

Package ‘SIAMCAT’

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Communities And host phenoTypes

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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
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SIAMCAT-package

SIAMCAT: Statistical Inference of Associations between Microbial Communities And host phenoTypes

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)
```

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	<i>Add metadata as predictors</i>
---------------	-----------------------------------

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 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

Details

This functions 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

```
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	<i>Visualize associations between features and classes</i>
------------------	--

Description

This function visualizes different measures of association between features and the label, computed previously with the [check.associations](#) function

Usage

```
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", "#ffc125"))

```

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<-	<i>Assign a new associations object to x</i>
----------------	--

Description

Assign a new associations object to x

Usage

```
associations(x) <- value

## 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

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

assoc_param	<i>Retrieve the list of parameters for association testing from a SIAMCAT object</i>
-------------	--

Description

Function to retrieve the list of parameters for association testing

Usage

```
assoc_param(siamcat, verbose=1)

## 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.

Value

A list of parameters for association testing or NULL

Examples

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

check.associations	<i>Calculate associations between features and labels</i>
--------------------	---

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.
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

```
# Example data
data(siamcat_example)

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

# Confounder-corrected testing (corrected for Sex)
#
# this is not run during checks
# siamcat_example <- check.associations(siamcat_example,
#   formula='feat~label+(1|Sex)', test='lm')

# Paired testing
#
# this is not run during checks
# siamcat_paired <- check.associations(siamcat_paired,
#   paired='Individual_ID')
```

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 metadata 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!

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
- `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	<i>Create a label list</i>
--------------	----------------------------

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 a negative label (testing one vs rest).
meta	metadata dataframe object or an object of class <code>sample_data-class</code>
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 will be used as negative. If the variable has multiple values, all the values not equal to cases will be used 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)
```

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<-	<i>Assign a new list containing a cross-validation split to a SIAMCAT object</i>
--------------	--

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

```
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:

- `$roc` - average ROC-curve across repeats or a single ROC-curve on complete dataset (see [roc](#));
- `$aucroc` - 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

- `$roc.all` - list of roc objects (see [roc](#)) for every repeat;
- `$aucroc.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:

- `r2` - the mean R squared value across repeats or a single R-squared value on the complete dataset;
- `mae` - them 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 `r2.all`, `mae.all`, and `mse.all`.

Examples

```
data(siamcat_example)

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

<code>eval_data</code>	<i>Retrieve the evaluation metrics from a SIAMCAT object</i>
------------------------	--

Description

Function to retrieve the evaluation metrics from a SIAMCAT object

Usage

```
eval_data(siamcat, verbose=1)

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

Arguments

<code>siamcat</code>	(Required). A siamcat-class object that contains evaluation data
<code>verbose</code>	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 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	<i>Retrieve the feature type used for model training from a SIAMCAT object</i>
--------------	--

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	<i>Retrieve the matrix of feature weights from a SIAMCAT object</i>
-----------------	---

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	<i>Perform unsupervised feature filtering.</i>
-----------------	--

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 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

Details

This function filters the features in a [siamcat-class](#) object in a 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	<i>Filter the label of a SIAMCAT object</i>
--------------	---

Description

This functions filters the label in a SIAMCAT 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 SIAMCAT 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	<i>Retrieve the information stored in the filt_feat slot within a SIAMCAT object</i>
-----------	--

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.fe` - 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

`x` an object of class [siamcat-class](#)

`value` an `filt_feat` object

Value

none

Examples

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

filt_params	<i>Retrieve the list of parameters for feature filtering from a SIAMCAT object</i>
-------------	--

Description

Function to retrieve the list of parameters for feature filtering

Usage

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

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

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.component.classes *Show the component objects classes and slot names.*

Description

Show the component objects classes and slot names.

Usage

```
get.component.classes(class)
```

Value

list of component classes

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.

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]
```

`get.orig_feat.matrix` *Retrieve the original features from a SIAMCAT object*

Description

Function to retrieve the original features from a SIAMCAT object

Usage

```
get.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]
```

`label` *Retrieve the label from a SIAMCAT object*

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` -> `LearnerClassifLiblineaR`

Methods

Public methods:

- `LearnerClassifLiblineaR$new()`
- `LearnerClassifLiblineaR$clone()`

Method `new()`: #' Creates a new instance of this [R6][R6::R6Class] class.

Usage:

```
LearnerClassifLiblineaR$new()
```

Method `clone()`: The objects of this class are cloneable with this method.

Usage:

```
LearnerClassifLiblineaR$clone(deep = FALSE)
```

Arguments:

`deep` Whether to make a deep clone.

make.predictions	<i>Make predictions on a test set</i>
------------------	---------------------------------------

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 <code>siamcat?</code> , 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	<i>Retrieve the metadata from a SIAMCAT object</i>
------	--

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)
```

meta.crc.zeller	<i>Example metadata matrix</i>
-----------------	--------------------------------

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(..., fn.plot = NULL,
  colours=NULL, show.all=FALSE, verbose = 1)
```

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

<code>siamcat</code>	object of class <code>siamcat-class</code>
<code>fn.plot</code>	string, filename for the pdf-plot
<code>color.scheme</code>	color scheme for the heatmap, defaults to 'BrBG'

<code>consens.thres</code>	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
<code>heatmap.type</code>	string, type of the heatmap, can be either 'fc' or 'zscore', defaults to 'zscore'
<code>limits</code>	vector, cutoff for extreme values in the heatmap, defaults to <code>c(-3, 3)</code>
<code>log.n0</code>	float, pseudocount to be added before log-transformation of features, defaults to <code>1e-06</code>
<code>max.show</code>	integer, maximum number of features to be shown in the model interpretation plot, defaults to 50
<code>prompt</code>	boolean, turn on/off prompting user input when not plotting into a pdf-file, defaults to TRUE
<code>verbose</code>	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]]
```

model_list	<i>Retrieve the information stored in the model_list slot within a SIAMCAT object</i>
------------	---

Description

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

Usage

```
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

```
data(siamcat_example)
temp <- model_list(siamcat_example)
names(temp)
```

`model_list<-` *Assign a new list containing trained models to a SIAMCAT object*

Description

Assign a new list containing trained models to a SIAMCAT object

Usage

```
model_list(x) <- value

## S4 replacement method for signature 'siamcat,list'
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

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

model_type	<i>Retrieve the machine learning method from a SIAMCAT object</i>
------------	---

Description

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

Usage

```
model_type(siamcat, verbose=1)

## S4 method for signature 'siamcat'
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	<i>Perform feature normalization</i>
--------------------	--------------------------------------

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 Details 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

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 methods 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

```
# 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.param=list(log.n0=1e-05, n.p=1, norm.margin=1))

# log.std example
siamcat_norm <- normalize.features(siamcat_example,
  norm.method='log.std',
  norm.param=list(log.n0=1e-05, sd.min.q=.1))

# Frozen normalization
# normalize the object siamcat with the same parameters as used in
# siamcat_reference
#
# this is not run
# siamcat_norm <- normalize.features(siamcat,
#   norm.param=norm_params(siamcat_reference))
```

norm_feat

Retrieve the information stored in the norm_feat slot within a SIAM-CAT 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 SIAM-CAT object. This list contains:

- norm.feats - normalized features as matrix, see [get.norm_feats.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)
```

<code>norm_feat<-</code>	<i>Assign a new list containing normalization parameters and normalized features to a SIAMCAT object</i>
-----------------------------	--

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

<code>x</code>	an object of class <code>siamcat-class</code>
<code>value</code>	a list containing normalization parameters and features

Value

none

Examples

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

norm_params	<i>Retrieve the list of parameters for feature normalization from a SIAM-CAT object</i>
-------------	---

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

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

orig_feat	<i>Retrieve a otu_table-class object from otu_table slot in the phyloseq slot in a siamcat object</i>
-----------	---

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)
```

orig_feat<- *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 *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 lables and information on what do they mean

physeq *Retrieve a [phyloseq-class](#) object from object.*

Description

Retrieve a [phyloseq-class](#) object from object.

Usage

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

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

```
## S4 method for signature 'phyloseq'  
physeq(siamcat)
```

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

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

```
physeq<-
```

Assign a new phyloseq object to x

Description

Assign a new phyloseq object to x

Usage

```
physeq(x) <- value

## 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

```
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

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

read.label	<i>Read label file</i>
------------	------------------------

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:

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

Examples

```
# 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	<i>read an input file in a LEfSe input format</i>
------------	---

Description

This reads an input file in a LEfSe input format

Usage

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

Arguments

filename	name of the input file in a LEfSe input format
rows.meta	specifies in which rows metadata variables are stored
row.samples	specifies in which row sample names are stored

Value

a list with two elements:

- feat a features matrix
- meta a metadata matrix

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	<i>Select samples based on metadata</i>
----------------	---

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

...	additional arguments
feat	feature information for SIAMCAT (see details)
label	label information for SIAMCAT (see details)
meta	(optional) metadata information for SIAMCAT (see details)
phyloseq	(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 functions creates a SIAMCAT object (see [siamcat-class](#)). In order to do so, the function needs

- `feat` 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.
- `meta` 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](#)
- `phyloseq` Alternatively to supplying both `feat` and `meta`, SIAMCAT can also work with a `phyloseq` object containing an `otu_table` and other optional slots (like `sample_data` for meta-variables).

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:

- `metadata column`: if there is metadata (either via `meta` or as `sample_data` in the `phyloseq` object), 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:

```
siamcat <- siamcat(feat=feat.matrix, meta=metadata, 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`.

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(feats=feat_crc_zeller,
  meta=meta_crc_zeller,
  label='Group',
  case='CRC')
```

<code>siamcat-class</code>	<i>The S4 SIAMCAT class</i>
----------------------------	-----------------------------

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_feats` 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

pred_matrix matrix of predictions, created by calling the [make.predictions](#) function

eval_data list containing different evaluation metrics, created by calling the [evaluate.predictions](#) 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

```
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.maaslin(siamcat_example)
```

siamcat_example	<i>SIAMCAT example</i>
-----------------	------------------------

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	<i>Summarize features</i>
--------------------	---------------------------

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. <code>g__</code> = genus)
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

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

```
## 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 the a machine learning model on the training data

Usage

```
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

<code>siamcat</code>	object of class <code>siamcat-class</code>
<code>method</code>	string, specifies the type of model to be trained, may be one of these: <code>c('lasso', 'enet', 'ridge', 'lasso_1l', 'ridge_1l', 'randomForest')</code>
<code>measure</code>	character, specifies the model selection criterion during internal cross-validation, see <code>mlr_measures</code> for more details, defaults to <code>'classif.acc'</code>
<code>param.set</code>	list, set of extra parameters for mlr, see below for details, defaults to <code>NULL</code>
<code>grid.size</code>	integer, grid size for internal tuning (needed for some machine learning methods, for example <code>lasso_1l</code>), 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 <code>list(thres.fs=100, method.fs="AUC", direction='absolute')</code>
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 functions 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 through 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.weights'=c(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)`

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')
```

<code>validate.data</code>	<i>Validate samples in labels, features, and metadata</i>
----------------------------	---

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

<code>siamcat</code>	an object of class siamcat-class
<code>verbose</code>	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

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 <code>siamcat-class</code>
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

Details

bla bla bla

Value

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

Examples

```
# Example data
data(siamcat_example)

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

weight_matrix	<i>Retrieve the weight matrix from a SIAMCAT object</i>
---------------	---

Description

Function to retrieve the feature weights from a SIAMCAT object

Usage

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

```
## 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

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


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