

# text mining

Tong Sun

11/29/2021

## Task one and two

### Download and Tidy

Jane Eyre is divided into 38 chapters. It was originally published in three volumes in the 19th century, comprising chapters 1 to 15, 16 to 27, and 28 to 38. The novel is a first-person narrative from the perspective of the title character. It has five distinct stages: Jane's childhood at Gateshead Hall, where she is emotionally and physically abused by her aunt and cousins; her education at Lowood School, where she gains friends and role models but suffers privations and oppression; her time as governess at Thornfield Hall, where she falls in love with her mysterious employer, Edward Fairfax Rochester; her time in the Moor House, during which her earnest but cold clergyman cousin, St. John Rivers, proposes to her; and ultimately her reunion with, and marriage to, her beloved Rochester. Throughout these sections, it provides perspectives on a number of important social issues and ideas, many of which are critical of the status quo.

```
## Determining mirror for Project Gutenberg from http://www.gutenberg.org/robot/harvest
```

```
## Using mirror http://aleph.gutenberg.org
```

```
## # A tibble: 188,399 x 4
##   gutenberg_id linenumber chapter word
##   <int>         <int>    <int> <chr>
## 1         1260         1      0 jane
## 2         1260         1      0 eyre
## 3         1260         2      0 an
## 4         1260         2      0 autobiography
## 5         1260         4      0 by
## 6         1260         4      0 charlotte
## 7         1260         4      0 brontë
## 8         1260         6      0 _illustrated
## 9         1260         6      0 by
## 10        1260         6      0 f
## # ... with 188,389 more rows
```

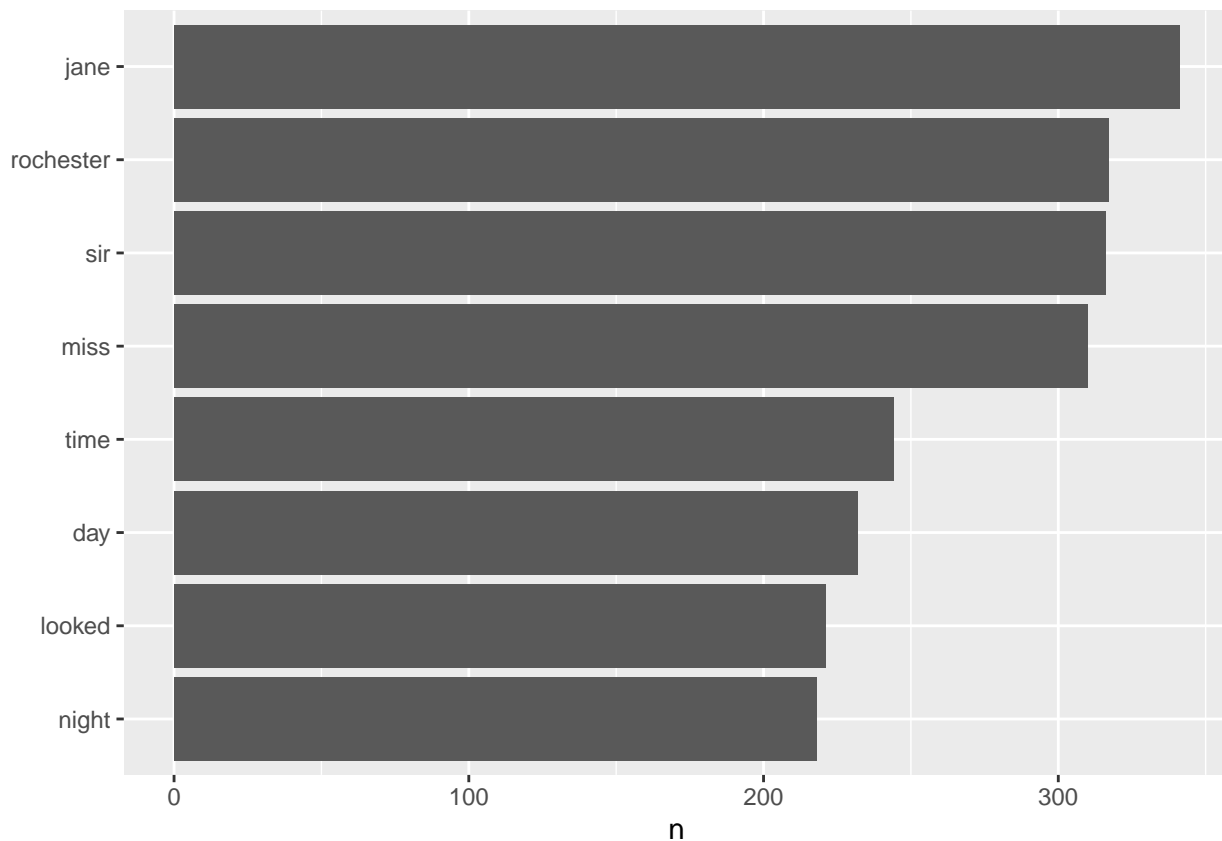
```
## [1] 39
```

```
## # A tibble: 39 x 2
##   chapter    n
##   <int> <int>
## 1      0  959
```

```
## 2      1 1954
## 3      2 2771
## 4      3 3230
## 5      4 5866
## 6      5 5046
## 7      6 2923
## 8      7 3604
## 9      8 3015
## 10     9 3276
## # ... with 29 more rows
```

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

```
## # A tibble: 12,309 x 2
##   word      n
##   <chr>   <int>
## 1 jane     341
## 2 rochester 317
## 3 sir      316
## 4 miss     310
## 5 time     244
## 6 day      232
## 7 looked   221
## 8 night    218
## 9 eyes     187
## 10 john    184
## # ... with 12,299 more rows
```



## Sentiment analysis with tidy data

In the previous part, I explored what changed the original book into the tidy text format and showed how this format can be used to approach questions about word frequency. Next I would like to address the topic of sentiment analysis. When human readers approach a text, we use our understanding of the emotional intent of words to infer whether a section of text is positive or negative. So here I use the tools of text mining to approach the emotional content of text.

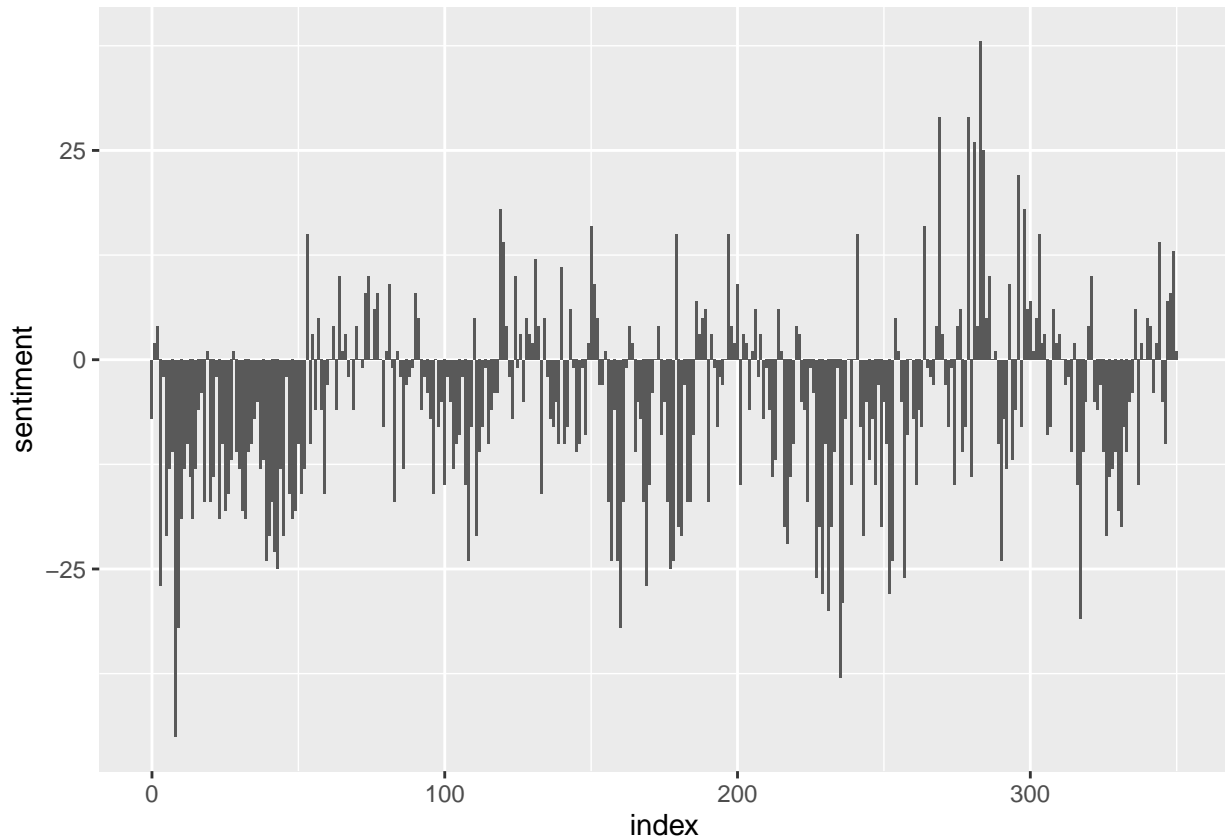
```
## Joining, by = "word"
```

```
## # A tibble: 404 x 2
##   word      n
##   <chr>  <int>
## 1 love    151
## 2 found   126
## 3 god     96
## 4 child   88
## 5 hope    75
## 6 feeling 67
## 7 happy   54
## 8 marry   49
## 9 smile   47
## 10 mother 45
## # ... with 394 more rows
```

I found that mostly positive, happy words about love, hope and happy here.

In addition, I want to examine how sentiment changes throughout each novel. First, I find a sentiment score for each word using the Bing lexicon and `'inner_join()'`. Next I count up how many positive and negative words there are in defined sections of the book and define an “index” here to keep track of where we are in the narrative; this index (using integer division) counts up sections of 60 lines of text. In addition, I use `'pivot_wider()'` so that we can have negative and positive sentiment in separate columns and finally calculate a net sentiment (positive - negative).

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

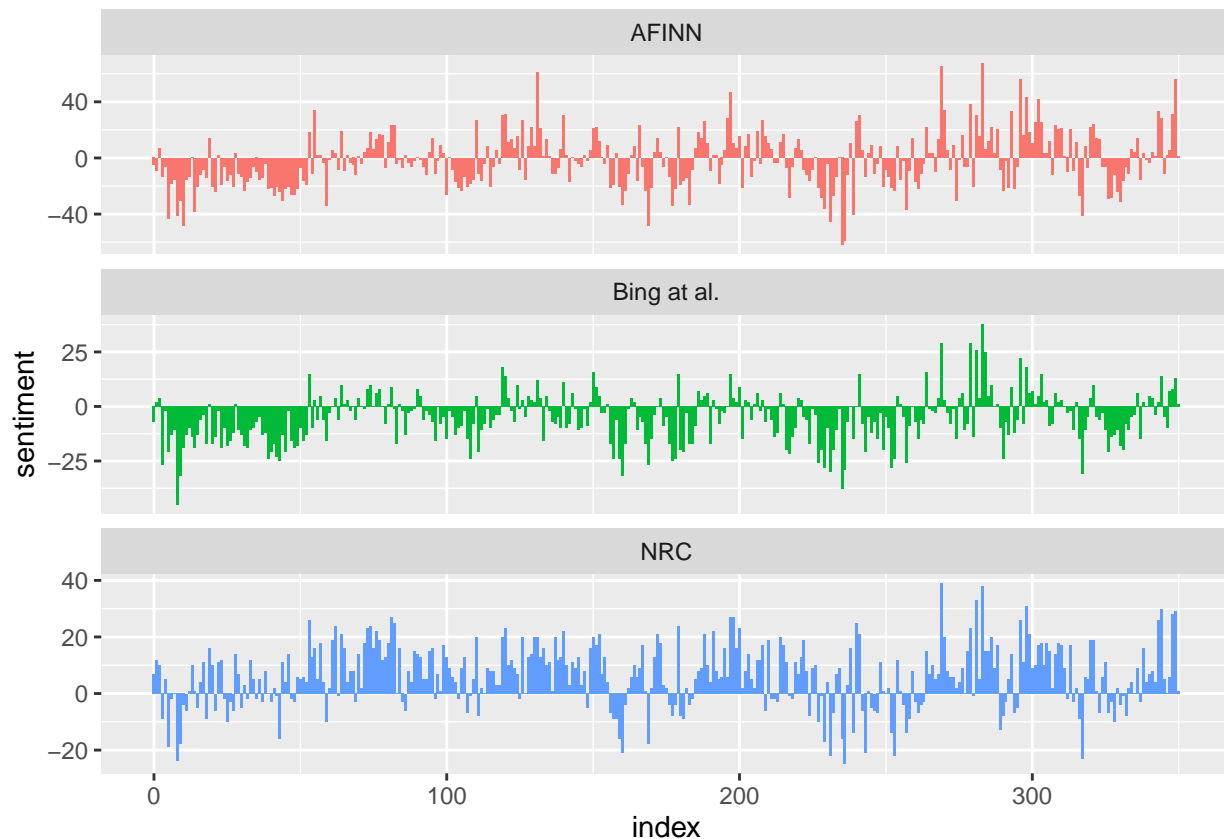


We can see how the plot of each novel changes toward more positive or negative sentiment over the trajectory of the story. From the plot above, I find that there are more negative sentiments during the early index and more positive sentiments during the late index. I think this fits the story. During Eyre’s childhood, she was abused by her aunt, Sarah Reed, she did not have a happy childhood in her early years. Therefore the sentiment is negative. But when she met Rochester as written in the late plots, she was married with him and had a sweet life. So there are more positive ones here.

## Compare the three sentiment dictionaries

- AFINN: the AFINN lexicon measures sentiment with a numeric score between -5 and 5;
- Bing: the Bing lexicon categorizes words in a binary fashion into positive and negative categories;
- nrc: the nrc lexicon categorizes words in a binary fashion (“yes”/“no”) into categories of positive, negative, anger, anticipation, disgust, fear, joy, sadness, surprise and trust.

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



The three different lexicons for calculating sentiment give results that are different in an absolute sense but have similar relative trajectories through the novel. We can see similar dips and peaks in sentiment at about the same places in the novel, but the absolute values are significantly different. The AFINN lexicon gives the largest absolute values, with high positive values. The Bing lexicon has lower absolute values. The NRC lexicon are shifted higher relative to the other two, labeling the text more positively.

```
## # A tibble: 2 x 2
##   sentiment      n
##   <chr>      <int>
## 1 negative   3318
## 2 positive   2308
```

```
## # A tibble: 2 x 2
##   sentiment      n
##   <chr>      <int>
## 1 negative   4781
## 2 positive   2005
```

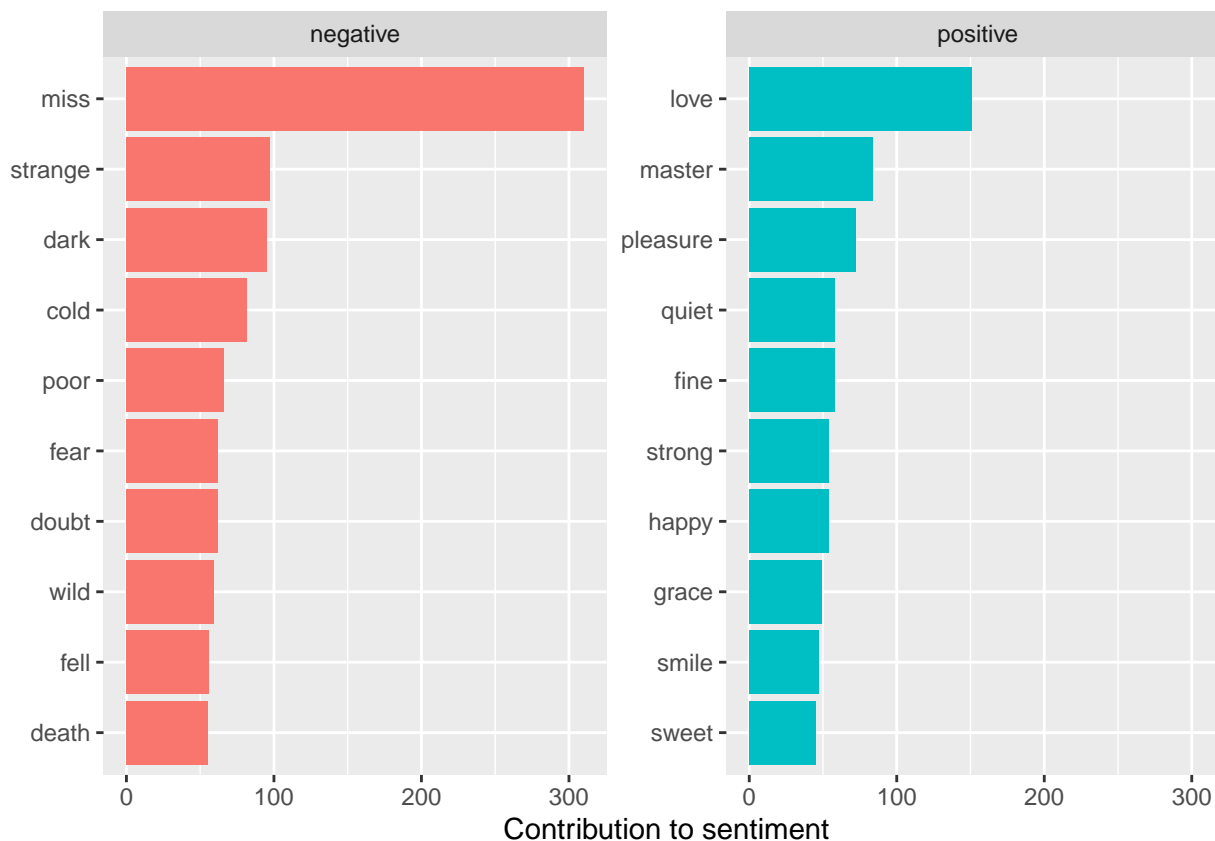
Both lexicons have more negative than positive words, but the ratio of negative to positive words is higher in the Bing lexicon than the NRC lexicon. This will contribute to the effect we see in the plot above, as will any systematic difference in word matches, e.g. if the negative words in the NRC lexicon do not match the words

that Jane Austen uses very well. Whatever the source of these differences, we see similar relative trajectories across the narrative arc, with similar changes in slope, but marked differences in absolute sentiment from lexicon to lexicon. This is all important context to keep in mind when choosing a sentiment lexicon for analysis.

**Most common positive and negative words – One advantage of having the data frame with both sentiment and word is that we can analyze word counts that contribute to each sentiment**

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

## # A tibble: 2,326 x 3
##   word      sentiment      n
##   <chr>    <chr>    <int>
## 1 miss      negative    310
## 2 love      positive    151
## 3 strange   negative     97
## 4 dark      negative     95
## 5 master     positive     84
## 6 cold      negative     82
## 7 pleasure  positive     72
## 8 poor      negative     66
## 9 doubt     negative     62
## 10 fear     negative     62
## # ... with 2,316 more rows
```



The plot above shows words that contribute to positive and negative sentiment in Eyre.

## Wordclouds





The size of a word's text is in proportion to its frequency within its sentiment. We can use this visualization to see the most important positive and negative words. Here the most important positive word is 'positive' and the most important negative word is 'miss'.