



ETC4500/ETC5450 Advanced R programming

Week 7: Reactive programming with targets and renv



Outline



Outline



Regular (imperative) programming

Consider how code is usually evaluated...

a <- 1 b <- 2 x <- a + b x

What is x?

a <- -1 x

What is x now?

Regular (imperative) programming

Predictable programming

All programming we've seen so far evaluates code in sequential order, line by line.

Since \times was not re-evaluated, its value stays the same even when its inputs have changed.

Within a reactive programming paradigm, objects *react* to changes in their inputs and automatically update their value!

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🛕 Disclaimer

Reactive programming is a broad and diverse paradigm, we'll focus only on the basic concepts and how they apply in shiny applications.

We can implement *reactivity* with functions & environments.

library(rlang)
react <- function(e) new_function(alist(), expr(eval(!!enexpr(e))))</pre>

We'll learn how this function works later (metaprogramming).

Reactive programming is also smarter about *'invalidation'*, results are **cached and reused** if the inputs aren't changed.

Reactive programming

How does reactive programming differ?

a <- 1 b <- 2 y <- react(a + b) y()

What is y?

a <- -1 y()

What is y now?

Reactive programming

• (Un)predictable programming?

Reactive programming can be disorienting!

Reactive objects *invalidate* whenever their inputs change, and so its value will be recalculated and stay up-to-date.

Reactive programming

🌢 Your turn!

```
a <- 1
b <- 2
y <- react(a + b)
y()
```

When was a + b evaluated?

How does this differ from ordinary (imperative) code?

Imperative and declarative programming

Imperative programming

- Specific commands are carried out immediately.
- Usually direct and exact instructions.
- e.g. read in data from this file.

Declarative programming

- Specific commands are carried out when needed.
- Expresses higher order goals / constraints.
- e.g. make sure this dataset is up to date every time I see it.

Use cases for reactive programming

Use-less cases

This paradigm is rarely needed or used in R for data analysis.

🍨 Useful cases

Reactive programming is useful for developing user applications (including web apps!).

In R, the shiny package uses reactive programming for writing app interactivity.

Outline



Caching: using rds

```
if (file.exists("results.rds")) {
  res <- readRDS("results.rds")
} else {
  res <- compute_it() # a time-consuming function
    saveRDS(res, "results.rds")
}</pre>
```

Caching: using rds

```
if (file.exists("results.rds")) {
  res <- readRDS("results.rds")
} else {
  res <- compute_it() # a time-consuming function
    saveRDS(res, "results.rds")
}</pre>
```

Equivalently...

```
res <- xfun::cache_rds(
    compute_it(), # a time-consuming function
    file = "results.rds"
)</pre>
```

Caching: using rds

```
compute <- function(...) {
   xfun::cache_rds(rnorm(6), file = "results.rds", ...)
}
compute()</pre>
```

[1] 1.113 -0.163 -0.557 -0.428 -0.444 -0.503
compute()

```
[1] 1.113 -0.163 -0.557 -0.428 -0.444 -0.503
```

compute(rerun = TRUE)

[1] -0.5011 1.0217 0.1034 -1.7602 -0.0269 2.1689
compute()

[1] -0.5011 1.0217 0.1034 -1.7602 -0.0269 2.1689

You often want to prevent downloads of the same data multiple times.

```
download_data <- function(url) {</pre>
  dest folder <- tempdir()</pre>
  sanitized_url <- stringr::str_replace_all(url, "/", "_")</pre>
  dest_file <- file.path(dest_folder, paste0(sanitized_url, ".rds"))</pre>
  if (file.exists(dest_file)) {
    data <- readRDS(dest file)</pre>
  } else {
    data <- read_tsv(url, show_col_types = FALSE)</pre>
    saveRDS(data, dest file)
  data
bulldozers <- download data("https://robihyndman.com/data/Bulldozers.csv")</pre>
```

Caching: memoise

Caching stores results of computations so they can be reused.

```
library(memoise)
sq <- function(x) {
    print("Computing square of 'x'")
    x**2
}
memo_sq <- memoise(sq)
memo_sq(2)</pre>
```

[1] "Computing square of 'x'"

[1] 4

memo_sq(2)

[1] 4

Caching: Rmarkdown

```
'``{r import-data, cache=TRUE}
d <- read.csv('my-precious.csv')
'``
{r analysis, dependson='import-data', cache=TRUE}
summary(d)
```</pre>
```

- Requires explicit dependencies or changes not detected.
- Changes to functions or packages not detected.
- Good practice to frequently clear cache to avoid problems.
- targets is a better solution

# **Caching: Quarto**

```
```{r}
#| label: import-data
  cache: true
#1
d <- read.csv('my-precious.csv')</pre>
- - -
···{r}
#| label: analysis
#| dependson: import-data
  cache: true
# I
summary(d)
- - -
```

Same problems as Rmarkdowntargets is a better solution

Outline



targets: reproducible computation at scale



- Supports a clean, modular, function-oriented programming style.
- Learns how your pipeline fits together.
- Runs only the necessary computation.
- Abstracts files as R objects.
- Similar to Makefiles, but with R functions.

Some images from https://wlandau.github.io/targets-tutorial



Interconnected tasks



Interconnected tasks



Dilemma: short runtimes or reproducible results?



Let a pipeline tool do the work



Save time while ensuring computational reproducibility.
 Automatically skip tasks that are already up to date.

Typical project structure

no_targets.R

library(tidyverse)
library(fable)
source("R/functions.R")
my_data <- read_csv("data/my_data.csv")
my_model <- model_function(my_data)</pre>

Typical project structure

no_targets.R

library(tidyverse)
library(fable)
source("R/functions.R")
my_data <- read_csv("data/my_data.csv")
my_model <- model_function(my_data)</pre>

_targets.R

```
library(targets)
tar_option_set(packages = c("tidyverse", "fable"))
tar_source() # source all files in R folder
list(
   tar_target(my_file, "data/my_data.csv", format = "file"),
   tar_target(my_data, read_csv(my_file)),
   tar_target(my_model, model_function(my_data))
```

Generate _targets.R in working directory

library(targets)
tar_script()

Activity

- Set up a project using targets: tar_script()
- Add targets to generate a plot from the mtcars dataset, and fit a linear regression model.
- Make the project using tar_make()
- Visualize the pipeline using tar_visnetwork()

Useful targets commands

- tar_make() to run the pipeline.
- tar_make(starts_with("fig")) to run only targets
 starting with "fig".
- tar_read(object) to read a target.
- tar_load(object) to load a target.
- tar_load_everything() to load all targets.
- tar_manifest() to list all targets
- tar_visnetwork() to visualize the pipeline.
- tar_destroy() to remove all targets.
- tar_outdated() to list outdated targets.

Errored targets to return NULL so pipeline continues.

tar_option_set(error = "null")

Errored targets to return NULL so pipeline continues.

tar_option_set(error = "null")

See error messages for all targets.

tar_meta(fields = error, complete_only = TRUE)

Errored targets to return NULL so pipeline continues.

tar_option_set(error = "null")

See error messages for all targets.

tar_meta(fields = error, complete_only = TRUE)

See warning messages for all targets.

tar_meta(fields = warnings, complete_only = TRUE)

- Try loading all available targets: tar_load_everything().
 Then run the command of the errored target in the console.
- Pause the pipeline with browser()
- Use the debug option: tar_option_set(debug =
 "target_name")
- Save the workspaces:
 - tar_option_set(workspace_on_error = TRUE)
 - tar_workspaces()
 - tar_workspace(target_name)

Random numbers

Each target runs with its own seed based on its name and the global seed from tar_option_set(seed = ???)
 So running only some targets, or running them in a different order, will not change the results.

Folder structure

- --- .git/
- --- .Rprofile
- -----.Renviron
- renv/
- index.Rmd
- _targets/
- _targets.R
- _targets.yaml
- R/
- _____ functions_data.R
- —— functions_analysis.R
- functions_visualization.R
- data/
 - input_data.csv

_targets.R with quarto

```
library(targets)
library(tarchetypes)
tar_source() # source all files in R folder
tar_option_set(packages = c("tidyverse", "fable"))
list(
    tar_target(my_file, "data/my_data.csv", format = "file"),
    tar_target(my_data, read_csv(my_file)),
    tar_target(my_model, model_function(my_data)),
    tar_quarto(report, "file.qmd", extra_files = "references.bib")
    )
```

Load tarchetypes package for quarto support.
 Add a quarto target.

Replace quarto chunks with tar_read() or tar_load().

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Chunk options

Chunk with regular R code

```
% {r}
# | label: fig-chunklabel
# | fig-caption: My figure
mtcars |>
ggplot(aes(x = mpg, y = wt)) +
geom_point()
```

Chunk options

Chunk with regular R code

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% {r}
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mtcars |>
ggplot(aes(x = mpg, y = wt)) +
geom_point()
```

Chunk with targets

```
% {r}
#| label: fig-chunklabel
#| fig-caption: My figure
tar_read(my_plot)
```

Exercise

Add a quarto document to your targets project that includes the plot and the output from the linear regression model.

Outline



Reproducible environments

To ensure that your code runs the same way on different machines and at different times, you need the computing environment to be the same.

- 1 Operating system
- 2 System components
- 3 R version
- 4 R packages

Solutions for 1–4: Docker, Singularity, containerit, rang

Solutions for 4: packrat, checkpoint, renv

Reproducible environments



- Creates project-specific R environments.
- Uses a package cache so you are not repeatedly installing the same packages in multiple projects.
- Does not ensure R itself, system
 - dependencies or the OS are the same.
- Not a replacement for Docker or

Apptainer.

Reproducible environments



- Can use packages from CRAN, Bioconductor, GitHub, Gitlab, Bitbucket, etc.
- renv::init() to initialize a new project.
- renv::snapshot() to save state of
 project to renv.lock.
- renv::restore() to restore project
 as saved in renv.lock.

renv package



renv package

- renv::install() can install from CRAN, Bioconductor, GitHub, Gitlab, Bitbucket, etc.
- renv uses a package cache so you are not repeatedly installing the same packages in multiple projects.
- renv::update() gets latest versions of all dependencies from wherever they were installed from.
- renv::deactivate(clean = TRUE) will remove the renv environment.



Add renv to your targets project.

Example paper

ISSN: 0140-5482

JOURNAL OF THE OPERATIONAL RESEARCH SOCIETY



Hyndman RJ, Rostami-Tabar B (2024) Forecasting interrupted time series, *Journal of the Operational Research Society*, in press.



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forecasting_interrupted_time_series



