Part 6: Putting your project on Github

In the previous chapters, we've set up a Github account, learned how to download the files from a Github repository, upload files into RStudio.Cloud, create and run R code, and commit changes using Git.

In this final chapter, we are going to create some figures and graphs for this project, then put these changes on Github in a way for people to find and share.

Moving R code into the .Rmd file

In the last chapter, we used the README.Rmd file to upload our data into RStudio. We'll continue using this file to add the contents from one additional .R scripts, 02-wrangle.R. We will then create new section and script for visualizing the data (03-visualize.R).

We've provided a lot of code and comments in the 02-wrangle.R script for you to explore, revise, and adapt to your liking. In the next few sections, we are going to move the code from the 02-wrangle.R script into two new sections of the README.Rmd (you probably guessed it "Wrangle").

A quick lesson in compassionate programming

Your code will always be communicating to at least two audiences: your computer, and your future self. Be nice to both of them!*

Things like the pipe %>% in R can help with clarity. The pipe is part of the magrittr package and it takes code written like this:

```
outer_function(inner_function(Data_X), Data_Y)
```

And makes it look like this:

```
Data_X %>% # do this
  inner_function() %>% # then do this
  outer_function(Data_Y)
```

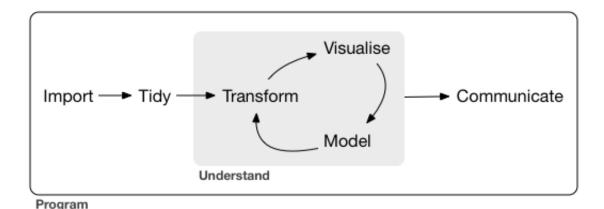
%>% is a form of syntactic sugar, which is a fancy way of saying "something that helps us communicate better."

You'll see the pipe throughout the project's R code files, and you can always read it as, "do this, then do this."

Wrangling code

Data wrangling (or cleaning, or munging) is whatever steps need to be taken to take the raw data (which we always remember not to change) into something we can use to create a table, visualization, model, etc.

The 02-wrangle.R script prepares the data from the 01-import.R so that they can be used for the visualizations. If you think back to the process outlined in the figure in R for Data Science, you will notice that wrangle isn't listed explicitly. This is because both 'Tidy' and 'Transform' would be considered wrangling steps (both of these need to happen before any visualizations or models can be properly run).

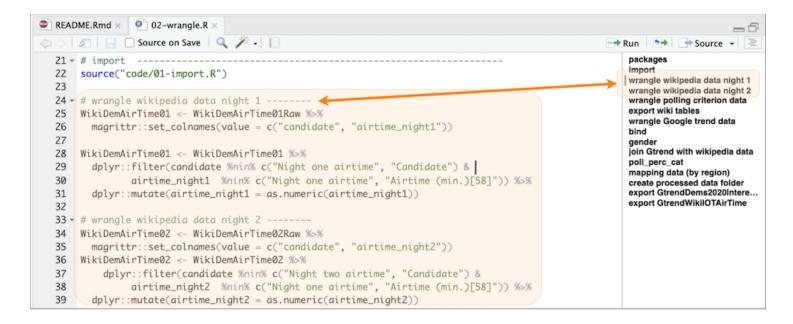


The simplest way to include this script in the README.Rmd file is to create a code chunk, insert the base::source() function, and enter the path to the 02-wrangle.R file.

However, we want to be nice to our future selves, so we will include some language that describes what the functions are doing above each code chunk.

Wrangling the data sets

The first data that need to be wrangled are the Wikipedia tables (seen in the 02-wrangle.R file on the section below).



These data were stored on the web in Wikipedia (.html) tables. We will create some new column names, remove some columns that used to be headers, and make the airtime variable numeric.

The polling criterion Wikipedia data starting at the section titled, wrangle polling criterion data. This section actually creates a list of candidates (in cand_names_wiki) and uses it to filter out the observations we want. Check out this webinar to get an understanding of how dplyr's verbs work.

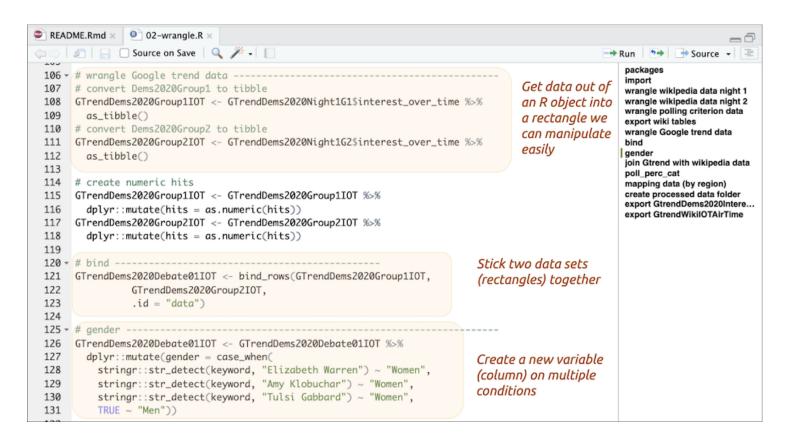
```
# wrangle polling criterion data -----
# create list from names using dput()
# dput(WikiPollCriterionRaw[ 1:11, 1])
```

After the Wikipedia tables have been wrangled, the script exports these files to a new processed/ folder. This helps ensure they won't be accidentally altered or mistaken for the data files in the raw/ data folder.

The export section also timestamps each file so we know the last time it was created. Read more about importing and exporting data in this RStudio cheatsheet.

The Google trend data are a little more complicated because they come into RStudio.Cloud as a list, which is a data container in R that doesn't have to be rectangular.

The image below outlines what each portion of code is doing. These are fairly common wrangling tasks, so we recommend going back or bookmarking these files as a reference.



We have different sources of data in RStudio right now (Wikipedia and Google trend data). They both have information on Candidates though. Often times we'll want to join two (or more) data sets on a common column (like candidates). We will perform an example of this in the section outlined below.

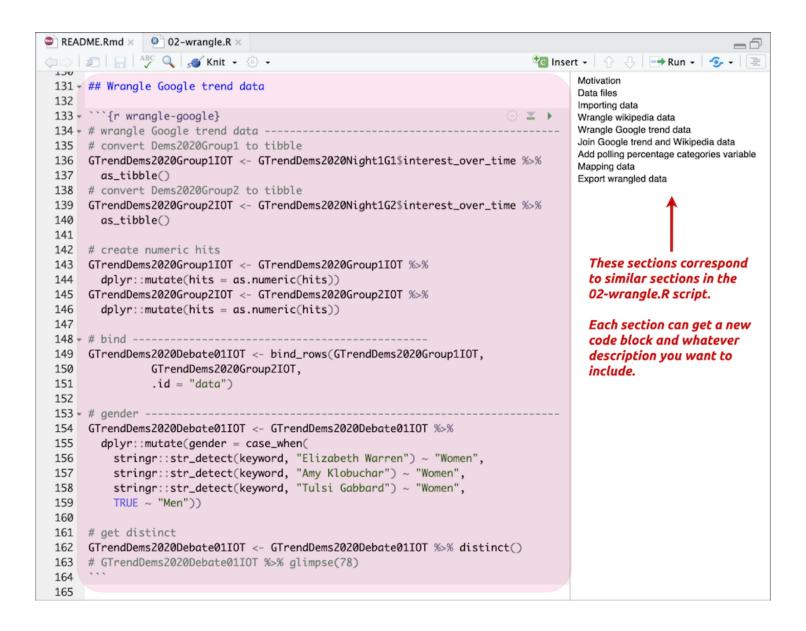
```
gender
# join Gtrend with wikipedia data -----
                                                                                                     join Gtrend with wikipedia data
# sort alphabetically, join on id
                                                                                                     poll perc cat
WikiDemAirTime01 <- WikiDemAirTime01 %>% dplyr::arrange(desc(candidate))
                                                                                                     mapping data (by region)
                                                                                                     create processed data folder
                                                                                                     export GtrendDems2020Intere...
WikiDemAirTime01 <- WikiDemAirTime01 %>%
                                                                                                     export GtrendWikiIOTAirTime
  mutate(candidate_id = row_number())
                                                                              Prep work to get
# create candidate_id for GTrendDems2020Debate01IOT
                                                                              a common id in
GTrendDems2020Debate01IOT <- GTrendDems2020Debate01IOT %>%
                                                                              each data set
  dplyr::mutate(candidate_id = case_when(
    stringr::str_detect(string = keyword, pattern = "Warren") ~ 1,
    stringr::str_detect(string = keyword, pattern = "Ryan") ~ 2,
    stringr::str_detect(string = keyword, pattern = "Beto") ~ 3,
    stringr::str_detect(string = keyword, pattern = "Klobuchar") ~ 4,
    stringr::str_detect(string = keyword, pattern = "Inslee") ~ 5,
    stringr::str_detect(string = keyword, pattern = "Gabbard") ~ 6,
    stringr::str_detect(string = keyword, pattern = "Delaney") ~ 7,
    stringr::str_detect(string = keyword, pattern = "de Blasio") ~ 8,
    stringr::str_detect(string = keyword, pattern = "Castro") ~ 9,
    stringr::str_detect(string = keyword, pattern = "Booker") ~ 10)) %>%
  dplyr::arrange(desc(candidate_id))
# Join WikiDemAirTime01 to GTrendDems2020Debate01IOT on candidate_id
                                                                              Joining two data
GtrendWikiIOTAirTime <- GTrendDems2020Debate01IOT %>%
                                                                              sets 'candidate id'
  dplyr::left_join(x = .,
                   y = WikiDemAirTime01,
                   by = "candidate_id")
```

The process usually isn't so involved, but we included extra to give more explicit instructions. Be sure to check out the relational data chapter the R for Data Science book.

We'll also be creating a map with the Google (or Twitter) data. Doing this requires another common task, which is loading a dataset from a package in R. The code below loads a state-level map into RStudio.Cloud and joins it to the Google trend data.

```
wrangle Google trend data
# mapping data (by region) -----
                                                                                                       bind
# convert to tibble (another data structure in R)
                                                                                                       gender
GtrendDems2020IBRGroup1 <- tibble::as_tibble(GTrendDems2020Night1G1$interest_by_region)</pre>
                                                                                                       join Gtrend with wikipedia data
                                                                                                       poll perc cat
GtrendDems2020IBRGroup2 <- tibble::as_tibble(GTrendDems2020Night1G2$interest_by_region)</pre>
                                                                                                       mapping data (by region)
# bind Dems2020IBRGroup1 Dems2020IBRGroup2 together
                                                                                                       create processed data folder
GtrendDems2020IBR <- bind_rows(GtrendDems2020IBRGroup1,</pre>
                                                                                                       export GtrendDems2020Intere...
                                                                                                       export GtrendWikilOTAirTime
                         GtrendDems2020IBRGroup2, .id = "data")
# convert the region to lowercase
GtrendDems2020InterestByRegion <- GtrendDems2020IBR %>%
  dplyr::mutate(region = stringr::str_to_lower(location))
# create a data set for the states in the US
                                                                            Load the data from the
statesMap = ggplot2::map_data("state")
                                                                            ggplot2 package and join
# now merge the two data sources together
                                                                            it to the Google trend
GtrendDems2020InterestByRegion <- GtrendDems2020InterestByRegion %>%
  dplyr::inner_join(x = .,
                                                                            data
                    y = statesMap,
                    by = "region")
```

The Google trend data are also exported with time-stamp into the processed/ folder. We should continue adding the code into the README.Rmd file until we're confident all the functions will run and we don't see any errors.

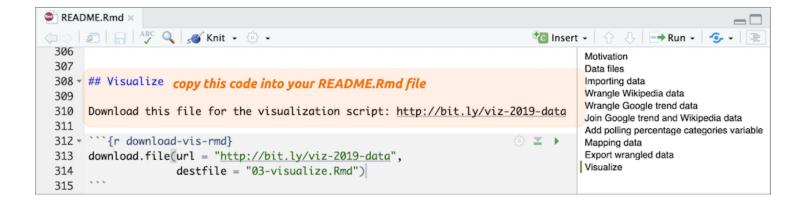


Note: The 02-wrangle.R file is in the code/ folder, but you won't have to alter the file paths because you're using an RStudio project file. Read more about how these are so helpful to your workflow here.

Visualizations

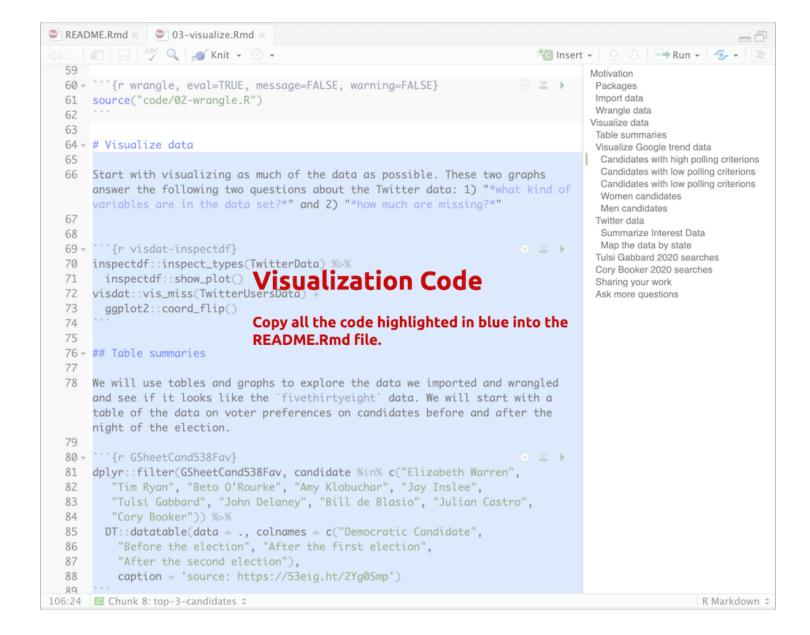
OK, we've completed our section for the wrangling the data. We are going to insert a divider (***) and start a new visualize section (## Visualize) in the README.Rmd file.

We've created a 03-visualize.Rmd file for you to download from Github. You can do this by typing the following code into your README.Rmd file:



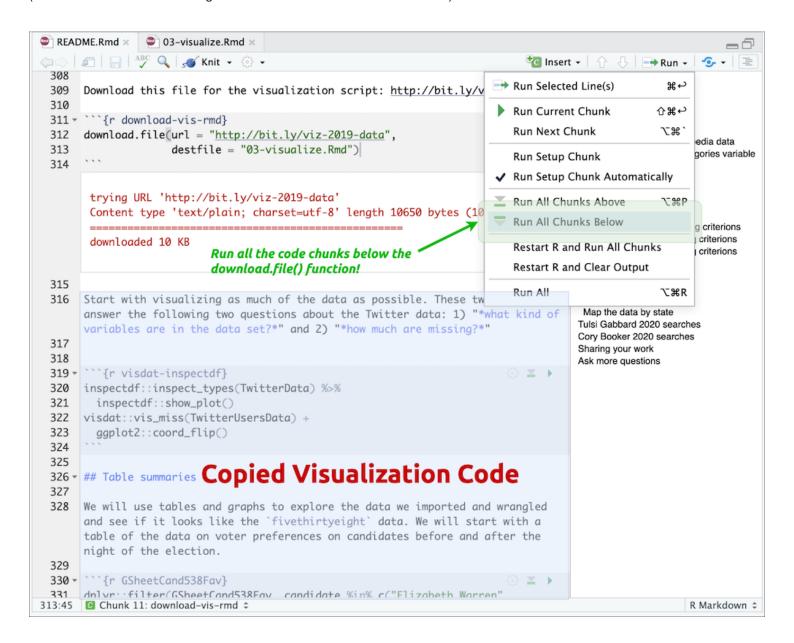
The hyperlink is here: http://bit.ly/viz-2019-data

After we've downloaded our 03-visualize.Rmd file, we can open this file and copy the code starting from the line just below the # Visualize data heading (it should be on about line 65) and extending all the way to the end of the file.



After selecting the code from 03-visualize.Rmd, we should click on the line directly under the previous code chunk we used to download the .Rmd file.

After pasting the code from 03-visualize.Rmd into the README.Rmd file, we can click on the Run > Run All Chunks Below (this will run all the code starting at line 319 until the end of the document).



Running the code will create multiple tables and figures in the README.Rmd file. We'll go over these in more depth below. For now, we'll follow the directions at the bottom of the pasted code and "Click knit to get the markdown file to share."

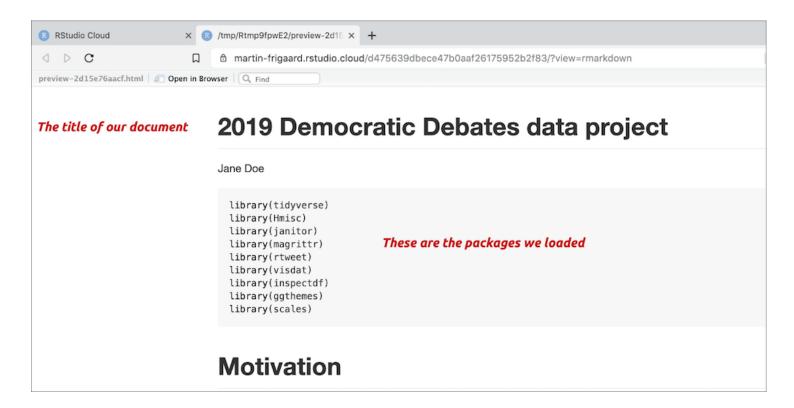
Knitting RMarkdown files

Clicking *Knit* (or clicking shift+cmd+k) activates the **Markdown** pane in RStudio.Cloud, and we see the code chunks being run for the entire document.

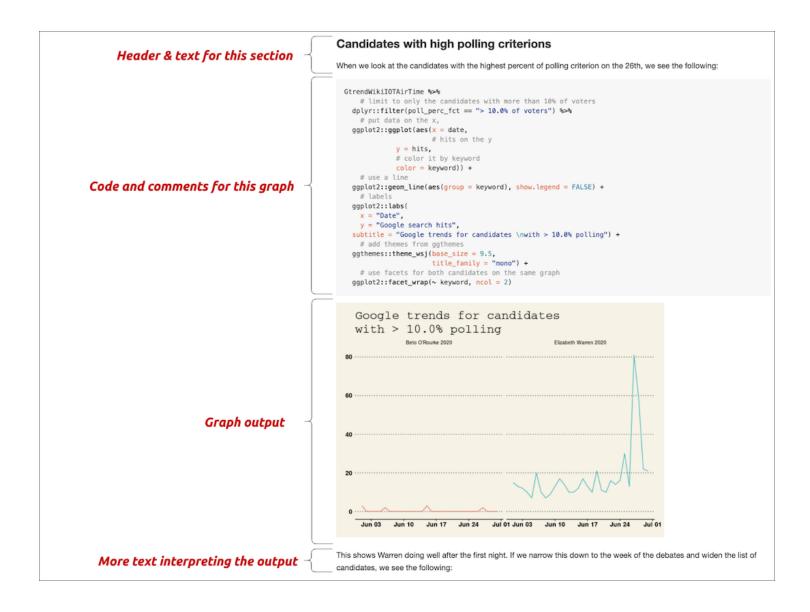


When the knitting process completes, a new browser window will pop up with our README.md document. The README.md will have sections of formatted text (from the Markdown), R code, and the various outputs.

The top of the file should list the title and the packages:

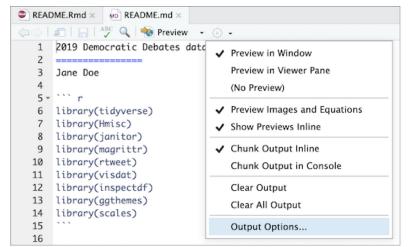


Lets scroll down to the visualize section and look at the section titled, **Candidates with high polling criterions**). We can see the different parts of the Rmarkdown file in the image below:



The file output is actually a *Preview* of our markdown file (README.md). Our browser renders the markdown as a webpage (README.html).

Let's keep a version of this file in .html. We can do this by opening the README.md file in the **Source** pane, and clicking on the small gear next to the *Preview* button. Follow the directions in the figure below to setup the .html output file.



Markdown document options

Click on the small gear icon to get the list of settings for the markdown options.

From this list, select the **Output Options**



Change the settings to match these options

When we click Ok and look at the top of the README.md file we see another YAML header.

```
output:
  html_document:
    df_print: kable
    fig_height: 5.5
    fig_width: 7.5
    highlight: kate
    keep_md: yes
    theme: simplex
    toc: yes
    toc_depth: 6
```

The settings displayed above are some of the ways we can customize our .Rmd files. Read more about the html documents here in the Rmarkdown guide.

To see what these settings do, we will click Preview on the README.md file.

This conversion is incredibly handy for weaving formatted text, code, and media (tables, images, and graphs).

Extracting the .R from the .Rmd

But now we have all our visualize code in the 03-visualize.Rmd file-what if we wanted this code in an .R script?

We can run the following code in the Console pane.

```
knitr::purl("03-visualize.Rmd")
```

We'll see the following script file is generated.

Push the changes to Github