Automated Web Scraping with R

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Who am I?

Resul Umit

- post-doctoral researcher in political science at the University of Oslo
- teaching and studying representation, elections, and parliaments
 - a recent publication: Parliamentary communication allowances do not increase electoral turnout or incumbents' vote share

- teaching workshops, also on
 - writing reproducible research papers
 - version control and collaboration
 - working with Twitter data
 - creating academic websites
- more information available at resulumit.com

The Workshop — Overview

- One day, on how to automate the process of extracting data from websites
 - 200+ slides, 75+ exercises
 - a demonstration website for practice
- Designed for researchers with basic knowledge of R programming language
 - does not cover programming with R
 - e.g., we will use existing functions and packages
 - ability to work with R will be very helpful
 - but not absolutely necessary this ability can be developed during and after the workshop as well

The Workshop — Motivation

- Data available on websites provide attractive opportunities for academic research
 - o e.g., parliamentary websites were the main source of data for my PhD
- Acquiring such data requires
 - either a lot of resources, such as time
 - or a set of skills, such as automated web scraping
- Typically, such skills are not part of academic training
 - for my PhD, I hand-visited close to 3000 webpages to collect data manually
 - on members of ten parliaments
 - multiple times, to update the dataset as needed

The Workshop — Motivation — Aims

- To provide you with an understanding of what is ethically possible
 - we will cover a large breath of issues, not all of it is for long-term memory
 - hence the slides are designed for self study as well
 - o awareness of what is ethical and possible, Google, and perseverance are all you need
- To start you with acquiring and practicing the skills needed
 - practice with the demonstration website
 - plenty of data, stable structure, and an ethical playground
 - start working on a real project

The Workshop — Contents

Part 1. Getting the Tools Ready

• e.g., installing packages

Part 2. Preliminary Considerations

• e.g., ethics of web scraping

Part 3. Webpage Source Code

• e.g., elements, tags, and classes

Part 4. Scraping Static Pages

• e.g., getting text from an element

Part 5. Scraping Dynamic Pages

• e.g., clicking before scraping

To the list of references.

The Workshop — Organisation

- Sit in groups of two Breakout in groups of two for exercises
 - participants learn as much from their partner as from instructors
 - switch partners after every other part
 - leave your breakout room manually, when everyone in the group is ready
- Type, rather than copy and paste, the code that you will find on these slides
 - typing is a part of the learning process
 - slides are, and will remain, available at resulumit.com/teaching/scrp_workshop.html
- When you have a question
 - ask your partner
 - google together
 - o ask me

The Workshop — Organisation — Slides

03:00

Slides with this background colour indicate that your action is required, for

- setting the workshop up
 - e.g., installing R
- completing the exercises
 - e.g., downloading tweets
 - there are 75+ exercises
 - these slides have countdown timers
 - as a guide, not to be followed strictly

The Workshop — Organisation — Slides

- Codes and texts that go in R console or scripts appear as such in a different font, on gray background
 - long codes and texts will have their own line(s)

```
# read in the tweets dataset
df <- read_rds("tweets.rds") %>%

# split the variable text, create a new variable called da_tweets
  unnest_tokens(output = da_tweets, input = text, token = "tweets") %>%

# remove rows that match any of the stop words as stored in the stop_words dataset
  anti_join(stop_words, by = c("da_tweets" = "word"))
```

The Workshop — Organisation — Slides

- Codes and texts that go in R console or scripts appear as such in a different font, on gray background
 - long codes and texts will have their own line(s)
- Results that come out as output appear as such in the same font, on green background
 - except very obvious results, such as figures and tables
- Specific sections are highlighted yellow as such for emphasis
 - these could be for anything codes and texts in input, results in output, and/or texts on slides
- The slides are designed for self-study as much as for the workshop
 - *accessible*, in substance and form, to go through on your own

Part 1. Getting the Tools Ready

Workshop Slides — Access on Your Browser

- Having the workshop slides* on your own machine might be helpful
 - flexibility to go back and forward on your own
 - especially while in a breakout room
 - ability to scroll across long codes on some slides
- Access at https://resulumit.com/teaching/scrp_workshop.html
 - will remain accessible after the workshop
 - might crash for some Safari users
 - if using a different browser application is not an option, view the PDF version of the slides on GitHub

^{*} These slides are produced in R, with the xaringan package (Xie, 2020).

Demonstration Website — Explore on Your Browser

05:00

- There is a demonstration website for this workshop
 - available at https://parliament-luzland.netlify.app/
 - includes fabricated data on the imaginary Parliament of Luzland
 - o provides us with plenty of data, stable structure, and an ethical playground
- Using this demonstration website for practice is recommended
 - tailored to exercises, no ethical concern
 - but not compulsory use a different one if you prefer so
- Explore the website now
 - see the four sections
 - o click on the links to see an individual page for
 - states, constituencies, members, and documents

R — Download from the Internet and Install

- Programming language of this workshop
 - created for data analysis, extending for other purposes
 - e.g., accessing websites
 - allows for all three steps in one environment
 - accessing websites; scraping and processing data
 - an alternative: python
- Download R from https://cloud.r-project.org
 - optional, if you have it already installed but then consider updating*
 - the R. version.string command checks the version of your copy
 - compare with the latest official release at https://cran.r-project.org/sources.html

^{*}The same applies to all software that follows — consider updating if you have them already installed. This ensures everyone works with the latest, exactly the same, tools.

RStudio — Download from the Internet and Install

- Optional, but highly recommended
 - facilitates working with R
- A popular integrated development environment (IDE) for R
 - an alternative: GNU Emacs

- Download RStudio from https://rstudio.com/products/rstudio/download
 - choose the free version
 - to check for any updates, follow from the RStudio menu:
 - Help -> Check for Updates

RStudio Project — Create from within RStudio

- RStudio allows for dividing your work with R into separate projects
 - each project gets dedicated workspace, history, and source documents
 - this page has more information on why projects are recommended
- Create a new RStudio project for for this workshop, following from the RStudio menu:
 - File -> New Project -> New Directory -> New Project
- Choose a location for the project with Browse...
 - avoid choosing a synced location, e.g., Dropbox
 - likely to cause warning and/or error messages
 - if you must, pause syncing, or add an sync exclusion

R Packages — Install from within RStudio*

02:00

Install the packages that we need

^{*} You may already have a copy of one or more of these packages. In that case, I recommend updating by re-installing them now.

R Packages — Install from within RStudio

Install the packages that we need

We will use

- rvest (Wickham, 2021), for scraping websites
- RSelenium (Harrison, 2020), for browsing the web programmatically
- robotstxt (Meissner & Ren, 2020), for checking permissions to scrape websites
- polite (Perepolkin, 2019), for compliance with permissions to scrape websites

R Packages — Install from within RStudio

- tidyverse (Wickham & RStudio 2019), for various tasks
 - including data manipulation, visualisation
 - alternative: e.g., base R

• tidytext (Robinson & Silge, 2021), for working with text as data

Java — Download from the Internet and Install

- A language and software that RSelenium needs
 - for automation scripts
- Download Java from https://www.java.com/en/download/
 - requires restarting any browser that you might have open

Chrome — Download from the Internet and Install

- A browser that facilitates web scraping
 - favoured by RSelenium and most programmers
- Download Chrome from https://www.google.com/chrome/

ScrapeMate — Add Extension to Browser

- An open source software extension to Chrome, Firefox
 - facilitates selecting what to scrape from a webpage
 - optional, but highly recommended
- Add the extension to your preferred browser
 - for Chrome, search at https://chrome.google.com/webstore/category/extensions
 - for Firefox, search at https://addons.mozilla.org/
- If you cannot use Chrome or Firefox
 - drag and drop the following link to your bookmarks bar: SelectorGadget
 - another similar but older open source software with the same functionality

Other Resources*

- RSelenium vignettes
 - available at https://cran.r-project.org/web/packages/RSelenium/vignettes/basics.html
- R for Data Science (Wickham & Grolemund, 2019)
 - open access at https://r4ds.had.co.nz
- Text Mining with R: A Tidy Approach (Silge & Robinson, 2017)
 - open access at tidytextmining.com
 - o comes with a course website where you can practice

^{*} I recommend these to be consulted not during but after the workshop.

Part 2. Preliminary Considerations

Considerations — the Law

- Web scraping might be illegal
 - depending on who is scraping what, why, how and under which jurisdiction
 - o reflect, and check, before you scrape
- Web scraping might be more likely to be illegal if, for example,
 - it is harmful to the source
 - commercially
 - e.g., scraping a commercial website to create a rival website
 - physically
 - e.g., scraping a website so hard and fast that it collapses
 - it gathers data that is
 - under copyright
 - not meant for the public to see
 - then used for financial gain

Considerations — the Ethics

- Web scraping might be unethical
 - even when it is legal
 - o depending on who is scraping what, why, and how
 - reflect before you scrape
- Web scraping might be more likely to be unethical if, for example,
 - ∘ it is edging towards illegal
 - it does not respect the restrictions
 - as defined in robots.txt files
 - it harvests data
 - that is otherwise available to download, e.g., through APIs
 - without purpose, at dangerous speed, repeatedly

Considerations — the Ethics — robots.txt

- Most websites declare a robots exclusion protocol
 - making their rules known with respect to programmatic access
 - who is (not) allowed to scrape what, and sometimes, at what speed
 - within robots.txt files
 - available at, e.g., www.websiteurl.com/robots.txt
- The rules in robots, txt cannot not enforced
 - but should be respected for ethical reasons

- The language in robots.txt files is specific but intuitive
 - easy to read and understand
 - the robotstxt package makes it even easier

- It has pre-defined keys, most importantly
 - User-agent indicates who the protocol is for
 - Allow indicates which part(s) of the website can be scraped
 - Disallow indicates which part(s) must not be scraped
 - Crawl-delay indicates how fast the website could be scraped
- In case you write your own protocol one day, note that
 - the keys start with capital letters
 - they are followed by a colon:

```
User-agent:
Allow:
Disallow:
Crawl-delay:
```

- Websites define their own values
 - after the colon and a white space
- Note that
 - * indicates the protocol is for everyone
 - / indicates all sections and pages
 - /about/ indicates a specific path
 - values for Crawl-delay indicate seconds
 - this website allows anyone to scrape, provided that
 - /about/ is left out, and
 - the website is accessed at 5-seconds intervals

```
User-agent: *
Allow: /
Disallow: /about/
Crawl-delay: 5
```

Files might include optional comments, written after he number sign #

```
# thank you for respecting our protocol

User-agent: *
Allow: /
Disallow: /about/
Crawl-delay: 5  # five second delay, to ensure our servers are not overloaded
```

The protocol of this website only applies to Google

- Google is allowed to scrape everything
- there is no defined rule for anyone else

User-agent: googlebot
Allow: /

The protocol of this website only applies to Google

- Google is disallowed to scrape two specific paths
 - with no limit on speed
- there is no defined rule for anyone else

User-agent: googlebot
Disallow: /about/
Disallow: /history/

This website has different protocols for different agents

- Google is allowed to scrape everything, with a 5-second delay
- Bing is not allowed to scrape anything
- everyone else can scrape the section or page located at www.websiteurl/about/

```
User-agent: googlebot
Allow: /
Crawl-delay: 5

User-agent: bing
Disallow: /
User-agent: *
Allow: /about/
```

Considerations — the Ethics — robotstxt

- The robotstxt packages facilitates checking website protocols
 - from within R no need to visit websites via browser
 - o provides functions to check, among others, the rules for specific paths and/or agents
- There are two main functions
 - robotstxt, which gets complete protocols
 - paths_allowed, which checks protocols for one or more specific paths

Considerations — the Ethics — robotstxt

Use the robotstxt function to get a protocol

- supply a base url with the domain argument
 - as a string
 - probably the only argument that you will need

```
robotstxt(
  domain = NULL,
  ...
)
```

Considerations — the Ethics — robotstxt

```
library(robotstxt)
robotstxt(domain = "https://parliament-luzland.netlify.app")
```

```
User-agent: googlebot
Disallow: /states/
User-agent: *
Allow: /
$robexclobj
<Robots Exclusion Protocol Object>
$bots [1] "googlebot" "*"
```

Considerations — the Ethics — robotstxt

Use the paths_allowed function to checks protocols for one or more specific paths

- supply a base url with the domain argument
- path and bot are the other important arguments
 - notice the default values
- leads to either TRUE (allowed to scrape) or FALSE (not allowed)

```
paths_allowed(
  domain = "auto",
  paths = "/",
  bot = "*",
  ...
)
```

Considerations — the Ethics — robotstxt

```
paths_allowed(domain = "https://parliament-luzland.netlify.app")
```

[1] TRUE

[1] TRUE TRUE

[1] FALSE TRUE

Exercises

07:30

- 1) Check the protocols for https://www.theguardian.com
 - via a browser and with the robotstxt function
 - compare what you see
- 2) Check a path with the paths_allowed function
 - such that it will return FALSE

- 3) Check the protocols for any website that you might wish to scrape
 - with the robotstxt function

Considerations — the Ethics — Speed

- Websites are designed for visitors with human-speed in mind
 - o computer-speed visits can overload servers, depending on their bandwidth
 - popular websites might have more bandwidth
 - but, they might attract multiple scrapers at the same time
- Waiting a little between two visits makes scraping more ethical
 - waiting time may or may not be defined in the protocol
 - lookout for, and respect, the Crawl-delay key in robots.txt
 - Part 4 covers how to wait

- Not waiting enough might lead to a ban
 - by site owners, administrators
 - for IP addresses with undesirably high number of visits in a short period of time

Considerations — the Ethics — Purpose

Ideally, we scrape for a purpose

- e.g., for academics, to answer one or more research questions, test hypotheses
 - developed prior to data collection, analysis
 - based on, e.g., theory, claims, observations
 - o perhaps, even pre-registered
 - e.g., at OSF Registries

Considerations — Data Storage

Scraped data frequently requires

- large amounts of digital storage space
 - internet data is typically big data
- private, safe storage spaces
 - due to local rules, institutional requirements

Part 3. Webpage Source Code

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Webpage Source Code — Overview

- Webpages include more than what is immediately visible to visitors
 - not only text, images, links
 - but also code for structure, style, and functionality interpreted by browsers first
 - HTML provides the structure
 - CSS provides the style
 - JavaScript provides functionality, if any
- Web scraping requires working with the source code
 - even when harvesting the visible only
 - but source code is rarely a nuisance
 - allows choosing one or more desired parts of the visible
 - e.g., text in table and/or bold only
 - offers more, invisible, data
 - e.g., URLs hidden under text

Webpage Source Code — View in Browser

The Ctrl + U shortcut is to display source code — alternatively, right click and View Page Source

Webpage Source Code — View in Browser — DOM

Browsers also offer putting source codes in a structure

• known as DOM (document object model), initiated by the F12 key on Chrome

Exercises

05:00

- 4) View the source code of a page
 - as plain code and as in DOM
 - compare the look of the two
- 5) Search for a word or a phrase in source code
 - copy from the front-end page
 - search in plain text code or in DOM
 - ∘ the Ctrl + F
 - compare the look of the front- and back-end

Webpage Source Code — HTML — Overview

- HTML stands for Hypertext Markup Language
 - it gives the structure to what is visible to visitors
 - text, images, links
 - would a piece of text appear in a paragraph or a list?
 - depends on the HTML code around that text

```
<!DOCTYPE html>
<html>
 <head>
   <style>
     h1 {color: blue;}
   </style>
 </head>
 <body>
   <h1>A header</h1>
   This is a paragraph.
   <l
     This
     is a
     list
   </body>
</html>
```

HTML consists of elements

```
This is a one sentence paragraph.
```

This is a one sentence paragraph.

- there is only one element on this page
 - a paragraph

Most elements have opening and closing tags

```
This is a one sentence paragraph.
```

This is a one sentence paragraph.

- tag name, in this case p, defines the structure of the element
- the closing tag has a forward slash / before the element name

Most elements have some content

```
This is a one sentence paragraph.
```

This is a one sentence paragraph.

Elements can be nested

```
This is a <strong>one</strong> sentence paragraph.
```

This is a **one** sentence paragraph.

- there are two elements above, a paragraph and a strong emphasis
- strong is said to be the child of the paragraph element
 - there could be more than one child
 - in that case, children are numbered from the left
- paragraph is said to be the parent of the strong element

Elements can have attributes

```
This is a <strong id="sentence-count">one</strong> sentence paragraph.
```

This is a **one** sentence paragraph.

- the id attribute is not visible to the visitors
- attribute string sentence-count could have been anything I could come up with
 - unlike the tag and attribute names e.g., strong, id as they are pre-defined
- there are some other attributes that are visible

There could be more than one attribute in an element

```
This is a <strong class="count" id="sentence-count">one</strong> sentence paragraph.
There are now <strong class="count" id="paragraph-count">two</strong> paragraphs.
```

This is a **one** sentence paragraph.

There are now **two** paragraphs.

- there could be more than one attribute in an element
 - with a white space in between them
- the class attribute can apply to multiple elements
 - while the id attibute must be unique on a page

Webpage Source Code — HTML — Important Elements & Attibutes

Links

```
Click <a href="https://www.google.com/">here</a> to google things.
```

Click here to google things.

- href (hypertext reference) is a required attribute for the the a (anchor) tag
- most attributes are optional, some are required

Webpage Source Code — HTML — Links

Links

```
Click <a title="This text appears when visitors hover over the link"
href="https://www.google.com/">here</a> to google things.
```

Click here to google things.

Note that

• the a (anchor) tag is used with href (hypertext reference)

Webpage Source Code — HTML — Lists

The <lu> tag introduces unordered lists, while the tag defines lists items

```
  books
  journal articles
  reports
```

- books
- journal articles
- reports

Note that

Ordered lists are introduced with the the <lo> tag instead

Webpage Source Code — HTML — Notes

By default, multiple spaces and/or lines breaks are not meaningful

- books
- journal articles
- reports

- plain source code may or may not be written in a readable manner
- this is one reason why DOM is helpful

Webpage Source Code — CSS — Overview

- CSS stands for Cascading Stylesheets
 - it gives the style to what is visible to visitors
 - text, images, links
 - would a piece of text appear in black or blue?
 - depends on the CSS for that text
- CSS can be defined
 - inline, as an attribute of an element
 - internally, as a child element of the head element
 - externally, but then linked in the head element

CSS is written in rules

```
p {font-size:12px;}
.count {background-color:yellow;}
#sentence-count {color:red;}
```

CSS is written in rules, with a syntax consisting of

• one or more selectors

p {font-size:14px;} h1 h2 {color:blue;} .count {background-color:yellow;} #sentence-count {color:red; font-size:14px;}

- selector type defines the syntax
 - elements are plain
 - e.g., p, h1, h2
 - classes start with a full stop
 - ids start with a number sign

CSS is written in rules, with a syntax consisting of

- one or more selectors
- a decleration

```
p {font-size:14px;}
h1 h2 {color:blue;}
.count {background-color:yellow;}
#sentence-count {color:red; font-size:14px;}
```

Note that

• declerations are written in between two curly brackets

CSS is written in rules, with a syntax consisting of

- one or more selectors
- a decleration, with one or more properties

Note that

• properties are followed by a colon

```
p { font-size: 14px;}
h1 h2 { color: blue;}
.count { background-color: yellow;}
#sentence-count { color: red; font-size: 14px;}
```

CSS is written in rules, with a syntax consisting of

- one or more selectors
- a decleration, with one or more properties and values

```
p {font-size: 14px;}
h1 h2 {color: blue;}
.count {background-color: yellow;}
#sentence-count {color: red; font-size: 14px;}
```

- values are followed by a semicolon
- property:value; pairs are seperated by a white space

Webpage Source Code — CSS — Internal

- CSS rules can be defined internally
 - within the style element
 - as a child of the head element
- Internally defined rules apply to all matching selectors
 - on the same page

```
<!DOCTYPE html>
<html>
 <head>
   <style>
    h1 {color: blue;}
   </style>
 </head>
 <body>
   <h1>A header</h1>
   This is a paragraph.
   <l
     This
     is a
     list
   </body>
</html>
```

Webpage Source Code — CSS — External

- CSS rules can be defined externally
 - saved somewhere linkable
 - defined with the the linked element
 - as a child of the head element
- Externally defined rules
 - are saved in a file with .css extension
 - apply to all matching selectors
 - on any page linked

```
<!DOCTYPF html>
<html>
 <head>
   <link rel="styles" href="simple.css">
 </head>
 <body>
   <h1>A header</h1>
   This is a paragraph.
   <l
     This
     is a
     list
   </body>
</html>
```

Webpage Source Code — CSS — Inline

CSS rules can also be defined inline

- with the style attribute
- does not require selector
- applies only to that element

```
This is a <strong style="color:blue;">one</strong> sentence paragraph.
```

This is a **one** sentence paragraph.

Part 3. Data Collection

Back to the contents slide.

Part 4. Data Preperation

Data Preperation — Overview

- The rtweet package does a very good job with data preparation to start with
 - returns data frames, with mostly tidy data
 - although Twitter APIs return nested lists
 - some variables are still lists
 - e.g., hastags
- Further data preparation depends on your research project
 - most importantly, on whether you will work with texts or not
 - we will cover some common preparation steps

Data Preperation — Overview — Strings

- Most researchers would be interested in textual Twitter data
 - tweets as a whole, but also specifically hashtags *etc*.
- There are many components of tweets as texts
 - e.g., mentions, hashtags, emojis, links etc.
 - but also punctuation, white spaces, upper case letters *etc*.
 - some of these may need to be taken out before analysis
- I use the stringr package (Wickham, 2019) for string operations
 - part of the tidyverse family
 - you might have another favourite already
 - no need to change as long as it does the job

Data Preperation — Overview — Numbers

- There is more to Twitter data than just tweets
 - e.g., the number of followers, likes *etc*.
 - see Silva and Proksch (2020) for a great example
- I use the dplyr package (Wickham et al, 2020 for most data operations
 - part of the tidyverse family
 - you might have another favourite already
 - no need to change as long as it does the job

Data Preperation — Remove Mentions

Data Preperation — Remove Hashtags

Data Preperation — Remove Links

```
tweet <- "These from @handle1 are #socool. A #mustsee, @handle2!

$\tilde{T} \tilde{t.co/aq7MJJ1}

$\tilde{T} \tilde{https:}/\tilde{t.co/aq7MJJ2"}

str_remove_all(string = tweet, pattern = "http\\S+\\s*")

[1] "These from @handle1 are. A, @handle2! \( \frac{T}{T} \tilde{t.co/aq7MJJ1"} \)
```

- Notice that
 - links come in various formats
 - you may need multiple or complicated regular expression patterns

Data Preperation — Remove Links — Alternative

08:00

Use the urls_t.co variable to remove all links

• if there are more than one link in a tweet, they are stored as a list in this variable

Data Preperation — Remove Emojis

[1] "These from @handle1 are #socool. A #mustsee, @handle2! t.co/aq7MJJ1 https://t.co/aq7MJJ2"

Data Preperation — Exercises — Notes

- The exercises in this part are best followed by
 - using tweets.rds or similar dataset
 - saving a new variable at every step of preparation
 - observing the newly created variables
 - to confirm whether the code works as intended

- The mutate function, from the dplyr package, can be helpful, as follows
 - recall that text is the variable for tweets

```
tweets <- read_rds("data/tweets.rds")

clean_tweets <- tweets %>%
    mutate(no_mentions = str_remove_all(string = text, pattern = "[@][\\w_-]+"))
```

Exercises

10:00

- 41) Remove mentions
 - hint: the pattern is "[@][\\w_-]+"
- 42) Remove hastags
 - hint: the pattern is "[#][\\w_-]+"
- 43) Remove links
 - by using the links from the urls_t.co variable
- 44) Remove emojis
 - pull the help file for the iconv function first

Data Preperation — Remove Punctuations

```
tweet <- "These from @handle1 are #socool. ♠ A #mustsee, @handle2!

☐ t.co/aq7MJJ1
☐ https://t.co/aq7MJJ2"

str_remove_all(string = tweet, pattern = "[[:punct:]]")
```

[1] "This from are socool 🕙 A mustsee handle2 👍 tcoaq7MJJ1 👍 httpst.coaq7MJJ2"

Notice that

- this removed all punctuation, including those in mentions, hashtags, and links
- if tweets are typed with no spaces after punctuation, this might lead to merged pieces of text
 - alternatively, try the str_replace function to replace punctuation with space

Data Preperation — Remove Punctuations — Alternative

```
tweet <- "This is a sentence.There is no space before this sentence."

str_remove_all(string = tweet, pattern = "[[:punct:]]")

[1] "This is a sentenceThere is no space before this sentence"</pre>
```

```
str_replace_all(string = tweet, pattern = "[[:punct:]]", replacement = " ")
```

[1] "This is a sentence There is no space before this sentence "

Data Preperation — Remove Punctuations — Alternative

```
tweet <- "This is a sentence.There is no space before this sentence."

str_replace_all(string = tweet, pattern = "[[:punct:]]", replacement = " ")</pre>
```

[1] "This is a sentence There is no space before this sentence "

Data Preperation — Remove Repeated Whitespace

```
tweet <- "There are too many spaces after this sentence. This is a new sentence."

str_squish(string = tweet)</pre>
```

[1] "There are too many spaces after this sentence. This is a new sentence."

Note that

- white spaces can be introduced not only by users on Twitter, but also by us, while cleaning the data
 - e.g., removing and/or replacing operations above
 - hence, this function might be useful after other operations

Data Preperation — Change Case

```
tweet <- "lower case. Sentence case. Title Case. UPPER CASE."

str_to_lower(string = tweet)

[1] "lower case. sentence case. title case. upper case."</pre>
```

Note that

- there are other functions in this family, including
 - str_to_sentence, str_to_title, str_to_upper

Exercises

10:00

- 45) Remove punctuations
 - by using the str_replace_all function
 - hint: the pattern is [[:punct:]]
- 46) Remove whitespace
 - hint: the function is called str_squish
- 47) Change case to lower case
 - hint: the function is called str_to_lower

Data Preperation — Change Unit of Observation

Research designs might require changing the unit of observation

- aggregation
 - e.g., at the level of users, locations, hashtags etc.
 - summarise with dplyr
- dis-aggregation
 - e.g., to the level of words
 - tokenise with tidytext

Aggregate at the level of users

• the number of tweets per user

```
# load the tweets dataset
df <- read_rds("tweets.rds") %>%

# group by users for aggregation
group_by(user_id) %>%

# create summary statistics for variables of interest
summarise(sum_tweets = n())
```

What is aggregated at which level depends on your research design, such as

- aggregate the tweets into a single text
- at the level of users by source

```
# load the tweets dataset
df <- read_rds("tweets.rds") %>%

# group by users for aggregation
group_by(user_id, source) %>%

# create summary statistics for variables of interest
summarise(merged_tweets = paste0(text, collapse = ". "))
```

Disaggregate the tweets, by splitting them into smaller units

• also called tokenisation

Note that

- by default sep = "[^[:alnum:].]+", which works well with separating tweets into words
 change this argument with a regular expression of your choice
- this creates a tidy dataset, where each observation is a word
 - all other tweet-level variables are repeated for each observation

```
# load the tweets dataset
df <- read_rds("tweets.rds") %>%

# split the variable text
    separate_rows(text)
```

The tidytext has a function that works better with tokenising tweets

with token = "tweets", it dis-aggregates text into words
 except that it respects usernames, hashtags, and URLS

```
# load the tweets dataset
df <- read_rds("tweets.rds") %>%

# split the variable text, create a new variable called da_tweets
    unnest_tokens(output = da_tweets, input = text, token = "tweets")
```

Tokenise variables to levels other than words

• e.g., characters, words (the default), sentences, lines

```
# load the tweets dataset
df <- read_rds("tweets.rds") %>%

# split the variable text into sentences, create a new variable called da_tweets
    unnest_tokens(output = da_tweets, input = text, token = "sentences")
```

Tokenise variables other than tweets

• recall that rtweet stores multiple hastags, mentions etc. as lists

```
# load the tweets dataset
df <- read_rds("tweets.rds") %>%

# unlist the lists of hashtags to create strings
group_by(status_id) %>%
mutate(tidy_hashtags = str_c(unlist(hashtags), collapse = " ")) %>%

# split the string, create a new variable called da_tweets
unnest_tokens(output = da_hashtags, input = tidy_hashtags, token = "words")
```

Data Preperation — Remove Stop Words

Remove the common, uninformative words

• e.g., the, a, i

Note that

- this operation requires a tokenised-to-word variable
- stop words for English are stored in the stop_words dataset in the tidytext variable
- list of words for other languages are available elsewhere, including
 - the stopwordslangs function from the rtweet package
 - the stopwords function from the tm package
 - e.g., use tm::stopwords("german") for German

```
# load the tweets dataset
df <- read_rds("tweets.rds") %>%

# split the variable text, create a new variable called da_tweets
  unnest_tokens(output = da_tweets, input = text, token = "tweets") %>%

# remove rows that match any of the stop words as stored in the stop_words dataset
  anti_join(stop_words, by = c("da_tweets" = "word"))
```

Exercises

10:00

- 48) Aggregate text to a higher level
 - e.g., if you are not using tweets.rds, to MP level
 if not, perhaps to source level
- 49) Dis-aggregate text to a lower level
 - e.g., to words
- 50) Dis-aggregate hashtags
 - i.e., make sure each row has at most one hashtag
- 51) Remove stop words

References

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References

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The workshop ends here.

Congradulations for making it this far, and thank you for joining me!