Topic 3: Sentiment Analysis I

Overview

Sentiment analysis is a tool for assessing the mood of a piece of text. For example, we can use sentiment analysis to understand public perceptions of topics in environmental policy like energy, climate, and conservation.

```
library(tidyr) #text analysis in R
library(lubridate) #working with date data
library(pdftools) #read in pdfs
library(tidyverse)
library(tidytext)
library(here)
library(LexisNexisTools) #Nexis Uni data wrangling
library(sentimentr)
library(readr)
```

Introduction to the Nexis Uni data source

```
#can we create a similar graph to Figure 3A from Froelich et al.?
mytext <- get_sentences(dat2$Headline)
sent <- sentiment(mytext)

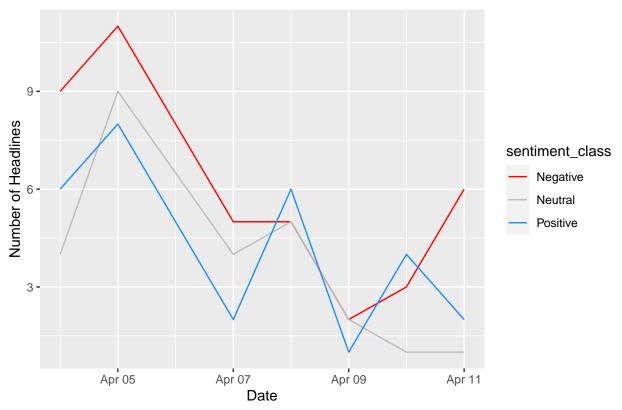
sent_df <- inner_join(dat2, sent, by = "element_id")
sentiment <- sentiment_by(sent_df$Headline)</pre>
```

```
sent_df %>%
 arrange(sentiment)
## # A tibble: 109 x 6
##
     element id Date
                           Headline
                                                  sentence_id word_count sentiment
                           <chr>
##
          <int> <date>
                                                        <int>
                                                                   <int>
                                                                             <dbl>
## 1
             66 2022-04-04 Scientists risk arres~
                                                                       7
                                                                            -0.756
                                                            1
## 2
             91 2022-04-07 The 'climate change' \sim
                                                            1
                                                                       9
                                                                           -0.75
## 3
            28 2022-04-09 The Dread 1.5 Degree ~
                                                                            -0.714
                                                            1
## 4
            43 2022-04-06 India's banks unprepa~
                                                                       7
                                                                           -0.510
                                                            1
## 5
             34 2022-04-08 Dangerous radicals ar~
                                                            1
                                                                            -0.449
                                                                       6
## 6
            14 2022-04-04 'Now or never' to avo~
                                                                           -0.442
                                                            1
                                                                      8
## 7
            78 2022-04-07 Statewide Gas Ban Bil~
                                                           1
                                                                      10
                                                                            -0.427
## 8
            50 2022-04-04 Guardian: Media 'Bare~
                                                                            -0.407
                                                            1
                                                                      8
             62 2022-04-06 Governor Youngkin's I~
## 9
                                                            1
                                                                      11
                                                                            -0.377
             7 2022-04-05 Narrow path to avoid ~
                                                            1
                                                                      8
                                                                            -0.354
## 10
## # ... with 99 more rows
sent_df_aggregate <- sent_df %>%
                     mutate(sentiment_class = case_when(sent_df$sentiment > 0 ~ 'Positive'
                                                        ,sent_df$sentiment == 0 ~ 'Neutral'
                                                     ,sent_df$sentiment < 0 ~ 'Negative')) %>%
                     group_by(Date, sentiment_class) %>%
```

0 Recreate Figure 1A from Froelich et al.

summarize(count_sentiment=n())

IPCC-related Articles Sentiment 4/4/22 to 4/11/22



1-3 Query on Nexis Uni

"Cloud seeding" was the search term used in on Nexis Uni database to grab the first 1000 articles.

4 Read in Nexis Uni data

5 Clean Nexis Uni data

```
#May be of use for assignment: using the full text from the articles
cloudseed_paragraphs_dat <- data_frame(element_id = cloudseed_paragraphs_df$Art_ID,</pre>
                                        Text = cloudseed_paragraphs_df$Paragraph)
cloudseed_dat3 <- inner_join(cloudseed_dat2,cloudseed_paragraphs_dat, by = "element_id") %%</pre>
                   janitor::clean names()
cloudseed_dat3 <- subset(cloudseed_dat3, text != " " )</pre>
cloudseed_dat3 <- cloudseed_dat3[!grepl("POSTED", cloudseed_dat3$text,ignore.case = TRUE),]</pre>
cloudseed_dat3 <- cloudseed_dat3[!grepl("GRAPHIC", cloudseed_dat3$text,ignore.case = TRUE),]</pre>
cloudseed_dat3 <- cloudseed_dat3[!grepl(":", cloudseed_dat3$text),]</pre>
cloudseed_dat3 <- cloudseed_dat3[!grepl("LINK TO", cloudseed_dat3$text,ignore.case = TRUE),]</pre>
cloudseed_dat3 <- cloudseed_dat3[grep1("[a-zA-Z]", cloudseed_dat3$text),]</pre>
bing_sent <- get_sentiments('bing') #qrab the bing sentiment lexicon from tidytext
\#head(binq\_sent, n = 20)
#unnest to word-level tokens, remove stop words, and join sentiment words
cloudseed_text_words <- cloudseed_dat3 %>%
  unnest_tokens(output = word, input = text, token = 'words')
cloudseed_sent_words <- cloudseed_text_words%>% #break text into individual words
  anti_join(stop_words, by = 'word') %>% #returns only the rows without stop words
  inner_join(bing_sent, by = 'word') #joins and retains only sentiment words
```

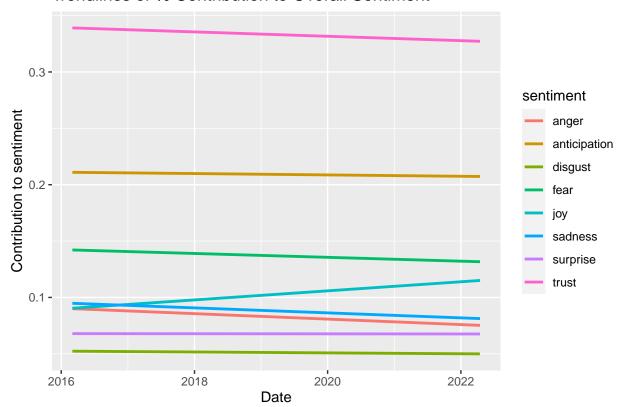
6 Explore dataset

7 Trends in Sentiment over time

```
cloudseed_sent_counts <- cloudseed_text_words %>%
        inner_join(nrc_sent) %>%
        group_by(date) %>%
        count(sentiment, sort = TRUE) %>%
        mutate(sentwords_per_day = sum(n)) %>%
        mutate(pct_contribution = (n/sentwords_per_day))

cloudseed_sent_counts %>%
    group_by(date) %>%
    group_by(date) %>%
    ggplot(aes(date, pct_contribution, group=sentiment, color=sentiment)) +
    geom_smooth( method="lm", se=F) +
    labs(x = "Date",
        y = "Contribution to sentiment",
        title = "Trendlines of % Contribution to Overall Sentiment")
```

Trendlines of % Contribution to Overall Sentiment



The concept of cloud seeding has always been very controversial. In recent years, it seems that the anticipation for the potential of the technology has stagnated. There seems to be a lot of trust (the dominating sentiment through all the years) in the technology but it is slowly waning as well. The decrease in sadness and increase in joy shows potential shifts in attitudes towards cloud seeding. Perhaps, we need to go further back in time to explore more drastic changes in sentiment.