Some Simple SPOOKY Data Analysis

Cindy Rush
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Introduction

This files contains some simple analysis of the SPOOKY data. The goal is to remind ourselves of some of our basic tools for working with text data in R and also to practice reproducibility. You should be able to put this file in the doc folder of your Project 1 repository and it should just run (provided you have multiplot.R in the libs folder and spooky.csv in the data folder). If you open to file from a forked Week1-GitHub repo, you should have no trouble running the code directly.

Setup the libraries

First we want to install and load libraries we need along the way. Note that the following code is completely reproducible – you don't need to add any code on your own to make it run.

```
packages.used=c("ggplot2", "dplyr", "tidytext", "wordcloud", "stringr", "ggridges")

# check packages that need to be installed.
packages.needed = setdiff(packages.used, intersect(installed.packages()[,1], packages.used))

# install additional packages
if(length(packages.needed) > 0) {
   install.packages(packages.needed, dependencies = TRUE, repos = 'http://cran.us.r-project.org')}

library(ggplot2)
library(dplyr)
library(tidytext)
library(wordcloud)
library(stringr)
library(ggridges)

source("../libs/multiplot.R")
```

Read in the data

The following code assumes that the dataset spooky.csv lives in a data folder (and that we are inside a docs folder).

```
spooky <- read.csv('../data/spooky.csv', as.is = TRUE)</pre>
```

An overview of the data structure and content

Let's first remind ourselves of the structure of the data.

```
head(spooky)
##
          id
## 1 id26305
## 2 id17569
## 3 id11008
## 4 id27763
## 5 id12958
## 6 id22965
##
## 1
## 2
## 3
## 4
## 5
## 6 A youth passed in solitude, my best years spent under your gentle and feminine fosterage, has so r
##
     author
## 1
        EAP
        HPL
## 2
## 3
        EAP
        MWS
## 4
## 5
        HPL
## 6
        MWS
summary(spooky)
##
         id
                             text
                                                author
## Length:19579
                        Length: 19579
                                             Length: 19579
## Class :character
                        Class :character
                                             Class : character
## Mode :character
                        Mode : character
                                             Mode : character
We see from the above that each row of our data contains a unique ID, a single sentence text excerpt, and an
abbreviated author name. HPL is Lovecraft, MWS is Shelly, and EAP is Poe. Here are a few example sentences:
spooky$text[1]
## [1] "This process, however, afforded me no means of ascertaining the dimensions of my dungeon; as I
spooky$text[13494]
## [1] "While my companion contemplated with a serious and satisfied spirit the magnificent appearances
spooky$text[666]
## [1] "What was it I paused to think what was it that so unnerved me in the contemplation of the House
We finally note that there are no missing values, and we change author name to be a factor variable, which
will help us later on.
sum(is.na(spooky))
## [1] 0
spooky$author <- as.factor(spooky$author)</pre>
```

An intro to tidytext

For my tutorials on Project 1, I will be using tidytext. If this is new to you, here's a textbook that can help: Text Mining with R; A Tidy Approach. It teaches the basic handling of natural language data in R using tools from the "tidyverse". The tidy text format is a table with one token per row, where a token is a word.

Data Cleaning

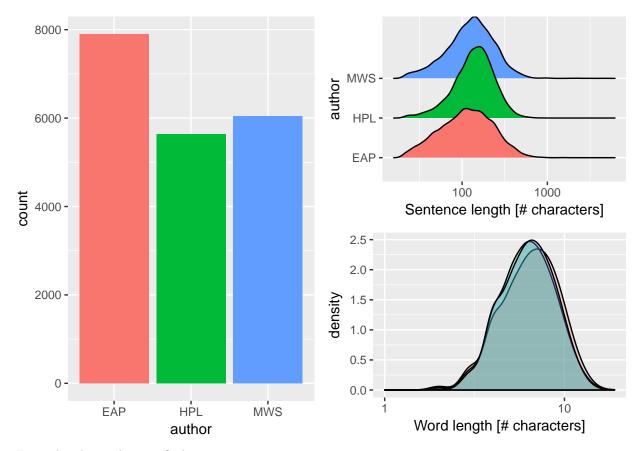
We first use the unnest_tokens() function to drop all punctuation and transform all words into lower case. At least for now, the punctuation isn't really important to our analysis – we want to study the words. In addition, tidytext contains a dictionary of stop words, like "and" or "next", that we will get rid of for our analysis, the idea being that the non-common words (... maybe the SPOOKY words) that the authors use will be more interesting.

```
spooky_wrd <- unnest_tokens(spooky, word, text)
spooky_wrd <- anti_join(spooky_wrd, stop_words, by = "word")</pre>
```

Data Visualization

First we'll do some simple numerical summaries of the data to provide some nice visualizations.

```
p1 <- ggplot(spooky) +
      geom_bar(aes(author, fill = author)) +
      theme(legend.position = "none")
spooky$sen_length <- str_length(spooky$text)</pre>
head(spooky$sen_length)
## [1] 231 71 200 206 174 468
p2 <- ggplot(spooky) +
      geom_density_ridges(aes(sen_length, author, fill = author)) +
      scale x log10() +
      theme(legend.position = "none") +
      labs(x = "Sentence length [# characters]")
spooky_wrd$word_length <- str_length(spooky_wrd$word)</pre>
head(spooky_wrd$word_length)
## [1] 7 8 5 12 10 7
p3 <- ggplot(spooky_wrd) +
      geom density(aes(word length, fill = author), bw = 0.05, alpha = 0.3) +
      scale_x_log10() +
      theme(legend.position = "none") +
      labs(x = "Word length [# characters]")
layout \leftarrow matrix(c(1, 2, 1, 3), 2, 2, byrow = TRUE)
multiplot(p1, p2, p3, layout = layout)
```



From the above plots we find:

- .
- •
- _

Now we study some of the most common words in the entire data set. With the below code we plot the fifty most common words in the entire datset. We see that "time", "life", and "night" all appear frequently.

```
words <- names(table(spooky_wrd$word))</pre>
freqs <- table(spooky_wrd$word)</pre>
head(sort(freqs, decreasing = TRUE))
##
##
          life found night
                                      day
    time
                              eyes
                                     516
##
     729
           563
                  559
                         559
                               540
wordcloud(words, freqs, max.words = 50, color = c("purple4", "red4", "black"))
```

passed words nature day
passed friend horror
spirit heart looked appeared
mind is earth father
hope half voice hand death
city night over soul or
headtoldhuman water
left days world time
raymond door moment