

In [109]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sb
%matplotlib inline
from sklearn.model_selection import train_test_split as tts
from sklearn.preprocessing import LabelEncoder
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import confusion_matrix
from sklearn.model_selection import cross_val_score
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from sklearn.cluster import KMeans
```

In [2]:

```
import warnings
warnings.simplefilter("ignore")
```

In [3]:

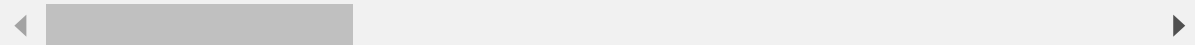
```
df = pd.read_csv('Python_Case_Studytop.csv')
```

In [4]:

```
df.head()
```

Out[4]:

| | State | Account_length | Area_code | International_plan | Voice_mail_plan | Number_vmail_messages |
|---|-------|----------------|-----------|--------------------|-----------------|-----------------------|
| 0 | KS | 128 | 415 | No | Yes | |
| 1 | OH | 107 | 415 | No | Yes | |
| 2 | NJ | 137 | 415 | No | No | |
| 3 | OH | 84 | 408 | Yes | No | |
| 4 | OK | 75 | 415 | Yes | No | |



In [5]:

```
df.columns
```

Out[5]:

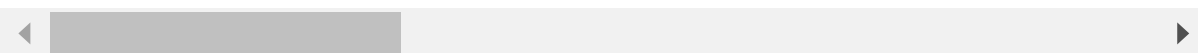
```
Index(['State', 'Account_length', 'Area_code', 'International_plan',  
      'Voice_mail_plan', 'Number_vmail_messages', 'Total_day_minutes',  
      'Total_day_calls', 'Total_day_charge', 'Total_eve_minutes',  
      'Total_eve_calls', 'Total_eve_charge', 'Total_night_minutes',  
      'Total_night_calls', 'Total_night_charge', 'Total_intl_minutes',  
      'Total_intl_calls', 'Total_intl_charge', 'Customer_service_calls',  
      'Churn'],  
      dtype='object')
```

In [6]:

```
df.describe()
```

Out[6]:

| | Account_length | Area_code | Number_vmail_messages | Total_day_minutes | Total_day_call |
|--------------|----------------|-------------|-----------------------|-------------------|----------------|
| count | 2666.000000 | 2666.000000 | 2666.000000 | 2666.000000 | 2666.000000 |
| mean | 100.620405 | 437.438860 | 8.021755 | 179.48162 | 100.31020 |
| std | 39.563974 | 42.521018 | 13.612277 | 54.21035 | 19.98816 |
| min | 1.000000 | 408.000000 | 0.000000 | 0.00000 | 0.00000 |
| 25% | 73.000000 | 408.000000 | 0.000000 | 143.40000 | 87.00000 |
| 50% | 100.000000 | 415.000000 | 0.000000 | 179.95000 | 101.00000 |
| 75% | 127.000000 | 510.000000 | 19.000000 | 215.90000 | 114.00000 |
| max | 243.000000 | 510.000000 | 50.000000 | 350.80000 | 160.00000 |



In [7]:

```
df.isnull().any()
```

Out[7]:

| | |
|------------------------|-------|
| State | False |
| Account_length | False |
| Area_code | False |
| International_plan | False |
| Voice_mail_plan | False |
| Number_vmail_messages | False |
| Total_day_minutes | False |
| Total_day_calls | False |
| Total_day_charge | False |
| Total_eve_minutes | False |
| Total_eve_calls | False |
| Total_eve_charge | False |
| Total_night_minutes | False |
| Total_night_calls | False |
| Total_night_charge | False |
| Total_intl_minutes | False |
| Total_intl_calls | False |
| Total_intl_charge | False |
| Customer_service_calls | False |
| Churn | False |

dtype: bool

In [8]:

```
df.isnull().sum()
```

Out[8]:

| | |
|------------------------|---|
| State | 0 |
| Account_length | 0 |
| Area_code | 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

In [9]:

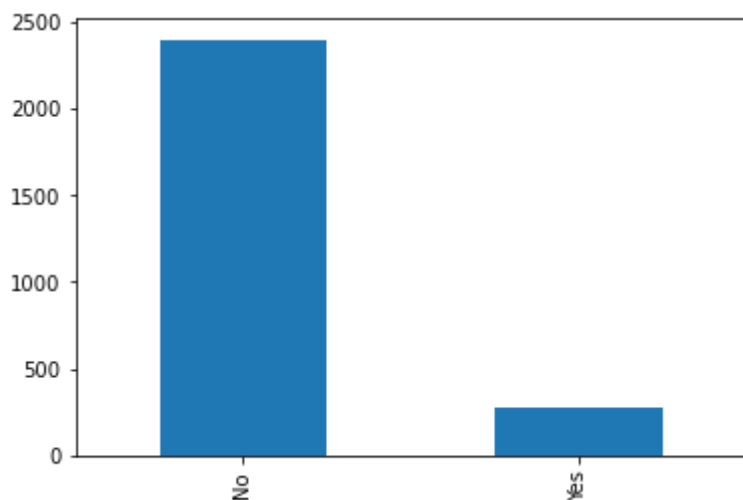
```
df['International_plan'].value_counts()
```

Out[9]:

```
No      2396
Yes      270
Name: International_plan, dtype: int64
```

In [10]:

```
df['International_plan'].value_counts().plot(kind='bar');
```



In [11]:

```
df['Voice_mail_plan'].value_counts()
```

Out[11]:

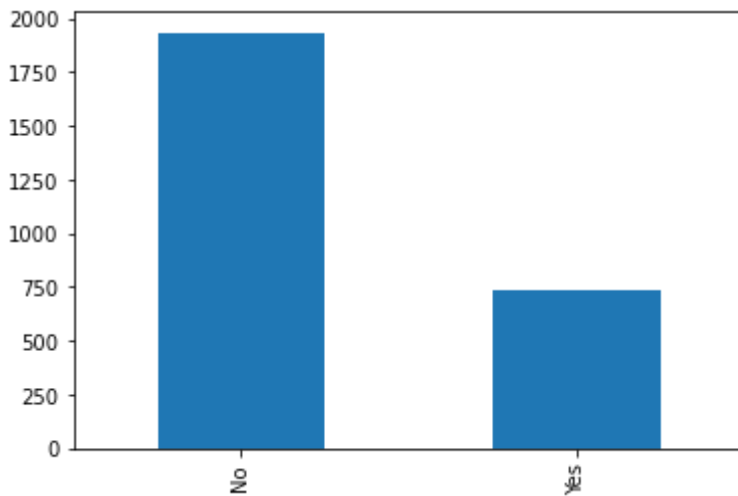
```
No      1933
Yes      733
Name: Voice_mail_plan, dtype: int64
```

In [12]:

```
df['Voice_mail_plan'].value_counts().plot(kind='bar')
```

Out[12]:

<AxesSubplot:>



In [13]:

```
c0_50 = df[(df.Total_day_calls < 50 )]
c51_75 = df[(df.Total_day_calls > 50 ) & (df.Total_day_calls <= 75)]
c76_100 = df[(df.Total_day_calls > 75 ) & (df.Total_day_calls <= 100)]
c101_125 = df[(df.Total_day_calls > 100 ) & (df.Total_day_calls <= 125)]
c126_150 = df[(df.Total_day_calls > 125 ) & (df.Total_day_calls <= 150)]
c151_175 = df[(df.Total_day_calls > 150 )]
```

In [14]:

```
print(c0_50.shape)
print(c51_75.shape)
print(c76_100.shape)
print(c101_125.shape)
print(c126_150.shape)
print(c151_175.shape)
```

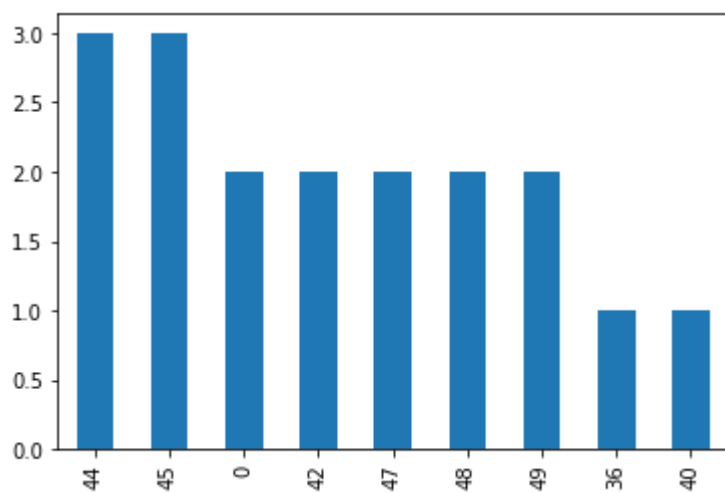
```
(18, 20)
(260, 20)
(1048, 20)
(1076, 20)
(252, 20)
(12, 20)
```

In [15]:

```
c0_50['Total_day_calls'].value_counts().plot(kind='bar')
```

Out[15]:

<AxesSubplot:>

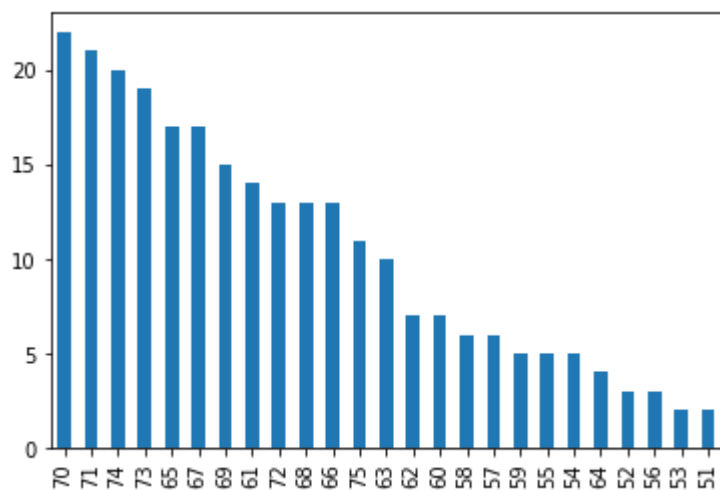


In [16]:

```
c51_75['Total_day_calls'].value_counts().plot(kind='bar')
```

Out[16]:

<AxesSubplot:>

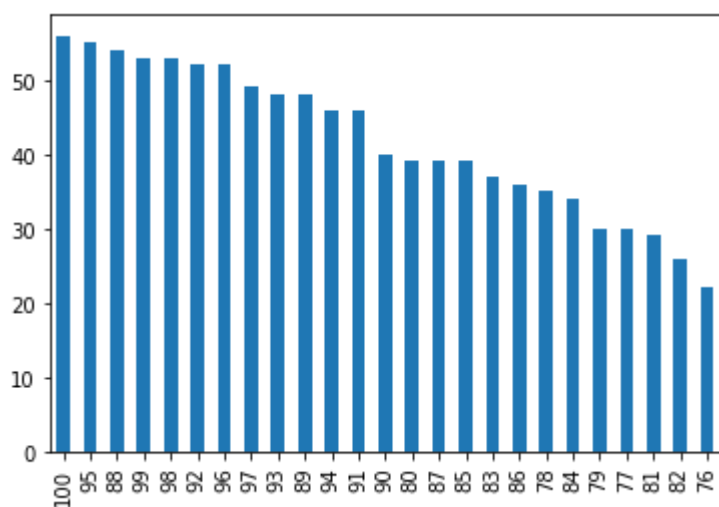


In [17]:

```
c76_100['Total_day_calls'].value_counts().plot(kind='bar')
```

Out[17]:

<AxesSubplot:>

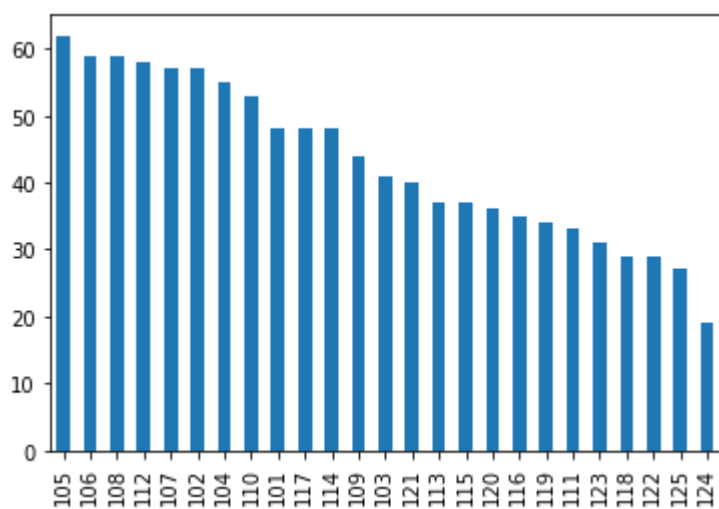


In [18]:

```
c101_125['Total_day_calls'].value_counts().plot(kind='bar')
```

Out[18]:

<AxesSubplot:>

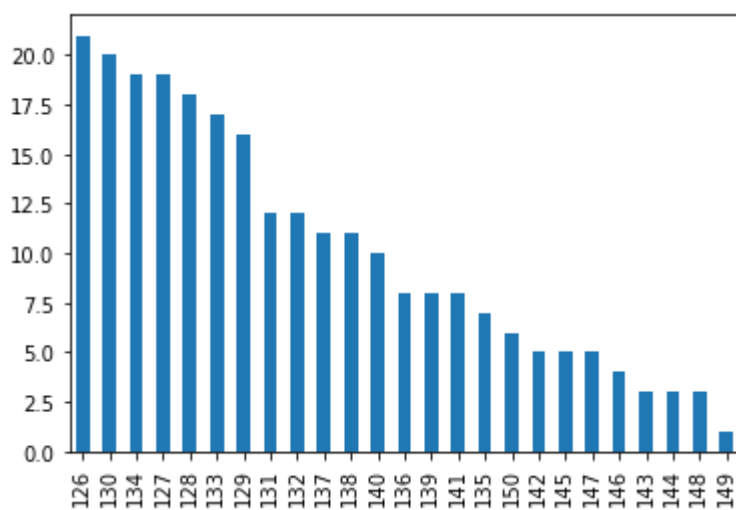


In [19]:

```
c126_150['Total_day_calls'].value_counts().plot(kind='bar')
```

Out[19]:

<AxesSubplot:>

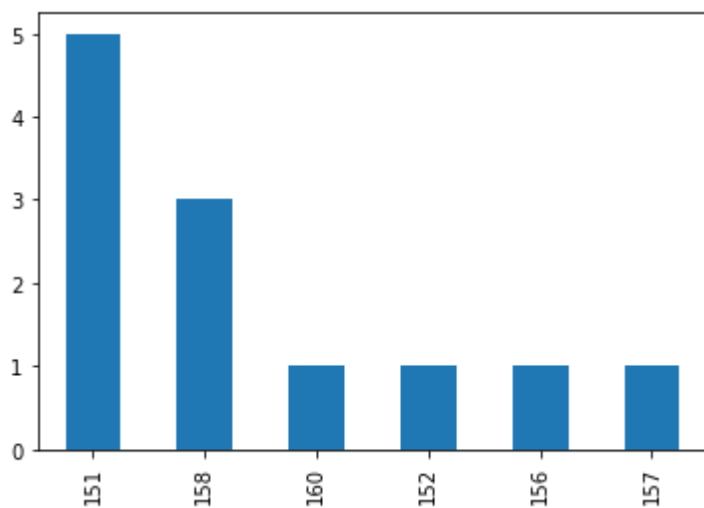


In [20]:

```
c151_175['Total_day_calls'].value_counts().plot(kind='bar')
```

Out[20]:

<AxesSubplot:>



In [21]:

```
c0_5 = df[(df.Total_day_charge <= 5 )]
c6_10 = df[(df.Total_day_charge > 5 ) & (df.Total_day_charge <= 10)]
c11_15 = df[(df.Total_day_charge > 10 ) & (df.Total_day_charge <= 15)]
c16_20 = df[(df.Total_day_charge > 15 ) & (df.Total_day_charge <= 20)]
c21_25 = df[(df.Total_day_charge > 20 ) & (df.Total_day_charge <= 25)]
c26_30 = df[(df.Total_day_charge > 25 ) & (df.Total_day_charge <= 30)]
c31_35 = df[(df.Total_day_charge > 30 ) & (df.Total_day_charge <= 35)]
c36_40 = df[(df.Total_day_charge > 35 ) & (df.Total_day_charge <= 40)]
c41_45 = df[(df.Total_day_charge > 40 ) & (df.Total_day_charge <= 45)]
c46_50 = df[(df.Total_day_charge > 45 ) & (df.Total_day_charge <= 50)]
c51_55 = df[(df.Total_day_charge > 50 ) & (df.Total_day_charge <= 55)]
c56_60 = df[(df.Total_day_charge > 55 ) & (df.Total_day_charge <= 60)]
```

In [22]:

```
print(c0_5.shape)
print(c6_10.shape)
print(c11_15.shape)
print(c16_20.shape)
print(c21_25.shape)
print(c26_30.shape)
print(c31_35.shape)
print(c36_40.shape)
print(c41_45.shape)
print(c46_50.shape)
print(c51_55.shape)
print(c56_60.shape)
```

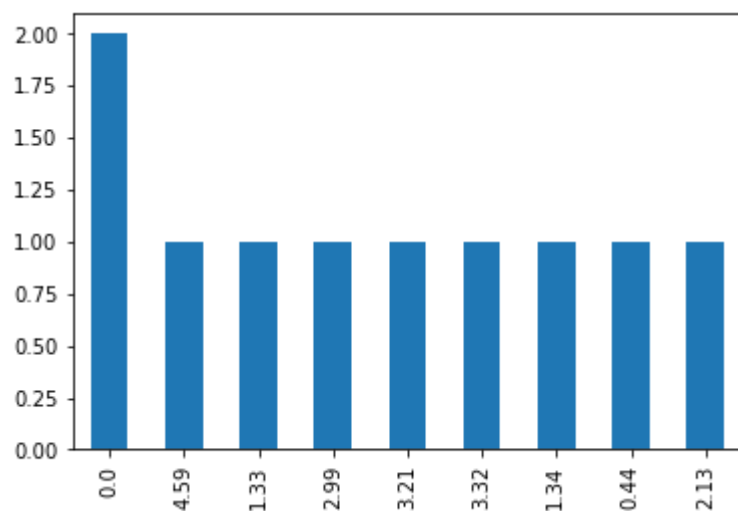
```
(10, 20)
(32, 20)
(80, 20)
(216, 20)
(400, 20)
(533, 20)
(551, 20)
(443, 20)
(243, 20)
(114, 20)
(36, 20)
(8, 20)
```

In [23]:

```
c0_5['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[23]:

<AxesSubplot:>

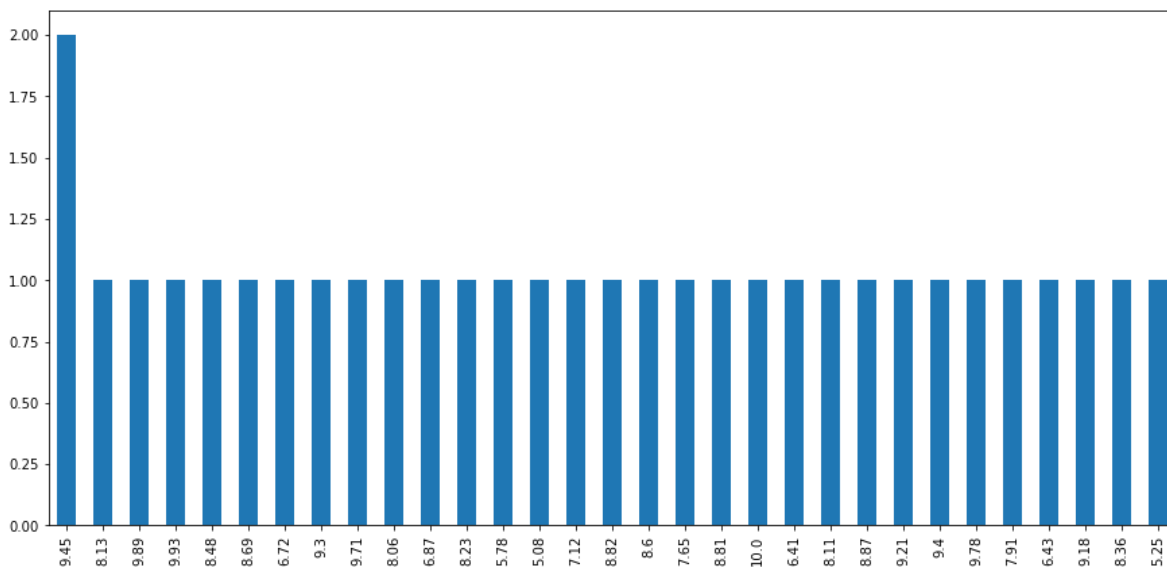


In [24]:

```
plt.figure(figsize=(15,7))  
c6_10['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[24]:

<AxesSubplot:>

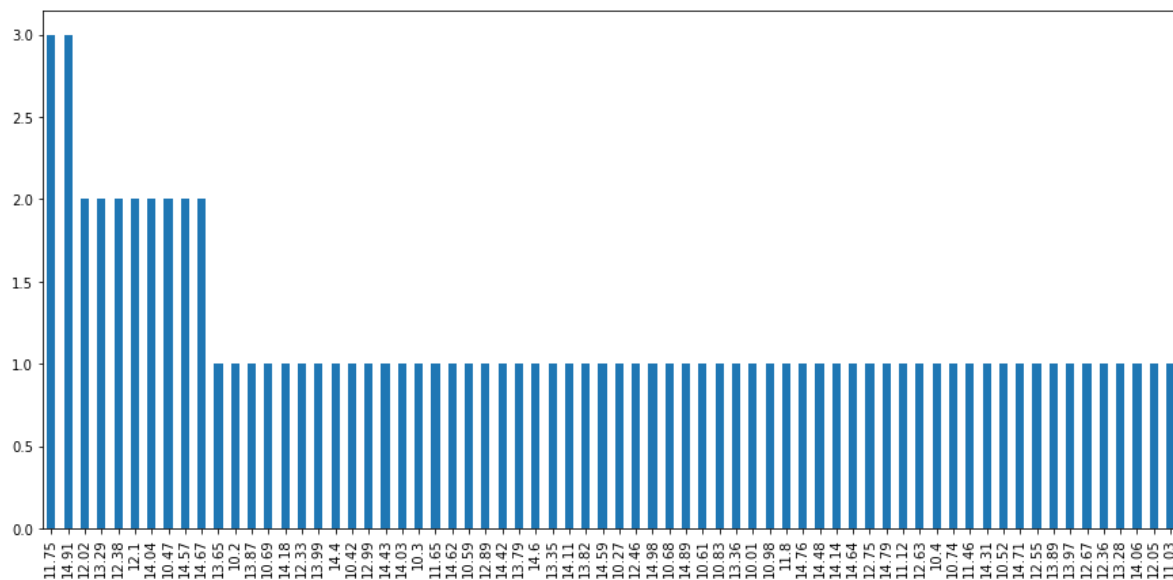


In [25]:

```
plt.figure(figsize=(15,7))
c11_15['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[25]:

<AxesSubplot:>

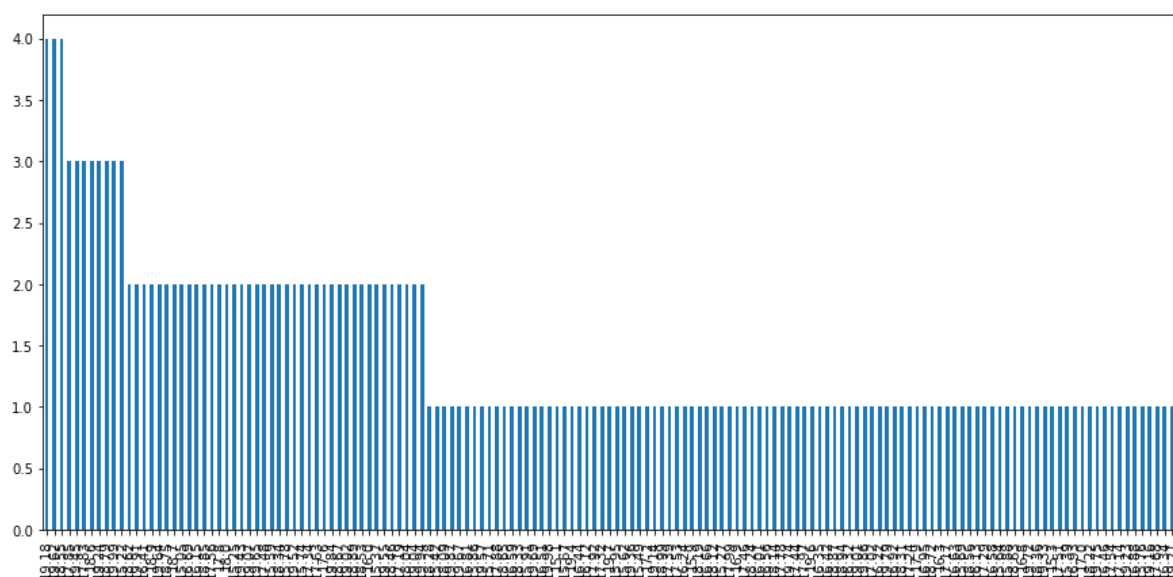


In [26]:

```
plt.figure(figsize=(15,7))
c16_20['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[26]:

<AxesSubplot:>

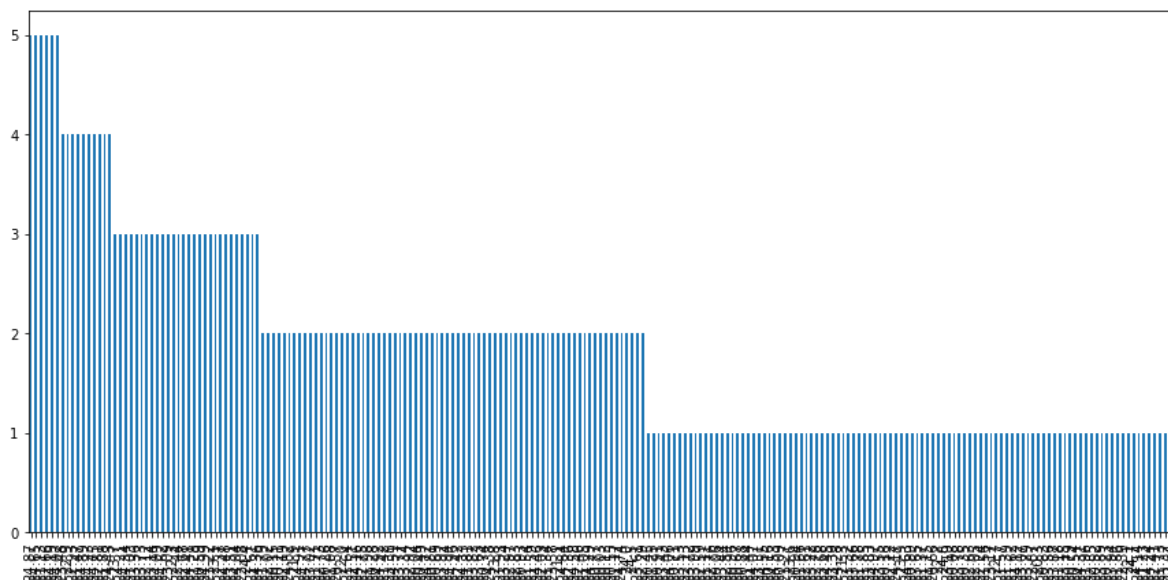


In [27]:

```
plt.figure(figsize=(15,7))
c21_25['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[27]:

<AxesSubplot:>

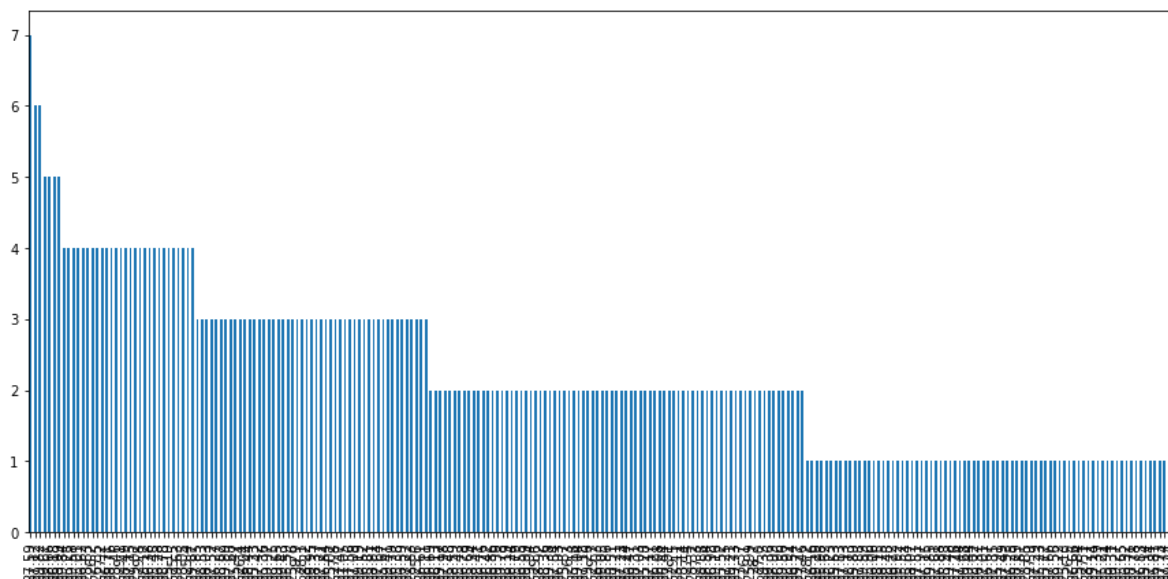


In [28]:

```
plt.figure(figsize=(15,7))
c26_30['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[28]:

<AxesSubplot:>

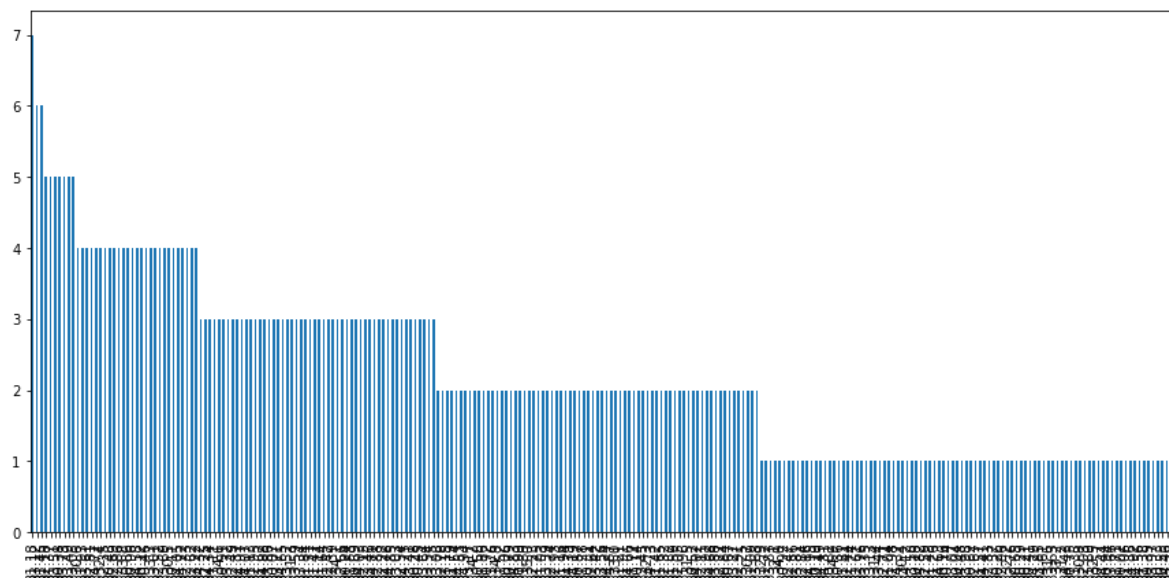


In [29]:

```
plt.figure(figsize=(15,7))  
c31_35['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[29]:

<AxesSubplot:>

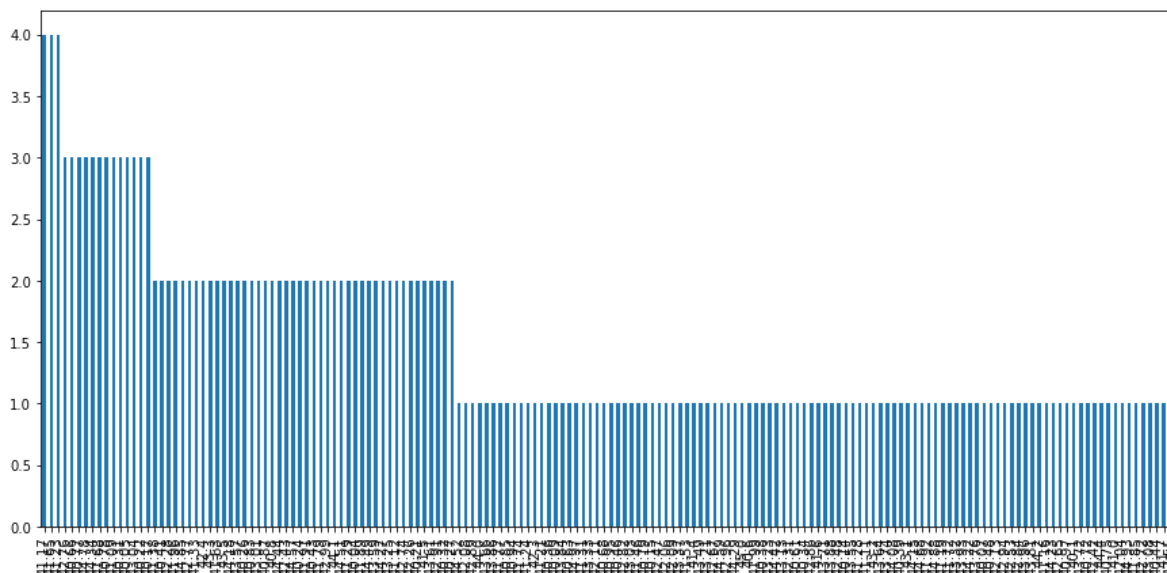


In [31]:

```
plt.figure(figsize=(15,7))
c41_45['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[31]:

<AxesSubplot:>

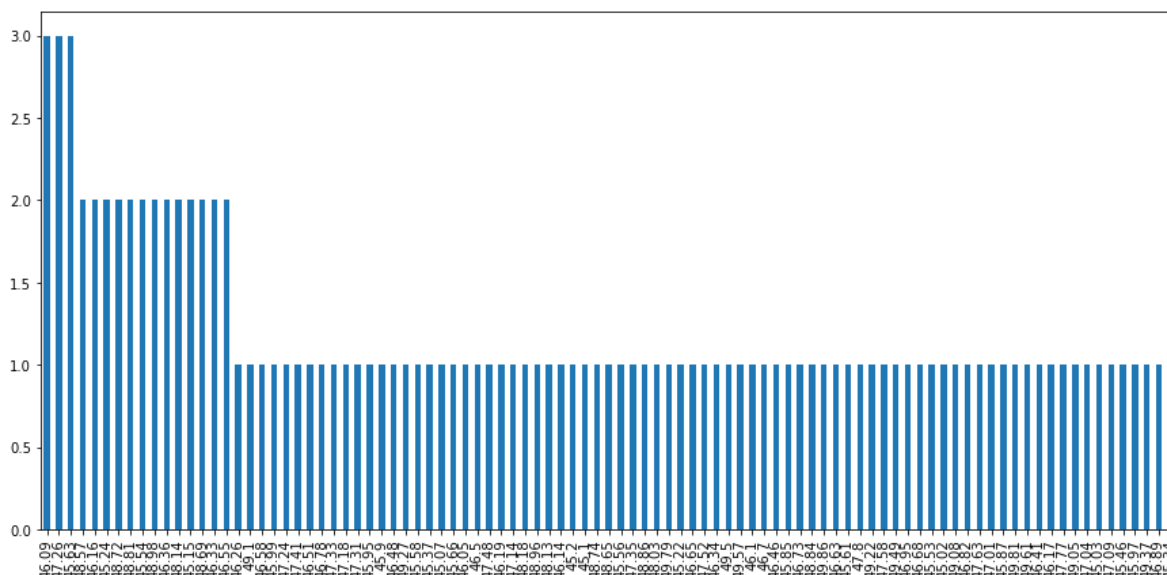


In [32]:

```
plt.figure(figsize=(15,7))
c46_50['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[32]:

<AxesSubplot:>

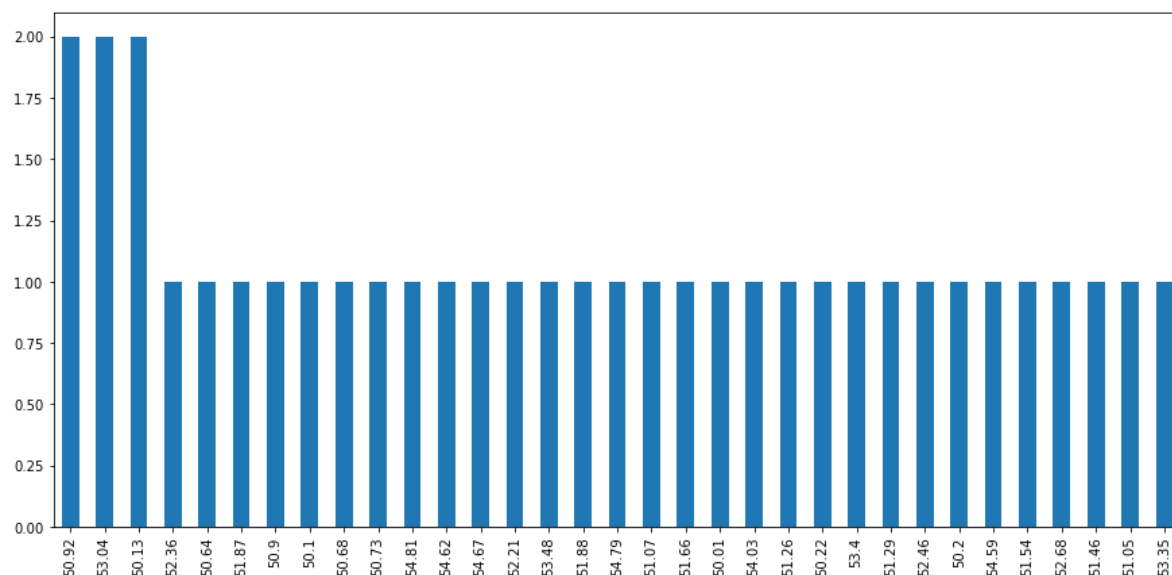


In [33]:

```
plt.figure(figsize=(15,7))  
c51_55['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[33]:

<AxesSubplot:>

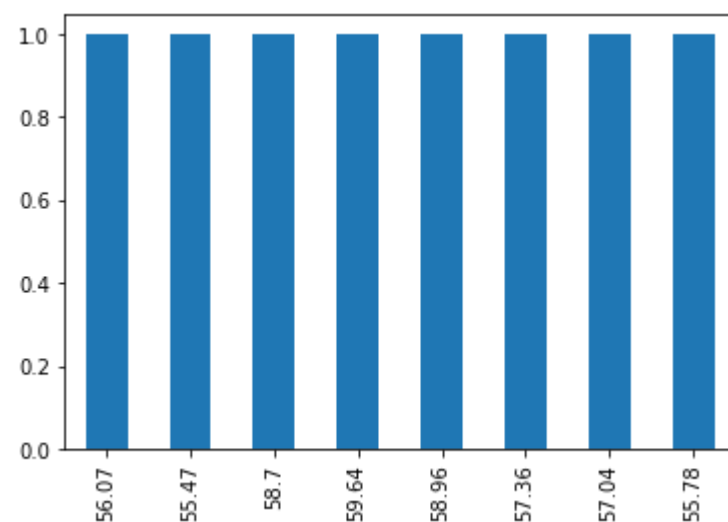


In [34]:

```
c56_60['Total_day_charge'].value_counts().plot(kind='bar')
```

Out[34]:

<AxesSubplot:>

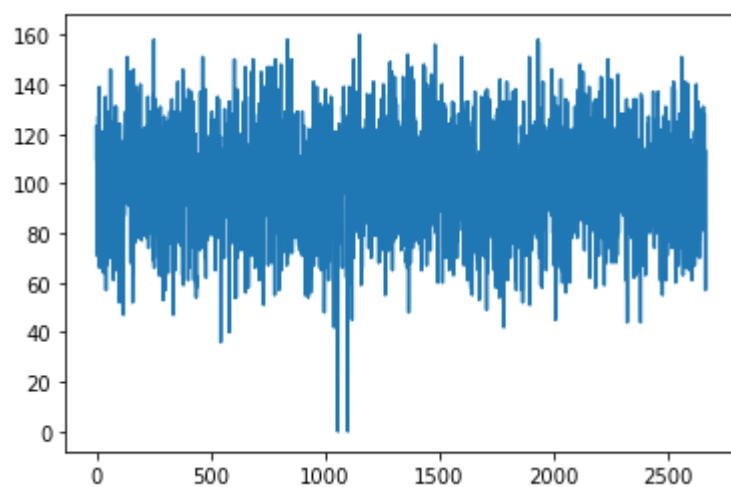


In [35]:

```
plt.plot(df['Total_day_calls'])
```

Out[35]:

[<matplotlib.lines.Line2D at 0xe787aa8>]

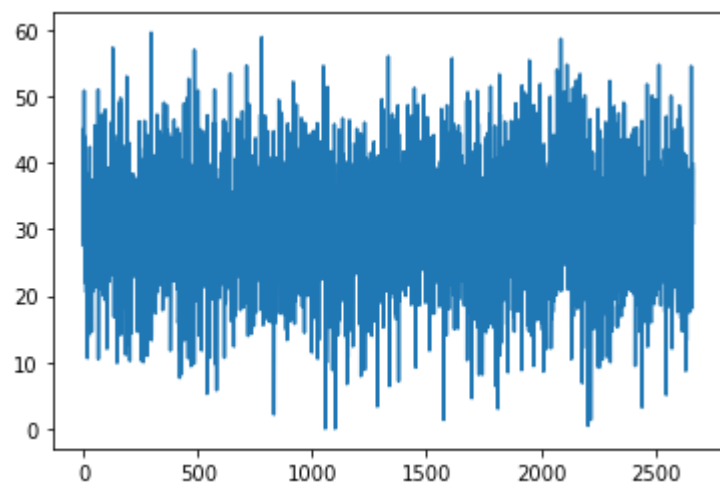


In [36]:

```
plt.plot(df['Total_day_charge'])
```

Out[36]:

[<matplotlib.lines.Line2D at 0xea16778>]



In [37]:

```
a = df['Churn'].value_counts()
a
```

Out[37]:

```
False    2278
True      388
Name: Churn, dtype: int64
```

In [38]:

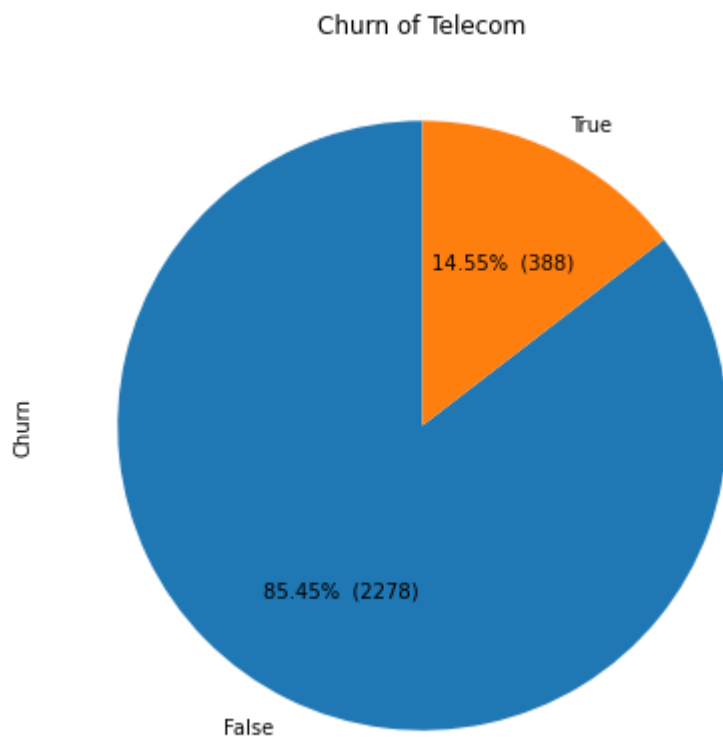
```
def make_autopct(a):
    def my_autopct(pct):
        total = sum(a)
        val = int(round(pct*total/100.0))
        return '{p:.2f}% ({v:d})'.format(p=pct,v=val)
    return my_autopct
```

In [39]:

```
plt.figure(figsize=(15,7))
plt.title('Churn of Telecom')
df['Churn'].value_counts().plot(kind='pie',autopct=make_autopct(a),startangle = 90)
```

Out[39]:

<AxesSubplot:title={'center':'Churn of Telecom'}, ylabel='Churn'>

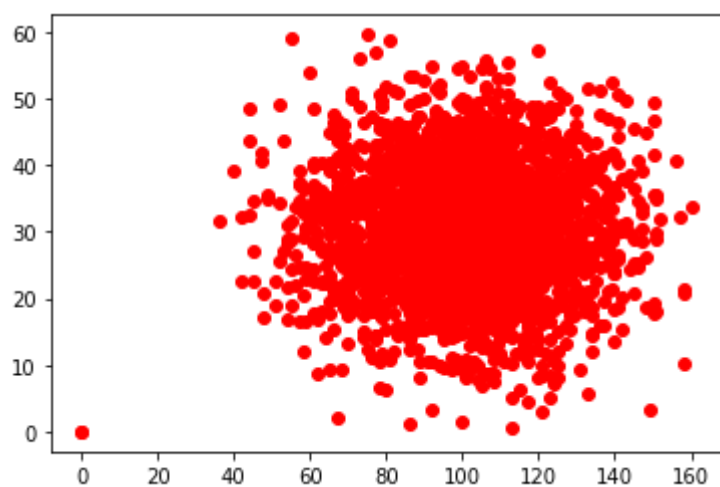


In [40]:

```
plt.scatter(x='Total_day_calls',y='Total_day_charge',data=df,color='red')
```

Out[40]:

<matplotlib.collections.PathCollection at 0xeb4c928>

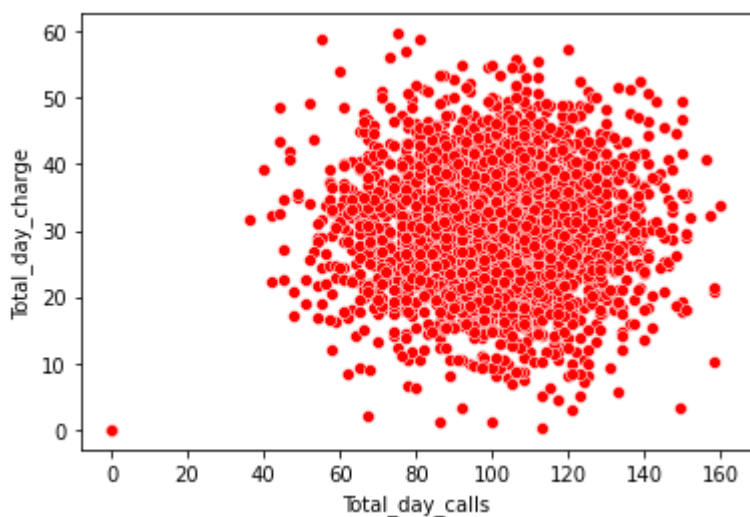


In [41]:

```
sb.scatterplot(x='Total_day_calls',y='Total_day_charge',data=df,color='red')
```

Out[41]:

<AxesSubplot:xlabel='Total_day_calls', ylabel='Total_day_charge'>

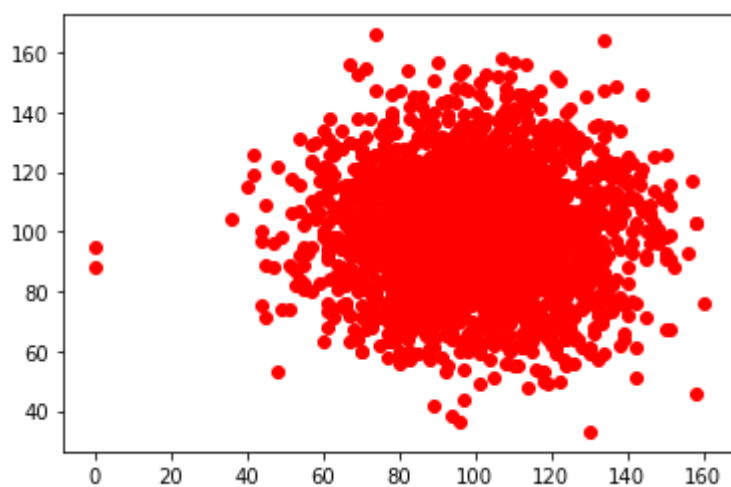


In [42]:

```
plt.scatter(x='Total_day_calls',y='Total_night_calls',data=df,color='red')
```

Out[42]:

<matplotlib.collections.PathCollection at 0xfeed4ca0>

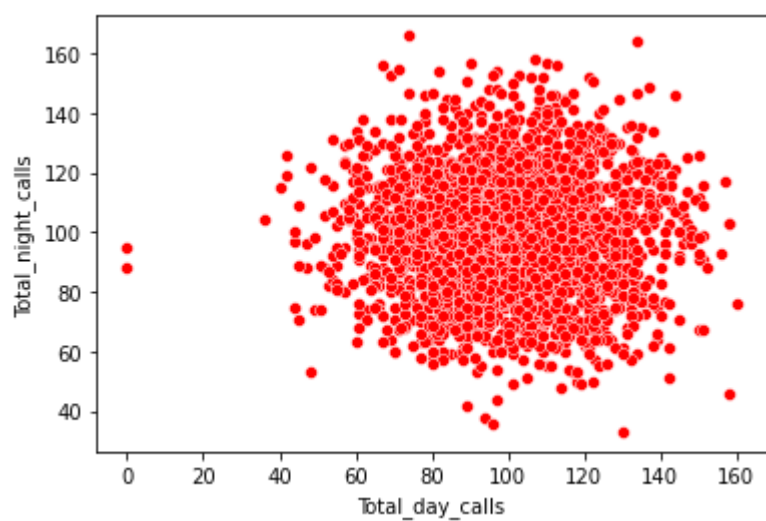


In [43]:

```
sb.scatterplot(x='Total_day_calls',y='Total_night_calls',data=df,color='red')
```

Out[43]:

<AxesSubplot:xlabel='Total_day_calls', ylabel='Total_night_calls'>

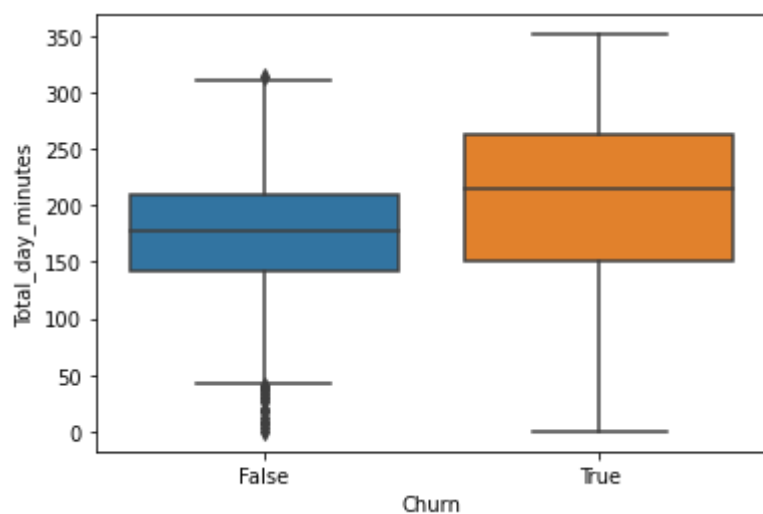


In [44]:

```
sb.boxplot(x = 'Churn',y = 'Total_day_minutes',data = df)
```

Out[44]:

<AxesSubplot:xlabel='Churn', ylabel='Total_day_minutes'>

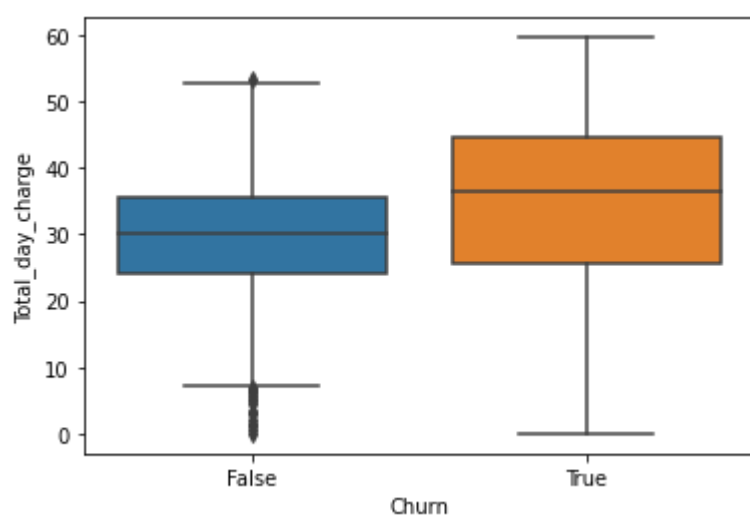


In [45]:

```
sb.boxplot(x = 'Churn',y = 'Total_day_charge',data = df)
```

Out[45]:

<AxesSubplot:xlabel='Churn', ylabel='Total_day_charge'>



In [46]:

```
LE = LabelEncoder()
```

In [47]:

```
df['Churn'] = LE.fit_transform(df['Churn'])
```

In [48]:

```
df.head()
```

Out[48]:

| | State | Account_length | Area_code | International_plan | Voice_mail_plan | Number_vmail_messages |
|---|-------|----------------|-----------|--------------------|-----------------|-----------------------|
| 0 | KS | 128 | 415 | No | Yes | |
| 1 | OH | 107 | 415 | No | Yes | |
| 2 | NJ | 137 | 415 | No | No | |
| 3 | OH | 84 | 408 | Yes | No | |
| 4 | OK | 75 | 415 | Yes | No | |

In [49]:

```
df['International_plan'] = LE.fit_transform(df['International_plan'])
```

In [50]:

```
df.head()
```

Out[50]:

| | State | Account_length | Area_code | International_plan | Voice_mail_plan | Number_vmail_messages |
|---|-------|----------------|-----------|--------------------|-----------------|-----------------------|
| 0 | KS | 128 | 415 | 0 | Yes | |
| 1 | OH | 107 | 415 | 0 | Yes | |
| 2 | NJ | 137 | 415 | 0 | No | |
| 3 | OH | 84 | 408 | 1 | No | |
| 4 | OK | 75 | 415 | 1 | No | |

In [51]:

```
df['Voice_mail_plan'] = LE.fit_transform(df['Voice_mail_plan'])
```

In [52]:

```
df.head()
```

Out[52]:

| | State | Account_length | Area_code | International_plan | Voice_mail_plan | Number_vmail_messag |
|---|-------|----------------|-----------|--------------------|-----------------|---------------------|
| 0 | KS | 128 | 415 | 0 | 1 | |
| 1 | OH | 107 | 415 | 0 | 1 | |
| 2 | NJ | 137 | 415 | 0 | 0 | |
| 3 | OH | 84 | 408 | 1 | 0 | |
| 4 | OK | 75 | 415 | 1 | 0 | |

In [53]:

```
df.corr()
```

Out[53]:

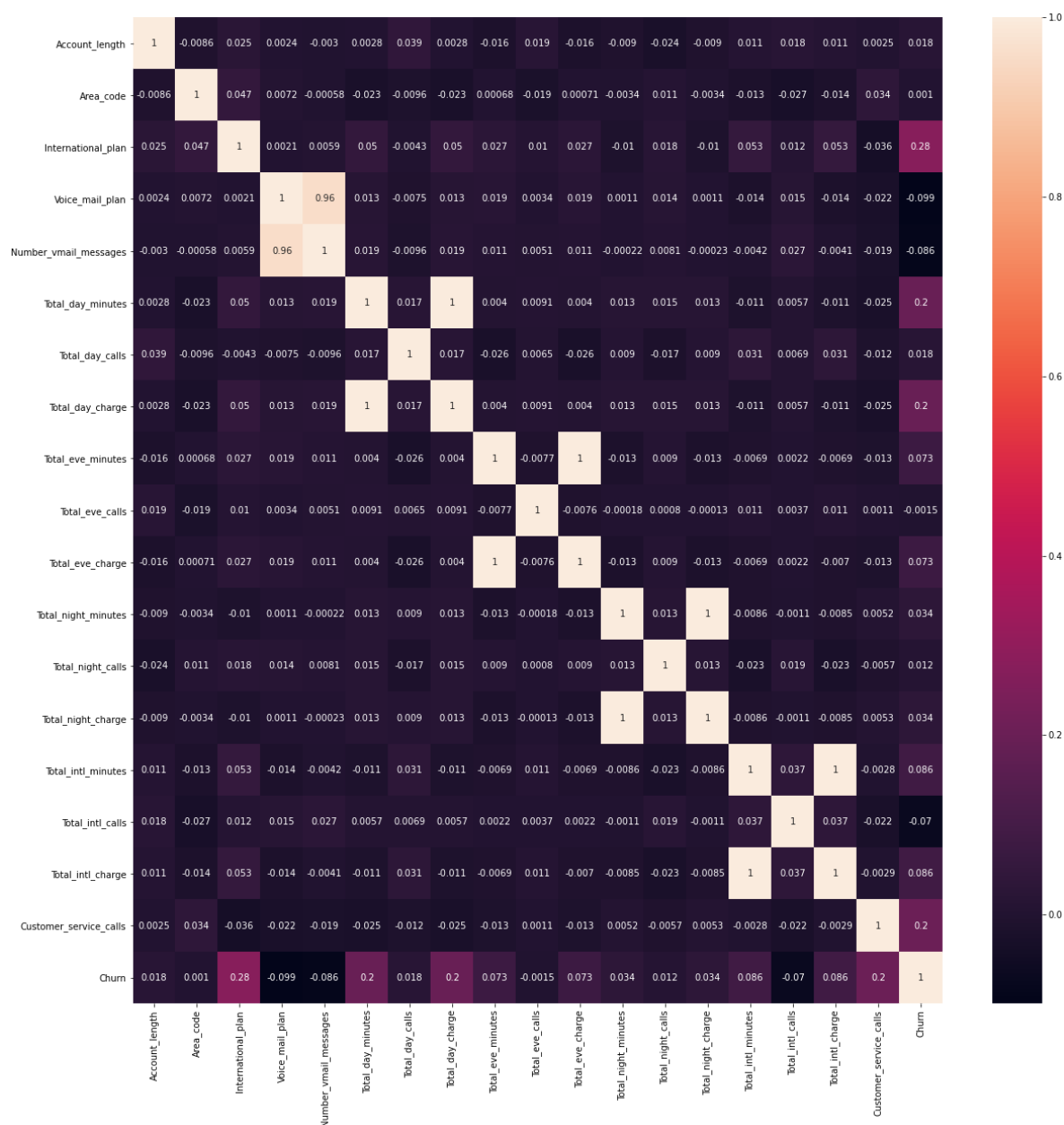
| | Account_length | Area_code | International_plan | Voice_mail_plan | Numb |
|------------------------|----------------|-----------|--------------------|-----------------|------|
| Account_length | 1.000000 | -0.008620 | 0.024500 | 0.002448 | |
| Area_code | -0.008620 | 1.000000 | 0.047099 | 0.007180 | |
| International_plan | 0.024500 | 0.047099 | 1.000000 | 0.002131 | |
| Voice_mail_plan | 0.002448 | 0.007180 | 0.002131 | 1.000000 | |
| Number_vmail_messages | -0.002996 | -0.000584 | 0.005858 | 0.957159 | |
| Total_day_minutes | 0.002847 | -0.023134 | 0.049550 | 0.013438 | |
| Total_day_calls | 0.038862 | -0.009629 | -0.004277 | -0.007541 | |
| Total_day_charge | 0.002843 | -0.023130 | 0.049555 | 0.013439 | |
| Total_eve_minutes | -0.015923 | 0.000679 | 0.026616 | 0.019132 | |
| Total_eve_calls | 0.018552 | -0.018602 | 0.010277 | 0.003404 | |
| Total_eve_charge | -0.015909 | 0.000707 | 0.026623 | 0.019147 | |
| Total_night_minutes | -0.008994 | -0.003353 | -0.010310 | 0.001065 | |
| Total_night_calls | -0.024007 | 0.011455 | 0.018081 | 0.013985 | |
| Total_night_charge | -0.008999 | -0.003382 | -0.010316 | 0.001066 | |
| Total_intl_minutes | 0.011369 | -0.013418 | 0.053162 | -0.013963 | |
| Total_intl_calls | 0.017627 | -0.027423 | 0.011549 | 0.015196 | |
| Total_intl_charge | 0.011383 | -0.013534 | 0.053037 | -0.013931 | |
| Customer_service_calls | 0.002455 | 0.034442 | -0.035955 | -0.022054 | |
| Churn | 0.017728 | 0.001019 | 0.277489 | -0.099291 | |

In [54]:

```
plt.figure(figsize=(20,20))
sb.heatmap(df.corr(),annot=True)
```

Out[54]:

<AxesSubplot:>



In [62]:

```
x = df[['International_plan', 'Total_day_minutes', 'Total_day_charge', 'Customer_service_calls']]
y = df['Churn']
```

In [64]:

```
x.head()
```

Out[64]:

| | International_plan | Total_day_minutes | Total_day_charge | Customer_service_calls | Total_eve_m |
|---|--------------------|-------------------|------------------|------------------------|-------------|
| 0 | 0 | 265.1 | 45.07 | 1 | |
| 1 | 0 | 161.6 | 27.47 | 1 | |
| 2 | 0 | 243.4 | 41.38 | 0 | |
| 3 | 1 | 299.4 | 50.90 | 2 | |
| 4 | 1 | 166.7 | 28.34 | 3 | |

In [66]:

```
y.head()
```

Out[66]:

```
0    0
1    0
2    0
3    0
4    0
Name: Churn, dtype: int32
```

In [68]:

```
x_train,x_test,y_train,y_test = tts(x,y,train_size = 0.7, random_state = 20)
```

In [69]:

```
print(x_train.shape)
print(y_train.shape)
print(x_test.shape)
print(y_test.shape)
```

```
(1866, 6)
(1866,)
(800, 6)
(800,)
```


In [84]:

```
LR = LogisticRegression(solver='liblinear')
LR.fit(x_train,y_train)
y_pred = LR.predict(x_test)
print("Train Score :",LR.score(x_train,y_train))
print("Test Score :",LR.score(x_test,y_test))
print("Prediction :", y_pred)
```

Train Score : 0.8590568060021436

Test Score : 0.85

[illegible][illegible]

Multiple model in one code

In [98]:

```
cross_val_score(LogisticRegression(solver='liblinear', multi_class='ovr'), x_train, y_train,
```

Out[98]:

```
array([0.86012862, 0.86012862, 0.85209003])
```

In [99]:

```
cross_val_score(LogisticRegression(solver='liblinear',multi_class='ovr'), x_test, y_test,cv
```

Out[99]:

```
array([0.84269663, 0.87640449, 0.86090226])
```

In [100]:

```
cross_val_score(SVC(gamma='auto'), x_train, y_train,cv=3)
```

Out[100]:

```
array([0.85369775, 0.85369775, 0.85369775])
```

In [101]:

```
cross_val_score(SVC(gamma='auto'), x_test, y_test,cv=3)
```

Out[101]:

```
array([0.8576779 , 0.85393258, 0.85338346])
```

In [102]:

```
cross_val_score(RandomForestClassifier(n_estimators=40),x_train, y_train,cv=3)
```

Out[102]:

```
array([0.89228296, 0.90032154, 0.90353698])
```

In [103]:

```
cross_val_score(RandomForestClassifier(n_estimators=40),x_test, y_test,cv=3)
```

Out[103]:

```
array([0.87265918, 0.8988764 , 0.89097744])
```

In [107]:

```
def get_score(model, x_train, x_test, y_train, y_test):  
    model.fit(x_train, y_train)  
    return model.score(x_test, y_test)  
  
print('Logistic Regration :',get_score(LogisticRegression(solver='liblinear',multi_class='o  
    x_train, x_test, y_train, y_test))  
  
print('SVM :',get_score(SVC(gamma='auto'),x_train, x_test, y_train, y_test))  
  
print("RandomForest :",get_score(RandomForestClassifier(n_estimators=45),x_train, x_test, y
```

Logistic Regration : 0.85

SVM : 0.85875

RandomForest : 0.89125

In [87]:

```
Confusion_matrix = confusion_matrix(y_test,y_pred)
Confusion_matrix
```

Out[87]:

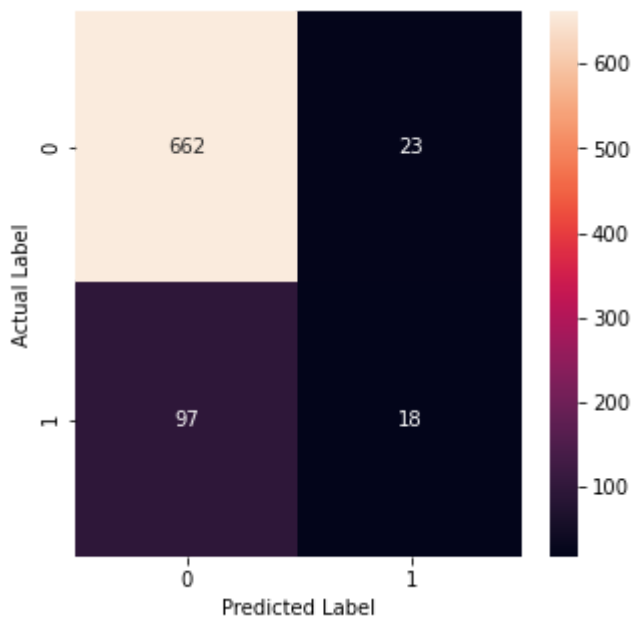
```
array([[662,  23],
       [ 97,  18]], dtype=int64)
```

In [89]:

```
plt.figure(figsize=(5,5))
sb.heatmap(Confusion_matrix,annot=True,fmt='g')
plt.xlabel('Predicted Label')
plt.ylabel('Actual Label')
```

Out[89]:

```
Text(24.0, 0.5, 'Actual Label')
```



Perform Hyper parameter tuning

In [111]:

```
def get_score(model, x_train, x_test, y_train, y_test):  
    model.fit(x_train, y_train)  
    return model.score(x_test, y_test)  
  
print('Logistic Regration :',get_score(LogisticRegression(solver='liblinear',multi_class='o  
    x_train, x_test, y_train, y_test))  
  
print('SVM :',get_score(SVC(gamma='auto'),x_train, x_test, y_train, y_test))  
  
print("RandomForest :",get_score(RandomForestClassifier(n_estimators=45),x_train, x_test, y
```

Logistic Regration : 0.85

SVM : 0.85875

RandomForest : 0.88875

In []: