Package 'important'

August 20, 2025

Title Supervised Feature Selection

```
Version 0.0.1
Description Interfaces for choosing important predictors in supervised
      regression, classification, and censored regression models. Permuted
      importance scores (Biecek and Burzykowski (2021)
      <doi:10.1201/9780429027192>) can be computed for 'tidymodels' model
      fits.
License MIT + file LICENSE
URL https://important.tidymodels.org/,
      https://github.com/tidymodels/important
BugReports https://github.com/tidymodels/important/issues
Depends R (>= 4.1.0)
Imports cli, dplyr, generics, ggplot2, hardhat (>= 1.4.1), purrr,
      rlang, tibble, tidyr, tune, vctrs, withr, workflows
Suggests censored, future, future.apply, mirai, modeldata, parsnip,
      recipes, spelling, survival, testthat (>= 3.0.0), yardstick
Config/Needs/website tidyverse/tidytemplate, tidymodels
Config/testthat/edition 3
Config/usethis/last-upkeep 2025-06-09
Encoding UTF-8
Language en-US
RoxygenNote 7.3.2
NeedsCompilation no
Author Max Kuhn [aut, cre] (ORCID: <a href="https://orcid.org/0000-0003-2402-136X">https://orcid.org/0000-0003-2402-136X</a>),
      Posit Software, PBC [cph, fnd] (ROR: <a href="https://ror.org/03wc8by49">https://ror.org/03wc8by49</a>)
Maintainer Max Kuhn <max@posit.co>
Repository CRAN
Date/Publication 2025-08-20 17:10:10 UTC
```

Contents

Description

Visualize importance scores

Usage

```
## $3 method for class 'importance_perm'
autoplot(
  object,
  top = Inf,
  metric = NULL,
  eval_time = NULL,
  type = "importance",
  std_errs = stats::qnorm(0.95),
  ...
)
```

Arguments

object	A tibble of results from importance_perm().
top	An integer for how many terms to show. To define importance when there are multiple metrics, the rankings of predictors are computed across metrics and the average rank is used. In the case of tied rankings, all the ties are included.
metric	A character vector or NULL for which metric to plot. By default, all metrics will be shown via facets. Possible options are the entries in .metric column of the object.
eval_time	For censored regression models, a vector of time points at which the survival probability is estimated.
type	A character value. The default is "importance" which shows the overall signal-to-noise ration (i.e., mean divided by standard error). Alternatively, "difference" shows the mean difference value with standard error bounds.
std_errs	The number of standard errors to plot (when type = "difference").
	Not used.

Value

A ggplot2 object.

Examples

```
# Pre-computed results. See code at
system.file("make_imp_example.R", package = "important")

# Load the results
load(system.file("imp_examples.RData", package = "important"))

# A classification model with two classes and highly correlated predictors.
# To preprocess them, PCA feature extraction is used.
#
# Let's first view the importance in terms of the original predictor set
# using 50 permutations:
imp_orig
autoplot(imp_orig, top = 10)

# Now assess the importance in terms of the PCA components
imp_derv
autoplot(imp_derv)
autoplot(imp_derv, metric = "brier_class", type = "difference")
```

importance_perm

Compute permutation-based predictor importance

Description

importance_perm() computes model-agnostic variable importance scores by permuting individual predictors (one at a time) and measuring how worse model performance becomes.

Usage

```
importance_perm(
  wflow,
  data,
  metrics = NULL,
  type = "original",
  size = 500,
  times = 10,
  eval_time = NULL,
  event_level = "first"
)
```

Arguments

wflow	A fitted workflows::workflow().
data	A data frame of the data passed to workflows::fit.workflow(), including the outcome and case weights (if any).
metrics	A yardstick::metric_set() or NULL.
type	A character string for which <i>level</i> of predictors to compute. A value of "original" (default) will return values in the same representation of data. Using "derived" will compute them for any derived features/predictors, such as dummy indicator columns, etc.
size	How many data points to predict for each permutation iteration.
times	How many iterations to repeat the calculations.
eval_time	For censored regression models, a vector of time points at which the survival probability is estimated. This is only needed if a dynamic metric is used, such as the Brier score or the area under the ROC curve.
event_level	A single string. Either "first" or "second" to specify which level of truth to consider as the "event". This argument is only applicable when estimator = "binary".

Details

The function can compute importance at two different levels.

- The "original" predictors are the unaltered columns in the source data set. For example, for a categorical predictor used with linear regression, the original predictor is the factor column.
- "Derived" predictors are the final versions given to the model. For the categorical predictor example, the derived versions are the binary indicator variables produced from the factor version.

This can make a difference when pre-processing/feature engineering is used. This can help us understand *how* a predictor can be important

Importance scores are computed for each predictor (at the specified level) and each performance metric. If no metric is specified, defaults are used:

- Classification: yardstick::brier_class(), yardstick::roc_auc(), and yardstick::accuracy().
- Regression: yardstick::rmse() and yardstick::rsq().
- Censored regression: yardstick::brier_survival()

For censored data, importance is computed for each evaluation time (when a dynamic metric is specified).

By default, no parallelism is used to process models in **tune**; you have to opt-in.

Using future to parallel process:

You should install the package and choose your flavor of parallelism using the plan function. This allows you to specify the number of worker processes and the specific technology to use. For example, you can use:

```
library(future)
plan(multisession, workers = 4)
```

and work will be conducted simultaneously (unless there is an exception; see the section below). See future::plan() for possible options other than multisession.

Using mirai to parallel process:

To configure parallel processing with **mirai**, use the mirai::daemons() function. The first argument, n, determines the number of parallel workers. Using daemons(0) reverts to sequential processing.

The arguments url and remote are used to set up and launch parallel processes over the network for distributed computing. See mirai::daemons() documentation for more details.

Value

A tibble with extra classes "importance_perm" and either "original_importance_perm" or "derived_importance_perm" The columns are:

- .metric the name of the performance metric:
- predictor: the predictor
- n: the number of usable results (should be the same as times)
- mean: the average of the differences in performance. For each metric, larger values indicate worse performance (i.e., higher importance).
- std_err: the standard error of the differences.
- importance: the mean divided by the standard error.
- For censored regression models, an additional .eval_time column may also be included (depending on the metric requested).

Examples

```
if (rlang::is_installed(c("modeldata", "recipes", "workflows", "parsnip"))) {
 library(modeldata)
 library(recipes)
 library(workflows)
 library(dplyr)
 library(parsnip)
 set.seed(12)
 dat_tr <-
   sim_logistic(250, ~.1 + 2 * A - 3 * B + 1 * A *B, corr = .7) |>
   dplyr::bind_cols(sim_noise(250, num_vars = 10))
 rec <-
   recipe(class ~ ., data = dat_tr) |>
   step_interact(~ A:B) |>
   step_normalize(all_numeric_predictors()) |>
   step_pca(contains("noise"), num_comp = 5)
 lr_wflow <- workflow(rec, logistic_reg())</pre>
```

Index

```
autoplot.importance_perm, 2
future::plan(), 5
importance_perm, 3
importance_perm(), 2, 3
mirai::daemons(), 5
plan, 4
workflows::fit.workflow(), 4
workflows::workflow(), 4
yardstick::accuracy(), 4
yardstick::brier_class(), 4
yardstick::brier_survival(), 4
yardstick::metric_set(), 4
yardstick::rmse(), 4
yardstick::roc_auc(), 4
yardstick::rsq(), 4
```