Reproducible science in R

with targets



A. Ginolhac | HPC School | 2020-12-16



Introduction to R

Not the scope of this session from the previous sessions

But an introduction to targets, a Make-like workflow manager for R

• if you are a beginner user, check out this lecture

What is R?

R is shorthand for "GNU R":



- An interactive programming language derived from S (J. Chambers, Bell Lab, 1976)
- Appeared in 1993, created by R. Ihaka and R. Gentleman, University of Auckland, NZ
- · Focus on data analysis and plotting
- R is also shorthand for the ecosystem around this language
 - Book authors
 - Package developers
 - Ordinary useRs

Learning to use R will make you more efficient and facilitate the use of advanced data analysis tools

• if you are an advanced user, interested in programming, check out this lecture

Evaluation in programming

tidyeval



A. Ginolhac | rworkshop | 2020-11-30



About this lecture

Learning objectives

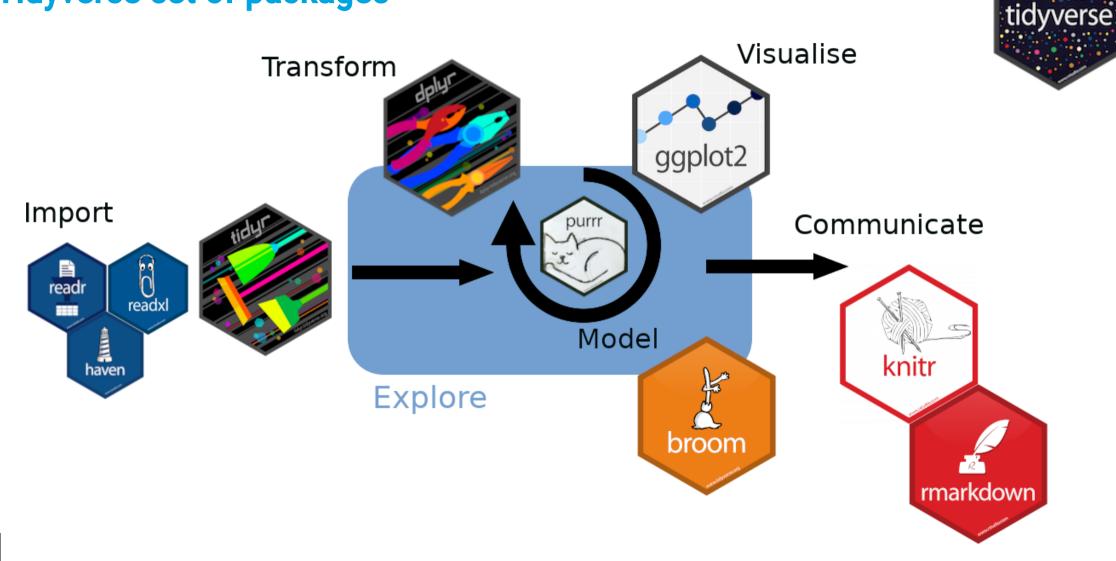
- Why a workflow manager is saving you time and stress
- Understand how it is implemented in targets
 - folder structure
 - define your targets
 - connect targets to create the dependencies
 - check dependencies with visnetwork
 - embrace dynamic branching
 - run only what needs to be executed
 - bundle dependencies in a Rmarkdown document with tar_render()
 - increase reproducibility with the package manager renv
- Practice on a dataset, running > 100 linear models in the <u>tidyverse</u> framework





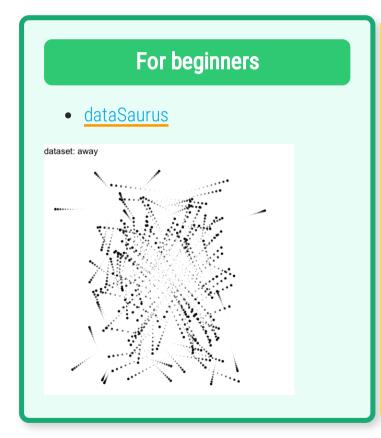


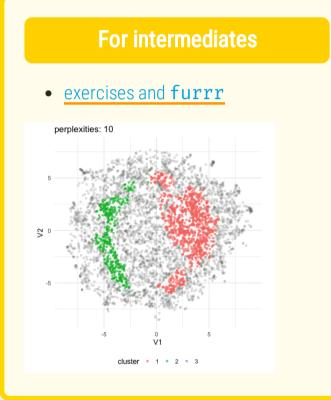
Tidyverse set of packages





Practical sessions









If you wonder about drake

• drake, is the most used workflow manager for R. 1,300 \uparrow on Github

Rationale for rewriting from scratch

- drake is still an excellent choice for pipeline management, but it has permanent user-side limitations.
- targets was created to overcome these limitations and create a smoother user experience.
- Stronger guardrails by design.
- A friendlier, lighter, more transparent data management system.
- Show which functions are up to date.
- More flexible dynamic branching.
- Improved parallel efficiency.
- Designed for custom user-side metaprogramming (tarchetypes/stantargets).
- <u>Debugging!</u> with <u>workspaces</u> (wishing this exists in **snakemake**)

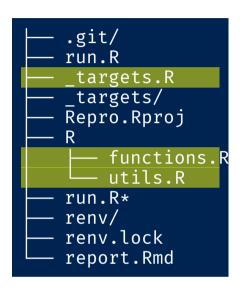
Not on CRAN yet

```
library(remotes)
# Landau is not using 'master'
install_github("wlandau/targets@main")
install_github("wlandau/tarchetypes@main")
```



Source: targets slides W. Landau. NY Meetup hackR

Folder structure



- with renv
- _targets.R is the only mandatory file
- use a R sub-folder, gets closer to a R package
- a run. R file for custom building in RStudio
- Rmd file allows to gather results in a report
- in a RStudio project
- version tracked with git











Main file _targets.R

- add library(tarchetypes) for extrautilities
- nice helpers in options like silencing tidyverse loading
- declare packages that will be loaded by default
- each target can have their own additional packages to load (avoid loading for nothing)
- using bar names of targets, create the dependencies
- pattern = map() creates dynamic branching

Guardrails

- all targets are fresh R processes
- _targets.R must be at the project root



Formats, Patterns, Iterations

Formats

- "rds": default
- "qs": faster binary objects (see qs::qsave())
- "fst": amazing speed for data.frame, (see fst::write_fst())
- "fst_dt": same for data.table objects
- "fst_tbl": same for tibbles objects
- "keras"
- "torch"
- "url"
- "file": automatic hashing
- "aws_rds/file": for AWS S3 cloud

Patterns

- map(): iterate over one or more targets in sequence.
- cross(): iterate over combinations of slices of targets.
- head(): restrict branching to the first few elements.
- tail(): restrict branching to the last few elements.
- sample(): restrict branching to a random subset of elements.

Iterations

- "vector": using vctrs
- "list": less strict than vctrs::vec c()
- "group":handy for dplyr::group_by() tibble

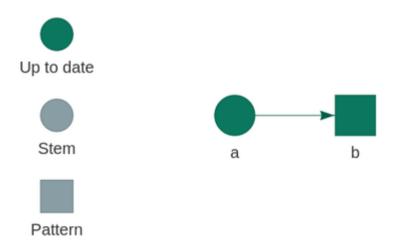
and tar_target reference page



Source: W. Landau, NYhackr

Toy example

re-run with tar_make(): all skipped





Toy example

Add one element to a, ② one more b branch

change **b all** branches re-run



Toy parallelism, using future

<u>future</u> by <u>Henrik Bengtsson</u> is available. Only one worker

```
library(future)
plan(multisession)
tar pipeline(
  tar_target(a, 1:3),
  tar target(b, {
   # fake a long process
   Sys.sleep(5)
   a + 1
  pattern = map(a)
system.time(tar_make_future(workers = 1L))
#> • run target a
#> ● run branch b 0a91b2ed
#> • run branch b 847c5ae2
#> • run branch b 1619f1a0
    user system elapsed
   2.156
           0.284 20.807
```

with 3 workers



```
library(future)
plan(multisession)
tar pipeline(
  tar target(a, 1:3),
  tar target(b, {
    # fake a long process
    Sys.sleep(5)
    a + 1
  pattern = map(a))
system.time(tar make future(workers = 3L))
#> • run target a
  • run branch b_0a91b2ed
  • run branch b 847c5ae2
#> • run branch b 1619f1a0
     user system elapsed
     2.126
             0.164 12.920
```

Still overhead, but individual target can be parallelised (deployment / resources args)

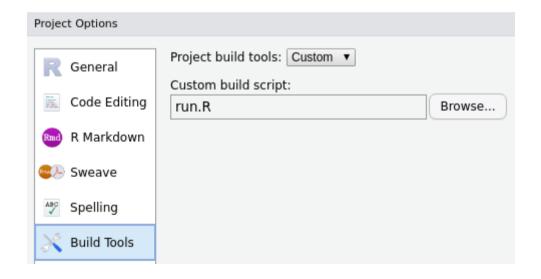


Custom Build in RStudio

run. R is essentially a call to tar_make() as seen below:

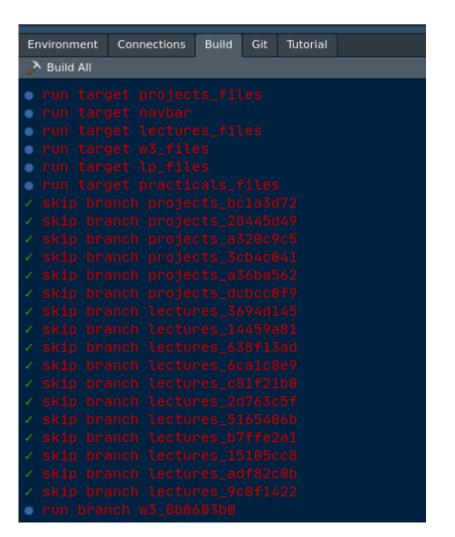
```
#!/usr/bin/env Rscript
targets::tar_make()
```

Then specified in RStudio project building option:





Example (also nice a keyboard shortcut)



A more realistic example

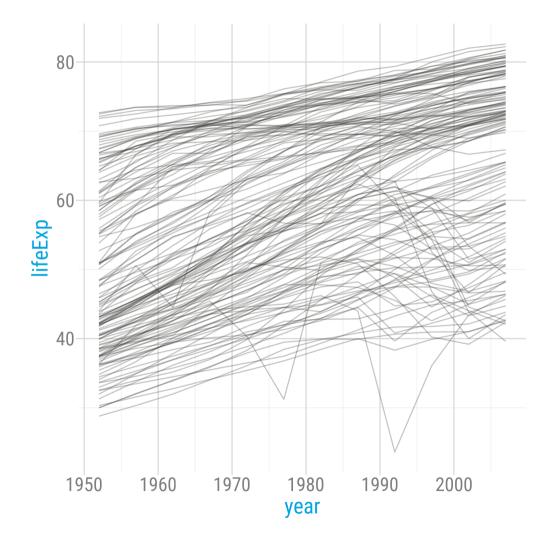
Linear modelling on the gapminder dataset

142 countries, 12 yearly observations of 3 parameters

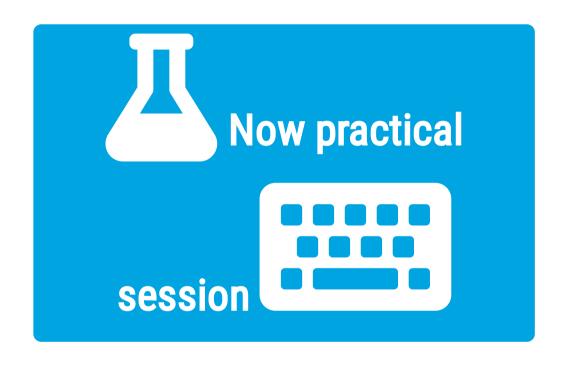
country	continent	year	lifeExp	pop	gdpPercap
Afghanistan	Asia	1952	28.8	8425333	779
Afghanistan	Asia	1957	30.3	9240934	821
Afghanistan	Asia	1962	32	10267083	853

• • •

country	continent	year	lifeExp	pop	gdpPercap
Zimbabwe	Africa	1997	46.8	11404948	792
Zimbabwe	Africa	2002	40	11926563	672
Zimbabwe	Africa	2007	43.5	12311143	470



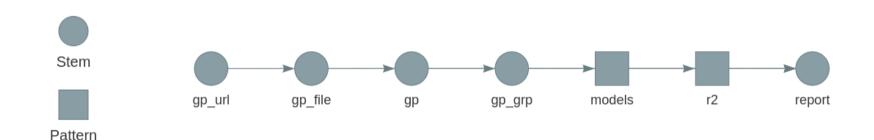


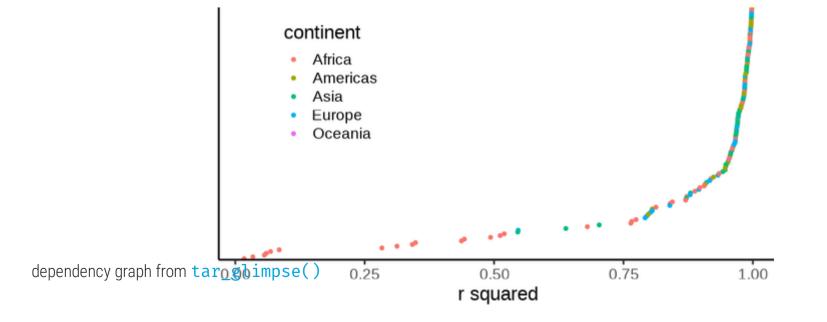




Aim: life expectancy explained by time for each country

- gapminder as tsv URL (Jenny Bryan)
- download the file
- read tsv as tibble
- group by country
- run 142 linear models
- extract 142 r^2
- bundle results in a Rmarkdown
- ullet snake plots the r^2







Step1: set-up and data URL

Set-up

- Create an RStudio project
- install tidyverse, fst, rmarkdown, visNetwork
- install targets, tarchetypes (slide 6)
- run.R and custom build (slide 13)
- Create an empty file R/functions.R
- Create a <u>targets</u>.R file with (copy icon on top right corner)



```
library(targets)
library(tarchetypes)
source("R/functions.R")
options(tidyverse.quiet = TRUE)
# necessary packages to load in isolated R sestar_option_set(packages = c("tidyverse", "fst"
# Define the pipeline
tar_pipeline(
  tar_url(gp_url, "https://raw.githubuserconte))
```

- Build
- check that the target is correct:

```
tar_read(gp_url)
#> [1] "https://raw.githubusercontent.com/jenr
```



Step 2: download and load the file

- download.file() does not return anything.
- we need a filename, so create a wrapper download_file()
 - format is **file**, tsv will be tracked
- load the file with readr::read tsv()
 - format is tibble via fst super fast
- run the pipeline
 - gp_url should be skipped
 - tar_read(gp) should return a tibble

lines for _targets.R

```
tar_file(gp_file, download_file(gp_url, "gapm:
tar_fst_tbl(gp, read_tsv(gp_file, col_types =
```

lines for R/functions.R

```
#' wrap download.file to return the filename
download_file <- function(url, out) {
   download.file(url, out, quiet = TRUE)
   out
}</pre>
```



Step 3: Group by country

```
    dplyr::group_by() by continent and country
    pipe to tar_group() which create extra column
    specify iteration = "group"
```

- don't forget the **comma** between targets.
- run the Build



Step 4: run the models

- add a function to wrap lm
- add the target models
 - pattern is map over the countries
 - o specify iteration = "list"

lines for R/functions.R

```
#' linear model on country, liefExp explained
#' so we get meaningful intercept
ml_lifeExp <- function(.data) {
   gp <- mutate(.data, year1950 = year - 1950)
   lm(lifeExp ~ year1950, data = gp)
}</pre>
```

lines for _targets.R

- run tar_visnetwork(labels = "branches"), 142 expected
- run the Build, should see 142 branches



Step 5: extract r^2 from models

- add a function to extract r^2
- add the target r2
 - pattern is map over the models
 - o specify iteration = "vector"

lines for R/functions.R

```
#' extract r.square from lm object
extract_r2 <- function(model) {
   summary(model)$r.squared
}</pre>
```

lines for _targets.R

```
tar_target(r2,
extract_r2(models),
pattern = map(models),
# now vector is enough
iteration = "vector")
```

- run tar_visnetwork(labels = "branches"), 142 expected
- run the Build, should see 142 branches for r2



Step 6: add the report

- edit the report.Rmd
- add the special target tar_render()

lines for report.Rmd

```
title: "gapminder report"
author: "your name"
date: "2020-12-16"
output: html_document
---
## Gapminder Data
we used the URL `` https://raw.githubusercont
```

```
#### lines for `_targets.R`
   ```r
tar_render(report, "report.Rmd")
```

- run tar\_visnetwork(labels = "branches")
- run the Build



### Before we stop

#### You learned to:

- apprehend targets
- discover the advantages of workflow managers
- Makefile-like approach and project design

#### Acknowledgments 👃 👋



- Willam Laudau obviously
- Eric Koncina early adopter of targets
- Hadley Wickham
- Henrik Bengtsson
- Xie Yihuie and Garrick Aden-Buie for xarigan/xaringanExtra
- Jennyfer Bryan

### Further reading

- targets manual by William Landau
- targets reference by William Landau
- tarchetypes reference by William Landau
- NY Open Statistical Meeup, Dec 2020, targets presentation by William Landau

### Thank you for your attention!

