task-1

August 3, 2023

1 Task 1

1. Churn Prediction in Telecom Industry using Logistic Regression

[1]:
```
import warnings
warnings.filterwarnings('ignore')
```

[2]:
```
import numpy as np
```

[3]:
```
import pandas as pd
```

[4]:
```
import matplotlib.pyplot as plt
```

[5]:
```
import seaborn as sns
```

[6]:
```
churn_data = pd.read_csv('churn_data.csv')
```

[7]:
```
churn_data.head()
```

[7]:

	customerID	tenure	PhoneService	Contract	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges	Churn
0	7590-VHVEG	1	No	Month-to-month	Yes	Electronic check	29.85	29.85	No
1	5575-GNVDE	34	Yes	One year	No	Mailed check	56.95	1889.5	No
2	3668-QPYBK	2	Yes	Month-to-month	Yes	Mailed check	53.85	108.15	Yes
3	7795-CFOCW	45	No	One year	No	Bank transfer (automatic)	42.30	1840.75	No
4	9237-HQITU	2	Yes	Month-to-month	Yes	Electronic check	70.70	151.65	Yes

[8]:
```
customer_data = pd.read_csv('customer_data.csv')
customer_data
```

[8]:

	customerID	gender	SeniorCitizen	Partner	Dependents
0	7590-VHVEG	Female	0	Yes	No

```
1     5575-GNVDE    Male              0      No         No
2     3668-QPYBK    Male              0      No         No
3     7795-CFOCW    Male              0      No         No
4     9237-HQITU  Female              0      No         No
…          …     …             …     …        …
7038  6840-RESVB    Male              0     Yes        Yes
7039  2234-XADUH  Female              0     Yes        Yes
7040  4801-JZAZL  Female              0     Yes        Yes
7041  8361-LTMKD    Male              1     Yes         No
7042  3186-AJIEK    Male              0      No         No

[7043 rows x 5 columns]
```

[9]:
```
internet_data = pd.read_csv('internet_data.csv')
internet_data
```

[9]:
```
      customerID     MultipleLines InternetService OnlineSecurity  \
0     7590-VHVEG  No phone service             DSL             No
1     5575-GNVDE                No             DSL            Yes
2     3668-QPYBK                No             DSL            Yes
3     7795-CFOCW  No phone service             DSL            Yes
4     9237-HQITU                No     Fiber optic             No
…          …                …             …             …
7038  6840-RESVB               Yes             DSL            Yes
7039  2234-XADUH               Yes     Fiber optic             No
7040  4801-JZAZL  No phone service             DSL            Yes
7041  8361-LTMKD               Yes     Fiber optic             No
7042  3186-AJIEK                No     Fiber optic            Yes

     OnlineBackup DeviceProtection TechSupport StreamingTV StreamingMovies
0             Yes                No          No          No              No
1              No               Yes          No          No              No
2             Yes                No          No          No              No
3              No               Yes         Yes          No              No
4              No                No          No          No              No
…             …                …           …           …               …
7038           No               Yes         Yes         Yes             Yes
7039          Yes               Yes          No         Yes             Yes
7040           No                No          No          No              No
7041           No                No          No          No              No
7042           No               Yes         Yes         Yes             Yes

[7043 rows x 9 columns]
```

2 Merging tables using customer id

[10]:
```
df = pd.merge(churn_data, customer_data, how='inner', on = 'customerID')
```

[11]:
```
telecom_df = pd.merge(df, internet_data, how='inner', on='customerID')
```

[12]:
```
telecom_df
```

[12]:

	customerID	tenure	PhoneService	Contract	PaperlessBilling	\
0	7590-VHVEG	1	No	Month-to-month	Yes	
1	5575-GNVDE	34	Yes	One year	No	
2	3668-QPYBK	2	Yes	Month-to-month	Yes	
3	7795-CFOCW	45	No	One year	No	
4	9237-HQITU	2	Yes	Month-to-month	Yes	
…	…	…	…	…	…	
7038	6840-RESVB	24	Yes	One year	Yes	
7039	2234-XADUH	72	Yes	One year	Yes	
7040	4801-JZAZL	11	No	Month-to-month	Yes	
7041	8361-LTMKD	4	Yes	Month-to-month	Yes	
7042	3186-AJIEK	66	Yes	Two year	Yes	

	PaymentMethod	MonthlyCharges	TotalCharges	Churn	gender	\
0	Electronic check	29.85	29.85	No	Female	
1	Mailed check	56.95	1889.5	No	Male	
2	Mailed check	53.85	108.15	Yes	Male	
3	Bank transfer (automatic)	42.30	1840.75	No	Male	
4	Electronic check	70.70	151.65	Yes	Female	
…	…	…	…	…	…	
7038	Mailed check	84.80	1990.5	No	Male	
7039	Credit card (automatic)	103.20	7362.9	No	Female	
7040	Electronic check	29.60	346.45	No	Female	
7041	Mailed check	74.40	306.6	Yes	Male	
7042	Bank transfer (automatic)	105.65	6844.5	No	Male	

	…	Partner	Dependents	MultipleLines	InternetService	\
0	…	Yes	No	No phone service	DSL	
1	…	No	No	No	DSL	
2	…	No	No	No	DSL	
3	…	No	No	No phone service	DSL	
4	…	No	No	No	Fiber optic	
…	…	…	…	…	…	
7038	…	Yes	Yes	Yes	DSL	
7039	…	Yes	Yes	Yes	Fiber optic	
7040	…	Yes	Yes	No phone service	DSL	
7041	…	Yes	No	Yes	Fiber optic	
7042	…	No	No	No	Fiber optic	

```
     OnlineSecurity OnlineBackup DeviceProtection TechSupport StreamingTV  \
0                No          Yes               No          No          No
1               Yes           No              Yes          No          No
2               Yes          Yes               No          No          No
3               Yes           No              Yes         Yes          No
4                No           No               No          No          No
…               …            …                …           …           …
7038            Yes           No              Yes         Yes         Yes
7039             No          Yes              Yes          No         Yes
7040            Yes           No               No          No          No
7041             No           No               No          No          No
7042            Yes           No              Yes         Yes         Yes

     StreamingMovies
0                 No
1                 No
2                 No
3                 No
4                 No
…                 …
7038              Yes
7039              Yes
7040               No
7041               No
7042              Yes

[7043 rows x 21 columns]
```

[13]:
```
for col in telecom_df.columns:
    print(col)
```

```
customerID
tenure
PhoneService
Contract
PaperlessBilling
PaymentMethod
MonthlyCharges
TotalCharges
Churn
gender
SeniorCitizen
Partner
Dependents
MultipleLines
InternetService
OnlineSecurity
```

```
OnlineBackup
DeviceProtection
TechSupport
StreamingTV
StreamingMovies
```

[14]:
```
telecom_df.shape
```

[14]:
```
(7043, 21)
```

[15]:
```
telecom_df.describe()
```

[15]:

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

[16]:
```
telecom_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 7043 entries, 0 to 7042
Data columns (total 21 columns):
 #   Column            Non-Null Count  Dtype
---  ------            --------------  -----
 0   customerID        7043 non-null   object
 1   tenure            7043 non-null   int64
 2   PhoneService      7043 non-null   object
 3   Contract          7043 non-null   object
 4   PaperlessBilling  7043 non-null   object
 5   PaymentMethod     7043 non-null   object
 6   MonthlyCharges    7043 non-null   float64
 7   TotalCharges      7043 non-null   object
 8   Churn             7043 non-null   object
 9   gender            7043 non-null   object
 10  SeniorCitizen     7043 non-null   int64
 11  Partner           7043 non-null   object
 12  Dependents        7043 non-null   object
 13  MultipleLines     7043 non-null   object
 14  InternetService   7043 non-null   object
 15  OnlineSecurity    7043 non-null   object
 16  OnlineBackup      7043 non-null   object
 17  DeviceProtection  7043 non-null   object
 18  TechSupport       7043 non-null   object
```

```
 19  StreamingTV       7043 non-null   object
 20  StreamingMovies   7043 non-null   object
dtypes: float64(1), int64(2), object(18)
memory usage: 1.2+ MB
```

3 Data Cleaning

[17]:
```
telecom_df.isnull().sum()*100/telecom_df.shape[0]
```

[17]:
```
customerID          0.0
tenure              0.0
PhoneService        0.0
Contract            0.0
PaperlessBilling    0.0
PaymentMethod       0.0
MonthlyCharges      0.0
TotalCharges        0.0
Churn               0.0
gender              0.0
SeniorCitizen       0.0
Partner             0.0
Dependents          0.0
MultipleLines       0.0
InternetService     0.0
OnlineSecurity      0.0
OnlineBackup        0.0
DeviceProtection    0.0
TechSupport         0.0
StreamingTV         0.0
StreamingMovies     0.0
dtype: float64
```

[18]:
```
telecom_df['TotalCharges'].describe()
```

[18]:
```
count     7043
unique    6531
top
freq        11
Name: TotalCharges, dtype: object
```

[19]:
```
print(telecom_df['MonthlyCharges'])
```

```
0        29.85
1        56.95
2        53.85
3        42.30
4        70.70
```

```
          …
7038     84.80
7039    103.20
7040     29.60
7041     74.40
7042    105.65
Name: MonthlyCharges, Length: 7043, dtype: float64
```

[20]:
```
telecom_df['TotalCharges'] = telecom_df['TotalCharges'].replace(' ', np.nan)
telecom_df['TotalCharges'] = pd.to_numeric(telecom_df['TotalCharges'])
```

[21]:
```
value = (telecom_df['TotalCharges']/telecom_df['MonthlyCharges']).
↪median()*telecom_df['MonthlyCharges']
```

[22]:
```
telecom_df['TotalCharges'].describe()
```

[22]:
```
count    7032.000000
mean     2283.300441
std      2266.771362
min        18.800000
25%       401.450000
50%      1397.475000
75%      3794.737500
max      8684.800000
Name: TotalCharges, dtype: float64
```

[23]:
```
telecom_df['TotalCharges'] = value.where(telecom_df['TotalCharges'] == np.nan,␣
↪other =telecom_df['TotalCharges'])
```

[24]:
```
telecom_df['TotalCharges'].describe()
```

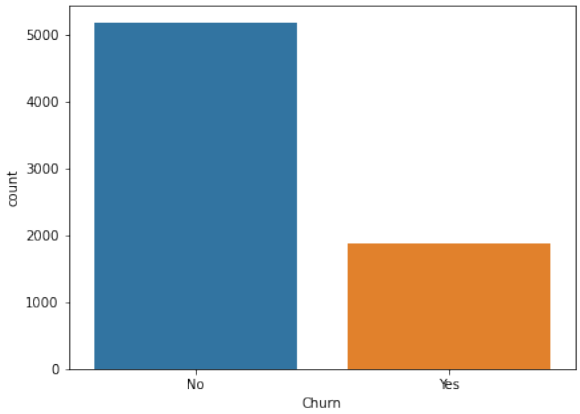
[24]:
```
count    7032.000000
mean     2283.300441
std      2266.771362
min        18.800000
25%       401.450000
50%      1397.475000
75%      3794.737500
max      8684.800000
Name: TotalCharges, dtype: float64
```

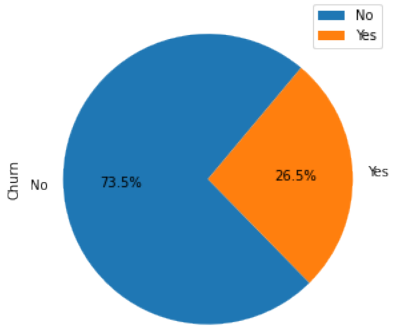
4 Data Analysis

[25]:
```
telecom_df.Churn.describe()
```

[25]:
```
count      7043
unique        2
top          No
freq       5174
Name: Churn, dtype: object
```

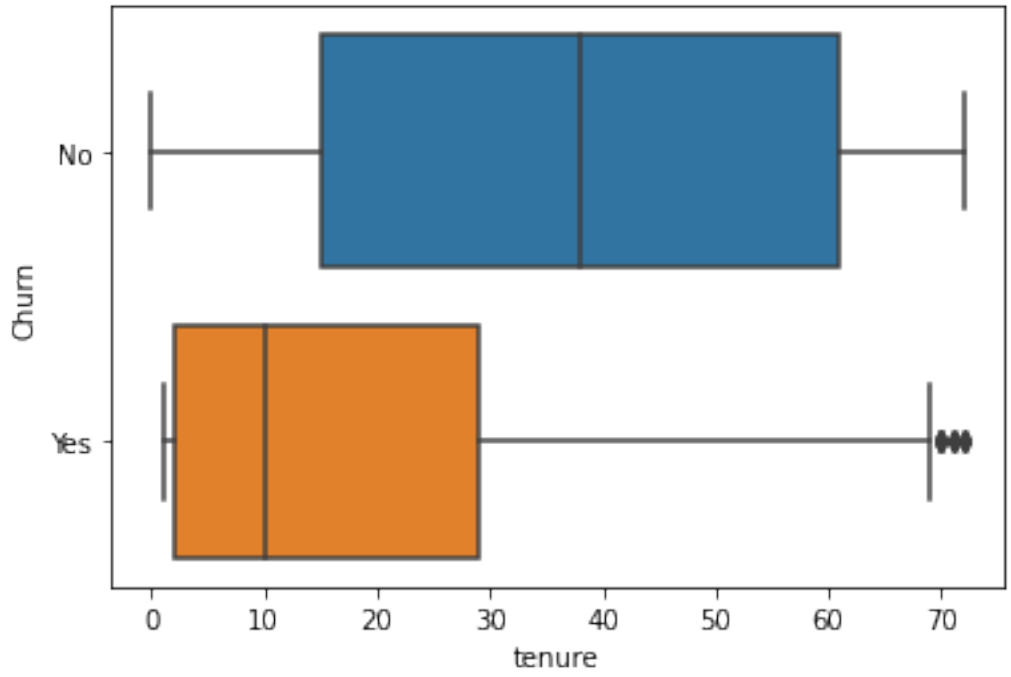
[26]:
```
fig, axs = plt.subplots(1,2, figsize = (15,5))
plt1 = sns.countplot(telecom_df['Churn'], ax = axs[0])

pie_churn = pd.DataFrame(telecom_df['Churn'].value_counts())
pie_churn.plot.pie( subplots=True,labels = pie_churn.index.values, autopct='%1.
 ↪1f%%', figsize = (15,5), startangle= 50, ax = axs[1])
# Unsquish the pie.

plt.gca().set_aspect('equal')
plt.show()
```





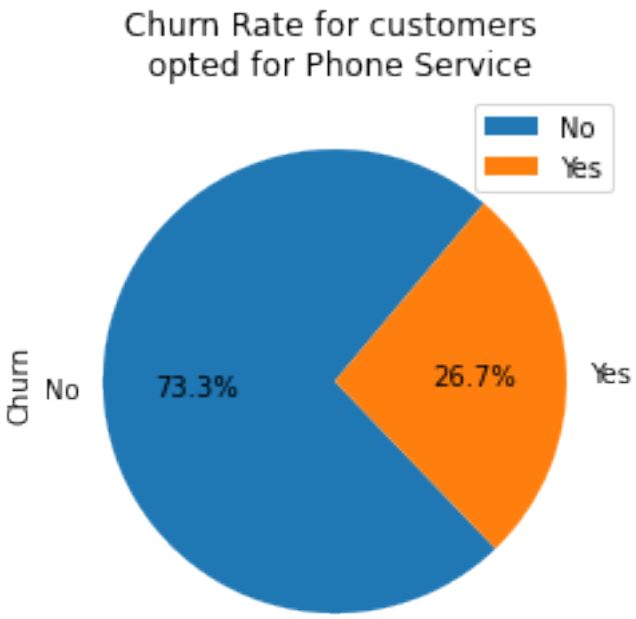
5 Tenure

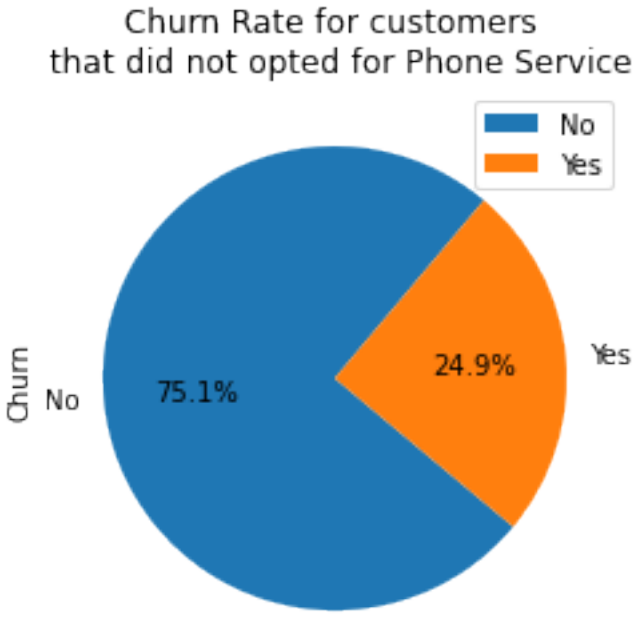
[27]:
```
sns.boxplot(x = 'tenure', y = 'Churn', data = telecom_df)
plt.show()
```



6 Phone Service

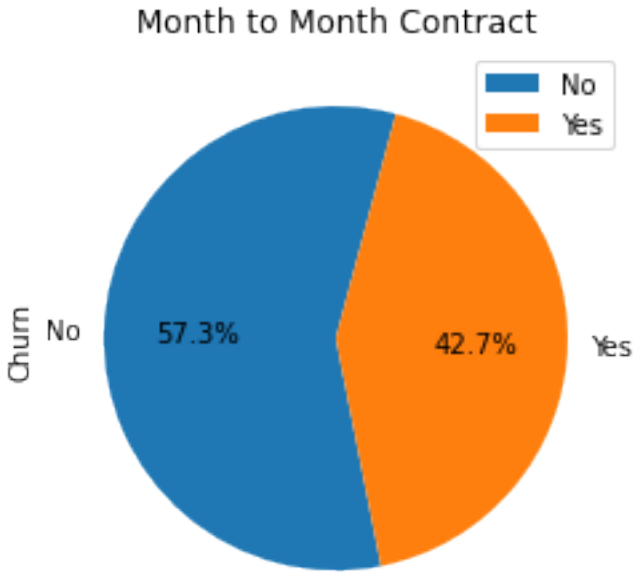
[28]:
```
pie_PhoneService_Yes = pd.DataFrame(telecom_df[telecom_df['PhoneService'] ==␣
 ↪"Yes"]['Churn'].value_counts())
pie_PhoneService_Yes.plot.pie(subplots=True, labels = pie_PhoneService_Yes.
 ↪index.values, autopct='%1.1f%%', startangle= 50 )
plt.title('Churn Rate for customers \n opted for Phone Service')
plt.gca().set_aspect('equal')

pie_PhoneService_No = pd.DataFrame(telecom_df[telecom_df['PhoneService'] ==␣
 ↪"No"]['Churn'].value_counts())
pie_PhoneService_No.plot.pie(subplots=True, labels = pie_PhoneService_Yes.index.
 ↪values, autopct='%1.1f%%', startangle= 50)
plt.title('Churn Rate for customers \n that did not opted for Phone Service')
plt.gca().set_aspect('equal')

plt.show()
```


Churn Rate for customers
opted for Phone Service
No
Yes
Churn
No
73.3%
26.7%
Yes

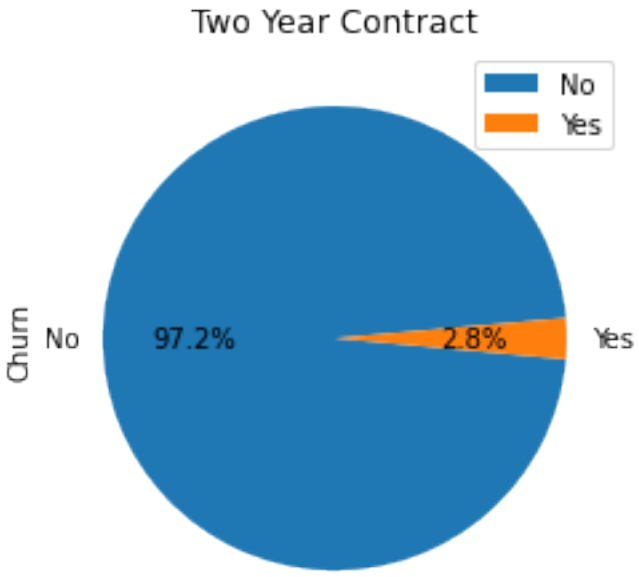

Churn Rate for customers
that did not opted for Phone Service
No
Yes
Churn
No
75.1%
24.9%
Yes

7 Contract

[29]:
```
pie_Contract_m2m = pd.DataFrame(telecom_df[telecom_df['Contract'] ==␣
↪"Month-to-month"]['Churn'].value_counts())
pie_Contract_m2m.plot.pie(subplots=True, labels = pie_Contract_m2m.index.
↪values, autopct='%1.1f%%', startangle= 75)
plt.title('Month to Month Contract')
plt.gca().set_aspect('equal')

pie_Contract_1y = pd.DataFrame(telecom_df[telecom_df['Contract'] == "One␣
↪year"]['Churn'].value_counts())
pie_Contract_1y.plot.pie(subplots=True, labels = pie_Contract_1y.index.values,␣
↪autopct='%1.1f%%', startangle= 20)
plt.title('One Year Contract')
plt.gca().set_aspect('equal')

pie_Contract_2y = pd.DataFrame(telecom_df[telecom_df['Contract'] == "Two␣
↪year"]['Churn'].value_counts())
pie_Contract_2y.plot.pie(subplots=True, labels = pie_Contract_2y.index.values,␣
↪autopct='%1.1f%%', startangle= 5)
plt.title('Two Year Contract')
plt.gca().set_aspect('equal')

plt.show()
```



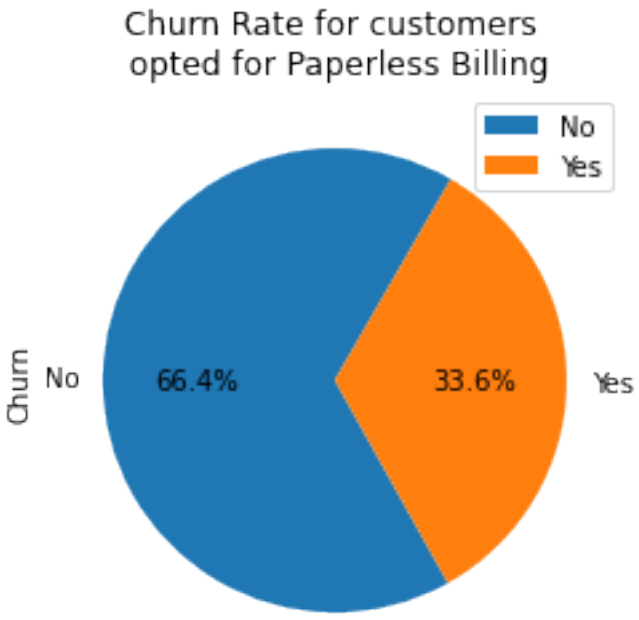

One Year Contract
No
Yes
Churn
No
88.7%
11.3%
Yes

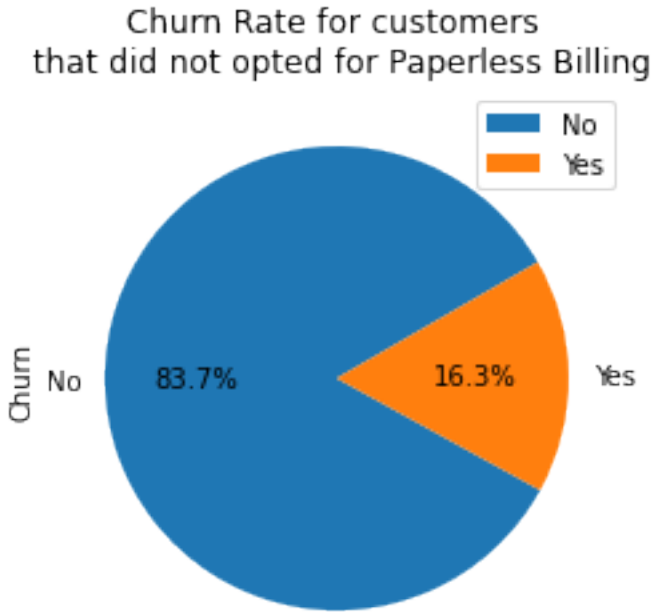

Two Year Contract
No
Yes
Churn
No
97.2%
2.8%
Yes

8 Paperless Bills

[30]:
```
plt.figure(figsize=(15,5))

pie_PaperlessBilling_Yes = pd.
 ↪DataFrame(telecom_df[telecom_df['PaperlessBilling'] == "Yes"]['Churn'].
 ↪value_counts())
pie_PaperlessBilling_Yes.plot.pie(subplots=True, labels =␣
 ↪pie_PaperlessBilling_Yes.index.values, autopct='%1.1f%%', startangle= 60)
plt.title('Churn Rate for customers \n opted for Paperless Billing')
plt.gca().set_aspect('equal')


pie_PaperlessBilling_No = pd.
 ↪DataFrame(telecom_df[telecom_df['PaperlessBilling'] == "No"]['Churn'].
 ↪value_counts())
pie_PaperlessBilling_No.plot.pie(subplots=True, labels =␣
 ↪pie_PaperlessBilling_No.index.values, autopct='%1.1f%%', startangle= 30)
plt.title('Churn Rate for customers \n that did not opted for Paperless␣
 ↪Billing')
plt.gca().set_aspect('equal')

plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```





9 Payment Method

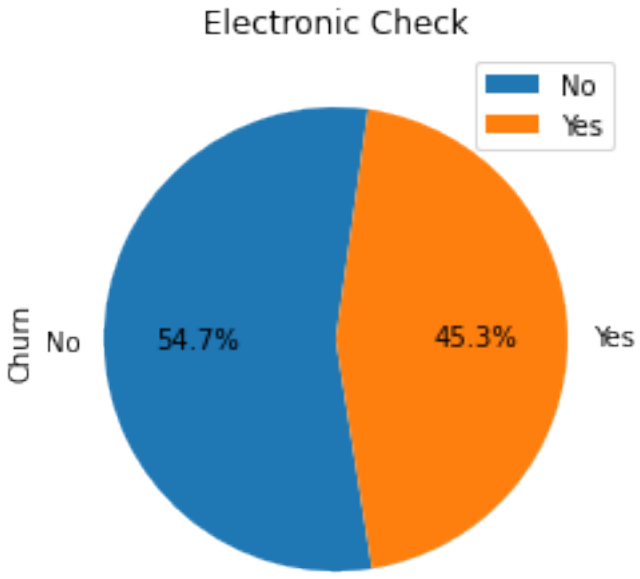
[31]:
```
telecom_df.PaymentMethod.describe()
```

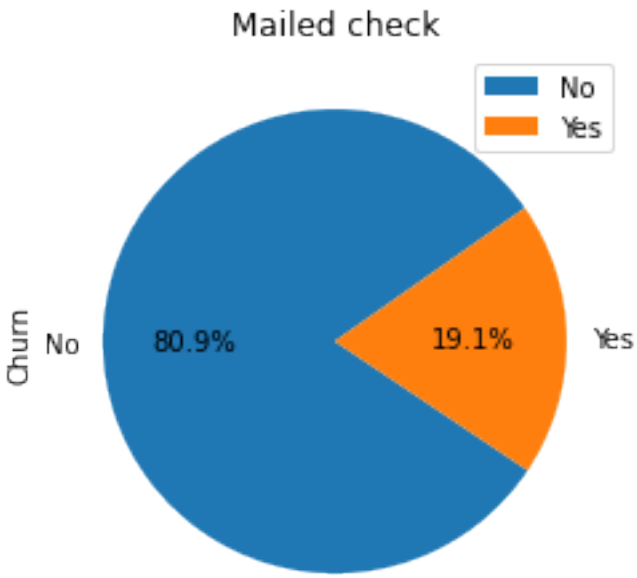
[31]:
```
count                7043
unique                  4
top       Electronic check
freq                 2365
Name: PaymentMethod, dtype: object
```

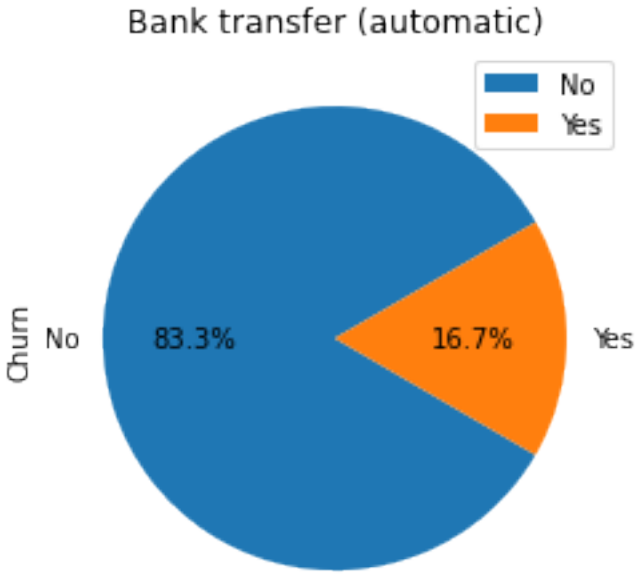
[32]:
```
plt.figure(figsize=(15,10))
pie_PaymentMethod_ec = pd.DataFrame(telecom_df[telecom_df['PaymentMethod'] ==␣
 ↪"Electronic check"]['Churn'].value_counts())
pie_PaymentMethod_ec.plot.pie(subplots=True, labels = pie_PaymentMethod_ec.
 ↪index.values, autopct='%1.1f%%', startangle= 82)
plt.title('Electronic Check')
plt.gca().set_aspect('equal')

pie_PaymentMethod_mc = pd.DataFrame(telecom_df[telecom_df['PaymentMethod'] ==␣
 ↪"Mailed check"]['Churn'].value_counts())
pie_PaymentMethod_mc.plot.pie(subplots=True, labels = pie_PaymentMethod_mc.
 ↪index.values, autopct='%1.1f%%', startangle= 35)
```

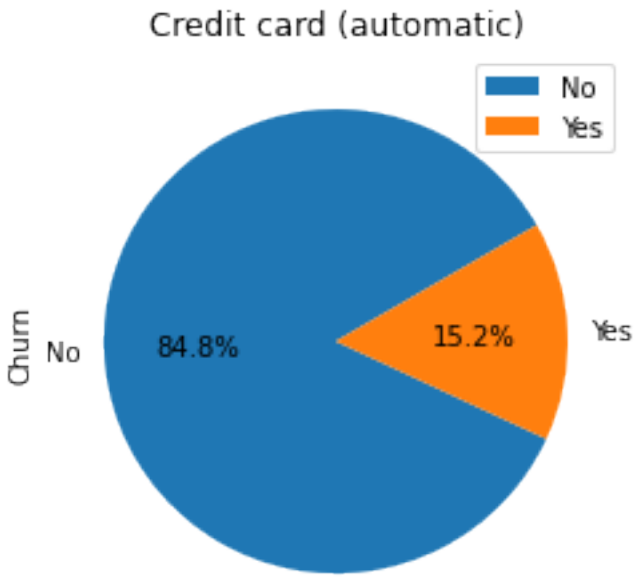
```
plt.title('Mailed check')
plt.gca().set_aspect('equal')

pie_PaymentMethod_bta = pd.DataFrame(telecom_df[telecom_df['PaymentMethod'] ==␣
 ↪"Bank transfer (automatic)"]['Churn'].value_counts())
pie_PaymentMethod_bta.plot.pie(subplots=True, labels = pie_PaymentMethod_bta.
 ↪index.values, autopct='%1.1f%%', startangle= 30)
plt.title('Bank transfer (automatic)')
plt.gca().set_aspect('equal')

pie_PaymentMethod_cca = pd.DataFrame(telecom_df[telecom_df['PaymentMethod'] ==␣
 ↪"Credit card (automatic)"]['Churn'].value_counts())
pie_PaymentMethod_cca.plot.pie(subplots=True, labels = pie_PaymentMethod_cca.
 ↪index.values, autopct='%1.1f%%', startangle= 30)
plt.title('Credit card (automatic)')
plt.gca().set_aspect('equal')

plt.show()
```

```
<Figure size 1080x720 with 0 Axes>
```



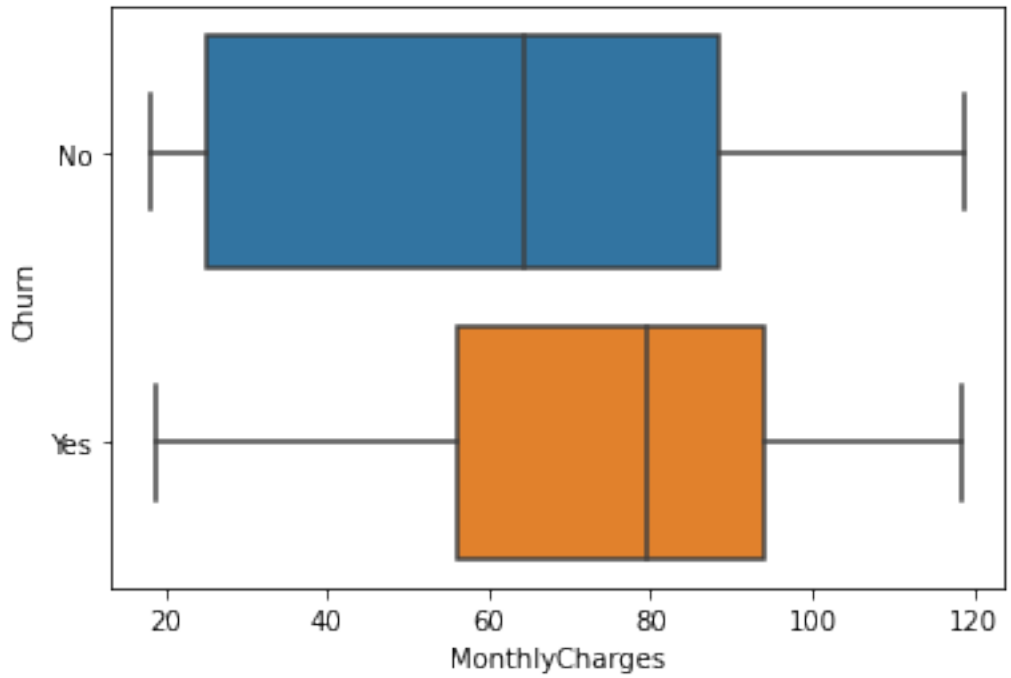

Mailed check
No
Yes
Churn
No
80.9%
19.1%
Yes


Bank transfer (automatic)
No
Yes
Churn
No
83.3%
16.7%
Yes



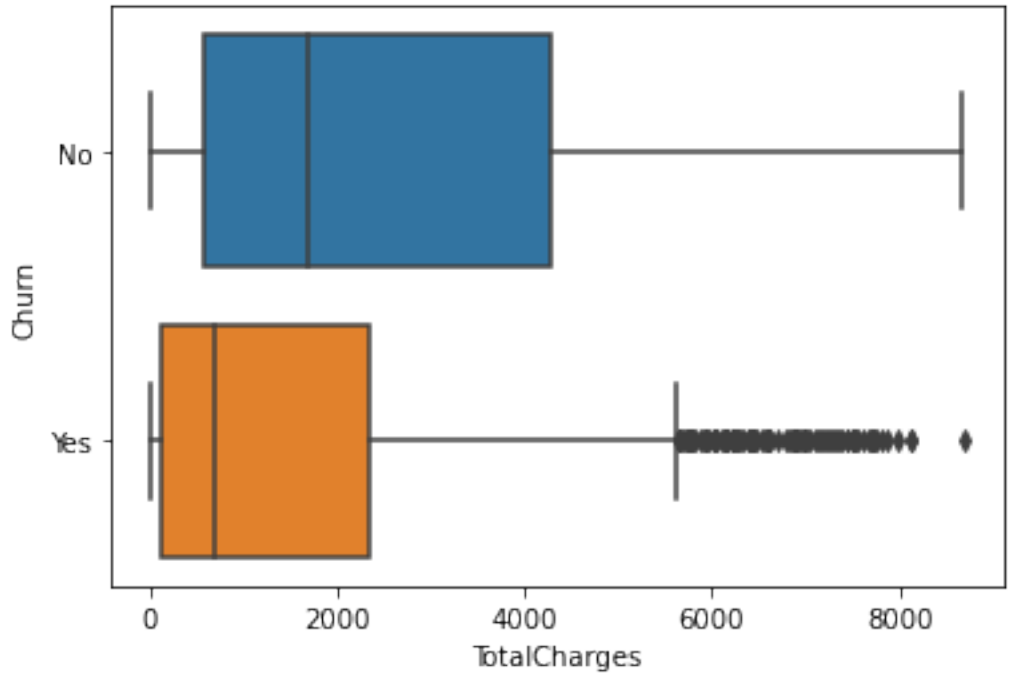
10 Monthly Charges

[33]:
```
sns.boxplot(x = 'MonthlyCharges', y = 'Churn', data = telecom_df)
plt.show()
```



11 Total Charges

[34]:
```
sns.boxplot(x = 'TotalCharges', y= 'Churn', data = telecom_df)
plt.show()
```

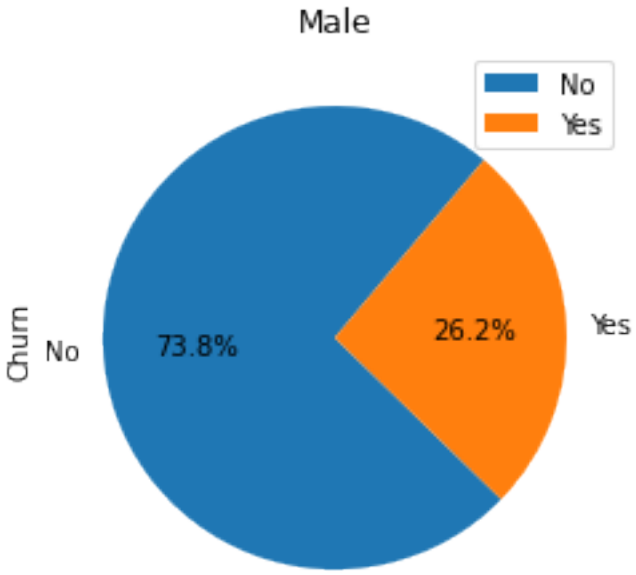


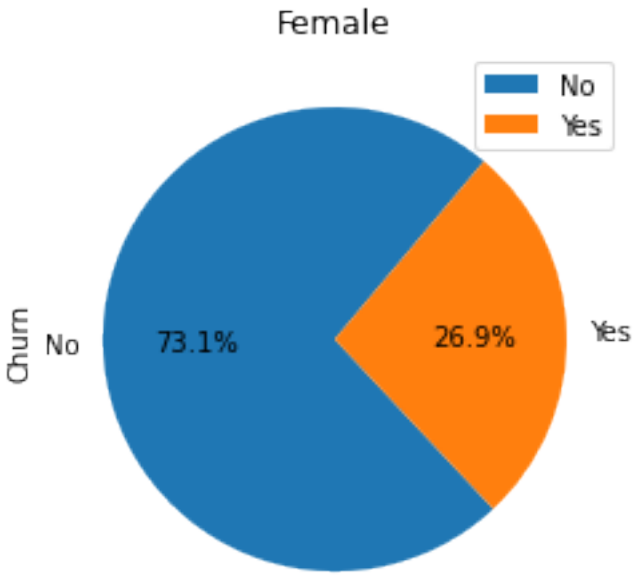
12 Gender

[35]:
```
plt.figure(figsize=(15,5))
pie_Gender_M = pd.DataFrame(telecom_df[telecom_df['gender'] == "Male"]['Churn'].
 ↪value_counts())
pie_Gender_M.plot.pie(subplots = True, labels = pie_Gender_M.index.values,␣
 ↪autopct='%1.1f%%', startangle= 50)
plt.title('Male')
plt.gca().set_aspect('equal')

pie_Gender_F = pd.DataFrame(telecom_df[telecom_df['gender'] ==␣
 ↪"Female"]['Churn'].value_counts())
pie_Gender_F.plot.pie(subplots = True,  labels = pie_Gender_F.index.values,␣
 ↪autopct='%1.1f%%', startangle= 50)
plt.title('Female')
```

```
plt.gca().set_aspect('equal')

plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```

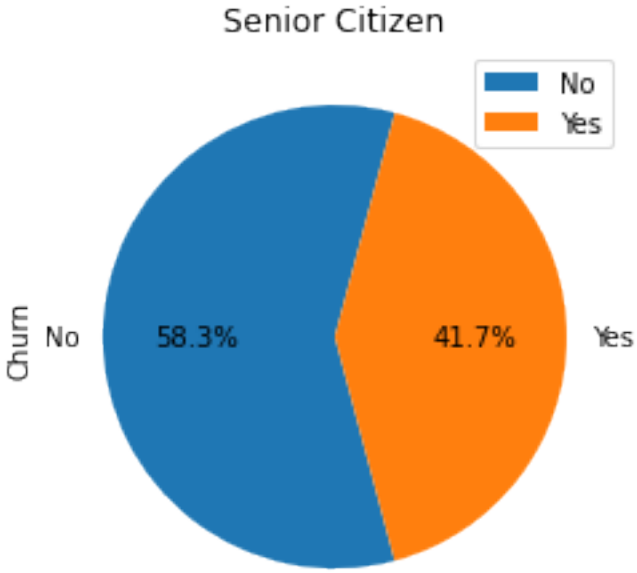


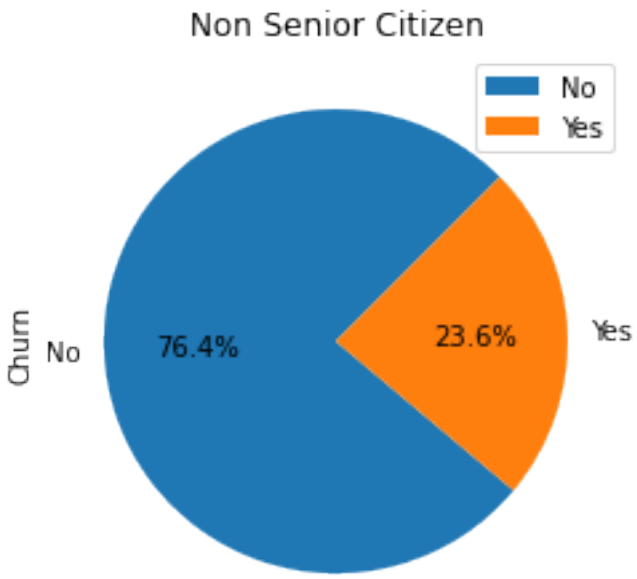


13 Senior Citizen

[36]:
```
plt.figure(figsize=(15,5))
pie_SeniorCitizen_Y = pd.DataFrame(telecom_df[telecom_df['SeniorCitizen'] ==␣
 ↪1]['Churn'].value_counts())
pie_SeniorCitizen_Y.plot.pie(subplots = True, labels = pie_SeniorCitizen_Y.
 ↪index.values, autopct='%1.1f%%', startangle= 75)
plt.title('Senior Citizen')
plt.gca().set_aspect('equal')

pie_SeniorCitizen_N = pd.DataFrame(telecom_df[telecom_df['SeniorCitizen'] ==␣
 ↪0]['Churn'].value_counts())
pie_SeniorCitizen_N.plot.pie(subplots = True, labels = pie_SeniorCitizen_N.
 ↪index.values, autopct='%1.1f%%', startangle= 45)
plt.title('Non Senior Citizen')

plt.gca().set_aspect('equal')
plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```

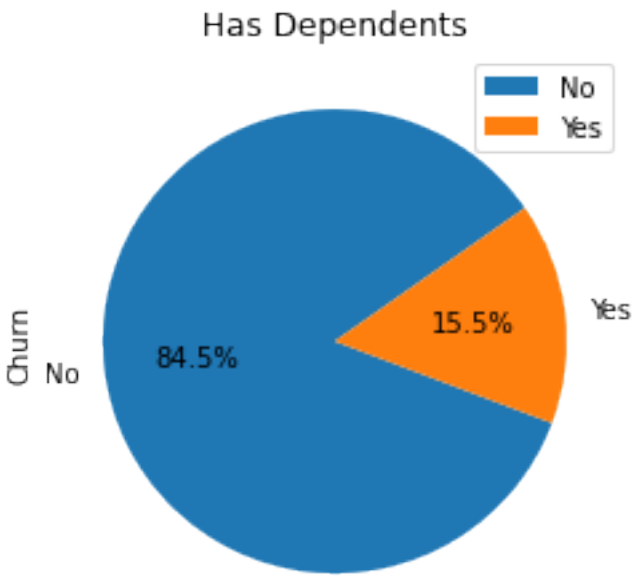


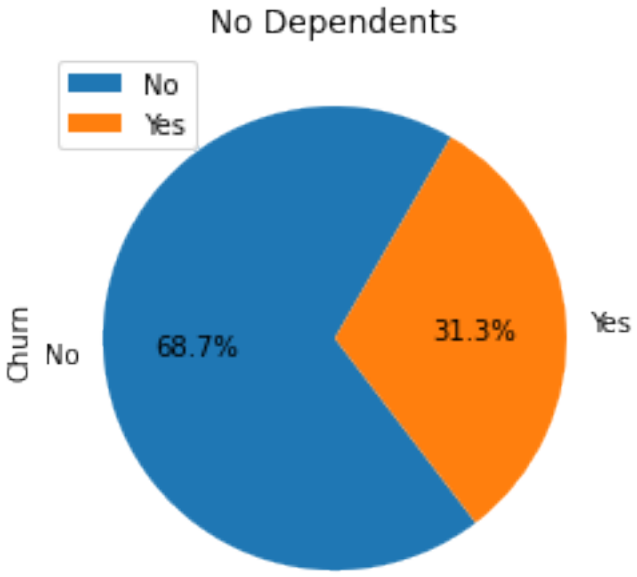


14 Dependents

[37]:
```
plt.figure(figsize=(15,5))
pie_Dependents_Y = pd.DataFrame(telecom_df[telecom_df['Dependents'] ==␣
 ↪'Yes']['Churn'].value_counts())
pie_Dependents_Y.plot.pie(subplots = True,  labels = pie_Dependents_Y.index.
 ↪values, autopct='%1.1f%%', startangle= 35)
plt.title('Has Dependents')
plt.gca().set_aspect('equal')

pie_Dependents_N = pd.DataFrame(telecom_df[telecom_df['Dependents'] ==␣
 ↪'No']['Churn'].value_counts())
pie_Dependents_N.plot.pie(subplots = True,  labels = pie_Dependents_N.index.
 ↪values, autopct='%1.1f%%', startangle= 60)
plt.title('No Dependents')

plt.gca().set_aspect('equal')
plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```

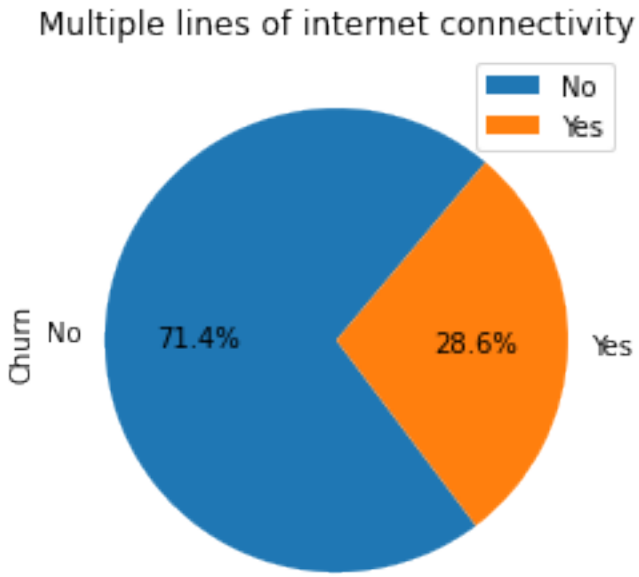

Has Dependents
No
Yes
Churn
No
Yes
84.5%
15.5%

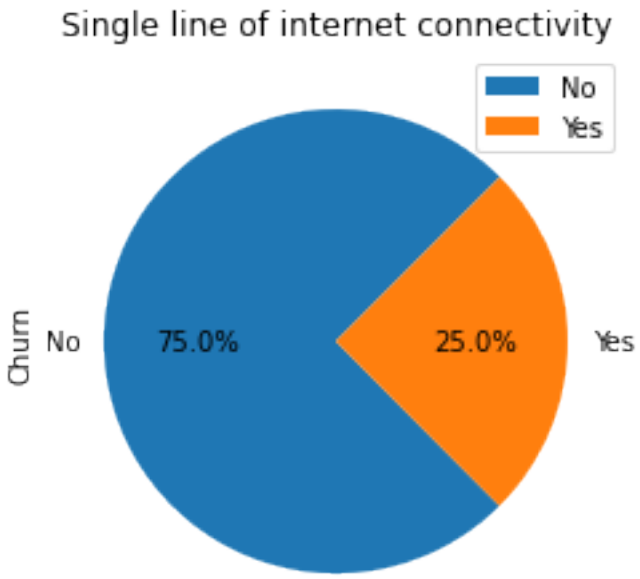

No Dependents
No
Yes
Churn
No
Yes
68.7%
31.3%

15 Multiple Lines

[38]:
```
plt.figure(figsize=(15,5))
pie_MultipleLines_Y = pd.DataFrame(telecom_df[telecom_df['MultipleLines'] ==␣
 ↪'Yes']['Churn'].value_counts())
pie_MultipleLines_Y.plot.pie(subplots = True,  labels = pie_MultipleLines_Y.
 ↪index.values, autopct='%1.1f%%', startangle= 50)
plt.title('Multiple lines of internet connectivity')
plt.gca().set_aspect('equal')

pie_MultipleLines_N = pd.DataFrame(telecom_df[telecom_df['MultipleLines'] ==␣
 ↪'No']['Churn'].value_counts())
pie_MultipleLines_N.plot.pie(subplots = True,  labels = pie_MultipleLines_N.
 ↪index.values, autopct='%1.1f%%', startangle= 45)
plt.title('Single line of internet connectivity')

plt.gca().set_aspect('equal')
plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```





[39]:
```
import jovian
```

[40]:
```
jovian.commit
```

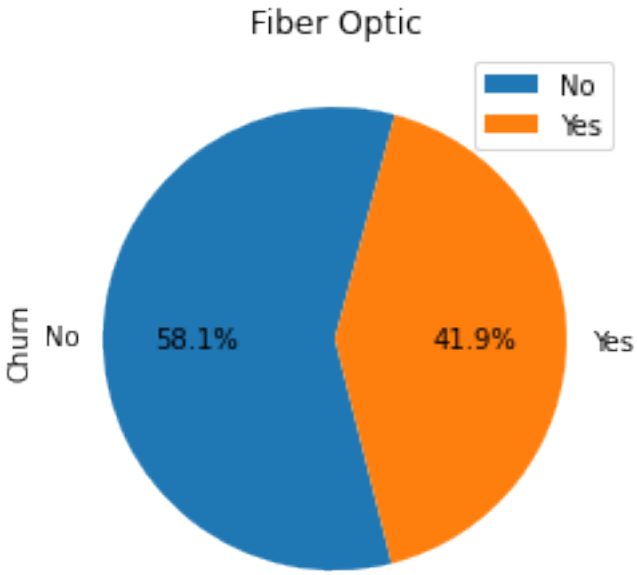
[40]:
```
<function jovian.utils.commit.commit(message=None, files=[], outputs=[],
environment=None, privacy='auto', filename=None, project=None, new_project=None,
git_commit=False, git_message='auto', require_write_access=False, **kwargs)>
```

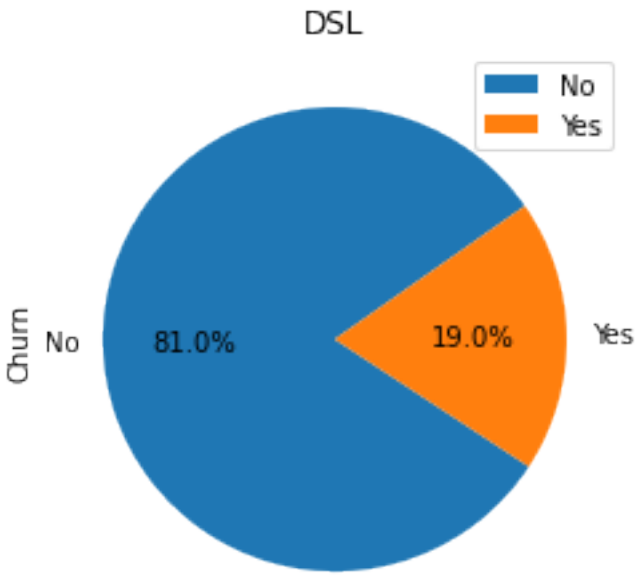
16 Internet service

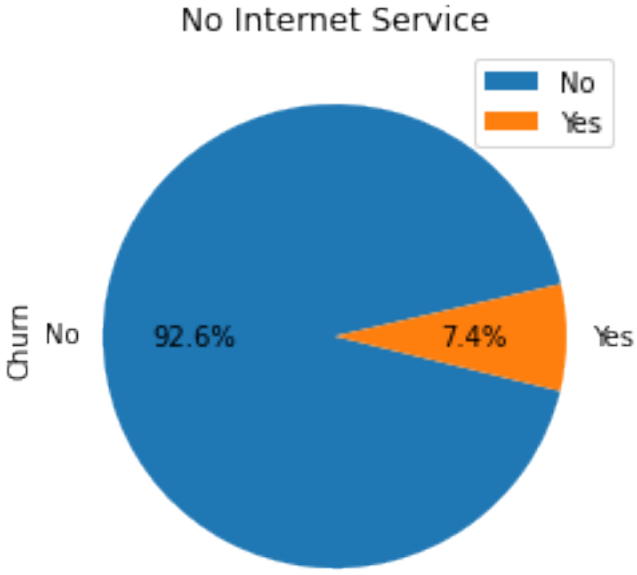
[41]:
```
plt.figure(figsize=(15,5))
pie_InternetService_fo = pd.DataFrame(telecom_df[telecom_df['InternetService']␣
 ↪== "Fiber optic"]['Churn'].value_counts())
pie_InternetService_fo.plot.pie(subplots = True, labels =␣
 ↪pie_InternetService_fo.index.values, autopct='%1.1f%%', startangle= 75)
plt.title('Fiber Optic')
plt.gca().set_aspect('equal')

pie_InternetService_dsl = pd.DataFrame(telecom_df[telecom_df['InternetService']␣
 ↪== "DSL"]['Churn'].value_counts())
pie_InternetService_dsl.plot.pie(subplots = True, labels =␣
 ↪pie_InternetService_dsl.index.values, autopct='%1.1f%%', startangle= 35)
plt.title('DSL')
plt.gca().set_aspect('equal')
```

```
pie_InternetService_no = pd.DataFrame(telecom_df[telecom_df['InternetService']␣
 ↪== "No"]['Churn'].value_counts())
pie_InternetService_no.plot.pie(subplots = True, labels =␣
 ↪pie_InternetService_no.index.values, autopct='%1.1f%%', startangle= 13)
plt.title('No Internet Service')
plt.gca().set_aspect('equal')

plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```



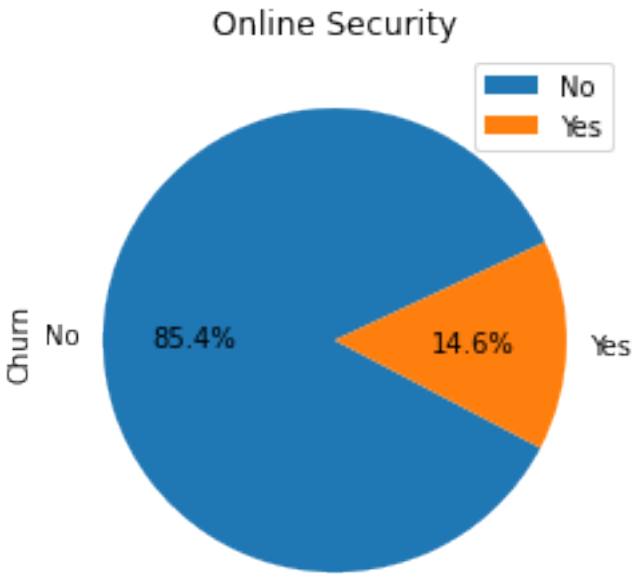

DSL
No
Yes
Churn
No
81.0%
19.0%
Yes

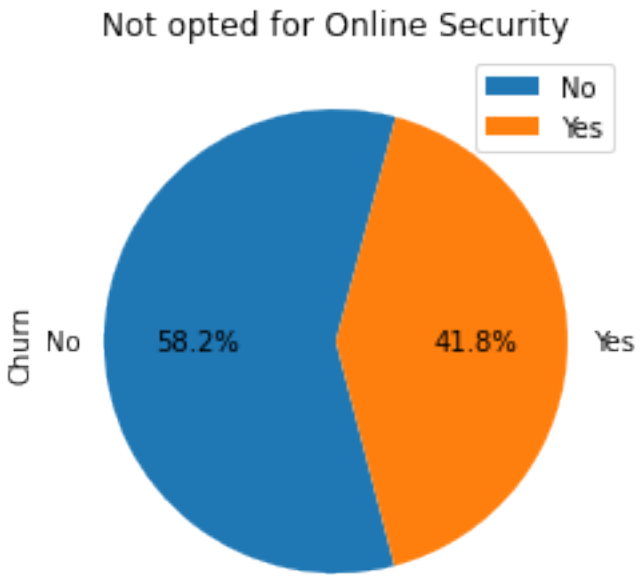

No Internet Service
No
Yes
Churn
No
92.6%
7.4%
Yes

17 Online Security

[42]:
```
plt.figure(figsize=(15,5))
pie_OnlineSecurity_Y = pd.DataFrame(telecom_df[telecom_df['OnlineSecurity'] ==␣
 ↪'Yes']['Churn'].value_counts())
pie_OnlineSecurity_Y.plot.pie(subplots = True,  labels = pie_OnlineSecurity_Y.
 ↪index.values, autopct='%1.1f%%', startangle= 25)
plt.title('Online Security')
plt.gca().set_aspect('equal')

pie_OnlineSecurity_N = pd.DataFrame(telecom_df[telecom_df['OnlineSecurity'] ==␣
 ↪'No']['Churn'].value_counts())
pie_OnlineSecurity_N.plot.pie(subplots = True, labels = pie_OnlineSecurity_N.
 ↪index.values, autopct='%1.1f%%', startangle= 75)
plt.title('Not opted for Online Security')
plt.gca().set_aspect('equal')
plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```

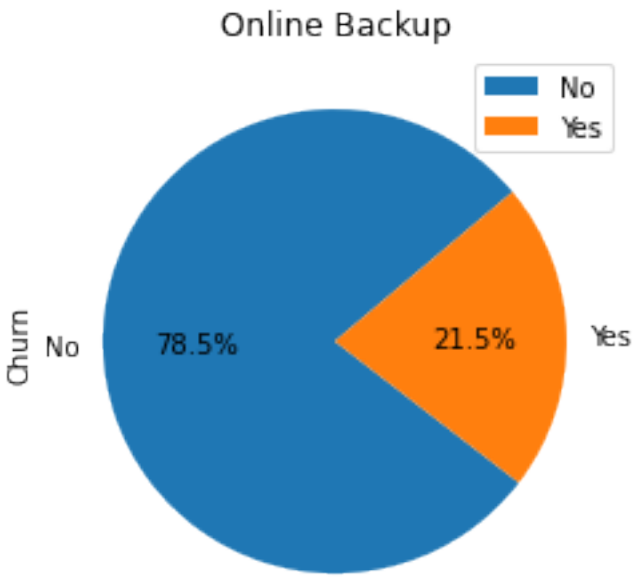


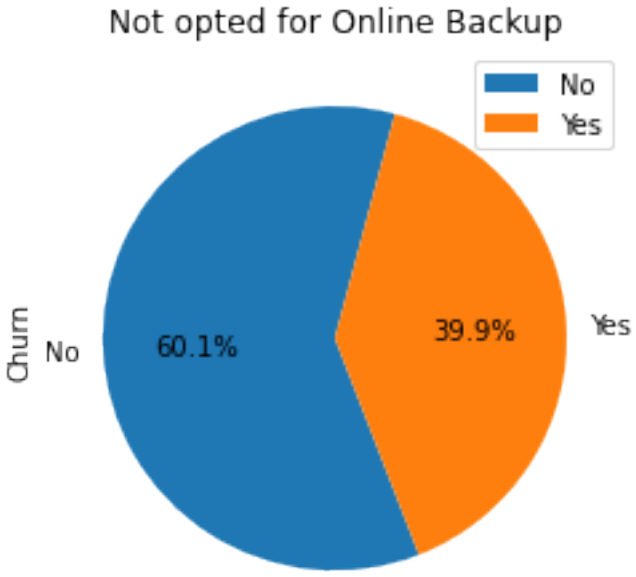


18 online backup

[43]:
```
plt.figure(figsize=(15,5))
pie_OnlineBackup_Y = pd.DataFrame(telecom_df[telecom_df['OnlineBackup'] ==␣
 ↪'Yes']['Churn'].value_counts())
pie_OnlineBackup_Y.plot.pie(subplots = True,  labels = pie_OnlineBackup_Y.index.
 ↪values, autopct='%1.1f%%', startangle= 40)
plt.title('Online Backup')
plt.gca().set_aspect('equal')

pie_OnlineBackup_N = pd.DataFrame(telecom_df[telecom_df['OnlineBackup'] ==␣
 ↪'No']['Churn'].value_counts())
pie_OnlineBackup_N.plot.pie(subplots = True, labels = pie_OnlineBackup_N.index.
 ↪values, autopct='%1.1f%%', startangle= 75)
plt.title('Not opted for Online Backup')
plt.gca().set_aspect('equal')

plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```

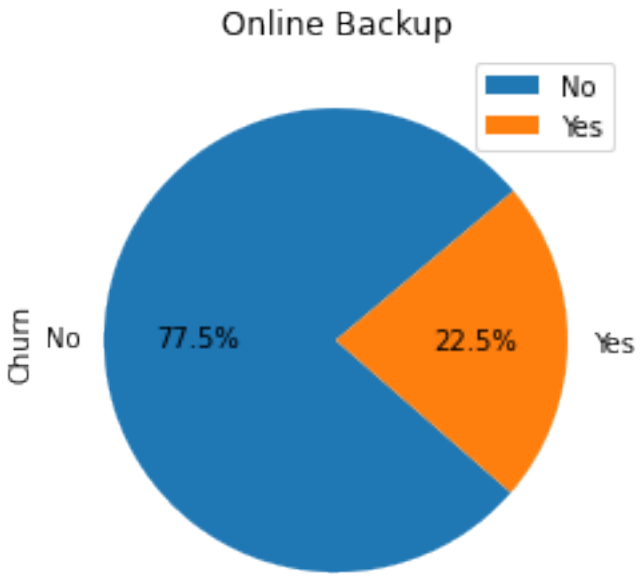

Online Backup
No
Yes
Churn
No
78.5%
21.5%
Yes

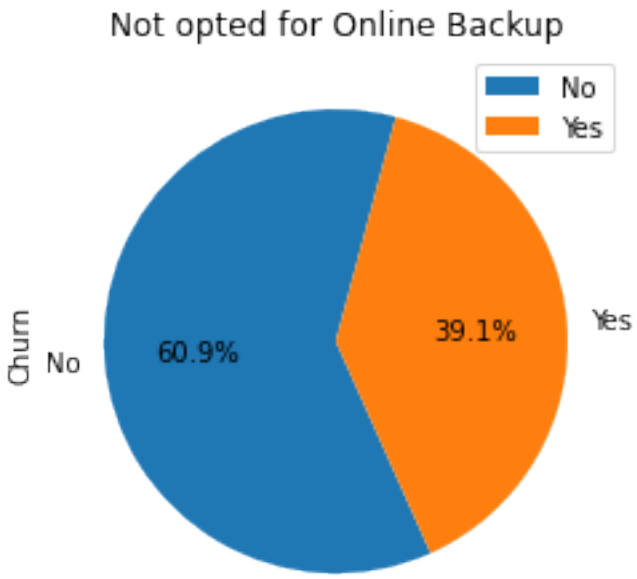

Not opted for Online Backup
No
Yes
Churn
No
60.1%
39.9%
Yes

19 Device Protection

[44]:
```
plt.figure(figsize=(15,5))

pie_DeviceProtection_Y = pd.DataFrame(telecom_df[telecom_df['DeviceProtection']␣
↪== 'Yes']['Churn'].value_counts())
pie_DeviceProtection_Y.plot.pie(subplots = True, labels =␣
↪pie_DeviceProtection_Y.index.values, autopct='%1.1f%%', startangle= 40)
plt.title('Online Backup')
plt.gca().set_aspect('equal')

pie_DeviceProtection_N = pd.DataFrame(telecom_df[telecom_df['DeviceProtection']␣
↪== 'No']['Churn'].value_counts())
pie_DeviceProtection_N.plot.pie(subplots = True, labels =␣
↪pie_DeviceProtection_N.index.values, autopct='%1.1f%%', startangle= 75)
plt.title('Not opted for Online Backup')
plt.gca().set_aspect('equal')
plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```

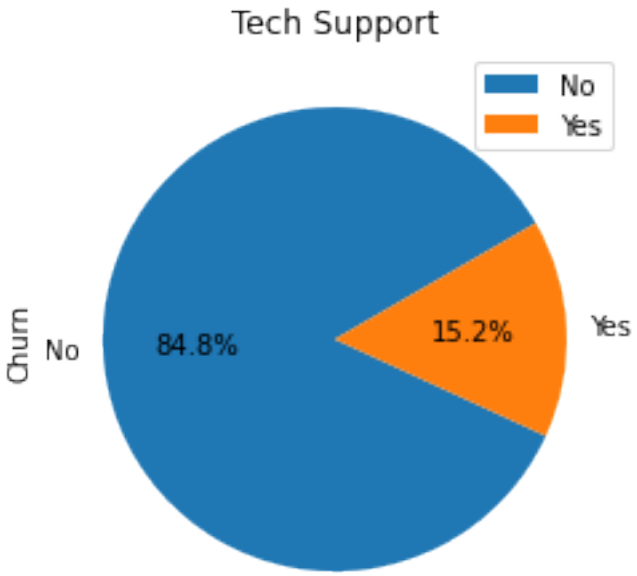


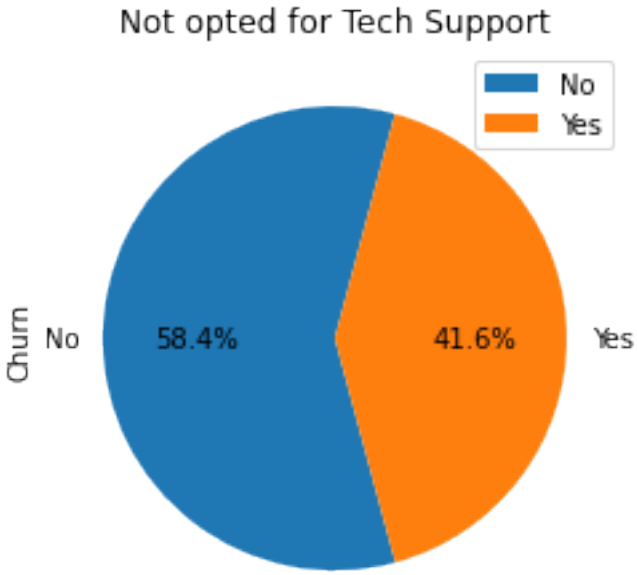


20 Tech Support

[45]:
```
plt.figure(figsize=(15,5))
pie_TechSupport_Y = pd.DataFrame(telecom_df[telecom_df['TechSupport'] ==␣
 ↪'Yes']['Churn'].value_counts())
pie_TechSupport_Y.plot.pie(subplots = True,labels = pie_TechSupport_Y.index.
 ↪values, autopct='%1.1f%%', startangle= 30)
plt.title('Tech Support')
plt.gca().set_aspect('equal')

pie_TechSupport_N = pd.DataFrame(telecom_df[telecom_df['TechSupport'] ==␣
 ↪'No']['Churn'].value_counts())
pie_TechSupport_N.plot.pie(subplots = True, labels = pie_TechSupport_N.index.
 ↪values, autopct='%1.1f%%', startangle= 75)
plt.title('Not opted for Tech Support')

plt.gca().set_aspect('equal')
plt.show()
```

```
<Figure size 1080x360 with 0 Axes>
```


Tech Support
No
Yes
Churn
No
84.8%
15.2%
Yes


Not opted for Tech Support
No
Yes
Churn
No
58.4%
41.6%
Yes

21 Streaming Tv and Movies doesn't make such impact on churning.

22 Model Building

DATA PREPRATION

[46]:
```
# List of variables to map

varlist =  ['PhoneService', 'PaperlessBilling', 'Churn', 'Partner',␣
 ↪'Dependents']

# Defining the map function
def binary_map(x):
    return x.map({'Yes': 1, "No": 0})

# Applying the function to the housing list
telecom_df[varlist] = telecom_df[varlist].apply(binary_map)
```

[47]:
```
telecom_df.head()
```

[47]:

	customerID	tenure	PhoneService	Contract	PaperlessBilling	\
0	7590-VHVEG	1	0	Month-to-month	1	
1	5575-GNVDE	34	1	One year	0	
2	3668-QPYBK	2	1	Month-to-month	1	
3	7795-CFOCW	45	0	One year	0	
4	9237-HQITU	2	1	Month-to-month	1	

	PaymentMethod	MonthlyCharges	TotalCharges	Churn	gender	\
0	Electronic check	29.85	29.85	0	Female	
1	Mailed check	56.95	1889.50	0	Male	
2	Mailed check	53.85	108.15	1	Male	
3	Bank transfer (automatic)	42.30	1840.75	0	Male	
4	Electronic check	70.70	151.65	1	Female	

	…	Partner	Dependents	MultipleLines	InternetService	OnlineSecurity	\
0	…	1	0	No phone service	DSL	No	
1	…	0	0	No	DSL	Yes	
2	…	0	0	No	DSL	Yes	
3	…	0	0	No phone service	DSL	Yes	
4	…	0	0	No	Fiber optic	No	

	OnlineBackup	DeviceProtection	TechSupport	StreamingTV	StreamingMovies
0	Yes	No	No	No	No
1	No	Yes	No	No	No
2	Yes	No	No	No	No
3	No	Yes	Yes	No	No

```
4           No                No           No           No                No

[5 rows x 21 columns]
```

23 For categorical variables with multiple levels, create dummy features (one-hot encoded)

[48]:
```
# Creating a dummy variable for some of the categorical variables and dropping␣
↪the first one.
dummy1 = pd.get_dummies(telecom_df[['Contract', 'PaymentMethod', 'gender',␣
↪'InternetService']], drop_first=True)

# Adding the results to the master dataframe
telecom_df = pd.concat([telecom_df, dummy1], axis=1)
```

[49]:
```
telecom_df.head()
```

[49]:
```
   customerID  tenure  PhoneService        Contract  PaperlessBilling  \
0  7590-VHVEG       1             0  Month-to-month                 1
1  5575-GNVDE      34             1        One year                 0
2  3668-QPYBK       2             1  Month-to-month                 1
3  7795-CFOCW      45             0        One year                 0
4  9237-HQITU       2             1  Month-to-month                 1

               PaymentMethod  MonthlyCharges  TotalCharges  Churn  gender  \
0           Electronic check           29.85         29.85      0  Female
1               Mailed check           56.95       1889.50      0    Male
2               Mailed check           53.85        108.15      1    Male
3  Bank transfer (automatic)           42.30       1840.75      0    Male
4           Electronic check           70.70        151.65      1  Female

   …  StreamingTV  StreamingMovies  Contract_One year Contract_Two year  \
0  …           No               No                  0                 0
1  …           No               No                  1                 0
2  …           No               No                  0                 0
3  …           No               No                  1                 0
4  …           No               No                  0                 0

  PaymentMethod_Credit card (automatic) PaymentMethod_Electronic check  \
0                                     0                              1
1                                     0                              0
2                                     0                              0
3                                     0                              0
4                                     0                              1
```

```
  PaymentMethod_Mailed check gender_Male InternetService_Fiber optic  \
0                            0           0                            0
1                            1           1                            0
2                            1           1                            0
3                            0           1                            0
4                            0           0                            1

  InternetService_No
0                  0
1                  0
2                  0
3                  0
4                  0

[5 rows x 29 columns]
```

[50]:
```
# Creating dummy variables for the remaining categorical variables and dropping␣
 ↪the level with big names.

# Creating dummy variables for the variable 'MultipleLines'
ml = pd.get_dummies(telecom_df['MultipleLines'], prefix='MultipleLines')
# Dropping MultipleLines_No phone service column
ml1 = ml.drop(['MultipleLines_No phone service'], 1)
#Adding the results to the master dataframe
telecom_df = pd.concat([telecom_df,ml1], axis=1)

# Creating dummy variables for the variable 'OnlineSecurity'.
os = pd.get_dummies(telecom_df['OnlineSecurity'], prefix='OnlineSecurity')
os1 = os.drop(['OnlineSecurity_No internet service'], 1)
# Adding the results to the master dataframe
telecom_df = pd.concat([telecom_df,os1], axis=1)

# Creating dummy variables for the variable 'OnlineBackup'.
ob = pd.get_dummies(telecom_df['OnlineBackup'], prefix='OnlineBackup')
ob1 = ob.drop(['OnlineBackup_No internet service'], 1)
# Adding the results to the master dataframe
telecom_df = pd.concat([telecom_df,ob1], axis=1)

# Creating dummy variables for the variable 'DeviceProtection'.
dp = pd.get_dummies(telecom_df['DeviceProtection'], prefix='DeviceProtection')
dp1 = dp.drop(['DeviceProtection_No internet service'], 1)
# Adding the results to the master dataframe
telecom_df = pd.concat([telecom_df,dp1], axis=1)

# Creating dummy variables for the variable 'TechSupport'.
ts = pd.get_dummies(telecom_df['TechSupport'], prefix='TechSupport')
ts1 = ts.drop(['TechSupport_No internet service'], 1)
```

```
# Adding the results to the master dataframe
telecom_df = pd.concat([telecom_df,ts1], axis=1)

# Creating dummy variables for the variable 'StreamingTV'.
st =pd.get_dummies(telecom_df['StreamingTV'], prefix='StreamingTV')
st1 = st.drop(['StreamingTV_No internet service'], 1)
# Adding the results to the master dataframe
telecom_df = pd.concat([telecom_df,st1], axis=1)

# Creating dummy variables for the variable 'StreamingMovies'.
sm = pd.get_dummies(telecom_df['StreamingMovies'], prefix='StreamingMovies')
sm1 = sm.drop(['StreamingMovies_No internet service'], 1)
# Adding the results to the master dataframe
telecom_df = pd.concat([telecom_df,sm1], axis=1)
```

[51]:
```
telecom_df.head()
```

[51]:

	customerID	tenure	PhoneService	Contract	PaperlessBilling	\
0	7590-VHVEG	1	0	Month-to-month	1	
1	5575-GNVDE	34	1	One year	0	
2	3668-QPYBK	2	1	Month-to-month	1	
3	7795-CFOCW	45	0	One year	0	
4	9237-HQITU	2	1	Month-to-month	1	

	PaymentMethod	MonthlyCharges	TotalCharges	Churn	gender	\
0	Electronic check	29.85	29.85	0	Female	
1	Mailed check	56.95	1889.50	0	Male	
2	Mailed check	53.85	108.15	1	Male	
3	Bank transfer (automatic)	42.30	1840.75	0	Male	
4	Electronic check	70.70	151.65	1	Female	

	…	OnlineBackup_No	OnlineBackup_Yes	DeviceProtection_No	\
0	…	0	1	1	
1	…	1	0	0	
2	…	0	1	1	
3	…	1	0	0	
4	…	1	0	1	

	DeviceProtection_Yes	TechSupport_No	TechSupport_Yes	StreamingTV_No	\
0	0	1	0	1	
1	1	1	0	1	
2	0	1	0	1	
3	1	0	1	1	
4	0	1	0	1	

	StreamingTV_Yes	StreamingMovies_No	StreamingMovies_Yes
0	0	1	0

```
1              0                 1                 0
2              0                 1                 0
3              0                 1                 0
4              0                 1                 0

[5 rows x 43 columns]
```

[52]:
```
# We have created dummies for the below variables, so we can drop them
telecom_df = telecom_df.
 ↪drop(['Contract','PaymentMethod','gender','MultipleLines','InternetService',␣
 ↪'OnlineSecurity', 'OnlineBackup', 'DeviceProtection',
       'TechSupport', 'StreamingTV', 'StreamingMovies'], 1)
```

[53]:
```
telecom_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 7043 entries, 0 to 7042
Data columns (total 32 columns):
 #   Column                                 Non-Null Count  Dtype
---  ------                                 --------------  -----
 0   customerID                             7043 non-null   object
 1   tenure                                 7043 non-null   int64
 2   PhoneService                           7043 non-null   int64
 3   PaperlessBilling                       7043 non-null   int64
 4   MonthlyCharges                         7043 non-null   float64
 5   TotalCharges                           7032 non-null   float64
 6   Churn                                  7043 non-null   int64
 7   SeniorCitizen                          7043 non-null   int64
 8   Partner                                7043 non-null   int64
 9   Dependents                             7043 non-null   int64
 10  Contract_One year                      7043 non-null   uint8
 11  Contract_Two year                      7043 non-null   uint8
 12  PaymentMethod_Credit card (automatic)  7043 non-null   uint8
 13  PaymentMethod_Electronic check         7043 non-null   uint8
 14  PaymentMethod_Mailed check             7043 non-null   uint8
 15  gender_Male                            7043 non-null   uint8
 16  InternetService_Fiber optic            7043 non-null   uint8
 17  InternetService_No                     7043 non-null   uint8
 18  MultipleLines_No                       7043 non-null   uint8
 19  MultipleLines_Yes                      7043 non-null   uint8
 20  OnlineSecurity_No                      7043 non-null   uint8
 21  OnlineSecurity_Yes                     7043 non-null   uint8
 22  OnlineBackup_No                        7043 non-null   uint8
 23  OnlineBackup_Yes                       7043 non-null   uint8
 24  DeviceProtection_No                    7043 non-null   uint8
 25  DeviceProtection_Yes                   7043 non-null   uint8
 26  TechSupport_No                         7043 non-null   uint8
```

```
 27  TechSupport_Yes                          7043 non-null   uint8
 28  StreamingTV_No                           7043 non-null   uint8
 29  StreamingTV_Yes                          7043 non-null   uint8
 30  StreamingMovies_No                       7043 non-null   uint8
 31  StreamingMovies_Yes                      7043 non-null   uint8
dtypes: float64(2), int64(7), object(1), uint8(22)
memory usage: 1014.6+ KB
```

[54]:
```
# Checking for outliers in the continuous variables
num_telecom =␣
 ↪telecom_df[['tenure','MonthlyCharges','SeniorCitizen','TotalCharges']]
```

[55]:
```
# Checking outliers at 25%, 50%, 75%, 90%, 95% and 99%
num_telecom.describe(percentiles=[.25, .5, .75, .90, .95, .99])
```

[55]:

	tenure	MonthlyCharges	SeniorCitizen	TotalCharges
count	7043.000000	7043.000000	7043.000000	7032.000000
mean	32.371149	64.761692	0.162147	2283.300441
std	24.559481	30.090047	0.368612	2266.771362
min	0.000000	18.250000	0.000000	18.800000
25%	9.000000	35.500000	0.000000	401.450000
50%	29.000000	70.350000	0.000000	1397.475000
75%	55.000000	89.850000	0.000000	3794.737500
90%	69.000000	102.600000	1.000000	5976.640000
95%	72.000000	107.400000	1.000000	6923.590000
99%	72.000000	114.729000	1.000000	8039.883000
max	72.000000	118.750000	1.000000	8684.800000

[56]:
```
# Adding up the missing values (column-wise)
telecom_df.isnull().sum()
```

[56]:
```
customerID                               0
tenure                                   0
PhoneService                             0
PaperlessBilling                         0
MonthlyCharges                           0
TotalCharges                            11
Churn                                    0
SeniorCitizen                            0
Partner                                  0
Dependents                               0
Contract_One year                        0
Contract_Two year                        0
PaymentMethod_Credit card (automatic)    0
PaymentMethod_Electronic check           0
PaymentMethod_Mailed check               0
gender_Male                              0
```

```
InternetService_Fiber optic              0
InternetService_No                       0
MultipleLines_No                         0
MultipleLines_Yes                        0
OnlineSecurity_No                        0
OnlineSecurity_Yes                       0
OnlineBackup_No                          0
OnlineBackup_Yes                         0
DeviceProtection_No                      0
DeviceProtection_Yes                     0
TechSupport_No                           0
TechSupport_Yes                          0
StreamingTV_No                           0
StreamingTV_Yes                          0
StreamingMovies_No                       0
StreamingMovies_Yes                      0
dtype: int64
```

[57]:
```
# Checking the percentage of missing values
round(100*(telecom_df.isnull().sum()/len(telecom_df.index)), 2)
```

[57]:
```
customerID                           0.00
tenure                               0.00
PhoneService                         0.00
PaperlessBilling                     0.00
MonthlyCharges                       0.00
TotalCharges                         0.16
Churn                                0.00
SeniorCitizen                        0.00
Partner                              0.00
Dependents                           0.00
Contract_One year                    0.00
Contract_Two year                    0.00
PaymentMethod_Credit card (automatic)  0.00
PaymentMethod_Electronic check       0.00
PaymentMethod_Mailed check           0.00
gender_Male                          0.00
InternetService_Fiber optic          0.00
InternetService_No                   0.00
MultipleLines_No                     0.00
MultipleLines_Yes                    0.00
OnlineSecurity_No                    0.00
OnlineSecurity_Yes                   0.00
OnlineBackup_No                      0.00
OnlineBackup_Yes                     0.00
DeviceProtection_No                  0.00
DeviceProtection_Yes                 0.00
```

```
TechSupport_No                          0.00
TechSupport_Yes                         0.00
StreamingTV_No                          0.00
StreamingTV_Yes                         0.00
StreamingMovies_No                      0.00
StreamingMovies_Yes                     0.00
dtype: float64
```

[58]:
```
# Removing NaN TotalCharges rows
telecom_df = telecom_df[~np.isnan(telecom_df['TotalCharges'])]
```

[59]:
```
# Checking percentage of missing values after removing the missing values
round(100*(telecom_df.isnull().sum()/len(telecom_df.index)), 2)
```

[59]:
```
customerID                          0.0
tenure                              0.0
PhoneService                        0.0
PaperlessBilling                    0.0
MonthlyCharges                      0.0
TotalCharges                        0.0
Churn                               0.0
SeniorCitizen                       0.0
Partner                             0.0
Dependents                          0.0
Contract_One year                   0.0
Contract_Two year                   0.0
PaymentMethod_Credit card (automatic)    0.0
PaymentMethod_Electronic check      0.0
PaymentMethod_Mailed check          0.0
gender_Male                         0.0
InternetService_Fiber optic         0.0
InternetService_No                  0.0
MultipleLines_No                    0.0
MultipleLines_Yes                   0.0
OnlineSecurity_No                   0.0
OnlineSecurity_Yes                  0.0
OnlineBackup_No                     0.0
OnlineBackup_Yes                    0.0
DeviceProtection_No                 0.0
DeviceProtection_Yes                0.0
TechSupport_No                      0.0
TechSupport_Yes                     0.0
StreamingTV_No                      0.0
StreamingTV_Yes                     0.0
StreamingMovies_No                  0.0
StreamingMovies_Yes                 0.0
dtype: float64
```

[60]:
```
from sklearn.model_selection import train_test_split
```

[62]:
```
# Putting feature variable to X
X = telecom_df.drop(['Churn','customerID'], axis=1)

X.head()
```

[62]:

	tenure	PhoneService	PaperlessBilling	MonthlyCharges	TotalCharges	\
0	1	0	1	29.85	29.85	
1	34	1	0	56.95	1889.50	
2	2	1	1	53.85	108.15	
3	45	0	0	42.30	1840.75	
4	2	1	1	70.70	151.65	

	SeniorCitizen	Partner	Dependents	Contract_One year	Contract_Two year	\
0	0	1	0	0	0	
1	0	0	0	1	0	
2	0	0	0	0	0	
3	0	0	0	1	0	
4	0	0	0	0	0	

	…	OnlineBackup_No	OnlineBackup_Yes	DeviceProtection_No	\
0	…	0	1	1	
1	…	1	0	0	
2	…	0	1	1	
3	…	1	0	0	
4	…	1	0	1	

	DeviceProtection_Yes	TechSupport_No	TechSupport_Yes	StreamingTV_No	\
0	0	1	0	1	
1	1	1	0	1	
2	0	1	0	1	
3	1	0	1	1	
4	0	1	0	1	

	StreamingTV_Yes	StreamingMovies_No	StreamingMovies_Yes
0	0	1	0
1	0	1	0
2	0	1	0
3	0	1	0
4	0	1	0

[5 rows x 30 columns]

[63]:
```
# Putting response variable to y
y = telecom_df['Churn']
```

```
y.head()
```

[63]:
```
0    0
1    0
2    1
3    0
4    1
Name: Churn, dtype: int64
```

[64]:
```
# Splitting the data into train and test
X_train, X_test, y_train, y_test = train_test_split(X, y, train_size=0.7,␣
↪test_size=0.3, random_state=100)
```

[65]:
```
    from sklearn.preprocessing import StandardScaler
```

[66]:
```
scaler = StandardScaler()

X_train[['tenure','MonthlyCharges','TotalCharges']] = scaler.
↪fit_transform(X_train[['tenure','MonthlyCharges','TotalCharges']])

X_train.head()
```

[66]:

	tenure	PhoneService	PaperlessBilling	MonthlyCharges	TotalCharges	\
879	0.019693	1	1	-0.338074	-0.276449	
5790	0.305384	0	1	-0.464443	-0.112702	
6498	-1.286319	1	1	0.581425	-0.974430	
880	-0.919003	1	1	1.505913	-0.550676	
2784	-1.163880	1	1	1.106854	-0.835971	

	SeniorCitizen	Partner	Dependents	Contract_One year	\
879	0	0	0	0	
5790	0	1	1	0	
6498	0	0	0	0	
880	0	0	0	0	
2784	0	0	1	0	

	Contract_Two year	…	OnlineBackup_No	OnlineBackup_Yes	\
879	0	…	0	1	
5790	0	…	0	1	
6498	0	…	0	1	
880	0	…	0	1	
2784	0	…	1	0	

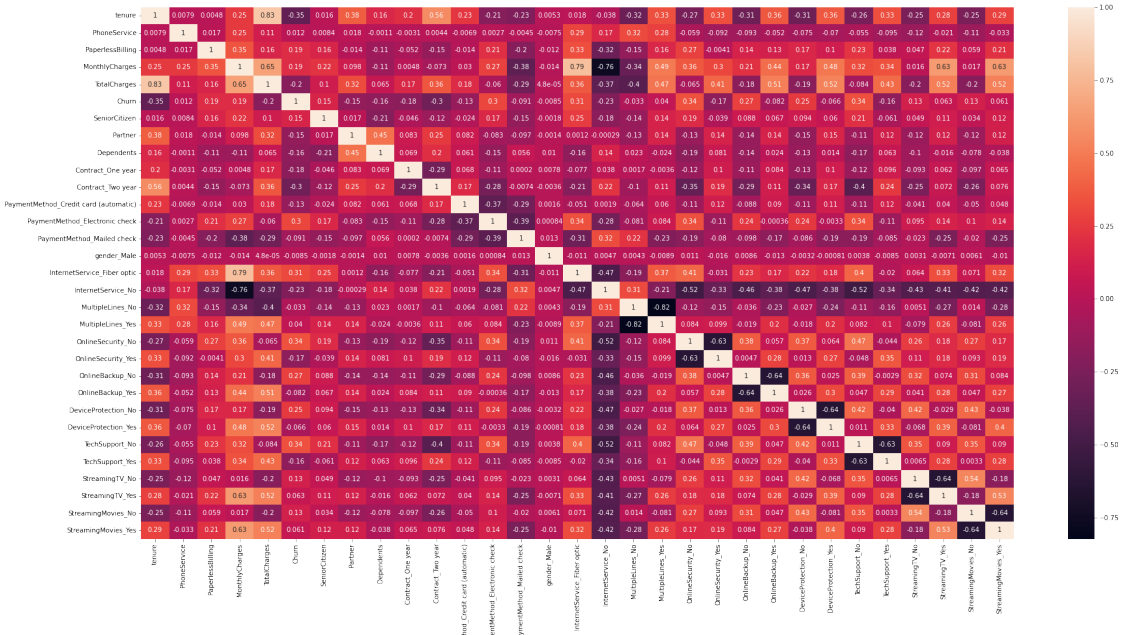
	DeviceProtection_No	DeviceProtection_Yes	TechSupport_No	\
879	1	0	1	
5790	1	0	1	
6498	0	1	1	

```
880                     0                     1               0
2784                    0                     1               0

      TechSupport_Yes  StreamingTV_No  StreamingTV_Yes  StreamingMovies_No  \
879                 0               1                0                   1
5790                0               0                1                   0
6498                0               1                0                   1
880                 1               0                1                   0
2784                1               0                1                   0

      StreamingMovies_Yes
879                     0
5790                    1
6498                    0
880                     1
2784                    1

[5 rows x 30 columns]
```

[67]:
```
### Checking the Churn Rate
churn = (sum(telecom_df['Churn'])/len(telecom_df['Churn'].index))*100
churn
```

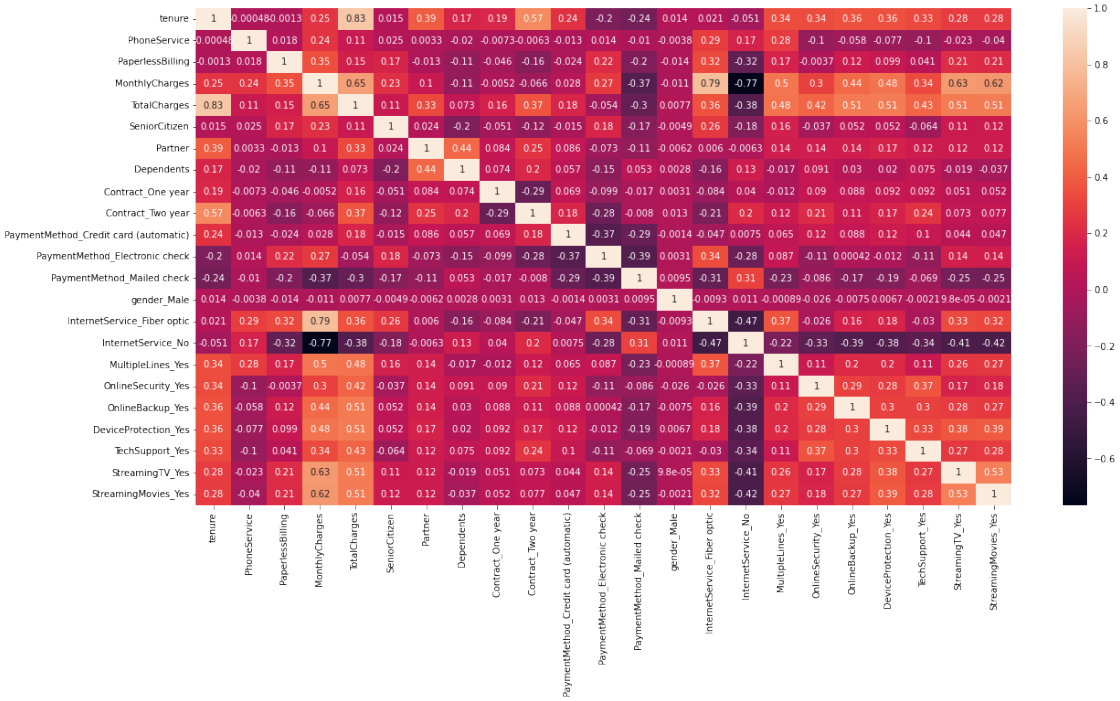
[67]:
```
26.578498293515356
```

[68]:
```
# Let's see the correlation matrix
plt.figure(figsize = (30,15))        # Size of the figure
sns.heatmap(telecom_df.corr(),annot = True)
plt.show()
```



[69]:
```
X_test = X_test.
 ↪drop(['MultipleLines_No','OnlineSecurity_No','OnlineBackup_No','DeviceProtection_No','TechS
                        'StreamingTV_No','StreamingMovies_No'], 1)
X_train = X_train.
 ↪drop(['MultipleLines_No','OnlineSecurity_No','OnlineBackup_No','DeviceProtection_No','TechS
                          'StreamingTV_No','StreamingMovies_No'], 1)
```

[70]:
```
plt.figure(figsize = (20,10))
sns.heatmap(X_train.corr(),annot = True)
plt.show()
```



[71]:
```
jovian.commit
```

[71]:
```
<function jovian.utils.commit.commit(message=None, files=[], outputs=[],
environment=None, privacy='auto', filename=None, project=None, new_project=None,
git_commit=False, git_message='auto', require_write_access=False, **kwargs)>
```

[ ]:
```

```