Reproducible Research in R

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Introduction

Introduction

Who are we?

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Introduction

Who are you?

- 1. Occupation, employer?
- 2. Previous knowledge and experience
 - with reproducible research
 - with R?
 - with other statistical software?
 - with other programming languages?
- 3. Specific interest/motivation for this workshop?

Motivation

Why care about reproducible research?

Agenda

- Conceptual things
- Writing Reproducible R Code
- RMarkdown
- Version Control/git

Reproducible Research

RStudio setup

RStudio setup

- 1. Save the course content to a directory on your machine
- 2. Open RStudio
- 3. Choose File < New Project ...
- 4. Choose Existing Directory
- Browse to the directory on your machine where you saved the course content and select the "R-programming" folder as the Project working directory
- 6. Click Open in new session
- 7. Click Create Project

RStudio setup - optional

- 1. Choose Tools < Global options
- 2. Under General
 - DON'T Restore .RData into workspace at startup
 - NEVER Save workspace to .Rdata on exit:
 - Save the code instead!
 - Use saveRDS() and readRDS() for objects that require a long time to computate
- 3. Further personalize RStudio

Writing Reproducible R Scripts

RMarkdown

Version Control via Git and Github

Good programming practices

"Write code for humans, not for machines!" $\,$

Code Style

Invest time in writing readable R-code.

- It will make collaborations easier
- It will make debugging easier
- It will make your analyses more reproducible

There is a complete *tidyverse* style-guide https://style.tidyverse.org/.

Go easy on your eyes

- with spaces before and after: + / * = <- < == >
- always use <- for assignments
- only use = in function calls
- use indentation (largely automatic in RStudio)
- CamelCaseNames vs snake_case_names
- be consistent!
- wrap long lines at column 70-80 (Rstudio)

White space

```
new_var=(var1*var2/2)-5/(var3+var4)

# versus
new_var <- (var1 * var2 / 2) - 5 / (var3 + var4)</pre>
```

Indentation

```
for(name in names){formula=as.formula(paste0("y~.-",name))
fit<-lm(formula, data=my_data)</pre>
coefs[["name"]]=coef(fit)
print(name)
print(summary(fit))}
 versus
for(name in names){
  formula <- as.formula(paste0("y~.-", name))</pre>
  fit <- lm(formula, data = my_data)
  coefs[["name"]] <- coef(fit)</pre>
  print(name)
  print(summary(fit))
```

Wrap long lines

```
final_results <- data.frame(first_variable =</pre>
sgrt(results$mean_squared_error), second_variable =
paste0(results$condition, results$class, sep = ":"),
third variable = results$bias)
# versus
final_results <- data.frame(</pre>
  first_variable = sqrt(results$mean_squared_error),
  second_variable = paste0(results$condition,
                            results$class, sep = ":"),
  third_variable = results$bias)
```

Go easy on your mind

- use meaningful names: "self-explainable"
- always write the formal arguments in function calls (except the first)
- benefit from autocompletion (<tab>) => embrace longer names
- use TRUE and FALSE not T and F
- comment, comment, comment
 - NOT what (should be clear from the code)
 - but why
 - explain the reasoning, not the code

Use meaningful names

```
V <- myFun(m1_B)
# versus

RMSE_age_gender <- get_RMSE(lm_age_gender)</pre>
```

Programming advice

Use verbs for functions and nouns for other objects.

Write formal arguments

Benefit from auto completion using tab

Comment, comment

```
## Start every Rscript with a comment that explains
   what the code in the script does, why it does
##
   this, and to which project it belongs.
##
   Your future self will be very thankful!
##
## Mention which packages you are using in this Rscript.
## Use sections to separate chunks -----
## Maybe even subsections =================
## Recode variables so that missings are coded as "NA"
dat[dat %in% c(99, 999)] <- NA # missings coded 99 or 999
```

Keep your code slim

Try to limit your package-dependencies.

Only load library() the packages that you absolutely need. If you are only using dplyr, it does not make sense to load the complete tidyverse.

Controversial: when possible, use the :: operator (and consider not loading the package). ckage::<function>

- explicit dependencies
- less name conflicts

Never Attach

Forget about attach()!

Don't use it, unless you completely understand what happens (see ?attach).

Use with(data.frame, expression) instead.

Testing R code

Writing code is error prone. Incorporate tests and checks in your workflow.

- minimal examples
- write tests and checks
- helpful packages: testthat, RUnit, testit, ...

Computing speed can become an issue. Avoid common pitfalls:

- don't grow, but replace
- vectorize where possible
- check the computing speed

?system.time, microbenchmark or profiling tools

Don't grow!

```
system.time({
  new_data <- NULL
  for(row_nr in seq_len(NROW(data))){
    new_data <- cbind(</pre>
      data[row_nr,],
      result = exp(data$x[row_nr]) /
        log(data$z[row_nr]) +
        5 * sqrt(data$y[row_nr]))
})
           system elapsed
     user
     2.20
             0.00
                      2.25
```

Replace!

```
system.time({
  n_rows <- dim(data)[1]</pre>
  data$result <- rep(NA, n_rows)</pre>
  for(row_nr in seq_len(n_rows)){
    data$result[row_nr] <- exp(data$x[row_nr]) /</pre>
      log(data$z[row_nr]) +
      5 * sqrt(data$y[row_nr])
})
            system elapsed
>
     user
                       0.33
     0.29
              0.01
```

Vectorize!

```
system.time({
  data$result <- exp(data$x) / log(data$z) +
    5 * sqrt(data$y)
})

> user system elapsed
> 0 0 0
```

Compare the speed of different implementations using:

microbenchmark::microbenchmark

```
get_mean1 <- function(x){</pre>
  weight <- 1/length(x)</pre>
  out <- 0
  for(i in seq_along(x)){
    out <- out + x[i] * weight
  return(out)
get_mean2 <- function(x){</pre>
  sum(x)/length(x)
```

Compare the speed of different implementations using:

microbenchmark::microbenchmark

```
x < - rnorm(500)
microbenchmark::microbenchmark(
 mean(x), get_mean1(x), get_mean2(x))
> Unit: nanoseconds
               min lq mean median
                                               max neval
>
          expr
                                         uq
       mean(x) 2100 2200 3202 2300 2400
                                                     100
>
                                              88200
>
  get_mean1(x) 13500 13700 43205 13800 14000 2931700 100
  get_mean2(x)
                600 700 10165 700
                                             935400
                                                   100
                                        800
```

Programming advice

Don't worry about speed before it becomes an issue.

Wrap Up

General Advice

- Investing time in learning R pays off
- It's a steady learning curve
- Learn from masters
- Rewrite important code the first attempt is usually not the best approach

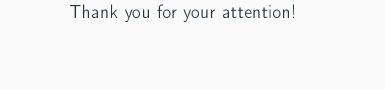
General R Advice

- Document well
- Use a consistent style
- Write functions
- Split long functions in smaller ones
- Write wrappers
- Use Iteration (don't copy paste)
- Use matrix operations and vectorized functions instead of loops
- Use git

Literature Recommendations

R Resources

- Avanced R Ed. 1 (http://adv-r.had.co.nz/)
- Avanced R Ed. 2 (https://adv-r.hadley.nz/)
- R Inferno (https: //www.burns-stat.com/pages/Tutor/R_inferno.pdf)
- R Packages (https://r-pkgs.org/)
- Clean Code (https://mooc.aptikom.or.id/pluginfile. php/1174/mod_resource/content/1/Clean%20Code_%20A% 20Handbook%20of%20Agile%20Software%20C%20-% 20Robert%20C.%20Martin.pdf)



Thank you for your attention!

Questions? Remarks?