Orange is an American global telecommunication company. The firm has been providing terrible customer service. They continue to fall short despite repeated promises to improve. Only last month (October 2016) the authority fined them a \$2.3 million, after receiving over 1000 consumer complaints.

The existing database will serve as a repository of public customer complaints filed against Orange.

It will help to pin down what is wrong with Orange's customer service.

Data Dictionary

- 1. Ticket #: Ticket number assigned to each complaint
- 2. Customer Complaint: Description of complaint
- 3. Date: Date of complaint
- 4. Time: Time of complaint
- 5. Received Via: Mode of communication of the complaint
- 6. City: Customer city
- 7. State: Customer state
- 8. Zipcode: Customer zip
- 9. Status: Status of complaint
- 10. Filing on behalf of someone

▼ Analysis Task

Import data into environment.

Provide the trend chart for the number of complaints at monthly and daily granularity levels.

Create a new categorical variable with value as Open and Closed. Open & Pending is to be categorized as Open and Closed & Solved is to be categorized as Closed.

Provide state wise status of complaints in a stacked bar chart. Use the categorized variable from Q3. Provide insights on:

Which state has the maximum complaints

Which state has the highest percentage of unresolved complaints

Provide a table with the frequency of complaint types. Which complaint types are maximum i.e., around internet, network issues, or across any other domains.

Provide the percentage of complaints resolved till date, which were received through the Internet and customer care calls.

The analysis results to be provided with insights wherever applicable.

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

#importing dataset
df = pd.read_csv("Orange_telecom_complaints_data.csv")

df.head() Zip Status Behalf of 22-Comcast Cable 3:53:50 Customer 0 250635 04-22-Apr-15 Abingdon Maryland 21009 Closed No PMInternet Speeds Care Call 15 #checking data types of all df.dtypes Ticket # Customer Complaint Date Date month year object Time object Received Via object object State object Zip code int64 Status Filing on Behalf of Someone object dtype: object #create time based indexing , so can perform time series analysis df["date_index"] = df["Date_month_year"] + " " + df["Time"] #convert necessary field to datatime before it was in object form that is our of no use df["date_index"] = pd.to_datetime(df["date_index"]) df["Date month year"] = pd.to datetime(df["Date month year"]) #checking if data dypes changed or not df.dtypes Ticket # object Customer Complaint object Date object Date_month_year datetime64[ns] Time object Received Via object object

#setting index to date_index
df = df.set_index(df["date_index"])

Filing on Behalf of Someone

object

object
datetime64[ns]

State Zip code

Status

date_index
dtype: object

```
Ticket Customer Date Date_month_year Time Via Via
                            Comcast
                              Cable
                                                          3:53:50 Customer
      2015-04-22
                                               2015-04-22
                250635
                                      04-
                                                                            Abingdon Maryland 21009
      15:53:50
                                                              PM Care Call
                             Internet
                                      15
                             Speeds
                                      18-
      2015-04-18
                          Speed and
                242732
                                      04-
                                                2015-04-18
                                                                    Internet Acworth Georgia 30101
                            Service
      09:55:47
                                      15
df["Date_month_year"].value_counts()[:6]
     2015-06-24
    2015-06-23
     2015-06-25
     2015-06-26
    2015-06-30
    2015-06-29
    Name: Date_month_year, dtype: int64
#checking complaints - date wise / daily basis
from matplotlib.pyplot import figure
figure(figsize=(14,4), dpi=80)
#fd = df["Date_month_year"].value_counts()
#print(fd.head())
#fd.plot();
# or
fd = df.groupby(pd.Grouper(freq="D")).size()
print(fd.head())
fd.plot();
```

```
200 -
      150 -
#checking on monthly basis
fm = df.groupby(pd.Grouper(freq="M")).size()
                                                       N I N
#plotting monthly wise
fm.plot();
      1000
      800
      600
      400
      200
             Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec
                             date_index
#checking status
df.Status.unique()
     array(['Closed', 'Open', 'Solved', 'Pending'], dtype=object)
## Create a new categorical variable with value as Open and Closed.
## Open & Pending is to be categorized as Open and Closed & Solved is to be categorized as Closed.
df["new_status"] = ["Open" if Status =="Open" or Status== "Pending" else "Closed" for Status in df["Status"]]
df.head(2)
```

Provide state wise status of complaints in a stacked bar chart.
Use the categorized variable from above data. Provide insights on:

1. Which state has the maximum complaints
2. Which state has the highest percentage of unresolved complaints

2. Which state has the highest percentage of unresolved complaints

df.groupby(["State"]).size().sort_values(ascending=False).to_frame().reset_index().rename({0: "Count"}, axis=1)[: 5]

f.groupby(["State"]).size().sort_values(ascending=False)[:5]

	State	Count
0	Georgia	288
1	Florida	240
2	California	220
4	Tennessee	143

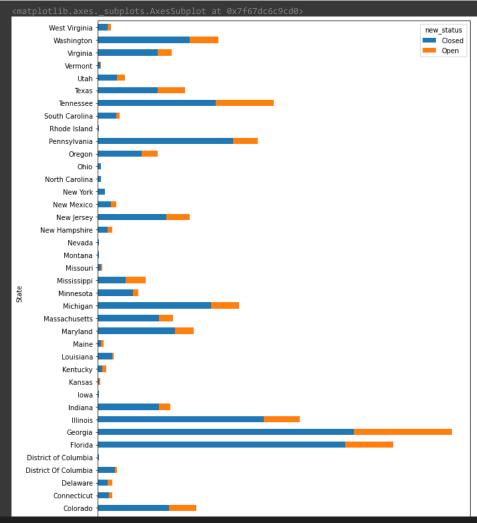
finding States wise - status of complaints
Status_complaint = df.groupby(["State","new_status"]).size().unstack().fillna(0)
Status_complaint



State

State		
Alabama	17.0	9.0
Arizona	14.0	6.0
Arkansas	6.0	0.0
California		
Colorado	58.0	22.0
Connecticut	9.0	3.0
Delaware	8.0	4.0
District Of Columbia		
District of Columbia	1.0	0.0
Florida		
Georgia	208.0	80.0
Illinois	135.0	29.0
Indiana	50.0	9.0
lowa	1.0	0.0
Kansas	1.0	1.0
Kentucky	4.0	3.0
Louisiana	12.0	1.0
Maine	3.0	2.0
Maryland	63.0	15.0
Massachusetts	50.0	11.0
Michigan	92.0	23.0
Minnesota	29.0	4.0
Mississippi	23.0	16.0
Missouri	3.0	1.0
Montana	1.0	0.0
Nevada	1.0	0.0
New Hampshire	8.0	4.0
Now Jorgov	56 N	10 N

#plotting state wise status of open and closed complaints
Status_complaint.plot(kind="barh",figsize=(10,15), stacked =True)



#state wise
df.groupby(["State"]).size().sort_values(ascending=False).to_frame().reset_index().rename({0:"count"},axis=1)[:5]

	State	count	
0	Georgia	288	
1	Florida	240	
2	California	220	
3	Illinois	164	
4	Tennessee	143	

#maxx state wise complanits open and closed
df.groupby(["State","new_status"]).size().unstack().fillna(0).max()

new_status Closed 208.0 Open 80.0 dtype: float64

Which state has the highest percentage of unresolved complaints

```
# now we do the sorting of the data and show the data in State-percent wise
Status_complaint = df.groupby(["State","new_status"]).size().unstack().fillna(0) #fillna for filling null values
Status_complaint['% Unresolved Complaints']=np.divide(Status_complaint['Open'],np.add(Status_complaint['Open'],Status_complaint['Closed']))*100
Status_complaint = Status_complaint.sort_values(by='% Unresolved Complaints',ascending=False)
Status_complaint.head(10)
```

State			
Kansas	1.0	1.0	50.000000
Kentucky	4.0	3.0	42.857143
Mississippi	23.0	16.0	41.025641
Maine	3.0	2.0	40.000000
Alabama	17.0	9.0	34.615385
Vermont	2.0	1.0	33.333333
Delaware	8.0	4.0	33.333333
New Hampshire	8.0	4.0	33.333333
Tennessee	96.0	47.0	32.867133
Texas	49.0	22.0	30.985915

```
#importing gensim and its necessary libraries
import gensim
from gensim import corpora
from print import pprint
from nltk.corpus import stopwords
from nltk.stem.wordnet import WordNetLemmatizer
import string
import nltk
nltk.download('stopwords')

stop = set(stopwords.words('english'))
exclude = set(string.punctuation)
lemma = WordNetLemmatizer()

[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

```
#we are removing punctuation and stop words from the complaints with lemmatize
def clean(doc):
    stop_free = " ".join([i for i in doc.lower().split() if i not in stop])
    punc_free = "" .join(ch for ch in stop_free if ch not in exclude)
    normalized = " ".join(lemma.lemmatize(word) for word in punc free.split())
    return normalized
#converting "Customer Complaint" to unique words(tokenID) and splitting
import nltk
nltk.download('wordnet')
nltk.download('omw-1.4')
doc_complete = df["Customer Complaint"].tolist()
doc_clean = [clean(doc).split() for doc in doc_complete]
     [nltk data] Downloading package wordnet to /root/nltk data...
     [nltk_data] Package wordnet is already up-to-date!
     [nltk_data] Downloading package omw-1.4 to /root/nltk_data...
     [nltk data] Package omw-1.4 is already up-to-date!
#importing corpora liberary
from gensim import corpora
#we have found 1412 unique tokens
dictionary = corpora.Dictionary(doc_clean)
print(dictionary)
     Dictionary(1412 unique tokens: ['cable', 'comcast', 'internet', 'speed', 'disappear']...)
#showing frequency of bag of words
doc term matrix = [dictionary.doc2bow(doc) for doc in doc clean]
doc term matrix[:5]
      [(4, 1), (5, 1), (6, 1), (7, 1), (8, 1)],
      [(3, 1), (8, 1)],
      [(1, 1), (8, 1), (16, 1), (17, 1)]]
#importing LdaModel and executing it
from gensim.models import LdaModel
Num Topic = 9
ldamodel = LdaModel(doc_term_matrix, num_topics= Num_Topic, id2word = dictionary, passes = 30)
topics = ldamodel.show_topics() #0.149*"comcast" + 0.145*"billing" + 0.136*"service"
for topic in topics:
 print(topic)
 print()
     (0, '0.169*"internet" + 0.147*"comcast" + 0.111*"service" + 0.057*"issue" + 0.034*"billing" + 0.024*"customer" + 0.010*"business" + 0.010*"year" + 0.009*"terrible" + 0.008*"modem"')
     (1, '0.239*"comcast" + 0.126*"data" + 0.102*"cap" + 0.045*"charge" + 0.034*"complaint" + 0.024*"usage" + 0.017*"xfinity" + 0.015*"fee" + 0.011*"fraudulent" + 0.011*"overage"')
     (2, '0.045*"comcast" + 0.042*"help" + 0.038*"access" + 0.031*"false" + 0.024*"lack" + 0.023*"rate" + 0.023*"please" + 0.020*"hbo" + 0.020*"go" + 0.017*"xfinity"')
```

```
(3, '0.131*"billing" + 0.110*"comcast" + 0.065*"service" + 0.052*"practice" + 0.037*"bill" + 0.031*"unfair" + 0.025*"cable" + 0.024*"complaint" + 0.019*"problem" + 0.018*"comcastxfin
     (4, '0.081*"comcast" + 0.035*"monopoly" + 0.031*"show" + 0.027*"day" + 0.026*"refund" + 0.024*"appointment" + 0.021*"call" + 0.021*"said" + 0.019*"throttled" + 0.017*"overcharge"')
     (5, '0.206*"service" + 0.056*"poor" + 0.043*"xfinity" + 0.042*"customer" + 0.028*"connection" + 0.025*"credit" + 0.021*"paying" + 0.021*"phone" + 0.019*"failure" + 0.018*"unreliable"
     (6, '0.062*"comcast" + 0.029*"pay" + 0.028*"bandwidth" + 0.025*"without" + 0.024*"email" + 0.020*"12" + 0.019*"one" + 0.017*"month" + 0.016*"scam" + 0.016*"throttle"')
     (7, '0.106*"comcast" + 0.069*"internet" + 0.056*"speed" + 0.055*"throttling" + 0.041*"pricing" + 0.039*"service" + 0.021*"promised" + 0.015*"unfair" + 0.014*"without" + 0.013*"low"')
     (8, '0.141*"internet" + 0.124*"speed" + 0.066*"comcast" + 0.054*"slow" + 0.029*"service" + 0.021*"outage" + 0.020*"switch" + 0.019*"high" + 0.018*"charged" + 0.013*"bait"')
word_dict = {}
for i in range(Num Topic):
    words = ldamodel.show_topic(i, topn =10)
    word_dict["topic # " + "{}".format(i)] = [i[0] for i in words]
#showing the main results
pd.DataFrame(word_dict)
                                             billing
           internet
                    comcast
                              comcast
                                                       comcast
                                                                   service
                                                                            comcast
                                                                                       comcast
```

speed comcast service cap access service show xfinity bandwidth bill billing complaint lack refund connection email pricing service business xfinity please cable paying one promised switch terrible fraudulent problem throttled failure without charged scam

This types of complaints came from the Telecoms consumers

