# Package 'reconsi'

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Type Package

**Title** Resampling Collapsed Null Distributions for Simultaneous Inference

Version 1.6.0

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**Description** Improves simultaneous inference under dependence of tests by estimating a collapsed null distribution through resampling. Accounting for the dependence between tests increases the power while reducing the variability of the false discovery proportion. This dependence is common in genomics applications, e.g. when combining flow cytometry measurements with microbiome sequence counts.

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**Encoding UTF-8** 

LazyData true

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**Imports** phyloseq, KernSmooth, reshape2, ggplot2, stats, methods, graphics, grDevices, matrixStats

Suggests knitr, rmarkdown, testthat

VignetteBuilder knitr

biocViews Metagenomics, Microbiome, MultipleComparison, FlowCytometry

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# Description

Obtain weights as posterior probabilities to calculate the consensus null

# Usage

```
calcWeights(logDensPerm, fdr)
```

# Arguments

logDensPerm A matrix with B rows of logged density estimates of the B permutation distributions, and p columns for the p observed test statistics

A vector of local false discovery rates for the observed tests statistics of length

p

# Value

A vector of weights of length B

estNormal 3

estNormal	Fast estimation of mean and standard deviation of a normal distrbution, optionally with weights

## **Description**

Fast estimation of mean and standard deviation of a normal distrbution, optionally with weights

## Usage

```
estNormal(y, w = NULL, p = length(y))
```

# Arguments

y vector of observations
w optional weight vector
p The number of features

## Value

A vector of length 2 with mean and standard deviation

Estimate the fraction of true null hypotheses.
Estimate the fraction of true null hypotheses.

## **Description**

Estimate the fraction of true null hypotheses.

## Usage

```
estP0(statObs, fitAll, z0quantRange, smooth.df)
```

## **Arguments**

statObs A vector of observed z-values fitAll the estimated normal null

z@quantRange a number of quantiles between 0 and 0.5 smooth.df degrees of freedom for the spline smoother

## **Details**

A natural spline is used over a range of intervals. Based on the qvalue::qvalue() function and Storey and Tibshirani, 2003

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#### Value

The estimated null fraction, the value of the spline evaluated at the first element of z0quantRange

getApproxCovar

Obtain a null covariance matrix of binned test statistics

## **Description**

Obtain a null covariance matrix of binned test statistics

# Usage

```
getApproxCovar(statsPerm, nBins = 82L, binEdges = c(-4.1, 4.1))
```

# **Arguments**

statsPerm The pxB matrix of permutation z-values in the columns

nBins an integer, the number of bins

binEdges A vector of length 2 with the outer bin edges

#### Value

The covariance matrix of binned z-values

## Note

This is not the covariance matrix of the p test statistic, nor of the data! It is an approximate covariance matrix of binned test statistics for visualization purposes.

```
p = 200; n = 50; B = 5e1
x = rep(c(0,1), each = n/2)
mat = cbind(
matrix(rnorm(n*p/10, mean = 5+x),n,p/10), #DA
matrix(rnorm(n*p*9/10, mean = 5),n,p*9/10) #Non DA
)
mat = mat = mat + rnorm(n, sd = 0.3) #Introduce some dependence
fdrRes = reconsi(mat, x, B = B)
corMat = getApproxCovar(fdrRes$statsPerm)
```

getFdr 5

getFdr	Calculate tail-area (Fdr) and local (fdr) false discovery rates, based on a certain null distribution

# Description

Calculate tail-area (Fdr) and local (fdr) false discovery rates, based on a certain null distribution

# Usage

```
getFdr(statObs, fitAll, fdr, zSeq, p, p0, zValsDensObs, smoothObs, ...)
```

# Arguments

stat0bs	Vector of observed z-values
fitAll	The parameters of the estimated random null
fdr	local false discovery rate, already estimated
zSeq	Support of the density estimation
р	the number of hypotheses
p0	The estimated fraction of null hypotheses
zValsDensObs	estimated densities of observed test statistics
smoothObs	A boolean, should estimated observed densities of the test statistics be used in estimating the Fdr
	more arguments, ignored

# Value

A list with components

Fdr	Tail are false discovery rate
fdr	Local false discovery rate

getG0	Obtain the consensus null	
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# Description

Obtain the consensus null

6 getG0

### Usage

```
getG0(
   statObs,
   statsPerm,
   z0Quant,
   gridsize,
   maxIter,
   tol,
   estP0args,
   testPargs,
   B,
   p,
   pi0
)
```

## **Arguments**

statObs A vector of length p with observed test statistics

statsPerm A pxB matrix with permuation z-values

z@Quant a vector of length of quantiles defining the central part R of the distribution. If a

single number is supplied, then (z0quant, 1-z0quant) will be used

gridsize An integer, the gridsize for the density estimation

maxIter An integer, the maximum number of iterations in determining R

tol The convergence tolerance.

estP0args A list of arguments passed on to the estP0args() function

testPargs A list of arguments passed on to quantileFun

B an integer, the number of permutations
p an integer, the number of hypotheses
pi0 A known fraction of true null hypotheses.

#### Value

A list with following entries

PermDensFits The permutation density fits

zSeq The support of the kernel for density estimation

zValsDensObs The estimated densities of the observed z-values

convergence A boolean, has the algorithm converged?

weights Vector of length B with weights for the permutation distributions fdr Estimated local false discovery rate along the support of the kernel

p0 The estimated fraction of true null hypotheses

iter The number of iterations fitAll The consensus null fit

getTestStats 7

getTestStats	A function to calculate observed and permuation z-statistics on a n-by-p matrix of observations

# Description

A function to calculate observed and permuation z-statistics on a n-by-p matrix of observations

# Usage

```
getTestStats(
   Y,
   center,
   test = "wilcox.test",
   X,
   B,
   argList,
   tieBreakRan,
   replace,
   scale
)
```

# Arguments

Υ	The nxp data matrix
center	a boolean, should data be centered prior to permuation
test	A function name, possibly user defined. See details.
x	A vector defining the groups. Will be coerced to factor.
В	an integer, the number of permuations
argList	A list of further arguments passed on to the test function
tieBreakRan	A boolean, should ties of permutation test statistics be broken randomly? If not, midranks are used
replace	A boolean. If FALSE, samples are permuted (resampled without replacement), if TRUE the samples are bootstrapped (resampled with replacement)
scale	a boolean, should data be scaled prior to resampling

# **Details**

For test "wilcox.test" and "t.test", fast custom implementations are used. Other functions can be supplied but must accept a y outcome variable, a x as grouping variable, and possibly a list of other arguments. It must return all arguments needed to evaluate its quantile function if z-values are to be used.

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#### Value

A list with components

statObs A vector of length p of observed test statistics statsPerm A p-by-B matrix of permutation test statistics

resamDesign The resampling design

getTstat

A function to obtain a t-test statistic efficiently. For internal use only

# Description

A function to obtain a t-test statistic efficiently. For internal use only

## Usage

```
getTstat(y1, y2, mm, nn)
```

# Arguments

y1, y2 vectors of observeed values in the two groups

mm, nn number of observations in the corresponding groups

#### Value

A list with items

tstat The t-test statistic

df The degrees of freedom (Welch approximation)

plotApproxCovar

Plot an approximatio of the correlation structure of the test statistics

# Description

Plot an approximatio of the correlation structure of the test statistics

plotApproxCovar 9

#### Usage

```
plotApproxCovar(
   reconsiFit,
   col = colorRampPalette(c("yellow", "blue"))(12),
   x = seq(-4.2, 4.2, 0.1),
   y = seq(-4.2, 4.2, 0.1),
   xlab = "Z-values",
   ylab = "Z-values",
   nBins = 82L,
   binEdges = c(-4.1, 4.1),
   ...
)
```

# Arguments

```
\begin{tabular}{ll} reconsifit & The reconsifit \\ col, x, y, xlab, ylab, ... & A list of arguments for the image() function. \\ nBins, binEdges & passed on to the getApproxCovar function \\ \end{tabular}
```

#### **Details**

By default, yellow indicates negative correlation between bin counts, blue positive correlation

#### Value

invisible()

## Note

This is not the covariance matrix of the p test statistic, nor of the data! It is an approximate covariance matrix of binned test statistics for visualization purposes.

```
p = 200; n = 50; B = 5e1
x = rep(c(0,1), each = n/2)
mat = cbind(
matrix(rnorm(n*p/10, mean = 5+x),n,p/10), #DA
matrix(rnorm(n*p*9/10, mean = 5),n,p*9/10) #Non DA
)
mat = mat = mat + rnorm(n, sd = 0.3) #Introduce some dependence
fdrRes = reconsi(mat, x, B = B)
plotApproxCovar(fdrRes)
```

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plotNull

Plot the obtained null distribution along with a histogram of observed test statistics

# Description

Plot the obtained null distribution along with a histogram of observed test statistics

## Usage

```
plotNull(
   fit,
   lowColor = "yellow",
   highColor = "blue",
   idDA = NULL,
   nResampleCurves = length(fit$Weights),
   hSize = 0.5
)
```

# **Arguments**

```
fit an object returned by the reconsi() (or testDAA()) function lowColor, highColor

The low and high ends of the colour scale

idDA indices of known null taxa

nResampleCurves

The number of resampling null distributions to plot

hSize A double, the size of the line of the collapsed null estimate
```

#### Value

```
a ggplot2 plot object
```

```
p = 175; n = 50; B = 1e2
#Low number of resamples keeps computation time down
x = rep(c(0,1), each = n/2)
mat = cbind(
matrix(rnorm(n*p/10, mean = 5+x),n,p/10), #DA
matrix(rnorm(n*p*9/10, mean = 5),n,p*9/10) #Non DA
)
fdrRes = reconsi(mat, x, B = B)
plotNull(fdrRes)
```

ptEdit 11

ptEdit	A custom function to calculate the distribution function of the t-test
	statistic. For internal use only

# Description

A custom function to calculate the distribution function of the t-test statistic. For internal use only

# Usage

```
ptEdit(q)
```

# **Arguments**

q

a vector with t-statistic and degrees of freedom

## Value

A value between 0 and 1, the evaluation of the cdf

qtEdit	A custom function to calculate the quantile function of the t-test statis-
	tic. For internal use only

# Description

A custom function to calculate the quantile function of the t-test statistic. For internal use only

# Usage

```
qtEdit(p)
```

# Arguments

p a vector with quantile and degrees of freedom

#### Value

the corresponding quantile

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quantCorrect

Correct quantiles by not returning 0 or 1

#### **Description**

Correct quantiles by not returning 0 or 1

## Usage

```
quantCorrect(quants)
```

## **Arguments**

quants

A vector of quantiles

#### Value

The same vector of quantiles but without 0 or 1 values

reconsi

Perform simultaneous inference through collapsed resampling null distributions

## Description

Perform simultaneous inference through collapsed resampling null distributions

## Usage

```
reconsi(
 Υ,
 x = NULL
 B = 1000L
  test = "wilcox.test",
 argList = list(),
 distFun = "pnorm",
  zValues = TRUE,
  testPargs = list(),
  z0Quant = pnorm(c(-1, 1)),
  gridsize = 801L,
 maxIter = 1000L,
  tol = 1e-08,
  center = FALSE,
  replace = is.null(x),
  zVals = NULL,
 estP0args = list(z0quantRange = seq(0.05, 0.45, 0.0125), smooth.df = 3),
```

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```
resamZvals = FALSE,
smoothObs = TRUE,
scale = FALSE,
tieBreakRan = FALSE,
pi0 = NULL
)
```

# Arguments

Υ	the matrix of sequencing counts
X	a grouping factor. If provided, this grouping factor is permuted. Otherwise a bootstrap procedure is performed
В	the number of resampling instances
test	Character string, giving the name of the function to test for differential absolute abundance. Must accept the formula interface. Features with tests resulting in observed NA test statistics will be discarded
argList	A list of arguments, passed on to the testing function
distFun	the distribution function of the test statistic, or its name. Must at least accept an argument named 'q', 'p' and 'x' respectively.
zValues	A boolean, should test statistics be converted to z-values. See details
testPargs	A list of arguments passed on to distFun
z0Quant	A vector of length 2 of quantiles of the null distribution, in between which only null values are expected
gridsize	The number of bins for the kernel density estimates
maxIter	An integer, the maximum number of iterations in the estimation of the null distribution
tol	The tolerance for the infinity norm of the central borders in the iterative procedure
center	A boolean, should observations be centered in each group prior to permuations? See details.
replace	A boolean. Should resampling occur with replacement (boostrap) or without replacement (permutation)
zVals	An optional list of observed (statObs) and resampling (statsPerm) z-values. If supplied, the calculation of the observed and resampling test statistics is skipped and the algorithm proceeds with calculation of the consensus null distribution
estP0args	A list of arguments passed on to the estP0 function
resamZvals	A boolean, should resampling rather than theoretical null distributions be used?
smoothObs	A boolean, should the fitted rather than estimated observed distribution be used in the Fdr calculation?
scale	a boolean, should data be scaled prior to resampling
tieBreakRan	A boolean, should ties of resampling test statistics be broken randomly? If not, midranks are used
pi0	A known fraction of true null hypotheses. If provided, the fraction of true null hypotheses will not be estimated. Mainly for oracle purposes.

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#### **Details**

Efron (2007) centers the observations in each group prior to permutation. As permutations will remove any genuine group differences anyway, we skip this step by default. If zValues = FALSE, the density is fitted on the original test statistics rather than converted to z-values. This unlocks the procedure for test statistics with unknown distributions, but may be numerically less stable.

#### Value

#### A list with entries

statsPerm Resampling Z-values statObs Observed Z-values

distFun Density, distribution and quantile function as given

testPargs Same as given

zValues A boolean, were z-values used?

resamZvals A boolean, were the resampling null distribution used?

cdfVal0bs Cumulative distribution function evaluation of observed test statistics

p@estimated A boolean, was the fraction of true null hypotheses estimated from the data?

Fdr, fdr Estimates of tail-area and local false discovery rates

p0 Estimated or supplied fraction of true null hypotheses

iter Number of iterations executed

fitAll Mean and standard deviation estimated collapsed null

PermDensFits Mean and standard deviations of resamples

convergence A boolean, did the iterative algorithm converge?

zSeq Basis for the evaluation of the densities weights weights of the resampling distributions

zValsDensObs Estimated overall densities, evaluated in zSeq

#### Note

Ideally, it would be better to only use unique resamples, to avoid unnecesarry replicated calculations of the same test statistics. Yet this issue is almost always ignored in practice; as the sample size grows it also becomes irrelevant. Notice also that this would require to place weights in case of the bootstrap, as some bootstrap samples are more likely than others.

```
#Important notice: low number of resamples B necessary to keep # computation time down, but not recommended. Pray set B at 200 or higher. p = 50; n = 20; B = 5e1 x = rep(c(0,1), each = n/2) mat = cbind( matrix(rnorm(n*p/10, mean = 5+x), n, p/10), #DA matrix(rnorm(n*p*9/10, mean = 5), n, p*9/10) #Non DA )
```

rowMultiply 15

```
fdrRes = reconsi(mat, x, B = B)
fdrRes$p0
#Indeed close to 0.9
estFdr = fdrRes\$Fdr
#The estimated tail-area false discovery rates.
#With another type of test. Need to supply quantile function in this case
fdrResLm = reconsi(mat, x, B = B,
test = function(x, y){
fit = lm(y^x)
c(summary(fit)$coef["x","t value"], fit$df.residual)},
distFun = function(q)\{pt(q = q[1], df = q[2])\})
#With a test statistic without known null distribution(for small samples)
fdrResKruskal = reconsi(mat, x, B = B,
test = function(x, y){}
kruskal.test(y^x)statistic}, zValues = FALSE)
#Provide an additional covariate through the 'argList' argument
z = rpois(n, lambda = 2)
fdrResLmZ = reconsi(mat, x, B = B,
test = function(x, y, z){
fit = lm(y^x+z)
c(summary(fit)$coef["x","t value"], fit$df.residual)},
distFun = function(q){pt(q = q[1], df = q[2])},
argList = list(z = z))
#When nog grouping variable is provided, a bootstrap is performed
matBoot = cbind(
matrix(rnorm(n*p/10, mean = 1), n, p/10), #DA
matrix(rnorm(n*p*9/10, mean = 0), n, p*9/10) #Non DA
fdrResBoot = reconsi(matBoot, B = B,
test = function(y, x){testRes = t.test(y, mu = 0, var.equal = TRUE);
c(testRes$statistic, testRes$parameter)},
distFun = function(q){pt(q = q[1], df = q[2])},
center = TRUE, replace = TRUE)
```

rowMultiply

A function to efficiently row multiply a a-by-b matrix by a vector of length b. More memory intensive but that does not matter with given matrix sizes

# Description

A function to efficiently row multiply a a-by-b matrix by a vector of length b. More memory intensive but that does not matter with given matrix sizes

# Usage

```
rowMultiply(matrix, vector)
```

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## **Arguments**

matrix a numeric matrix of dimension a-by-b

vector a numeric vector of length b

## **Details**

t(t(matrix)\*vector) but then faster

#### Value

a matrix, row multplied by the vector

 ${\it stabExp} \qquad \qquad {\it A function to numerically stabilize an exponentiation. For internal use}$ 

only

## **Description**

A function to numerically stabilize an exponentiation. For internal use only

## Usage

stabExp(exps)

## Arguments

exps the vector to be exponentiated

## Value

the vector with the maximum subtracted

testDAA A function to test for differential absolute abundance on a phyloseq

object

# Description

A function to test for differential absolute abundance on a phyloseq object

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#### **Usage**

```
testDAA(Y, ...)
## S4 method for signature 'phyloseq'
testDAA(Y, groupName, FCname, ...)
## S4 method for signature 'matrix'
testDAA(Y, FC, x, S = rowSums(Y), tieBreakRan = TRUE, ...)
```

#### **Arguments**

Y A phyloseq object, or a data matrix with samples in the rows and OTUs in the

columns

... passed on to the reconsi() function

groupName A character string, the name of a variable in physeq indicating the grouping

factor

FCname A character string, the name of a variable in physeq containing the total cell

count

FC a vector of length n with total flow cytometry cell counts

x a grouping factor of length n

S a vector of library sizes. Will be calculated if not provided

tieBreakRan A boolean, should ties be broken at random.

#### Value

See the reconsi() function

```
#Test for phyloseq object
library(phyloseq)
VandeputtePruned = prune_samples(Vandeputte,
samples = sample_names(Vandeputte)[20:40])
testVanDePutte = testDAA(VandeputtePruned, "Health.status", "absCountFrozen",
B = 15)
#Test for matrix
testMat = testDAA(as(otu_table(VandeputtePruned), "matrix"),
get_variable(VandeputtePruned, "Health.status"),
get_variable(VandeputtePruned, "absCountFrozen"), B = 15)
```

Vandeputte

Vandeputte

Microbiomes of Crohn's disease patients and healthy controls

# Description

Microbiome sequencing data of Crohn's disease patients, and healthy controls, together with other baseline covariates. Both sequencing and flow cytometry data are available.

# Usage

Vandeputte

#### **Format**

A phyloseq object with an OTU-table and sample data

otu\_table Count data matrix of 234 taxa in 135 samples

sample\_data Data frame of patient covariates

## Source

https://www.ncbi.nlm.nih.gov/pubmed/29143816

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