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HMP-package

Hypothesis Testing and Power Calculations for Comparing Metagenomics Samples

Description

This package provides tools for generating data matrices following Multinomial and Dirichlet-Multinomial distributions, computing the following test-statistics and their corresponding p-values, and computing the power and size of the tests described above using Monte-Carlo simulations.

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In addition to hypothesis testing and power calculations you can:

1. Perform basic data management to exclude samples with fewer than pre-specified number of reads, collapse rare taxa and order the taxa by frequency. This is useful to exclude failed samples (i.e. samples with very few reads) - Data.filter
2. Plot your data - Barchart.data
3. Generate random sample of Dirichlet Multinomial data with pre-specified parameters - Dirichlet.multinomial

Note: Though the description of the functions refer its application to ranked abundance distributions (RAD) data, every function is also applicable to model species abundance data. See references for a discussion and application to both type of ecological data.

Author(s)
Patricio S. La Rosa, Elena Deych, Berkley Shands, Sharina Carter, William D. Shannon

References

Barchart.data

A Graphical Representation of Taxa Proportions

Description
Creates a bar plot of taxonomic proportions.

Usage
Barchart.data(data, title = "Taxa Proportions")

Arguments
- data: A matrix of taxonomic counts(columns) for each sample(rows).
- title: A string to be used as the plots title. The default is "Taxa Proportions".

Value
A bar plot of taxonomic proportions for all samples at a given taxonomic level.

Author(s)
Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon
**Examples**

```r
data(saliva)
Barchart.data(saliva)
```

---

**C.alpha.multinomial**  
*C(α) - Optimal Test for Assessing Multinomial Goodness of Fit Versus Dirichlet-Multinomial Alternative*

**Description**

A function to compute the $C(\alpha)$-optimal test statistics of Kim and Margolin (1992) for evaluating the Goodness-of-Fit of a Multinomial distribution (null hypothesis) versus a Dirichlet-Multinomial distribution (alternative hypothesis).

**Usage**

```r
C.alpha.multinomial(data)
```

**Arguments**

- `data` A matrix of taxonomic counts(columns) for each sample(rows).

**Details**

In order to test if a set of ranked-abundance distribution (RAD) from microbiome samples can be modeled better using a multinomial or Dirichlet-Multinomial distribution, we test the hypothesis $H : \theta = 0$ versus $H : \theta \neq 0$, where the null hypothesis implies a multinomial distribution and the alternative hypothesis implies a DM distribution. Kim and Margolin (Kim and Margolin, 1992) proposed a $C(\alpha)$-optimal test statistics given by,

$$T = \sum_{j=1}^{K} \sum_{i=1}^{P} \frac{1}{\sum_{i=1}^{P} x_{ij}} \left( x_{ij} - \frac{N_i \sum_{i=1}^{P} x_{ij}}{N_g} \right)^2$$

Where $K$ is the number of taxa, $P$ is the number of samples, $x_{ij}$ is the taxon $j$, $j = 1, \ldots, K$ from sample $i$, $i = 1, \ldots, P$, $N_i$ is the number of reads in sample $i$, and $N_g$ is the total number of reads across samples.

As the number of reads increases, the distribution of the $T$ statistic converges to a Chi-square with degrees of freedom equal to $(P - 1)(K - 1)$, when the number of sequence reads is the same in all samples. When the number of reads is not the same in all samples, the distribution becomes a weighted Chi-square with a modified degree of freedom (see (Kim and Margolin, 1992) for more details).

Note: Each taxa in `data` should be present in at least 1 sample, a column with all 0’s may result in errors and/or invalid results.
**Description**

This function creates a new dataset from an existing one by ordering taxa in order of decreasing abundance, collapsing less-abundant taxa into one category as specified by the user and excluding samples with a total number of reads fewer than the user-specified value.

**Usage**

```r
Data.filter(data, order.type = "data", minReads = 0, numTaxa = NULL, perTaxa = NULL, K = NULL, reads.crit = NULL)
```

**Arguments**

- `data`: A matrix of taxonomic counts(columns) for each sample(rows).
- `order.type`: If "sample": Rank taxa based on its taxonomic frequency. If "data": Rank taxa based on cumulative taxonomic counts across all samples (default).
- `minReads`: Samples with a total number of reads less than read.crit value will be deleted.
- `numTaxa`: The number of taxa to keep, while collapsing the other (less abundant) taxa. Only one argument, numTaxa or perTaxa should be specified.
- `perTaxa`: The combined percentage of data to keep, while collapsing the remaining taxa. Only one argument, numTaxa or perTaxa should be specified.
- `K`: Deprecated. Replaced with numTaxa for clarity.
- `reads.crit`: Deprecated. Replaced with minReads for clarity.
Value

A data frame of taxa and samples with a total number of reads greater than the minimum value. The last taxon labeled 'Other' contains the sum of the least abundant taxa collapsed by setting 'numTaxa' or 'perTaxa'.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

Examples

data(saliva)

### Excludes all samples with fewer than 1000 reads and collapses
### taxa with 11th or smaller abundance into one category
filterDatNum <- Data.filter(saliva, "data", 1000, numTaxa=10)

### Excludes all samples with fewer than 1000 reads and collapses
### the least abundant taxa to keep as close to 85% of the data as
### possible
filterDatPer <- Data.filter(saliva, "data", 1000, perTaxa=.95)

dim(saliva)
dim(filterDatNum)
dim(filterDatPer)

Dirichlet.multinomial  Generation of Dirichlet-Multinomial Random Samples

Description

Random generation of Dirichlet-Multinomial samples.

Usage

Dirichlet.multinomial(Nrs, shape)

Arguments

Nrs  A vector specifying the number of reads or sequence depth for each sample.
shape  A vector of Dirichlet parameters for each taxa.
Details

The Dirichlet-Multinomial distribution is given by (Mosimann, J. E. (1962); Tvedebrink, T. (2010)),

\[ P\left(X_i = x_i;\{\pi_j\},\theta\right) = \frac{N_i!}{x_{i1}!\cdots x_{iK}!} \prod_{j=1}^{K} \frac{\pi_{ij}}{x_{ij}} \left(\frac{\pi_j}{1-\theta} + (r-1)\theta\right)^{x_{ij}} \prod_{r=1}^{N_i} \left(\frac{1-\theta}{1-\theta} + (r-1)\theta\right) \]

where \( x_i = [x_{i1},\ldots,x_{iK}] \) is the random vector formed by \( K \) taxa (features) counts (RAD vector), \( N_i = \sum_{j=1}^{K} x_{ij} \) is the total number of reads (sequence depth), \( \{\pi_j\} \) are the mean of taxa-proportions (RAD-probability mean), and \( \theta \) is the overdispersion parameter.

Note: Though the test statistic supports an unequal number of reads across samples, the performance has not yet been fully tested.

Value

A data matrix of taxa counts where the rows are samples and columns are the taxa.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

References


Examples

data(saliva)

### Generate a the number of reads per sample
### The first number is the number of reads and the second is the number of subjects
nrs <- rep(15000, 20)

### Get gamma from the dirichlet-multinomial parameters
shape <- dirmult(saliva)$gamma
dmData <- Dirichlet.multinomial(nrs, shape)
dmData[1:5, 1:5]

---

**DM.MoM**

*Method-of-Moments (MoM) Estimators of the Dirichlet-Multinomial Parameters*

Description

Usage

DM.MoM(data)

Arguments

data A matrix of taxonomic counts(columns) for each sample(rows).

Details

Given a set of taxa-count vectors \( \{x_i, \ldots, x_P\} \), the methods of moments (MoM) estimator of the set of parameters \( \theta \) and \( \{\pi_j\}_{j=1}^K \) is given as follows (Mosimann, 1962; Tvedebrink, 2010):

\[
\hat{\pi}_j = \frac{\sum_{i=1}^P x_{ij}}{\sum_{i=1}^P N_i},
\]

and

\[
\hat{\theta} = \sum_{j=1}^K \frac{S_j - G_j}{\sum_{j=1}^K (S_j + (N_c - 1) G_j)},
\]

where \( N_c = (P - 1)^{-1} \left( \sum_{i=1}^P N_i - \left( \sum_{i=1}^P N_i \right)^{-1} \sum_{i=1}^P N_i^2 \right) \), and \( S_j = (P - 1)^{-1} \sum_{i=1}^P N_i (\hat{\pi}_{ij} - \hat{\pi}_j)^2 \),

and \( G_j = \left( \sum_{i=1}^P (N_i - 1) \right)^{-1} \sum_{i=1}^P N_i \hat{\pi}_{ij} (1 - \hat{\pi}_{ij}) \) with \( \hat{\pi}_{ij} = \frac{x_{ij}}{N_i} \).

Value

A list providing the MoM estimator for overdispersion, the MoM estimator of the RAD-probability mean vector, and the corresponding loglikelihood value for the given data set and estimated parameters.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

References


Examples

data(throat)

fit.throat <- DM.MoM(throat)
fit.throat
**Description**

Uses custom recursive partitioning (rpart) methods to find groups based on the Dirichlet-Multinomial distribution.

**Usage**

```r
DM.Rpart(data, covars, plot = TRUE, main = "", minsplit = 1, minbucket = 1, cp = 0)
```

**Arguments**

- `data` A matrix of taxonomic counts (columns) for each sample (rows).
- `covars` A matrix of covariates (columns) for each sample (rows).
- `plot` When 'TRUE' a tree plot of the results will be generated.
- `main` A string to be used as the plot title.
- `minsplit` The minimum number of observations to split on, see `rpart.control`.
- `minbucket` The minimum number of observations in any terminal node, see `rpart.control`.
- `cp` The complexity parameter, see `rpart.control`.

**Value**

An rpart object and a plot of the tree created if plot = 'TRUE'

**Author(s)**

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon

**Examples**

```r
data(saliva)
data(throat)
data(tonsils)

# Create some covariates for our data set
site <- c(rep("Saliva", nrow(saliva)), rep("Throat", nrow(throat)),
          rep("Tonsils", nrow(tonsils)))
covars <- data.frame(Group=site)

# Combine our data into a single object
data <- rbind(saliva, throat, tonsils)

rpartRes <- DM.Rpart(data, covars)
```
Dirichlet-Multinomial RPart Permutations

Description

Uses custom rpart methods to find the optimal tree splits.

Usage

```r
dmNrpartNpermHdataL covarsL plotL trueL numpermsL 1PPPL parallelL falseL coresL SL minsplitL 1L minbucketL 1L cpL PI
```

Arguments

- `data`: A matrix of taxonomic counts(columns) for each sample(rows).
- `covars`: A matrix of covariates(columns) for each sample(rows).
- `plot`: When 'TRUE' a tree plot of the results will also be generated.
- `numperms`: Number of permutations. In practice this should be at least 1,000.
- `parallel`: When this is 'TRUE' it allows for parallel calculation of the permutations. Requires the package `doParallel`.
- `cores`: The number of parallel processes to run if parallel is 'TRUE'.
- `minsplit`: The minimum number of observations to split on, see `rpart.control`. This should generally stay at 1.
- `minbucket`: The minimum number of observations in any terminal node, see `rpart.control`. This should generally stay at 1.
- `cp`: The complexity parameter, see `rpart.control`. This should generally stay at 0.

Value

A list containing the raw rpart tree, the pvalues for different splits, and data from the rpart permuting.

Author(s)

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon

Examples

```r
data(saliva)
data(throat)
data(tonsils)

### Create some covariates for our data set
site <- c(rep("Saliva", nrow(saliva)), rep("Throat", nrow(throat)), rep("Tonsils", nrow(tonsils)))
covars <- data.frame(\Group=site)
```
### Combine our data into a single object

data <- rbind(saliva, throat, tonsils)

### We use 1 for speed, should be at least 1,000

numPerms <- 1

rpartPerm <- DM.Rpart.Perm(data, covars, numPerms=numPerms)

### Pull out a tree

rpartPerm$pvals

selectedNum <- 3

bestTree <- rpart::prune(rpartPerm$rawTree, cp=rpartPerm$rawPrune[selectedNum, 2])

### Plot the best tree

rpart.plot::rpart.plot(bestTree, main="Best Tree", extra=1)

---

**Est. PI**

*Estimate the Pi Vector*

**Description**

Calculates Dirichlet-Multinomial parameters for every group using Maximum Likelihood and Method of Moments estimates: Taxa proportion estimates (Pi vector) with standard errors and Confidence intervals, as well as theta values with standard errors.

**Usage**

Est.PI(group.data, conf = .95)

**Arguments**

- **group.data**: A list of matrices of taxonomic counts(columns) for each sample(rows).
- **conf**: The desired confidence limits. The default is 95%

**Value**

A list containing the parameters: PI, SE and the upper/lower bounds of the confidence interval for every taxa, and the theta values with standard errors for both MLE and MOM.

**Author(s)**

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon
Examples

```r
## Not run:
data(saliva)
data(throat)
data(tonsils)

### Combine the data sets into a single list
group.data <- list(saliva, throat, tonsils)

### Get PI using MLE and MOM with CI
piEsts <- Est.PI(group.data)
mle <- piEsts$MLE
mom <- piEsts$MOM

## End(Not run)
```

formatDataSets (Format Data)

Description

For a list of datasets, this function finds the union of taxa across all datasets and transforms them such that they all have the same columns of taxa.

Usage

```r
formatDataSets(group.data, data)
```

Arguments

- `group.data`: A list where each element is a matrix of taxonomic counts (columns) for each sample (rows). Note that the row names should correspond to sample names.
- `data`: Deprecated. Replaced with `group.data` for clarity.

Details

This function will also sort all the columns into the same order for every dataset and remove any columns that have 0's for every sample.

E.g. For two datasets, any taxa present in dataset1 but not dataset2 will be added to dataset2 with a 0 count for all samples and vice versa.

Value

The list given, but modified so every data set has the same ordering and number of columns.

Author(s)

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon
Examples

data(saliva)
data(throat)

### Set each data set to have 10 different columns
saliva2 <- saliva[,1:10]
throat2 <- throat[,11:20]

### Combine the data sets into a single list
group.data <- list(saliva2, throat2)

formattedData <- formatDataSets(group.data)
formattedData[[1]][1:5, 1:5]

---

Gen.Alg  Find Taxa Separating Two Groups using Genetic Algorithm (GA)

Description

GA-Mantel is a fully multivariate method that uses a genetic algorithm to search over possible taxa subsets using the Mantel correlation as the scoring measure for assessing the quality of any given taxa subset.

Usage

Gen.Alg(data, covars, iters = 50, popSize = 200, earlyStop = 0, 
dataDist = "euclidean", covarDist = "gower", verbose = FALSE, 
plot = TRUE, minSolLen = NULL, maxSolLen = NULL, custCovDist = NULL)

Arguments

data  A matrix of taxonomic counts(columns) for each sample(rows).
covars  A matrix of covariates(columns) for each sample(rows).
iters  The number of times to run through the GA.
popSize  The number of solutions to test on each iteration.
earlyStop  The number of consecutive iterations without finding a better solution before stopping regardless of the number of iterations remaining. A value of '0' will prevent early stopping.
dataDist  The distance metric to use for the data. Either "euclidean" or "gower".
covarDist  The distance metric to use for the covariates. Either "euclidean" or "gower".
verbose  While 'TRUE’ the current status of the GA will be printed periodically.
plot  A boolean to plot the progress of the scoring statistics by iteration.
minSolLen  The minimum number of columns to select.
maxSolLen  The maximum number of columns to select.
custCovDist  A custom covariate distance matrix to use in place of calculating one from covars.
Details

Use a GA approach to find taxa that separate subjects based on group membership or set of covariates.

The data and covariates should be normalized BEFORE use with this function because of distance functions.

This function uses modified code from the rbga function in the genalg package. rbga

Because the GA looks at combinations and uses the raw data, taxa with a small difference in their PIs may be selected and large differences may not be.

The distance calculations use the vegdist package. vegdist

Value

A list containing

- scoreSumm: A matrix summarizing the score of the population. This can be used to figure out if the ga has come to a final solution or not. This data is also plotted if plot is ‘TRUE’.
- solutions: The final set of solutions, sorted with the highest scoring first.
- scores: The scores for the final set of solutions.
- time: How long in seconds the ga took to run.
- selected: The selected columns by name.
- nonSelected: The columns that were NOT selected by name.
- selectedIndex: The selected taxa by column number.

Author(s)

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon

Examples

```R
## Not run:
data(saliva)
data(throat)

### Combine the data into a single data frame
group.data <- list(saliva, throat)
group.data <- formatDataSets(group.data)
data <- do.call("rbind", group.data)

### Normalize the data by subject
dataNorm <- t(apply(data, 1, function(x){x/sum(x)}))

### Set covars to just be group membership
memb <- c(rep(0, nrow(saliva)), rep(1, nrow(throat)))
covars <- matrix(memb, length(memb), 1)

### We use low numbers for speed. The exact numbers to use depend
```
### Description

GA-Mantel is a fully multivariate method that uses a genetic algorithm to search over possible taxa subsets using the Mantel correlation as the scoring measure for assessing the quality of any given taxa subset.

### Usage

```r
Gen.Alg.Consensus(data, covars, consensus = .5, numRuns = 10, parallel = FALSE, cores = 3, ...)
```

### Arguments

- `data`: A matrix of taxonomic counts(columns) for each sample(rows).
- `covars`: A matrix of covariates(columns) for each sample(rows).
- `consensus`: The required fraction (0, 1] of solutions containing an edge in order to keep it.
- `numRuns`: Number of runs to do. In practice the number of runs needed varies based on data set size and the GA parameters set.
- `parallel`: When this is 'TRUE' it allows for parallel calculation of the bootstraps. Requires the package `doParallel`.
- `cores`: The number of parallel processes to run if parallel is 'TRUE'.
- `...`: Other arguments for the GA function see `Gen.Alg`

### Details

Use a GA consensus approach to find taxa that separate subjects based on group membership or set of covariates if you cannot run the GA long enough to get a final solution.
Value

A list containing

- solutions: The best solution from each run.
- consSol: The consensus solution.
- selectedIndex: The selected taxa by column number.

Author(s)

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon

Examples

```r
### Not run:
data(saliva)
data(throat)

### Combine the data into a single data frame
group.data <- list(saliva, throat)
group.data <- formatDataSets(group.data)
data <- do.call("rbind", group.data)

### Normalize the data by subject
dataNorm <- t(apply(data, 1, function(x){x/sum(x)}))

### Set covars to just be group membership
memb <- c(rep(0, nrow(saliva)), rep(1, nrow(throat)))
covars <- matrix(memb, length(memb), 1)

### We use low numbers for speed. The exact numbers to use depend
### on the data being used, but generally the higher iters and popSize
### the longer it will take to run. earlyStop is then used to stop the
### run early if the results aren't improving.
iters <- 500
popSize <- 200
earlyStop <- 250
numRuns <- 3

### End(Not run)
```

Description

Calculates Kullback Leibler divergence for all pairs of the datasets.
Kullback.Leibler

Usage

Kullback.Leibler(group.data, plot = TRUE, main="Kullback Leibler Divergences", parallel = FALSE, cores = 3)

Arguments

group.data A list where each element is a matrix of taxonomic counts(columns) for each sample(rows).
plot When 'TRUE' a heatmap of the results will also be generated.
main A string to be used as the plots title.
parallel When this is 'TRUE' it allows for parallel calculation of the KL distances. Requires the package doParallel.
cores The number of parallel processes to run if parallel is 'TRUE'.

Value

A matrix of Kullback Leibler divergence values and a heatmap if plot is TRUE.

Author(s)

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon

References


Examples

data(saliva)
data(throat)
data(tonsils)

### Combine the data sets into a single list
group.data <- list(saliva, throat, tonsils)

## Not run:
kl <- Kullback.Leibler(group.data)
kl

## End(Not run)
MC.Xdc.statistics  Size and Power for the Several-Sample DM Parameter Test Comparison

Description

This Monte-Carlo simulation procedure provides the power and size of the several sample Dirichlet-Multinomial parameter test comparison, using the likelihood-ratio-test statistics.

Usage

MC.Xdc.statistics(group.Nrs, numMC = 10, alphap, type = "ha", siglev = 0.05, est = "mom", MC = NULL, Nrs = NULL)

Arguments

- **group.Nrs**: A list specifying the number of reads/sequence depth for each sample in a group with one group per list entry.
- **numMC**: Number of Monte-Carlo experiments. In practice this should be at least 1,000.
- **alphap**: If "hnull": A matrix where rows are vectors of alpha parameters for the reference group.
  If "ha": A matrix consisting of vectors of alpha parameters for each taxa in each group.
- **type**: If "hnull": Computes the size of the test.
  If "ha": Computes the power of the test. (default)
- **siglev**: Significance level for size of the test / power calculation. The default is 0.05.
- **est**: The type of parameter estimator to be used with the Likelihood-ratio-test statistics, 'mle’ or ‘mom’. Default value is 'mom’. (See Note 2 in details)
- **MC**: Deprecated. Replaced with numMC for clarity.
- **Nrs**: Deprecated. Replaced with group.Nrs for clarity.

Details

1. Note 1: Though the test statistic supports an unequal number of reads across samples, the performance has not yet been fully tested.
2. Note 2: 'mle’ will take significantly longer time and may not be optimal for small sample sizes; ‘mom’ will provide a more conservative result in such a case.
3. Note 3: All components of alphap should be non-zero or it may result in errors and/or invalid results.

Value

Size of the test statistics (under "hnull") or power (under "ha") of the test.
MC.Xmc.statistics

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

Examples

data(saliva)
data(throat)
data(tonsils)

### Get a list of dirichlet-multinomial parameters for the data
fit.saliva <- DM.MM(saliva)
fit.throat <- DM.MM(throat)
fit.tonsils <- DM.MM(tonsils)

### Set up the number of Monte-Carlo experiments
### We use 1 for speed, should be at least 1,000
numMC <- 1

### Generate the number of reads per sample
### The first number is the number of reads and the second is the number of subjects
nrsGrp1 <- rep(12000, 9)
nrsGrp2 <- rep(12000, 11)
nrsGrp3 <- rep(12000, 12)
group.Nrs <- list(nrsGrp1, nrsGrp2, nrsGrp3)

### Computing size of the test statistics (Type I error)
alphap <- fit.saliva$gamma
pval1 <- MC.Xdc.statistics(group.Nrs, numMC, alphap, "hnull")
pval1

### Computing Power of the test statistics (Type II error)
alphap <- rbind(fit.saliva$gamma, fit.throat$gamma, fit.tonsils$gamma)
pval2 <- MC.Xdc.statistics(group.Nrs, numMC, alphap)
pval2

---

MC.Xmc.statistics  Size and Power of Several Sample RAD-Probability Mean Test Comparison

Description

This Monte-Carlo simulation procedure provides the power and size of the several sample RAD-probability mean test comparison with known reference vector of proportions, using the Generalized Wald-type statistics.

Usage

MC.Xmc.statistics(group.Nrs, numMC = 10, pi0, group.pi, group.theta, type = "ha", siglev = 0.05, MC = NULL, Nrs = NULL)
Arguments

group.Nrs A list specifying the number of reads/sequence depth for each sample in a group with one group per list entry.

numMC Number of Monte-Carlo experiments. In practice this should be at least 1,000.

pi0 The RAD-probability mean vector.

group.pi If "hnull": This argument is ignored.
If "ha": A matrix where each row is a vector pi values for each group.

group.theta A vector of overdispersion values for each group.

type If "hnull": Computes the size of the test. 
If "ha": Computes the power of the test. (default)

siglev Significance level for size of the test / power calculation. The default is 0.05.

MC Deprecated. Replaced with numMC for clarity.

Nrs Deprecated. Replaced with group.Nrs for clarity.

Details

Note: Though the test statistic supports an unequal number of reads across samples, the performance has not yet been fully tested.

Value

Size of the test statistics (under "hnull") or power (under "ha") of the test.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

Examples

data(saliva)
data(throat)
data(tonsils)

### Get a list of dirichlet-multinomial parameters for the data
fit.saliva <- DM.MoM(saliva)
fit.throat <- DM.MoM(throat)
fit.tonsils <- DM.MoM(tonsils)

### Set up the number of Monte-Carlo experiments
### We use 1 for speed, should be at least 1,000
numMC <- 1

### Generate the number of reads per sample
### The first number is the number of reads and the second is the number of subjects
nrsGrp1 <- rep(12000, 9)
nrsGrp2 <- rep(12000, 11)
group.Nrs <- list(nrsGrp1, nrsGrp2)
MC.Xmcupo.statistics

Description

This Monte-Carlo simulation procedure provides the power and size of the several sample RAD-probability mean test comparisons without reference vector of proportions, using the Generalized Wald-type statistics.

Usage

MC.Xmcupo.statistics(group.Nrs, numMC, pi0, group.theta, type = "hnull", siglev = 0.05, MC = NULL, Nrs = NULL)

Arguments

group.Nrs: A list specifying the number of reads/sequence depth for each sample in a group with one group per list entry.

numMC: Number of Monte-Carlo experiments. In practice this should be at least 1,000.

pi0: The RAD-probability mean vector.

group.pi: If "hnull": This argument is ignored. If "ha": A matrix where each row is a vector pi values for each group.

group.theta: A vector of overdispersion values for each group.

type: If "hnull": Computes the size of the test. If "ha": Computes the power of the test. (default)

siglev: Significance level for size of the test / power calculation. The default is 0.05.

MC: Deprecated. Replaced with numMC for clarity.

Nrs: Deprecated. Replaced with group.Nrs for clarity.

Details

Note: Though the test statistic supports an unequal number of reads across samples, the performance has not yet been fully tested.
Value

Size of the test statistics (under "hnull") or power (under "ha") of the test.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

Examples

data(saliva)
data(throat)
data(tonsils)

### Get a list of dirichlet-multinomial parameters for the data
fit.saliva <- DM.MoM(saliva)
fit.throat <- DM.MoM(throat)
fit.tonsils <- DM.MoM(tonsils)

### Set up the number of Monte-Carlo experiments
### We use 1 for speed, should be at least 1,000
numMC <- 1

### Generate the number of reads per sample
### The first number is the number of reads and the second is the number of subjects
Nrs1 <- rep(12000, 10)
Nrs2 <- rep(12000, 19)
group.Nrs <- list(Nrs1, Nrs2)

pi0 <- fit.saliva$p

### Computing size of the test statistics (Type I error)
group.theta <- c(fit.throat$theta, fit.tonsils$theta)
pval1 <- MC.Xmcupo.statistics(group.Nrs, numMC, pi0, group.theta=group.theta, type="hnull")
pval1

### Computing Power of the test statistics (Type II error)
group.pi <- rbind(fit.throat$pi, fit.tonsils$pi)
pval2 <- MC.Xmcupo.statistics(group.Nrs, numMC, group.pi=group.pi, group.theta=group.theta)
pval2

---

**MC.Xoc.statistics**

**Size and Power of Several Sample-Overdispersion Test Comparisons**

Description

This Monte-Carlo simulation procedure provides the power and size of the several sample-overdispersion test comparison, using the likelihood-ratio-test statistics.
Usage

MC.Xoc.statistics(group.Nrs, numMC = 10, group.alphap, type = "ha", siglev = 0.05, MC = NULL, Nrs = NULL)

Arguments

group.Nrs  A list specifying the number of reads/sequence depth for each sample in a group with one group per list entry.
numMC  Number of Monte-Carlo experiments. In practice this should be at least 1,000.
group.alphap  If "hnul": A vector of alpha parameters for each taxa.
               If "ha": A list consisting of vectors of alpha parameters for each taxa.
type  If "hnul": Computes the size of the test.
       If "ha": Computes the power of the test. (default)
siglev  Significance level for size of the test / power calculation. The default is 0.05.
MC  Deprecated. Replaced with numMC for clarity.
Nrs  Deprecated. Replaced with group.Nrs for clarity.

Details

1. Note 1: Though the test statistic supports an unequal number of reads across samples, the performance has not yet been fully tested.
2. Note 2: All components of group.alphap should be non-zero or it may result in errors and/or invalid results.

Value

Size of the test statistics (under "hnul") or power (under "ha") of the test.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

Examples

data(saliva)
data(throat)
data(tonsils)

### Get a list of dirichlet-multinomial parameters for the data
fit.saliva <- DM.MoM(saliva)
fit.throat <- DM.MoM(throat)
fit.tonsils <- DM.MoM(tonsils)

### Set up the number of Monte-Carlo experiments
### We use 1 for speed, should be at least 1,000
numMC <- 1

### Generate the number of reads per sample
### The first number is the number of reads and the second is the number of subjects
nrsGrp1 <- rep(12000, 9)
nrsGrp2 <- rep(12000, 11)
nrsGrp3 <- rep(12000, 12)
group.Nrs <- list(nrsGrp1, nrsGrp2, nrsGrp3)

### Computing size of the test statistics (Type I error)
alphap <- fit.tonsils$gamma
pval1 <- MC.Xsc.statistics(group.Nrs, numMC, alphap, "hnull")
pval1

## Not run:
### Computing Power of the test statistics (Type II error)
alphap <- rbind(fit.saliva$gamma, fit.throat$gamma, fit.tonsils$gamma)
pval2 <- MC.Xsc.statistics(group.Nrs, numMC, alphap, "ha")
pval2

## End(Not run)

---

**MC.Xsc.statistics**

### Size and Power for the One Sample RAD Probability-Mean Test Comparison

**Description**

This Monte-Carlo simulation procedure provides the power and size of the one sample RAD probability-mean test, using the Generalized Wald-type statistic.

**Usage**

MC.Xsc.statistics(Nrs, numMC = 10, fit, pi0 = NULL, type = "ha", siglev = 0.05, MC = NULL)

**Arguments**

- **Nrs**
  
  A vector specifying the number of reads/sequence depth for each sample.

- **numMC**
  
  Number of Monte-Carlo experiments. In practice this should be at least 1,000.

- **fit**
  
  A list (in the format of the output of dirmult function) containing the data parameters for evaluating either the size or power of the test.

- **pi0**
  
  The RAD-probability mean vector. If the type is set to "hnull" then pi0 is set by the sample in fit.

- **type**
  
  If "hnull": Computes the size of the test.
  If "ha": Computes the power of the test. (default)

- **siglev**
  
  Significance level for size of the test / power calculation. The default is 0.05.

- **MC**
  
  Deprecated. Replaced with numMC for clarity.
Details

Note: Though the test statistic supports an unequal number of reads across samples, the performance has not yet been fully tested.

Value

Size of the test statistics (under "hnull") or power (under "ha") of the test.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

Examples

data(saliva)
data(throat)
data(tonsils)

### Get a list of dirichlet-multinomial parameters for the data
fit.saliva <- DM.MoM(saliva)
fitted.throat <- DM.MoM(throat)
fitted.tonsils <- DM.MoM(tonsils)

### Set up the number of Monte-Carlo experiments
### We use 1 for speed, should be at least 1,000
numMC <- 1

### Generate the number of reads per sample
### The first number is the number of reads and the second is the number of subjects
nrs <- rep(15000, 25)

### Computing size of the test statistics (Type I error)
pval1 <- MC.Xsc.statistics(nrs, numMC, fitted.tonsils, fitted.saliva$pi, "hnull")
pval1

### Computing Power of the test statistics (Type II error)
pval2 <- MC.Xsc.statistics(nrs, numMC, fitted.throat, fitted.tonsils$pi)
pval2

---

**MC.ZT.statistics**

**Size and Power of Goodness of Fit Test: Multinomial vs. Dirichlet-Multinomial**

Description

This Monte-Carlo simulation procedure provides the power and size of the Multinomial vs. Dirichlet-Multinomial goodness of fit test, using the C(α)-optimal test statistics of Kim and Margolin (1992) (t statistics) and the C(α)-optimal test statistics of (Paul et al., 1989).
Usage

MC.ZT.statistics(Nrs, numMC = 10, fit, type = "ha", siglev = 0.05, MC = NULL)

Arguments

Nrs A vector specifying the number of reads/sequence depth for each sample.
numMC Number of Monte-Carlo experiments. In practice this should be at least 1,000.
fit A list (in the format of the output of dirmult function) containing the data parameters for evaluating either the size or power of the test.
type If "hnull": Computes the size of the test.
If "ha": Computes the power of the test. (default)
siglev Significance level for size of the test / power calculation. The default is 0.05.
MC Deprecated. Replaced with numMC for clarity.

Details

Note: Though the test statistic supports an unequal number of reads across samples, the performance has not yet been fully tested.

Value

A vector containing both the size of the test statistics (under "hnull") or power (under "ha") of the test for both the z and t statistics.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

Examples

data(saliva)

### Get a list of dirichlet-multinomial parameters for the data
fit.saliva <- DM.MoM(saliva)

### Set up the number of Monte-Carlo experiments
### We use 1 for speed, should be at least 1,000
numMC <- 1

### Generate the number of reads per sample
### The first number is the number of reads and the second is the number of subjects
nrs <- rep(15000, 25)

### Computing size of the test statistics (Type I error)
pval1 <- MC.ZT.statistics(nrs, numMC, fit.saliva, "hnull")
pval1

### Computing Power of the test statistics (Type II error)
pval2 <- MC.ZT.statistics(nrs, numMC, fit.saliva)
pval2
Description

It generates a data matrix with random samples from a multinomial distribution where the rows are the samples and the columns are the taxa.

Usage

Multinomial(nrs, probs)

Arguments

- nrs: A vector specifying the number of reads or sequence depth for each sample.
- probs: A vector specifying taxa probabilities.

Details

Note: Though the test statistic supports an unequal number of reads across samples, the performance has not yet been fully tested.

Value

A data matrix of taxa counts where the rows are the samples and the columns are the taxa.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

Examples

```r
## Generate the number of reads per sample
## The first number is the number of reads and the second is the number of subjects
nrs <- rep(15000, 25)

## Create a probability vector
probs <- c(0.4, 0.3, 0.2, 0.05, 0.04, 0.01)

mData <- Multinomial(nrs, probs)
mData[1:5, 1:5]
```
Plot.MDS  

*Multidimensional Scaling Plot of Microbiome Data*

**Description**

Plots any number of data sets on an MDS plot.

**Usage**

Plot.MDS(group.data, main = "Group MDS", retCords = FALSE)

**Arguments**

- `group.data`: A list of matrices of taxonomic counts(columns) for each sample(rows).
- `main`: A string to be used as the plots title.
- `retCords`: A boolean to return the mds coordinates or not.

**Value**

A MDS plot and possibly the x-y coordinates for every point.

**Author(s)**

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon

**Examples**

```r
data(saliva)
data(throat)
data(tonsils)

### Combine the data sets into a single list
group.data <- list(saliva, throat, tonsils)

Plot.MDS(group.data)
```

Plot.PI  

*Plot the Pi Vector*

**Description**

Plots the taxa proportions for every group.

**Usage**

Plot.PI(estPi, errorBars = TRUE, logScale = FALSE, main = "PI Vector", ylab = "Fractional Abundance")
Argsuemts

- `estPi`: The results for either MLE or MOM from the function 'Est.Pi'.
- `errorBars`: A boolean to display the error bars or not.
- `logScale`: A boolean to log the y scale or not.
- `main`: A string to be used as the plots title.
- `ylab`: A string to be used as the plots y-axis title.

Value

A plot of the pi vectors for every group.

Author(s)

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon

Examples

```r
## Not run:
data(saliva)
data(throat)
data(tonsils)

### Combine the data sets into a single list
group.data <- list(saliva, throat, tonsils)

### Get PI using MLE with CI
mle <- Est.PI(group.data)$MLE

### Plot with Error Bars
Plot.PI(mle)

### Plot without Error Bars
Plot.PI(mle, FALSE)

### Plot with Error Bars and scaling
Plot.PI(mle, TRUE, TRUE)

## End(Not run)
```

Description

Plots the taxa proportions for every group/time as a barchart.
Usage

Plot.RM.Barchart(group.data, groups, times, plotByGrp = TRUE, 
col = NULL, conf = .95)

Arguments

- **group.data**: A list of matrices of taxonomic counts(columns) for each sample(rows).
- **groups**: A vector indicating group membership.
- **times**: A vector indicating time.
- **plotByGrp**: When 'TRUE', the plot will be split by group rather than time.
- **col**: A vector of colors to use to denote taxa.
- **conf**: The desired confidence limits. The default is 95%

Value

A barchart of the pi vectors for every group/time.

Author(s)

Elena Deych, Berkley Shands, William D. Shannon, Dake Yang

Examples

```r
## Not run:
data(saliva)
data(throat)

### Reduce the size of the data
saliva <- Data.filter(saliva, numTaxa=9)
throat <- Data.filter(throat, numTaxa=9)

### Get the gamma value for the data
saliva.gamma <- DM.MoM(saliva)$gamma
throat.gamma <- DM.MoM(throat)$gamma
mid.gamma <- (saliva.gamma + throat.gamma)/2

### Generate a the number of reads per sample
### The first number is the number of reads and the second is the number of subjects
nrs <- rep(10000, 20)

### Create data sets to be our time series in a list
group.data <- list(
  Dirichlet.multinomial(nrs, saliva.gamma),
  Dirichlet.multinomial(nrs, saliva.gamma),
  Dirichlet.multinomial(nrs, saliva.gamma),
  Dirichlet.multinomial(nrs, saliva.gamma),
  Dirichlet.multinomial(nrs, mid.gamma),
  Dirichlet.multinomial(nrs, throat.gamma)
)
names(group.data) <- c(
```
### Description

Plots the taxa proportions for every group/time as a dot plot.

### Usage

```r
Plot.RM.Dotplot(group.data, groups, times, errorBars = TRUE,
                 col = NULL, conf = .95, alpha = 1)
```

### Arguments

- `group.data`: A list of matrices of taxonomic counts(columns) for each sample(rows).
- `groups`: A vector indicating group membership.
- `times`: A vector indicating time.
- `errorBars`: When 'TRUE', error bars will also be displayed.
- `col`: A vector of colors to use to denote taxa.
- `conf`: The desired confidence limits. The default is 95%
- `alpha`: The desired alpha level for the colors.

### Value

A plot of the pi vectors for every group/time.

### Author(s)

Elena Deych, Berkley Shands, William D. Shannon, Dake Yang
Examples

```r
## Not run:
data(saliva)
data(throat)

### Reduce the size of the data
saliva <- Data.filter(saliva, numTaxa=9)
throat <- Data.filter(throat, numTaxa=9)

### Get the gamma value for the data
saliva.gamma <- DM.MoM(saliva)$gamma
throat.gamma <- DM.MoM(throat)$gamma
mid.gamma <- (saliva.gamma + throat.gamma)/2

### Generate a the number of reads per sample
### The first number is the number of reads and the second is the number of subjects
nrs <- rep(10000, 20)

### Create data sets to be our time series in a list
group.data <- list(
  Dirichlet.multinomial(nrs, saliva.gamma),
  Dirichlet.multinomial(nrs, saliva.gamma),
  Dirichlet.multinomial(nrs, saliva.gamma),
  Dirichlet.multinomial(nrs, saliva.gamma),
  Dirichlet.multinomial(nrs, mid.gamma),
  Dirichlet.multinomial(nrs, throat.gamma)
)
names(group.data) <- c(
  "Group 1, Time 1", "Group 2, Time 1",
  "Group 1, Time 2", "Group 2, Time 2",
  "Group 1, Time 3", "Group 2, Time 3"
)

### Set the group and time information for each list element
groups <- c(1, 2, 1, 2, 1, 2)
times <- c(1, 2, 3, 1, 2, 3)

### Plot the data with error bars
Plot.RM.Dotplot(group.data, groups, times)

### Plot the data without error bars
Plot.RM.Dotplot(group.data, groups, times, FALSE)

## End(Not run)
```
Description

The saliva data set formed by the Ranked-abundance distribution vectors of 24 subjects. The RAD vectors contains 21 elements formed by the 20 most abundant taxa at the genus level and additional taxa containing the sum of the remaining less abundant taxa per sample. Note that the incorporation of the additional taxon (taxon 21) in the analysis allows for estimating the RAD proportional-mean of taxa with respect to all the taxa within the sample.

Usage

data(saliva)

Format

The format is a matrix of 24 rows by 21 columns, with the each row being a separate subject and each column being a different taxa.

Examples

data(saliva)

Test.Paired

Test Paired Data Sets

Description

Tests two paired data sets for similarity.

Usage

Test.Paired(group.data, numPerms = 1000, parallel = FALSE, cores = 3)

Arguments

group.data A list of 2 matrices of taxonomic counts(columns) for each sample(rows).
numPerms Number of permutations. In practice this should be at least 1,000.
parallel When this is ’TRUE’ it allows for parallel calculation of the permutations. Requires the package doParallel.
cores The number of parallel processes to run if parallel is ’TRUE’.

Value

A pvalue.

Author(s)

Sharina Carter, Elena Deych, Berkley Shands, William D. Shannon
Examples

data(saliva)
data(throat)

### Since saliva and throat come from same subjects, the data is paired
salival <- saliva[-24,] # Make saliva 23 subjects to match throat
group.data <- list(throat, salival)

### We use 1 for speed, should be at least 1,000
umPerms <- 1

pval <- Test.Paired(group.data, numPerms)
pval

---

**throat**

**Throat Data Set**

Description

The throat data set formed by the Ranked-abundance distribution vectors of 24 subjects. The RAD vectors contains 21 elements formed by the 20 most abundant taxa at the genus level and additional taxa containing the sum of the remaining less abundant taxa per sample. Note that the incorporation of the additional taxon (taxon 21) in the analysis allows for estimating the RAD proportional-mean of taxa with respect to all the taxa within the sample.

Usage

data(throat)

Format

The format is a matrix of 24 rows by 21 columns, with the each row being a separate subject and each column being a different taxa.

Examples

data(throat)
**tongue**

*Tongue Data Set*

**Description**

The tongue data set formed by the Ranked-abundance distribution vectors of 24 subjects. The RAD vectors contains 21 elements formed by the 20 most abundant taxa at the genus level and additional taxa containing the sum of the remaining less abundant taxa per sample. Note that the incorporation of the additional taxon (taxon 21) in the analysis allows for estimating the RAD proportional-mean of taxa with respect to all the taxa within the sample.

**Usage**

data(tongue)

**Format**

The format is a matrix of 24 rows by 21 columns, with the each row being a separate subject and each column being a different taxa.

**Examples**

data(tongue)

---

**tonsils**

*Palatine Tonsil Data Set*

**Description**

The palatine tonsil data set formed by the Ranked-abundance distribution vectors of 24 subjects. The RAD vectors contains 21 elements formed by the 20 most abundant taxa at the genus level and additional taxa containing the sum of the remaining less abundant taxa per sample. Note that the incorporation of the additional taxon (taxon 21) in the analysis allows for estimating the RAD proportional-mean of taxa with respect to all the taxa within the sample.

**Usage**

data(tonsils)

**Format**

The format is a matrix of 24 rows by 21 columns, with the each row being a separate subject and each column being a different taxa.

**Examples**

data(tonsils)
Description

This routine provides the value of the Likelihood-Ratio-Test Statistics and the corresponding p-value for evaluating the several sample Dirichlet-Multinomial parameter test comparison.

Usage

$\text{Xdc.sevsample(group.data, epsilon }= 10^\epsilon(-4), \text{ est } = \text{"mom"})$

Arguments

- \text{group.data} A list where each element is a matrix of taxonomic counts(columns) for each sample(rows). (See Notes 1 and 2 in details)
- \text{epsilon} Convergence tolerance. To terminate, the difference between two succeeding log-likelihoods must be smaller than epsilon. Default value is $10^\epsilon(-4)$.
- \text{est} The type of parameter estimator to be used with the Likelihood-ratio-test statistics, 'mle' or 'mom'. Default value is 'mom'. (See Note 3 in details)

Details

To assess whether the Dirichlet parameter vector, $\alpha_m = \pi_m \frac{1-\theta_m}{\theta_m}$ (a function of the RAD probability-mean vector and overdispersion), observed in $J$ groups of microbiome samples are equal to each other, the following hypothesis $H_0: \alpha_1 = \cdots = \alpha_m = \cdots = \alpha_J = \alpha_o$ versus $H_a: \alpha_m \neq \alpha_o, m = 1, \ldots, J$ can be tested. The null hypothesis implies that the HMP samples across groups have the same mean and overdispersion, indicating that the RAD models are identical. In particular, the likelihood-ratio test statistic is used, which is given by,

$$x_{dc} = -2 \log \left\{ \frac{L(\alpha_o; X_1, \ldots, X_J)}{L(\alpha_1, \ldots, \alpha_J; X_1, \ldots, X_J)} \right\}.$$

The asymptotic null distribution of $x_{dc}$ follows a Chi-square with degrees of freedom equal to $(J-1)K$, where $K$ is the number of taxa (Wilks, 1938).

1. Note 1: The matrices in \text{group.data} must contain the same taxa, in the same order.
2. Note 2: Each taxa should be present in at least 1 sample, a column with all 0’s may result in errors and/or invalid results.
3. Note 3: 'mle' will take significantly longer time and may not be optimal for small sample sizes; 'mom' will provide more conservative results in such a case.

Value

A list containing the Xdc statistics and p-value.
Author(s)
Patrick S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

References

Examples

```r
data(saliva)
data(throat)

# Combine the data sets into a single list
group.data <- list(saliva, throat)

xdc <- xdc.sevsample(group.data)
xdc
```

---

Xmc.sevsample

Generalized Wald-type Statistics: Several Sample RAD Probability-Mean Test Comparison with a Known Common Vector

Description

This function computes the Generalized Wald-type test statistic (Wilson and Koehler, 1984) and corresponding p-value to assess whether the sample RAD probability-means from multiple populations are the same or different. The statistics assumes that a common RAD probability-mean vector for comparison under the null hypothesis is known.

Usage

```r
Xmc.sevsample(group.data, pi0)
```

Arguments

- `group.data`: A list where each element is a matrix of taxonomic counts(columns) for each sample(rows).
- `pi0`: The RAD-probability mean vector.

Details

Note: The matrices in `group.data` must contain the same taxa, in the same order.

Value

A list containing the Generalized Wald-type statistics and p-value.
Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

References


Examples

data(saliva)
data(throat)
data(tonsils)

### Get pi from the dirichlet-multinomial parameters
pi0 <- dirmult(saliva)$pi

### Combine the data sets into a single list
group.data <- list(throat, tonsils)

xmc <- Xmc.sevsample(group.data, pi0)
xmc

---

Xmcupo.efffectsize  Effect Size for Xmcupo Statistic

Description

This function computes the Cramer’s Phi and Modified Cramer’s Phi Criterion for the test statistic
Xmcupo.sevsample.

Usage

Xmcupo.efffectsize(group.data)

Arguments

group.data  A list where each element is a matrix of taxonomic counts(columns) for each
sample(rows).

Details

Note: The matrices in group.data must contain the same taxa, in the same order.

Value

A vector containing the Chi-Squared statistic value, the Cramer’s Phi Criterion, and the modified
Cramer’s Phi Criterion.
**Author(s)**

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

**Examples**

```r
data(saliva)
data(throat)

### Combine the data sets into a single list
group.data <- list(saliva, throat)
effect <- xmcupo.effectsize(group.data)
effect
```

---

**Description**

This function computes the Generalized Wald-type test statistic (Wilson and Koehler, 1984) and corresponding p-value to assess whether the sample RAD probability-means from multiple populations are same or different. The statistics assumes that a common RAD probability-mean vector for comparison under the null hypothesis is unknown.

**Usage**

```r
xmcupo.sevsample(group.data)
```

**Arguments**

- `group.data` A list where each element is a matrix of taxonomic counts(columns) for each sample(rows).

**Details**

Note: The matrices in `group.data` must contain the same taxa, in the same order.

**Value**

A list containing the Generalized Wald-type statistics and p-value.

**Author(s)**

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon
References


Examples

data(saliva)
data(tonsils)
data(throat)

### Combine the data sets into a single list
group.data <- list(saliva, throat, tonsils)

xmcupo <- Xmcupo.sevsample(group.data)
xmcupo

---

**Xoc.sevsample**  
*Likelihood-Ratio-Test Statistics: Several Sample Overdispersion Test Comparison*

**Description**

This routine provides the value of the likelihood-ratio-test statistic and the corresponding p-value to
assess whether the overdispersion observed in multiple groups of microbiome samples are equal.

**Usage**

Xoc.sevsample(group.data, epsilon = 10^(-4))

**Arguments**

- `group.data`: A list where each element is a matrix of taxonomic counts(columns) for each
  sample(rows). (See Notes 1 and 2 in details)
- `epsilon`: Convergence tolerance. To terminate, the difference between two succeeding
  log-likelihoods must be smaller than epsilon. Default value is 10^(-4).

**Details**

To assess whether the over dispersion parameter vectors \( \theta_m \) observed in \( J \) groups of microbiome
samples are equal to each other, the following hypothesis \( H_0 : \theta_1 = \cdots = \theta_m = \cdots = \theta_J = \theta_o \)
versus \( H_a : \theta_m \neq \theta_o, m = 1, \ldots, J \) can be tested. In particular, the likelihood-ratio test statistic is
used (Tvedebrink, 2010), which is given by,

\[
x_{oc} = -2 \log \left\{ \frac{L(\theta_o; X_1, \ldots, X_J)}{L(\theta_1, \ldots, \theta_J; X_1, \ldots, X_J)} \right\}.
\]

The asymptotic null distribution of \( x_{oc} \) follows a Chi-square with degrees of freedom equal to \( J-1 \)
(Wilks, 1938).
1. Note 1: The matrices in group.data must contain the same taxa, in the same order.
2. Note 2: Each taxa should be present in at least 1 sample, a column with all 0’s may result in errors and/or invalid results.

Value

A list containing the Xoc statistics and p-value.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

References


Examples

data(saliva)
data(tonsils)

### Combine the data sets into a single list
group.data <- list(saliva, tonsils)

### Not run:
xoc <- Xoc.sevsample(group.data)
xoc

### End(Not run)

---

### Xsc.onesample

**Generalized Wald-Type Statistics: One Sample RAD Probability-Mean Test Comparison**

**Description**

This routine provides the value of the Generalized Wald-type statistic to assess whether the RAD probability-mean observed in one group of samples is equal to a known RAD probability-mean.

**Usage**

Xsc.onesample(data, pi0)
Arguments

data A matrix of taxonomic counts(columns) for each sample(rows).
piθ The RAD-probability mean vector.

Value

A list containing Generalized Wald-type statistics and p-value.

Author(s)

Patricio S. La Rosa, Elena Deych, Berkley Shands, William D. Shannon

Examples

data(saliva)
data(throat)

### Get pi from the dirichlet-multinomial parameters
piθ <- dirmult(saliva)$pi

gxsc <- Xsc.onesample(throat, piθ)
gxsc
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