Intro to text analysis

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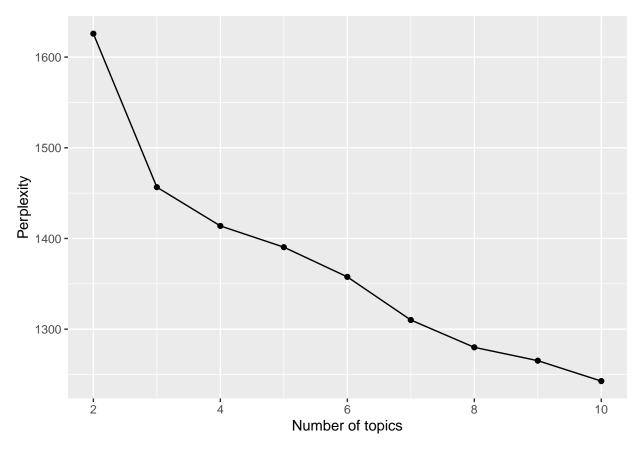
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```
# install packages
\# installed.packages(c("quanteda", "ggplot2", "dplyr", "smart", "quanteda.dictionaries", "quanteda.text")
# load packages
library(quanteda)
## Package version: 3.2.1
## Unicode version: 13.0
## ICU version: 69.1
## Parallel computing: 4 of 4 threads used.
## See https://quanteda.io for tutorials and examples.
library(ggplot2)
library(dplyr)
## Warning: package 'dplyr' was built under R version 4.2.3
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
      filter, lag
## The following objects are masked from 'package:base':
      intersect, setdiff, setequal, union
##
library(quanteda.dictionaries)
library(quanteda.textplots)
library(topicmodels)
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.2
## --
## v tibble 3.2.1
                    v purrr
                               0.3.4
## v tidyr 1.2.0
                     v stringr 1.5.0
## v readr
          2.1.2
                     v forcats 0.5.1
## Warning: package 'tibble' was built under R version 4.2.3
## Warning: package 'stringr' was built under R version 4.2.3
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
```

```
library(tidyr)
library(tidytext)
# load data
df <- read.csv("platforms_1928-2020.csv")</pre>
table(df$years)
##
## 1928 1932 1936 1940 1944 1948 1952 1956 1960 1964 1968 1972 1976 1980 1984 1988
## 1 1 1 1 1 1
                                  1
                                     1 1 1 1 1 1 1
## 1992 1996 2000 2004 2008 2012 2016
          1
              1
                   1
                        1
# create corpus and preprocessing
corpus1 <- corpus(df,</pre>
                 text_field = "textR") %>%
  tokens(remove_punct = TRUE, # remove punctuation
        remove_numbers = TRUE, # remove numbers
        remove_symbols = TRUE, # remove special symbols
        padding = TRUE) %>%
 tokens_remove(c(quanteda::stopwords(language = "en", source = "smart")), padding = TRUE) %>% # remove
 tokens_tolower() %>% # make all words into lower cases
 tokens wordstem()
# create DFM
# trim DFM, delete words with counts less than 10 or more than 1000
dfm1 <- dfm(corpus1) %>%
 dfm_trim(min_termfreq = 1, max_termfreq = 10000)
nfeat(dfm1)
## [1] 10134
# make a word cloud
set.seed(132)
textplot_wordcloud(dfm1, max_words = 100)
```

```
leadership
                                                                                                                    million children
                                                                                      creat communiti trade technolog
                                                                        assist privat free war opportun
                                                                         militari increas protectiob
                                                                  S countri administr educ bush
energi en
                                                                                law
                                                                                                                                                                                                                                                  citizen right
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             futur
                                                                                                                                                                                                                                work public a
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                                                                                                                                                                                          develop encourag
                                                        congress secur partiservic reform provid democration
                                                                                                                                                                           market
democrat<sup>polit</sup>
                                                                                problem individu import women
```

```
## LDA Model ##
dtm1 <- quanteda::convert(dfm1, to = "topicmodels")</pre>
## as(<dgCMatrix>, "dgTMatrix") is deprecated since Matrix 1.5-0; do as(., "TsparseMatrix") instead
# fitting an LDA model is determining the size of k.
n_{topics} \leftarrow c(2, 3, 4, 5, 6, 7, 8, 9, 10)
lda_compare1 <- n_topics %>%
 map(LDA, x = dfm1, control = list(seed = 1109))
tibble(k = n_topics,
       perplex = map_dbl(lda_compare1, perplexity)) %>%
  ggplot(aes(k, perplex)) +
  geom_point() +
  geom_line() +
 labs(# title = "Evaluating LDA topic models",
    # subtitle = "Optimal number of topics (smaller is better)",
    x = "Number of topics",
   y = "Perplexity")
```

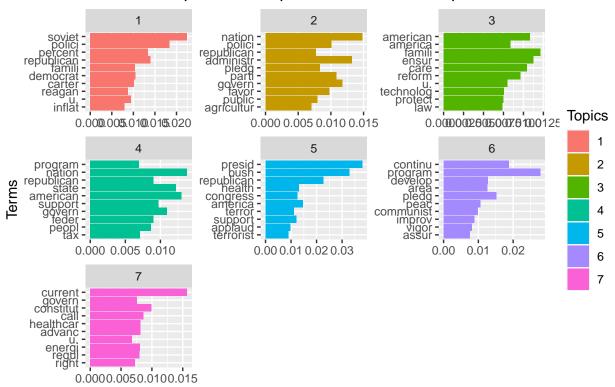


```
### K=7
set.seed(122)
lda1 <- LDA(dtm1, method = "Gibbs", k = 7, control = list(alpha = 0.1))
terms(lda1, 10)</pre>
```

```
Topic 2
                                     Topic 3
##
         Topic 1
                                                  Topic 4
                                                                Topic 5
                       "nation"
                                     "famili"
                                                                "presid"
##
    [1,] "soviet"
                                                  "nation"
    [2,] "polici"
                       "administr"
                                     "ensur"
                                                  "american"
                                                                "bush"
##
##
    [3,] "republican"
                       "govern"
                                     "american"
                                                  "state"
                                                                "republican"
##
    [4,] "percent"
                       "parti"
                                     "care"
                                                  "govern"
                                                                "america"
    [5,] "democrat"
                       "polici"
                                     "reform"
                                                  "support"
                                                                "health"
    [6,] "famili"
                       "favor"
##
                                     "america"
                                                  "feder"
                                                                "congress"
    [7,] "carter"
                       "pledg"
                                     "u."
                                                  "republican"
                                                                "support"
##
    [8,] "u."
                                     "technolog"
                                                  "peopl"
                                                                "terror"
##
                       "public"
##
    [9,] "reagan"
                       "republican"
                                     "protect"
                                                  "tax"
                                                                "applaud"
## [10,] "inflat"
                       "agricultur" "law"
                                                                "terrorist"
                                                  "program"
##
         Topic 6
                      Topic 7
    [1,] "program"
                      "current"
##
    [2,] "continu"
                      "constitut"
##
                      "call"
##
    [3,] "pledg"
##
    [4,] "develop"
                      "advanc"
   [5,] "area"
                      "healthcar"
##
    [6,] "peac"
                      "energi"
##
##
    [7,] "communist"
                      "regul"
##
    [8,] "improv"
                      "govern"
##
    [9,] "vigor"
                      "right"
```

```
## [10,] "assur"
topics1 <- tidy(lda1, matrix="beta")</pre>
head(topics1, 10)
## # A tibble: 10 x 3
##
     topic term
                             beta
##
      <int> <chr>
                            <dbl>
##
         1 republican 0.0139
## 2
         2 republican 0.00770
## 3
         3 republican 0.00620
## 4
         4 republican 0.00898
## 5
         5 republican 0.0226
         6 republican 0.00000957
## 6
## 7
         7 republican 0.0000115
## 8
         1 parti
                       0.00136
## 9
          2 parti
                       0.0108
## 10
          3 parti
                       0.0000360
# Reshape this by grouping values by topic
top_terms1 <- topics1 %>%
  group_by(topic) %>%
  top_n(10, beta) %>% # We just keep top 20 words
 ungroup() %>%
  arrange(topic, -beta) # We arrange them by descending beta values
# Let's see what this looks like
head(top_terms1, 25)
## # A tibble: 25 x 3
##
     topic term
                          beta
##
      <int> <chr>
                         <dbl>
## 1
         1 soviet
                       0.0224
## 2
         1 polici
                       0.0184
## 3
         1 republican 0.0139
## 4
        1 percent
                       0.0134
                       0.0105
## 5
         1 democrat
## 6
         1 famili
                       0.0104
## 7
         1 carter
                      0.0101
## 8
                      0.00946
         1 u.
## 9
         1 reagan
                      0.00875
## 10
         1 inflat
                       0.00797
## # i 15 more rows
top_terms1 %>%
 mutate(term = reorder(term, beta)) %>%
  ggplot(aes(term, beta, fill = factor(topic))) +
  geom_col(show.legend = T) +
  labs(title = "Proportion of Top 10 Terms in Each Topic",
      x = "Terms", y = "Term Distribution Per Topic") +
  theme(plot.title = element_text(hjust = 0.5)) +
  guides(fill = guide_legend(title = "Topics", title.position = "top")) +
  # scale fill discrete(limits=c("1", "2", "3"), labels=c("")) +
  facet_wrap(~ topic, scales = "free") +
  coord_flip()
```

Proportion of Top 10 Terms in Each Topic



Term Distribution Per Topic