Topic modeling

Quarto

Quarto enables you to weave together content and executable code into a finished document. To learn more about Quarto see https://quarto.org.

Running Code

```
# Load Data
library(readr)
library(dplyr)

'dplyr'

The following objects are masked from 'package:stats':
    filter, lag

The following objects are masked from 'package:base':
    intersect, setdiff, setequal, union

library(tm)
```

```
library(tidytext)
library(textclean)
library(SnowballC)
library(Matrix)
library(topicmodels)
library(ldatuning)
library(ggplot2)
   'ggplot2'
The following object is masked from 'package:NLP':
   annotate
library(wordcloud)
    RColorBrewer
# Load the data
movie_data <- read_csv("C:/Users/Owner/Downloads/movie_plots_with_genres.csv")</pre>
Rows: 1077 Columns: 4
-- Column specification ------
Delimiter: ","
chr (3): Movie Name, Genre, Plot
dbl (1): row
i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
head(movie_data)
```

```
# A tibble: 6 x 4
   row 'Movie Name'
                                               Genre Plot
 <dbl> <chr>
                                               <chr>
                                                       <chr>
   31 "Pioneers of the West"
                                               western "Pioneers of the West ~
   87 "The Infiltrators"
                                               action "The Infiltrators : A~
  146 "\"Graviton: The Ghost Particle\""
                                               sci-fi "\"Graviton: The Ghost ~
4 197 "Moses: Fallen. In the City of Angels." action "Moses: Fallen. In the ~
  314 "The Slave Trade"
                                               history "The Slave Trade : Be~
  448 "The 303rd"
                                               history "The 303rd : Ret. Col~
corpus <- Corpus(VectorSource(movie_data$Plot)) # Replace 'plot' with your actual column na</pre>
# Preprocess the text (e.g., remove punctuation, stop words, etc.)
corpus <- tm_map(corpus, content_transformer(tolower))</pre>
```

Warning in tm_map.SimpleCorpus(corpus, content_transformer(tolower)): transformation drops documents

```
corpus <- tm_map(corpus, removePunctuation)</pre>
```

Warning in tm_map.SimpleCorpus(corpus, removePunctuation): transformation drops documents

```
corpus <- tm_map(corpus, removeNumbers)</pre>
```

Warning in tm_map.SimpleCorpus(corpus, removeNumbers): transformation drops documents

```
corpus <- tm_map(corpus, removeWords, stopwords("english"))</pre>
```

Warning in tm_map.SimpleCorpus(corpus, removeWords, stopwords("english")): transformation drops documents

```
corpus <- tm_map(corpus, stripWhitespace)</pre>
```

Warning in tm_map.SimpleCorpus(corpus, stripWhitespace): transformation drops documents

```
# Create Document Term Matrix
dtm <- DocumentTermMatrix(corpus, control = list(bounds = list(global = c(2, Inf))))
# Convert DTM to a matrix if it has non-zero dimensions
if (ncol(dtm) > 0) {
   dtm_matrix <- as.matrix(dtm)
} else {
   stop("The DTM is empty after preprocessing. Try adjusting preprocessing steps.")
}
# Check if the DTM has content
print(dim(dtm_matrix))</pre>
```

[1] 1077 6359

scree plots for k

Purpose:

A scree plot helps to determine the optimal number of topics k in LDA. It visualizes how well different values of k explain the data, allowing you to choose an appropriate number of topics based on metrics like coherence or perplexity.

Interpretation:

• The "elbow point" or a plateau on the plot often indicates the best value for k, where increasing k further provides diminishing returns

```
library(Matrix)
dtm_matrix <- as.matrix(dtm)

# Now convert the matrix to a sparse format
dtm_sparse <- Matrix(dtm_matrix, sparse = TRUE)

results <- FindTopicsNumber(
    dtm_sparse,
    topics = seq(from = 2, to = 15, by = 1),
    metrics = c("CaoJuan2009", "Arun2010", "Deveaud2014"), # Avoid Griffiths2004 if it's problemethod = "Gibbs",
    control = list(seed = 1234),
    verbose = 5
)</pre>
```

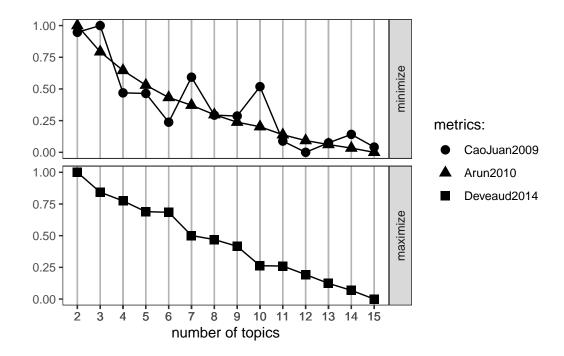
```
fit models... done.
calculate metrics:
   CaoJuan2009... done.
   Arun2010... done.
   Deveaud2014... done.
```

```
# Plot the results
FindTopicsNumber_plot(results)
```

Warning: The `<scale>` argument of `guides()` cannot be `FALSE`. Use "none" instead as of ggplot2 3.3.4.

i The deprecated feature was likely used in the ldatuning package.

Please report the issue at https://github.com/nikita-moor/ldatuning/issues.



cluster plot

Purpose:

A cluster plot shows how documents (in this case, movies) group together based on their topic distributions. Each point represents a document, and colors indicate different clusters. After dimensionality reduction (using techniques like PCA or t-SNE), the plot displays relationships between documents in a 2D space.

Interpretation:

- Documents in the same cluster are close to each other, indicating they have similar topic distributions.
- Points that are far from each other represent documents with distinct topic profiles.

```
k <- 5  # Replace with the optimal number from previous step

# Fit LDA model
lda_model <- LDA(dtm, k = k, control = list(seed = 1234))

# Get the topics and terms
topics <- terms(lda_model, 6)  # 6 terms per topic
print(topics)</pre>
```

```
Topic 1 Topic 2 Topic 3 Topic 4 Topic 5
[1,] "town" "war" "will" "will" "gang"
[2,] "ranch" "one" "love" "world" "man"
[3,] "gang" "new" "life" "life" "bill"
[4,] "horse" "life" "young" "new" "one"
[5,] "jim" "world" "one" "one" "daughter"
[6,] "tom" "story" "film" "time" "john"
```

```
# Get gamma matrix
gamma <- posterior(lda_model)$topics

# Perform k-means clustering
set.seed(1234)
kmeans_result <- kmeans(gamma, centers = k)

# Add cluster results to the original data
movie_data$cluster <- as.factor(kmeans_result$cluster)</pre>
```

```
pca <- prcomp(gamma, center = TRUE, scale. = TRUE)

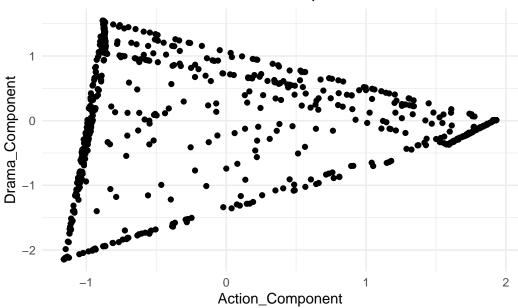
# Extract the first two principal components
gamma_pca <- as.data.frame(pca$x[, 1:2])

# Rename columns based on identified themes
colnames(gamma_pca) <- c("Action_Component", "Drama_Component") # Replace with appropriate relations.</pre>
```

```
# Now `gamma_pca` has descriptive names for the two dimensions
cluster_df <- data.frame(gamma, cluster = kmeans_result$cluster)

ggplot(gamma_pca, aes(x = Action_Component, y = Drama_Component), color= cluster) +
    geom_point() +
    labs(title = "Cluster Plot of Movies Based on Topics") +
    theme_minimal()</pre>
```

Cluster Plot of Movies Based on Topics



gamma plots

Purpose:

The gamma plot represents the document-topic distribution. For each document (movie), it shows the proportion (or probability) of each topic being present in that document. This is extracted from the gamma matrix in LDA.

Interpretation:

- A high gamma value for a specific topic indicates that the topic is strongly represented in the document.
- By plotting gamma distributions for documents, you can understand which topics dominate and how evenly distributed topics are across documents.

```
# Convert the gamma matrix (document-topic distribution) to a data frame
gamma_df <- as.data.frame(posterior(lda_model)$topics)

# Add document IDs
gamma_df$document <- 1:nrow(gamma_df)

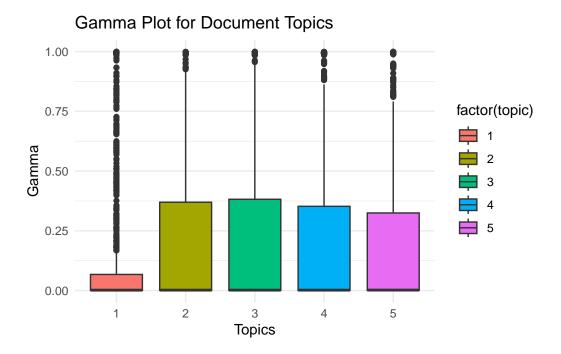
# Reshape data to long format
library(tidyr)</pre>
```

'tidyr'

The following objects are masked from 'package:Matrix':

```
expand, pack, unpack
```

```
gamma_long <- pivot_longer(gamma_df, cols = -document, names_to = "topic", values_to = "gamma"
# Plot the gamma values for each topic
ggplot(gamma_long, aes(x = factor(topic), y = gamma, fill = factor(topic))) +
    geom_boxplot() +
    labs(title = "Gamma Plot for Document Topics", x = "Topics", y = "Gamma") +
    theme_minimal()</pre>
```



beta plots

Purpose:

The beta plot visualizes the term-topic distribution, showing the probability of each word being associated with a particular topic. This is derived from the beta matrix in LDA.

Interpretation:

- High beta values for certain terms within a topic indicate that those terms are highly representative of that topic.
- By examining beta values, you can understand the key words that define each topic, which aids in naming and interpreting topics.

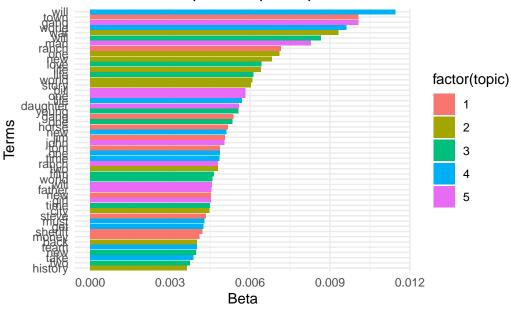
```
beta_df <- tidy(lda_model, matrix = "beta")

# Select the top 10 terms per topic based on the beta values
top_terms <- beta_df %>%
    group_by(topic) %>%
    slice_max(beta, n = 10) %>%
    ungroup()

# Plot the beta values for the top terms in each topic
```

```
ggplot(top_terms, aes(x = reorder_within(term, beta, topic), y = beta, fill = factor(topic))
geom_bar(stat = "identity") +
scale_x_reordered() +
labs(title = "Beta Plot for Top Terms per Topic", x = "Terms", y = "Beta") +
coord_flip() +
theme_minimal()
```

Beta Plot for Top Terms per Topic



word cloud

```
# Assuming lda_model is your LDA model object and k is the number of topics
library(tidytext)
topic_words <- tidy(lda_model, matrix = "beta")

# Filter for a specific topic if necessary, e.g., topic 1
topic1_words <- topic_words %>%
   filter(topic == 1) %>%
   arrange(desc(beta)) %>%
   slice_max(order_by = beta, n = 100) # Top 100 words
library(wordcloud)
library(RColorBrewer)
```

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : series could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : bob could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : rancher could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : horses could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : order could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : friend could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : brother could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : brothers could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : jack could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : sam could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : takes could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : goes could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : gene could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : gets could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : daughter could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : john could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : war could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : footage could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : work could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : friends could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : serial could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : family could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : keep could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : company could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : just could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : prison could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : fight could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : stage could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : bud could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : way could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : bank could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : game could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : ranger could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : morgan could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : use could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : plans could not be fit on page. It will not be plotted.

Warning in wordcloud(words = topic1_words\$term, freq = topic1_words\$beta, : soon could not be fit on page. It will not be plotted.

