

Assignment 3

Sungjoo Cho, Catherine (Kate) Lamoreaux

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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 (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(robotstxt)
library(janeaugstenr)
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-population" ))]')

# storing html
tbl <- html_text(nds)
historical_pop <- tbl[5:44] %>% matrix(ncol=4, byrow = TRUE) %>% as.data.frame()

# change the variable names
names(historical_pop) <- tbl[1:4]

# 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, Washington_Park_(community_area), Chicago)
#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-population" ))]')
  names2 <- html_text(nds2)

  # storing html
```

```
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)
}
```

3. Scraping and Analyzing Text Data

(1) Creating a corpus

```
#xpath
nds_text <- html_elements(url, xpath = '//p')
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"
```

```
description <- html_text(nds_text)
description
```

```
description <- description %>% paste(collapse = ' ')
```

```
# creating a corpus
directions <- c("Grand_Boulevard,_Chicago", "Oakland,_Chicago", "Kenwood,_Chicago", "Hyde_Park,_Chicago")
descriptions <- data.frame()

for (i in directions) {
  url2 <- paste0("https://en.wikipedia.org/wiki/", i)
  src2 <- read_html(url2)

  #xpath
  nds_text <- html_elements(src2, xpath = '//p')

  description <- html_text(nds_text)
  description <- description %>% paste(collapse = ' ') %>%
    as.data.frame()

  descriptions <- rbind(descriptions, description)
}
```

```
# create tokens using unnest_tokens
text_df <- tibble(location = c("Grand_Boulevard", "Oakland", "Kenwood", "Hyde_Park"), text = description)

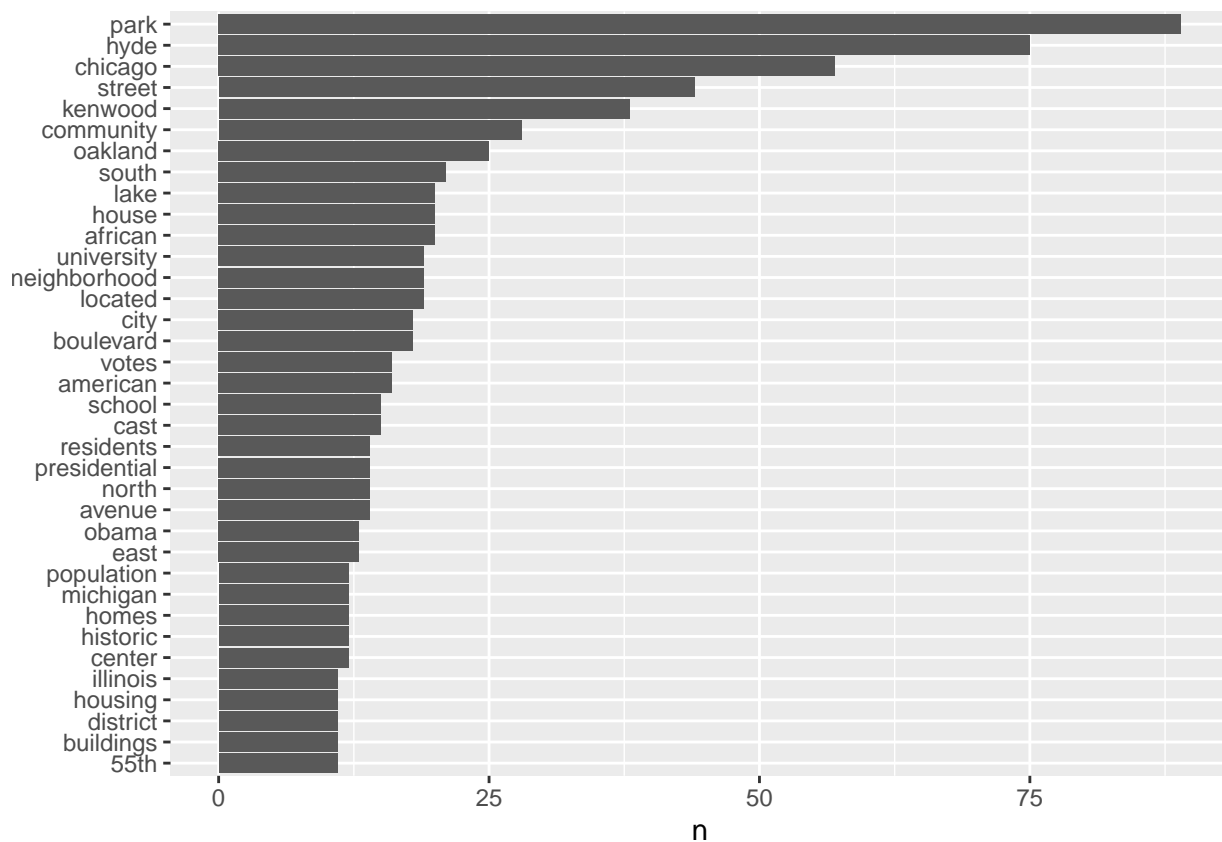
token <- text_df %>%
  unnest_tokens(word, text)
```

```
# 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)

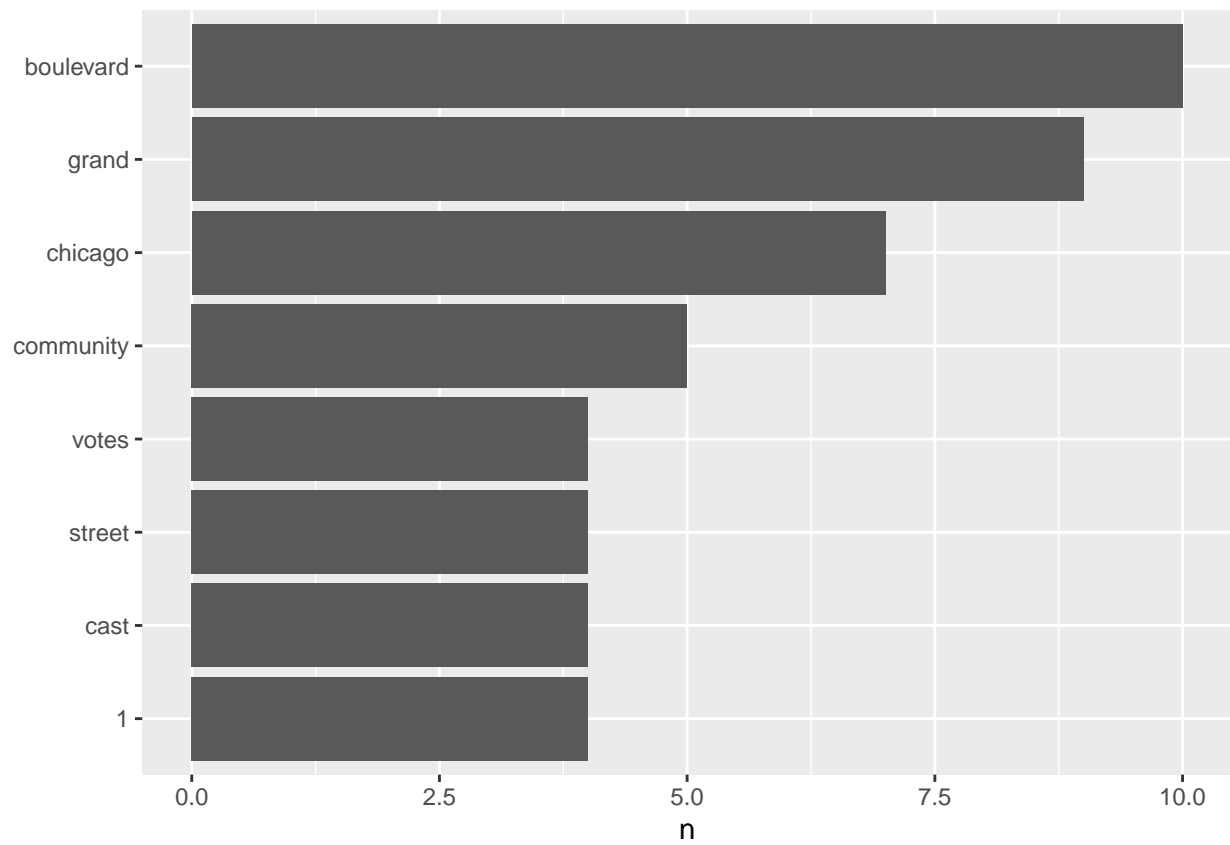
# plot the most common words within Oakland
token_Oakland <- token %>%
  filter(location == "Oakland") %>%
  count(word, sort = TRUE) %>%
  filter(n > 3) %>%
  mutate(word = reorder(word, n)) %>%
  ggplot(aes(n, word)) +
  geom_col() +
  labs(y = NULL)

# 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)

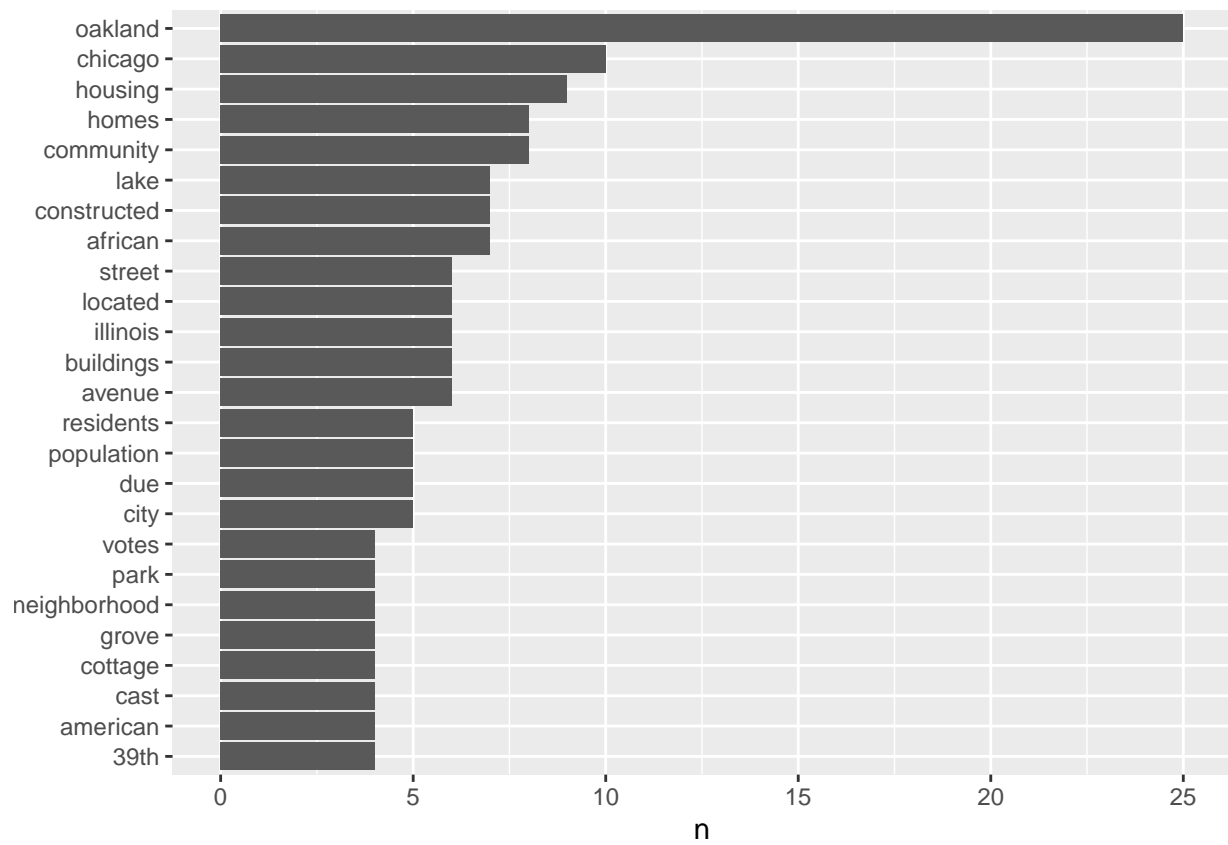
# plot the most common words within Hyde_Park
token_Hyde_Park <- token %>%
  filter(location == "Hyde_Park") %>%
  count(word, sort = TRUE) %>%
  filter(n > 5) %>%
  mutate(word = reorder(word, n)) %>%
  ggplot(aes(n, word)) +
  geom_col() +
  labs(y = NULL)

par(mfrow=c(2,2))
token_Grand_Boulevard

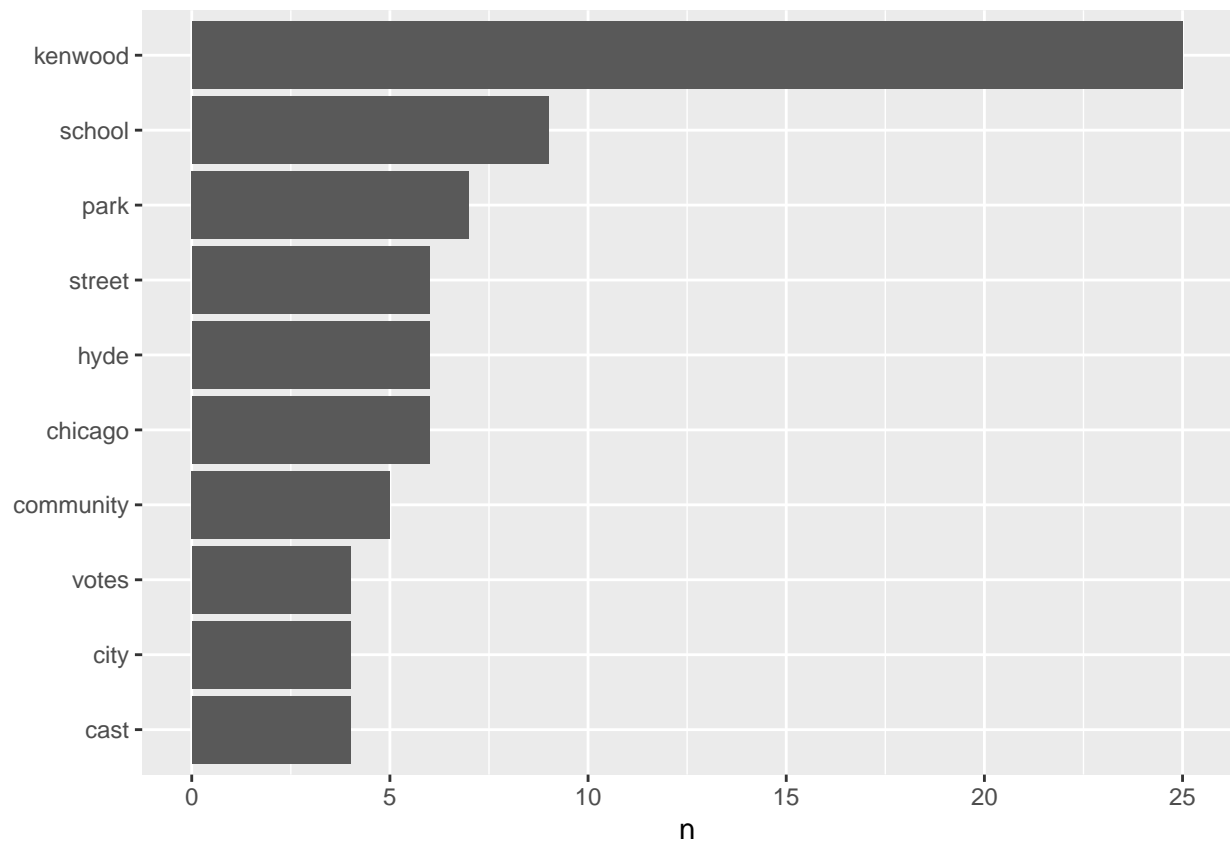
```



token_Oakland



token_Kenwood



token_Hyde_Park

