## Peer-graded Assignment: Milestone Report

Megha

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# Introduction

This milestone report is for the Coursera Capstone Project for Data Scince Speciaization Course. The report provides a short overview of the exploratory analysis of the text data to be used together with a briefly summary of the the plans for creating the prediction algorithm and Shiny app in a way that would be understandable to a non-data scientist manager. The motivation for this project is to:

1. Demonstrate that i have downloaded the data and have successfully loaded it in 2. Create a basic report of summary statistics about the data sets. 3. Report any interesting findings that you amassed so far. 4. Get feedback on your plans for creating a prediction algorithm and Shiny app. Initiatization, Data Reading and Summarization

Install and load requisite packages that are not already installed and loaded.

### knitr::opts\_chunk\$set(echo = FALSE)

new.packages <- list.of.packages[!(list.of.packages %in% installed.packages()[,"Package"])]</pre>

list.of.packages <- c("dplyr", "ggplot2", "stringi", "knitr", "tm", "pwr", "wordcloud", "RColorBrewer", "kableExt ra", "survey", "RSQLite", "SnowballC", "RWeka", "plotly", "devtools", "slam", "tokenizers", "quanteda", "ngram")

if(length(new.packages)) install.packages(new.packages, repos="http://cran.rstudio.com/") library(dplyr) library(ggplot2) library(stringi) library(tm) library(wordcloud) library(RColorBrewer) library(SnowballC) library(RWeka) library(knitr) library(kableExtra) library(pwr) library(survey) library(RSQLite) library(plotly) library(devtools) library(slam) library(tokenizers) library(quanteda) library(ngram) library(NLP) # devtools::install\_github("ropensci/plotly") We start by reading the file from the source https://d396qusza40orc.cloudfront.net/dsscapstone/dataset/Coursera-SwiftKey.zip We then unzip the file to get the individual files in the zip file. We note that the data is from three different text sources, i.e. blogs, news, and twitter feeds and are given in four (4) different languages: German, English (US), Finnish, and Russian. For this project, we shall use the English (US) data sets. We then get a summary of the English Data: file size, number of lines and number of words.

196.28 1010242 34762395 News 30093410 Twitter 159.36 2360148

No. of Lines

899288

No. of Words

37546246

**Data Sampling** 

Clearly, the data sets are very large and it is only logical that we take a non-biased sample to conduct the exploratory data analysis at this stage. We sample 2% of the number lines in each data set, then combine them into one data set that shall be used in building the corpus.

## [1] "The computed and combined sample size (lines) is 42695 with 1023145 words."

File size, Mb

200.42

**Exploratory Data Analysis Building the Corpus** The combined sample is used to build the corpus and the following preprocessing is done to: 1. remove emoticons 2. Remove punctuation

#### 4. Remove stopwords 5. Stem the words (eg. "worked" will be changed to "work") 6. Convert to lower case

7. Remove white space

3. Remove numbers

**Engilsh Data Sets Summary** 

Blogs

We identify all combinations of adjacent words (ngrams) in the corpus. We limit ourselves to one, two and three words (unigrams, bigrams and trigrams). We then generate the word clouds.

**Unigram Word Cloud** 

best littl lol mani hope also first great state inthank now good follow today work but dont go one make you home want see want way so school last there last the last there last the last there last the last the last there last the night need the get rt Syour

Bigram Word Cloud

good morn mother day one day even though lol i guess but i first time never last week year old

you can now iright now make sure i hate i ha

i made i got

two year ago ∂

N-Gram Frequency

jnathe saidgetjust like onegotime -

Unigrams

imdayyearlove makenew goodknownowthankworkdontwantsee-

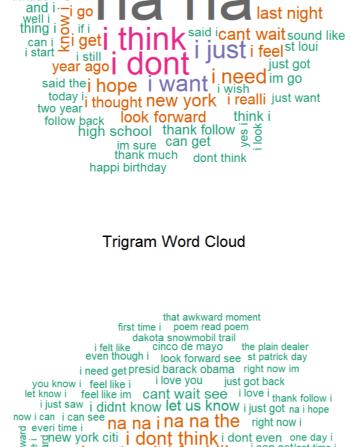
> na na i love · i thinki donti just na the -

you can now iright now make sure

Tokenization (Identifying ngrams) and Word Clouds

and new said a year us show feelcome timecan triwell what two we peopl back Freallithis even if that got F game way week stillplay

dont know



i cant wait i feel like i can getlast time i

I dont know i wish i to na na in

much i love to be iknow im think i just want im sorri is to happi mother day last night i i know i i guess i just want happi mother day last night i i dont want i think in ith a

have great day i dont get

We conduct ngram analysis. The analysis is limited to the top twenty (20) ngrams in the corpus.

na na this i dont see i knew i north dakota snowmobil know i can just want say

i dont want i think im i thought i i dont care i dont like i realli like na na we dont know i

Most frequent Unigrams

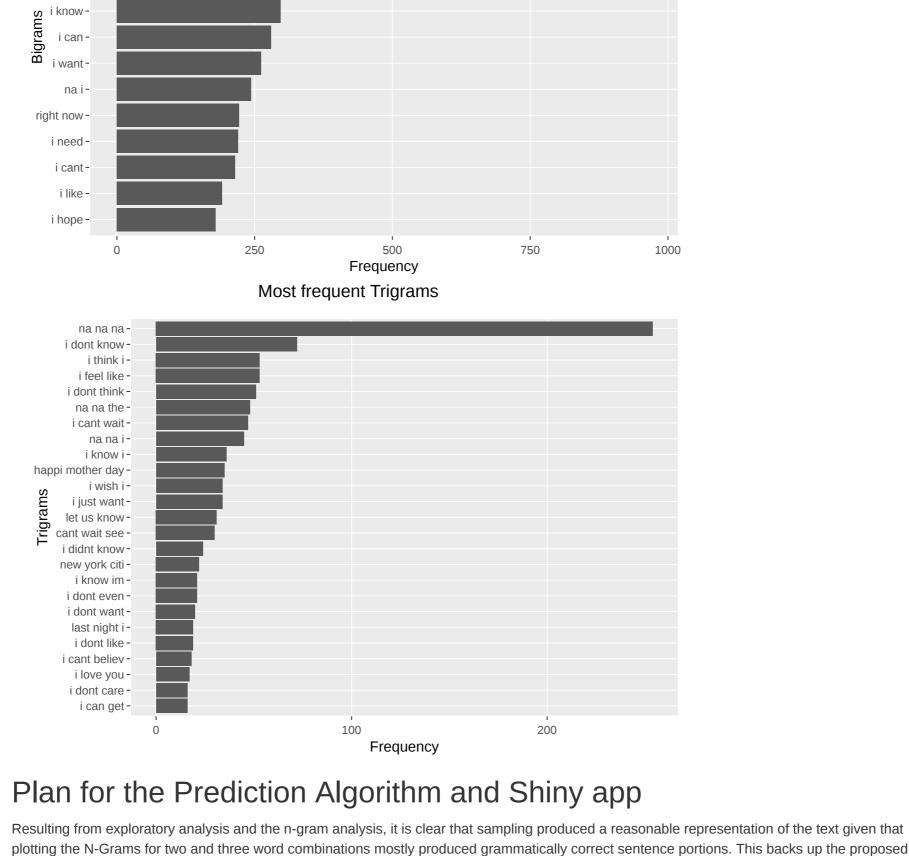
have great weekend la la la<sub>world war ii</sub>

i want go

i dont need

10000

Most frequent Bigrams



plan for for the prediction algorithm to utilize a frequency look up using the n-gram models.

In developing and testing the prediction algorithm, the following shall specifically be done:

5. Build a shiny app which is interactive and allows users to enter text for prediction of the next word(s)

new.packages <- list.of.packages[!(list.of.packages %in% installed.packages()[,"Package"])] if(length(new.packages)) install.packages(new.packages, repos="http://cran.rstudio.com/")

6. Experiment to optimize the model to handle unseen n-grams, increasing model performance.

4. Test the performance of the algorithm on test data using some of the following metrics: accuracy, precision, confusion matrix, recall and f1

list.of.packages <- c("dplyr", "ggplot2", "stringi", "knitr", "tm", "pwr", "wordcloud", "RColorBrewer", "kableExt ra", "survey", "RSQLite", "SnowballC", "RWeka", "plotly", "devtools", "slam", "tokenizers", "quanteda", "ngram")

2. Better sample size selection technique for the training and testing data sets 3. Build the prediction model basing on the n-gram frequancy on test data

7. Get feedback ion the plan for creating a prediction algorithm and Shiny app

1. Additional data cleaning

**Appendices** 

Appendix - Setup

library(dplyr) library(ggplot2) library(stringi) library(tm)

library(wordcloud) library(RColorBrewer) library(SnowballC) library(RWeka) library(knitr)

library(kableExtra)

library(tokenizers) library(quanteda) library(ngram) library(NLP)

Appendix - Data Source

unzip("Coursera-SwiftKey.zip")

"No. of Lines", "No. of Words")))

con <- file(file.list[i], "rb")</pre>

en\_US.summary[i,2] <- length(text[[i]])</pre>

for (i in 1:3) {

close(con)

Appendix - Read Data

# devtools::install\_github("ropensci/plotly")

text <- list(blogs = "", news = "", twitter = "")</pre>

text[[i]] <- readLines(con, encoding = "UTF-8", skipNul = TRUE)</pre>

en\_US.summary[i,3] <- sum(stri\_count\_words(text[[i]]))</pre>

en\_US.summary[i,1] <- round(file.info(file.list[i])\$size / 1024^2, 2)</pre>

kable(en\_US.summary, "html", caption = "Engilsh Data Sets Summary") %>%

tm\_map(removeWords, stopwords("english")) %>% # remove stopwords

unigram <- NGramTokenizer(corpusNew, Weka\_control(min = 1, max = 1))</pre> bigram <- NGramTokenizer(corpusNew, Weka\_control(min = 2, max = 2))</pre> trigram <- NGramTokenizer(corpusNew, Weka\_control(min = 3, max = 3))</pre>

unigram.df <- unigram.df[order(unigram.df\$Freq, decreasing = TRUE),]</pre>

bigram.df <- bigram.df[order(bigram.df\$Freq, decreasing = TRUE),]</pre>

trigram.df <- trigram.df[order(trigram.df\$Freq, decreasing = TRUE),]</pre>

tm\_map(stemDocument) %>% # Stem words

knitr::opts\_chunk\$set(echo = TRUE)

unigram.df <- data.frame(table(unigram))</pre>

bigram.df <- data.frame(table(bigram))</pre>

trigram.df <- data.frame(table(trigram))</pre>

## Unigram

## Bigram

## Trigram

tm\_map(tolower) %>% # convert to lower case

Appendix - Tokenise and Word Cloud

tm\_map(stripWhitespace)%>% # Strip white spaces tm\_map(PlainTextDocument) # Convert to plain text

library(pwr) library(survey) library(RSQLite) library(plotly) library(devtools) library(slam)

knitr::opts\_chunk\$set(echo = FALSE)

#### setwd("P:/NITA/New/Personal/Training/Data Science/John Hopkins/Course 10 - Capstone Project/Week 2/Peer-Graded") sourceUrl <-"https://d396qusza40orc.cloudfront.net/dsscapstone/dataset/Coursera-SwiftKey.zip"</pre> if (!file.exists("Coursera-SwiftKey.zip")){ download.file(sourceUrl, destfile = "Coursera-SwiftKey.zip")

kable\_styling(bootstrap\_options = c("striped", "hover")) Appendix - Sample Data set.seed(12345) ## Get population sizes (lines) blogs\_pop <- length(text\$blogs)</pre> news\_pop <- length(text\$news)</pre> twitter\_pop <- length(text\$twitter)</pre> ## Create the sample blogs\_sample <- sample(text\$blogs, 0.01\*length(text\$blogs))</pre> news\_sample <- sample(text\$news, 0.01\*length(text\$news))</pre> twitter\_sample <- sample(text\$twitter, 0.01\*length(text\$twitter))</pre> sample\_corpus\_data <- c(blogs\_sample, news\_sample, twitter\_sample)</pre> sumLines <- length(sample\_corpus\_data)</pre> sumWords <- sum(stri\_count\_words(sample\_corpus\_data))</pre> paste("The computed and combined sample size (lines) is", sumLines, "with", sumWords, "words.") ## Write the file to disc writeLines(sample\_corpus\_data, "sample\_corpus\_data.txt") Appendix - Clean Corpus sample\_corpus\_data <- iconv(sample\_corpus\_data, 'UTF-8', 'ASCII')</pre> corpusNew <- Corpus(VectorSource(as.data.frame(sample\_corpus\_data, na.rm=TRUE, stringsAsFactors = FALSE)))</pre> ## Pre-process the corpus corpusNew <- corpusNew %>% tm\_map(removePunctuation) %>% # remove punctuation tm\_map(removeNumbers) %>% # Remove numbers

file.list = c("final/en\_US/en\_US.blogs.txt", "final/en\_US/en\_US.news.txt", "final/en\_US/en\_US.twitter.txt")

en\_US.summary <- matrix(0, nrow = 3, ncol = 3, dimnames = list(c("Blogs", "News", "Twitter"),c("File size, Mb",

### col.palette <- brewer.pal(8, "Dark2") # Colour Palette to be used for all Word Clouds # Unigram Word Cloud layout(matrix(c(1, 2), nrow=2), heights=c(1, 4)) par(mar=rep(0, 4))plot.new()

text(x=0.5, y=0.5, "Unigram Word Cloud")

text(x=0.5, y=0.5, "Bigram Word Cloud")

text(x=0.5, y=0.5, "Trigram Word Cloud")

layout(matrix(c(1, 2), nrow=2), heights=c(1, 4))

layout(matrix(c(1, 2), nrow=2), heights=c(1, 4))

## Generate Word Clouds

# Bigram Word Cloud

par(mar=rep(0, 4))

## Trigram Word Cloud

par(mar=rep(0, 4))

plot.new()

plot.new()

```
Appendix - Tokenise and Word Cloud
#set up the plot area
layout(matrix(c(1,1,2,3), 2, 2, byrow = TRUE))
## Unigram Frequency
ggUni <- ggplot(head(unigram.df,25), aes(reorder(unigram,Freq), Freq)) +
  geom_bar(stat="identity") + coord_flip() +
  xlab("Unigrams") + ylab("Frequency") +
  ggtitle("Most frequent Unigrams")
ggplotly(ggUni)
# Bigram Frequency
ggBi <- ggplot(head(bigram.df,15), aes(reorder(bigram,Freq), Freq)) +</pre>
  geom_bar(stat="identity") + coord_flip() +
  xlab("Bigrams") + ylab("Frequency") +
  ggtitle("Most frequent Bigrams")
ggplotly(ggBi)
## Tigram Frequency
ggTri <- ggplot(head(trigram.df,25), aes(reorder(trigram,Freq), Freq)) +</pre>
  geom_bar(stat="identity") + coord_flip() +
  xlab("Trigrams") + ylab("Frequency") +
  ggtitle("Most frequent Trigrams")
ggplotly(ggTri)
```

wordcloud(unigram.df\$unigram, unigram.df\$Freq, max.words=90, random.order = F, colors=col.palette, main="Title")

wordcloud(bigram.df\$bigram, bigram.df\$Freq, max.words=90, random.order = F, colors=col.palette, main="Title")

wordcloud(trigram.df\$trigram, trigram.df\$Freq, max.words=90, random.order = F, colors=col.palette, main="Title")