Telecommunication Project - Resume

February 8, 2024

Analyzed the dataset of an European Telecom Company.

This Telecom's Churn Dataset, consists of cleaned customer activity data (features), along with a churn label specifying whether a customer cancelled the subscription or not.

Analyzed the data to discover key factors responsible for customer churn and come up with ways/recommendations to ensure customer retention.

1 Business Understanding Of A Telecom Industry Customer Churn:

Customer churn is a big problem in any industry and one of the most important concerns for the Telecom industry.

The effect on the revenues of the companies, because of this customer churns is huge, especially in the telecom field, that's why these companies are seeking to develop a predictive potential customer churn.

In this highly competitive market, the telecommunications industry experiences an average of 15-25% annual churn rate, and it costs 5-10 times more to acquire a new customer than to retain an existing one, that's why customer retention has now become even more important than customer acquisition.

Therefore, finding those factors that increase customer churn is important to take necessary actions to reduce this churn.

The main goal of this project is to develop an understanding of the cause of customer churn which assists telecom operators to predict customers who are most likely subject to churn, and what to do to retain the most valuable customer.

1.0.1 I will find how I can maximize the profit by retaining customer, and, how I can reduce the churn rate by identifying the issues.

[2]:
```
# Importing the required libraries

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

[3]:
```
# importing the dataset

data = pd.read_csv("Telecom Dataset.csv")
```

---

2 Exploring the Dataset

[3]:
```
data                    # To have a look at our dataset ; it will show Top5 &
↪Bottom5 rows, and about 20 columns
```

[3]:
```
     state  account length  area code phone number international plan  \
0       KS             128        415     382-4657                 no
1       OH             107        415     371-7191                 no
2       NJ             137        415     358-1921                 no
3       OH              84        408     375-9999                yes
4       OK              75        415     330-6626                yes
…     …               …         …           …                    …
3328    AZ             192        415     414-4276                 no
3329    WV              68        415     370-3271                 no
3330    RI              28        510     328-8230                 no
3331    CT             184        510     364-6381                yes
3332    TN              74        415     400-4344                 no

     voice mail plan  number vmail messages  total day minutes  \
0                yes                     25              265.1
1                yes                     26              161.6
2                 no                      0              243.4
3                 no                      0              299.4
4                 no                      0              166.7
…               …                       …                …
3328             yes                     36              156.2
3329              no                      0              231.1
3330              no                      0              180.8
3331              no                      0              213.8
3332             yes                     25              234.4

      total day calls  total day charge  …  total eve calls  \
0                 110             45.07  …               99
1                 123             27.47  …              103
2                 114             41.38  …              110
```

```
3                 71             50.90  …               88
4                113             28.34  …              122
…                …               …  …                …
3328              77             26.55  …              126
3329              57             39.29  …               55
3330             109             30.74  …               58
3331             105             36.35  …               84
3332             113             39.85  …               82

      total eve charge  total night minutes  total night calls  \
0                16.78                244.7                 91
1                16.62                254.4                103
2                10.30                162.6                104
3                 5.26                196.9                 89
4                12.61                186.9                121
…                …                   …                   …
3328             18.32                279.1                 83
3329             13.04                191.3                123
3330             24.55                191.9                 91
3331             13.57                139.2                137
3332             22.60                241.4                 77

      total night charge  total intl minutes  total intl calls  \
0                  11.01                10.0                 3
1                  11.45                13.7                 3
2                   7.32                12.2                 5
3                   8.86                 6.6                 7
4                   8.41                10.1                 3
…                  …                   …                   …
3328               12.56                 9.9                 6
3329                8.61                 9.6                 4
3330                8.64                14.1                 6
3331                6.26                 5.0                10
3332               10.86                13.7                 4

      total intl charge  customer service calls  churn
0                  2.70                       1  False
1                  3.70                       1  False
2                  3.29                       0  False
3                  1.78                       2  False
4                  2.73                       3  False
…                 …                          …    …
3328               2.67                       2  False
3329               2.59                       3  False
3330               3.81                       2  False
3331               1.35                       2  False
3332               3.70                       0  False
```

```
[3333 rows x 21 columns]
```

[4]:
```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3333 entries, 0 to 3332
Data columns (total 21 columns):
 #   Column                  Non-Null Count  Dtype
---  ------                  --------------  -----
 0   state                   3333 non-null   object
 1   account length          3333 non-null   int64
 2   area code               3333 non-null   int64
 3   phone number            3333 non-null   object
 4   international plan      3333 non-null   object
 5   voice mail plan         3333 non-null   object
 6   number vmail messages   3333 non-null   int64
 7   total day minutes       3333 non-null   float64
 8   total day calls         3333 non-null   int64
 9   total day charge        3333 non-null   float64
 10  total eve minutes       3333 non-null   float64
 11  total eve calls         3333 non-null   int64
 12  total eve charge        3333 non-null   float64
 13  total night minutes     3333 non-null   float64
 14  total night calls       3333 non-null   int64
 15  total night charge      3333 non-null   float64
 16  total intl minutes      3333 non-null   float64
 17  total intl calls        3333 non-null   int64
 18  total intl charge       3333 non-null   float64
 19  customer service calls  3333 non-null   int64
 20  churn                   3333 non-null   bool
dtypes: bool(1), float64(8), int64(8), object(4)
memory usage: 524.2+ KB
```

[7]:
```
data.shape              # To show the number of Rows & Columns in the DataFrame
```

[7]:
```
(3333, 21)
```

[8]:
```
data.nunique()          # To show the number of unique values in each column
```

[8]:
```
state                     51
account length           212
area code                  3
phone number            3333
international plan         2
voice mail plan            2
number vmail messages     46
total day minutes       1667
```

```
total day calls            119
total day charge          1667
total eve minutes         1611
total eve calls            123
total eve charge          1440
total night minutes       1591
total night calls          120
total night charge         933
total intl minutes         162
total intl calls            21
total intl charge          162
customer service calls      10
churn                        2
dtype: int64
```

[9]:
```
data.columns              # To show all the column names of the DataFrame
```

[9]:
```
Index(['state', 'account length', 'area code', 'phone number',
       'international plan', 'voice mail plan', 'number vmail messages',
       'total day minutes', 'total day calls', 'total day charge',
       'total eve minutes', 'total eve calls', 'total eve charge',
       'total night minutes', 'total night calls', 'total night charge',
       'total intl minutes', 'total intl calls', 'total intl charge',
       'customer service calls', 'churn'],
      dtype='object')
```

[10]:
```
data.dtypes               # To show the data-type of each column
```

[10]:
```
state                    object
account length            int64
area code                 int64
phone number             object
international plan       object
voice mail plan          object
number vmail messages     int64
total day minutes       float64
total day calls           int64
total day charge        float64
total eve minutes       float64
total eve calls           int64
total eve charge        float64
total night minutes     float64
total night calls         int64
total night charge      float64
total intl minutes      float64
total intl calls          int64
total intl charge       float64
```

```
customer service calls      int64
churn                        bool
dtype: object
```

[5]:
```
data.describe()
```

[5]:

	account length	area code	number vmail messages	total day minutes	\
count	3333.000000	3333.000000	3333.000000	3333.000000	
mean	101.064806	437.182418	8.099010	179.775098	
std	39.822106	42.371290	13.688365	54.467389	
min	1.000000	408.000000	0.000000	0.000000	
25%	74.000000	408.000000	0.000000	143.700000	
50%	101.000000	415.000000	0.000000	179.400000	
75%	127.000000	510.000000	20.000000	216.400000	
max	243.000000	510.000000	51.000000	350.800000	

	total day calls	total day charge	total eve minutes	total eve calls	\
count	3333.000000	3333.000000	3333.000000	3333.000000	
mean	100.435644	30.562307	200.980348	100.114311	
std	20.069084	9.259435	50.713844	19.922625	
min	0.000000	0.000000	0.000000	0.000000	
25%	87.000000	24.430000	166.600000	87.000000	
50%	101.000000	30.500000	201.400000	100.000000	
75%	114.000000	36.790000	235.300000	114.000000	
max	165.000000	59.640000	363.700000	170.000000	

	total eve charge	total night minutes	total night calls	\
count	3333.000000	3333.000000	3333.000000	
mean	17.083540	200.872037	100.107711	
std	4.310668	50.573847	19.568609	
min	0.000000	23.200000	33.000000	
25%	14.160000	167.000000	87.000000	
50%	17.120000	201.200000	100.000000	
75%	20.000000	235.300000	113.000000	
max	30.910000	395.000000	175.000000	

	total night charge	total intl minutes	total intl calls	\
count	3333.000000	3333.000000	3333.000000	
mean	9.039325	10.237294	4.479448	
std	2.275873	2.791840	2.461214	
min	1.040000	0.000000	0.000000	
25%	7.520000	8.500000	3.000000	
50%	9.050000	10.300000	4.000000	
75%	10.590000	12.100000	6.000000	
max	17.770000	20.000000	20.000000	

	total intl charge	customer service calls

```
count        3333.000000              3333.000000
mean            2.764581                 1.562856
std             0.753773                 1.315491
min             0.000000                 0.000000
25%             2.300000                 1.000000
50%             2.780000                 1.000000
75%             3.270000                 2.000000
max             5.400000                 9.000000
```

[12]:
```
data.describe(include='object')       # It gives the summary of all categorical columns
# It shows the count of non-null and unique values in each column, Top value with its occurance in each column
```

[12]:
```
        state phone number international plan voice mail plan
count    3333         3333                3333            3333
unique     51         3333                   2               2
top        WV     341-9443                  no              no
freq      106            1                3010            2411
```

---

3 Checking Missing and Duplicate Values

[13]:
```
data.head(2)            # To show the Top2 records of the DataFrame
```

[13]:
```
  state  account length  area code phone number international plan  \
0    KS             128        415     382-4657                 no
1    OH             107        415     371-7191                 no

  voice mail plan  number vmail messages  total day minutes  total day calls  \
0             yes                     25              265.1              110
1             yes                     26              161.6              123

   total day charge  …  total eve calls  total eve charge  \
0             45.07  …               99             16.78
1             27.47  …              103             16.62

   total night minutes  total night calls  total night charge  \
0                244.7                 91               11.01
1                254.4                103               11.45

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3                2.7
1                13.7                 3                3.7
```

```
   customer service calls  churn
0                       1  False
1                       1  False

[2 rows x 21 columns]
```

isna( )

[14]:
```
data.isna().sum()           # To show the count of missing (null) values in␣
 ↪each column
```

[14]:
```
state                     0
account length            0
area code                 0
phone number              0
international plan        0
voice mail plan           0
number vmail messages     0
total day minutes         0
total day calls           0
total day charge          0
total eve minutes         0
total eve calls           0
total eve charge          0
total night minutes       0
total night calls         0
total night charge        0
total intl minutes        0
total intl calls          0
total intl charge         0
customer service calls    0
churn                     0
dtype: int64
```

isnull( )

[15]:
```
data.isnull().sum()          # Alternatively, To show the count of missing␣
 ↪(null) values in each column
```

[15]:
```
state                     0
account length            0
area code                 0
phone number              0
international plan        0
voice mail plan           0
number vmail messages     0
total day minutes         0
```

```
total day calls           0
total day charge          0
total eve minutes         0
total eve calls           0
total eve charge          0
total night minutes       0
total night calls         0
total night charge        0
total intl minutes        0
total intl calls          0
total intl charge         0
customer service calls    0
churn                     0
dtype: int64
```

notnull( )

[16]:
```
data.notnull().sum()          # To show the count of non-null values in each column
```

[16]:
```
state                     3333
account length            3333
area code                 3333
phone number              3333
international plan        3333
voice mail plan           3333
number vmail messages     3333
total day minutes         3333
total day calls           3333
total day charge          3333
total eve minutes         3333
total eve calls           3333
total eve charge          3333
total night minutes       3333
total night calls         3333
total night charge        3333
total intl minutes        3333
total intl calls          3333
total intl charge         3333
customer service calls    3333
churn                     3333
dtype: int64
```

notna( )

[18]:
```
data[data.duplicated()]          # To show the duplicate records present in the DataFrame
```

```
                                    # In this dataframe, no duplicate records␣
↪present
```

[18]:
```
Empty DataFrame
Columns: [state, account length, area code, phone number, international plan,
voice mail plan, number vmail messages, total day minutes, total day calls,
total day charge, total eve minutes, total eve calls, total eve charge, total
night minutes, total night calls, total night charge, total intl minutes, total
intl calls, total intl charge, customer service calls, churn]
Index: []

[0 rows x 21 columns]
```

[19]:
```
data = data.drop_duplicates()     # To drop all the duplicate records from the␣
↪DataFrame, and saving the updated DF
```

[20]:
```
data                              # To have a look at the DataFrame
```

[20]:
```
     state  account length  area code phone number international plan  \
0       KS             128        415     382-4657                 no
1       OH             107        415     371-7191                 no
2       NJ             137        415     358-1921                 no
3       OH              84        408     375-9999                yes
4       OK              75        415     330-6626                yes
…    …              …         …           …                  …
3328    AZ             192        415     414-4276                 no
3329    WV              68        415     370-3271                 no
3330    RI              28        510     328-8230                 no
3331    CT             184        510     364-6381                yes
3332    TN              74        415     400-4344                 no

     voice mail plan  number vmail messages  total day minutes  \
0                yes                     25              265.1
1                yes                     26              161.6
2                 no                      0              243.4
3                 no                      0              299.4
4                 no                      0              166.7
…              …                       …                …
3328             yes                     36              156.2
3329              no                      0              231.1
3330              no                      0              180.8
3331              no                      0              213.8
3332             yes                     25              234.4

      total day calls  total day charge  …  total eve calls  \
0                 110             45.07  …               99
1                 123             27.47  …              103
```

```
2                 114             41.38  …              110
3                  71             50.90  …               88
4                 113             28.34  …              122
…                 …                 …  …                …
3328               77             26.55  …              126
3329               57             39.29  …               55
3330              109             30.74  …               58
3331              105             36.35  …               84
3332              113             39.85  …               82

      total eve charge  total night minutes  total night calls  \
0                16.78                244.7                 91
1                16.62                254.4                103
2                10.30                162.6                104
3                 5.26                196.9                 89
4                12.61                186.9                121
…                 …                   …                   …
3328             18.32                279.1                 83
3329             13.04                191.3                123
3330             24.55                191.9                 91
3331             13.57                139.2                137
3332             22.60                241.4                 77

      total night charge  total intl minutes  total intl calls  \
0                  11.01                10.0                 3
1                  11.45                13.7                 3
2                   7.32                12.2                 5
3                   8.86                 6.6                 7
4                   8.41                10.1                 3
…                   …                   …                   …
3328               12.56                 9.9                 6
3329                8.61                 9.6                 4
3330                8.64                14.1                 6
3331                6.26                 5.0                10
3332               10.86                13.7                 4

      total intl charge  customer service calls  churn
0                  2.70                       1  False
1                  3.70                       1  False
2                  3.29                       0  False
3                  1.78                       2  False
4                  2.73                       3  False
…                  …                          …  …
3328               2.67                       2  False
3329               2.59                       3  False
3330               3.81                       2  False
3331               1.35                       2  False
```

```
3332              3.70                         0  False

[3333 rows x 21 columns]
```

---

4 1. Analyzing the 'Churn' Variable

[21]:
```
data.head()                    # To show Top5 records of the DataFrame
```

[21]:
```
  state  account length  area code phone number international plan  \
0    KS             128        415     382-4657                 no
1    OH             107        415     371-7191                 no
2    NJ             137        415     358-1921                 no
3    OH              84        408     375-9999                yes
4    OK              75        415     330-6626                yes

  voice mail plan  number vmail messages  total day minutes  total day calls  \
0             yes                     25              265.1              110
1             yes                     26              161.6              123
2              no                      0              243.4              114
3              no                      0              299.4               71
4              no                      0              166.7              113

   total day charge  …  total eve calls  total eve charge  \
0             45.07  …               99             16.78
1             27.47  …              103             16.62
2             41.38  …              110             10.30
3             50.90  …               88              5.26
4             28.34  …              122             12.61

   total night minutes  total night calls  total night charge  \
0                244.7                 91               11.01
1                254.4                103               11.45
2                162.6                104                7.32
3                196.9                 89                8.86
4                186.9                121                8.41

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3               2.70
1                13.7                 3               3.70
2                12.2                 5               3.29
3                 6.6                 7               1.78
4                10.1                 3               2.73
```

```
   customer service calls  churn
0                       1  False
1                       1  False
2                       0  False
3                       2  False
4                       3  False

[5 rows x 21 columns]
```

[6]:
```
data['churn'].unique()                # To show the unique values present in the column 'churn'
```

[6]:
```
array([False,  True])
```

[23]:
```
A = data['churn'].value_counts()      # To show the count of the unique values of the column 'churn' ...
print(A)                              # ... and saving the result in variable 'A' and printing variable 'A'
```

```
False    2850
True      483
Name: churn, dtype: int64
```

[24]:
```
str(483/3333*100) + " %"              # Just to check the percentage of True (churned customers) from the dataset ...
                                      # ... and showing the result in string format, after adding % sign at the end
```

[24]:
```
'14.491449144914492 %'
```

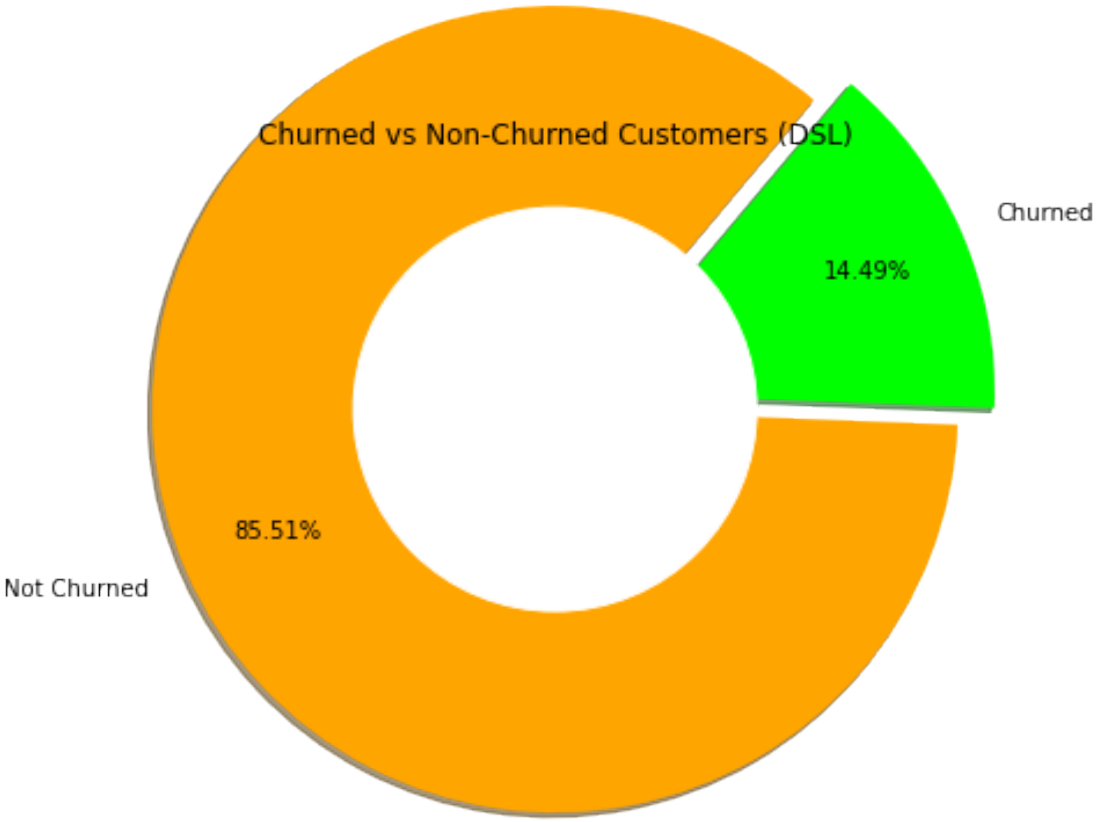
[25]:
```
type(A)                               # To check the type of variable 'A' ; it's a Series
```

[25]:
```
pandas.core.series.Series
```

4.0.1 Donut Chart

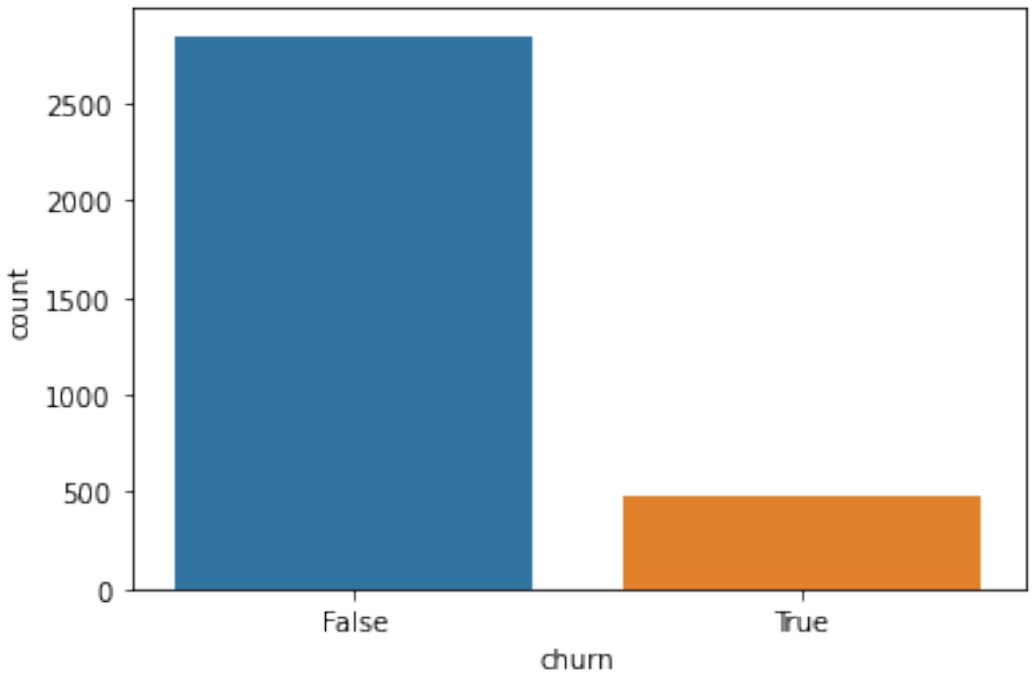
[26]:
```
# Creating a Donut Chart for Churn vs Non-Churn Customers

plt.pie(A, labels=['Not Churned', 'Churned'], colors=['orange','lime'],startangle=50, shadow=True, radius=2,
        explode=(0,0.2), autopct='%1.2f%%', pctdistance=0.75);

circle = plt.Circle((0,0), 1, color='white')
c = plt.gcf()

c.gca().add_artist(circle)
```

```
plt.title("Churned vs Non-Churned Customers (DSL)")
plt.show()

# Donut chart is a modified Pie chart, with an area from center cut out
#1 First, drawing a pie plot, using Matplotlib library
# The data is taken from the variable 'A' .... labels are given as 'Churned' &␣
 ↪'Not Churned' ...
# ... color is given to each label 'orange' & 'lime' ... startangle means the␣
 ↪angle for slicing, set as 50 ...
# ... shadow is True means it will drop some shadow of the chart ... radius of␣
 ↪the circle is set as 2 ...
# ... explode is used to cut the slice out of the figure ...
# ... autopct is used to show the % on the chart upto required decimal points ..
 ↪.
# ... pctdistance is given for distance of % from the center
#2 Second, using plt.Circle...we will create a circle and save it in the␣
 ↪variable name 'circle' ...
# ... putting (0,0) by default ... 1 is radius of circle ... and color of␣
 ↪circle is white
# plt.gcf() is used to get the current figure ... and we are saving it in␣
 ↪variable 'c'
#3 Third, we will add the 'circle' at the center of pie chart ... using gca().
 ↪add_artist()
# We have given the title to the chart using plt.title()
# plt.show() - To show the chart
```



4.0.2 Countplot

[27]:
```
# Countplot for Churn vs Non-Churn Customers

sns.countplot(data['churn']);

# Drawing a countplot for the column 'churn' from the datset using seaborn␣
↪library
```



Outcome - After analyzing the Churn column, we notice that almost 15% customers have churned. Now, we will analyze other columns to find out the possible reasons for this churn.

---

5 2. Analyzing the 'State' Variable

[28]:
```
data.head(2)                    # To show Top 2 records of the dataset
```

[28]:
```
  state  account length  area code phone number international plan  \
0    KS             128        415     382-4657                 no
1    OH             107        415     371-7191                 no

  voice mail plan  number vmail messages  total day minutes  total day calls  \
0             yes                     25              265.1              110
1             yes                     26              161.6              123

   total day charge  …  total eve calls  total eve charge  \
0             45.07  …               99             16.78
1             27.47  …              103             16.62
```

```
   total night minutes  total night calls  total night charge  \
0                244.7                 91               11.01
1                254.4                103               11.45

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3                2.7
1                13.7                 3                3.7

   customer service calls  churn
0                       1  False
1                       1  False

[2 rows x 21 columns]
```

[29]:
```
data.state.nunique()          # To show the total number of unique values present in column 'state'
```

[29]:
```
51
```

unique( )

[30]:
```
data.state.unique()           # To show all the unique values of the column 'state'
                              # it shows the output in the form of 1-D array
```

[30]:
```
array(['KS', 'OH', 'NJ', 'OK', 'AL', 'MA', 'MO', 'LA', 'WV', 'IN', 'RI',
       'IA', 'MT', 'NY', 'ID', 'VT', 'VA', 'TX', 'FL', 'CO', 'AZ', 'SC',
       'NE', 'WY', 'HI', 'IL', 'NH', 'GA', 'AK', 'MD', 'AR', 'WI', 'OR',
       'MI', 'DE', 'UT', 'CA', 'MN', 'SD', 'NC', 'WA', 'NM', 'NV', 'DC',
       'KY', 'ME', 'MS', 'TN', 'PA', 'CT', 'ND'], dtype=object)
```

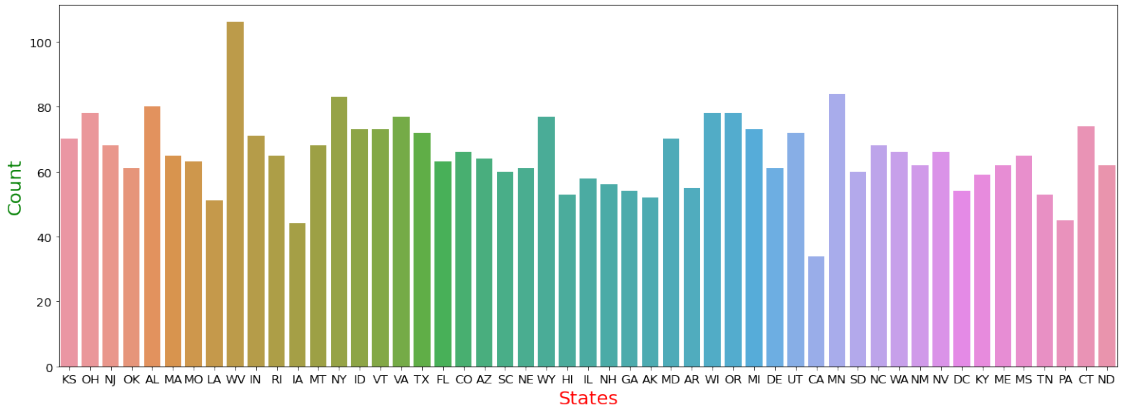
[7]:
```
data.state.value_counts()     # To show the occurance/count of all unique values of the column 'state'

# By default , it shows result in descending order
```

[7]:
```
state
WV    106
MN     84
NY     83
AL     80
WI     78
OH     78
OR     78
WY     77
VA     77
CT     74
MI     73
```

```
ID     73
VT     73
TX     72
UT     72
IN     71
MD     70
KS     70
NC     68
NJ     68
MT     68
CO     66
NV     66
WA     66
RI     65
MA     65
MS     65
AZ     64
FL     63
MO     63
NM     62
ME     62
ND     62
NE     61
OK     61
DE     61
SC     60
SD     60
KY     59
IL     58
NH     56
AR     55
GA     54
DC     54
HI     53
TN     53
AK     52
LA     51
PA     45
IA     44
CA     34
Name: count, dtype: int64
```

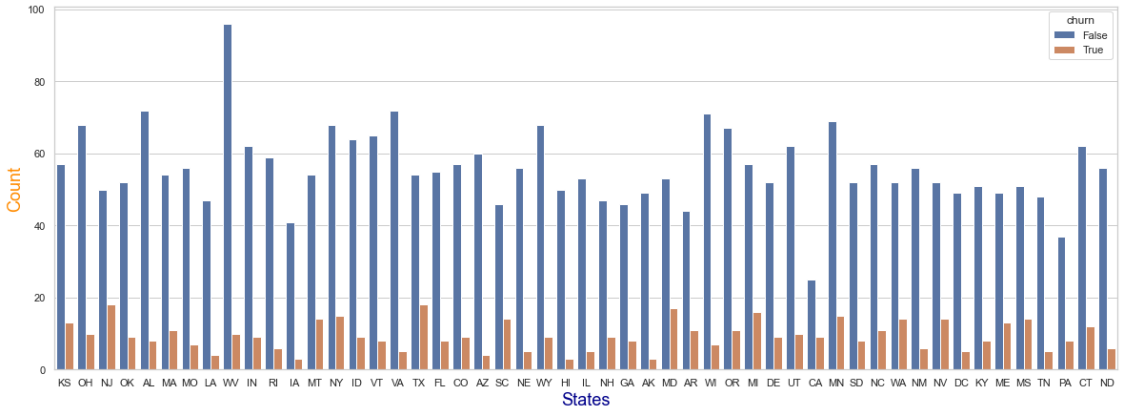
5.0.1 Countplot

[32]:
```
plt.figure(figsize=(20,7))        # Setting the size of the figure as 20 x 7, using matplotlib library

sns.countplot(data['state'])     # Drawing a countplot for the column 'state' from the datset using seaborn library
plt.xticks(fontsize=13)          # Setting the fontsize of states on x-axis
plt.yticks(fontsize=13)          # Setting the fontsize of numbers on y-axis
plt.xlabel('States', fontsize=20, color='Red')        # Setting the label on x-axis as States , its fontsize , its color
plt.ylabel('Count', fontsize=20, color='Green')       # Setting the label on y-axis as Count , its fontsize , its color
plt.show()                                            # To display the figure

# The x-axis showing the names of the states ... and y-axis showing their counts (occurance) ...
# ... means how many times a state is present in the column
```



[33]:
```
# Showing Churn/Not Churn State-wise

sns.set(style="whitegrid")           # Using seaborn library, setting the background of the figure as "whitegrid"

plt.figure(figsize=(20,7))           # Setting the size of the figure as 20 x 7, using matplotlib library

sns.countplot(data.state, hue = data.churn)

plt.xlabel('States', fontsize=18, color='DarkBlue')   # Setting the Label on x-axis as States, its fontsize & color
```

```
plt.ylabel('Count', fontsize=18, color='DarkOrange'); # Setting the Label on␣
 ↪y-axis as Count , its fontsize & color

# Drawing a countplot for the column 'state' from the datset using seaborn␣
 ↪library ...
# ... selecting the Churn column for hue
# 'hue = data.churn' represents that we want to use column "churn" for color␣
 ↪encoding
# i.e. color the bars for Churn and Not-Churn differently
# Blue color representing False (Not Churn) and Orange color representing True␣
 ↪(Churn) for each state separately
```



[34]:
```
data.head(2)                # To show Top2 records of the dataset
```

[34]:

	state	account length	area code	phone number	international plan	\
0	KS	128	415	382-4657	no	
1	OH	107	415	371-7191	no	

	voice mail plan	number vmail messages	total day minutes	total day calls	\
0	yes	25	265.1	110	
1	yes	26	161.6	123	

	total day charge	…	total eve calls	total eve charge	\
0	45.07	…	99	16.78	
1	27.47	…	103	16.62	

	total night minutes	total night calls	total night charge	\
0	244.7	91	11.01	
1	254.4	103	11.45	

	total intl minutes	total intl calls	total intl charge	\
0	10.0	3	2.7	

```
1                13.7                3                3.7

   customer service calls  churn
0                       1  False
1                       1  False

[2 rows x 21 columns]
```

[35]:
```
a = data.groupby("state")['churn'].mean()*100

a.sort_values()

# Here, we will consider two columns while using groupby()
# Used groupby on ("state") column ... and showing the mean of values of␣
 ↪['churn'] column wrt to each state ...
# ... and multiplying the result by 100
# By default, it will consider 'True' values from ['churn'] column to calculate␣
 ↪mean ...
# ... bcz the data-type of column ['churn'] is boolean type
```

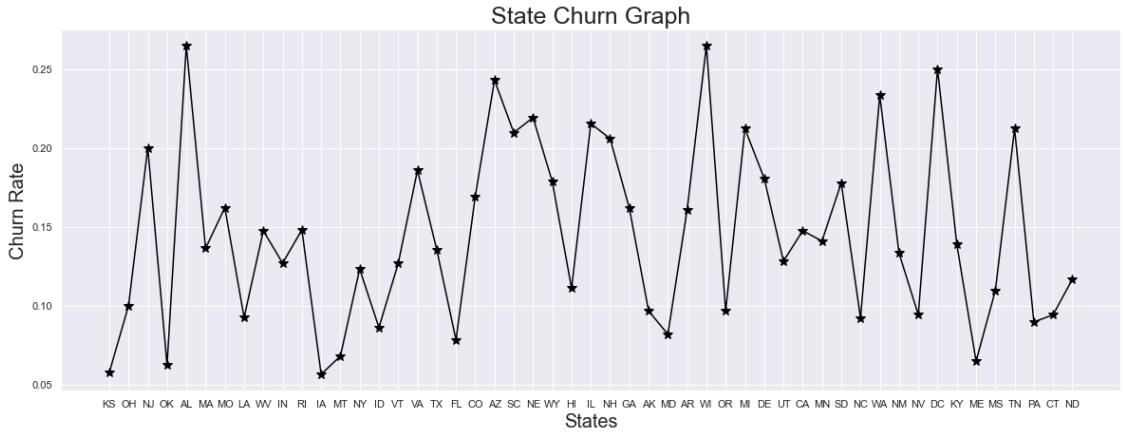
[35]:
```
state
HI     5.660377
AK     5.769231
AZ     6.250000
VA     6.493506
IA     6.818182
LA     7.843137
NE     8.196721
IL     8.620690
WI     8.974359
RI     9.230769
DC     9.259259
TN     9.433962
WV     9.433962
NM     9.677419
ND     9.677419
AL    10.000000
VT    10.958904
MO    11.111111
WY    11.688312
ID    12.328767
IN    12.676056
FL    12.698413
OH    12.820513
SD    13.333333
KY    13.559322
CO    13.636364
```

```
UT    13.888889
OR    14.102564
OK    14.754098
DE    14.754098
GA    14.814815
NH    16.071429
NC    16.176471
CT    16.216216
MA    16.923077
PA    17.777778
MN    17.857143
NY    18.072289
KS    18.571429
AR    20.000000
MT    20.588235
ME    20.967742
WA    21.212121
NV    21.212121
MS    21.538462
MI    21.917808
SC    23.333333
MD    24.285714
TX    25.000000
NJ    26.470588
CA    26.470588
Name: churn, dtype: float64
```

5.0.2 Line Chart

[36]:
```
# Drawing a Line Chart showing State-wise Churn rate

X = data.state.unique()      # Considering all unique values of 'state' column,␣
 ↪and saving the output in variable X
Y = data.groupby("state")['churn'].mean()    # Using groupby on 'state' column,␣
 ↪and calculating the mean of churn ...
                                             # ... values against each state,␣
 ↪and saving the output in variable Y
sns.set(style="darkgrid")               # Using seaborn library, setting the␣
 ↪background of the figure as "whitegrid"

plt.rcParams['figure.figsize'] = (20,7)    # Using rcParams from matplotlib␣
 ↪library ...
                                           # ... setting the size of the figure␣
 ↪as 20 x 7
plt.rcParams['lines.linestyle'] = '-'      # Using rcParams from matplotlib␣
 ↪library ...
```

```
                                                # ... setting the style of the line␣
 ↪as ' - '

plt.plot(X, Y, color='black', marker = '*', markersize='10')     # Plotting a␣
 ↪Line chart with X and Y ...
# ... States showing on x-axis & Churn Rate showing on y-axis ... color of line␣
 ↪set as 'black' ...
# ... marker set as '*' ... markersize set as '10'

plt.title("State Churn Graph", fontsize=25)  # Using matplotlib, setting the␣
 ↪Title of figure, with fontsize of 25
plt.xlabel("States", fontsize=20)            # Using matplotlib, setting the␣
 ↪Label of x-axis, with fontsize of 20
plt.ylabel("Churn Rate", fontsize=20);       # Using matplotlib, setting the␣
 ↪Label of y-axis, with fontsize of 20

# rcParams - run command parameters
# matplotlib.rcParams contains some properties in matplotlibrc file. We can use␣
 ↪it to control the defaults ...
# ... of almost every property in Matplotlib: figure size and DPI, line width,␣
 ↪color and style, axes, axis ...
# ... and grid properties, text and font properties and so on. In order to use␣
 ↪matplotlib.
```



[37]:
```
#x = data.state
#x = data.groupby("state")['churn'].mean().sort_values(ascending=False).head(5);
#y = data.groupby("state")['churn'].mean().sort_values(ascending=False);

#plt.bar(x,y)

data.groupby("state")['churn'].mean().sort_values(ascending=False).head(5)*100
```

[37]:
```
state
NJ    26.470588
CA    26.470588
TX    25.000000
MD    24.285714
SC    23.333333
Name: churn, dtype: float64
```

[38]:
```
data.head()              # To show Top5 records of the dataset
```

[38]:

	state	account length	area code	phone number	international plan	voice mail plan	number vmail messages	total day minutes	total day calls	total day charge	…	total eve calls	total eve charge	total night minutes	total night calls	total night charge	total intl minutes	total intl calls	total intl charge	customer service calls	churn
0	KS	128	415	382-4657	no	yes	25	265.1	110	45.07	…	99	16.78	244.7	91	11.01	10.0	3	2.70	1	False
1	OH	107	415	371-7191	no	yes	26	161.6	123	27.47	…	103	16.62	254.4	103	11.45	13.7	3	3.70	1	False
2	NJ	137	415	358-1921	no	no	0	243.4	114	41.38	…	110	10.30	162.6	104	7.32	12.2	5	3.29	0	False
3	OH	84	408	375-9999	yes	no	0	299.4	71	50.90	…	88	5.26	196.9	89	8.86	6.6	7	1.78	2	False
4	OK	75	415	330-6626	yes	no	0	166.7	113	28.34	…	122	12.61	186.9	121	8.41	10.1	3	2.73	3	False

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

[5 rows x 21 columns]
```

[39]:
```
# Calculatiog State vs Churn Percentage

c = pd.crosstab(data.state, data.churn ,margins=True)
                                # Using crosstab function from pandas library␣
 ↪to compute a simple cross-tabulation...
                                # ...of two columns i.e., 'state' & 'churn'␣
 ↪and saving the output in varibale 'c'
                                # ... margins=True means - to show the sum of␣
 ↪both values in a new column 'All'
c['Percentage_Churn'] = c[True]/(c['All']) * 100        # Creating a new column␣
 ↪'Percantage_Churn', which shows the...
                                                        # ...percentage of churn␣
 ↪customers in each state
print(c.head())                                         # Printing top 5 records␣
 ↪of the result
print(type(c))                                          # Printing type of 'c' ;␣
 ↪it's is a DataFrame
```

```
churn  False  True  All  Percentage_Churn
state
AK        49     3   52          5.769231
AL        72     8   80         10.000000
AR        44    11   55         20.000000
AZ        60     4   64          6.250000
CA        25     9   34         26.470588
<class 'pandas.core.frame.DataFrame'>
```

[40]:
```
c.info()            # To show the basic information of the DataFrame 'c'
#It shows indexes, columns counts, each column name with its non-null values␣
 ↪count & data-type and, memory of DataFrame
# Now, in this dataframe 'c', we have 4 columns␣
 ↪'False,True,All,Percentage_Churn' each having 52 non-null values
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 52 entries, AK to All
Data columns (total 4 columns):
 #   Column            Non-Null Count  Dtype
---  ------            --------------  -----
 0   False             52 non-null     int64
 1   True              52 non-null     int64
 2   All               52 non-null     int64
```

```
 3   Percentage_Churn  52 non-null     float64
dtypes: float64(1), int64(3)
memory usage: 2.0+ KB
```

[41]:
```
c.sort_values(by='Percentage_Churn', ascending = False).head()
# sorting the records of the dataframe 'c' wrt the column 'Percentage_Churn' in␣
 ↪descending order...
#...and showing top 5 records using head function
```

[41]:
```
churn  False  True  All  Percentage_Churn
state
CA        25     9   34         26.470588
NJ        50    18   68         26.470588
TX        54    18   72         25.000000
MD        53    17   70         24.285714
SC        46    14   60         23.333333
```

---

6 3. Analyzing the 'Area Code' Column

[8]:
```
data.head()
```

[8]:
```
  state  account length  area code phone number international plan  \
0    KS             128        415     382-4657                 no
1    OH             107        415     371-7191                 no
2    NJ             137        415     358-1921                 no
3    OH              84        408     375-9999                yes
4    OK              75        415     330-6626                yes

  voice mail plan  number vmail messages  total day minutes  total day calls  \
0             yes                     25              265.1              110
1             yes                     26              161.6              123
2              no                      0              243.4              114
3              no                      0              299.4               71
4              no                      0              166.7              113

   total day charge  …  total eve calls  total eve charge  \
0             45.07  …               99             16.78
1             27.47  …              103             16.62
2             41.38  …              110             10.30
3             50.90  …               88              5.26
4             28.34  …              122             12.61
```

```
   total night minutes  total night calls  total night charge  \
0                244.7                 91               11.01
1                254.4                103               11.45
2                162.6                104                7.32
3                196.9                 89                8.86
4                186.9                121                8.41

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3               2.70
1                13.7                 3               3.70
2                12.2                 5               3.29
3                 6.6                 7               1.78
4                10.1                 3               2.73

   customer service calls  churn
0                       1  False
1                       1  False
2                       0  False
3                       2  False
4                       3  False

[5 rows x 21 columns]
```

[43]:
```
data['area code'].nunique()          # To show the total number of unique␣
 ↪values present in the column 'area code'
```

[43]:
```
3
```

[44]:
```
data['area code'].unique()           # To show all the unique values of the␣
 ↪column 'area code'...
                                     # it shows the output in the form of 1-D␣
 ↪array
```

[44]:
```
array([415, 408, 510], dtype=int64)
```

[45]:
```
data['area code'].value_counts()     # To show the occurance/count of all␣
 ↪unique values of the column 'area code'

# By default , it shows result in descending order
# To show the results in ascending order, we can write data.state.
 ↪value_counts(ascending=True)
```

[45]:
```
415    1655
510     840
408     838
Name: area code, dtype: int64
```

[46]:
```
data.groupby('area code')['churn'].mean() * 100

# Here, we will consider two columns while using groupby()
# Used groupby on 'area_code' column ... and showing the mean of values of␣
 ↪'churn' column wrt to each area code...
# ...and multiplying the result by 100
```

[46]:
```
area code
408    14.558473
415    14.259819
510    14.880952
Name: churn, dtype: float64
```

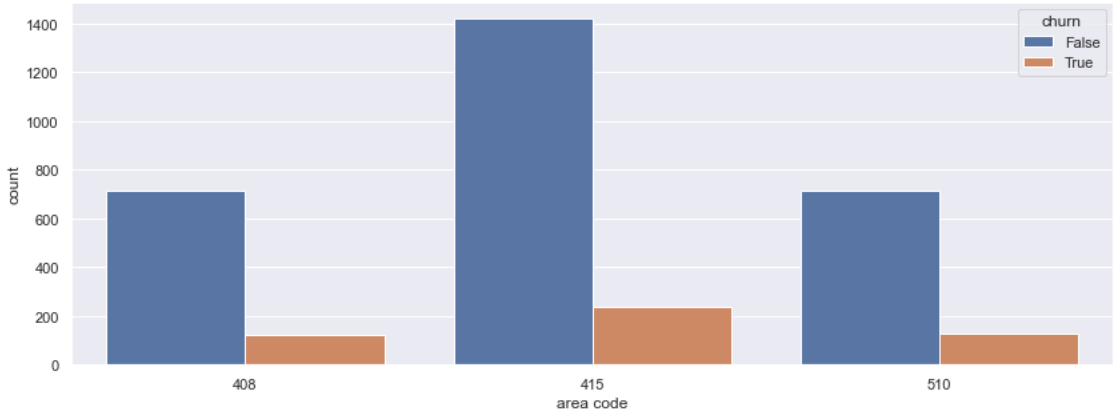
[47]:
```
# Calculatiog Area Code vs Churn Percentage

d = pd.crosstab(data['area code'], data['churn'], margins= True)
                # Using crosstab function from pandas library to compute a␣
 ↪simple...
                #...cross-tabulation of two columns i.e., 'area code' & 'churn'␣
 ↪and saving the output in varibale 'd'
                # margins=True means - to show sum of both values in a new␣
 ↪column 'All'
d['Churn Percentage'] = d[True]/(d['All']) * 100                    # Creating␣
 ↪a new column 'Percantage_Churn'...
                                                    #...which shows the␣
 ↪percentage of churn customers in each area code
d                                                   # showing result of 'd'
```

[47]:

churn	False	True	All	Churn Percentage
area code				
408	716	122	838	14.558473
415	1419	236	1655	14.259819
510	715	125	840	14.880952
All	2850	483	3333	14.491449

[48]:
```
# Drawing a Bar Graph to show Area Code wise Churn

plt.figure(figsize=(14,5))
# Setting the size of the figure as 14x5, using matplotlib library

sns.countplot(x='area code', hue='churn', data=data);
# Drawing a countplot for the column 'area code' from the dataset using seaborn␣
 ↪library...
# ...selecting the Churn column for hue
# 'hue=churn' represents that we want to use column 'churn' for color encoding
# i.e. color the bars for Churn and Not-Churn differently
```

```
# Blue color representing False (Not Churn) and Orange color representing True␣
↪(Churn) for each area code separately
```



---

7 4. Analyzing the 'Account Length' Column

[9]:
```
data.head()
```

[9]:

	state	account length	area code	phone number	international plan	voice mail plan	number vmail messages	total day minutes	total day calls	total day charge	…	total eve calls	total eve charge
0	KS	128	415	382-4657	no	yes	25	265.1	110	45.07	…	99	16.78
1	OH	107	415	371-7191	no	yes	26	161.6	123	27.47	…	103	16.62
2	NJ	137	415	358-1921	no	no	0	243.4	114	41.38	…	110	10.30
3	OH	84	408	375-9999	yes	no	0	299.4	71	50.90	…	88	5.26
4	OK	75	415	330-6626	yes	no	0	166.7	113	28.34	…	122	12.61

```
   total night minutes  total night calls  total night charge  \
0                244.7                 91               11.01
1                254.4                103               11.45
2                162.6                104                7.32
3                196.9                 89                8.86
4                186.9                121                8.41

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3               2.70
1                13.7                 3               3.70
2                12.2                 5               3.29
3                 6.6                 7               1.78
4                10.1                 3               2.73

   customer service calls  churn
0                       1  False
1                       1  False
2                       0  False
3                       2  False
4                       3  False

[5 rows x 21 columns]
```

[50]:
```
data['account length'].nunique()   # To show the total number of unique values present in the column 'account length'
```

[50]:
```
212
```

[11]:
```
# Creating two different dataframes - one for 'churned customers' & second for 'non-churned customers'

churn_data = data[data['churn'] == True]
#creating a new dataframe "churn_data", in which we are considering all the records where churn is 'True'

not_churn_data = data[data['churn'] == False]
#creating a new dataframe "not_churn_data", in which we are considering all the records where churn is 'False'
```

[52]:
```
churn_data.head(2)              # To show Top2 records of the dataframe
```

[52]:
```
   state  account length  area code phone number international plan  \
10    IN              65        415     329-6603                 no
15    NY             161        415     351-7269                 no

   voice mail plan  number vmail messages  total day minutes  total day calls  \
```

```
10              no                     0              129.1              137
15              no                     0              332.9               67

    total day charge  …  total eve calls  total eve charge  \
10             21.95  …              83             19.42
15             56.59  …              97             27.01

    total night minutes  total night calls  total night charge  \
10                208.8                111                9.40
15                160.6                128                7.23

    total intl minutes  total intl calls  total intl charge  \
10                12.7                 6               3.43
15                 5.4                 9               1.46

    customer service calls  churn
10                       4   True
15                       4   True

[2 rows x 21 columns]
```

[53]:
```
not_churn_data.head(2)              # To show Top2 records of the dataframe
```

[53]:
```
  state  account length  area code phone number international plan  \
0    KS             128        415     382-4657                 no
1    OH             107        415     371-7191                 no

  voice mail plan  number vmail messages  total day minutes  total day calls  \
0             yes                     25              265.1              110
1             yes                     26              161.6              123

   total day charge  …  total eve calls  total eve charge  \
0             45.07  …               99             16.78
1             27.47  …              103             16.62

   total night minutes  total night calls  total night charge  \
0                244.7                 91               11.01
1                254.4                103               11.45

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3                2.7
1                13.7                 3                3.7

   customer service calls  churn
0                       1  False
1                       1  False
```

```
[2 rows x 21 columns]
```

[12]:
```
churn_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 483 entries, 10 to 3323
Data columns (total 21 columns):
 #   Column                  Non-Null Count  Dtype
---  ------                  --------------  -----
 0   state                   483 non-null    object
 1   account length          483 non-null    int64
 2   area code               483 non-null    int64
 3   phone number            483 non-null    object
 4   international plan      483 non-null    object
 5   voice mail plan         483 non-null    object
 6   number vmail messages   483 non-null    int64
 7   total day minutes       483 non-null    float64
 8   total day calls         483 non-null    int64
 9   total day charge        483 non-null    float64
 10  total eve minutes       483 non-null    float64
 11  total eve calls         483 non-null    int64
 12  total eve charge        483 non-null    float64
 13  total night minutes     483 non-null    float64
 14  total night calls       483 non-null    int64
 15  total night charge      483 non-null    float64
 16  total intl minutes      483 non-null    float64
 17  total intl calls        483 non-null    int64
 18  total intl charge       483 non-null    float64
 19  customer service calls  483 non-null    int64
 20  churn                   483 non-null    bool
dtypes: bool(1), float64(8), int64(8), object(4)
memory usage: 79.7+ KB
```

[13]:
```
not_churn_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 2850 entries, 0 to 3332
Data columns (total 21 columns):
 #   Column                  Non-Null Count  Dtype
---  ------                  --------------  -----
 0   state                   2850 non-null   object
 1   account length          2850 non-null   int64
 2   area code               2850 non-null   int64
 3   phone number            2850 non-null   object
 4   international plan      2850 non-null   object
 5   voice mail plan         2850 non-null   object
 6   number vmail messages   2850 non-null   int64
 7   total day minutes       2850 non-null   float64
```

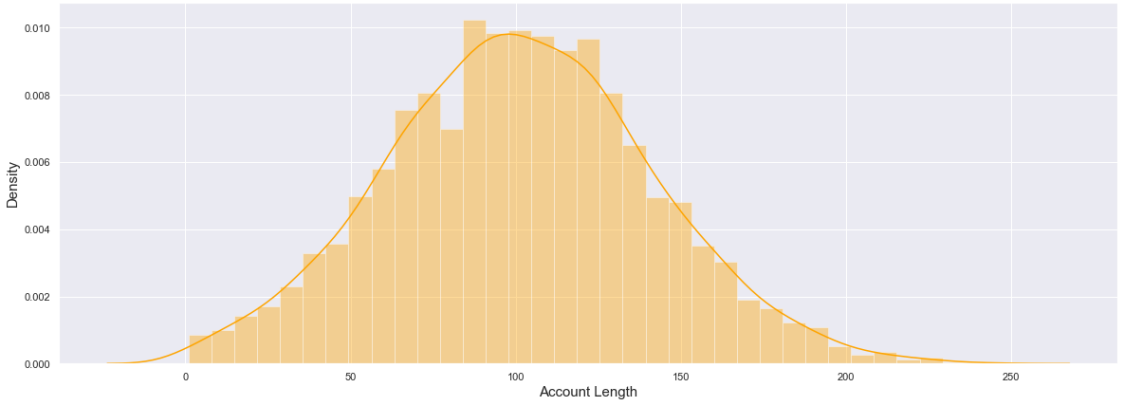
```
 8   total day calls         2850 non-null   int64
 9   total day charge        2850 non-null   float64
 10  total eve minutes       2850 non-null   float64
 11  total eve calls         2850 non-null   int64
 12  total eve charge        2850 non-null   float64
 13  total night minutes     2850 non-null   float64
 14  total night calls       2850 non-null   int64
 15  total night charge      2850 non-null   float64
 16  total intl minutes      2850 non-null   float64
 17  total intl calls        2850 non-null   int64
 18  total intl charge       2850 non-null   float64
 19  customer service calls  2850 non-null   int64
 20  churn                   2850 non-null   bool
dtypes: bool(1), float64(8), int64(8), object(4)
memory usage: 470.4+ KB
```

[14]:
```
data.shape
```

[14]:
```
(3333, 21)
```

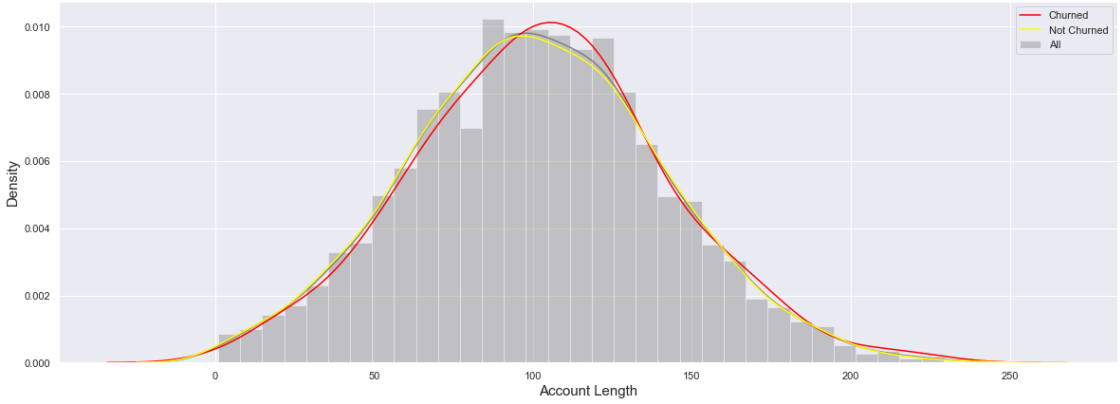
7.0.1 Distribution Plot

[57]:
```
# Creating a Distribution Plot for Account Length column

sns.distplot(data['account length'], color='orange')
plt.xlabel("Account Length", fontsize=15)
plt.ylabel("Density", fontsize=15)
plt.show()

# The distplot shows a histogram with a line on it. It represents the␣
 ↪distribution of a variable against the...
#...density distribution.
# Here, using Seaborn library we are plotting a distribution plot for the␣
 ↪column 'account length'...
# ...with 'orange' color.
# Using matplotlib, setting the label of x-axis as "Account Length", with␣
 ↪fontsize of 15
# Using matplotlib, setting the label of x-axis as "Density", with fontsize of␣
 ↪15
# plt.show() - To show the plot
```



[58]:
```
# Comparing Account Length of 'churned' and 'not churned' customers, using␣
 ↪distplot

sns.distplot(data['account length'], color='grey', label='All')
# Using Seaborn library, we are plotting a distribution plot for the column␣
 ↪'account length' from original dataframe...
# ...with 'grey' color ... and setting the label as 'All'

sns.distplot(churn_data['account length'], hist=False, color='red',␣
 ↪label='Churned')
# Using Seaborn library, we are plotting a distribution plot for the column␣
 ↪'account length' from 'churn_data'...
#...dataframe with 'red' color ... and setting the label as 'Churned'. Here,␣
 ↪hist=False is given because...
#...we don't want to show a histogram for this, we want to show a line only

sns.distplot(not_churn_data['account length'], hist=False, color='yellow',␣
 ↪label='Not Churned')
# Using Seaborn library, we are plotting a distribution plot for the column␣
 ↪'account length' from 'non_churn_data'...
#...dataframe with 'yellow' color ... and setting the label as 'Not Churned'.␣
 ↪Here, hist=False is given because...
#...we don't want to show a histogram for this, we want to show a line only

plt.xlabel("Account Length", fontsize=15)
# Using matplotlib, setting the label of x-axis as "Account Length", with␣
 ↪fontsize of 15

plt.ylabel("Density", fontsize=15)
# Using matplotlib, setting the label of y-axis as "Density", with fontsize of␣
 ↪15
```

```
plt.rcParams['figure.figsize'] = (15,7);
# setting the size of the figure as 15x7 using matplotlib's rcParams

plt.show()       # To show the plot
```



---

8 5. Analyzing the 'International Plan' Column

[15]:
```
data.head()
```

[15]:

	state	account length	area code	phone number	international plan	voice mail plan	number vmail messages	total day minutes	total day calls	total day charge	…	total eve calls	total eve charge	\
0	KS	128	415	382-4657	no	yes	25	265.1	110	45.07	…	99	16.78	
1	OH	107	415	371-7191	no	yes	26	161.6	123	27.47	…	103	16.62	
2	NJ	137	415	358-1921	no	no	0	243.4	114	41.38	…	110	10.30	
3	OH	84	408	375-9999	yes	no	0	299.4	71					
4	OK	75	415	330-6626	yes	no	0	166.7	113					

```
3             50.90  …                88              5.26
4             28.34  …               122             12.61

   total night minutes  total night calls  total night charge  \
0                244.7                 91               11.01
1                254.4                103               11.45
2                162.6                104                7.32
3                196.9                 89                8.86
4                186.9                121                8.41

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3               2.70
1                13.7                 3               3.70
2                12.2                 5               3.29
3                 6.6                 7               1.78
4                10.1                 3               2.73

   customer service calls  churn
0                       1  False
1                       1  False
2                       0  False
3                       2  False
4                       3  False

[5 rows x 21 columns]
```

[60]:
```
data['international plan'].unique()       # To show all the unique values of the column 'international plan'
                                          # Two unique values present - 'no' and 'yes'
```

[60]:
```
array(['no', 'yes'], dtype=object)
```

[61]:
```
data['international plan'].value_counts()

# To show the occurance/count of all unique values of the column 'international plan'
# By default , it shows result in descending order
# To show the results in ascending order, we can write data.state.value_counts(ascending=True)
```

[61]:
```
no     3010
yes     323
Name: international plan, dtype: int64
```

[18]:
```
# Creating a new dataframe 'yes_int_plan',...
```

```
#...containing the records of those Churned customers who were having␣
 ↪international plan

yes_int_plan = data[(data['international plan'] == 'yes') &␣
 ↪(data['churn']==True)]
yes_int_plan

# Here, we have used filtering with '&' operator, to select records satisfying␣
 ↪two conditions -
# 1. 'international plan' == 'yes'
# 2. 'churn' == 'True'

# it shows there are total 137 records satisfying these two conditions
```

[18]:

	state	account length	area code	phone number	international plan	\
41	MD	135	408	383-6029	yes	
115	ME	36	510	363-1069	yes	
144	VT	117	408	390-2390	yes	
198	ME	131	510	353-7292	yes	
214	FL	70	510	366-6345	yes	
…	…	…	…	…	…	
3246	NC	77	408	334-6129	yes	
3255	RI	138	510	411-6823	yes	
3291	MI	119	510	335-7324	yes	
3304	IL	71	510	330-7137	yes	
3320	GA	122	510	411-5677	yes	

	voice mail plan	number vmail messages	total day minutes	\
41	yes	41	173.1	
115	yes	42	196.8	
144	no	0	167.1	
198	yes	26	292.9	
214	no	0	226.7	
…	…	…	…	
3246	yes	44	103.2	
3255	no	0	286.2	
3291	yes	22	172.1	
3304	no	0	186.1	
3320	no	0	140.0	

	total day calls	total day charge	…	total eve calls	\
41	85	29.43	…	107	
115	89	33.46	…	122	
144	86	28.41	…	87	
198	101	49.79	…	97	
214	98	38.54	…	115	
…	…	…	…	…	

```
3246              117             17.54  …               86
3255               61             48.65  …               60
3291              119             29.26  …              133
3304              114             31.64  …              140
3320              101             23.80  …               77

      total eve charge  total night minutes  total night calls  \
41               17.33                122.2                 78
115              21.67                138.3                126
144              15.09                249.4                132
198              16.97                255.3                127
214              19.39                 73.2                 93
…                …                    …                    …
3246             20.09                203.5                101
3255             15.91                146.2                114
3291             19.01                150.0                 94
3304             16.88                206.5                 80
3320             16.69                120.1                133

      total night charge  total intl minutes  total intl calls  \
41                  5.50                14.6                15
115                 6.22                20.0                 6
144                11.22                14.1                 7
198                11.49                13.8                 7
214                 3.29                17.6                 4
…                   …                   …                    …
3246                9.16                11.9                 2
3255                6.58                11.0                 4
3291                6.75                13.9                20
3304                9.29                13.8                 5
3320                5.40                 9.7                 4

      total intl charge  customer service calls  churn
41                 3.94                       0   True
115                5.40                       0   True
144                3.81                       2   True
198                3.73                       4   True
214                4.75                       2   True
…                  …                          …   …
3246               3.21                       0   True
3255               2.97                       2   True
3291               3.75                       1   True
3304               3.73                       4   True
3320               2.62                       4   True

[137 rows x 21 columns]
```

[19]:
```
# Creating a new dataframe 'no_int_plan',...
#...containing the records of those Churned customers who don't have␣
 ↪international plan

no_int_plan = data[(data['international plan'] == 'no') & (data['churn']==True)]
no_int_plan

# Here, we have used filtering with '&' operator, to select records satisfying␣
 ↪two conditions -
# 1. 'international plan' == 'no'
# 2. 'churn' == 'True'

# it shows there are total 346 records satisfying these two conditions
```

[19]:

	state	account length	area code	phone number	international plan	\
10	IN	65	415	329-6603	no	
15	NY	161	415	351-7269	no	
21	CO	77	408	393-7984	no	
33	AZ	12	408	360-1596	no	
48	ID	119	415	398-1294	no	
…	…	…	…	…	…	
3280	AR	76	408	345-3614	no	
3287	KS	170	415	404-5840	no	
3301	CA	84	415	417-1488	no	
3322	MD	62	408	409-1856	no	
3323	IN	117	415	362-5899	no	

	voice mail plan	number vmail messages	total day minutes	\
10	no	0	129.1	
15	no	0	332.9	
21	no	0	62.4	
33	no	0	249.6	
48	no	0	159.1	
…	…	…	…	
3280	no	0	107.3	
3287	yes	42	199.5	
3301	no	0	280.0	
3322	no	0	321.1	
3323	no	0	118.4	

	total day calls	total day charge	…	total eve calls	\
10	137	21.95	…	83	
15	67	56.59	…	97	
21	89	10.61	…	121	
33	118	42.43	…	119	
48	114	27.05	…	117	
…	…	…	…	…	

```
3280              140             18.24  …              133
3287              119             33.92  …               90
3301              113             47.60  …               90
3322              105             54.59  …              122
3323              126             20.13  …               97

      total eve charge  total night minutes  total night calls  \
10               19.42                208.8                111
15               27.01                160.6                128
21               14.44                209.6                 64
33               21.45                280.2                 90
48               19.66                143.2                 91
…                    …                    …                  …
3280             20.25                271.8                116
3287             11.48                184.6                 49
3301             17.19                156.8                103
3322             22.57                180.5                 72
3323             21.19                227.0                 56

      total night charge  total intl minutes  total intl calls  \
10                  9.40                12.7                 6
15                  7.23                 5.4                 9
21                  9.43                 5.7                 6
33                 12.61                11.8                 3
48                  6.44                 8.8                 3
…                    …                    …                  …
3280               12.23                10.0                 3
3287                8.31                10.9                 3
3301                7.06                10.4                 4
3322                8.12                11.5                 2
3323               10.22                13.6                 3

      total intl charge  customer service calls  churn
10                 3.43                       4   True
15                 1.46                       4   True
21                 1.54                       5   True
33                 3.19                       1   True
48                 2.38                       5   True
…                     …                       …    …
3280               2.70                       4   True
3287               2.94                       4   True
3301               2.81                       0   True
3322               3.11                       4   True
3323               3.67                       5   True

[346 rows x 21 columns]
```

[64]:
```
# we noticed that 323 customers were using international plan, and 137 customers churned out of them
# means a churn percentage of 42.4

137/323*100
```

[64]:
```
42.414860681114554
```

[65]:
```
# we noticed that 3010 customers were not using international plan, and 346 customers churned out of them
# means a churn percentage of 11.4

346/3010*100
```

[65]:
```
11.495016611295682
```

crosstab( )

[21]:
```
int_plan_data = pd.crosstab(data['international plan'], data['churn'], margins=True)
            # Using crosstab function from pandas library to compute a simple cross-tabulation...
            #...of two columns i.e., 'international plan' & 'churn' and saving the output in varibale 'int_plan_data'
            # ... margins=True means - to show the sum of both values in a new column 'All'
int_plan_data
```

[21]:

churn	False	True	All
international plan			
no	2664	346	3010
yes	186	137	323
All	2850	483	3333

[22]:
```
#creating a new column 'churn percent'- to show the churn percentage of the customers who were using international plan

int_plan_data['churn percent'] = int_plan_data[True]/int_plan_data['All']*100
int_plan_data
```

[22]:

churn	False	True	All	churn percent
international plan				
no	2664	346	3010	11.495017
yes	186	137	323	42.414861
All	2850	483	3333	14.491449

[23]:
```
int_plan_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 3 entries, no to All
Data columns (total 4 columns):
 #   Column          Non-Null Count  Dtype
---  ------          --------------  -----
 0   False           3 non-null      int64
 1   True            3 non-null      int64
 2   All             3 non-null      int64
 3   churn percent   3 non-null      float64
dtypes: float64(1), int64(3)
memory usage: 120.0+ bytes
```

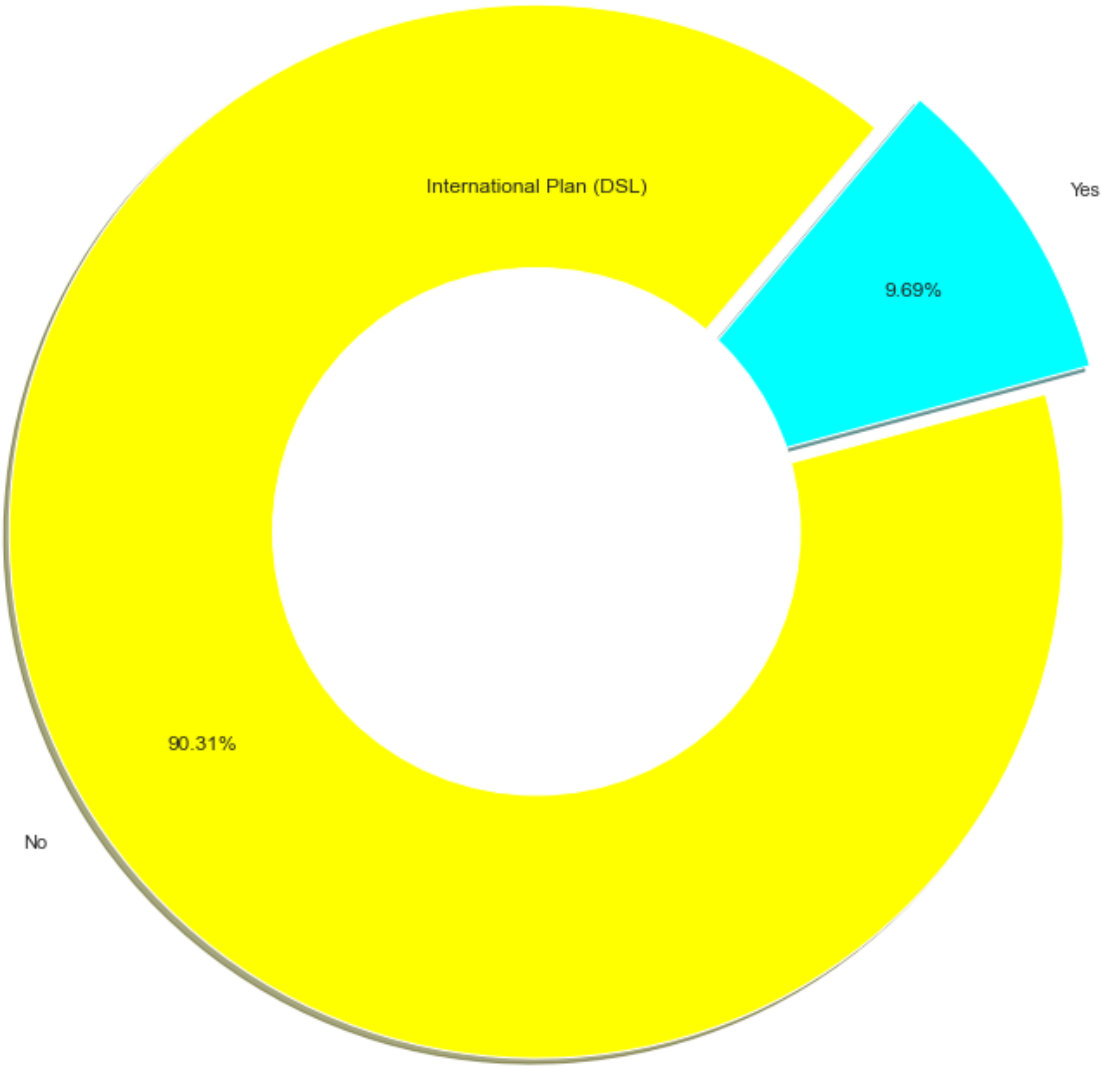
[69]:
```
i = data['international plan'].value_counts()
i

# To show the count of the unique values of the column 'international plan',
 ↪and saving in variable 'i'
```

[69]:
```
no     3010
yes     323
Name: international plan, dtype: int64
```

8.0.1 Donut Chart

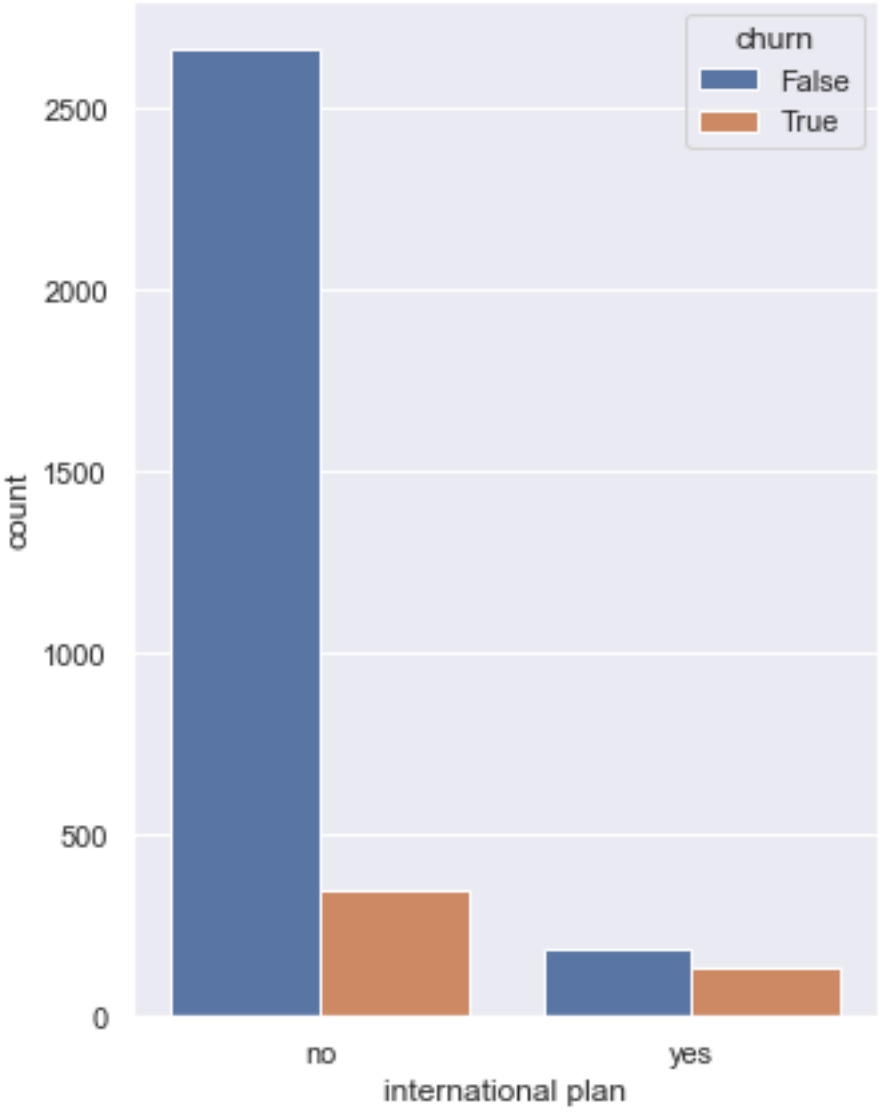
[70]:
```
# Creating a Donut Chart of customers count 'having international plan' and
 ↪'not having international plan'

plt.pie(i, labels=['No', 'Yes'], colors=['yellow','cyan'], startangle=50,
 ↪shadow=True, radius=2,
        explode=(0,0.2), autopct='%1.2f%%',  pctdistance=0.75);

circle = plt.Circle((0,0), 1, color='white')
c = plt.gcf()

c.gca().add_artist(circle)

plt.title("International Plan (DSL)")
plt.rcParams['figure.figsize'] = (5,7)
plt.show()

# Donut chart is a modified Pie chart, with an area from center cut out
#1 First, drawing a pie plot, using Matplotlib library
# The data is taken from the variable 'i' .... labels are given as 'No' & 'Yes'
 ↪...
# ... color is given to each label 'yellow' & 'cyan' ... startangle means the
 ↪angle for slicing, set as 50 ...
```

```
# ... shadow is True means it will drop some shadow of the chart ... radius of␣
 ↪the circle is set as 2 ...
# ... explode is used to cut the slice out of the figure ...
# ... autopct is used to show the % on the chart upto required decimal points ..
 ↪.
# ... pctdistance is given for distance of % from the center
#2 Second, using plt.Circle...we will create a circle and save it in the␣
 ↪variable name 'circle' ...
# ... putting (0,0) by default ... 1 is radius of circle ... and color of␣
 ↪circle is 'white'
# plt.gcf() is used to get the current figure ... and we are saving it in␣
 ↪variable 'c'
#3 Third, we will add the 'circle' at the center of pie chart ... using gca().
 ↪add_artist()
# We have given the title to the chart using plt.title()
# Using rcParams from matplotlib library, setting the size of the figure as 5x7
# plt.show() - To show the chart
```



Countplot

[71]:
```
sns.countplot(data['international plan'], hue='churn', data=data)
plt.rcParams['figure.figsize'] = (10,4)
plt.show()

# Drawing a countplot for the column 'international plan' from the dataset␣
 ↪using seaborn library...
# ...selecting the Churn column for hue
# 'hue=churn' represents that we want to use column 'churn' for color encoding
# i.e. color the bars for Churn and Not-Churn differently
# Blue color representing False (Not Churn) and Orange color representing True␣
 ↪(Churn)
```



9 6. Analyzing the 'Voice Mail Plan' Column

[24]:
```
data.head()
```

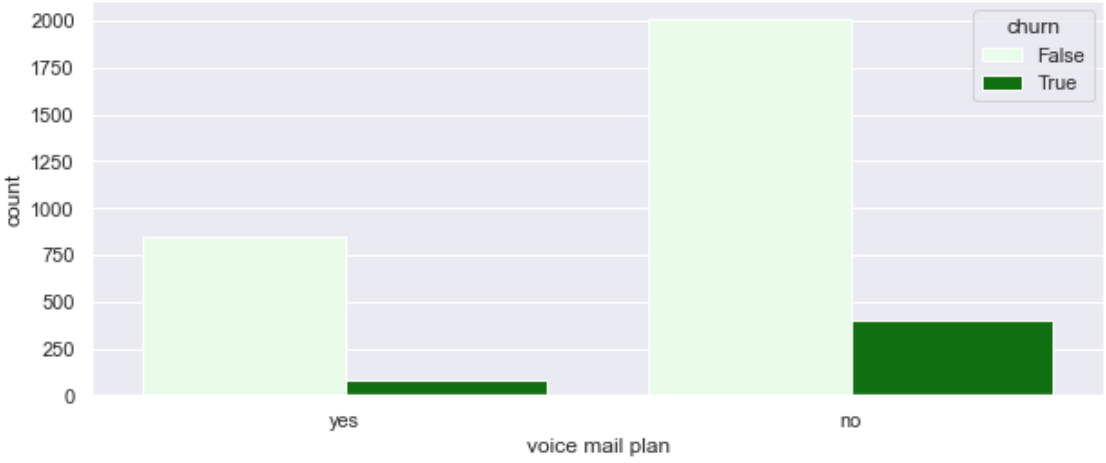
[24]:
```
  state  account length  area code phone number international plan  \
0    KS             128        415     382-4657                 no
1    OH             107        415     371-7191                 no
```

```
2    NJ             137        415     358-1921                 no
3    OH              84        408     375-9999                yes
4    OK              75        415     330-6626                yes

  voice mail plan  number vmail messages  total day minutes  total day calls  \
0             yes                     25              265.1              110
1             yes                     26              161.6              123
2              no                      0              243.4              114
3              no                      0              299.4               71
4              no                      0              166.7              113

   total day charge  …  total eve calls  total eve charge  \
0             45.07  …               99             16.78
1             27.47  …              103             16.62
2             41.38  …              110             10.30
3             50.90  …               88              5.26
4             28.34  …              122             12.61

   total night minutes  total night calls  total night charge  \
0                244.7                 91                11.01
1                254.4                103                11.45
2                162.6                104                 7.32
3                196.9                 89                 8.86
4                186.9                121                 8.41

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3               2.70
1                13.7                 3               3.70
2                12.2                 5               3.29
3                 6.6                 7               1.78
4                10.1                 3               2.73

   customer service calls  churn
0                       1  False
1                       1  False
2                       0  False
3                       2  False
4                       3  False

[5 rows x 21 columns]
```

Countplot

[73]:
```
sns.countplot(data['voice mail plan'], hue="churn", data=data, color='green')
plt.rcParams['figure.figsize']=(10,5)
plt.show()
```

```
# Drawing a countplot for the column 'voice mail plan' from the dataset using␣
 ↪seaborn library...
# ...selecting the 'churn' column for hue .... color is set as 'green' for bars
# 'hue=churn' represents that we want to use column 'churn' for color encoding
# i.e. color the bars for Churn and Not-Churn differently
# Light Green color representing False (Not Churn) and Dark Green color␣
 ↪representing True (Churn)
```



[74]:
```
vmplan = pd.crosstab(data['voice mail plan'], data['churn'], margins=True)
                     # Using crosstab function from pandas library to compute␣
 ↪a simple cross-tabulation...
                     # ...of two columns i.e., 'voice mail plan' & 'churn'␣
 ↪and saving the output in varibale 'vmplan'
                     # ... margins=True means - to show the sum of both␣
 ↪values in a new column 'All'
vmplan
```

[74]:

churn	False	True	All
voice mail plan			
no	2008	403	2411
yes	842	80	922
All	2850	483	3333

[75]:
```
vmplan['churn percent'] = vmplan[True] / vmplan['All'] * 100
vmplan

#creating a new column 'churn percent'- to show the churn percentage of the␣
 ↪customers who were using voice mail plan
```

[75]:
```
churn            False  True   All  churn percent
voice mail plan
no                2008   403  2411      16.715056
yes                842    80   922       8.676790
All               2850   483  3333      14.491449
```

[76]:
```
data['voice mail plan'].unique()              # To show all the unique values␣
 ↪of the column 'voice mail plan'
```

[76]:
```
array(['yes', 'no'], dtype=object)
```

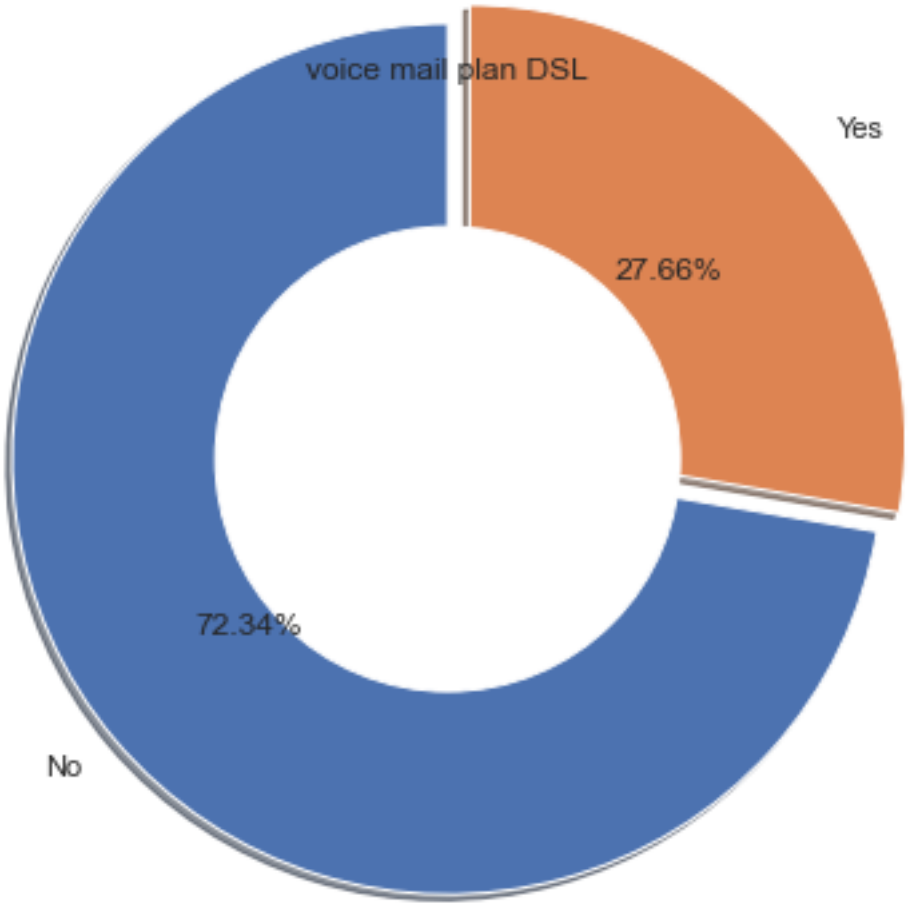
[77]:
```
v = data['voice mail plan'].value_counts()
v
#To show the occurance/count of all unique values of the column 'voice mail␣
 ↪plan', and saving the output in variable 'v'
```

[77]:
```
no     2411
yes     922
Name: voice mail plan, dtype: int64
```

9.0.1 Donut Chart

[78]:
```
# Creating a Donut Chart of customers count 'having voice mail plan' and 'not␣
 ↪having voice mail plan'

plt.pie(v, labels=['No','Yes'], startangle=90, shadow=True, radius=1.5,␣
 ↪explode=(0,0.1), autopct='%1.2f%%')

circle = plt.Circle((0,0),0.8, color='white')
c = plt.gcf()

c.gca().add_artist(circle)

plt.title("voice mail plan DSL")
plt.show()

# Donut chart is a modified Pie chart, with an area from center cut out
#1 First, drawing a pie plot, using Matplotlib library
# The data is taken from the variable 'v' .... labels are given as 'No' & 'Yes'␣
 ↪...
# ... startangle means the angle for slicing, set as 90 ...
# ... shadow is True means it will drop some shadow of the chart ... radius of␣
 ↪the circle is set as 1.5 ...
# ... explode is used to cut the slice out of the figure ...
# ... autopct is used to show the % on the chart upto required decimal points ..
 ↪.
```

```
#2 Second, using plt.Circle...we will create a circle and save it in the␣
 ↪variable name 'circle' ...
# ... putting (0,0) by default ... 0.8 is radius of circle ... and color of␣
 ↪circle is 'white'
# plt.gcf() is used to get the current figure ... and we are saving it in␣
 ↪variable 'c'
#3 Third, we will add the 'circle' at the center of pie chart ... using gca().
 ↪add_artist()
# We have given the title to the chart using plt.title()
# plt.show() - To show the chart
```



10 7. Analyzing the 'Number Vmail Messages' Column

[25]:
```
data.head()
```

[25]:

	state	account length	area code	phone number	international plan	voice mail plan	number vmail messages	total day minutes	total day calls	total day charge	…	total eve calls	total eve charge	total night minutes	total night calls	total night charge	total intl minutes	total intl calls	total intl charge	customer service calls	churn
0	KS	128	415	382-4657	no	yes	25	265.1	110	45.07	…	99	16.78	244.7	91	11.01	10.0	3	2.70	1	False
1	OH	107	415	371-7191	no	yes	26	161.6	123	27.47	…	103	16.62	254.4	103	11.45	13.7	3	3.70	1	False
2	NJ	137	415	358-1921	no	no	0	243.4	114	41.38	…	110	10.30	162.6	104	7.32	12.2	5	3.29	0	False
3	OH	84	408	375-9999	yes	no	0	299.4	71	50.90	…	88	5.26	196.9	89	8.86	6.6	7	1.78	2	False
4	OK	75	415	330-6626	yes	no	0	166.7	113	28.34	…	122	12.61	186.9	121	8.41	10.1	3	2.73	3	False

[5 rows x 21 columns]

[80]:
```
data['number vmail messages'].unique()      # To show the unique values present␣
 ↪in the column 'number vmail messages'
```

[80]:
```
array([25, 26,  0, 24, 37, 27, 33, 39, 30, 41, 28, 34, 46, 29, 35, 21, 32,
       42, 36, 22, 23, 43, 31, 38, 40, 48, 18, 17, 45, 16, 20, 14, 19, 51,
       15, 11, 12, 47,  8, 44, 49,  4, 10, 13, 50,  9], dtype=int64)
```

[81]:
```
data['number vmail messages'].value_counts().head()

# Showing the occurance/count of top 5 unique values of the column 'number␣
 ↪vmail messages'
```

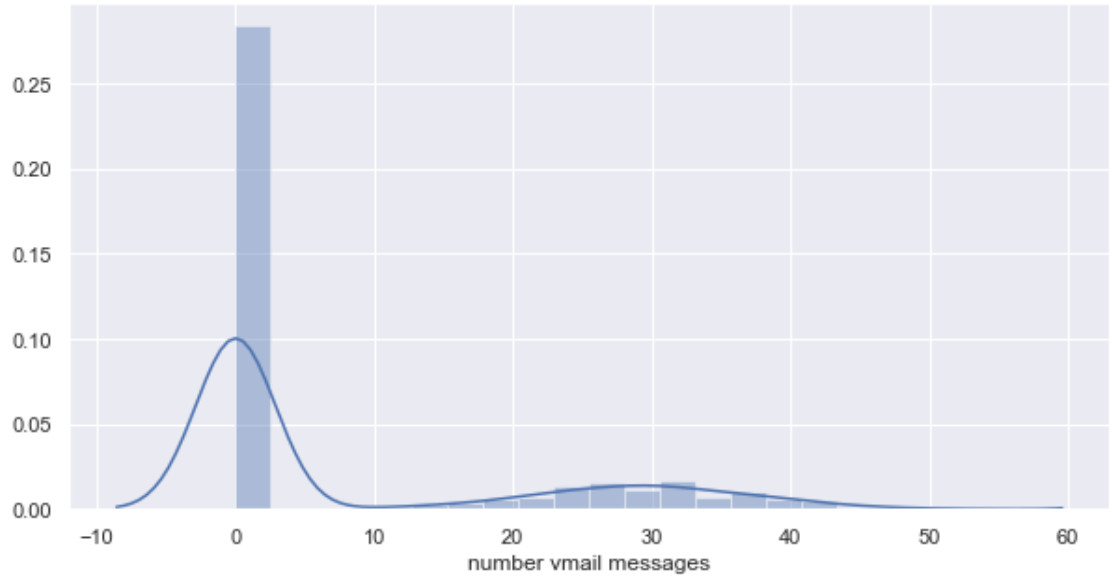
[81]:
```
0     2411
31      60
29      53
28      51
33      46
Name: number vmail messages, dtype: int64
```

[82]:
```
data['number vmail messages'].describe()

# To check the summary statistics of the column. It checks extreme outliers and␣
 ↪large deviations etc.
# It shows the count of non-null values,mean,std,minimum,maximum,25th,50th,75th␣
 ↪percentile value.
# Percentile means - how many of the values are less than the given percentile.
```

[82]:
```
count    3333.000000
mean        8.099010
std        13.688365
min         0.000000
25%         0.000000
50%         0.000000
75%        20.000000
max        51.000000
Name: number vmail messages, dtype: float64
```

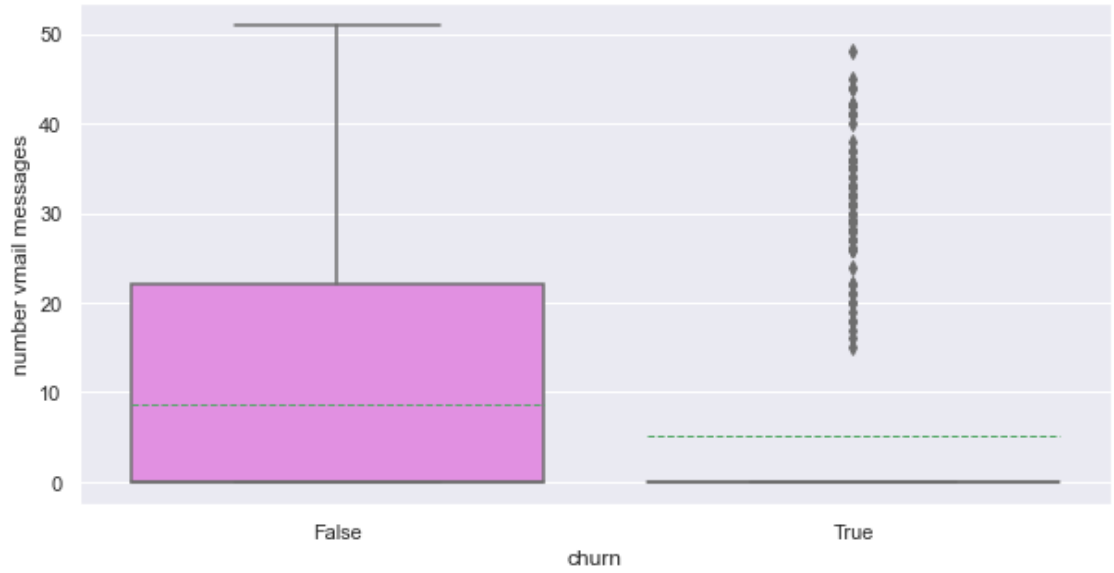
Distribution Plot

[83]:
```
sns.distplot(data['number vmail messages']);

# From Seaborn library we are generating a distribution plot for the column␣
 ↪'number vmail messages'
```



10.0.1 Box Plot

[84]:
```
sns.boxplot(x='churn', y='number vmail messages', color='violet',␣
 ↪meanline=True, showmeans=True, sym='r+', data=data)
                            # creating a box plot for the cdolumn 'number␣
 ↪vmail messages' ... grouped by column 'churn'
plt.figure(figsize =(5, 5))  # Setting the size of the figure as 5x5, using␣
 ↪matplotlib library
plt.show()                   # to show the box plot

# In box plot, a box is plotted from Ist quartile to IIIrd quartile
# here, in this boxplot ... we have values from 'churn' column on x-axis and␣
 ↪distribution at y-axis ...
# there are no outliers ... the boundaries are at 0 and 51 ... the minimum␣
 ↪value is at 0 ...
# the first quartile (25%) is at 0 ... the second quartile (50%) is also at 0 ..
 ↪.
# the third quartile (75th percentile) is at 20 ... the maximum value is at 51 .
 ↪..
# color is set as 'violet' ... meanline=True indicates the mean line ...␣
 ↪showmeans=True indicates the mean point ...
# sym='r+' to change the symbol to + sign with red color ... data is taken from␣
 ↪our original dataframe 'data'
```




```
<Figure size 360x360 with 0 Axes>
```

11 8. Analyzing the 'Customer Service Calls' Column

[26]:
```
data.head()
```

[26]:

	state	account length	area code	phone number	international plan	voice mail plan	number vmail messages	total day minutes	total day calls	total day charge	…	total eve calls	total eve charge	\
0	KS	128	415	382-4657	no	yes	25	265.1	110	45.07	…	99	16.78	
1	OH	107	415	371-7191	no	yes	26	161.6	123	27.47	…	103	16.62	
2	NJ	137	415	358-1921	no	no	0	243.4	114					
3	OH	84	408	375-9999	yes	no	0	299.4	71					
4	OK	75	415	330-6626	yes	no	0	166.7	113					

```
2             41.38  …              110             10.30
3             50.90  …               88              5.26
4             28.34  …              122             12.61

   total night minutes  total night calls  total night charge  \
0                244.7                 91               11.01
1                254.4                103               11.45
2                162.6                104                7.32
3                196.9                 89                8.86
4                186.9                121                8.41

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3               2.70
1                13.7                 3               3.70
2                12.2                 5               3.29
3                 6.6                 7               1.78
4                10.1                 3               2.73

   customer service calls  churn
0                       1  False
1                       1  False
2                       0  False
3                       2  False
4                       3  False

[5 rows x 21 columns]
```

[86]:
```
data['customer service calls'].nunique()

# To show the total number of unique values present in column 'customer service␣
 ↪calls'
```

[86]:
```
10
```

[87]:
```
data['customer service calls'].unique()     # To show all the unique values of␣
 ↪the column 'customer service calls'
```

[87]:
```
array([1, 0, 2, 3, 4, 5, 7, 9, 6, 8], dtype=int64)
```

[88]:
```
data['customer service calls'].value_counts()

# To show the count of the unique values of the column 'customer service calls'␣
 ↪from original dataframe
```

[88]:
```
1    1181
2     759
0     697
```

```
3     429
4     166
5      66
6      22
7       9
8       2
9       2
Name: customer service calls, dtype: int64
```

[89]:
```
churn_data = data[data['churn']]    # creating a new dataframe 'churn_data',␣
 ↪with the records of all churned customers
churn_data
```

[89]:

	state	account length	area code	phone number	international plan	voice mail plan	number vmail messages	total day minutes	total day calls	total day charge	…	total eve calls
10	IN	65	415	329-6603	no	no	0	129.1	137	21.95	…	83
15	NY	161	415	351-7269	no	no	0	332.9	67	56.59	…	97
21	CO	77	408	393-7984	no	no	0	62.4	89	10.61	…	121
33	AZ	12	408	360-1596	no	no	0	249.6	118	42.43	…	119
41	MD	135	408	383-6029	yes	yes	41	173.1	85	29.43	…	107
…	…	…	…	…	…	…	…	…	…	…	…	…
3301	CA	84	415	417-1488	no	no	0	280.0	113	47.60	…	90
3304	IL	71	510	330-7137	yes	no	0	186.1				
3320	GA	122	510	411-5677	yes	no	0	140.0				
3322	MD	62	408	409-1856	no	no	0	321.1				
3323	IN	117	415	362-5899	no	no	0	118.4				

```
3304              114             31.64  …              140
3320              101             23.80  …               77
3322              105             54.59  …              122
3323              126             20.13  …               97

      total eve charge  total night minutes  total night calls  \
10               19.42                208.8                111
15               27.01                160.6                128
21               14.44                209.6                 64
33               21.45                280.2                 90
41               17.33                122.2                 78
…                    …                    …                  …
3301             17.19                156.8                103
3304             16.88                206.5                 80
3320             16.69                120.1                133
3322             22.57                180.5                 72
3323             21.19                227.0                 56

      total night charge  total intl minutes  total intl calls  \
10                  9.40                12.7                 6
15                  7.23                 5.4                 9
21                  9.43                 5.7                 6
33                 12.61                11.8                 3
41                  5.50                14.6                15
…                    …                   …                  …
3301                7.06                10.4                 4
3304                9.29                13.8                 5
3320                5.40                 9.7                 4
3322                8.12                11.5                 2
3323               10.22                13.6                 3

      total intl charge  customer service calls  churn
10                 3.43                       4   True
15                 1.46                       4   True
21                 1.54                       5   True
33                 3.19                       1   True
41                 3.94                       0   True
…                     …                       …    …
3301               2.81                       0   True
3304               3.73                       4   True
3320               2.62                       4   True
3322               3.11                       4   True
3323               3.67                       5   True

[483 rows x 21 columns]
```

[90]:
```
churn_data['customer service calls'].value_counts()

# To show the count/occurance of the unique values of the column 'customer␣
 ↪service calls'
```

[90]:
```
1    122
0     92
2     87
4     76
3     44
5     40
6     14
7      5
9      2
8      1
Name: customer service calls, dtype: int64
```

crosstab( )

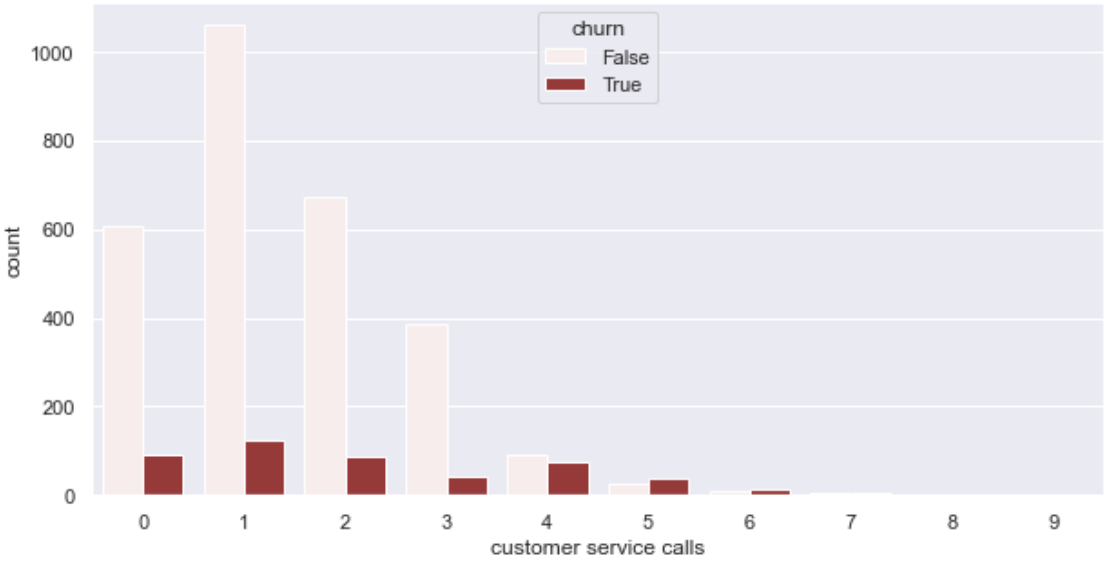
[91]:
```
cscalls = pd.crosstab(data['customer service calls'], data['churn'],␣
 ↪margins=True)
                    # Using crosstab function from pandas library to compute a␣
 ↪simple cross-tabulation...
                    # ...of two columns i.e., 'customer service calls' &␣
 ↪'churn' and saving the output in varibale 'c'
                    # ... margins=True means - to show the sum of both values␣
 ↪in a new column 'All'
cscalls['churn percent'] = cscalls[True] / cscalls['All'] * 100
                                    # Creating a new column 'churn percent',␣
 ↪which shows the...
                                    # ...percentage of churn customers against␣
 ↪each count of customer service calls
cscalls
```

[91]:

churn	False	True	All	churn percent
customer service calls				
0	605	92	697	13.199426
1	1059	122	1181	10.330229
2	672	87	759	11.462451
3	385	44	429	10.256410
4	90	76	166	45.783133
5	26	40	66	60.606061
6	8	14	22	63.636364
7	4	5	9	55.555556
8	1	1	2	50.000000
9	0	2	2	100.000000
All	2850	483	3333	14.491449

Count Plot

[92]:
```
sns.countplot('customer service calls', hue='churn', color='brown', data=data)

# Drawing a countplot for the column 'customer service calls' from the dataset␣
 ↪using seaborn library...
# ...selecting the 'churn' column for hue .... color is set as 'black' for bars
# 'hue=churn' represents that we want to use column 'churn' for color encoding
# i.e. color the bars for Churn and Not-Churn differently
# Light color representing False (Not Churn) and Dark color representing True␣
 ↪(Churn)

plt.show()   # to show the plot
```



---

12 9. Analyzing the 'Per Minute Charge'

[27]:
```
data.head()
```

[27]:

	state	account length	area code	phone number	international plan	\
0	KS	128	415	382-4657	no	
1	OH	107	415	371-7191	no	
2	NJ	137	415	358-1921	no	
3	OH	84	408	375-9999	yes	

```
4    OK              75        415     330-6626                yes

  voice mail plan  number vmail messages  total day minutes  total day calls  \
0             yes                    25              265.1              110
1             yes                    26              161.6              123
2              no                     0              243.4              114
3              no                     0              299.4               71
4              no                     0              166.7              113

   total day charge  …  total eve calls  total eve charge  \
0             45.07  …               99             16.78
1             27.47  …              103             16.62
2             41.38  …              110             10.30
3             50.90  …               88              5.26
4             28.34  …              122             12.61

   total night minutes  total night calls  total night charge  \
0                244.7                 91                11.01
1                254.4                103                11.45
2                162.6                104                 7.32
3                196.9                 89                 8.86
4                186.9                121                 8.41

   total intl minutes  total intl calls  total intl charge  \
0                10.0                 3               2.70
1                13.7                 3               3.70
2                12.2                 5               3.29
3                 6.6                 7               1.78
4                10.1                 3               2.73

   customer service calls  churn
0                       1  False
1                       1  False
2                       0  False
3                       2  False
4                       3  False

[5 rows x 21 columns]
```

[94]:
```
dc_pm = data['total day charge'].mean()/data['total day minutes'].mean()
# checking per minute day charge
ec_pm = data['total eve charge'].mean()/data['total eve minutes'].mean()
# checking per minute evening charge
nc_pm = data['total night charge'].mean()/data['total night minutes'].mean()
# checking per minute night charge
intc_pm = data['total intl charge'].mean()/data['total intl minutes'].mean()
# checking per minute international charge
```

[95]:
```
print(dc_pm)      # printing per minute day charge
print(ec_pm)      # printing per minute evening charge
print(nc_pm)      # printing per minute night charge
print(intc_pm)    # printing per minute international charge
```

```
0.1700030073913066
0.08500104871485774
0.04500041448440013
0.2700500279887098
```

Bar Plot

[96]:
```
sns.barplot(x=['day','evening', 'night', 'int'], y=[dc_pm, ec_pm, nc_pm,␣
↪intc_pm])
plt.show()

# drawing a bar plot to represent different 'per minute call chagres'
# on x-axis we have given the naming .... on y-axis we have given the charges
# plt.show() - to show the plot
```