Module 1: Basic programming in R

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Methods and computing camp

This summer we will together learn and review materials of statistical computing and methods.

What do we do during lectures?

Materials will be available at course website. Lecture notes are created by Rmarkdown.

 If you have questions, feel free to interrupt or send a message in the chat.

We will cover 10 modules of statistical methods and computing.

Each module takes ~1 hour.

What contents we will cover?

| Module | Topics | References |
|--------|---|----------------|
| 1 | R, Rstudio, and Rmarkdown | - |
| | Basic data types and structures | |
| 2 | Reporting, data wrangling and graphing (I) | - |
| | LaTeX, tidyverse | |
| 3 | Reporting, data wrangling and graphing (II) | - |
| | Elementary data analysis | |
| | ggplot and R style guide | |
| 4 | Probability distributions | AoS Chp 1-5 |
| | Statistical inference (I) | AoS Chp 6 |
| | Fundamental concepts in inference | |
| 5 | Statistical inference (II) | C&B Chp 6.3, 7 |
| | Maximum likelihood estimation | AoS Chp 3-4 |
| 6 | Statistical inference (III) | AoS Chp 8 |
| | Hypothesis testing | C&B Chap 8 |
| 7 | Statistical models (I) | AoS Chp 13 |
| | Linear regression models | C&B Chp 11 |
| 8 | Statistical models (II) | C&B Chp 12 |
| | Generalized linear models | AoS Chp 13 |
| 9 | Simulation and parallel computing | C&B Chap 10 |
| | | AoS Chp 24 |
| 10 | Bootstrap | AoS Chp 5 |
| | | |

What contents we will not cover?

- Nonparametric inference (STA3000)
- Bayesian inference (STA3000, STA2201)
- Computing of Bayesian inference (STA2201)
- Shiny and blogdown in R

Exercises!

- Available before each module.
- Exercises can be difficult.
- Group discussion is recommended.
- Solutions are released during the last 10-20 mins.

Module 1: Basic programming in R

We will review R, Rstudio, and Syntax of R together.

- Rstudio (Knit)
- Basic data types
- Basic data structures
- Functions
- For loops

Useful resources:

- Tidyverse style guide
- The R Inferno

Introduction to R and Rstudio

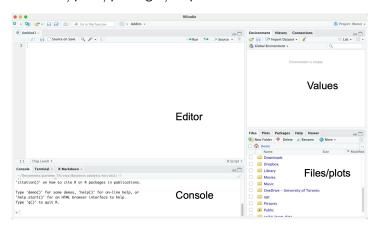
 $\ensuremath{\mathsf{R}}$ is a free statistical software. We use $\ensuremath{\mathsf{R}}$ frequently/intensively during our study.

First please download R and its IDE Rstudio (if you haven't).

- https://www.r-project.org/
- https://www.rstudio.com/

Studio

- Editor: edit or save the file.
- Console: outputs.
- Values: store values of assigned.
- Files/plots/packages/help.



How to set working directory?

Working directory is important since you might want to read and import data from other files. It is recommended to put these files under the same directory with your scripts.

- Method 1: Session -> set working directory.
- Method 2: Files -> Navigate to your directory.
- Method 3: setwd().

Alternatives: Set R.project.

How to install packages?

Common packages in statistical analysis with R:

- tidyverse/dpylr
- ggplot
- kableExtra or gridExtra
- glm

Several options to install packages:

- Method 1: Tools -> install packages
- Method 2: Packages window.
- Method 3: install.packages().

Ready with your Rstudio?

Let's code!

Vector

Can contain numerical, string, or Boolean value.

Store data = make an assignment. c() is for component.

```
v <- c(TRUE, FALSE, TRUE, TRUE, FALSE)
v <- c("python", "mathlab", "R")
v <- c(6, 5, 4, 3, 2, 1)</pre>
```

Are you familiar with the output of these?

```
which(v == 3)
## [1] 4
v[2]
v[-2]
v[2:3]
v[v < 4]
which(v == 3)</pre>
```

Matrix

```
mymat \leftarrow matrix(c(1:10), nrow = 2, ncol = 5,
                 byrow = TRUE)
mymat[2, ]
## [1] 6 7 8 9 10
mymat \leftarrow matrix(c(1:10), nrow = 2, ncol = 5,
                 byrow = TRUE)
mymat
mymat[1, 5]
mymat[2,]
mymat[c(1:2), c(1:2)]
```

Data frame

```
studentID \leftarrow c(1, 2, 3, 4)
age \leftarrow c(17, 18, 16, 19)
gender <-c("M", "F", "M", "M")
studentData <- data.frame(studentID, age, gender)</pre>
rownames(studentData) <- c("A", "B", "C", "D")
colnames(studentData) <- c("ID", "age", "gender")</pre>
studentData
## ID age gender
## A 1 17 M
## B 2 18 F
## C 3 16 M
## D 4 19 M
studentID \leftarrow c(1, 2, 3, 4)
age \leftarrow c(17, 18, 16, 19)
gender <-c("M", "F", "M", "M")
```

Linear Regression

$$y_i = \beta^T x_i + \epsilon_i$$

$$x \leftarrow c(1:5)$$

$$eps \leftarrow rnorm(5)$$

$$y \leftarrow 2*x + eps$$

$$mod \leftarrow lm(y \sim x)$$

Linear Regression

summary(mod)

```
##
## Call:
## lm(formula = v \sim x)
##
## Residuals:
##
## -0.9224 -0.7459 2.2121 1.5028 -2.0467
##
## Coefficients:
##
             Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.9669 2.1620 -0.447 0.6850
             2.4125 0.6519 3.701 0.0343 *
## x
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.061 on 3 degrees of freedom
## Multiple R-squared: 0.8203, Adjusted R-squared: 0.7604
## F-statistic: 13.7 on 1 and 3 DF, p-value: 0.03426
```

```
List
g <- "My List"
h \leftarrow c(2, 3, 5, 7)
j <- matrix(1:10, nrow = 5, byrow = FALSE)</pre>
k <- c("one", "two", "three")</pre>
mylist <- list(title = g, ages = h, j, k)
mylist[[2]]
## [1] 2 3 5 7
```

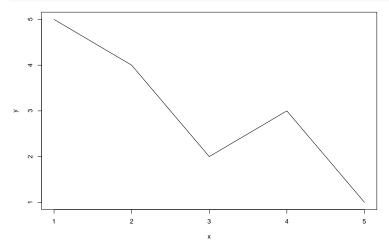
```
## [1] 2 3 5 7

g <- "My List"
h <- c(2, 3, 5, 7)
j <- matrix(1:10, nrow = 5, byrow = FALSE)
k <- c("one", "two", "three")
mylist <- list(title = g, ages = h, j, k)

mylist$ages
mylist[[1]]</pre>
```

Plotting

```
y <- c(5, 4, 2, 3, 1)
x <- c(1, 2, 3, 4, 5)
plot(x, y, type = "l")
```



Other

Important for stats research.

floor(v), ceiling(v)

```
• round(v, 2)
  • rnorm(), rexp(), rbinom(), etc generate random variable
  Function
rnorm(5, mean = 0, sd = 1) # 5 random numbers from N(0, 1)
## [1] -1.31882144 -0.29859589 -0.07072269 -0.76864930 2.24406137
round(2.333333, 2)
## [1] 2.33
func <- function(x1, x2) {</pre>
   y \leftarrow 3 * x1^2 + 3 * x2^2 - 2* x1^2 *x2
   return(y)
func(1, 2)
```

[1] 11

If else

What is the output?

```
p <- 3
if (p <= 2) {
  print("p <= 2!")
} else {
  print("p > 2!")
}
```

for loop

```
v \leftarrow c(1, 2, 4, 3)
w \leftarrow c(0, 0, 0, 0)
t <- 0
for (i in v) {
  t <- t + 1
  w[t] <- w[t] + i
  print(w)
}
## [1] 1 0 0 0
## [1] 1 2 0 0
## [1] 1 2 4 0
## [1] 1 2 4 3
W
## [1] 1 2 4 3
```

while loop

```
i <- 1
while (i <= 10) {
  print(i)
  i <- i + 1
}
## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
## [1] 6
```

[1] 7 ## [1] 8 ## [1] 9 ## [1] 10

next

```
alphabet <- LETTERS[1:6]</pre>
for (i in alphabet) {
  if(i == 'D') {
    next
 print(i)
   [1] "A"
   [1]
       "B"
```

"C"

"E" [1] "F"

[1]

[1]

break

```
alphabet <- LETTERS[7:12]</pre>
for (i in alphabet) {
  if (i == 'K') {
    break
  print(i)
## [1]
       "G"
   [1]
```

[1] "I" ## [1] "J"

apply

It can be applied on matrix, vector, data frame, and loop through row or column (defined by the second input - MARGIN).

```
Syntax: apply(X, MARGIN, FUN, ...)

f <- function(x) {
   ts <- 2*x^2
   return(ts)
}

ii <- matrix(1:4, nrow = 1)
apply(ii, 1, f)

## [,1]
## [1,] 2</pre>
```

```
## [1,] 2
## [2,] 8
## [3,] 18
## [4,] 32
```

Knit in Rstudio

Common types of R files: .R, .Rmd

- Simulations, e.g. for loops, functions, I use .R
- Reporting, analysis, plotting, I use .Rmd

Rmd file can be converted to pdf, html through Knit.

• yaml style.

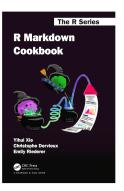
```
title: 'Module 1: Basic programming in R'
date: "04/15/2022"
output:
beamer_presentation
```

More yaml style

```
title: "A summary of xx"
date: "02/14/2022"
output:
  pdf_document:
    toc: true
    number_sections: true
```

Resource

- "R Markdown Cookbook" by Yihui Xie.
- https://bookdown.org/yihui/rmarkdown-cookbook/



Code style

- Google's R Style Guide
- styler software embedded in Rstudio allows you to interactively restyle selected text, files, or entire projects.
- Lintr software embedded in Rstudio performs automated checks to confirm that you conform to the style guide.

Exercise

Available on course website.

- \bullet Matrix and vector operations generate a 100×100 matrix for matrix inverse calculation.
- For loops and plotting understand for loops by plotting a complex polygon.