

Workshop 'Computational Text Analysis' Session 1: Web-Scraping

Mirko Wegemann

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Mirko Wegemann

Workshop 'Computational Text Analysis'



About us I

Mirko Wegemann (he/him)

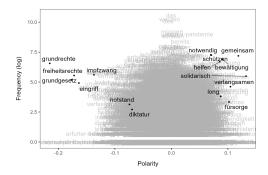
- Postdoctoral Researcher at the University of Münster, Germany
- Research Agenda
 - Political Parties
 - Gender & Politics
 - Political Communication



About us II

Application of CTA in my own work

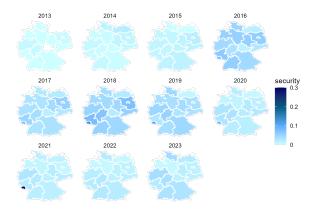
• scaling the Covid-19 discourse using Latent Semantic Scaling (joint work with Rebecca Kittel)





About us III

 classification of gender appeals by the radical right using a multilingual transformer model (joint work with Leon Küstermann)

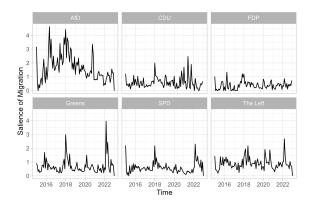


Workshop 'Computational Text Analysis'



About us ${\sf IV}$

• tracing the migration discourse in Germany using large language models





About us V

Eva Krejcova (she/her)

- Junior Lecturer at the Swiss Graduate School of Public Administration (IDHEAP) at the University of Lausanne, Switzerland
- introduces you to images-as-data



About us VI

Sara Dybesland (she/her)

- 2nd-year SPS PhD Researcher
- regularly uses web-scraping and topic models in R
- analyses parliamentary speeches
- TA for the lab sessions



• your name (if you want pronoun)

- research topic
- what's text analysis doing for your studies?
- do you have any experiences with text analysis?
- do you have any experiences with images-as-data?



Goals of this workshop

- 1. Automation of data collection
- 2. Analysis of textual data
 - unsupervised approaches (e.g., topic models)
 - supervised approaches (e.g. text classification)
- 3. Analysis of images-as-data

 \rightarrow understanding and dealing with key challenges, implementing first own analyses, getting a 'glimpse' into what's possible



Basic structure

- Session 1 & 2 on web-scraping and basic text-analysis techniques
 - 1. Seminar session: Introduction into the topic with first parallel application
 - 2. Lab session: Application on your own data
- Session 3 & 4 on transformer models and images-as-data are showcase sections
- \rightarrow to get the most out of it: bring your own data to the lab sessions!



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What is web-scraping?



The process of systematically collecting data from the Internet to store it in a structured data format.



How can we utilize it in research? I

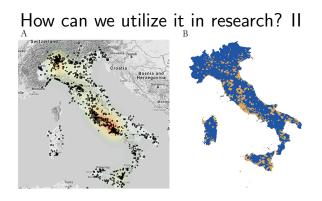


Figure 1.1 Location of UNESCO World Heritage Sites in danger (as of March 2014). Cultural sites are marked with triangles, natural sites with dots

World Heritage in Danger? (Munzert 2015, p. 5)

From a table on a Wikipedia page to a nice visualization





Bischof and Kurer (2023) show how local campaigning of M5S mobilizes voters



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How can we utilize it in research? III

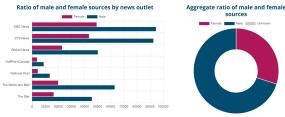


Fig 1. The Gender Gap Tracker online dashboard page. Reprinted from https://gendergaptracker.informedopinions.org/ under a CC BY license, with permission from Informed Opinions, original copyright 2018.

Asr et al. (2021) on how Canadian newspaper quote men more often than women (still online!)



How can we utilize it in research? IV

Web scraping is a first step in our text-as-data query: we access the data. However, neither is scraping the only way to access data nor is text analysis the only type of analysis we can apply.



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Types of web-scraping

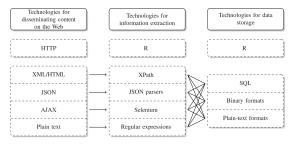
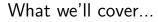


Figure 1.4 Technologies for disseminating, extracting, and storing web data

Munzert (2015, p. 10)



- 1. 'static' HTML structures (rvest)
 - all content is available by parsing the HTML
- 2. 'dynamic' webpages (RSelenium)
 - content is not visible in HTML source code but becomes available after interacting with the webpage



Before you scrape data

- 1. carefully consider whether this is the data you need for your research question
 - Do the data allow you to measure what you need to?
 - Do they comply with academic quality standards?
 - Is it really worth the effort?
- 2. check whether the data is directly available (e.g., someone else downloaded it, there is a direct download link, there is API access)
- 3. consider legal constraints



Disclaimer: Legal considerations I

Is web-scraping even allowed?

It depends...

• violation against terms of service?



Disclaimer: Legal considerations II

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Terms of Service of 'The Guardian'

Disclaimer: Legal considerations III

Is web-scraping even allowed?

It depends...

- violation against terms of service?
- gathering of personally-identifiable information may violate EU General Data Protection Regulation (GDPR)
- even if you do not collect personal data, you may infringe copyright regulations
- check robots.txt



Disclaimer: Legal considerations IV



- hiQ scraped data from LinkedIn public profiles
- LinkedIn tried to prevent that
- initial ruling: scraping allowed; later rulings: hiQ breached User Agreement \rightarrow settlement

In the end, "[l]egalities depend a lot on where you live. However, as a general principle, if the data is public, non-personal, and factual, you're likely to be ok" (Wickham et al. 2023)



HTML Basics

Websites are based on HyperText Markup Language

- HTML contains information about the structure of a web-page
- $\bullet\,$ it's responsible for how content is graphically displayed

An example



HTML Elements and Attributes

- HTML consists of elements, tags and attributes
 - elements are the different components of a webpage (e.g. headlines, text, images)
 - elements are mostly embedded into tags
 (<element>content</element>) but some come
 without start and end tags
 - attributes are additional information of an element (e.g., the size of an image, the font type etc.)

We will only cover HTML in a superficial way, but try this tutorial



1

2

5

7

8

9

Static webpages

HTML head vs body

```
< DOCTYPE html>
   <html>
3
   <head>
4
     <title>"EUI Website"</title>
   </head>
6
   <body>
     <h1>"New Academic Year"</h1>
   </body>
   </html>
```

HTML documents consist of a head (with meta information on the webpage) and a body (with content) \rightarrow mainly interested in the body!



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Important elements

- h1, h2, h3, etc.: headline elements
- *p*: paragraph elements
- a: hyperlink elements
- *img*: image elements



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Important attributes

- href: weblink, always comes with a element
- *src*: source of an image



How to use HTML for scraping?

We need to identify the CSS selector of an element of interest. There are \mathbf{two} options:

- manual approach: cursor on element of interest > right-click
 > Inspect
- semi-automatic approach: download **SelectorGadget** or save as bookmark



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Basic pipeline

Setup

- install SelectorGadget
- R library: rvest



Step 1

Download HTML source with rvest

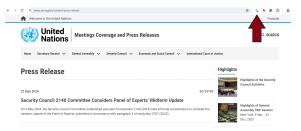
```
1
  > library(rvest)
2
  > url <- "https://press.un.org/en/content/press-release"</pre>
3
  > html <- read_html(url)</pre>
4
  > html
5
  {html_document}
6
   <html lang="en" dir="ltr">
7
   [1] <head>\n<meta http-equiv="Content-Type"
       content="text/html; charset=UTF-8">\n<meta
       charset="utf-8">\n<link rel="canonical" href="https:
       . . .
8
   [2] <body class="layout-one-sidebar
       layout-sidebar-first page-view-home-press
       path-content">\n <div
       class="visually-hidden-focusable bg- ..
```



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Step 2

Example: UN press releases





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Step 3

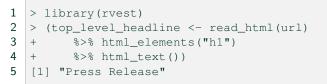


Try it out yourself!



Headline

Here, we retrieve every level-1 headline of the webpage.





Text

Here, we retrieve every paragraph on the webpage.

```
1
  > library(rvest)
2
  > (paragraphs <- read_html(url) %>%
3
         html_elements("p") %>%
  +
4
  +
        html text())
5
   [1] "On 3 May 2024, the Security Council Committee
      established pursuant to resolution 2140 (2014) held
      informal consultations to consider the midterm
      update of the Panel of Experts, submitted in
      accordance with paragraph 3 of resolution 2707
      (2023)."
6
   [2] "United Nations Secretary-General Antonio Guterres
      flew from Muscat, Oman, to Manama, Bahrain, in the
      early evening of Wednesday, 15 May."
```



Links

If we want to access links, we need to retrieve the a element first, and then call its attribute ${\tt href}$



Tables

$\tt rvest$ has a pre-defined function $\tt read_table()$ to extract information from HTML tables

```
1 > html <- read_html(url2)
2 > table <- html %>%
3 + html_element(".wikitable:nth-child(4)") %>%
4 + html_table()
```



Images

For images, it is a bit more complicated.

- 1. open a session
- 2. retrieve image source link
- 3. download the image to your directory



Images II

```
1
   > session <- session(url)</pre>
2
   >
3
   > # access links for image sources
4
   > imgsrc <- session %>%
5
          read_html() %>%
   +
6
          html_nodes("img") %>%
   +
7
          html attr("src")
   +
8
   >
9
   > # access page of source image (only retrieve the
       first image here)
10
   > img <- session_jump_to(session, paste0(root_url,</pre>
       imgsrc[[1]]))
11
   >
12
   > # write to our project's directory
13
   > writeBin(img$response$content, basename(imgsrc[1]))
```

Let's do it in R



Loops I

Often, we want to automatize these steps for multiple pages. Two options:

- 1. creating empty objects and filling them in a for-loop
- 2. defining a function, apply it and retrieve objects of interest from list

 \rightarrow usually, functions are more versatile and can be easier run in parallel

Loops II

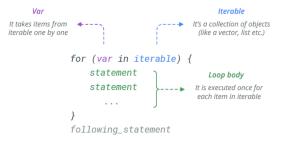
Before creating a loop/function

- check the pagination structure of a website (e.g. United Nations use '?page=#' to list results)
- 2. make sure which elements you need to retrieve (often, you just want links but some information, such as dates, may not be available on sub-pages, so you want to gather these as well)
- 3. test the pipeline on a single element before using it in a loop



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For-loops



Tutorial on for-loops Graph and guide to for-loops



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For-loops for automation

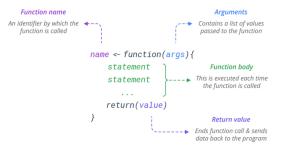
```
1
2
   > urls <- c("https://www.eui.eu/en/home",</pre>
        "https://www.ceu.edu/")
3
   > links <- c()
4
   > for(url in 1:length(urls)) {
5
            html <- read_html(urls[[url]])</pre>
      +
6
      +
            links[url] <- html %>%
 7
                 html_node("h1") %>%
      +
8
      +
                html text()
9
      + }
10
   > links
11
    [1] "\nOpening doors to the future\n" "Apply Now for
        2024-2025!"
```

Let's do it in R



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Functions in R



Tutorial on Functions Graph and guide to functions



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Functions for automation

```
1
   > h1 scrape <- function(url) {
2
     +
           html <- read_html(urls[[url]])</pre>
3
           links[url] <- html %>%
     +
4
                html_node("h1") %>%
     +
5
               html text()
     +
6
     + }
7
   >
8
   > (links <- sapply(1:length(urls), h1_scrape))</pre>
9
   [1] "\nOpening doors to the future\n" "Apply Now for
       2024-2025!"
```

Let's do it in R



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Post processing

Regular expressions

Often, the data you retrieve are not in the desired format. Regular expressions are handy for data processing, they can match/extract/remove a pattern in text

```
1 > library(stringr)
2 > date_time <- "2009/03/12 20:12:31"
3 > str_extract(date_time, "\\d{4}/\\d{2}/\\d{2}")
4 [1] "2009/03/12"
```



Regular expressions in R

- base R includes regular expressions in functions like gsub (replacement), regmatches (extraction) or regexpr (detection)
- stringr is the tidyverse approach to regex and provides equivalent functions such as str_replace, str_extract or str_detect



More information on regular expressions

- You can try out regular expressions in online tools like these
- StringR cheat sheet
- A guide to regular expressions



Dynamic Webpages

Sometimes, webpages change only when we interact with them in a browser session (e.g. by clicking on specific objects). For these webpages, rvest is not applicable.

Usually, you can identify these pages by their usage of active classes



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Functionality of RSelenium

Solution?



RSelenium

RSelenium was initially built to test webpages, we use it to send commands to a virtual browser window (e.g. clicking a button)



Before using RSelenium

Whenever possible, try to use the static approach, it's easier and less prone to error

- check other webpages which may contain the same information
- look for other sub-directories that may store the same content
- check the Wayback Machine
- use the 'Search' bar of a webpage to list results

Pipeline of RSelenium and wdman

A typical pipeline looks as follows:

- configuration of a session (now more complex, we need wdman for this)
- opening browser
- navigate to URL
- accept/decline cookies
- identify css selector for button, click on it (repeat if necessary)
- retrieve object of interest (usually links)



Setup

- configure driver
- extract client
- navigate to webpage

```
> url <- "https://www.europarl.europa.eu/news/en"</pre>
1
2
   > rd <- rsDriver(browser = "firefox",</pre>
3
                      chromever = NULL,
   +
4
                      port = sample(1:65535, 1),
   +
5
                      verbose = FALSE)
   +
6
   > browser <- rd[["client"]]</pre>
7
   > browser$navigate(url)
```



Identifying and clicking on a button

- 1. Find an element by its css selector
- 2. store it as an object
- 3. use the clickElement() function



How to extract the text of a dynamic webpage

- identify elements by css selector
- elements contain lots of other information
- to extract plain text, loop through the elements and extract only the text

```
1
  > paragraphs <- browser$findElements(using = 'css
      selector', value='p')
2
  > text <- c()
3
  > for(i in 1:length(paragraphs)) {
4
          text[i] <- paragraphs[[i]]$getElementText()[[1]]</pre>
     +
5
     + }
6
  > text[1]
7
   [1] "On Thursday afternoon, the Eurovision debate
      between the lead candidates for the presidency of
      the European Commission took place in the European
      Parliament."
```



How to extract the link of a dynamic webpage

```
1 > urls_euparl <- browser$findElements(using = 'css
selector', value='.ep_title > a')
2 > urls_euparl2 <- c()
3 > for(i in 1:length(urls_euparl)){
4 + urls_euparl2[i] <-
urls_euparl[[i]]$getElementAttribute('href')[[1]]
5 }
```

Let's do it in R



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Processing data

Once we have downloaded individual links to pages, we can often proceed in rvest (which is easier to handle).



Concluding remarks

That's it for today. We covered the first aspect of text analysis on how to retrieve data.

- in the lab session: practice scraping on your own projects
- tomorrow: bags-of-words approaches
 - topic modelling
 - prediction tasks

Thank you for your attention!

References I

Asr, F. T., Mazraeh, M., Lopes, A., Gautam, V., Gonzales, J., Rao, P., & Taboada, M. (2021). The Gender Gap Tracker: Using Natural Language Processing to measure gender bias in media. PLOS ONE, 16(1), e0245533. https://doi.org/10.1371/journal.pone.0245533 Bischof, D., & Kurer, T. (2023). Place-Based Campaigning: The Political Impact of Real Grassroots Mobilization. The Journal of Politics, 85(3), 984–1002. https://doi.org/10.1086/723985 Munzert, S. (2015). Automated Data Collection with R: A Practical Guide to Web Scraping and Text Mining (1st ed.). Wiley. Wickham, H., Çetinkaya-Rundel, M., & Grolemund, G. (2023). R for Data Science. O'Reilly Media, Inc. https://r4ds.hadley.nz/