EDS 231 Final Project

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Sentiment of Sustainable Agriculture over 30 year period

This text sentiment analysis was completed as an assignment for the course, Environmental Data Science 231: Text and Sentiment Analysis for Environmental Problems. The data was sourced from Nexis Uni database and the Food and Agriculture Organization (FAO) of the United Nations.

Original assignment instructions can be found here

Load Libraries

```
#install packages as necessary, then load libraries
if (!require(librarian)){
  install.packages("librarian")
  library(librarian)
}
librarian::shelf(here,
                 igraph,
                 janitor,
                 kableExtra,
                 ldatuning,
                 LDAvis,
                 LexisNexisTools,
                 lubridate,
                 pdftools,
                 quanteda,
                 quanteda.textplots,
                 quanteda.textstats,
                 readr,
                 reshape2,
                 sentimentr,
                 tidyr,
                 tidytext,
                  tidyverse,
                  topicmodels,
                  tsne)
```

Nexis Uni Data - Sustainable Agriculture Sentiment past 30 years

Data consists of news articles from the Nexis Uni database, and published between the years 1990-2021 using the search terms "Integrated Pest Management" and "sustainable farming." Due to Nexis Uni limiting article returns to 100 documents per search, we collected the 100 most relevant articles from each year. When a year returned less than 100 articles for the search terms, all returned articles were collected. Articles were not geographically restricted, but they were all published in English.

```
my_files <- list.files(pattern = ".docx", path = here("data", "text_data"),</pre>
                         full.names = TRUE, recursive = TRUE, ignore.case = TRUE)
dat <- Int_read(my_files) #Object of class 'LNT output'</pre>
meta df <- dat@meta
articles_df <- dat@articles</pre>
paragraphs_df <- dat@paragraphs</pre>
dat2<- data_frame(element_id = seq(1:length(meta_df$Headline)),</pre>
                   Date = meta df$Date,
                   Headline = meta df$Headline) %>%
  clean_names() %>%
  drop_na()
dat2$date[dat2$date == "0008-05-13"] <- "2008-05-13"
paragraphs_dat <- data_frame(element_id = paragraphs_df$Art_ID,</pre>
                                Text = paragraphs_df$Paragraph)
# join the headlines with the paragraphs
dat3 <- inner_join(dat2, paragraphs_dat, by = "element_id") %>%
  clean_names() %>%
  drop_na()
dat3$date[dat3$date == "0008-05-13"] <- "2008-05-13"
dat3 <- subset(dat3, text != " ")</pre>
dat3 <- dat3[!duplicated(dat3$text),]</pre>
dat3 <- dat3[!grepl("http", dat3$text),]</pre>
dat3 <- dat3[!grep1("www.", dat3$text),]</pre>
dat3 <- dat3[!grepl("Article Rating", dat3$text),]</pre>
dat3 <- dat3[!grepl("ZRL: Zero Risk Level", dat3$text),]</pre>
dat3 <- dat3[!grep1("ZOI: Zone Of Incorporation", dat3$text),]</pre>
dat3 <- dat3[!grepl("Contact", dat3$text),]</pre>
dat3 <- dat3[!grepl("--", dat3$text),]</pre>
dat3 <- dat3[!grepl("By", dat3$text),]</pre>
dat3 <- dat3[!grepl("Language:", dat3$text),]</pre>
dat3 <- dat3[!grepl("Share on", dat3$text),]</pre>
dat3 <- dat3[!grepl("Mar ", dat3$text),]</pre>
dat3 <- dat3[!grepl("Apr ", dat3$text),]</pre>
dat3 <- dat3[!grepl("zones", dat3$text),]</pre>
dat3 <- dat3[!grepl("-", dat3$text),]</pre>
dat3 <- dat3[!grepl("WASHINGTON, D.C.", dat3$text),]</pre>
dat3 <- dat3[!grepl("No19. Pest Controllers", dat3$text),]</pre>
dat3 <- dat3[!grepl("YOUR TOWN", dat3$text),]</pre>
dat3 <- dat3[!grepl("NO HEADLINE", dat3$text),]</pre>
dat3 <- dat3 %>%
```

```
drop_na()
```

Calculate the Polarity of the Headlines

```
mytext <- get_sentences(dat2$headline)
sent <- sentiment(mytext)

sent_df <- inner_join(dat2, sent, by = "element_id")

sentiment <- sentiment_by(sent_df$headline)

sent_df <- sent_df %>%
    arrange(sentiment)

sent_df$polarity <- ifelse(sent_df$sentiment < 0, -1, ifelse(sent_df$sentiment > 0, 1, 0))

sent_df <- sent_df %>%
    mutate(polarity = factor(polarity, levels = c("1", "0", "-1"))) %>%
    clean_names()

sent_df_group <- sent_df %>%
    mutate(year = lubridate::year(date)) %>%
    count(polarity, year) %>%
    drop_na()
```

Create a Plot of Headline Sentiment Polarity

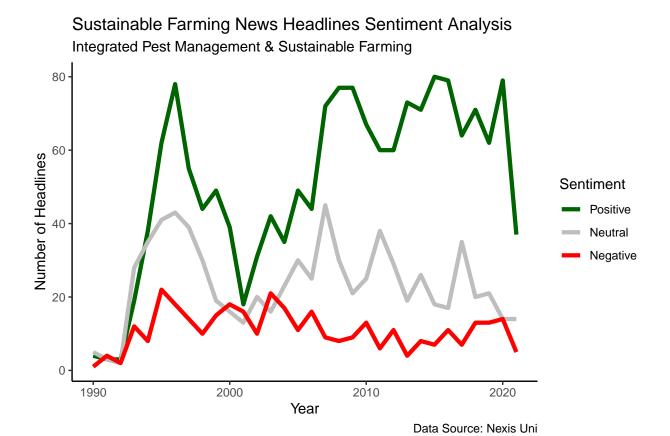


Figure 1: News media sentiment from 1990 to 2021 based on headlines with positive (green), negative (red), and neutral (grey) sentiment. Data Source: Nexis Uni.

#ggsave("graphs/headline_sentiment.png", width = 10, height = 6, units = "in")

Unnest Data into Words

```
bing_sent <- get_sentiments('bing')

#unnest to word-level tokens, remove stop words, and join sentiment words

text_words <- dat3 %>%

unnest_tokens(output = word, input = text, token = 'words') %>%

anti_join(stop_words, by = 'word') %>%

inner_join(bing_sent, by = "word")
```

Calculate the Sentiment Score of Text

This calculation was done by counting the number of sentiment words occurring per day.

Create a Plot of Sentiment Scores

News Article Text Sentiment Analysis Integrated Pest Management & Sustainable Farming

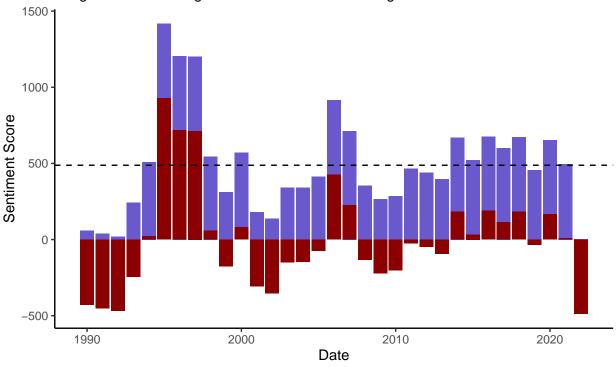


Figure 2: News article text sentiment analysis from 1990 to 2021. The raw sentiment score is in blue and offset sentiment score is red. Data Source: Nexis Uni.

Data Source: Nexis Uni

```
\#ggsave("graphs/text\_sentiment.png", width = 10, height = 6, units = "in")
```

Perform Sentiment Analysis using the NRC Sentiment Words

```
nrc_sent <- get_sentiments('nrc')</pre>
# restart with the initial data
# unnest into words again before inner joining with nrc sentiment words
slr_nrc_sent <- dat3 %>%
  mutate(year = lubridate::year(date)) %>%
  unnest_tokens(output = word, input = text, token = 'words') %>%
  inner_join(nrc_sent) %>%
  count(word, sentiment, year, sort = TRUE) %>%
  ungroup()
# remove the positive and negative sentiments from the list
slr_nrc_sent <- subset(slr_nrc_sent, !(sentiment %in% c("positive", "negative")))</pre>
slr_nrc_sent_graph <- slr_nrc_sent %>%
  group_by(sentiment) %>%
  slice_max(n, n = 20) %>%
  ungroup() %>%
  mutate(word = reorder(word, n)) %>%
```

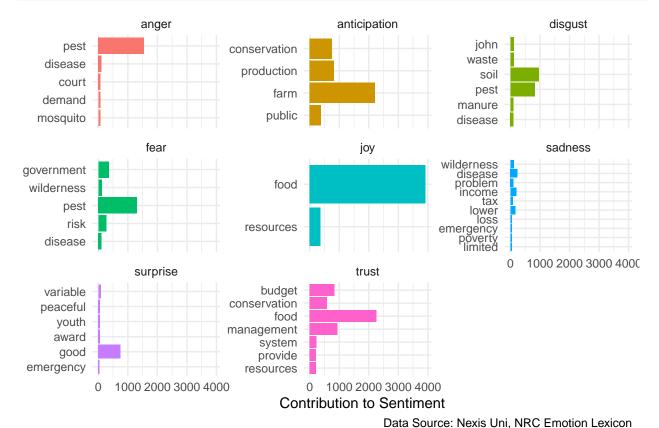


Figure 3: Individual words associated with the emotions within our sentiment analysis. The key takeaways from this figure are that most people associate food with joy and trust, there is a lot of anticipation around farming and production/conservation. However, many people are angry and fearful of pests, disease, and the government. Data Source: Nexis Uni, NRC Emotion Lexicon

```
#ggsave("graphs/nrc_sentiment.png", width = 10, height = 6, units = "in")
```

Calculate the Distribution of Sentiment

```
slr_nrc_sent2 <- slr_nrc_sent %>%
  group_by(year, sentiment) %>%
  summarise(total_n = sum(n)) %>%
  spread(sentiment, total_n) %>%
  ungroup()
```

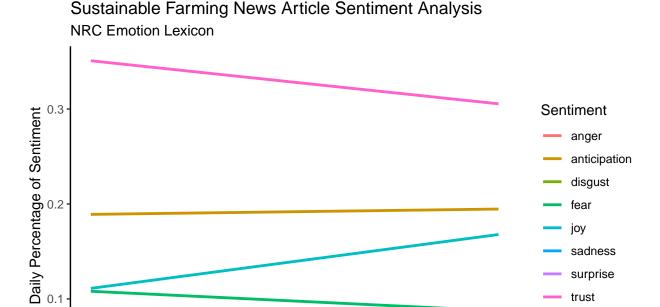
```
slr_nrc_sent2[is.na(slr_nrc_sent2)] = 0

slr_nrc_sent2 <- slr_nrc_sent2 %>%
  mutate(totals = anger + anticipation + disgust + fear + joy + sadness + surprise + trust)

slr_nrc_sent2 <- slr_nrc_sent2 %>%
  pivot_longer(cols = !c("year", "totals"), names_to = "sentiment", values_to = "n")

slr_nrc_sent2 <- slr_nrc_sent2 %>%
  mutate(percentage = n / totals)
```

Daily Percentage of Sentiment Words



Data Source: Nexis Uni, NRC Emotion Lexicon

2020

trust

Figure 4: This analysis shows how emotions are changing over time. Trust had an overwhelming percentage in the articles, across the entire time period, but decreased as time went on. The second most used emotion was anticipation, which had a very slight increase as time went on. Joy increased the most, and was the third most used emotion. All the other emotions were very similar in percentage used. There was also an interesting increase in disgust over time. Data Source: Nexis Uni, NRC Emotion Lexicon

2010

```
#ggsave("graphs/sentiment_words.png", width = 10, height = 6, units = "in")
```

Number of Sustainable Farming News Articles: 1990 - 2021

Date

2000

1990

Here we are looking at the number of articles that were published from 1990-2021. These articles were found using the Nexis Uni database using the search terms "Integrated Pest Management" and "sustainable farming".

```
# Create a table that contains the total number of published articles for each year.
articles_by_year <- data.frame(year = 1990:2021, n_articles = 1)
# Add number of publications for each year.
articles_by_year[1,2] <- 9
articles_by_year[2,2] <- 10</pre>
articles_by_year[3,2] <- 7</pre>
articles_by_year[4,2] <- 36
articles_by_year[5,2] <- 57</pre>
articles_by_year[6,2] <- 76
articles by year[7,2] \leftarrow 89
articles_by_year[8,2] <- 76
```

```
articles_by_year[9,2] <- 67</pre>
articles_by_year[10,2] <- 70
articles_by_year[11,2] <- 52</pre>
articles_by_year[12,2] <- 47
articles_by_year[13,2] <- 56</pre>
articles_by_year[14,2] <- 69</pre>
articles_by_year[15,2] <- 70</pre>
articles_by_year[16,2] <- 78
articles_by_year[17,2] <- 82</pre>
articles_by_year[18,2] <- 100</pre>
articles_by_year[19,2] <- 176
articles_by_year[20,2] <- 207</pre>
articles by year[21,2] \leftarrow 227
articles_by_year[22,2] <- 234
articles_by_year[23,2] <- 386</pre>
articles_by_year[24,2] <- 381</pre>
articles_by_year[25,2] <- 411</pre>
articles_by_year[26,2] <- 485</pre>
articles_by_year[27,2] <- 696
articles_by_year[28,2] <- 606</pre>
articles_by_year[29,2] <- 710
articles_by_year[30,2] <- 905</pre>
articles_by_year[31,2] <- 1000</pre>
articles_by_year[32,2] <- 1118</pre>
# Now plot the above data.
articles_plot <- ggplot(articles_by_year, aes(x = year, y = n_articles)) +
  geom_line(size = 1.5, col = "darkgreen") +
  labs(title = "Number of Articles Published by Year",
       subtitle = 'Search terms "Integrated Pest Management" and "Sustainable Farming".',
       caption = "Data: Nexis Uni",
       x = "Year",
       y = "Number of Articles Published") +
  theme_classic()
articles_plot
```

Number of Articles Published by Year Search terms "Integrated Pest Management" and "Sustainable Farming".

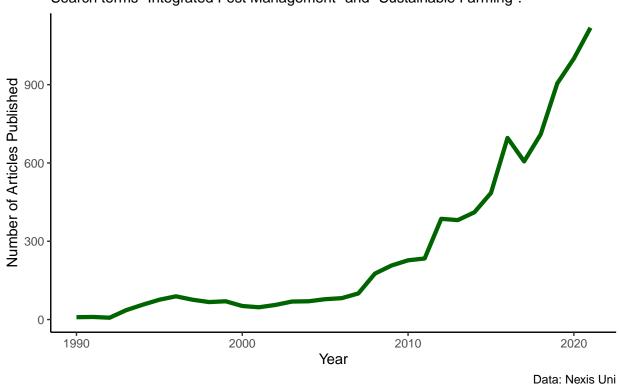


Figure 5: Trend of news articles that include the search terms 'Intergrated Pest Managment' and 'sustainable farming' over time from 1990 to 2021. Data Source: Nexis Uni

#ggsave("graphs/articles_plot.png", width = 10, height = 6, units = "in")

Global FAO data on farming practices over time

Data was collected from the Food and Agriculture Organization (FAO) of the United Nations on global farming practices since 1960. The data was collected from 1990 to 2021 from the FAO data service website, and combined into one csy file.

```
farming_practices_full <- list.files(path = "data/farming_practices/", full.names = T)</pre>
farming_practices <- list.files(path = "data/farming_practices/", full.names = F)</pre>
practice names <- str replace(farming practices, "fao ", "") %>% str replace(".csv", "")
practices_df <- map(farming_practices_full, read_csv) %>%
        map(~ select(., Area, Element, Item, Year, Unit, Value)) %>%
        reduce(rbind)
summary practices <- practices df %>%
        group_by(Item, Year) %>%
        filter(Unit != "%") %>%
        summarize(Unit = Unit,
                  value = sum(Value))
graph_practices <- function(topic) {</pre>
filtered_practice <- summary_practices %>%
        filter(Item == topic,
               Year > 1989)
filtered_practice %>%
        ggplot(aes(x = Year, y = value)) +
        geom line(size = 1.5, color = "darkgreen") +
        expand_limits(y = 0) +
        ylab(paste(filtered_practice$Unit[1], topic)) +
        theme_classic()
}
topics <- summary_practices %>%
        ungroup() %>%
        select(Item) %>%
        distinct() %>%
        pull()
practices graphs <- map(topics, graph practices)</pre>
# filenames_practices_graphs <- pasteO(topics, ".jpg")</pre>
#
# ggsave_med <- partial(ggsave, device = "jpg", width = 10, height = 6, units = "in")
# map2(filenames_practices_graphs, practices_graphs, ggsave_med)
Let's adjust the summary we made earlier by adding a country grouping
summary_practices <- practices_df %>%
        group_by(Item, Year, Unit, Area) %>%
        filter(Unit != "%") %>%
```

summarize(value = sum(Value))

Now, let's filter our data before we graph to only include the U.S.

```
graph_practices <- function(topic, country) {</pre>
filtered_practice <- summary_practices %>%
        filter(Item == topic, Area == country,
               Year > 1989)
filtered practice %>%
        ggplot(aes(x = Year, y = value)) +
        geom_line(size = 1.5, color = "darkgreen") +
        expand_limits(y = 0) +
        ylab(paste(filtered_practice$Unit[1], topic)) +
        theme_classic()
}
topics <- summary_practices %>%
        ungroup() %>%
        select(Item) %>%
        distinct() %>%
        pull()
practices_graphs <- map2(topics, "United States of America", graph_practices)</pre>
# filenames_practices_graphs <- pasteO(topics, "us_only", ".jpg")</pre>
\# ggsave\_med \leftarrow partial(ggsave, device = "jpg", width = 10, height = 6, units = "in")
# map2(filenames_practices_graphs, practices_graphs, ggsave_med)
```