

# Common text mining visuals

TEXT MINING WITH BAG-OF-WORDS IN R

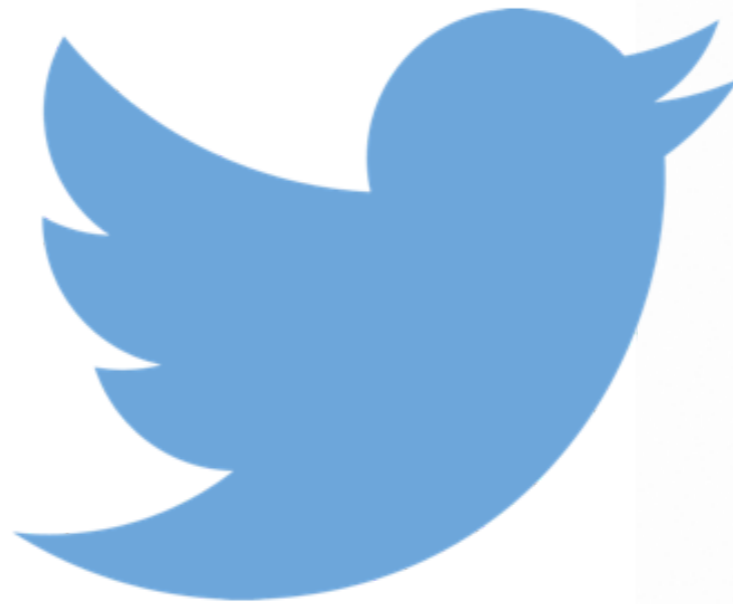


**Ted Kwartler**  
Instructor

# Why make visuals?

- Good visuals lead to quick conclusions
- The brain efficiently processes visual information

# Setting the scene



# Setting the scene

*Term Document Matrix (TDM)*

	Tweet1	Tweet2	Tweet3	...	Tweet_N
Term1	0	0	0	0	0
Term2	1	1	0	0	0
Term3	1	0	0	2	0
...	0	0	3	1	1
Term_N	0	0	1	1	0

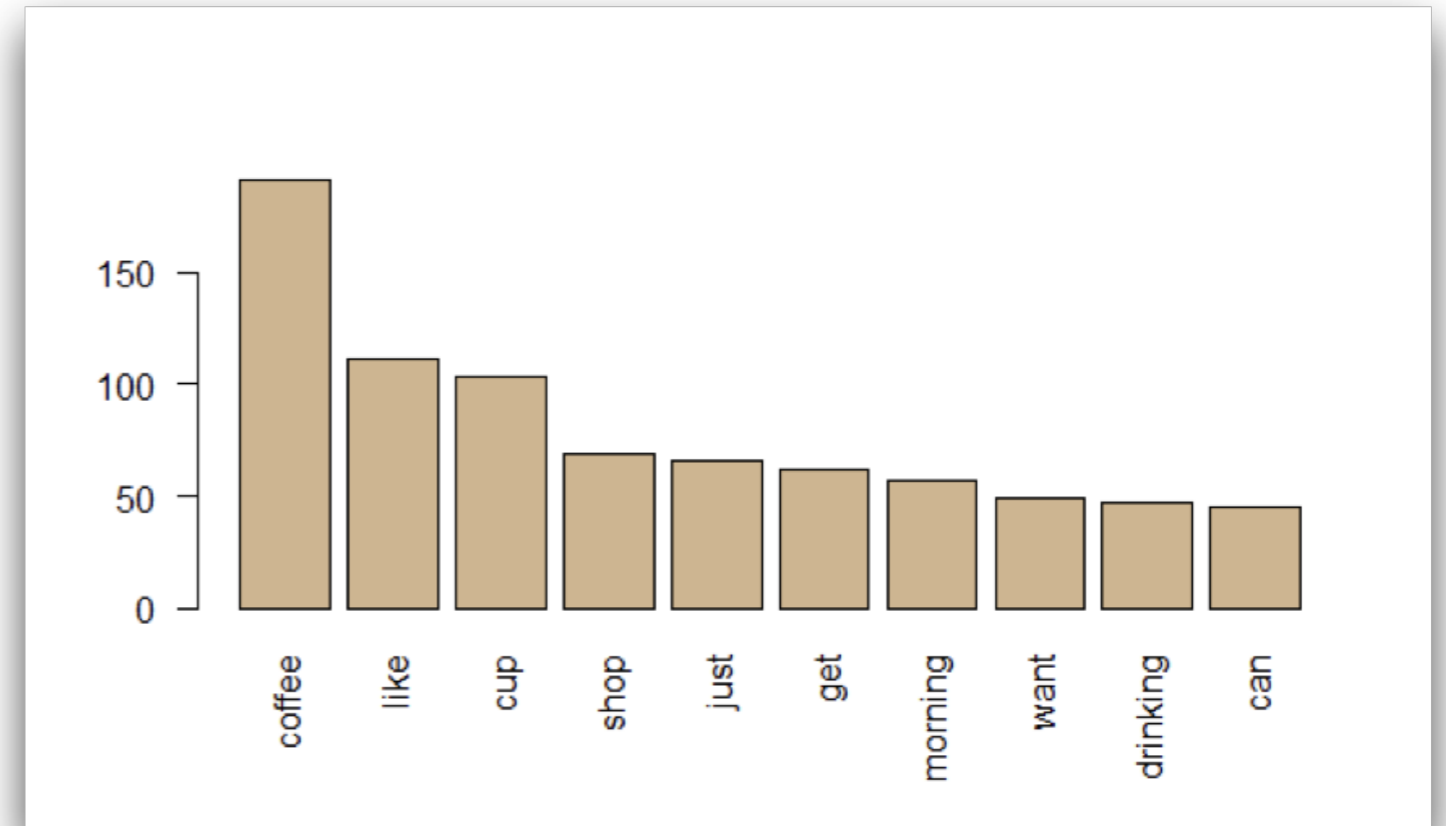


*Summed vector*

Sum
0
2
3
5
2

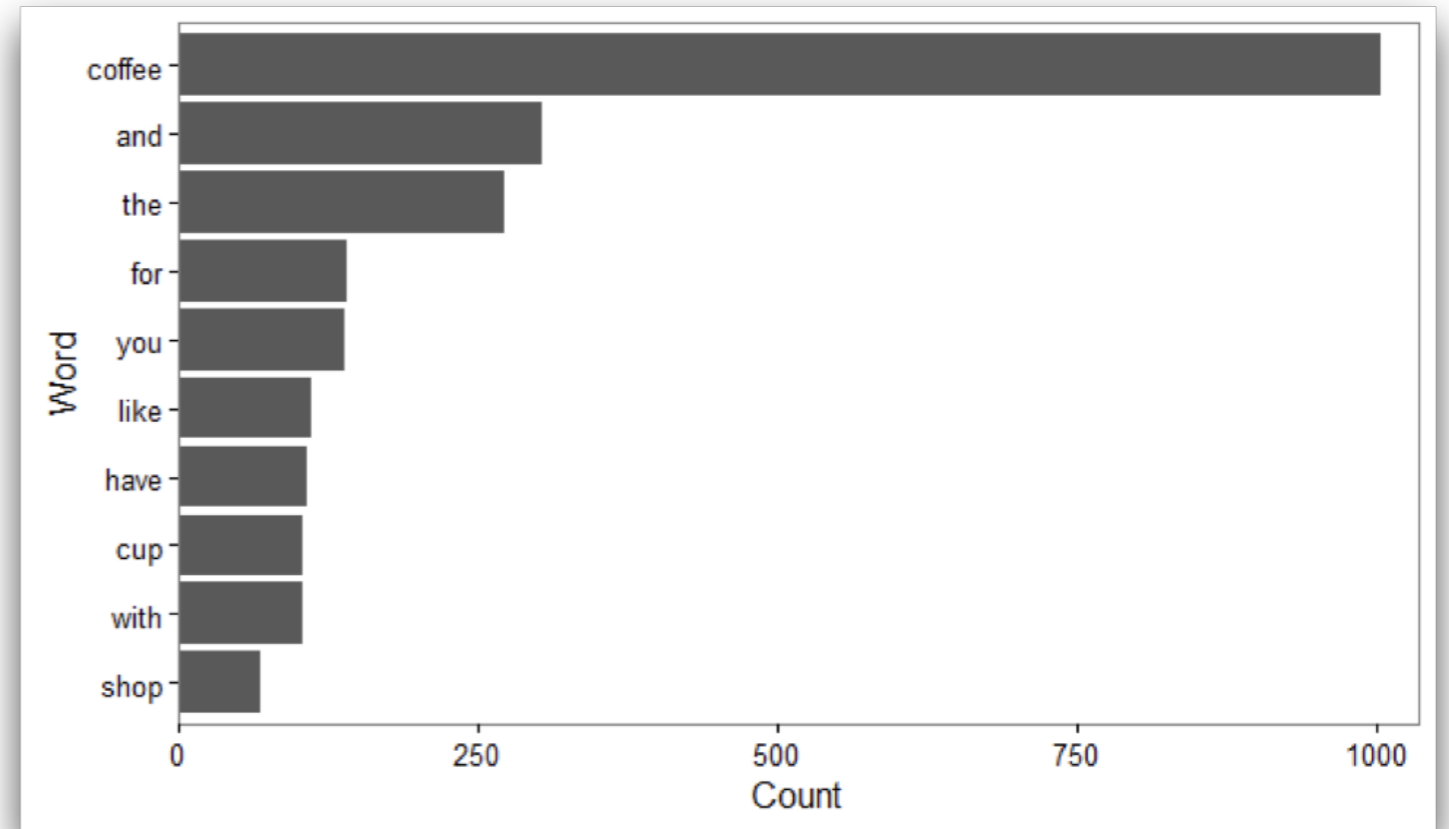
# Term frequency plots with tm

```
# Convert TDM to matrix
coffee_m <- as.matrix(coffee_tdm)
# Sum rows and sort by frequency
term_frequency <- rowSums(coffee_m)
term_frequency <- sort(term_frequency,
                      decreasing = TRUE)
# Create a barplot
barplot(term_frequency[1:10],
        col = "tan",
        las = 2)
```



# Term frequency plots with qdap

```
# Load qdap package
library(qdap)
# Find term frequencies
frequency <- freq_terms(
  tweets$text,
  top = 10,
  at.least = 3,
  stopwords = "Top200Words"
)
# Plot term frequencies
plot(frequency)
```



# Let's practice!

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# Intro to word clouds

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# A simple word cloud

```
# Convert TDM to matrix
chardonnay_tdm <- TermDocumentMatrix(clean_chardonnay)
chardonnay_m <- as.matrix(chardonnay_tdm)

# Sum rows and sort by frequency
term_frequency <- rowSums(chardonnay_m)
word_freqs <- data.frame(term = names(term_frequency),
                          num = term_frequency)

# Make word cloud
wordcloud(word_freqs$term, word_freqs$num,
          max.words = 100, colors = "red")
```



# The impact of stop words

```
clean_corpus <- function(corpus){  
  corpus <- tm_map(corpus, removePunctuation)  
  corpus <- tm_map(corpus, stripWhitespace)  
  corpus <- tm_map(corpus, removeNumbers)  
  corpus <- tm_map(corpus,  
                    content_transformer(tolower))  
  corpus <- tm_map(corpus, removeWords,  
                    c(stopwords("en"), "amp"))  
  return(corpus)  
}
```



# Removing uninformative words

```
clean_corpus <- function(corpus){  
  corpus <- tm_map(corpus, removePunctuation)  
  corpus <- tm_map(corpus, stripWhitespace)  
  corpus <- tm_map(corpus, removeNumbers)  
  corpus <- tm_map(corpus,  
                    content_transformer(tolower))  
  corpus <- tm_map(corpus, removeWords,  
                    c(stopwords("en"), "amp",  
                      "chardonnay", "wine", "glass"))  
  return(corpus)  
}
```



# Let's practice!

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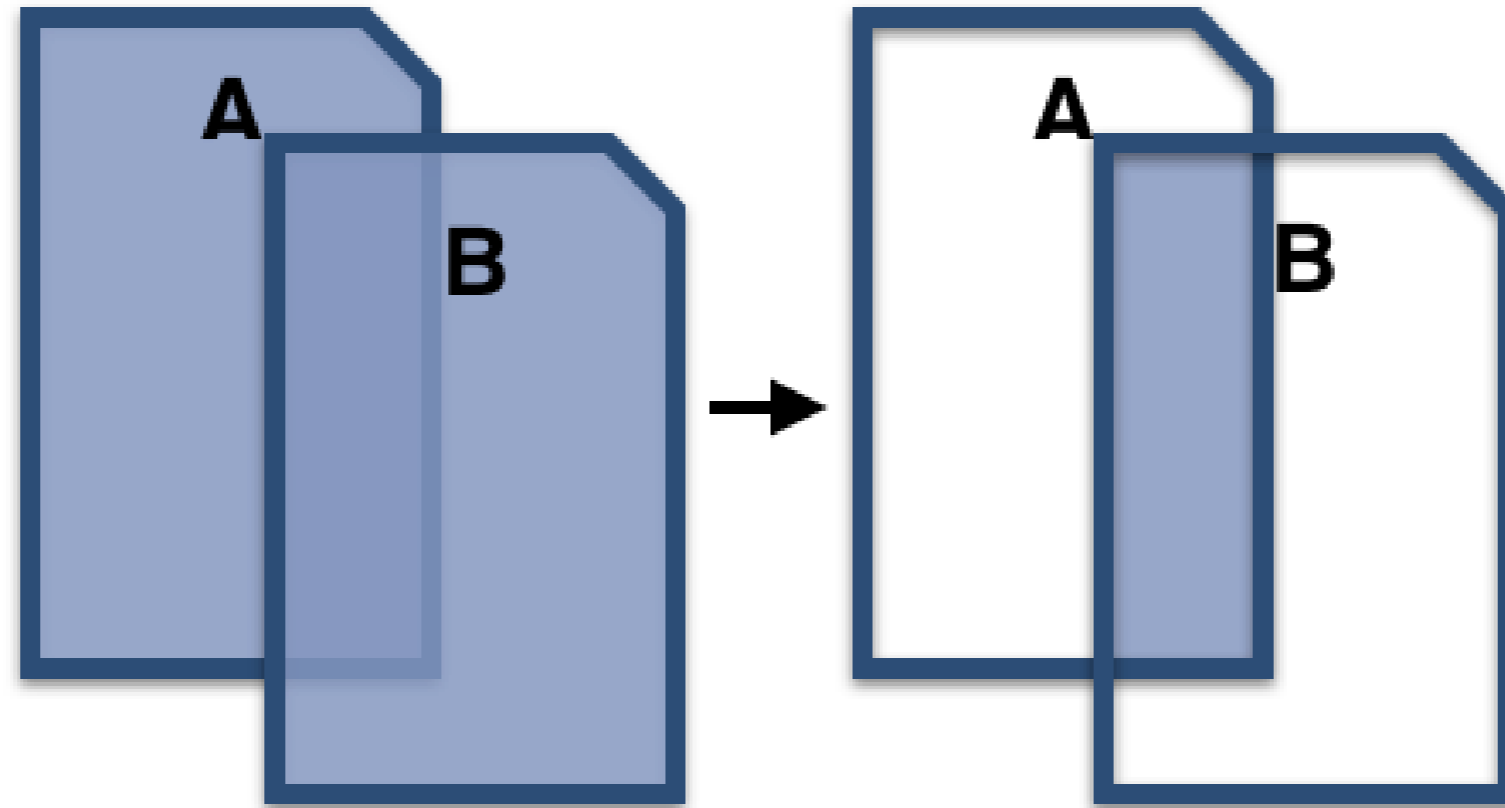
# Other word clouds and word networks

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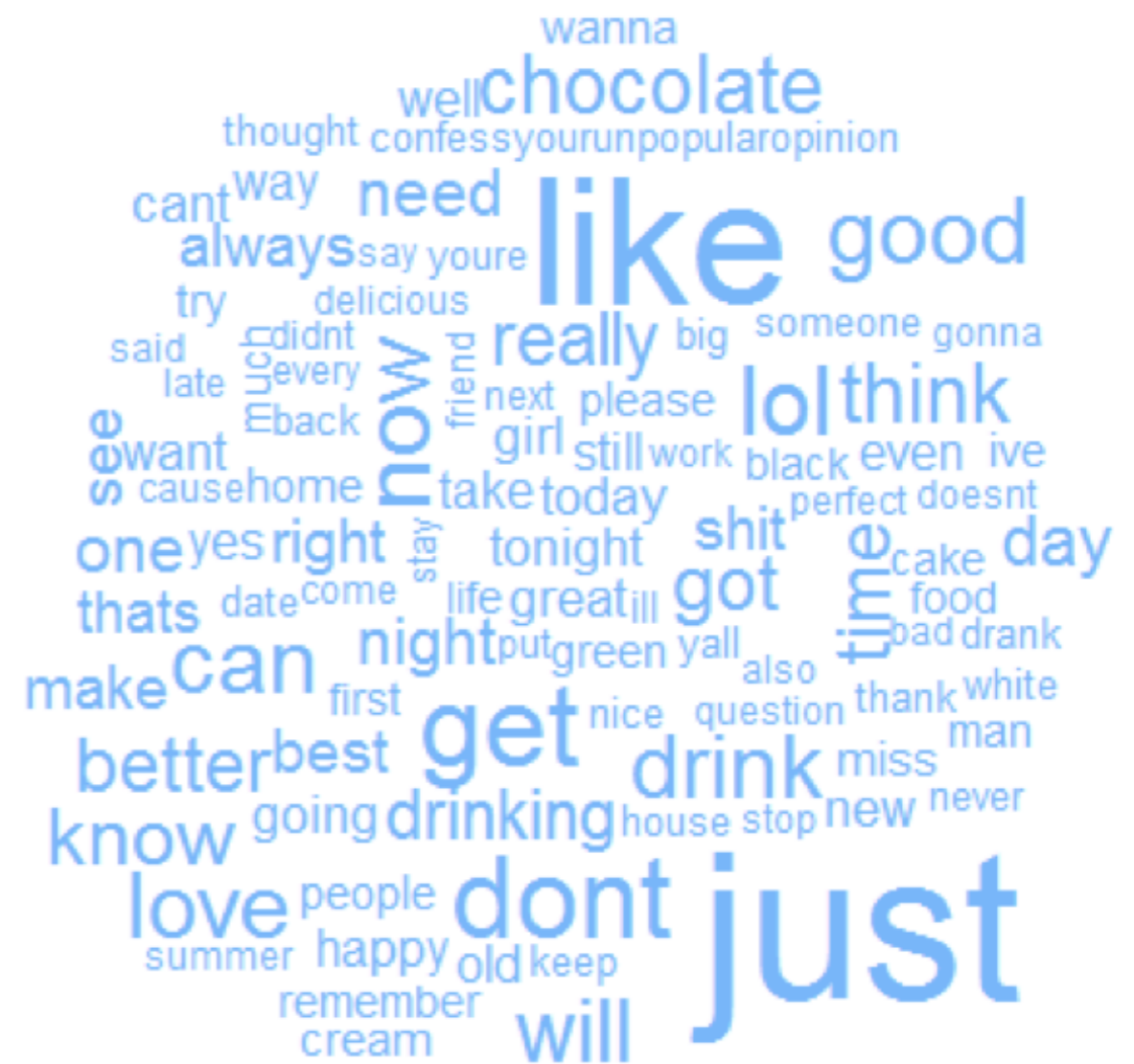
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# Commonality clouds

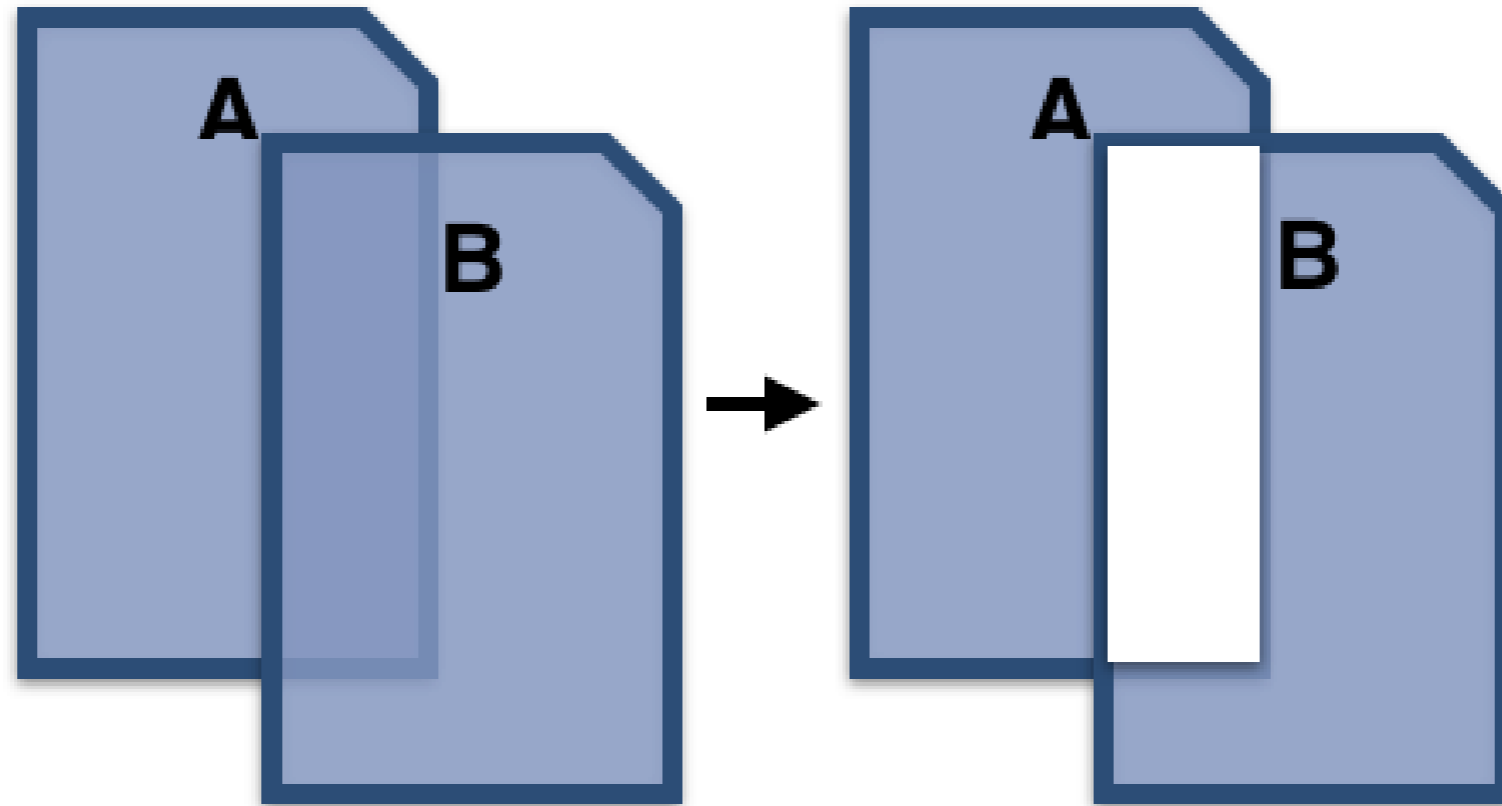


# Commonality clouds

```
# Combine both corpora: all_tweets
all_coffee <- paste(coffee_tweets$text,
                    collapse = "")
all_chardonnay <- paste(chardonnay_tweets$text,
                       collapse = "")
all_tweets <- c(all_coffee, all_chardonnay)
# Clean all_tweets
all_tweets <- VectorSource(all_tweets)
all_corpus <- VCorpus(all_tweets)
all_clean <- clean_corpus(all_corpus)
all_dm <- TermDocumentMatrix(all_clean)
all_m <- as.matrix(all_tdm)
# Make commonality cloud
commonality.cloud(all_m, colors = "steelblue1",
                  max.words = 100)
```



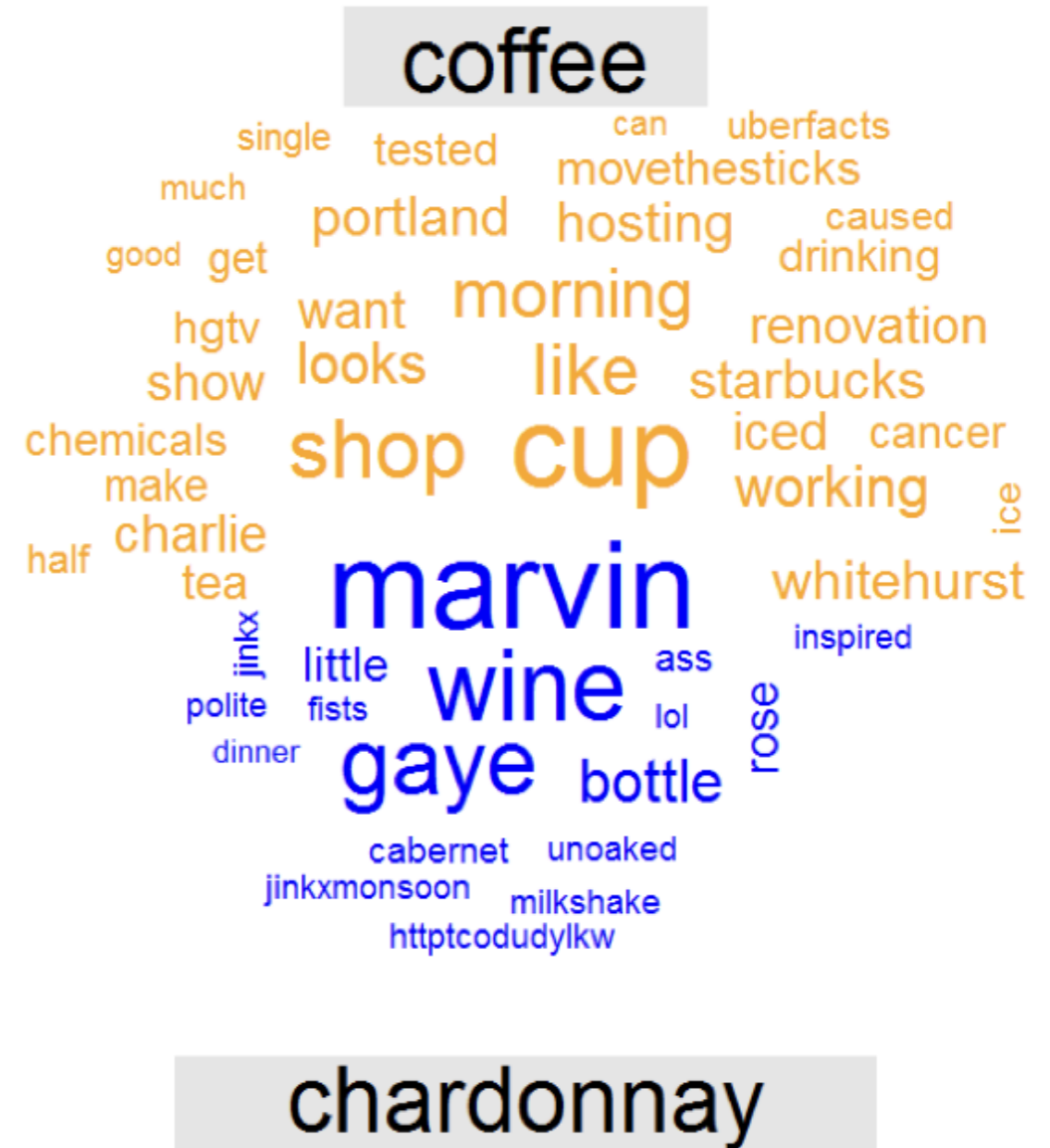
# Comparison clouds





# Comparison clouds

```
# Combine both corpora: all_tweets
all_coffee <- paste(coffee_tweets$text,
                   collapse = "")
all_chardonnay <- paste(chardonnay_tweets$text,
                      collapse = "")
all_tweets <- c(all_coffee, all_chardonnay)
# Clean all_tweets
all_tweets <- VectorSource(all_tweets)
all_corpus <- VCorpus(all_tweets)
all_clean <- clean_corpus(all_corpus)
all_tdm <- TermDocumentMatrix(all_clean)
colnames(all_tdm) <- c("coffee", "chardonnay")
all_m <- as.matrix(all_tdm)
# Make comparison cloud
comparison.cloud(all_m,
                colors = c("orange", "blue"), max.words = 50)
```

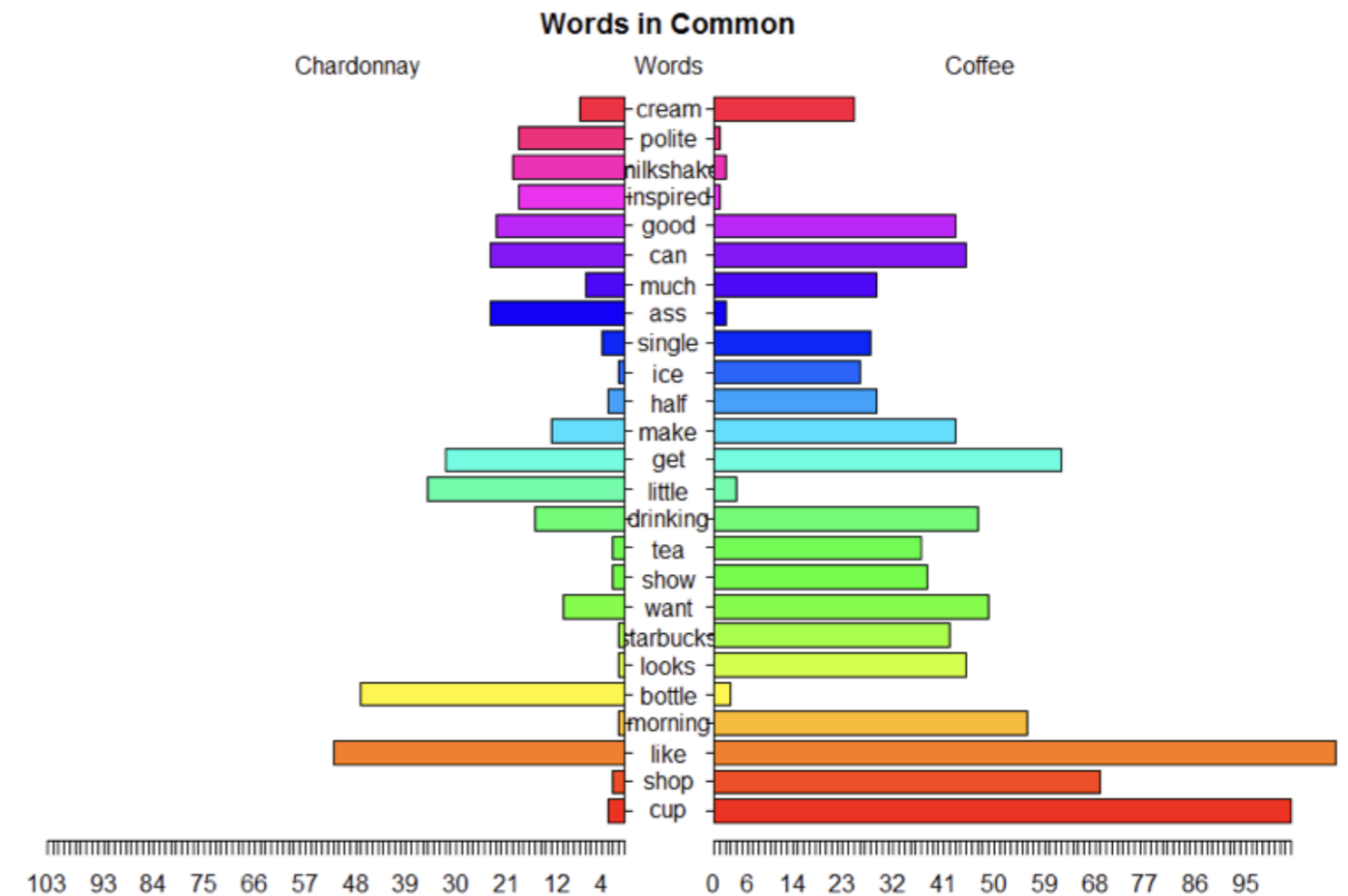


# Pyramid plots

```
# Identify terms shared by both documents
common_words <- subset(
  all_tdm_m,
  all_tdm_m[, 1] > 0 & all_tdm_m[, 2] > 0
)
# Find most commonly shared words
difference <- abs(common_words[, 1] - common_words[, 2])
common_words <- cbind(common_words, difference)
common_words <- common_words[order(common_words[, 3],
                                   decreasing = TRUE), ]
top25_df <- data.frame(x = common_words[1:25, 1],
                      y = common_words[1:25, 2],
                      labels = rownames(common_words[1:25, ]))
```

# Pyramid plots

```
# Make pyramid plot
pyramid.plot(top25_df$x, top25_df$y,
             labels = top25_df$labels,
             main = "Words in Common",
             gap = 8, laxly = NULL,
             raxlab = NULL, unit = NULL,
             top.labels = c("Chardonnay",
                           "Words",
                           "Coffee"))
)
```

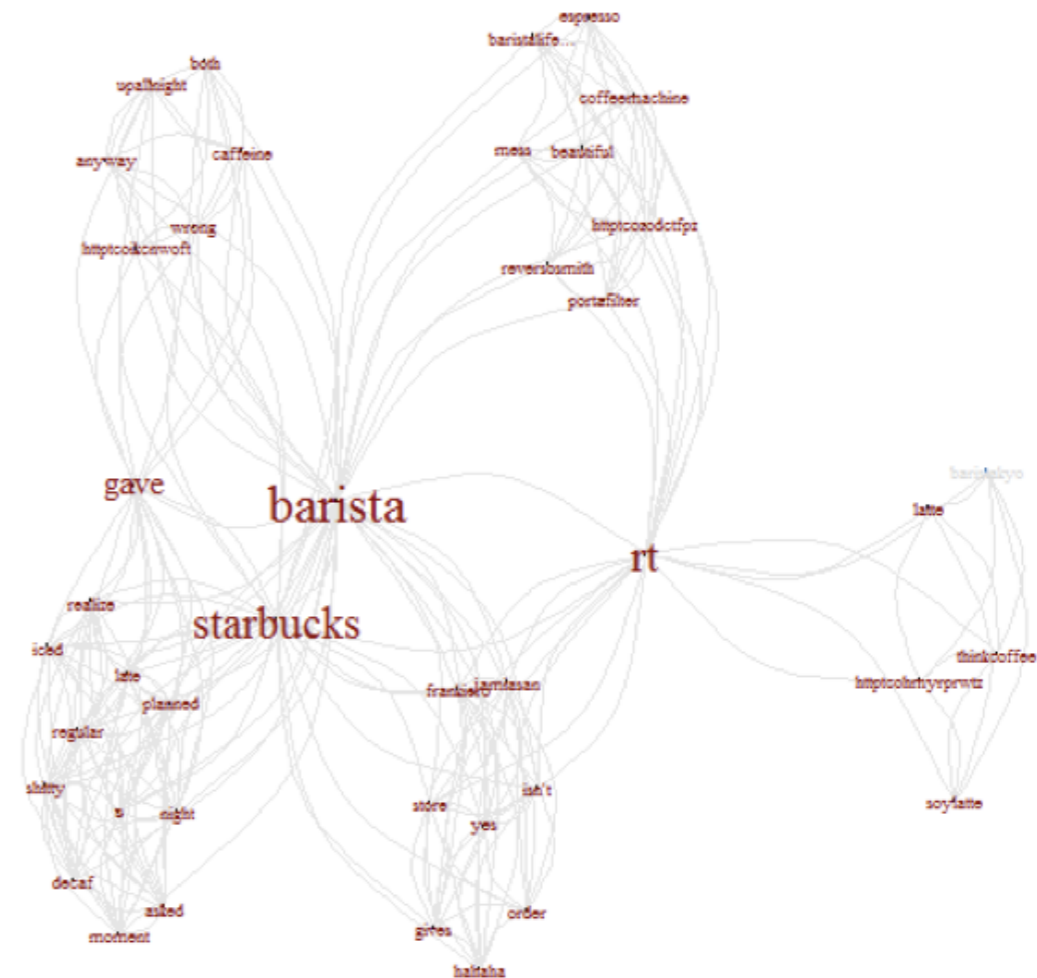


# Word networks

```
# Create word network
word_associate(coffee_tweets$text,
  match.string = c("barista"),
  stopwords = c(Top200Words, "coffee", "amp"),
  network.plot = TRUE,
  cloud.colors = c("gray85", "darkred"))

# Add title
title(main = "Barista Coffee Tweet Associations")
```

Coffee Tweets Associated with Barista



# Let's practice!

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