# Package 'tboot'

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Title Tilted Bootstrap

Version 0.2.0

<b>Description</b> Creates simulated clinical trial data with realistic correlation structures and assumed efficacy levels by using a tilted bootstrap resampling approach. Samples are drawn from observed data with some samples appearing more frequently than others. May also be used for simulating from a joint Bayesian distribution along with clinical trials based on the Bayesian distribution.
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tboot-package

tboot: tilted bootstrapping and Bayesian marginal reconstruction.

# Description

tboot: tilted bootstrapping and Bayesian marginal reconstruction.

# Author(s)

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# References

https://github.com/njm18/tboot

post\_bmr

Function post\_bmr

# 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
```

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## **Examples**

tboot

Function thoot

## **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' calll.

fillMissingAug fill in missing augmentation with primary weights resampling.

## **Details**

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

#### Value

A simulated dataset with 'nrow' rows.

## See Also

tweights

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## **Examples**

```
target=c(Sepal.Length=5.5, Sepal.Width=2.9, Petal.Length=3.4)
w = tweights(dataset = iris, target = target, silent = TRUE)
simulated_data = tboot(nrow = 1000, weights = w)
```

tboot\_bmr

Function tboot\_bmr

#### **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

releative 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 useing attr(ret, "post\_bmr") where 'ret' is the matrix.

# See Also

```
tweights
```

## **Examples**

```
#Use winsorized marginal to keep marginal simulation within feasible bootstrap region
winsor=function(marginalSims,y) {
    l=min(y)
    u=max(y)
    ifelse(marginalSims<1,1,ifelse(marginalSims>u,u, marginalSims))
}
#Create an example marginal posterior
marginal = list(Sepal.Length=winsor(rnorm(10000,mean=5.8, sd=.2),iris$Sepal.Length),
```

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tweights

Function tweights

# 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 'euchlidean,' 'klqp' or 'klpq' (i.e. Kullback-Leibler). 'klqp' which is expontential tilting is recomneded.
maxit	Defines the maximum number of iterations for optimizing 'kl' distance.
tol	Tolerance. If the achieved mean is to 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 warnint.
silent	Allows silencing some messages.
Nindependent	Assumes the input also includes 'Nindependent'samples with independent columns. See details.

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#### **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:

$$q = argmin_p d(q, p)$$

Subject to the constraint that:

$$sum_i q_i = 1$$

and

$$dataset'q = target$$

where dataset is a n x K matrix of variables input to the function.

$$d_euclidian(q, p) = sqrt(sum_i(p_i - q_i)^2)$$
$$d_k l(q, p) = sum_i(log(p_i) - log(q_i))$$

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 porbability (p)

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

## Value

An object of type tweights. This object conains 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 Inputed dataset.

X Reformated dataset.

**Nindependent** Inputed 'Nindependent' option.

## See Also

tboot

## **Examples**

```
target=c(Sepal.Length=5.5, Sepal.Width=2.9, Petal.Length=3.4)
w = tweights(dataset = iris, target = target, silent = TRUE)
simulated_data = tboot(nrow = 1000, weights = w)
```

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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 recomneded.
maxit	Defines the maximum number of iterations for optimizing 'kl' distance.
tol	Tolerance. If the achieved mean is to 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 warnint.
silent	Allows silencing some messages.
Nindependent	Assumes the input also includes 'Nindependent'samples with independent columns. See details.

# **Details**

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

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#### Value

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

Csqrt Matrix square root of the covariance.

tweights Result from the call to tweigths.

marginal Input marginal simulations.

dataset Formatted dataset.

target Attempted target.

distance,maxit,tol, Nindependent, warningcut Inputed values to 'tweights\_bmr'.

Nindependent Inputed '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
```

## **Examples**

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