



Hit your reproducibility {targets}

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Reproducevangelism

		Data	
		Same	Different
Analysis	Same	Reproducible	Replicable
	Different	Robust	Generalisable

From **The Turing Way** by The Alan Turing Institute

Can I recreate what you did:

- from scratch?
- on a different machine?
- in the future?
- without you present?

R has many reproducibility tools, like:

- RStudio Projects to keep everything together
- R Markdown for reproducible docs
- packages for reusable functions
- {here} for relative filepaths
- {renv} for dependency management

Reproducible Analytical Pipelines



An example: Can {drake} RAP?

Today's focus:

1. Make *workflows* reproducible
2. Try {targets}

1. Make workflows reproducible

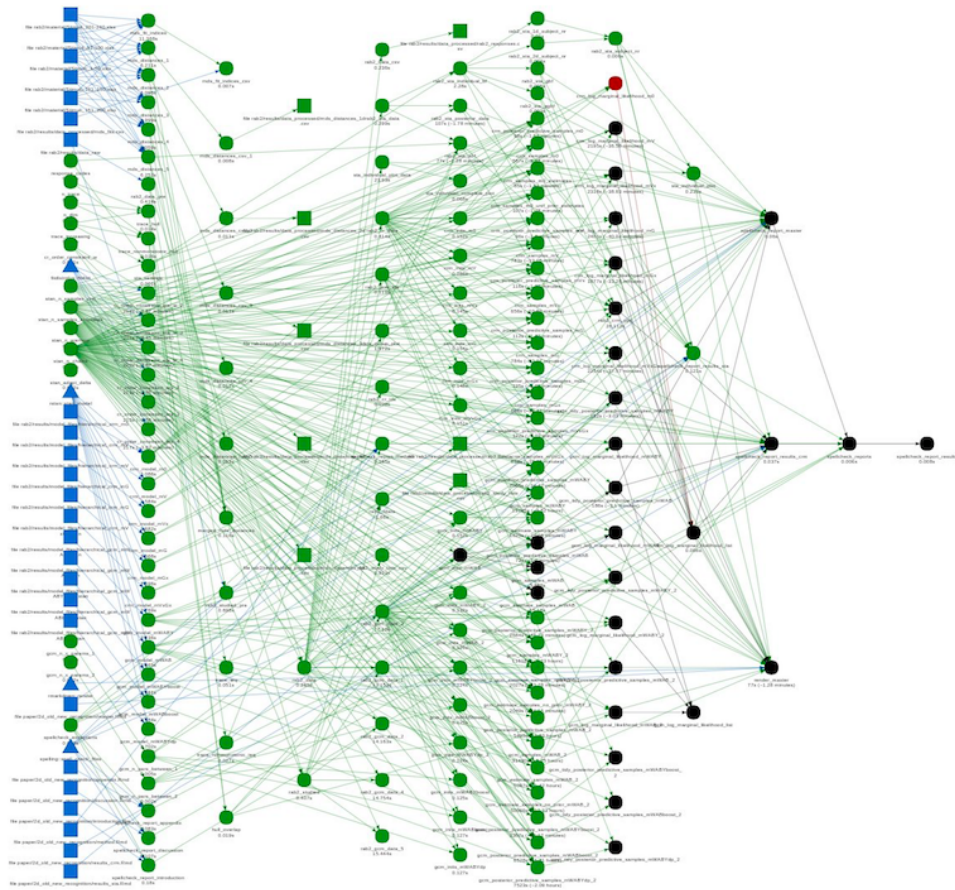
What about your analytical *workflow* itself?

How do you keep track of function, file and object relationships?

What if:

- you haven't recorded the steps?
- the interdependencies become complex?
- some steps are computationally intensive?
- something changes?

You can't remember this



Let a workflow manager handle it

- [Dagobah](#) - Simple DAG-based job scheduler in Python.
- [Dagr](#) - A scala based DSL and framework for writing and executing bioinformatics pipelines as Directed Acyclic GRaphs.
- [Dagster](#) - Python-based API for defining DAGs that interfaces with popular workflow managers for building data applications.
- [DataJoint](#) - an open-source relational framework for scientific data pipelines.
- [Dask](#) - Dask is a flexible parallel computing library for analytics.
- [Dockerflow](#) - Workflow runner that uses Dataflow to run a series of tasks in Docker.
- [Doit](#) - Task management & automation tool.
- [Drake](#) - Robust DSL akin to Make, implemented in Clojure.
- [Drake R package](#) - Reproducibility and high-performance computing with an easy R-focused interface. Unrelated to Factual's Drake.
- [Dray](#) - An engine for managing the execution of container-based workflows.
- [eHive](#) - System for creating and running pipelines on a distributed compute resource.
- [Fission Workflows](#) - A fast, lightweight workflow engine for serverless/FaaS functions.
- [Flex](#) - Language agnostic framework for building flexible data science pipelines (Python/Shell/Gnuplot).
- [Flower](#) - Robust and efficient workflows using a simple language agnostic approach (R package).



[pditommaso/awesome-pipeline](#)

2. Try {targets}



{targets} by Will Landau



Supersedes {drake} by Will Landau

{targets} is compelling because it's:

- R-specific
- free
- under active development
- got great documentation and examples

At its simplest:

1. Make a `tar_pipeline()`
2. Run `tar_make()`
3. Change stuff
4. Go to 2

Small demo



Generate `_targets.R` in working directory

```
tar_script()
```

You can [view the file](#) for this example online

Set up functions and options in `_targets.R`

```
b_plot <- function(data, image) {  
  ggplot(data, aes(id, temp)) +  
    geom_boxplot() +  
    labs(title = "Beaver temperatures") +  
    add_phylopic(image)  
}
```

```
tar_option_set(  
  packages = c(  
    "targets", "tarchetypes", "knitr",  
    "dplyr", "ggplot2", "rphylopic",  
  )  
)
```

Set up pipeline in `_targets.R`

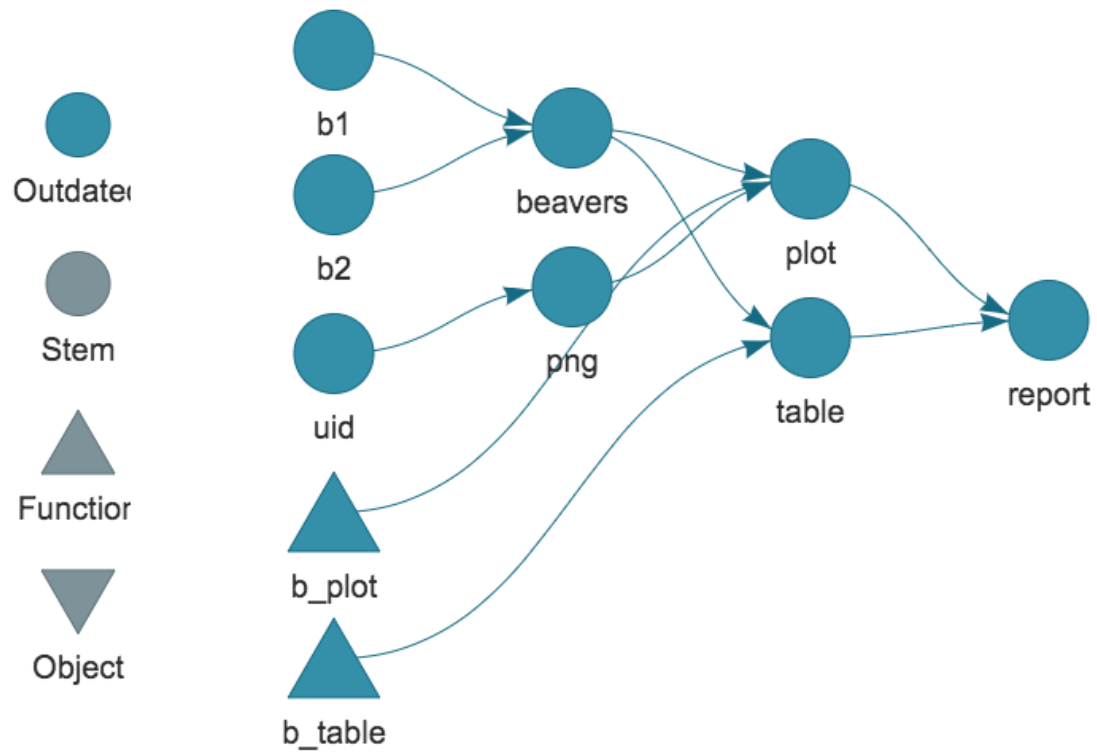
```
targets <- list(  
  tar_target(b1, mutate(beaver1, id = "A")),  
  tar_target(b2, mutate(beaver2, id = "B")),  
  tar_target(beavers, bind_rows(b1, b2)),  
  tar_target(uid, "be8670c2-a5bd-4b44-88e8-92f8b0c7f4"),  
  tar_target(png, image_data(uid, size = "512")[[1]]),  
  tar_target(plot, b_plot(beavers, png)),  
  tar_target(table, b_table(beavers)),  
  tarchetypes::tar_render(report, "beavers-report.Rmd")  
)  
  
tar_pipeline(targets)
```

View targets

```
tar_manifest(fields = "command")
```

	name	command
	<chr>	<chr>
1	uid	"\"be8670c2-a5bd-4b44-88e8-92f8b0c7f4c6\""
2	report	"tarchetypes::tar_render_run(path = \"beave
3	beavers	"bind_rows(b1, b2)"
4	table	"b_table(beavers)"
5	plot	"b_plot(beavers, png)"
6	b1	"mutate(beaver1, id = \"A\")"
7	b2	"mutate(beaver2, id = \"B\")"
8	png	"image_data(uid, size = \"512\")[[1]]"

```
tar_visnetwork()
```

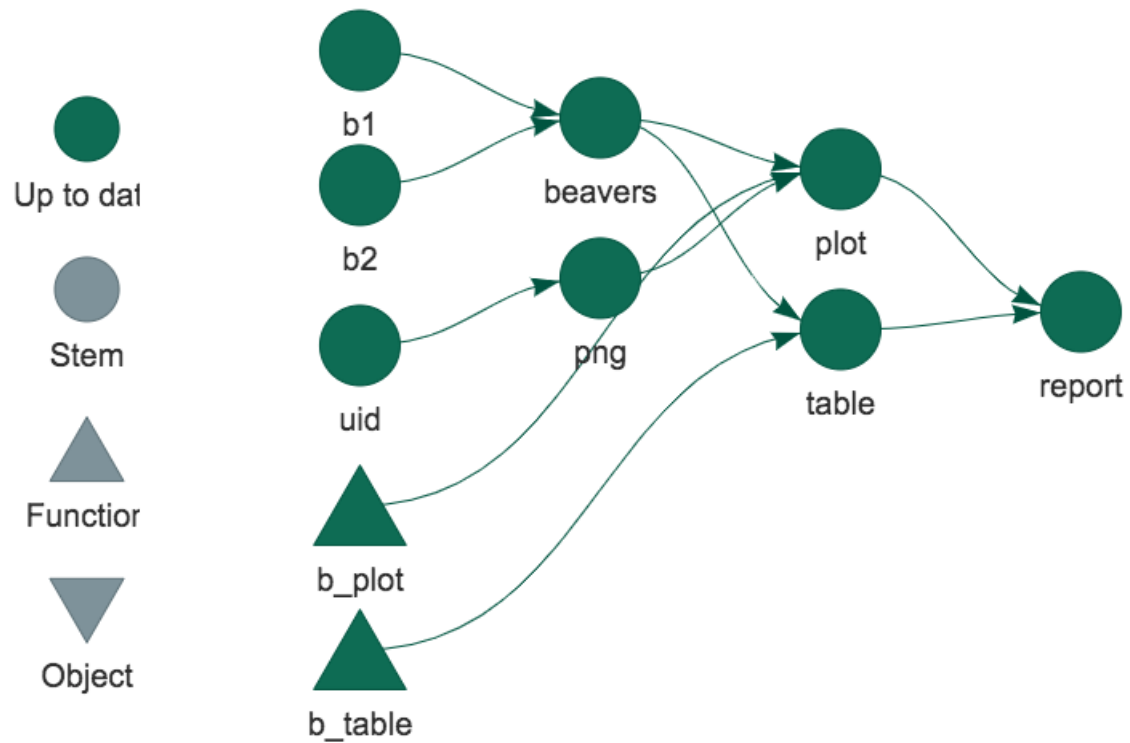


Execute the pipeline

```
tar_make()
```

- run target uid
- run target b1
- run target b2
- run target png
- run target beavers
- run target plot
- run target table
- run target report


```
tar_visnetwork()
```



Now make your changes

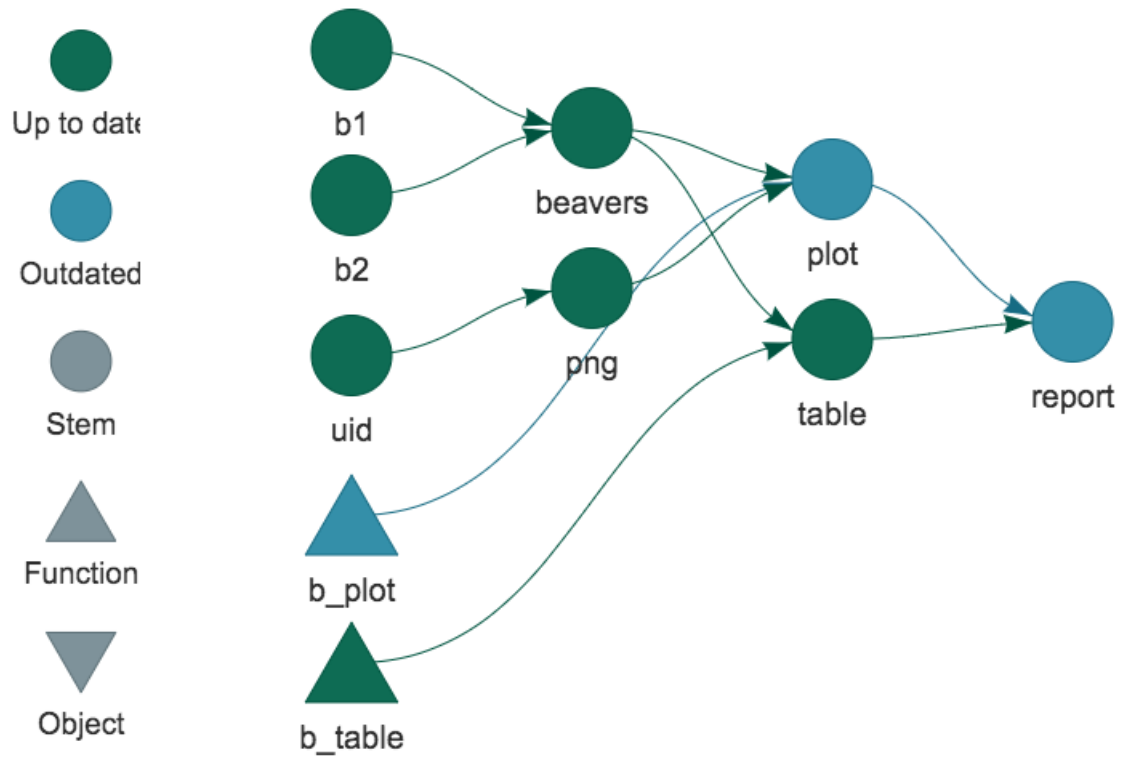
Example: alter the title in the `b_plot()` function

Check what has become outdated as a result

```
tar_outdated()
```

```
[1] "report" "plot"
```

```
tar_visnetwork()
```

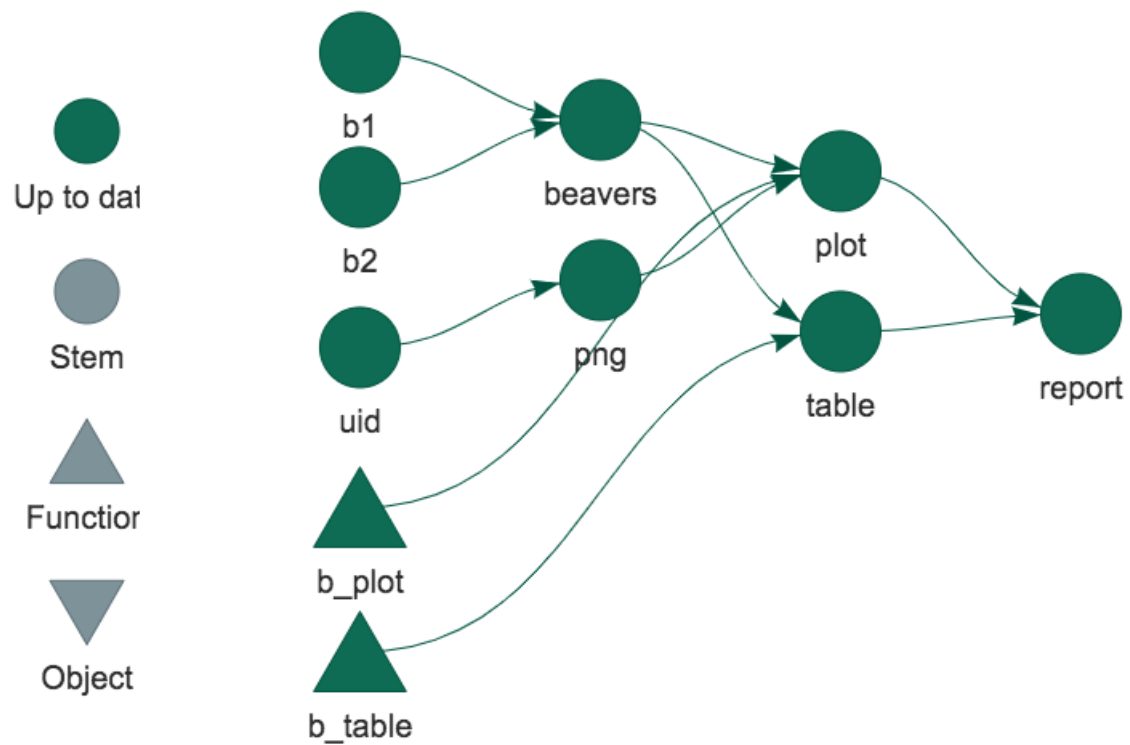


Only outdated targets are re-run

```
tar_make()
```

- ✓ skip target uid
- ✓ skip target b1
- ✓ skip target b2
- ✓ skip target png
- ✓ skip target beavers
- run target plot
- ✓ skip target table
- run target report

```
tar_visnetwork()
```



To recap:

1. `tar_script()` creates `_targets.R`
2. Add `tar_targets()`
3. Check `tar_manifest()`
4. `tar_visnetwork()` to visualise
5. Execute with `tar_make()`
6. Change stuff, check `tar_outdated()`
7. Go to 4

What now?

Check out official {targets} materials:

- the [site](#) and [source](#)
- the {targets} R package user manual
- [minimal example](#) (can be run in RStudio Cloud)
- [targetsketch](#): a Shiny app for learning {targets} and setup new projects

More available from [the {targets} README](#).

Revisit today's materials:

- matt-dray.github.io/targets-dsfest/
- github.com/matt-dray/targets-dsfest

Find me:

 [mattdray](#)  [matt-dray](#)  [rostrum.blog](#)



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Sources

- {targets} hex logo
- {drake} hex logo
- {drake} hairball by Frederik Aust
- *Castor canadensis* from PhyloPic