

SYBERIA: A DEVELOPMENT FRAMEWORK FOR R

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- Packages work well for recording abstract solutions to problems, but not for large projects maintained by multiple users tied to solving problems in a specific domain

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- Packages work well for recording abstract solutions to problems, but not for large projects maintained by multiple users tied to solving problems in a specific domain
- Test-driven development is difficult for modeling work that is intended for real-time systems

An Industry-wide Problem



Machine Learning: The High Interest Credit Card of Technical Debt (2014)

NIPS 2014 Workshop proceedings

D. Sculley, Gary Holt, Daniel Golovin, et al

"Risk factors include boundary erosion, entanglement, hidden feedback loops, undeclared consumers, data dependencies, changes in the external world, and a variety of system-level anti-patterns."

Solution: Think like a developer

Inspiration from the Competition

Developers are good at simplifying work that needs to be done down to its core abstractions





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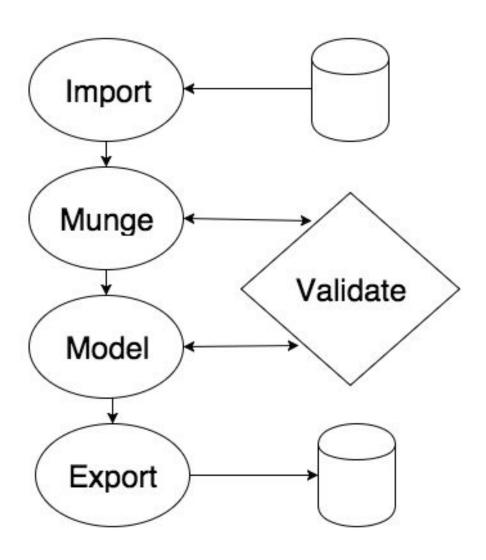
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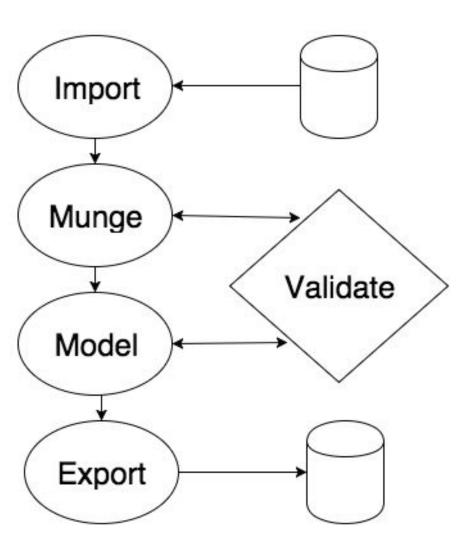
Syberia is a framework for building complex projects in R.

The modular design unit is an engine.

Today we are releasing the **modeling engine** for building and deploying production-ready machine learning products in R.

How Does It Work?

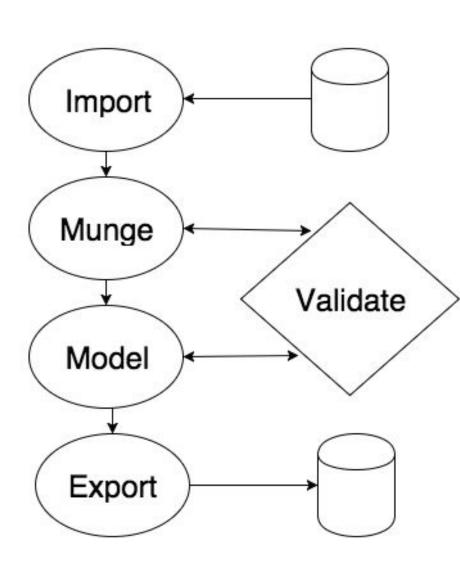




The modeling engine is the boss.

Mungebits package powers feature engineering.

Tundra package powers model containers.



The modeling engine is the boss.

Director package holds the project together.

Stagerunner package is the workflow and execution system.

The Basic Structure

- Script-driven workflow makes it harder to re-use components
- Testing is not built in unless you move to a package

resource ("lib/adapters/s3")

- Basic idea: Everything is a resource
- All resources must be tested (test/lib/adapters/s3)
- Each resource type can have its own "grammar"

Task: Let's import some data using Syberia.

```
# lib/adapters/s3.R
read <- function(name) {</pre>
  s3mpi::s3read(name)
write <- function(object, name) {</pre>
  s3mpi::s3store(object, name)
 package: github.com/robertzk/s3mpi
```

```
# config/routes.R
list(
  "config/global"
                    = "globals",
  "lib/adapters"
                    = "adapters",
  "lib/classifiers" = "classifiers",
  "lib/mungebits"
                    = "mungebits",
         "models"
                     = "models",
    "test/models"
                    = "test/models",
                    = "data")
           "data"
```

The Basic Structure

```
a <- resource("lib/adapters/s3")</pre>
a$write(iris, "tmp/iris")
 From a new R session
a <- resource("lib/adapters/s3")</pre>
identical (
  a$read("tmp/iris"),
  iris
```

```
# test/lib/adapters/s3.R
 test that("it can write a data set to S3", {
   env <- new.env()
   package_stub("s3mpi", "s3store", function(...) { env[[..2]] <- ..1 }, {</pre>
     adapter <- resource()
     adapter$write(iris, "test key", prefix = "")
     expect identical(env$test key, iris,
       info = "iris should have been stored in the test_key in env")
   })
 })
 test_that("it can read a data set from S3", {
   env <- list2env(list(test_key = iris))</pre>
   package stub("s3mpi", "s3read", function(...) { env[[..1]] }, {
     adapter <- resource()
     expect_identical(adapter$read("test_key", prefix = ""), env$test_key,
       info = "iris should have been read from the test_key in env")
   })
```

})

```
# config/routes.R
list(
  "config/global"
                    = "globals",
  "lib/adapters"
                    = "adapters",
  "lib/classifiers" = "classifiers",
  "lib/mungebits"
                    = "mungebits",
         "models"
                     = "models",
    "test/models"
                    = "test/models",
                    = "data")
           "data"
```

```
# lib/controllers/adapters.R
function (input) {
 adapter class <- function(r, w) {
    list(read = r, write = w)
  }
  # Construct the adapter object.
  adapter class (
   input$read,
   input$write
```

```
# lib/adapters/s3.R
read <- function(name) {</pre>
  s3mpi::s3read(name)
write <- function(object, name) {</pre>
  s3mpi::s3store(object, name)
 package: github.com/robertzk/s3mpi
```

Think of how you might write adapters for other storage backends:

- Reading from and writing to a CSV file
- Reading from and writing to a database
- Reading from and writing to a JSON service
- Et cetera

- Defining "adapters" abstracted away the storage backend from the underlying implementation
- Each adapter has the same interface

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- Each adapter has the same interface

```
adapter <- resource("lib/adapters/s3")
adapter$read("some_key")
adapter$write(object, "some key")</pre>
```

- Defining "adapters" abstracted away the storage backend from the underlying implementation
- Each adapter has the same interface

```
adapter <- resource("lib/adapters/s3")
adapter$read("some_key")
adapter$write(object, "some_key")

adapter <- resource("lib/adapters/file")
adapter$read("some_file.rds")
adapter$write(iris, "some file.csv")</pre>
```

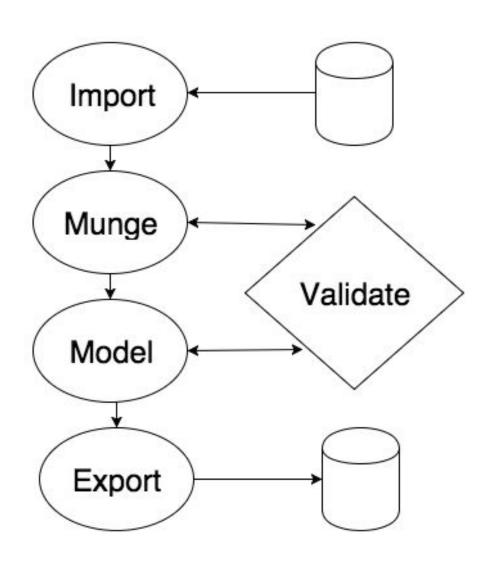
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- Each resource produces a single R object
- Encourages finding common interfaces and abstractions
- Work becomes easily re-usable instead of locked away in scripts

The Modeling Engine



The Modeling Engine

```
list(
 # Titanic dataset is fairly popular. Here we're downloading it from a public github repo
 import = list(url = "https://raw.githubusercontent.com/haven-jeon/.../master/bicdata/data/titanic.csv"),
 data = list(
   # The left-hand side defines the informal name of a mungebit that you will see when you run this model.
   # The right-hand side is the mungebit invocation.
   "Factor to character" = list(column_transformation(as.character), is.factor)
                        = list(multi_column_transformation(function(name) grepl("(", fixed = TRUE, name)), "name", "has_paren")
   ,"has paren in name"
   ,"Name length variable" = list(new variable, function(name) nchar(name), "name length")
   # ~40 removed
   ,"Restore levels"
                          = list(restore_categorical_variables)
   ,"Rename dep_var"
                          = list(renamer, c("survived" = "dep_var"))
 ),
 model = list('gbm'
   , .id_var
                        = 'X'
   , distribution
                     = 'bernoulli'
   , number_of_trees
                     = 3000
   , shrinkage_factor
                        = 0.005
 ),
 export = list(R = 'titanic')
```

The Modeling Engine

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list(
 # Titanic dataset is fairly popular. Here we're downloading it from a public github repo
 import = list(url = "https://raw.githubusercontent.com/haven-jeon/.../master/bicdata/data/titanic.csv"),
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export = list(R = 'titanic')

Mungebits

```
# config/routes.R
list(
  "config/global"
                    = "globals",
  "lib/adapters"
                     = "adapters",
  "lib/classifiers" = "classifiers",
  "lib/mungebits"
                     = "mungebits",
         "models"
                     = "models",
    "test/models"
                     = "test/models",
                     = "data")
           "data"
```

Mungebits

Sort data in ascending order by a given column.

```
# lib/mungebits/orderer.R
train <- predict <- function(dataframe, col) {
   dataframe[order(dataframe[[col]]), ]
}

# From R console
m <- resource("lib/mungebits/orderer")
stopifnot(all.equal(
   m$run(iris, 1), iris[order(iris[[1]]), ]
))</pre>
```

Mean imputation for one column.

```
# lib/mungebits/simple impute.R
train <- function(dataframe, col) {</pre>
  input$col <- col</pre>
  input$mean <- mean(dataframe[[col]], na.rm=T)</pre>
  dataframe[is.na(dataframe[[col]]), col] <-</pre>
    input$mean
  dataframe
predict <- function(dataframe, ...) {</pre>
  col <- input$col
  dataframe[is.na(dataframe[[col]]), col] <-</pre>
    input$mean
  dataframe
```

Mungebits

```
# In R console
m <- resource("lib/mungebits/simple_impute")
iris2 <- iris; iris2[1, 1] <- NA
m$run(iris2, 1)
stopifnot(all.equal(
    m$run(iris2)[[1]],
    c(mean(iris[-1, 1]), iris2[-1, 1])
))</pre>
```

Mungebits

```
# lib/controllers/mungebits.R
function (input)
  if (isTRUE(input$column transformation)) {
    mungebits2::mungebit$new(
mungebits2::column transformation(input$train),
mungebits2::column transformation(input$predict))
  } else {
    mungebits2::mungebit$new(
      input$train, input$predict)
```

Classifiers

lib/controllers/classifiers.R

```
function(input) {
force(input)
function(munge_procedure = list(), default_args = list(), internal = list()) {
  input <- lapply(as.list(input), full_deflate)</pre>
  container <- tundra::tundra_container$new(resource, input$train, input$predict,
    munge_procedure, full_deflate(default_args), full_deflate(internal))
  container$hooks <- lapply(container$hooks, function(fn) {</pre>
    environment(fn) <- globalenv(); fn</pre>
  })
  if (!is.null(input$read) ||
      !is.null(input$write)) {
    attr(container, "s3mpi.serialize") <- list(read = input$read, write = input$write)
   container
}
```

The Modeling Engine

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 # Titanic dataset is fairly popular. Here we're downloading it from a public github repo
 import = list(url = "https://raw.githubusercontent.com/haven-jeon/.../master/bicdata/data/titanic.csv"),
 data = list(
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   ,"has paren in name"
                          = list(multi_column_transformation(function(name) grepl("(", fixed = TRUE, name)), "name", "has_paren")
   ,"Name length variable" = list(new variable, function(name) nchar(name), "name length")
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   ,"Restore levels"
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   , .id_var
   , distribution
                        = 'bernoulli'
   , number_of_trees
   , shrinkage_factor
                        = 0.005
 export = list(R = 'titanic')
```

The Modeling Engine

```
list(
 # Import and data stage
 model = list('gbm'
   , .id_var
                         = 'X'
                         = 'bernoulli'
   , distribution
   , number_of_trees = 3000
   , shrinkage_factor = 0.005
 # Export stage
```

The Models Controllers

lib/controllers/models.R

```
construct stage runner <- Ramd::define("construct stage runner")[[1]](resource)</pre>
preprocessor <- Ramd::define("preprocessor")[[1]]</pre>
function(args, resource, output, director, modified, any dependencies modified) {
parent.env(parent.env(environment(construct stage runner))) <- environment()</pre>
if (is.element("raw", names(args))) return(output)
require(objectdiff)
message("Loading model: ", resource)
tests <- file.path('test', resource)
has tests <- director$exists(tests)
has tests <- FALSE
if (has_tests) {
  testrunner <- stageRunner$new(new.env(), director$resource(tests))</pre>
  testrunner$transform(function(fn) {
    library(testthat); force(fn)
    function(after) fn(cached env, after)
  })
model version <- gsub("^\\w+/", "", resource)</pre>
if (isTRUE(args$fresh) | | !identical(resource, director$cache get("last model"))) {
   stagerunner <- construct stage runner(output, model version)
} else if (modified | any dependencies modified) {
  message(crayon::yellow("Copying cached environments..."))
  stagerunner <- construct stage runner(output, model version)</pre>
  stagerunner$coalesce(director$cache get("last model runner"))
} else if (!director$cache exists("last model runner")) {
  stagerunner <- construct stage runner(output, model version)
  stagerunner <- director$cache get("last model runner")</pre>
if (has tests) stagerunner$overlay(testrunner, "tests", flat = TRUE)
director$cache set("last model", resource)
director$cache set("last model runner", stagerunner)
stagerunner
```

../models/preprocessor.R

```
preprocessor <- function(resource, director, source env) {</pre>
source env$extending <- function(model version, expr) {</pre>
   eval.parent(substitute(within(resource(file.path("models/", model version), raw = TRUE), { expr })))
source env\$model version <- version <- gsub("^[^/]+\/[^/]+\/", "", resource)
source env$model name
                         <- basename(version)</pre>
 source env$output <-
   function(suffix = "", create = TRUE, dir = file.path(director$root(), "tmp")) {
     filename <- file.path(dir, version, suffix)
     if (create && !file.exists(dir <- dirname(filename)))</pre>
       dir.create(dir, recursive = TRUE)
     filename
lexicals <- director$resource("lib/shared/lexicals")</pre>
for (x in ls(lexicals)) source env[[x]] <- lexicals[[x]]</pre>
director$resource("lib/shared/source mungebits")(source env, director)
model <- source()</pre>
if (nzchar(Sys.getenv("CI"))) {
     model$import <- NULL
}
model
```

../models/ construct_stage_runner.R

```
model_env <- function() {
  if (identical(getOption("environment_type"), "environment")) {</pre>
```

Other Controller Ideas

- Other useful abstractions with their own grammar:
 - Modules: Nested bundles of related R functions.
 - Indicators: $y \sim x1 + x2$ tied to an ETL backend
 - Stages: Execution tasks for our stage runner.
 - Jobs: Build reports, monitoring checks, etc.
 - Queries: "Object-relational mapper" for R

run("example") # Runs models/dev/example.R

```
list(
  import = list(R = "iris"),
  data = list(
    "Create dep var" = list(renamer,
        c("Sepal.Length" = "dep_var")),
    "Example var" = list(example),
    "Create ID var" = list(
multi_column_transformation(seq_along),
        "dep_var", "id")
),
  model = list("lm", .id_var = "id"),
  export = list(R = "model")
)
```

```
3. R
> run('example')
Loading model: models/dev/example1
Beginning 1. import stage...
  Running 1. Import from R...
  Running 2. (Internal) Verify data was loaded...
Ending 1. import stage...
Beginning 2. data stage...
  Running 1. Create dep var...
  Running 2. Example var...
  Running 3. Create ID var...
Ending 2. data stage...
Running 3. model...
Beginning 4. export stage...
  Running 1. Export to R...
Ending 4. export stage...
```

run("example") # Runs models/dev/example.R

```
3. R
> run('example')
Loading model: models/dev/example1
Beginning 1. import stage...
  Running 1. Import from R...
  Running 2. (Internal) Verify data was loaded...
Ending 1. import stage...
Beginning 2. data stage...
  Running 1. Create dep var...
  Running 2. Example var...
  Running 3. Create ID var...
Ending 2. data stage...
Running 3. model...
Beginning 4. export stage...
  Running 1. Export to R...
Ending 4. export stage...
```

run(, "data/3") # Re-runs one munge step

```
list(
import = list(R = "iris"),
data = list(
    "Create dep var" = list(renamer
        c("Sepal.Length" = "dep_var")),
    "Example var" = list(example),
    "Create ID var" = list(
multi_column_transformation(seq_along),
        "dep_var", "id")
),
model = list("lm", .id_var = "id"),
export = list(R = "model")
)
```

```
> run(,"data/3")
Loading model: models/dev/example1
Beginning 2. data stage...
Running 3. Create ID var...
Ending 2. data stage...
> dim(B)
[1] 150 5
> dim(A)
[1] 150 6
> setdiff(ls(A), ls(B))
[1] "id"
> ■
```

setdiff(ls(B), ls(A)) # [1] "id"

Testing Built In

Each resource requires an accompanying test

test_project()

```
0 0
                             2. R
> test_engine()
Running tests...
lib/adapters/example:
lib/adapters/file: ...
lib/adapters/R: ...
lib/adapters/s3 : ...
lib/classifiers/gbm : .
lib/classifiers/lm: ....
lib/mungebits/trivial_column_transform : ...
lib/mungebits/trivial : ...
lib/shared/full_deflate: ....
lib/shared/gbm_parameters : ...
lib/shared/simple_deflate: .....
lib/stages/data: .
lib/stages/export : ....
lib/stages/import : ...
lib/stages/model : ...
```

Dependency Management

Everyone working on the project has the same set of dependencies

```
2. R
Using dependency stringi 1.0.1
Using dependency magnitur 1.5
Using dependency codetools 0.2.14
Using dependency XML 3.98.1.4
Using dependency Rcpp 0.12.4
Using dependency stringr 1.0.0
Using dependency pryr 0.1.2
Using dependency praise 1.0.0
Using dependency memoise 1.0.0
Using dependency whisker 0.4
Using dependency plyr 1.8.3
Using dependency R6 2.1.2
Using dependency crayon 1.3.1
Using dependency bettertrace 0.1.1
Using dependency testthat 0.11.0
Using dependency lazyeval 0.1.10.9000
Using dependency cacher 0.0.7
Using dependency testthatsomemore 0.2.4
Using dependency AWS.tools 0.1.0
Using s3mpi 0.2.19
Using stagerunner 0.5.3
Using Ramd 0.3.8
Using statsUtils 0.1.3
Using mungebits2 0.1.0.9002
Using syberiaMungebits2 0.1.0.9001
Using director 0.3.0.5.9000
Using tundra 0.3.0.9000
Using syberia 0.6.1.9007
Loading R profile...
```

Dependency Management



lockfile.yml

packages:

name: devtools

version: 1.12.0

repo: hadley/devtools

ref: v1.12.0

name: checkr

version: 0.1.4

repo: syberia/checkr

name: s3mpi

version: 0.2.40

repo: robertzk/s3mpi

name: objectdiff

version: 0.2.3.9003

repo: robertzk/objectdiff

name: stagerunner

version: 0.5.6

repo: syberia/stagerunner

name: Ramd
version: 0.3.8

repo: robertzk/Ramd

name: statsUtils
version: 0.1.4

repo: robertzk/statsUtils

name: mungebits2
version: 0.1.0.9014

repo: syberia/mungebits2

name: syberiaMungebits2

version: 0.1.2.9002

repo: syberia/syberiaMungebits2

name: director

version: 0.3.0.5.9000 repo: syberia/director

name: tundra

version: 0.3.0.9000 repo: syberia/tundra

name: syberia

version: 0.6.1.9009 repo: syberia/syberia

ref: 0.6.1.9009

Arbitrarily Large Projects

Scales to large teams of contributors working on **thousands of R models**



Currently 2 main engines are open sourced:

- Modeling engine
- Base engine (dependency of modeling engine)

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 Define a backtesting engine & grammar (built on top of base engine)

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- Use it to visualize + test strategies in R

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- Define resources to transpile R -> VHDL

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- Modeling engine
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Using Syberia for HFT automation:

- Define a backtesting engine & grammar (built on top of base engine)
- Use it to visualize + test strategies in R
- Define resources to transpile R -> VHDL
- Push to FPGA arrays and trade realtime

Shoutouts!

A huge **thank you** to all Avantees that contributed to making Syberia happen:

- Abel Castillo
- David Feldman
- Jason French
- Kirill Sevastyanenko
- Peter Hurford
- Tong Lu

syberia.io

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