Package ‘SIAMCAT’

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Type Package

Title Statistical Inference of Associations between Microbial Communities And host phenoTypes

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Description Pipeline for Statistical Inference of Associations between Microbial Communities And host phenoTypes (SIAMCAT). A primary goal of analyzing microbiome data is to determine changes in community composition that are associated with environmental factors. In particular, linking human microbiome composition to host phenotypes such as diseases has become an area of intense research. For this, robust statistical modeling and biomarker extraction toolkits are crucially needed. SIAMCAT provides a full pipeline supporting data preprocessing, statistical association testing, statistical modeling (LASSO logistic regression) including tools for evaluation and interpretation of these models (such as cross validation, parameter selection, ROC analysis and diagnostic model plots).

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Description
Pipeline for Statistical Inference of Associations between Microbial Communities And host phenotypes (SIAMCAT). A primary goal of analyzing microbiome data is to determine changes in community composition that are associated with environmental factors. In particular, linking human microbiome composition to host phenotypes such as diseases has become an area of intense research. For this, robust statistical modeling and biomarker extraction toolkits are crucially needed. SIAMCAT provides a full pipeline supporting data preprocessing, statistical association testing, statistical modeling (LASSO logistic regression) including tools for evaluation and interpretation of these models (such as cross validation, parameter selection, ROC analysis and diagnostic model plots).

Details
SIAMCAT is a pipeline for Statistical Inference of Associations between Microbial Communities And host phenotypes. A primary goal of analyzing microbiome data is to determine changes in community composition that are associated with environmental factors. In particular, linking human microbiome composition to host phenotypes such as diseases has become an area of intense research. For this, robust statistical modeling and biomarker extraction toolkits are crucially needed!

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accessSlot
Universal slot accessor function for siamcat-class.

Description
This function is used internally by many accessors.

Usage
accessSlot(siamcat, slot, verbose=1)
add.meta.pred

Arguments

siamcat an object of siamcat-class.
slot A character string indicating the slot (not data class) of the component data type that is desired.
verbose If the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Value

Returns the component object specified by the argument slot. Returns NULL if slot does not exist.

Examples

#
data(siamcat_example)
accessSlot(siamcat_example, "label")
accessSlot(siamcat_example, "model_list")

add.meta.pred

Add metadata as predictors

Description

This function adds metadata to the feature matrix to be later used as predictors

Usage

add.meta.pred(siamcat, pred.names, std.meta = TRUE,
feature.type='normalized', verbose = 1)

Arguments

siamcat object of class siamcat-class
pred.names vector of names of the variables within the metadata to be added to the feature matrix as predictors
std.meta boolean, should added metadata features be standardized?, defaults to TRUE
feature.type string, on which type of features should the function work? Can be either "original", "filtered", or "normalized". Please only change this parameter if you know what you are doing!
verbose integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1
Details

This function adds one or several metadata variables to the set of features, so that they can be included for model training.

Usually, this function should be called before `train.model`.

Numerical meta-variables are added as z-scores to the feature matrix unless specified otherwise. Please be aware, that non-numerical metadata variables will be converted to numerical values by using `as.numeric()` and could therefore lead to errors. Thus, it makes sense to encode non-numerical metadata variables to numerically before you start the SIAMCAT workflow.

Value

an object of class `siamcat-class` with metadata added to the features

Examples

```r
data(siamcat_example)

# Add the Age of the patients as potential predictor
siamcat_age_added <- add.meta.pred(siamcat_example, pred.names=c('Age'))

# Add Age and BMI as potential predictors
# Additionally, prevent standardization of the added features
siamcat_meta_added <- add.meta.pred(siamcat_example,
   pred.names=c('Age', 'BMI'), std.meta=FALSE)
```

**association.plot**

*Visualize associations between features and classes*

Description

This function visualizes different measures of association between features and the label, computed previously with the `check.associations` function

Usage

```r
association.plot(siamcat, fn.plot=NULL, color.scheme = "RdYlBu",
   sort.by = "fc", max.show = 50, plot.type = "quantile.box",
   panels = c("fc", "auroc"), prompt=TRUE, verbose = 1)
```

Arguments

- `siamcat`: object of class `siamcat-class`
- `fn.plot`: string, filename for the pdf-plot. If `fn.plot` is `NULL`, the plot will be produced in the active graphics device.
- `color.scheme`: valid R color scheme or vector of valid R colors (must be of the same length as the number of classes), defaults to 'RdYlBu'
sort.by string, sort features by p-value ("p.val"), by fold change ("fc") or by prevalence shift ("pr.shift"), defaults to "fc"
max.show integer, how many associated features should be shown, defaults to 50
plot.type string, specify how the abundance should be plotted, must be one of these: c("bean", "box", "quantile.box", "quantile.rect"), defaults to "quantile.box"
panels vector, name of the panels to be plotted next to the abundances, possible entries are c("fc", "auroc", "prevalence"), defaults to c("fc", "auroc")
prompt boolean, turn on/off prompting user input when not plotting into a pdf-file, defaults to TRUE
verbose integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details
This function visualizes the results of the computations carried out in the check.associations function. It produces a plot of the top max.show associated features at a user-specified significance level alpha.

For binary classification problems, the plot will show the distribution of the log10-transformed abundances for both classes, a P-value from the significance test, and user-selected panels for the effect size (AU-ROC, prevalence shift, or generalized fold change). For regression problems, the plot will show the Spearman correlation, the significance, and the linear model effect size.

Value
Does not return anything, but instead produces association plot

Examples

# Example data
data(siamcat_example)

# Simple example
association.plot(siamcat_example, fn.plot = "./assoc_plot.pdf")

# Plot associations as box plot
association.plot(siamcat_example,
    fn.plot = "./assoc_plot_box.pdf",
    plot.type = "box")

# Additionally, sort by p-value instead of by fold change
association.plot(siamcat_example,
    fn.plot = "./assoc_plot_fc.pdf",
    plot.type = "box", sort.by = "p.val")

# Custom colors
association.plot(siamcat_example,
    fn.plot = "./assoc_plot_blue_yellow.pdf",
    plot.type = "box", color.scheme = c("cornflowerblue", "#fffc125"))
associations

Retrieve the results of association testing from a SIAMCAT object

Description

Function to retrieve the results of association testing

Usage

associations(siamcat, verbose = 1)

## S4 method for signature 'siamcat'
associations(siamcat, verbose = 1)

Arguments

- **siamcat**: (Required). An instance of `siamcat-class` containing the results of association testing
- **verbose**: integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details

The function returns the results of the association testing procedure as dataframe. See `check.associations` for more details.

Value

A `data.frame` of association testing results or NULL

Examples

data(siamcat_example)
temp <- associations(siamcat_example)
head(temp)

associations<- Assign a new associations object to x

Description

Assign a new associations object to x
Usage

`associations(x) <- value`

```r
## S4 replacement method for signature 'siamcat,list'
associations(x) <- value
```

Arguments

- **x**: an object of class `siamcat-class`
- **value**: an associations object

Value

none

Examples

```r
data(siamcat_example)
associations(siamcat_example) <- list(
  'assoc.results'=associations(siamcat_example),
  'assoc.param'=assoc_param(siamcat_example))
```

---

### assoc_param

**Retrieve the list of parameters for association testing from a SIAMCAT object**

**Description**

Function to retrieve the list of parameters for association testing

**Usage**

```r
assoc_param(siamcat, verbose=1)
```

```r
## S4 method for signature 'siamcat'
assoc_param(siamcat, verbose = 1)
```

**Arguments**

- **siamcat**: (Required). An instance of `siamcat-class` containing the results from association testing
- **verbose**: integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

**Details**

The function returns the list of parameters used in association testing. See `check.associations` for more details.
check.associations

Value

A list of parameters for association testing or NULL

Examples

data(siamcat_example)
temp <- assoc_param(siamcat_example)
names(temp)

check.associations

Calculate associations between features and labels

Description

This function computes different measures of association between features and the label and stores the results in the association slot of the SIAMCAT object

Usage

check.associations(siamcat, formula="feat~label", test='wilcoxon', alpha=0.05, mult.corr="fdr", log.n0=1e-06, pr.cutoff=1e-06, probs.fc=seq(.1, .9, .05), paired=NULL, feature.type='filtered', verbose = 1)

Arguments

siamcat object of class siamcat-class
formula string, formula used for testing, see Details for more information, defaults to "feat~label"
test string, statistical test used for the association testing, can be either 'wilcoxon' or 'lm', see Details for more information, defaults to 'wilcoxon'
alpha float, significance level, defaults to 0.05
mult.corr string, multiple hypothesis correction method, see p.adjust, defaults to "fdr"
log.n0 float, pseudo-count to be added before log-transformation of the data, defaults to 1e-06. Will be ignored if feature.type is "normalized".
pr.cutoff float, cutoff for the prevalence computation, defaults to 1e-06
probs.fc numeric vector, quantiles used to calculate the generalized fold change between groups, see Details for more information, defaults to seq(.1, .9, .05)
paired character, column name of the meta-variable containing information for a paired test, defaults to NULL
feature.type string, on which type of features should the function work? Can be either c("original", "filtered", or "normalized"). Please only change this parameter if you know what you are doing! If feature.type is "normalized", the normalized abundances will not be log10-transformed.
check.associations

verbose integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Value object of class `siamcat-class` with the slot `associations` filled

Statistical testing

The function uses the Wilcoxon test as default statistical test for binary classification problems. Alternatively, a simple linear model (as implemented in `lm`) can be used as well. For regression problems, the function defaults to the linear model.

Effect sizes

The function calculates several measures for the effect size of the associations between microbial features and the label. For binary classification problems, these associations are:

- AUROC (area under the Receiver Operating Characteristics curve) as a non-parametric measure of enrichment,
- the generalized fold change (gFC), a pseudo-fold change which is calculated as geometric mean of the differences between quantiles across both groups,
- prevalence shift (difference in prevalence between the two groups).

For regression problems, the effect sizes are:
- Spearman correlation between the feature and the label.

Confounder-corrected testing

To correct for possible confounders while testing for association, the function uses linear mixed effect models as implemented in the `lmerTest` package. To do so, the test formula needs to be adjusted to include the confounder. For example, when correcting for the metadata information `Sex`, the formula would be: `feat~label+(1|Sex)` (see also the example below).

Please note that modifying the formula parameter in this function might lead to unexpected results!

Paired testing

For paired testing, e.g. when the same patient has been sampled before and after an intervention, the `paired` parameter can be supplied to the function. This indicated a column in the metadata table that holds the information about pairing.

Examples

```r
# Example data
data(siamcat_example)

# Simple example
siamcat_example <- check.associations(siamcat_example)
```
check.confounders

Check for potential confounders in the metadata

Description
Checks potential confounders in the metadata and visualize the results

Usage
check.confounders(siamcat, fn.plot, meta.in = NULL, feature.type='filtered', verbose = 1)

Arguments
siamcat an object of class siamcat-class
fn.plot string, filename for the pdf-plot
meta.in vector, specific metadata variable names to analyze, defaults to NULL (all meta-
data variables will be analyzed)
feature.type string, on which type of features should the function work? Can be either c("original", "filtered", or "normalized"). Please only change this parameter if you know what you are doing!
verbose integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details
This function checks for associations between class labels and potential confounders (e.g. Age, Sex, or BMI) that are present in the metadata. Statistical testing is performed with Fisher’s exact test or Wilcoxon test, while associations are visualized either as barplot or Q-Q plot, depending on the type of metadata.
Additionally, it evaluates associations among metadata variables using conditional entropy and associations with the label using generalized linear models, producing a correlation heatmap and appropriate quantitative barplots, respectively.
Please note that the confounder check is currently only available for binary classification problems!
create.data.split

Value

Does not return anything, but outputs plots to specified pdf file

Examples

# Example data
data(siamcat_example)

# Simple working example
check.confounders(siamcat_example, './conf_plot.pdf')

create.data.split  Split a dataset into training and a test sets.

Description

This function prepares the cross-validation by splitting the data into num.folds training and test folds for num.resample times.

Usage

create.data.split(siamcat, num.folds = 2, num.resample = 1, stratify = TRUE, inseparable = NULL, verbose = 1)

Arguments

siamcat  object of class siamcat-class
num.folds  integer number of cross-validation folds (needs to be >=2), defaults to 2
num.resample  integer, resampling rounds (values <= 1 deactivate resampling), defaults to 1
stratify  boolean, should the splits be stratified so that an equal proportion of classes are present in each fold?, will be ignored for regression tasks, defaults to TRUE
inseparable  string, name of metadata variable to be inseparable, defaults to NULL, see Details below
verbose  integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details

This function splits the labels within a siamcat-class object and prepares the internal cross-validation for the model training (see train.model).

The function saves the training and test instances for the different cross-validation folds within a list in the data_split-slot of the siamcat-class object, which is a list with four entries:

- num.folds - the number of cross-validation folds
- num.resample - the number of repetitions for the cross-validation
create.label

- training.folds - a list containing the indices for the training instances
- test.folds - a list containing the indices for the test instances

If provided, the data split will take into account a metadata variable for the data split (by providing the inseparable argument). For example, if the data contains several samples for the same individual, it makes sense to keep data from the same individual within the same fold.

If inseparable is given, the stratify argument will be ignored.

Value

object of class siamcat-class with the data_split-slot filled

Examples

data(siamcat_example)

# simple working example
siamcat_split <- create.data.split(siamcat_example, num.folds=10, num.resample=5, stratify=TRUE)

create.label

Create a label list

Description

This function creates a label object from metadata or an atomic vector

Usage

create.label(label, case, meta=NULL, control=NULL, p.lab = NULL, n.lab = NULL, remove.meta.column=FALSE, verbose=1)

Arguments

label
 named vector to create the label or the name of the metadata column that will be used to create the label

case
 name of the group that will be used as a positive label. If the variable is binary, the other label will be used as a negative one. If the variable has multiple values, all the other values will be used as a negative label (testing one vs rest).

meta
 metadata dataframe object or an object of class sample_data-class

control
 name of a label or vector with names that will be used as a negative label. All values that are not equal to case and control will be dropped. Default to NULL in which case: If the variable is binary, the value not equal to case and control will be used as negative. If the variable has multiple values, all the values not equal to cases will be used as a negative label (testing one vs rest).

p.lab
 name of the positive group (useful mostly for visualizations). Default to NULL in which case the value of the positive group will be used.
n.lab
name of the negative group (useful mostly for visualizations). Default to NULL in which case the value of the negative group will be used for binary variables and "rest" will be used for variables with multiple values.

remove.meta.column
boolean indicating if the label column in the metadata should be retained. Please note that if this is set to TRUE, the function will return a list as result. Defaults to FALSE

verbose
integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details
The function creates a list to be used as label in a SIAMCAT object. Mainly for internal use, but it can be used to customize your label (p.lab and n.lab will be used as labels during plotting, for example).

The input for the function can be either a named vector encoding the label or the name of a column in the metadata (needs to be provided as well) which contains the label information.

Value
return either

• a list to be used in a SIMCAT object OR
• a list with entries meta and label, if remove.meta.column is set to TRUE

Examples

data('meta_crc_zeller')

label <- create.label(label='Group', case='CRC', meta=meta_crc_zeller)

---

data_split

Retrieve the data split from a SIAMCAT object

Description
Function to retrieve the data split stored in the data_split slot within a SIAMCAT object

Usage

data_split(siamcat, verbose=1)

## S4 method for signature 'siamcat'
data_split(siamcat, verbose = 1)
data_split<-  

Arguments

siamcat (Required). An instance of siamcat-class containing a data split
verbose integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details

The function returns a list containing information about the data split. See create.data.split for more details.

Value

A list containing the data split information or NULL

Examples

data(siamcat_example)
temp <- data_split(siamcat_example)
names(temp)

data_split<- Assign a new list containing a cross-validation split to a SIAMCAT object

Description

Assign a new list containing a cross-validation split to a SIAMCAT object

Usage

data_split(x) <- value

## S4 replacement method for signature 'siamcat,list'
data_split(x) <- value

Arguments

x an object of class siamcat-class
value list containing a cross-validation split

Value

none

Examples

data(siamcat_example)
data_split(siamcat_example) <- data_split(siamcat_example)
**evaluate.predictions**  
*Evaluate prediction results*

### Description
This function compares the predictions (from `make.predictions()`) and true labels for all samples and evaluates the results.

### Usage
```r
evaluate.predictions(siamcat, verbose = 1)
```

### Arguments
- **siamcat**: object of class `siamcat-class`
- **verbose**: integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

### Value
object of class `siamcat-class` with the slot `eval_data` filled

### Binary classification problems
This function calculates several metrics for the predictions in the `pred_matrix`-slot of the `siamcat-class`-object. The Area Under the Receiver Operating Characteristic (ROC) Curve (AU-ROC) and the Precision-Recall Curve will be evaluated and the results will be saved in the `eval_data`-slot of the supplied `siamcat-class`-object. The `eval_data`-slot contains a list with several entries:

- `$roc` - average ROC-curve across repeats or a single ROC-curve on complete dataset (see roc);
- `$auroc` - AUC value for the average ROC-curve;
- `$prc` - list containing the positive predictive value (precision) and true positive rate (recall) values used to plot the mean PR curve;
- `$auprc` - AUC value for the mean PR curve;
- `$ev` - list containing for different decision thresholds the number of false positives, false negatives, true negatives, and true positives.

For the case of repeated cross-validation, the function will additionally return

- `$roc.all` - list of roc objects (see roc) for every repeat;
- `$auroc.all` - vector of AUC values for the ROC curves for every repeat;
- `$prc.all` - list of PR curves for every repeat;
- `$auprc.all` - vector of AUC values for the PR curves for every repeat;
- `$ev.all` - list of `ev` lists (see above) for every repeat.
Regression problems

This function calculates several metrics for the evaluation of predictions and will store the results in the eval_data-slot of the supplied siamcat-class objects. The eval_data-slot will contain:

- \( r^2 \) - the mean R squared value across repeats or a single R-squared value on the complete dataset;
- mae - the mean absolute error of the predictions;
- mse - the mean squared error of the predictions.

For the case of repeated cross-validation, the function will additionally compute all three of these measures for the individual cross-validation repeats and will store the results in the eval_data slot as \( r^2 \text{.all}, \text{mae}\text{.all}, \text{and} \text{mse}\text{.all}.\)

Examples

```r
data(siamcat_example)
siamcat_evaluated <- evaluate.predictions(siamcat_example)
```

### eval_data

**Retrieve the evaluation metrics from a SIAMCAT object**

#### Description

Function to retrieve the evaluation metrics from a SIAMCAT object

#### Usage

```r
eval_data(siamcat, verbose=1)
```

```r
## S4 method for signature 'siamcat'
eval_data(siamcat, verbose = 1)
```

#### Arguments

- **siamcat** *(Required).* A siamcat-class object that contains evaluation data
- **verbose** integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

#### Details

The function returns a list containing the evaluation metrics from a SIAMCAT object. See evaluate.predictions for more information on evaluation data.

#### Value

The list of evaluation data or NULL
Examples

data(siamcat_example)
temp <- eval_data(siamcat_example)
names(temp)
temp$auroc

---

eval_data <- Assign a new list with evaluation data to a SIAMCAT object

Description

Assign a new list with evaluation data to a SIAMCAT object

Usage

eval_data(x) <- value

## S4 replacement method for signature 'siamcat,list'
eval_data(x) <- value

Arguments

x an object of class siamcat-class
value a list of evaluation data

Value

none

Examples

data(siamcat_example)
eval_data(siamcat_example) <- eval_data(siamcat_example)

---

feat.crc.zeller Example feature matrix

Description

Feature matrix (as data.frame) of the CRC dataset from Zeller et al. MSB 2014 (see http://msb.embopress.org/content/10/11/766), containing 141 samples and 1754 bacterial species (features).

Source

http://msb.embopress.org/content/10/11/766
feature_type

Retrieve the feature type used for model training from a SIAMCAT object

Description
Function to retrieve information on which type of features the models were trained

Usage
feature_type(siamcat, verbose=1)

## S4 method for signature 'siamcat'
feature_type(siamcat, verbose = 1)

Arguments
siamcat (Required). An instance of siamcat-class that contains trained models
verbose integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details
The function extracts the information on which type of features the models were trained.

Value
The string describing type of feature used for the model training or NULL

Examples
data(siamcat_example)
feature_type(siamcat_example)

---

feature_weights

Retrieve the matrix of feature weights from a SIAMCAT object

Description
Function to extract the feature weights from a SIAMCAT object

Usage
feature_weights(siamcat, verbose=1)

## S4 method for signature 'siamcat'
feature_weights(siamcat, verbose = 1)
Arguments

siamcat  (Required). A siamcat-class object that contains trained models
verbose integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details

The function extracts the weight matrix from all trained models (see weight_matrix) and computes several metrics on the feature weights:

- mean.weight - mean weight across trained models
- median.weight - median weight across trained models
- sd.weight - standard deviation of the weight across trained models
- mean.rel.weight - mean relative weight across trained models (each model is normalized by the absolute of all weights)
- median.rel.weight - median relative weight across trained models
- sd.rel.weight - standard deviation of the relative weight across trained models
- percentage - percentage of models in which this feature was selected (i.e. non-zero)

Value

A dataframe containing mean/median feature weight and additional info or NULL

Examples

data(siamcat_example)
temp <- feature_weights(siamcat_example)
head(temp)

filter.features Perform unsupervised feature filtering.

Description

This function performs unsupervised feature filtering.

Usage

filter.features(siamcat, filter.method = "abundance",
cutoff = 0.001, rm.unmapped = TRUE, feature.type='original', verbose = 1)
Arguments

- **siamcat**: an object of class `siamcat-class`
- **filter.method**: string, method used for filtering the features, can be one of these: c('abundance', 'cum.abundance', 'prevalence', 'variance', 'pass'). defaults to 'abundance'
- **cutoff**: float, abundance, prevalence, or variance cutoff, defaults to 0.001 (see Details below)
- **rm.unmapped**: boolean, should unmapped reads be discarded?, defaults to TRUE
- **feature.type**: string, on which type of features should the function work? Can be either "original", "filtered", or "normalized". Please only change this parameter if you know what you are doing!
- **verbose**: integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details

This function filters the features in a `siamcat-class` object in an unsupervised manner.

The different filter methods work in the following way:

- 'abundance' - remove features whose maximum abundance is never above the threshold value in any of the samples
- 'cum.abundance' - remove features with very low abundance in all samples, i.e. those that are never among the most abundant entities that collectively make up (1-cutoff) of the reads in any sample
- 'prevalence' - remove features with low prevalence across samples, i.e. those that are undetected (relative abundance of 0) in more than 1 - cutoff percent of samples.
- 'variance' - remove features with low variance across samples, i.e. those that have a variance lower than cutoff
- 'pass' - pass-through filtering will not change the features

Features can also be filtered repeatedly with different methods, e.g. first using the maximum abundance filtering and then using prevalence filtering. However, if a filtering method has already been applied to the dataset, SIAMCAT will default back on the original features for filtering.

Value

- siamcat an object of class `siamcat-class`

Examples

```
# Example dataset
data(siamcat_example)

# Simple examples
siamcat_filtered <- filter.features(siamcat_example,
    filter.method='abundance',
    cutoff=1e-03)
```
# 5% prevalence filtering
siamcat_filtered <- filter.features(siamcat_example,
filter.method='prevalence',
cutoff=0.05)

# filter first for abundance and then for prevalence
siamcat_filt <- filter.features(siamcat_example,
filter.method='abundance', cutoff=1e-03)
siamcat_filt <- filter.features(siamcat_filt, filter.method='prevalence',
cutoff=0.05, feature.type='filtered')

---

**filter.label**

*Filter the label of a SIMACAT object*

**Description**

This function filters the label in a SIMACAT object.

**Usage**

```
filter.label(siamcat, ids, verbose = 1)
```

**Arguments**

- **siamcat**: an object of class `siamcat-class`
- **ids**: vector, can contain either names or indices of samples to be retained
- **verbose**: integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

**Details**

This function filters the label contained in a SIMACAT object, based on the provided `ids`. The IDs can be either sample names or indices to be retained.

Predominantly for internal use...

**Please note**: It makes sense to run `validate.data` after filtering the label.

**Value**

- **siamcat**: an object of class `siamcat-class`

**Examples**

```
data(siamcat_example)

# simple working example
siamcat_filtered <- filter.label(siamcat_example, ids=c(1:20))
```
filt_feat

Retrieve the information stored in the filt_feat slot within a SIAM-CAT object

Description
Function to retrieve the information stored in the filt_feat slot within a SIAMCAT object

Usage
filt_feat(siamcat, verbose=1)

## S4 method for signature 'siamcat'
filt_feat(siamcat, verbose = 1)

Arguments
siamcat (Required). An instance of siamcat-class that contains filtered features
verbose integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details
The function will return a list containing the information stored in the filt_feat slot of a SIAMCAT object. This list contains:

- filt.feat - filtered features as matrix, see get.filt_feat.matrix
- filt.param - parameters used for feature filtering, see get.filt_feat.matrix

Value
The list stored in the filt_feat slot of the SIAMCAT object or NULL

Examples
data(siamcat_example)
temp <- filt_feat(siamcat_example)
names(temp)
filt_feat<-  Assign a new filt_feat object to x

Description
Assign a new filt_feat object to x

Usage
filt_feat(x) <- value

## S4 replacement method for signature 'siamcat,list'
filt_feat(x) <- value

Arguments

<table>
<thead>
<tr>
<th></th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>x</td>
<td>an object of class siamcat-class</td>
</tr>
<tr>
<td>value</td>
<td>an filt_feat object</td>
</tr>
</tbody>
</table>

Value

none

Examples

data(siamcat_example)
filt_feat(siamcat_example) <- list(
   filt.feat=filt_feat(siamcat_example),
   filt.param=filt_params(siamcat_example))

filt_params  Retrieve the list of parameters for feature filtering from a SIAMCAT object

Description
Function to retrieve the list of parameters for feature filtering

Usage

class(siamcat_example) <- list(
   filt.feat=filt_feat(siamcat_example),
   filt.param=filt_params(siamcat_example))

class(siamcat_example) <- list(
   filt.feat=filt_feat(siamcat_example),
   filt.param=filt_params(siamcat_example))
get.component.classes

Arguments

siamcat (Required). An instance of siamcat-class containing filtered features

verbose integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details

The function returns the list of feature filtering parameters. See filter.features for more details.

Value

A list of feature filtering parameters or NULL

Examples

data(siamcat_example)
temp <- filt_params(siamcat_example)
names(temp)
get.filt.feat.matrix

Retrieve the filtered features from a SIAMCAT object

Description

Function to retrieve the filtered features from a SIAMCAT object

Usage

get.filt.feat.matrix(siamcat)

Arguments

siamcat (Required). An instance of siamcat-class containing filtered features

Details

The function returns the filtered features as matrix. See filter.features for more details.

Value

A matrix containing the filtered features

Examples

data(siamcat_example)
feat.filt <- get.filt.feat.matrix(siamcat_example)
feat.filt[1:3, 1:3]

get.norm.feat.matrix

Retrieve the normalized features from a SIAMCAT object

Description

Function to retrieve the normalized features from a SIAMCAT object

Usage

get.norm.feat.matrix(siamcat)

Arguments

siamcat (Required). An instance of siamcat-class containing normalized features

Details

The function returns the normalized features as matrix. See normalize.features for more details.
get.orig_feat.matrix

Value

A matrix containing the normalized features

Examples

data(siamcat_example)
feat.norm <- get.norm_feat.matrix(siamcat_example)
feat.norm[1:3, 1:3]

give.orig_feat.matrix

Retrieve the original features from a SIAMCAT object

Description

Function to retrieve the original features from a SIAMCAT object

Usage

give.orig_feat.matrix(siamcat)

Arguments

siamcat (Required). An instance of siamcat-class

Details

The function returns the original features as matrix.

Value

A matrix containing the original features

Examples

data(siamcat_example)
feat.original <- get.orig_feat.matrix(siamcat_example)
feat.original[1:3, 1:3]
Description

Retrieve the label from a SIAMCAT object

Usage

label(siamcat, verbose=1)

## S4 method for signature 'siamcat'
label(siamcat, verbose = 1)

Arguments

siamcat (Required). A siamcat-class object
verbose integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details

This function will retrieve the label information from a SIAMCAT object. The label will contain three entries:

- label: The label as named vector, in which the classes are encoded numerically
- info: Information about the different classes
- type: What kind of label is it?

Value

The label or NULL.

Examples

data(siamcat_example)
temp <- label(siamcat_example)
head(temp$label)
temp$info
temp$type
label<- Assign a new label object to a SIAMCAT object

Description
Assign a new label object to a SIAMCAT object

Usage
label(x) <- value

## S4 replacement method for signature 'siamcat,list'
label(x) <- value

Arguments

x an object of class siamcat-class
value an list (in label format)

Value
none

Examples
data(siamcat_example)
label(siamcat_example) <- label(siamcat_example)

LearnerClassifLiblineaR
LiblineaR Classification Learner

Description
LiblineaR Classification Learner
LiblineaR Classification Learner

Details
Type of SVC depends on type argument:

- 0 L2-regularized logistic regression (primal)
- 1 L2-regularized L2-loss support vector classification (dual)
- 3 L2-regularized L1-loss support vector classification (dual)
- 2 L2-regularized L2-loss support vector classification (primal)
• 4 Support vector classification by Crammer and Singer
• 5 L1-regularized L2-loss support vector classification
• 6 L1-regularized logistic regression
• 7 L2-regularized logistic regression (dual)

If number of records > number of features, type = 2 is faster than type = 1 (Hsu et al. 2003).
Note that probabilistic predictions are only available for types 0, 6, and 7. The default epsilon value depends on the type parameter, see [Liblinear::Liblinear].

Super classes

mlr3::Learner -> mlr3::LearnerClassif -> LearnerClassifLiblinearR

Methods

Public methods:
• LearnerClassifLiblinearR$new()
• LearnerClassifLiblinearR$clone()

Method $new()$: #’ Creates a new instance of this [R6][R6::R6Class] class.
Usage:
LearnerClassifLiblinearR$new()

Method $clone()$: The objects of this class are cloneable with this method.
Usage:
LearnerClassifLiblinearR$clone(deep = FALSE)

Arguments:
dee

make.predictions

Make predictions on a test set

Description

This function takes a siamcat-class-object containing a model trained by train.model and performs predictions on a given test-set.

Usage

make.predictions(siamcat, siamcat.holdout = NULL,
  normalize.holdout = TRUE, verbose = 1)
Arguments

siamcat object of class siamcat-class
siamcat.holdout optional, object of class siamcat-class on which to make predictions, defaults to NULL
normalize.holdout boolean, should the holdout features be normalized with a frozen normalization (see normalize.features) using the normalization parameters in siamcat?, defaults to TRUE
verbose integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details

This functions uses the model in the model_list-slot of the siamcat object to make predictions on a given test set. The test set can either consist of the test instances in the cross-validation, saved in the data_split-slot of the same siamcat object, or a completely external feature set, given in the form of another siamcat object (siamcat.holdout).

Value

object of class siamcat-class with the slot pred_matrix filled

Examples

data(siamcat_example)

# Simple example
siamcat_example <- train.model(siamcat_example, method='lasso')
siamcat.pred <- make.predictions(siamcat_example)

# Predictions on a holdout-set (not run)
# pred.mat <- make.predictions(siamcat.trained, siamcat.holdout,
#     #   normalize.holdout=TRUE)

meta Retrieve the metadata from a SIAMCAT object

Description

Retrieve the metadata from a SIAMCAT object
Usage

meta(siamcat)

## S4 method for signature 'siamcat'
meta(siamcat)

## S4 method for signature 'sample_data'
meta(siamcat)

Arguments

siamcat (Required). A siamcat-class object

Details

This function will retrieve the metadata from a SIAMCAT object. The metadata is a object of the sample_data-class.

Value

The metadata as sample_data-class object

Examples

data(siamcat_example)
temp <- meta(siamcat_example)
head(temp)

Description

Metadata (as data.frame) of the CRC dataset from Zeller et al. MSB 2014 (see http://msb.embopress.org/content/10/11/766), containing 6 metadata variables variables (e.g. Age or BMI) for 141 samples.

Source

http://msb.embopress.org/content/10/11/766
meta<-

Assign a new sam_data object to x

Description

Assign a new sam_data object to x

Usage

meta(x) <- value

## S4 replacement method for signature 'siamcat,sample_data'
meta(x) <- value

Arguments

x

an object of class siamcat-class

value

an object of class sample_data-class

Value

none

Examples

data(siamcat_example)
meta(siamcat_example) <- meta(siamcat_example)

model.evaluation.plot  Model Evaluation Plot

Description

Produces plots for model evaluation.

Usage

model.evaluation.plot(...)
model.evaluation.plot

Arguments

... one or more object of class siamcat-class, can be named
fn.plot string, filename for the pdf-plot
colours colour specification for the different siamcat-class- objects, defaults to NULL which will cause the colours to be picked from the 'Set1' palette
show.all boolean, Should the results from repeated cross-validation models be plotted? Defaults to FALSE, leading to a single line for the mean across cross-validation repeats
verbose control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Value

Does not return anything, but produces the model evaluation plot.

Binary classification problems

The first plot shows the Receiver Operating Characteristic (ROC)-curve, the other plot the Precision-recall (PR)-curve for the model. If show.all == FALSE (which is the default), a single line representing the mean across cross-validation repeats will be plotted, otherwise the individual cross-validation repeats will be included as lightly shaded lines.

Regression problems

For regression problems, this function will produce a scatter plot between the real and predicted values. If several siamcat-class-objects are supplied, a single plot for each object will be produced.

Examples

data(siamcat_example)

# simple working example
model.evaluation.plot(siamcat_example, fn.plot='.eval.pdf')

# plot several named SIAMCAT object
# although we use only one example object here
model.evaluation.plot('Example_1'=siamcat_example,
  'Example_2'=siamcat_example, colours=c('red', 'blue'),
  fn.plot='.eval.pdf')

# show individual cross-validation repeats
model.evaluation.plot(siamcat_example, fn.plot='.eval.pdf', show.all=TRUE)
model.interpretation.plot

Model Interpretation Plot

Description

This function produces a plot for model interpretation.

Usage

model.interpretation.plot(siamcat, fn.plot = NULL, color.scheme = "BrBG", consens.thres = 0.5, heatmap.type = "zscore", limits = c(-3, 3), log.n0 = 1e-06, max.show = 50, prompt=TRUE, verbose = 1)

Arguments

siamcat object of class siamcat-class
fn.plot string, filename for the pdf-plot
color.scheme color scheme for the heatmap, defaults to 'BrBG'
consens.thres float, minimal ratio of models incorporating a feature in order to include it into the heatmap, defaults to 0.5
Note that for 'randomForest' models, this cutoff specifies the minimum median Gini coefficient for a feature to be included and should therefore be much lower, e.g. 0.01
heatmap.type string, type of the heatmap, can be either 'fc' or 'zscore', defaults to 'zscore'
limits vector, cutoff for extreme values in the heatmap, defaults to c(-3, 3)
log.n0 float, pseudocount to be added before log-transformation of features, defaults to 1e-06
max.show integer, maximum number of features to be shown in the model interpretation plot, defaults to 50
prompt boolean, turn on/off prompting user input when not plotting into a pdf-file, defaults to TRUE
verbose control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details

Produces a plot consisting of

- a barplot showing the feature weights and their robustness (i.e. in what proportion of models have they been incorporated)
- a heatmap showing the z-scores of the metagenomic features across samples
- another heatmap displaying the metadata categories (if applicable)
- a boxplot displaying the proportion of weight per model that is actually shown for the features that are incorporated into more than consens.thres percent of the models.
Value

Does not return anything, but produces the model interpretation plot.

Examples

data(siamcat_example)

# simple working example
siamcat_example <- train.model(siamcat_example, method = 'lasso')
model.interpretation.plot(siamcat_example, fn.plot = './interpretion.pdf',
                           heatmap.type = 'zscore')

models

Retrieve list of trained models from a SIAMCAT object

Description

Function to retrieve the list of trained models

Usage

models(siamcat, verbose=1)

## S4 method for signature 'siamcat'
models(siamcat, verbose = 1)

Arguments

siamcat (Required). An instance of siamcat-class that contains trained models
verbose integer, if the slot is empty, should a message be printed? values can be either 0
(no output) or 1 (print message)

Details

The function extracts the list of trained models.

Value

The list of models or NULL

Examples

data(siamcat_example)
temp <- models(siamcat_example)
temp[1]
Retrieve the information stored in the `model_list` slot within a SIAMCAT object

Description

Function to retrieve the information stored in the `model_list` slot within a SIAMCAT object

Usage

```r
model_list(siamcat, verbose=1)
## S4 method for signature 'siamcat'
model_list(siamcat, verbose = 1)
```

Arguments

- `siamcat` (Required). An instance of `siamcat-class` that contains trained models
- `verbose` integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details

The function will return a list containing the information stored in the `model_list` slot of a SIAMCAT object. This list contains:

- `models` - list of trained models
- `model_type` - machine learning method used for training
- `feature_type` - string describing on which type of features the models were trained

Value

The list stored in the `model_list` slot of the SIAMCAT object or `NULL`

Examples

```r
data(siamcat_example)
temp <- model_list(siamcat_example)
names(temp)
```
Assign a new list containing trained models to a SIAMCAT object

**Description**

Assign a new list containing trained models to a SIAMCAT object

**Usage**

```r
model_list(x) <- value
```

## S4 replacement method for signature 'siamcat,list'
```r
model_list(x) <- value
```

**Arguments**

- `x`: an object of class `siamcat-class`
- `value`: list containing trained models, type of models and of features

**Value**

none

**Examples**

```r
data(siamcat_example)
siamcat_example <- train.model(siamcat_example, method='lasso')
model_list(siamcat_example) <- model_list(siamcat_example)
```

Retrieve the machine learning method from a SIAMCAT object

**Description**

Function to retrieve information on which type of machine learning method was used for model training

**Usage**

```r
model_type(siamcat, verbose=1)
```

## S4 method for signature 'siamcat'
```r
model_type(siamcat, verbose = 1)
```
Arguments

siamcat  (Required). An instance of siamcat-class that contains trained models
verbose  integer, if the slot is empty, should a message be printed? values can be either 0
(no output) or 1 (print message)

Details

The function extracts the information on which type of machine learning method was used for
model training.

Value

The string describing the machine learning method or NULL

Examples

data(siamcat_example)
model_type(siamcat_example)

normalize.features  Perform feature normalization

Description

This function performs feature normalization according to user-specified parameters.

Usage

normalize.features(siamcat, norm.method = c("rank.unit", "rank.std",
"log.std", "log.unit", "log.clr", "std", "pass"),
norm.param = list(log.n0 = 1e-06, sd.min.q = 0.1, n.p = 2, norm.margin = 1),
feature.type='filtered', verbose = 1)

Arguments

siamcat  an object of class siamcat-class
norm.method  string, normalization method, can be one of these: c('rank.unit', 'rank.std',
'log.std', 'log.unit', 'log.clr', 'std', 'pass')
norm.param  list, specifying the parameters of the different normalization methods, see De-
tails for more information
feature.type  string, on which type of features should the function work? Can be either
"original", "filtered", or "normalized". Please only change this paramter
if you know what you are doing!
verbose  integer, control output: 0 for no output at all, 1 for only information about
progress and success, 2 for normal level of information and 3 for full debug
information, defaults to 1
normalize.features

Value

an object of class `siamcat-class` with normalized features

Implemented methods

There are seven different normalization methods available, which might need additional parameters, which are passed via the `norm.param` list:

- `'rank.unit'` - converts features to ranks and normalizes each column (=sample) by the square root of the sum of ranks. This method does not require additional parameters.
- `'rank.std'` - converts features to ranks and applies z-score standardization. This method requires `sd.min.q` (minimum quantile of the standard deviation to be added to all features in order to avoid underestimation of standard deviation) as additional parameter.
- `'log.clr'` - centered log-ratio transformation. This method requires a pseudocount (`log.n0`) before log-transformation.
- `'log.std'` - log-transforms features and applies z-score standardization. This method requires both a pseudocount (`log.n0`) and `sd.min.q`.
- `'log.unit'` - log-transforms features and normalizes by features or samples with different norms. This method requires a pseudocount (`log.n0`) and then additionally the parameters `norm.maring` (margin over which to normalize, similarly to the `apply`-syntax: Allowed values are 1 for normalization over features, 2 over samples, and 3 for normalization by the global maximum) and the parameter `n.p` (vector norm to be used, can be either 1 for `x/sum(x)` or 2 for `x/sqrt(sum(x^2))`.
- `'std'` - z-score standardization without any other transformation. This method only requires the `sd.min.q` parameter.
- `'pass'` - pass-through normalization will not change the features.

Frozen normalization

The function additionally allows to perform a frozen normalization on a different dataset. After normalizing the first dataset, the `norm_feat` slot in the SIAMCAT object contains all parameters of the normalization, which you can access via the `norm_params` accessor.

In order to perform a frozen normalization of a new dataset, you can run the function supplying the normalization parameters as argument to `norm.param`: `norm.param=norm_params(siamcat_reference)`.

See also the example below.

Examples

```r
# Example data
data(siamcat_example)

# Simple example
siamcat_norm <- normalize.features(siamcat_example, norm.method='rank.unit')

# log.unit example
siamcat_norm <- normalize.features(siamcat_example, norm.method='log.unit'),
```
norm.feat

Retrieve the information stored in the norm.feat slot within a SIAMCAT object

Description

Function to retrieve the information stored in the norm.feat slot within a SIAMCAT object

Usage

norm.feat(siamcat, verbose=1)

## S4 method for signature 'siamcat'
norm.feat(siamcat, verbose = 1)

Arguments

siamcat (Required). An instance of siamcat-class that contains normalized features
verbose integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details

The function will return a list containing the information stored in the norm.feat slot of a SIAMCAT object. This list contains:

- norm.feat - normalized features as matrix, see get.norm.feat.matrix
- norm.param - parameters used for normalization, see normalize.features

Value

The list stored in the norm.feat slot of the SIAMCAT object or NULL
Examples

data(siamcat_example)
temp <- norm_feat(siamcat_example)
names(temp)

---

norm_feat<-  

Assign a new list containing normalization parameters and normalized features to a SIAMCAT object

Description

Assign a new list containing normalization parameters and normalized features to a SIAMCAT object

Usage

norm_feat(x) <- value

## S4 replacement method for signature 'siamcat,list'
norm_feat(x) <- value

Arguments

x  an object of class `siamcat-class`
value  a list containing normalization parameters and features

Value

none

Examples

data(siamcat_example)
norm_feat(siamcat_example) <- norm_feat(siamcat_example)

---

norm_params

Retrieve the list of parameters for feature normalization from a SIAMCAT object

Description

Function to retrieve the list of parameters for feature normalization
Usage

norm_params(siamcat, verbose = 1)

## S4 method for signature 'siamcat'
norm_params(siamcat, verbose = 1)

Arguments

siamcat (Required). An instance of siamcat-class containing normalized features
verbose integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details

The function returns the list of normalization parameters used in the feature normalization procedure. See normalize.features for more details.

Value

A list of normalization parameters or NULL

Examples

```r
data(siamcat_example)
temp <- norm_params(siamcat_example)
names(temp)
```

Description

Retrieve a otu_table-class object from otu_table slot in the phyloseq slot in a siamcat object

Usage

orig_feat(siamcat)

## S4 method for signature 'siamcat'
orig_feat(siamcat)

## S4 method for signature 'otu_table'
orig_feat(siamcat)
Arguments

siamcat (Required). An instance of `siamcat-class` that contains a label or instance of `otu_table-class`.

Value

The `otu_table-class` object or NULL.

Examples

data(siamcat_example)
temp <- orig_feat(siamcat_example)

---

Assign a new `otu_table` object to x `orig_feat` slot

Description

Assign a new `otu_table` object to x `orig_feat` slot

Usage

`orig_feat(x) <- value`

```
## S4 replacement method for signature 'siamcat,otu_table'
orig_feat(x) <- value
```

Arguments

x an object of class `siamcat-class`

value an object of class `otu_table-class`

Value

none

Examples

data(siamcat_example)
orig_feat(siamcat_example) <- orig_feat(siamcat_example)
### Parse label header

**Description**

This function parses the header of a label file.

**Usage**

```
parse.label.header(label.header)
```

**Arguments**

- `label.header` - string in the format: 
  `#<TYPE>:<L1>=<class1>; <L2>=<class2>[;<L3>=<class3>]`

  where `<TYPE>` is a string specifying the type of label variable such as BINARY (for binary classification), CATEGORICAL (for multi-class classification), or CONTINUOUS (for regression) `<L1>` is a short numeric label for the first class with description `<class1>` (similarly for the other classes).

**Value**

a list with tow items:

- `$type` type of the label: BINARY, CONTINUOUS or CATEGORICAL
- `$class.descr` labels and information on what they mean

### Retrieve phyloseq-class object from object.

**Description**

Retrieve a phyloseq-class object from object.

**Usage**

```
physeq(siamcat, verbose=1)
```

```r
## S4 method for signature 'ANY'
physeq(siamcat, verbose = 1)
```

```r
## S4 method for signature 'phyloseq'
physeq(siamcat)
```
**physeq<-**

**Arguments**

- **siamcat** (Required). An instance of `siamcat-class` that contains a label or instance of `phyloseq-class`.
- **verbose** If the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

**Value**

The `phyloseq-class` object or NULL.

**Examples**

```r
data(siamcat_example)
physeq(siamcat_example)
```

---

---

**physeq<-**  
*Assign a new phyloseq object to x*

**Description**

Assign a new phyloseq object to x

**Usage**

```r
physeq(x) <- value
```

```r
## S4 replacement method for signature 'siamcat,phyloseq'
physeq(x) <- value
```

**Arguments**

- **x** an object of class `siamcat-class`
- **value** an object of class `phyloseq-class`

**Value**

none

**Examples**

```r
data(siamcat_example)
physeq(siamcat_example) <- physeq(siamcat_example)
```
pred_matrix <- Retrieve the prediction matrix from a SIAMCAT object

Description
Function to retrieve the prediction matrix from a SIAMCAT object

Usage
pred_matrix(siamcat, verbose = 1)

## S4 method for signature 'siamcat'
pred_matrix(siamcat, verbose = 1)

Arguments
siamcat (Required). A siamcat-class object that contains a prediction matrix
verbose integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

Details
The functions returns a matrix containing the predictions for all samples across the different cross-validation repeats. See make.predictions for more information.

Value
A matrix containing predictions or NULL

Examples
data(siamcat_example)
temp <- pred_matrix(siamcat_example)
head(temp)

pred_matrix<- Assign a new matrix with predictions to a SIAMCAT object

Description
Assign a new matrix with predictions to a SIAMCAT object
Usage
pred_matrix(x) <- value

## S4 replacement method for signature 'siamcat,matrix'
pred_matrix(x) <- value

Arguments
- x: an object of class `siamcat-class`
- value: a matrix containing predictions

Value
none

Examples
```r
data(siamcat_example)
pred_matrix(siamcat_example) <- pred_matrix(siamcat_example)
```

---

read.label  Read label file

description
Read label information from a file

Usage
read.label(fn.in.label)

Arguments
- fn.in.label: name of the tsv file containing labels

Details
This function reads in a tsv file with labels and converts it into a label.
First row is expected to be
# BINARY: 1=[label for cases]; -1=[label for controls].
Second row should contain the sample identifiers as tab-separated list (consistent with feature and metadata).
Third row is expected to contain the actual class labels (tab-separated): 1 for each case and -1 for each control.
Note: Labels can take other numeric values (but not characters or strings); importantly, the label for cases has to be greater than the one for controls.
Value

label object containing several entries:

- $\textit{\$label}$ named vector containing the numerical labels from the file;
- $\textit{\$info}$ information about the classes in the label;
- $\textit{\$type}$ information about the label type (e.g. \textit{\textsc{binary}});

Examples

```r
# run with example data
fn.label <- system.file('extdata',
  'label_crc_zeller_msb_mocat_specI.tsv',
  package = 'SIAMCAT')

crc.zeller.label <- read.label(fn.label)
```

---

read.lefse  
\textit{\textbf{read an input file in a LEfSe input format}}

Description

This reads an input file in a LEfSe input format

Usage

```r
read.lefse(filename = "data.txt", rows.meta = 1, row.samples = 2)
```

Arguments

- \textit{\textbf{filename}}  
  name of the input file in a LEfSe input format
- \textit{\textbf{rows.meta}}  
  specifies in which rows medata variables are stored
- \textit{\textbf{row.samples}}  
  specifies in which row sample names are stored

Value

a list with two elements:

- \textit{\textbf{feat}}  
  a features matrix
- \textit{\textbf{meta}}  
  a metadata matrix
select.samples

Examples

fn.in.lefse<- system.file("extdata", "LEfSe_crc_zeller_msb_mocat_specI.tsv", package = "SIAMCAT")
meta.and.features <- read.lefse(fn.in.lefse, rows.meta = 1:6, row.samples = 7)
meta <- meta.and.features$meta
feat <- meta.and.features$feat
label <- create.label(meta=meta, label="label", case = "cancer")
siamcat <- siamcat(feat=feat, label=label, meta=meta)

select.samples

Select samples based on metadata

Description

This function select samples based on information given in the metadata

Usage

select.samples(siamcat, filter, allowed.set = NULL, allowed.range = NULL, verbose = 1)

Arguments

siamcat: an object of class siamcat-class
filter: string, name of the meta variable on which the selection should be done
allowed.set: a vector of allowed values
allowed.range: a range of allowed values
verbose: integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details

This functions selects labels and metadata based on a specific column in the metadata. Provided with a column-name in the metadata and a range or a set of allowed values, the function will filter the siamcat-class object accordingly.

Value

an object of class siamcat-class with labels and metadata filtered in order to contain only allowed values
Examples

data(siamcat_example)

# Select all samples that fall into an Age-range between 25 and 80 years
siamcat_selected <- select.samples(siamcat_example,
   filter = 'Age',
   allowed.range = c(25, 80))

# Select only female samples
siamcat_female <- select.samples(siamcat_example,
   filter = 'Gender',
   allowed.set = c('F'))

show, siamcat-method  Show method for siamcat class object

Description

Show method for siamcat class object

Usage

## S4 method for signature 'siamcat'
show(object)

Value

none

siamcat  SIAMCAT constructor function

Description

Function to construct an object of class siamcat-class

Usage

siamcat(..., feat=NULL, label=NULL, meta=NULL,
   phyloseq=NULL, validate=TRUE, verbose=3)
Arguments

- `feats`: feature information for SIAMCAT (see details)
- `labels`: label information for SIAMCAT (see details)
- `metas`: (optional) metadata information for SIAMCAT (see details)
- `phyloseqs`: (optional) a phyloseq object for the creation of an SIAMCAT object (see details)
- `validate`: boolean, should the newly constructed SIAMCAT object be validated? defaults to TRUE (we strongly recommend against setting this parameter to FALSE)
- `verbose`: control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details

Build siamcat-class objects from their components.

This function creates a SIAMCAT object (see siamcat-class). In order to do so, the function needs

- `feats` the feature information for SIAMCAT, should be either a matrix, a data.frame, or a otu_table-class. The columns should correspond to the different samples (e.g. patients) and the rows the different features (e.g. taxa). Columns and rows should be named.
- `metas` metadata information for the different samples in the feature matrix. Metadata is optional for the SIAMCAT workflow. Should be either a data.frame (with the rownames corresponding to the sample names of the feature matrix) or an object of class sample_data-class
- `phyloseqs` alternatively to supplying both feats and metas, SIAMCAT can also work with a phyloseq object containing an otu_table and other optional slots (like sample_data for metadata).

Notice: do supply **either** the feature information as matrix/data.frame/otu_table (and optionally metadata) or a phyloseq object, but not both.

The label information for SIAMCAT can take several forms:

- `label`: the label object can be created by taking the information in a specific metadata column. In order to do so, `label` should be the name of the column, and `case` should indicate which group(s) should be the positive group(s). A typical example could look like that:

  ```r
  siamcat <- siamcat(feat=feat.matrix, meta=metas, label='DiseaseState', case='CRC')
  ```

  for the construction of a label to predict CRC status (which is encoded in the column "DiseaseState" of the metadata). For more control (e.g. specific labels for plotting or specific control state), the label can also be created outside of the `siamcat` function using the create.label function.

- `named vector`: the label can also be supplied as named vector which encodes the label either as characters (e.g. "Healthy" and "Diseased"), as factor, or numerically (e.g. -1 and 1). The vector must be named with the names of samples (corresponding to the samples in features). Also here, the information about the positive group(s) is needed via the `case` parameter. Internally, the vector is given to the create.label function.

- `label object`: A label object can be created with the create.label function or by reading a dedicated label file with read.label.
siamcat-class

Value

A new siamcat-class object

Examples

# example with package data
data("feat_crc_zeller", package="SIAMCAT")
data("meta_crc_zeller", package="SIAMCAT")

siamcat <- siamcat(feat=feat_crc_zeller,
                   meta=meta_crc_zeller,
                   label='Group',
                   case='CRC')

---

siamcat-class  The S4 SIAMCAT class

Description

The SIAMCAT class

Details

The S4 SIAMCAT class stores the results from the SIAMCAT workflow in different slots. The different slots will be filled by different functions (referenced in the description below).

In order to construct a SIAMCAT class object, please refer to the documentation of the construction function siamcat.

The SIAMCAT class is based on the phyloseq-class. Therefore, you can easily import a phyloseq object into SIAMCAT.

Slots

phyloseq  object of class phyloseq-class

label  list containing the label information for the samples and some metadata about the label, created by create.label or when creating the siamcat-class object by calling siamcat

filt_feat  list containing the filtered features as matrix and the list of filtering parameters, created by calling the filter.features function

associations  list containing the parameters for association testing and the results of association testing with these parameters in a dataframe, created by calling the check.associations function

norm_feat  list containing the normalized features as matrix and the list of normalization parameters (for frozen normalization), created by calling the normalize.features function

data_split  list containing cross-validation instances, created by calling the create.data.split function

model_list  list containing the trained models, the type of model that was trained, and on which kind of features it was trained, created by calling the train.model function
**siamcat.to.lefse**  
create a LEfSe input file from SIAMCAT object

**Description**  
This function creates a LEfSe input file from SIAMCAT object

**Usage**  
siamcat.to.lefse(siamcat, filename = "siamcat_output.txt")

**Arguments**  
- `siamcat`: object of class `siamcat-class`
- `filename`: name of the input file to which data will be save

**Value**  
nothing but data is written to a file

**Examples**  
```r
data(siamcat_example)  
siamcat.to.lefse(siamcat_example)
```

---

**siamcat.to.maaslin**  
create a MaAsLin input file from SIAMCAT object

**Description**  
This function creates a MaAsLin merged PCL single input file from SIAMCAT object

**Usage**  
siamcat.to.maaslin(siamcat, filename = "siamcat_output.pcl")

**Arguments**  
- `siamcat`: object of class `siamcat-class`
- `filename`: name of the input file to which data will be save
Value

nothing but data is written to a file

Examples

data(siamcat_example)
siamcat.to.maaaslin(siamcat_example)

siamcat_example  SIAMCAT example

Description

Reduced version of the CRC dataset from Zeller et al. MSB 2014 (see http://msb.embopress.org/content/10/11/766), containing 100 features (15 associated features at 5% FDR in the original dataset and 85 random other features) and 141 samples, saved after the complete SIAMCAT pipeline has been run.

Thus, the example dataset contains entries in every slot of the SIAMCAT object (see siamcat-class), e.g, eval_data or data_split.

Mainly used for running the examples in the function documentation.

Source

http://msb.embopress.org/content/10/11/766

summarize.features  Summarize features

Description

This function summarize features on a specific taxonomic level

Usage

summarize.features(siamcat, level = 'g__',
feature.type='original', verbose=1)

Arguments

siamcat  object of class siamcat-class
level  string, at which level to summarize (e.g. g__ = genus)
feature.type  string, on which type of features should the function work? Can be either "original", "filtered", or "normalized". Please only change this parameter if you know what you are doing!
verbose  integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1
**train.model**

**Details**

This function will summarize features at different taxonomic levels, e.g. transform species-level relative abundance into genus-level taxonomic profiles.

The function expects a SIAMCAT object that either contains an entry in the `tax_table` slot of its `phyloseq` object, OR a set of feature names which encode taxonomic information, e.g. `k__Bacteria;p__Actinobacteria;c__Actinobacteria;o__Acidimicrobiales;..`

Then, for a given taxonomic level (e.g. `g__`), the function will sum up all the relative abundances of features belonging to the same group at that specific taxonomic level.

**Please note that this function is currently maturing and not necessarily reliable!!!**

**Value**

object of class `siamcat-class` with a summarized feature table

**Examples**

```r
## load the phyloseq example data
data("GlobalPatterns")
## create an example label
label <- create.label(meta=sample_data(GlobalPatterns),
                      label = "SampleType",
                      case = c("Freshwater", "Freshwater (creek)", "Ocean"))
# run the constructor function
siamcat <- siamcat(phyloseq=GlobalPatterns, label=label, verbose=1)
siamcat <- summarize.features(siamcat, level='Genus', verbose=3)
```

---

**train.model  Model training**

**Description**

This function trains a machine learning model on the training data

**Usage**

```r
train.model(siamcat, method = "lasso", measure = "classif.acc",
            param.set = NULL, grid.size=11, min.nonzero=5, perform.fs = FALSE,
            param.fs = list(no_features = 100, method = "AUC", direction="absolute"),
            feature.type='normalized', verbose = 1)
```

**Arguments**

- `siamcat` object of class `siamcat-class`
- `method` string, specifies the type of model to be trained, may be one of these: c('lasso', 'enet', 'ridge', 'lasso_ll', 'ridge_ll','randomForest')
measure character, specifies the model selection criterion during internal cross-validation, see `mlr_measures` for more details, defaults to 'classif.acc'
param.set list, set of extra parameters for mlr, see below for details, defaults to NULL
grid.size integer, grid size for internal tuning (needed for some machine learning methods, for example `lasso_ll`), defaults to 11
min.nonzero integer number of minimum nonzero coefficients that should be present in the model (only for 'lasso', 'ridge', and 'enet'), defaults to 5
perform.fs boolean, should feature selection be performed? Defaults to FALSE
param.fs list, parameters for the feature selection, see Details, defaults to list(thres.fs=100, method.fs="AUC", direction='absolute')
feature.type string, on which type of features should the function work? Can be either "original", "filtered", or "normalized". Please only change this parameter if you know what you are doing!
verbose integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Value

object of class `siamcat-class` with added model_list

Machine learning methods

This function performs the training of the machine learning model and functions as an interface to the mlr3-package.

The function expects a `siamcat-class`-object with a prepared cross-validation (see `create.data.split`) in the data_split-slot of the object. It then trains a model for each fold of the data split. The different machine learning methods are implemented as Learners from the mlr3learners package:

- 'lasso', 'enet', and 'ridge' use the 'classif.cv_glmnet' or 'regr.cv_glmnet' Learners, which interface to the glmnet package,
- 'lasso_ll' and 'ridge_ll' use a custom Learner, which is only available for classification tasks. The underlying package is the LiblineaR package.
- 'randomForest' is implemented via the 'classif.ranger' or regr.ranger Learners available trough the ranger package.

Hyperparameter tuning

There is additional control over the machine learning procedure by supplying information through the param.set parameter within the function. We encourage you to check out the excellent mlr documentation for more in-depth information.

Here is a short overview which parameters you can supply in which form:
- enet The alpha parameter describes the mixture between lasso and ridge penalty and is per default determined using internal cross-validation (the default would be equivalent to param.set=list('alpha'=c(0,1))). You can supply either the limits of the hyperparameter exploration (e.g. with limits 0.2 and 0.8: param.set=list('alpha'=c(0.2,0.8))) or you can supply a fixed alpha value as well (param.set=list('alpha'=0.5)).
- lasso_ll/ridge_ll You can supply both class.weights and the cost parameter (cost of the constraints violation, see LiblineaR for more info). The default values would be equal to param.set=list('class.weight'=c(5,1),'cost'=c(-2,3)).
- randomForest You can supply the two parameters num.trees (Number of trees to grow) and mtry (Number of variables randomly sampled as candidates at each split). See also ranger for more info. The default values correspond to param.set=list('num.trees'=c(100,1000), 'mtry'=c(round(sqrt.mdim / 2), round(sqrt.mdim), round(sqrt.mdim * 2))) with sqrt.mdim=sqrt(nrow(data)).

Feature selection

If feature selection should be performed (for example for functional data with a large number of features), the param.fs list should contain:

- no_features - Number of features to be retained after feature selection,
- method - method for the feature selection, may be AUC, gFC, or Wilcoxon for binary classification problems or spearman, pearson, or MI (mutual information) for regression problems
- direction - indicates if the feature selection should be performed in a single direction only. Can be either
  - absolute - select the top associated features (independent of the sign of enrichment),
  - positive the top positively associated featured (enriched in the case group for binary classification or enriched in higher values for regression),
  - negative the top negatively associated features (inverse of positive)
Direction will be ignored for Wilcoxon and MI.

Examples

data(siamcat_example)

# simple working example
siamcat_example <- train.model(siamcat_example, method='lasso')

validate.data

validate samples in labels, features, and metadata

Description

This function checks if labels are available for all samples in features. Additionally validates metadata, if available.

Usage

validate.data(siamcat, verbose = 1)
Arguments

siamcat an object of class siamcat-class
verbose integer, control output: 0 for no output at all, 1 for only information about progress and success, 2 for normal level of information and 3 for full debug information, defaults to 1

Details

This function validates the data by checking that labels are available for all samples in the feature matrix. Furthermore, the number of samples per class is checked to ensure a minimum number. If metadata is available, the overlap between labels and metadata is checked as well.

This function is run when a siamcat-class object is created.

Value

an object of class siamcat-class

Examples

data(siamcat_example)

# validate.data should be run before completing the pipeline
# since the complete pipeline had been run on siamcat_example, we
# construct a new siamcat object for the example
feat <- orig_feat(siamcat_example)
label <- label(siamcat_example)
siamcat <- siamcat(feat=feat, label=label, validate=FALSE)
siamcat <- validate.data(siamcat, verbose=2)

volcano.plot

volcano.plot

Visualize associations between features and classes as volcano plot

Description

This function creates a volcano plot to visualize the association between features and the label

Usage

volcano.plot(siamcat, fn.plot=NULL, color.scheme="RdYlBu",
annotate=5)

Arguments

siamcat object of class siamcat-class
fn.plot string, filename for the pdf-plot. If fn.plot is NULL, the plot will be produced in the active graphics device.
color.scheme valid R color scheme or vector of valid R colors (must be of the same length as the number of classes), defaults to 'RdYlBu'
annotate integer, number of features to annotate with the name
**weight_matrix**

**Details**

bla bla bla

**Value**

Does not return anything, but produces a volcano plot based on association measures

**Examples**

```r
# Example data
data(siamcat_example)

# Simple example
volcano.plot(siamcat_example, fn.plot='./volcano.pdf')
```

---

**Description**

Function to retrieve the feature weights from a SIAMCAT object

**Usage**

```r
weight_matrix(siamcat, verbose=1)
```

```r
## S4 method for signature 'siamcat'
weight_matrix(siamcat, verbose = 1)
```

**Arguments**

- `siamcat` (Required). An instance of `siamcat-class` that contains trained models
- `verbose` integer, if the slot is empty, should a message be printed? values can be either 0 (no output) or 1 (print message)

**Details**

The function extracts the feature weights from all trained models across all cross-validation folds and repeats.

**Value**

A matrix containing the feature weights or NULL

**Examples**

```r
data(siamcat_example)
temp <- weight_matrix(siamcat_example)
temp[1:3, 1:3]
```
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