

Chapter 3

John Peach

3/23/2021

```
book_words <- austen_books() %>%
  unnest_tokens(word, text) %>%
  count(book, word, sort = TRUE) %>%
  ungroup()

total_words <- book_words %>%
  group_by(book) %>%
  summarise(total = sum(n))

## `summarise()` ungrouping output (override with `.groups` argument)
book_words <- left_join(book_words, total_words)

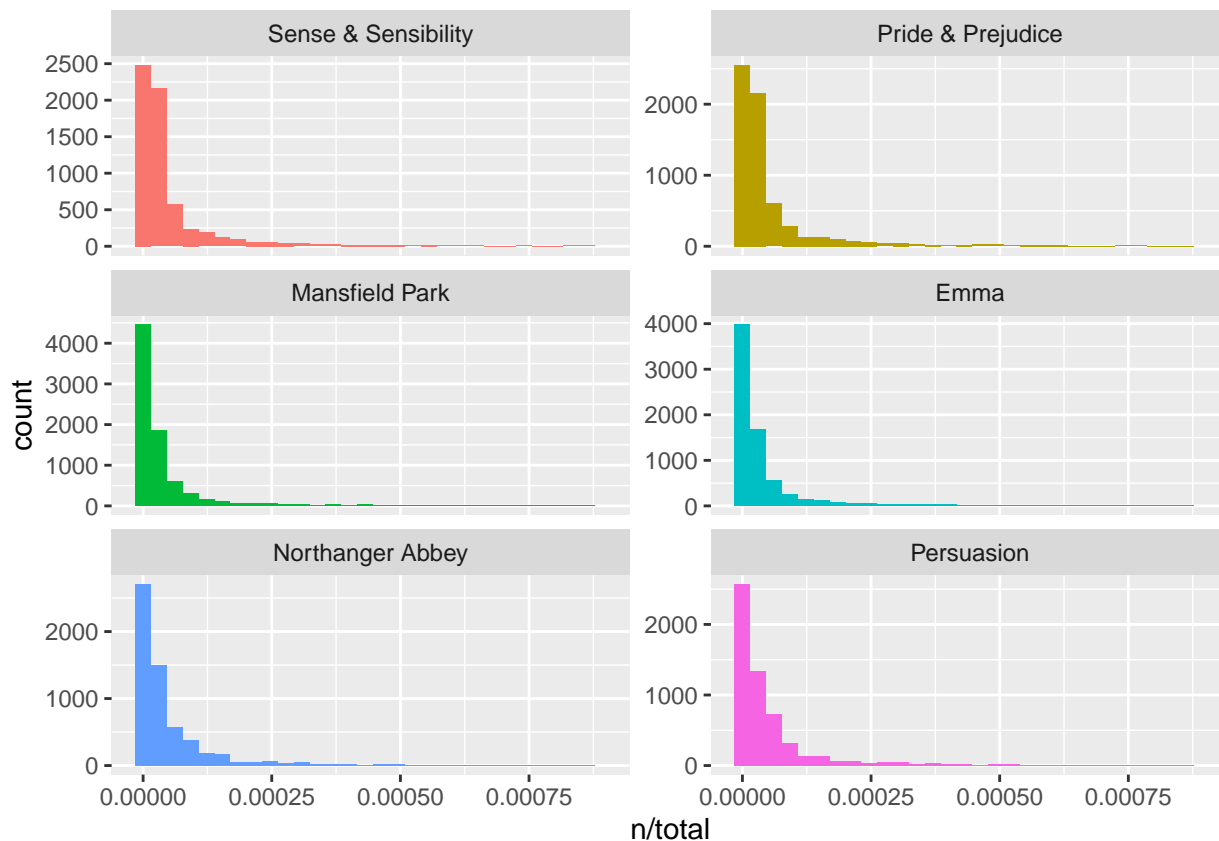
## Joining, by = "book"
book_words

## # A tibble: 40,379 x 4
##   book          word      n total
##   <fct>         <chr> <int> <int>
## 1 Mansfield Park the      6206 160460
## 2 Mansfield Park to       5475 160460
## 3 Mansfield Park and      5438 160460
## 4 Emma          to       5239 160996
## 5 Emma          the      5201 160996
## 6 Emma          and      4896 160996
## 7 Mansfield Park of       4778 160460
## 8 Pride & Prejudice the     4331 122204
## 9 Emma          of       4291 160996
## 10 Pride & Prejudice to     4162 122204
## # ... with 40,369 more rows

library(ggplot2)

ggplot(book_words, aes(n/total, fill = book)) +
  geom_histogram(show.legend = FALSE) +
  xlim(NA, 0.0009) +
  facet_wrap(~book, ncol = 2, scales = 'free_y')

## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## Warning: Removed 896 rows containing non-finite values (stat_bin).
## Warning: Removed 6 rows containing missing values (geom_bar).
```

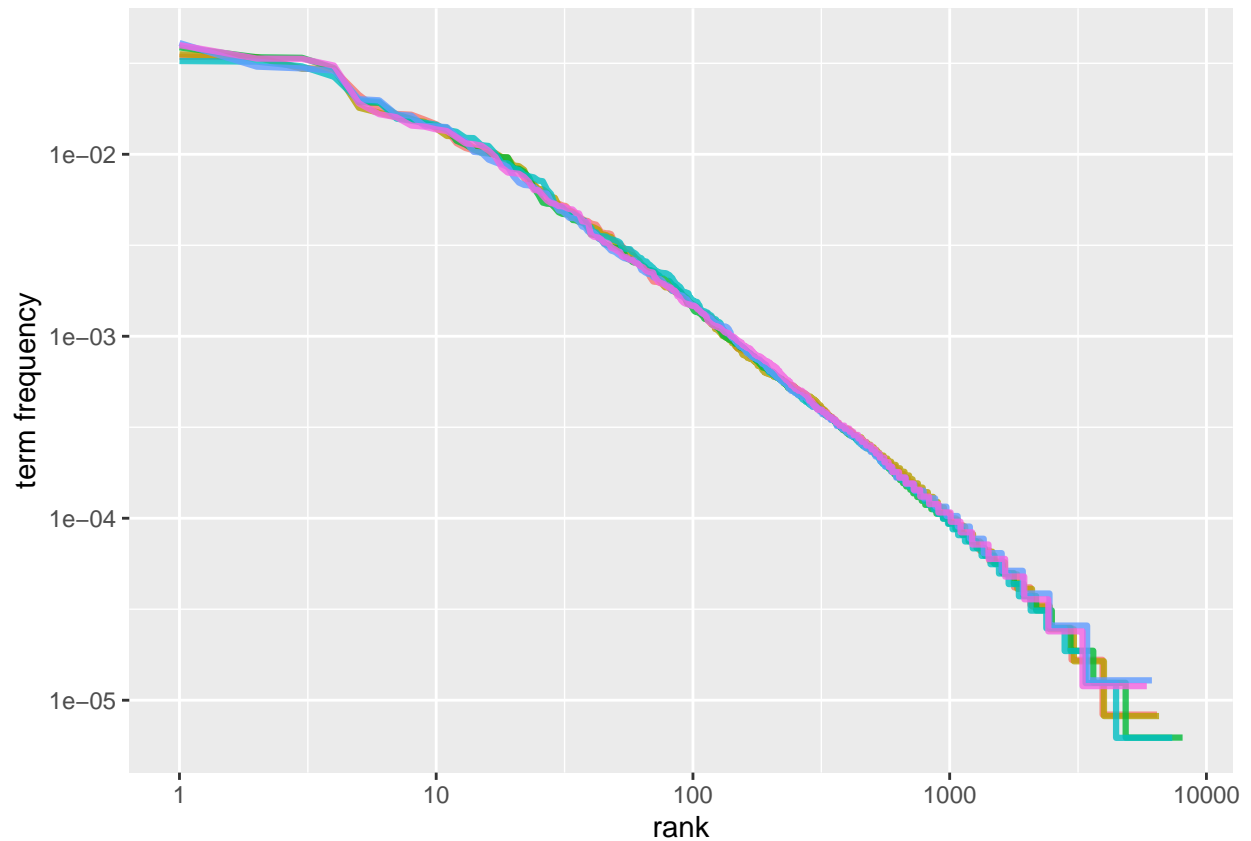


```
freq_by_rank <- book_words %>%
  group_by(book) %>%
  mutate(rank = row_number(), `term frequency` = n/total)
```

```
freq_by_rank
```

```
## # A tibble: 40,379 x 6
## # Groups:   book [6]
##   book      word      n  total  rank `term frequency`
##   <fct>    <chr> <int> <int> <int>      <dbl>
## 1 Mansfield Park the      6206 160460 1         0.0387
## 2 Mansfield Park to       5475 160460 2         0.0341
## 3 Mansfield Park and      5438 160460 3         0.0339
## 4 Emma      to       5239 160996 1         0.0325
## 5 Emma      the      5201 160996 2         0.0323
## 6 Emma      and      4896 160996 3         0.0304
## 7 Mansfield Park of       4778 160460 4         0.0298
## 8 Pride & Prejudice the     4331 122204 1         0.0354
## 9 Emma      of       4291 160996 4         0.0267
## 10 Pride & Prejudice to     4162 122204 2         0.0341
## # ... with 40,369 more rows
```

```
freq_by_rank %>%
  ggplot(aes(rank, `term frequency`, color = book)) +
  geom_line(size = 1.1, alpha = 0.8, show.legend = FALSE) +
  scale_x_log10() +
  scale_y_log10()
```

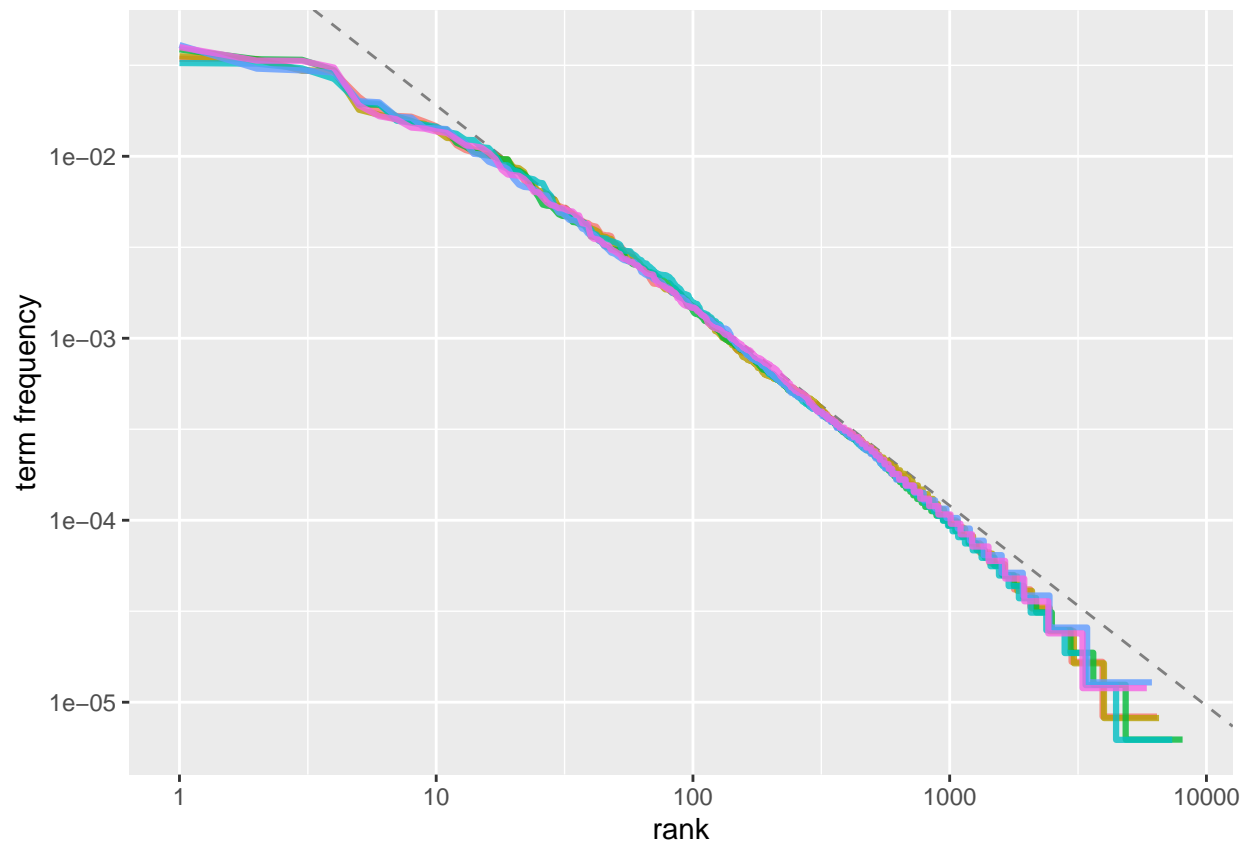


```
rank_subset <- freq_by_rank %>%
  dplyr::filter(rank < 500, rank > 10)

lm(log10(`term frequency`) ~ log10(rank), data = rank_subset)

##
## Call:
## lm(formula = log10(`term frequency`) ~ log10(rank), data = rank_subset)
##
## Coefficients:
## (Intercept) log10(rank)
##      -0.6226      -1.1125

freq_by_rank %>%
  ggplot(aes(rank, `term frequency`, color = book)) +
    geom_abline(intercept = -0.62, slope = -1.1, color = 'gray50', linetype = 2) +
    geom_line(size = 1.1, alpha = 0.8, show.legend = FALSE) +
    scale_x_log10() +
    scale_y_log10()
```



```
book_words <- book_words %>%
  bind_tf_idf(word, book, n)
```

```
book_words
```

```
## # A tibble: 40,379 x 7
```

	book	word	n	total	tf	idf	tf_idf
	<fct>	<chr>	<int>	<int>	<dbl>	<dbl>	<dbl>
##	1	Mansfield Park	the	6206	160460	0.0387	0
##	2	Mansfield Park	to	5475	160460	0.0341	0
##	3	Mansfield Park	and	5438	160460	0.0339	0
##	4	Emma	to	5239	160996	0.0325	0
##	5	Emma	the	5201	160996	0.0323	0
##	6	Emma	and	4896	160996	0.0304	0
##	7	Mansfield Park	of	4778	160460	0.0298	0
##	8	Pride & Prejudice	the	4331	122204	0.0354	0
##	9	Emma	of	4291	160996	0.0267	0
##	10	Pride & Prejudice	to	4162	122204	0.0341	0

... with 40,369 more rows

```
book_words %>%
  select(-total) %>%
  arrange(desc(tf_idf))
```

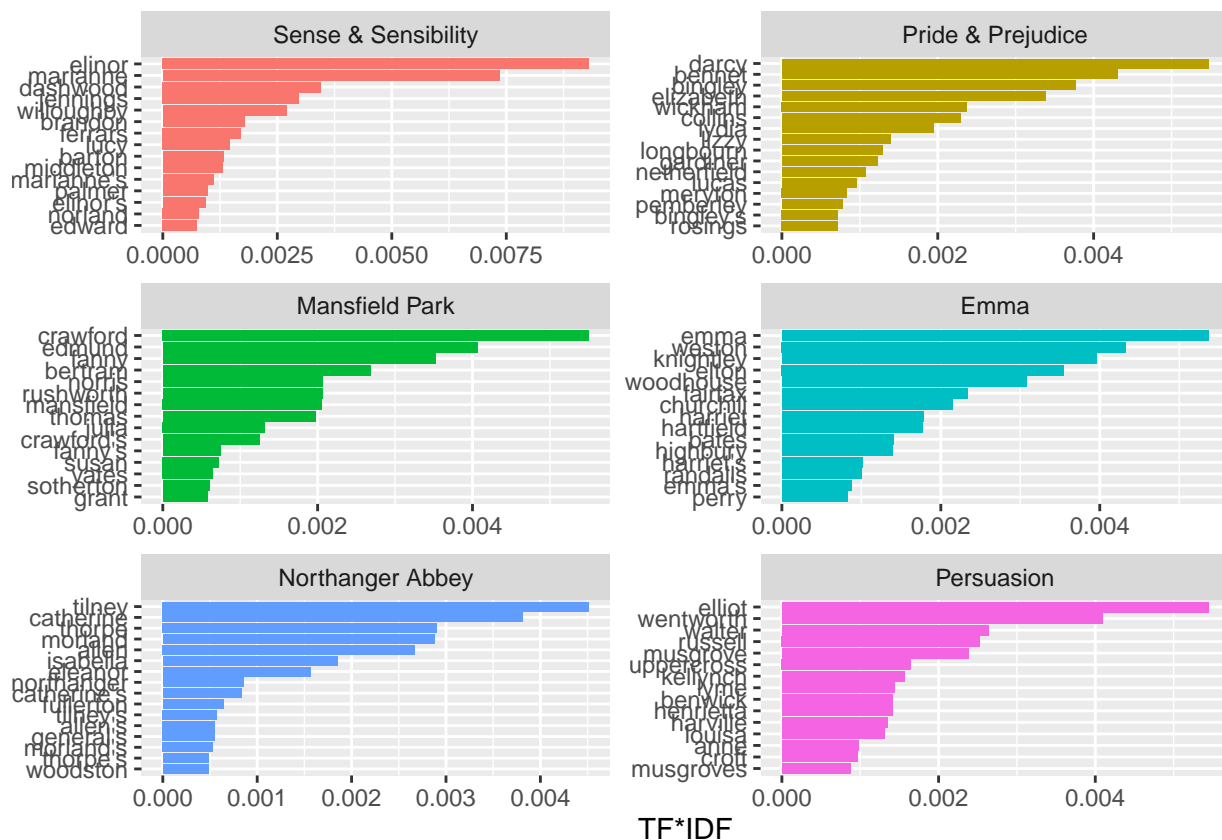
```
## # A tibble: 40,379 x 6
```

	book	word	n	tf	idf	tf_idf
	<fct>	<chr>	<int>	<dbl>	<dbl>	<dbl>
##	1	Sense & Sensibility	elinor	623	0.00519	1.79

```
## 2 Sense & Sensibility marianne 492 0.00410 1.79 0.00735
## 3 Mansfield Park crawford 493 0.00307 1.79 0.00551
## 4 Pride & Prejudice darcy 373 0.00305 1.79 0.00547
## 5 Persuasion elliot 254 0.00304 1.79 0.00544
## 6 Emma emma 786 0.00488 1.10 0.00536
## 7 Northanger Abbey tilney 196 0.00252 1.79 0.00452
## 8 Emma weston 389 0.00242 1.79 0.00433
## 9 Pride & Prejudice bennet 294 0.00241 1.79 0.00431
## 10 Persuasion wentworth 191 0.00228 1.79 0.00409
## # ... with 40,369 more rows
```

```
book_words %>%
  arrange(desc(tf_idf)) %>%
  mutate(word = factor(word, levels = rev(unique(word)))) %>%
  group_by(book) %>%
  top_n(15) %>%
  ungroup %>%
  ggplot(aes(word, tf_idf, fill = book)) +
  geom_col(show.legend = FALSE) +
  labs(x = NULL, y = 'TF*IDF') +
  facet_wrap(~book, ncol = 2, scales = 'free') +
  coord_flip()
```

Selecting by tf_idf



```
library(gutenbergr)
physics <- gutenberg_download(c(37729, 14725, 13476, 5001),
                              meta_fields = 'author',
```

```
mirror = 'http://eremita.di.uminho.pt/gutenberg/')
```

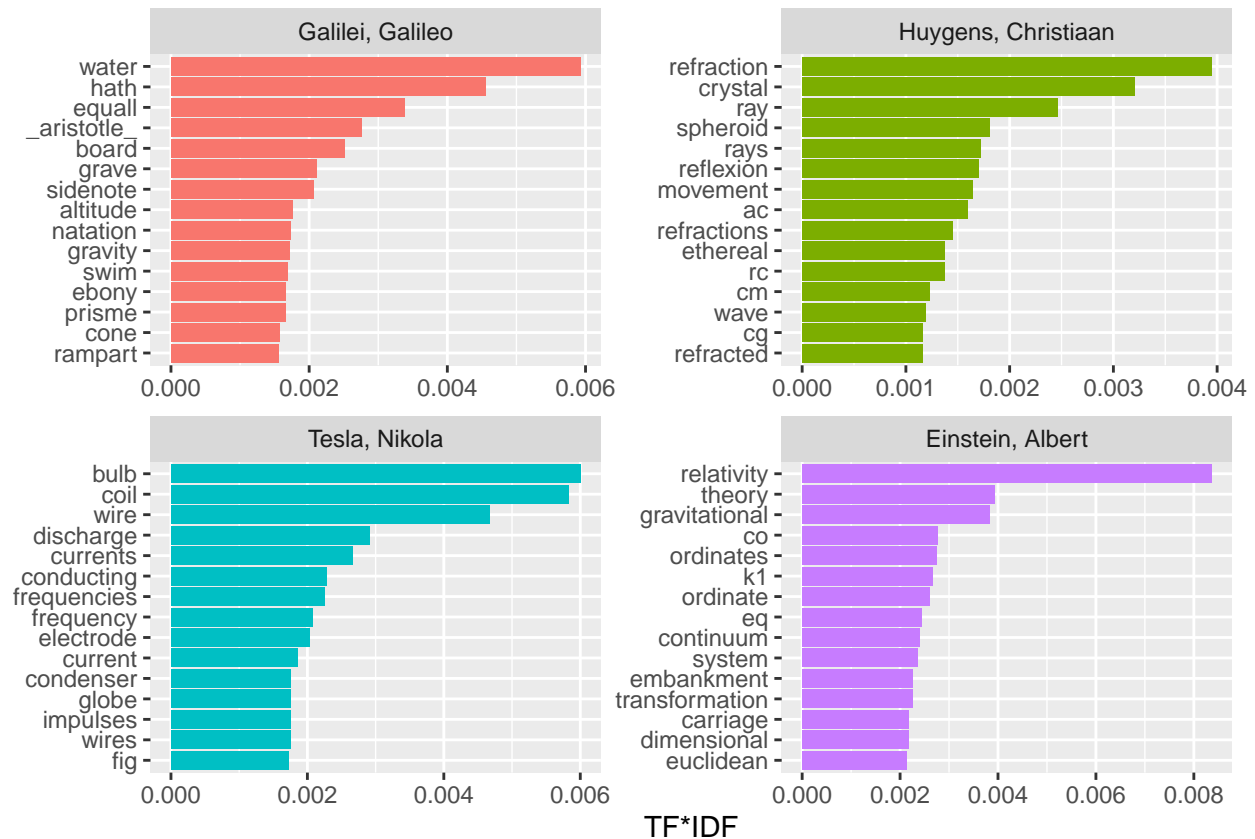
```
physics_word <- physics %>%  
  unnest_tokens(word, text) %>%  
  count(author, word, sort = TRUE) %>%  
  ungroup()
```

```
physics_word
```

```
## # A tibble: 12,592 x 3  
##   author      word      n  
##   <chr>      <chr> <int>  
## 1 Galilei, Galileo the    3760  
## 2 Tesla, Nikola the    3604  
## 3 Huygens, Christiaan the    3553  
## 4 Einstein, Albert the    2994  
## 5 Galilei, Galileo of     2049  
## 6 Einstein, Albert of     2030  
## 7 Tesla, Nikola of     1737  
## 8 Huygens, Christiaan of     1708  
## 9 Huygens, Christiaan to     1207  
## 10 Tesla, Nikola a       1176  
## # ... with 12,582 more rows
```

```
plot_physics <- physics_word %>%  
  bind_tf_idf(word, author, n) %>%  
  arrange(desc(tf_idf)) %>%  
  mutate(word = factor(word, levels = rev(unique(word)))) %>%  
  mutate(author = factor(author, levels = c('Galilei, Galileo',  
                                            'Huygens, Christiaan',  
                                            'Tesla, Nikola',  
                                            'Einstein, Albert')))
```

```
plot_physics %>%  
  group_by(author) %>%  
  top_n(15, tf_idf) %>%  
  ungroup() %>%  
  mutate(word = reorder(word, tf_idf)) %>%  
  ggplot(aes(word, tf_idf, fill = author)) +  
    geom_col(show.legend = FALSE) +  
    labs(x = NULL, y = 'TF*IDF') +  
    facet_wrap(~author, ncol = 2, scales = 'free') +  
    coord_flip()
```



```
library(stringr)
```

```
physics %>%
  dplyr::filter(str_detect(text, "eq\\\\")) %>%
  select(text)
```

```
## # A tibble: 55 x 1
##   text
##   <chr>
## 1 "eq. 1: file eq01.gif"
## 2 "eq. 2: file eq02.gif"
## 3 "eq. 3: file eq03.gif"
## 4 "eq. 4: file eq04.gif"
## 5 "eq. 05a: file eq05a.gif"
## 6 "eq. 05b: file eq05b.gif"
## 7 "the distance between the points being eq. 06 ."
## 8 "direction of its length with a velocity v is eq. 06 of a metre."
## 9 "velocity v=c we should have eq. 06a ,"
## 10 "the rod as judged from K1 would have been eq. 06 ;"
## # ... with 45 more rows
```

```
physics %>%
  dplyr::filter(str_detect(text, "K1")) %>%
  select(text)
```

```
## # A tibble: 59 x 1
##   text
##   <chr>
```

```
## 1 to a second co-ordinate system K1 provided that the latter is
## 2 condition of uniform motion of translation. Relative to K1 the
## 3 tenet thus: If, relative to K, K1 is a uniformly moving co-ordinate
## 4 with respect to K1 according to exactly the same general laws as with
## 5 does not hold, then the Galileian co-ordinate systems K, K1, K2, etc.,
## 6 Relative to K1, the same event would be fixed in respect of space and
## 7 to K1, when the magnitudes x, y, z, t, of the same event with respect
## 8 of light (and of course for every ray) with respect to K and K1. For
## 9 reference-body K and for the reference-body K1. A light-signal is sent
## 10 immediately follows. If referred to the system K1, the propagation of
## # ... with 49 more rows
```

```
physics %>%
  dplyr::filter(str_detect(text, "AK")) %>%
  select(text)
```

```
## # A tibble: 34 x 1
##   text
##   <chr>
## 1 Now let us assume that the ray has come from A to C along AK, KC; the
## 2 be equal to the time along KMN. But the time along AK is longer than
## 3 that along AL: hence the time along AKN is longer than that along ABC.
## 4 And KC being longer than KN, the time along AKC will exceed, by as
## 5 line which is comprised between the perpendiculars AK, BL. Then it
## 6 ordinary refraction. Now it appears that AK and BL dip down toward the
## 7 side where the air is less easy to penetrate: for AK being longer than
## 8 than do AK, BL. And this suffices to show that the ray will continue
## 9 surface AB at the points AK_k_B. Then instead of the hemispherical
## 10 along AL, LB, and along AK, KB, are always represented by the line AH,
## # ... with 24 more rows
```

```
mystopwords <- data_frame(word = c('eq', 'co', 'rc', 'ac', 'ak', 'bn',
                                   'fig', 'file', 'cg', 'cb', 'cm'))
```

```
## Warning: `data_frame()` is deprecated as of tibble 1.1.0.
## Please use `tibble()` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_warnings()` to see where this warning was generated.
```

```
physics_word <- anti_join(physics_word, mystopwords, by = 'word')
plot_physics <- physics_word %>%
  bind_tf_idf(word, author, n) %>%
  arrange(desc(tf_idf)) %>%
  mutate(word = factor(word, levels = rev(unique(word)))) %>%
  group_by(author) %>%
  top_n(15, tf_idf) %>%
  ungroup %>%
  mutate(author = factor(author, levels = c('Galilei, Galileo',
                                             'Huygens, Christiaan',
                                             'Tesla, Nikola',
                                             'Einstein, Albert')))
```

```
ggplot(plot_physics, aes(word, tf_idf, fill = author)) +
  geom_col(show.legend = FALSE) +
  labs(x = NULL, y = "TF*IDF") +
  facet_wrap(~author, ncol = 2, scales = 'free') +
```


coord_flip()

