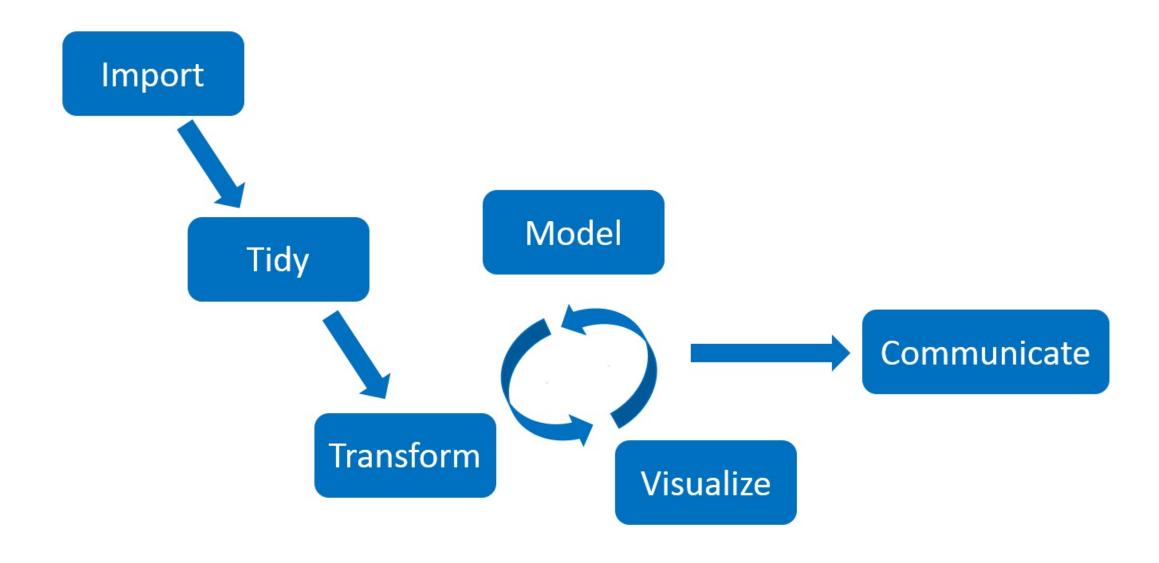
Machine learning workflow management in R

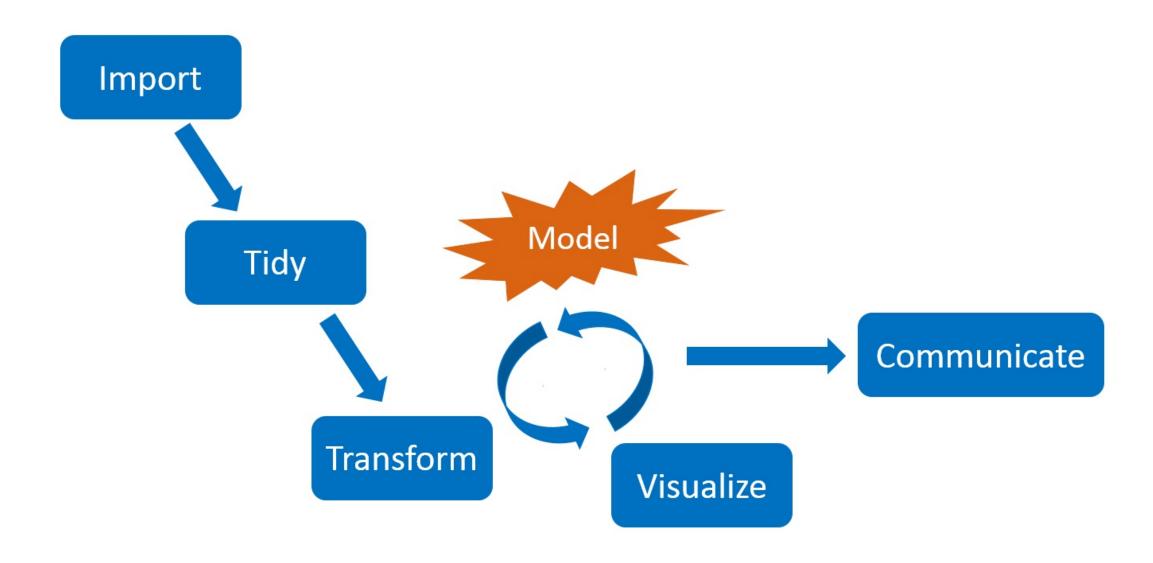


Will Landau

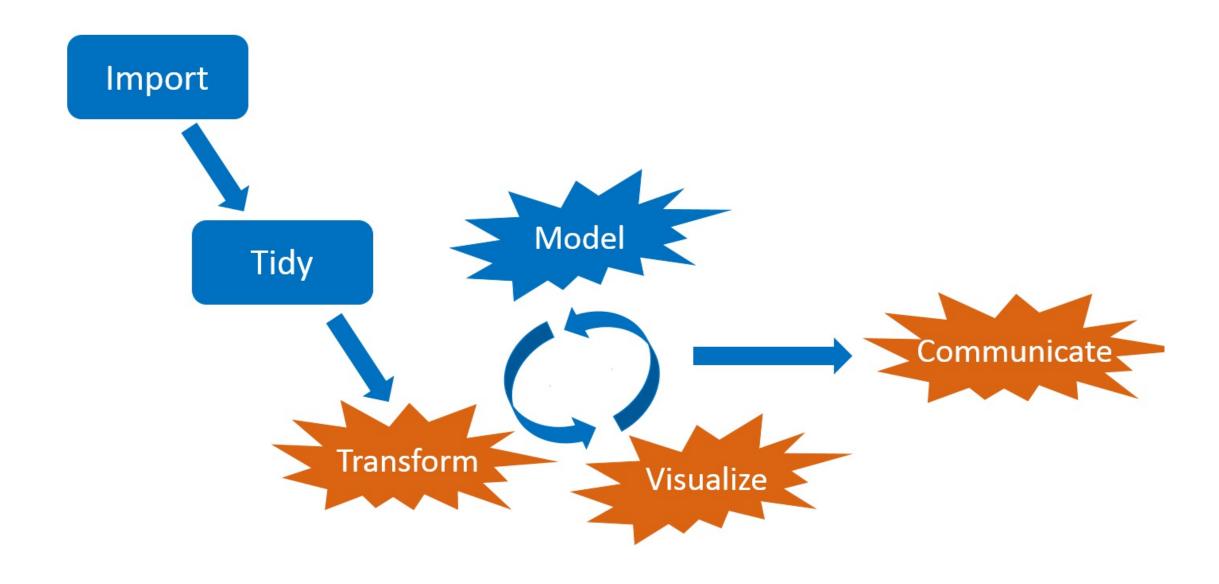
Workflows have interconnected steps.



When you change something...



...the downstream steps are no longer valid.



Do you rerun everything from scratch?

• Not if you deal with long runtimes!



https://openclipart.org/detail/275842/sisyphus-overcoming-silhouette

Do you pick and choose what to update?

- Messy.
- Prone to human error.
- Not reproducible.



https://openclipart.org/detail/216179/messy-desk

When do we face these issues?

- Long computation!
 - Clinical trial modeling and simulation
 - Subgroup identification
 - Bayesian data analysis
 - Graph-based multiple comparison procedures
 - Bayesian networks in genomics
 - PK/PD modeling (e.g. mrgsolve)
 - Deep learning
 - 0 ...

Solution: pipeline tools





See evidence of reproducibility.

- Tons exist already: github.com/pditommaso/awesome-pipeline.
- Most are language-agnostic or designed for Python or the shell.

What makes drake different?



- Aggressively designed for R.
 - Think **functions**, not script files.
 - Think **variables**, not output files.
 - Think data frames, not Makefiles.
- Major improvements in late 2018 and early 2019:
 - A domain-specific language for workflows.
 - Increased speed and reduced memory usage.
 - More reproducibility safeguards.
 - History and provenance.
 - Reproducible data recovery and renaming (experimental).

Example: a deep learning workflow

- Goal: predict customers who cancel their subscriptions with a telecom company.
- Data: IBM Watson Telco Customer Churn dataset.
- Workflow principles generalize to other industries.



https://openclipart.org/detail/90739/newplus, https://github.com/rstudio/keras

File structure

```
make.R
R/
I packages.R
I functions.R
I plan.R
data/
Customer_churn.csv
```

packages.R

```
library(drake)
library(keras)
library(recipes)
library(rsample)
library(tidyverse)
library(yardstick)
```

functions.R

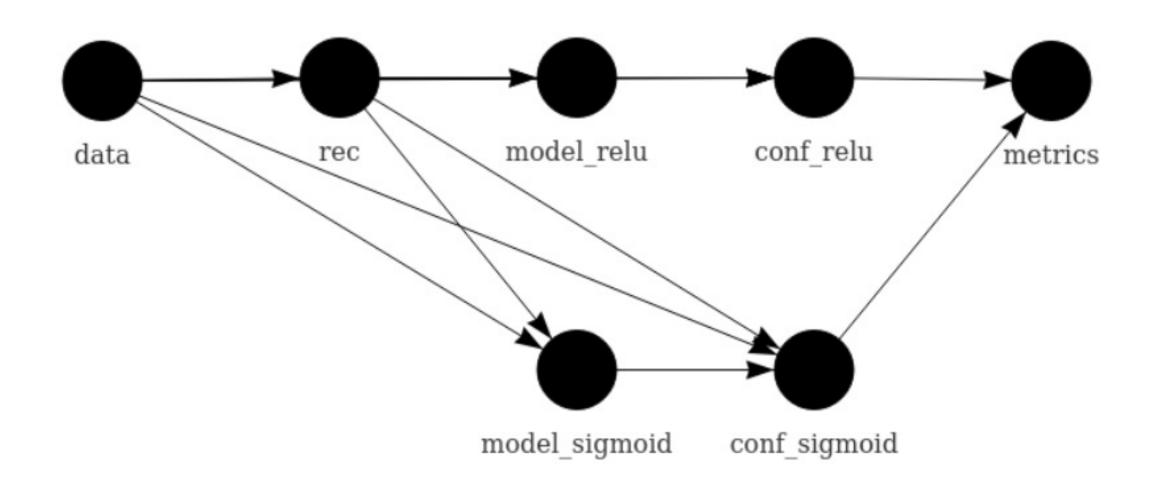
```
prepare_recipe <- function(data) {</pre>
define_model <- function(rec, units1, units2, act1, act2, act3) {</pre>
  # . . .
train_model <- function(data, rec, units1, units2, act1, act2, act3) {</pre>
  # . . .
confusion_matrix <- function(data, rec, serialized_model) {</pre>
  # . . .
compare_models <- function(...) {</pre>
  # ...
```

plan.R

```
activations <- c("relu", "sigmoid")</pre>
plan <- drake_plan(</pre>
 data = read_csv(file_in("data/customer_churn.csv"), col_types = cols()) %>;
    initial_split(prop = 0.3),
  rec = prepare_recipe(data),
 model = target(
   train_model(data, rec, act1 = act),
   transform = map(act = !!activations)
 conf = target(
    confusion_matrix(data, rec, model),
   transform = map(model, .id = act)
 metrics = target(
    compare_models(conf),
   transform = combine(conf)
```

Data frame of workflow steps

The workflow



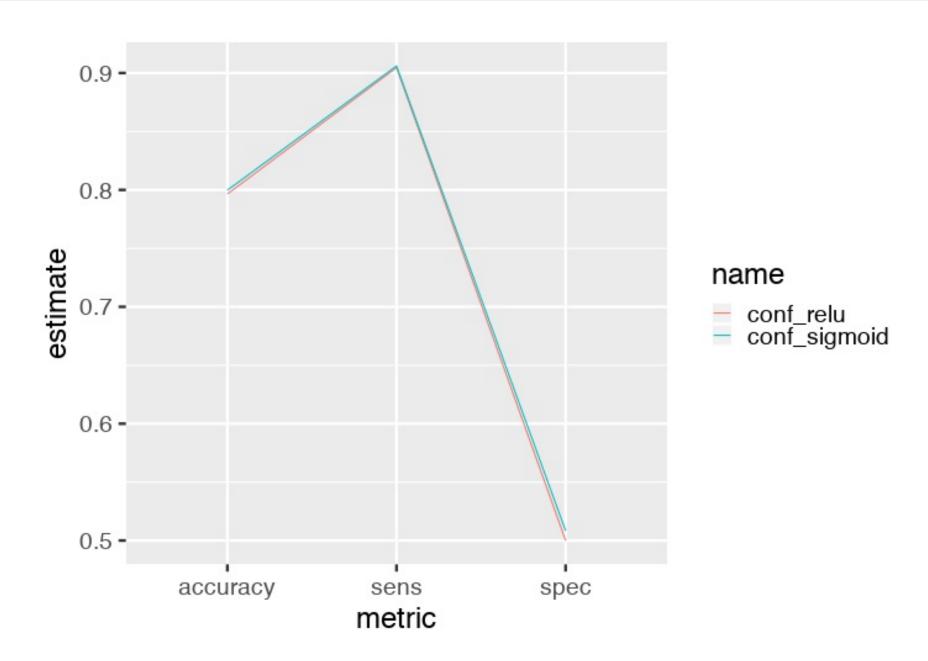
Run the project in make.R.

```
source("R/packages.R")
source("R/functions.R")
source("R/plan.R")

make(plan)
## target data
## target rec
## target model_relu
## target model_sigmoid
## target conf_relu
## target conf_sigmoid
## target metrics
```

Compare models.

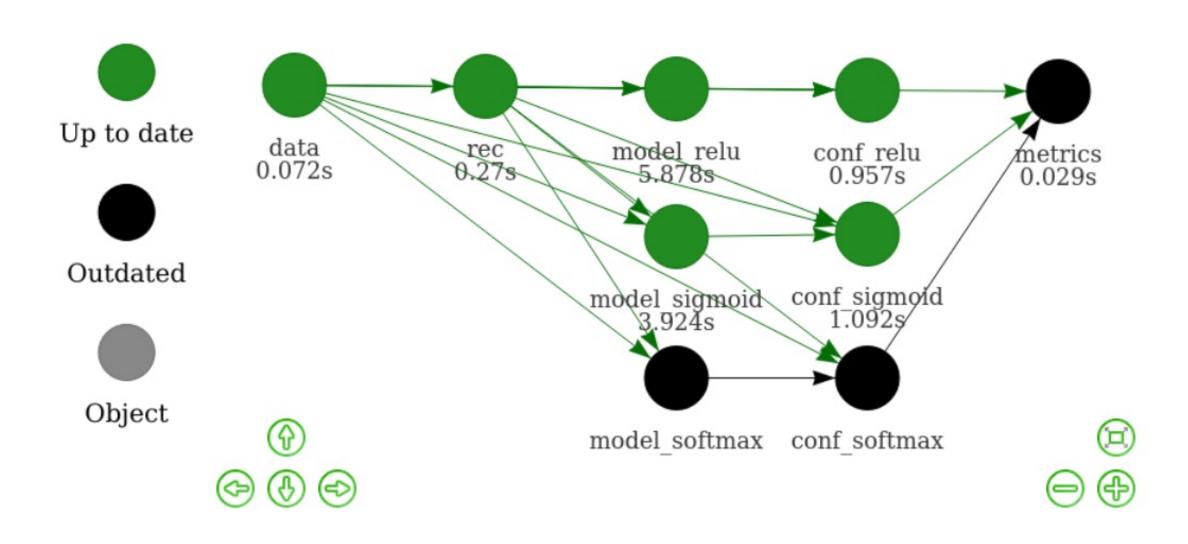
readd(metrics) # See also loadd()



Try another activation function.

```
activations <- c("relu", "sigmoid", "softmax")</pre>
plan <- drake_plan(</pre>
  data = read_csv(file_in("data/customer_churn.csv"), col_types = cols()) %>9
    initial_split(prop = 0.3),
  rec = prepare_recipe(data),
 model = target(
  train_model(data, rec, act1 = act),
   transform = map(act = !!activations)
 conf = target(
    confusion_matrix(data, rec, model),
   transform = map(model, .id = act)
 metrics = target(
    compare_models(conf),
    transform = combine(conf)
```

vis_drake_graph()



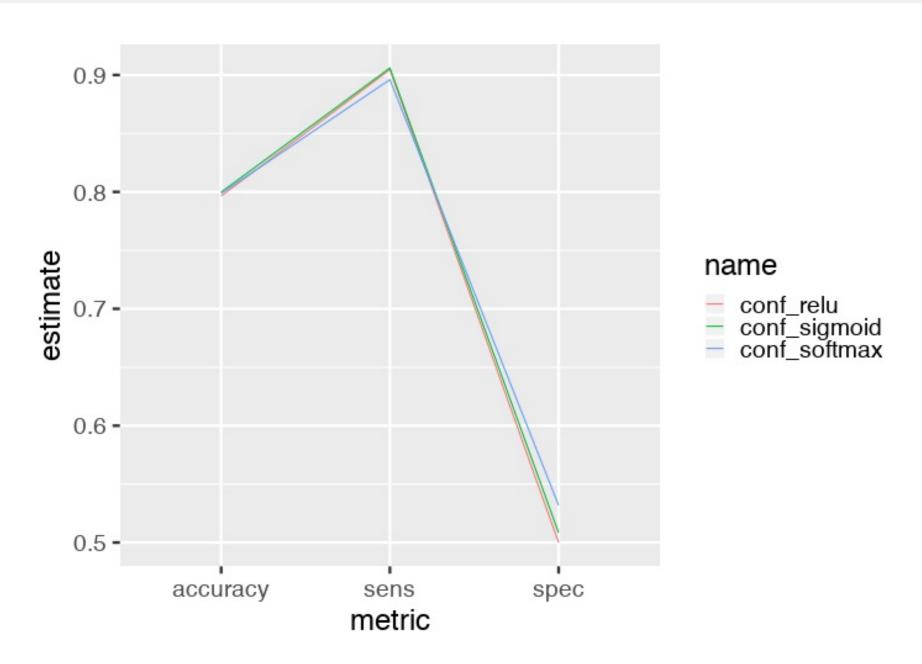
Refresh the results in make.R.

```
source("R/packages.R")
source("R/functions.R")
source("R/plan.R") # modified

make(plan)
## target model_softmax
## target conf_softmax
## target metrics
```

Compare models.

readd(metrics)



Evidence of reproducibility

```
source("R/packages.R")
source("R/functions.R")
source("R/plan.R")

make(plan)
## All targets are already up to date.
```

• See also outdated().

History and provenance

```
drake_history()
## # A tibble: 10 x 10
##
      target current built
                               exists hash command
                                                           seed runtime
                                                                            prop act.
               <lgl>
                                                                  <dbl> <dbl> <ch
    <chr>
                       <chr> <lgl>
                                       <chr> <chr>
##
                                                           <int>
    1 conf_r... TRUE
##
                        2019-... TRUE
                                        3cc5... confusio... 4.05e8
                                                                   1.65
                                                                            NA
                                                                                 <NA:
##
    2 conf_s... TRUE
                        2019-... TRUE
                                        9c3e... confusio... 1.93e9
                                                                            NA
                                                                                 <NA:
                                                                   1.78
                        2019-... TRUE
                                                                                 <NA:
##
    3 conf s... TRUE
                                        527e... confusio... 1.80e9
                                                                   2.40
                                                                            NA
    4 data
               TRUE
                        2019-... TRUE
                                        ca84... "read cs... 1.29e9
                                                                            0.3 <NA
##
                                                                   0.118
    5 metrics FALSE
                        2019-... TRUE
                                        6cbb... compare_... 1.21e9
                                                                                 <NA:
##
                                                                   0.0250
                                                                            NA
                        2019-... TRUE
    6 metrics TRUE
                                        5103... compare_... 1.21e9
                                                                                 <NA:
##
                                                                   0.021
                                                                            NA
    7 model_... TRUE
                                        ba33... "train_m... 1.47e9 35.6
                        2019-... TRUE
                                                                                 reli
##
                                                                            NA
##
    8 model ... TRUE
                        2019-... TRUE
                                        c86c... "train m... 1.26e9
                                                                   6.09
                                                                            NA
                                                                                 sign
    9 model_... TRUE
                        2019-... TRUE
                                        19d9... "train m... 8.05e8
                                                                  7.34
                                                                                 sof.
                                                                            NA
                        2019-... TRUE
##
  10 rec
               TRUE
                                        40e5... prepare_... 6.29e8
                                                                   0.191
                                                                            NA
                                                                                  <NA:
                                                                                  ₩F
```

Reproducible data recovery (experimental)

```
clean() # Oops!
start <- proc.time()</pre>
make(plan, recover = TRUE)
## recover data
## recover rec
## recover model relu
## recover model_sigmoid
## recover model_softmax
## recover conf_relu
## recover conf_sigmoid
## recover conf_softmax
## recover metrics
proc.time() - start
## user system elapsed
## 0.089 0.043 0.256
```

• Details + how to rename a target: https://ropenscilabs.github.io/drake-manual/walkthrough.html#automated-recovery-and-renaming

High-performance computing

```
# template file with configuration
drake_hpc_template_file("slurm_clustermq.tmpl")

# Use SLURM resource manager with the template.
options(
   clustermq.scheduler = "slurm",
   clustermq.template = "slurm_clustermq.tmpl"
)

# make() is the basically the same.
make(plan, jobs = 2, parallelism = "clustermq")
```

High-performance computing

Resources

• Get drake:

```
install.packages("drake")
```

• Workshop materials:

```
remotes::install_github("wlandau/learndrake")
```

• Example code from these slides:

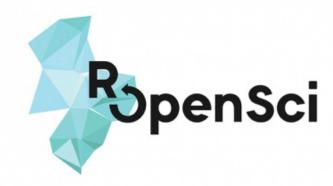
```
drake::drake_example("customer-churn-simple")
```

Links

- Development repository: https://github.com/ropensci/drake
- Full user manual https://ropenscilabs.github.io/drake-manual
- Reference website: https://ropensci.github.io/drake
- Code examples: https://github.com/wlandau/drake-examples
- Discuss at rOpenSci.org: https://discuss.ropensci.org

rOpenSci use cases

• Use drake? Share your use case at https://ropensci.org/usecases.



Thanks



Edgar Ruizexample code



Matt Danchoblog post

Thanks



- Maëlle Salmon
- Ben Marwick
- Julia Lowndes
- Peter Slaughter
- Jenny Bryan
- Rich FitzJohn
- Stefanie Butland

- Jarad Niemi
- Kirill Müller
- Henrik Bengtsson
- Michael Schubert
- Kendon Bell
- Miles McBain
- Patrick Schratz
- Alex Axthelm
- Jasper Clarkberg
- Tiernan Martin
- Ben Listyg
- TJ Mahr
- Ben Bond-Lamberty
- Tim Mastny
- Bill Denney
- Amanda Dobbyn
- Daniel Falster
- Rainer Krug
- Brianna McHorse
- Chan-Yub Park