

Package ‘stabs’

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Title Stability Selection with Error Control

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Description Resampling procedures to assess the stability of selected variables with additional finite sample error control for high-dimensional variable selection procedures such as Lasso or boosting. Both, standard stability selection (Meinshausen & Bühlmann, 2010, [doi:10.1111/j.1467-9868.2010.00740.x](https://doi.org/10.1111/j.1467-9868.2010.00740.x)) and complementary pairs stability selection with improved error bounds (Shah & Samworth, 2013, [doi:10.1111/j.1467-9868.2011.01034.x](https://doi.org/10.1111/j.1467-9868.2011.01034.x)) are implemented. The package can be combined with arbitrary user specified variable selection approaches.

VignetteBuilder knitr

Depends R (>= 2.14.0), methods, stats, parallel

Imports graphics, grDevices, utils

Suggests glmnet, lars, mboost (> 2.3-0), gamboostLSS (>= 1.2-0), TH.data, hdi, testthat, knitr, rmarkdown

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URL <https://github.com/hofnerb/stabs>

BugReports <https://github.com/hofnerb/stabs/issues>

NeedsCompilation no

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check_folds	<i>Check if folds result from subsampling with p = 0.5</i>
-------------	--

Description

(Internal) function that checks if folds result from subsampling with $p = 0.5$ and adds complementary pairs if needed.

Usage

```
check_folds(folds, B, n, sampling.type)
```

Arguments

folds	a weight matrix that represents the subsamples.
B	number of subsampling replicates.
n	the number of observations; needed for internal checks.
sampling.type	sampling type to be used.

Details

This is an internal function used to check if `folds` are specified correctly. For details (e.g. on arguments) see [stabsel](#).

Value

A matrix containing the `folds`, possibly after adding the complementary pairs.

References

B. Hofner, L. Boccuto and M. Goeker (2015), Controlling false discoveries in high-dimensional situations: Boosting with stability selection. *BMC Bioinformatics*, 16:144.
[doi:10.1186/s1285901505753](https://doi.org/10.1186/s1285901505753).

See Also

For details see [stabsel](#).

Fitting Functions

Fit Functions for Stability Selection

Description

Functions that fit a model until q variables are selected and that returns the indices (and names) of the selected variables.

Usage

```
## package lars:
lars.lasso(x, y, q, ...)
lars.stepwise(x, y, q, ...)

## package glmnet:
glmnet.lasso(x, y, q, type = c("conservative", "anticonservative"), ...)
glmnet.lasso_maxCoef(x, y, q, ...)
```

Arguments

- `x` a matrix containing the predictors or an object of class "mboost".
- `y` a vector or matrix containing the outcome.
- `q` number of (unique) selected variables (or groups of variables depending on the model) that are selected on each subsample.
- `type` a character vector specifying if the number of selected variables per subsample is $\leq q$ (`type = "conservative"`) or $\geq q$ (`type = "anticonservative"`). The conservative version *ensures* that the PFER is controlled.
- `...` additional arguments passed to the underlying fitting function. See the example on `glmnet.lasso_maxCoef` in [stabsel](#) for the specification of additional arguments via [stabsel](#).

Details

All fitting functions are named after the package and the type of model that is fitted: `package_name.model`, e.g., `glmnet.lasso` stands for a lasso model that is fitted using the package `glmnet`.

`glmnet.lasso_maxCoef` fits a lasso model with a given penalty parameter and returns the q variables with the largest (absolute) coefficients. If one wants to use `glmnet.lasso_maxCoef`, one must specify the penalty parameter `lambda` (via the `...` argument) or in [stabsel](#) via `args.fitfun(lambda =)`. Note that usually, the penalty parameter cannot be specified but is chosen such that q variables are selected. For an example on how to use `glmnet.lasso_maxCoef` see [stabsel](#).

Value

A named list with elements

selected	logical. A vector that indicates which variable was selected.
path	logical. A matrix that indicates which variable was selected in which step. Each row represents one variable, the columns represent the steps.

See Also

[stabsel](#) for stability selection itself.

Examples

```
if (require("TH.data")) {
  ## make data set available
  data("bodyfat", package = "TH.data")
} else {
  ## simulate some data if TH.data not available.
  ## Note that results are nonsense with this data.
  bodyfat <- matrix(rnorm(720), nrow = 72, ncol = 10)
}

if (require("lars")) {
  ## selected variables
  lars.lasso(bodyfat[, -2], bodyfat[,2], q = 3)$selected
  lars.stepwise(bodyfat[, -2], bodyfat[,2], q = 3)$selected
}

if (require("glmnet")) {
  glmnet.lasso(bodyfat[, -2], bodyfat[,2], q = 3)$selected
  ## selection path
  glmnet.lasso(bodyfat[, -2], bodyfat[,2], q = 3)$path

  ## Using the anticonservative glmnet.lasso (see args.fitfun):
  stab.glmnet <- stabsel(x = bodyfat[, -2], y = bodyfat[,2],
                         fitfun = glmnet.lasso,
                         args.fitfun = list(type = "anticonservative"),
                         cutoff = 0.75, PFER = 1)

  ## Using glmnet.lasso_maxCoef with penalty parameter specified via args.fitfun:
  stab.maxcoef <- stabsel(x = bodyfat[, -2], y = bodyfat[,2],
                           cutoff = 0.75, PFER = 1, fitfun = glmnet.lasso_maxCoef,
                           args.fitfun = list(lambda = 0.3))
  ## this selectes the q variables with the largest (absolute) effects rather
  ## than the first q variables which are selected in the lasso model
  ## hence leading to a very different selection
}
```

parameters	<i>Method to Extract Parameters</i>
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Description

Extract stability selection parameters, i.e., tuning parameters, from a `stabsel` object.

Usage

```
parameters(object)

## extract parameters from a stabsel model
## (same as parameters(p) )
## S3 method for class 'stabsel'
stabsel_parameters(p, ...)
```

Arguments

object	an object of class "stabsel" or "stabsel_parameters".
p	an object of class "stabsel".
...	additional arguments, currently not used.

Value

An object of class `stabsel_parameters` with a special `print` method. See there for details.

See Also

`stabsel` to run stability selection and `stabsel_parameters` for details on the parameters.

plot.stabsel	<i>Plot and Print Methods for Stability Selection</i>
--------------	---

Description

Display results of stability selection.

Usage

```
## S3 method for class 'stabsel'
plot(x, main = deparse(x$call), type = c("maxsel", "paths"),
      xlab = NULL, ylab = NULL, col = NULL, ymargin = 10, np = sum(x$max > 0),
      labels = NULL, ...)
## S3 method for class 'stabsel'
print(x, decreasing = FALSE, print.all = TRUE, ...)
```

Arguments

<code>x</code>	object of class <code>stabsel</code> .
<code>main</code>	main title for the plot.
<code>type</code>	plot type; either stability paths ("paths") or a plot of the maximum selection frequency ("maxsel").
<code>xlab, ylab</code>	labels for the x- and y-axis of the plot. Per default, sensible labels are used depending on the type of the plot.
<code>col</code>	a vector of colors; Typically, one can specify a single color or one color for each variable. Per default, colors depend on the maximal selection frequency of the variable and range from grey to red.
<code>ymargin</code>	(temporarily) specifies the y margin of of the plot in lines (see argument "mar" of function <code>par</code>). This only affects the right margin for <code>type = "paths"</code> and the left margin for <code>type = "maxsel"</code> . Explicit user specified margins are kept and are not overwritten.
<code>np</code>	number of variables to plot for the maximum selection frequency plot (<code>type = "maxsel"</code>); the first <code>np</code> variables with highest selection frequency are plotted.
<code>labels</code>	variable labels for the plot; one label per variable / effect must be specified. Per default, the names of <code>x\$max</code> are used.
<code>decreasing</code>	logical. Should the selection frequencies be printed in descending order (TRUE) or in ascending order (FALSE)?
<code>print.all</code>	logical. Should all selection frequencies be displayed or only those that are greater than zero?
<code>...</code>	additional arguments to <code>plot</code> and <code>print</code> functions.

Details

This function implements the stability selection procedure by Meinshausen and Buehlmann (2010) and the improved error bounds by Shah and Samworth (2013).

Two of the three arguments `cutoff`, `q` and PFER *must* be specified. The per-family error rate (PFER), i.e., the expected number of false positives $E(V)$, where V is the number of false positives, is bounded by the argument PFER.

As controlling the PFER is more conservative as controlling the family-wise error rate (FWER), the procedure also controls the FWER, i.e., the probability of selecting at least one non-influential variable (or model component) is less than PFER.

Value

An object of class `stabsel` with a special `print` method. The object has the following elements:

<code>phat</code>	selection probabilities.
<code>selected</code>	elements with maximal selection probability greater <code>cutoff</code> .
<code>max</code>	maximum of selection probabilities.
<code>cutoff</code>	<code>cutoff</code> used.
<code>q</code>	average number of selected variables used.

PFER per-family error rate.
 sampling.type the sampling type used for stability selection.
 assumption the assumptions made on the selection probabilities.
 call the call.

References

B. Hofner, L. Boccuto and M. Goeker (2015), Controlling false discoveries in high-dimensional situations: Boosting with stability selection. *BMC Bioinformatics*, 16:144.
[doi:10.1186/s1285901505753](https://doi.org/10.1186/s1285901505753).

N. Meinshausen and P. Bühlmann (2010), Stability selection. *Journal of the Royal Statistical Society, Series B*, **72**, 417–473.

R.D. Shah and R.J. Samworth (2013), Variable selection with error control: another look at stability selection. *Journal of the Royal Statistical Society, Series B*, **75**, 55–80.

See Also

[stabsel](#)

Examples

```

if (require("TH.data")) {
  ## make data set available
  data("bodyfat", package = "TH.data")
} else {
  ## simulate some data if TH.data not available.
  ## Note that results are nonsense with this data.
  bodyfat <- matrix(rnorm(720), nrow = 72, ncol = 10)
}

## set seed
set.seed(1234)

#####
#### using stability selection with Lasso methods:

if (require("lars")) {
  (stab.lasso <- stabsel(x = bodyfat[, -2], y = bodyfat[, 2],
                         fitfun = lars.lasso, cutoff = 0.75,
                         PFER = 1, mc.cores = 1L))
  ## set mc.cores = 1L only to avoid parallelization on CRAN;
  ## In practice you usually do not need set this value unless you want to
  ## avoid parallelization via mclapply or restrict it to a specific number
  ## of available cores. See documentation of mclapply for more details.

  par(mfrow = c(2, 1))
  plot(stab.lasso, ymargin = 6)
  opar <- par(mai = par("mai") * c(1, 1, 1, 2.7))
  plot(stab.lasso, type = "paths")
}

```

run_stabsel*Run Stability Selection*

Description

(Internal) function that is used to run stability selection (i.e. to apply the fit-function to the subsamples. This function is not intended to be directly called.

Usage

```
run_stabsel(fitter, args.fitter, n, p, cutoff, q, PFER, folds, B, assumption,
            sampling.type, papply, verbose, FWER, eval, names,
            mc.preschedule = FALSE, ...)
```

Arguments

fitter	a function to fit the model on subsamples. See argument <code>fitfun</code> of stabsel for details.
args.fitter	a named list containing additional arguments that are passed to <code>fitter</code> . See argument <code>args.fitfun</code> stabsel for details.
n	the number of observations; needed for internal checks.
p	number of possible predictors (including intercept if applicable).
cutoff	cutoff between 0.5 and 1 or q/p and 1 in case of r-concavity.
q	number of (unique) selected variables (or groups of variables depending on the model) that are selected on each subsample.
PFER	upper bound for the per-family error rate.
folds	a weight matrix that represents the subsamples.
B	number of subsampling replicates.
assumption	distributional assumption.
sampling.type	sampling type to be used.
papply	(parallel) apply function.
verbose	logical (default: TRUE) that determines whether warnings should be issued.
FWER	deprecated. Only for compatibility with older versions, use <code>PFER</code> instead.
eval	logical. Determines whether stability selection is evaluated.
names	variable names that are used to label the results.
mc.preschedule	preschedule tasks?
...	additional arguments to be passed to next function.

Details

This is an internal function that fits the actual models to the subsamples, i.e., this is the work horse that runs stability selection. Usually, one should use [stabsel](#), which internally calls `run_stabsel`. `run_stabsel` can be used by expert users to implement stability selection methods for new model types.

For details (e.g. on arguments) see [stabsel](#).

Value

An object of class `stabsel` with the following elements:

<code>phat</code>	selection probabilities.
<code>selected</code>	elements with maximal selection probability greater cutoff.
<code>max</code>	maximum of selection probabilities.
<code>cutoff</code>	cutoff used.
<code>q</code>	average number of selected variables used.
<code>PFER</code>	per-family error rate.
<code>p</code>	the number of effects subject to selection.
<code>sampling.type</code>	the sampling type used for stability selection.
<code>assumption</code>	the assumptions made on the selection probabilities.

References

B. Hofner, L. Boccuto and M. Goeker (2015), Controlling false discoveries in high-dimensional situations: Boosting with stability selection. *BMC Bioinformatics*, 16:144.
[doi:10.1186/s1285901505753](https://doi.org/10.1186/s1285901505753).

See Also

For details see [stabsel](#).

`selected`

Method to Extract Selected Variables

Description

Extract selected variables from a `stabsel` object.

Usage

```
selected(object, ...)
## S3 method for class 'stabsel'
selected(object, ...)
```

Arguments

- object an object of class "stabsel".
- ... additional arguments passed to specific selected methods.

Details

The ids of variables selected during the stability selection process can be extracted using `selected()`.

stabsel	<i>Stability Selection</i>
---------	----------------------------

Description

Selection of influential variables or model components with error control.

Usage

```
## generic stability selection function
stabsel(x, ...)

## a method to fit models with stability selection
## S3 method for class 'matrix'
stabsel(x, y, fitfun = lars.lasso,
        args.fitfun = list(), cutoff, q, PFER,
        folds = subsample(rep(1, nrow(x)), B = B),
        B = ifelse(sampling.type == "MB", 100, 50),
        assumption = c("unimodal", "r-concave", "none"),
        sampling.type = c("SS", "MB"),
        papply = mclapply, mc.preschedule = FALSE,
        verbose = TRUE, FWER, eval = TRUE, ...)

## essentially a wrapper for data.frames (see details)
## S3 method for class 'data.frame'
stabsel(x, y, intercept = FALSE, ...)
```

Arguments

- x a `matrix` or a `data.frame` containing the predictors.
- y a vector or matrix containing the outcome.
- intercept logical. If x is a `data.frame`, this argument determines if the resulting model matrix should contain a separate intercept or not.
- fitfun a function that takes the arguments x, y as above, and additionally the number of variables to include in each model q. The function then needs to fit the model and to return a logical vector that indicates which variable was selected (among the q selected variables).

args.fitfun	a named list containing additional arguments that are passed to the fitting function; see also argument args in do.call .
cutoff	cutoff between 0.5 and 1. Preferably a value between 0.6 and 0.9 should be used. If assumption = "r-concave" the cutoff can be below 0.5 but needs to be larger than q/p.
q	number of (unique) selected variables (or groups of variables depending on the model) that are selected on each subsample.
PFER	upper bound for the per-family error rate. This specifies the amount of falsely selected base-learners, which is tolerated. See details.
folds	a weight matrix with number of rows equal to the number of observations, see subsample . Usually one should not change the default here as subsampling with a fraction of 1/2 is needed for the error bounds to hold. One usage scenario where specifying the folds by hand might be the case when one has dependent data (e.g. clusters) and thus wants to draw clusters (i.e., multiple rows together) not individuals.
assumption	Defines the type of assumptions on the distributions of the selection probabilities and simultaneous selection probabilities. Only applicable for sampling.type = "SS". Per default, "unimodality" is assumed. For sampling.type = "MB" we always use "none".
sampling.type	use sampling scheme of of Shah & Samworth (2013), i.e., with complementary pairs (sampling.type = "SS"), or the original sampling scheme of Meinshausen & Buehlmann (2010).
B	number of subsampling replicates. Per default, we use 50 complementary pairs for the error bounds of Shah & Samworth (2013) and 100 for the error bound derived in Meinshausen & Buehlmann (2010). As we use B complementary pairs in the former case this leads to $2B$ subsamples.
papply	(parallel) apply function, defaults to mclapply . Alternatively, parLapply can be used. In the latter case, usually more setup is needed (see example of cvrisk for some details).
mc.preschedule	preschedule tasks if papply = mclapply (default: mc.preschedule = FALSE)? For details see mclapply .
verbose	logical (default: TRUE) that determines whether warnings should be issued.
FWER	deprecated. Only for compatibility with older versions, use PFER instead.
eval	logical. Determines whether stability selection is evaluated (eval = TRUE; default) or if only the parameter combination is returned.
...	additional arguments to parallel apply methods such as mclapply .

Details

This function implements the stability selection procedure by Meinshausen and Buehlmann (2010) and the improved error bounds by Shah and Samworth (2013). For details see also Hofner et al. (2014). The error bounds are implemented in the function [stabsel_parameters](#). Two of the three arguments cutoff, q and PFER *must* be specified. The per-family error rate (PFER), i.e., the expected number of false positives $E(V)$, where V is the number of false positives, is bounded by the argument PFER.

As controlling the PFER is more conservative as controlling the family-wise error rate (FWER), the procedure also controls the FWER, i.e., the probability of selecting at least one non-influential variable (or model component) is less than PFER.

Predefined `fitfuns` functions exist but more can be easily implemented. Note that stepwise regression methods are usually not advised as they tend to be relatively unstable. See example below.

The function `stabsel` for `data.frames` is essentially just a wrapper to the `matrix` function with the same arguments. The only difference is that in a pre-processing step, the data set is converted to a model matrix using the function `model.matrix`. The additional argument `intercept` determines if an explicit intercept should be added to the model matrix. This is often not necessary but depends on the `fitfun`.

Value

An object of class `stabsel` with a special `print` method. The object has the following elements:

<code>phat</code>	selection probabilities.
<code>selected</code>	elements with maximal selection probability greater cutoff.
<code>max</code>	maximum of selection probabilities.
<code>cutoff</code>	cutoff used.
<code>q</code>	average number of selected variables used.
<code>PFER</code>	(realized) upper bound for the per-family error rate.
<code>specifiedPFER</code>	specified upper bound for the per-family error rate.
<code>p</code>	the number of effects subject to selection.
<code>B</code>	the number of subsamples.
<code>sampling.type</code>	the sampling type used for stability selection.
<code>assumption</code>	the assumptions made on the selection probabilities.
<code>call</code>	the call.

References

B. Hofner, L. Boccuto and M. Goeker (2015), Controlling false discoveries in high-dimensional situations: Boosting with stability selection. *BMC Bioinformatics*, 16:144.
[doi:10.1186/s1285901505753](https://doi.org/10.1186/s1285901505753).

N. Meinshausen and P. Buehlmann (2010), Stability selection. *Journal of the Royal Statistical Society, Series B*, **72**, 417–473.

R.D. Shah and R.J. Samworth (2013), Variable selection with error control: another look at stability selection. *Journal of the Royal Statistical Society, Series B*, **75**, 55–80.

See Also

`stabsel_parameters` for the computation of error bounds, `stabsel.stabsel` for the fast re-computation of parameters of a fitted `stabsel` object, `fitfun` for available fitting functions and `plot.stabsel` for available plot functions

Examples

```

if (require("TH.data")) {
  ## make data set available
  data("bodyfat", package = "TH.data")
} else {
  ## simulate some data if TH.data not available.
  ## Note that results are nonsense with this data.
  bodyfat <- matrix(rnorm(720), nrow = 72, ncol = 10)
}

## set seed
set.seed(1234)

#####
#### using stability selection with Lasso methods:

if (require("lars")) {
  (stab.lasso <- stabsel(x = bodyfat[, -2], y = bodyfat[, 2],
                        fitfun = lars.lasso, cutoff = 0.75,
                        PFER = 1, , mc.cores = 1L))
  ## set mc.cores = 1L only to avoid parallelization on CRAN;
  ## In practice you usually do not need set this value unless you want to
  ## avoid parallelization via mclapply or restrict it to a specific number
  ## of available cores. See documentation of mclapply for more details.

  (stab.stepwise <- stabsel(x = bodyfat[, -2], y = bodyfat[, 2],
                            fitfun = lars.stepwise, cutoff = 0.75,
                            PFER = 1, , mc.cores = 1L))
  ## set mc.cores = 1L only to avoid parallelization on CRAN;
  ## In practice you usually do not need set this value unless you want to
  ## avoid parallelization via mclapply or restrict it to a specific number
  ## of available cores. See documentation of mclapply for more details.

  par(mfrow = c(2, 1))
  plot(stab.lasso, main = "Lasso")
  plot(stab.stepwise, main = "Stepwise Selection")
  ## --> stepwise selection seems to be quite unstable even in this low
  ##      dimensional example!
}

## set seed (again to make results comparable)
set.seed(1234)
if (require("glmnet")) {
  (stab.glmnet <- stabsel(x = bodyfat[, -2], y = bodyfat[, 2],
                        fitfun = glmnet.lasso, cutoff = 0.75,
                        PFER = 1, mc.cores = 1L))
  ## set mc.cores = 1L only to avoid parallelization on CRAN;
  ## In practice you usually do not need set this value unless you want to
  ## avoid parallelization via mclapply or restrict it to a specific number
  ## of available cores. See documentation of mclapply for more details.

  par(mfrow = c(2, 1))
}

```

```

plot(stab.glmnet, main = "Lasso (glmnet)")
if (exists("stab.lasso"))
  plot(stab.lasso, main = "Lasso (lars)")
}

## Select variables with maximum coefficients based on lasso estimate
set.seed(1234) # reset seed
if (require("glmnet")) {
  ## use cross-validated lambda
  lambda.min <- cv.glmnet(x = as.matrix(bodyfat[, -2]), y = bodyfat[,2])$lambda.min
  (stab.maxCoef <- stabsel(x = bodyfat[, -2], y = bodyfat[,2],
                            fitfun = glmnet.lasso_maxCoef,
                            # specify additional parameters to fitfun
                            args.fitfun = list(lambda = lambda.min),
                            cutoff = 0.75, PFER = 1))

  ## WARNING: Using a fixed penalty (lambda) is usually not permitted and
  ##           not sensible. See ?fitfun for details.

  ## now compare standard lasso with "maximal parameter estimates" from lasso
  par(mfrow = c(2, 1))
  plot(stab.maxCoef, main = "Lasso (glmnet; Maximum Coefficients)")
  plot(stab.glmnet, main = "Lasso (glmnet)")
  ## --> very different results.
}

#####
#### using stability selection directly on computed boosting models
#### from mboost

if (require("mboost")) {
  ### low-dimensional example
  mod <- glmboost(DEXfat ~ ., data = bodyfat)

  ## compute cutoff ahead of running stabsel to see if it is a sensible
  ## parameter choice.
  ##   p = ncol(bodyfat) - 1 (= Outcome) + 1 (= Intercept)
  stabsel_parameters(q = 3, PFER = 1, p = ncol(bodyfat) - 1 + 1,
                     sampling.type = "MB")
  ## the same:
  stabsel(mod, q = 3, PFER = 1, sampling.type = "MB", eval = FALSE)

  ## now run stability selection
  (sbody <- stabsel(mod, q = 3, PFER = 1, sampling.type = "MB"))
  opar <- par(mai = par("mai") * c(1, 1, 1, 2.7))
  plot(sbody)
  par(opar)
  plot(sbody, type = "maxsel", ymargin = 6)
}

}

```

 stabsel.stabsel *Change Parameters of Stability Selection*

Description

Method to change the parameters cutoff, PFER and assumption of stability selection that can be altered without the need to re-run the subsampling process.

Usage

```
## S3 method for class 'stabsel'
stabsel(x, cutoff, PFER, assumption = x$assumption, ...)
```

Arguments

x	an object that results from a call to stabsel .
cutoff	cutoff between 0.5 and 1. Preferably a value between 0.6 and 0.9 should be used. If <code>assumption = "r-concave"</code> the cutoff can be below 0.5 but needs to be larger than q/p .
PFER	upper bound for the per-family error rate. This specifies the amount of falsely selected base-learners, which is tolerated. See details.
assumption	Defines the type of assumptions on the distributions of the selection probabilities and simultaneous selection probabilities. Only applicable for <code>sampling.type = "SS"</code> . For <code>sampling.type = "MB"</code> we always use "none".
...	additional arguments that are currently ignored.

Details

This function allows to alter the parameters cutoff, PFER and assumption of a fitted stability selection result. All other parameters are re-used from the original stability selection results. The missing parameter is computed and the selected variables are updated accordingly.

Value

An object of class [stabsel](#). For details see there.

See Also

[stabsel](#) for the generic function, [stabsel_parameters](#) for the computation of error bounds, [fitfun](#) for available fitting functions and [plot.stabsel](#) for available plot functions

Examples

```

if (require("TH.data")) {
  ## make data set available
  data("bodyfat", package = "TH.data")
} else {
  ## simulate some data if TH.data not available.
  ## Note that results are nonsense with this data.
  bodyfat <- matrix(rnorm(720), nrow = 72, ncol = 10)
}

## set seed
set.seed(1234)

#####
#### using stability selection with Lasso methods:

if (require("lars")) {
  (stab.lasso <- stabsel(x = bodyfat[, -2], y = bodyfat[, 2],
                        fitfun = lars.lasso, cutoff = 0.75,
                        PFER = 1, mc.cores = 1L))
  ## set mc.cores = 1L only to avoid parallelization on CRAN;
  ## In practice you usually do not need set this value unless you want to
  ## avoid parallelization via mclapply or restrict it to a specific number
  ## of available cores. See documentation of mclapply for more details.

  par(mfrow = c(2, 1))
  plot(stab.lasso)

  ## now change the PFER and the assumption:
  (stab.lasso_cf0.93_rconc <- stabsel(stab.lasso, cutoff = 0.93,
                                         assumption = "r-concave", mc.cores = 1L))
  ## In practice you usually do not need set this value unless you want to
  ## avoid parallelization via mclapply or restrict it to a specific number
  ## of available cores. See documentation of mclapply for more details.

  plot(stab.lasso_cf0.93_rconc)
  ## the cutoff did change and hence the PFER and the selected
  ## variables
}

```

Description

Compute the missing parameter from the two given parameters in order to assess suitability of the parameter constellation

Usage

```
stabsel_parameters(p, ...)

## Default S3 method:
stabsel_parameters(p, cutoff, q, PFER,
                    B = ifelse(sampling.type == "MB", 100, 50),
                    assumption = c("unimodal", "r-concave", "none"),
                    sampling.type = c("SS", "MB"),
                    verbose = FALSE, FWER, ...)

## S3 method for class 'stabsel_parameters'
print(x, heading = TRUE, ...)
```

Arguments

<code>p</code>	number of possible predictors (including intercept if applicable).
<code>cutoff</code>	cutoff between 0.5 and 1. Preferably a value between 0.6 and 0.9 should be used. If <code>assumption = "r-concave"</code> the cutoff can be below 0.5 but needs to be larger than <code>q/p</code> .
<code>q</code>	number of (unique) selected variables (or groups of variables depending on the model) that are selected on each subsample.
<code>PFER</code>	upper bound for the per-family error rate. This specifies the amount of falsely selected base-learners, which is tolerated. See details.
<code>B</code>	number of subsampling replicates. Per default, we use 50 complementary pairs for the error bounds of Shah & Samworth (2013) and 100 for the error bound derived in Meinshausen & Buehlmann (2010). As we use B complementary pairs in the former case this leads to $2B$ subsamples.
<code>assumption</code>	Defines the type of assumptions on the distributions of the selection probabilities and simultaneous selection probabilities. Only applicable for <code>sampling.type = "SS"</code> . For <code>sampling.type = "MB"</code> we always use "none".
<code>sampling.type</code>	use sampling scheme of Shah & Samworth (2013), i.e., with complementary pairs (<code>sampling.type = "SS"</code>), or the original sampling scheme of Meinshausen & Buehlmann (2010).
<code>verbose</code>	logical (default: <code>TRUE</code>) that determines whether warnings should be issued.
<code>FWER</code>	deprecated. Only for compatibility with older versions, use <code>PFER</code> instead.
<code>x</code>	an object of class <code>"stabsel_parameters"</code> .
<code>heading</code>	logical. Specifies if a heading line should be printed.
<code>...</code>	additional arguments to be passed to next function.

Details

This function implements the error bounds for stability selection by Meinshausen and Buehlmann (2010) and the improved error bounds by Shah and Samworth (2013). For details see also Hofner et al. (2014).

Two of the three arguments `cutoff`, `q` and `PFER` *must* be specified. The per-family error rate (PFER), i.e., the expected number of false positives $E(V)$, where V is the number of false positives, is bounded by the argument `PFER`.

For more details see also [stabsel](#).

Value

An object of class `stabsel_parameters` with a special `print` method. The object has the following elements:

<code>cutoff</code>	cutoff used.
<code>q</code>	average number of selected variables used.
<code>PFER</code>	(realized) upper bound for the per-family error rate.
<code>specifiedPFER</code>	specified upper bound for the per-family error rate.
<code>p</code>	the number of effects subject to selection.
<code>B</code>	the number of subsamples.
<code>sampling.type</code>	the sampling type used for stability selection.
<code>assumption</code>	the assumptions made on the selection probabilities.

References

B. Hofner, L. Boccuto and M. Goeker (2015), Controlling false discoveries in high-dimensional situations: Boosting with stability selection. *BMC Bioinformatics*, 16:144.
[doi:10.1186/s1285901505753](https://doi.org/10.1186/s1285901505753).

N. Meinshausen and P. Buehlmann (2010), Stability selection. *Journal of the Royal Statistical Society, Series B*, **72**, 417–473.

R.D. Shah and R.J. Samworth (2013), Variable selection with error control: another look at stability selection. *Journal of the Royal Statistical Society, Series B*, **75**, 55–80.

See Also

For more details see also [stabsel](#).

<code>subsample</code>	<i>Draw Random Subsamples</i>
------------------------	-------------------------------

Description

Set up weight matrix for subsampling with sample proportion 1/2 to be used with [stabsel](#).

Usage

```
subsample(weights, B = 100, strata = NULL)
```

Arguments

weights	a numeric vector of weights for the model to be cross-validated.
B	number of folds, per default 25 for bootstrap and subsampling and 10 for kfold.
strata	a factor of the same length as <code>weights</code> for stratification.

Details

The function `subsample` can be used to build an appropriate weight matrix to be used with [stabsel](#). See there for more details.

If `strata` is defined sampling is performed in each stratum separately thus preserving the distribution of the `strata` variable in each fold.

See Also

[stabsel](#)

Examples

```
## just a low-dimensional example
subsample(weights = rep(1, 10), B = 50)
```

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