Statistical Computing with R Masters in Data Science 503 (S6) Second Batch, SMS, TU, 2023

Shital Bhandary

Associate Professor

Statistics/Bio-statistics, Demography and Public Health Informatics
Patan Academy of Health Sciences, Lalitpur, Nepal

Faculty, Data Analysis and Decision Modeling, MBA, Pokhara University, Nepal Faculty, FAIMER Fellowship in Health Professions Education, India/USA.

Review Preview (Unit 2, Part 1)

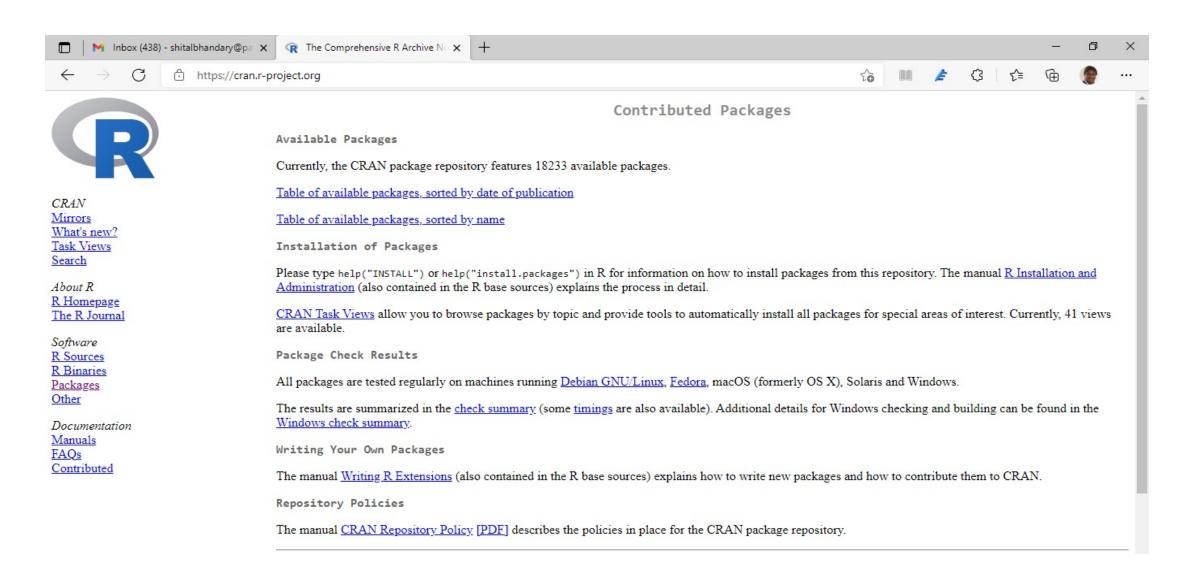
- Package in R
- Using package in R
- Reading data in R
- Reviewing data in R
- Cleaning data in R

What is a "package" in R?

- In R, the fundamental unit of shareable code/function is the package.
- A package bundles together code, data, documentation, and tests, and is easy to share with others.
- As of June 2019, there were over 14,000 packages available on the Comprehensive R Archive Network, or CRAN, the public clearing house for R packages.
- This huge variety of packages is one of the reasons that R is so successful: the chances are that someone has already solved a problem that you're working on, and you can benefit from their work by downloading their package.

Link: https://r-pkgs.org/intro.html from R-packages book: https://r-pkgs.org/index.html

Available "packages" from CRAN:



"Packages" details at CRAN:

Related Directories

Archive

Previous versions of the packages listed above, and other packages formerly available.

<u>Orphaned</u>

Packages with no active maintainer, see the corresponding README.

bin/windows/contrib

Windows binaries of contributed packages

bin/macosx/contrib

macOS High Sierra binaries of contributed packages

bin/macosx/el-capitan/contrib

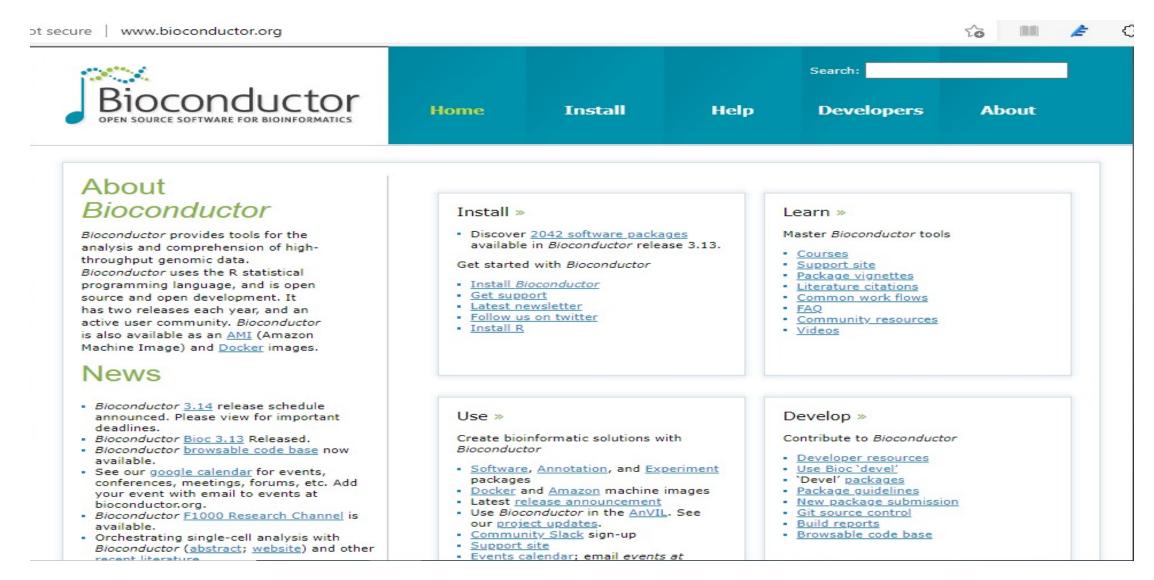
OS X El Capitan binaries of contributed packages

How to install, use and get help about any package from CRAN?

```
    We can install packages of CRAN in R using:
install.packages("packagename")
    e.g. install.packages("dplyr")
```

- We can then use the installed packages in R using: library(packagename)
 i.e. library(dplyr)
- We can get help on the installed packages in R using: ?packagement or help(package = "packagement")
 e.g. ?dplyr OR help(package = "dplyr")

"Packages" from Bioconductor repository: To work with Bioinformatics!



https://hilaryparker.com/2014/04/29/writing-an-r-package-from-scratch/

- Step 0: Packages you will need
- Step 1: Creating your package directory
- Step 2: Add functions
- Step 3: Add documentation
- Step 4: Process your documentation
- Step 5: Install
- Step 6: Make a package GitHub repo (Bonus!)
- Step 7: Infinity- Iterate

You do not need to create package in R for this course but it is required to know how to do it so that you can do it if required.

https://hilaryparker.com/2014/04/29/writing-an-r-package-from-scratch/

Step 0: Packages you will need

```
install.packages("devtools")
library("devtools")
devtools::install_github("klutometis/roxygen")
library(roxygen2)
```

https://hilaryparker.com/2014/04/29/writing-an-r-package-from-scratch/

Step 1: Creating your package directory

setwd("parent_directory")
create("cats")



- If you look in your parent directory, you will now have a folder called cats, and in it you will have two folders: man & R and, one file called DESCRIPTION.
- You should edit the DESCRIPTION file to include all of your contact information

https://hilaryparker.com/2014/04/29/writing-an-r-package-from-scratch/

```
Step 2: Add functions
For example:
                                                   DESCRIPTION
cat_function <- function(love=TRUE){</pre>
                                                   man
    if(love==TRUE){
         print("I love cats!")
                                                     cat_function.R
                                                   cats-package.r
    else {
         print("I am not a cool person.")
```

Save this as a cat_function.R to your R directory.

https://hilaryparker.com/2014/04/29/writing-an-r-package-from-scratch/

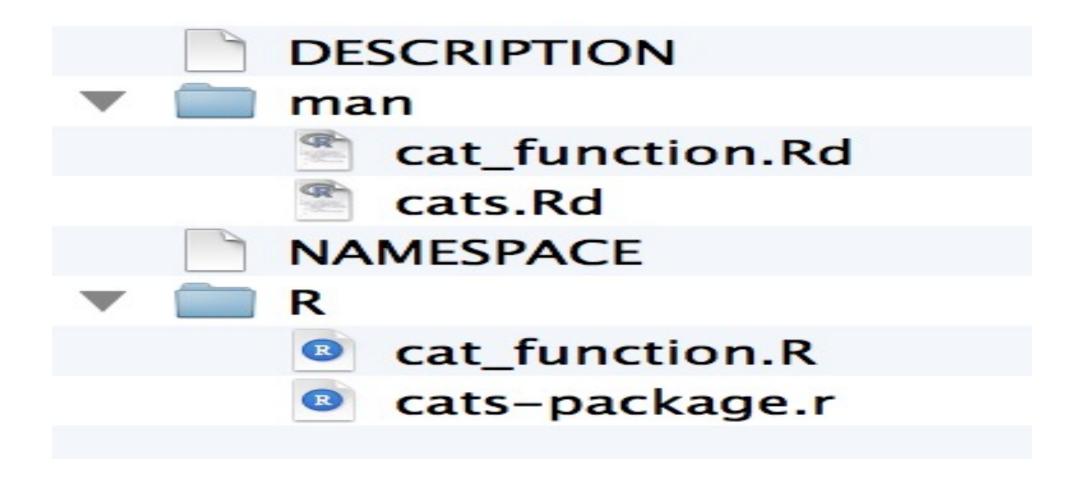
Step 3: Add documentation

```
A Cat Function
       This function allows you to express your love of cats.
       @param love Do you love cats? Defaults to TRUE.
       @keywords cats
       @export
    #' @examples
       cat_function()
10
    cat_function <- function(love=TRUE){</pre>
         if(love==TRUE){
             print("I love cats!")
13
        else {
             print("I am not a cool person.")
```

https://hilaryparker.com/2014/04/29/writing-an-r-package-from-scratch/

- Step 4: Process your documentation
- Now you need to create the documentation from your
- setwd("./cats")
- document()
- This automatically adds in the .Rd files to the man directory, and adds a NAMESPACE file to the main directory.
- More here: https://r-pkgs.org/description.html

This is what you will see:



https://hilaryparker.com/2014/04/29/writing-an-r-package-from-scratch/

- Step 5: Install
- Now it is as simple as installing the package! You need to run this from the parent working directory that contains the cats folder.

- setwd("..")
- install("cats")
- Now you have a real, live, functioning R package.

For example, try typing ?cat_function

cat_function {cats}

A Cat Function

Description

This function allows you to express your love of cats.

Usage

cat function(love = TRUE)

Arguments

love Do you love cats? Defaults to TRUE.

Examples

cat_function()

Step 6: Make the package a GitHub repo (Bonus!)

• The benefit, however, to putting your package onto GitHub is that you can use the devtools install_github() function to install your new package directly from the GitHub page.

install_github('cats','github_username')

Step 7.1: Infinity

- This is where the benefit of having the package pulled together really helps.
- You can flesh out the documentation as you use and share the package.
- You can add new functions the moment you write them, rather than waiting to see if you'll reuse them.
- You can divide up the functions into new packages.
- The possibilities are endless!

Note: "... best products are built in small steps, not by waiting for a perfect final product to be created"

Step 7.2: Iterate

The best products are built in small steps, not by waiting for a perfect final product to be created

This concept is called the <u>minimum viable product</u> — it's best to get a project started and improve it through iteration.

R packages can seem like a big, intimidating feat, and they really shouldn't be. **The minimum viable R package is a package with just one function!**

Reading data in R/R Studio:

- Text files: R base, readr etc. #Already covered in Unit 1
- Excel files: readXL, openxls etc. #Already covered in Unit 1
- SPSS, Stata, SAS files: foreign, haven etc. #Already covered in Unit 1
- JSON files: rjason, jsonlite, RJSONIO etc.

Where, JSON = JavaScript Object Notation, used a lot in websites!

Creating a JSON file using a Text Editor (Notepad): https://www.tutorialspoint.com/r/r_json_files.htm

• Example of a JavaScript Object Notation (JSON) file:

```
"ID":["1","2","3","4","5","6","7","8"],
   "Name":["Rick","Dan","Michelle","Ryan","Gary","Nina","Simon","Guru"],
   "Salary":["623.3","515.2","611","729","843.25","578","632.8","722.5"],
   "StartDate":["1/1/2012","9/23/2013","11/15/2014","5/11/2014","3/27/2015","5/2 1/2013","7/30/2013","6/17/2014"],
   "Dept":["IT","Operations","IT","HR","Finance","IT","Operations","Finance"]
}
```

- It can be typed in text editor and saved with .json extension e.g. jason_data.json
- Install.packages("rjson")

Read the created JSON file in R and Convert it as data.frame for further manipulation in R:

- library("rjson")
- data <- fromJSON(file = "jason_data.json")
 # jason_data.json must be in the working directory of R!
- print(data)
- Covert to data frame:
- jason_data_frame <- as.data.frame(data)
- print(jason_data_frame) #Get summary, histogram of salary,
 # Average salary by department
 #Frequency distribution of all variables

Reading JSON file from URL: Web API

https://www.geeksforgeeks.org/working-with-json-files-in-r-programming/

- install.packages("jsonlite") #Package "RJSONIO" also works!
- library(jsonlite)
- Raw <- from JSON ("https://data.ny.gov/api/views/9a8cvfzj/rows.json?accessType=DOWNLOAD") #Large list!
- food market <- Raw[['data']]
- str(food_market)
- head(food_market)
- Names <- food_market[,14]
- heads(Names)

#Large Matrix, 28472 rows and 24 columns!

#Large characters, Col 14 only!

#Few names from Column 14!

What more can you do with the food_market data?

Try: table(Names)

Try: table(V19)

Try: table(food_market\$V19) #What is "atomic vector"?

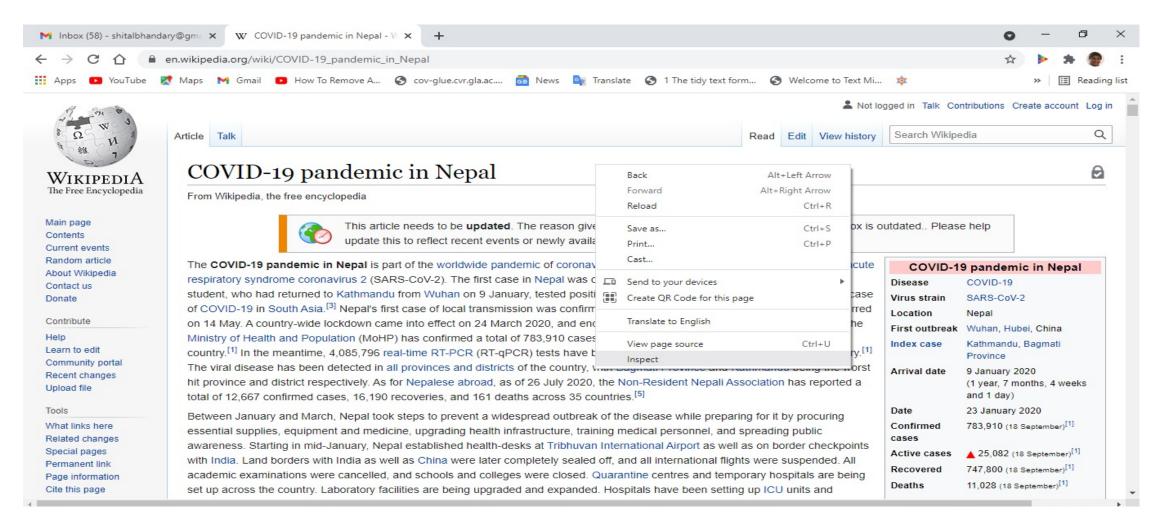
Try: table(food_market[,19]) #What do you get?

Convert the food_market data to data.frame and get summary, create plots
of all the "useful" variables and compute appropriate averages too!

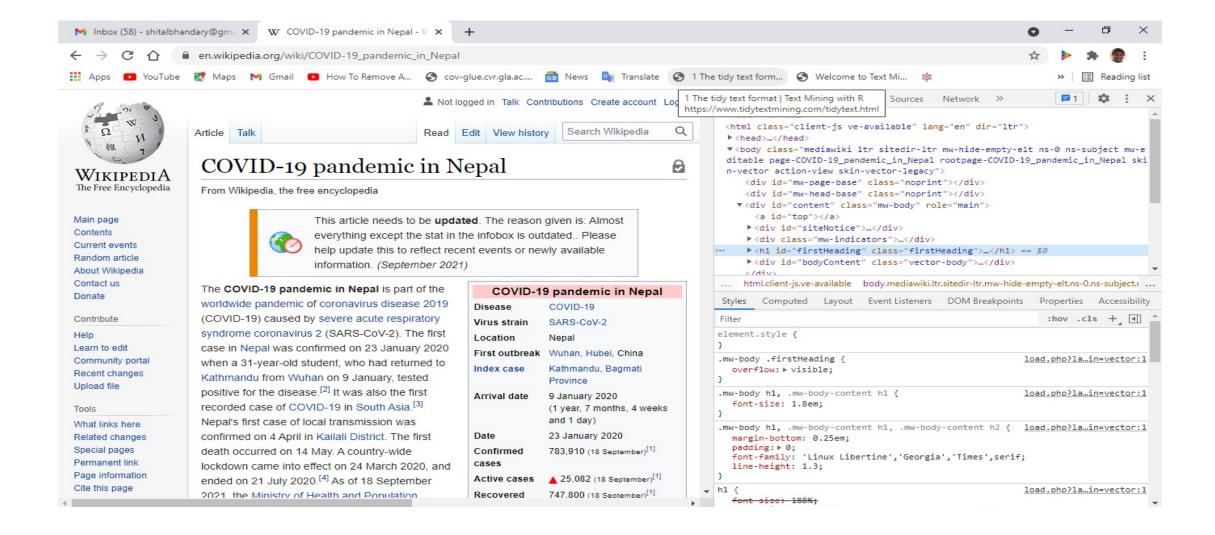
Web scrapping in R: A Simple (barebones) Example https://www.dataquest.io/blog/web-scraping-in-r-rvest

- The recommended package for web scrapping in R is "rvest"
- install.packages("rvest")
- library(rvest)
- simple <- read_html("https://dataquestio.github.io/web-scraping-pages/simple.html")
- simple %>%html_nodes("p") %>%html text()

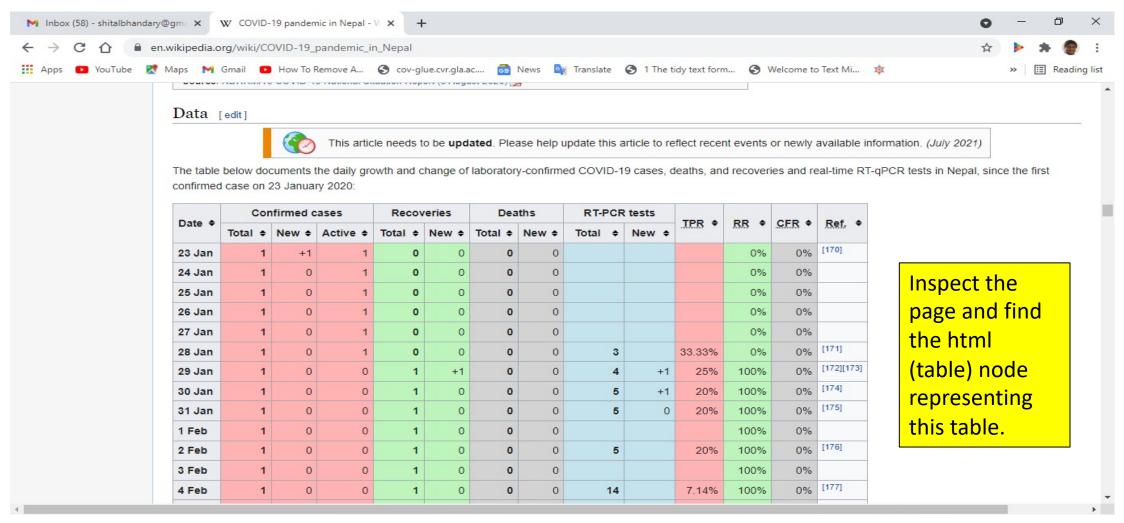
How to do complex web scrapping? Inspect the HTML elements!



What to do now?



We/you need to scrap this data (table) in R: And create plots, get summaries etc.



We can do as follows:

- library(rvest)
- wiki_link <- "https://en.wikipedia.org/wiki/COVID-19_pandemic_in_Nepal"
- wiki_page <- read_html(wiki_link) #rvest package
- str(wiki_page)
- wiki_page %>% html_nodes("table") #to get tables with index
- Find the node number with class="wikitable sortable" as per inspect
 - [7] \n<caption>\n</caption>\n<tbod ...

Let's grab this node in R as "covid_table"

- covid_table <- wiki_page %>% html_nodes("table") %>% .[7] %>% html_table() %>% .[[1]]
- str(covid_table) #tibble [496 x 14]; tibble = fast data frame!
- #Changing names of the columns by adding values of first row
 - names(covid_table) <- paste(names(covid_table), covid_table[1,], sep = "_")
- #Removing the first row thereafter
 - covid_table <- covid_table[-1,]
- #Check the structure of data again
- str(covid_table)

Now do as follows in R for "covid-table" data:

- Change "Date_Date" variable as "Date"
- Change "Confirmed cases_Total" variable as "Confirmed_Cases_Total"
- Change "Confirmed cases_New" variable as "Confirmed_Cases_New"
- Change "Confirmed cases_Active" variable as "Confirmed_Cases_Active"
- Change "RT-PCR tests_Total" variable as "RT-PCR_tests_Total"
- Change "RT-PCR tests_New" variable as "RT-PCR_tests_New"
- Change "TPR_TPR" variable as "TPR"
- Change "RR_RR" variable as "RR"
- Change "CFR_CFR" variable as "CFR"
- Change "Ref. _Ref." variable as "Ref"

Check if this works or not! You can use other method too!

```
    colnames(covid_table) <- c("Date", "Confirmed_Cases_Total",
        "Confirmed_Cases_New", "Confirmed_Cases_Active",
        "Recoveries_Total", "Recoveries_New", "Deaths_Total",
        "Deaths_New", "PCR_Total", "PCR_New", "TPR", "RR", "CFR", "Ref")</li>
```

• str(covid table)

OR, will this also work?

```
15
    colnames(covid_table)
    names(covid_table)[names(covid_table) == "Date_Date"] = "Date"
16
    names(covid_table)[names(covid_table) == "Confirmed cases_Total"] = "Confirmed_Cases_Total"]
17
    names(covid_table)[names(covid_table) == "Confirmed cases_New"] = "Confirmed_Cases_New"]
18
19
    names(covid_table)[names(covid_table) == "Confirmed cases_Active"] = "Confirmed_Cases_
    names(covid_table)[names(covid_table) == "RT-PCR tests_Total"] = "PCR_Total"
20
    names(covid_table)[names(covid_table) == "RT-PCR tests_New"] = "PCR_New"
21
    names(covid_table)[names(covid_table) == "TPR_TPR"] = "TPR"
22
    names(covid_table)[names(covid_table) == "RR_RR"] = "RR"
23
    names(covid_table)[names(covid_table) == "CFR_CFR"] = "CFR"
24
    names(covid_table)[names(covid_table) == "Ref._Ref."] = "Ref"
25
    colnames(covid_table)
26
```

Removing "+" and "%" from variables:

#Removing + from four variables

- covid_table\$Confirmed_Cases_New <- gsub('[+]', '', covid_table\$Confirmed_Cases_New)
- covid_table\$Recoveries_New <- gsub('[+]', '', covid_table\$Recoveries_New)
- covid_table\$Deaths_New <- gsub('[+]', '', covid_table\$Deaths_New)
- covid_table\$PCR_New <- gsub('[+]', '', covid_table\$PCR_New)

#Removing % from three variables

- covid_table\$TPR <- gsub('[%]', '', covid_table\$TPR)
- covid_table\$RR <- gsub('[%]', '', covid_table\$RR)
- covid_table\$CFR <- gsub('[%]', '', covid_table\$CFR)

#Converting "chr" variables as integers 1

- covid_table\$Confirmed_Cases_Total <as.integer(covid_table\$Confirmed_Cases_Total)
- covid_table\$Confirmed_Cases_New <as.integer(covid_table\$Confirmed_Cases_New)
- covid_table\$Confirmed_Cases_Active <as.integer(covid_table\$Confirmed_Cases_Active)
- covid_table\$Recoveries_Total <as.integer(covid_table\$Recoveries_Total)
- covid_table\$Recoveries_New <as.integer(covid_table\$Recoveries_New)

#Converting "chr" variables as integers 2

- covid_table\$Deaths_Total <- as.integer(covid_table\$Deaths_Total)
- covid_table\$Deaths_New <- as.integer(covid_table\$Deaths_New)
- covid_table\$PCR_Total <- as.integer(covid_table\$PCR_Total)
- covid_table\$PCR_New <- as.integer(covid_table\$PCR_New)

#Converting "chr" variables as numbers

- covid_table\$TPR <- as.numeric(covid_table\$TPR)
- covid_table\$RR <- as.numeric(covid_table\$RR)
- covid_table\$CFR <- as.numeric(covid_table\$CFR)

How to change "date" variable?

- The date is shown as "23 Jan", "24 Jan", "25 Jan" etc.
- You need to use as.Date function
- What is the default Date values to use this function?
- Can you use different format to covert?
- This is an assignment for you!

#OR, create "date2" variable as follows:

- date2 = seq(as.Date('2020-1-23'), by='days', length.out = 495)
- covid_table = cbind(covid_table, date2)
- str(covid_table)
- summary(covid_table)
- Save this cleaned data as "covid_table.csv" file for next class!

Also see these posts to know more on web scrapping with "rvest" in R:

• https://kyleake.medium.com/wikipedia-data-scraping-with-r-rvest-in-action-3c419db9af2d

• https://www.engineeringbigdata.com/web-scraping-wikipedia-world-population-rvest-r/

• https://stackoverflow.com/questions/33360634/how-to-scrape-data-from-wikipedia-using-r

 Web scraping is an extremely popular amongst researchers and web developers. The best example may be Google Search.

• When you use Google to find information, you are (in highly over simplified terms) not actually searching the "live" internet, but rather a database of webpages that Google has mapped.

If Google is allowed to do it, why can't you!?

- Before assuming that your project is both legal and ethical, you need to ask yourself a lot of questions regarding the nature, scope, and purpose of your web scraping activities.
- If your web scraping project is focused on collecting data that is in the public-domain from publicly available websites and individual datum do not connect directly to individual humans in the real world and your web scraping activities represent minimal technical burden on the part of the website owner, then you are in the clear!

 However, in truth, most web scraping research projects are not so clearly defined.

 While you will have to make the determination yourself regarding whether your project is both legal and ethical, the questions in the subsequent slides are meant to prompt the kind of thinking that may not be immediately obvious the first time you start scraping the internet for data.

1. Research Ethics

- Is the data you are collecting potentially sensitive information?
- If you are scraping user-comments from a social media website, are the users aware that their comments are visible to you or others?
- Are the users fully or partially aware of how their comments and data may be used?
- Do the users have an expectation of anonymity or confidentiality?
- Do the users represent marginalized or at-risk group?
- Does your research pose any form of potential risk to the users who supplied the data you are using?

2. Public vs. Protected Content

- Are the webpages you are collecting data from freely and publicly visible? OR...
- Are the webpages you are collecting data password-protected, requiring you to log into the website?
- If webpages and content are password-protected, does the website require you to adhere to a "Terms of Service", "Terms of Use" or other type of agreement in order to access and use the website?
 - Often these agreements explicitly forbid systematic web-scraping activities

3. Copyright & Commercial Activity

 Are you violating copyright as part of your overall as part of your web-scraping activities?

 Are you reproducing the data or contents of webpages on your own website or in another medium?

• If you are reproducing or embedding the content in some way, do you have the site owner's permission?

4. Sustainability

- Are you systematically collecting large volumes of webpages at a high rate from the target website?
- Are you systematically collecting on a repeating schedule at a rapid rate?
- Are you collecting webpages from the website in such a way that poses commercial and/or technical risk to the technical operations of the target website?

More on "ethical issues" with the use of web scrapping/web APIs are here:

• https://towardsdatascience.com/ethics-in-web-scraping-b96b18136f01

• https://blogs.mulesoft.com/api-integration/strategy/ethics-of-apis/

Self-learning!

Question/Queries?

Thank you!

@shitalbhandary