Reproducible pipelines in R



High Performance Computing & Big Data Services







LU EMBOURG

with targets



HPC SChool 2021 PS11: R session

A. Ginolhac

DLSM University of Luxembourg

















Introduction to R

Not the scope of this session

Beginner user, check out this lecture

What is **Q** really?



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

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





Evaluation in programming

tidyeval



A. Ginolhac | rworkshop | 2020-11-30



targets a Make-like workflow manager for @



targets and companion package tarchetypes

A workflow manager for R

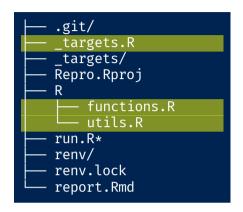
- Saving you time and stress
- Understand how it is implemented in targets
 - Define your targets
 - Connect targets to create the dependencies
 - Check **dependencies** with **visnetwork**
 - Embrace dynamic branching
 - Run only what needs to be executed
 - Bundle **dependencies** in a Rmarkdown document with <u>tar render()</u>
 - Increase reproducibility with the package manager <u>renv</u>
- Example with RNA-seq data from Wendkouni Nadège MINOUNGOU







Folder structure



- With <u>renv</u>. Snapshot your package environment (and restore! 😌)
- _targets.R is the only mandatory file
- Use a R sub-folder for functions, gets closer to a R package
- In a RStudio project
- Version tracked with git
- Rmarkdown file allows to gather results in a report
- Optional: an executable run. sh allows to use Build Tools in RStudio

Targets Markdown



- Makes development easier
- Documentation can be embedded
- targets engine recognizes by knitr and takes care of writing an scripts









renv features

- hydrate() parses your code and finds library calls
- install() from **CRAN** with dependencies (also from \bigcirc)
- snapshot() registers changes, hashes and origin
- restore() to a certain point in time



renv.lock file after a snapshot

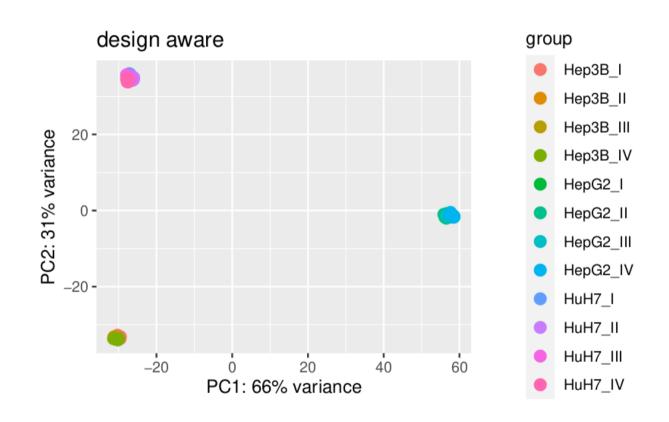
```
"R": {
  "Version": "4.0.3",
  "Repositories": [
      "Name": "CRAN".
      "URL": "https://cloud.r-project.org"
"Bioconductor": {
  "Version": "3.12"
"Packages": {
  "AnnotationDbi": {
    "Package": "AnnotationDbi",
    "Version": "1.52.0",
    "Source": "Bioconductor",
    "Hash": "ca5106b296b3aa6af713ce197be547c1"
  "BH": {
    "Package": "BH",
    "Version": "1.75.0-0",
    "Source": "Repository",
    "Repository": "CRAN",
    "Hash": "e4c04affc2cac20c8fec18385cd14691"
  "targets": {
    "Package": "targets",
    "Version": "0.1.0.9000",
    "Source": "GitHub",
    "RemoteType": "github",
    "RemoteUsername": "ropensci",
```

Example with RNA-seq data across 3 cell lines

PCA shows that differences between cells >> biological effect (roman numbers)

Solution: Split counts and metadata for each cell

Do we copy code 3 times?





Define targets = explicit dependencies



_targets.R, define 4 targets

Last target depends on the 3 upstreams

```
library(targets)
source("R/functions.R")
source("R/plotting.R")
list(
  tar target(cells, c("HepG2", "HuH7", "Hep3B")),
  tar_qs(dds, read_rds(here::here("data", "all.rds")),
         packages = "DESeq2"),
  tar fst tbl(annotation, gtf to tbl(here::here("data"
                                                 "gencode.v36.ar
              packages = c("tibble", "rtracklayer")),
 tar qs(sub dds, subset dds(dds,
                             filter(annotation, type == "gene"
                             .cell = cells),
         pattern = map(cells), # dynamic branching
         packages = c("DESeq2", "tidyverse"))
[\ldots]
```

Dynamic branching makes dependencies easier to read.

Of course, someone has to write for loops, it doesn't have to be you

Jenny Bryan

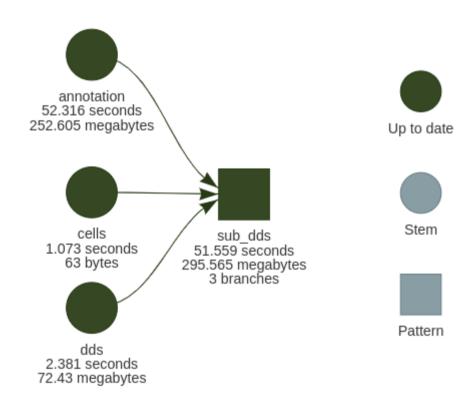


Running targets



```
• run target annotation
• run target cells
• run target dds
• run branch sub dds 3078b1e0
         condition time h
HepG2 I1 control
HepG2 I2
             HTI 6
using pre-existing size factors
estimating dispersions
gene-wise dispersion estimates: 2 workers
mean-dispersion relationship
final dispersion estimates, fitting model and testing: 2 worker
• run branch sub dds d05c5da7
        condition time h
HuH7 I1 control
HuH7 I2
            HTI 6
using pre-existing size factors
estimating dispersions
gene-wise dispersion estimates: 2 workers
mean-dispersion relationship
final dispersion estimates, fitting model and testing: 2 worker
• run branch sub dds c60d7096
         condition time h
Hep3B I1 control
Hep3B I2
              HTI 6
using pre-existing size factors
estimating dispersions
gene-wise dispersion estimates: 2 workers
mean-dispersion relationship
final dispersion estimates, fitting model and testing: 2 worker
• end pipeline
```

Options to display time and object sizes

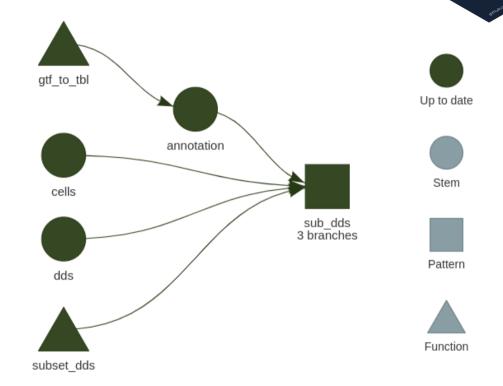


Re-running

- skip target annotation
 skip target cells
 skip target dds
 skip branch sub_dds_3078b1e0
 skip branch sub_dds_d05c5da7
 skip branch sub_dds_c60d7096
 skip pipeline
- All good, nothing to be done \checkmark .

Actually targets tracks all objects and so functions

A more complete dependency graph shows **functions**



Let's add the PCA per cell type now



targets

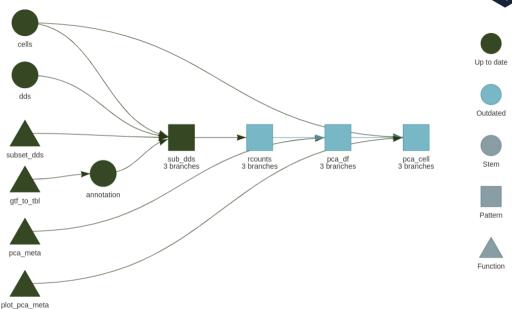
PCA, add 4 targets



Smaller targets avoid unnecessary rerunning steps

Translate into:

- For every cell data, compute regularized counts (vst: variance stabilization)
- For every regularized counts, compute PCA (df: data.frame, i. e a table)
- For every cell names / PCA tables, plot PCA in a table for easier labeling



PCA results





Awesome feature: load results IN a Rmarkdown document

Separate code from content

How to display a plot





The full picture

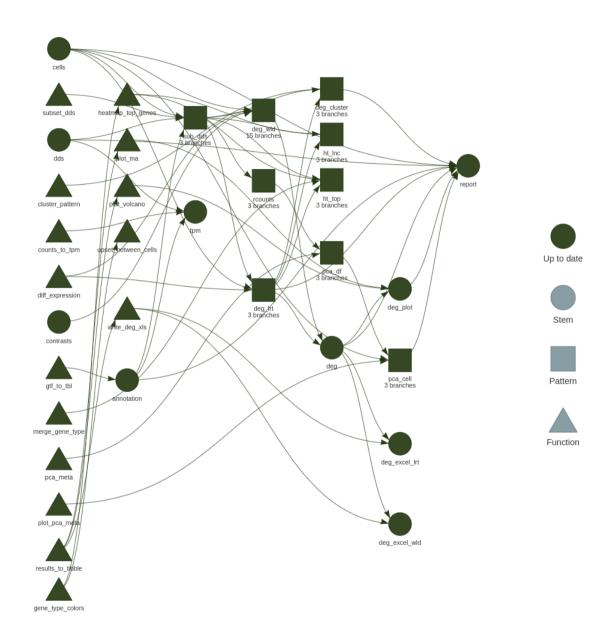
Adding step by step

desired analyses

Whole analysis takes 24 minutes and 4.54 seconds

Of course, someone has to remember the dependencies, it doesn't have to be you

— could be William Landau via **Jenny Bryan**







Is it worth the effort?

Yes

For you

- Autonomy
- Skills
- Free time
- Confidence over results
- Reproducibility
- Fun 🥳

Better project design

Thinking at what is a good targets helps tremendously the coding

- 1. Are large enough to subtract a decent amount of runtime when skipped.
- 2. Are small enough that some targets can be skipped even if others need to run.
- 3. Invoke no side effects (tar_target(format = "file") can save files.) 4. Return a single value that is
 - Easy to understand and introspect.
 - Meaningful to the project [...]

William Landau

Reproducibility

Both thanks to targets and renv via git

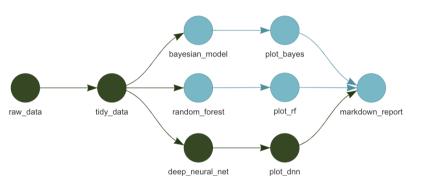


Scalability and parallelization

- Scale-up with **dynamic** branching
- Parallelization on **HPC** using:
 - o tar_make_clustermq(workers = 3L)
 (clustermq by Michael Schubert)
 - o tar_make_future(workers = 3L)
 (future by Henrik Bengtsson)
- Static branching

to get explicit branch names.















Source: William Landau: talk at Bayes Lund

Reports as Rmarkdown documents

targets, written by William Landau (pictured), is flexible, robust and still allows for a customized report.

All computing is done only when needed, and code is away from writing content.

Pipelines can now also be a **Rmd!**

Once **knitted** the report can be sent to the inquirer.

Targets Markdown

New in targets > **0.6**. Instructions at William bookdown

Test it as the Rmd template (and excellent <u>video</u> from R Lille meetup by **Landau**):



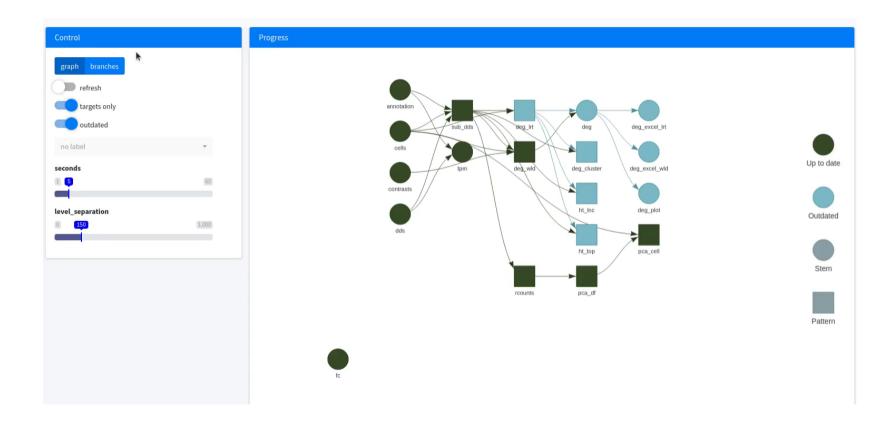


Bonus: watch the pipeline running live

Shiny, targets

- targets events watched live
- Here, after changing a threshold in the LRT step
- **branches** can be monitored too
- 2 videos joined as I fixed an **error** at 1'42"
- Option to display functions (unset here)

tar_watch() shiny app from targets





Before we stop

Highlights

- targets: a Makefile-like approach for project design
 - dependencies manager
 - re-run only what's needed

Acknowledgments 👃 👋



- Eric Koncina early adopter of targets
- Wendkouni N. Minoungou for the RNA-seq data
- William Landau main developer of targets
- Xie Yihui and Garrick Aden-Buie for xarigan/xaringanExtra
- Jennifer Bryan

Further reading

- Main website
- Targetopia Landau universe of targets-derived
- Video from R Lille meetup by William Landau. June 2021 45"
- Video from Bayes Lund by William Landau. October 2021
- <u>Documentation</u> as bookdown by **Landau**

Thank you for your attention!

