

Package: tboot (via r-universe)

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Title Tilted Bootstrap

Version 0.2.0

Description The tilted bootstrap as implemented in the 'tboot' package is an approach to resampling where samples are drawn from the observed data with some samples appearing more frequently than others. Tilted bootstrapping may be used to create simulated clinical trial data with realistic correlation structures and assumed efficacy levels. The 'tboot' package may also be used for simulating a joint Bayesian distribution along along with clinical trials based on the Bayesian distribution.

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Depends R (>= 3.2.0)

Imports stats, quadprog, kernlab

Suggests knitr, rmarkdown, testthat, MASS, ggplot2

VignetteBuilder knitr

URL <https://github.com/nmorris-lilly/tboot>

BugReports <https://github.com/nmorris-lilly/tboot/issues>

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Repository <https://njm18.r-universe.dev>

RemoteUrl <https://github.com/njm18/tboot>

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tboot-package	<i>tboot: tilted bootstrapping and Bayesian marginal reconstruction.</i>
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Description

tboot: tilted bootstrapping and Bayesian marginal reconstruction.

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References

<https://github.com/wlandau-lilly/tboot>

post_bmr	<i>Function post_bmr</i>
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Description

Simulates the joint posterior based upon a dataset and specified marginal posterior distribution of the mean of selected variables.

Usage

```
post_bmr(nsims, weights_bmr)
```

Arguments

nsims The number of posterior simulations to draw.

weights_bmr An object of class 'tweights_bmr' created using the 'tweights_bmr' function.

Value

A matrix of simulations from the posterior.

See Also

[tweights_bmr](#)

tboot	<i>Function tboot</i>
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Description

Bootstrap nrow rows of dataset using the given row-level weights.

Usage

```
tboot(nrow, weights, dataset = weights$dataset, fillMissingAug = TRUE)
```

Arguments

nrow	number of rows in the new bootstrapped dataset.
weights	an object of class 'tweights' output from the 'tweights' function.
dataset	Data frame or matrix to bootstrap. Rows of the dataset must be in the same order as was used for the 'tweights' call. However the dataset may include additional columns not included in the 'tweights' call.
fillMissingAug	fill in missing augmentation with primary weights resampling.

Details

Bootstrap simulates a dataset using the tilted weights. Details are further documented in the vignette.

Value

A simulated dataset with 'nrow' rows.

See Also

[tweights](#)

tboot_bmr	<i>Function tboot_bmr</i>
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Description

Bootstrap nrow rows of dataset using the given row-level weights.

Usage

```
tboot_bmr(nrow, weights_bmr, tol_rel_sd = 0.01)
```

Arguments

nrow	number of rows in the new bootstrapped dataset.
weights_bmr	an object of class 'tweights' output from the 'tweights' function.
tol_rel_sd	An error will be called if for some simulation if the target is not achievable with the data. However, the error will only be called if max absolute difference relative to the marginal standard is greater than specified.

Details

Simulates a dataset by first simulating from the posterior distribution of the column means and then simulating a dataset with that underlying mean. Details a further documented in the vignette.

Value

A simulated dataset with 'nrow' rows. The underlying 'true' posterior parameter value is an attribute which can be extracted using `attr(ret, "post_bmr")` where 'ret' is the matrix.

See Also

[tweights](#)

tweights	<i>Function</i> tweights
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Description

Returns a vector `p` of resampling probabilities such that the column means of `tboot(dataset = dataset, p = p)` equals target on average.

Usage

```
tweights(
  dataset,
  target = apply(dataset, 2, mean),
  distance = "klqp",
  maxit = 1000,
  tol = 1e-08,
  warningcut = 0.05,
  silent = FALSE,
  Nindependent = 0
)
```

Arguments

dataset	Data frame or matrix to use to find row weights.
target	Numeric vector of target column means. If the 'target' is named, then all elements of names(target) should be in the dataset.
distance	The distance to minimize. Must be either 'euclidean,' 'klqp' or 'klpq' (i.e. Kullback-Leibler). 'klqp' which is exponential tilting is recommended.
maxit	Defines the maximum number of iterations for optimizing 'kl' distance.
tol	Tolerance. If the achieved mean is too far from the target (i.e. as defined by tol) an error will be thrown.
warningcut	Sets the cutoff for determining when a large weight will trigger a warning.
silent	Allows silencing some messages.
Nindependent	Assumes the input also includes 'Nindependent' samples with independent columns. See details.

Details

Let $p_i = 1/n$ be probability of sampling subject i from a dataset with n individuals (i.e. rows of the dataset) in the classic resampling with replacement scheme. Also, let q_i be the probability of sampling subject i from a dataset with n individuals in our new resampling scheme. Let $d(q, p)$ represent a distance between the two resampling schemes. The tweights function seeks to solve the problem:

$$q = \operatorname{argmin}_p d(q, p)$$

Subject to the constraint that:

$$\sum_i q_i = 1$$

and

$$\text{dataset}'q = \text{target}$$

where dataset is a $n \times K$ matrix of variables input to the function.

$$d_{\text{euclidian}}(q, p) = \sqrt{\sum_i (p_i - q_i)^2}$$

$$d_{\text{kl}}(q, p) = \sum_i (\log(p_i) - \log(q_i))$$

Optimization for euclidean distance is a quadratic program and utilizes the ipop function in kernLab. The euclidean based solution helps form a starting value which is used along with the constOptim function and lagrange multipliers to solve the Kullback-Leibler distance optimization. Output is the optimal probability (p)

The 'Nindependent' option augments the dataset by assuming some additional specified number of patients. These patients are assumed to be made up of a random bootstrapped sample from the dataset for each variable marginally leading to independent variables.

Value

An object of type `tweights`. This object contains the following components:

- weights** tilted weights for resampling
- originalTarget** Will be null if target was not changed.
- target** Actual target that was attempted.
- achievedMean** Achieved mean from tilting.
- dataset** Inputted dataset.
- X** Reformatted dataset.
- Nindependent** Inputted 'Nindependent' option.

See Also

[tboot](#)

<code>tweights_bmr</code>	<i>Function <code>tweights_bmr</code></i>
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Description

Setup the needed pre-requisites in order to prepare for bayesian marginal reconstruction (including a call to `tweights`). Takes as input simulations from the posterior marginal distribution of variables in a dataset.

Usage

```
tweights_bmr(
  dataset,
  marginal,
  distance = "klqp",
  maxit = 1000,
  tol = 1e-08,
  warningcut = 0.05,
  silent = FALSE,
  Nindependent = 1
)
```

Arguments

- `dataset` Data frame or matrix to use to find row weights.
- `marginal` Must be a named list with each element a vector of simulations of the marginal distribution of the posterior mean of data in the dataset.
- `distance` The distance measure to minimize. Must be either 'euchlidean' or 'kl' (i.e. Kullback-Leibler). 'klqp' is recommended.

<code>maxit</code>	Defines the maximum number of iterations for optimizing 'kl' distance.
<code>tol</code>	Tolerance. If the achieved mean is too far from the target (i.e. as defined by <code>tol</code>) an error will be thrown.
<code>warningcut</code>	Sets the cutoff for determining when a large weight will trigger a warning.
<code>silent</code>	Allows silencing some messages.
<code>Nindependent</code>	Assumes the input also includes 'Nindependent' samples with independent columns. See details.

Details

Reconstructs a correlated joint posterior from simulations from a marginal posterior. Algorithm is summarized more fully in the vignettes. The 'Nindependent' option augments the dataset by assuming some additional specified number of patients. These patients are assumed to be made up of a random bootstrapped sample from the dataset for each variable marginally leading to independent variables.

Value

An object of type `tweights`. This object contains the following components:

Csqr Matrix square root of the covariance.

tweights Result from the call to `tweights`.

marginal Input marginal simulations.

dataset Formatted dataset.

target Attempted target.

distance,maxit,tol, Nindependent, warningcut Inputted values to 'tweights_bmr'.

Nindependent Inputted 'Nindependent' option.

augmentWeights Used for 'Nindependent' option weights for each variable.

weights tilted weights for resampling

originalTarget Will be null if target was not changed.

marginal_sd Standard deviation of the marginals.

See Also

[tweights](#)

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