0 | Yes | Yes | 11 | No | No phone service | |
| 7042 | 3186-AJIEK | Male | 0 | No | No | 66 | Yes | No | |

2666 rows × 21 columns

In [23]:

```
df[df['PaymentMethod'].isin(['Electronic check','Mailed check'])]
```

Out[23]:

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | Intl |
|------|------------|--------|---------------|---------|------------|--------|--------------|------------------|------|
| 0 | 7590-VHVEG | Female | 0 | Yes | No | 1 | No | No phone service | |
| 1 | 5575-GNVDE | Male | 0 | No | No | 34 | Yes | No | |
| 2 | 3668-QPYBK | Male | 0 | No | No | 2 | Yes | No | |
| 4 | 9237-HQITU | Female | 0 | No | No | 2 | Yes | No | |
| 5 | 9305-CDSKC | Female | 0 | No | No | 8 | Yes | Yes | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 7032 | 6894-LFHLY | Male | 1 | No | No | 1 | Yes | Yes | |
| 7036 | 7750-EYXWZ | Female | 0 | No | No | 12 | No | No phone service | |
| 7038 | 6840- | Male | 0 | Yes | Yes | 24 | Yes | Yes | |

| customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService |
|------------|------------|---------------|---------|------------|--------|--------------|---------------|-----------------|
| RESVB | | | | | | | | |
| 7040 | 4801-JZAZL | Female | | 0 | Yes | Yes | 11 | No |
| 7041 | 8361-LTMKD | Male | | 1 | Yes | No | 4 | Yes |

3977 rows × 21 columns

remove duplicates

In [24]:

```
df[~(df['gender'] == 'Male')]
```

Out[24]:

| customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService |
|------------|------------|---------------|---------|------------|--------|--------------|---------------|-----------------|
| 0 | 7590-VHVEG | Female | | 0 | Yes | No | 1 | No |
| 4 | 9237-HQITU | Female | | 0 | No | No | 2 | Yes |
| 5 | 9305-CDSKC | Female | | 0 | No | No | 8 | Yes |
| 7 | 6713-OKOMC | Female | | 0 | No | No | 10 | No |
| 8 | 7892-POOKP | Female | | 0 | Yes | No | 28 | Yes |
| ... | ... | ... | | ... | ... | ... | ... | ... |
| 7034 | 0639-TSIQW | Female | | 0 | No | No | 67 | Yes |
| 7036 | 7750-EYXWZ | Female | | 0 | No | No | 12 | No |
| 7037 | 2569-WGERO | Female | | 0 | No | No | 72 | Yes |
| 7039 | 2234-XADUH | Female | | 0 | Yes | Yes | 72 | Yes |
| 7040 | 4801-JZAZL | Female | | 0 | Yes | Yes | 11 | No |

3488 rows × 21 columns

In [31]:

```
df['tenure'] = df['tenure'].apply(lambda x: x+3)
```

In [32]:

```
df
```

Out[32]:

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService |
|------|------------|--------|---------------|---------|------------|--------|--------------|---------------|------------------|
| 0 | 7590-VHVEG | Female | 0 | Yes | No | 4 | No | No | No phone service |
| 1 | 5575-GNVDE | Male | 0 | No | No | 37 | Yes | No | No |
| 2 | 3668-QPYBK | Male | 0 | No | No | 5 | Yes | No | No |
| 3 | 7795-CFOCW | Male | 0 | No | No | 48 | No | No | No phone service |
| 4 | 9237-HQITU | Female | 0 | No | No | 5 | Yes | No | No |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 7038 | 6840-RESVB | Male | 0 | Yes | Yes | 27 | Yes | Yes | Yes |
| 7039 | 2234-XADUH | Female | 0 | Yes | Yes | 75 | Yes | Yes | Yes |
| 7040 | 4801-JZAZL | Female | 0 | Yes | Yes | 14 | No | No | No phone service |
| 7041 | 8361-LTMKD | Male | 1 | Yes | No | 7 | Yes | Yes | Yes |
| 7042 | 3186-AJIEK | Male | 0 | No | No | 69 | Yes | No | No |

7043 rows × 21 columns

In [33]:

```
df[df.PaymentMethod=='Mailed check'].gender.value_counts()
```

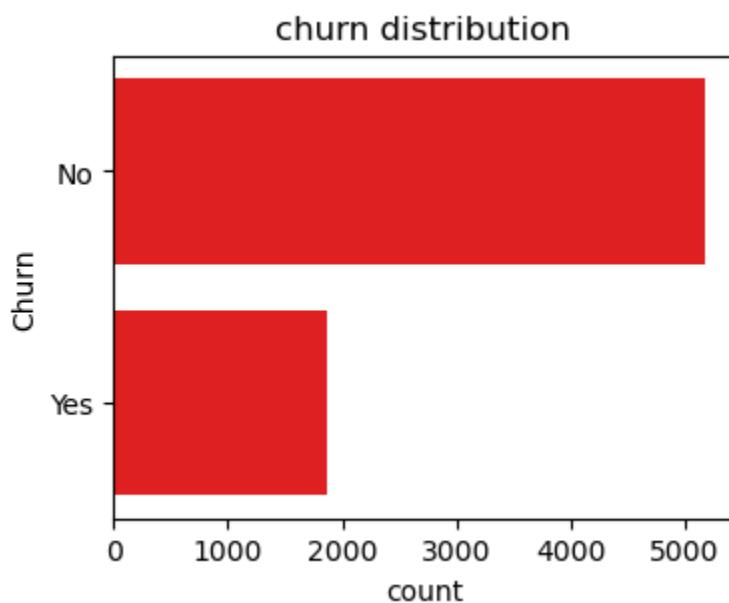
Out[33]:

```
gender
Male      834
Female    778
Name: count, dtype: int64
```

what is the distribution of churn(yes/no)?

In [41]:

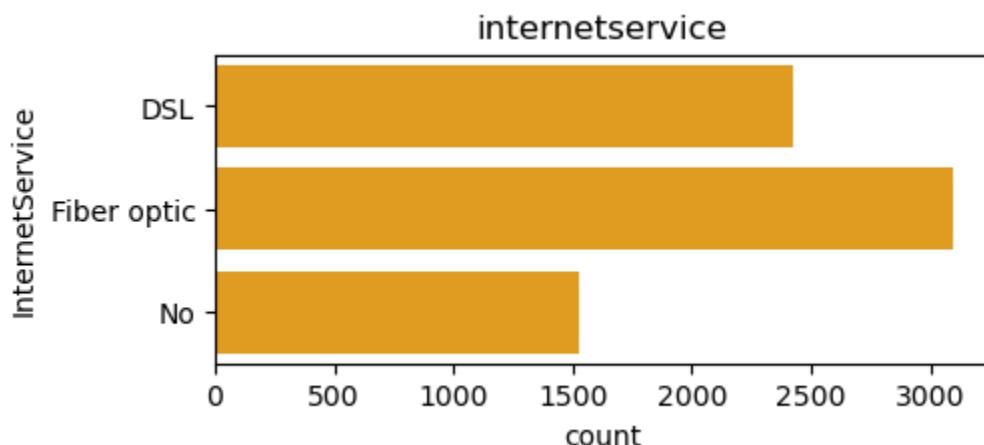
```
plt.figure(figsize=(4,3))
sns.countplot(df['Churn'], color='red')
plt.title('churn distribution')
plt.show()
```



what is the distribution of internet service types?

In [42]:

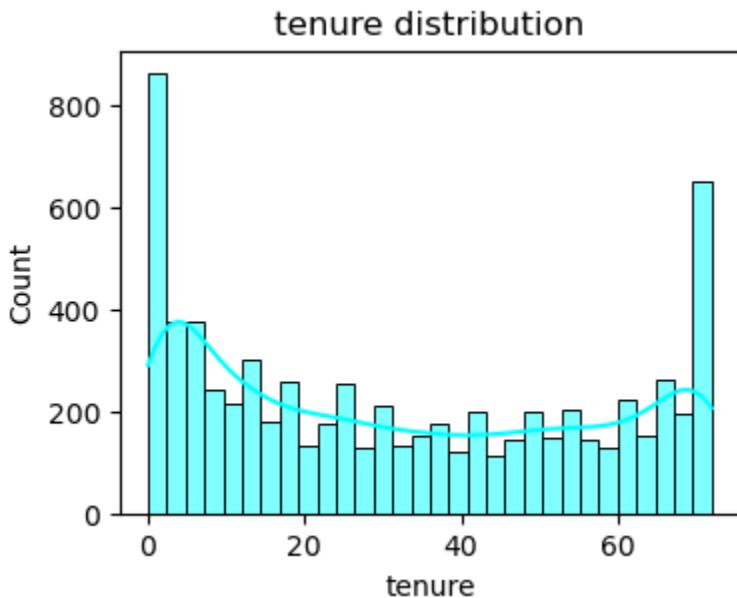
```
plt.figure(figsize=(5,2))
sns.countplot(df['InternetService'],color='orange')
plt.title('internetservice')
plt.show()
```



how is the tenure distribution among customers?

In [43]:

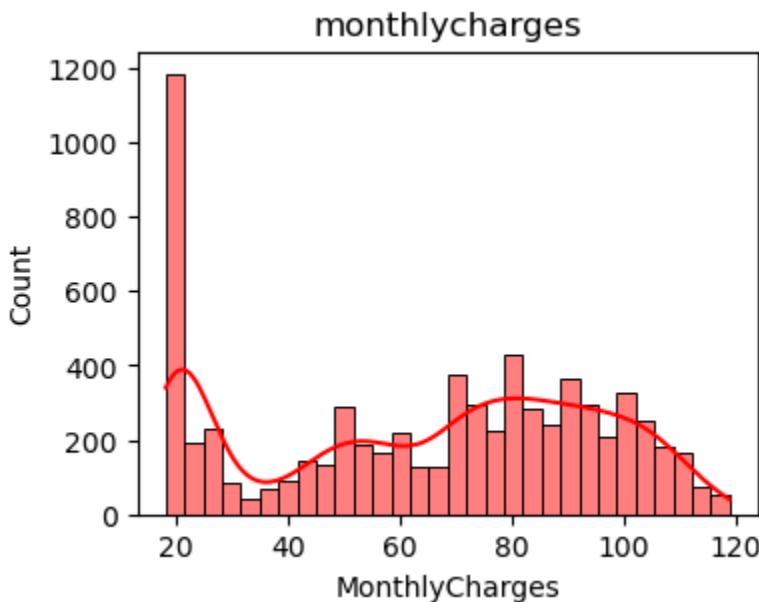
```
plt.figure(figsize=(4,3))
sns.histplot(df['tenure'],color='cyan',kde=True,bins=30)
plt.title('tenure distribution')
plt.show()
```



what is the distribution of monthly charges?

In [44]:

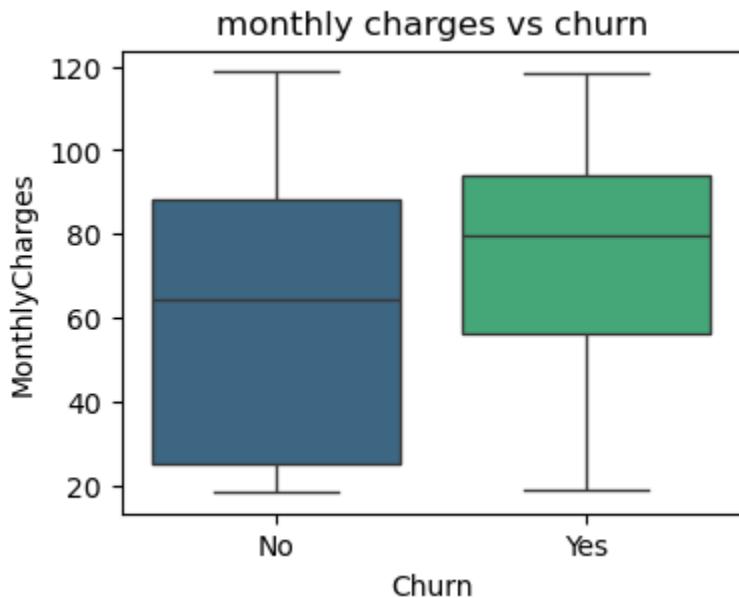
```
plt.figure(figsize=(4,3))
sns.histplot(df['MonthlyCharges'], color='red', bins=30, kde=True, )
plt.title('monthlycharges')
plt.show()
```



do monthly charges differ between churned and non-churned customers?

In [46]:

```
plt.figure(figsize=(4,3))
sns.boxplot(df,x='Churn',y='MonthlyCharges',hue='Churn',palette='viridis')
plt.title('monthly charges vs churn')
plt.show()
```



In [24]:

```
print("Duplicate Customer IDs:")
duplicate_ids = df[df['customerID'].duplicated(keep=False)]
print(f"Found {len(duplicate_ids)} duplicate customer IDs")
```

Duplicate Customer IDs:

Found 0 duplicate customer IDs

In [4]:

```
print("\nduplicates:")
df[df.duplicated()]
```

duplicates:

Out[4]:

| customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService |
|------------|--------|---------------|---------|------------|--------|--------------|---------------|-----------------|
|------------|--------|---------------|---------|------------|--------|--------------|---------------|-----------------|

0 rows × 21 columns

In [8]:

```
print("\ndrop duplicates :")
print(df.drop_duplicates())
```

drop duplicates :

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines | InternetService | OnlineSecurity | ... \ |
|------|------------|--------|---------------|---------|------------|--------|--------------|---------------|-----------------|----------------|-------|
| 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 | | | | | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 7038 | 6840-RESVB | Male | 0 | Yes | Yes | 24 | | | | | |
| 7039 | 2234-XADUH | Female | 0 | Yes | Yes | 72 | | | | | |
| 7040 | 4801-JZAZL | Female | 0 | Yes | Yes | 11 | | | | | |
| 7041 | 8361-LTMKD | Male | 1 | Yes | No | 4 | | | | | |
| 7042 | 3186-AJIEK | Male | 0 | No | No | 66 | | | | | |

| | PhoneService | MultipleLines | InternetService | OnlineSecurity | ... \ |
|---|--------------|------------------|-----------------|----------------|-------|
| 0 | No | No phone service | DSL | No | ... |
| 1 | Yes | No | DSL | Yes | ... |

| | | | | | |
|------|-----|------------------|-------------|-----|-----|
| 2 | Yes | No | DSL | Yes | ... |
| 3 | No | No phone service | DSL | Yes | ... |
| 4 | Yes | No | Fiber optic | No | ... |
| ... | ... | ... | ... | ... | ... |
| 7038 | Yes | Yes | DSL | Yes | ... |
| 7039 | Yes | Yes | Fiber optic | No | ... |
| 7040 | No | No phone service | DSL | Yes | ... |
| 7041 | Yes | Yes | Fiber optic | No | ... |
| 7042 | Yes | No | Fiber optic | Yes | ... |

| | DeviceProtection | TechSupport | StreamingTV | StreamingMovies | Contract | \ |
|------|------------------|-------------|-------------|-----------------|----------------|-----|
| 0 | No | No | No | No | Month-to-month | |
| 1 | Yes | No | No | No | One year | |
| 2 | No | No | No | No | Month-to-month | |
| 3 | Yes | Yes | No | No | One year | |
| 4 | No | No | No | No | Month-to-month | |
| ... | ... | ... | ... | ... | ... | ... |
| 7038 | Yes | Yes | Yes | Yes | One year | |
| 7039 | Yes | No | Yes | Yes | One year | |
| 7040 | No | No | No | No | Month-to-month | |
| 7041 | No | No | No | No | Month-to-month | |
| 7042 | Yes | Yes | Yes | Yes | Two year | |

| | PaperlessBilling | | PaymentMethod | MonthlyCharges | TotalCharges | \ |
|------|------------------|---------------------------|------------------|----------------|--------------|-----|
| 0 | Yes | | Electronic check | 29.85 | 29.85 | |
| 1 | No | | Mailed check | 56.95 | 1889.50 | |
| 2 | Yes | | Mailed check | 53.85 | 108.15 | |
| 3 | No | Bank transfer (automatic) | | 42.30 | 1840.75 | |
| 4 | Yes | | Electronic check | 70.70 | 151.65 | |
| ... | ... | | ... | ... | ... | ... |
| 7038 | Yes | | Mailed check | 84.80 | 1990.50 | |
| 7039 | Yes | Credit card (automatic) | | 103.20 | 7362.90 | |
| 7040 | Yes | | Electronic check | 29.60 | 346.45 | |
| 7041 | Yes | | Mailed check | 74.40 | 306.60 | |
| 7042 | Yes | Bank transfer (automatic) | | 105.65 | 6844.50 | |

| | Churn |
|------|-------|
| 0 | No |
| 1 | No |
| 2 | Yes |
| 3 | No |
| 4 | Yes |
| ... | ... |
| 7038 | No |
| 7039 | No |
| 7040 | No |
| 7041 | Yes |
| 7042 | No |

[7043 rows x 21 columns]

In [16]:

```
print("\n transpose the data:")
print(df.T)
```

transpose the data:

| | 0 | 1 | 2 | \ |
|------------|------------|------------|------------|---|
| customerID | 7590-VHVEG | 5575-GNVDE | 3668-QPYBK | |
| gender | Female | Male | Male | |

| | | | |
|------------------|------------------|--------------|----------------|
| SeniorCitizen | 0 | 0 | 0 |
| Partner | Yes | No | No |
| Dependents | No | No | No |
| tenure | 1 | 34 | 2 |
| PhoneService | No | Yes | Yes |
| MultipleLines | No phone service | No | No |
| InternetService | DSL | DSL | DSL |
| OnlineSecurity | No | Yes | Yes |
| OnlineBackup | Yes | No | Yes |
| DeviceProtection | No | Yes | No |
| TechSupport | No | No | No |
| StreamingTV | No | No | No |
| StreamingMovies | No | No | No |
| Contract | Month-to-month | One year | Month-to-month |
| PaperlessBilling | Yes | No | Yes |
| PaymentMethod | Electronic check | Mailed check | Mailed check |
| MonthlyCharges | 29.85 | 56.95 | 53.85 |
| TotalCharges | 29.85 | 1889.5 | 108.15 |
| Churn | No | No | Yes |

| | | | |
|------------------|---------------------------|------------------|---|
| customerID | 3 | 4 | \ |
| gender | 7795-CFOCW | 9237-HQITU | |
| SeniorCitizen | Male | Female | |
| Partner | 0 | 0 | |
| Dependents | No | No | |
| tenure | 45 | 2 | |
| PhoneService | No | Yes | |
| MultipleLines | No phone service | No | |
| InternetService | DSL | Fiber optic | |
| OnlineSecurity | Yes | No | |
| OnlineBackup | No | No | |
| DeviceProtection | Yes | No | |
| TechSupport | Yes | No | |
| StreamingTV | No | No | |
| StreamingMovies | No | No | |
| Contract | One year | Month-to-month | |
| PaperlessBilling | No | Yes | |
| PaymentMethod | Bank transfer (automatic) | Electronic check | |
| MonthlyCharges | 42.3 | 70.7 | |
| TotalCharges | 1840.75 | 151.65 | |
| Churn | No | Yes | |

| | | | | |
|------------------|-------------|-------------|------------------|---|
| customerID | 5 | 6 | 7 | \ |
| gender | 9305-CDSKC | 1452-KIOVK | 6713-OKOMC | |
| SeniorCitizen | Female | Male | Female | |
| Partner | 0 | 0 | 0 | |
| Dependents | No | No | No | |
| tenure | 8 | 22 | 10 | |
| PhoneService | Yes | Yes | No | |
| MultipleLines | Yes | Yes | No phone service | |
| InternetService | Fiber optic | Fiber optic | DSL | |
| OnlineSecurity | No | No | Yes | |
| OnlineBackup | No | Yes | No | |
| DeviceProtection | Yes | No | No | |
| TechSupport | No | No | No | |
| StreamingTV | Yes | Yes | No | |
| StreamingMovies | Yes | No | No | |

| | | | |
|------------------|------------------|-------------------------|----------------|
| Contract | Month-to-month | Month-to-month | Month-to-month |
| PaperlessBilling | Yes | Yes | No |
| PaymentMethod | Electronic check | Credit card (automatic) | Mailed check |
| MonthlyCharges | 99.65 | 89.1 | 29.75 |
| TotalCharges | 820.5 | 1949.4 | 301.9 |
| Churn | Yes | No | No |

| | | | |
|------------------|------------------|---------------------------|-----|
| | 8 | 9 | ... |
| customerID | 7892-P00KP | 6388-TABGU | ... |
| gender | Female | Male | ... |
| SeniorCitizen | 0 | 0 | ... |
| Partner | Yes | No | ... |
| Dependents | No | Yes | ... |
| tenure | 28 | 62 | ... |
| PhoneService | Yes | Yes | ... |
| MultipleLines | Yes | No | ... |
| InternetService | Fiber optic | DSL | ... |
| OnlineSecurity | No | Yes | ... |
| OnlineBackup | No | Yes | ... |
| DeviceProtection | Yes | No | ... |
| TechSupport | Yes | No | ... |
| StreamingTV | Yes | No | ... |
| StreamingMovies | Yes | No | ... |
| Contract | Month-to-month | One year | ... |
| PaperlessBilling | Yes | No | ... |
| PaymentMethod | Electronic check | Bank transfer (automatic) | ... |
| MonthlyCharges | 104.8 | 56.15 | ... |
| TotalCharges | 3046.05 | 3487.95 | ... |
| Churn | Yes | No | ... |

| | | | |
|------------------|-------------------------|-------------------------|---|
| | 7033 | 7034 | \ |
| customerID | 9767-FFLEM | 0639-TSIQW | |
| gender | Male | Female | |
| SeniorCitizen | 0 | 0 | |
| Partner | No | No | |
| Dependents | No | No | |
| tenure | 38 | 67 | |
| PhoneService | Yes | Yes | |
| MultipleLines | No | Yes | |
| InternetService | Fiber optic | Fiber optic | |
| OnlineSecurity | No | Yes | |
| OnlineBackup | No | Yes | |
| DeviceProtection | No | Yes | |
| TechSupport | No | No | |
| StreamingTV | No | Yes | |
| StreamingMovies | No | No | |
| Contract | Month-to-month | Month-to-month | |
| PaperlessBilling | Yes | Yes | |
| PaymentMethod | Credit card (automatic) | Credit card (automatic) | |
| MonthlyCharges | 69.5 | 102.95 | |
| TotalCharges | 2625.25 | 6886.25 | |
| Churn | No | Yes | |

| | | | |
|---------------|------------|------------|---|
| | 7035 | 7036 | \ |
| customerID | 8456-QDAVC | 7750-EYXWZ | |
| gender | Male | Female | |
| SeniorCitizen | 0 | 0 | |
| Partner | No | No | |
| Dependents | No | No | |

| | | |
|------------------|---------------------------|------------------|
| tenure | 19 | 12 |
| PhoneService | Yes | No |
| MultipleLines | No | No phone service |
| InternetService | Fiber optic | DSL |
| OnlineSecurity | No | No |
| OnlineBackup | No | Yes |
| DeviceProtection | No | Yes |
| TechSupport | No | Yes |
| StreamingTV | Yes | Yes |
| StreamingMovies | No | Yes |
| Contract | Month-to-month | One year |
| PaperlessBilling | Yes | No |
| PaymentMethod | Bank transfer (automatic) | Electronic check |
| MonthlyCharges | 78.7 | 60.65 |
| TotalCharges | 1495.1 | 743.3 |
| Churn | No | No |

| | | | |
|------------------|---------------------------|--------------|---|
| customerID | 7037 | 7038 | \ |
| | 2569-WGERO | 6840-RESVB | |
| gender | Female | Male | |
| SeniorCitizen | 0 | 0 | |
| Partner | No | Yes | |
| Dependents | No | Yes | |
| tenure | 72 | 24 | |
| PhoneService | Yes | Yes | |
| MultipleLines | No | Yes | |
| InternetService | No | DSL | |
| OnlineSecurity | No internet service | Yes | |
| OnlineBackup | No internet service | No | |
| DeviceProtection | No internet service | Yes | |
| TechSupport | No internet service | Yes | |
| StreamingTV | No internet service | Yes | |
| StreamingMovies | No internet service | Yes | |
| Contract | Two year | One year | |
| PaperlessBilling | Yes | Yes | |
| PaymentMethod | Bank transfer (automatic) | Mailed check | |
| MonthlyCharges | 21.15 | 84.8 | |
| TotalCharges | 1419.4 | 1990.5 | |
| Churn | No | No | |

| | | | | |
|------------------|-------------------------|------------------|----------------|---|
| customerID | 7039 | 7040 | 7041 | \ |
| | 2234-XADUH | 4801-JZAZL | 8361-LTMKD | |
| gender | Female | Female | Male | |
| SeniorCitizen | 0 | 0 | 1 | |
| Partner | Yes | Yes | Yes | |
| Dependents | Yes | Yes | No | |
| tenure | 72 | 11 | 4 | |
| PhoneService | Yes | No | Yes | |
| MultipleLines | Yes | No phone service | Yes | |
| InternetService | Fiber optic | DSL | Fiber optic | |
| OnlineSecurity | No | Yes | No | |
| OnlineBackup | Yes | No | No | |
| DeviceProtection | Yes | No | No | |
| TechSupport | No | No | No | |
| StreamingTV | Yes | No | No | |
| StreamingMovies | Yes | No | No | |
| Contract | One year | Month-to-month | Month-to-month | |
| PaperlessBilling | Yes | Yes | Yes | |
| PaymentMethod | Credit card (automatic) | Electronic check | Mailed check | |

| | | | |
|------------------|---------------------------|--------|-------|
| MonthlyCharges | 103.2 | 29.6 | 74.4 |
| TotalCharges | 7362.9 | 346.45 | 306.6 |
| Churn | No | No | Yes |
| | 7042 | | |
| customerID | 3186-AJIEK | | |
| gender | Male | | |
| SeniorCitizen | 0 | | |
| Partner | No | | |
| Dependents | No | | |
| tenure | 66 | | |
| PhoneService | Yes | | |
| MultipleLines | No | | |
| InternetService | Fiber optic | | |
| OnlineSecurity | Yes | | |
| OnlineBackup | No | | |
| DeviceProtection | Yes | | |
| TechSupport | Yes | | |
| StreamingTV | Yes | | |
| StreamingMovies | Yes | | |
| Contract | Two year | | |
| PaperlessBilling | Yes | | |
| PaymentMethod | Bank transfer (automatic) | | |
| MonthlyCharges | 105.65 | | |
| TotalCharges | 6844.5 | | |
| Churn | No | | |

[21 rows x 7043 columns]

In [38]:

```
# Create a summary table and transpose it
summary_by_gender = df.groupby('gender').agg({
    'customerID': 'count',
    'MonthlyCharges': 'mean',
    'TotalCharges': 'sum',
}).round(2)

print("Summary by Gender:")
print(summary_by_gender)
print("\n" + "="*60)

# Transpose the table
transposed = summary_by_gender.T
print("\nTransposed Summary (Gender as columns):")
print(transposed)
```

Summary by Gender:

| | customerID | MonthlyCharges | TotalCharges |
|--------|------------|----------------|--------------|
| gender | | | |
| Female | 3488 | 65.20 | 7952354.2 |
| Male | 3555 | 64.33 | 8103814.5 |

Transposed Summary (Gender as columns):

| gender | Female | Male |
|----------------|-----------|------------|
| customerID | 3488.0 | 3555.00 |
| MonthlyCharges | 65.2 | 64.33 |
| TotalCharges | 7952354.2 | 8103814.50 |

```
In [7]:
```

```
print(df.tenure.value_counts())

tenure
1      613
72     362
2      238
3      200
4      176
...
28      57
39      56
44      51
36      50
0       11
Name: count, Length: 73, dtype: int64
```

```
In [6]:
```

```
# Basic value_counts
print("Value Counts - Contract Type:")
contract_counts = df['Contract'].value_counts()
print(contract_counts)
print("\n" + "="*60)

# Value_counts with percentages
print("\nValue Counts with Percentages - Payment Method:")
payment_counts = df['PaymentMethod'].value_counts(normalize=True).round(4) * 100
print(payment_counts)
print("\n" + "="*60)

# Value_counts with custom sorting
print("\nValue Counts Sorted by Values - Online Services:")
online_services = ['OnlineSecurity', 'OnlineBackup', 'DeviceProtection',
                    'TechSupport', 'StreamingTV', 'StreamingMovies']

for service in online_services:
    counts = df[service].value_counts()
    print(f"\n{service}:")
    print(counts)

# Value_counts for multiple columns
print("\n" + "="*60)
print("\nValue Counts for Multiple Columns (first 3):")
multi_counts = {}
for col in ['Partner', 'Dependents', 'PhoneService']:
    multi_counts[col] = df[col].value_counts()

for key, value in multi_counts.items():
    print(f"\n{key}:")
    print(value)
```

```
Value Counts - Contract Type:
Contract
Month-to-month    3875
Two year         1695
One year          1473
Name: count, dtype: int64
```

Value Counts with Percentages - Payment Method:

```
PaymentMethod
Electronic check      33.58
Mailed check         22.89
Bank transfer (automatic) 21.92
Credit card (automatic) 21.61
Name: proportion, dtype: float64
```

Value Counts Sorted by Values - Online Services:

OnlineSecurity:

```
OnlineSecurity
No                  3498
Yes                 2019
No internet service 1526
Name: count, dtype: int64
```

OnlineBackup:

```
OnlineBackup
No                  3088
Yes                 2429
No internet service 1526
Name: count, dtype: int64
```

DeviceProtection:

```
DeviceProtection
No                  3095
Yes                 2422
No internet service 1526
Name: count, dtype: int64
```

TechSupport:

```
TechSupport
No                  3473
Yes                 2044
No internet service 1526
Name: count, dtype: int64
```

StreamingTV:

```
StreamingTV
No                  2810
Yes                 2707
No internet service 1526
Name: count, dtype: int64
```

StreamingMovies:

```
StreamingMovies
No                  2785
Yes                 2732
No internet service 1526
Name: count, dtype: int64
```

Value Counts for Multiple Columns (first 3):

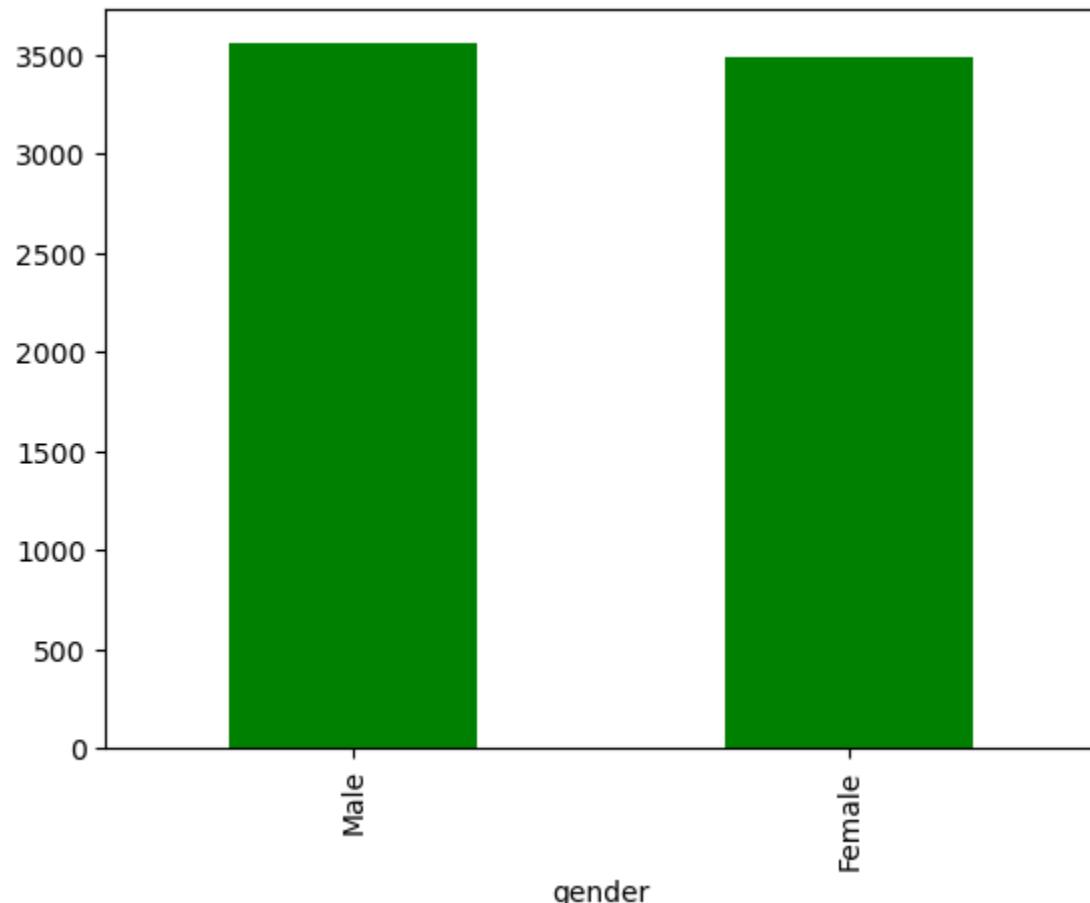
```
Partner:  
Partner  
No      3641  
Yes     3402  
Name: count, dtype: int64
```

```
Dependents:  
Dependents  
No      4933  
Yes     2110  
Name: count, dtype: int64
```

```
PhoneService:  
PhoneService  
Yes     6361  
No      682  
Name: count, dtype: int64
```

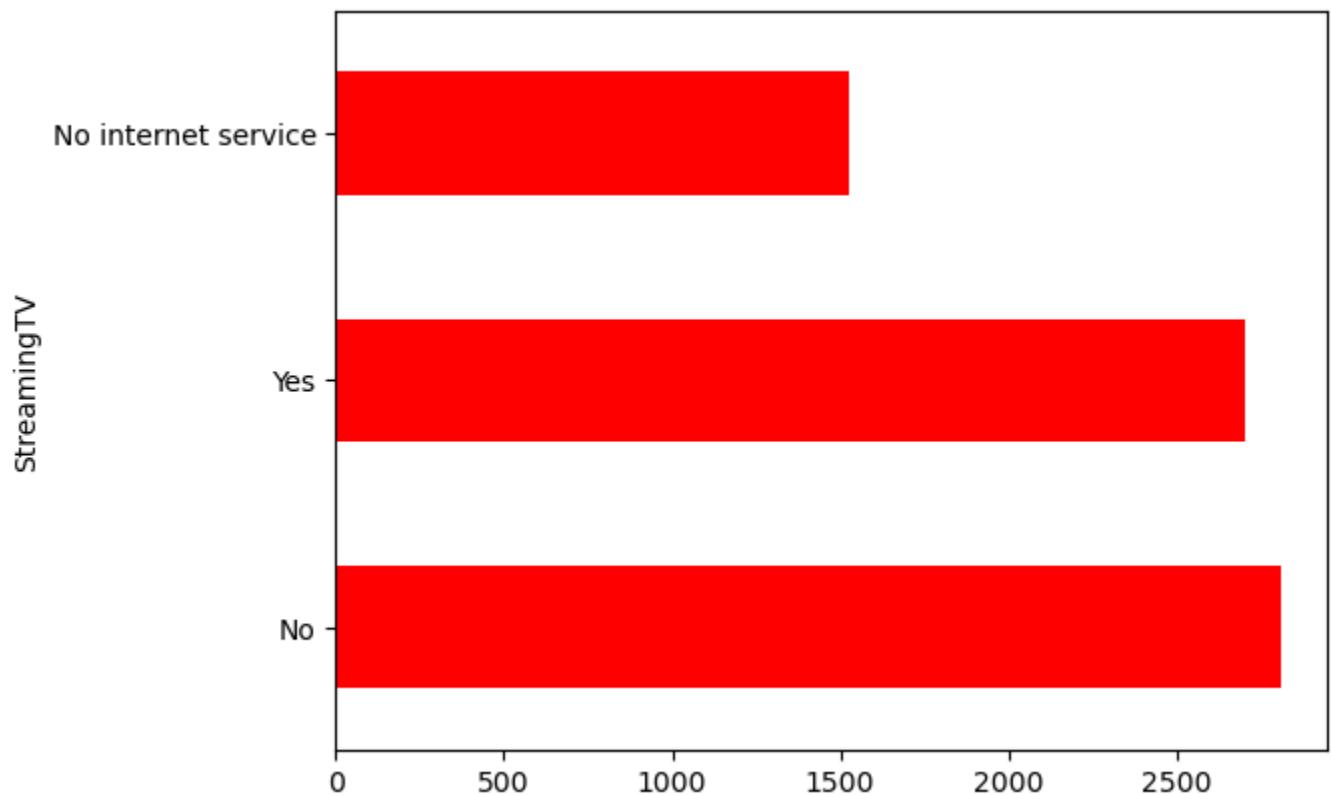
```
In [9]:
```

```
print("\n value count by gender:")  
c=df['gender'].value_counts()  
print(c)  
c.plot(kind='bar',color='g')  
plt.show()  
print("\n" + "="*60)  
print("\n value count by streaming tv:")  
d=df['StreamingTV'].value_counts()  
print(d)  
d.plot(kind='barh',color='r')  
plt.show()  
print("\n" + "="*60)  
print("\n value count by phoneservie :")  
k=df['PhoneService'].value_counts()  
print(k)  
  
df['PhoneService'].value_counts().plot(kind='pie',autopct='%1.1f%%',colors='myr',shadow=  
value count by gender:  
gender  
Male      3555  
Female    3488  
Name: count, dtype: int64
```



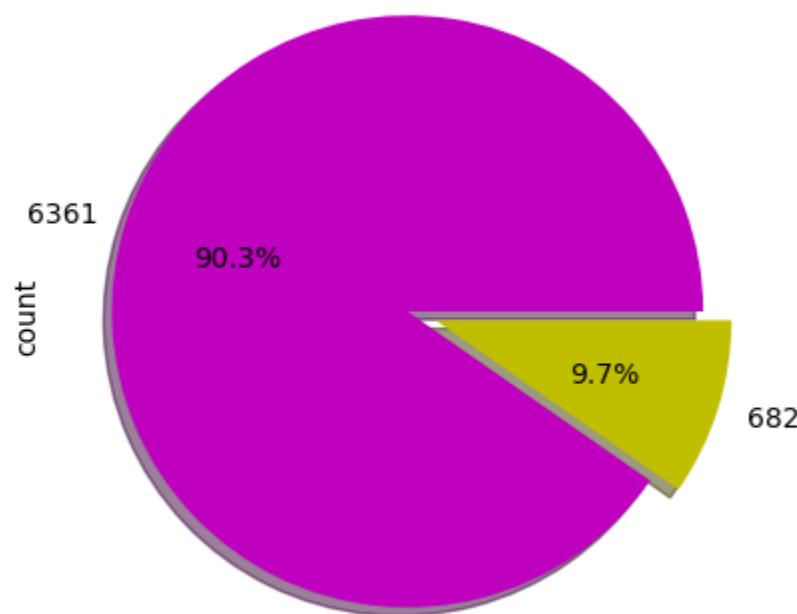
```
=====
```

value count by streaming tv:
StreamingTV
No 2810
Yes 2707
No internet service 1526
Name: count, dtype: int64



```
=====
```

```
value count by phoneservie :  
PhoneService  
Yes    6361  
No     682  
Name: count, dtype: int64  
Out[9]:  
<Axes: ylabel='count'>
```



```
In [12]:
```

```
print("\n unique values by internetservice : \n")
print(df['InternetService'].unique())
print("\n" + "*60)
print("\n unique values by internetservice and its count: \n")
print(df['InternetService'].nunique())
print("\n" + "*60)
print("\n value count by internetservice: \n")
print(df['InternetService'].value_counts())
df['InternetService'].value_counts().plot(kind='pie', autopct='%1.1f%%', colors='myr', shad
```

unique values by internetservice :

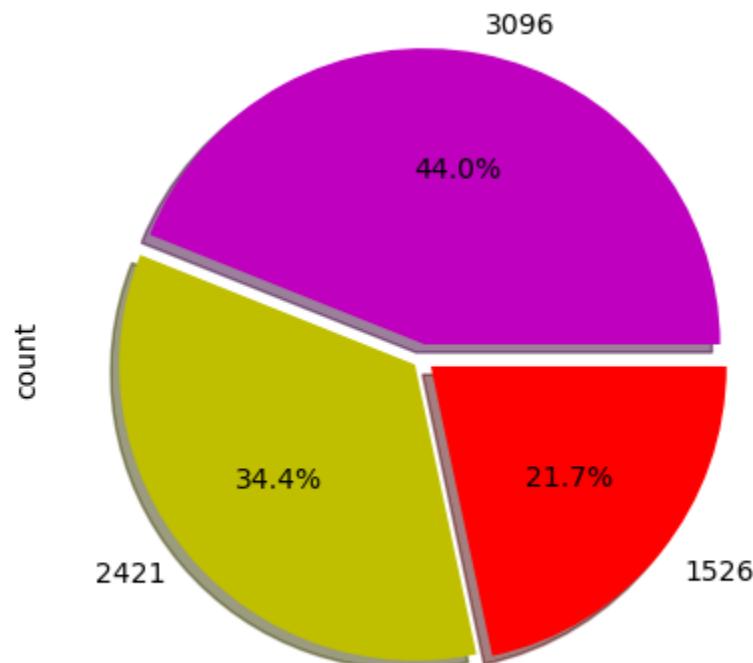
```
['DSL' 'Fiber optic' 'No']
```

unique values by internetservice and its count:

```
3
```

value count by internetservice:

```
InternetService
Fiber optic    3096
DSL           2421
No            1526
Name: count, dtype: int64
Out[12]:
<Axes: ylabel='count'>
```



In [23]:

```
# Get unique values
print("Unique Payment Methods:")
```

```

unique_payments = df['PaymentMethod'].unique()
print(unique_payments)
print("\n" + "*60")

# Number of unique values
print("\nNumber of unique values in each categorical column:")
categorical_cols=df.select_dtypes(include=['object','category']).columns.tolist()
for col in categorical_cols:
    unique_count = df[col].nunique()
    print(f"{col}: {unique_count} unique values")
print("\n" + "*60")

# Unique combinations
print("\nUnique Combinations of Contract and PaperlessBilling:")
unique_combinations = df[['Contract', 'PaperlessBilling']].drop_duplicates()
print(unique_combinations.sort_values(['Contract', 'PaperlessBilling']))

```

Unique Payment Methods:
['Electronic check' 'Mailed check' 'Bank transfer (automatic)'
'Credit card (automatic)']

Number of unique values in each categorical column:
customerID: 7043 unique values
gender: 2 unique values
Partner: 2 unique values
Dependents: 2 unique values
PhoneService: 2 unique values
MultipleLines: 3 unique values
InternetService: 3 unique values
OnlineSecurity: 3 unique values
OnlineBackup: 3 unique values
DeviceProtection: 3 unique values
TechSupport: 3 unique values
StreamingTV: 3 unique values
StreamingMovies: 3 unique values
Contract: 3 unique values
PaperlessBilling: 2 unique values
PaymentMethod: 4 unique values
Churn: 2 unique values

Unique Combinations of Contract and PaperlessBilling:

| | Contract | PaperlessBilling |
|----|----------------|------------------|
| 7 | Month-to-month | No |
| 0 | Month-to-month | Yes |
| 1 | One year | No |
| 54 | One year | Yes |
| 11 | Two year | No |
| 23 | Two year | Yes |

In [55]:

```
print(df.sort_values(by='MonthlyCharges', ascending=False))
```

| | customerID | gender | SeniorCitizen | Partner | Dependents | tenure | \ |
|------|------------|--------|---------------|---------|------------|--------|----|
| 4586 | 7569-NMZQ | Female | | 0 | Yes | Yes | 72 |
| 2115 | 8984-HPEMB | Female | | 0 | No | No | 71 |
| 3894 | 5989-AXPUC | Female | | 0 | Yes | No | 68 |

| | | | | | | |
|------|------------|--------|-----|-----|-----|-----|
| 4804 | 5734-EJKXG | Female | 0 | No | No | 61 |
| 5127 | 8199-ZLLSA | Male | 0 | No | No | 67 |
| ... | ... | ... | ... | ... | ... | ... |
| 6906 | 9945-PSVIP | Female | 0 | Yes | Yes | 25 |
| 1156 | 0621-CXBKL | Female | 0 | No | No | 53 |
| 6652 | 0827-ITJPH | Male | 0 | No | No | 36 |
| 1529 | 9764-REAFF | Female | 0 | Yes | No | 59 |
| 3719 | 6823-SIDFQ | Male | 0 | No | No | 28 |

| | PhoneService | MultipleLines | InternetService | OnlineSecurity | ... | \ |
|------|--------------|---------------|-----------------|---------------------|-----|---|
| 4586 | Yes | Yes | Fiber optic | Yes | ... | |
| 2115 | Yes | Yes | Fiber optic | Yes | ... | |
| 3894 | Yes | Yes | Fiber optic | Yes | ... | |
| 4804 | Yes | Yes | Fiber optic | Yes | ... | |
| 5127 | Yes | Yes | Fiber optic | Yes | ... | |
| ... | ... | ... | ... | ... | ... | |
| 6906 | Yes | No | No | No internet service | ... | |
| 1156 | Yes | No | No | No internet service | ... | |
| 6652 | Yes | No | No | No internet service | ... | |
| 1529 | Yes | No | No | No internet service | ... | |
| 3719 | Yes | No | No | No internet service | ... | |

| | DeviceProtection | TechSupport | StreamingTV | \ |
|------|---------------------|---------------------|---------------------|---|
| 4586 | Yes | Yes | Yes | |
| 2115 | Yes | Yes | Yes | |
| 3894 | Yes | Yes | Yes | |
| 4804 | Yes | Yes | Yes | |
| 5127 | Yes | Yes | Yes | |
| ... | ... | ... | ... | |
| 6906 | No internet service | No internet service | No internet service | |
| 1156 | No internet service | No internet service | No internet service | |
| 6652 | No internet service | No internet service | No internet service | |
| 1529 | No internet service | No internet service | No internet service | |
| 3719 | No internet service | No internet service | No internet service | |

| | StreamingMovies | Contract | PaperlessBilling | \ |
|------|---------------------|----------|------------------|---|
| 4586 | Yes | Two year | Yes | |
| 2115 | Yes | Two year | Yes | |
| 3894 | Yes | Two year | No | |
| 4804 | Yes | One year | Yes | |
| 5127 | Yes | One year | Yes | |
| ... | ... | ... | ... | |
| 6906 | No internet service | Two year | Yes | |
| 1156 | No internet service | Two year | No | |
| 6652 | No internet service | Two year | Yes | |
| 1529 | No internet service | Two year | No | |
| 3719 | No internet service | One year | No | |

| | PaymentMethod | MonthlyCharges | TotalCharges | Churn |
|------|---------------------------|----------------|--------------|-------|
| 4586 | Bank transfer (automatic) | 118.75 | 8672.45 | No |
| 2115 | Electronic check | 118.65 | 8477.60 | No |
| 3894 | Mailed check | 118.60 | 7990.05 | No |
| 4804 | Electronic check | 118.60 | 7365.70 | No |
| 5127 | Bank transfer (automatic) | 118.35 | 7804.15 | Yes |
| ... | ... | ... | ... | |
| 6906 | Mailed check | 18.70 | 383.65 | No |
| 1156 | Mailed check | 18.70 | 1005.70 | No |
| 6652 | Credit card (automatic) | 18.55 | 689.00 | No |
| 1529 | Bank transfer (automatic) | 18.40 | 1057.85 | No |

```
3719    Credit card (automatic)      18.25      534.70      No
```

[7043 rows x 21 columns]

In [9]:

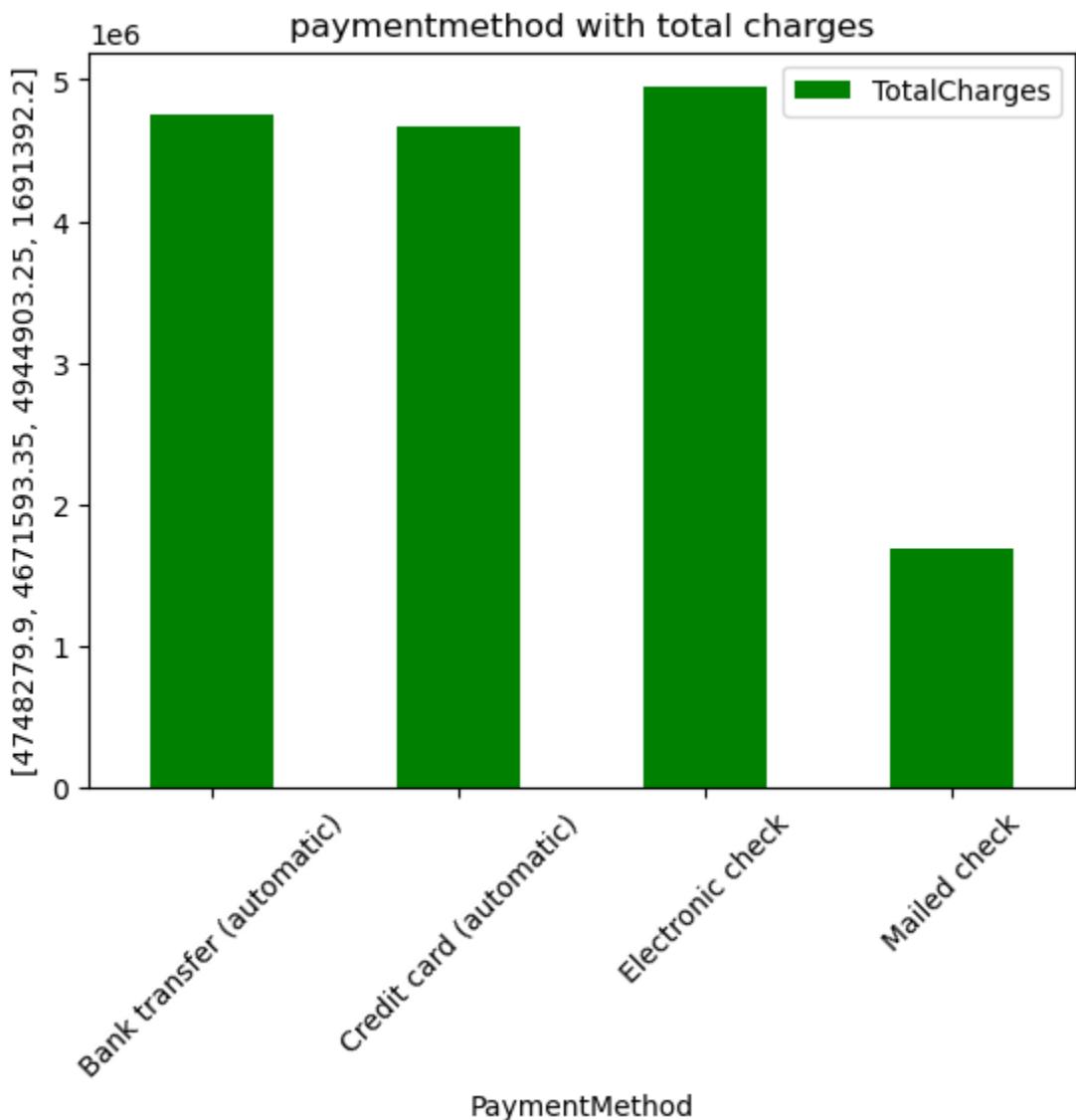
```
print("\n group by paymentmethod with totalcharges :\n")
print(df.groupby('PaymentMethod')['TotalCharges'].sum())
df.groupby('PaymentMethod')['TotalCharges'].sum().plot(kind='bar', rot=45, ylabel=df.groupby('PaymentMethod').sum().index)
plt.title('paymentmethod with total charges')
plt.legend()
```

group by paymentmethod with totalcharges :

```
PaymentMethod
Bank transfer (automatic)      4748279.90
Credit card (automatic)        4671593.35
Electronic check                4944903.25
Mailed check                   1691392.20
Name: TotalCharges, dtype: float64
```

Out[9]:

```
<matplotlib.legend.Legend at 0x1ab97bfe480>
```

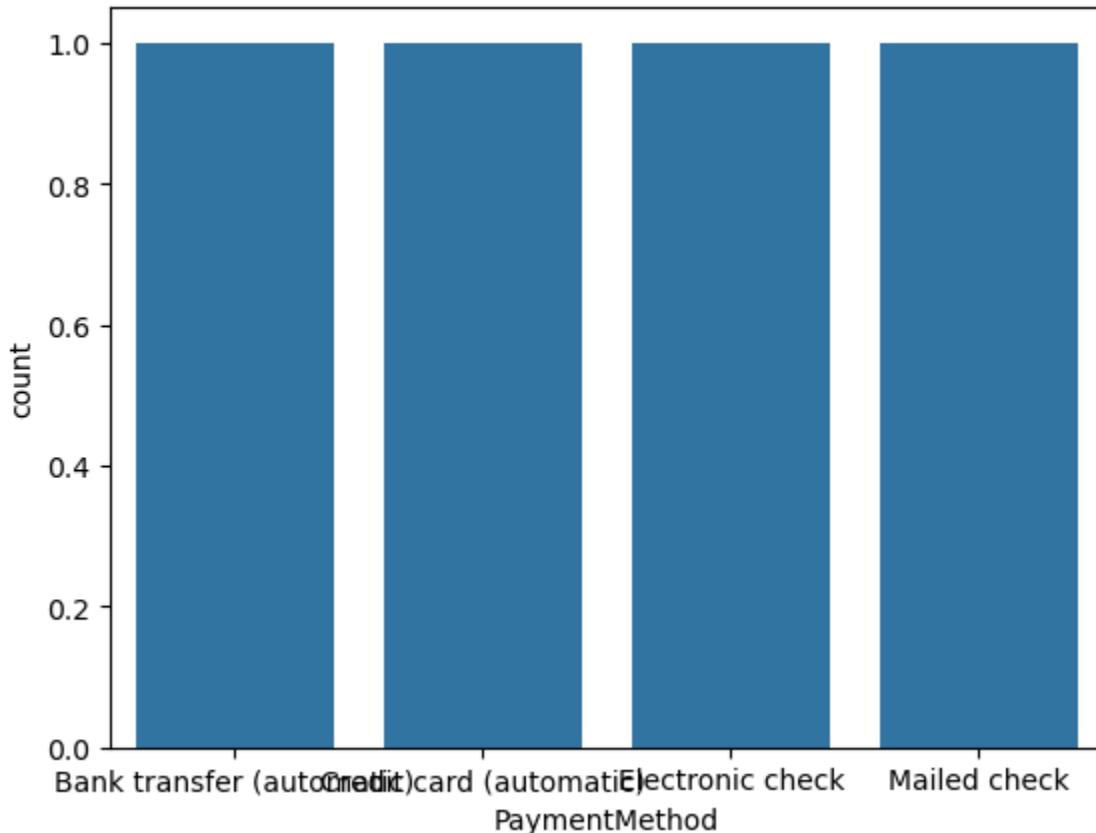


In [14]:

```
import seaborn as sns
sns.countplot(df.groupby('PaymentMethod')['TotalCharges'].sum())
```

Out[14]:

```
<Axes: xlabel='PaymentMethod', ylabel='count'>
```



In [4]:

```
print(df.groupby('PaymentMethod')['TotalCharges'].sum().reset_index())
```

| | PaymentMethod | TotalCharges |
|---|---------------------------|--------------|
| 0 | Bank transfer (automatic) | 4748279.90 |
| 1 | Credit card (automatic) | 4671593.35 |
| 2 | Electronic check | 4944903.25 |
| 3 | Mailed check | 1691392.20 |

In [22]:

```
print("\n group by paymentmethod with totalcharges along with maximum value with name :\\n")
print(df.groupby('PaymentMethod')['TotalCharges'].sum().sort_values(ascending=False).hea
```

group by paymentmethod with totalcharges along with maximum value with name :

```
PaymentMethod
Electronic check    4944903.25
Name: TotalCharges, dtype: float64
```

In [105]:

```
print(df.groupby(['PaymentMethod', 'InternetService'])['TotalCharges'].sum())
```

| PaymentMethod | InternetService | TotalCharges |
|---------------------------|-----------------|--------------|
| Bank transfer (automatic) | DSL | 1655766.90 |
| | Fiber optic | 2783830.65 |
| | No | 308682.35 |
| Credit card (automatic) | DSL | 1730501.05 |
| | Fiber optic | 2647442.45 |
| | No | 293649.85 |
| Electronic check | DSL | 914329.00 |
| | Fiber optic | 3964264.10 |
| | No | 66310.15 |

```
Mailed check          DSL           820813.90
                      Fiber optic   528085.75
                      No            342492.55
```

Name: TotalCharges, dtype: float64

In [9]:

```
# Fix 1: Correct way to handle missing values
df['TotalCharges'] = pd.to_numeric(df['TotalCharges'], errors='coerce')
df['TotalCharges'] = df['TotalCharges'].fillna(df['TotalCharges'].median())

# Fix 2: Create binary churn column
df['Churn_binary'] = df['Churn'].apply(lambda x: 1 if x == 'Yes' else 0)

# Multiple aggregation functions with groupby
groupby_results = df.groupby('Churn').agg({
    'tenure': ['mean', 'median', 'min', 'max', 'count'],
    'MonthlyCharges': ['mean', 'median', 'std'],
    'TotalCharges': ['mean', 'sum']
})

print("GroupBy with Multiple Aggregations:")
print(groupby_results)
print("\n" + "="*60)

# Group by multiple columns
multi_group = df.groupby(['InternetService', 'Contract']).agg({
    'customerID': 'count',
    'MonthlyCharges': 'mean',
    'Churn_binary': 'mean'
}).round(2)

multi_group = multi_group.rename(columns={
    'customerID': 'Customer_Count',
    'MonthlyCharges': 'Avg_Monthly_Charge',
    'Churn_binary': 'Churn_Rate'
})

print("\nGroupBy InternetService and Contract:")
print(multi_group)
```

GroupBy with Multiple Aggregations:

| Churn | tenure | | | | | MonthlyCharges | | | \ |
|-------|-----------|--------|-----|-----|-------|----------------|--------|-----------|---|
| | mean | median | min | max | count | mean | median | std | |
| No | 37.569965 | 38.0 | 0 | 72 | 5174 | 61.265124 | 64.425 | 31.092648 | |
| Yes | 17.979133 | 10.0 | 1 | 72 | 1869 | 74.441332 | 79.650 | 24.666053 | |

| Churn | TotalCharges | |
|-------|--------------|--------------|
| | mean | sum |
| No | 2552.882494 | 1.320861e+07 |
| Yes | 1531.796094 | 2.862927e+06 |

=====

GroupBy InternetService and Contract:

| InternetService | Contract | Customer_Count | Avg_Monthly_Charge | Churn_Rate |
|-----------------|----------------|----------------|--------------------|------------|
| DSL | Month-to-month | 1223 | 50.22 | 0.32 |

| | | | | |
|-------------|----------------|------|--------|------|
| | One year | 570 | 61.40 | 0.09 |
| | Two year | 628 | 70.46 | 0.02 |
| Fiber optic | Month-to-month | 2128 | 87.02 | 0.55 |
| | One year | 539 | 98.78 | 0.19 |
| | Two year | 429 | 104.57 | 0.07 |
| No | Month-to-month | 524 | 20.41 | 0.19 |
| | One year | 364 | 20.82 | 0.02 |
| | Two year | 638 | 21.78 | 0.01 |

In [24]:

```
# Create a pivot table first, then transpose
pivot_complex = pd.pivot_table(
    df,
    values='MonthlyCharges',
    index=['Contract', 'PaperlessBilling'],
    columns='Churn',
    aggfunc=['sum', 'count']
)

print("\nComplex Pivot Table:")
print(pivot_complex)
print("\n" + "="*60)

print("\nTransposed Complex Pivot Table:")
transposed_complex = pivot_complex.T
print(transposed_complex)
```

Complex Pivot Table:

| Churn | Contract | PaperlessBilling | sum | | count | |
|----------|----------------|------------------|----------|----------|-------|------|
| | | | No | Yes | No | Yes |
| One year | Month-to-month | No | 44372.70 | 24910.65 | 883 | 406 |
| | | Yes | 92074.35 | 95936.45 | 1337 | 1249 |
| Two year | Month-to-month | No | 31492.70 | 3673.30 | 625 | 48 |
| | | Yes | 50205.45 | 10445.15 | 682 | 118 |
| Two year | One year | No | 43837.55 | 1036.90 | 895 | 15 |
| | | Yes | 55003.00 | 3128.40 | 752 | 33 |

Transposed Complex Pivot Table:

| Contract | PaperlessBilling | Month-to-month | | One year | | Two year | | \ |
|----------|------------------|----------------|----------|----------|----------|----------|-----|---|
| | | No | Yes | No | Yes | No | Yes | |
| sum | No | 44372.70 | 92074.35 | 31492.7 | 50205.45 | 43837.55 | | |
| | Yes | 24910.65 | 95936.45 | 3673.3 | 10445.15 | 1036.90 | | |
| count | No | 883.00 | 1337.00 | 625.0 | 682.00 | 895.00 | | |
| | Yes | 406.00 | 1249.00 | 48.0 | 118.00 | 15.00 | | |

Contract

| PaperlessBilling | Yes |
|------------------|---------|
| Churn | |
| sum | No |
| | 55003.0 |
| | Yes |
| | 3128.4 |
| count | No |
| | 752.0 |
| | Yes |
| | 33.0 |

In [17]:

```
# Select a subset of columns for melt/unpivot demonstration
melt_df = df[['customerID', 'InternetService', 'Contract', 'PaymentMethod', 'MonthlyChar
```

```

print("Original Data (before melt):")
print(melt_df)
print("\n" + "*60)

# Melt/Unpivot the data
melted = pd.melt(
    melt_df,
    id_vars=['customerID', 'MonthlyCharges'],
    value_vars=['InternetService', 'Contract', 'PaymentMethod'],
    var_name='Service_Category',
    value_name='Service_Value'
)

print("\nMelted/Unpivoted Data:")
print(melted)
print("\n" + "*60)

# More practical melt example
# Create a summary table first, then melt it
summary_df = df.groupby(['InternetService', 'Churn']).size().reset_index(name='Count')

pivot_summary = summary_df.pivot(index='InternetService', columns='Churn', values='Count')
print("\nPivot Summary (before melt):")
print(pivot_summary)

# Melt the pivot table
melted_summary = pivot_summary.reset_index().melt(
    id_vars='InternetService',
    value_vars=['No', 'Yes'],
    var_name='Churn',
    value_name='Customer_Count'
)

print("\nMelted Summary:")
print(melted_summary.sort_values(['InternetService', 'Churn']))

```

Original Data (before melt):

| | customerID | InternetService | Contract | PaymentMethod | \ |
|---|------------|-----------------|----------------|---------------------------|---|
| 0 | 7590-VHVEG | DSL | Month-to-month | Electronic check | |
| 1 | 5575-GNVDE | DSL | One year | Mailed check | |
| 2 | 3668-QPYBK | DSL | Month-to-month | Mailed check | |
| 3 | 7795-CFOCW | DSL | One year | Bank transfer (automatic) | |
| 4 | 9237-HQITU | Fiber optic | Month-to-month | Electronic check | |

MonthlyCharges

| | |
|---|-------|
| 0 | 29.85 |
| 1 | 56.95 |
| 2 | 53.85 |
| 3 | 42.30 |
| 4 | 70.70 |

Melted/Unpivoted Data:

| | customerID | MonthlyCharges | Service_Category | Service_Value |
|---|------------|----------------|------------------|---------------|
| 0 | 7590-VHVEG | 29.85 | InternetService | DSL |
| 1 | 5575-GNVDE | 56.95 | InternetService | DSL |
| 2 | 3668-QPYBK | 53.85 | InternetService | DSL |

```

3  7795-CFOCW      42.30 InternetService          DSL
4  9237-HQITU      70.70 InternetService          Fiber optic
5  7590-VHVEG      29.85 Contract                Month-to-month
6  5575-GNVDE      56.95 Contract                One year
7  3668-QPYBK      53.85 Contract                Month-to-month
8  7795-CFOCW      42.30 Contract                One year
9  9237-HQITU      70.70 Contract                Month-to-month
10 7590-VHVEG      29.85 PaymentMethod          Electronic check
11 5575-GNVDE      56.95 PaymentMethod          Mailed check
12 3668-QPYBK      53.85 PaymentMethod          Mailed check
13 7795-CFOCW      42.30 PaymentMethod          Bank transfer (automatic)
14 9237-HQITU      70.70 PaymentMethod          Electronic check
=====
```

Pivot Summary (before melt):

| Churn | No | Yes |
|-----------------|------|------|
| InternetService | | |
| DSL | 1962 | 459 |
| Fiber optic | 1799 | 1297 |
| No | 1413 | 113 |

Melted Summary:

| | InternetService | Churn | Customer_Count |
|---|-----------------|-------|----------------|
| 0 | DSL | No | 1962 |
| 3 | DSL | Yes | 459 |
| 1 | Fiber optic | No | 1799 |
| 4 | Fiber optic | Yes | 1297 |
| 2 | | No | 1413 |
| 5 | | Yes | 113 |

In [18]:

```

# Basic crosstab
ct_basic = pd.crosstab(df['InternetService'], df['Churn'])
print("Crosstab - InternetService vs Churn:")
print(ct_basic)
print("\n" + "="*60)

# Crosstab with margins and percentages
ct_margins = pd.crosstab(
    df['InternetService'],
    df['Churn'],
    margins=True,
    margins_name='Total',
    normalize='index' # Row percentages
).round(4) * 100

print("\nCrosstab with Percentages (Row %):")
print(ct_margins)
print("\n" + "="*60)

# Multi-dimensional crosstab
ct_multi = pd.crosstab(
    [df['Contract'], df['PaperlessBilling']],
    [df['Churn'], df['SeniorCitizen']],
    margins=True
)

print("\nMulti-dimensional Crosstab:")
```

```

print(ct_multi)
print("\n" + "="*60)

# Crosstab with aggregation
ct_agg = pd.crosstab(
    df['Contract'],
    df['Churn'],
    values=df['MonthlyCharges'],
    aggfunc='mean'
).round(2)

print("\nCrosstab with Average Monthly Charges:")
print(ct_agg)

```

Crosstab - InternetService vs Churn:

| Churn | No | Yes |
|-----------------|------|------|
| InternetService | | |
| DSL | 1962 | 459 |
| Fiber optic | 1799 | 1297 |
| No | 1413 | 113 |

Crosstab with Percentages (Row %):

| Churn | No | Yes |
|-----------------|-------|-------|
| InternetService | | |
| DSL | 81.04 | 18.96 |
| Fiber optic | 58.11 | 41.89 |
| No | 92.60 | 7.40 |
| Total | 73.46 | 26.54 |

Multi-dimensional Crosstab:

| Churn | SeniorCitizen | Contract | PaperlessBilling | No | Yes | All | | |
|----------------|---------------|----------------|------------------|------|-----|------|-----|------|
| | | | | 0 | 1 | 0 | 1 | |
| Month-to-month | No | Month-to-month | No | 795 | 88 | 335 | 71 | 1289 |
| | Yes | | Yes | 1059 | 278 | 879 | 370 | 2586 |
| One year | No | One year | No | 573 | 52 | 43 | 5 | 673 |
| | Yes | | Yes | 573 | 109 | 94 | 24 | 800 |
| Two year | No | Two year | No | 847 | 48 | 13 | 2 | 910 |
| | Yes | | Yes | 661 | 91 | 29 | 4 | 785 |
| All | | | | 4508 | 666 | 1393 | 476 | 7043 |

Crosstab with Average Monthly Charges:

| Churn | No | Yes |
|----------------|-------|-------|
| Contract | | |
| Month-to-month | 61.46 | 73.02 |
| One year | 62.51 | 85.05 |
| Two year | 60.01 | 86.78 |

In [39]:

```

# Check for duplicate rows based on specific columns
print("\nDuplicate Rows based on key service columns:")
key_columns = ['InternetService', 'Contract', 'PaymentMethod', 'MonthlyCharges']
duplicate_rows = df[df.duplicated(subset=key_columns, keep=False)]

```

```

print(f"Found {len(duplicate_rows)} rows with duplicate service combinations")
print(f"That's {len(duplicate_rows)/len(df)*100:.2f}% of total data")
print("\n" + "*60")

# Find and display some duplicate examples
if len(duplicate_rows) > 0:
    print("\nSample Duplicate Rows:")
    sample_duplicates = duplicate_rows.sort_values(key_columns).head(10)
    print(sample_duplicates[['customerID'] + key_columns])

# Remove duplicates (creating a new dataframe for demonstration)
df_no_duplicates = df.drop_duplicates(subset=key_columns)
print(f"\nOriginal shape: {df.shape}")
print(f"After removing service duplicates: {df_no_duplicates.shape}")
print(f"Rows removed: {len(df) - len(df_no_duplicates)}")

```

Duplicate Rows based on key service columns:
 Found 3634 rows with duplicate service combinations
 That's 51.60% of total data

Sample Duplicate Rows:

| | customerID | InternetService | Contract | PaymentMethod | \ |
|------|------------|-----------------|----------------|---------------------------|---|
| 4024 | 1329-VHWNP | DSL | Month-to-month | Bank transfer (automatic) | |
| 5665 | 6345-HOVES | DSL | Month-to-month | Bank transfer (automatic) | |
| 3940 | 1559-DTODC | DSL | Month-to-month | Bank transfer (automatic) | |
| 6204 | 3058-WQDRE | DSL | Month-to-month | Bank transfer (automatic) | |
| 667 | 5533-RJFTJ | DSL | Month-to-month | Bank transfer (automatic) | |
| 4897 | 2533-TIBIX | DSL | Month-to-month | Bank transfer (automatic) | |
| 5329 | 2894-Q0JRX | DSL | Month-to-month | Bank transfer (automatic) | |
| 6522 | 7233-I0QNP | DSL | Month-to-month | Bank transfer (automatic) | |
| 4961 | 6954-00YZZ | DSL | Month-to-month | Bank transfer (automatic) | |
| 5463 | 6142-VSJQ0 | DSL | Month-to-month | Bank transfer (automatic) | |

MonthlyCharges

| | MonthlyCharges |
|------|----------------|
| 4024 | 25.05 |
| 5665 | 25.05 |
| 3940 | 25.15 |
| 6204 | 25.15 |
| 667 | 30.20 |
| 4897 | 30.20 |
| 5329 | 34.00 |
| 6522 | 34.00 |
| 4961 | 44.35 |
| 5463 | 44.35 |

Original shape: (7043, 22)
 After removing service duplicates: (4666, 22)
 Rows removed: 2377

In [25]:

```

# Basic sorting
print("Top 10 Highest Monthly Charges:")
top_charges = df.sort_values('MonthlyCharges', ascending=False).head(10)
print(top_charges[['customerID', 'MonthlyCharges', 'Contract', 'InternetService']])
print("\n" + "*60")

# Sorting by multiple columns

```

```

print("\nCustomers sorted by Tenure (desc) and Monthly Charges (desc):")
multi_sort = df.sort_values(['tenure', 'MonthlyCharges'], ascending=[False, False]).head
print(multi_sort[['customerID', 'tenure', 'MonthlyCharges', 'Churn']])
print("\n" + "*60)

# Sorting with groupby results
group_sorted = df.groupby('InternetService').agg({
    'MonthlyCharges': 'mean',
    'Churn_binary': 'mean',
    'customerID': 'count'
}).round(2)

group_sorted = group_sorted.rename(columns={
    'MonthlyCharges': 'Avg_Monthly_Charge',
    'Churn_binary': 'Churn_Rate',
    'customerID': 'Customer_Count'
})

print("\nInternet Service Analysis (sorted by Churn Rate):")
sorted_by_churn = group_sorted.sort_values('Churn_Rate', ascending=False)
print(sorted_by_churn)
print("\n" + "*60)

# Sorting with custom order (using categorical)
contract_order = ['Month-to-month', 'One year', 'Two year']
df['Contract_ordered'] = pd.Categorical(df['Contract'], categories=contract_order, order=True)

print("\nData sorted by Custom Contract Order:")
contract_sorted = df.sort_values('Contract_ordered').head(10)
print(contract_sorted[['customerID', 'Contract', 'MonthlyCharges']])

```

Top 10 Highest Monthly Charges:

| | customerID | MonthlyCharges | Contract | InternetService |
|------|------------|----------------|----------------|-----------------|
| 4586 | 7569-NMZYQ | 118.75 | Two year | Fiber optic |
| 2115 | 8984-HPEMB | 118.65 | Two year | Fiber optic |
| 3894 | 5989-AXPUC | 118.60 | Two year | Fiber optic |
| 4804 | 5734-EJKXG | 118.60 | One year | Fiber optic |
| 5127 | 8199-ZLLSA | 118.35 | One year | Fiber optic |
| 6118 | 9924-JPRMC | 118.20 | Two year | Fiber optic |
| 4610 | 2889-FPWRM | 117.80 | One year | Fiber optic |
| 3205 | 3810-DVDQQ | 117.60 | Two year | Fiber optic |
| 6768 | 9739-JLPQJ | 117.50 | Two year | Fiber optic |
| 4875 | 2302-ANTDP | 117.45 | Month-to-month | Fiber optic |

=====

Customers sorted by Tenure (desc) and Monthly Charges (desc):

| | customerID | tenure | MonthlyCharges | Churn |
|------|------------|--------|----------------|-------|
| 4586 | 7569-NMZYQ | 72 | 118.75 | No |
| 6118 | 9924-JPRMC | 72 | 118.20 | No |
| 4610 | 2889-FPWRM | 72 | 117.80 | Yes |
| 3205 | 3810-DVDQQ | 72 | 117.60 | No |
| 6768 | 9739-JLPQJ | 72 | 117.50 | No |
| 4155 | 6904-JLBGY | 72 | 117.35 | No |
| 2368 | 6650-BWFRT | 72 | 117.15 | No |
| 5347 | 9788-HNGUT | 72 | 116.95 | No |
| 2025 | 1488-PBLJN | 72 | 116.85 | No |
| 4206 | 0017-IUDMW | 72 | 116.80 | No |

```
=====
Internet Service Analysis (sorted by Churn Rate):
      Avg_Monthly_Charge  Churn_Rate  Customer_Count
InternetService
Fiber optic                 91.50       0.42          3096
DSL                         58.10       0.19          2421
No                           21.08       0.07          1526
=====
```

Data sorted by Custom Contract Order:

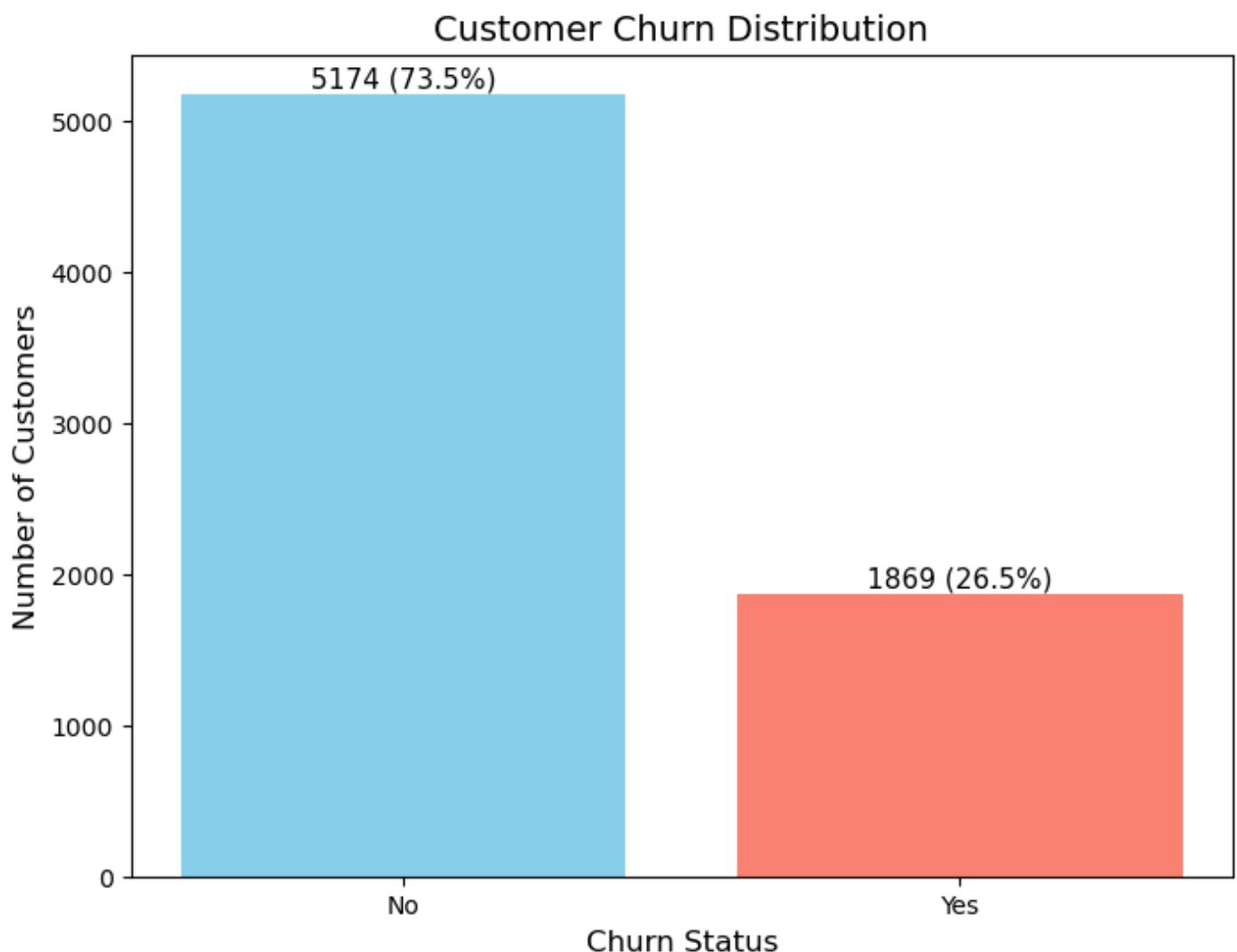
| | customerID | Contract | MonthlyCharges |
|------|------------|----------------|----------------|
| 0 | 7590-VHVEG | Month-to-month | 29.85 |
| 5322 | 8731-WBBMB | Month-to-month | 81.90 |
| 5317 | 1213-NGCUN | Month-to-month | 49.65 |
| 5315 | 2082-CEFLT | Month-to-month | 45.60 |
| 2856 | 2740-TVLFN | Month-to-month | 50.15 |
| 5313 | 1935-IMVBB | Month-to-month | 89.70 |
| 2858 | 9512-PHSMG | Month-to-month | 20.55 |
| 5312 | 9564-KCLHR | Month-to-month | 51.25 |
| 2860 | 2452-KDRRH | Month-to-month | 101.40 |
| 2861 | 2004-OCQXK | Month-to-month | 81.95 |

In [38]:

```
# Calculate overall churn rate
churn_counts = df['Churn'].value_counts()
churn_rate = (churn_counts['Yes'] / len(df)) * 100

plt.figure(figsize=(8, 6))
plt.bar(churn_counts.index, churn_counts.values, color=['skyblue', 'salmon'])
plt.title('Customer Churn Distribution', fontsize=14)
plt.xlabel('Churn Status', fontsize=12)
plt.ylabel('Number of Customers', fontsize=12)
plt.text(0, churn_counts['No'] + 50, f"{churn_counts['No']} ({100-churn_rate:.1f}%)",
         ha='center', fontsize=11)
plt.text(1, churn_counts['Yes'] + 50, f"{churn_counts['Yes']} ({churn_rate:.1f}%)",
         ha='center', fontsize=11)
plt.show()

print(f"Overall Churn Rate: {churn_rate:.2f}%")
```



Overall Churn Rate: 26.54%

In [8]:

```
# Compare numerical features for churned vs non-churned customers
churned = df[df['Churn'] == 'Yes']
not_churned = df[df['Churn'] == 'No']

fig, axes = plt.subplots(1, 3, figsize=(15, 5))

# Tenure comparison
axes[0].hist([not_churned['tenure'], churned['tenure']],
            bins=20, label=['Not Churned', 'Churned'],
            alpha=0.7, color=['blue', 'red'])
axes[0].set_title('Tenure Distribution')
axes[0].set_xlabel('Tenure (months)')
axes[0].set_ylabel('Count')
axes[0].legend()

# Monthly Charges comparison
axes[1].hist([not_churned['MonthlyCharges'], churned['MonthlyCharges']],
            bins=20, label=['Not Churned', 'Churned'],
            alpha=0.7, color=['blue', 'red'])
axes[1].set_title('Monthly Charges Distribution')
axes[1].set_xlabel('Monthly Charges ($)')
axes[1].set_ylabel('Count')
axes[1].legend()
```

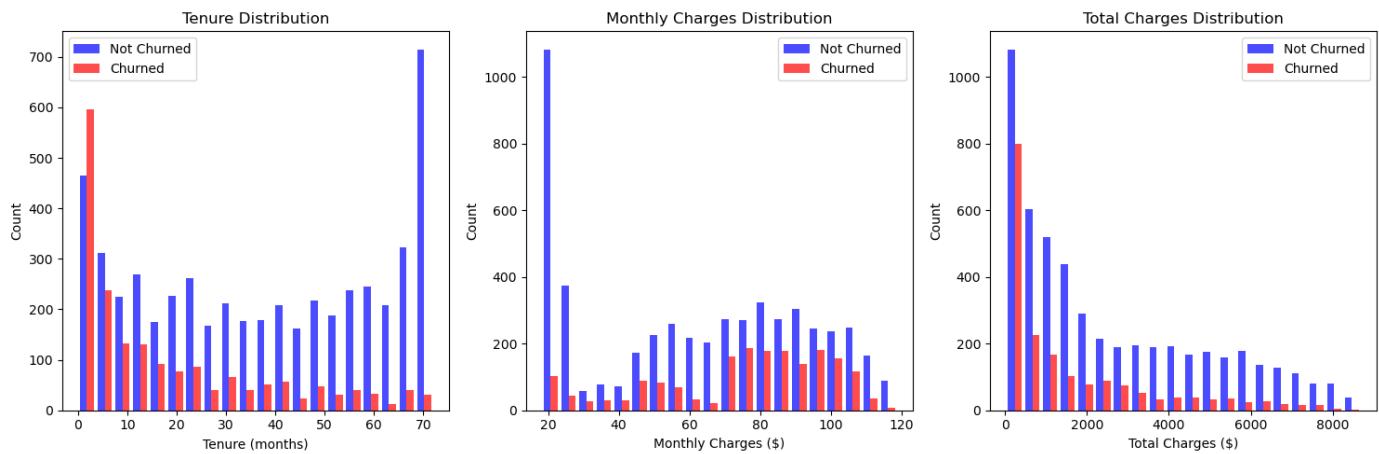
```

# Total Charges comparison
axes[2].hist([not_churned['TotalCharges'], churned['TotalCharges']],
            bins=20, label=['Not Churned', 'Churned'],
            alpha=0.7, color=['blue', 'red'])
axes[2].set_title('Total Charges Distribution')
axes[2].set_xlabel('Total Charges ($)')
axes[2].set_ylabel('Count')
axes[2].legend()

plt.tight_layout()
plt.show()

# Calculate average values
print("\nAverage Values Comparison:")
print(f"Average Tenure - Churned: {churned['tenure'].mean():.1f} months")
print(f"Average Tenure - Not Churned: {not_churned['tenure'].mean():.1f} months")
print(f"Average Monthly Charges - Churned: ${churned['MonthlyCharges'].mean():.2f}")
print(f"Average Monthly Charges - Not Churned: ${not_churned['MonthlyCharges'].mean():.2f}")

```



Average Values Comparison:
 Average Tenure - Churned: 18.0 months
 Average Tenure - Not Churned: 37.6 months
 Average Monthly Charges - Churned: \$74.44
 Average Monthly Charges - Not Churned: \$61.27

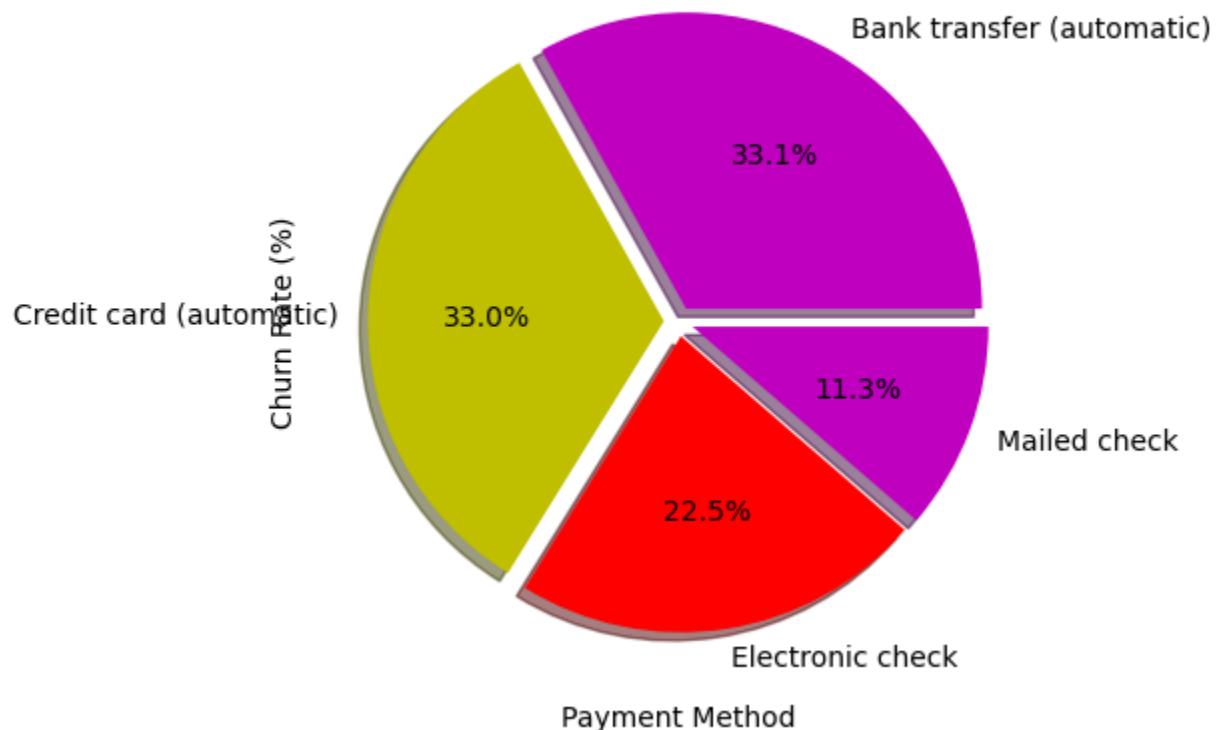
In [14]:

```

# Churn by Payment Method
payment_churn = df.groupby('PaymentMethod')['TotalCharges'].mean() * 100
plt.figure(figsize=(6, 6))
payment_churn.sort_values(ascending=False).plot(kind='pie', autopct='%1.1f%%', colors='myr'
plt.title('Churn Rate by Payment Method')
plt.xlabel('Payment Method')
plt.ylabel('Churn Rate (%)')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()

```

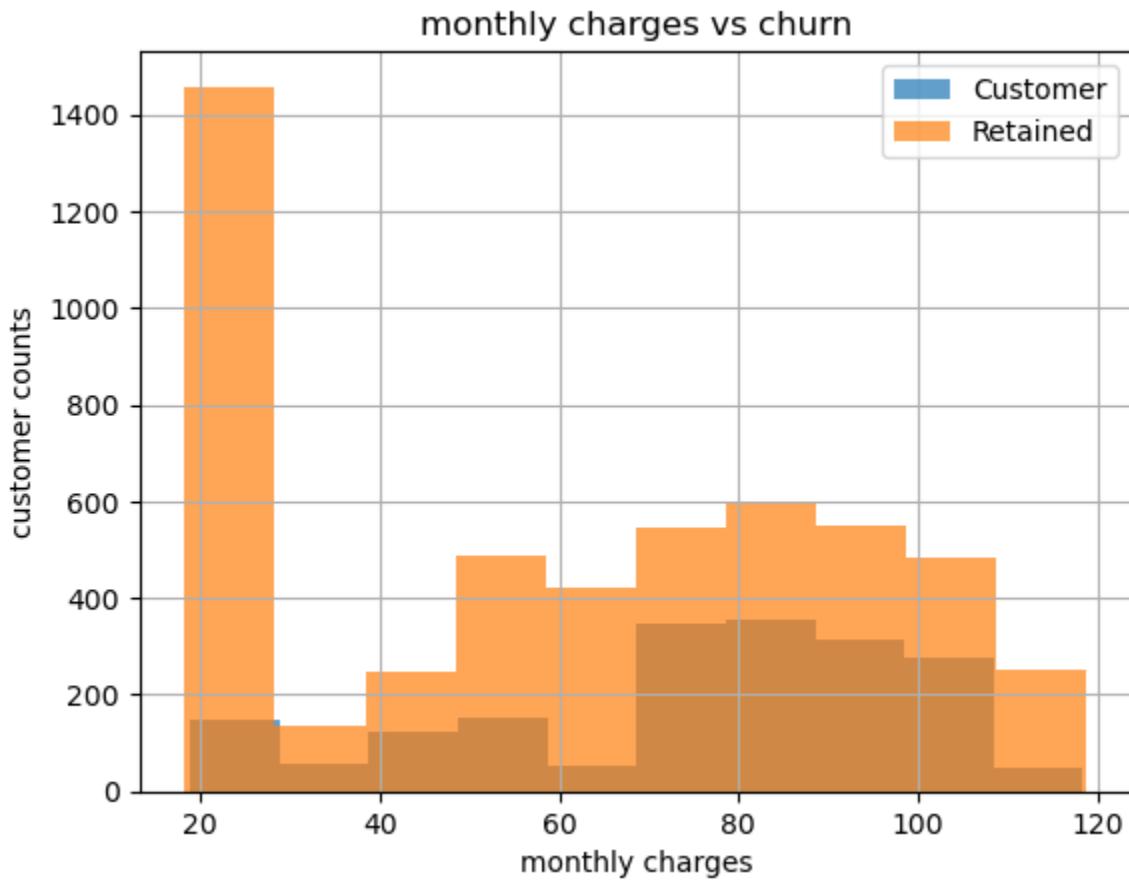
Churn Rate by Payment Method



does monthly charge impact churn

In [37]:

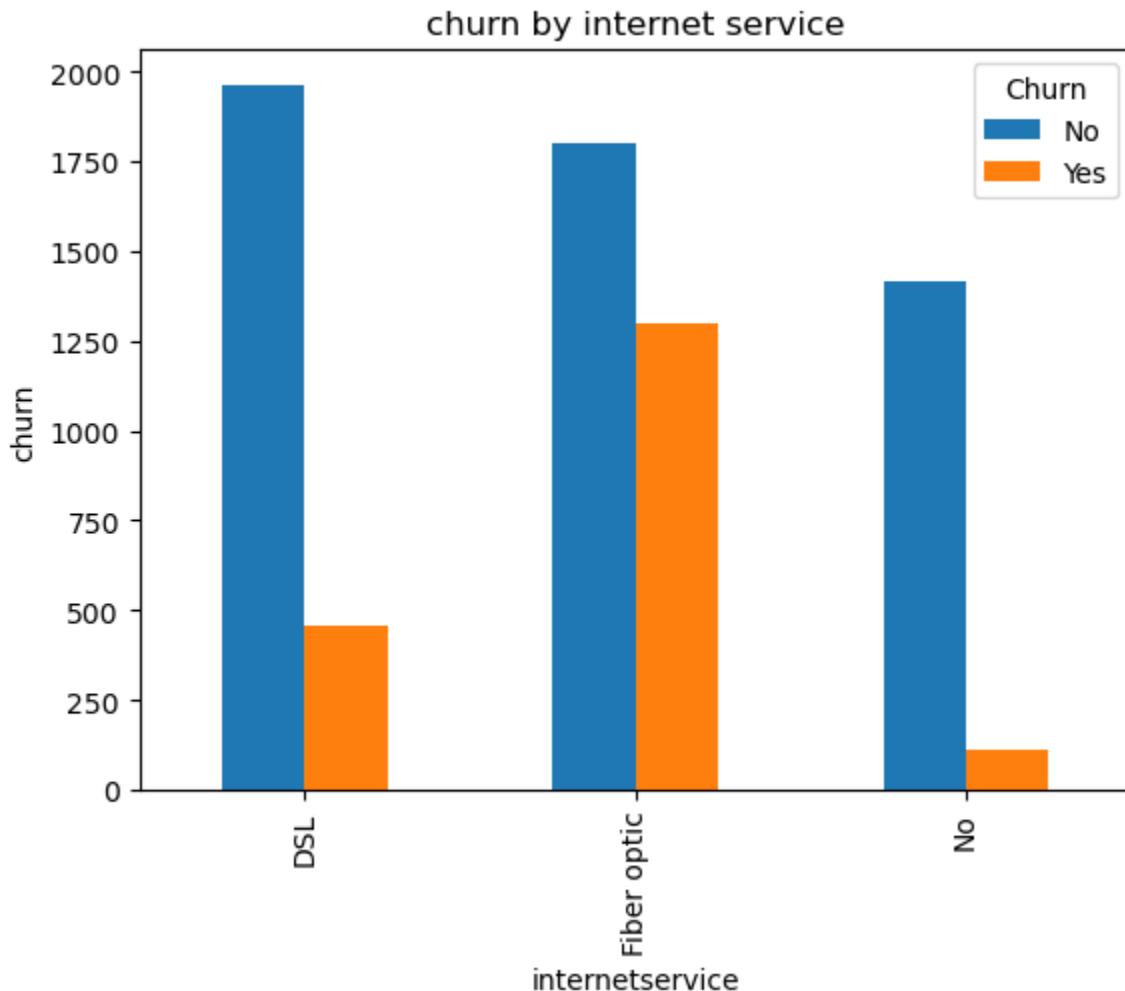
```
plt.figure()
df[df['Churn']=='Yes']['MonthlyCharges'].hist(alpha=0.7)
df[df['Churn']=='No']['MonthlyCharges'].hist(alpha=0.7)
plt.legend(['Customer','Retained'])
plt.xlabel('monthly charges')
plt.ylabel('customer counts')
plt.title('monthly charges vs churn')
plt.show()
```



which internet service has highest churn

In [52]:

```
internet_churn=df.groupby('InternetService')['Churn'].value_counts().unstack()
internet_churn.plot(kind='bar')
plt.title('churn by internet service')
plt.xlabel('internetservice')
plt.ylabel('churn')
plt.show()
```



In [10]:

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

# Load and prepare data
df = pd.read_csv("C:\\\\Users\\\\prems\\\\Videos\\\\power bi\\\\Telco_Cusomer_Churn.csv")

# Data preparation
df['TotalCharges'] = pd.to_numeric(df['TotalCharges'], errors='coerce')
df['TotalCharges'] = df['TotalCharges'].fillna(df['TotalCharges'].median())
df['Churn_binary'] = df['Churn'].apply(lambda x: 1 if x == 'Yes' else 0)
churned = df[df['Churn'] == 'Yes']
not_churned = df[df['Churn'] == 'No']

# Create figure with all visualizations
plt.figure(figsize=(22, 30))

# ===== 1. CHURN RATE BY INTERNET SERVICE =====
plt.subplot(5, 4, 1)
churn_by_internet = df.groupby('InternetService')['Churn_binary'].mean() * 100
bars1 = plt.bar(churn_by_internet.index, churn_by_internet.values,
                 color=['#FF9999', '#66B2FF', '#99FF99'])
plt.title('1. Churn Rate by Internet Service', fontsize=12, fontweight='bold')
plt.xlabel('Internet Service')
plt.ylabel('Churn Rate (%)')
plt.ylim(0, 50)
plt.grid(True, alpha=0.3)

```

```

for i, v in enumerate(churn_by_internet.values):
    plt.text(i, v + 1, f'{v:.1f}%', ha='center', fontsize=10)

# ===== 2. CHURN RATE BY CONTRACT TYPE =====
plt.subplot(5, 4, 2)
churn_by_contract = df.groupby('Contract')['Churn_binary'].mean() * 100
bars2 = plt.bar(churn_by_contract.index, churn_by_contract.values,
                 color=['#FFCC99', '#CC99FF', '#99CCFF'])
plt.title('2. Churn Rate by Contract Type', fontsize=12, fontweight='bold')
plt.xlabel('Contract Type')
plt.ylabel('Churn Rate (%)')
plt.ylim(0, 50)
plt.xticks(rotation=45)
plt.grid(True, alpha=0.3)
for i, v in enumerate(churn_by_contract.values):
    plt.text(i, v + 1, f'{v:.1f}%', ha='center', fontsize=10)

# ===== 3. TENURE DISTRIBUTION COMPARISON =====
plt.subplot(5, 4, 3)
plt.hist([not_churned['tenure'], churned['tenure']],
         bins=15, label=['Not Churned', 'Churned'],
         alpha=0.7, color=['#66CC66', '#FF6666'], edgecolor='black')
plt.title('3. Tenure Distribution', fontsize=12, fontweight='bold')
plt.xlabel('Tenure (months)')
plt.ylabel('Count')
plt.legend()
plt.grid(True, alpha=0.3)

# Add text with averages
avg_tenure_churned = churned['tenure'].mean()
avg_tenure_not = not_churned['tenure'].mean()
plt.text(0.05, 0.95, f'Avg Churned: {avg_tenure_churned:.1f} months\nAvg Not Churned: {a
    transform=plt.gca().transAxes, fontsize=9,
    verticalalignment='top', bbox=dict(boxstyle='round', facecolor='wheat', alpha=0

# ===== 4. MONTHLY CHARGES COMPARISON =====
plt.subplot(5, 4, 4)
plt.hist([not_churned['MonthlyCharges'], churned['MonthlyCharges']],
         bins=20, label=['Not Churned', 'Churned'],
         alpha=0.7, color=['#66CC66', '#FF6666'], edgecolor='black')
plt.title('4. Monthly Charges Distribution', fontsize=12, fontweight='bold')
plt.xlabel('Monthly Charges ($)')
plt.ylabel('Count')
plt.legend()
plt.grid(True, alpha=0.3)

# Add text with averages
avg_monthly_churned = churned['MonthlyCharges'].mean()
avg_monthly_not = not_churned['MonthlyCharges'].mean()
plt.text(0.05, 0.95, f'Avg Churned: ${avg_monthly_churned:.2f}\nAvg Not Churned: ${avg_m
    transform=plt.gca().transAxes, fontsize=9,
    verticalalignment='top', bbox=dict(boxstyle='round', facecolor='wheat', alpha=0

# ===== 5. CHURN RATE BY PAYMENT METHOD =====
plt.subplot(5, 4, 5)
payment_churn = df.groupby('PaymentMethod')['Churn_binary'].mean() * 100
sorted_payments = payment_churn.sort_values(ascending=False)
bars5 = plt.bar(range(len(sorted_payments)), sorted_payments.values,
                color=['#FF9999', '#66B2FF', '#99FF99', '#FFCC66'])

```

```

plt.title('5. Churn Rate by Payment Method', fontsize=12, fontweight='bold')
plt.xlabel('Payment Method')
plt.ylabel('Churn Rate (%)')
plt.xticks(range(len(sorted_payments)), sorted_payments.index, rotation=45, ha='right')
plt.ylim(0, 50)
plt.grid(True, alpha=0.3)
for i, v in enumerate(sorted_payments.values):
    plt.text(i, v + 1, f'{v:.1f}%', ha='center', fontsize=9)

# ===== 6. CHURN RATE BY SENIOR CITIZEN STATUS =====
plt.subplot(5, 4, 6)
senior_churn = df.groupby('SeniorCitizen')['Churn_binary'].mean() * 100
bars6 = plt.bar(['Not Senior', 'Senior'], senior_churn.values,
               color=['#66CCCC', '#FF9966'])
plt.title('6. Churn Rate by Senior Citizen', fontsize=12, fontweight='bold')
plt.xlabel('Senior Citizen Status')
plt.ylabel('Churn Rate (%)')
plt.ylim(0, 35)
plt.grid(True, alpha=0.3)
for i, v in enumerate(senior_churn.values):
    plt.text(i, v + 1, f'{v:.1f}%', ha='center', fontsize=10)

# ===== 7. ADD-ON SERVICES ADOPTION =====
plt.subplot(5, 4, 7)
services = ['OnlineSecurity', 'OnlineBackup', 'DeviceProtection',
            'TechSupport', 'StreamingTV', 'StreamingMovies']

service_adoption = []
for service in services:
    churned_adoption = (churned[service] == 'Yes').mean() * 100
    not_churned_adoption = (not_churned[service] == 'Yes').mean() * 100
    service_adoption.append([churned_adoption, not_churned_adoption])

service_adoption = np.array(service_adoption)
x = np.arange(len(services))
width = 0.35

plt.bar(x - width/2, service_adoption[:, 0], width, label='Churned', color='#FF6666')
plt.bar(x + width/2, service_adoption[:, 1], width, label='Not Churned', color='#66CC66')
plt.title('7. Add-on Services Adoption', fontsize=12, fontweight='bold')
plt.xlabel('Service')
plt.ylabel('Adoption Rate (%)')
plt.xticks(x, services, rotation=45, ha='right')
plt.legend()
plt.grid(True, alpha=0.3)

# ===== 8. PAPERLESS BILLING + CONTRACT CHURN =====
plt.subplot(5, 4, 8)
# Create pivot table
pivot_churn = df.pivot_table(values='Churn_binary',
                              index='Contract',
                              columns='PaperlessBilling',
                              aggfunc='mean') * 100

x = np.arange(len(pivot_churn.index))
width = 0.35

plt.bar(x - width/2, pivot_churn['No'].values, width, label='Paperless: No', color='#66B'
plt.bar(x + width/2, pivot_churn['Yes'].values, width, label='Paperless: Yes', color='F

```

```

plt.title('8. Churn: Contract × Paperless Billing', fontsize=12, fontweight='bold')
plt.xlabel('Contract Type')
plt.ylabel('Churn Rate (%)')
plt.xticks(x, pivot_churn.index, rotation=45, ha='right')
plt.legend()
plt.grid(True, alpha=0.3)

# ===== 9. TOTAL CHARGES VS TENURE SCATTER =====
plt.subplot(5, 4, 9)
# Sample for better visualization
sample_df = df.sample(n=300, random_state=42)

plt.scatter(sample_df[sample_df['Churn'] == 'No']['tenure'],
            sample_df[sample_df['Churn'] == 'No']['TotalCharges'],
            alpha=0.6, c='#66CC66', label='Not Churned', s=30)

plt.scatter(sample_df[sample_df['Churn'] == 'Yes']['tenure'],
            sample_df[sample_df['Churn'] == 'Yes']['TotalCharges'],
            alpha=0.6, c='FF6666', label='Churned', s=30)

plt.title('9. Total Charges vs Tenure', fontsize=12, fontweight='bold')
plt.xlabel('Tenure (months)')
plt.ylabel('Total Charges ($)')
plt.legend()
plt.grid(True, alpha=0.3)

# ===== 10. SERVICE BUNDLES CHURN RATE =====
plt.subplot(5, 4, 10)
# Create service bundles
def create_service_bundle(row):
    services_count = 0
    for service in ['PhoneService', 'StreamingTV', 'StreamingMovies']:
        if row[service] == 'Yes':
            services_count += 1
    if row['InternetService'] != 'No':
        services_count += 1
    return services_count

df['ServiceBundle'] = df.apply(create_service_bundle, axis=1)
bundle_churn = df.groupby('ServiceBundle')['Churn_binary'].mean() * 100

plt.bar(bundle_churn.index, bundle_churn.values,
        color=['#FF9999', '#66B2FF', '#99FF99', '#FFCC66', '#CC99FF'])
plt.title('10. Churn Rate by Service Bundle Size', fontsize=12, fontweight='bold')
plt.xlabel('Number of Services')
plt.ylabel('Churn Rate (%)')
plt.xticks(range(len(bundle_churn)))
plt.grid(True, alpha=0.3)

for i, v in enumerate(bundle_churn.values):
    plt.text(i, v + 1, f'{v:.1f}%', ha='center', fontsize=9)

# ===== 11. OVERALL CHURN DISTRIBUTION =====
plt.subplot(5, 4, 11)
churn_counts = df['Churn'].value_counts()
colors = ['#66CC66', '#FF6666']
plt.pie(churn_counts.values, labels=['Not Churned', 'Churned'],
        colors=colors, autopct='%.1f%%', startangle=90)
plt.title('11. Overall Churn Distribution', fontsize=12, fontweight='bold')

```

```

# ===== 12. CHURN BY DEPENDENTS =====
plt.subplot(5, 4, 12)
dependents_churn = df.groupby('Dependents')['Churn_binary'].mean() * 100
bars12 = plt.bar(['No Dependents', 'Has Dependents'], dependents_churn.values,
                 color=['#FF9999', '#66B2FF'])
plt.title('12. Churn Rate by Dependents', fontsize=12, fontweight='bold')
plt.xlabel('Dependents Status')
plt.ylabel('Churn Rate (%)')
plt.ylim(0, 35)
plt.grid(True, alpha=0.3)
for i, v in enumerate(dependents_churn.values):
    plt.text(i, v + 1, f'{v:.1f}%', ha='center', fontsize=10)

# ===== 13. CHURN BY PARTNER STATUS =====
plt.subplot(5, 4, 13)
partner_churn = df.groupby('Partner')['Churn_binary'].mean() * 100
bars13 = plt.bar(['No Partner', 'Has Partner'], partner_churn.values,
                 color=['#99FF99', '#FFCC66'])
plt.title('13. Churn Rate by Partner Status', fontsize=12, fontweight='bold')
plt.xlabel('Partner Status')
plt.ylabel('Churn Rate (%)')
plt.ylim(0, 35)
plt.grid(True, alpha=0.3)
for i, v in enumerate(partner_churn.values):
    plt.text(i, v + 1, f'{v:.1f}%', ha='center', fontsize=10)

# ===== 14. MONTHLY CHARGES BY CONTRACT =====
plt.subplot(5, 4, 14)
contract_charges = df.groupby('Contract')['MonthlyCharges'].mean().sort_values()
plt.bar(contract_charges.index, contract_charges.values,
        color=['#FF9999', '#66B2FF', '#99FF99'])
plt.title('14. Avg Monthly Charges by Contract', fontsize=12, fontweight='bold')
plt.xlabel('Contract Type')
plt.ylabel('Avg Monthly Charges ($)')
plt.xticks(rotation=45, ha='right')
plt.grid(True, alpha=0.3)

for i, v in enumerate(contract_charges.values):
    plt.text(i, v + 1, f'${v:.1f}', ha='center', fontsize=9)

# ===== 15. TENURE BY CONTRACT TYPE =====
plt.subplot(5, 4, 15)
tenure_by_contract = df.groupby('Contract')['tenure'].mean()
bars15 = plt.bar(tenure_by_contract.index, tenure_by_contract.values,
                 color=['#FF9999', '#66B2FF', '#99FF99'])
plt.title('15. Avg Tenure by Contract Type', fontsize=12, fontweight='bold')
plt.xlabel('Contract Type')
plt.ylabel('Avg Tenure (months)')
plt.xticks(rotation=45, ha='right')
plt.grid(True, alpha=0.3)

for i, v in enumerate(tenure_by_contract.values):
    plt.text(i, v + 1, f'{v:.1f}', ha='center', fontsize=9)

# ===== 16. CHURN HEATMAP (SIMULATED) =====
plt.subplot(5, 4, 16)
# Create correlation matrix for key numerical features
numerical_cols = ['tenure', 'MonthlyCharges', 'TotalCharges', 'Churn_binary']

```

```

corr_matrix = df[numerical_cols].corr()

# Create heatmap using imshow
plt.imshow(corr_matrix, cmap='coolwarm', aspect='auto')
plt.colorbar(label='Correlation')
plt.title('16. Feature Correlation Heatmap', fontsize=12, fontweight='bold')
plt.xticks(range(len(numerical_cols)), numerical_cols, rotation=45)
plt.yticks(range(len(numerical_cols)), numerical_cols)

# Add correlation values
for i in range(len(numerical_cols)):
    for j in range(len(numerical_cols)):
        plt.text(j, i, f'{corr_matrix.iloc[i, j]:.2f}', ha='center', va='center',
                 color='white' if abs(corr_matrix.iloc[i, j]) > 0.5 else 'black',
                 fontsize=9)

# ===== 17. CUSTOMER DISTRIBUTION BY CONTRACT =====
plt.subplot(5, 4, 17)
contract_dist = df['Contract'].value_counts()
colors17 = ['#FF9999', '#66B2FF', '#99FF99']
plt.pie(contract_dist.values, labels=contract_dist.index,
         colors=colors17, autopct='%.1f%%', startangle=90)
plt.title('17. Customer Distribution by Contract', fontsize=12, fontweight='bold')

# ===== 18. CHURN RATE BY GENDER =====
plt.subplot(5, 4, 18)
gender_churn = df.groupby('gender')['Churn_binary'].mean() * 100
bars18 = plt.bar(gender_churn.index, gender_churn.values,
                  color=['#FFB6C1', '#ADD8E6']) # Pink for Female, Light Blue for Male
plt.title('18. Churn Rate by Gender', fontsize=12, fontweight='bold')
plt.xlabel('Gender')
plt.ylabel('Churn Rate (%)')
plt.ylim(0, 30)
plt.grid(True, alpha=0.3)
for i, v in enumerate(gender_churn.values):
    plt.text(i, v + 0.5, f'{v:.1f}%', ha='center', fontsize=10)

# ===== 19. TOP 5 HIGHEST CHURN COHORTS =====
plt.subplot(5, 4, 19)
# Create cohorts based on multiple factors
cohort_data = []
cohorts = [
    ('Fiber Optic', 'Month-to-month'),
    ('Fiber Optic', 'One year'),
    ('DSL', 'Month-to-month'),
    ('No Internet', 'Month-to-month'),
    ('Fiber Optic', 'Two year')
]

for internet, contract in cohorts:
    cohort_df = df[(df['InternetService'] == internet) & (df['Contract'] == contract)]
    if len(cohort_df) > 0:
        churn_rate = cohort_df['Churn_binary'].mean() * 100
        cohort_data.append((f'{internet}\n{contract}', churn_rate, len(cohort_df)))

# Sort by churn rate
cohort_data.sort(key=lambda x: x[1], reverse=True)
top_5 = cohort_data[:5]

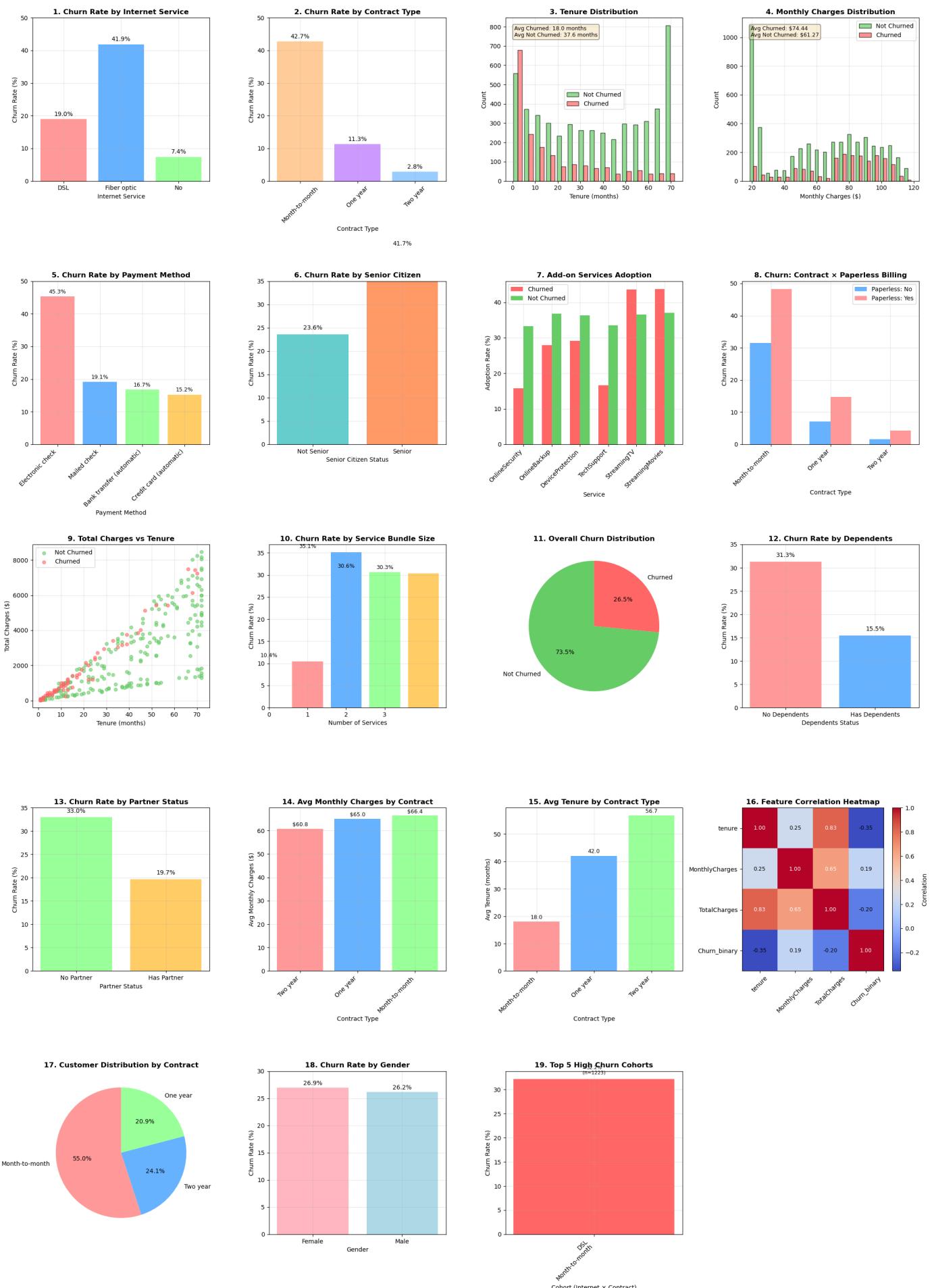
```

```
cohort_names = [x[0] for x in top_5]
churn_rates = [x[1] for x in top_5]
sizes = [x[2] for x in top_5]

bars19 = plt.bar(cohort_names, churn_rates, color="#FF6666")
plt.title('19. Top 5 High Churn Cohorts', fontsize=12, fontweight='bold')
plt.xlabel('Cohort (Internet × Contract)')
plt.ylabel('Churn Rate (%)')
plt.xticks(rotation=45, ha='right')
plt.grid(True, alpha=0.3)

for i, (v, s) in enumerate(zip(churn_rates, sizes)):
    plt.text(i, v + 1, f'{v:.1f}%\n(n={s})', ha='center', fontsize=8)

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



In [11]:

```

import pandas as pd
import matplotlib.pyplot as plt

# Load and prepare data
df = pd.read_csv("C:\\\\Users\\\\prems\\\\Videos\\\\power bi\\\\Telco_Cusomer_Churn.csv")

# Simple data prep
df['TotalCharges'] = pd.to_numeric(df['TotalCharges'], errors='coerce')
df['TotalCharges'] = df['TotalCharges'].fillna(0)
df['Churn_binary'] = df['Churn'].map({'Yes': 1, 'No': 0})

# Create 6 simple visualizations
fig, axes = plt.subplots(2, 3, figsize=(15, 8))
fig.suptitle('Telco Customer Churn Analysis', fontsize=16, fontweight='bold')

# 1. Overall Churn Rate
churn_counts = df['Churn'].value_counts()
axes[0,0].pie(churn_counts.values, labels=['Not Churned', 'Churned'],
              colors=['lightgreen', 'lightcoral'], autopct='%1.1f%%')
axes[0,0].set_title('Overall Churn Rate')

# 2. Churn by Contract
contract_churn = df.groupby('Contract')['Churn_binary'].mean() * 100
axes[0,1].bar(contract_churn.index, contract_churn.values,
              color=['lightcoral', 'gold', 'lightgreen'])
axes[0,1].set_title('Churn Rate by Contract')
axes[0,1].set_ylabel('Churn Rate (%)')
axes[0,1].tick_params(axis='x', rotation=45)
# Add labels
for i, v in enumerate(contract_churn.values):
    axes[0,1].text(i, v + 0.5, f'{v:.1f}%', ha='center', fontsize=9)

# 3. Monthly Charges Comparison
churned_avg = df[df['Churn'] == 'Yes']['MonthlyCharges'].mean()
not_churned_avg = df[df['Churn'] == 'No']['MonthlyCharges'].mean()
axes[0,2].bar(['Churned', 'Not Churned'], [churned_avg, not_churned_avg],
              color=['lightcoral', 'lightgreen'])
axes[0,2].set_title('Avg Monthly Charges')
axes[0,2].set_ylabel('Amount ($)')
# Add labels
axes[0,2].text(0, churned_avg + 2, f'${churned_avg:.0f}', ha='center', fontsize=10)
axes[0,2].text(1, not_churned_avg + 2, f'${not_churned_avg:.0f}', ha='center', fontsize=10)

# 4. Tenure Distribution
tenure_churned = df[df['Churn'] == 'Yes']['tenure'].mean()
tenure_not = df[df['Churn'] == 'No']['tenure'].mean()
axes[1,0].bar(['Churned', 'Not Churned'], [tenure_churned, tenure_not],
              color=['lightcoral', 'lightgreen'])
axes[1,0].set_title('Avg Tenure (Months)')
axes[1,0].set_ylabel('Months')
# Add labels
axes[1,0].text(0, tenure_churned + 2, f'{tenure_churned:.0f}m', ha='center', fontsize=10)
axes[1,0].text(1, tenure_not + 2, f'{tenure_not:.0f}m', ha='center', fontsize=10)

# 5. Payment Method Churn
payment_churn = df.groupby('PaymentMethod')['Churn_binary'].mean() * 100
payment_churn = payment_churn.sort_values()
bars = axes[1,1].barh(range(len(payment_churn)), payment_churn.values,
                      color=['lightblue', 'lightgreen', 'gold', 'lightcoral'])

```

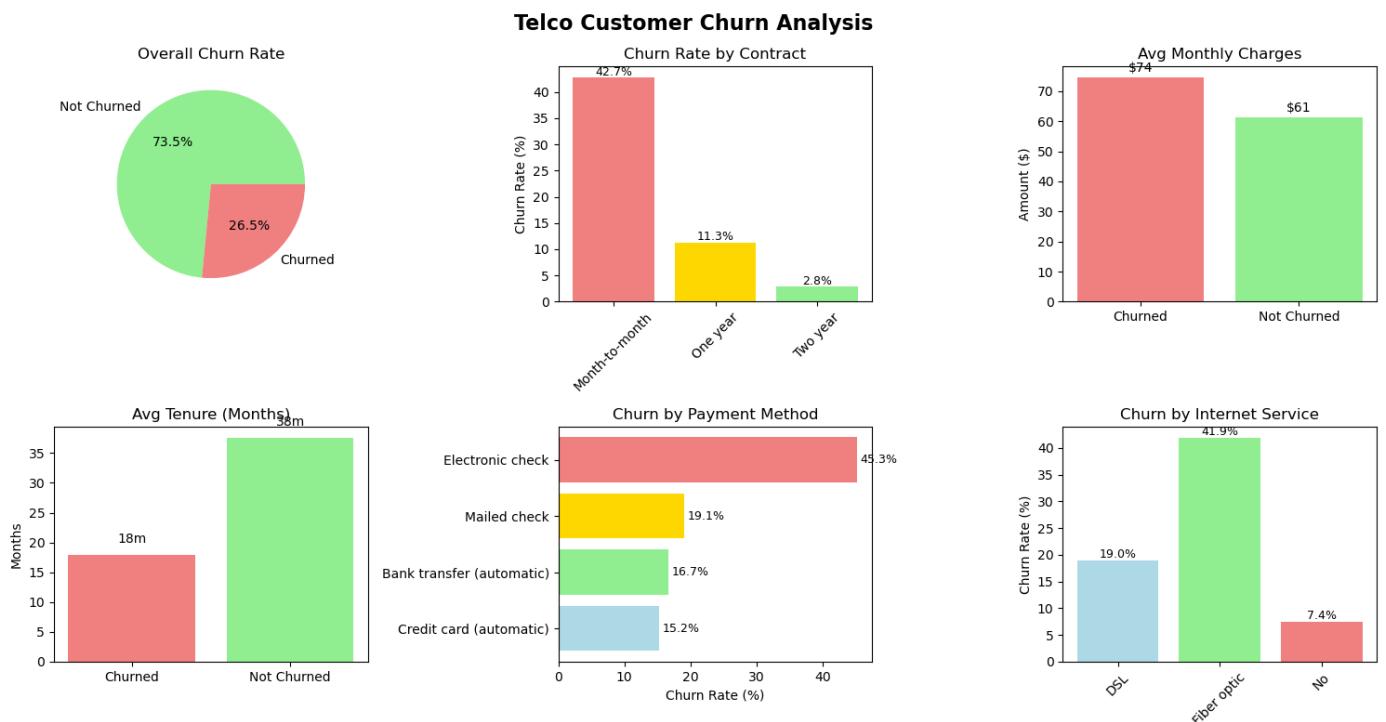
```

axes[1,1].set_title('Churn by Payment Method')
axes[1,1].set_xlabel('Churn Rate (%)')
axes[1,1].set_yticks(range(len(payment_churn)))
axes[1,1].set_yticklabels(payment_churn.index)
# Add labels
for i, v in enumerate(payment_churn.values):
    axes[1,1].text(v + 0.5, i, f'{v:.1f}%', va='center', fontsize=9)

# 6. Internet Service Churn
internet_churn = df.groupby('InternetService')['Churn_binary'].mean() * 100
axes[1,2].bar(internet_churn.index, internet_churn.values,
              color=['lightblue', 'lightgreen', 'lightcoral'])
axes[1,2].set_title('Churn by Internet Service')
axes[1,2].set_ylabel('Churn Rate (%)')
axes[1,2].tick_params(axis='x', rotation=45)
# Add labels
for i, v in enumerate(internet_churn.values):
    axes[1,2].text(i, v + 0.5, f'{v:.1f}%', ha='center', fontsize=9)

plt.tight_layout()
plt.show()
# Print key insights
print("*50")
print("KEY INSIGHTS")
print("*50")
print(f"1. Overall Churn Rate: {(df['Churn_binary'].mean()*100):.1f}%")
print(f"2. Highest Churn by Contract: {contract_churn.idxmax()} ({contract_churn.max():.1f}%")
print(f"3. Avg Monthly Charges: Churned pay ${churned_avg - not_churned_avg:.0f} more")
print(f"4. Avg Tenure: Churned customers stay {tenure_not - tenure_churned:.0f} months longer")
print(f"5. Highest Churn Payment: {payment_churn.idxmax()} ({payment_churn.max():.1f}%")
print(f"6. Highest Churn Internet: {internet_churn.idxmax()} ({internet_churn.max():.1f}%")
print("*50")

```



KEY INSIGHTS

1. Overall Churn Rate: 26.5%
 2. Highest Churn by Contract: Month-to-month (42.7%)
 3. Avg Monthly Charges: Churned pay \$13 more
 4. Avg Tenure: Churned customers stay 20 months less
 5. Highest Churn Payment: Electronic check (45.3%)
 6. Highest Churn Internet: Fiber optic (41.9%)
-

In [3]:

```
# 1. Setup and Load Data
import pandas as pd
import matplotlib.pyplot as plt

# Load data - FIXED PATH
df = pd.read_csv("C:\\Users\\prems\\Videos\\power bi\\Telco_Cusomer_Churn.csv")

# Convert SeniorCitizen to Yes/No
df['SeniorCitizen'] = df['SeniorCitizen'].map({0: 'No', 1: 'Yes'})

# Convert TotalCharges to numeric and handle missing values
df['TotalCharges'] = pd.to_numeric(df['TotalCharges'], errors='coerce')
df['TotalCharges'] = df['TotalCharges'].fillna(0)

# Create one big figure with all subplots
fig, axes = plt.subplots(4, 3, figsize=(18, 16))
fig.suptitle('Telco Customer Churn Analysis', fontsize=20, fontweight='bold', y=1.02)

# 1. Overall Churn Rate (Top Left)
churn_counts = df['Churn'].value_counts()
axes[0,0].pie(churn_counts.values, labels=['No Churn', 'Churned'],
               autopct='%.1f%%', colors=['lightgreen', 'lightcoral'],
               startangle=90)
axes[0,0].set_title('Overall Churn Rate')

# 2. Customer Gender Distribution
gender_counts = df['gender'].value_counts()
axes[0,1].bar(gender_counts.index, gender_counts.values, color=['lightblue', 'lightpink'])
axes[0,1].set_title('Customer Gender')
axes[0,1].set_ylabel('Count')
# Add numbers on bars
for i, value in enumerate(gender_counts.values):
    axes[0,1].text(i, value + 50, str(value), ha='center')

# 3. Monthly Charges Distribution
axes[0,2].hist(df['MonthlyCharges'], bins=20, color='skyblue', edgecolor='black')
axes[0,2].set_title('Monthly Charges Distribution')
axes[0,2].set_xlabel('Monthly Charges ($)')
axes[0,2].set_ylabel('Count')

# 4. Churn by Gender
churn_gender = pd.crosstab(df['gender'], df['Churn'])
churn_gender.plot(kind='bar', ax=axes[1,0], color=['lightgreen', 'lightcoral'])
axes[1,0].set_title('Churn by Gender')
axes[1,0].set_ylabel('Count')
axes[1,0].legend(['No', 'Yes'], title='Churn')

# 5. Churn by Contract Type
churn_contract = pd.crosstab(df['Contract'], df['Churn'])
churn_contract.plot(kind='bar', ax=axes[1,1], color=['lightgreen', 'lightcoral'])
```

```

axes[1,1].set_title('Churn by Contract Type')
axes[1,1].set_ylabel('Count')
axes[1,1].legend(['No', 'Yes'], title='Churn')

# 6. Churn by Payment Method
churn_payment = pd.crosstab(df['PaymentMethod'], df['Churn'])
churn_payment.plot(kind='bar', ax=axes[1,2], color=['lightgreen', 'lightcoral'])
axes[1,2].set_title('Churn by Payment Method')
axes[1,2].set_ylabel('Count')
axes[1,2].set_xticklabels(churn_payment.index, rotation=45, ha='right')
axes[1,2].legend(['No', 'Yes'], title='Churn')

# 7. Monthly Charges by Churn (Box Plot) - FIXED VERSION
churned_data = df[df['Churn'] == 'Yes']['MonthlyCharges']
not_churned_data = df[df['Churn'] == 'No']['MonthlyCharges']
box_data = [churned_data, not_churned_data]

# Using boxplot with correct parameter name
bp = axes[2,0].boxplot(box_data, patch_artist=True)
axes[2,0].set_xticklabels(['Churned', 'Not Churned'])
bp['boxes'][0].set_facecolor('lightcoral')
bp['boxes'][1].set_facecolor('lightgreen')
axes[2,0].set_title('Monthly Charges by Churn')
axes[2,0].set_ylabel('Monthly Charges ($)')

# 8. Senior Citizen Churn
churn_senior = pd.crosstab(df['SeniorCitizen'], df['Churn'])
churn_senior.plot(kind='bar', ax=axes[2,1], color=['lightgreen', 'lightcoral'])
axes[2,1].set_title('Churn by Senior Citizen')
axes[2,1].set_ylabel('Count')
axes[2,1].legend(['No', 'Yes'], title='Churn')

# 9. Partner Status Distribution
partner_counts = df['Partner'].value_counts()
axes[2,2].bar(partner_counts.index, partner_counts.values, color=['lightblue', 'orange'])
axes[2,2].set_title('Customers with Partner')
axes[2,2].set_ylabel('Count')
# Add numbers on bars
for i, value in enumerate(partner_counts.values):
    axes[2,2].text(i, value + 50, str(value), ha='center')

# 10. Dependents Distribution
dependents_counts = df['Dependents'].value_counts()
axes[3,0].bar(dependents_counts.index, dependents_counts.values, color=['lightgreen', 'yellow'])
axes[3,0].set_title('Customers with Dependents')
axes[3,0].set_ylabel('Count')
# Add numbers on bars
for i, value in enumerate(dependents_counts.values):
    axes[3,0].text(i, value + 50, str(value), ha='center')

# 11. Tenure Distribution
axes[3,1].hist(df['tenure'], bins=20, color='purple', alpha=0.7, edgecolor='black')
axes[3,1].set_title('Customer Tenure (Months)')
axes[3,1].set_xlabel('Tenure (Months)')
axes[3,1].set_ylabel('Count')

# 12. Internet Service Type
internet_counts = df['InternetService'].value_counts()
axes[3,2].pie(internet_counts.values, labels=internet_counts.index,

```

```

        autopct='%.1f%%', startangle=90,
        colors=['lightblue', 'lightgreen', 'lightcoral'])
axes[3,2].set_title('Internet Service Type')

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

# Print summary statistics
print("=*60)
print("SUMMARY STATISTICS")
print("=*60)
print(f"Total Customers: {len(df)}")
print(f"Churn Rate: {(df['Churn'] == 'Yes').mean() * 100:.1f}%" )
print(f"Average Monthly Charge: ${df['MonthlyCharges'].mean():.2f}")
print(f"Average Tenure: {df['tenure'].mean():.1f} months")
print(f"Female Customers: {len(df[df['gender'] == 'Female'])} ({(df['gender'] == 'Female').mean() * 100:.1f} %)" )
print(f"Male Customers: {len(df[df['gender'] == 'Male'])} ({(df['gender'] == 'Male').mean() * 100:.1f} %)" )
print(f"Senior Citizens: {len(df[df['SeniorCitizen'] == 'Yes'])} ({(df['SeniorCitizen'].mean() * 100:.1f} %)" )
print("=*60)

```

Telco Customer Churn Analysis



SUMMARY STATISTICS

Total Customers: 7043
Churn Rate: 26.5%
Average Monthly Charge: \$64.76
Average Tenure: 32.4 months
Female Customers: 3488 (49.5%)
Male Customers: 3555 (50.5%)
Senior Citizens: 1142 (16.2%)

In [2]:

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

# Load the dataset
df = pd.read_csv("C:\\\\Users\\\\prems\\\\Videos\\\\power bi\\\\Telco_Cusomer_Churn.csv")

# Data cleaning
df['TotalCharges'] = pd.to_numeric(df['TotalCharges'], errors='coerce')
df['TotalCharges'] = df['TotalCharges'].fillna(0)

# Convert Churn to binary
df['ChurnBinary'] = df['Churn'].map({'Yes': 1, 'No': 0})

# 1. Overall Churn Rate
total_customers = len(df)
churned_customers = df['ChurnBinary'].sum()
churn_rate = (churned_customers / total_customers) * 100

# 2. Tenure Impact Analysis
short_tenure = df[df['tenure'] < 12]
long_tenure = df[df['tenure'] >= 24]
short_tenure_churn_rate = (short_tenure['ChurnBinary'].sum() / len(short_tenure)) * 100
long_tenure_churn_rate = (long_tenure['ChurnBinary'].sum() / len(long_tenure)) * 100
tenure_churn_ratio = short_tenure_churn_rate / long_tenure_churn_rate

# 3. Contract Type Analysis
contract_churn = df.groupby('Contract')['ChurnBinary'].mean() * 100

# 4. Internet Service Analysis
internet_churn = df.groupby('InternetService')['ChurnBinary'].mean() * 100

# 5. Payment Method Analysis
payment_churn = df.groupby('PaymentMethod')['ChurnBinary'].mean() * 100

# 6. Monthly Charges Impact
avg_monthly_charge = df['MonthlyCharges'].mean()
high_charge_churn = df[df['MonthlyCharges'] > avg_monthly_charge]['ChurnBinary'].mean()
low_charge_churn = df[df['MonthlyCharges'] <= avg_monthly_charge]['ChurnBinary'].mean()

# 7. Senior Citizens Analysis
senior_churn = df[df['SeniorCitizen'] == 1]['ChurnBinary'].mean() * 100
non_senior_churn = df[df['SeniorCitizen'] == 0]['ChurnBinary'].mean() * 100

# 8. Service Add-ons Analysis
def analyze_service_feature(feature):
```

```

with_service = df[df[feature] == 'Yes']['ChurnBinary'].mean() * 100
without_service = df[df[feature] == 'No']['ChurnBinary'].mean() * 100
return with_service, without_service

tech_support_churn = analyze_service_feature('TechSupport')
online_security_churn = analyze_service_feature('OnlineSecurity')
online_backup_churn = analyze_service_feature('OnlineBackup')

# 9. Paperless Billing Analysis
paperless_churn = df[df['PaperlessBilling'] == 'Yes']['ChurnBinary'].mean() * 100
non_paperless_churn = df[df['PaperlessBilling'] == 'No']['ChurnBinary'].mean() * 100

# 10. Customer Lifetime Value Analysis
avg_total_charges_churned = df[df['Churn'] == 'Yes']['TotalCharges'].mean()
avg_total_charges_retained = df[df['Churn'] == 'No']['TotalCharges'].mean()
revenue_difference_ratio = avg_total_charges_retained / avg_total_charges_churned

# Print Insights
print("=*60)
print("TELCO CUSTOMER CHURN ANALYSIS - KEY INSIGHTS")
print("=*60)

print(f"\n1. Overall Churn Rate: {churn_rate:.1f}%")
print(f"    Total Customers: {total_customers:,}")
print(f"    Churned Customers: {churned_customers:,}")

print(f"\n2. Tenure Impact:")
print(f"    Short-tenure (<12 months) churn: {short_tenure_churn_rate:.1f}%")
print(f"    Long-tenure (24+ months) churn: {long_tenure_churn_rate:.1f}%")
print(f"    Short-tenure customers are {tenure_churn_ratio:.1f}x more likely to churn")

print(f"\n3. Contract Type Analysis:")
for contract, rate in contract_churn.items():
    print(f"    {contract}: {rate:.1f}% churn")

print(f"\n4. Internet Service Analysis:")
for service, rate in internet_churn.items():
    print(f"    {service}: {rate:.1f}% churn")

print(f"\n5. Payment Method Analysis:")
for method, rate in payment_churn.items():
    print(f"    {method}: {rate:.1f}% churn")

print(f"\n6. Monthly Charges Impact:")
print(f"    Average monthly charge: ${avg_monthly_charge:.2f}")
print(f"    Above-average charge churn: {high_charge_churn:.1f}%")
print(f"    Below-average charge churn: {low_charge_churn:.1f}%")

print(f"\n7. Senior Citizen Analysis:")
print(f"    Senior citizens: {senior_churn:.1f}% churn")
print(f"    Non-seniors: {non_senior_churn:.1f}% churn")

print(f"\n8. Service Add-ons Impact:")
print(f"    Tech Support - With: {tech_support_churn[0]:.1f}%, Without: {tech_support_chu
print(f"    Online Security - With: {online_security_churn[0]:.1f}%, Without: {online_sec
print(f"    Online Backup - With: {online_backup_churn[0]:.1f}%, Without: {online_backup_"

service_ratio = tech_support_churn[1] / tech_support_churn[0]
print(f"    Customers without tech support are {service_ratio:.1f}x more likely to churn"

```

```

print(f"\n9. Paperless Billing:")
print(f"    With paperless billing: {paperless_churn:.1f}% churn")
print(f"    Without paperless billing: {non_paperless_churn:.1f}% churn")

print(f"\n10. Customer Lifetime Value:")
print(f"    Average total charges (churned): ${avg_total_charges_churned:.2f}")
print(f"    Average total charges (retained): ${avg_total_charges_retained:.2f}")
print(f"    Retained customers generate {revenue_difference_ratio:.1f}x more revenue")

# Additional insights
print(f"\n" + "="*60)
print("ADDITIONAL INSIGHTS")
print("=".*60)

# Partner/Dependents analysis
partner_churn = df.groupby('Partner')['ChurnBinary'].mean() * 100
dependents_churn = df.groupby('Dependents')['ChurnBinary'].mean() * 100

print(f"\nPartner Status:")
print(f"    With partner: {partner_churn['Yes']:.1f}% churn")
print(f"    Without partner: {partner_churn['No']:.1f}% churn")

print(f"\nDependents:")
print(f"    With dependents: {dependents_churn['Yes']:.1f}% churn")
print(f"    Without dependents: {dependents_churn['No']:.1f}% churn")

# Gender analysis
gender_churn = df.groupby('gender')['ChurnBinary'].mean() * 100
print(f"\nGender Analysis:")
print(f"    Female: {gender_churn['Female']:.1f}% churn")
print(f"    Male: {gender_churn['Male']:.1f}% churn")

# Streaming services analysis
streaming_tv_churn = analyze_service_feature('StreamingTV')
streaming_movies_churn = analyze_service_feature('StreamingMovies')

print(f"\nStreaming Services:")
print(f"    Streaming TV - With: {streaming_tv_churn[0]:.1f}%, Without: {streaming_tv_churn[1]:.1f}%")
print(f"    Streaming Movies - With: {streaming_movies_churn[0]:.1f}%, Without: {streaming_movies_churn[1]:.1f}%")
=====
```

TELCO CUSTOMER CHURN ANALYSIS - KEY INSIGHTS

1. Overall Churn Rate: 26.5%

Total Customers: 7,043
 Churned Customers: 1,869

2. Tenure Impact:

Short-tenure (<12 months) churn: 48.3%
 Long-tenure (24+ months) churn: 14.3%
 Short-tenure customers are 3.4x more likely to churn

3. Contract Type Analysis:

Month-to-month: 42.7% churn
 One year: 11.3% churn
 Two year: 2.8% churn

4. Internet Service Analysis:

DSL: 19.0% churn

Fiber optic: 41.9% churn

No: 7.4% churn

5. Payment Method Analysis:

Bank transfer (automatic): 16.7% churn

Credit card (automatic): 15.2% churn

Electronic check: 45.3% churn

Mailed check: 19.1% churn

6. Monthly Charges Impact:

Average monthly charge: \$64.76

Above-average charge churn: 34.5%

Below-average charge churn: 16.5%

7. Senior Citizen Analysis:

Senior citizens: 41.7% churn

Non-seniors: 23.6% churn

8. Service Add-ons Impact:

Tech Support - With: 15.2%, Without: 41.6%

Online Security - With: 14.6%, Without: 41.8%

Online Backup - With: 21.5%, Without: 39.9%

Customers without tech support are 2.7x more likely to churn

9. Paperless Billing:

With paperless billing: 33.6% churn

Without paperless billing: 16.3% churn

10. Customer Lifetime Value:

Average total charges (churned): \$1531.80

Average total charges (retained): \$2549.91

Retained customers generate 1.7x more revenue

=====

ADDITIONAL INSIGHTS

Partner Status:

With partner: 19.7% churn

Without partner: 33.0% churn

Dependents:

With dependents: 15.5% churn

Without dependents: 31.3% churn

Gender Analysis:

Female: 26.9% churn

Male: 26.2% churn

Streaming Services:

Streaming TV - With: 30.1%, Without: 33.5%

Streaming Movies - With: 29.9%, Without: 33.7%

In [32]:

```

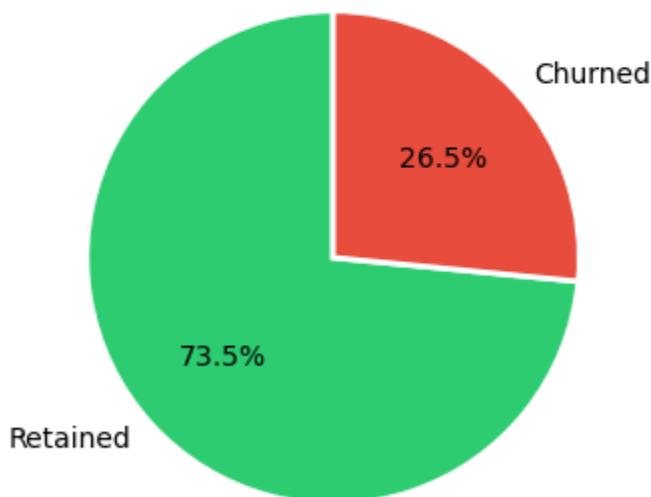
# 1. Overall Churn Rate - Pie Chart
plt.figure(figsize=(7, 4))
churn_counts = df['Churn'].value_counts()
colors = ['#e74c3c', '#2ecc71']
plt.pie(churn_counts, labels=['Retained', 'Churned'], autopct='%.1f%%',
        colors=colors, startangle=90, wedgeprops={'edgecolor': 'white', 'linewidth': 2})
plt.title('Overall Customer Churn Distribution\nTotal Customers: {}'.format(total_cust),
          fontsize=16, fontweight='bold', pad=20)
plt.show()
# 6. Monthly Charges Analysis - Histogram
plt.figure(figsize=(6, 4))
churned = df[df['Churn'] == 'Yes']['MonthlyCharges']
retained = df[df['Churn'] == 'No']['MonthlyCharges']

plt.hist([churned, retained], bins=20, label=['Churned', 'Retained'],
         color=['#e74c3c', '#2ecc71'], alpha=0.7, edgecolor='black')
plt.title('Monthly Charges Distribution: Churned vs Retained', fontsize=16, fontweight='bold')
plt.xlabel('Monthly Charges ($)', fontsize=12)
plt.ylabel('Number of Customers', fontsize=12)
plt.legend()
plt.grid(alpha=0.3)
plt.tight_layout()
plt.show()

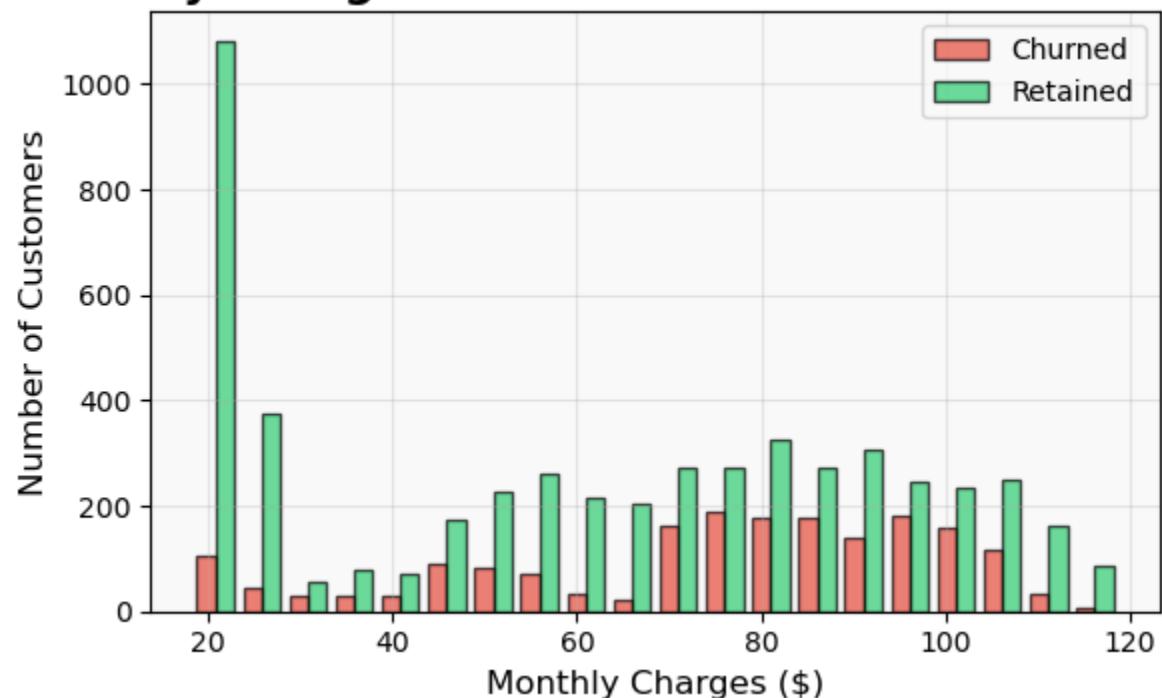
```

Overall Customer Churn Distribution

Total Customers: 7,043



Monthly Charges Distribution: Churned vs Retained



In []: