## Package 'bundle'

November 12, 2024

Title Serialize Model Objects with a Consistent Interface

Version 0.1.2

**Description** Typically, models in 'R' exist in memory and can be saved via regular 'R' serialization. However, some models store information in locations that cannot be saved using 'R' serialization alone. The goal of 'bundle' is to provide a common interface to capture this information, situate it within a portable object, and restore it for use in new settings.

License MIT + file LICENSE

```
URL https://github.com/rstudio/bundle,
    https://rstudio.github.io/bundle/
```

BugReports https://github.com/rstudio/bundle/issues

**Depends** R (>= 3.6)

Imports glue, purrr, rlang, utils, withr

**Suggests** bonsai, butcher, callr, caret, covr, dbarts, embed, h2o, keras, kernlab, knitr, luz, MASS, modeldata, parsnip, recipes, renv, rmarkdown, stacks, tensorflow, testthat (>= 3.0.0), torch, torchvision, uwot, vetiver, workflows, xgboost (>= 1.6.0.1)

VignetteBuilder knitr

Config/Needs/website tidyverse/tidytemplate

Config/testthat/edition 3

**Encoding** UTF-8

RoxygenNote 7.3.2

NeedsCompilation no

Author Julia Silge [aut, cre] (<a href="https://orcid.org/0000-0002-3671-836X">https://orcid.org/0000-0002-3671-836X</a>), Simon Couch [aut], Qiushi Yan [aut],

Max Kuhn [aut],

Posit Software, PBC [cph, fnd]

2 bundle

Maintainer Julia Silge <julia.silge@posit.co>

**Repository** CRAN

**Date/Publication** 2024-11-12 10:30:02 UTC

## **Contents**

bund1	Bundling	
Index		27
	sandio_description	_`
	bundle_description	
	bundle.xgb.Booster	24
	bundle.workflow	22
	bundle.train	20
	bundle.step_umap	18
	bundle.recipe	
	bundle.model_stack	
	bundle.model_fit	
	bundle.luz_module_fitted	
	bundle.keras.engine.training.Model	
	bundle.H2OAutoML	
	bundle.bart	
	bundle	2

## Description

bundle() methods provide a consistent interface to serialization methods for statistical model objects. The created bundle can be saved, then re-loaded and unbundle()d in a new R session for use in prediction.

## Usage

```
bundle(x, ...)
unbundle(x)
```

## Arguments

x A model object to bundle.

. . . Additional arguments to bundle methods.

bundle.bart 3

#### Value

A bundle object with subclass referencing the modeling function. If a bundle method is not defined for the supplied object, bundle.default is the identity function.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

#### See Also

```
Other bundles: bundle.H2OAutoML(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.model_stack(), bundle.recipe(), bundle.step_umap(), bundle.train(), bundle.workflow(), bundle.xgb.Booster()
```

bundle.bart

Bundle a bart object

## Description

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

4 bundle.bart

#### Usage

```
## S3 method for class 'bart'
bundle(x, ...)
```

#### Arguments

x A bart object returned from dbarts::bart(). Notably, this ought not to be the

output of parsnip::bart().

... Not used in this bundler and included for compatibility with the generic only.

Additional arguments passed to this method will return an error.

#### Value

A bundle object with subclass bundled\_bart.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

#### bundle and butcher

The butcher package allows you to remove parts of a fitted model object that are not needed for prediction.

This bundle method is compatible with pre-butchering. That is, for a fitted model x, you can safely call:

bundle.H2OAutoML 5

```
res <-
  x %>%
  butcher() %>%
  bundle()
```

and predict with the output of unbundle(res) in a new R session.

#### See Also

```
Other bundlers: bundle(), bundle.H2OAutoML(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.model_stack(), bundle.recipe(), bundle.step_umap(), bundle.train(), bundle.workflow(), bundle.xgb.Booster()
```

## **Examples**

```
# fit model and bundle ------
library(dbarts)

mtcars$vs <- as.factor(mtcars$vs)

set.seed(1)
fit <- dbarts::bart(mtcars[c("disp", "hp")], mtcars$vs, keeptrees = TRUE)

fit_bundle <- bundle(fit)

# then, after saveRDS + readRDS or passing to a new session -------
fit_unbundled <- unbundle(fit_bundle)

fit_unbundled_preds <- predict(fit_unbundled, mtcars)</pre>
```

bundle.H2OAutoML

Bundle an h2o object

## **Description**

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

## Usage

```
## $3 method for class 'H2OAutoML'
bundle(x, id = NULL, n = NULL, ...)
## $3 method for class 'H2OMultinomialModel'
bundle(x, ...)
## $3 method for class 'H2OBinomialModel'
```

6 bundle.H2OAutoML

```
bundle(x, ...)
## S3 method for class 'H2ORegressionModel'
bundle(x, ...)
```

#### **Arguments**

x An object returned from modeling functions in the h2o package.

id A single character. The model\_id entry in the leaderboard. Applies to AutoML

output only. Supply only one of this argument or n.

n An integer giving the position in the leaderboard of the model to bundle. Applies

to AutoML output only. Will be ignored if id is supplied.

.. Not used in this bundler and included for compatibility with the generic only.

Additional arguments passed to this method will return an error.

#### Value

A bundle object with subclass bundled\_h2o.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

#### See Also

```
These methods wrap h2o::h2o.save_mojo() and h2o::h2o.saveModel().

Other bundlers: bundle(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.model_stack(), bundle.recipe(), bundle.step_umap(), bundle.train(), bundle.workflow(), bundle.xgb.Booster()
```

## **Examples**

```
# fit model and bundle ------
library(h2o)
set.seed(1)
h2o.init()
cars_h2o <- as.h2o(mtcars)</pre>
cars_fit <-
 h2o.glm(
   x = colnames(cars_h2o)[2:11],
   y = colnames(cars_h2o)[1],
   training_frame = cars_h2o
 )
cars_bundle <- bundle(cars_fit)</pre>
# then, after saveRDS + readRDS or passing to a new session -----
cars_unbundled <- unbundle(cars_fit)</pre>
predict(cars_unbundled, cars_h2o[, 2:11])
h2o.shutdown(prompt = FALSE)
```

```
{\it bundle.keras.engine.training.Model} \\ {\it Bundle~a~keras~object}
```

## Description

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

#### Usage

```
## S3 method for class 'keras.engine.training.Model' bundle(x, ...)
```

#### **Arguments**

x An object returned from modeling functions in the keras package.

... Not used in this bundler and included for compatibility with the generic only.

Additional arguments passed to this method will return an error.

#### **Details**

This bundler does not currently support custom keras extensions, such as use of a keras::new\_layer\_class() or custom metric function. In such situations, consider using keras::with\_custom\_object\_scope().

#### Value

A bundle object with subclass bundled\_keras.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

## See Also

```
This method wraps keras::save_model_tf() and keras::load_model_tf().

Other bundlers: bundle(), bundle.H2OAutoML(), bundle.bart(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.model_stack(), bundle.recipe(), bundle.step_umap(), bundle.train(), bundle.workflow(), bundle.xgb.Booster()
```

```
# fit model and bundle ------
library(keras)
set.seed(1)
mnist <- dataset_mnist()</pre>
x_train <- mnist$train$x</pre>
y_train <- mnist$train$y</pre>
x_test <- mnist$test$x</pre>
y_test <- mnist$test$y</pre>
x_train <- array_reshape(x_train, c(nrow(x_train), 784))</pre>
x_test <- array_reshape(x_test, c(nrow(x_test), 784))</pre>
x_{train} <- x_{train} / 255
x_{\text{test}} <- x_{\text{test}} / 255
y_train <- to_categorical(y_train, 10)</pre>
y_test <- to_categorical(y_test, 10)</pre>
mod <- keras_model_sequential()</pre>
mod %>%
  layer_dense(units = 128, activation = 'relu', input_shape = c(784)) %>%
  layer_dropout(rate = 0.4) %>%
  layer_dense(units = 64, activation = 'relu') %>%
  layer_dropout(rate = 0.3) %>%
  layer_dense(units = 10, activation = 'softmax')
mod %>% compile(
  loss = 'categorical_crossentropy',
  optimizer = optimizer_rmsprop(),
  metrics = c('accuracy')
)
mod %>% fit(
  x_train, y_train,
  epochs = 5, batch_size = 128,
  validation_split = 0.2,
  verbose = 0
)
mod_bundle <- bundle(mod)</pre>
# then, after saveRDS + readRDS or passing to a new session ------
mod_unbundled <- unbundle(mod_bundle)</pre>
predict(mod_unbundled, x_test)
```

```
bundle.luz_module_fitted 
 Bundle \; a \; \texttt{luz_module_fitted} \; object
```

## **Description**

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

## Usage

```
## S3 method for class 'luz_module_fitted'
bundle(x, ...)
```

## **Arguments**

A luz\_module\_fitted object returned from luz::fit.luz\_module\_generator().
 ... Not used in this bundler and included for compatibility with the generic only.
 Additional arguments passed to this method will return an error.

#### **Details**

For now, bundling methods for torch are only available via the luz package, "a higher level API for torch providing abstractions to allow for much less verbose training loops."

#### Value

A bundle object with subclass bundled\_luz\_module\_fitted.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict()

on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

#### See Also

```
This method wraps luz::luz_save() and luz::luz_load().

Other bundlers: bundle(), bundle.H2OAutoML(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.model_fit(), bundle.model_stack(), bundle.recipe(), bundle.step_umap(), bundle.train(), bundle.workflow(), bundle.xgb.Booster()
```

```
if (torch::torch_is_installed()) {
# fit model and bundle -----
library(torch)
library(torchvision)
library(luz)
set.seed(1)
# example adapted from luz pkgdown article "Autoencoder"
dir <- tempdir()</pre>
mnist_dataset2 <- torch::dataset(</pre>
  inherit = mnist_dataset,
  .getitem = function(i) {
    output <- super$.getitem(i)</pre>
    output$v <- output$x
    output
  }
)
train_ds <- mnist_dataset2(</pre>
  download = TRUE,
  transform = transform_to_tensor
)
test_ds <- mnist_dataset2(</pre>
  train = FALSE,
  transform = transform_to_tensor
)
train_dl <- dataloader(train_ds, batch_size = 128, shuffle = TRUE)</pre>
```

bundle.model\_fit

```
test_dl <- dataloader(test_ds, batch_size = 128)</pre>
net <- nn_module(</pre>
  "Net",
  initialize = function() {
    self$encoder <- nn_sequential(</pre>
      nn_conv2d(1, 6, kernel_size=5),
      nn_relu(),
      nn_conv2d(6, 16, kernel_size=5),
      nn_relu()
    )
    self$decoder <- nn_sequential(</pre>
      nn_conv_transpose2d(16, 6, kernel_size = 5),
      nn_relu(),
      nn_conv_transpose2d(6, 1, kernel_size = 5),
      nn_sigmoid()
    )
  },
  forward = function(x) {
    x %>%
      self$encoder() %>%
      self$decoder()
  },
  predict = function(x) {
    self$encoder(x) %>%
      torch_flatten(start_dim = 2)
  }
)
mod <- net %>%
  setup(
    loss = nn_mse_loss(),
    optimizer = optim_adam
  ) %>%
  fit(train_dl, epochs = 1, valid_data = test_dl)
mod_bundle <- bundle(mod)</pre>
# then, after saveRDS + readRDS or passing to a new session -----
mod_unbundled <- unbundle(mod_bundle)</pre>
mod_unbundled_preds <- predict(mod_unbundled, test_dl)</pre>
}
```

bundle.model\_fit 13

#### **Description**

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

#### Usage

```
## S3 method for class 'model_fit'
bundle(x, ...)
```

## **Arguments**

x A model\_fit object returned from parsnip or other tidymodels packages.

... Not used in this bundler and included for compatibility with the generic only. Additional arguments passed to this method will return an error.

#### **Details**

Primarily, these methods call bundle() on the output of parsnip::extract\_fit\_engine(). See the class of the output of that function for more details on the bundling method for that object.

#### Value

A bundle object with subclass bundled\_model\_fit.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

14 bundle.model\_fit

See vignette("bundle") for more information on bundling and its motivation.

#### bundle and butcher

The butcher package allows you to remove parts of a fitted model object that are not needed for prediction.

This bundle method is compatible with pre-butchering. That is, for a fitted model x, you can safely call:

```
res <-
  x %>%
  butcher() %>%
  bundle()
```

and predict with the output of unbundle(res) in a new R session.

#### See Also

```
Other bundles: bundle(), bundle.H2OAutoML(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_stack(), bundle.recipe(), bundle.step_umap(), bundle.train(), bundle.workflow(), bundle.xgb.Booster()
```

```
# fit model and bundle -----
library(parsnip)
library(xgboost)

set.seed(1)

mod <-
    boost_tree(trees = 5, mtry = 3) %>%
    set_mode("regression") %>%
    set_engine("xgboost") %>%
    fit(mpg ~ ., data = mtcars)

mod_bundle <- bundle(mod)

# then, after saveRDS + readRDS or passing to a new session ------
mod_unbundled <- unbundle(mod_bundle)

mod_unbundled_preds <- predict(mod_unbundled, new_data = mtcars)</pre>
```

bundle.model\_stack 15

bundle.model\_stack

Bundle a tidymodels model\_stack object

## **Description**

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

## Usage

```
## S3 method for class 'model_stack'
bundle(x, ...)
```

#### Arguments

x A model\_stack object returned from fit\_members().

Not used in this bundler and included for compatibility with the generic only. Additional arguments passed to this method will return an error.

#### **Details**

This bundler wraps bundle.model\_fit() and bundle.workflow(). Both the fitted members and the meta-learner (in x\$coefs) are bundled.

## Value

A bundle object with subclass bundled\_model\_stack.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

16 bundle.recipe

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

#### See Also

```
Other bundles: bundle(), bundle.H2OAutoML(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.recipe(), bundle.step_umap(), bundle.train(), bundle.workflow(), bundle.xgb.Booster()
```

## **Examples**

```
# fit model and bundle -----
library(stacks)

set.seed(1)

mod <-
    stacks() %>%
    add_candidates(reg_res_lr) %>%
    add_candidates(reg_res_svm) %>%
    blend_predictions(times = 10) %>%
    fit_members()

mod_bundle <- bundle(mod)

# then, after saveRDS + readRDS or passing to a new session ------
mod_unbundled <- unbundle(mod_bundle)</pre>
```

bundle.recipe

Bundle a recipe object

## **Description**

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

## Usage

```
## S3 method for class 'recipe' bundle(x, ...)
```

bundle.recipe 17

## Arguments

x A recipe object returned from recipes.

... Not used in this bundler and included for compatibility with the generic only.

Additional arguments passed to this method will return an error.

#### **Details**

The method call bundle() on every step in the recipe object. See the classes of individual steps for more details on the bundling method for that object.

#### Value

A bundle object with subclass bundled\_recipe.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

## See Also

```
Other bundless: bundle(), bundle.H2OAutoML(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.model_stack(), bundle.step_umap(), bundle.train(), bundle.workflow(), bundle.xgb.Booster()
```

18 bundle.step\_umap

bundle.step\_umap

Bundle a step\_umap object

## **Description**

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

## Usage

```
## S3 method for class 'step_umap'
bundle(x, ...)
```

## Arguments

x A step\_umap object returned from embed.

Not used in this bundler and included for compatibility with the generic only. Additional arguments passed to this method will return an error.

#### Value

A bundle object with subclass bundled\_step\_umap.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle

bundle.step\_umap 19

package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

#### bundle and butcher

The butcher package allows you to remove parts of a fitted model object that are not needed for prediction.

This bundle method is compatible with pre-butchering. That is, for a fitted model x, you can safely call:

```
res <-
  x %>%
  butcher() %>%
  bundle()
```

and predict with the output of unbundle (res) in a new R session.

#### See Also

```
This method wraps uwot::save_uwot() and uwot::load_uwot().

Other bundlers: bundle(), bundle.H2OAutoML(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.model_stack(), bundle.recipe(), bundle.train(), bundle.workflow(), bundle.xgb.Booster()
```

```
# fit model and bundle ------
library(recipes)
library(embed)

set.seed(1)

rec <- recipe(Species ~ ., data = iris) %>%
    step_normalize(all_predictors()) %>%
    step_umap(all_predictors(), outcome = vars(Species), num_comp = 2) %>%
    prep()

rec_bundle <- bundle(rec)

# then, after saveRDS + readRDS or passing to a new session ------
rec_unbundled <- unbundle(rec_bundle)

bake(rec_unbundled, new_data = iris)</pre>
```

20 bundle.train

bundle.train

Bundle a caret train object

## **Description**

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

## Usage

```
## S3 method for class 'train' bundle(x, ...)
```

#### Arguments

x A train object returned from caret::train().

... Not used in this bundler and included for compatibility with the generic only. Additional arguments passed to this method will return an error.

#### **Details**

Primarily, these methods call bundle() on the output of train\_model\_object\$finalModel. See the class of the output of that slot for more details on the bundling method for that object.

## Value

A bundle object with subclass bundled\_train.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

bundle.train 21

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

#### bundle and butcher

The butcher package allows you to remove parts of a fitted model object that are not needed for prediction.

This bundle method is compatible with pre-butchering. That is, for a fitted model x, you can safely call:

```
res <-
   x %>%
   butcher() %>%
   bundle()
```

and predict with the output of unbundle(res) in a new R session.

## See Also

```
Other bundle: bundle(), bundle.H2OAutoML(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.model_stack(), bundle.recipe(), bundle.step_umap(), bundle.workflow(), bundle.xgb.Booster()
```

```
# fit model and bundle ------
library(caret)

predictors <- mtcars[, c("cyl", "disp", "hp")]

set.seed(1)

mod <-
    train(
        x = predictors,
        y = mtcars$mpg,
        method = "glm"
    )

mod_bundle <- bundle(mod)

# then, after saveRDS + readRDS or passing to a new session ------
mod_unbundled <- unbundle(mod_bundle)

mod_unbundled_preds <- predict(mod_unbundled, new_data = mtcars)</pre>
```

22 bundle.workflow

bundle.workflow

Bundle a tidymodels workflow object

#### **Description**

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

## Usage

```
## S3 method for class 'workflow'
bundle(x, ...)
```

## **Arguments**

x A workflow object returned from workflows or other tidymodels packages.

Not used in this bundler and included for compatibility with the generic only. Additional arguments passed to this method will return an error.

#### **Details**

This bundler wraps bundle.model\_fit() and bundle.recipe().

#### Value

A bundle object with subclass bundled\_workflow.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

bundle.workflow 23

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

#### bundle and butcher

The butcher package allows you to remove parts of a fitted model object that are not needed for prediction.

This bundle method is compatible with pre-butchering. That is, for a fitted model x, you can safely call:

```
res <-
  x %>%
  butcher() %>%
  bundle()
```

and predict with the output of unbundle(res) in a new R session.

#### See Also

```
Other bundles: bundle(), bundle.H2OAutoML(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.model_stack(), bundle.recipe(), bundle.step_umap(), bundle.train(), bundle.xgb.Booster()
```

```
# fit model and bundle -----
library(workflows)
library(recipes)
library(parsnip)
library(xgboost)
set.seed(1)
spec <-
 boost_tree(trees = 5, mtry = 3) %>%
 set_mode("regression") %>%
 set_engine("xgboost")
rec <-
 recipe(mpg ~ ., data = mtcars) %>%
 step_log(hp)
mod <-
 workflow() %>%
 add_model(spec) %>%
 add_recipe(rec) %>%
```

24 bundle.xgb.Booster

```
fit(data = mtcars)
mod_bundle <- bundle(mod)

# then, after saveRDS + readRDS or passing to a new session -----
mod_unbundled <- unbundle(mod_bundle)</pre>
```

bundle.xgb.Booster

Bundle an xgb. Booster object

#### **Description**

Bundling a model prepares it to be saved to a file and later restored for prediction in a new R session. See the 'Value' section for more information on bundles and their usage.

## Usage

```
## S3 method for class 'xgb.Booster'
bundle(x, ...)
```

#### **Arguments**

x An xgb.Booster object returned from xgboost::xgboost() or xgboost::xgb.train().

Not used in this bundler and included for compatibility with the generic only. Additional arguments passed to this method will return an error.

#### Value

A bundle object with subclass bundled\_xgb.Booster.

Bundles are a list subclass with two components:

object An R object. Gives the output of native serialization methods from the model-

supplying package, sometimes with additional classes or attributes that aid porta-

bility. This is often a raw object.

situate A function. The situate() function is defined when bundle() is called, though

is a loose analogue of an unbundle() S3 method for that object. Since the function is defined on bundle(), it has access to references and dependency information that can be saved alongside the object component. Calling unbundle() on a bundled object x calls x\$situate(x\$object), returning the unserialized version of object. situate() will also restore needed references, such as server

instances and environmental variables.

Bundles are R objects that represent a "standalone" version of their analogous model object. Thus, bundles are ready for saving to a file; saving with base::saveRDS() is our recommended serialization strategy for bundles, unless documented otherwise for a specific method.

bundle.xgb.Booster 25

To restore the original model object x in a new environment, load its bundle with base::readRDS() and run unbundle() on it. The output of unbundle() is a model object that is ready to predict() on new data, and other restored functionality (like plotting or summarizing) is supported as a side effect only.

The bundle package wraps native serialization methods from model-supplying packages. Between versions, those model-supplying packages may change their native serialization methods, possibly introducing problems with re-loading objects serialized with previous package versions. The bundle package does not provide checks for these sorts of changes, and ought to be used in conjunction with tooling for managing and monitoring model environments like vetiver or renv.

See vignette("bundle") for more information on bundling and its motivation.

#### bundle and butcher

The butcher package allows you to remove parts of a fitted model object that are not needed for prediction.

This bundle method is compatible with pre-butchering. That is, for a fitted model x, you can safely call:

```
res <-
   x %>%
  butcher() %>%
  bundle()
```

and predict with the output of unbundle(res) in a new R session.

#### See Also

```
This method adapts the xgboost functions xgboost::xgb.save.raw() and xgboost::xgb.load.raw().

Other bundlers: bundle(), bundle.H2OAutoML(), bundle.bart(), bundle.keras.engine.training.Model(), bundle.luz_module_fitted(), bundle.model_fit(), bundle.model_stack(), bundle.recipe(), bundle.step_umap(), bundle.train(), bundle.workflow()
```

26 bundle\_description

```
xgb_unbundled <- unbundle(xgb_bundle)
xgb_unbundled_preds <- predict(xgb_unbundled, agaricus.test$data)</pre>
```

 $bundle\_description$ 

bundle: Serialize Model Objects With A Consistent Interface

## **Description**

Typically, models in 'R' exist in memory and can be saved via regular 'R' serialization. However, some models store information in locations that cannot be saved using 'R' serialization alone. The goal of 'bundle' is to provide a common interface to capture this information, situate it within a portable object, and restore it for use in new settings.

## Author(s)

**Maintainer**: Julia Silge <julia.silge@posit.co> (ORCID)

Authors:

- Simon Couch <simonpatrickcouch@gmail.com>
- Qiushi Yan <qiushi.yann@gmail.com>
- Max Kuhn <max@posit.co>

Other contributors:

• Posit Software, PBC [copyright holder, funder]

## See Also

Useful links:

- https://github.com/rstudio/bundle
- https://rstudio.github.io/bundle/
- Report bugs at https://github.com/rstudio/bundle/issues

# Index

* bundlers bundle, 2	bundle.model_stack, 3, 5, 7, 8, 11, 14, 15, 17, 19, 21, 23, 25
bundle.bart, 3	bundle.recipe, 3, 5, 7, 8, 11, 14, 16, 16, 19,
bundle.H2OAutoML, 5	21, 23, 25
bundle.keras.engine.training.Model,	bundle.recipe(), 22
7	bundle.step_umap, 3, 5, 7, 8, 11, 14, 16, 17,
bundle.luz_module_fitted, 10	18, 21, 23, 25
bundle.model_fit, 12	bundle.train, 3, 5, 7, 8, 11, 14, 16, 17, 19,
bundle.model_stack, 15	20, 23, 25
bundle.recipe, 16	bundle.workflow, 3, 5, 7, 8, 11, 14, 16, 17,
bundle.step_umap, 18	19, 21, 22, 25
bundle.train, 20	bundle.workflow(), 15
bundle.workflow, 22	bundle.xgb.Booster, 3, 5, 7, 8, 11, 14, 16,
bundle.xgb.Booster, 24	17, 19, 21, 23, 24
,	bundle_description, 26
base::readRDS(), 3, 4, 6, 8, 10, 13, 15, 17,	<pre>bundle_model_fit (bundle.model_fit), 12</pre>
18, 20, 22, 25	bundle_model_stack
base::saveRDS(), 3, 4, 6, 8, 10, 13, 15, 17,	(bundle.model_stack), 15
18, 20, 22, 24	<pre>bundle_recipe (bundle.recipe), 16</pre>
bundle, 2, 5, 7, 8, 11, 14, 16, 17, 19, 21, 23, 25	<pre>bundle_step_umap (bundle.step_umap), 18</pre>
bundle(), 3, 4, 6, 8, 10, 13, 15, 17, 18, 20, 22,	bundle_train(bundle.train), 20
24	<pre>bundle_workflow(bundle.workflow), 22</pre>
bundle-package (bundle_description), 26	
bundle.bart, 3, 3, 7, 8, 11, 14, 16, 17, 19, 21, 23, 25	<pre>caret::train(), 20</pre>
bundle.H2OAutoML, 3, 5, 5, 8, 11, 14, 16, 17, 19, 21, 23, 25	dbarts::bart(),4
bundle.H2OBinomialModel	embed, 18
(bundle.H2OAutoML), 5	Chibed, 10
bundle.H2OMultinomialModel	<pre>fit_members(), 15</pre>
(bundle.H2OAutoML), 5	
bundle.H2ORegressionModel	h2o, 6
(bundle.H2OAutoML), 5	h2o::h2o.save_mojo(), 7
bundle.keras.engine.training.Model, 3,	h2o::h2o.saveModel(), 7
5, 7, 7, 11, 14, 16, 17, 19, 21, 23, 25	
bundle.luz_module_fitted, 3, 5, 7, 8, 10,	keras, 8
14, 16, 17, 19, 21, 23, 25	<pre>keras::load_model_tf(), 8</pre>
bundle.model_fit, 3, 5, 7, 8, 11, 12, 16, 17,	keras::new_layer_class(),8
19, 21, 23, 25	keras::save_model_tf(),8
<pre>bundle.model_fit(), 15, 22</pre>	${\tt keras::with\_custom\_object\_scope(), 8}$

28 INDEX

```
luz::fit.luz_module_generator(), 10
luz::luz_load(), 11
luz::luz_save(), 11
model_fit, 13
model_stack, 15
parsnip, 13
parsnip::bart(),4
parsnip::extract_fit_engine(), 13
predict(), 3, 4, 6, 8, 10, 13, 15, 17, 18, 20,
         22, 25
raw, 3, 4, 6, 8, 10, 13, 15, 17, 18, 20, 22, 24
recipe, 17
recipes, 17
renv, 3, 4, 6, 8, 11, 13, 16, 17, 19, 21, 23, 25
step_umap, 18
train, 20
unbundle (bundle), 2
unbundle(), 3, 4, 6, 8, 10, 13, 15, 17, 18, 20,
         22, 24, 25
uwot::load_uwot(), 19
uwot::save_uwot(), 19
vetiver, 3, 4, 6, 8, 11, 13, 16, 17, 19, 21, 23,
         25
workflow, 22
workflows, 22
xgboost::xgb.load.raw(), 25
xgboost::xgb.save.raw(), 25
xgboost::xgb.train(), 24
xgboost::xgboost(), 24
```