## R for Korean Studies

A Gentle Introduction to Computational Social Science

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### **Preface**

Korean Studies is traditionally dominated by scholars of history and literature. It's relatively rare to see R, Python, or other computational social science tools being used or taught in this field.

I believe computational social science offers huge opportunities for Korean Studies, not only for quantitative research but also for qualitative studies, including those on history and literature!

In this book, I aim to increase data literacy and convince as many Korean Studies scholars and students as possible about the relative ease of learning R with code samples, and motivational case studies about Korea.

This book is supposed to be a gentle introduction, so I do not go into the details of the R language. You can refer to the links that I provide in this book for more information. Furthermore, I also strongly encourage you to use Github's Copilot which is free for academic use, Chatgpt which is not necessarily a coding bot, but still helpful especially for simple tasks, Stackoverflow, and Google for help whenever you are stuck or come across an error.

I also encourage you to join our bootcamps for problem solving! You can sign up for my newsletter to get updates on the workshops.

# 1 Introduction

# 2 Setting Up

Downloading R, and Rstudio.

# 3 Korean Studies Data Sources

- 3.1 Statistical Data
- 3.2 Text Data

## 4 Basics of R

# 5 Data Wrangling

# 6 Data Visualization: Figures

## 7 Data Visualization: Plots

# 8 Data Visualization: Maps

## 9 Korean Text Analysis

In this chapter, we will learn how to analyze Korean text data using R. We will use the tidyverse, pdftools, and bitNLP packages to extract text from a pdf file and analyze it. We will use Korea's 2022 Diplomatic White Paper (, waegyo baekseo) as an example text.

We will learn the following things in order:

- Extracting text and tables from a PDF file.
- Extracting text and tables from the internet.
- Ensuring accurate spacing between words in Korean text.
- Analyzing morphemes in Korean text.
- Analyzing word frequency in Korean text.
- Analyzing the noun word network in Korean text.
- Analyzing the sentiment of Korean text.
- Topic modeling of Korean text.

#### 9.1 Libraries

First, we need to install bitNLP which requires us to install the MeCab library for Korean text analysis. Uncomment the following lines in your first usage. After the first usage, you can comment out the installation lines.

```
# install.packages("remotes")
# remotes::install_github("bit2r/bitNLP")
library(bitNLP)
# install_mecab_ko()
# install.packages("RcppMeCab")
```

Now let's load the necessary libraries. If you are missing any of the following packages, you can install them by uncommenting the install.packages lines.

```
# install.packages("tidyverse")
# install.packages("pdftools")
# install.packages("rvest")
# install.packages("tidytext")
```

```
# install.packages("igraph")
# install.packages("ggraph")
# install.packages("extrafont")
library(tidyverse)
library(pdftools)
library(rvest)
library(tidytext)
library(igraph)
library(ggraph)
library(extrafont)
```

#### 9.2 Loading pdf Data

Let's analyze the text from Korea's 2024 Public Diplomacy Comprehensive Implementation Plan (2024  $\,$  ) which is available as a pdf file on the Ministry of Foreign Affairs' (MOFA) website  $^1$ .

If the pdf file is in your local directory, you can load it using the following code.

```
# Load PDF
pdf_path <- "data/2024 .pdf"</pre>
```

Alternatively, you can download the pdf file from the MOFA's website using the download.file function. You can then load the pdf file using the pdf\_path variable. Working with the online pdf file and the local pdf file is the same. We can do either. For now, I will use the local pdf file since the MOFA might change the url for the pdf later. That is why I commented the download code. You can comment the earlier code for the local pdf file and uncomment the following code for the online pdf file.

```
# Download PDF
#file <- tempfile()

# This url works for now. But MOFA might change it later. You can replace the link with an
#url <- "https://www.mofa.go.kr/cntntsDown.do?path=www&physic=2024%EB%85%84%EB%8F%84_%EA%E
# download.file(url, pdf_path, headers = c("User-Agent" = "My Custom User Agent"))</pre>
```

<sup>&</sup>lt;sup>1</sup>Please bear in mind that MOFA website's url might change later, making this hyperlink broken. In that case, you can download the pdf file on the MOFA's website by searching for "2024".

Now let's extract the text from the pdf file using the pdf\_text function from the pdftools package.

```
# Extract text
pdf_text_all <- pdf_text(pdf_path)</pre>
```

Now, pdf\_text\_all is a list of character vectors, where each element corresponds to a page in the pdf file. For example, we can look at the 4<sup>th</sup> page of the pdf file in the following way.

```
# Let's look at the 4th page
pdf_text_all[4]

[1] " \n[ ]\n '24 '23 '24 '23 \n
```

Oh, this is too long even for an example. But you can realize that there are many \n characters in the text. Let's split the text by the newline character and look at the first 10 lines of the 4th page. \n refers to a new line in the text. We can split the text into lines by using the str\_split function from the stringr package, which is part of tidyverse. So, we don't need to load it separately. Let's look at the first six lines of the 4th page.

```
# Look at the first 10 lines of the 4th page
pdf_text_all[4] |>
    # Split by newline character.
    str_split("\n") |>
    # Unlist
    unlist() |>
    # Take the first 10 lines
    head(10)
```

```
"
 [1] "
 [2] "[
            7 "
                               <sup>24</sup>
                                         23
                                                      '24
                                                                         <sup>23</sup>
 [3] "
                  11
 [4] "
 [5] "
                                                      ( )
                                                                         ( )"
 [6] " 1
                                  16
                                             16
                                                             194,996
                                                                                   94,963"
 [7] " 2
                               6
                                           6
                                                          32,852
                                                                               40,283"
 [8] " 3
                                  73
                                                                                   39,419"
                                             63
                                                              40,215
 [9] "3-1
                               37
                                          41
                                                           42,514
                                                                                44,664"
[10] " 4
                                                                                    2,386"
                                   6
                                              6
                                                               1,831
```

The 4th page in the pdf file looks like this:

## 참고 기관별 사업규모 및 예산

[중앙행정기관]

기관명		'24년 사업수	'23년 사업수	'24년 예산 (백만원)	'23년 예산 (백만원)
1	교육부	16	16	194,996	94,963
2	과학기술정보통신부	6	6	32,852	40,283
3	외교부	73	63	40,215	39,419
3-1	한국국제교류재단	37	41	42,514	44,664
4	통일부	6	6	1,831	2,386
5	법무부	3	3	15,068	14,346
6	국방부	7	8	6,165	7,221
7	행정안전부	3	3	594	574
8	문화체육관광부	21	22	185,478	145,049
9	농림축산식품부	6	7	3,048	4,268
10	보건복지부	7	7	6,497	8,557
11	환경부	1	1	1,888	1,427
12	고용노동부	1	1	1,264	1,529
13	여성가 <del>족</del> 부	6	7	1,531	2,748
14	국토교통부	4	4	2,394	2,394
15	중소벤처기업부	5	5	7,246	5,548
16	국가보훈부	1	1	8,774	3,637
17	법제처	2	2	327	327
18	해양수산부	1	1	100	100
19	재외동포청	5	-	22,289	=
합계		211	204	475,038	419,440

[지자체]

E: 1: 1: 1: 1					
기관명		'24년 사업수	'23년 사업수	'24년 예산 (백만원)	′23년 예산 (백만원)
1	경기도	25	14	21,558	3,899
2	강원특별자치도	10	11	78,593	11,024
3	충청북도	7	8	789	736
4	충청남도	10	10	2,508	1,731
5	전라북도	19	19	2,626	10,703
6	전라남도	13	13	2,962	6,917
7	경상북도	18	18	2709	3,314
8	경상남도	8	10	미정	1,408
9	제주특별자치도	23	24	4,433	7,343
10	서울특별시	31	31	10,005	9,628
11	부산광역시	36	35	3,017	2,355
12	대구광역시	11	11	316	321
13	인천광역시	26	25	5,516	5,008
14	광주광역시	22	26	3,487	6,459
15	대전광역시	38	44	3685	3,848
16	울산광역시	17	14	1,302	660
17	세종특별자치시	8	9	96	373
합계		322	322	143,602	75,727

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Figure 9.1: 2024 Public Diplomacy Comprehensive Implementation Plan, p.  $4\,$ 

#### 9.3 pdf Table Extraction

Let's try to extract the second table on page 4 of the pdf file. The table has the number of public diplomacy projects and budgets for first-tier local administration unit (hereafter, province\_city for short) in Korea. We will unlist each line as we did earlier so that we can see the table in a more readable way.

```
# Look at the first 10 lines of the 4th page
lines_pdf_4 <- pdf_text_all[4] |>
    # Split by newline character.
    str_split("\n") |>
    # Unlist
unlist()
```

First, let's look at the 29<sup>th</sup> and 30<sup>th</sup> lines for the column names in the pdf file.

```
lines_pdf_4[29:30]
[1] " '24 '23 '24 '23 "
[2] " "
```

The column names are the line number, province or city's name, project numbers for 2024 and 2023 respectively, and the budget for 2024 and 2023 in million Korean Won respectively. Let's use the following English column names that correspond to the Korean column names in the pdf file.

```
# Column names
col_names <- c("no", "province_city", "project_no_2024", "project_no_2023", "budget_2024",</pre>
```

By observing the lines\_pdf\_4 object using view(lines\_pdf\_4), we can see that the second table starts from the 32<sup>nd</sup> line and ends on the 48<sup>th</sup>. We will extract only those lines. We will use str\_trim "removes whitespace from start and end of string". We will also use str\_replace\_all to remove commas from each line to convert entries into numbers. We will then split each line based on two or more consecutive spaces (our string is "\s{2,}") using str\_split and simplify the result into a matrix. We will convert this matrix into a data frame with non-factor columns using data.frame(stringsAsFactors = FALSE). We will set the column names of the data frame using the col\_names vector that we created above. These explanations are also available in each step in the following code chunk.

```
# Select lines 32 to 48 from the lines_pdf_4 data frame
province_city_pd <- lines_pdf_4[32:48] |>
```

```
# Trim whitespace from both ends of each element in the selected rows
str_trim() |>
# Replace all commas with an empty string in each element
str_replace_all(",", "") |>
# Split each element based on 2 or more consecutive spaces and simplify into a matrix
str_split("\\s{2,}", simplify = TRUE) |>
# Convert the matrix into a data frame with non-factor columns
data.frame(stringsAsFactors = FALSE) |>
# Set column names for the data frame using the provided 'col_names' vector
setNames(col_names)
```

Let's rearrange the table (which is originally in alphabetical order) by descending order based on public diplomacy budgets in 2024.

```
province_city_pd |>
  arrange(desc(budget_2024))
```

	no	<pre>province_city project_no_2024</pre>	<pre>project_no_</pre>	2023 budget_2	2024 budget_2023
1	8	8	10		1408
2	17	8	9	96	373
3	3	7	8	789	736
4	2	10	11	78593	11024
5	13	26	25	5516	5008
6	9	23	24	4433	7343
7	15	38	44	3685	3848
8	14	22	26	3487	6459
9	12	11	11	316	321
10	11	36	35	3017	2355
11	6	13	13	2962	6917
12	7	18	18	2709	3314
13	5	19	19	2626	10703
14	4	10	10	2508	1731
15	1	25	14	21558	3899
16	16	17	14	1302	660
17	10	31	31	10005	9628

But these province\_city names are in Korean since the document was in Korean. Let's practice extracting a table from internet then to find English names for these Korean provinces or cities. As of May 6, 2024, Wikipedia's list of South Korea's administrative divisions seems to be correct. Let's extract the table there.

#### 9.4 html Table Extraction

We will use the <code>rvest</code> package to extract the table from the Wikipedia page. We will use the <code>read\_html</code> function to read the html content of the Wikipedia page. We will then use the <code>html\_node</code> function to select the table we want to extract. You can refer to <code>rvest</code> package for more information on how to extract what you want. We can use the xpath of the table we want to extract. You can find the xpath of the table by right-clicking on the table on the Wikipedia page and selecting "Inspect" or "Inspect Element" depending on your browser. You can then right-click on the highlighted html element in the "Elements" tab of the "Developer Tools" and select "Copy" -> "Copy XPath". The xpath of the table we want to extract is <code>//\*[@id="mw-content-text"]/div[1]/table[5]</code>. We will use the <code>html\_table</code> function to extract the table as a data frame. We will use the <code>fill = TRUE</code> argument to fill in the missing values in the table.

```
html <- read_html("https://en.wikipedia.org/wiki/Administrative_divisions_of_South_Korea")

table <- html |>
  html_node(xpath = '//*[@id="mw-content-text"]/div[1]/table[5]') |>
  html_table(fill = TRUE)
```

Let's look at the first 10 rows of the table.

```
head(table)
```

```
# A tibble: 6 x 9
  Code Emblem Name
                      Official English nam~1 Hangul Hanja Population 2020 Cens~2
  <chr> <lgl>
               <chr>
                      <chr>
                                              <chr>
                                                    <chr> <chr>
                                                  .mw-~ 9,586,195
1 KR-11 NA
               Seoul~ Seoul
2 KR-26 NA
               Busan~ Busan
                                                   ~ 3,349,016
3 KR-27 NA
                                                   ~ 2,410,700
               Daegu~ Daegu
4 KR-28 NA
               Inche~ Incheon
                                                   ~ 2,945,454
5 KR-29 NA
               Gwang~ Gwangju
                                                   ~ 1,477,573
6 KR-30 NA
               Daeje~ Daejeon
                                                   ~ 1,488,435
# i abbreviated names: 1: `Official English name[5]`,
    2: `Population 2020 Census`
# i 2 more variables: `Area (km2)` <chr>,
    `Population density 2022 (per km2)` <chr>
```

Perfect! Now, let's keep only the columns that we will need.

```
# Select columns 4 and 5 from the table
table <- table |>
    select(4:5)

# Let's change the English province_city column name.

table <- table |>
    rename(province_city_eng = `Official English name[5]`)
```

Let's hope that the Korean names in the Wikipedia table and the MOFA's pdf file are the same. Let's merge the two tables based on the Korean names.

```
# Merge the two tables based on the Korean names
province_city_pd_joined <- province_city_pd |>
   left_join(table, by = c("province_city" = "Hangul"))
```

Let's see if we have any missing values in the English names.

```
# Check for missing values in the English names
province_city_pd_joined |>
  filter(is.na(province_city_eng))
```

We almost got it! The only difference is (North Jeolla Province) in the MOFA's pdf file which is written as (Jeonbuk State) in the Wikipedia table. Let's fix this.

```
# Move the English name column next to the Korean name column, and remove the 'no' column
province_city_pd_joined <- province_city_pd_joined |>
    select(province_city, province_city_eng, everything(), -no)

# Fix the English name of

province_city_pd_joined <- province_city_pd_joined |>
    mutate(province_city_eng = ifelse(province_city == " ", "North Jeolla province_city", pd_joined |>
    mutate(province_city_eng = ifelse(province_city == " ", "North Jeolla province_city", pd_joined |>
```

#### 9.5 Text Analysis

#### 9.5.1 Word Frequency

This time let's look at all of the text in the 2024 Public Diplomacy Comprehensive Implementation Plan. We will combine all the text into a single character vector.

```
# Combine text
pdf_text <- str_c(pdf_text_all, collapse = " ")</pre>
```

We will now split the text into words using the str\_split function from the stringr package. We will then convert the result into a data frame with non-factor columns using the data.frame(stringsAsFactors = FALSE) function. We will set the column name of the data frame as word.

```
# Split the text into words
words <- pdf_text |>
    # Split the text into words
str_split("\\s+") |>
    # Convert the result into a data frame with non-factor columns
data.frame(stringsAsFactors = FALSE) |>
    # Set the column name of the data frame as "word"
setNames("word")
```

Let's look at the first 10 rows of the data frame.

```
head(words, 10)

word

1
2
2024
3
4
5
6
7
8
9
10
```

Now, let's count the frequency of each word in the text using the count function from the

dplyr package package. We will then arrange the result in descending order based on the frequency of the words.

```
# Count the frequency of each word
word_freq <- words |>
count(word, sort = TRUE)
```

Let's look at the first 10 rows of the data frame

```
head(word_freq, 10)
```

```
word n
1
            72
2
           - 55
3
       40
4
            33
5
            28
6
        , 22
7
          22
8
          18
          18
9
10
          17
```

This is not very useful. There are two main issues with Korean text. First, Korean text does not have consistent spacing between words. Second, Korean text has particles and other morphemes that are not words. We will address these issues now.

#### 9.5.2 Spacing in Korean Text

Let's get the spacing right in Korean text using the bitNLP package's get\_spacing function, which will add spaces between words in the Korean text. So, for example " " will become " ...

```
# Get the spacing right in Korean text
pdf_text_ko <- get_spacing(pdf_text)</pre>
```

Now, let's split the text into words again using the str\_split function from the stringr package.

```
# Split the text into words
words_ko <- pdf_text_ko |>
```

```
# Split the text into words
str_split("\\s+") |>
# Convert the result into a data frame with non-factor columns
data.frame(stringsAsFactors = FALSE) |>
# Set the column name of the data frame as "word"
setNames("word")
```

Let's analyze the word frequency in the text again.

```
# Count the frequency of each word
  word_freq_ko <- words_ko |>
    count(word, sort = TRUE)
  head(word_freq_ko, 10)
   word
          n
1
       175
2
      (97
3
      - 80
4
       73
5
      67
6
      62
      36
7
8
      35
9
       33
10
      30
```

We have many special characters in the text. Let's remove all characters except for Korean characters, spaces, English letters, and numbers using the str\_replace\_all function from the stringr package.

```
# Remove all characters except for Korean characters, spaces, English letters, and numbers
word_freq_ko <- pdf_text_ko |>
    # Remove all characters except Korean characters, English letters, numbers, and spaces
    str_replace_all("[^ - a-zA-Z0-9\\s]", "") |>
    # Split the cleaned text into words based on one or more spaces
    str_split("\\s+") |>
    # Convert the list result into a data frame with non-factor columns
    data.frame(stringsAsFactors = FALSE) |>
    # Set the column name of the data frame as "word"
    setNames("word")
```

Let's analyze the word frequency in the text again.

```
# Count the frequency of each word
  word_freq_ko <- word_freq_ko |>
    count(word, sort = TRUE)
  head(word_freq_ko, 10)
   word n
1
      73
2
     67
3
     62
4
     44
5
     37
     36
6
7
      35
8
     30
9
     29
10
     28
```

This is much better! We have removed the special characters and have more meaningful words in the text. Let's move on to morpheme analysis which makes more sense in Korean text analysis context.

#### 9.5.3 Morpheme Analysis in Korean Text

Let's analyze the morphemes in the Korean text using the morpho\_mecab function from the bitNLP package, which will extract morphemes from the Korean text.

```
# Analyze the morphemes in the Korean text
morphemes <- morpho_mecab(pdf_text_ko)</pre>
```

This creates a list of character vectors, where each element corresponds to a morpheme in the text. We can also combine all of the morphemes and tokenize them into a single character vector.

```
# Combine all the morphemes into a single character vector
morphemes_single <- morpho_mecab(pdf_text_ko, indiv = FALSE)</pre>
```

Now, let's split the text into words again this time by converting morphemes\_single into a data frame using the as.data.frame function. We will set the column name of the data frame as "word".

```
# Split the text into words
words_morphemes <- morphemes_single |>
   as.data.frame() |>
   # Set the column name of the data frame as "word"
   setNames("word")
```

We will now count the frequency of each morpheme in the text using the **count** function from the **dplyr** package package. We will then arrange the result in descending order based on the frequency of the morphemes.

```
# Count the frequency of each morpheme
  morpheme freq <- words morphemes |>
    count(word, sort = TRUE)
  head(morpheme_freq, 10)
   word
         n
1
     68
2
     62
3
     46
4
     39
5
     37
6
     30
7
     29
8
     28
9
     26
10
     25
```

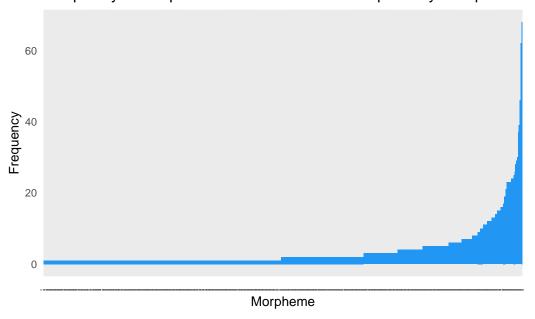
Now, this is more like it!

Let's visualize the frequency of the morphemes in the text using a bar plot. We will use the ggplot function from the ggplot2 package to create the plot. We will use the geom\_col function to add the bars to the plot. We will use the theme\_minimal function to set the theme of the plot to minimal. We will use the theme function to adjust the font size in the plot. We will set the font size to 10. We will use the labs function to add the title and labels to the plot.

```
# Visualize the frequency of the morphemes in the text

morpheme_freq |>
    # Create a bar plot
    ggplot(aes(x = reorder(word, n), y = n)) +
    geom_col(fill = "#2196f3") +
    theme_minimal() +
    theme(text = element_text(size = 10)) +
    labs(title = "Frequency of Morphemes in the 2024 Public Diplomacy Comprehensive Implement
        x = "Morpheme",
        y = "Frequency")
```

#### Frequency of Morphemes in the 2024 Public Diplomacy Comprehens



#### 9.5.4 Word Network in Korean Text

Let's analyze the word network in the Korean text using the tokenize\_noun\_ngrams function from the bitNLP package which builds on tidytext package. We will use the tokenize\_noun\_grams function to extract the noun word network from the Korean text.

```
# We can use a user-defined dictionary to improve the accuracy of the tokenization. We wil
dic_path <- system.file("dic", package = "bitNLP")</pre>
```

```
dic_file <- glue::glue("{dic_path}/buzz_dic.dic")

word_network <- tokenize_noun_ngrams(pdf_text_ko, simplify = TRUE, user_dic = dic_file, n
    as_tibble() |>
    setNames("paired_words")
```

Now, let's separate the paired words into two columns using the **separate** function from the **tidyr** package which is loaded as part of the **tidyverse** package. This will allow us to create bigrams from the paired words.

```
word_network_separated <- word_network |>
  separate(paired_words, c("word1", "word2"), sep = " ")
```

We will now count the frequency of each bigram in the text using the count function from the dplyr package package, which is also party of the tidyverse. We will then arrange the result in descending order based on the frequency of the bigrams.

```
# new bigram counts:
word_network_counts <- word_network_separated |>
count(word1, word2, sort = TRUE)
```

Korean text sometimes is not visible in the graph due to the font issue. This was the case in my Macbook. Let's set the font to one that supports Korean characters. We will use the extrafont package to set the font to one that supports Korean characters. We will use the font\_import function to import the fonts from the system. This may take some time. You only need to do it once. That's why I commented it. You can uncomment it in first usage.

```
# Load extrafont and register fonts
#font_import() # This might take a while if it's the first time you're running it
```

We will then use the <code>loadfonts</code> function to load the fonts. We will use the <code>fonts</code> function to display the available fonts and find one that supports Korean characters. We will set the font to one that supports Korean characters. For now, I have chosen "Arial Unicode MS" as the Korean font. You can replace it with a font from your system that supports Korean characters if necessary.

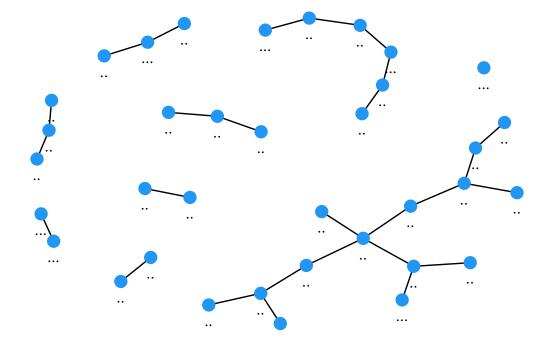
```
#loadfonts(device = "all")
# Display available fonts, find one that supports Korean
#fonts()
```

```
# Set the font to one that supports Korean characters
korean_font <- "Arial Unicode MS" # Replace with a font from your system that supports Korean_font</pre>
```

We will now create a graph from the bigram counts using the <code>graph\_from\_data\_frame</code> function from the <code>igraph</code> package. We will use the <code>ggraph</code> function from the <code>ggraph</code> package to create the graph. We will use the <code>geom\_edge\_link</code> function to add the edges to the graph. We will use the <code>geom\_node\_point</code> function to add the nodes to the graph. We will use the <code>geom\_node\_text</code> function to add the labels to the nodes in the graph. We will set the font to the Korean font that we set earlier. We will then adjust the font in the graph. Here, <code>n >= 6</code> is used to filter out bigrams that appear less than 6 times. You can adjust this number as needed. You can check out <code>ggraph</code> layout options here.

```
word_network_select <- word_network_counts |>
  filter(n >= 6) |>
  graph_from_data_frame() |>
  ggraph(layout = "fr") +
  geom_edge_link(aes()) +
  geom_node_point(color = "#2196f3", size = 4) +
  geom_node_text(aes(label = name), family = korean_font, vjust = 2, size = 4) + # Set fatheme_void()
```

word\_network\_select



#### 9.5.5 Sentiment Analysis

#### 9.5.6 Topic Modeling

#### 9.6 Korean Tweet Analysis

#### 9.7 Further Readings

#### 9.8 References

#### 9.9 Session Info

```
sessionInfo()
```

```
R version 4.4.0 (2024-04-24)
Platform: aarch64-apple-darwin20
Running under: macOS Sonoma 14.4.1
Matrix products: default
        /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRblas.0.dylib
LAPACK: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRlapack.dylib;
locale:
[1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
time zone: Asia/Seoul
tzcode source: internal
attached base packages:
[1] stats
             graphics grDevices utils
                                            datasets methods
                                                                base
other attached packages:
 [1] extrafont_0.19
                       ggraph_2.2.1
                                                           tidytext_0.4.2
                                         igraph_2.0.3
                       pdftools_3.4.0
 [5] rvest_1.0.4
                                         lubridate_1.9.3
                                                           forcats_1.0.0
 [9] stringr_1.5.1
                       dplyr_1.1.4
                                         purrr_1.0.2
                                                           readr_2.1.5
[13] tidyr_1.3.1
                       tibble_3.2.1
                                         ggplot2_3.5.1
                                                           tidyverse_2.0.0
[17] bitNLP_1.4.3.9000
```

loaded via a namespace (and not attached):

[1]	tidyselect_1.2.1	viridisLite_0.4.2	farver_2.1.1
[4]	viridis_0.6.5	fastmap_1.1.1	tweenr_2.0.3
[7]	<pre>janeaustenr_1.0.0</pre>	promises_1.3.0	shinyjs_2.1.0
[10]	digest_0.6.35	timechange_0.3.0	mime_0.12
[13]	lifecycle_1.0.4	qpdf_1.3.3	tokenizers_0.3.0
[16]	magrittr_2.0.3	compiler_4.4.0	rlang_1.1.3
[19]	sass_0.4.9	tools_4.4.0	utf8_1.2.4
[22]	knitr_1.46	labeling_0.4.3	askpass_1.2.0
[25]	<pre>graphlayouts_1.1.1</pre>	htmlwidgets_1.6.4	curl_5.2.1
[28]	xm12_1.3.6	miniUI_0.1.1.1	ngram_3.2.3
[31]	withr_3.0.0	grid_4.4.0	polyclip_1.10-6
[34]	fansi_1.0.6	xtable_1.8-4	<pre>colorspace_2.1-0</pre>
[37]	extrafontdb_1.0	scales_1.3.0	MASS_7.3-60.2
[40]	tinytex_0.50	cli_3.6.2	rmarkdown_2.26
[43]	generics_0.1.3	RcppParallel_5.1.7	rstudioapi_0.16.0
[46]	httr_1.4.7	tzdb_0.4.0	cachem_1.0.8
[49]	ggforce_0.4.2	<pre>RcppMeCab_0.0.1.2</pre>	parallel_4.4.0
[52]	${\tt rhandsontable\_0.3.8}$	vctrs_0.6.5	Matrix_1.7-0
[55]	jsonlite_1.8.8	hms_1.1.3	ggrepel_0.9.5
[58]	jquerylib_0.1.4	shinyBS_0.61.1	glue_1.7.0
[61]	stringi_1.8.3	gtable_0.3.5	later_1.3.2
[64]	munsell_0.5.1	pillar_1.9.0	${\tt htmltools\_0.5.8.1}$
[67]	R6_2.5.1	tidygraph_1.3.1	evaluate_0.23
[70]	shiny_1.8.1.1	lattice_0.22-6	SnowballC_0.7.1
[73]	memoise_2.0.1	DataEditR_0.1.5	httpuv_1.6.15
[76]	bslib_0.7.0	Rcpp_1.0.12	Rttf2pt1_1.3.12
[79]	<pre>gridExtra_2.3</pre>	xfun_0.43	pkgconfig_2.0.3

# 10 Statistical Analysis

# 11 Storytelling with Quarto

# 12 Productivity Tools

Setting up Github.

Creating a new Github project.

 ${\bf Copilot\ etc.}$ 

# 13 Working with API to get Korean Data

WDI etc. readily available packages

Creating your own API

https://httr2.r-lib.org/articles/wrapping-apis.html

https://www.andrewheiss.com/blog/2024/01/12/diy-api-plumber-quarto-ojs/\_book/

# 14 #kdiplo and #kdata

# 15 R for Korean Studies Bootcamps: 3-Days to Jumpstart R with Korean Studies-based Examples

## References