Assignment 10 by Team

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1. Convert the html to text files and separate the individual news items. The individual press release items serve as documents.

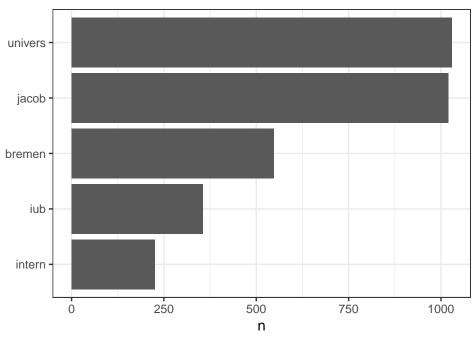
```
colnames(text_df)
[1] "id"
           "text" "year"
text_df[2]
# A tibble: 638 x 1
                                                                           text
 1 Wissenschaft jenseits von Science Fiction: Jacobs University beteiligt sich
 2 Sozialer Mehrwert durch Musik? Begleitstudie der Jacobs University zur Symb
3 Leibniz-Preis für Jacobs-Professorin Antje Boetius Dec , Antje Boetius, sei
4 Neuer Förderpreis der Stiftung Mercator für Studierende der Jacobs Universi
 5 Management mit Zukunft: TiasNimbas Business School und Jacobs University st
 6 Deutscher Hochschulverband ernennt Katja Windt zur »Hochschullehrerin des J
7 Der persönliche Eindruck zählt: Studienberater aus vier Kontinenten informi
8 Spintronik: Physikerteam gelingt Nachweis eines nano-mechanischen Torsionse
9 "Neue malerische Wendungen": University Club der Jacobs University zeigt ab
10 RWE startet CO-Konversions-Pilotanlage auf Basis einer von der Jacobs Unive
# ... with 628 more rows
```

2. Remove stop words and perform stemming.

```
t <- text_df %>%
   unnest_tokens(word, text) %>%
    anti_join(tibble(word = c(stopwords("de"), stopwords("en")))) %>%
   mutate(stemmed_word = wordStem(word))
Joining, by = "word"
head(t)
# A tibble: 6 x 4
     id year
                      word stemmed_word
  <int> <chr>
                     <chr>
                                  <chr>>
      1 2008 wissenschaft wissenschaft
      1 2008
                  jenseits
                                jenseit
3
     1 2008
                   science
                                 scienc
4
     1 2008
                   fiction
                                fiction
5
     1 2008
                    jacobs
                                  jacob
     1 2008
               university
                                univers
```

3. Perform a frequency analysis to compute the term-document (TD) matrix. What are the most common terms?

```
top_5_words <- t %>%
    group_by(stemmed_word) %>%
    count(sort = TRUE) %>%
    ungroup() %>%
    slice(1:5)
```



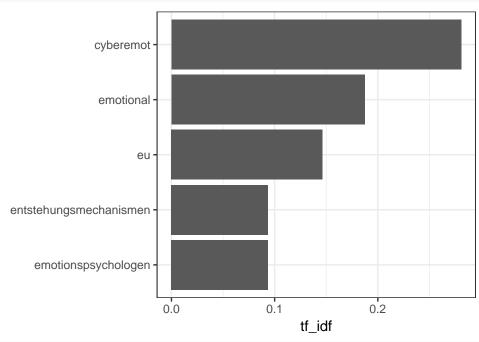
```
word_counts <- t %>%
    group_by(id, stemmed_word) %>%
    count() %>%
    arrange(id, -n) %>%
    ungroup()

td <- word_counts %>% spread(stemmed_word, n, fill = 0) %>%
    select(-id) %>%
    as.matrix()
```

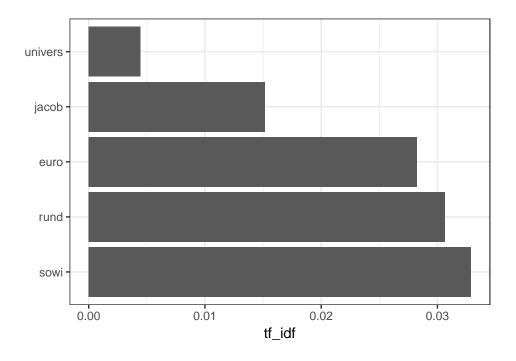
```
univers jacob bremen iub intern
          3
                3
                            0
1
                        0
                                    0
2
          2
                2
                        5
                             0
                                    0
3
                2
                        2
                                    0
          1
                             0
                2
4
          1
                        0
                             0
                                    1
          2
                2
                        2
5
                            0
                                    0
6
          1
                1
                        1
                            0
                                    0
7
          2
                3
                        1
                            0
                                    0
8
          2
                            0
                                    0
                1
                        1
9
          4
                3
                        1
                            0
                                    0
                                    0
10
          3
                3
                        1
                            0
```

4. Compute inverse-document frequency (IDF) and term importance (TI). What are now the most common terms?

```
tf_idf <- word_counts %>%
   bind_tf_idf(term = stemmed_word, document = id, n = n)
```



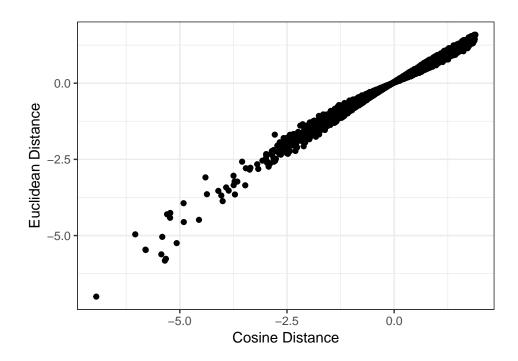
```
tf_idf %>%
  filter(id == 1) %>%
  arrange(tf_idf) %>%
  slice(1:5) %>%
  ggplot() +
  geom_col(aes(x = reorder(stemmed_word, -tf_idf), y = tf_idf)) +
  xlab(NULL) +
  coord_flip() +
  theme_bw()
```



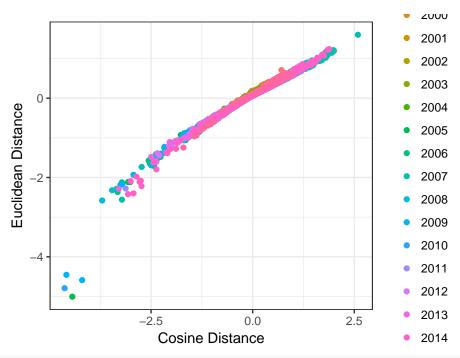
5. Compute pairwise cosine and Euclidean distance between all documents.

```
cos_dist <- dist2(td, method = 'cosine')</pre>
euc_dist <- dist2(td, method = 'euclidean')</pre>
cos_dist[1:3, 1:3]
                            2
                                           3
             1
1 9.992007e-16 8.578515e-01 8.455115e-01
2 8.578515e-01 -4.440892e-16 7.659177e-01
3 8.455115e-01 7.659177e-01 -1.332268e-15
euc_dist[1:3, 1:3]
         [,1]
                   [,2]
                            [,3]
[1,] 0.000000 1.309848 1.300393
[2,] 1.309848 0.000000 1.237673
[3,] 1.300393 1.237673 0.000000
```

6. Apply a multi-dimensional scaling approach to the distance matrix and render a 2D scatter- plot. Compare the two distance metrics.



7. Capture the year of release during parsing and color code the scatterplot by time. Produce a Word Cloud for each year.



```
create_wordcloud <- function(year) {
  d <- t %>%
    filter(year == year) %>%
      group_by(stemmed_word) %>%
      count() %>%
      ungroup()
```

create_wordcloud(2015)

