# Assignment 3

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#### **Packages**

```
library(xml2)
library(rvest)
library(tidyverse)
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr 1.1.3 v readr
                                 2.1.4
## v forcats 1.0.0 v stringr 1.5.0
## v ggplot2 3.4.3 v tibble 3.2.1
## v lubridate 1.9.3
                    v tidyr
                                  1.3.0
## v purrr
             1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x readr::guess_encoding() masks rvest::guess_encoding()
## x dplyr::lag()
masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(robotstxt)
library(janeaustenr)
library(dplyr)
library(stringr)
library(tidytext)
library(dplyr)
library(tidytext)
```

# Web Scraping

#### Wikipedia

```
# To look at whether we are allowed to do web scraping
paths_allowed("https://en.wikipedia.org/wiki/Grand_Boulevard,_Chicago")

## en.wikipedia.org

## [1] TRUE
```

```
# give URL and read HTML
url <- read_html("https://en.wikipedia.org/wiki/Grand_Boulevard,_Chicago")
str(url)

## List of 2
## $ node:<externalptr>
## $ doc :<externalptr>
## - attr(*, "class") = chr [1:2] "xml_document" "xml_node"
```

#### 1. Grab the html elements

```
#xpath
nds <- html_elements(url, xpath = '//*[contains(concat( " ", @class, " " ), concat( " ", "us-census-pop</pre>
# storing html
tbl <- html_text(nds)</pre>
historical_pop <- tbl[5:44] %>% matrix(ncol=4, byrow = TRUE) %>% as.data.frame()
# change the variable names
names(historical_pop) <- tbl[1:4]</pre>
# drop the third column
historical_pop <- historical_pop[, c(1, 2, 4)]
# by Sungjoo: X path is different (Oakland, Kenwood, Hyde_Park & Armour_Square, Douglas, Fuller_Park,
#https://en.wikipedia.org/wiki/Armour_Square,_Chicago
#https://en.wikipedia.org/wiki/Douglas,_Chicago
#https://en.wikipedia.org/wiki/Oakland,_Chicago
#https://en.wikipedia.org/wiki/Fuller_Park,_Chicago
#https://en.wikipedia.org/wiki/Washington_Park_(community_area),_Chicago
#https://en.wikipedia.org/wiki/Hyde_Park,_Chicago
#https://en.wikipedia.org/wiki/Kenwood,_Chicago
```

## 2. Expanding to More Pages

```
historical_pops <- historical_pop
directions <- c("Oakland,_Chicago", "Kenwood,_Chicago", "Hyde_Park,_Chicago")
population <- data.frame()

# for loop 1 (Oakland, Kenwood, Hyde_Park)
for (i in directions) {
    url2 <- paste0("https://en.wikipedia.org/wiki/", i)
    src2 <- read_html(url2)

# xpath
    nds2 <- html_elements(src2, xpath = '//*[contains(concat( " ", @class, " " ), concat( " ", "us-census names2 <- html_text(nds2)

# storing html</pre>
```

```
tbl2 <- html_text(nds2)
population <- tbl2[5:44] %>% matrix(ncol=4, byrow = TRUE) %>% as.data.frame()

# change the variable names
names(population) <- tbl2[1:4]

# drop the third column
population <- population[, c(2, 4)]

#part <- data.frame(names2, population)
historical_pops <- cbind(historical_pops, population)
}</pre>
```

## 3. Scraping and Analyzing Text Data

(1) Creating a corpus

## \$ :List of 2

```
#xpath
nds_text <- html_elements(url, xpath = '//p')</pre>
str(nds_text)
## List of 8
## $ :List of 2
## ..$ node:<externalptr>
   ..$ doc :<externalptr>
## ..- attr(*, "class")= chr "xml_node"
## $ :List of 2
##
   ..$ node:<externalptr>
   ..$ doc :<externalptr>
    ..- attr(*, "class")= chr "xml_node"
##
## $ :List of 2
##
   ..$ node:<externalptr>
##
    ..$ doc :<externalptr>
    ..- attr(*, "class")= chr "xml_node"
##
##
   $ :List of 2
##
   ..$ node:<externalptr>
    ..$ doc :<externalptr>
##
##
    ..- attr(*, "class")= chr "xml_node"
## $ :List of 2
    ..$ node:<externalptr>
##
     ..$ doc :<externalptr>
    ..- attr(*, "class")= chr "xml_node"
## $ :List of 2
    ..$ node:<externalptr>
    ..$ doc :<externalptr>
##
    ..- attr(*, "class")= chr "xml_node"
##
## $ :List of 2
   ..$ node:<externalptr>
##
##
    ..$ doc :<externalptr>
   ..- attr(*, "class")= chr "xml_node"
```

```
..$ node:<externalptr>
##
   ..$ doc :<externalptr>
    ..- attr(*, "class")= chr "xml_node"
## - attr(*, "class")= chr "xml_nodeset"
description <- html_text(nds_text)</pre>
description
## [1] "\n"
## [2] "Grand Boulevard on the South Side of Chicago, Illinois, is one of the city's Community Areas. T.
## [3] "This is one of the two community areas that encompass the Bronzeville neighborhood, with the ot
## [4] "The Harold Washington Cultural Center is one of its newer and more famous buildings. It arose of
## [5] "According to a 2018 US Census American Community Survey, there were 22,784 people and 10,383 ho
## [6] "Grand Boulevard is part of City of Chicago School District #299 and City Colleges of Chicago Di
## [7] "The Chicago Transit Authority operates the Chicago \"L\" system in the Grand Boulevard communit
## [8] "The Grand Boulevard community area has supported the Democratic Party in the past two president
description <- description %>% paste(collapse = ' ')
```

(2) Grab the descriptions of the various communities areas using for loop

```
# creating a corpus
directions <- c("Grand_Boulevard,_Chicago", "Oakland,_Chicago", "Kenwood,_Chicago", "Hyde_Park,_Chicago
descriptions <- data.frame()</pre>
for (i in directions) {
  url2 <- paste0("https://en.wikipedia.org/wiki/", i)</pre>
  src2 <- read_html(url2)</pre>
  nds_text <- html_elements(src2, xpath = '//p')</pre>
  description <- html_text(nds_text)</pre>
  description <- description %>% paste(collapse = ' ') %>%
    as.data.frame()
  descriptions <- rbind(descriptions, description)</pre>
```

(3) Create tokens using unnest tokens. Make sure the data is in one token per row format. Remove any stop words within the data. What are the most common words used overall?

The most common words used overall is 'park' (n=89).

##

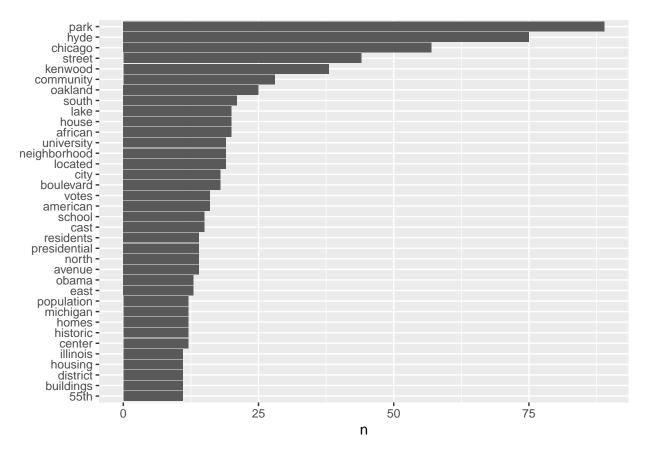
```
# create tokens using unnest_tokens
text_df <- tibble(location = c("Grand_Boulevard", "Oakland", "Kenwood", "Hyde_Park"), text = description</pre>
token <- text_df %>%
 unnest_tokens(word, text)
```

```
# remove stop words
data("stop_words")
token <- token %>%
  anti_join(stop_words)
```

## Joining with 'by = join\_by(word)'

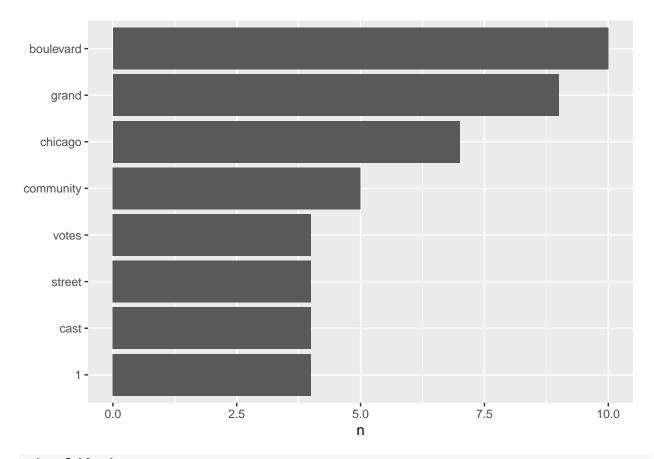
```
# find the most common words in all the books as a whole
token_count <- token %>%
    count(word, sort = TRUE)

# plot the most common words overall
token %>%
    count(word, sort = TRUE) %>%
    filter(n > 10) %>%
    mutate(word = reorder(word, n)) %>%
    ggplot(aes(n, word)) +
    geom_col() +
    labs(y = NULL)
```

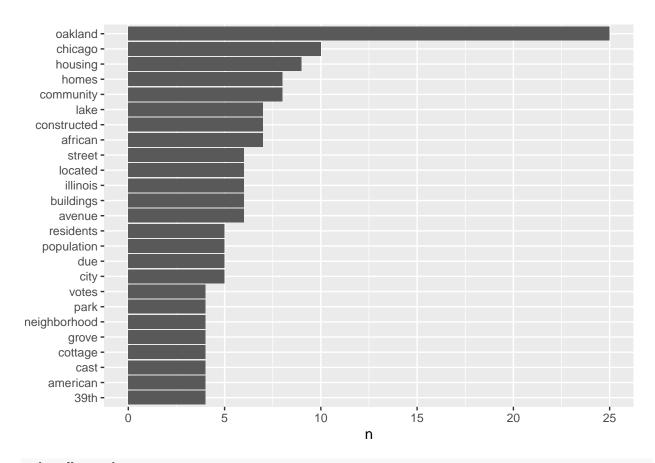


(4) Plot the most common words within each location. What are some of the similarities between the locations? What are some of the differences?

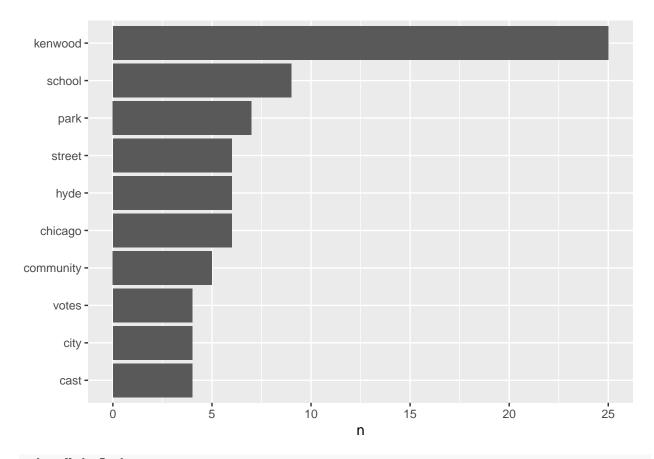
```
# plot the most common words within Grand_Boulevard
token_Grand_Boulevard <- token %>%
 filter(location == "Grand_Boulevard") %>%
  count(word, sort = TRUE) %>%
 filter(n > 3) %>%
 mutate(word = reorder(word, n)) %>%
 ggplot(aes(n, word)) +
 geom col() +
 labs(y = NULL)
# plot the most common words within Oakland
token_Oakland <- token %>%
 filter(location == "Oakland") %>%
  count(word, sort = TRUE) %>%
 filter(n > 3) %>%
 mutate(word = reorder(word, n)) %>%
  ggplot(aes(n, word)) +
 geom_col() +
 labs(y = NULL)
# plot the most common words within Kenwood
token_Kenwood <- token %>%
  filter(location == "Kenwood") %>%
  count(word, sort = TRUE) %>%
  filter(n > 3) %>%
  mutate(word = reorder(word, n)) %>%
  ggplot(aes(n, word)) +
  geom_col() +
 labs(y = NULL)
# plot the most common words within Hyde_Park
token_Hyde_Park <- token %>%
 filter(location == "Hyde_Park") %>%
  count(word, sort = TRUE) %>%
 filter(n > 5) %>%
 mutate(word = reorder(word, n)) %>%
 ggplot(aes(n, word)) +
 geom col() +
 labs(y = NULL)
par(mfrow=c(2,2))
token_Grand_Boulevard
```



token\_Oakland



token\_Kenwood



token\_Hyde\_Park

