REPORT PROBLEM #5

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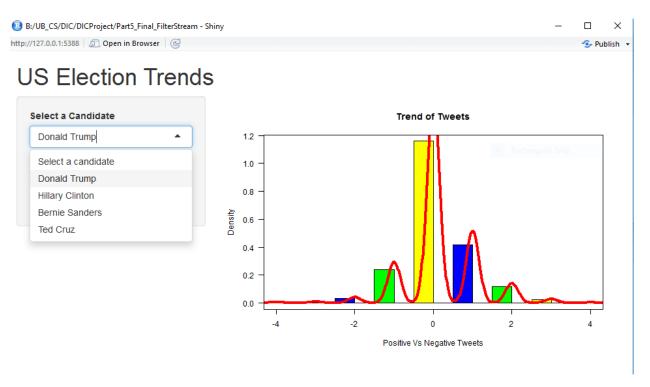
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ANALYSIS OF TWITTER DATA FOR THE WEEK

In this part of the project we have made a UI that based upon the election candidates in USA displays the trends of their popularity among people. For this we had to do sentiment analysis of the tweets and their nature if they were positive or negative.

The following is an image of the website developed:



To analyze this trend I used a 'Sentiment_new.R' that, I downloaded from 'https://github.com/mjhea0/twitter-sentiment-analysis'. Now this file basically compares the tweets against a list of positive and negative words. Based on that, this separates the tweets into positive and negative tweets. The following R script is used to figure out the trend of tweets. This also forms the basis of our server side coding.

```
a = readLines('DonaldTrump.json', warn = FALSE)
    x = jsonlite::fromJSON(a)
    pos = scan('positive-words.txt', what = 'character', comment.char = ';')
    neg = scan('negative-words.txt', what = 'character', comment.char = ';')
    #This file sentiment new.R along with the list of positive and negative words has been taken from url: https://github.com/mjhea0/twitter-sentiment-
    source('sentiment new.R')
    y = x$text
    cleaned = iconv(y, 'UTF-8', 'ASCII')
    analysis = score.sentiment(cleaned, pos.words = pos, neg.words = neg)
    hist(analysis$score, xlab = 'Positive Vs Negative Tweets', las = 1, main = 'Trend of Tweets', col = c('blue', 'yellow', 'green'), probability = T)
    lines(density(analysis$score), col = 2, lwd = 4)
    box()
  }
if(candidateName == 'Donald Trump' & radioSelect == 'Now'){
    load('my oauth.Rdata')
    #serching TWitter
    filterStream(file.name = "Donald.json", # Save tweets in a json file
              track = c("Donald Trump", "president", "election"), # Collect tweets mentioning either Affordable Care Act, ACA, or Obamacare
              language = "en",
              timeout = 10, # Keep connection alive for 60 seconds
              oauth = my oauth) # Use my oauth file as the OAuth credentials
    tweets df <- parseTweets('Donald.json', simplify = FALSE)
    pos <- scan('positive-words.txt' , what = 'character', comment.char = ';')
    neg <- scan('negative-words.txt' ,what = 'character', comment.char = ';')</pre>
    y = tweets df$text
    cleaned = iconv(y,'UTF-8','ASCII')
    analysis = score.sentiment(cleaned, pos.words = pos, neq.words = neq)
    #This file sentiment new.R along with the list of positive and negative words has been taken from url: https://github.com/mjhea0/twitter-sentiment
    source('sentiment new.R')
    hist(analysis$score, xlab = 'Positive Vs Negative Tweets', las = 1, main = 'Trend of Tweets', col = c('blue', 'yellow', 'green'), probability = T)
```

if(candidateName == 'Donald Trump' & radioSelect == 'Weekly Trend'){

lines(density(analysis\$score), col = 2, lwd = 4)

box()

The above is the server side code. For weekly analysis we already have tweets that will be used to identify who is getting the most negative feedback. But for current trend 'streamR' package of twitter will be used to pull in the latest data from twitter Stream API. This has been done above by the use of 'filterStream()' API.

This filterStream() API is used to fetch the current tweets from twitter. After parsing them into JSON format the text field can be extracted from them and hence, they can be used for analyzing if they are positive or negative in nature.

From there a graph is plotted that will show the count and the probability of tweets as positive or negative. Also, the above code is repeated for all the other four candidates.

EXPLAINATION OF THE CURVE:

The curve is a histogram of probability. It has in X-axis, two divisions based on the position of 0. If the bar is on the right side of 0 that means that the bar represents a set of positive tweets with the score of tweets representing the value at which the bar itself stands. The more the score of tweets the more positive they are.

Similarly, the lesser the score of tweets from zero the more negative they are. If the tweet bar stands at 0, that means that people related to those tweets are either indifferent or neutral. Hence, they will not be counted towards winning or losing trend.

Also, there is a line that is the probability curve. The more the curve bulges towards the positive side the better it is for the candidate. Hence, after all considerations we can say that the trend will be calculated as

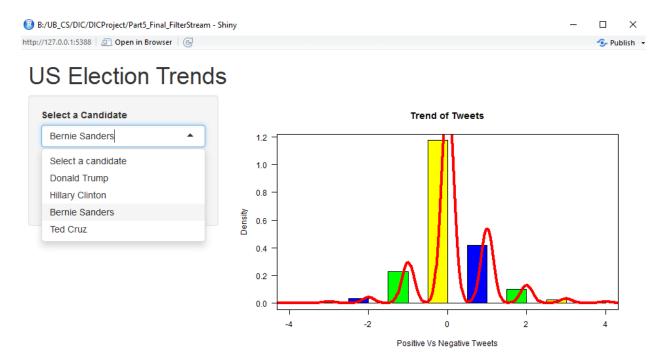
TREND = Number of positive tweets – Number of negative tweets (1)

The more the value of trend the better it is for candidates. Now let us analyze the real weekly data obtained for our four election candidates.

CLOSE ANALYSIS OF WEEKLY TRENDS

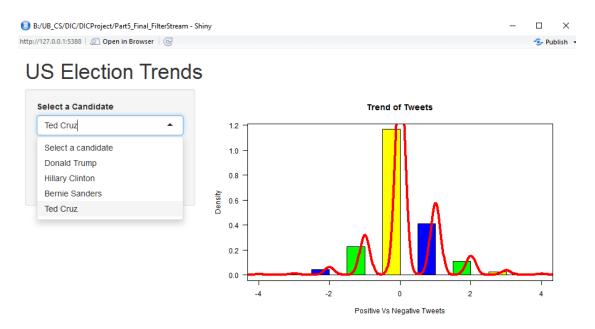
Weekly Trends Plots:

• Bernie Sanders:



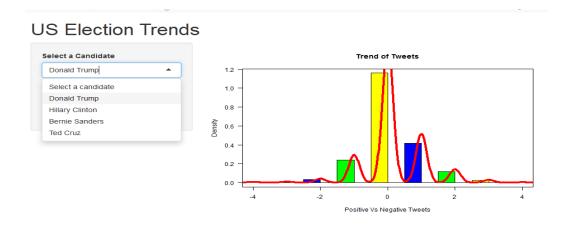
The graph has a lot of people who are confused about his candidacy, thus a huge graph at zero represents their tweets. Now overall, the blue graph at right side of 0 seems taller than the negative graph on left side of 0. Hence, for Bernie Sanders, the trend seems positive at least for the last week.

• Ted Cruz:



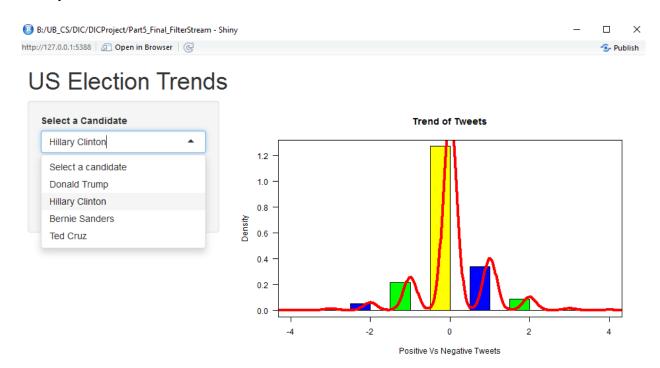
For Ted Cruz, as well the trend for the last week represents a positive vote. Although according to (1) the larger the value of trends the better it is for the candidate. Hence, between Bernie Sanders and Ted Cruz we can from there decide who has an upper hand.

Donald Trump



The same is the case for Donald Trump. However, the number of neutral tweets for Donald Trump remains lesser than the above two.

Hillary Clinton



For Hillary Clinton, as well the same situation persists. Thus, from the above graphs we have at least got an idea on how to analyze the trends based on the understanding of the graph.

SUMMARY

From the analysis of the week almost all the candidates for now have an equal trend for winning. The graph not only divides the tweets based upon strong and week sentiments but also gives a chance to know how many among the followers of a candidate are string and how many are weak supporters. Thus, it provides an extended analysis of public sentiments and also, provides us with a probability curve whose peak will mold in the direction of trend.