Introduction to R Programming GSND 5340Q, BMDA

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R Tutorials (GitHub and YouTube)

Dr. Johnson will provide an online R tutorial on GitHub: https://github.com/wevanjohnson/2024_04_R_tutorial

R Tutorials (GitHub and YouTube)

Please complete Lectures 1-6 ASAP (required). In addition, plan to complete Lectures 7-12 by the end of the course.

Lecture	Topics
Lecture 1	Installing R, RStudio, and R packages
Lecture 2	Introduction to R/RStudio
Lecture 3	R basics, Part 1
Lecture 4	R basics, Part 2
Lecture 5	R basics, Part 3
Lecture 6	Programming Basics
Lecture 7	R Markdown
Lecture 8	Input/output data, Data structures
Lecture 9	The tidyverse
Lecture 10	Visualization with ggplot2, Part 1
Lecture 11	Visualization with ggplot2, Part 2
Lecture 12	Visualization with ggplot2, Part 3
Lecture 13	Creating R Packages
Lecture 14	Shiny Programming, Part 1
Lecture 15	Shiny Programming, Part 2

Section 1

Installation Details

Important installations

You will need to install the following:

Mac Users

- R and R Studio
- Know how to access a terminal (Rstudio or Terminal)
- git (type git --version in the terminal)

Windows Users:

- R and R Studio
- A terminal app (Git Bash, MobaXterm, Putty)
- Git for Windows

R and Rstudio

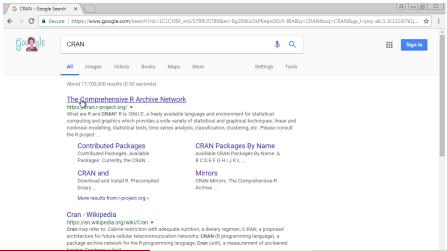
R is a language for statistical computing and graphics. RStudio is an interactive desktop environment (IDE), but it is not R, nor does it include R when you download and install it. Therefore, to use RStudio, we first need to install R.





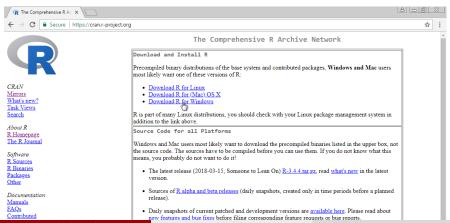
Installing R (Windows and Mac)

You can download R from the Comprehensive R Archive Network (CRAN) 1 . Search for CRAN on your browser:



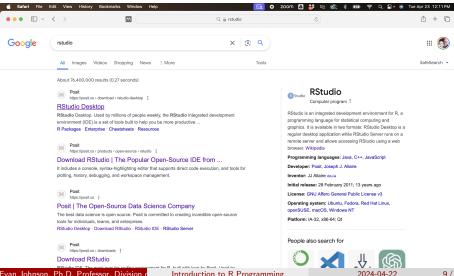
Installing R (Windows and Mac)

Once on the CRAN page, select the version for your operating system: Linux, Mac OS X, or Windows. Here we show screenshots for Windows, but the process is similar for the other platforms. When they differ, we will also show screenshots for Mac OS X.



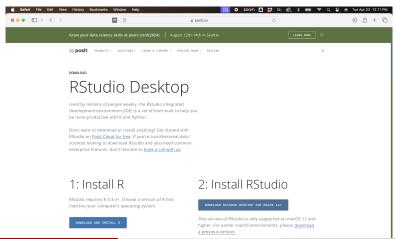
Installing RStudio (Windows and Mac)

To install RStudio, start by searching for "RStudio" on your browser:



Installing RStudio (Mac)

You should find the Posit/RStudio website as shown above. Once there, click on "Download RStudio Desktop for Mac OS 12+" below the 2: Install RStudio header.



More on R and Rstudio

See more detailed instructions at in Lecture 1 at: $https://github.com/wevanjohnson/2024_04_R_tutorial$

Accessing the terminal and installing Git

For Wednesday (4/24), you will also need to install a terminal and install Git (and get a GitHub account).

Detailed instructions: https://rafalab.github.io/dsbook/accessing-the-terminal-and-installing-git.html

Why R?

R is not a programming language for software development like C or Java. It was created by statisticians as an environment for data analysis. A history of R is summarized here: A Brief History of S.

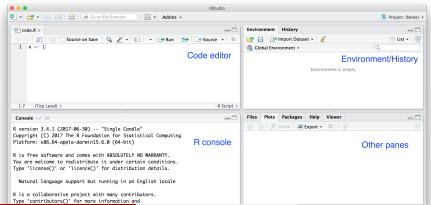




The **interactivity** of R (more later), is an indispensable feature in data science because, as you will learn, the ability to quickly explore data is a necessity for success in this field.

RStudio

One of the great advantages of R over point-and-click analysis software is that you can save your work as scripts. You can edit and save these scripts using a text editor. We will use the interactive *Integrated Development Environment* (IDE) RStudio.



Objects

Suppose we are asked to solve the quadratic equation $x^2 + x - 1 = 0$, we can define:

```
a <- 1
b <- 1
c <- -1
```

which stores the values for later use. We use <- to assign values to the variables. We can also assign values using = instead of <-, but we recommend against using = to avoid confusion.

Objects

To see the value stored in a variable, we simply ask R to evaluate a and it shows the stored value:

а

```
## [1] 1
```

A more explicit way to ask R to show us the value stored in a is using print like this:

```
print(a)
```

```
## [1] 1
```

The Workspace

Now since these values are saved in variables, to obtain a solution to our equation, we use the quadratic formula:

```
(-b + sqrt(b^2 - 4*a*c)) / (2*a)

## [1] 0.618034

(-b - sqrt(b^2 - 4*a*c)) / (2*a)

## [1] -1.618034
```

Scripts

To solve another equation such as $3x^2 + 2x - 1$, we can copy and paste the code above and then redefine the variables and recompute the solution:

```
a <- 3
b <- 2
c <- -5
(-b + sqrt(b^2 - 4*a*c)) / (2*a)
(-b - sqrt(b^2 - 4*a*c)) / (2*a)
```

By creating and saving a script with the code above, we would not need to retype everything each time and, instead, simply change the variable names. Try writing the script above into an editor and notice how easy it is to change the variables and receive an answer.

Functions

Once you define variables, the data analysis process can usually be described as a series of **functions** applied to the data. R includes several predefined functions and most of the analysis pipelines we construct make extensive use of these.

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Functions

Most functions require one or more **arguments**. Below is an example of how we assign an object to the argument of the function log. Remember that we earlier defined a to be 1:

```
log(8)
## [1] 2.079442
log(a)
```

[1] 0

Functions

You can change the default values by simply assigning another object:

$$log(8, base = 2)$$

[1] 3

Note that we have not been specifying the argument x as such:

$$log(x = 8, base = 2)$$

[1] 3

Installing R packages

The functionality provided by a fresh install of R is only a small fraction of what is possible. In fact, we refer to what you get after your first install as **base R**. The extra functionality comes from add-ons available from developers.

There are currently hundreds of these available from CRAN and many others shared via other repositories such as GitHub. However, because not everybody needs all available functionality, R instead makes different components available via **packages**.

Installing R packages

R makes it very easy to install packages from within R. For example, to install the **dslabs** package, which we use to share datasets and code related to this book, you would type:

```
install.packages("dslabs")
```

We can install more than one package at once by feeding a character vector to this function:

```
install.packages(c("tidyverse", "dslabs"))
```

Vectors

In R, the most basic objects available to store data are **vectors**. As we have seen, complex datasets can usually be broken down into components that are vectors. For example, in a data frame, each column is a vector. Here we learn more about this important class.

Creating Vectors

We can create vectors using the function c, which stands for **concatenate**. We use c to concatenate entries in the following way:

```
codes <- c(380, 124, 818)
codes
```

```
## [1] 380 124 818
```

Creating Vectors

We can also create character vectors. We use the quotes to denote that the entries are characters rather than variable names.

```
country <- c("italy", "canada", "egypt")</pre>
```

Names

Sometimes it is useful to name the entries of a vector. For example, when defining a vector of country codes, we can use the names to connect the two:

```
codes <- c("italy" = 380, "canada" = 124, "egypt" = 818)
codes</pre>
```

```
## italy canada egypt
## 380 124 818
```

Subsetting elements

The elements of vectors can be obtained using the following:

```
codes[1]
  italy
     380
##
codes["italy"]
  italy
     380
##
codes[2:3]
## canada
           egypt
```

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818

124

##

Data Frames

The most common way of storing a dataset in R is in a **data frame**, which is a combination of several (columns of) vectors.

Data frames are particularly useful for datasets because we can combine different data types into one object.

Data Frames

A large proportion of data analysis challenges start with data stored in a data frame. For example, we stored the data for our motivating example in a data frame. You can access this dataset by loading the **dslabs** library and loading the murders dataset using the data function:

```
library(dslabs)
data(murders)
```

Data Frames

We can show the first six lines using the function head:

```
head(murders)
```

```
##
          state abb region population total
                              4779736
## 1
       Alabama
                AT.
                    South
                                        135
                              710231
                                         19
## 2
        Alaska AK
                     West
                             6392017
                                       232
## 3
       Arizona AZ
                     West
                AR
                             2915918
## 4
      Arkansas
                    South
                                        93
## 5 California
                CA West 37253956
                                       1257
                             5029196
## 6
      Colorado
                CO
                     West
                                         65
```

The Accessor: \$

For our analysis, we will need to access the different variables represented by columns included in this data frame. To do this, we use the accessor operator \$ in the following way:

```
murders$population
```

```
##
    [1]
         4779736
                     710231
                              6392017
                                        2915918
                                                37253956
                                                            5029196
                                                                      3574097
                                                                                 89
##
    [9]
           601723
                   19687653
                              9920000
                                        1360301
                                                  1567582 12830632
                                                                      6483802
                                                                                304
   Γ17]
                                                  5773552
##
         2853118
                    4339367
                              4533372
                                        1328361
                                                            6547629
                                                                      9883640
                                                                                530
   [25]
         2967297
                               989415
                                                  2700551
                                                            1316470
                                                                      8791894
##
                    5988927
                                        1826341
                                                                                205
   [33]
        19378102
                    9535483
                               672591
                                      11536504
                                                  3751351
                                                            3831074
                                                                     12702379
                                                                                105
##
   Γ417
                                                                                672
##
         4625364
                     814180
                              6346105 25145561
                                                  2763885
                                                             625741
                                                                      8001024
   [49]
          1852994
                    5686986
                               563626
##
```

Subsetting Columns and Rows

In addtion the columns and rows of a data frame can be subsetted using the following syntax:

```
murders[1,1]
## [1] "Alabama"
murders[1,]
##
       state abb region population total
              AI. South
                            4779736
## 1 Alabama
                                       135
murders[,4]
    [1]
         4779736
                    710231
                            6392017
                                      2915918 37253956
                                                         5029196
                                                                  3574097
                                                                             89
##
```

```
[9]
##
                             9920000
                                       1360301
                                                                    6483802
                                                                              304
   [17]
         2853118
                   4339367
                             4533372
                                       1328361
                                                 5773552
                                                          6547629
                                                                    9883640
                                                                              530
##
   Γ251
         2967297
                   5988927
                              989415
                                       1826341
                                                 2700551
                                                          1316470
                                                                    8791894
                                                                              205
##
   [33]
        19378102
                   9535483
                              672591 11536504
                                                3751351
                                                          3831074
                                                                   12702379
                                                                              105
##
   [41]
         4625364
                    814180
                             6346105 25145561
                                                 2763885
                                                           625741
                                                                    8001024
                                                                              672
```

1852994

[49]

601723

19687653

5686986

563626

1567582 12830632

Programming Basics

By coding in R, we can efficiently perform exploratory data analysis, build data analysis pipelines, and prepare data visualization to communicate results. However, R is not just a data analysis environment but a programming language.

You should also understdn the following three key programming concepts: **conditional expressions**, **for-loops**, and **functions**. These are not just key building blocks for advanced programming, but are sometimes useful during data analysis.

R markdown

R markdown is a format for **literate programming** documents. Literate programming weaves instructions, documentation, equations, and detailed comments among executable code.

It is based on **markdown**, a markup language that is widely used to generate html pages. You can learn more about markdown with the following tutorial: **click here**

R markdown

You can start an R markdown document in RStudio by clicking on **File**, **New File**, then **R markdown**. You will then be asked for a title and author.

Once you gain experience with R markdown, you will be able to do this without the template and can simply start from a blank template.

As a convention, we use the .Rmd suffix for R markdown files

Compiling the document using knitR

With R markdown, you need to **compile** the document into the final report. We use the knitR package to to do this. The specific function used to compile is the knit function, which takes a filename as input.

RStudio provides a Knit button that makes it easy and convenient to compile the document.

The Header

At the top of the document is the R markdown header:

```
title: "Nanostring Analysis"
author: "Evan Johnson"
date: "12/5/2019"
output: html_document:
```

R markdown reports can be to be in HTML, PDF, Microsoft Word, or presentation formats. By changing the output to, for example pdf_document or word_document, we can control the type of output that is produced.

The Header

Other output options include code folding, themes, etc.:

```
title: "Nanostring Analysis"
author: "Evan Johnson"
date: "12/5/2019"
output:
  html document:
    code folding: hide
    toc: true
    toc float: true
    theme: "flatly"
editor_options:
  chunk_output_type: console
```

In various places in the document, we see something like this:

```
\``{r}
summary(pressure)
\``
```

These are the **code chunks**. When you compile the document, the R code inside the chunk, in this case summary(pressure), will be evaluated and the result included in that position in the final document.

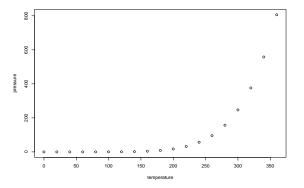
Its a good habit to adding a label to the R code chunks. This will be very useful when debugging, among other situations. You do this by adding a descriptive word like this:

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```
```{r pressure-summary}
summary(pressure)
```

We can also add plots to our report:

```
```{r, out.width = '60%', fig.align = 'center'}
plot(pressure)
```



The options out.width and fig.align adjust the figrue size and loction.

By default, the code will show up as well. To avoid having the code show up, you can use an argument. To avoid this, you can use the argument echo=FALSE. For example:

```
```{r, echo=FALSE}
summary(pressure)
```
```

If you want to include the code in the document, but not run the code and/or include results, you can use the argument eval=FALSE:

```
"\{r, eval=FALSE}

# You can install the tidyverse using:
install.packages(tidyverse)
```

R Code Chunks (global knitr options)

One of the R chunks may contain a complex looking call:

```
```{r setup, include=FALSE}
knitr::opts_chunk$set(echo = TRUE)
```

The include=FALSE option in the chunk call will tell the system to run the R code, but not include it in the html/pdf/word report.

The R code in the chunk sets the knitr default to echo=TRUE in the call for any R chunks following this one.

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### Other Code Chunks

knitr can execute code in many languages besides R:

```
x = [2, 1, 3]
print(x[0], "Hello Python!")
```

```
2 Hello Python!
```

Some of the available language engines include: Python, SQL, Bash, Rcpp, Stan, JavaScript, and CSS.

# LATEX Equations

One exciting feature about R markdown is that it allows you to include LATEX equations to be rendered in html, pdf, or Word.

For example, to show the probability under the normal curve in the interval (a, b), you can use the LATEX code:

```
\mbox{Pr}(a < x < b) =
\int_a^b \frac{1}{\sqrt{2\pi}\sigma}
e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma} \right)^2} \, dx</pre>
```

This will be rendered by knitr as:

$$\Pr(a < x < b) = \int_a^b \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

# Other Options (tabsets)

One very useful tool in Markdown is to add tabs to the html document:

```
This is a header {.tabset}
This is Tab 1
This is Tab 2
This is a new header (not in the tabs)
```

The {.tabset} option will make all lower level subheadings tabs in the html. The next same or higher level header will break out of the tabset.

# Other Options (tabsets)

**Pro tip:** tabs can be generated dynamically:

```
My Header {.tabset}
```{r, results = 'asis'}
n <- 10
for (i in 1:n){
   cat("###" , "Tab", i, "\n")
    print(i)
   cat("\n\n")
}</pre>
```

Other Options

We will explore other R markdown options: headers, tabsets, and LATEX equations in class and in your Exercises.

<u>Note:</u> From now on, all R-based Exercises should preferably be done using an R markdown document. Your document should include headers, descriptive text, equations, R code, and plots/figures!

More on R markdown

There is a lot more you can do with R markdown. We highly recommend you continue learning as you gain more experience writing reports in R. There are many free resources on the internet including:

- RStudio's tutorial: https://rmarkdown.rstudio.com
- The cheat sheet: https://www.rstudio.com/wp-content/uploads/2015/02/rmarkdown-cheatsheet.pdf
- The knitR book: https://yihui.name/knitr/

Exercises

R markdown is a powerful tool for literate prgramming. To gain more practice, recreate the .Rmd file for the example file:

"Rmarkdown_example.html."

Session info

```
sessionInfo()
## R version 4.3.2 (2023-10-31)
## Platform: aarch64-apple-darwin20 (64-bit)
## Running under: macOS Sonoma 14.2.1
##
## Matrix products: default
         /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/lib/libRlapack.dylib; LAPACK version 3
##
## locale:
## [1] en US.UTF-8/en US.UTF-8/en US.UTF-8/C/en US.UTF-8/en US.UTF-8
##
## time zone: America/Denver
## tzcode source: internal
##
## attached base packages:
                                               datasets methods
## [1] stats
                graphics grDevices utils
                                                                   base
##
## other attached packages:
## [1] dslabs_0.7.6
##
## loaded via a namespace (and not attached):
   [1] digest_0.6.35
                          fastmap_1.1.1
                                            xfun_0.43
                                                              Matrix_1.6-5
  [5] lattice_0.22-5
                          reticulate_1.34.0 knitr_1.45
                                                              htmltools_0.5.8
  [9] png_0.1-8
                          rmarkdown 2.26
                                            cli 3.6.2
                                                              grid 4.3.2
##
## [13] compiler_4.3.2
                          rprojroot_2.0.4 here_1.0.1
                                                              rstudioapi_0.16.0
```

[17] tools_4.3.2

[21] isonlite 1.8.8

yam1_2.3.8

Rcpp_1.0.12

evaluate_0.23

rlang 1.1.3