

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-CF0CW	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]:

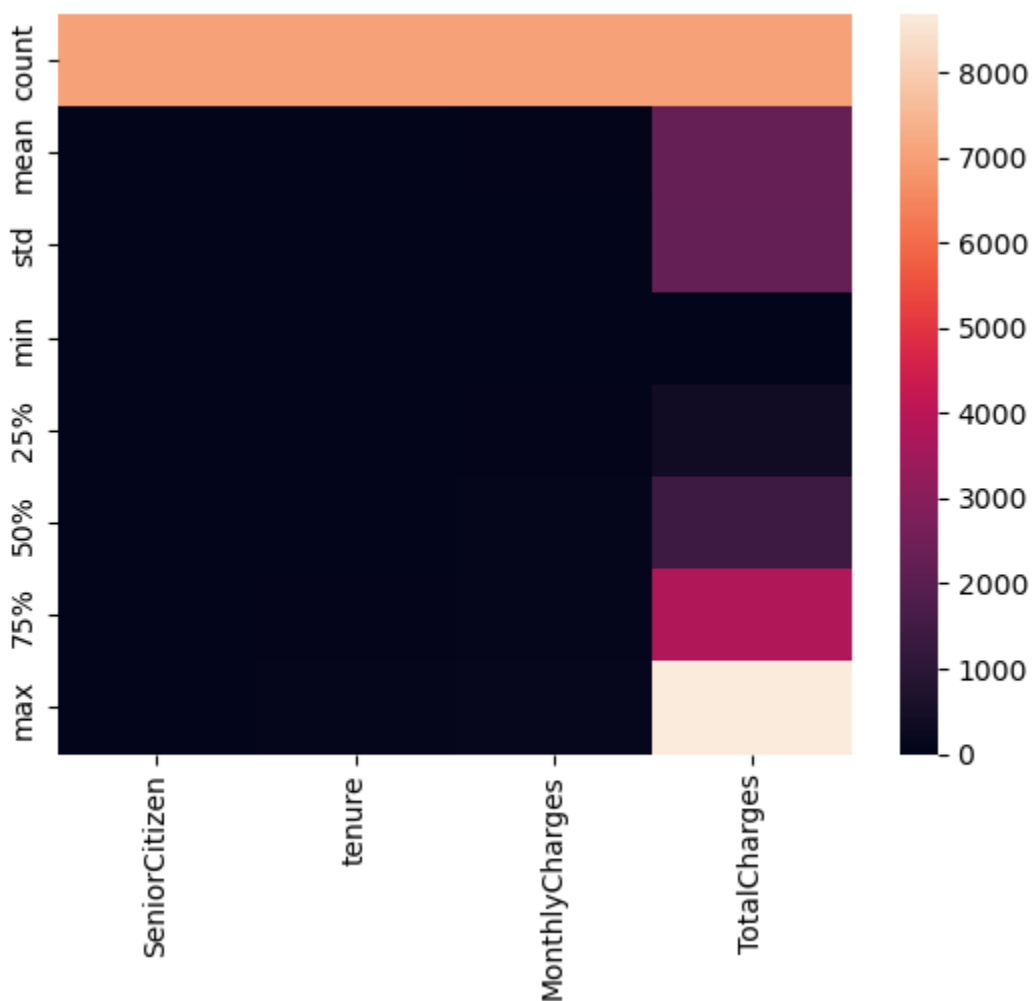
```
# Check basic statistics
print("\nBasic Statistics:")
print(df.describe())
sns.heatmap(df.describe())
```

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	
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	

	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges	Churn
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

[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_righttone	gender_righttone	SeniorCitizen_righttone	Partner_righttone	Dependents_righttone
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-JJAZL	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	Int
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	
6	1452-KIOVK	Male	0	No	Yes	22	Yes	Yes	
9	6388-TABGU	Male	0	No	Yes	62	Yes	No	
...
7033	9767-FFLEM	Male	0	No	No	38	Yes	No	
7035	8456-QDAVC	Male	0	No	No	19	Yes	No	
7038	6840-RESVB	Male	0	Yes	Yes	24	Yes	Yes	
7041	8361-LTMKD	Male	1	Yes	No	4	Yes	Yes	
7042	3186-AJIEK	Male	0	No	No	66	Yes	No	

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	Intern
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-JJAZL	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	Inter
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	Inter
18	4190-MFLUW	Female	0	Yes	Yes	10	Yes	No	
20	8779-QRDMV	Male	1	No	No	1	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	Int
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	
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 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	Int
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-JAZL	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	Int
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	Int
	RESVB								
7040	4801-JAZL	Female	0	Yes	Yes	11	No	No phone service	
7041	8361-LTMKD	Male	1	Yes	No	4	Yes	Yes	

3977 rows × 21 columns

remove duplicates

In [24]:

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

Out[24]:

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

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	Int
0	7590-VHVEG	Female	0	Yes	No	4	No	No phone service	
1	5575-GNVDE	Male	0	No	No	37	Yes	No	
2	3668-QPYBK	Male	0	No	No	5	Yes	No	
3	7795-CFOCW	Male	0	No	No	48	No	No phone service	
4	9237-HQITU	Female	0	No	No	5	Yes	No	
...
7038	6840-RESVB	Male	0	Yes	Yes	27	Yes	Yes	
7039	2234-XADUH	Female	0	Yes	Yes	75	Yes	Yes	
7040	4801-JAZZL	Female	0	Yes	Yes	14	No	No phone service	
7041	8361-LTMKD	Male	1	Yes	No	7	Yes	Yes	
7042	3186-AJIEK	Male	0	No	No	69	Yes	No	

7043 rows × 21 columns

In [33]:

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

Out[33]:

gender

Male 834

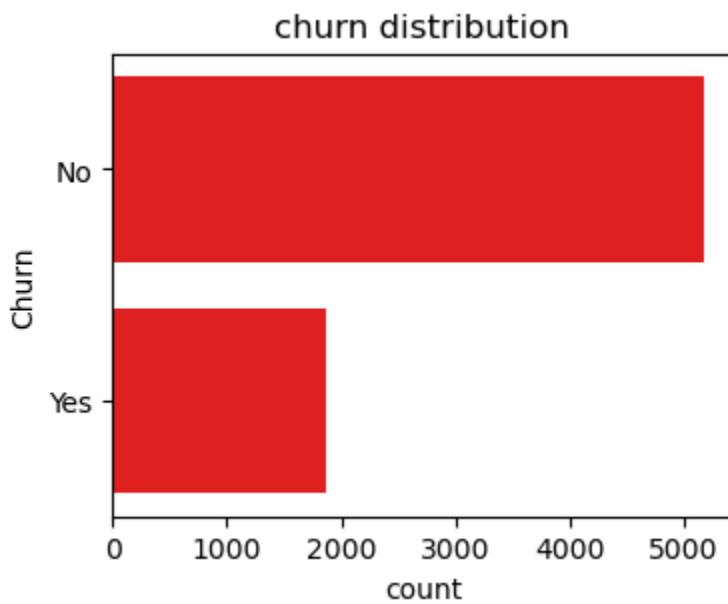
Female 778

Name: count, dtype: int64

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

In [41]:

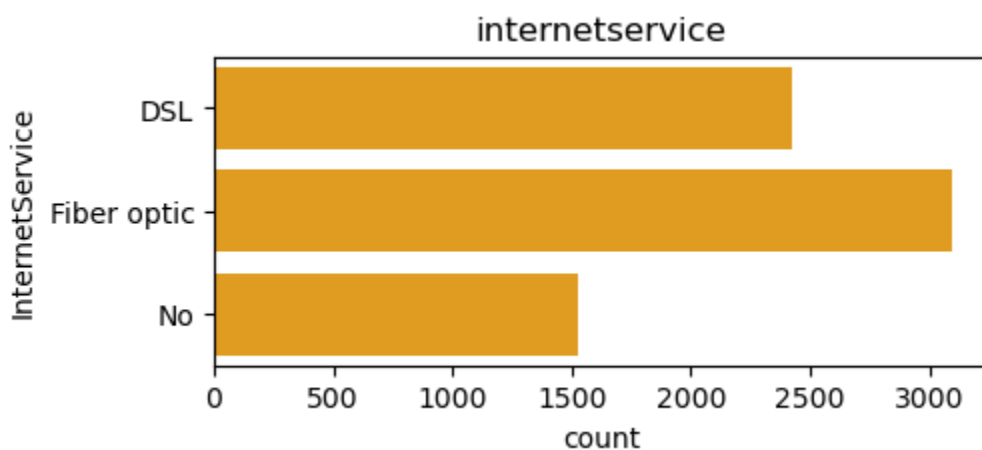
```
plt.figure(figsize=(4,3))
sns.countplot(df['Churn'],color='red')
plt.title('churn distribution')
plt.show()
```



what is the distribution of internet service types?

In [42]:

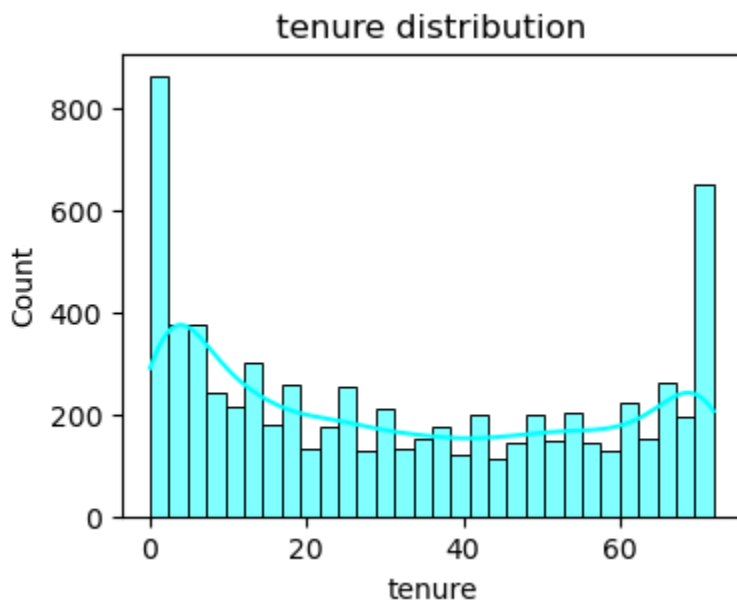
```
plt.figure(figsize=(5,2))
sns.countplot(df['InternetService'],color='orange')
plt.title('internetservice')
plt.show()
```



how is the tenure distribution among customers?

In [43]:

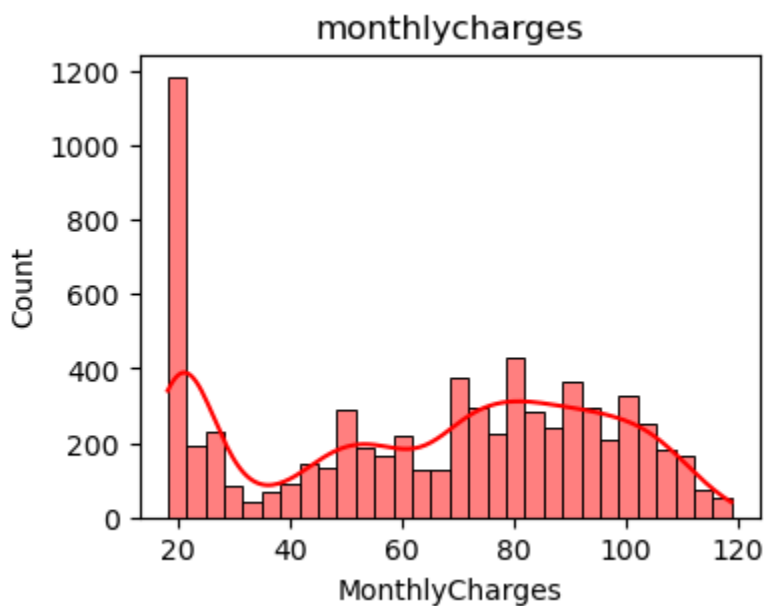
```
plt.figure(figsize=(4,3))
sns.histplot(df['tenure'],color='cyan',kde=True,bins=30)
plt.title('tenure distribution')
plt.show()
```



what is the distribution of monthly charges?

In [44]:

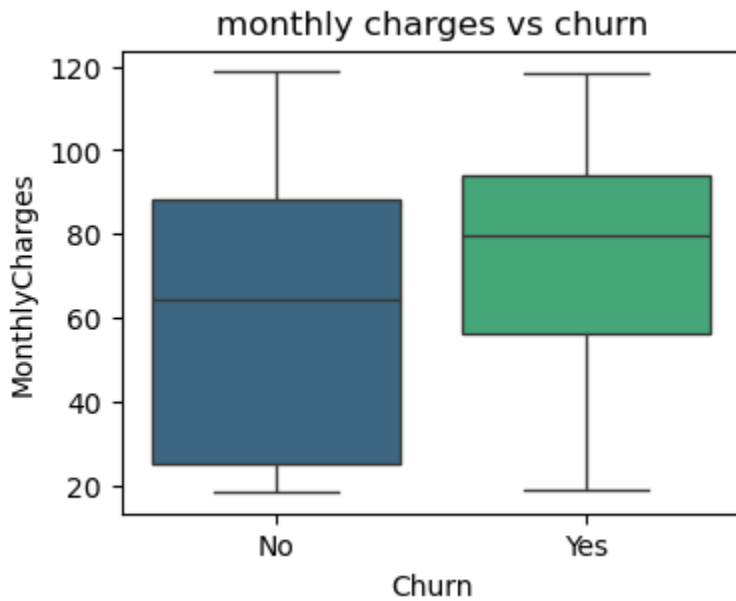
```
plt.figure(figsize=(4,3))
sns.histplot(df['MonthlyCharges'],color='red',bins=30,kde=True,)
plt.title('monthlycharges')
plt.show()
```



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	Interne
------------	--------	---------------	---------	------------	--------	--------------	---------------	---------

0 rows × 21 columns

In [8]:

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

drop duplicates :

	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-CF0CW	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-JJAZL	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

	3	4	\
customerID	7795-CF0CW	9237-HQITU	
gender	Male	Female	
SeniorCitizen	0	0	
Partner	No	No	
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	

	5	6	7	\
customerID	9305-CDSKC	1452-KIOVK	6713-OK0MC	
gender	Female	Male	Female	
SeniorCitizen	0	0	0	
Partner	No	No	No	
Dependents	No	Yes	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

	7037	7038 \
customerID	2569-WGER0	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

	7039	7040	7041 \
customerID	2234-XADUH	4801-JJAZL	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
```

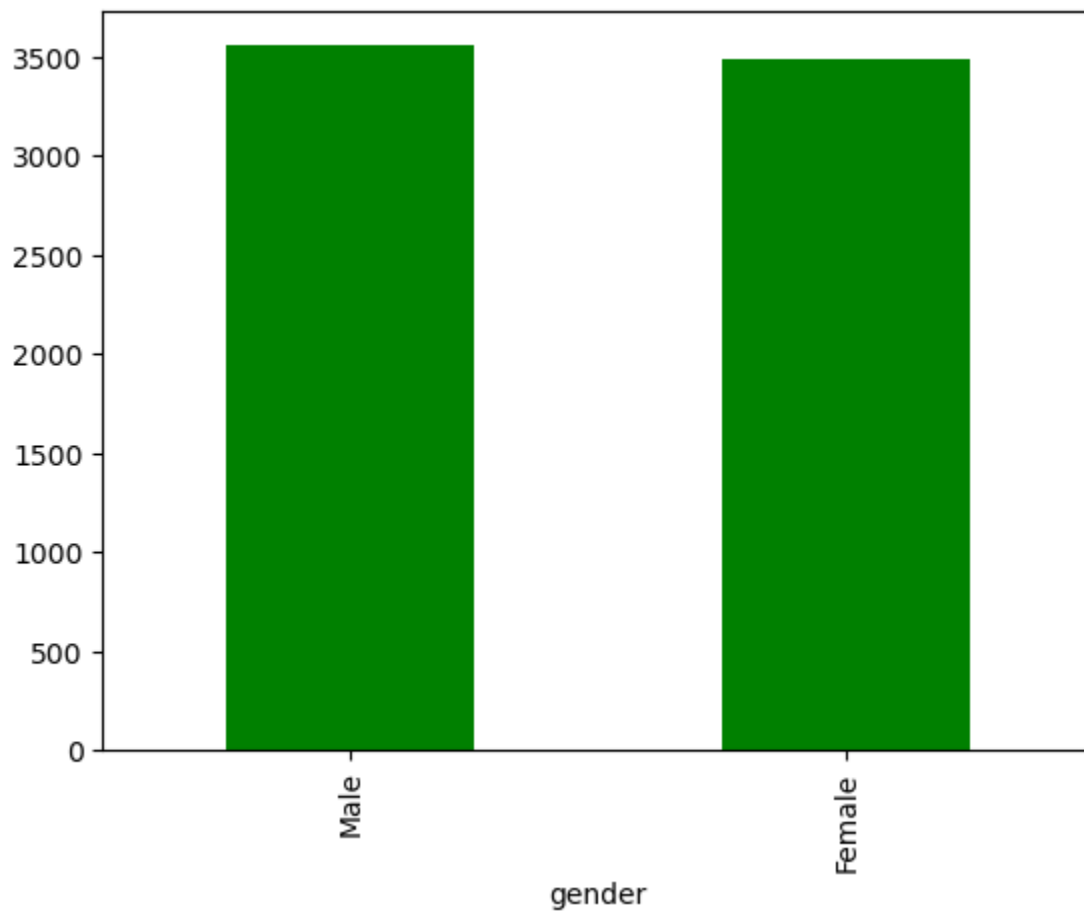
```
PhoneService:
PhoneService
Yes     6361
No       682
Name: count, dtype: int64
```

In [9]:

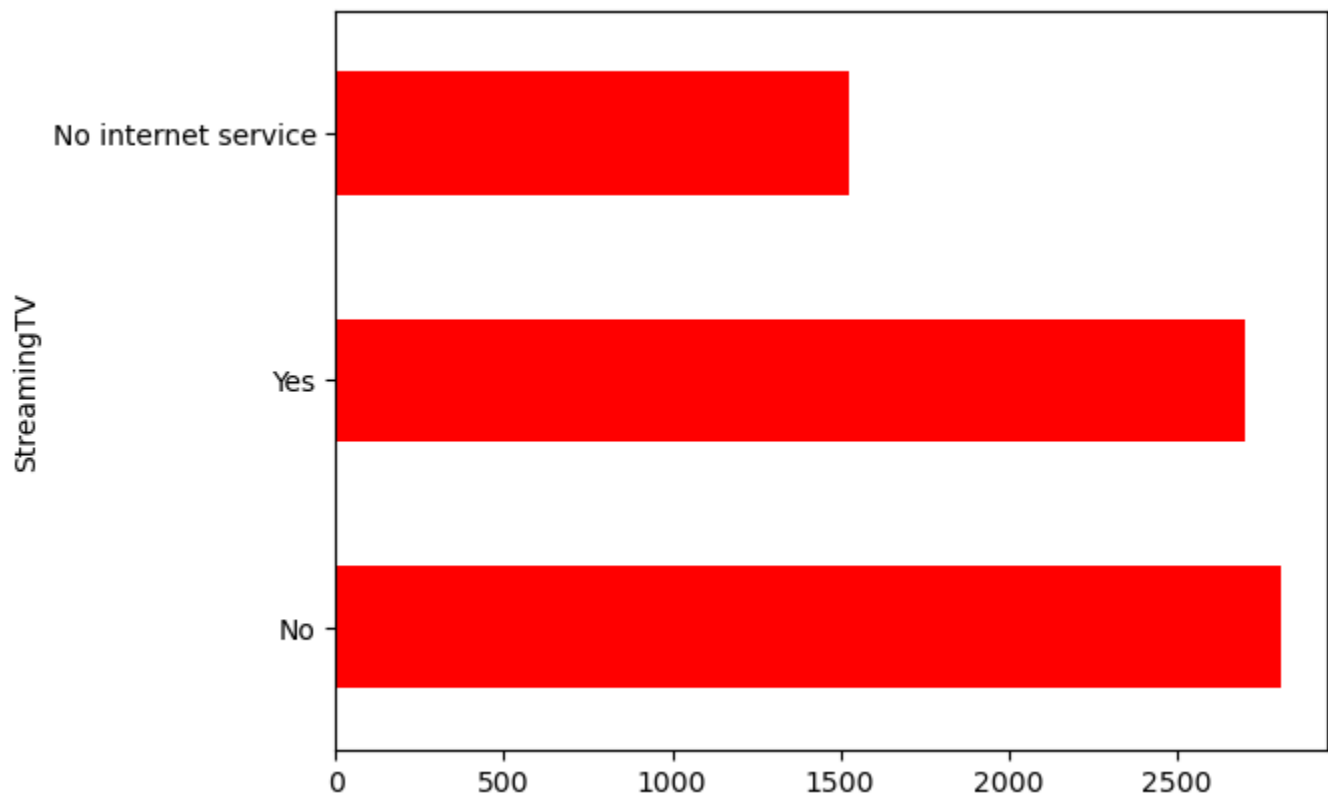
```
print("\n value count by gender:")
c=df['gender'].value_counts()
print(c)
c.plot(kind='bar',color='g')
plt.show()
print("\n" + "="*60)
print("\n value count by streaming tv:")
d=df['StreamingTV'].value_counts()
print(d)
d.plot(kind='barh',color='r')
plt.show()
print("\n" + "="*60)
print("\n value count by phoneservie :")
k=df['PhoneService'].value_counts()
print(k)

df['PhoneService'].value_counts().plot(kind='pie',autopct='%1.1f%%',colors='myr',shadow=
```

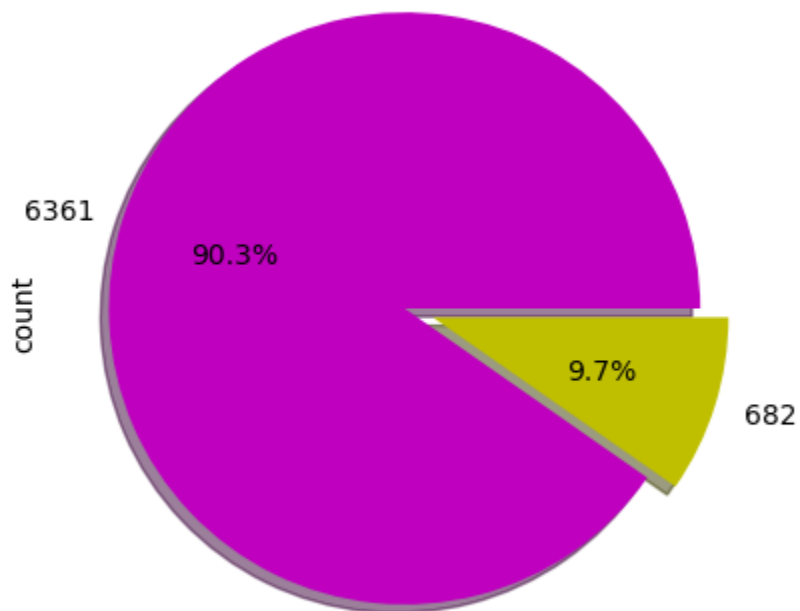
```
value count by gender:
gender
Male      3555
Female    3488
Name: count, dtype: int64
```



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

```

unique values by internet service :

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

=====

unique values by internet service and its count:

```
3
```

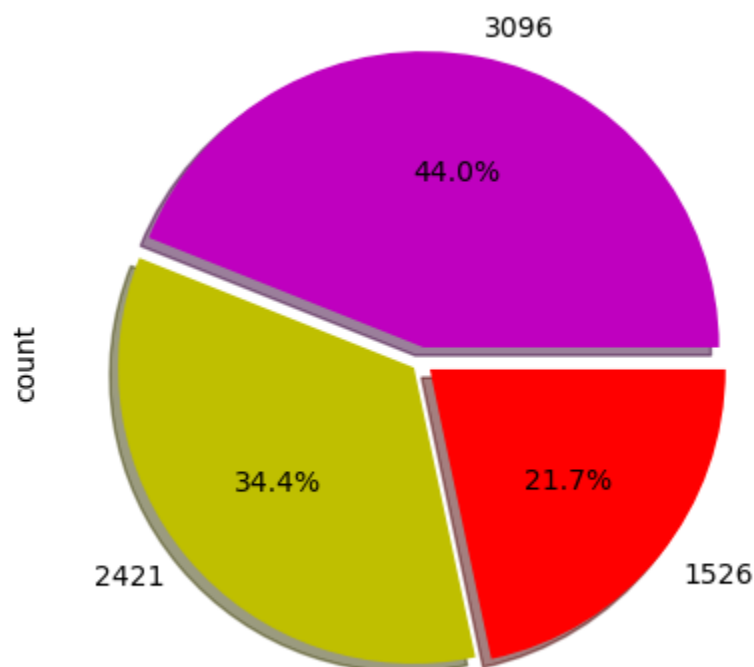
=====

value count by internet service:

```

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-NMZYQ	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]:

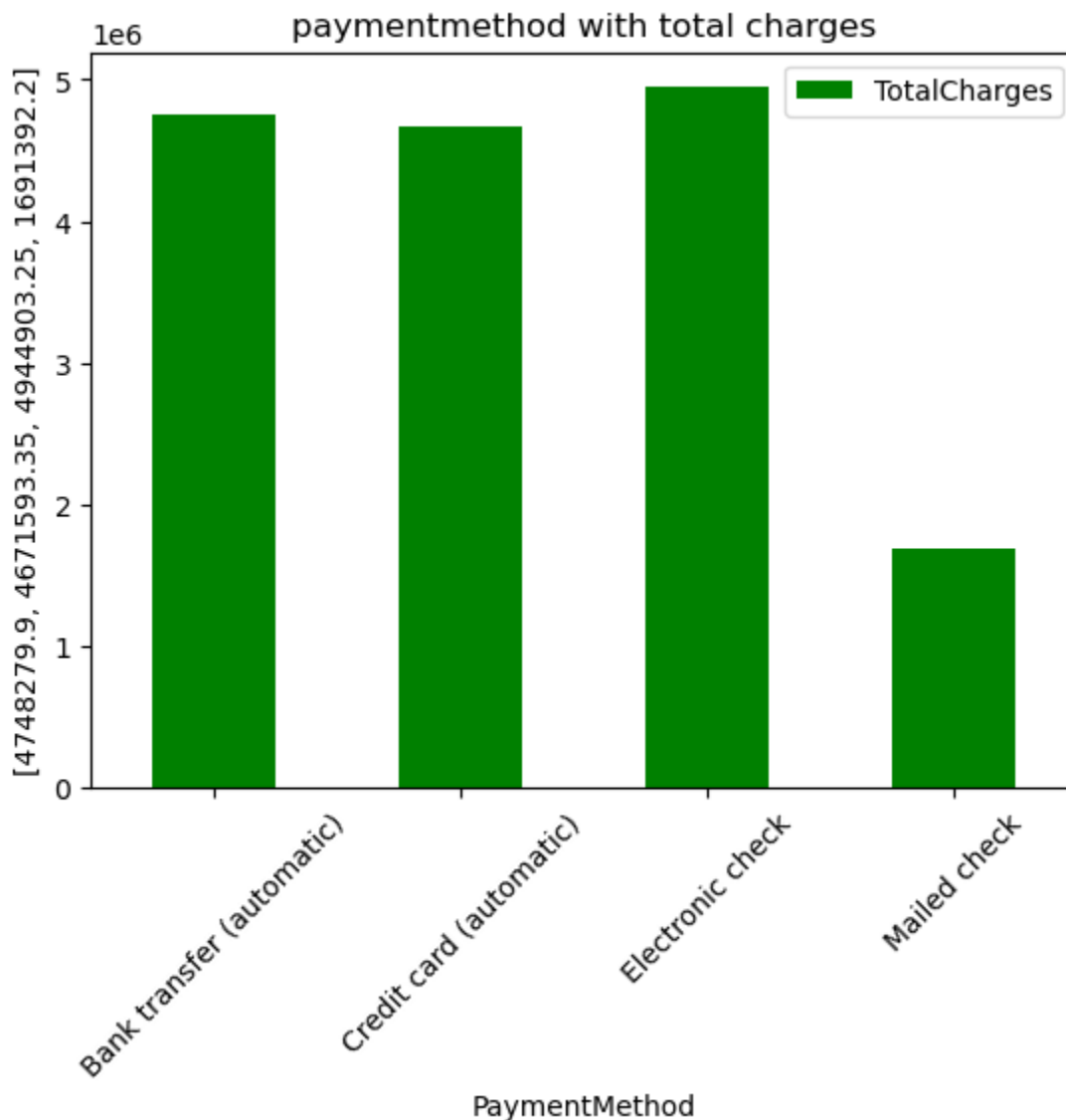
```
print("\n group by paymentmethod with totalcharges :\n")
print(df.groupby('PaymentMethod')['TotalCharges'].sum())
df.groupby('PaymentMethod')['TotalCharges'].sum().plot(kind='bar',rot=45,ylabel=df.group
plt.title('paymentmethod with total charges')
plt.legend()
```

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]:

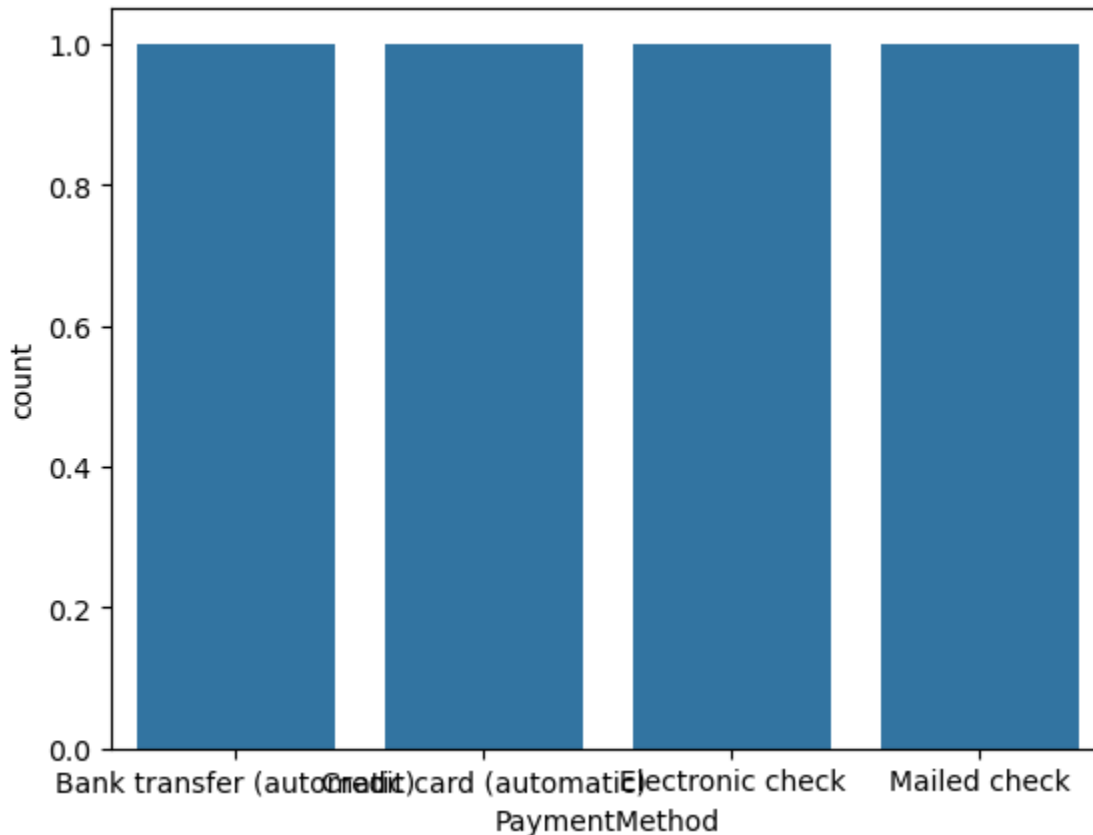
<matplotlib.legend.Legend at 0x1ab97bfe480>



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 :")
print(df.groupby('PaymentMethod')['TotalCharges'].sum().sort_values(ascending=False).head(1))
```

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:

	tenure					MonthlyCharges		
	mean	median	min	max	count	mean	median	std
Churn								
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

	TotalCharges	
	mean	sum
Churn		
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

Fiber optic	One year	570	61.40	0.09
	Two year	628	70.46	0.02
	Month-to-month	2128	87.02	0.55
No	One year	539	98.78	0.19
	Two year	429	104.57	0.07
	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:

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

=====

Transposed Complex Pivot Table:

Contract		Month-to-month		One year		Two year		\
PaperlessBilling		No	Yes	No	Yes	No	Yes	
Churn								
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-CF0CW             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-CF0CW	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-CF0CW	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-CF0CW	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	No	1413
5	No	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		No	1	Yes	0	1	All
SeniorCitizen		0	1	0	1		
Contract	PaperlessBilling						
Month-to-month	No	795	88	335	71	1289	
	Yes	1059	278	879	370	2586	
One year	No	573	52	43	5	673	
	Yes	573	109	94	24	800	
Two year	No	847	48	13	2	910	
	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-QOJRX	DSL	Month-to-month	Bank transfer (automatic)	
6522	7233-IOQNP	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
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

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]:

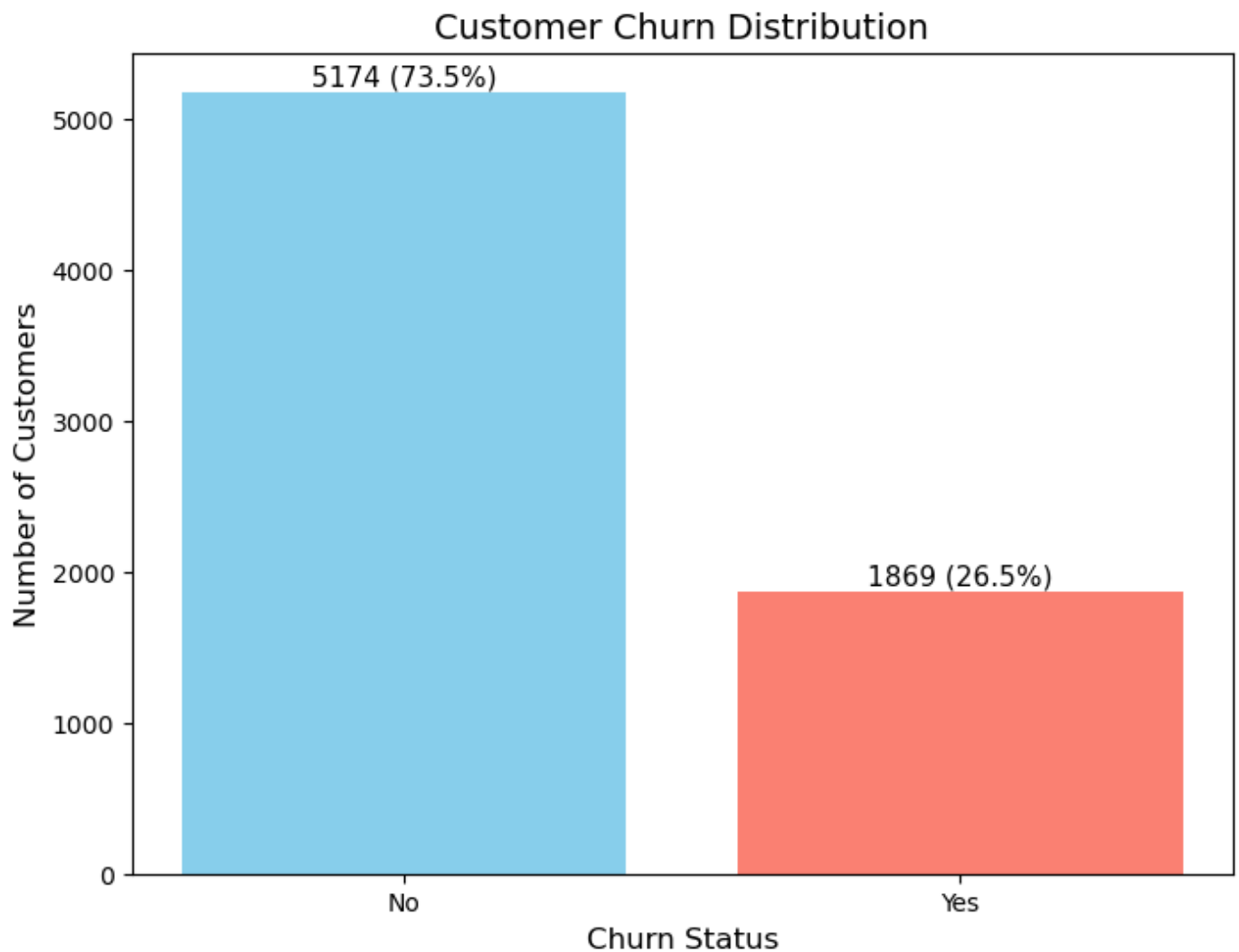
```

# 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))

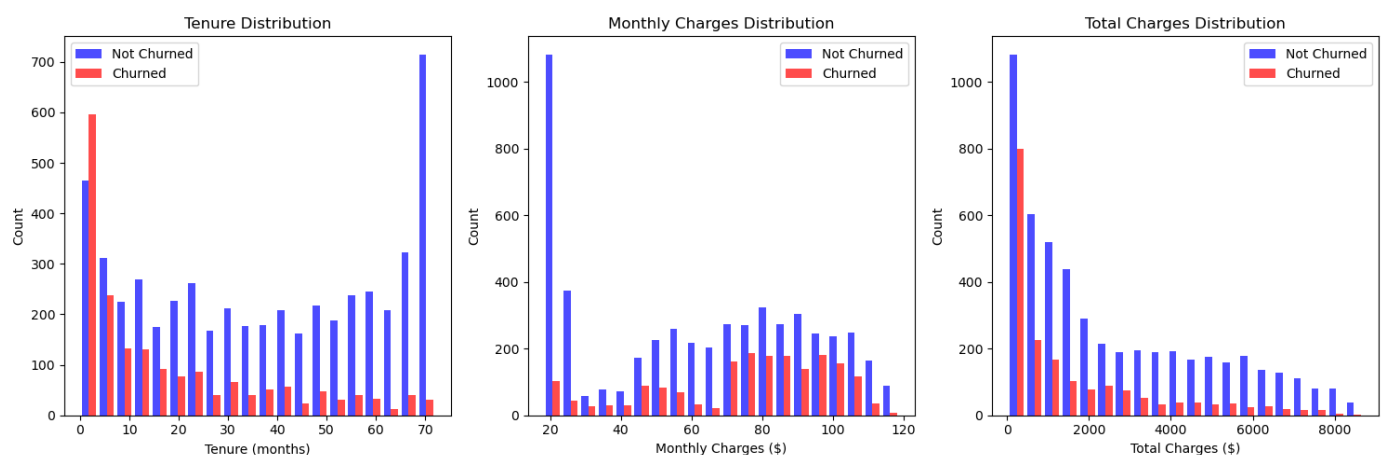
# Tenure comparison
axes[0].hist([not_churned['tenure'], churned['tenure']],
             bins=20, label=['Not Churned', 'Churned'],
             alpha=0.7, color=['blue', 'red'])
axes[0].set_title('Tenure Distribution')
axes[0].set_xlabel('Tenure (months)')
axes[0].set_ylabel('Count')
axes[0].legend()

# 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}")
```

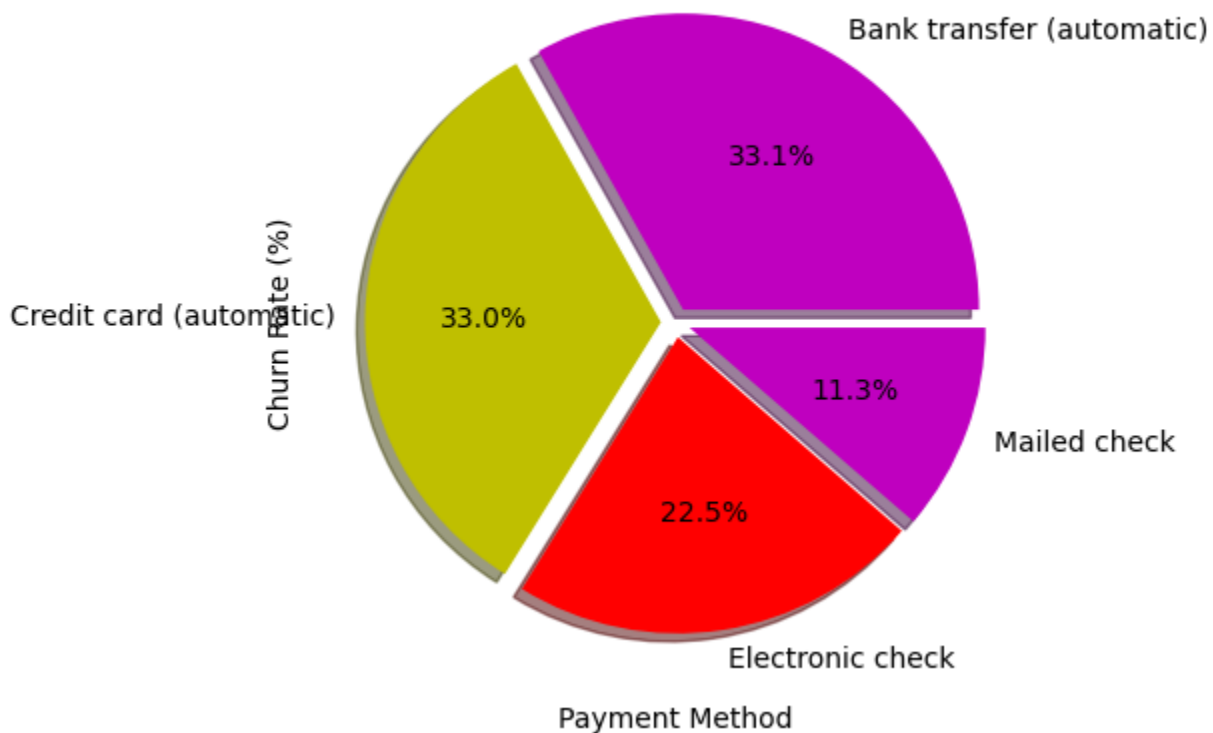


Average Values Comparison:
Average Tenure - Churned: 18.0 months
Average Tenure - Not Churned: 37.6 months
Average Monthly Charges - Churned: \$74.44
Average Monthly Charges - Not Churned: \$61.27

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()
```

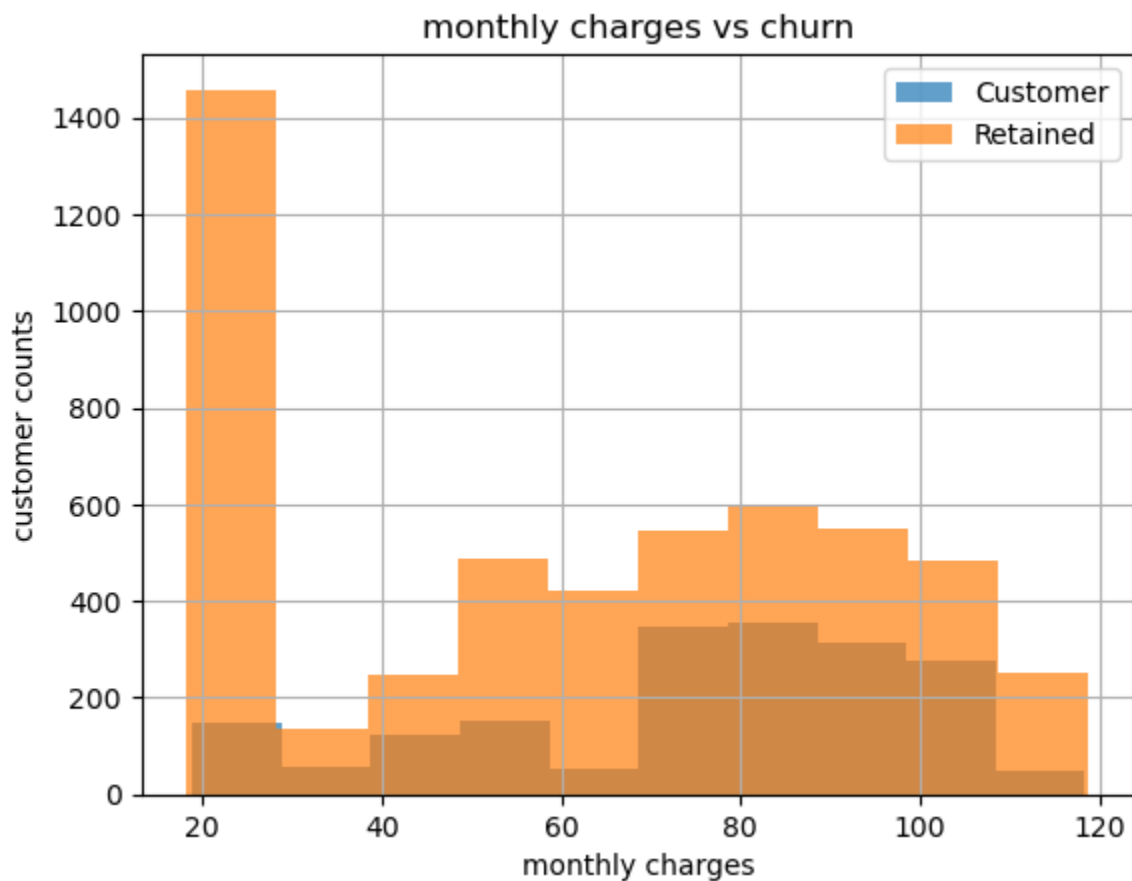

Churn Rate by Payment Method



does monthly charge impact churn

In [37]:

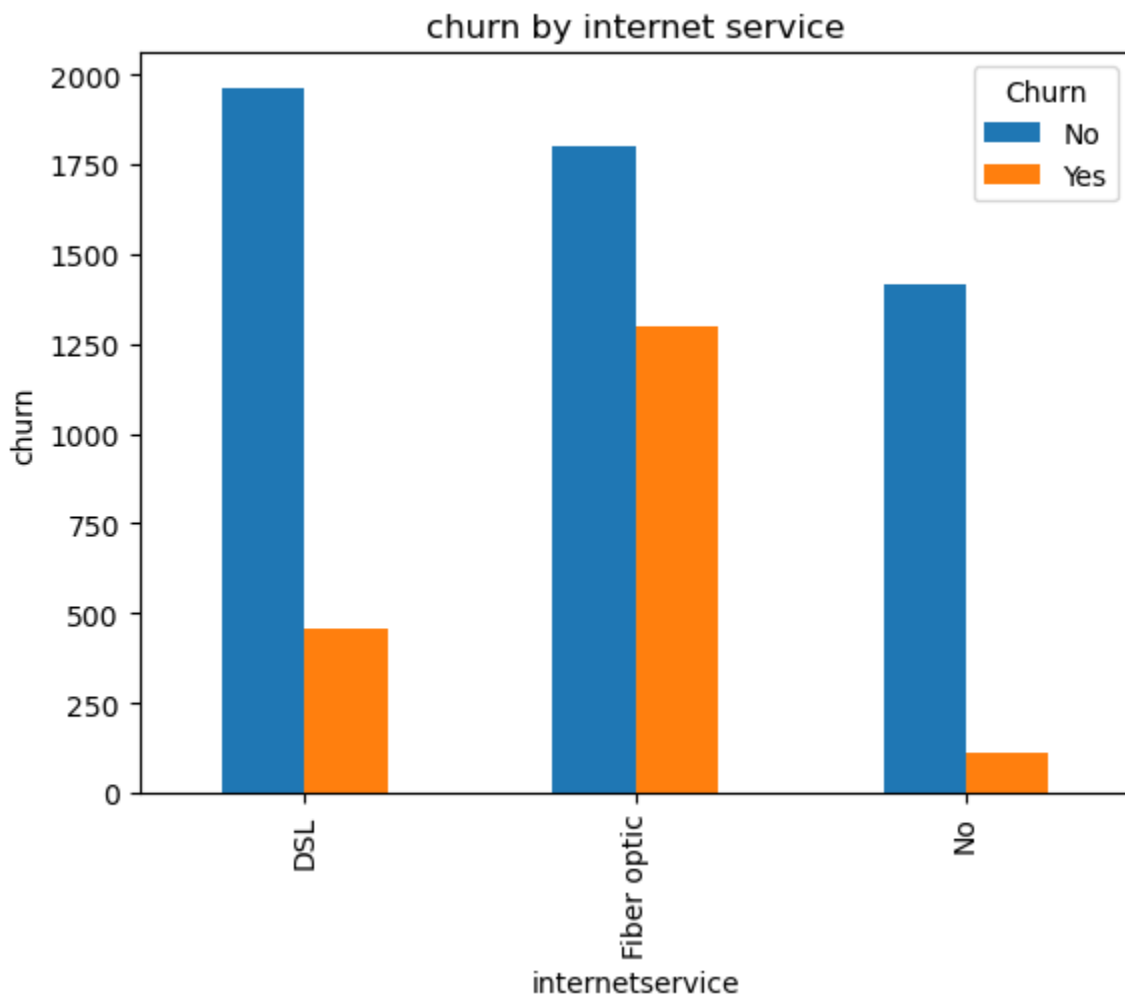
```
plt.figure()
df[df['Churn']=='Yes']['MonthlyCharges'].hist(alpha=0.7)
df[df['Churn']=='No']['MonthlyCharges'].hist(alpha=0.7)
plt.legend(['Customer', 'Retained'])
plt.xlabel('monthly charges')
plt.ylabel('customer counts')
plt.title('monthly charges vs churn')
plt.show()
```



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='%1.1f%%', 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='%1.1f%%', 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]

```

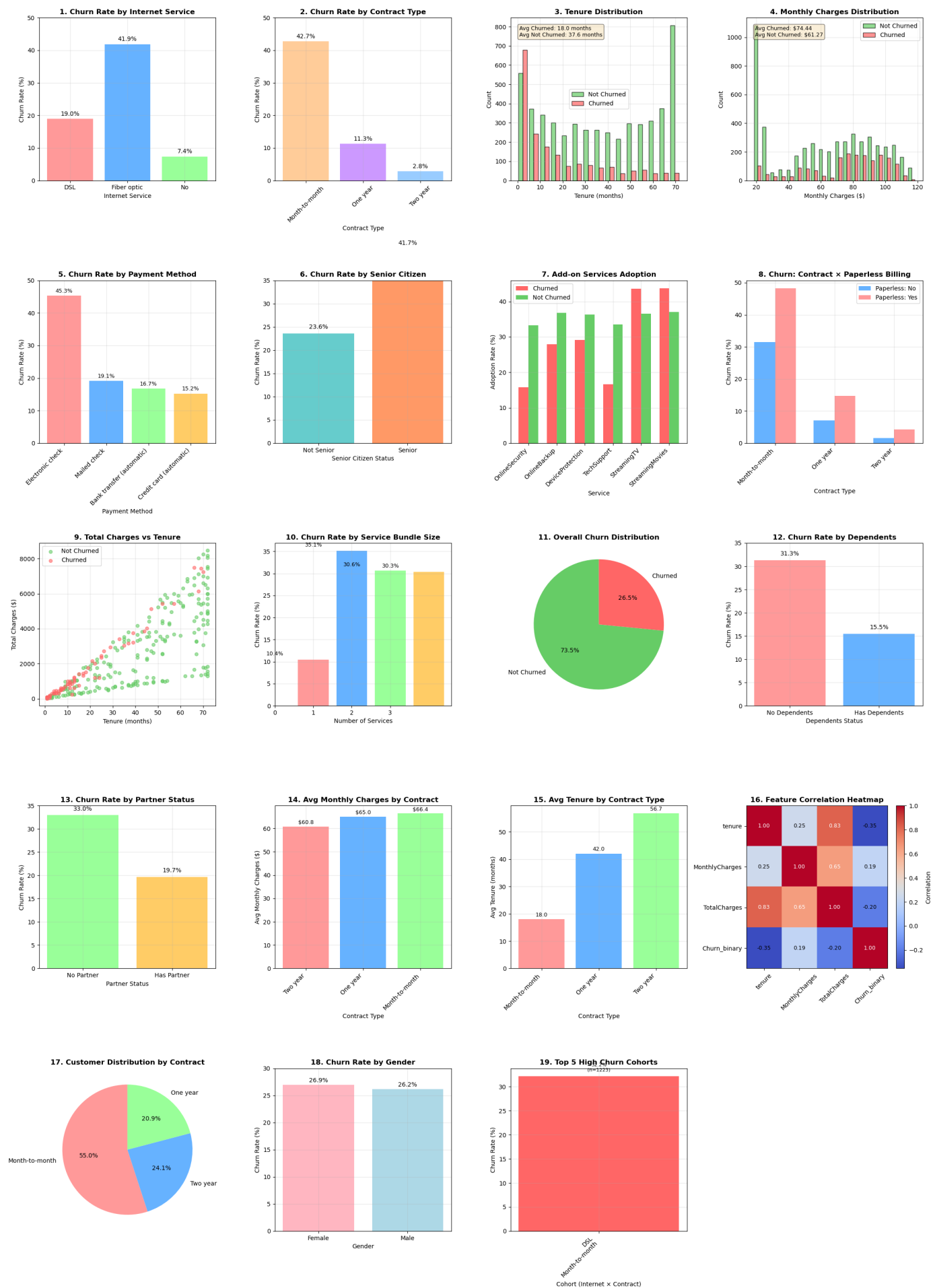


```
cohort_names = [x[0] for x in top_5]
churn_rates = [x[1] for x in top_5]
sizes = [x[2] for x in top_5]

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'])

```

```

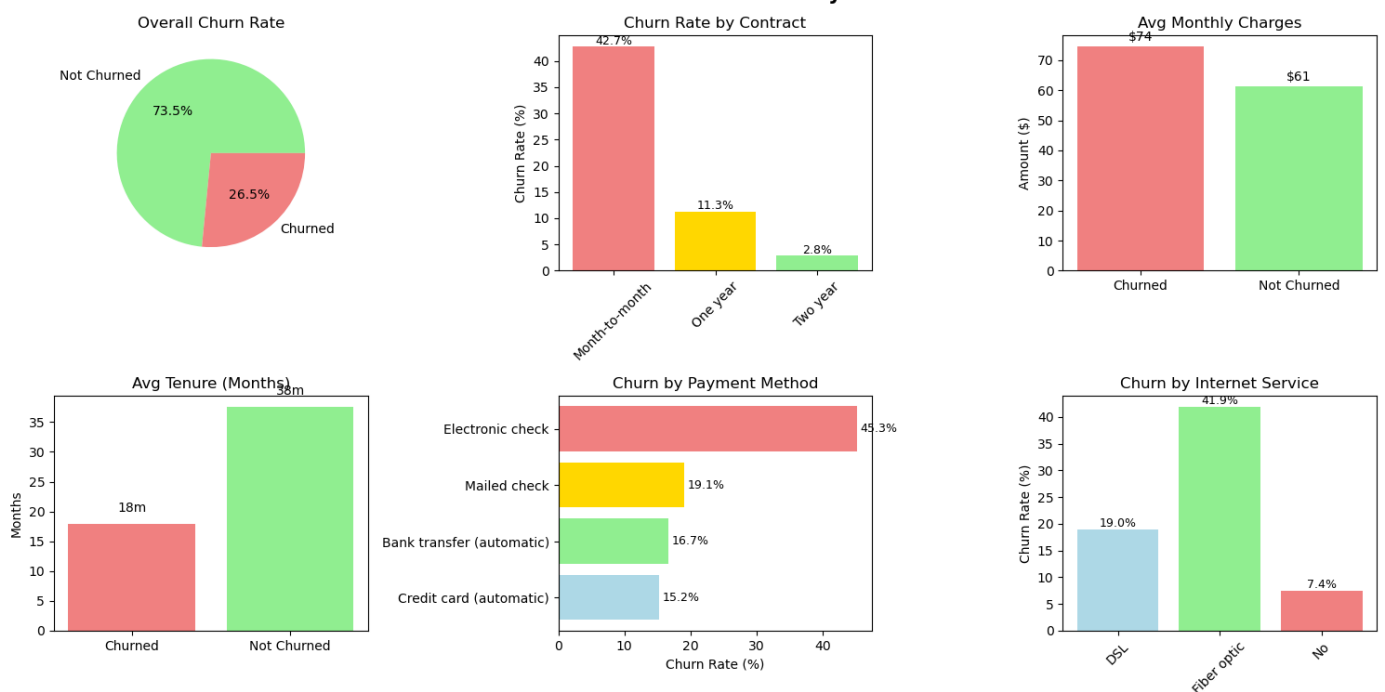
axes[1,1].set_title('Churn by Payment Method')
axes[1,1].set_xlabel('Churn Rate (%)')
axes[1,1].set_yticks(range(len(payment_churn)))
axes[1,1].set_yticklabels(payment_churn.index)
# Add labels
for i, v in enumerate(payment_churn.values):
    axes[1,1].text(v + 0.5, i, f'{v:.1f}%', va='center', fontsize=9)

# 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)

```

Telco Customer Churn Analysis



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='%1.1f%%', 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', 'y'])
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='%1.1f%%', 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'] == 'Yes').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_churn[1]:.1f}%")
print(f"    Online Security - With: {online_security_churn[0]:.1f}%, Without: {online_security_churn[1]:.1f}%")
print(f"    Online Backup - With: {online_backup_churn[0]:.1f}%, Without: {online_backup_churn[1]:.1f}%")

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}× 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}%")

```

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TELCO CUSTOMER CHURN ANALYSIS - KEY INSIGHTS

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- Overall Churn Rate: 26.5%
Total Customers: 7,043
Churned Customers: 1,869
- Tenure Impact:
Short-tenure (<12 months) churn: 48.3%
Long-tenure (24+ months) churn: 14.3%
Short-tenure customers are 3.4× more likely to churn
- 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 2.7× 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 1.7× more revenue

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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%

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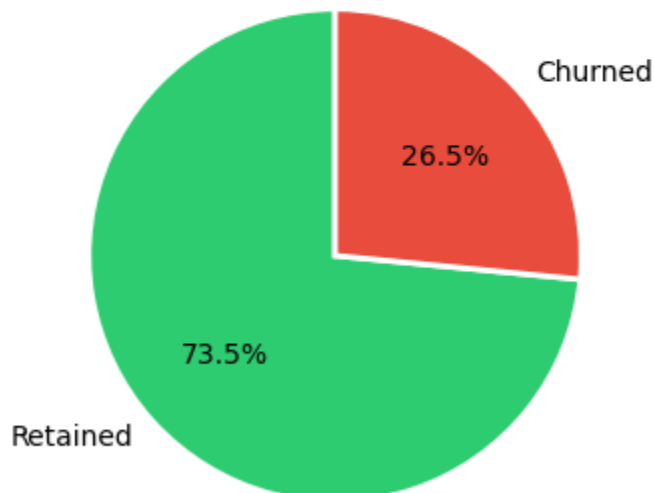
In [32]:

```
# 1. Overall Churn Rate - Pie Chart
plt.figure(figsize=(7, 4))
churn_counts = df['Churn'].value_counts()
colors = ['#2ecc71', '#e74c3c']
plt.pie(churn_counts, labels=['Retained', 'Churned'], autopct='%1.1f%%',
        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()

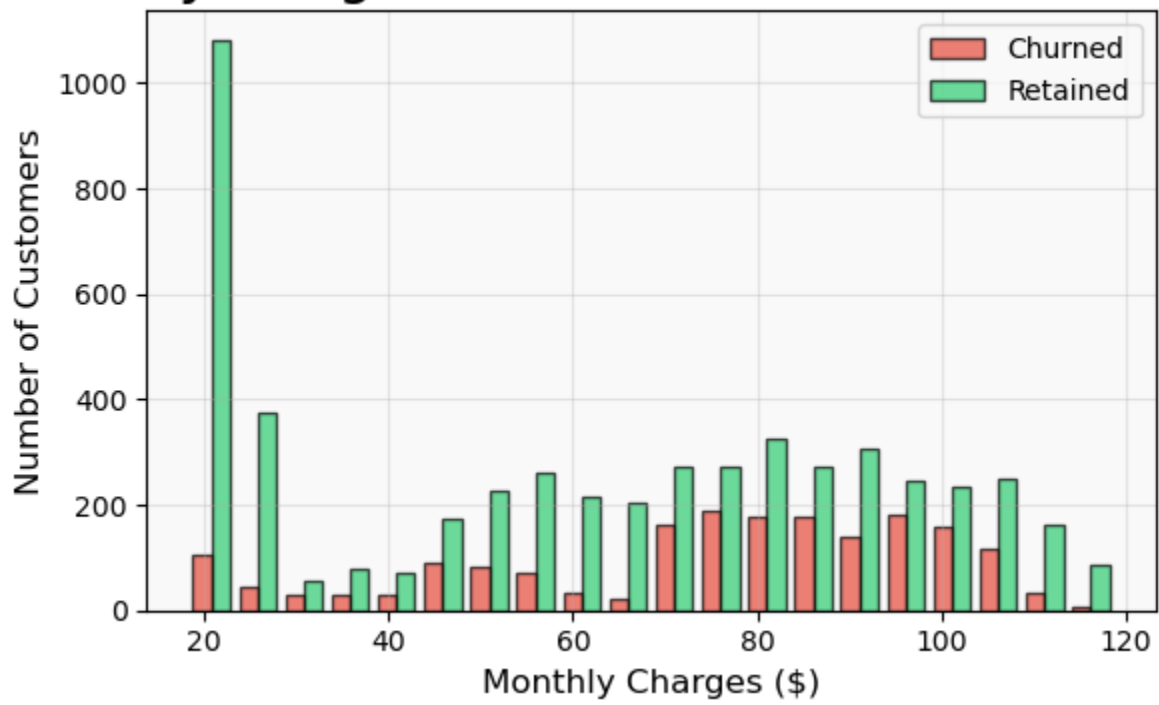
# 6. Monthly Charges Analysis - Histogram
plt.figure(figsize=(6, 4))
churned = df[df['Churn'] == 'Yes']['MonthlyCharges']
retained = df[df['Churn'] == 'No']['MonthlyCharges']

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='
plt.xlabel('Monthly Charges ($)')
plt.ylabel('Number of Customers')
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 []:

