## **Explaining the newAbbrvReport Function**

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I will discuss the newAbbrvReport function in this document. The function can be loaded by running the newAbbrvReport.R document, and the newAbbrvReportGuide.pdf file explains how to operate it. In case people are interested in improving, changing, or otherwise understanding the code, I will explain here line-by-line what the code means. I will first present a segment of code and then explain it below.

## The newAbbrvReport function

```
newAbbrvReport <- function(channel, month, year) {</pre>
```

This first line names the function newAbbrvReport and specifies the three different arguments that it will take (see newAbbrvReportGuide.pdf for a description of these arguments).

```
suppressMessages(require(RODBC))
suppressMessages(require(dplyr))
suppressMessages(require(stringr))
suppressMessages(require(stringi))
suppressMessages(require(tidytext))
```

This chunk loads the necessary packages, while suppressing annoying startup messages that they might tell the user.

```
year_ch <- str_sub(year,start=-2)</pre>
```

The user provides a four digit year (e.g., "2017"); however, we will want to shorten this to just "17" for variable naming purposes. This line only takes the last two characters of the year argument. So what was "2017" now is saved as "17" in the object named year\_ch.

The user also provides a two digit number that represents the month. For variable naming purposes, we want to convert this to a month name. These lines of code create a small data frame where one variable is the month number, and the other variable is a month name. This is a "key" for the function to convert month numbers into month names. Here's what this data frame looks like:

| month | ch  |
|-------|-----|
| 01    | jan |
| 02    | feb |
| 03    | mar |
| 04    | apr |
| 05    | may |
| 06    | jun |
| 07    | jul |
| 80    | aug |
| 09    | spt |
| 10    | oct |
| 11    | nov |
| 12    | dec |

The last part of the code above takes whatever row includes the correct month number and returns the column that holds the month names. This is saved as a character into the month ch object. This will come in handy in a bit.

This is a SQL query that pulls down the BioSense ID, admit reason, and chief complaint for every record appearing in the month and year specified and saves it into a new data frame called newraw.

This chunk cleans up that new bit of data called newraw. All NA values in the admit reason and chief complaint fields will be replaced with nothing (i.e., ""), which will help in the next line when we create a new variable, called textraw, that concatenates the two fields (if NAs

are not replaced, then the character "NA" would have been pasted). The next line takes the textraw variable, removes everything that isn't a letter, and saves it in a new variable called text. Lastly, the column names for this data frame are renamed to be easier to work with.

```
newword <- unnest_tokens(newraw, word, text)
```

This function takes the text variable and replaces it with a word variable. For every word in the text field, a new row is created that contains just one word. So if the text field read, "pt is in pain", the text variable would be replaced with a word variable, and that case would have four rows: one for "pt", another for "is", another for "in", and another for "pain". This allows us to do analyses for each word. This new data frame is saved into an object called newword.

```
newword <- filter(group_by(newword, id), !duplicated(word))</pre>
```

This line first groups the newword data by id (i.e., the BioSense ID). For each group, the data are then filtered to remove any word that has been duplicated. This trimmed data file is reassigned to the object newword.

```
data("stop_words")
newword <- anti_join(newword, stop_words, by="word")</pre>
```

These two lines read in the stop\_words data from the tidytext package. The next line then removes any of these stop words from the newword data. Stop words are things we are generally uninterested in: Words like, "of", "the", "and", etc.

```
newcounts <- count(ungroup(newword), word, sort=TRUE)</pre>
```

This line ungroups the data by id so that counts are done acros the whole dataset. We count up words and sort them from most used to least used. This new dataset, which includes a column word and a column n (i.e., how many times it appeared), is saved into the newcounts object.

Now we are ready to start joining these new counts with the old data we had!

```
oldcounts$new <- factor("not new")
```

The first thing that happens is a new variable named new is created in the oldcounts data. Since none of these words will be considered new (because they all appeared in the old data), every entry for this is saved as a factor, with the level being "not new".

```
allcounts <- full_join(oldcounts, newcounts, by="word")
```

This code joins the old and new counts by the word column.

```
colnames(allcounts)[which(names(allcounts)=="n")] <- paste0("n_", month_ch,
year_ch)</pre>
```

Merging them left the count column for the current month named n. We can change this by looking for which column name is named "n", and then we replace that with "n\_" and then the month and year characters we extracted at the very beginning of the code.

```
allcounts$new <- ifelse(is.na(allcounts$new)==TRUE, "new", "not new")</pre>
```

The new column was not present in the newcounts data frame, so all words that were not present in the old data frame that *are* present now (i.e., the new words!) are registered as NA values. This code looks for NA values in the allcounts\$new column and replaces them with the word "new"; otherwise, it is kept as being called "not new".

```
allcounts[is.na(allcounts)] <- 0
```

Any other NA values are blank counts. These can simply be replaced with the number zero.

```
allcounts$length <- stri_length(allcounts$word)
newabrv <- c(allcounts[which(allcounts$new=="new" &
allcounts$length<4),"word"]$word)</pre>
```

We only want to include words that are short (i.e., under four words), as the reports can get *very* long if we don't. The first line here creates a new variable that simply calculates the length of the word. The second line creates a list: It takes data from allcounts, but only words that are new and where the length is under four words. We then extract just what is in the word column. This is the list of new abbreviations.

Now how do we get the ID and context to match it?

```
report <- data_frame()
for (i in 1:length(newabrv)) {
   report <- bind_rows(report, unique(newword[which(newword$word %in%
newabrv[i]),c("id","word","textraw")]))
}</pre>
```

The first line creates an empty data frame that we will fill with the results of our report. The next three lines are a for loop. For every abbreviation in the list of new abbreviations, we go back to the newword data and look for the row in which the new abbreviation appears. We return the id, word, and textraw columns from these rows, and we bind them to the report data frame. This loops through for every abbreviation.

```
output <- list(report, subset(allcounts, select=-c(new,length)))
names(output) <- c("report", "allcounts")
return(output)
}</pre>
```

Here is where we organize and return the output. The first line here takes the report and allcounts data (removing the new and length columns), and it puts them into a list named output. The next line names these two elements of the output: "report" and "allcounts". Next, we return the output, and the closed curly bracket indicates that we are done specifying our function!