Introduction

Today we will be extracting some useful data from websites. There's a bunch of different ways to web-scrape, but we'll be exploring using the rvest package in R, that helps you to deal with parsing html.

Why is web scraping useful? If our research involves getting data from a website that isn't already in a easily downloadable form, it improves the reproducibility of our research. Once you get a scraper working, it's less prone to human error than copy-pasting, for example, and much easier for someone else to see what you did.

A note on responsibility

Seven principles for web-scraping responsibly:

- 1. Try to use an API.
- 2. Check robots.txt. (e.g. https://www.utoronto.ca/robots.txt)
- 3. Slow down (why not only visit the website once a minute if you can just run your data collection in the background while you're doing other things?).
- 4. Consider the timing (if it's a retailer then why not set your script to run overnight?).
- 5. Only scrape once (save the data as you go and monitor where you are up to).
- 6. Don't republish the data you scraped (cf datasets that create based off it).
- 7. Take ownership (add contact details to your scripts, don't hide behind VPNs, etc)

Extracting data on opioid prescriptions from CDC

We're going to grab some data on opioid prescription rates from the CDC website. While the data are nicely presented and mapped, there's no nice way of downloading the data for each year as a csv or similar form. So let's use rvest to extract the data. We'll also load in janitor to clean up column names etc later on.

```
library(tidyverse)
```

```
----- tidyverse 1.3.2 --
## -- Attaching packages -----
## v ggplot2 3.4.0
                      v purrr
                               1.0.1
## v tibble 3.1.8
                      v dplyr
                               1.0.10
           1.2.1
## v tidyr
                      v stringr 1.5.0
## v readr
           2.1.3
                      v forcats 0.5.2
## -- Conflicts -----
                                      ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
library(rvest)
```

```
##
## Attaching package: 'rvest'
##
## The following object is masked from 'package:readr':
##
## guess_encoding
```

library(janitor)

```
##
## Attaching package: 'janitor'
##
## The following objects are masked from 'package:stats':
##
## chisq.test, fisher.test
```

Getting the data for 2008

Have a look at the website at the url below. It shows a map of state prescription rates in 2008. Let's read in the html of this page.

```
cdcpage <- "https://www.cdc.gov/drugoverdose/rxrate-maps/state2008.html"
cdc <- read_html(cdcpage)
cdc

## {html_document}
## <html lang="en-us" class="cdc-2022 theme-purple cdc-page-type-content">
## [1] <head>\n<meta http-equiv="Content-Type" content="text/html; charset=UTF-8 ...
## [2] <body class="no-js cdc-page">\r\n\t<div id="skipmenu">\r\n\t\t<a class="s ...</pre>
```

Note that it has two main parts, a head and body. For the majority of use cases, you will probably be interested in the body. You can select a node using html_node() and then see its child nodes using html_children().

```
body_nodes <- cdc |>
html_node("body") |>
html_children()
body_nodes
```

```
## {xml nodeset (20)}
  [1] <div id="skipmenu">\r\n\t\t<a class="skippy sr-only-focusable" href="#co ...
##
   [2] <div class="header-language-bar container text-right pt-1 pb-1 fs0875">\ ...
  [3] <header id="page_banner" role="banner" aria-label="Banner"><div class="c ...
##
  [4] <div class="container d-flex flex-wrap body-wrapper bg-white">\r\n\t\t\t ...
   [5] <footer class="" role="contentinfo" aria-label="Footer"><div class="cont ...
##
##
   [6] <script src="https://www.cdc.gov/config/cdc_config.js"></script>
  [7] <script>var CDC_POST={"id":"299_24637","type":"cdc_page","tax":[]};</scr ...
## [8] <script src="/TemplatePackage/contrib/libs/jquery/latest/jquery.min.js?_ ...
  [9] <script src="/TemplatePackage/contrib/libs/bootstrap/latest/js/bootstrap ...
## [10] <script src="/TemplatePackage/contrib/libs/cdc/ab/4.0.0/ab.js"></script>
## [11] <script src="/TemplatePackage/4.0/assets/js/app.min.js?_=97791"></script>
## [12] <script src="/TemplatePackage/4.0/assets/js/tp-cookie-policy.js"></script>
## [13] <svg viewbox="0 0 40 40" class="d-none"><radialgradient id="svg_ig_1" cx ...
## [14] <svg id="multicolor_icons" style="display:none" xmlns="http://www.w3.org ...
## [15] <script>\r\n\t\r\n\ts.pageName=document.title; \r\ns.channel="Drug Overd ...
## [16] <script>\r\n\twindow.shortTitle = "U.S. State Opioid Dispensing Rates, 2 ...
## [17] <script>\r\n\t<B>Error processing SSI file</B><BR>\r\n</script>
## [18] <script>\r\n\twindow.CDC.tp4.public.appInit( window.pageOptions );\r\n</ ...
## [19] <div class="modal fade noindex" id="cdcExtLink" tabindex="-1" role="dial ...
## [20] <div role="dialog" aria-labelledby="privacy-policy-label" aria-modal="tr ...
```

Inspecting elements of a website

The above is still fairly impenetrable. But we can get hints from the website itself. Using Chrome (or Firefox) you can highlight a part of the website of interest (say, 'Alabama'), right click and choose 'Inspect'. That gives you info on the underlying html of the webpage on the right hand side. Alternatively, and probably easier to find what we want, right click on the webpage and choose View Page Source. This opens a new window with all the html. Do a search for the world 'Alabama'. Now we can see the code for the table. We can see that the data we want are all within tr. So let's extract those nodes:

```
cdc |>
     html_nodes("tr")
## {xml nodeset (52)}
##
            [1] \nState\nState Abbreviation\nOpioid Dispensing ...
            [2] <tr>\nAlabama\n\n\n\n\n\n\n
##
            [3] <tr>\nAlaska\nAK\n68.5\n\n
            [4] \frac{d}{dx} \cdot \frac{d}{n \cdot dx} \cdot \frac{d}
##
            [5] <tr>\nArkansas\nAR\112.1\n\n
##
            [6] <tr>\nCalifornia\nCA\n55.1\n\n
##
            [7] <tr>\nColorado\nCO\n67.7\n\n
            [8] <tr>\nConnecticut\n\n\n\n\n\n
            [9] \nDelaware\nDE\n95.4\n\n
##
## [10] <tr>\nDistrict of Columbia\nDC\n34.5\n\n
## [11] \nFlorida\nFL\n84.3\n\n
## [12] \nGeorgia\nGA\n86.3\n\n
## [13] \nHawaii\nHI\n46.6\n\n
## [14] <tr>\nIdaho\nID\n82.7\n\n
        [15] <tr>\nIllinois\n\l\n60.2\n\n
        [16] \nIndiana\n\n\n\n\n3.3\n\n
        [17] <tr>\nIowa\nIA\n59.1\n\n
## [18] <tr>\nKansas\nKS\n82.7\n\n
## [19] \nKentucky\n\n136.6\n\n
## [20] \nLA\n13.7\n\n
## ...
```

Great, now we're getting somewhere. We only want the text, not the html rubbish, so let's extract that:

```
table_text <- cdc |>
  html_nodes("tr") |>
  html_text()
table_text
```

```
##
    [1] "State\nState Abbreviation\nOpioid Dispensing Rate per 100\n"
##
    [2] "Alabama\nAL\n126.1\n"
##
    [3] "Alaska\nAK\\n68.5\\n"
    [4] "Arizona\nAZ\n80.9\n"
##
##
       "Arkansas\nAR\n112.1\n"
##
        "California\nCA\n55.1\n"
##
    [7]
        "Colorado\nCO\n67.7\n"
        "Connecticut\nCT\n68.7\n"
##
    [8]
##
    [9]
        "Delaware\nDE\n95.4\n"
  [10] "District of Columbia\nDC\n34.5\n"
  [11] "Florida\nFL\n84.3\n"
```

```
## [12] "Georgia\nGA\n86.3\n"
       "Hawaii\nHI\n46.6\n"
  [13]
  [14] "Idaho\nID\n82.7\n"
  [15] "Illinois\nIL\n60.2\n"
  [16]
       "Indiana\nIN\n103.3\n"
  [17]
       Iowa\nIA\n59.1\n''
##
## [18]
       "Kansas\nKS\n82.7\n"
## [19]
        "Kentucky\nKY\n136.6\n"
##
  [20]
       "Louisiana\nLA\n113.7\n"
##
  [21]
       "Maine\nME\n88.7\n"
  [22] Maryland\nMD\n65.5\n
  [23] "Massachusetts\nMA\n69.2\n"
  [24]
        "Michigan\nMI\n89.9\n"
##
        "Minnesota\nMN\n56.5\n"
  [25]
  [26]
        \verb"Mississippi\nMS\n113.2\n"
   [27]
        "Missouri\nMO\n86.8\n"
   [28]
##
        "Montana\nMT\n85.3\n"
   [29] "Nebraska\nNE\n66.2\n"
  [30] "Nevada\nNV\n97.0\n"
   [31]
        "New Hampshire\nNH\n81.7\n"
##
  [32]
        "New Jersey\nNJ\n59.5\n"
##
  [33]
        "New Mexico\nNM\n71.4\n"
        "New York\nY\n48.4\n"
  [34]
##
##
  [35]
        "North Carolina\nNC\n88.6\n"
##
  [36]
       "North Dakota\nND\n61.7\n"
  [37] "Ohio\nOH\n97.5\n"
   [38] "Oklahoma\nOK\n111.3\n"
##
##
   [39]
       "Oregon\nOR\n99.1\n"
  [40]
        "Pennsylvania\nPA\n76.5\n"
  [41]
        "Rhode Island\nRI\n82.9\n"
  [42]
        "South Carolina\nSC\n94.1\n"
##
   [43]
       "South Dakota\nSD\n52.1\n"
   [44]
       "Tennessee\nTN\n132.9\n"
  [45] "Texas\nTX\n71.3\n"
   [46]
        "Utah\nUT\n91.3\n"
        "Vermont\nVT\n56.5\n"
##
  [47]
  [48]
       "Virginia\nVA\n73.0\n"
  [49]
        "Washington\nWA\n86.6\n"
        "West Virginia\nWV\n145.5\n"
##
  [50]
  [51]
        "Wisconsin\nWI\n70.6\n"
##
## [52] "Wyoming\nWY\n81.0\n"
```

This is almost useful! Turning it into a tibble and using **separate** to get the variables into separate columns gets us almost there:

```
rough_table <- table_text |>
  as_tibble() |>
  separate(value, into = c("state", "abbrev", "rate"), sep = "\n", extra = "drop")
rough_table
```

```
##
   1 State
                           State Abbreviation Opioid Dispensing Rate per 100
##
                           AL
                                               126.1
  2 Alabama
                                               68.5
##
  3 Alaska
                           AK
                                              80.9
## 4 Arizona
                           ΑZ
##
   5 Arkansas
                           AR
                                               112.1
##
  6 California
                           CA
                                               55.1
  7 Colorado
                           CO
                                               67.7
##
  8 Connecticut
                           CT
                                              68.7
## 9 Delaware
                           DE
                                               95.4
## 10 District of Columbia DC
                                               34.5
## # ... with 42 more rows
```

Now we can just divert to our standard tidyverse cleaning skills (janitor functions help here) to tidy it up:

```
d_prescriptions <- rough_table |>
  janitor::row_to_names(1) |>
  janitor::clean_names() |>
  rename(prescribing_rate = opioid_dispensing_rate_per_100) |>
  mutate(prescribing_rate = as.numeric(prescribing_rate))
d_prescriptions
```

```
## # A tibble: 51 x 3
##
      state
                            state_abbreviation prescribing_rate
##
      <chr>
                            <chr>>
                                                            <dbl>
##
                                                           126.
   1 Alabama
                            AL
##
    2 Alaska
                            AK
                                                            68.5
                                                            80.9
##
   3 Arizona
                            ΑZ
##
   4 Arkansas
                            AR
                                                           112.
## 5 California
                            CA
                                                            55.1
##
   6 Colorado
                            CO
                                                            67.7
##
  7 Connecticut
                            CT
                                                            68.7
## 8 Delaware
                            DE
                                                            95.4
## 9 District of Columbia DC
                                                            34.5
## 10 Florida
                            FI.
                                                            84.3
## # ... with 41 more rows
```

Now we have clean data for 2008!

Take-aways

This example showed you how to extract a particular table from a particular website. The take-away is to inspect the page html, find where what you want is hiding, and then use the tools in rvest (html_nodes() and html_text() particularly useful) to extract it.

Question 1

Add a year column to d_prescriptions.

```
d_prescriptions <- d_prescriptions |> mutate(Year = 2008)
d_prescriptions
```

```
## # A tibble: 51 x 4
##
                            state_abbreviation prescribing_rate
      state
                                                                   Year
##
      <chr>
                                                            <dbl> <dbl>
                                                            126.
                                                                    2008
##
    1 Alabama
                            AT.
##
    2 Alaska
                            AK
                                                              68.5
                                                                    2008
##
                                                              80.9
    3 Arizona
                            ΑZ
                                                                    2008
##
   4 Arkansas
                            AR
                                                            112.
                                                                    2008
##
    5 California
                            CA
                                                              55.1
                                                                    2008
##
    6 Colorado
                            CO
                                                              67.7
                                                                    2008
##
   7 Connecticut
                            CT
                                                              68.7
                                                                    2008
   8 Delaware
                            DE
                                                              95.4
                                                                    2008
## 9 District of Columbia DC
                                                              34.5
                                                                    2008
## 10 Florida
                                                              84.3
                                                                    2008
## # ... with 41 more rows
```

Getting all the other years

Now I want you to get data for 2008-2019 and save it into one big tibble. If you go to cdc.gov/drugoverdose/rxrate-maps/index.html, on the right hand side there's hyperlinks to all the years under "U.S. State Opioid Dispensing Rate Maps".

Click on 2009. Look at the url. Confirm that it's exactly the same format as the url for 2008, except the year has changed. This is useful, because we can just loop through in an automated way, changing the year as we go.

Question 2

Make a vector of the urls for each year, storing them as strings.

```
base_url = "https://www.cdc.gov/drugoverdose/rxrate-maps/state"
urls = numeric(0)
for(i in 2008:2019){
   urls[i-2007] = paste(base_url, as.character(i), ".html", sep="")
}
urls
```

```
##
    [1] "https://www.cdc.gov/drugoverdose/rxrate-maps/state2008.html"
##
    [2] "https://www.cdc.gov/drugoverdose/rxrate-maps/state2009.html"
##
    [3] "https://www.cdc.gov/drugoverdose/rxrate-maps/state2010.html"
    [4] "https://www.cdc.gov/drugoverdose/rxrate-maps/state2011.html"
##
##
    [5] "https://www.cdc.gov/drugoverdose/rxrate-maps/state2012.html"
##
       "https://www.cdc.gov/drugoverdose/rxrate-maps/state2013.html"
##
       "https://www.cdc.gov/drugoverdose/rxrate-maps/state2014.html"
##
    [8] "https://www.cdc.gov/drugoverdose/rxrate-maps/state2015.html"
    [9]
        "https://www.cdc.gov/drugoverdose/rxrate-maps/state2016.html"
##
   [10] "https://www.cdc.gov/drugoverdose/rxrate-maps/state2017.html"
##
  [11] "https://www.cdc.gov/drugoverdose/rxrate-maps/state2018.html"
## [12] "https://www.cdc.gov/drugoverdose/rxrate-maps/state2019.html"
```

Question 3

Extract the prescriptions data for the years 2008-2019, and store in the one tibble. Make sure you have a column for state, state abbreviation, prescription rate and year. Note if you are looping over years/urls (which is probably the easiest thing to do), it's good practice to include a Sys.sleep(1) at the end of your loop, so R waits for a second before trying again.

Plot prescriptions by state over time.

```
cdcpage <- urls[1]</pre>
  d_prescriptions_full <- read_html(cdcpage) |> html_nodes("tr") |> html_text() |>
    as tibble() |>
    separate(value, into = c("state", "abbrev", "rate"), sep = "\n", extra = "drop") |>
    mutate(Year = 2008)
for(i in 2009:2019){
  cdcpage <- urls[i-2007]</pre>
  new tibble <- read html(cdcpage) |> html nodes("tr") |> html text() |>
    as_tibble() |>
    separate(value, into = c("state", "abbrev", "rate"), sep = "\n", extra = "drop") |>
    mutate(Year = i)
  d_prescriptions_full <- full_join(d_prescriptions_full, new_tibble)</pre>
  Sys.sleep(1)
## Joining, by = c("state", "abbrev", "rate", "Year")
```

```
## Joining, by = c("state", "abbrev", "rate", "Year")
```

```
d_prescriptions_full
```

```
## # A tibble: 627 x 4
##
      state
                            abbrev
                                                rate
                                                                                  Year
##
      <chr>
                            <chr>>
                                                <chr>
                                                                                 <dbl>
##
   1 State
                            State Abbreviation Opioid Dispensing Rate per 100
                                                                                  2008
                                                126.1
   2 Alabama
                                                                                  2008
##
                            AT.
##
   3 Alaska
                            AK
                                                68.5
                                                                                  2008
## 4 Arizona
                            ΑZ
                                                80.9
                                                                                  2008
## 5 Arkansas
                            AR
                                                112.1
                                                                                  2008
  6 California
                            CA
                                                55.1
                                                                                  2008
##
   7 Colorado
                            CO
                                                67.7
                                                                                  2008
##
##
  8 Connecticut
                            CT
                                                68.7
                                                                                  2008
## 9 Delaware
                            DF.
                                                95.4
                                                                                  2008
## 10 District of Columbia DC
                                                34.5
                                                                                  2008
## # ... with 617 more rows
```

Question 4: Install rstan and brms

We will be using the packages rstan and brms from next week. Please install these. Here's some instructions:

- https://github.com/paul-buerkner/brms
- https://github.com/stan-dev/rstan/wiki/RStan-Getting-Started

In most cases it will be straightforward and may not need much more than install.packages(), but you might run into issues. Every Stan update seems to cause problems for different OS.

To make sure it works, run the following code:

```
# library(brms)
# x <- rnorm(100)
# y <- 1 + 2*x + rnorm(100)
# d <- tibble(x = x, y= y)
# mod <- brm(y~x, data = d)
# summary(mod)</pre>
```