Russian interference in the U.S. 2016 presidential elections

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Background

In January 2017, major US agencies such as the Central Intelligence Agency(CIA), Federal Bureau of Investigations (FBI), and the National Security Agency(NSA) reported that there were evidences that the Russian government had conducted a campaign to influence the 2016 Presidential election. The campaign was aimed at damaging Hillary Clinton's presidential campaign, in the process, undermining the US democratic process ¹.

According to the report, it was found that the tweets were originating from the Russian "Trolls Factory" called the Internet Research Agency. In this assignment, we analyse the 3 million trolls tweets dataset obtained by Clemson University to explore the strategy used by the trolls².

Hypothesis

A strategy utilized by the trolls was to divide the democratic party supporters by expressing support for Bernie Sanders and deriding Hillary Clinton in an effort to drive low voter turnouts overall and encouraging them to vote against Clinton. At the same time, tweets were rallying the right wing votes.

We will analyze the tweets around a specific event, the Democratic National Convention (DNC), held between July 25, 2016 - July 28, 2016. The DNC is a series of presidential nomination conventions held by the Democratic Party. The main purpose of the convention is to confirm the nomination of the president and vice president candidates. In 2016, DNC had two main candidates competed for the presidential nomination - Hillary Clinton and Bernie Sanders. At the same time, in june and july 2016, hackers released emails by DNC members suggesting that the party leadership had undermined Bernie Sanders campaign and in turn favouring the Clinton campaign³.

Research Questions

- 1. What behaviour do the left leaning and right leaning trolls exhibit during the period of observation? Is it different from the general trend?
- 2. Comparing words and hashtags used by left and right leaning trolls around the same topic
- 3. Can we identify certain characteristics of a troll?

Data

The data used for this analysis were the 3 million tweets sent from Twitter handles connected to the Internet Research Agency, a Russian "troll factory" and a defendant in an indictment filed by the Justice Department in February 2018, as part of special counsel Robert Mueller's Russia investigation. The tweets in this database were sent between February 2012 and May 2018, with the vast majority posted from 2015 through 2017. The data was published by FiveThirtyEight in their story "Why We're Sharing 3 Million Russian Troll Tweets"²

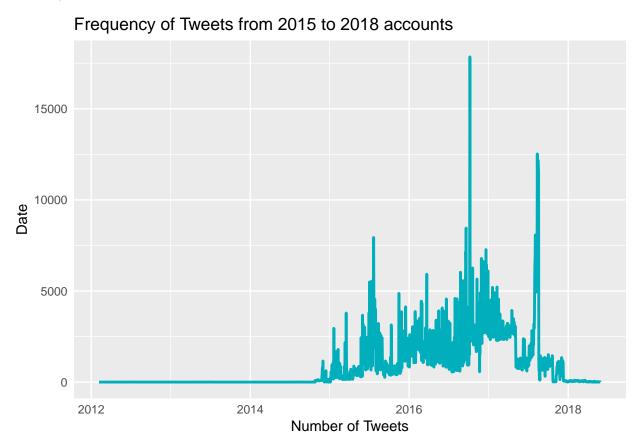
We first conducted general data analysis on the whole set of tweets to draw out general trends. We further analyzed tweets from June 1 2016 to August 31 2016 to validate our hypothesis and find other insights that

might arise out of the data. Due to the focus of our analysis being on how these tweets affect sentiment during the election, we have used a subset tweets that are in English

Findings

1 - Frequency of tweets

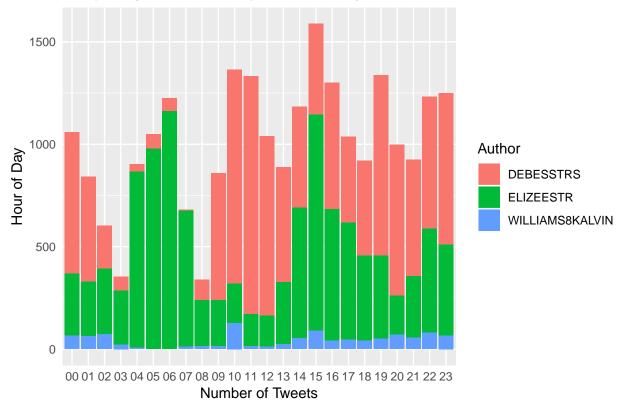
When the whole set is considered, we see that the frequency of tweets is higher between 2015 and 2018, with it peaking just before the end of 2016 around the elections. Interestingly, the tweets don't subside post elections, but continue till the end of 2017.



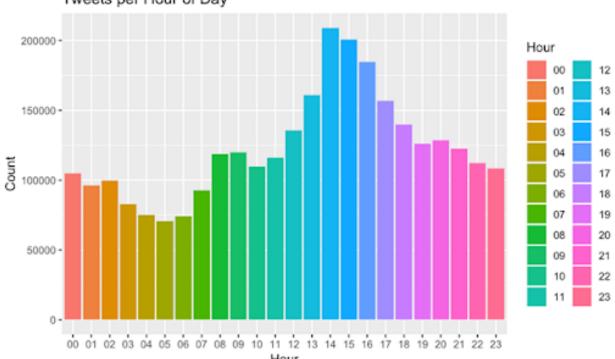
2 Timing of tweets

As the figures above show, there is more activity in the afternoon and early-evening hours. The exact cause of the heightened activity in the afternoon/evening would need further analysis; however, it is clear that these times are more active.

Frequency of Tweets as per time of day for three most active accounts

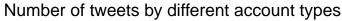


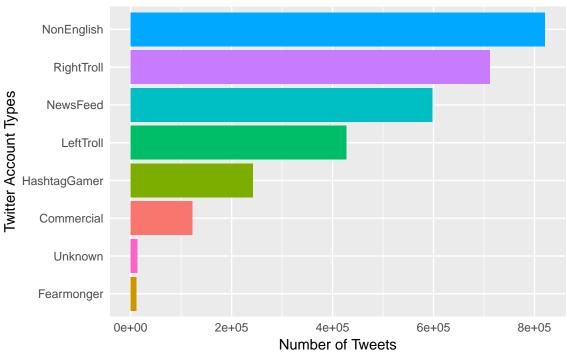
Tweets per Hour of Day



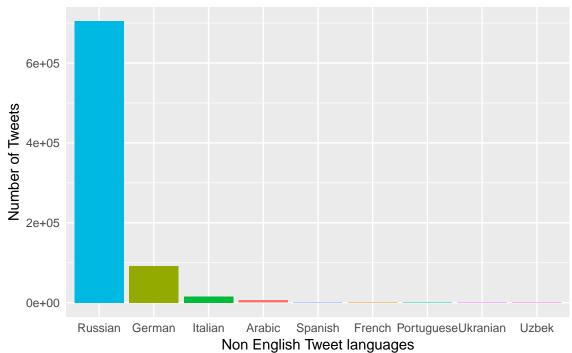
3 Tweets according to account types

The tweets were in 56 different languages, out of which 71.9% of tweets were in English and the tweets came from 37 different regions. The number of non-english tweets was substantially higher than english tweets, with more right leaning trolls than left leaning trolls.



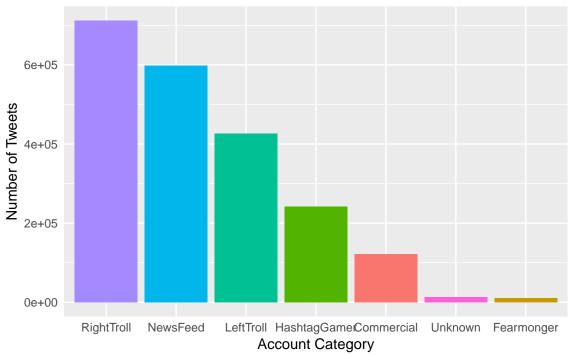


Number of tweets by non english Language



If we analyze the non-english tweets further, we see that russian tweets dominate followed by German, Italian, Arabic and Spanish. This might indicate to the fact that the IRA was not only looking to interfer in the US election, but also probably spreading some kind of propaganda in other countries as well





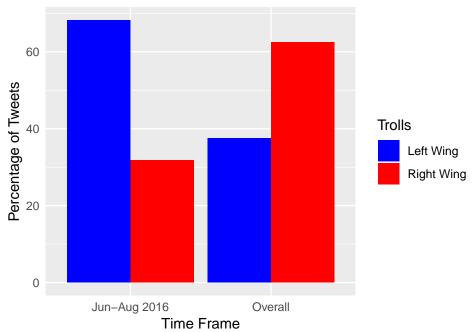
Most of the Tweets are Right Wing/ Left wing trolls, or are news feeds.

4. Analysis During DNC (tweets from the period 1st of June 2016 to 31st of August 2016):

4.1. Frequency of tweets:

Here, we will look at the number of tweets by accounts classified as right or left wing trolls around the period when DNC emails were leaked and compare it with the overall number.

Number of tweets for different time periods More left Troll accounts working in the DNC time period



An interesting insight is that though the overall data shows that the Right Troll tweets were about twice of Left Wing trolls, the left troll accounts were more active in the period of June- August 2016

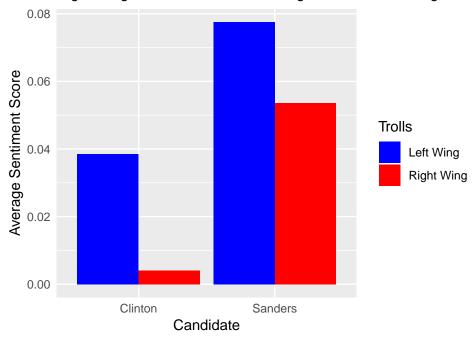
4.2. Sentiment Analysis of the tweets:

Here, we look at the sentiment of the tweets tweeted by the right and the left wing trolls suring the DNC period. We hypothesize that the Russian trolls were trying to divide the democratic party by cashing in on the email leaks and trying to display negative sentiment on Hillary Clinton, while trying to play up Bernie Sanders Supporters by encouraging them to vote against Hillary.

We use the R 'SentimentAnalysis' package to calculate sentiment of the Tweeters (Both Left and Right Trolls) for Bernie Sanders and Hillary Clinton during the DNC. We will be using SentimentGI calculator of the R's 'SentimentAnalysis', which is based on sentiment on the words of Harvard-IV Dictionary.

We subset the tweets where Keywords 'Bernie' and 'Sanders' (referencing Bernie Sanders) and 'Hillary' and 'Clinton' (referencing Hillary Clinton) were used. This is done seperately for Right and left wing Trolls. Each Tweet receives a score from -1 to +1 depending on the sentiment expressed in the tweet. (Negetive to Positive). The Average Sentiment Score is calculated by taking the average of the tweet sentient score for the particular group.

Sentiment in the period Jun 1 – Aug 31, 2016
Right Wing Sentiment on Sanders is greater than Left wing on Clin



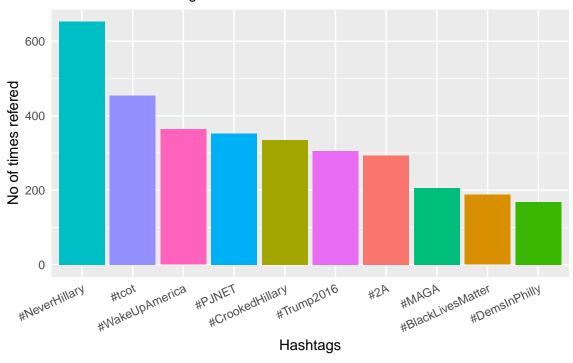
As expected, the average sentiment score of the accounts classified as left wing trolls for both Clinton and Sanders are higher than the average sentiment score by the right wing trolls. But the interesting fact to be noted is that the Sentiment for Clinton in tweets by both the trolls is quite less than the sentiment for Sanders. Shockingly, the Sentiment score by left trolls for hillary is so low, that the sentiment of Sanders among the Right is higher that it. From this, we can come to the conclusion that,

- The trolls delibrately posted anti Hillary Clinton tweets so that pro-Sanders people would not vote her.
- The accounts classfied as Left trolls probably posted pro left tweets in the beginning to get pro-left people to follow them and then posted anti Hillary tweets in an attempt to diminish her popularity and divide the left.
- The Trolls tried to play up Sanders so that people would not vote for Hillary in the election in response to the email leak

4.3. Popular Hashtags during the period

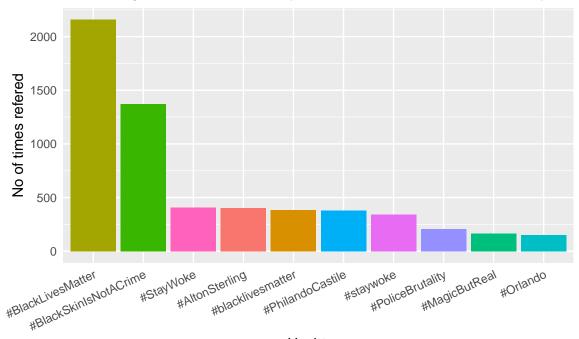
The Most Popular Hashtags referenced by the Trolls during that time period were

Popular hashtag among Right wing Trolls Jun 1 – Aug 31, 2016 Most of the the hastags are Anti–Clinton



Most of the Hashtags used by Right wing trolls were Anti Hillary Clinton, in an attempt to undermine her.

Popular hashtag among Left wing Trolls Jun 1 – Aug 31, 2016 The Hastags refer to Police Brutality towards African–American community

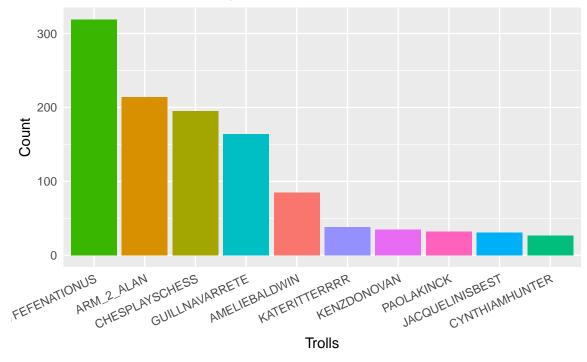


The Hashtags used by IRA Left trolls were regards to the Black Lives matter movement, possibly in an attempt to divide the country

5 Speed of Tweets by a user / Bot detection

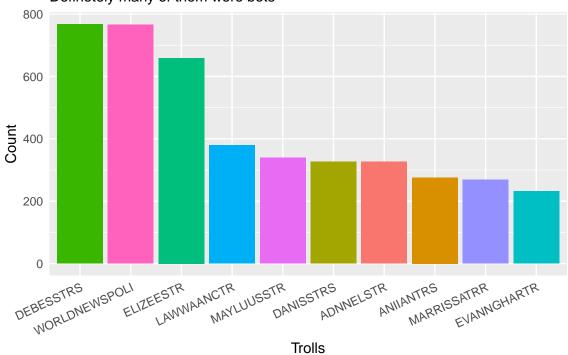
As the tweets had a time stamp to the minute, we conducted analysis on the hypothesis that some of the accounts might by bots. Lets look at the Accounts that retweeted more than 10 times in a minute.

Number of times trolls had more that 10 retweets in a min It can be infered that many of them were bots



Now lets look at the Accounts that tweeted more than 5 times in a minute (which may be impossible for a human).

Number of times trolls had more that 5 tweets in a min Definetely many of them were bots



Top 10 instances where accounts had high number of tweets in a minute

##		<pre>publish_date</pre>	author	count
##	1:	4/19/2017 10:56	WILLIAMS8KALVIN	116
##	2:	8/4/2017 13:36	ELIZEESTR	26
##	3:	8/16/2017 1:31	DEBESSTRS	25
##	4:	8/13/2017 1:11	MARRISSATRR	24
##	5:	8/12/2017 18:52	ALANISSTRS	23
##	6:	8/16/2017 14:28	ANIIANTRS	22
##	7:	8/13/2017 16:38	EISSYT56T	22
##	8:	10/2/2017 16:07	DANISSTRS	21
##	9:	8/3/2017 11:53	DEBESSTRS	21
##	10:	8/5/2017 23:53	DEBESSTRS	21

Time count was greater than 5 per min 12065

12065 times has more that 5 'Orignal Tweets' been posted per min. WILLIAMS8KALVIN has 116 tweets per min, which makes a case for some of the accounts being bots and not human accounts

Policy Recommendations & Future Research Questions

- The burst of activity on twitter around certain events like the DNC convention bring to light the strategy used on twitter by russian trolls more sharply. Looking at tweets around the event also help focus the analysis.
- Identifying bots based on certain characteristics of a tweet includes the text (like specific words used and patterns of capitalizing letters), how many hashtags it includes and how often the accounts tweet or retweet.
- Changing Twitter Policies to reduce the number of tweets possible in a minute/day can possibly reduce
 the use of bots.
- We could further look at how twitter activity affects real-life sentiments by also considering opinion polls conducted by agencies like Gallup

Contributions

- The final approach was decided collaboratively through group discussion and initial analysis of the data.
- Alexandra DeKinder and Ameya Karnad worked on data analysis using R
- Kulkanya Lekhyananda and Nitasha Nair worked on the presentation and policy report

References

- 1 Jonathan Masters, "Russia, Trump, and the 2016 U.S. Election" (Council on Foreign Relations, 2018), https://www.cfr.org/backgrounder/russia-trump-and-2016-us-election.
- 2 "Why We're Sharing 3 Million Russian Troll Tweets," FiveThiryEight, 2018, https://fivethirtyeight.com/features/why-were-sharing-3-million-russian-troll-tweets/.
- 3 The Washington Post, "Here Are the Latest, Most Damaging Things in the DNC's Leaked Emails," 2016, https://www.washingtonpost.com/news/the-fix/wp/2016/07/24/here-are-the-latest-most-damaging-things-in-the-dncs-leaked?noredirect=on&utm_term=.43968de0b481.