

Package ‘mia’

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Description mia implements tools for microbiome analysis based on the SummarizedExperiment, SingleCellExperiment and TreeSummarizedExperiment infrastructure. Data wrangling and analysis in the context of taxonomic data is the main scope. Additional functions for common task are implemented such as community indices calculation and summarization.

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Description

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See Also

[TreeSummarizedExperiment](#)

addAlpha

Estimate alpha diversity indices

Description

The function estimates alpha diversity indices optionally using rarefaction, then stores results in [colData](#).

Usage

```
addAlpha(
  x,
  assay.type = "counts",
  index = c("coverage_diversity", "fisher_diversity", "faith_diversity",
    "gini_simpson_diversity", "inverse_simpson_diversity",
    "log_modulo_skewness_diversity", "shannon_diversity", "absolute_dominance",
    "dbp_dominance", "core_abundance_dominance", "gini_dominance", "dmn_dominance",
    "relative_dominance", "simpson_lambda_dominance", "camargo_evenness",
    "pielou_evenness", "simpson_evenness", "evar_evenness", "bulla_evenness",
    "ace_richness", "chao1_richness", "hill_richness", "observed_richness"),
  name = index,
  niter = NULL,
  ...
)

## S4 method for signature 'SummarizedExperiment'
addAlpha(
  x,
  assay.type = "counts",
  index = c("coverage_diversity", "fisher_diversity", "faith_diversity",
    "gini_simpson_diversity", "inverse_simpson_diversity",
    "log_modulo_skewness_diversity", "shannon_diversity", "absolute_dominance",
    "dbp_dominance", "core_abundance_dominance", "gini_dominance", "dmn_dominance",
    "relative_dominance", "simpson_lambda_dominance", "camargo_evenness",
    "pielou_evenness", "simpson_evenness", "evar_evenness", "bulla_evenness",
    "ace_richness", "chao1_richness", "hill_richness", "observed_richness"),
  name = index,
  niter = NULL,
  ...
)
```

Arguments

x	a SummarizedExperiment object.
assay.type	Character scalar. Specifies the name of assay used in calculation. (Default: "counts")
index	Character vector. Specifies the alpha diversity indices to be calculated.
name	Character vector. A name for the column of the colData where results will be stored. (Default: index)
niter	Integer scalar. Specifies the number of rarefaction rounds. Rarefaction is not applied when niter=NULL (see Details section). (Default: NULL)
...	optional arguments: <ul style="list-style-type: none"> • sample: Integer scalar. Specifies the rarefaction depth i.e. the number of counts drawn from each sample. (Default: $\min(\text{colSums2}(\text{assay}(x, \text{assay.type})))$) • tree.name: Character scalar. Specifies which rowTree will be used. (Faith's index). (Default: "phylo") • node.label: Character vector or NULL Specifies the links between rows and node labels of phylogeny tree specified by tree.name. If a certain row is not linked with the tree, missing instance should be noted as NA. When NULL, all the rownames should be found from the tree. (Faith's index). (Default: NULL) • only.tips: (Faith's index). Logical scalar. Specifies whether to remove internal nodes when Faith's index is calculated. When only.tips=TRUE, those rows that are not tips of tree are removed. (Default: FALSE) • threshold: (Coverage and all evenness indices). Numeric scalar. From 0 to 1, determines the threshold for coverage and evenness indices. When evenness indices are calculated values under or equal to this threshold are denoted as zeroes. For coverage index, see details. (Default: 0.5 for coverage, 0 for evenness indices) • quantile: (log modulo skewness index). Numeric scalar. Arithmetic abundance classes are evenly cut up to to this quantile of the data. The assumption is that abundances higher than this are not common, and they are classified in their own group. (Default: 0.5) • nclasses: (log modulo skewness index). Integer scalar. The number of arithmetic abundance classes from zero to the quantile cutoff indicated by quantile. (Default: 50) • ntaxa: (absolute and relative indices). Integer scalar. The n-th position of the dominant taxa to consider. (Default: 1) • aggregate: (absolute, dbp, dmn, and relative indices). Logical scalar. Aggregate the values for top members selected by ntaxa or not. If TRUE, then the sum of relative abundances is returned. Otherwise the relative abundance is returned for the single taxa with the indicated rank (default: aggregate = TRUE). • detection: (observed index). Numeric scalar Selects detection threshold for the abundances (Default: 0)

Details

Diversity:

Alpha diversity is a joint quantity that combines elements or community richness and evenness. Diversity increases, in general, when species richness or evenness increase.

The following diversity indices are available:

- 'coverage': Number of species needed to cover a given fraction of the ecosystem (50 percent by default). Tune this with the threshold argument.
- 'faith': Faith's phylogenetic alpha diversity index measures how long the taxonomic distance is between taxa that are present in the sample. Larger values represent higher diversity. Using this index requires rowTree. (Faith 1992)
If the data includes features that are not in tree's tips but in internal nodes, there are two options. First, you can keep those features, and prune the tree to match features so that each tip can be found from the features. Other option is to remove all features that are not tips. (See only. tips parameter)
- 'fisher': Fisher's alpha; as implemented in `vegan::fisher.alpha`. (Fisher et al. 1943)
- 'gini_simpson': Gini-Simpson diversity i.e. $1 - \lambda$, where λ is the Simpson index, calculated as the sum of squared relative abundances. This corresponds to the diversity index 'simpson' in `vegan::diversity`. This is also called Gibbs–Martin, or Blau index in sociology, psychology and management studies. The Gini-Simpson index (1-lambda) should not be confused with Simpson's dominance (λ), Gini index, or inverse Simpson index ($1/\lambda$).
- 'inverse_simpson': Inverse Simpson diversity: $1/\lambda$ where $\lambda = \sum(p^2)$ and p refers to relative abundances. This corresponds to the diversity index 'invsimpson' in `vegan::diversity`. Don't confuse this with the closely related Gini-Simpson index
- 'log_modulo_skewness': The rarity index characterizes the concentration of species at low abundance. Here, we use the skewness of the frequency distribution of arithmetic abundance classes (see Magurran & McGill 2011). These are typically right-skewed; to avoid taking log of occasional negative skews, we follow Locey & Lennon (2016) and use the log-modulo transformation that adds a value of one to each measure of skewness to allow logarithmization.
- 'shannon': Shannon diversity (entropy).

Dominance:

A dominance index quantifies the dominance of one or few species in a community. Greater values indicate higher dominance.

Dominance indices are in general negatively correlated with alpha diversity indices (species richness, evenness, diversity, rarity). More dominant communities are less diverse.

The following community dominance indices are available:

- 'absolute': Absolute index equals to the absolute abundance of the most dominant n species of the sample (specify the number with the argument `ntaxa`). Index gives positive integer values.
- 'dbp': Berger-Parker index (See Berger & Parker 1970) calculation is a special case of the 'relative' index. `dbp` is the relative abundance of the most abundant species of the sample. Index gives values in interval 0 to 1, where bigger value represent greater dominance.

$$dbp = \frac{N_1}{N_{tot}}$$

where N_1 is the absolute abundance of the most dominant species and N_{tot} is the sum of absolute abundances of all species.

- 'core_abundance': Core abundance index is related to core species. Core species are species that are most abundant in all samples, i.e., in whole data set. Core species are defined as those species that have prevalence over 50\ species must be prevalent in 50\ calculate the core abundance index. Core abundance index is sum of relative abundances of core species in the sample. Index gives values in interval 0 to 1, where bigger value represent greater dominance.

$$core_abundance = \frac{N_{core}}{N_{tot}}$$

where N_{core} is the sum of absolute abundance of the core species and N_{tot} is the sum of absolute abundances of all species.

- 'gini': Gini index is probably best-known from socio-economic contexts (Gini 1921). In economics, it is used to measure, for example, how unevenly income is distributed among population. Here, Gini index is used similarly, but income is replaced with abundance. If there is small group of species that represent large portion of total abundance of microbes, the inequality is large and Gini index closer to 1. If all species has equally large abundances, the equality is perfect and Gini index equals 0. This index should not be confused with Gini-Simpson index, which quantifies diversity.
- 'dmn': McNaughton's index is the sum of relative abundances of the two most abundant species of the sample (McNaughton & Wolf, 1970). Index gives values in the unit interval:

$$dmn = (N_1 + N_2)/N_{tot}$$

where N_1 and N_2 are the absolute abundances of the two most dominant species and N_{tot} is the sum of absolute abundances of all species.

- 'relative': Relative index equals to the relative abundance of the most dominant n species of the sample (specify the number with the argument ntaxa). This index gives values in interval 0 to 1.

$$relative = N_1/N_{tot}$$

where N_1 is the absolute abundance of the most dominant species and N_{tot} is the sum of absolute abundances of all species.

- 'simpson_lambda': Simpson's (dominance) index or Simpson's lambda is the sum of squared relative abundances. This index gives values in the unit interval. This value equals the probability that two randomly chosen individuals belongs to the same species. The higher the probability, the greater the dominance (See e.g. Simpson 1949).

$$lambda = \sum(p^2)$$

where p refers to relative abundances.

There is also a more advanced Simpson dominance index (Simpson 1949). However, this is not provided and the simpler squared sum of relative abundances is used instead as the alternative index is not in the unit interval and it is highly correlated with the simpler variant implemented here.

Evenness:

Evenness is a standard index in community ecology, and it quantifies how evenly the abundances of different species are distributed. The following evenness indices are provided:

By default, this function returns all indices.

The available evenness indices include the following (all in lowercase):

- 'camargo': Camargo's evenness (Camargo 1992)
- 'simpson_evenness': Simpson's evenness is calculated as inverse Simpson diversity ($1/\lambda$) divided by observed species richness S : $(1/\lambda)/S$.
- 'pielou': Pielou's evenness (Pielou, 1966), also known as Shannon or Shannon-Weaver/Wiener/Weiner evenness; $H/\ln(S)$. The Shannon-Weaver is the preferred term; see Spellerberg and Fedor (2003).
- 'evar': Smith and Wilson's Evar index (Smith & Wilson 1996).
- 'bulla': Bulla's index (O) (Bulla 1994).

Desirable statistical evenness metrics avoid strong bias towards very large or very small abundances; are independent of richness; and range within the unit interval with increasing evenness (Smith & Wilson 1996). Evenness metrics that fulfill these criteria include at least camargo, simpson, smith-wilson, and bulla. Also see Magurran & McGill (2011) and Beisel et al. (2003) for further details.

Richness:

The richness is calculated per sample. This is a standard index in community ecology, and it provides an estimate of the number of unique species in the community. This is often not directly observed for the whole community but only for a limited sample from the community. This has led to alternative richness indices that provide different ways to estimate the species richness.

Richness index differs from the concept of species diversity or evenness in that it ignores species abundance, and focuses on the binary presence/absence values that indicate simply whether the species was detected.

The function takes all index names in full lowercase. The user can provide the desired spelling through the argument `name` (see examples).

The following richness indices are provided.

- 'ace': Abundance-based coverage estimator (ACE) is another nonparametric richness index that uses sample coverage, defined based on the sum of the probabilities of the observed species. This method divides the species into abundant (more than 10 reads or observations) and rare groups in a sample and tends to underestimate the real number of species. The ACE index ignores the abundance information for the abundant species, based on the assumption that the abundant species are observed regardless of their exact abundance. We use here the bias-corrected version (O'Hara 2005, Chiu et al. 2014) implemented in `estimateR`. For an exact formulation, see `estimateR`. Note that this index comes with an additional column with standard error information.
- 'chao1': This is a nonparametric estimator of species richness. It assumes that rare species carry information about the (unknown) number of unobserved species. We use here the bias-corrected version (O'Hara 2005, Chiu et al. 2014) implemented in `estimateR`. This index implicitly assumes that every taxa has equal probability of being observed. Note that it gives a lower bound to species richness. The bias-corrected for an exact formulation, see `estimateR`. This estimator uses only the singleton and doubleton counts, and hence it gives more weight to the low abundance species. Note that this index comes with an additional column with standard error information.

- 'hill': Effective species richness aka Hill index (see e.g. Chao et al. 2016). Currently only the case 1D is implemented. This corresponds to the exponent of Shannon diversity. Intuitively, the effective richness indicates the number of species whose even distribution would lead to the same diversity than the observed community, where the species abundances are unevenly distributed.
- 'observed': The *observed richness* gives the number of species that is detected above a given detection threshold in the observed sample (default 0). This is conceptually the simplest richness index. The corresponding index in the **vegan** package is "richness".

Value

x with additional colData column(s) named code

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Simpson EH (1949) Measurement of Diversity. *Nature* 163(688). doi: 10.1038/163688a0

Smith B and Wilson JB. (1996) A Consumer's Guide to Evenness Indices. *Oikos* 76(1):70-82.

Spellerberg and Fedor (2003). A tribute to Claude Shannon (1916 –2001) and a plea for more rigorous use of species richness, species diversity and the 'Shannon–Wiener' Index. *Alpha Ecology & Biogeography* 12, 177–197.

See Also

- [plotColData](#)
- [estimateR](#)
- [diversity](#)

Examples

```
data("GlobalPatterns")
tse <- GlobalPatterns

# Calculate the default Shannon index with no rarefaction
tse <- addAlpha(tse, index = "shannon")

# Shows the estimated Shannon index
tse$shannon

# Calculate observed richness with 10 rarefaction rounds
tse <- addAlpha(tse,
  assay.type = "counts",
  index = "observed_richness",
  sample = min(colSums(assay(tse, "counts")), na.rm = TRUE),
  niter=10)

# Shows the estimated observed richness
tse$observed_richness
```

addCluster

Clustering wrapper

Description

This function returns a SummarizedExperiment with clustering information in its colData or row-Data

Usage

```

addCluster(
  x,
  BLUSPARAM,
  assay.type = assay_name,
  assay_name = "counts",
  by = MARGIN,
  MARGIN = "rows",
  full = FALSE,
  name = "clusters",
  clust.col = "clusters",
  ...
)

## S4 method for signature 'SummarizedExperiment'
addCluster(
  x,
  BLUSPARAM,
  assay.type = assay_name,
  assay_name = "counts",
  by = MARGIN,
  MARGIN = "rows",
  full = FALSE,
  name = "clusters",
  clust.col = "clusters",
  ...
)

```

Arguments

<code>x</code>	A SummarizedExperiment object.
<code>BLUSPARAM</code>	A BlusterParam object specifying the algorithm to use.
<code>assay.type</code>	Character scalar. Specifies the name of the assay used in calculation. (Default: "counts")
<code>assay_name</code>	Deprecated. Use <code>assay.type</code> instead.
<code>by</code>	Character scalar. Determines if association is calculated row-wise / for features ('rows') or column-wise / for samples ('cols'). Must be 'rows' or 'cols'.
<code>MARGIN</code>	Deprecated. Use <code>by</code> instead.
<code>full</code>	Logical scalar indicating whether the full clustering statistics should be returned for each method.
<code>name</code>	Character scalar. The name to store the result in metadata
<code>clust.col</code>	Character scalar. Indicates the name of the <code>rowData</code> (or <code>colData</code>) where the data will be stored. (Default: "clusters")
<code>...</code>	Additional parameters to use <code>altExps</code> for example

Details

This is a wrapper for the `clusterRows` function from the [bluster](#) package.

When setting `full = TRUE`, the clustering information will be stored in the metadata of the object.

By default, clustering is done on the features.

Value

`addCluster` returns an object of the same type as the `x` parameter with clustering information named `clusters` stored in `colData` or `rowData`.

Examples

```
library(bluster)
data(GlobalPatterns, package = "mia")
tse <- GlobalPatterns

# Cluster on rows using Kmeans
tse <- addCluster(tse, KmeansParam(centers = 3))

# Clustering done on the samples using Hclust
tse <- addCluster(tse,
  by = "samples",
  HclustParam(metric = "bray", dist.fun = vegan::vegdist))

# Getting the clusters
colData(tse)$clusters
```

addDivergence

Estimate divergence

Description

Estimate divergence against a given reference sample.

Usage

```
addDivergence(x, name = "divergence", ...)

## S4 method for signature 'SummarizedExperiment'
addDivergence(x, name = "divergence", ...)

getDivergence(
  x,
  assay.type = assay_name,
  assay_name = "counts",
  reference = "median",
```

```

    method = "bray",
    ...
)

## S4 method for signature 'SummarizedExperiment'
getDivergence(
  x,
  assay.type = assay_name,
  assay_name = "counts",
  reference = "median",
  method = "bray",
  ...
)

```

Arguments

x	a SummarizedExperiment object.
name	Character scalar. The name to be used to store the result in metadata of the output. (Default: method)
...	optional arguments passed to addDissimilarity . Additionally: <ul style="list-style-type: none"> • dimred: Character scalar. Specifies the name of dimension reduction result from <code>reducedDim(x)</code>. If used, these values are used to calculate divergence instead of the assay. Can be disabled with NULL. (Default: NULL)
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
assay_name	Deprecated. Use <code>assay.type</code> instead.
reference	Character scalar. A column name from <code>colData(x)</code> or either "mean" or "median". (Default: "median")
method	Character scalar. Specifies which dissimilarity to calculate. (Default: "bray")

Details

Microbiota divergence (heterogeneity / spread) within a given sample set can be quantified by the average sample dissimilarity or beta diversity with respect to a given reference sample.

The calculation makes use of the function `getDissimilarity()`. The divergence measure is sensitive to sample size. Subsampling or bootstrapping can be applied to equalize sample sizes between comparisons.

Value

x with additional `colData` named name

See Also

- [addAlpha](#)
- [addDissimilarity](#)
- [plotColData](#)

Examples

```

data(GlobalPatterns)
tse <- GlobalPatterns

# By default, reference is median of all samples. The name of column where
# results is "divergence" by default, but it can be specified.
tse <- addDivergence(tse)

# The method that are used to calculate distance in divergence and
# reference can be specified. Here, euclidean distance is used. Reference is
# the first sample. It is recommended # to add reference to colData.
tse[["reference"]] <- rep(colnames(tse)[[1]], ncol(tse))
tse <- addDivergence(
  tse, name = "divergence_first_sample",
  reference = "reference",
  method = "euclidean")

# Here we compare samples to global mean
tse <- addDivergence(tse, name = "divergence_average", reference = "mean")

# All three divergence results are stored in colData.
colData(tse)

```

addLDA

Latent Dirichlet Allocation

Description

These functions perform Latent Dirichlet Allocation on data stored in a [TreeSummarizedExperiment](#) object.

Usage

```

getLDA(x, ...)

addLDA(x, ...)

## S4 method for signature 'SummarizedExperiment'
getLDA(x, k = 2, assay.type = "counts", eval.metric = "perplexity", ...)

## S4 method for signature 'SummarizedExperiment'
addLDA(x, k = 2, assay.type = "counts", name = "LDA", ...)

```

Arguments

x a [TreeSummarizedExperiment](#) object.
... optional arguments passed to [LDA](#)

k	Integer vector. A number of latent vectors/topics. (Default: 2)
assay.type	Character scalar. Specifies which assay to use for LDA ordination. (Default: "counts")
eval.metric	Character scalar. Specifies evaluation metric that will be used to select the model with the best fit. Must be either "perplexity" (<code>topicmodels::perplexity</code>) or "coherence" (<code>topicdoc::topic_coherence</code> , the best model is selected based on mean coherence). (Default: "perplexity")
name	Character scalar. The name to be used to store the result in the <code>reducedDims</code> of the output. (Default: "LDA")

Details

The functions `getLDA` and `addLDA` internally use [LDA](#) to compute the ordination matrix and feature loadings.

Value

For `getLDA`, the ordination matrix with feature loadings matrix as attribute "loadings".

For `addLDA`, a [TreeSummarizedExperiment](#) object is returned containing the ordination matrix in `reducedDim(..., name)` with feature loadings matrix as attribute "loadings".

Examples

```
data(GlobalPatterns)
tse <- GlobalPatterns

# Reduce the number of features
tse <- agglomerateByPrevalence(tse, rank="Phylum")

# Run LDA and add the result to reducedDim(tse, "LDA")
tse <- addLDA(tse)

# Extract feature loadings
loadings <- attr(reducedDim(tse, "LDA"), "loadings")
head(loadings)

# Estimate models with number of topics from 2 to 10
tse <- addLDA(tse, k = c(2, 3, 4, 5, 6, 7, 8, 9, 10), name = "LDA_10")
# Get the evaluation metrics
tab <- attr(reducedDim(tse, "LDA_10"), "eval_metrics")
# Plot
plot(tab[["k"]], tab[["perplexity"]], xlab = "k", ylab = "perplexity")
```

addNMF *Non-negative Matrix Factorization*

Description

These functions perform Non-negative Matrix Factorization on data stored in a [TreeSummarizedExperiment](#) object.

Usage

```
getNMF(x, ...)

addNMF(x, ...)

## S4 method for signature 'SummarizedExperiment'
getNMF(x, k = 2, assay.type = "counts", eval.metric = "evar", ...)

## S4 method for signature 'SummarizedExperiment'
addNMF(
  x,
  k = 2,
  assay.type = "counts",
  eval.metric = "evar",
  name = "NMF",
  ...
)
```

Arguments

x	a TreeSummarizedExperiment object.
...	optional arguments passed to <code>nmf::NMF</code> .
k	numeric vector. A number of latent vectors/topics. (Default: 2)
assay.type	Character scalar. Specifies which assay to use for NMF ordination. (Default: "counts")
eval.metric	Character scalar. Specifies the evaluation metric that will be used to select the model with the best fit. Must be one of the following options: "evar" (explained variance; maximized), "sparseness.basis" (degree of sparsity in the basis matrix; maximized), "sparseness.coef" (degree of sparsity in the coefficient matrix; maximized), "rss" (residual sum of squares; minimized), "silhouette.coef" (quality of clustering based on the coefficient matrix; maximized), "silhouette.basis" (quality of clustering based on the basis matrix; maximized), "cophenetic" (correlation between cophenetic distances and original distances; maximized), "dispersion" (spread of data points within clusters; minimized). (Default: "evar")
name	Character scalar. The name to be used to store the result in the reducedDims of the output. (Default: "NMF")

Details

The functions `getNMF` and `addNMF` internally use `nmf::NMF` compute the ordination matrix and feature loadings.

If `k` is a vector of integers, NMF output is calculated for all the rank values contained in `k`, and the best fit is selected based on `eval.metric` value.

Value

For `getNMF`, the ordination matrix with feature loadings matrix as attribute `"loadings"`.

For `addNMF`, a `TreeSummarizedExperiment` object is returned containing the ordination matrix in `reducedDims(x, name)` with the following attributes:

- `"loadings"` which is a matrix containing the feature loadings
- `"NMF_output"` which is the output of function `nmf::NMF`
- `"best_fit"` which is the result of the best fit if `k` is a vector of integers

Examples

```
data(GlobalPatterns)
tse <- GlobalPatterns

# Reduce the number of features
tse <- agglomerateByPrevalence(tse, rank = "Phylum")

# Run NMF and add the result to reducedDim(tse, "NMF").
tse <- addNMF(tse, k = 2, name = "NMF")

# Extract feature loadings
loadings_NMF <- attr(reducedDim(tse, "NMF"), "loadings")
head(loadings_NMF)

# Estimate models with number of topics from 2 to 4. Perform 2 runs.
tse <- addNMF(tse, k = c(2, 3, 4), name = "NMF_4", nrun = 2)

# Extract feature loadings
loadings_NMF_4 <- attr(reducedDim(tse, "NMF_4"), "loadings")
head(loadings_NMF_4)
```

Description

Agglomeration functions can be used to sum-up data based on specific criteria such as taxonomic ranks, variables or prevalence.

`agglomerateByRank` can be used to sum up data based on associations with certain taxonomic ranks, as defined in `rowData`. Only available `taxonomyRanks` can be used.

`agglomerateByVariable` merges data on rows or columns of a `SummarizedExperiment` as defined by a factor alongside the chosen dimension. This function allows agglomeration of data based on other variables than taxonomy ranks. Metadata from the `rowData` or `colData` are retained as defined by archetype. `assay` are agglomerated, i.e. summed up. If the assay contains values other than counts or absolute values, this can lead to meaningless values being produced.

`agglomerateByRanks` takes a `SummarizedExperiment`, splits it along the taxonomic ranks, aggregates the data per rank, converts the input to a `SingleCellExperiment` objects and stores the aggregated data as alternative experiments. `unsplitByRanks` takes these alternative experiments and flattens them again into a single `SummarizedExperiment`.

Usage

```
agglomerateByRank(x, ...)
```

```
## S4 method for signature 'TreeSummarizedExperiment'
```

```
agglomerateByRank(
  x,
  rank = taxonomyRanks(x)[1],
  update.tree = agglomerateTree,
  agglomerate.tree = agglomerateTree,
  agglomerateTree = FALSE,
  ...
)
```

```
## S4 method for signature 'SingleCellExperiment'
```

```
agglomerateByRank(
  x,
  rank = taxonomyRanks(x)[1],
  altexp = NULL,
  altexp.rm = strip_altexp,
  strip_altexp = TRUE,
  ...
)
```

```
## S4 method for signature 'SummarizedExperiment'
```

```
agglomerateByRank(
  x,
  rank = taxonomyRanks(x)[1],
  empty.rm = TRUE,
  empty.fields = c(NA, "", " ", "\t", "-", "_"),
  ...
)
```

```
agglomerateByVariable(x, ...)  
  
## S4 method for signature 'TreeSummarizedExperiment'  
agglomerateByVariable(  
  x,  
  by,  
  group = f,  
  f,  
  update.tree = mergeTree,  
  mergeTree = FALSE,  
  ...  
)  
  
## S4 method for signature 'SummarizedExperiment'  
agglomerateByVariable(x, by, group = f, f, ...)  
  
agglomerateByRanks(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
agglomerateByRanks(  
  x,  
  ranks = taxonomyRanks(x),  
  na.rm = TRUE,  
  as.list = FALSE,  
  ...  
)  
  
## S4 method for signature 'SingleCellExperiment'  
agglomerateByRanks(  
  x,  
  ranks = taxonomyRanks(x),  
  na.rm = TRUE,  
  as.list = FALSE,  
  ...  
)  
  
## S4 method for signature 'TreeSummarizedExperiment'  
agglomerateByRanks(  
  x,  
  ranks = taxonomyRanks(x),  
  na.rm = TRUE,  
  as.list = FALSE,  
  ...  
)  
  
splitByRanks(x, ...)
```

```

unsplitByRanks(x, ...)

## S4 method for signature 'SingleCellExperiment'
unsplitByRanks(
  x,
  ranks = taxonomyRanks(x),
  keep.dimred = keep_reducedDims,
  keep_reducedDims = FALSE,
  ...
)

## S4 method for signature 'TreeSummarizedExperiment'
unsplitByRanks(
  x,
  ranks = taxonomyRanks(x),
  keep.dimred = keep_reducedDims,
  keep_reducedDims = FALSE,
  ...
)

```

Arguments

x	TreeSummarizedExperiment .
...	arguments passed to <code>agglomerateByRank</code> function for <code>SummarizedExperiment</code> objects and other functions. See agglomerateByRank for more details.
rank	Character scalar. Defines a taxonomic rank. Must be a value of <code>taxonomyRanks()</code> function.
update.tree	Logical scalar. Should <code>rowTree()</code> also be merged? (Default: FALSE)
agglomerate.tree	Deprecated. Use <code>update.tree</code> instead.
agglomerateTree	Deprecated. Use <code>update.tree</code> instead.
altexp	Character scalar or integer scalar. Specifies an alternative experiment containing the input data.
altexp.rm	Logical scalar. Should alternative experiments be removed prior to agglomeration? This prevents too many nested alternative experiments by default. (Default: TRUE)
strip_altexp	Deprecated. Use <code>altexp.rm</code> instead.
empty.rm	Logical scalar. Defines whether rows including empty fields in specified rank will be excluded. (Default: TRUE)
empty.fields	Character vector. Defines which values should be regarded as empty. (Default: <code>c(NA, "", " ", "\t")</code>). They will be removed if <code>na.rm = TRUE</code> before agglomeration.
by	Character scalar. Determines if data is merged row-wise / for features ('rows') or column-wise / for samples ('cols'). Must be 'rows' or 'cols'.

group	Character scalar, character vector or factor vector. A column name from <code>rowData(x)</code> or <code>colData(x)</code> or alternatively a vector specifying how the merging is performed. If vector, the value must be the same length as <code>nrow(x)/ncol(x)</code> . Rows/Cols corresponding to the same level will be merged. If <code>length(levels(group)) == nrow(x)/ncol(x)</code> , <code>x</code> will be returned unchanged.
f	Deprecated. Use <code>group</code> instead.
mergeTree	Deprecated. Use <code>update.tree</code> instead.
ranks	Character vector. Defines taxonomic ranks. Must all be values of <code>taxonomyRanks()</code> function.
na.rm	Logical scalar. Should NA values be omitted? (Default: TRUE)
as.list	Logical scalar. Should the list of <code>SummarizedExperiment</code> objects be returned by the function <code>agglomerateByRanks</code> as a <code>SimpleList</code> or stored in <code>altExps</code> ? (Default: FALSE)
keep.dimred	Logical scalar. Should the <code>reducedDims(x)</code> be transferred to the result? Please note, that this breaks the link between the data used to calculate the reduced dims. (Default: FALSE)
keep_reducedDims	Deprecated. Use <code>keep.dimred</code> instead.

Details

Agglomeration sums up the values of assays at the specified taxonomic level. With certain assays, e.g. those that include binary or negative values, this summing can produce meaningless values. In those cases, consider performing agglomeration first, and then applying the transformation afterwards.

`agglomerateByVariable` works similarly to `sumCountsAcrossFeatures`. However, additional support for `TreeSummarizedExperiment` was added and science field agnostic names were used. In addition the `archetype` argument lets the user select how to preserve row or column data.

For merge data of assays the function from `scuttle` are used.

`agglomerateByRanks` will use by default all available taxonomic ranks, but this can be controlled by setting `ranks` manually. NA values are removed by default, since they would not make sense, if the result should be used for `unsplitByRanks` at some point. The input data remains unchanged in the returned `SingleCellExperiment` objects.

`unsplitByRanks` will remove any NA value on each taxonomic rank so that no ambiguous data is created. In addition, a column `taxonomicLevel` is created or overwritten in the `rowData` to specify from which alternative experiment this originates from. This can also be used for `splitAltExps` to split the result along the same factor again. The input data from the base objects is not returned, only the data from the `altExp()`. Be aware that changes to `rowData` of the base object are not returned, whereas only the `colData` of the base object is kept.

Value

`agglomerateByRank` returns a taxonomically-agglomerated, optionally-pruned object of the same class as `x`. `agglomerateByVariable` returns an object of the same class as `x` with the specified entries merged into one entry in all relevant components. `agglomerateByRank` returns a taxonomically-agglomerated, optionally-pruned object of the same class as `x`.

For `agglomerateByRanks`: If `as.list = TRUE` : SummarizedExperiment objects in a SimpleList
 If `as.list = FALSE` : The SummarizedExperiment passed as a parameter and now containing the SummarizedExperiment objects in its `altExps`

For `unsplitByRanks`: `x`, with `rowData` and assay data replaced by the unsplit data. `colData` of `x` is kept as well and any existing `rowTree` is dropped as well, since existing `rowLinks` are not valid anymore.

See Also

[splitOn](#), [unsplitOn](#), [agglomerateByVariable](#), [sumCountsAcrossFeatures](#), [agglomerateByRank](#), [altExps](#), [splitAltExps](#)

Examples

```
### Agglomerate data based on taxonomic information

data(GlobalPatterns)
# print the available taxonomic ranks
colnames(rowData(GlobalPatterns))
taxonomyRanks(GlobalPatterns)

# agglomerate at the Family taxonomic rank
x1 <- agglomerateByRank(GlobalPatterns, rank="Family")
## How many taxa before/after agglomeration?
nrow(GlobalPatterns)
nrow(x1)

# agglomerate the tree as well
x2 <- agglomerateByRank(GlobalPatterns, rank="Family",
                        update.tree = TRUE)
nrow(x2) # same number of rows, but
rowTree(x1) # ... different
rowTree(x2) # ... tree

# If assay contains binary or negative values, summing might lead to
# meaningless values, and you will get a warning. In these cases, you might
# want to do agglomeration again at chosen taxonomic level.
tse <- transformAssay(GlobalPatterns, method = "pa")
tse <- agglomerateByRank(tse, rank = "Genus")
tse <- transformAssay(tse, method = "pa")

# Removing empty labels by setting empty.rm = TRUE
sum(is.na(rowData(GlobalPatterns)$Family))
x3 <- agglomerateByRank(GlobalPatterns, rank="Family", empty.rm = TRUE)
nrow(x3) # different from x2

# Because all the rownames are from the same rank, rownames do not include
# prefixes, in this case "Family:".
print(rownames(x3[1:3,]))

# To add them, use getTaxonomyLabels function.
rownames(x3) <- getTaxonomyLabels(x3, with.rank = TRUE)
```

```

print(rownames(x3[1:3,]))

# use 'empty.ranks.rm' to remove columns that include only NAs
x4 <- agglomerateByRank(
  GlobalPatterns, rank="Phylum", empty.ranks.rm = TRUE)
head(rowData(x4))

# If the assay contains NAs, you might want to specify na.rm=TRUE,
# since summing-up NAs lead to NA
x5 <- GlobalPatterns
# Replace first value with NA
assay(x5)[1,1] <- NA
x6 <- agglomerateByRank(x5, "Kingdom")
head( assay(x6) )
# Use na.rm=TRUE
x6 <- agglomerateByRank(x5, "Kingdom", na.rm = TRUE)
head( assay(x6) )

## Look at enterotype dataset...
data(enterotype)
## Print the available taxonomic ranks. Shows only 1 available rank,
## not useful for agglomerateByRank
taxonomyRanks(enterotype)

### Merge TreeSummarizedExperiments on rows and columns

data(esophagus)
esophagus
plot(rowTree(esophagus))
# Get a factor for merging
f <- factor(regmatches(rownames(esophagus),
  regexpr("[0-9]*_[0-9]*", rownames(esophagus))))
merged <- agglomerateByVariable(
  esophagus, by = "rows", f, update.tree = TRUE)
plot(rowTree(merged))
#
data(GlobalPatterns)
GlobalPatterns
merged <- agglomerateByVariable(
  GlobalPatterns, by = "cols", colData(GlobalPatterns)$SampleType)
merged

data(GlobalPatterns)
# print the available taxonomic ranks
taxonomyRanks(GlobalPatterns)

# agglomerateByRanks
#
tse <- agglomerateByRanks(GlobalPatterns)
altExps(tse)
altExp(tse, "Kingdom")
altExp(tse, "Species")

```



```
# unsplitByRanks
tse <- unsplitByRanks(tse)
tse
```

agglomerateByPrevalence

Agglomerate data based on population prevalence

Description

Agglomerate data based on population prevalence

Usage

```
agglomerateByPrevalence(x, ...)
```

```
## S4 method for signature 'SummarizedExperiment'
```

```
agglomerateByPrevalence(
  x,
  rank = NULL,
  other.name = other_label,
  other_label = "Other",
  ...
)
```

```
## S4 method for signature 'TreeSummarizedExperiment'
```

```
agglomerateByPrevalence(
  x,
  rank = NULL,
  other.name = other_label,
  other_label = "Other",
  update.tree = FALSE,
  ...
)
```

Arguments

x	TreeSummarizedExperiment .
...	arguments passed to <code>agglomerateByRank</code> function for <code>SummarizedExperiment</code> objects and other functions. See agglomerateByRank for more details.
rank	Character scalar. Defines a taxonomic rank. Must be a value of <code>taxonomyRanks()</code> function.
other.name	Character scalar. Used as the label for the summary of non-prevalent taxa. (default: "Other")
other_label	Deprecated. use <code>other.name</code> instead.
update.tree	Logical scalar. Should <code>rowTree()</code> also be merged? (Default: FALSE)

Details

agglomerateByPrevalence sums up the values of assays at the taxonomic level specified by rank (by default the highest taxonomic level available) and selects the summed results that exceed the given population prevalence at the given detection level. The other summed values (below the threshold) are agglomerated in an additional row taking the name indicated by other.name (by default "Other").

Value

agglomerateByPrevalence returns a taxonomically-agglomerated object of the same class as x and based on prevalent taxonomic results.

Examples

```
## Data can be aggregated based on prevalent taxonomic results
data(GlobalPatterns)
tse <- GlobalPatterns
tse <- transformAssay(tse, method = "relabundance")
tse <- agglomerateByPrevalence(
  tse,
  rank = "Phylum",
  assay.type = "relabundance",
  detection = 1/100,
  prevalence = 50/100)

tse

# Here data is aggregated at the taxonomic level "Phylum". The five phyla
# that exceed the population prevalence threshold of 50/100 represent the
# five first rows of the assay in the aggregated data. The sixth and last row
# named by default "Other" takes the summed up values of all the other phyla
# that are below the prevalence threshold.

assay(tse)[,1:5]
```

calculateDMN

Dirichlet-Multinomial Mixture Model: Machine Learning for Microbiome Data

Description

These functions are accessors for functions implemented in the [DirichletMultinomial](#) package

Usage

```
calculateDMN(x, ...)

## S4 method for signature 'ANY'
```

```
calculateDMN(  
  x,  
  k = 1,  
  BPPARAM = SerialParam(),  
  seed = runif(1, 0, .Machine$integer.max),  
  ...  
)  
  
## S4 method for signature 'SummarizedExperiment'  
calculateDMN(  
  x,  
  assay.type = assay_name,  
  assay_name = exprs_values,  
  exprs_values = "counts",  
  transposed = FALSE,  
  ...  
)  
  
runDMN(x, name = "DMN", ...)  
  
getDMN(x, name = "DMN", ...)  
  
## S4 method for signature 'SummarizedExperiment'  
getDMN(x, name = "DMN")  
  
bestDMNFit(x, name = "DMN", type = c("laplace", "AIC", "BIC"), ...)  
  
## S4 method for signature 'SummarizedExperiment'  
bestDMNFit(x, name = "DMN", type = c("laplace", "AIC", "BIC"))  
  
getBestDMNFit(x, name = "DMN", type = c("laplace", "AIC", "BIC"), ...)  
  
## S4 method for signature 'SummarizedExperiment'  
getBestDMNFit(x, name = "DMN", type = c("laplace", "AIC", "BIC"))  
  
calculateDMNgroup(x, ...)  
  
## S4 method for signature 'ANY'  
calculateDMNgroup(  
  x,  
  variable,  
  k = 1,  
  seed = runif(1, 0, .Machine$integer.max),  
  ...  
)  
  
## S4 method for signature 'SummarizedExperiment'  
calculateDMNgroup(  
  x,  
  variable,  
  k = 1,  
  seed = runif(1, 0, .Machine$integer.max),  
  ...  
)
```

```

    x,
    variable,
    assay.type = assay_name,
    assay_name = exprs_values,
    exprs_values = "counts",
    transposed = FALSE,
    ...
)

performDMNgroupCV(x, ...)

## S4 method for signature 'ANY'
performDMNgroupCV(
  x,
  variable,
  k = 1,
  seed = runif(1, 0, .Machine$integer.max),
  ...
)

## S4 method for signature 'SummarizedExperiment'
performDMNgroupCV(
  x,
  variable,
  assay.type = assay_name,
  assay_name = exprs_values,
  exprs_values = "counts",
  transposed = FALSE,
  ...
)

```

Arguments

x	a numeric matrix with samples as rows or a SummarizedExperiment object.
...	optional arguments not used.
k	Numeric scalar. The number of Dirichlet components to fit. See dmn . (Default: 1)
BPPARAM	A BiocParallelParam object specifying whether the calculation should be parallelized.
seed	Numeric scalar. Random number seed. See dmn
assay.type	Character scalar. Specifies the name of the assay used in calculation. (Default: "counts")
assay_name	Deprecated. Use <code>assay.type</code> instead.
exprs_values	Deprecated. Use <code>assay.type</code> instead.
transposed	Logical scalar. Is x transposed with samples in rows? (Default: FALSE)
name	Character scalar. The name to store the result in metadata

type	Character scalar. The type of measure used for the goodness of fit. One of 'laplace', 'AIC' or 'BIC'.
variable	Character scalar. A variable from colData to use as a grouping variable. Must be a character of factor.

Value

calculateDMN and getDMN return a list of DMN objects, one element for each value of k provided.

bestDMNFit returns the index for the best fit and getBestDMNFit returns a single DMN object.

calculateDMNgroup returns a [DMNGroup](#) object

performDMNgroupCV returns a data.frame

See Also

[DMN-class](#), [DMNGroup-class](#), [dmn](#), [dmn](#)group, [cvdmn](#)group, [accessors for DMN objects](#)

Examples

```
f1 <- system.file(package="DirichletMultinomial", "extdata", "Twins.csv")
counts <- as.matrix(read.csv(f1, row.names=1))
f1 <- system.file(package="DirichletMultinomial", "extdata", "TwinStudy.t")
pheno0 <- scan(f1)
lvls <- c("Lean", "Obese", "Overwt")
pheno <- factor(lvls[pheno0 + 1], levels=lvls)
colData <- DataFrame(pheno = pheno)

tse <- TreeSummarizedExperiment(assays = list(counts = counts),
                                colData = colData)

library(bluster)

# Compute DMM algorithm and store result in metadata
tse <- addCluster(tse, name = "DMM", DmmParam(k = 1:3, type = "laplace"),
                 by = "samples", full = TRUE)

# Get the list of DMN objects
metadata(tse)$DMM$dmm

# Get and display which objects fits best
bestFit <- metadata(tse)$DMM$best
bestFit

# Get the model that generated the best fit
bestModel <- metadata(tse)$DMM$dmm[[bestFit]]
bestModel

# Get the sample-cluster assignment probability matrix
head(metadata(tse)$DMM$prob)

# Get the weight of each component for the best model
bestModel@mixture$Weight
```

convertFromDADA2	<i>Create a TreeSummarizedExperiment object from 'DADA2' results</i>
------------------	--

Description

Create a TreeSummarizedExperiment object from 'DADA2' results

Usage

```
convertFromDADA2(...)
```

Arguments

... Additional arguments. For convertFromDADA2, see mergePairs function for more details.

Details

convertFromDADA2 is a wrapper for the mergePairs function from the dada2 package. A count matrix is constructed via makeSequenceTable(mergePairs(...)) and rownames are dynamically created as ASV(N) with N from 1 to nrow of the count tables. The colnames and rownames from the output of makeSequenceTable are stored as colnames and in the referenceSeq slot of the TreeSummarizedExperiment, respectively.

Value

convertFromDADA2 returns an object of class [TreeSummarizedExperiment](#)

Examples

```
### Coerce DADA2 results to a TreeSE object
if(requireNamespace("dada2")) {
  fnF <- system.file("extdata", "sam1F.fastq.gz", package="dada2")
  fnR = system.file("extdata", "sam1R.fastq.gz", package="dada2")
  dadaF <- dada2::dada(fnF, selfConsist=TRUE)
  dadaR <- dada2::dada(fnR, selfConsist=TRUE)

  tse <- convertFromDADA2(dadaF, fnF, dadaR, fnR)
  tse
}
```

convertFromPhyloseq *Create a TreeSummarizedExperiment object from a phyloseq object*

Description

Create a `TreeSummarizedExperiment` object from a `phyloseq` object

Create a `phyloseq` object from a `TreeSummarizedExperiment` object

Usage

```
convertFromPhyloseq(x)
```

```
convertToPhyloseq(x, ...)
```

```
## S4 method for signature 'SummarizedExperiment'
convertToPhyloseq(x, assay.type = "counts", assay_name = NULL, ...)
```

```
## S4 method for signature 'TreeSummarizedExperiment'
convertToPhyloseq(x, tree.name = tree_name, tree_name = "phylo", ...)
```

Arguments

<code>x</code>	a <code>TreeSummarizedExperiment</code> object
<code>...</code>	Additional arguments. Not used currently.
<code>assay.type</code>	Character scalar. Specifies the name of assay used. (Default: "counts")
<code>assay_name</code>	Deprecated. Use <code>assay.type</code> instead.
<code>tree.name</code>	Character scalar. Specifies the name of the tree to be included in the <code>phyloseq</code> object that is created, (Default: "phylo")
<code>tree_name</code>	Deprecated. Use <code>tree.name</code> instead.

Details

`convertFromPhyloseq` converts `phyloseq` objects into `TreeSummarizedExperiment` objects. All data stored in a `phyloseq` object is transferred.

`convertToPhyloseq` creates a `phyloseq` object from a `TreeSummarizedExperiment` object. By using `assay.type`, it is possible to specify which table from assay is added to the `phyloseq` object.

Value

`convertFromPhyloseq` returns an object of class `TreeSummarizedExperiment`

`convertToPhyloseq` returns an object of class `phyloseq`

Examples

```
### Coerce a phyloseq object to a TreeSE object
if (requireNamespace("phyloseq")) {
  data(GlobalPatterns, package="phyloseq")
  convertFromPhyloseq(GlobalPatterns)
  data(enterotype, package="phyloseq")
  convertFromPhyloseq(enterotype)
  data(esophagus, package="phyloseq")
  convertFromPhyloseq(esophagus)
}

### Coerce a TreeSE object to a phyloseq object
# Get tse object
data(GlobalPatterns)
tse <- GlobalPatterns

# Create a phyloseq object from it
phy <- convertToPhyloseq(tse)
phy

# By default the chosen table is counts, but if there are other tables,
# they can be chosen with assay.type.

# Counts relative abundances table
tse <- transformAssay(tse, method = "relabundance")
phy2 <- convertToPhyloseq(tse, assay.type = "relabundance")
phy2
```

deprecate

These functions will be deprecated. Please use other functions instead.

Description

These functions will be deprecated. Please use other functions instead.

Usage

```
cluster(x, ...)

## S4 method for signature 'SummarizedExperiment'
cluster(x, ...)

addTaxonomyTree(x, ...)

## S4 method for signature 'SummarizedExperiment'
addTaxonomyTree(x, ...)

taxonomyTree(x, ...)
```



```
## S4 method for signature 'SummarizedExperiment'  
taxonomyTree(x, ...)  
  
mergeRows(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
mergeRows(x, ...)  
  
## S4 method for signature 'TreeSummarizedExperiment'  
mergeRows(x, ...)  
  
mergeCols(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
mergeCols(x, ...)  
  
## S4 method for signature 'TreeSummarizedExperiment'  
mergeCols(x, ...)  
  
mergeFeatures(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
mergeFeatures(x, ...)  
  
## S4 method for signature 'TreeSummarizedExperiment'  
mergeFeatures(x, ...)  
  
mergeSamples(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
mergeSamples(x, ...)  
  
## S4 method for signature 'TreeSummarizedExperiment'  
mergeSamples(x, ...)  
  
mergeFeaturesByRank(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
mergeFeaturesByRank(x, ...)  
  
## S4 method for signature 'SingleCellExperiment'  
mergeFeaturesByRank(x, ...)  
  
mergeFeaturesByPrevalence(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
mergeFeaturesByPrevalence(x, ...)
```

```
getExperimentCrossAssociation(x, ...)  
  
## S4 method for signature 'MultiAssayExperiment'  
getExperimentCrossAssociation(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
getExperimentCrossAssociation(x, ...)  
  
## S4 method for signature 'TreeSummarizedExperiment'  
mergeFeaturesByRank(x, ...)  
  
testExperimentCrossAssociation(x, ...)  
  
## S4 method for signature 'ANY'  
testExperimentCrossAssociation(x, ...)  
  
testExperimentCrossCorrelation(x, ...)  
  
## S4 method for signature 'ANY'  
testExperimentCrossCorrelation(x, ...)  
  
getExperimentCrossCorrelation(x, ...)  
  
## S4 method for signature 'ANY'  
getExperimentCrossCorrelation(x, ...)  
  
loadFromBiom(...)  
  
loadFromQIIME2(...)  
  
readQZA(...)  
  
loadFromMothur(...)  
  
loadFromMetaphlan(...)  
  
loadFromHumann(...)  
  
countDominantFeatures(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
countDominantFeatures(x, ...)  
  
subsetByRareTaxa(x, ...)  
  
## S4 method for signature 'ANY'  
subsetByRareTaxa(x, ...)
```

```
subsetByRareFeatures(x, ...)  
  
## S4 method for signature 'ANY'  
subsetByRareFeatures(x, ...)  
  
subsetByPrevalentTaxa(x, ...)  
  
## S4 method for signature 'ANY'  
subsetByPrevalentTaxa(x, ...)  
  
subsetByPrevalentFeatures(x, ...)  
  
## S4 method for signature 'ANY'  
subsetByPrevalentFeatures(x, ...)  
  
countDominantTaxa(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
countDominantTaxa(x, ...)  
  
full_join(x, ...)  
  
## S4 method for signature 'ANY'  
full_join(x, ...)  
  
inner_join(x, ...)  
  
## S4 method for signature 'ANY'  
inner_join(x, ...)  
  
left_join(x, ...)  
  
## S4 method for signature 'ANY'  
left_join(x, ...)  
  
right_join(x, ...)  
  
## S4 method for signature 'ANY'  
right_join(x, ...)  
  
plotNMDS(x, ...)  
  
estimateDivergence(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
estimateDivergence(x, ...)
```

```
meltAssay(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
meltAssay(x, ...)  
  
transformSamples(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
transformSamples(x, ...)  
  
ZTransform(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
ZTransform(x, ...)  
  
relAbundanceCounts(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
relAbundanceCounts(x, ...)  
  
transformCounts(x, ...)  
  
transformFeatures(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
transformFeatures(x, ...)  
  
getUniqueFeatures(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
getUniqueFeatures(x, ...)  
  
getUniqueTaxa(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
getUniqueTaxa(x, ...)  
  
getTopFeatures(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
getTopFeatures(x, ...)  
  
getTopTaxa(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
getTopTaxa(x, ...)  
  
getRareFeatures(x, ...)
```

```
## S4 method for signature 'ANY'
getRareFeatures(x, ...)

## S4 method for signature 'SummarizedExperiment'
getRareFeatures(x, ...)

getRareTaxa(x, ...)

## S4 method for signature 'ANY'
getRareTaxa(x, ...)

## S4 method for signature 'SummarizedExperiment'
getRareTaxa(x, ...)

getPrevalentFeatures(x, ...)

## S4 method for signature 'ANY'
getPrevalentFeatures(x, ...)

## S4 method for signature 'SummarizedExperiment'
getPrevalentFeatures(x, ...)

getPrevalentTaxa(x, ...)

## S4 method for signature 'ANY'
getPrevalentTaxa(x, ...)

## S4 method for signature 'SummarizedExperiment'
getPrevalentTaxa(x, ...)

subsampleCounts(x, ...)

## S4 method for signature 'SummarizedExperiment'
subsampleCounts(x, ...)

addPerSampleDominantFeatures(x, ...)

## S4 method for signature 'SummarizedExperiment'
addPerSampleDominantFeatures(x, ...)

addPerSampleDominantTaxa(x, ...)

## S4 method for signature 'SummarizedExperiment'
addPerSampleDominantTaxa(x, ...)

perSampleDominantFeatures(x, ...)
```

```
## S4 method for signature 'SummarizedExperiment'  
perSampleDominantFeatures(x, ...)  
  
perSampleDominantTaxa(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
perSampleDominantTaxa(x, ...)  
  
makePhyloseqFromTreeSE(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
makePhyloseqFromTreeSE(x)  
  
## S4 method for signature 'TreeSummarizedExperiment'  
makePhyloseqFromTreeSE(x)  
  
makePhyloseqFromTreeSummarizedExperiment(x)  
  
## S4 method for signature 'ANY'  
makePhyloseqFromTreeSummarizedExperiment(x)  
  
makeTreeSummarizedExperimentFromPhyloseq(x)  
  
## S4 method for signature 'ANY'  
makeTreeSummarizedExperimentFromPhyloseq(x)  
  
makeTreeSEFromBiom(...)  
  
makeTreeSummarizedExperimentFromBiom(...)  
  
makeTreeSEFromDADA2(...)  
  
makeTreeSummarizedExperimentFromDADA2(...)  
  
makeTreeSEFromPhyloseq(x)  
  
estimateEvenness(x, ...)  
  
## S4 method for signature 'ANY'  
estimateEvenness(x, ...)  
  
estimateRichness(x, ...)  
  
## S4 method for signature 'ANY'  
estimateRichness(x, ...)  
  
estimateDiversity(x, ...)
```

```
## S4 method for signature 'ANY'
estimateDiversity(x, ...)

estimateFaith(x, ...)

## S4 method for signature 'ANY'
estimateFaith(x, ...)

estimateDominance(x, ...)

## S4 method for signature 'ANY'
estimateDominance(x, ...)

subsetSamples(x, ...)

subsetFeatures(x, ...)

subsetTaxa(x, ...)

## S4 method for signature 'SummarizedExperiment'
subsetSamples(x, ...)

## S4 method for signature 'SummarizedExperiment'
subsetFeatures(x, ...)

## S4 method for signature 'SummarizedExperiment'
subsetTaxa(x, ...)

relabundance(x, ...)

relabundance(x) <- value

## S4 method for signature 'SummarizedExperiment'
relabundance(x)

## S4 replacement method for signature 'SummarizedExperiment'
relabundance(x) <- value

runOverlap(x, ...)

## S4 method for signature 'SummarizedExperiment'
runOverlap(x, ...)

calculateOverlap(x, ...)

## S4 method for signature 'ANY'
calculateOverlap(x, ...)
```

```

calculateJSD(x, ...)

## S4 method for signature 'ANY'
calculateJSD(x, ...)

runJSD(x, ...)

## S4 method for signature 'SummarizedExperiment'
runJSD(x, ...)

calculateUnifrac(x, ...)

## S4 method for signature 'ANY'
calculateUnifrac(x, ...)

runUnifrac(mat, tree, ...)

```

Arguments

x	A SummarizedExperiment object.
...	Additional parameters. See dedicated function.
value	a matrix to store as the ‘relabundance’ assay
mat	An abundance matrix
tree	A phylogenetic tree

dmn_se

Twins’ microbiome data from 278 individuals

Description

dmn_se is a dataset on twins’ microbiome where samples are stratified by their community composition through Dirichlet Multinomial Mixtures (DMM). It was derived from the **DirichletMultinomial** package.

Usage

```
data(dmn_se)
```

Format

A SummarizedExperiment with 130 features and 278 samples. The rowData contains no taxonomic information. The colData includes:

pheno participant’s weight condition (Lean, Overwt and Obese)

Author(s)

Turnbaugh, PJ et al.

References

Holmes I, Harris K, Quince C (2012). Dirichlet Multinomial Mixtures: Generative Models for Microbial Metagenomics. PLoS ONE 7(2): e30126. <https://doi.org/10.1371/journal.pone.0030126>

Turnbaugh PJ, Hamady M, Yatsunencko T, Cantarel BL, Duncan A, et al. (2009). A core gut microbiome in obese and lean twins. Nature 457: 480–484. <https://doi.org/10.1038/nature07540>

See Also

[mia-datasets calculateDMN](#)

enterotype	<i>Human gut microbiome dataset from 22 subjects based on shotgun DNA sequencing</i>
------------	--

Description

The enterotype data of the human gut microbiome includes taxonomic profiling for 280 fecal samples from 22 subjects based on shotgun DNA sequencing. The authors claimed that the data naturally clumps into three community-level clusters, or "enterotypes", that are not immediately explained by sequencing technology or demographic features of the subjects. In a later addendum from 2014 the authors stated that enterotypes should not be seen as discrete clusters, but as a way of stratifying samples to reduce complexity. It was converted into a TreeSummarizedExperiment from the **phyloseq** package.

Usage

```
data(enterotype)
```

Format

A TreeSummarizedExperiment with 553 features and 280 samples. The rowData contains taxonomic information at Genus level. The colData includes:

Enterotype enterotype the sample belongs to (1, 2 and 3)

Sample_ID sample ID of samples from all studies

SeqTech sequencing technology

SampleID sample ID of complete samples

Project original project from which sample was obtained (gill06, turnbaugh09, MetaHIT, MicroObes, MicroAge and kurokawa07)

Nationality participant's nationality (american, danish, spanish, french, italian and japanese)

Gender participant's gender (F or M)

Age participant's age (0.25 – 87)

ClinicalStatus participant's clinical status (healthy, obese, CD, UC and elderly)

Author(s)

Arumugam, M., Raes, J., et al.

Source

http://www.bork.embl.de/Docu/Arumugam_et_al_2011/downloads.html

References

Arumugam, M., et al. (2011). Enterotypes of the human gut microbiome. *Nature*, 473(7346), 174-180. <https://doi.org/10.1038/nature09944>

Arumugam, M., et al. (2014). Addendum: Enterotypes of the human gut microbiome. *Nature* 506, 516 (2014). <https://doi.org/10.1038/nature13075>

See Also

[mia-datasets](#)

esophagus

Human esophageal community from 3 individuals

Description

This small dataset from a human esophageal community includes 3 samples from 3 human adults based on biopsies analysed with 16S rDNA PCR. The 16S rRNA sequence processing is provided in the mothur wiki from the link below. It was converted into a TreeSummarizedExperiment from the **phyloseq** package.

Usage

```
data(esophagus)
```

Format

A TreeSummarizedExperiment with 58 features and 3 samples. The rowData contains no taxonomic information. The colData is empty.

Author(s)

Pei et al. <zhiheng.pei@med.nyu.edu>.

Source

http://www.mothur.org/wiki/Esophageal_community_analysis

References

Pei, Z., Bini, E. J., Yang, L., Zhou, M., Francois, F., & Blaser, M. J. (2004). Bacterial biota in the human distal esophagus. *Proceedings of the National Academy of Sciences of the United States of America*, 101(12), 4250-4255. <https://doi.org/10.1073/pnas.0306398101>

McMurdie, J. & Holmes, S. (2013) *phyloseq*: An R Package for reproducible interactive analysis and graphics of microbiome census data. *PLoS ONE*. 8(4):e61217. <https://doi.org/10.1371/journal.pone.0061217>

See Also

[mia-datasets](#)

getCrossAssociation *Calculate correlations between features of two experiments.*

Description

Calculate correlations between features of two experiments.

Usage

```
getCrossAssociation(x, ...)  
  
## S4 method for signature 'MultiAssayExperiment'  
getCrossAssociation(  
  x,  
  experiment1 = 1,  
  experiment2 = 2,  
  assay.type1 = assay_name1,  
  assay_name1 = "counts",  
  assay.type2 = assay_name2,  
  assay_name2 = "counts",  
  altexp1 = NULL,  
  altexp2 = NULL,  
  col.var1 = colData_variable1,  
  colData_variable1 = NULL,  
  col.var2 = colData_variable2,  
  colData_variable2 = NULL,  
  by = MARGIN,  
  MARGIN = 1,  
  method = c("kendall", "spearman", "categorical", "pearson"),  
  mode = "table",
```

```

p.adj.method = p_adj_method,
p_adj_method = c("fdr", "BH", "bonferroni", "BY", "hochberg", "holm", "hommel", "none"),
p.adj.threshold = p_adj_threshold,
p_adj_threshold = NULL,
cor.threshold = cor_threshold,
cor_threshold = NULL,
sort = FALSE,
filter.self.cor = filter_self_correlations,
filter_self_correlations = FALSE,
verbose = TRUE,
test.signif = test_significance,
test_significance = FALSE,
show.warnings = show_warnings,
show_warnings = TRUE,
paired = FALSE,
...
)

## S4 method for signature 'SummarizedExperiment'
getCrossAssociation(x, experiment2 = x, ...)

```

Arguments

x	A MultiAssayExperiment or SummarizedExperiment object.
...	Additional arguments: <ul style="list-style-type: none"> • <code>symmetric</code>: Logical scalar. Specifies if measure is symmetric or not. When <code>symmetric = TRUE</code>, associations are calculated only for unique variable-pairs, and they are assigned to corresponding variable-pair. This decreases the number of calculations in 2-fold meaning faster execution. (By default: <code>FALSE</code>) • <code>association.fun</code>: A function that is used to calculate (dis-)similarity between features. Function must take matrix as an input and give numeric values as an output. Adjust method and other parameters correspondingly. Supported functions are, for example, <code>stats::dist</code> and <code>vegan::vegdist</code>.
experiment1	Character scalar or numeric scalar. Selects the experiment 1 from <code>experiments(x)</code> of <code>MultiassayExperiment</code> object. (Default: 1)
experiment2	Character scalar or numeric scalar. Selects the experiment 2 from <code>experiments(x)</code> of <code>MultiAssayExperiment</code> object or <code>altExp(x)</code> of <code>TreeSummarizedExperiment</code> object. Alternatively, <code>experiment2</code> can also be <code>TreeSE</code> object when <code>x</code> is <code>TreeSE</code> object. (Default: 2 when <code>x</code> is MAE and <code>x</code> when <code>x</code> is <code>TreeSE</code>)
assay.type1	Character scalar. Specifies the name of the assay in experiment 1 to be transformed.. (Default: "counts")
assay_name1	Deprecated. Use <code>assay.type1</code> instead.
assay.type2	Character scalar. Specifies the name of the assay in experiment 2 to be transformed.. (Default: "counts")
assay_name2	Deprecated. Use <code>assay.type2</code> instead.

altexp1	Character scalar or numeric scalar. Specifies alternative experiment from the altExp of experiment 1. If NULL, then the experiment is itself and altExp option is disabled. (Default: NULL)
altexp2	Character scalar or numeric scalar. Specifies alternative experiment from the altExp of experiment 2. If NULL, then the experiment is itself and altExp option is disabled. (Default: NULL)
col.var1	Character scalar. Specifies column(s) from colData of experiment 1. If col.var1 is used, assay.type1 is disabled. (Default: NULL)
colData_variable1	Deprecated. Use col.var1 instead.
col.var2	Character scalar. Specifies column(s) from colData of experiment 2. If col.var2 is used, assay.type2 is disabled. (Default: NULL)
colData_variable2	Deprecated. Use col.var2 instead.
by	ACharacter scalar. Determines if association are calculated row-wise / for features ('rows') or column-wise / for samples ('cols'). Must be 'rows' or 'cols'.
MARGIN	Deprecated. Use by instead.
method	Character scalar. Defines the association method ('kendall', 'pearson', or 'spearman' for continuous/numeric; 'categorical' for discrete) (Default: "kendall")
mode	Character scalar. Specifies the output format Available formats are 'table' and 'matrix'. (Default: "table")
p.adj.method	Character scalar. Specifies adjustment method of p-values. Passed to p.adjust function. (Default: "fdr")
p_adj_method	Deprecated. Use p.adj.method instead.
p.adj.threshold	Numeric scalar. From 0 to 1, specifies adjusted p-value threshold for filtering. (Default: NULL)
p_adj_threshold	Deprecated. Use p.dj.threshold instead.
cor.threshold	Numeric scalar. From 0 to 1, specifies correlation threshold for filtering. (Default: NULL)
cor_threshold	Deprecated. Use cor.threshold instead.
sort	Logical scalar. Specifies whether to sort features or not in result matrices. Used method is hierarchical clustering. (Default: FALSE)
filter.self.cor	Logical scalar. Specifies whether to filter out correlations between identical items. Applies only when correlation between experiment itself is tested, i.e., when assays are identical. (Default: FALSE)
filter_self_correlations	Deprecated. Use filter.self.cor instead.
verbose	Logical scalar. Specifies whether to get messages about progress of calculation. (Default: FALSE)

<code>test.signif</code>	Logical scalar. Specifies whether to test statistical significance of associations. (Default: FALSE)
<code>test_significance</code>	Deprecated. Use <code>test.signif</code> instead.
<code>show.warnings</code>	Logical scalar. specifies whether to show warnings that might occur when correlations and p-values are calculated. (Default: FALSE)
<code>show_warnings</code>	Deprecated. use <code>show.warnings</code> instead.
<code>paired</code>	Logical scalar. Specifies if samples are paired or not. <code>colnames</code> must match between two experiments. <code>paired</code> is disabled when <code>by = 1</code> . (Default: FALSE)

Details

The function `getCrossAssociation` calculates associations between features of two experiments. By default, it not only computes associations but also tests their significance. If desired, setting `test.signif` to `FALSE` disables significance calculation.

We recommend the non-parametric Kendall's tau as the default method for association analysis. Kendall's tau has desirable statistical properties and robustness at lower sample sizes. Spearman rank correlation can provide faster solutions when running times are critical.

Value

This function returns associations in table or matrix format. In table format, returned value is a data frame that includes features and associations (and p-values) in columns. In matrix format, returned value is a one matrix when only associations are calculated. If also significances are tested, then returned value is a list of matrices.

Examples

```
data(HintikkaX0Data)
mae <- HintikkaX0Data

# Subset so that less observations / quicker to run, just for example
mae[[1]] <- mae[[1]][1:20, 1:10]
mae[[2]] <- mae[[2]][1:20, 1:10]
# Several rows in the counts assay have a standard deviation of zero
# Remove them, since they do not add useful information about
# cross-association
mae[[1]] <- mae[[1]][rowSds(assay(mae[[1]])) > 0, ]
# Transform data
mae[[1]] <- transformAssay(mae[[1]], method = "rclr")

# Calculate cross-correlations
result <- getCrossAssociation(mae, method = "pearson", assay.type2 = "nmr")
# Show first 5 entries
head(result, 5)

# Use altExp option to specify alternative experiment from the experiment
altExp(mae[[1]], "Phylum") <- agglomerateByRank(mae[[1]], rank = "Phylum")
# Transform data
altExp(mae[[1]], "Phylum") <- transformAssay(
```

```

    altExp(mae[[1]], "Phylum"), method = "relabundance")
# When mode = "matrix", the return value is a matrix
result <- getCrossAssociation(
  mae, experiment2 = 2, assay.type1 = "relabundance", assay.type2 = "nmr",
  altexp1 = "Phylum", method = "pearson", mode = "matrix")
# Show first 5 entries
head(result, 5)

# If test.signif = TRUE, then getCrossAssociation additionally returns
# significances
# filter.self.cor = TRUE filters self correlations
# p.adj.threshold can be used to filter those features that do not
# have any correlations whose p-value is lower than the threshold
result <- getCrossAssociation(
  mae[[1]], experiment2 = mae[[1]], method = "pearson",
  filter.self.cor = TRUE, p.adj.threshold = 0.05, test.signif = TRUE)
# Show first 5 entries
head(result, 5)

# Returned value is a list of matrices
names(result)

# Calculate Bray-Curtis dissimilarity between samples. If dataset includes
# paired samples, you can use paired = TRUE.
result <- getCrossAssociation(
  mae[[1]], mae[[1]], by = 2, paired = FALSE,
  association.fun = getDissimilarity, method = "bray")

# If experiments are equal and measure is symmetric
# (e.g., taxa1 vs taxa2 == taxa2 vs taxa1),
# it is possible to speed-up calculations by calculating association only
# for unique variable-pairs. Use "symmetric" to choose whether to measure
# association for only other half of of variable-pairs.
result <- getCrossAssociation(
  mae, experiment1 = "microbiota", experiment2 = "microbiota",
  assay.type1 = "counts", assay.type2 = "counts", symmetric = TRUE)

# For big data sets, the calculations might take a long time.
# To speed them up, you can take a random sample from the data.
# When dealing with complex biological problems, random samples can be
# enough to describe the data. Here, our random sample is 30 % of whole data.
sample_size <- 0.3
tse <- mae[[1]]
tse_sub <- tse[ sample( seq_len( nrow(tse) ), sample_size * nrow(tse) ), ]
result <- getCrossAssociation(tse_sub)

# It is also possible to choose variables from colData and calculate
# association between assay and sample metadata or between variables of
# sample metadata
mae[[1]] <- addAlpha(mae[[1]])
# col.var works similarly to assay.type. Instead of fetching
# an assay named assay.type from assay slot, it fetches a column named

```

```
# col.var from colData.
result <- getCrossAssociation(
  mae[[1]], assay.type1 = "counts",
  col.var2 = c("shannon_diversity", "coverage_diversity"),
  test.signif = TRUE)
```

getDissimilarity *Calculate dissimilarities*

Description

These functions are designed to calculate dissimilarities on data stored within a [TreeSummarizedExperiment](#) object. For overlap, Unifrac, and Jensen-Shannon Divergence (JSD) dissimilarities, the functions use mia internal functions, while for other types of dissimilarities, they rely on [vegdist](#) by default.

Usage

```
addDissimilarity(x, method, ...)

## S4 method for signature 'SummarizedExperiment'
addDissimilarity(x, method = "bray", name = method, ...)

getDissimilarity(x, method, ...)

## S4 method for signature 'SummarizedExperiment'
getDissimilarity(
  x,
  method = "bray",
  assay.type = "counts",
  niter = NULL,
  transposed = FALSE,
  ...
)

## S4 method for signature 'TreeSummarizedExperiment'
getDissimilarity(
  x,
  method = "bray",
  assay.type = "counts",
  niter = NULL,
  transposed = FALSE,
  ...
)

## S4 method for signature 'ANY'
getDissimilarity(x, method = "bray", niter = NULL, ...)
```


Arguments

x	TreeSummarizedExperiment or matrix.
method	Character scalar. Specifies which dissimilarity to calculate. (Default: "bray")
...	other arguments passed into avgdist , vegdist , or into mia internal functions: <ul style="list-style-type: none"> • <code>sample</code>: The sampling depth in rarefaction. (Default: <code>min(rowSums2(x))</code>) • <code>dis.fun</code>: Character scalar. Specifies the dissimilarity function to be used. • <code>tree.name</code>: (Unifrac) Character scalar. Specifies the name of the tree from <code>rowTree(x)</code> that is used in calculation. Disabled when <code>tree</code> is specified. (Default: "phylo") • <code>tree</code>: (Unifrac) phylo. A phylogenetic tree used in calculation. (Default: NULL) • <code>weighted</code>: (Unifrac) Logical scalar. Should use weighted-Unifrac calculation? Weighted-Unifrac takes into account the relative abundance of species/taxa shared between samples, whereas unweighted-Unifrac only considers presence/absence. Default is FALSE, meaning the unweighted-Unifrac dissimilarity is calculated for all pairs of samples. (Default: FALSE) • <code>node.label</code>: (Unifrac) character vector. Used only if <code>x</code> is a matrix. Specifies links between rows/columns and tips of tree. The length must equal the number of rows/columns of <code>x</code>. Furthermore, all the node labs must be present in tree. • <code>chunkSize</code>: (JSD) Integer scalar. Defines the size of data send to the individual worker. Only has an effect, if <code>BPPARAM</code> defines more than one worker. (Default: <code>nrow(x)</code>) • <code>BPPARAM</code>: (JSD) BiocParallelParam. Specifies whether the calculation should be parallelized. • <code>detection</code>: (Overlap) Numeric scalar. Defines detection threshold for absence/presence of features. Feature that has abundance under threshold in either of samples, will be discarded when evaluating overlap between samples. (Default: 0)
name	Character scalar. The name to be used to store the result in metadata of the output. (Default: <code>method</code>)
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
niter	The number of iterations performed. If NULL, rarefaction is disabled. (Default: NULL)
transposed	Logical scalar. Specifies if <code>x</code> is transposed with cells in rows. (Default: FALSE)

Details

Overlap reflects similarity between sample-pairs. When overlap is calculated using relative abundances, the higher the value the higher the similarity is. When using relative abundances, overlap value 1 means that all the abundances of features are equal between two samples, and 0 means that samples have completely different relative abundances.

Unifrac is calculated with `rbiom:unifrac()`.

If rarefaction is enabled, `vegan:avgdist()` is utilized.

For JSD implementation: Susan Holmes <susan@stat.stanford.edu>. Adapted for phyloseq by Paul J. McMurdie. Adapted for mia by Felix G.M. Ernst

Value

`getDissimilarity` returns a sample-by-sample dissimilarity matrix.

`addDissimilarity` returns `x` that includes dissimilarity matrix in its metadata.

References

For unifrac dissimilarity: <http://bmf.colorado.edu/unifrac/>

See also additional descriptions of Unifrac in the following articles:

Lozupone, Hamady and Knight, "Unifrac - An Online Tool for Comparing Microbial Community Diversity in a Phylogenetic Context.", BMC Bioinformatics 2006, 7:371

Lozupone, Hamady, Kelley and Knight, "Quantitative and qualitative (beta) diversity measures lead to different insights into factors that structure microbial communities." Appl Environ Microbiol. 2007

Lozupone C, Knight R. "Unifrac: a new phylogenetic method for comparing microbial communities." Appl Environ Microbiol. 2005 71 (12):8228-35.

For JSD dissimilarity: Jensen-Shannon Divergence and Hilbert space embedding. Bent Fuglede and Flemming Topsøe University of Copenhagen, Department of Mathematics <http://www.math.ku.dk/~topsoe/ISIT2004JSD.pdf>

See Also

http://en.wikipedia.org/wiki/Jensen-Shannon_divergence

Examples

```
library(mia)
library(scater)

# load dataset
data(GlobalPatterns)
tse <- GlobalPatterns

### Overlap dissimilarity

tse <- addDissimilarity(tse, method = "overlap", detection = 0.25)
metadata(tse)[["overlap"]][1:6, 1:6]

### JSD dissimilarity

tse <- addDissimilarity(tse, method = "jsd")
metadata(tse)[["jsd"]][1:6, 1:6]
```

```

# Multi Dimensional Scaling applied to JSD dissimilarity matrix
tse <- runMDS(tse, FUN = getDissimilarity, method = "overlap",
             assay.type = "counts")
metadata(tse)[["MDS"]][1:6, ]

### Unifrac dissimilarity

res <- getDissimilarity(tse, method = "unifrac", weighted = FALSE)
dim(as.matrix((res)))

tse <- addDissimilarity(tse, method = "unifrac", weighted = TRUE)
metadata(tse)[["unifrac"]][1:6, 1:6]

```

getDominant

Get dominant taxa

Description

These functions return information about the most dominant taxa in a [SummarizedExperiment](#) object.

Usage

```

getDominant(
  x,
  assay.type = assay_name,
  assay_name = "counts",
  group = rank,
  rank = NULL,
  other.name = "Other",
  n = NULL,
  complete = TRUE,
  ...
)

## S4 method for signature 'SummarizedExperiment'
getDominant(
  x,
  assay.type = assay_name,
  assay_name = "counts",
  group = rank,
  rank = NULL,
  other.name = "Other",
  n = NULL,
  complete = TRUE,
  ...
)

```

```

addDominant(x, name = "dominant_taxa", other.name = "Other", n = NULL, ...)

## S4 method for signature 'SummarizedExperiment'
addDominant(
  x,
  name = "dominant_taxa",
  other.name = "Other",
  n = NULL,
  complete = FALSE,
  ...
)

```

Arguments

x	TreeSummarizedExperiment .
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
assay_name	Deprecated. Use assay.type instead.
group	Character scalar. Defines a group. Must be one of the columns from rowData(x). (Default: NULL)
rank	Deprecated. Use group instead.
other.name	Character scalar. A name for features that are not included in n the most frequent dominant features in the data. (Default: "Other")
n	Numeric scalar. The number of features that are the most frequent dominant features. Default is NULL, which defaults that each sample is assigned a dominant taxon. (Default: NULL)
complete	Logical scalar. A value to manage multiple dominant taxa for a sample. Default for getDominant is TRUE to include all equally dominant taxa for each sample. complete = FALSE samples one taxa for the samples that have multiple. Default for addDominant is FALSE to add a column with only one dominant taxon assigned for each sample into colData. complete = TRUE adds a list that includes all dominant taxa for each sample into colData.
...	Additional arguments passed on to agglomerateByRank() when rank is specified.
name	Character scalar. A name for the column of the colData where results will be stored. (Default: "dominant_taxa")

Details

addDominant extracts the most abundant taxa in a [SummarizedExperiment](#) object, and stores the information in the colData. It is a wrapper for getDominant.

With group parameter, it is possible to agglomerate rows based on groups. If the value is one of the columns in taxonomyRanks(), agglomerateByRank() is applied. Otherwise, agglomerateByVariable() is utilized. E.g. if 'Genus' rank is used, all abundances of same Genus are added together, and agglomerated features are returned. See corresponding functions for additional arguments to deal with missing values or special characters.

Value

getDominant returns a named character vector `x` while addDominant returns `SummarizedExperiment` with additional column in `colData` named `*name*`.

Examples

```
data(GlobalPatterns)
x <- GlobalPatterns

# Finds the dominant taxa.
sim.dom <- getDominant(x, group = "Genus")

# Add information to colData
x <- addDominant(x, group = "Genus", name = "dominant_genera")
colData(x)
```

getMediation	<i>Perform mediation analysis</i>
--------------	-----------------------------------

Description

getMediation and addMediation provide a wrapper of `mediate` for `SummarizedExperiment`.

Usage

```
addMediation(x, ...)

## S4 method for signature 'SummarizedExperiment'
addMediation(
  x,
  outcome,
  treatment,
  name = "mediation",
  mediator = NULL,
  assay.type = NULL,
  dimred = NULL,
  family = gaussian(),
  covariates = NULL,
  p.adj.method = "holm",
  add.metadata = FALSE,
  verbose = TRUE,
  ...
)

getMediation(x, ...)

## S4 method for signature 'SummarizedExperiment'
```

```

getMediation(
  x,
  outcome,
  treatment,
  mediator = NULL,
  assay.type = "counts",
  dimred = NULL,
  family = gaussian(),
  covariates = NULL,
  p.adj.method = "holm",
  add.metadata = FALSE,
  verbose = TRUE,
  ...
)

```

Arguments

x	a SummarizedExperiment .
...	additional parameters that can be passed to mediate .
outcome	Character scalar. Indicates the colData variable used as outcome in the model.
treatment	Character scalar. Indicates the colData variable used as treatment in the model.
name	Character scalar. A name for the column of the colData where results will be stored. (Default: "mediation")
mediator	Character scalar. Indicates the colData variable used as mediator in the model. (Default: NULL)
assay.type	Character scalar. Specifies the assay used for feature-wise mediation analysis. (Default: "counts")
dimred	Character scalar. Indicates the reduced dimension result in <code>reducedDims(object)</code> for component-wise mediation analysis. (Default: NULL)
family	Character scalar. A specification for the outcome model link function. (Default: <code>gaussian("identity")</code>)
covariates	Character scalar or character vector. Indicates the colData variables used as covariates in the model. (Default: NULL)
p.adj.method	Character scalar. Selects adjustment method of p-values. Passed to <code>p.adjust</code> function. (Default: "holm")
add.metadata	Logical scalar. Should the model metadata be returned. (Default: FALSE)
verbose	Logical scalar. Should execution messages be printed. (Default: TRUE)

Details

This wrapper of [mediate](#) for [SummarizedExperiment](#) provides a simple method to analyse the effect of a treatment variable on an outcome variable found in `colData(se)` through the mediation of either another variable in `colData` (argument `mediator`) or an assay (argument `assay.type`) or a `reducedDim` (argument `dimred`). Importantly, those three arguments are mutually exclusive.

Value

getMediation returns a data.frame of adjusted p-values and effect sizes for the ACMEs and ADEs of every mediator given as input, whereas addMediation returns an updated [SummarizedExperiment](#) instance with the same data.frame stored in the metadata as "mediation". Its columns include:

Mediator the mediator variable

ACME_estimate the Average Causal Mediation Effect (ACME) estimate

ADE_estimate the Average Direct Effect (ADE) estimate

ACME_pval the adjusted p-value for the ACME estimate

ADE_pval the adjusted p-value for the ADE estimate

Examples

```
## Not run:
# Import libraries
library(mia)
library(scater)

# Load dataset
data(hitchip1006, package = "miaTime")
tse <- hitchip1006

# Agglomerate features by family (merely to speed up execution)
tse <- agglomerateByRank(tse, rank = "Phylum")
# Convert BMI variable to numeric
tse$bmi_group <- as.numeric(tse$bmi_group)

# Analyse mediated effect of nationality on BMI via alpha diversity
# 100 permutations were done to speed up execution, but ~1000 are recommended
med_df <- getMediation(tse,
  outcome = "bmi_group",
  treatment = "nationality",
  mediator = "diversity",
  covariates = c("sex", "age"),
  treat.value = "Scandinavia",
  control.value = "CentralEurope",
  boot = TRUE, sims = 100,
  add.metadata = TRUE)

# Visualise model statistics for 1st mediator
plot(attr(med_df, "metadata")[[1]])

# Apply clr transformation to counts assay
tse <- transformAssay(tse,
  method = "clr",
  pseudocount = 1)

# Analyse mediated effect of nationality on BMI via clr-transformed features
# 100 permutations were done to speed up execution, but ~1000 are recommended
tse <- addMediation(tse, name = "assay_mediation",
```

```

        outcome = "bmi_group",
        treatment = "nationality",
        assay.type = "clr",
        covariates = c("sex", "age"),
        treat.value = "Scandinavia",
        control.value = "CentralEurope",
        boot = TRUE, sims = 100,
        p.adj.method = "fdr")

# Show results for first 5 mediators
head(metadata(tse)$assay_mediation, 5)

# Perform ordination
tse <- runMDS(tse, name = "MDS",
             method = "euclidean",
             assay.type = "clr",
             ncomponents = 3)

# Analyse mediated effect of nationality on BMI via NMDS components
# 100 permutations were done to speed up execution, but ~1000 are recommended
tse <- addMediation(tse, name = "reddim_mediation",
                  outcome = "bmi_group",
                  treatment = "nationality",
                  dimred = "MDS",
                  covariates = c("sex", "age"),
                  treat.value = "Scandinavia",
                  control.value = "CentralEurope",
                  boot = TRUE, sims = 100,
                  p.adj.method = "fdr")

# Show results for first 5 mediators
head(metadata(tse)$reddim_mediation, 5)

## End(Not run)

```

getPERMANOVA

Calculate PERMANOVA (Permutational Multivariate Analysis of Variance)

Description

These functions perform PERMANOVA to assess the significance of group differences based on a specified dissimilarity matrix. The results can be returned directly or added to metadata in an object of class `TreeSummarizedExperiment`.

Usage

```
getPERMANOVA(x, ...)
```



```

addPERMANOVA(x, ...)

## S4 method for signature 'SingleCellExperiment'
getPERMANOVA(x, ...)

## S4 method for signature 'SummarizedExperiment'
getPERMANOVA(x, assay.type = "counts", formula = NULL, col.var = NULL, ...)

## S4 method for signature 'ANY'
getPERMANOVA(x, formula, data, method = "bray", test.homogeneity = TRUE, ...)

## S4 method for signature 'SummarizedExperiment'
addPERMANOVA(x, name = "permanova", ...)

```

Arguments

x	TreeSummarizedExperiment .
...	additional arguments passed to <code>vegan::adonis2</code> . <ul style="list-style-type: none"> • <code>by</code>: Character scalar. Specifies how significance is calculated. (Default: "margin") • <code>na.action</code>: function. Action to take when missing values for any of the variables in formula are encountered. (Default: <code>na.fail</code>) • <code>full</code>: Logical scalar. should all the results from the homogeneity calculations be returned. When FALSE, only summary tables are returned. (Default: FALSE) • <code>homogeneity.test</code>: Character scalar. Specifies the significance test used to analyse vegan::betadisper results. Options include 'permanova' (vegan::permutest), 'anova' (stats::anova) and 'tukeyhsd' (stats::TukeyHSD). (Default: "permanova")
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
formula	formula. If x is a SummarizedExperiment a formula can be supplied. Based on the right-hand side of the given formula <code>colData</code> is subset to <code>col.var</code> . <code>col.var</code> and <code>formula</code> can be missing, which turns the CCA analysis into a CA analysis and dbrDA into PCoA/MDS.
col.var	Character scalar. When x is a SummarizedExperiment , <code>col.var</code> can be used to specify variables from <code>colData</code> .
data	<code>data.frame</code> or coercible to one. The covariance table including covariates defined by formula.
method	Character scalar. A dissimilarity metric used in PERMANOVA and group dispersion calculation. (Default: "bray")
test.homogeneity	Logical scalar. Should the homogeneity of group dispersions be evaluated? (Default: TRUE)
name	Character scalar. A name for the results that will be stored to metadata. (Default: "permanova")

Details

PERMANOVA is a non-parametric method used to test whether the centroids of different groups (as defined by the formula or covariates) are significantly different in terms of multivariate space.

PERMANOVA relies on the assumption of group homogeneity, meaning the groups should be distinct and have similar variances within each group. This assumption is essential as PERMANOVA is sensitive to differences in within-group dispersion, which can otherwise confound results. This is why the functions return homogeneity test results by default.

The functions utilize `vegan::adonis2` to compute PERMANOVA. For homogeneity testing, `vegan::betadisper` along with `vegan::permutest` are utilized by default, which allow testing for equal dispersion across groups and validate the homogeneity assumption.

PERMANOVA and distance-based redundancy analysis (dbRDA) are closely related methods for analyzing multivariate data. PERMANOVA is non-parametric, making fewer assumptions about the data. In contrast, dbRDA assumes a linear relationship when constructing the ordination space, although it also employs a PERMANOVA-like approach to test the significance of predictors within this space. dbRDA offers a broader scope overall, as it provides visualization of the constrained ordination, which can reveal patterns and relationships. However, when the underlying data structure is non-linear, the results from the two methods can differ significantly due to dbRDA's reliance on linear assumptions.

Value

`getPERMANOVA` returns the PERMANOVA results or a list containing the PERMANOVA results and homogeneity test results if `test.homogeneity = TRUE`. `addPERMANOVA` adds these results to metadata of `x`.

See Also

For more details on the actual implementation see `vegan::adonis2`, `vegan::betadisper`, and `vegan::permutest`. See also `addCCA` and `addRDA`

Examples

```
data(GlobalPatterns)
tse <- GlobalPatterns

# Apply relative transformation
tse <- transformAssay(tse, method = "relabundance")
# Perform PERMANOVA
tse <- addPERMANOVA(
  tse,
  assay.type = "relabundance",
  method = "bray",
  formula = x ~ SampleType,
  permutations = 99
)
# The results are stored to metadata
metadata(tse)[["permanova"]]

# Calculate dbRDA
```

```
rda_res <- getRDA(  
  tse, assay.type = "relabundance", method = "bray",  
  formula = x ~ SampleType, permutations = 99)  
# Significance results are similar to PERMANOVA  
attr(rda_res, "significance")
```

getPrevalence

Calculation prevalence information for features across samples

Description

These functions calculate the population prevalence for taxonomic ranks in a [SummarizedExperiment-class](#) object.

Usage

```
getPrevalence(x, ...)  
  
## S4 method for signature 'ANY'  
getPrevalence(  
  x,  
  detection = 0,  
  include.lowest = include_lowest,  
  include_lowest = FALSE,  
  sort = FALSE,  
  na.rm = TRUE,  
  ...  
)  
  
## S4 method for signature 'SummarizedExperiment'  
getPrevalence(  
  x,  
  assay.type = assay_name,  
  assay_name = "counts",  
  rank = NULL,  
  ...  
)  
  
getPrevalent(x, ...)  
  
## S4 method for signature 'ANY'  
getPrevalent(  
  x,  
  prevalence = 50/100,  
  include.lowest = include_lowest,  
  include_lowest = FALSE,
```

```
    ...
  )

## S4 method for signature 'SummarizedExperiment'
getPrevalent(
  x,
  rank = NULL,
  prevalence = 50/100,
  include.lowest = include_lowest,
  include_lowest = FALSE,
  ...
)

getRare(x, ...)

## S4 method for signature 'ANY'
getRare(
  x,
  prevalence = 50/100,
  include.lowest = include_lowest,
  include_lowest = FALSE,
  ...
)

## S4 method for signature 'SummarizedExperiment'
getRare(
  x,
  rank = NULL,
  prevalence = 50/100,
  include.lowest = include_lowest,
  include_lowest = FALSE,
  ...
)

subsetByPrevalent(x, ...)

## S4 method for signature 'SummarizedExperiment'
subsetByPrevalent(x, rank = NULL, ...)

## S4 method for signature 'TreeSummarizedExperiment'
subsetByPrevalent(x, update.tree = FALSE, ...)

subsetByRare(x, ...)

## S4 method for signature 'SummarizedExperiment'
subsetByRare(x, rank = NULL, ...)

## S4 method for signature 'TreeSummarizedExperiment'
```

```

subsetByRare(x, update.tree = FALSE, ...)

getPrevalentAbundance(
  x,
  assay.type = assay_name,
  assay_name = "relabundance",
  ...
)

## S4 method for signature 'ANY'
getPrevalentAbundance(
  x,
  assay.type = assay_name,
  assay_name = "relabundance",
  ...
)

## S4 method for signature 'SummarizedExperiment'
getPrevalentAbundance(x, assay.type = assay_name, assay_name = "counts", ...)

```

Arguments

x	TreeSummarizedExperiment .
...	additional arguments <ul style="list-style-type: none"> • If <code>!is.null(rank)</code> arguments are passed on to agglomerateByRank. See ?agglomerateByRank for more details. • for <code>getPrevalent</code>, <code>getRare</code>, <code>subsetByPrevalent</code> and <code>subsetByRare</code> additional parameters passed to <code>getPrevalence</code> • for <code>getPrevalentAbundance</code> additional parameters passed to <code>getPrevalent</code>
detection	Numeric scalar. Detection threshold for absence/presence. If <code>as_relative = FALSE</code> , it sets the counts threshold for a taxon to be considered present. If <code>as_relative = TRUE</code> , it sets the relative abundance threshold for a taxon to be considered present. (Default: 0)
include.lowest	Logical scalar. Should the lower boundary of the detection and prevalence cutoffs be included? (Default: FALSE)
include_lowest	Deprecated. Use <code>include.lowest</code> instead.
sort	Logical scalar. Should the result be sorted by prevalence? (Default: FALSE)
na.rm	Logical scalar. Should NA values be omitted? (Default: TRUE)
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
assay_name	Deprecated. Use <code>assay.type</code> instead.
rank	Character scalar. Defines a taxonomic rank. Must be a value of <code>taxonomyRanks()</code> function.
prevalence	Prevalence threshold (in 0 to 1). The required prevalence is strictly greater by default. To include the limit, set <code>include.lowest</code> to TRUE.
update.tree	Logical scalar. Should <code>rowTree()</code> also be agglomerated? (Default: FALSE)

Details

getPrevalence calculates the frequency of samples that exceed the detection threshold. For SummarizedExperiment objects, the prevalence is calculated for the selected taxonomic rank, otherwise for the rows. The absolute population prevalence can be obtained by multiplying the prevalence by the number of samples (`ncol(x)`).

The core abundance index from `getPrevalentAbundance` gives the relative proportion of the core species (in between 0 and 1). The core taxa are defined as those that exceed the given population prevalence threshold at the given detection level as set for `getPrevalent`.

`subsetPrevalent` and `subsetRareFeatures` return a subset of `x`. The subset includes the most prevalent or rare taxa that are calculated with `getPrevalent` or `getRare` respectively.

`getPrevalent` returns taxa that are more prevalent with the given detection threshold for the selected taxonomic rank.

`getRare` returns complement of `getPrevalent`.

Value

`subsetPrevalent` and `subsetRareFeatures` return subset of `x`.

All other functions return a named vectors:

- `getPrevalence` returns a numeric vector with the names being set to either the row names of `x` or the names after agglomeration.
- `getPrevalentAbundance` returns a numeric vector with the names corresponding to the column name of `x` and include the joint abundance of prevalent taxa.
- `getPrevalent` and `getRare` return a character vector with only the names exceeding the threshold set by prevalence, if the rownames of `x` is set. Otherwise an integer vector is returned matching the rows in `x`.

References

A Salonen et al. The adult intestinal core microbiota is determined by analysis depth and health status. *Clinical Microbiology and Infection* 18(S4):16 20, 2012. To cite the R package, see `citation('mia')`

See Also

[agglomerateByRank](#), [getTop](#)

Examples

```
data(GlobalPatterns)
tse <- GlobalPatterns
# Get prevalence estimates for individual ASV/OTU
prevalence.frequency <- getPrevalence(tse,
                                     detection = 0,
                                     sort = TRUE)

head(prevalence.frequency)

# Get prevalence estimates for phyla
```

```

# - the getPrevalence function itself always returns population frequencies
prevalence.frequency <- getPrevalence(tse,
                                     rank = "Phylum",
                                     detection = 0,
                                     sort = TRUE)

head(prevalence.frequency)

# - to obtain population counts, multiply frequencies with the sample size,
# which answers the question "In how many samples is this phylum detectable"
prevalence.count <- prevalence.frequency * ncol(tse)
head(prevalence.count)

# Detection threshold 1 (strictly greater by default);
# Note that the data (GlobalPatterns) is here in absolute counts
# (and not compositional, relative abundances)
# Prevalence threshold 50 percent (strictly greater by default)
prevalent <- getPrevalent(
  tse,
  rank = "Phylum",
  detection = 10,
  prevalence = 50/100)
head(prevalent)

# Add relative abundance data
tse <- transformAssay(tse, assay.type = "counts", method = "relabundance")

# Gets a subset of object that includes prevalent taxa
altExp(tse, "prevalent") <- subsetByPrevalent(tse,
                                             rank = "Family",
                                             assay.type = "relabundance",
                                             detection = 0.001,
                                             prevalence = 0.55)

altExp(tse, "prevalent")

# getRare returns the inverse
rare <- getRare(tse,
               rank = "Phylum",
               assay.type = "relabundance",
               detection = 1/100,
               prevalence = 50/100)
head(rare)

# Gets a subset of object that includes rare taxa
altExp(tse, "rare") <- subsetByRare(
  tse,
  rank = "Class",
  assay.type = "relabundance",
  detection = 0.001,
  prevalence = 0.001)
altExp(tse, "rare")

# Names of both experiments, prevalent and rare, can be found from slot
# altExpNames

```

```
tse

data(esophagus)
getPrevalentAbundance(esophagus, assay.type = "counts")
```

GlobalPatterns	<i>Global patterns of 16S rRNA diversity at a depth of millions of sequences per sample.</i>
----------------	--

Description

GlobalPatterns compared the microbial communities from 25 environmental samples and three known "mock communities" at an average depth of 3.1 million reads per sample. Authors reproduced diversity patterns seen in many other published studies, while investigating technical bias by applying the same techniques to simulated microbial communities of known composition. Special thanks are given to J. Gregory Caporaso for providing the OTU-clustered data files for inclusion in the **phyloseq** package, from which this data was converted to TreeSummarizedExperiment.

Usage

```
data(GlobalPatterns)
```

Format

A TreeSummarizedExperiment with 19216 features and 26 samples. The rowData contains taxonomic information at Kingdom, Phylum, Class, Order, Family, Genus and Species levels. The colData includes:

X.SampleID Sample ID taken from the corresponding study

Primer primer used for sequencing

Final_Barcode final barcode (6 nucleotides)

Barcode_truncated_plus_T truncated barcode with an added tyrosine (6 nucleotides)

Barcode_full_length complete barcode with a length of 11 nucleotides

SampleType sampling type by collection site (Soil, Feces, Skin, Tongue, Freshwater, Creek Freshwater, Ocean, Estuary Sediment and Mock)

Description additional information (sampling location, environmental factors and study type)

Author(s)

Caporaso, J. G., et al.

References

Caporaso, J. G., et al. (2011). Global patterns of 16S rRNA diversity at a depth of millions of sequences per sample. PNAS, 108, 4516-4522. <https://doi.org/10.1073/pnas.1000080107>

See Also[mia-datasets](#)

hierarchy-tree	<i>Calculate hierarchy tree</i>
----------------	---------------------------------

Description

These functions generate a hierarchy tree using taxonomic information from a [SummarizedExperiment](#) object and add this hierarchy tree into the rowTree.

Usage

```
getHierarchyTree(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
getHierarchyTree(x, ...)  
  
addHierarchyTree(x, ...)  
  
## S4 method for signature 'SummarizedExperiment'  
addHierarchyTree(x, ...)
```

Arguments

x [TreeSummarizedExperiment](#).
... optional arguments not used currently.

Details

addHierarchyTree calculates a hierarchy tree from the available taxonomic information and add it to rowTree.

getHierarchyTree generates a hierarchy tree from the available taxonomic information. Internally it uses [toTree](#) and [resolveLoop](#) to sanitize data if needed.

Please note that a hierarchy tree is not an actual phylogenetic tree. A phylogenetic tree represents evolutionary relationships among features. On the other hand, a hierarchy tree organizes species into a hierarchical structure based on their taxonomic ranks.

Value

- addHierarchyTree: a [TreeSummarizedExperiment](#) whose phylo tree represents the hierarchy among available taxonomy information.
- getHierarchyTree: a phylo tree representing the hierarchy among available taxonomy information.

Examples

```
# Generate a tree based on taxonomic rank hierarchy (a hierarchy tree).
data(GlobalPatterns)
tse <- GlobalPatterns
getHierarchyTree(tse)

# Add a hierarchy tree to a TreeSummarizedExperiment.
# Please note that any tree already stored in rowTree() will be overwritten.
tse <- addHierarchyTree(tse)
tse
```

HintikkaXOData

Multiomics dataset from 40 rat samples

Description

HintikkaXO is a multiomics dataset from a rat experiment studying effect of fat and prebiotics in diet. It contains high-throughput profiling data from 40 rat samples, including 39 biomarkers, 38 metabolites (NMR), and 12706 OTUs from 318 species, measured from Cecum. This is diet comparison study with High/Low fat diet and xylo-oligosaccharide supplementation. Column metadata is common for all experiments (microbiota, metabolites, biomarkers) and is described below.

Usage

```
data(HintikkaXOData)
```

Format

A MultiAssayExperiment with 3 experiments (microbiota, metabolites and biomarkers). rowData of the microbiota experiment contains taxonomic information at Phylum, Class, Order, Family, Genus, Species and OTU levels. The metabolites and biomarkers experiments contain 38 NMR metabolites and 39 biomarkers, respectively. The colData includes:

Sample Sample ID (character)

Rat Rat ID (factor)

Site Site of measurement ("Cecum"); single value

Diet Diet group (factor; combination of the Fat and XOS fields)

Fat Fat in Diet (factor; Low/High)

XOS XOS Diet Supplement (numeric; 0/1)

Author(s)

Hintikka L et al.

References

Hintikka L et al. (2021): Xylo-oligosaccharides in prevention of hepatic steatosis and adipose tissue inflammation: associating taxonomic and metabolomic patterns in fecal microbiota with bi-clustering. *International Journal of Environmental Research and Public Health* 18(8):4049. <https://doi.org/10.3390/ijerph18084049>

See Also

[mia-datasets](#)

importBIOM	<i>Convert a TreeSummarizedExperiment object to/from 'BIOM' results</i>
------------	---

Description

For convenience, a few functions are available to convert BIOM, DADA2 and phyloseq objects to [TreeSummarizedExperiment](#) objects, and [TreeSummarizedExperiment](#) objects to phyloseq objects.

Usage

```
importBIOM(file, ...)  
  
convertFromBIOM(  
  x,  
  prefix.rm = removeTaxaPrefixes,  
  removeTaxaPrefixes = FALSE,  
  rank.from.prefix = rankFromPrefix,  
  rankFromPrefix = FALSE,  
  artifact.rm = remove.artifacts,  
  remove.artifacts = FALSE,  
  ...  
)  
  
convertToBIOM(x, assay.type = "counts", ...)  
  
## S4 method for signature 'SummarizedExperiment'  
convertToBIOM(x, assay.type = "counts", ...)
```

Arguments

file	BIOM file location
...	Additional arguments. Not used currently.
x	TreeSummarizedExperiment

<code>prefix.rm</code>	Logical scalar. Should taxonomic prefixes be removed? The prefixes is removed only from detected taxa columns meaning that <code>rank.from.prefix</code> should be enabled in the most cases. (Default: FALSE)
<code>removeTaxaPrefixes</code>	Deprecated. Use <code>prefix.rm</code> instead.
<code>rank.from.prefix</code>	Logical scalar. If file does not have taxonomic ranks on feature table, should they be scraped from prefixes? (Default: FALSE)
<code>rankFromPrefix</code>	Deprecated. Use <code>rank.from.prefix</code> instead.
<code>artifact.rm</code>	Logical scalar. If file have some taxonomic character naming artifacts, should they be removed. (default (Default: FALSE)
<code>remove.artifacts</code>	Deprecated. Use <code>artifact.rm</code> instead.
<code>assay.type</code>	Character scalar. The name of assay. (Default: "counts")

Details

`convertFromBIOM` coerces a `biom` object to a `TreeSummarizedExperiment` object.

`convertToBIOM` coerces a `TreeSummarizedExperiment` object to a `biom` object.

`importBIOM` loads a BIOM file and creates a `TreeSummarizedExperiment` object from the BIOM object contained in the loaded file.

Value

`convertFromBIOM` returns an object of class `TreeSummarizedExperiment`

`importBIOM` returns an object of class `TreeSummarizedExperiment`

See Also

[importQIIME2](#) [importMothur](#)

[importMetaPhlan](#) [convertFromPhyloseq](#) [convertFromBIOM](#) [convertFromDADA2](#) [importQIIME2](#)
[importMothur](#) [importHUMANN](#)

Examples

```
# Convert BIOM results to a TreeSE
# Load biom file
library(biomformat)
biom_file <- system.file("extdata", "rich_dense_otu_table.biom",
                        package = "biomformat")

# Make TreeSE from BIOM object
biom_object <- biomformat::read_biom(biom_file)
tse <- convertFromBIOM(biom_object)

# Convert TreeSE object to BIOM
biom <- convertToBIOM(tse)
```

```

# Load biom file
library(biomformat)
biom_file <- system.file(
  "extdata", "rich_dense_otu_table.biom", package = "biomformat")

# Make TreeSE from biom file
tse <- importBIOM(biom_file)

# Get taxonomyRanks from prefixes and remove prefixes
tse <- importBIOM(
  biom_file, rank.from.prefix = TRUE, prefix.rm = TRUE)

# Load another biom file
biom_file <- system.file(
  "extdata", "Aggregated_humanization2.biom", package = "mia")

# Clean artifacts from taxonomic data
tse <- importBIOM(biom_file, artifact.rm = TRUE)

```

importHUMANn

Import HUMANn results to TreeSummarizedExperiment

Description

Import HUMANn results to TreeSummarizedExperiment

Arguments

file	Character scalar. Defines the file path of the HUMANn file. The file must be in merged HUMANn format.
col.data	a DataFrame-like object that includes sample names in rownames, or a single character value defining the file path of the sample metadata file. The file must be in tsv format (Default: NULL).
colData	Deprecated. Use col.data instead.
...	additional arguments: <ul style="list-style-type: none"> • assay.type: Character scalar. Specifies the name of the assay used in calculation. (Default: "counts") • prefix.rm: Logical scalar. Should taxonomic prefixes be removed? (Default: FALSE) • remove.suffix: Logical scalar. Should suffixes of sample names be removed? HUMANn pipeline adds suffixes to sample names. Suffixes are formed from file names. By selecting remove.suffix = TRUE, you can remove pattern from end of sample names that is shared by all. (Default: FALSE)

Details

Import HUMAnN (currently version 3.0 supported) results of functional predictions based on metagenome composition (e.g. pathways or gene families). The input must be in merged HUMAnN format. (See [the HUMAnN documentation and humann_join_tables method.](#))

The function parses gene/pathway information along with taxonomy information from the input file. This information is stored to rowData. Abundances are stored to assays.

Usually the workflow includes also taxonomy data from Metaphlan. See [importMetaPhlAn](#) to load the data to TreeSE.

Value

A [TreeSummarizedExperiment](#) object

References

Beghini F, McIver LJ, Blanco-Míguez A, Dubois L, Asnicar F, Maharjan S, Mailyan A, Manghi P, Scholz M, Thomas AM, Valles-Colomer M, Weingart G, Zhang Y, Zolfo M, Huttenhower C, Franzosa EA, & Segata N (2021) Integrating taxonomic, functional, and strain-level profiling of diverse microbial communities with bioBakery 3. *eLife*. 10:e65088.

See Also

[importMetaPhlAn](#) [convertFromPhyloseq](#) [convertFromBIOM](#) [convertFromDADA2](#) [importQIIME2](#)
[importMothur](#)

Examples

```
# File path
file_path <- system.file("extdata", "humann_output.tsv", package = "mia")
# Import data
tse <- importHUMAN(file_path)
tse
```

importMetaPhlAn

Import Metaphlan results to TreeSummarizedExperiment

Description

Import Metaphlan results to TreeSummarizedExperiment

Arguments

<code>file</code>	NULL or DataFrame-like object. Defines the file path of the Metaphlan file. The file must be in merged Metaphlan format.
<code>col.data</code>	a DataFrame-like object that includes sample names in rownames, or a single character value defining the file path of the sample metadata file. The file must be in tsv format (Default: NULL).
<code>colData</code>	Deprecated. use <code>col.data</code> instead.
<code>sample_meta</code>	Deprecated. Use <code>col.data</code> instead.
<code>tree.file</code>	Character scalar. Defines the file path of the phylogenetic tree. (Default: NULL).
<code>phy_tree</code>	Deprecated. Use <code>tree.file</code> instead.
<code>...</code>	additional arguments: <ul style="list-style-type: none"> • <code>assay.type</code>: Character scalar. Specifies the name of the assay used in calculation. (Default: "counts") • <code>prefix.rm</code>: Logical scalar. Should taxonomic prefixes be removed? (Default: FALSE) • <code>remove.suffix</code>: Logical scalar. Should suffixes of sample names be removed? Metaphlan pipeline adds suffixes to sample names. Suffixes are formed from file names. By selecting <code>remove.suffix = TRUE</code>, you can remove pattern from end of sample names that is shared by all. (Default: FALSE) • <code>set.ranks</code>: Logical scalar. Should the columns in the rowData that are treated as taxonomy ranks be updated according to the ranks found in the imported data? (Default: FALSE)

Details

Import Metaphlan (versions 2, 3 and 4 supported) results. Input must be in merged Metaphlan format. (See [the Metaphlan documentation and merge_metaphlan_tables method](#).) Data is imported so that data at the lowest rank is imported as a `TreeSummarizedExperiment` object. Data at higher rank is imported as a `SummarizedExperiment` objects which are stored to `altExp` of `TreeSummarizedExperiment` object.

Currently Metaphlan versions 2, 3, and 4 are supported.

Value

A `TreeSummarizedExperiment` object

References

Beghini F, McIver LJ, Blanco-Míguez A, Dubois L, Asnicar F, Maharjan S, Mailyan A, Manghi P, Scholz M, Thomas AM, Valles-Colomer M, Weingart G, Zhang Y, Zolfo M, Huttenhower C, Franzosa EA, & Segata N (2021) Integrating taxonomic, functional, and strain-level profiling of diverse microbial communities with bioBakery 3. *eLife*. 10:e65088. doi: 10.7554/eLife.65088

See Also

[convertFromPhyloseq](#) [convertFromBIOM](#) [convertFromDADA2](#) [importQIIME2](#) [importMothur](#)

Examples

```
# (Data is from tutorial
# https://github.com/biobakery/biobakery/wiki/metaphlan3#merge-outputs)

# File path
file_path <- system.file(
  "extdata", "merged_abundance_table.txt", package = "mia")
# Import data
tse <- importMetaPhlan(file_path)
# Data at the lowest rank
tse
# Data at higher rank is stored in altExp
altExps(tse)
# Higher rank data is in SE format, for example, Phylum rank
altExp(tse, "phylum")
```

importMothur

Import Mothur results as a TreeSummarizedExperiment

Description

This method creates a TreeSummarizedExperiment object from Mothur files provided as input.

Usage

```
importMothur(
  assay.file = sharedFile,
  sharedFile,
  taxonomyFile = NULL,
  row.file = taxonomyFile,
  designFile = NULL,
  col.file = designFile
)
```

Arguments

assay.file	Character scalar. Defines the file path of the feature table to be imported. The File has to be in shared file format as defined in Mothur documentation.
sharedFile	Deprecated. Use assay.file instead.
taxonomyFile	Deprecated. Use row.file instead.
row.file	Character scalar. Defines the file path of the taxonomy table to be imported. The File has to be in taxonomy file or constaxonomy file format as defined in Mothur documentation. (Default: NULL).

designFile	Deprecated. Use col.file instead.
col.file	Character scalar. Defines file path of the sample metadata to be imported. The File has to be in design file format as defined in Mothur documentation. (Default: NULL).

Details

Results exported from Mothur can be imported as a SummarizedExperiment using importMothur. Except for the assay.file, the other data types, row.file, and col.file, are optional, but are highly encouraged to be provided.

Value

A `TreeSummarizedExperiment` object

References

<https://mothur.org/> https://mothur.org/wiki/shared_file/ https://mothur.org/wiki/taxonomy_file/ https://mothur.org/wiki/constaxonomy_file/ https://mothur.org/wiki/design_file/

See Also

[convertFromPhyloseq](#) [convertFromBIOM](#) [convertFromDADA2](#) [importQIIME2](#)

Examples

```
# Abundance table
counts <- system.file("extdata", "mothur_example.shared", package = "mia")
# Taxa table (in "cons.taxonomy" or "taxonomy" format)
taxa <- system.file("extdata", "mothur_example.cons.taxonomy", package = "mia")
#taxa <- system.file("extdata", "mothur_example.taxonomy", package = "mia")
# Sample meta data
meta <- system.file("extdata", "mothur_example.design", package = "mia")

# Creates se object from files
se <- importMothur(assay.file = counts, row.file = taxa, col.file = meta)
# Convert SE to TreeSE
tse <- as(se, "TreeSummarizedExperiment")
tse
```

importQIIME2

Import QIIME2 results to TreeSummarizedExperiment

Description

Results exported from QIIME2 can be imported as a TreeSummarizedExperiment using importQIIME2. Except for the assay.file, the other data types, row.file, refseq.file and tree.file, are optional, but are highly encouraged to be provided.

Import the QIIME2 artifacts to R.

Usage

```

importQIIME2(
  assay.file = featureTableFile,
  featureTableFile,
  row.file = taxonomyTableFile,
  taxonomyTableFile = NULL,
  col.file = sampleMetaFile,
  sampleMetaFile = NULL,
  as.refseq = featureNamesAsRefSeq,
  featureNamesAsRefSeq = TRUE,
  refseq.file = refSeqFile,
  refSeqFile = NULL,
  tree.file = phyTreeFile,
  phyTreeFile = NULL,
  ...
)

importQZA(file, temp.dir = temp, temp = tempdir(), ...)

```

Arguments

<code>assay.file</code>	Character scalar. Defines the file path of the feature table to be imported.
<code>featureTableFile</code>	Deprecated. use <code>assay.file</code> instead.
<code>row.file</code>	Character scalar or NULL. Defines the file path of the taxonomy table to be imported. (default: NULL).
<code>taxonomyTableFile</code>	Deprecated. use <code>row.file</code> instead.
<code>col.file</code>	Character scalar or NULL. Defines the file path of the sample metadata to be imported. The file has to be in tsv format. (Default: NULL).
<code>sampleMetaFile</code>	Deprecated. Use <code>col.file</code> instead.
<code>as.refseq</code>	Logical scalar or NULL. Should the feature names of the feature table be regarded as reference sequences? This setting will be disregarded, if <code>refseq.file</code> is not NULL. If the feature names do not contain valid DNA characters only, the reference sequences will not be set.
<code>featureNamesAsRefSeq</code>	Deprecated. Use <code>as.refseq</code> instead.
<code>refseq.file</code>	Character scalar or NULL. Defines the file path of the reference sequences for each feature. (Default: NULL).
<code>refSeqFile</code>	Deprecated. Use <code>refseq.file</code> instead.
<code>tree.file</code>	Character scalar. Defines the file path of the phylogenetic tree. (Default: NULL).
<code>phyTreeFile</code>	Deprecated. Use <code>tree.file</code> instead.
<code>...</code>	additional arguments:

	<ul style="list-style-type: none"> • <code>temp.dir</code>: the temporary directory used for decompressing the data. (default: <code>tempdir()</code>) • <code>prefix.rm</code>: TRUE or FALSE: Should taxonomic prefixes be removed? (default: <code>prefix.rm = FALSE</code>)
<code>file</code>	character, path of the input qza file. Only files in format of <code>BIOMV210DirFmt</code> (feature table), <code>TSVTaxonomyDirectoryFormat</code> (taxonomic table), <code>NewickDirectoryFormat</code> (phylogenetic tree) and <code>DNASequencesDirectoryFormat</code> (representative sequences) are supported right now.
<code>temp.dir</code>	character, a temporary directory in which the qza file will be decompressed to, default <code>tempdir()</code> .
<code>temp</code>	Deprecated. Use <code>temp.dir</code> instead.

Details

Both arguments `as.refseq` and `refseq.file` can be used to define reference sequences of features. `as.refseq` is only taken into account, if `refseq.file` is NULL. No reference sequences are tried to be created, if `featureNameAsRefSeq` is FALSE and `refseq.file` is NULL.

Value

A `TreeSummarizedExperiment` object

matrix object for feature table, `DataFrame` for taxonomic table, `ape::phylo` object for phylogenetic tree, `Biostrings::DNAStringSet` for representative sequences of taxa.

Author(s)

Yang Cao

References

Bolyen E et al. 2019: Reproducible, interactive, scalable and extensible microbiome data science using QIIME 2. *Nature Biotechnology* 37: 852–857. <https://doi.org/10.1038/s41587-019-0209-9>
<https://qiime2.org>

See Also

[convertFromPhyloseq](#) [convertFromBIOM](#) [convertFromDADA2](#) [importMothur](#)

Examples

```
assay.file <- system.file("extdata", "table.qza", package = "mia")
row.file <- system.file("extdata", "taxonomy.qza", package = "mia")
col.file <- system.file("extdata", "sample-metadata.tsv", package = "mia")
tree.file <- system.file("extdata", "tree.qza", package = "mia")
refseq.file <- system.file("extdata", "refseq.qza", package = "mia")
tse <- importQIIME2(
  assay.file = assay.file,
  row.file = row.file,
  col.file = col.file,
```

```

    refseq.file = refseq.file,
    tree.file = tree.file
  )

  tse
  # Read individual files
  assay.file <- system.file("extdata", "table.qza", package = "mia")
  row.file <- system.file("extdata", "taxonomy.qza", package = "mia")
  col.file <- system.file("extdata", "sample-metadata.tsv", package = "mia")

  assay <- importQZA(assay.file)
  rowdata <- importQZA(row.file, prefix.rm = TRUE)
  coldata <- read.table(col.file, header = TRUE, sep = "\t", comment.char = "")

  # Assign rownames
  rownames(coldata) <- coldata[, 1]
  coldata[, 1] <- NULL

  # Order coldata based on assay
  coldata <- coldata[match(colnames(assay), rownames(coldata)), ]

  # Create SE from individual files
  se <- SummarizedExperiment(assays = list(assay), rowData = rowdata, colData = coldata)
  se

```

importTaxpasta	<i>Import</i>	<i>taxpasta-specific</i>	<i>BIOM</i>	<i>results</i>	<i>to</i>
	TreeSummarizedExperiment				

Description

Import taxpasta-specific BIOM results to [TreeSummarizedExperiment](#)

Usage

```
importTaxpasta(file, add.tree = TRUE, ...)
```

Arguments

file	Character scalar. Defines the file path to a BIOM file.
add.tree	Logical scalar. Specifies whether to calculate and add hierarchy tree using addHierarchyTree . (Default: TRUE)
...	additional arguments <ul style="list-style-type: none"> • set.ranks: Logical scalar. Should column names of taxonomy table be treated as taxonomy ranks? (Default: FALSE)

Details

importTaxpasta imports data that is returned from Taxonomic Profile Aggregation and Standardization (taxpasta) pipeline. See more information on taxpasta from [taxpasta documentation](#) .

Value

A [TreeSummarizedExperiment](#) object.

See Also

[importBIOM](#) [convertFromBIOM](#)

Examples

```
## Not run:
# File path to BIOM file
file_path <- system.file("extdata", "complete.biom", package = "mia")
# Import BIOM as TreeSE, and set ranks.
tse <- importTaxpasta(file_path, set.ranks = TRUE)
# Import BIOM as TreeSE without adding hierarchy tree
tse <- importTaxpasta(file_path, add.tree = FALSE)

## End(Not run)
```

isContaminant	<i>decontam functions</i>
---------------	---------------------------

Description

The decontam functions isContaminant and isNotContaminant are made available for [SummarizedExperiment](#) objects.

Usage

```
## S4 method for signature 'SummarizedExperiment'
isContaminant(
  seqtab,
  assay.type = assay_name,
  assay_name = "counts",
  name = "isContaminant",
  concentration = NULL,
  control = NULL,
  batch = NULL,
  threshold = 0.1,
  normalize = TRUE,
  detailed = TRUE,
  ...
)
```

```

)

## S4 method for signature 'SummarizedExperiment'
isNotContaminant(
  seqtab,
  assay.type = assay_name,
  assay_name = "counts",
  name = "isNotContaminant",
  control = NULL,
  threshold = 0.5,
  normalize = TRUE,
  detailed = FALSE,
  ...
)

addContaminantQC(x, name = "isContaminant", ...)

## S4 method for signature 'SummarizedExperiment'
addContaminantQC(x, name = "isContaminant", ...)

addNotContaminantQC(x, name = "isNotContaminant", ...)

## S4 method for signature 'SummarizedExperiment'
addNotContaminantQC(x, name = "isNotContaminant", ...)

```

Arguments

seqtab, x	a SummarizedExperiment
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
assay_name	Deprecated. Use assay.type instead.
name	Character scalar. A name for the column of the colData where results will be stored. (Default: "isContaminant")
concentration	Character scalar or NULL. Defining a column with numeric values from the colData to use as concentration information. (Default: NULL)
control	Character scalar or NULL. Defining a column with logical values from the colData to define control and non-control samples. (Default: NULL)
batch	Character scalar or NULL. Defining a column with values interpretable as a factor from the colData to use as batch information. (Default: NULL)
threshold	Numeric scalar.. See decontam:isContaminant or decontam:isNotContaminant
normalize	Logical scalar. See decontam:isContaminant or decontam:isNotContaminant
detailed	Logical scalar. If TRUE, the return value is a data.frame containing diagnostic information on the contaminant decision. If FALSE, the return value is a logical vector containing the binary contaminant classifications. (Default: TRUE)
...	<ul style="list-style-type: none"> for isContaminant/ isNotContaminant: arguments passed on to decontam:isContaminant or decontam:isNotContaminant

- for addContaminantQC/addNotContaminantQC: arguments passed on to isContaminant/ isNotContaminant

Value

for isContaminant/ isNotContaminant a DataFrame or for addContaminantQC/addNotContaminantQC a modified object of class(x)

See Also

[decontam:isContaminant](#), [decontam:isNotContaminant](#)

Examples

```
data(esophagus)
# setup of some mock data
colData(esophagus)$concentration <- c(1,2,3)
colData(esophagus)$control <- c(FALSE,FALSE,TRUE)

isContaminant(esophagus,
              method = "frequency",
              concentration = "concentration")
esophagus <- addContaminantQC(esophagus,
                             method = "frequency",
                             concentration = "concentration")

colData(esophagus)

isNotContaminant(esophagus, control = "control")
esophagus <- addNotContaminantQC(esophagus, control = "control")
colData(esophagus)
```

meltSE

Converting a [SummarizedExperiment](#) object into a long data.frame

Description

meltSE Converts a [SummarizedExperiment](#) object into a long data.frame which can be used for tidyverse-tools.

Usage

```
meltSE(x, ...)
```

```
## S4 method for signature 'SummarizedExperiment'
meltSE(
  x,
  assay.type = assay_name,
  assay_name = "counts",
  add.row = add_row_data,
```

```

    add_row_data = NULL,
    add_col = add_col_data,
    add_col_data = NULL,
    row.name = feature_name,
    feature_name = "FeatureID",
    col.name = sample_name,
    sample_name = "SampleID",
    ...
  )

```

Arguments

x	TreeSummarizedExperiment .
...	optional arguments: <ul style="list-style-type: none"> • <code>check.names</code>: Logical scalar. Passed to <code>data.frame</code> function's <code>check.name</code> argument. Determines if sample names are checked that they are syntactically valid variable names and are not duplicated. If they are not, sample names are modified. (Default: TRUE)
<code>assay.type</code>	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
<code>assay_name</code>	Deprecated. Use <code>assay.type</code> instead.
<code>add.row</code>	Logical scalar or Character vector. To select information from the <code>rowData</code> to add to the molten assay data. If <code>add.row = NULL</code> no data will be added, if <code>add.row = TRUE</code> all data will be added and if <code>add.row</code> is a character vector, it will be used to subset to given column names in <code>rowData</code> . (Default: NULL)
<code>add_row_data</code>	Deprecated. Use <code>add.row</code> instead.
<code>add.col</code>	Logical scalar. NULL, or character vector. Used to select information from the <code>colData</code> to add to the molten assay data. If <code>add.col = NULL</code> no data will be added, if <code>add.col = TRUE</code> all data will be added and if <code>add.col</code> is a character vector, it will be used to subset to given column names in <code>colData</code> . (Default: NULL)
<code>add_col_data</code>	Deprecated. Use <code>add.col</code> instead.
<code>row.name</code>	Character scalar. To use as the output's name for the feature identifier. (Default: "FeatureID")
<code>feature_name</code>	Deprecated. Use <code>row.name</code> instead.
<code>col.name</code>	Character scalar. To use as the output's name for the sample identifier. (Default: "SampleID")
<code>sample_name</code>	Deprecated. Use <code>col.name</code> instead.

Details

If the `colData` contains a column "SampleID" or the `rowData` contains a column "FeatureID", they will be renamed to "SampleID_col" and "FeatureID_row", if row names or column names are set.

Value

A tibble with the molten data. The assay values are given in a column named like the selected assay `assay.type`. In addition, a column “FeatureID” will contain the rownames, if set, and analogously a column “SampleID” with the colnames, if set.

Examples

```
data(GlobalPatterns)
molten_tse <- meltSE(
  GlobalPatterns,
  assay.type = "counts",
  add.row = TRUE,
  add.col = TRUE
)
molten_tse
```

mergeSEs

Merge SE objects into single SE object.

Description

Merge SE objects into single SE object.

Usage

```
mergeSEs(x, ...)
```

```
## S4 method for signature 'SimpleList'
mergeSEs(
  x,
  assay.type = "counts",
  assay_name = NULL,
  join = "full",
  missing.values = missing_values,
  missing_values = NA,
  collapse.cols = collapse_samples,
  collapse_samples = FALSE,
  collapse.rows = collapse_features,
  collapse_features = TRUE,
  verbose = TRUE,
  ...
)
```

```
## S4 method for signature 'SummarizedExperiment'
mergeSEs(x, y = NULL, ...)
```

```
## S4 method for signature 'list'
mergeSEs(x, ...)
```

Arguments

x	TreeSummarizedExperiment .
...	optional arguments (not used).
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
assay_name	Deprecated. Use assay.type instead.
join	Character scalar. A value for selecting the joining method. Must be 'full', 'inner', 'left', or 'right'. 'left' and 'right' are disabled when more than two objects are being merged. (Default: "full")
missing.values	NA, \emptyset or Character scalar. Specifies the notation of missing values. (By default: NA)
missing_values	Deprecated. Use missing.values instead.
collapse.cols	Logical scalar. Determines whether to collapse identically named samples to one. (Default: FALSE)
collapse_samples	Deprecated. Use collapse.cols instead.
collapse.rows	Logical scalar. Selects whether to collapse identically named features to one. Since all taxonomy information is taken into account, this concerns rownames-level (usually strain level) comparison. Often OTU or ASV level is just an arbitrary number series from sequencing machine meaning that the OTU information is not comparable between studies. With this option, it is possible to specify whether these strains are combined if their taxonomy information along with OTU number matches. (Default: TRUE)
collapse_features	Deprecated. Use collapse.rows instead.
verbose	Logical scalar. Choose whether to show messages. (Default: TRUE)
y	a SummarizedExperiment object when x is a SummarizedExperiment object. Disabled when x is a list.

Details

This function merges multiple `SummarizedExperiment` objects. It combines `rowData`, `assays`, and `colData` so that the output includes each unique row and column ones. The merging is done based on `rownames` and `colnames`. `rowTree` and `colTree` are preserved if linkage between rows/cols and the tree is found.

Equally named rows are interpreted as equal. Further matching based on `rowData` is not done. For samples, collapsing is disabled by default meaning that equally named samples that are stored in different objects are interpreted as unique. Collapsing can be enabled with `collapse.cols = TRUE` when equally named samples describe the same sample.

If, for example, all rows are not shared with individual objects, there are missing values in assays. The notation of missing can be specified with the `missing.values` argument. If input consists of `TreeSummarizedExperiment` objects, also `rowTree`, `colTree`, and `referenceSeq` are preserved if possible. The data is preserved if all the rows or columns can be found from it.

Compared to `cbind` and `rbind` `mergeSEs` allows more freely merging since `cbind` and `rbind` expect that rows and columns are matching, respectively.

You can choose joining methods from `'full'`, `'inner'`, `'left'`, and `'right'`. In all the methods, all the samples are included in the result object. However, with different methods, it is possible to choose which rows are included.

- `full` – all unique features
- `inner` – all shared features
- `left` – all the features of the first object
- `right` – all the features of the second object

The output depends on the input. If the input contains `SummarizedExperiment` object, then the output will be `SummarizedExperiment`. When all the input objects belong to `TreeSummarizedExperiment`, the output will be `TreeSummarizedExperiment`.

Value

A single `SummarizedExperiment` object.

See Also

- `TreeSummarizedExperiment::cbind`
- `TreeSummarizedExperiment::rbind`
- [full_join](#)
- [inner_join](#)
- [left_join](#)
- [right_join](#)

Examples

```
data(GlobalPatterns)
data(esophagus)
data(enterotype)

# Take only subsets so that it wont take so long
tse1 <- GlobalPatterns[1:100, ]
tse2 <- esophagus
tse3 <- enterotype[1:100, ]

# Merge two TreeSEs
tse <- mergeSEs(tse1, tse2)

# Merge a list of TreeSEs
list <- SimpleList(tse1, tse2, tse3)
tse <- mergeSEs(list, assay.type = "counts", missing.values = 0)
tse

# With 'join', it is possible to specify the merging method. Subsets are used
```

```

# here just to show the functionality
tse_temp <- mergeSEs(tse[1:10, 1:10], tse[5:100, 11:20], join = "left")
tse_temp

# If your objects contain samples that describe one and same sample,
# you can collapse equally named samples to one by specifying 'collapse.cols'
tse_temp <- mergeSEs(list(tse[1:10, 1], tse[1:20, 1], tse[1:5, 1]),
                     collapse.cols = TRUE,
                     join = "inner")
tse_temp

# Merge all available assays
tse <- transformAssay(tse, method="relabundance")
tse1 <- transformAssay(tse1, method="relabundance")
tse_temp <- mergeSEs(tse, tse1, assay.type = assayNames(tse))

```

mia-datasets

mia datasets

Description

mia provides various datasets derived from independent experimental studies. The datasets represent instances of the `TreeSummarizedExperiment` and `MultiAssayExperiment` containers and can serve as tools to practice the *mia* functionality.

Details

Currently, the following datasets are available:

- [dmn_se](#): A `SummarizedExperiment` with 130 features and 278 samples
- [enterotype](#): A `TreeSummarizedExperiment` with 553 features and 280 samples
- [esophagus](#): A `TreeSummarizedExperiment` with 58 features and 3 samples
- [GlobalPatterns](#): A `TreeSummarizedExperiment` with 19216 features and 26 samples
- [HintikkaX0Data](#): A `MultiAssayExperiment` with 3 experiments (microbiota, metabolites and biomarkers)
- [peerj13075](#): A `TreeSummarizedExperiment` with 674 features and 58 samples
- [Tengeler2020](#): A `TreeSummarizedExperiment` with 151 features and 27 samples

Examples

```

# Load dataset from mia
library(mia)
data("GlobalPatterns", package = "mia")

# In this case, the dataset is a TreeSE, so it is renamed as tse
tse <- GlobalPatterns

# Print summary
tse

```

Description

peerj13075 includes skin microbial profiles of 58 volunteers with multiple factors. 16S r-RNA sequencing of V3-V4 regions was done to generate millions of read using illumina platform. A standard bioinformatic and statistical analysis done to explore skin bacterial diversity and its association with age, diet, geographical locations. The authors investigated significant association of skin microbiota with individual's geographical location.

Usage

```
data(peerj13075)
```

Format

A TreeSummarizedExperiment with 674 features and 58 samples. The rowData contains taxonomic information at kingdom, phylum, class, order, family and genus level. The colData includes:

Sample sample ID

Geographical_location city where participant lives (Ahmednagar, Pune and Nashik)

Gender participant's gender (Male or Female)

Age participant's age group (Middle_age, Adult and Elderly)

Diet participant's diet (Veg or Mixed)

Author(s)

Potbhare, R., et al.

References

Potbhare, R., RaviKumar, A., Munukka, E., Lahti, L., & Ashma, R. (2022). Skin microbiota diversity among genetically unrelated individuals of Indian origin. PeerJ, 10, e13075. <https://doi.org/10.7717/peerj.13075> Supplemental information includes OTU table and taxonomy table publicly-accessible from: <https://www.doi.org/10.7717/peerj.13075/supp-1> <https://www.doi.org/10.7717/peerj.13075/supp-2>

See Also

[mia-datasets](#)

rarefyAssay

Subsample Counts

Description

rarefyAssay randomly subsamples counts within a SummarizedExperiment object and returns a new SummarizedExperiment containing the original assay and the new subsampled assay.

Usage

```

rarefyAssay(
  x,
  assay.type = assay_name,
  assay_name = "counts",
  sample = min_size,
  min_size = min(colSums2(assay(x))),
  replace = TRUE,
  name = "subsamped",
  verbose = TRUE,
  ...
)

## S4 method for signature 'SummarizedExperiment'
rarefyAssay(
  x,
  assay.type = assay_name,
  assay_name = "counts",
  sample = min_size,
  min_size = min(colSums2(assay(x, assay.type))),
  replace = TRUE,
  name = "subsamped",
  verbose = TRUE,
  ...
)

```

Arguments

x	TreeSummarizedExperiment .
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
assay_name	Deprecated. Use assay.type instead.
sample	Integer scalar. Indicates the number of counts being simulated i.e. rarefying depth. This can equal to lowest number of total counts found in a sample or a user specified number.
min_size	Deprecated. Use sample instead.

replace	Logical scalar. The default is with replacement (replace=TRUE). See phyloseq::rarefy_even_depth for details on implications of this parameter. (Default: TRUE)
name	Character scalar. A name for the column of the colData where results will be stored. (Default: "method")
verbose	Logical scalar. Choose whether to show messages. (Default: TRUE)
...	additional arguments not used

Details

Although the subsampling approach is highly debated in microbiome research, we include the `rarefyAssay` function because there may be some instances where it can be useful. Note that the output of `rarefyAssay` is not the equivalent as the input and any result have to be verified with the original dataset.

Subsampling/Rarefying may undermine downstream analyses and have unintended consequences. Therefore, make sure this normalization is appropriate for your data.

To maintain the reproducibility, please define the seed using `set.seed()` before implement this function.

Value

`rarefyAssay` return x with subsampled data.

References

McMurdie PJ, Holmes S. Waste not, want not: why rarefying microbiome data is inadmissible. *PLoS computational biology*. 2014 Apr 3;10(4):e1003531.

Gloor GB, Macklaim JM, Pawlowsky-Glahn V & Egozcue JJ (2017) Microbiome Datasets Are Compositional: And This Is Not Optional. *Frontiers in Microbiology* 8: 2224. doi: 10.3389/fmicb.2017.02224

Weiss S, Xu ZZ, Peddada S, Amir A, Bittinger K, Gonzalez A, Lozupone C, Zaneveld JR, Vázquez-Baeza Y, Birmingham A, Hyde ER. Normalization and microbial differential abundance strategies depend upon data characteristics. *Microbiome*. 2017 Dec;5(1):1-8.

Examples

```
# When samples in TreeSE are less than specified sample, they will be removed.
# If after subsampling features are not present in any of the samples,
# they will be removed.
data(GlobalPatterns)
tse <- GlobalPatterns
set.seed(123)
tse_subsampled <- rarefyAssay(tse, sample = 60000, name = "subsampled")
tse_subsampled
dim(tse)
dim(assay(tse_subsampled, "subsampled"))
```

runCCA

Canonical Correspondence Analysis and Redundancy Analysis

Description

These functions perform Canonical Correspondence Analysis on data stored in a SummarizedExperiment.

Usage

```

getCCA(x, ...)

addCCA(x, ...)

getRDA(x, ...)

addRDA(x, ...)

calculateCCA(x, ...)

runCCA(x, ...)

## S4 method for signature 'ANY'
getCCA(x, formula, data, ...)

## S4 method for signature 'SummarizedExperiment'
getCCA(
  x,
  formula = NULL,
  col.var = variables,
  variables = NULL,
  test.signif = TRUE,
  assay.type = assay_name,
  assay_name = exprs_values,
  exprs_values = "counts",
  ...
)

## S4 method for signature 'SingleCellExperiment'
addCCA(x, altexp = NULL, name = "CCA", ...)

calculateRDA(x, ...)

runRDA(x, ...)

## S4 method for signature 'ANY'
getRDA(x, formula, data, ...)

```



```
## S4 method for signature 'SummarizedExperiment'
getRDA(
  x,
  formula = NULL,
  col.var = variables,
  variables = NULL,
  test.signif = TRUE,
  assay.type = assay_name,
  assay_name = exprs_values,
  exprs_values = "counts",
  ...
)

## S4 method for signature 'SingleCellExperiment'
addRDA(x, altexp = NULL, name = "RDA", ...)
```

Arguments

x	TreeSummarizedExperiment .
...	additional arguments passed to <code>vegan::cca</code> or <code>vegan::dbrda</code> and other internal functions. <ul style="list-style-type: none"> • method a dissimilarity measure to be applied in dbrDA and possible following homogeneity test. (Default: "euclidean") • scale: Logical scalar. Should the expression values be standardized? scale is disabled when using *RDA functions. Please scale before performing RDA. (Default: TRUE) • na.action: function. Action to take when missing values for any of the variables in formula are encountered. (Default: na.fail) • full Logical scalar. Should all the results from the significance calculations be returned. When FALSE, only summary tables are returned. (Default: FALSE) • homogeneity.test: Character scalar. Specifies the significance test used to analyse <code>vegan::betadisper</code> results. Options include 'permanova' (<code>vegan::permutest</code>), 'anova' (<code>stats::anova</code>) and 'tukeyhsd' (<code>stats::TukeyHSD</code>). (Default: "permanova") • permutations a numeric value specifying the number of permutations for significance testing in <code>vegan::anova.cca</code>. (Default: 999)
formula	formula. If x is a SummarizedExperiment a formula can be supplied. Based on the right-hand side of the given formula colData is subset to col.var. col.var and formula can be missing, which turns the CCA analysis into a CA analysis and dbrDA into PCoA/MDS.
data	data.frame or coercible to one. The covariance table including covariates defined by formula.
col.var	Character scalar. When x is a SummarizedExperiment , col.var can be used to specify variables from colData.
variables	Deprecated. Use col.var instead.

<code>test.signif</code>	Logical scalar. Should the PERMANOVA and analysis of multivariate homogeneity of group dispersions be performed. (Default: TRUE)
<code>assay.type</code>	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
<code>assay_name</code>	Deprecated. Use <code>assay.type</code> instead.
<code>exprs_values</code>	Deprecated. Use <code>assay.type</code> instead.
<code>altexp</code>	Character scalar or integer scalar. Specifies an alternative experiment containing the input data.
<code>name</code>	Character scalar. A name for the <code>reducedDim()</code> where results will be stored. (Default: "CCA")

Details

*CCA functions utilize `vegan::cca` and *RDA functions `vegan::dbRDA`. By default, `dbRDA` is done with euclidean distances, which is equivalent to RDA. `col.var` and `formula` can be missing, which turns the CCA analysis into a CA analysis and `dbRDA` into PCoA/MDS.

Significance tests are done with `vegan::anova.cca` (PERMANOVA). Group dispersion, i.e., homogeneity within groups is analyzed with `vegan::betadisper` (multivariate homogeneity of groups dispersions (variances)) and statistical significance of homogeneity is tested with a test specified by `homogeneity.test` parameter.

Value

For `getCCA` a matrix with samples as rows and CCA dimensions as columns. Attributes include output from `scores`, eigenvalues, the `cca/rda` object and significance analysis results.

For `addCCA` a modified `x` with the results stored in `reducedDim` as the given name.

See Also

For more details on the actual implementation see [cca](#) and [dbrda](#)

Examples

```
library(miaViz)
data("enterotype", package = "mia")
tse <- enterotype

# Perform CCA and exclude any sample with missing ClinicalStatus
tse <- addCCA(
  tse,
  formula = data ~ ClinicalStatus,
  na.action = na.exclude
)

# Plot CCA
plotCCA(tse, "CCA", colour_by = "ClinicalStatus")

# Fetch significance results
attr(reducedDim(tse, "CCA"), "significance")
```

```
tse <- transformAssay(tse, method = "relabundance")

# Specify dissimilarity measure
tse <- addRDA(
  tse,
  formula = data ~ ClinicalStatus,
  assay.type = "relabundance",
  method = "bray",
  name = "RDA_bray",
  na.action = na.exclude
)

# To scale values when using *RDA functions, use
# transformAssay(MARGIN = "features", ...)
tse <- transformAssay(tse, method = "standardize", MARGIN = "features")

# Data might include taxa that do not vary. Remove those because after
# z-transform their value is NA
tse <- tse[rowSums(is.na(assay(tse, "standardize"))) == 0, ]

# Calculate RDA
tse <- addRDA(
  tse,
  formula = data ~ ClinicalStatus,
  assay.type = "standardize",
  name = "rda_scaled",
  na.action = na.omit
)

# Plot RDA
plotRDA(tse, "rda_scaled", colour_by = "ClinicalStatus")

# A common choice along with PERMANOVA is ANOVA when statistical significance
# of homogeneity of groups is analysed. Moreover, full significance test
# results can be returned.
tse <- addRDA(
  tse,
  formula = data ~ ClinicalStatus,
  homogeneity.test = "anova",
  full = TRUE
)

# Example showing how to pass extra parameters, such as 'permutations',
# to anova.cca
tse <- addRDA(
  tse,
  formula = data ~ ClinicalStatus,
  permutations = 500
)
```

runDPCoA

Calculation of Double Principal Correspondance analysis

Description

Double Principal Correspondance analysis is made available via the `ade4` package in typical fashion. Results are stored in the `reducedDims` and are available for all the expected functions.

Usage

```
getDPCoA(x, y, ...)

## S4 method for signature 'ANY,ANY'
getDPCoA(
  x,
  y,
  ncomponents = 2,
  ntop = NULL,
  subset.row = subset_row,
  subset_row = NULL,
  scale = FALSE,
  transposed = FALSE,
  ...
)

## S4 method for signature 'TreeSummarizedExperiment,missing'
getDPCoA(
  x,
  ...,
  assay.type = assay_name,
  assay_name = exprs_values,
  exprs_values = "counts",
  tree.name = tree_name,
  tree_name = "phylo"
)

calculateDPCoA(x, ...)

addDPCoA(x, ..., altexp = NULL, name = "DPCoA")

runDPCoA(x, ...)
```

Arguments

`x` [TreeSummarizedExperiment](#).

`y` a dist or a symmetric matrix compatible with `ade4:dpcoa`

...	Currently not used.
ncomponents	Numeric scalar. Indicates the number of DPCoA dimensions to obtain. (Default: 2)
ntop	Numeric scalar. Specifies the number of features with the highest variances to use for dimensionality reduction. Alternatively NULL, if all features should be used. (Default: NULL)
subset.row	Character Vector. Specifies the subset of features to use for dimensionality reduction. This can be a character vector of row names, an integer vector of row indices or a logical vector. (Default: NULL)
subset_row	Deprecated. Use subset.row instead.
scale	Logical scalar. Should the expression values be standardized? (Default: FALSE)
transposed	Logical scalar. Specifies if x is transposed with cells in rows. (Default: FALSE)
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
assay_name	Deprecated. Use assay.type instead.
exprs_values	Deprecated. Use assay.type instead.
tree.name	Character scalar. Specifies the name of the tree to be included in the phyloseq object that is created, (Default: "phylo")
tree_name	Deprecated. Use tree.name instead.
altexp	Character scalar or integer scalar. Specifies an alternative experiment containing the input data. (Default: NULL)
name	Character scalar. A name for the column of the colData where results will be stored. (Default: "DPCoA")

Details

For addDPCoA a [TreeSummarizedExperiment](#) containing the expression values as well as a rowTree to calculate y using [cophenetic.phylo](#).

In addition to the reduced dimension on the features, the reduced dimension for samples are returned as well as sample_red attribute. eig, feature_weights and sample_weights are returned as attributes as well.

Value

For getDPCoA a matrix with samples as rows and CCA dimensions as columns

For addDPCoA a modified x with the results stored in reducedDim as the given name

See Also

[plotReducedDim](#) [reducedDims](#)

Examples

```
data(esophagus)
dpcoa <- getDPCoA(esophagus)
head(dpcoa)

esophagus <- addDPCoA(esophagus)
reducedDims(esophagus)

library(scater)
plotReducedDim(esophagus, "DPCoA")
```

runNMDS

Perform non-metric MDS on sample-level data

Description

Perform non-metric multi-dimensional scaling (nMDS) on samples, based on the data in a `SingleCellExperiment` object.

Usage

```
getNMDS(x, ...)

## S4 method for signature 'ANY'
getNMDS(
  x,
  FUN = vegdist,
  nmDS.fun = nmDSFUN,
  nmDSFUN = c("isoMDS", "monoMDS"),
  ncomponents = 2,
  ntop = 500,
  subset.row = subset_row,
  subset_col = NULL,
  scale = FALSE,
  transposed = FALSE,
  keep.dist = keep_dist,
  keep_col = FALSE,
  ...
)

## S4 method for signature 'SummarizedExperiment'
getNMDS(
  x,
  ...,
  assay.type = assay_name,
  assay_name = exprs_values,
  exprs_values = "counts",
```

```

    FUN = vegdist
  )

  ## S4 method for signature 'SingleCellExperiment'
  getNMDS(
    x,
    ...,
    assay.type = assay_name,
    assay_name = exprs_values,
    exprs_values = "counts",
    dimred = NULL,
    ndimred = n_dimred,
    n_dimred = NULL,
    FUN = vegdist
  )

  calculateNMDS(x, ...)

  addNMDS(x, ..., altexp = NULL, name = "NMDS")

  runNMDS(x, ...)

```

Arguments

x	TreeSummarizedExperiment .
...	additional arguments to pass to FUN and nmds.fun.
FUN	Function or Character scalar. A value with a function name returning a dist object
nmds.fun	Character scalar. A value to choose the scaling implementation, either “isoMDS” for MASS::isoMDS or “monoMDS” for vegan::monoMDS
nmdsFUN	Deprecated. Use nmds.fun instead.
ncomponents	Numeric scalar. Indicates the number of DPCoA dimensions to obtain. (Default: 2)
ntop	Numeric scalar. Specifies the number of features with the highest variances to use for dimensionality reduction. Alternatively NULL, if all features should be used. (Default: NULL)
subset.row	Character Vector. Specifies the subset of features to use for dimensionality reduction. This can be a character vector of row names, an integer vector of row indices or a logical vector. (Default: NULL)
subset_row	Deprecated. Use subset.row instead.
scale	Logical scalar. Should the expression values be standardized? (Default: FALSE)
transposed	Logical scalar. Specifies if x is transposed with cells in rows. (Default: FALSE)

keep.dist	Logical scalar. Indicates whether the dist object calculated by FUN should be stored as 'dist' attribute of the matrix returned/stored by getNMDS/ addNMDS. (Default: FALSE)
keep_dist	Deprecated. Use keep.dist instead.
assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
assay_name	Deprecated. Use assay.type instead.
exprs_values	Deprecated. Use assay.type instead.
dimred	Character scalar or integer scalar. Specifies the existing dimensionality reduction results to use.
ndimred	integer vector. Specifies the dimensions to use if dimred is specified.
n_dimred	Deprecated. Use ndimred instead.
altexp	Character scalar or integer scalar. Specifies an alternative experiment containing the input data. (Default: NULL)
name	Character scalar. A name for the column of the colData where results will be stored. (Default: "NMDS")

Details

For addNMDS a [SingleCellExperiment](#)

Either `MASS::isoMDS` or `vegan::monoMDS` are used internally to compute the NMDS components. If you supply a custom FUN, make sure that the arguments of FUN and `nmds.fun` do not collide.

Value

For getNMDS, a matrix is returned containing the MDS coordinates for each sample (row) and dimension (column).

See Also

[MASS::isoMDS](#), [vegan::monoMDS](#) for NMDS component calculation.
[plotMDS](#), to quickly visualize the results.

Examples

```
# generate some example data
mat <- matrix(1:60, nrow = 6)
df <- DataFrame(n = c(1:6))
tse <- TreeSummarizedExperiment(assays = list(counts = mat),
                               rowData = df)

#
getNMDS(tse)

#
data(esophagus)
esophagus <- addNMDS(esophagus, FUN = vegan::vegdist, name = "BC")
esophagus <- addNMDS(esophagus, FUN = vegan::vegdist, name = "euclidean",
```



```

                                method = "euclidean")
reducedDims(esophagus)

```

splitOn	<i>Split TreeSummarizedExperiment column-wise or row-wise based on grouping variable</i>
---------	--

Description

Split TreeSummarizedExperiment column-wise or row-wise based on grouping variable

Usage

```

splitOn(x, ...)

## S4 method for signature 'SummarizedExperiment'
splitOn(x, group = f, f = NULL, ...)

## S4 method for signature 'SingleCellExperiment'
splitOn(x, group = f, f = NULL, ...)

## S4 method for signature 'TreeSummarizedExperiment'
splitOn(
  x,
  group = f,
  f = NULL,
  update.tree = update_rowTree,
  update_rowTree = FALSE,
  ...
)

unsplitOn(x, ...)

## S4 method for signature 'list'
unsplitOn(x, update.tree = update_rowTree, update_rowTree = FALSE, ...)

## S4 method for signature 'SimpleList'
unsplitOn(x, update.tree = update_rowTree, update_rowTree = FALSE, ...)

## S4 method for signature 'SingleCellExperiment'
unsplitOn(
  x,
  altexp = altExpNames,
  altExpNames = names(altExps(x)),
  keep.dimred = keep_reducedDims,
  keep_reducedDims = FALSE,
  ...
)

```

Arguments

x	TreeSummarizedExperiment .
...	Arguments passed to agglomerateByVariable function for SummarizedExperiment objects and other functions. See agglomerateByVariable for more details. <ul style="list-style-type: none"> • <code>use.names</code>: Logical scalar. Specifies whether to name elements of list by their group names. (Default: TRUE)
group	Character scalar, character vector or factor vector. A column name from <code>rowData(x)</code> or <code>colData(x)</code> or alternatively a vector specifying how the merging is performed. If vector, the value must be the same length as <code>nrow(x)/ncol(x)</code> . Rows/Cols corresponding to the same level will be merged. If <code>length(levels(group)) == nrow(x)/ncol(x)</code> , x will be returned unchanged. If group matches with both dimensions, by must be specified. (Default: NULL)
f	Deprecated. Use <code>group</code> instead.
<code>update.tree</code>	Logical scalar. Should <code>rowTree()</code> also be merged? (Default: FALSE)
<code>update_rowTree</code>	Deprecated. Use <code>update.tree</code> instead.
<code>altexp</code>	Character vector. Specify the alternative experiments to be unsplit. (Default: <code>names(altExps(x))</code>)
<code>altExpNames</code>	Deprecated. Use <code>altexp</code> instead.
<code>keep.dimred</code>	Logical scalar. Should the <code>reducedDims(x)</code> be transferred to the result? Please note, that this breaks the link between the data used to calculate the reduced dims. (Default: FALSE)
<code>keep_reducedDims</code>	Deprecated. Use <code>keep.dimred</code> instead.

Details

`splitOn` split data based on grouping variable. Splitting can be done column-wise or row-wise. The returned value is a list of SummarizedExperiment objects; each element containing members of each group.

Value

For `splitOn`: SummarizedExperiment objects in a SimpleList.

For `unsplitOn`: x, with `rowData` and assay data replaced by the unsplit data. `colData` of x is kept as well and any existing `rowTree` is dropped as well, since existing `rowLinks` are not valid anymore.

See Also

[agglomerateByRanks](#) [agglomerateByVariable](#), [sumCountsAcrossFeatures](#), [agglomerateByRank](#), [altExps](#), [splitAltExps](#)

Examples

```
data(GlobalPatterns)
tse <- GlobalPatterns
# Split data based on SampleType.
se_list <- splitOn(tse, group = "SampleType")

# List of SE objects is returned.
se_list

# Create arbitrary groups
rowData(tse)$group <- sample(1:3, nrow(tse), replace = TRUE)
colData(tse)$group <- sample(1:3, ncol(tse), replace = TRUE)

# Split based on rows
# Each element is named based on their group name. If you don't want to name
# elements, use use_name = FALSE. Since "group" can be found from rowData and
# colData you must use `by`.
se_list <- splitOn(tse, group = "group", use.names = FALSE, by = 1)

# When column names are shared between elements, you can store the list to
# altExps
altExps(tse) <- se_list

altExps(tse)

# If you want to split on columns and update rowTree, you can do
se_list <- splitOn(tse, group = colData(tse)$group, update.tree = TRUE)

# If you want to combine groups back together, you can use unsplitBy
unsplitOn(se_list)
```

summaries

Summarizing microbiome data

Description

To query a SummarizedExperiment for interesting features, several functions are available.

Usage

```
getTop(
  x,
  top = 5L,
  method = c("mean", "sum", "median"),
  assay.type = assay_name,
  assay_name = "counts",
  na.rm = TRUE,
  ...
```

```

)

## S4 method for signature 'SummarizedExperiment'
getTop(
  x,
  top = 5L,
  method = c("mean", "sum", "median", "prevalence"),
  assay.type = assay_name,
  assay_name = "counts",
  na.rm = TRUE,
  ...
)

getUnique(x, ...)

## S4 method for signature 'SummarizedExperiment'
getUnique(x, rank = NULL, ...)

summarizeDominance(x, group = NULL, name = "dominant_taxa", ...)

## S4 method for signature 'SummarizedExperiment'
summarizeDominance(x, group = NULL, name = "dominant_taxa", ...)

## S4 method for signature 'SummarizedExperiment'
summary(object, assay.type = assay_name, assay_name = "counts")

```

Arguments

<code>x</code>	A TreeSummarizedExperiment .
<code>top</code>	Numeric scalar. Determines how many top taxa to return. Default is to return top five taxa. (Default: 5)
<code>method</code>	Character scalar. Specify the method to determine top taxa. Either sum, mean, median or prevalence. (Default: "mean")
<code>assay.type</code>	Character scalar. Specifies the name of the assay used in calculation. (Default: "counts")
<code>assay_name</code>	Deprecated. Use <code>assay.type</code> instead.
<code>na.rm</code>	Logical scalar. Should NA values be omitted? (Default: TRUE)
<code>...</code>	Additional arguments passed on to <code>agglomerateByRank()</code> when rank is specified for <code>summarizeDominance</code> .
<code>rank</code>	Character scalar. Defines a taxonomic rank. Must be a value of the output of <code>taxonomyRanks()</code> . (Default: NULL)
<code>group</code>	With <code>group</code> , it is possible to group the observations in an overview. Must be one of the column names of <code>colData</code> .
<code>name</code>	Character scalar. A name for the column of the <code>colData</code> where results will be stored. (Default: "dominant_taxa")
<code>object</code>	A SummarizedExperiment object.


```

# With group, it is possible to group observations based on specified groups
# Gets the overview of dominant taxa
dominant_taxa <- summarizeDominance(GlobalPatterns,
                                   rank = "Genus",
                                   group = "SampleType",
                                   na.rm = TRUE)

dominant_taxa

# Get an overview of sample and taxa counts
summary(GlobalPatterns, assay.type= "counts")

# Get unique taxa at a particular taxonomic rank
# sort = TRUE means that output is sorted in alphabetical order
# With na.rm = TRUE, it is possible to remove NAs
# sort and na.rm can also be used in function getTop
getUnique(GlobalPatterns, "Phylum", sort = TRUE)

```

taxonomy-methods

Functions for accessing taxonomic data stored in rowData.

Description

These function work on data present in `rowData` and define a way to represent taxonomic data alongside the features of a `SummarizedExperiment`.

Usage

```

taxonomyRanks(x)

## S4 method for signature 'SummarizedExperiment'
taxonomyRanks(x)

taxonomyRankEmpty(
  x,
  rank = taxonomyRanks(x)[1L],
  empty.fields = c(NA, "", " ", "\t", "-", "_")
)

## S4 method for signature 'SummarizedExperiment'
taxonomyRankEmpty(
  x,
  rank = taxonomyRanks(x)[1],
  empty.fields = c(NA, "", " ", "\t", "-", "_")
)

checkTaxonomy(x, ...)

```

```

## S4 method for signature 'SummarizedExperiment'
checkTaxonomy(x)

setTaxonomyRanks(ranks)

getTaxonomyRanks()

getTaxonomyLabels(x, ...)

## S4 method for signature 'SummarizedExperiment'
getTaxonomyLabels(
  x,
  empty.fields = c(NA, "", " ", "\t", "-", "_"),
  with.rank = with_rank,
  with_rank = FALSE,
  make.unique = make_unique,
  make_unique = TRUE,
  resolve.loops = resolve_loops,
  resolve_loops = FALSE,
  ...
)

mapTaxonomy(x, ...)

## S4 method for signature 'SummarizedExperiment'
mapTaxonomy(
  x,
  taxa = NULL,
  from = NULL,
  to = NULL,
  use.grepl = use_grepl,
  use_grepl = FALSE
)

IdTaxaToDataFrame(from)

```

Arguments

x	TreeSummarizedExperiment .
rank	Character scalar. Defines a taxonomic rank. Must be a value of <code>taxonomyRanks()</code> function.
empty.fields	Character vector. Defines which values should be regarded as empty. (Default: <code>c(NA, "", " ", "\t")</code>). They will be removed if <code>na.rm = TRUE</code> before agglomeration.
...	optional arguments not used currently.
ranks	Character vector. A vector of ranks to be set.

<code>with.rank</code>	Logical scalar. Should the level be add as a suffix? For example: "Phylum:Crenarchaeota". (Default: FALSE)
<code>with_rank</code>	Deprecated. Use <code>with.rank</code> instead.
<code>make.unique</code>	Logical scalar. Should the labels be made unique, if there are any duplicates? (Default: TRUE)
<code>make_unique</code>	Deprecated. Use <code>make.unique</code> instead.
<code>resolve.loops</code>	Logical scalar. Should <code>resolveLoops</code> be applied to the taxonomic data? Please note that has only an effect, if the data is unique. (Default: TRUE)
<code>resolve_loops</code>	Deprecated. Use <code>resolve.loops</code> instead.
<code>taxa</code>	Character vector. Used for subsetting the taxonomic information. If no information is found, NULL is returned for the individual element. (Default: NULL)
<code>from</code>	<ul style="list-style-type: none"> • For <code>mapTaxonomy</code>: character scalar. A value which must be a valid taxonomic rank. (Default: NULL) • otherwise a <code>Taxa</code> object as returned by <code>IdTaxa</code>
<code>to</code>	Character Scalar. Must be a valid taxonomic rank. (Default: NULL)
<code>use.grepl</code>	Logical. Should pattern matching via <code>grepl</code> be used? Otherwise literal matching is used. (Default: FALSE)
<code>use_grepl</code>	Deprecated. Use <code>use.grepl</code> instead.

Details

`taxonomyRanks` returns, which columns of `rowData(x)` are regarded as columns containing taxonomic information.

`taxonomyRankEmpty` checks, if a selected rank is empty of information.

`checkTaxonomy` checks, if taxonomy information is valid and whether it contains any problems. This is a soft test, which reports some diagnostic and might mature into a data validator used upon object creation.

`getTaxonomyLabels` generates a character vector per row consisting of the lowest taxonomic information possible. If data from different levels, is to be mixed, the taxonomic level is prepended by default.

`IdTaxaToDataFrame` extracts taxonomic results from results of `IdTaxa`.

`mapTaxonomy` maps the given features (taxonomic groups; `taxa`) to the specified taxonomic level (`to` argument) in `rowData` of the `SummarizedExperiment` data object (i.e. `rowData(x)[, taxonomyRanks(x)]`). If the argument `to` is not provided, then all matching taxonomy rows in `rowData` will be returned. This function allows handy conversions between different

Taxonomic information from the `IdTaxa` function of DECIPHER package are returned as a special class. With `as(taxa, "DataFrame")` the information can be easily converted to a `DataFrame` compatible with storing the taxonomic information a `rowData`. Please note that the assigned confidence information are returned as `metatdata` and can be accessed using `metadata(df)$confidence`.

Value

- `taxonomyRanks`: a character vector with all the taxonomic ranks found in `colnames(rowData(x))`
- `taxonomyRankEmpty`: a logical value
- `mapTaxonomy`: a list per element of taxa. Each element is either a `DataFrame`, a character or `NULL`. If all character results have the length of one, a single character vector is returned.

See Also

[agglomerateByRank](#), [toTree](#), [resolveLoop](#)

Examples

```
data(GlobalPatterns)
GlobalPatterns
taxonomyRanks(GlobalPatterns)

checkTaxonomy(GlobalPatterns)

table(taxonomyRankEmpty(GlobalPatterns, "Kingdom"))
table(taxonomyRankEmpty(GlobalPatterns, "Species"))

getTaxonomyLabels(GlobalPatterns[1:20,])

# mapTaxonomy
## returns the unique taxonomic information
mapTaxonomy(GlobalPatterns)
# returns specific unique taxonomic information
mapTaxonomy(GlobalPatterns, taxa = "Escherichia")
# returns information on a single output
mapTaxonomy(GlobalPatterns, taxa = "Escherichia", to="Family")

# setTaxonomyRanks
tse <- GlobalPatterns
colnames(rowData(tse))[1] <- "TAXA1"

setTaxonomyRanks(colnames(rowData(tse)))
# Taxonomy ranks set to: taxa1 phylum class order family genus species

# getTaxonomyRanks is to get/check if the taxonomic ranks is set to "TAXA1"
getTaxonomyRanks()
```

Description

Tengeler2020 includes gut microbiota profiles of 27 persons with ADHD. A standard bioinformatic and statistical analysis done to demonstrate that altered microbial composition could be a driver of altered brain structure and function and concomitant changes in the animals' behavior. This was investigated by colonizing young, male, germ-free C57BL/6JOLA^{Hsd} mice with microbiota from individuals with and without ADHD.

Usage

```
data(Tengeler2020)
```

Format

A TreeSummarizedExperiment with 151 features and 27 samples. The rowData contains taxonomic information at Kingdom, Phylum, Class, Order, Family and Genus level. The colData includes:

patient_status clinical status of the patient (ADHD or Control)

cohort cohort to which the patient belongs (Cohort_1, Cohort_2 and Cohort_3)

patient_status_vs_cohort combination of patient_status and cohort

sample_name unique sample ID

Author(s)

A.C. Tengeler, et al.

References

Tengeler, A.C., Dam, S.A., Wiesmann, M. et al. Gut microbiota from persons with attention-deficit/hyperactivity disorder affects the brain in mice. *Microbiome* 8, 44 (2020). <https://doi.org/10.1186/s40168-020-00816-x>

Supplemental information includes Home-cage activity, methods, results and imaging parameters and publicly-accessible from: https://static-content.springer.com/esm/art%3A10.1186%2Fs40168-020-00816-x/MediaObjects/40168_2020_816_MOESM1_ESM.docx https://static-content.springer.com/esm/art%3A10.1186%2Fs40168-020-00816-x/MediaObjects/40168_2020_816_MOESM2_ESM.docx https://static-content.springer.com/esm/art%3A10.1186%2Fs40168-020-00816-x/MediaObjects/40168_2020_816_MOESM3_ESM.docx

See Also

[mia-datasets](#)

Tito2024QMP

Fecal microbiota samples from 589 patients across different colorectal cancer stages

Description

The study combined Quantitative Microbiome Profiling (QMP) with extensive patient phenotyping from a group of 589 colorectal cancer (CRC) patients, advanced adenoma (AA) patients, and healthy controls. By implementing confounder control and quantitative profiling methods, the study was able to reveal potential misleading associations between microbial markers and colorectal cancer development that were driven by other factors like intestinal inflammation, rather than the cancer diagnosis itself.

Usage

```
data(Tito2024QMP)
```

Format

A TreeSummarizedExperiment with 676 features and 589 samples. The rowData contains species. The colData includes:

sampleID (character) Sample ID from the corresponding study

diagnosis (factor) Diagnosis type, with possible values: "ADE" (advanced adenoma), "CRC" (colorectal cancer), "CTL" (control)

colonoscopy (factor) Colonoscopy result, with possible values: "FIT_Positive", "familial_risk_familial_CRC_FCC", "familial_risk_no", "abdomil_complaints"

Author(s)

Shadman Ishraq

References

Raúl Y. Tito, Sara Verbandt, Marta Aguirre Vazquez, Leo Lahti, Chloe Verspecht, Verónica Lloréns-Rico, Sara Vieira-Silva, Janine Arts, Gwen Falony, Evelien Dekker, Joke Reumers, Sabine Tejpar & Jeroen Raes (2024). Microbiome confounders and quantitative profiling challenge predicted microbial targets in colorectal cancer development. *Nature Medicine*,30, 1339-1348. <https://doi.org/10.1038/s41591-024-02963-2>

See Also

[mia-datasets](#)

transformAssay	<i>Transform assay</i>
----------------	------------------------

Description

Variety of transformations for abundance data, stored in assay. See details for options.

Usage

```
transformAssay(x, ...)

## S4 method for signature 'SummarizedExperiment'
transformAssay(
  x,
  assay.type = "counts",
  assay_name = NULL,
  method = c("alr", "chi.square", "clr", "css", "frequency", "hellinger", "log", "log10",
    "log2", "max", "normalize", "pa", "range", "rank", "rclr", "relabundance", "rrank",
    "standardize", "total", "z"),
  MARGIN = "samples",
  name = method,
  pseudocount = FALSE,
  ...
)

## S4 method for signature 'SingleCellExperiment'
transformAssay(x, altexp = altExpNames(x), ...)
```

Arguments

- x** [TreeSummarizedExperiment](#).
- ...** additional arguments passed e.g. on to `vegan:decostand`.
- **reference**: Character scalar. Used to fill reference sample's column in returned assay when calculating alr. (Default: NA)
 - **ref_vals** Deprecated. Use **reference** instead.
 - **percentile**: Numeric scalar or NULL (`css`). Used to set the percentile value that calculates the scaling factors in the `css` normalization. If NULL, percentile is estimated from the data by calculating the portion of samples that exceed the threshold. (Default: NULL)
 - **scaling**: Numeric scalar. Adjusts the normalization scale by dividing the calculated scaling factors, effectively changing the magnitude of the normalized counts. (Default: 1000).
 - **threshold**: Numeric scalar. Specifies relative difference threshold and determines the first point where the relative change in differences between consecutive quantiles exceeds this threshold. (Default: 0.1).

assay.type	Character scalar. Specifies which assay to use for calculation. (Default: "counts")
assay_name	Deprecated. Use assay.type instead.
method	Character scalar. Specifies the transformation method.
MARGIN	Character scalar. Determines whether the transformation is applied sample (column) or feature (row) wise. (Default: "samples")
name	Character scalar. A name for the column of the colData where results will be stored. (Default: "method")
pseudocount	Logical scalar or numeric scalar. When TRUE, automatically adds half of the minimum positive value of assay.type (missing values ignored by default: na.rm = TRUE). When FALSE, does not add any pseudocount (pseudocount = 0). Alternatively, a user-specified numeric value can be added as pseudocount. (Default: FALSE).
altexp	Character vector or NULL. Specifies the names of alternative experiments to which the transformation should also be applied. If NULL, the transformation is only applied to the main experiment. (Default: altExpNames(x)).

Details

transformAssay function provides a variety of options for transforming abundance data. The transformed data is calculated and stored in a new assay.

The transformAssay provides sample-wise (column-wise) or feature-wise (row-wise) transformation to the abundance table (assay) based on specified MARGIN.

The available transformation methods include:

- 'alr', 'chi.square', 'clr', 'frequency', 'hellinger', 'log', 'normalize', 'pa', 'rank', 'rclr' relabundance', 'rrank', 'standardize', 'total': please refer to [decostand](#) for details.
- 'css': Cumulative Sum Scaling (CSS) can be used to normalize count data by accounting for differences in library sizes. By default, the function determines the normalization percentile for summing and scaling counts. If you want to specify the percentile value, good default value might be 0.5. The method is inspired by the CSS methods in [metagenomeSeq](#) package.
- 'log10': log10 transformation can be used for reducing the skewness of the data.

$$\log_{10} = \log_{10} x$$

where x is a single value of data.

- 'log2': log2 transformation can be used for reducing the skewness of the data.

$$\log_2 = \log_2 x$$

where x is a single value of data.

Value

transformAssay returns the input object x , with a new transformed abundance table named name added in the [assay](#).

References

Paulson, J., Stine, O., Bravo, H. et al. (2013) Differential abundance analysis for microbial marker-gene surveys *Nature Methods* 10, 1200–1202. doi:10.1038/nmeth.2658

Examples

```

data(GlobalPatterns)
tse <- GlobalPatterns

# By specifying 'method', it is possible to apply different transformations,
# e.g. compositional transformation.
tse <- transformAssay(tse, method = "relabundance")

# The target of transformation can be specified with "assay.type"
# Pseudocount can be added by specifying 'pseudocount'.

# Perform CLR with smallest positive value as pseudocount
tse <- transformAssay(
  tse, assay.type = "relabundance", method = "clr",
  pseudocount = TRUE
)

head(assay(tse, "clr"))

# Perform CSS normalization.
tse <- transformAssay(tse, method = "css")
head(assay(tse, "css"))

# With MARGIN, you can specify the if transformation is done for samples or
# for features. Here Z-transformation is done feature-wise.
tse <- transformAssay(tse, method = "standardize", MARGIN = "features")
head(assay(tse, "standardize"))

# Name of the stored table can be specified.
tse <- transformAssay(tse, method="hellinger", name="test")
head(assay(tse, "test"))

# pa returns presence absence table.
tse <- transformAssay(tse, method = "pa")
head(assay(tse, "pa"))

# rank returns ranks of taxa.
tse <- transformAssay(tse, method = "rank")
head(assay(tse, "rank"))

# In order to use other ranking variants, modify the chosen assay directly:
assay(tse, "rank_average", withDimnames = FALSE) <- colRanks(
  assay(tse, "counts"), ties.method = "average", preserveShape = TRUE)

# Using altexp parameter. First agglomerate the data and then apply
# transformation.
tse <- GlobalPatterns

```

```
tse <- agglomerateByRanks(tse)
tse <- transformAssay(tse, method = "relabundance")
# The transformation is applied to all alternative experiments
altExp(tse, "Species")
```

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