Homework 9

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Project Update

For homework 9 of our project, we decided to focus on the airport reviews and sentiment analysis. We got the top 2,000 words in the reviews and scraped synonyms of each word. We then calculated the frequency of the word and its synonyms. Using these counts, we can then simplify the reviews we have which will make sentiment analysis easier.

```
library(RSQLite)
library(stringr)
library(XML)

dcon <- dbConnect(SQLite(), dbname = "group10.db")
dbListTables(dcon)</pre>
```

```
## [1] "flights" "reviews" "top2kdf" "zillow"
```

Query all of the flight review data into a data frame.

```
res <- dbSendQuery(conn = dcon, "
SELECT *
FROM reviews;")
reviews <- dbFetch(res, -1)
dbClearResult(res)</pre>
```

Gather all of words from all of the reviews. We eliminate some words such as "the" because these words provide no business value.

```
# Parse the review into a list of words
content = reviews$content
content_split = str_split(reviews$content, " ")
content_split_lower = sapply(content_split, tolower)
content_split_grep = sapply(content_split_lower, gsub, pattern="[[:punct:]]", replacement="")
words = unlist(content_split_grep, recursive = FALSE)

# Ignore non-meaningful words and words that do not have synonyms
words_to_ignore <- c("the", "this", "to", "a", "of", "in", "he", "she", "an", "and", "is", "with", "but
actual_words <- words[!words %in% words_to_ignore]
allContent <- as.vector(actual_words)

# Count the number of words
counts_of_words = table(allContent)

# Sort the top words
top_words = sort(counts_of_words, decreasing = TRUE)[1:2000]
word_vector = rep(names(top_words)[1:5])</pre>
```

Next we scrape thesaurus.com for each top 2,000 word. For example, for the word "airport", the URL we are

scraping is "https://www.thesaurus.com/browse/airport?s=t". We scraped once and saved the output of the scraping into the top2kdf table in our database.

```
top2k = word vector
top2kdf = data.frame(row.names=top2k)
# Iterate through the top 2,000 words and scrape the thesaurus website for each word.
for (word in top2k)
{
  # Create the url
  url = paste("https://www.thesaurus.com/browse/", word, "?s=t", sep="")
  # Try to scrape the url, if it exists
  result <- try(download.file(url, destfile = "advfn.html", quiet = TRUE))
  # Parse the information from the scrapped url
  if (result == 0) {
    doc <- htmlParse("advfn.html")</pre>
    tmp <- getNodeSet(doc, "//a[@data-linkid='nn1ov4']")</pre>
    new_line <- append(word, as.character(xmlToDataFrame(tmp)$text[1:10]))</pre>
    top2kdf <- rbind(top2kdf, data.frame(matrix(new_line, nrow=1)))</pre>
  }
}
top2kdf
```

Below is a snippet of the created SQL table. Here we query all of the scraped top2kdf table into a data frame. Column 1 is the word from the review (one of the top 20,000), and columns 2-11 are the 10 most closely related columns.

```
res <- dbSendQuery(conn = dcon, "
SELECT *
FROM top2kdf;")
top2kdf <- dbFetch(res, -1)
dbClearResult(res)
head(top2kdf)</pre>
```

```
##
           X1
                     X2
                                  ХЗ
                                                Х4
                                                          Х5
                                                                     Х6
## 1
      airport airfield
                           airstrip installation
                                                               airdrome
                                                      runway
## 2
          not
              no more
                         not at all
                                       not either
                                                     neither
                                                                no more
## 3 security
                   bond
                                care
                                          freedom guarantee insurance
## 4 terminal
                  fatal
                                           lethal
                          incurable
                                                     closing
                                                                extreme
## 5
         very
                 actual appropriate
                                        authentic
                                                        bare bona fide
## 6
                                            never
                                                             negation
           no
                    nay
                                nix
                                                         not
##
                X7
                             Х8
                                         χ9
                                                      X10
                                                                X11
## 1
           hangar
                       heliport
                                                 aerodome
                                                           helipad
                                      strip
              not
                     not at all not either
                                                       no
                                                                nay
## 3 preservation surveillance
                                      aegis
                                                agreement armament
          killing
                            lag
## 4
                                       last
                                                   latest
                                                            latter
## 5
                                                  genuine
          correct
                       especial
                                    express
                                                              ideal
## 6
       antithesis
                        antonym
                                      blank cancellation contrary
```

The number of rows of data that have been produced is:

```
print(dim(top2kdf)[1])
```

```
## [1] 1749
```

The plot shows that there are multiple occurrences of a word, whether synonym or not. For example, the word "move" occurs 21 times, so we can expect that ~200 words in the top2kdf are synonyms of "move". We can then condense all the other synonyms of "move" to "move" itself, creating a more simplified and condensed

dictionary of review words. This will help standardize the way we compare reviews, and make sentiment analysis more reliable from a simplified dictionary. We will then use sentiment analysis to determine whether an airport is liked or not, which will help us compare airports.

```
all_words = as.vector(t(top2kdf))
word_table = table(all_words)
word_table = sort(word_table, decreasing=TRUE)
word_df = data.frame(word_table[2:26])
```

We decided to plot the frequency of the top 25 words, to help visualize the relative frequency of a word compared to others.

```
library(ggplot2)

ggplot(data = word_df, aes(x=all_words, y=Freq)) +
  geom_bar(stat="identity", fill="orange") +
  labs(x="Word", y="Count", title="Top 25 word counts") +
  theme(axis.text.x=element_text(angle=90, hjust=1, size=12))
```

Top 25 word counts

