tigeRs: Princeton's R Group

Kim Kreiss & Angela Li

Workshop 1



Introductions

- ► Kim
- ► Angela
- ► DDSS

Today

- An overview of R/RStudio/RMarkdown
- Framework for reproducible analysis
- ► The tidyverse
- Apply these by going through a brief class project

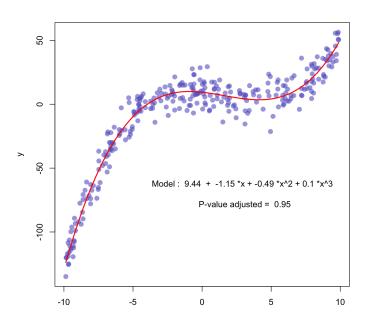
Overview of R/RStudio/RMarkdown

R/RStudio/RMarkdown and Why?

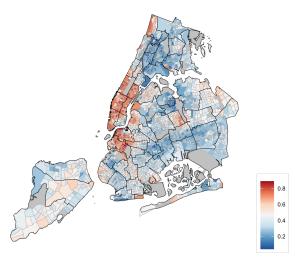
- ▶ R is a statistical programming language great for data analysis and data science applications
- RStudio is an Integrated Development Environment (IDE)-basically just a nice interface for using R, writing/running code, and interacting with data and files
- RMarkdown is a file format that lets you combine R code, data, and text that outputs a document, report, slideshow, etc.

Cool things you can do with R

- Easy and intuitive management of code and data
- Excellent visualization capabilities, including charts, maps, interactive dashboards etc.
- Easily output analysis into a digestible format
- can handle a wide arrange of statistical analysis, data analysis, and data science applications



Share of Creative Class Workers in NYC Census Tracts (by PUMA/Community District)



Source: 2019 American Community Survey 5-year estimates Tracts with less than 30 people are shaded grey.

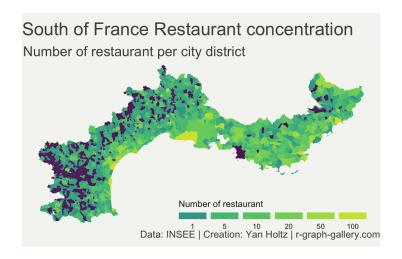


Figure 2: 'Source: r-graph-gallery.com'



Figure 3: 'Source: r-graph-gallery.com'

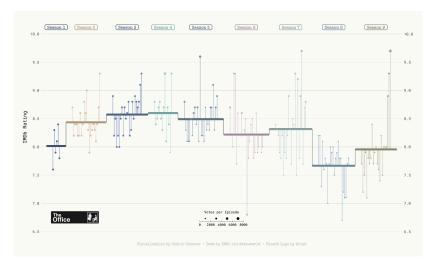


Figure 4: 'Source: r-graph-gallery.com'

A Pandoc Markdown Article Starter and Template *

Steven V. Miller Clemson University

This document provides an introduction to R Markdown, argues for its benefits, and presents a sample manuscript template intended for an academic audience. I include basic syntax to R Markdown and a minimal working example of how the analysis itself can be conducted within R with the knitr package.

Keywords: pandoc, r markdown, knitr

Introduction

Academic workflow, certainly in political science, is at a crossroads. The American Journal of Prelitial Science (Alga) announced a (pw woods) "show your work" initiative in which authors who wish cathors who which authors who will be a considered the substance of the produced the results shown in the manuscript. The editorial team at Afg Ps hen reproduces the code from the manuscript Pending successful replication, the manuscript howes toward publication. The Afg Ps might be at the ore of this movement, and it could be the most aggressive and among political science journals, but other journals in our field have signed the joint Data Access so the Research Transparren () (DAR) initiative. This, at a bare minimum, requires uploading code from quantitatively-oriented published articles to in-house directories hosted by the journal or to services like Dataverse.

There are workflow implications to the Lacour controversy as well. Political science, for the foreseeable future, will struggle with the extent of the data fraud prepertated by Michael Lacour in an article co-authored with Donald P. Green in Science, the general scientific journal of record in the United States. A failure to reproduce LaCour's results with different samples uncovered a comprehensive effort by a Caour to "false" data that provided results to what we felt or believed to be true (t.e. "truthines"). However, false data can have real consequences for both the researcher and those who want to learn from it and use if for various purposes. Even research done honestly may suffer the same false if the researchers are not diligatent in their workflow.

These recent events underscore the DART push and cost a shadow over our workflow. However, good workflow has always been an issue in our discipline. Cloud storage services like Dropbox are still relatively new among political scientists. Without cloud storage, previous workflow led ropen the possibility that work between a home computer and an office computer was lost as a function of a corrupted thumb drive, an overheated power supply, or, among other things, the wave of viruses that would particularly affect Microsoft unes every summer. Scalal sciences, uriliate engineering, have traditionally relied on software like Microsoft Word for manuscrip repearation through any word processor reduces workflow to a series of clicks and strokes on a legitation through any word processor reduces workflow to a series of clicks and strokes on a legitation through any word processor reduces workflow to a series of clicks and strokes on a legitation through any word processor reduces workflow to a series of clicks and strokes on a legitation of the stroke of the s

I think there is reason for optimism. We only struggle with it now because we have tools like R Markdown and Pandoc, more generally, that make significant strides in workflow. LafeX resolved earlier issues of corrupted binary files by reducing documents to raw markup that was little more

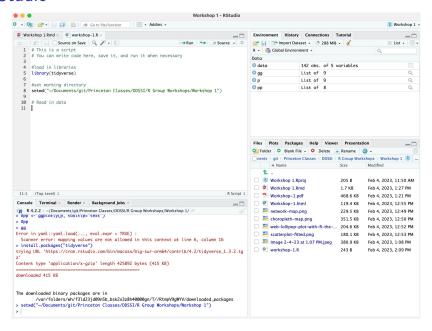
^{*}Replication files are available on the author's Github account (http://github.com/svmiller). Current version: July 26, 2019; Corresponding author: svmille@clemson.edu.

Goals

Today's workshop

- Learn the basics and fundamentals of working in RStudio and RMarkdown
- ► Learn a framework for reproducible analysis that can be applied to homework, work assignments, etc.
- Go through an analysis together to combine these
- Leave with a file that can serve as a template for future work

RStudio



R Scripts and R Markdown

- ► R Script (workshop-1.R)
 - ► A file that only runs normal R code
- RMarkdown Script (workshop-1-solutions.Rmd)
 - ► A different file type that combines code and text to produce a document that includes both
- Both a .R and .Rmd file are provided

Reproducible Analysis

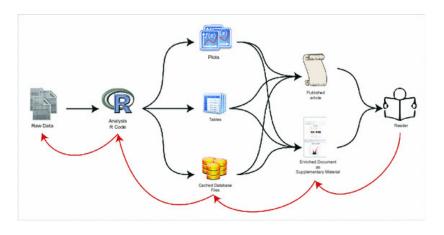


Figure 6: Source: Russo, Righeli, Angelini

Why you should care

- It will save you time and make analyses easy to update/run again
- Eliminates room for mistakes
- Easily shared and validated by others—especially crucial for analysis informing public policy

What to keep in mind for our purposes

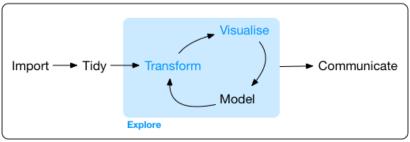
- ► The ideal: someone else (including and especially you, at a later date) should be able to
 - (1) Read your code file and understand what you did and why (code comments are everyone's friend)
 - (2) Re-run your code without making any edits and produce the same results
- In practice:
 - Any data cleaning, transformations to the data, edits, analyses, etc. should be documented in an R script
 - No edits should be made manually!!! (don't edit your data in excel, don't manually enter or copy/paste output, etc.)

Tidyverse



- ▶ A collection of R packages made for data/statistical analysis with the same underlying structure, intuitive syntax and philosophy
- Great for working with and manipulating a wide variety of datasets
- We will use tidyverse today

Tidyverse



Program

Code

What we will do today

- ► Go through and reproduce a shortened version of Project 2 from last semester's 507c class
- use tidyverse to:
 - read in data
 - clean/transform data
 - generate summary statistics
 - run a regression model
 - output results in a summary memo document

The task: replicate a project assignment

- You are working for a non-profit organization advocating for safe and healthy lifestyles and you received a request from the director to provide analysis and information on the effect of the minimum legal drinking age on alcohol consumption:
 - Does crossing the MLDA make it more likely someone will drink?
 - Is there a jump in consumption in the likelihood of drinking at age 21?
 - Are there important differences by demographic groups?
 - ► Can you something about the jump at age 21 for different levels of alcohol use (ie, heavy drinkers?)?

Files

- workshop-1.R
 - this is just an R script with all the code for your reference later
- workshop-1-skeleton.Rmd
 - an RMarkdown code skeleton for you to fill out as we go
- workshop-1-solutions.Rmd
 - the same RMarkdown code file as above, but with all the code and solutions

RMarkdown File Structure: Overview

```
-\Box
(iii) In Knit on Save ABC Q SKnit + 6 +
                                                               * Run - - - +
Source Visual

■ Outline

  1 - ---
  2 title: "Project 2"
  3 author: "Your Name"
  4 date: "2023-02-05"
  5 output: html_document
  6 - ---
     ```{r echo=FALSE, eval=TRUE, warning=FALSE, message=FALSE}
 library(tidyverse)
 10 -
 11
 12 - ## Overview
 13
 14
 This document provides high-level information on whether the minimum legal drinking age
 (MLDA) makes any difference in the likelihood of consuming alcohol. It includes:
 15
 16
 + Code to read in and clean data
 17
 18
 + Summary Statistics
 19
 20
 + Regression output
 21
 22 +
 `{r echo=FALSE, eval=TRUE, warning=FALSE, message=FALSE}
 # read in MLDA data and assian MLDA data
 mlda <- read_csv("mlda.csv")</pre>
 24
 25
 26 -
 27
 28
 29
```

#### RMarkdown File Structure: YAML

```
-\Box
(iii) In Knit on Save ABC Q SKnit • 💮 •
 * Run - - - +
Source Visual

■ Outline

 1 ----
 title: "Project 2"
 YAML
 3 author: "Your Name"
 Some settings for your document
 4 date: "2023-02-05"
 output: html document
 You will probably not do much with
 6 ---
 this beyond what is here
      ```{r echo=FALSE, eval=TRUE, warning=FALSE, message=FALSE}
     library(tidyverse)
 10 -
 11
 12 - ## Overview
 13
 14
     This document provides high-level information on whether the minimum legal drinking age
     (MLDA) makes any difference in the likelihood of consuming alcohol. It includes:
 15
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 17
 18
       + Summary Statistics
 19
 20
       + Regression output
 21
 22 -
        {r echo=FALSE, eval=TRUE, warning=FALSE, message=FALSE}
     # read in MLDA data and assian MLDA data
     mlda <- read_csv("mlda.csv")</pre>
 24
 25
 26 -
 27
 28
 29
```

RMarkdown File Structure: Code

```
-\Box
(iii) In Knit on Save ABC Q SKnit + 6 +
                                                                * Run - - - +
Source Visual

■ Outline

  1 + ---
  2 title: "Project 2"
                                                   CODE CHUNKS
  3 author: "Your Name"
  4 date: "2023-02-05"
                                                 All R Code Goes into a
                                                    CODE CHUNK
  5 output: html_document
  6 - ---
        {r echo=FALSE, eval=TRUE, warning=FALSE, message=FALSE}
     library(tidyverse)
 10 -
 11
 12 - ## Overview
 13
 14
     This document provides high-level information on whether the minimum legal drinking age
     (MLDA) makes any difference in the likelihood of consuming alcohol. It includes:
 15
 16
       + Code to read in and clean data
 17
 18
       + Summary Statistics
 19
 20
       + Regression output
 21
 22 -
        {r echo=FALSE, eval=TRUE, warning=FALSE, message=FALSE}
                                                                                 (6) ≥ >
     # read in MLDA data and assian MLDA data
 23
 24
     mlda <- read_csv("mlda.csv")</pre>
 25
 26
 27
 28
 29
```

RMarkdown File Structure: Text

```
-\Box
(iii) In Knit on Save ABC Q SKnit • 💮 •
                                                                 * Run - - - +
Source Visual

■ Outline

  1 + ---
                                                       TEXT
  2 title: "Project 2"
  3 author: "Your Name"
                                                This is where you write text
  4 date: "2023-02-05"
                                              It uses a syntax called 'Markdown'
  5 output: html document
                                                 which is very easy to learn
  6 - ---
       ``{r echo=FALSE, eval=TRUE, warning=FALSE, message=FALSE}
                                                                                   library(tidyverse)
 10 -
 11
 12 - ## Overview
 13
     This document provides high-level information on whether the minimum legal drinking age
 14
     (MLDA) makes any difference in the likelihood of consuming alcohol. It includes:
 15
 16
       + Code to read in and clean data
 17
 18
       + Summary Statistics
 19
 20
       + Regression output
 21
 22 -
       `{r echo=FALSE, eval=TRUE, warning=FALSE, message=FALSE}
                                                                                   £ ≥ 1
     # read in MLDA data and assian MLDA data
     mlda <- read_csv("mlda.csv")</pre>
 24
 25
 26 -
 27
 28
 29
```

Rmarkdown: Output

Project 2

Your Name 2023-02-05

Overview

This document provides high-level information on whether the minimum legal drinking age (MLDA) makes any difference in the likelihood of consuming alcohol. It includes:

- . Code to read in and clean data
- · Summary Statistics
- Regression output

R Programming Basics

```
# object assignment
# (strings, numbers, dataframes, lists, etc.)
this_is_an_object <- "object"
# this is how you inspect an object
this_is_an_object
## [1] "object"
# another assignment
x <- 2<sup>3</sup>
# another inspection
Х
## [1] 8
      Naming is case sensitive
      Must start with a letter and have no spaces
       i_suggest_using_this_format
       ButSomePeopleDoThis
      objects will be loaded into your environment on the right!
```

Set Up: Working Directory, Package Loading

```
#set working directory
setwd("~/Documents/workshop-1")
# you will only need to install packages once
install.packages("tidyverse")
install.packages("kableExtra")
install.packages("stargazer")
install.packages("knitr")
install.packages("rdrobust")
# you will do this whenever
# you need to load in and use tidyverse
library(tidyverse)
library(kableExtra)
library(stargazer)
library(knitr)
library(rdrobust)
```

Read in and inspect our data

```
mlda <- read_csv("mlda.csv")
mlda
## # A tibble: 61,263 x 11
     hs_diploma hispanic white black emplo~1 married male days_21 perc_~2 drink~3
          <dbl>
                <dbl> <dbl> <dbl>
                                    <dbl>
                                           <dbl> <dbl>
                                                        <dbl>
                                                               <dbl>
                                                                      <dbl>
##
##
                                                       1601
                                                               0.548
                                                       1024 1.10
                                                      2455 6.58
                                                      -590 0
                                                      1815 0
                                                      2424 0
                                        1 0 0 1116 3.29
## 8
                                                      2942 0.548
                                                         2516
                                                               3.29
## 10
                                                         2516 57.0
## # ... with 61.253 more rows, 1 more variable: student <dbl>, and abbreviated
      variable names 1: employed, 2: perc_days_drink, 3: drinks_alcohol
```

Use the viewer to view more of the data (click on it or run code below)

View(mlda)

Data Dictionary

Documentation for MLDA_507

The data file contains data on over 61,000 individuals drawn from the National Health Interview Sample Audit Files from 1997-2007. The data were collected by Professor Carlos Dobkin of UC Santa Cruz, who has done important research on the minimum legal drinking age.

Variable Name	Description
Age and Drinking Variables	
days_21	Days from 21 st birthday. (positive values indicate age ≥ 21; negative values indicate age
	<21)
drinks alcohol	Binary variable = 1 if person reports that he/she drinks alcohol, 0 otherwise
perc_days_drink	Percent of days on which he/she reports drinking alcohol
Other Variables	
HS Diploma	Binary variable = 1 if person has a high school diploma, 0 otherwise
Hispanic	Binary variable = 1 if person is Hispanic, 0 otherwise
Black	Binary variable = 1 if person is Black, 0 otherwise
Employed	Binary variable = 1 if person is employed, 0 otherwise
Student	Binary variable = 1 if person is a student, 0 otherwise
Male	Binary variable = 1 if person is a male, 0 otherwise
Married	Binary variable = 1 if person is married, 0 otherwise

Note: I eliminated all observations with $|days_21| \le 21$ to eliminate alcohol consumption associated with someone's 21^{st} birthday.

Some quick data exploration

[1] 0.6249449

```
mean(mlda$student)

## [1] 0.1004848

mean(mlda$drinks_alcohol)
```

Transform, explore, and visualize

Dplyr

The most useful tool in the tidyverse is dplyr. It's a swiss-army knife for data wrangling. dplyr has many handy functions that we recommend incorporating into your analysis:

- select() extracts columns and returns a tibble.
- arrange() changes the ordering of the rows.
- filter() picks cases based on their values.
- mutate() adds new variables that are functions of existing variables.
- rename() easily changes the name of a column(s)
- summarise() reduces multiple values down to a single summary.
- pull() extracts a single column as a vector.
- _join() group of functions that merge two data frames together, includes (inner_join(), left_join(), right_join(), and full_join()).

Figure 7: "Source: https://hbctraining.github.io/Intro-to-R/lessons/tidyverse_data_wrangling.html"

Tidyverse Function Syntax

Table 1: Useful dplyr functions and syntax

Function	Syntax
select()	select(df, var1, var2,)
mutate()	$mutate(df, new_var = old_var + 5)$
filter()	filter(df, var1 == value)
rename()	$rename(df, new_name = old_name)$
summarise()	$summarise(df,mean_var1 = mean(var1))$
if_else()	<pre>if_else(condition, true, false)</pre>

Transform and explore

```
## # A tibble: 1 x 5
## min_age mean_age median_age mean_perc_days_drink median_perc_days_drink
## <dbl> <dbl> <dbl> <dbl> <dbl> = 17.7 19.8 13.3 6.58
```

The Pipe

► The pipe operate %>% takes output from on function and 'pipes' it into another function

```
x <- paste("A", "String")
print(x)

print(paste("A", "String"))

paste("A", "String") %>%
    print(.)
```

Use the pipe operator to combine all of these into one 'dplyr chain'

► The period denotes where the previous output should be the argument in the new function

```
## # A tibble: 1 x 5

## min_age mean_age median_age mean_perc_days_drink median_perc_days_drink

## <dbl> <dbl> <dbl> <dbl> <dbl> 17.7 19.8 13.3 6.58
```

More Summary Statistics

```
# now let's look at summary stats of heavy drinkers us non heavy drinkers:

# demographic characteristics of heavy drinkers us non-heavy drinkers
summ_stats_heavy <- mlda %>%
filter(drinks_alcohol ==1) %>%
group_by(heavy_drinker) %>%
summarise(min_age = min(age_years),
mean_age = mean(age_years),
median_age = median(age_years),
mean_student = mean(student),
n = n())
summ_stats_heavy
```

```
## # A tibble: 2 x 6
    heavy_drinker min_age mean_age median_age mean_student
##
           <dbl>
                  <db1>
                                   <db1>
                                              <dbl> <int>
                        <dbl>
## 1
                   17.8
                          24.7
                               24.9
                                              0.0874 37322
## 2
                   17.7
                          25.3
                                    25.7
                                             0.0425
                                                      964
```

Try yourself

- Create a variable that denotes someone as an underage drinker (they are less than 21 years old and drink alcohol)
- ► Use the skeleton code below to calculate the share of underage drinkers that are a student, married, or male

```
summ_stats2 <- mutate(mlda, underage_drinker = if_else()) %>%
filter() %>%
summarise()
```

Export as a table

- use kable to export as a table
- can make aesthetic changes to the table, add labels, etc.

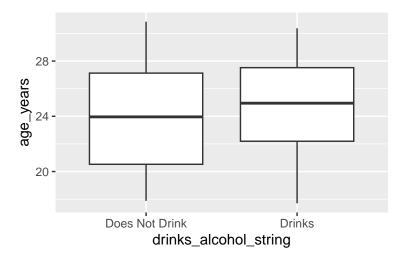
library(kableExtra)

Table 2: Drinking Patterns among Underage Drinkers

Minimum Age	Mean Age	Median Age	Avg Percent Days Drank	Median Percent Days Drank
17.7	19.7	19.8	13.3	6.6

Visual: age distribution of drinkers vs non-drinkers?

```
viz <- mlda %>%
mutate(drinks_alcohol_string = if_else(drinks_alcohol == 1, "Drinks", "Does Not Drink")) %>%
ggplot(aes(x=drinks_alcohol_string, y=age_years)) +
geom_boxplot()
viz
```



Run a Model

- ► Use a linear probability model to predict the effect of age in years on the likelihood of drinking alcohol
- we can use lm()

```
# general form

lpm1 <- lm(formula = y - x_1 + x_2 + ... x_n, data = df)
```

v is the dependent variable and the x variables are the independent variables

Model our data

```
# lpm of years in age on likelihood of drinking alcohol
lpm1 <- lm(formula = drinks_alcohol ~ age_years, data = mlda )</pre>
summary(lpm1)
##
## Call:
## lm(formula = drinks alcohol ~ age years, data = mlda)
##
## Residuals:
##
      Min
              10 Median
                                      Max
## -0.7365 -0.5766 0.3128 0.3852 0.4922
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 0.1999271 0.0140917 14.19 <2e-16 ***
## age years 0.0173927 0.0005712 30.45 <2e-16 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4805 on 61261 degrees of freedom
## Multiple R-squared: 0.01491, Adjusted R-squared: 0.01489
## F-statistic: 927.3 on 1 and 61261 DF, p-value: < 2.2e-16
```

Exercise

Now, add an age squared and age cubed variable to your dataset and rerun the model

```
mlda2 <- mlda %>%
mutate()
lpm2 <- lm(formula = , data = )
```

▶ Then add additional demographic controls to your model and rerun the model

```
lpm3 <- lm(formula = drinks_alcohol ~ , data =)</pre>
```

inspect your results

```
summary(1pm2)
summary(1pm3)
```

Use Stargazer to output your results

Note:

 Stargazer is an R package that creates LATEX code, HTML code and ASCII text for well-formatted regression tables, with multiple models side-by-side, and summary stats, etc.; makes changing models and updating tables very easy

Table 3:

	Dependent variable:				
	Probabilty of Drinking				
	(1)	(2)			
age_years	0.017***	1.697***			
	(0.001)	(0.104)			
age_sq		-0.065***			
		(0.004)			
age_cu		0.001***			
		(0.0001)			
Constant	0.200***	-13.996***			
	(0.014)	(0.817)			
Observations	61,263	61,263			
R^2	0.015	0.031			
Adjusted R ²	0.015	0.031			
Residual Std. Error	0.481 (df = 61261)	0.476 (df = 61259)			
F Statistic	927.281*** (df = 1; 61261)	661.294*** (df = 3; 61259)			

*p<0.1; **p<0.05; ***p<0.01

Output model results

Table 4:

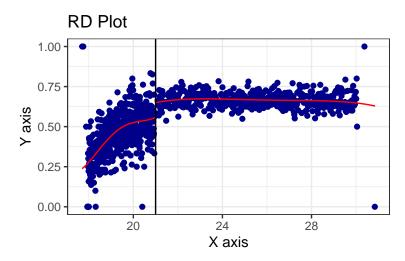
	Dependent variable: Probabilty of Drinking with Controls				
	(1)	(2)	(3)		
age_years	0.017***	1.697***	1.710***		
-	(0.001)	(0.104)	(0.102)		
age_sq		-0.065***	-0.066***		
· -		(0.004)	(0.004)		
age_cu		0.001***	0.001***		
		(0.0001)	(0.0001)		
student			-0.025***		
			(0.007)		
male			0.179***		
			(0.004)		
Constant	0.200***	-13.996***	-14.167***		
	(0.014)	(0.817)	(0.803)		
Observations	61,263	61,263	61,263		
R^2	0.015	0.031	0.065		
Adjusted R ²	0.015	0.031	0.065		
Residual Std. Error	0.481 (df = 61261)	0.476 (df = 61259)	0.468 (df = 61257)		
F Statistic	927.281*** (df = 1; 61261)	661.294*** (df = 3; 61259)	855.417*** (df = 5; 61257)		

Note: *p<0.1; **p<0.05; ***p<0.01

Visualize the discontinuity

```
library(rdrobust)
rdplot(mlda$drinks_alcohol,mlda$age_years,c=21)
```

[1] "Mass points detected in the running variable."





Today's workshop

- We covered a lot!
 - reproducible research practices
 - R Markdown
 - the tidyverse
 - data transformations
 - regression
 - data viz
- Future events
 - Workshop from Angela (Maps!)
 - lightning talks
 - speaker series
- Email us with ideas
 - kk9870@princeton.edu
 - al49@princeton.edu

Citations

Citations

Russo, Francesco & Righelli, Dario & Angelini, Claudia. (2016). Advantages and Limits in the Adoption of Reproducible Research and R-Tools for the Analysis of Omic Data. Lecture Notes in Computer Science. 9874. 245-258. 10.1007/978-3-319-44332-4 19.