

Canzone dei Dodici Mesi

Canzone dei Dodici Mesi di Guccini

As a bonus to the Caparezza's songs analysis, here is an analogue notebook which focuses on the lyrics of "Canzone dei Dodici Mesi" by Francesco Guccini.

```
library(geniusr)
library(tidyverse)
```

```
## Warning: il pacchetto 'lubridate' è stato creato con R versione 4.2.3
```

```
## — Attaching core tidyverse packages — tidyverse 2.0.0 —
## ✓ dplyr      1.1.0      ✓ readr      2.1.4
## ✓ forcats    1.0.0      ✓ stringr    1.5.0
## ✓ ggplot2     3.4.1      ✓ tibble     3.2.0
## ✓ lubridate  1.9.2      ✓ tidyr      1.3.0
## ✓ purrr       1.0.1
## — Conflicts — tidyverse_conflicts() —
## X dplyr::filter() masks stats::filter()
## X dplyr::lag()     masks stats::lag()
## i Use the [8];http://conflicted.r-lib.org/[8];[8] to force all conflict
s to become errors
```

```
library(tidytext)
```

```
## Warning: il pacchetto 'tidytext' è stato creato con R versione 4.2.3
```

```
library(quanteda)
```

```
## Warning: il pacchetto 'quanteda' è stato creato con R versione 4.2.3
```

```
## Package version: 3.3.1
## Unicode version: 13.0
## ICU version: 69.1
## Parallel computing: 8 of 8 threads used.
## See https://quanteda.io for tutorials and examples.
```

```
library(quanteda.textstats)
```

```
## Warning: il pacchetto 'quanteda.textstats' è stato creato con R versione 4.2.3
```

```
library(udpipe)
```

```
## Warning: il pacchetto 'udpipe' è stato creato con R versione 4.2.3
```

```
library(wordcloud)
```

```
## Warning: il pacchetto 'wordcloud' è stato creato con R versione 4.2.3
```

```
## Caricamento del pacchetto richiesto: RColorBrewer
```

```
library(textdata)
```

```
## Warning: il pacchetto 'textdata' è stato creato con R versione 4.2.3
```

```
library(reshape2)
```

```
## Warning: il pacchetto 'reshape2' è stato creato con R versione 4.2.3
```

```
##  
## Caricamento pacchetto: 'reshape2'  
##  
## Il seguente oggetto è mascherato da 'package:tidyr':  
##  
##      smiths
```

```
library(igraph)
```

```
## Warning: il pacchetto 'igraph' è stato creato con R versione 4.2.3
```

```
##
## Caricamento pacchetto: 'igraph'
##
## I seguenti oggetti sono mascherati da 'package:lubridate':
##
##    %--%, union
##
## I seguenti oggetti sono mascherati da 'package:dplyr':
##
##    as_data_frame, groups, union
##
## I seguenti oggetti sono mascherati da 'package:purrr':
##
##    compose, simplify
##
## Il seguente oggetto è mascherato da 'package:tidyr':
##
##    crossing
##
## Il seguente oggetto è mascherato da 'package:tibble':
##
##    as_data_frame
##
## I seguenti oggetti sono mascherati da 'package:stats':
##
##    decompose, spectrum
##
## Il seguente oggetto è mascherato da 'package:base':
##
##    union
```

Get lyrics

```
Sys.setenv(GENIUS_API_TOKEN = "VWN0BhFGVa9kdw9iY1dqwmcgo1LHICGF-U5B89k3h0g7szhSergcUNueRJpfck
aM")

lyrics <- get_lyrics_search(artist_name = "Francesco Guccini",
                           song_title = "Canzone dei dodici mesi")

months <- c("Gennaio", "Febbraio", "Marzo", "Aprile", "Maggio", "Giugno",
           "Luglio", "Agosto", "Settembre", "Ottobre", "Novembre", "Dicembre")

song_months <- c(
  rep(months[1], 4), rep(months[2], 4), rep(months[3], 4), rep(NA, 4),
  rep(months[4], 4), rep(months[5], 4), rep(months[6], 4), rep(NA, 4),
  rep(months[7], 2), rep(months[8], 2), rep(months[9], 4), rep(months[10], 4),
  rep(NA, 4), rep(months[11], 4), rep(months[12], 4), rep(NA, 6)
)

lyrics$month <- song_months

lyrics_by_month <- lyrics %>%
  group_by(month) %>%
  summarise(month_lyrics = paste(line, collapse = " ")) %>%
  arrange(match(month, months))
```

```
lyrics_by_month %>% head()
```

```
## # A tibble: 6 × 2
##   month      month_lyrics
##   <chr>      <chr>
## 1 Gennaio  Viene Gennaio, silenzioso e lieve, un fiume addormentato Fra le cui ...
## 2 Febbraio Viene Febbraio e il mondo è a capo chino, ma nei convitti e in piazz...
## 3 Marzo    Cantando, Marzo porta le sue piogge, la nebbia squarcia il velo Port...
## 4 Aprile    Con giorni lunghi, al sonno dedicati, il dolce Aprile viene Quali se...
## 5 Maggio  Ben venga Maggio e il gonfalone amico, ben venga primavera Il nuovo ...
## 6 Giugno   Giugno, che sei maturità dell'anno, di te ringrazio Dio In un tuo gi...
```

```
doc_ids <- vector()
for(i in 1:nrow(lyrics_by_month)){
  id <- paste("doc", toString(i), sep = "")
  doc_ids <- doc_ids %>% append(id)
}

lyrics_by_month <- lyrics_by_month %>% mutate(doc_id = doc_ids, .before = 1)
```

Text pre-processing

```
corpus <- corpus(lyrics_by_month$month_lyrics, docnames = lyrics_by_month$doc_id)
summary(corpus)
```

```
## Corpus consisting of 13 documents, showing 13 documents:
```

```
##
```

```
##   Text Types Tokens Sentences
```

```
## doc1    37    56         2
```

```
## doc2    40    62         2
```

```
## doc3    39    65         2
```

```
## doc4    41    61         2
```

```
## doc5    36    64         2
```

```
## doc6    42    67         2
```

```
## doc7    22    33         1
```

```
## doc8    22    33         1
```

```
## doc9    34    49         2
```

```
## doc10   37    63         2
```

```
## doc11   42    63         2
```

```
## doc12   42    66         2
```

```
## doc13   34   199         3
```

```
cat(as.character(corpus[1]))
```

```
## Viene Gennaio, silenzioso e lieve, un fiume addormentato Fra le cui rive giace come neve i  
l mio corpo malato, il mio corpo malato... Sono distese, lungo la pianura, bianche file di ca  
mpi Son come amanti dopo l'avventura, neri alberi stanchi, neri alberi stanchi...
```

```
corpus_tokens <- corpus %>%
```

```
  quantda::tokens(remove_punct = TRUE, remove_numbers = TRUE, remove_symbols = TRUE) %>%
```

```
  tokens_tolower()
```

```
txt <- sapply(corpus_tokens, FUN=function(x) paste(x, collapse = "\n"))
```

```
udpipe_download_model(language = "italian-isdt", model_dir = "resources/")
```

```
## Downloading udpipes model from https://raw.githubusercontent.com/jwijnffels/udpipe.models.u  
d.2.5/master/inst/udpipe-ud-2.5-191206/italian-isdt-ud-2.5-191206.udpipe to resources//italia  
n-isdt-ud-2.5-191206.udpipe
```

```
## - This model has been trained on version 2.5 of data from https://universaldependencies.o  
rg
```

```
## - The model is distributed under the CC-BY-SA-NC license: https://creativecommons.org/lic  
enses/by-nc-sa/4.0
```

```
## - Visit https://github.com/jwijnffels/udpipe.models.ud.2.5 for model license details.
```

```
## - For a list of all models and their licenses (most models you can download with this pac  
kage have either a CC-BY-SA or a CC-BY-SA-NC license) read the documentation at ?udpipe_down  
load_model. For building your own models: visit the documentation by typing vignette('udpipe-t  
rain', package = 'udpipe')
```

```
## Downloading finished, model stored at 'resources//italian-isdt-ud-2.5-191206.udpipe'
```

```
##           language                               file_model
## 1 italian-isdt resources//italian-isdt-ud-2.5-191206.udpipe
##
url
## 1 https://raw.githubusercontent.com/jwijffels/udpipe.models.ud.2.5/master/inst/udpipe-ud-2.5-191206/italian-isdt-ud-2.5-191206.udpipe
##   download_failed download_message
## 1                FALSE              OK
```

```
lang_model <- udpipe_load_model(file = "resources/italian-isdt-ud-2.5-191206.udpipe")
outL <- udpipe_annotate(lang_model, x = txt, tokenizer = "vertical", trace = TRUE) %>%
  as.data.frame()
```

```
## 2023-05-31 10:51:59 Annotating text fragment 1/13
## 2023-05-31 10:51:59 Annotating text fragment 2/13
## 2023-05-31 10:51:59 Annotating text fragment 3/13
## 2023-05-31 10:51:59 Annotating text fragment 4/13
## 2023-05-31 10:51:59 Annotating text fragment 5/13
## 2023-05-31 10:51:59 Annotating text fragment 6/13
## 2023-05-31 10:51:59 Annotating text fragment 7/13
## 2023-05-31 10:51:59 Annotating text fragment 8/13
## 2023-05-31 10:51:59 Annotating text fragment 9/13
## 2023-05-31 10:51:59 Annotating text fragment 10/13
## 2023-05-31 10:51:59 Annotating text fragment 11/13
## 2023-05-31 10:51:59 Annotating text fragment 12/13
## 2023-05-31 10:51:59 Annotating text fragment 13/13
```

```
it_stopwords <- readLines("https://raw.githubusercontent.com/stopwords-iso/stopwords-it/master/stopwords-it.txt")
```

```
## Warning in
## readLines("https://raw.githubusercontent.com/stopwords-iso/stopwords-it/master/stopwords-it.txt"):
## riga finale incompleta in
## 'https://raw.githubusercontent.com/stopwords-iso/stopwords-it/master/stopwords-it.txt'
```

```
outL <- outL %>% filter(!(token %in% it_stopwords) & !(lemma %in% it_stopwords))
```

```
outL %>% select(doc_id, token, lemma, upos) %>% sample_n(5)
```

```
##   doc_id  token  lemma upos
## 1   doc1   campi  campo NOUN
## 2  doc13 giocare giocare VERB
## 3   doc1  malato  malato ADJ
## 4  doc13 giocare giocare VERB
## 5   doc3   porta portare VERB
```

```
outL_reduced <- outL %>% filter(upos %in% c("NOUN", "PROPN", "ADJ", "VERB"))
```

```

lemmatized_lyrics <- outL_reduced %>% group_by(doc_id = fct_inorder(doc_id)) %>%
  summarise(lemmatized = paste(lemma, collapse = " "))
lyrics_by_month <- lyrics_by_month %>% right_join(lemmatized_lyrics, by = "doc_id")

corpus <- lyrics_by_month$lemmatized %>% corpus(docnames = lyrics_by_month$doc_id)

```

```
DTM <- corpus %>% tokens() %>% dfm()
```

```
DTM
```

```

## Document-feature matrix of: 13 documents, 197 features (91.53% sparse) and 0 docvars.
##           features
## docs  venire gennaio silenzioso lieve fiume addormentato riva giacere neve
## doc1      1      1      1      1      1      1      1      1      1
## doc2      1      0      0      0      0      0      0      0      0
## doc3      0      0      0      0      0      0      0      0      1
## doc4      1      0      0      0      0      0      0      0      0
## doc5      4      0      0      0      0      0      0      0      0
## doc6      0      0      0      0      0      0      0      0      0
##           features
## docs  corpo
## doc1      2
## doc2      0
## doc3      0
## doc4      0
## doc5      0
## doc6      0
## [ reached max_ndoc ... 7 more documents, reached max_nfeat ... 187 more features ]

```

Lexical Analysis

```
DTM %>% dim()
```

```
## [1] 13 197
```

```

words <- colnames(DTM)
freqs <- colSums(DTM)
wordlist <- data.frame(words, freqs)
wordlist %>% arrange(-freqs) %>% head()

```

```

##           words freqs
## sapere  sapere   13
## giocare giocare   12
## venire  venire    7
## mano     mano     5
## sole     sole     4
## nascere  nascere   4

```

Data visualization

```
par(mar=c(1,1,0.5,1))
wordcloud(words = wordlist$words, freq = wordlist$freqs, scale = c(3.5, 0.35), max.words = 5
0, random.order = F,
         colors = RColorBrewer::brewer.pal(name = "Dark2", n = 4))
text(0.5, 1, "wordcloud with TF ponderation", font = 2)
```

wordcloud with TF ponderation



TF-IDF ponderation

```
tf_idf <- dfm_tfidf(DTM)
freqs_tf_idf <- colSums(tf_idf)
words_tf_idf <- colnames(tf_idf)
wordlist_tf_idf <- data.frame(words = words_tf_idf, freqs = freqs_tf_idf)
wordlist_tf_idf %>% arrange(-freqs) %>% head(10)
```



```
##          words      freqs
## giocare giocare 13.367320
## sapere  sapere 10.567874
## andare  andare  4.455773
## simile  simile  4.455773
## tarocco tarocco  4.455773
## mano     mano   4.064567
## venire  venire  3.583184
## nascere nascere  3.251653
## sole     sole   2.547288
## malato   malato  2.438740
```

```
par(mar=c(1,1,0.2,1))
wordcloud(words = wordlist_tf_idf$words, freq = wordlist_tf_idf$freqs,
          scale = c(3.5, 0.35), max.words = 50, random.order = F,
          colors = RColorBrewer::brewer.pal(name = "Dark2", n = 4))
text(0.5, 1, "wordcloud with TF-IDF ponderation", font = 2)
```

wordcloud with TF-IDF ponderation



Co-occurrence analysis

```
binDTM <- DTM %>% dfm_weight("boolean")
coocCounts <- t(binDTM) %*% binDTM
as.matrix(coocCounts[16:18, 16:18])
```

```
##          son amante l'avventura
## son          1      1      1
## amante       1      1      1
## l'avventura  1      1      1
```

```
coocTerm <- "sole"
k <- nrow(binDTM)
ki <- sum(binDTM[, coocTerm])
kj <- colSums(binDTM)
names(kj) <- colnames(binDTM)
kij <- coocCounts[coocTerm, ]

mutualInformationSig <- log(k * kij / (ki * kj))
mutualInformationSig <- mutualInformationSig[order(mutualInformationSig, decreasing = TRUE)]

dicesig <- 2 * kij / (ki + kj)
dicesig <- dicesig[order(dicesig, decreasing=TRUE)]

logsig <- 2 * ((k * log(k)) - (ki * log(ki)) - (kj * log(kj)) + (kij * log(kij))
              + (k - ki - kj + kij) * log(k - ki - kj + kij)
              + (ki - kij) * log(ki - kij) + (kj - kij) * log(kj - kij)
              - (k - ki) * log(k - ki) - (k - kj) * log(k - kj))
logsig <- logsig[order(logsig, decreasing=T)]

resultOverView <- data.frame(
  names(sort(kij, decreasing=T)[1:10]), sort(kij, decreasing=T)[1:10],
  names(mutualInformationSig[1:10]), mutualInformationSig[1:10],
  names(dicesig[1:10]), dicesig[1:10],
  names(logsig[1:10]), logsig[1:10],
  row.names = NULL)
colnames(resultOverView) <- c("Freq-terms", "Freq", "MI-terms", "MI", "Dice-Terms", "Dice",
"LL-Terms", "LL")
print(resultOverView)
```

##	Freq-terms	Freq	MI-terms	MI	Dice-Terms	Dice	LL-Terms	LL
## 1	sole	3	febbraio	1.466337	sole	1.0000000	venire	2.2211525
## 2	venire	2	capo	1.466337	venire	0.5714286	malato	0.8416541
## 3	malato	1	chare	1.466337	febbraio	0.5000000	lasciare	0.8416541
## 4	febbraio	1	convitto	1.466337	capo	0.5000000	l'inverno	0.8416541
## 5	capo	1	piazza	1.466337	chare	0.5000000	apparire	0.8416541
## 6	chare	1	dolore	1.466337	convitto	0.5000000	poeta	0.8416541
## 7	convitto	1	vedere	1.466337	piazza	0.5000000	bello	0.8416541
## 8	piazza	1	arleccare	1.466337	dolore	0.5000000	terra	0.8416541
## 9	lasciare	1	carnevale	1.466337	vedere	0.5000000	nascere	0.8416541
## 10	dolore	1	impazzare	1.466337	arleccare	0.5000000	mano	0.8416541

Co-occurrence visualization

```
source("resources/calculateCoocStatistics.R")
numberOfCoocs <- 10
coocTerm <- "sapere"
coocs <- calculateCoocStatistics(coocTerm, binDTM, measure="LOGLIK")
```

```
## Caricamento del pacchetto richiesto: Matrix
```

```
##
## Caricamento pacchetto: 'Matrix'
```

```
## I seguenti oggetti sono mascherati da 'package:tidyr':
##
##      expand, pack, unpack
```

```
print(coocs[1:numberOfCoocs])
```

```
##      mano      triste      venire      gennaio      silenzioso      lieve
##      1.687816      1.687816      NaN      NaN      NaN      NaN
##      fiume addormentato      riva      giacere
##      NaN      NaN      NaN      NaN
```

```

resultGraph <- data.frame(from = character(), to = character(), sig = numeric(0))
tmpGraph <- data.frame(from = character(), to = character(), sig = numeric(0))

# Fill the data.frame to produce the correct number of lines
tmpGraph[1:numberOfCoocs, 3] <- coocs[1:numberOfCoocs]
# Entry of the search word into the first column in all lines
tmpGraph[, 1] <- coocTerm
# Entry of the co-occurrences into the second column of the respective line
tmpGraph[, 2] <- names(coocs)[1:numberOfCoocs]
# Set the significances
tmpGraph[, 3] <- coocs[1:numberOfCoocs]

# Attach the triples to resultGraph
resultGraph <- rbind(resultGraph, tmpGraph)

# Iteration over the most significant numberOfCoocs co-occurrences of the search term
for (i in 1:numberOfCoocs){

  # Calling up the co-occurrence calculation for term i from the search words co-occurrences
  newCoocTerm <- names(coocs)[i]
  coocs2 <- calculateCoocStatistics(newCoocTerm, binDTM, measure="LOGLIK")

  #print the co-occurrences
  coocs2[1:10]

  # Structure of the temporary graph object
  tmpGraph <- data.frame(from = character(), to = character(), sig = numeric(0))
  tmpGraph[1:numberOfCoocs, 3] <- coocs2[1:numberOfCoocs]
  tmpGraph[, 1] <- newCoocTerm
  tmpGraph[, 2] <- names(coocs2)[1:numberOfCoocs]
  tmpGraph[, 3] <- coocs2[1:numberOfCoocs]

  #Append the result to the result graph
  resultGraph <- rbind(resultGraph, tmpGraph[2:length(tmpGraph[, 1]), ])
}

# Sample of some examples from resultGraph
resultGraph[sample(nrow(resultGraph), 6), ]

```

```

##      from      to      sig
## 91  mano      fiume    NaN
## 79  riva      giacere    NaN
## 43  venire    l'inverno 0.3847397
## 87  fiume      neve     NaN
## 10  sapere    giacere    NaN
## 57  fiume     addormentato NaN

```

```

# set seed for graph plot
set.seed(1)

# Create the graph object as undirected graph
graphNetwork <- graph.data.frame(resultGraph, directed = F)

# Identification of all nodes with less than 2 edges
verticesToRemove <- V(graphNetwork)[degree(graphNetwork) < 2]
# These edges are removed from the graph
graphNetwork <- delete.vertices(graphNetwork, verticesToRemove)

# Assign colors to nodes (search term blue, others orange)
V(graphNetwork)$color <- ifelse(V(graphNetwork)$name == coocTerm, 'cornflowerblue', 'orange')

# Set edge colors
E(graphNetwork)$color <- adjustcolor("DarkGray", alpha.f = .5)
# scale significance between 1 and 10 for edge width
E(graphNetwork)$width <- scales::rescale(E(graphNetwork)$sig, to = c(1, 10))

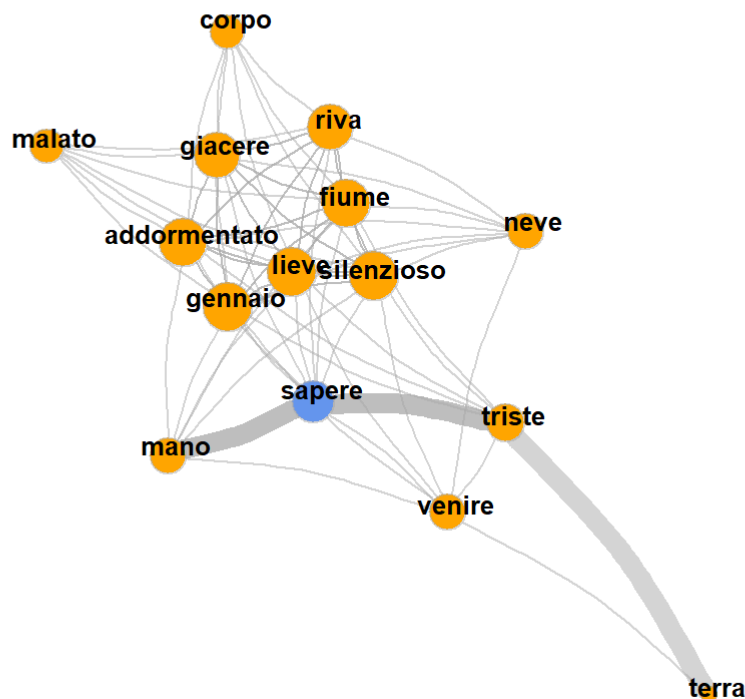
# Set edges with radius
E(graphNetwork)$curved <- 0.15
# Size the nodes by their degree of networking (scaled between 5 and 15)
V(graphNetwork)$size <- scales::rescale(log(degree(graphNetwork)), to = c(5, 15))

# Define the frame and spacing for the plot
par(mai=c(0,0,1,0))

# Final Plot
plot(
  graphNetwork,
  layout = layout.fruchterman.reingold, # Force Directed Layout
  main = paste(coocTerm, ' Graph'),
  vertex.label.family = "sans",
  vertex.label.cex = 0.8,
  vertex.shape = "circle",
  vertex.label.dist = 0.5, # Labels of the nodes moved slightly
  vertex.frame.color = adjustcolor("darkgray", alpha.f = .5),
  vertex.label.color = 'black', # Color of node names
  vertex.label.font = 2, # Font of node names
  vertex.label = V(graphNetwork)$name, # node names
  vertex.label.cex = 1 # font size of node names
)

```

sapere Graph



Sentiment analysis

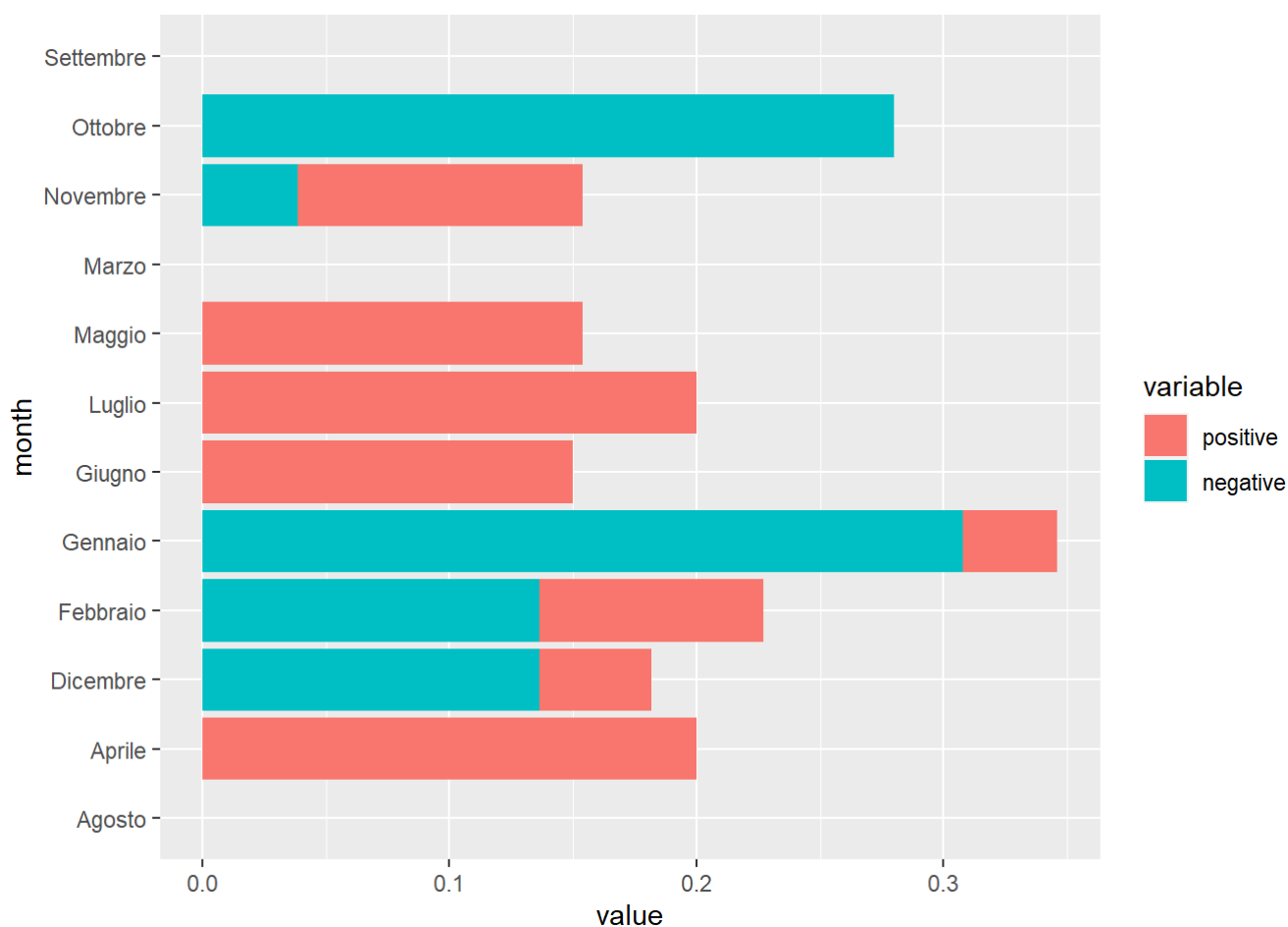
```
sentiment_lexicon <- read.table("resources/Italian-NRC-EmoLex.txt",
                                header = TRUE, sep = "\t")
sentiment_lexicon_corpus <- sentiment_lexicon %>% filter(Italian.Word %in% colnames(DTM))
positive_terms <- sentiment_lexicon_corpus %>%
  filter(positive == 1) %>%
  select(Italian.Word) %>% pull()
negative_terms <- sentiment_lexicon_corpus %>%
  filter(negative == 1) %>%
  select(Italian.Word) %>% pull()
counts_positive <- rowSums(DTM[, positive_terms])
counts_negative <- rowSums(DTM[, negative_terms])
counts_all_terms <- rowSums(DTM)
relative_sentiment_frequencies <- data.frame(
  positive = counts_positive / counts_all_terms,
  negative = counts_negative / counts_all_terms
)
```

```
sentiments_by_month <- aggregate(relative_sentiment_frequencies,
                                  by = list(month = lyrics_by_month$month), mean)

sentiments_by_month %>% head()
```

```
##      month  positive  negative
## 1  Agosto 0.00000000 0.00000000
## 2  Aprile 0.20000000 0.00000000
## 3 Dicembre 0.04545455 0.1363636
## 4 Febbraio 0.09090909 0.1363636
## 5 Gennaio 0.03846154 0.3076923
## 6  Giugno 0.15000000 0.00000000
```

```
df_sentiment <- melt(sentiments_by_month, id.vars = "month")
ggplot(data = df_sentiment, aes(x = month, y = value, fill = variable)) +
  geom_bar(stat="identity", position="stack") + coord_flip()
```



```
positive_months <- aggregate(
  relative_sentiment_frequencies, by = list(month = lyrics_by_month$month),
  mean) %>% filter(positive > negative)
positive_months
```

```
##      month  positive  negative
## 1  Aprile 0.20000000 0.00000000
## 2  Giugno 0.15000000 0.00000000
## 3  Luglio 0.20000000 0.00000000
## 4  Maggio 0.1538462 0.00000000
## 5 Novembre 0.1153846 0.03846154
```

```
negative_months <- aggregate(
  relative_sentiment_frequencies, by = list(month = lyrics_by_month$month),
  mean) %>% filter(negative > positive)
negative_months
```

```
##      month  positive  negative
## 1 Dicembre 0.04545455 0.1363636
## 2 Febbraio 0.09090909 0.1363636
## 3 Gennaio  0.03846154 0.3076923
## 4 Ottobre  0.00000000 0.2800000
```

```
neutral_months <- aggregate(
  relative_sentiment_frequencies, by = list(month = lyrics_by_month$month),
  mean) %>% filter(positive == negative)
neutral_months
```

```
##      month positive negative
## 1   Agosto         0         0
## 2   Marzo          0         0
## 3 Settembre        0         0
```

Emotion analysis

```
anger_terms <- sentiment_lexicon_corpus %>%
  filter(anger == 1) %>%
  select(Italian.Word) %>% pull()
fear_terms <- sentiment_lexicon_corpus %>%
  filter(fear == 1) %>%
  select(Italian.Word) %>% pull()
joy_terms <- sentiment_lexicon_corpus %>%
  filter(joy == 1) %>%
  select(Italian.Word) %>% pull()
sadness_terms <- sentiment_lexicon_corpus %>%
  filter(sadness == 1) %>%
  select(Italian.Word) %>% pull()

counts_anger <- rowSums(DTM[, anger_terms])
counts_fear <- rowSums(DTM[, fear_terms])
counts_joy <- rowSums(DTM[, joy_terms])
counts_sadness <- rowSums(DTM[, sadness_terms])

relative_emotion_frequencies <- data.frame(
  anger = counts_anger / counts_all_terms,
  fear = counts_fear / counts_all_terms,
  joy = counts_joy / counts_all_terms,
  sadness = counts_sadness / counts_all_terms
)
```



```
emotions_by_month <- aggregate(relative_emotion_frequencies,
                                by = list(month = lyrics_by_month$month), mean)
```

```
head(emotions_by_month)
```

```
##      month      anger      fear      joy      sadness
## 1  Agosto 0.00000000 0.00000000 0.00000000 0.00000000
## 2  Aprile 0.00000000 0.00000000 0.15000000 0.00000000
## 3 Dicembre 0.09090909 0.09090909 0.00000000 0.09090909
## 4 Febbraio 0.00000000 0.09090909 0.04545455 0.13636364
## 5 Gennaio 0.00000000 0.15384615 0.03846154 0.23076923
## 6  Giugno 0.00000000 0.00000000 0.15000000 0.05000000
```

```
df_emotions <- melt(emotions_by_month, id.vars = "month")
ggplot(data = df_emotions, aes(x = month, y = value, fill = variable)) +
  geom_bar(stat="identity", position="stack") + coord_flip()
```

