Package: sumSome (via r-universe)

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Type Package

Title True Discovery Guarantee by Sum-Based Tests

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Description It allows to quickly perform closed testing by sum-based global tests, and construct lower confidence bounds for the TDP, simultaneously over all subsets of hypotheses. As main features, it produces permutation-based simultaneous lower confidence bounds for the proportion of active voxels in clusters for fMRI data, and for the proportion of differentially expressed genes in pathways for gene expression data. Details may be found in Vesely at al. (2021) <arXiv:2102.11759> and Tian at al. (2021) <arXiv:2102.11253>.

License GPL (>= 2)

Imports Rcpp (>= 1.0.5), stats, pARI, ARIbrain, RNifti, FMStable

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BugReports https://github.com/annavesely/sumSome/issues

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Repository https://annavesely.r-universe.dev

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Contents

`sumSome-package` *True Discovery Guarantee by Sum-Based Tests*

Description

It provides true discovery guarantees, using sum-based global statistics (sum of t-scores, p-value combinations, etc.). As main features, it produces permutation-based simultaneous lower confidence bounds for the proportion of active voxels in clusters for fMRI data, and for the proportion of differentially expressed genes in pathways for gene expression data.

Author(s)

Anna Vesely and Xu Chen.

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References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/1-STS356.

Tian J., Chen X., Katsevich E., Goeman J. J. and Ramdas A. (2022). Large-scale simultaneous inference under dependence. Scandinavian Journal of Statistics, doi: 10.1111/sjos.12614.

Vesely A., Finos L., and Goeman J. J. (2023). Permutation-based true discovery guarantee by sum tests. Journal of the Royal Statistical Society, Series B (Statistical Methodology), doi: 10.1093/jrsssb/qkad019.

See Also

fMRI cluster analysis: brainScores, brainPvals, brainClusters, brainAnalysis

Gene expression pathway analysis: geneScores, genePvals, geneAnalysis

General setting: sumStats and sumPvals (permutations), sumStatsPar and sumPvalsPar (parametric)

`brainAnalysis` *True Discovery Guarantee for Cluster Analysis of Brain Imaging Data*

Description

This function uses permutation t-statistics/p-values to determine a true discovery guarantee for fMRI cluster analysis. It computes confidence bounds for the number of true discoveries and the true discovery proportion within each cluster. The bounds are simultaneous over all sets, and remain valid under post-hoc selection.

Usage

```
brainAnalysis(sumBrain, clusters = NULL, nMax = 50, silent = FALSE)
```

Arguments

`sumBrain`	an object of class sumBrain, as returned by the functions brainScores and brainPvals.
`clusters`	3D numeric array of cluster indices, or character for a Nifti file name. If NULL, the whole brain is considered.
`nMax`	maximum number of iterations per cluster.
`silent`	logical, FALSE to print a summary of active clusters.

Value

`brainAnalysis` returns a list containing `summary` (data frame) and `TDPmap` (3D numeric array of the true discovery proportions). The data frame `summary` contains, for each cluster,

- `size`: size
- `TD`: lower (1-alpha)-confidence bound for the number of true discoveries
- `maxTD`: maximum value of `TD` that could be found under convergence of the algorithm
- `TDP`: lower (1-alpha)-confidence bound for the true discovery proportion
- `maxTD`: maximum value of `TDP` that could be found under convergence of the algorithm
- `dim1`, `dim2`, `dim3`: coordinates of the center of mass.

Author(s)

Anna Vesely.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Vesely A., Finos L., and Goeman J. J. (2023). Permutation-based true discovery guarantee by sum tests. Journal of the Royal Statistical Society, Series B (Statistical Methodology), doi: 10.1093/jrsssb/qkad019.

See Also

Permutation statistics for brain imaging: `brainScores`, `brainPvals`

Suprathreshold clusters: `brainClusters`

Examples

```
# simulate 20 copes with dimensions 10x10x10
set.seed(42)
copes <- list()
for(i in seq(20)){copes[[i]] <- array(rnorm(10^3, mean = -10, sd = 30), dim=c(10,10,10))}

# cluster map where t scores are grater than 2.8, in absolute value
thr <- 2.8
cl <- brainClusters(copes = copes, thr = thr)

# create object of class sumBrain
res <- brainScores(copes = copes, alpha = 0.2, seed = 42, truncFrom = thr)
res
summary(res)

# confidence bound for the number of true discoveries and the TDP within clusters
out <- brainAnalysis(res, clusters = cl$clusters)
out$summary
```

`brainClusters` *Suprathreshold Clusters for Brain Imaging*

Description

This function determines spatially connected clusters, where t-scores are more extreme than a given threshold.

Usage

```
brainClusters(copes, mask = NULL, thr = 3.2, alternative = "two.sided", silent = FALSE)
```

Arguments

copes	list of 3D numeric arrays (contrasts maps for each subject).
mask	3D logical array, where TRUE values correspond to voxels inside the brain, or character for a Nifti file name.
thr	threshold.
alternative	direction of the alternative hypothesis (greater, lower, two.sided).
silent	logical, FALSE to print the number of clusters.

Value

brainClusters returns a 3D numeric array, with integer values corresponding to clusters, and 0 to other voxels.

Author(s)

Anna Vesely.

See Also

Permutation statistics for brain imaging: brainScores, brainPvals

True discovery guarantee for cluster analysis: brainAnalysis

Examples

```
# simulate 20 copes with dimensions 10x10x10
set.seed(42)
copes <- list()
for(i in seq(20)){copes[[i]] <- array(rnorm(10^3, mean = -10, sd = 30), dim=c(10,10,10))}

# cluster map where t scores are grater than 2.8, in absolute value
thr <- 2.8
cl <- brainClusters(copes = copes, thr = thr)

# create object of class sumBrain
res <- brainScores(copes = copes, alpha = 0.2, seed = 42, truncFrom = thr)
res
summary(res)

# confidence bound for the number of true discoveries and the TDP within clusters
out <- brainAnalysis(res, clusters = cl$clusters)
out$summary
```

brainPvals *Permutation p-Values for Brain Imaging*

Description

This function computes p-value combinations for different permutations of brain imaging data. A voxel's p-value is calculated by performing the one-sample t test for the null hypothesis that its mean contrast over the different subjects is zero.

Usage

```
brainPvals(copes, mask = NULL, alternative = "two.sided", alpha = 0.05, B = 200,
           seed = NULL, truncFrom = NULL, truncTo = 0.5,
           type = "vovk.wang", r = 0, rand = FALSE)
```

Arguments

copes	list of 3D numeric arrays (contrasts maps for each subject).
mask	3D logical array, where TRUE values correspond to voxels inside the brain, or character for a Nifti file name.
alternative	direction of the alternative hypothesis (greater, lower, two.sided).
alpha	significance level.
B	number of permutations, including the identity.
seed	seed.
truncFrom	truncation parameter: values greater than truncFrom are truncated. If NULL, it is set to alpha.
truncTo	truncation parameter: truncated values are set to truncTo. If NULL, p-values are not truncated.
type	p-value combination among edgington, fisher, pearson, liptak, cauchy, harmonic, vovk.wang (see details).
r	parameter for Vovk and Wang's p-value combination.
rand	logical, TRUE to compute p-values by permutation distribution.

Details

A p-value p is transformed as following.

- Edgington: p (Edgington, 1972)
- Fisher: -2log(p) (Fisher, 1925)
- Pearson: 2log(1-p) (Pearson, 1933)
- Liptak: qnorm(1-p) (Liptak, 1958; Stouffer et al., 1949)
- Cauchy: tan[(0.5-p)pi] with pi=3.142 (Liu and Xie, 2020)
- Harmonic mean: 1/p (Wilson, 2019)

- Vovk and Wang: `p^r` (`log(p)` for `r=0`) (Vovk and Wang, 2020)

An error message is returned if the transformation produces infinite values.

For Vovk and Wang, `r=0` corresponds to Fisher, and `r=-1` to the harmonic mean.

Truncation parameters should be such that `truncTo` is not smaller than `truncFrom`. As Pearson's and Liptak's transformations produce infinite values in 1, for such methods `truncTo` should be strictly smaller than 1.

The significance level `alpha` should be in the interval [1/`B`, 1).

Value

`brainPvals` returns an object of class `sumBrain`, containing

- `statistics`: numeric matrix of p-values, where columns correspond to voxels inside the brain, and rows to permutations. The first permutation is the identity
- `mask`: 3D logical array, where `TRUE` values correspond to voxels inside the brain
- `alpha`: significance level
- `truncFrom`: transformed first truncation parameter
- `truncTo`: transformed second truncation parameter

Author(s)

Anna Vesely.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Vesely A., Finos L., and Goeman J. J. (2023). Permutation-based true discovery guarantee by sum tests. Journal of the Royal Statistical Society, Series B (Statistical Methodology), doi: 10.1093/jrsssb/qkad019.

See Also

Permutation statistics for brain imaging using t scores: `brainScores`

True discovery guarantee for cluster analysis: `brainAnalysis`

Suprathreshold clusters: `brainClusters`

Examples

```
# simulate 20 copes with dimensions 10x10x10
set.seed(42)
copes <- list()
for(i in seq(20)){copes[[i]] <- array(rnorm(10^3, mean = -10, sd = 30), dim=c(10,10,10))}

# cluster map where t scores are grater than 2.8, in absolute value
thr <- 2.8
cl <- brainClusters(copes = copes, thr = thr)
```

```
# create object of class sumBrain (combination: Cauchy)
res <- brainPvals(copes = copes, alpha = 0.2, seed = 42, type = "cauchy")
res
summary(res)

# confidence bound for the number of true discoveries and the TDP within clusters
out <- brainAnalysis(res, clusters = cl$clusters)
out$summary
```

`brainScores` *Permutation t-Scores for Brain Imaging*

Description

This function computes t-scores for different permutations of brain imaging data. A voxel's score is calculated by performing the one-sample t test for the null hypothesis that its mean contrast over the different subjects is zero.

Usage

```
brainScores(copes, mask = NULL, alternative = "two.sided", alpha = 0.05, B = 200,
            seed = NULL, truncFrom = 3.2, truncTo = 0, squares = FALSE)
```

Arguments

`copes`	list of 3D numeric arrays (contrasts maps for each subject).
`mask`	3D logical array, where `TRUE` values correspond to voxels inside the brain, or character for a Nifti file name.
`alternative`	direction of the alternative hypothesis (`greater`, `lower`, `two.sided`).
`alpha`	significance level.
`B`	number of permutations, including the identity.
`seed`	seed.
`truncFrom`	truncation parameter: values less extreme than `truncFrom` are truncated. If `NULL`, statistics are not truncated.
`truncTo`	truncation parameter: truncated values are set to `truncTo`. If `NULL`, statistics are not truncated.
`squares`	logical, `TRUE` to use squared t-scores.

Details

Truncation parameters should be such that `truncTo` is not more extreme than `truncFrom`.

The significance level `alpha` should be in the interval [1/`B`, 1).

Value

`brainScores` returns an object of class `sumBrain`, containing

- `statistics`: numeric matrix of t-scores, where columns correspond to voxels inside the brain, and rows to permutations. The first permutation is the identity
- `mask`: 3D logical array, where `TRUE` values correspond to voxels inside the brain
- `alpha`: significance level
- `truncFrom`: transformed first truncation parameter
- `truncTo`: transformed second truncation parameter

Author(s)

Anna Vesely.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Vesely A., Finos L., and Goeman J. J. (2023). Permutation-based true discovery guarantee by sum tests. Journal of the Royal Statistical Society, Series B (Statistical Methodology), doi: 10.1093/jrsssb/qkad019.

See Also

Permutation statistics for brain imaging using p-values: `brainPvals`

True discovery guarantee for cluster analysis: `brainAnalysis`

Suprathreshold clusters: `brainClusters`

Examples

```
# simulate 20 copes with dimensions 10x10x10
set.seed(42)
copes <- list()
for(i in seq(20)){copes[[i]] <- array(rnorm(10^3, mean = -10, sd = 30), dim=c(10,10,10))}

# cluster map where t scores are grater than 2.8, in absolute value
thr <- 2.8
cl <- brainClusters(copes = copes, thr = thr)

# create object of class sumBrain
res <- brainScores(copes = copes, alpha = 0.2, seed = 42, truncFrom = thr)
res
summary(res)

# confidence bound for the number of true discoveries and the TDP within clusters
out <- brainAnalysis(res, clusters = cl$clusters)
out$summary
```

`discoveries` *Confidence Bound for the Number of True Discoveries*

Description

This function determines a lower confidence bound for the number of true discoveries within a set of interest. The bound remains valid under post-hoc selection.

Usage

```
discoveries(object)

## S3 method for class 'sumObj'
discoveries(object)
```

Arguments

`object`	an object of class `sumObj`, as returned by the functions `sumStats` and `sumPvals`.

Value

`discoveries` returns a lower (1-`alpha`)-confidence bound for the number of true discoveries in the set.

Author(s)

Anna Vesely.

See Also

Create a `sumObj` object: `sumStats`, `sumPvals`

Lower confidence bound for the TDP: `tdp`

Upper confidence bound for the FDP: `fdp`

Examples

```
# generate matrix of p-values for 5 variables and 10 permutations
G <- simData(prop = 0.6, m = 5, B = 10, alpha = 0.4, seed = 42)

# subset of interest (variables 1 and 2)
S <- c(1,2)

# create object of class sumObj
# combination: harmonic mean (Vovk and Wang with r = -1)
res <- sumPvals(G, S, alpha = 0.4, r = -1)
res
summary(res)
```

```
# lower confidence bound for the number of true discoveries in S
discoveries(res)

# lower confidence bound for the true discovery proportion in S
tdp(res)

# upper confidence bound for the false discovery proportion in S
fdp(res)
```

`fdp` *Confidence Bound for the FDP*

Description

This function determines an upper confidence bound for the false discovery proportion within a set of interest. The bound remains valid under post-hoc selection.

Usage

```
fdp(object)

## S3 method for class 'sumObj'
fdp(object)
```

Arguments

`object`	an object of class `sumObj`, as returned by the functions `sumStats` and `sumPvals`.

Value

`fdp` returns an upper (1-`alpha`)-confidence bound for the false discovery proportion in the set.

Author(s)

Anna Vesely.

See Also

Create a `sumObj` object: `sumStats`, `sumPvals`

Lower confidence bound for the number of true discoveries: `discoveries`

Lower confidence bound for the TDP: `tdp`

Examples

```
# generate matrix of p-values for 5 variables and 10 permutations
G <- simData(prop = 0.6, m = 5, B = 10, alpha = 0.4, seed = 42)

# subset of interest (variables 1 and 2)
S <- c(1,2)

# create object of class sumObj
# combination: harmonic mean (Vovk and Wang with r = -1)
res <- sumPvals(G, S, alpha = 0.4, r = -1)
res
summary(res)

# lower confidence bound for the number of true discoveries in S
discoveries(res)

# lower confidence bound for the true discovery proportion in S
tdp(res)

# upper confidence bound for the false discovery proportion in S
fdp(res)
```

`geneAnalysis` *True Discovery Guarantee for Pathway Analysis of Gene Expression Data*

Description

This function uses permutation t-statistics/p-values to determine a true discovery guarantee for gene pathway analysis. It computes confidence bounds for the number of true discoveries and the true discovery proportion within each cluster. The bounds are simultaneous over all sets, and remain valid under post-hoc selection.

Usage

```
geneAnalysis(sumGene, pathways = NULL, nMax = 50, silent = FALSE)
```

Arguments

`sumGene`	an object of class sumGene, as returned by the functions `geneScores` and `genePvals`.
`pathways`	list of character vectors containing gene names (one vector per pathway). If NULL, the whole gene set is considered.
`nMax`	maximum number of iterations per cluster.
`silent`	logical, `FALSE` to print a summary of active pathways.

Value

`geneAnalysis` returns a data frame containing, for each pathway,

- `size`: size
- `TD`: lower (1-`alpha`)-confidence bound for the number of true discoveries
- `maxTD`: maximum value of `TD` that could be found under convergence of the algorithm
- `TDP`: lower (1-`alpha`)-confidence bound for the true discovery proportion
- `maxTD`: maximum value of `TDP` that could be found under convergence of the algorithm.

Author(s)

Anna Vesely.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Vesely A., Finos L., and Goeman J. J. (2023). Permutation-based true discovery guarantee by sum tests. Journal of the Royal Statistical Society, Series B (Statistical Methodology), doi: 10.1093/jrsssb/qkad019.

See Also

Permutation statistics for gene expression: `geneScores`, `genePvals`

Examples

```
# simulate 20 samples of 100 genes
set.seed(42)
expr <- matrix(c(rnorm(1000, mean = 0, sd = 10), rnorm(1000, mean = 13, sd = 10)), ncol = 20)
rownames(expr) <- seq(100)
labels <- rep(c(1,2), each = 10)

# simulate pathways
pathways <- lapply(seq(3), FUN = function(x) sample(rownames(expr), 3*x))

# create object of class sumGene
res <- geneScores(expr = expr, labels = labels, alpha = 0.2, seed = 42)
res
summary(res)

# confidence bound for the number of true discoveries and the TDP within pathways
out <- geneAnalysis(res, pathways = pathways)
out
```

genePvals *Permutation p-Values for Gene Expression*

Description

This function computes p-value combinations for different permutations of gene expression data. A gene's p-value is calculated by performing the two-sample t test for the null hypothesis that the mean expression value is the same between two populations.

Usage

```
genePvals(expr, labels, alternative = "two.sided", alpha = 0.05, B = 200, seed = NULL,
        truncFrom = NULL, truncTo = 0.5, type = "vovk.wang", r = 0, rand = FALSE)
```

Arguments

expr	matrix where rows correspond to genes, and columns to samples.
labels	numeric/character vector with two levels, denoting the population of each sample.
alternative	direction of the alternative hypothesis (greater, lower, two.sided).
alpha	significance level.
B	number of permutations, including the identity.
seed	seed.
truncFrom	truncation parameter: values greater than truncFrom are truncated. If NULL, it is set to alpha.
truncTo	truncation parameter: truncated values are set to truncTo. If NULL, p-values are not truncated.
type	p-value combination among edgington, fisher, pearson, liptak, cauchy, harmonic, vovk.wang (see details).
r	parameter for Vovk and Wang's p-value combination.
rand	logical, TRUE to compute p-values by permutation distribution.

Details

A p-value p is transformed as following.

- Edgington: p (Edgington, 1972)
- Fisher: -2log(p) (Fisher, 1925)
- Pearson: 2log(1-p) (Pearson, 1933)
- Liptak: qnorm(1-p) (Liptak, 1958; Stouffer et al., 1949)
- Cauchy: tan[(0.5-p)pi] with pi=3.142 (Liu and Xie, 2020)
- Harmonic mean: 1/p (Wilson, 2019)

- Vovk and Wang: `p^r` (`log(p)` for `r=0`) (Vovk and Wang, 2020)

An error message is returned if the transformation produces infinite values.

For Vovk and Wang, `r=0` corresponds to Fisher, and `r=-1` to the harmonic mean.

Truncation parameters should be such that `truncTo` is not smaller than `truncFrom`. As Pearson's and Liptak's transformations produce infinite values in 1, for such methods `truncTo` should be strictly smaller than 1.

The significance level `alpha` should be in the interval [1/`B`, 1).

Value

`genePvals` returns an object of class `sumGene`, containing

- `statistics`: numeric matrix of p-values, where columns correspond to genes, and rows to permutations. The first permutation is the identity
- `alpha`: significance level
- `truncFrom`: transformed first truncation parameter
- `truncTo`: transformed second truncation parameter

Author(s)

Anna Vesely.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Vesely A., Finos L., and Goeman J. J. (2023). Permutation-based true discovery guarantee by sum tests. Journal of the Royal Statistical Society, Series B (Statistical Methodology), doi: 10.1093/jrsssb/qkad019.

See Also

Permutation statistics for gene expression using t scores: `geneScores`

True discovery guarantee for cluster analysis: `geneAnalysis`

Examples

```
# simulate 20 samples of 100 genes
set.seed(42)
expr <- matrix(c(rnorm(1000, mean = 0, sd = 10), rnorm(1000, mean = 13, sd = 10)), ncol = 20)
rownames(expr) <- seq(100)
labels <- rep(c(1,2), each = 10)

# simulate pathways
pathways <- lapply(seq(3), FUN = function(x) sample(rownames(expr), 3*x))

# create object of class sumGene
res <- genePvals(expr = expr, labels = labels, alpha = 0.2, seed = 42, type = "liptak")
res
```

```
summary(res)

# confidence bound for the number of true discoveries and the TDP within pathways
out <- geneAnalysis(res, pathways = pathways)
out
```

---

geneScores *Permutation t-Scores for Gene Expression*

Description

This function computes t-scores for different permutations of gene expression data. A gene's score is calculated by performing the two-sample t test for the null hypothesis that the mean expression value is the same between two populations.

Usage

```
geneScores(expr, labels, alternative = "two.sided", alpha = 0.05, B = 200, seed = NULL,
           truncFrom = 3.2, truncTo = 0, squares = FALSE)
```

Arguments

expr	matrix where rows correspond to genes, and columns to samples.
labels	numeric/character vector with two levels, denoting the population of each sample.
alternative	direction of the alternative hypothesis (greater, lower, two.sided).
alpha	significance level.
B	number of permutations, including the identity.
seed	seed.
truncFrom	truncation parameter: values less extreme than truncFrom are truncated. If NULL, statistics are not truncated.
truncTo	truncation parameter: truncated values are set to truncTo. If NULL, statistics are not truncated.
squares	logical, TRUE to use squared t-scores.

Details

Truncation parameters should be such that truncTo is not more extreme than truncFrom.

The significance level alpha should be in the interval [1/B, 1).

Value

geneScores returns an object of class sumGene, containing

- statistics: numeric matrix of scores, where columns correspond to genes, and rows to permutations. The first permutation is the identity
- alpha: significance level
- truncFrom: transformed first truncation parameter
- truncTo: transformed second truncation parameter

Author(s)

Anna Vesely.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Vesely A., Finos L., and Goeman J. J. (2023). Permutation-based true discovery guarantee by sum tests. Journal of the Royal Statistical Society, Series B (Statistical Methodology), doi: 10.1093/jrsssb/qkad019.

See Also

Permutation statistics for gene expression using p-values: genePvals

True discovery guarantee for cluster analysis: geneAnalysis

Examples

```
# simulate 20 samples of 100 genes
set.seed(42)
expr <- matrix(c(rnorm(1000, mean = 0, sd = 10), rnorm(1000, mean = 13, sd = 10)), ncol = 20)
rownames(expr) <- seq(100)
labels <- rep(c(1,2), each = 10)

# simulate pathways
pathways <- lapply(seq(3), FUN = function(x) sample(rownames(expr), 3*x))

# create object of class sumGene
res <- geneScores(expr = expr, labels = labels, alpha = 0.2, seed = 42)
res
summary(res)

# confidence bound for the number of true discoveries and the TDP within pathways
out <- geneAnalysis(res, pathways = pathways)
out
```

simData *Simulating Matrix of Statistics*

Description

This function simulates a matrix of permutation statistics, by performing a t test on normal data.

Usage

```
simData(prop, m, B = 200, rho = 0, n = 50, alpha = 0.05, pw = 0.8, p = TRUE, seed = NULL)
```

Arguments

prop	proportion of non-null hypotheses.
m	total number of variables.
B	number of permutations, including the identity.
rho	level of equicorrelation between pairs of variables.
n	number of observations.
alpha	significance level.
pw	power of the t test.
p	logical, TRUE to compute p-values, FALSE to compute t-scores.
seed	seed.

Details

The function applies the one-sample two-sided t test to a matrix of simulated data, for B data permutations. Data is obtained by simulating n independent observations from a multivariate normal distribution, where a proportion prop of the variables has non-null mean. This mean is such that the one-sample t test with significance level alpha has power equal to pw. Each pair of distinct variables has equicorrelation rho.

Value

simData returns a matrix where the B rows correspond to permutations (the first is the identity), and the m columns correspond to variables. The matrix contains p-values if p is TRUE, and t-scores otherwise. The first columns (a proportion prop) correspond to non-null hypotheses.

Author(s)

Anna Vesely.

See Also

True discovery guarantee: sumStats, sumPvals

Examples

```
# generate matrix of p-values for 5 variables and 10 permutations
G <- simData(prop = 0.6, m = 5, B = 10, alpha = 0.4, seed = 42)

# subset of interest (variables 1 and 2)
S <- c(1,2)

# create object of class sumObj
# combination: harmonic mean (Vovk and Wang with r = -1)
res <- sumPvals(G, S, alpha = 0.4, r = -1)
res
summary(res)

# lower confidence bound for the number of true discoveries in S
discoveries(res)

# lower confidence bound for the true discovery proportion in S
tdp(res)

# upper confidence bound for the false discovery proportion in S
fdp(res)
```

`sumPvals` *True Discovery Guarantee for p-Value Combinations - Permutation*

Description

This function uses permutation p-values to determine confidence bounds for the number of true discoveries, the true discovery proportion and the false discovery proportion within a set of interest. The bounds are simultaneous over all sets, and remain valid under post-hoc selection.

Usage

```
sumPvals(G, S = NULL, alpha = 0.05, truncFrom = NULL, truncTo = 0.5,
         type = "vovk.wang", r = 0, nMax = 50)
```

Arguments

`G`	numeric matrix of p-values, where columns correspond to variables, and rows to data transformations (e.g. permutations). The first transformation is the identity.
`S`	vector of indices for the variables of interest (if not specified, all variables).
`alpha`	significance level.
`truncFrom`	truncation parameter: values greater than `truncFrom` are truncated. If `NULL`, it is set to `alpha`.
`truncTo`	truncation parameter: truncated values are set to `truncTo`. If `NULL`, p-values are not truncated.

type	p-value combination among edgington, fisher, pearson, liptak, cauchy, harmonic, vovk.wang (see details).
r	parameter for Vovk and Wang's p-value combination.
nMax	maximum number of iterations.

Details

A p-value p is transformed as following.

- Edgington: p (Edgington, 1972)
- Fisher: -2log(p) (Fisher, 1925)
- Pearson: 2log(1-p) (Pearson, 1933)
- Liptak: qnorm(1-p) (Liptak, 1958; Stouffer et al., 1949)
- Cauchy: tan[(0.5-p)pi] with pi=3.142 (Liu and Xie, 2020)
- Harmonic mean: 1/p (Wilson, 2019)
- Vovk and Wang: p^r (log(p) for r=0) (Vovk and Wang, 2020)

An error message is returned if the transformation produces infinite values.

For Vovk and Wang, r=0 corresponds to Fisher, and r=-1 to the harmonic mean.

Truncation parameters should be such that truncTo is not smaller than truncFrom. As Pearson's and Liptak's transformations produce infinite values in 1, for such methods truncTo should be strictly smaller than 1.

The significance level alpha should be in the interval [1/B, 1), where B is the number of data transformations (rows in G).

Value

sumPvals returns an object of class sumObj, containing

- total: total number of variables (columns in G)
- size: size of S
- alpha: significance level
- TD: lower (1-alpha)-confidence bound for the number of true discoveries in S
- maxTD: maximum value of TD that could be found under convergence of the algorithm
- iterations: number of iterations of the algorithm

Author(s)

Anna Vesely.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Vesely A., Finos L., and Goeman J. J. (2023). Permutation-based true discovery guarantee by sum tests. Journal of the Royal Statistical Society, Series B (Statistical Methodology), doi: 10.1093/jrsssb/qkad019.

See Also

True discovery guarantee using generic statistics: sumStats

Access a sumObj object: discoveries, tdp, fdp

Examples

```
# generate matrix of p-values for 5 variables and 10 permutations
G <- simData(prop = 0.6, m = 5, B = 10, alpha = 0.4, seed = 42)

# subset of interest (variables 1 and 2)
S <- c(1,2)

# create object of class sumObj
# combination: harmonic mean (Vovk and Wang with r = -1)
res <- sumPvals(G, S, alpha = 0.4, r = -1)
res
summary(res)

# lower confidence bound for the number of true discoveries in S
discoveries(res)

# lower confidence bound for the true discovery proportion in S
tdp(res)

# upper confidence bound for the false discovery proportion in S
fdp(res)
```

---

sumPvalsPar *True Discovery Guarantee for p-Value Combinations - Parametric*

Description

This function uses p-values to determine confidence bounds for the number of true discoveries, the true discovery proportion and the false discovery proportion within a set of interest. The bounds are simultaneous over all sets, and remain valid under post-hoc selection.

Usage

```
sumPvalsPar(g, S = NULL, alpha = 0.05, type = "vovk.wang", r = 0, independence = NULL)
```

Arguments

g	numeric vector of p-values.
S	vector of indices for the variables of interest (if not specified, all variables).
alpha	significance level.
type	p-value combination among fisher, pearson, liptak, cauchy, harmonic, vovk.wang (see details).

r	parameter for Vovk and Wang's p-value combination.
independence	logical, TRUE to assume independence, FALSE for general dependence structure. If not specified, it is set to FALSE for vovk.wang, and TRUE otherwise.

Details

A p-value p is transformed as following.

- Fisher: -2log(p) (Fisher, 1925)
- Pearson: 2log(1-p) (Pearson, 1933)
- Liptak: qnorm(1-p) (Liptak, 1958; Stouffer et al., 1949)
- Cauchy: tan[(0.5-p)pi] with pi=3.142 (Liu and Xie, 2020)
- Harmonic mean: 1/p (Wilson, 2019)
- Vovk and Wang: p^r (log(p) for r=0) (Vovk and Wang, 2020)

An error message is returned if the transformation produces infinite values.

For Vovk and Wang, r=-Inf corresponds to the minimum p-value, r=Inf to the maximum p-value, r=0 to Fisher, and r=-1 to the harmonic mean.

Under independence, for Vovk and Wang the test is defined only for r=0 and r=1. Under general dependence, the test is defined only for Fisher, the harmonic mean and Vovk and Wang.

For combinations that are not implemented, if the vector of critical values is known the method can be applied through sumStatsPar. Please contact us to implement other known vectors of critical values that do not currently appear.

Value

sumPvalsPar returns an object of class sumObj, containing

- total: total number of variables (length of g)
- size: size of S
- alpha: significance level
- TD: lower (1-alpha)-confidence bound for the number of true discoveries in S
- maxTD: maximum value of TD that could be found under convergence of the algorithm
- iterations: number of iterations of the algorithm (NULL)

Author(s)

Xu Chen.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Tian J., Chen X., Katsevich E., Goeman J. J. and Ramdas A. (2022). Large-scale simultaneous inference under dependence. Scandinavian Journal of Statistics, doi: 10.1111/sjos.12614.

See Also

True discovery guarantee using generic statistics (parametric): sumStatsPar

Access a sumObj object: discoveries, tdp, fdp

Examples

```
# generate vector of p-values for 5 variables
g <- as.vector(simData(prop = 0.6, m = 5, B = 1, alpha = 0.4, seed = 42))

# subset of interest (variables 1 and 2)
S <- c(1,2)

# create object of class sumObj
# combination: harmonic mean under general dependence
res <- sumPvalsPar(g, S, alpha = 0.4, type = "harmonic", independence = FALSE)
res
summary(res)

# lower confidence bound for the number of true discoveries in S
discoveries(res)

# lower confidence bound for the true discovery proportion in S
tdp(res)

# upper confidence bound for the false discovery proportion in S
fdp(res)
```

---

sumStats *True Discovery Guarantee for Generic Statistics - Permutation*

Description

This function uses generic permutation statistics to determine confidence bounds for the number of true discoveries, the true discovery proportion and the false discovery proportion within a set of interest. The bounds are simultaneous over all sets, and remain valid under post-hoc selection.

Usage

```
sumStats(G, S = NULL, alternative = "greater", alpha = 0.05,
         truncFrom = NULL, truncTo = NULL, nMax = 50)
```

Arguments

G	numeric matrix of statistics, where columns correspond to variables, and rows to data transformations (e.g. permutations). The first transformation is the identity.
S	vector of indices for the variables of interest (if not specified, all variables).
alternative	direction of the alternative hypothesis (greater, lower, two.sided).

alpha	significance level.
truncFrom	truncation parameter: values less extreme than truncFrom are truncated. If NULL, statistics are not truncated.
truncTo	truncation parameter: truncated values are set to truncTo. If NULL, statistics are not truncated.
nMax	maximum number of iterations.

Details

Truncation parameters should be such that truncTo is not more extreme than truncFrom.

The significance level alpha should be in the interval [1/B, 1), where B is the number of data transformations (rows in G).

Value

sumStats returns an object of class sumObj, containing

- total: total number of variables (columns in G)
- size: size of S
- alpha: significance level
- TD: lower (1-alpha)-confidence bound for the number of true discoveries in S
- maxTD: maximum value of TD that could be found under convergence of the algorithm
- iterations: number of iterations of the algorithm

Author(s)

Anna Vesely.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Vesely A., Finos L., and Goeman J. J. (2023). Permutation-based true discovery guarantee by sum tests. Journal of the Royal Statistical Society, Series B (Statistical Methodology), doi: 10.1093/jrsssb/qkad019.

See Also

True discovery guarantee using p-values: sumPvals

Access a sumObj object: discoveries, tdp, fdp

Examples

```
# generate matrix of t-scores for 5 variables and 10 permutations
G <- simData(prop = 0.6, m = 5, B = 10, alpha = 0.4, p = FALSE, seed = 42)

# subset of interest (variables 1 and 2)
S <- c(1,2)
```

```
# create object of class sumObj
res <- sumStats(G, S, alpha = 0.4, truncFrom = 0.7, truncTo = 0)
res
summary(res)

# lower confidence bound for the number of true discoveries in S
discoveries(res)

# lower confidence bound for the true discovery proportion in S
tdp(res)

# upper confidence bound for the false discovery proportion in S
fdp(res)
```

`sumStatsPar` *True Discovery Guarantee for Generic Statistics - Parametric*

Description

This function uses generic statistics and a suitable vector of critical values to determine confidence bounds for the number of true discoveries, the true discovery proportion and the false discovery proportion within a set of interest. The bounds are simultaneous over all sets, and remain valid under post-hoc selection.

Usage

```
sumStatsPar(g, S = NULL, alpha = 0.05, cvs)
```

Arguments

`g`	numeric vector of statistics.
`S`	vector of indices for the variables of interest (if not specified, all variables).
`alpha`	significance level.
`cvs`	numeric vector of critical values for summed statistics considering `1:m` hypotheses.

Value

`sumStatsPar` returns an object of class `sumObj`, containing

- `total`: total number of variables (length of `g`)
- `size`: size of `S`
- `alpha`: significance level
- `TD`: lower (1-`alpha`)-confidence bound for the number of true discoveries in `S`
- `maxTD`: maximum value of `TD` that could be found under convergence of the algorithm
- `iterations`: number of iterations of the algorithm (`NULL`)

Author(s)

Xu Chen.

References

Goeman J. J. and Solari A. (2011). Multiple testing for exploratory research. Statistical Science, doi: 10.1214/11-STS356.

Tian J., Chen X., Katsevich E., Goeman J. J. and Ramdas A. (2022). Large-scale simultaneous inference under dependence. Scandinavian Journal of Statistics, doi: 10.1111/sjos.12614.

See Also

True discovery guarantee using p-values (parametric): `sumPvalsPar`

Access a `sumObj` object: `discoveries`, `tdp`, `fdp`

Examples

```
# generate vector of statistics for 5 variables (Fisher transformation of p-values)
g <- as.vector(simData(prop = 0.6, m = 5, B = 1, alpha = 0.4, seed = 42))
g <- -2 * log(g)

# subset of interest (variables 1 and 2)
S <- c(1,2)

# vector of critical values
cvs <- qchisq(p = 0.4, df = 2 * seq(5), lower.tail=FALSE)

# create object of class sumObj
res <- sumStatsPar(g, S, alpha = 0.4, cvs = cvs)
res
summary(res)

# lower confidence bound for the number of true discoveries in S
discoveries(res)

# lower confidence bound for the true discovery proportion in S
tdp(res)

# upper confidence bound for the false discovery proportion in S
fdp(res)
```

---

tdp *Confidence Bound for the TDP*

Description

This function determines a lower confidence bound for the true discovery proportion within a set of interest. The bound remains valid under post-hoc selection.

Usage

```
tdp(object)

## S3 method for class 'sumObj'
tdp(object)
```

Arguments

object	an object of class sumObj, as returned by the functions sumStats and sumPvals.

Value

tdp returns a lower (1-alpha)-confidence bound for the true discovery proportion in the set.

Author(s)

Anna Vesely.

See Also

Create a sumObj object: sumStats, sumPvals

Lower confidence bound for the number of true discoveries: discoveries

Upper confidence bound for the FDP: fdp

Examples

```
# generate matrix of p-values for 5 variables and 10 permutations
G <- simData(prop = 0.6, m = 5, B = 10, alpha = 0.4, seed = 42)

# subset of interest (variables 1 and 2)
S <- c(1,2)

# create object of class sumObj
# combination: harmonic mean (Vovk and Wang with r = -1)
res <- sumPvals(G, S, alpha = 0.4, r = -1)
res
summary(res)

# lower confidence bound for the number of true discoveries in S
discoveries(res)

# lower confidence bound for the true discovery proportion in S
tdp(res)

# upper confidence bound for the false discovery proportion in S
fdp(res)
```

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