Camille Text Mining

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```
Book: Camille (LA DAME AUX CAMILIAS)
Author: Alexandre Dumas
Cami <- gutenberg_download(1608)</pre>
## Determining mirror for Project Gutenberg from http://www.gutenberg.org/robot/harvest
## Using mirror http://aleph.gutenberg.org
newCami <- Cami %>%
  mutate(linenumber = row_number()) %>%
  select(-gutenberg_id) %>%
  mutate(chapter = cumsum(str_detect(text,
                                     regex("^chapter [\\divxlc]",
                                           ignore_case = TRUE))))
tidy_Cami <- newCami %>%
  unnest_tokens(word, text) %>%
  anti_join(stop_words)
## Joining, by = "word"
tidy_Cami %>%
  count(word, sort = TRUE)
## # A tibble: 4,134 x 2
##
     word
                    n
##
      <chr>
                <int>
## 1 marguerite 453
## 2 love
                  220
## 3 day
                  175
## 4 time
                  149
## 5 woman
                 147
## 6 prudence
                 143
## 7 life
                  117
## 8 father
                  113
                 106
## 9 armand
## 10 paris
                  87
## # ... with 4,124 more rows
```

Sentiment Analysis

nrc

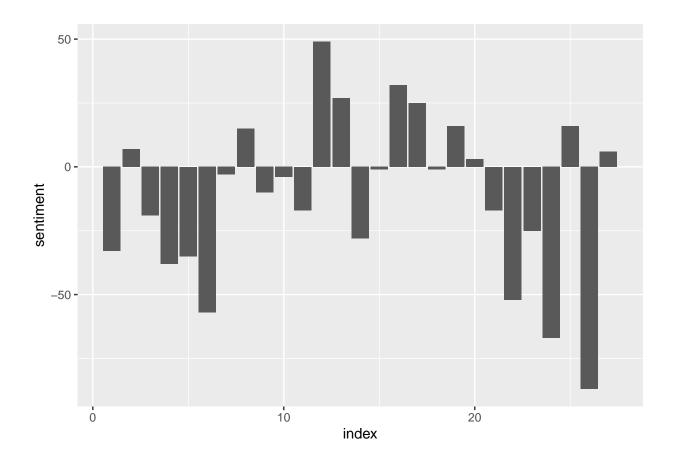
```
#textdata::lexicon nrc(delete = TRUE)
#nrc <- textdata::lexicon_nrc()</pre>
nrc_joy <- get_sentiments("nrc") %>%
 filter(sentiment == "joy")
tidy_Cami %>%
 inner_join(nrc_joy) %>%
 count(word, sort = TRUE)
## Joining, by = "word"
## # A tibble: 200 x 2
##
     word n
     <chr> <int>
##
## 1 love
              220
## 2 friend 71
## 3 money 50
## 4 happy 49
             38
## 5 pay
## 6 lover 37
## 7 child 34
## 8 found
           34
             34
## 9 god
## 10 true
              29
## # ... with 190 more rows
```

bing

```
get_sentiments("bing")
```

```
## # A tibble: 6,786 x 2
## word sentiment
## <chr> <chr> <chr>
## 1 2-faces negative
## 2 abnormal negative
## 3 abolish negative
## 4 abominable negative
## 5 abominably negative
## 6 abominate negative
## 7 abomination negative
## 8 abort negative
## 9 aborted negative
## 10 aborts negative
## # ... with 6,776 more rows
```

```
bing_neg <- get_sentiments("bing") %>%
 filter(sentiment == "negative")
tidy_Cami %>%
 inner_join(bing_neg) %>%
count(word, sort = TRUE)
## Joining, by = "word"
## # A tibble: 567 x 2
##
     word n
     <chr> <int>
##
## 1 poor 54
## 2 mistress 47
              33
## 3 dead
## 4 rue
                33
              32
## 5 doubt
## 6 sad
               29
## 7 death
               28
## 8 die
                28
## 9 spite
                26
## 10 fear
## # ... with 557 more rows
Cami_sentiment <- tidy_Cami %>%
 inner_join(get_sentiments("bing")) %>%
 count(index = chapter, sentiment) %>%
 pivot_wider(names_from = sentiment, values_from = n, values_fill = 0) %%
 mutate(sentiment = positive - negative)
## Joining, by = "word"
ggplot(Cami_sentiment, aes(index, sentiment)) +
 geom_col(show.legend = FALSE)
```

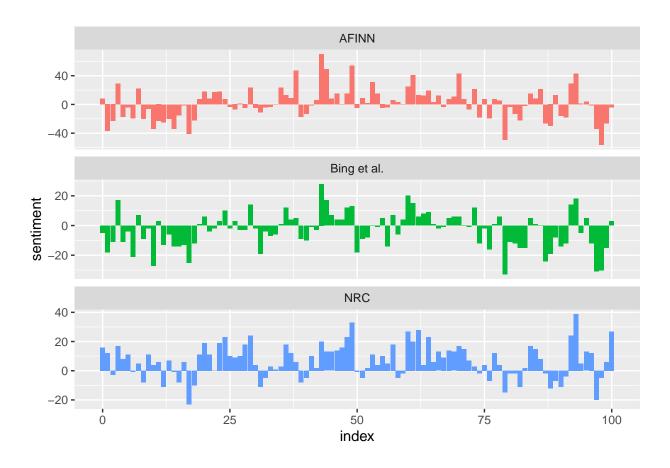


afinn

```
afinn <- tidy_Cami %>%
  inner_join(get_sentiments("afinn")) %>%
  group_by(index = linenumber %/% 80) %>%
  summarise(sentiment = sum(value)) %>%
  mutate(method = "AFINN")
```

Joining, by = "word"

compare the three sentiment dictionaries



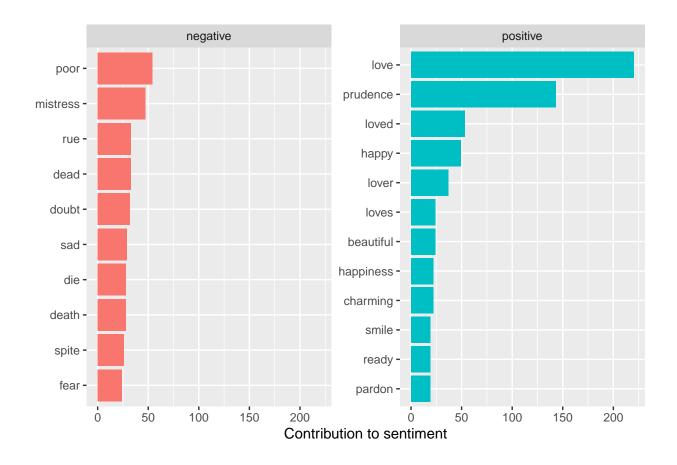
later

```
get_sentiments("nrc") %>%
filter(sentiment %in% c("positive", "negative")) %>%
count(sentiment)
```

```
## # A tibble: 2 x 2
##
    sentiment
            <int>
##
    <chr>
## 1 negative 3318
## 2 positive
              2308
get_sentiments("bing") %>%
 count(sentiment)
## # A tibble: 2 x 2
    sentiment
##
    <chr>
              <int>
## 1 negative
              4781
## 2 positive
               2005
```

2.4 Most common positive and negative words

```
bing_word_counts <- tidy_Cami %>%
 inner_join(get_sentiments("bing")) %>%
 count(word, sentiment, sort = TRUE) %>%
 ungroup()
## Joining, by = "word"
bing_word_counts
## # A tibble: 920 x 3
##
     word
              sentiment
                            n
##
     <chr>
              <chr>
                        <int>
## 1 love positive
                          220
                         143
## 2 prudence positive
## 3 poor negative
                          54
## 4 loved positive
                          53
## 5 happy positive
                         49
                         47
## 6 mistress negative
                          37
## 7 lover positive
## 8 dead
                          33
              negative
## 9 rue
              negative
                           33
## 10 doubt
              negative
                           32
## # ... with 910 more rows
bing_word_counts %>%
 group_by(sentiment) %>%
 slice_max(n, n = 10) %>%
 ungroup() %>%
 mutate(word = reorder(word, n)) %>%
 ggplot(aes(n, word, fill = sentiment)) +
 geom_col(show.legend = FALSE) +
 facet_wrap(~sentiment, scales = "free_y") +
 labs(x = "Contribution to sentiment",
      y = NULL
```



2.5 Wordclouds

```
tidy_Cami %>%
    anti_join(stop_words) %>%
    count(word) %>%
    with(wordcloud(word, n, max.words = 100))

## Joining, by = "word"

## Warning in strwidth(words[i], cex = size[i], ...): conversion failure on 'don't'

## warning in strwidth(words[i], cex = size[i], ...): conversion failure on 'don't'

## in 'mbcsToSbcs': dot substituted for <80>

## Warning in strwidth(words[i], cex = size[i], ...): conversion failure on 'don't'

## in 'mbcsToSbcs': dot substituted for <90>

## Warning in text.default(x1, y1, words[i], cex = size[i], offset = 0, srt =

## rotWord *: conversion failure on 'don't' in 'mbcsToSbcs': dot substituted for

## <e2>
```

```
## Warning in text.default(x1, y1, words[i], cex = size[i], offset = 0, srt =
## rotWord * : conversion failure on 'don't' in 'mbcsToSbcs': dot substituted for
## <80>
## Warning in text.default(x1, y1, words[i], cex = size[i], offset = 0, srt =
## rotWord * : conversion failure on 'don't' in 'mbcsToSbcs': dot substituted for
## <99>
## Warning in text.default(x1, y1, words[i], cex = size[i], offset = 0, srt =
## rotWord * : font metrics unknown for Unicode character U+2019
## Warning in strwidth(words[i], cex = size[i], ...): conversion failure on
## 'marguerite's' in 'mbcsToSbcs': dot substituted for <e2>
## Warning in strwidth(words[i], cex = size[i], ...): conversion failure on
## 'marguerite's' in 'mbcsToSbcs': dot substituted for <80>
## Warning in strwidth(words[i], cex = size[i], ...): conversion failure on
## 'marguerite's' in 'mbcsToSbcs': dot substituted for <99>
## Warning in text.default(x1, y1, words[i], cex = size[i], offset = 0, srt
## = rotWord * : conversion failure on 'marguerite's' in 'mbcsToSbcs': dot
## substituted for <e2>
## Warning in text.default(x1, y1, words[i], cex = size[i], offset = 0, srt
## = rotWord * : conversion failure on 'marguerite's' in 'mbcsToSbcs': dot
## substituted for <80>
## Warning in text.default(x1, y1, words[i], cex = size[i], offset = 0, srt
## = rotWord * : conversion failure on 'marguerite's' in 'mbcsToSbcs': dot
## substituted for <99>
## Warning in text.default(x1, y1, words[i], cex = size[i], offset = 0, srt =
## rotWord * : font metrics unknown for Unicode character U+2019
```

marguerite

```
country paris count woman loved continued word past friend tears told death mistress carriage adjusted to the poor people as ir pay child rue poor people as i
```

Joining, by = "word"

negative



positive

Find the number of negative words in each chapter and divide by the total words in each chapter. Which chapter has the highest proportion of negative words?

```
wordcounts <- tidy_Cami %>%
  group_by(chapter) %>%
  summarize(words = n())

tidy_Cami %>%
  semi_join(bing_neg) %>%
  group_by(chapter) %>%
  summarize(negativewords = n()) %>%
  summarize(negativewords, by = c("chapter")) %>%
  mutate(ratio = negativewords/words) %>%
  filter(chapter != 0) %>%
  slice_max(ratio, n = 1) %>%
  ungroup()
```

```
## Joining, by = "word"

## # A tibble: 1 x 4

## chapter negativewords words ratio
## <int> <int> <int> <dbl>
## 1 24 139 906 0.153
```

Chapter 24 has the highest proportion of negative words.