

Package ‘coda4microbiome’

December 19, 2022

Title Compositional Data Analysis for Microbiome Studies

Version 0.1.3

Description Functions for microbiome data analysis that take into account its compositional nature. Performs variable selection through penalized regression for both, cross-sectional and longitudinal studies, and for binary and continuous outcomes.

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URL <https://malucalle.github.io/coda4microbiome/>

BugReports <https://github.com/malucalle/coda4microbiome/issues>

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VignetteBuilder knitr

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R topics documented:

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| | |
|-----------------|--|
| coda4microbiome | <i>coda4microbiome: Compositional Data Analysis for Microbiome Studies</i> |
|-----------------|--|

Description

This package provides a set of functions to explore and study microbiome data within the CoDA framework, with a special focus on identification of microbial signatures (variable selection).

| | |
|-------------|--------------------|
| coda_glmnet | <i>coda_glmnet</i> |
|-------------|--------------------|

Description

Microbial signatures in cross-sectional studies. The algorithm performs variable selection through penalized regression on the set of all pairwise log-ratios. The result is expressed as the (weighted) balance between two groups of taxa. It allows the use of non-compositional covariates.

Usage

```
coda_glmnet(
  x,
  y,
  covar = NULL,
  lambda = "lambda.1se",
  nvar = NULL,
  alpha = 0.9,
  nfolds = 10,
  showPlots = TRUE,
  coef_threshold = 0
)
```

Arguments

| | |
|-----------------------------|--|
| <code>x</code> | abundance matrix or data frame (rows are samples, columns are variables (taxa)) |
| <code>y</code> | outcome (binary or continuous); data type: numeric, character or factor vector |
| <code>covar</code> | data frame with covariates (default = NULL) |
| <code>lambda</code> | penalization parameter (default = "lambda.1se") |
| <code>nvar</code> | number of variables to use in the glmnet.fit function (default = NULL) |
| <code>alpha</code> | elastic net parameter (default = 0.9) |
| <code>nfolds</code> | number of folds |
| <code>showPlots</code> | if TRUE, shows the plots (default = TRUE) |
| <code>coef_threshold</code> | coefficient threshold, minimum absolute value of the coefficient for a variable to be included in the model (default =0) |

Value

if `y` is binary: list with "taxa.num", "taxa.name", "log-contrast coefficients", "predictions", "apparent AUC", "mean cv-AUC", "sd cv-AUC", "predictions plot", "signature plot" if not: list with "taxa.num", "taxa.name", "log-contrast coefficients", "predictions", "apparent Rsq", "mean cv-MSE", "sd cv-MSE", "predictions plot", "signature plot"

Author(s)

M. Calle - T. Susin

Examples

```
data(Crohn, package = "coda4microbiome")  
  
set.seed(123)  
  
coda_glmnet(x_Crohn[, (1:10)], y_Crohn, showPlots= FALSE)
```

| | |
|--------------|---------------------|
| coda_glmnet0 | <i>coda_glmnet0</i> |
|--------------|---------------------|

Description

Internal function for the permutational test

Usage

```
coda_glmnet0(  
  x,  
  lrX,  
  idlrX,  
  nameslrX,  
  y,  
  covar = NULL,  
  lambda = "lambda.1se",  
  alpha = 0.9  
)
```

Arguments

| | |
|----------|---|
| x | . |
| lrX | . |
| idlrX | . |
| nameslrX | . |
| y | . |
| covar | . |
| lambda | . |
| alpha | . |

Value

.

Author(s)

M. Calle - T. Susin

```
coda_glmnet_longitudinal  
      coda_glmnet_longitudinal
```

Description

Microbial signatures in longitudinal studies. Identification of a set of microbial taxa whose joint dynamics is associated with the phenotype of interest. The algorithm performs variable selection through penalized regression over the summary of the log-ratio trajectories (AUC). The result is expressed as the (weighted) balance between two groups of taxa.

Usage

```
coda_glmnet_longitudinal(  
  x,  
  y,  
  x_time,  
  subject_id,  
  ini_time,  
  end_time,  
  covar = NULL,  
  lambda = "lambda.1se",  
  nvar = NULL,  
  alpha = 0.9,  
  nfolds = 10,  
  showPlots = TRUE,  
  coef_threshold = 0  
)
```

Arguments

| | |
|----------------|--|
| x | abundance matrix or data frame in long format (several rows per individual) |
| y | outcome (binary); data type: numeric, character or factor vector |
| x_time | observation times |
| subject_id | subject id |
| ini_time | initial time to be analyzed |
| end_time | end time to be analyzed |
| covar | data frame with covariates (default = NULL) |
| lambda | penalization parameter (default = "lambda.1se") |
| nvar | number of variables to use in the glmnet.fit function (default = NULL) |
| alpha | elastic net parameter (default = 0.9) |
| nfolds | number of folds (default = 10) |
| showPlots | if TRUE, shows the plots (default = FALSE) |
| coef_threshold | coefficient threshold, minimum absolute value of the coefficient for a variable to be included in the model (default =0) |

Value

in case of binary outcome: list with "taxa.num", "taxa.name", "log-contrast coefficients", "predictions", "apparent AUC", "mean cv-AUC", "sd cv-AUC", "predictions plot", "signature plot", "trajectories plot"

Author(s)

M. Calle - T. Susin

Examples

```
data(ecam_filtered, package = "coda4microbiome") # load the data

ecam_results<-coda_glmnet_longitudinal (x=x_ecam[, (1:4)], y= metadata$diet,
x_time= metadata$day_of_life, subject_id = metadata$studyid, ini_time=0,
end_time=60, lambda="lambda.min", nfolds=4, showPlots=FALSE)

ecam_results$taxa.num
```

coda_glmnet_longitudinal0
coda_glmnet_longitudinal0

Description

internal function

Usage

```
coda_glmnet_longitudinal0(  
  x,  
  lrX,  
  idlrX,  
  nameslrX,  
  y,  
  x_time,  
  subject_id,  
  ini_time,  
  end_time,  
  covar = NULL,  
  ktop = NULL,  
  lambda = "lambda.1se",  
  alpha = 0.9,  
  nfolds = 10  
)
```

Arguments

| | |
|------------|--|
| x | abundance matrix or data frame in long format (several rows per individual) |
| lrX | log-ratio matrix |
| idlrX | indices table in the log-ratio matrix |
| nameslrX | colnames of the log-ratio matrix |
| y | outcome (binary); data type: numeric, character or factor vector |
| x_time | observation times |
| subject_id | subject id |
| ini_time | initial time to be analyzed |
| end_time | end time to be analyzed |
| covar | data frame with covariates (default = NULL) |
| ktop | given number of selected taxa or compute the best number in case it is NULL (default = NULL) |
| lambda | penalization parameter (default = "lambda.1se") |
| alpha | elastic net parameter (default = 0.9) |
| nfolds | number of folds |

Value

.

Author(s)

M. Calle - T. Susin

`coda_glmnet_longitudinal_null`*coda_glmnet_longitudinal_null*

Description

Performs a permutational test for the `coda_glmnet_longitudinal()` algorithm: It provides the distribution of results under the null hypothesis by implementing the `coda_glmnet_longitudinal()` on different rearrangements of the response variable.

Usage

```
coda_glmnet_longitudinal_null(  
  x,  
  y,  
  x_time,  
  subject_id,  
  ini_time,  
  end_time,  
  niter = 100,  
  covar = NULL,  
  alpha = 0.9,  
  lambda = "lambda.1se",  
  nfold = 10,  
  sig = 0.05  
)
```

Arguments

| | |
|------------|---|
| x | abundance matrix or data frame in long format (several rows per individual) |
| y | outcome (binary); data type: numeric, character or factor vector |
| x_time | observation times |
| subject_id | subject id |
| ini_time | initial time to be analyzed |
| end_time | end time to be analyzed |
| niter | number of sample iterations |
| covar | data frame with covariates (default = NULL) |
| alpha | elastic net parameter (default = 0.9) |
| lambda | penalization parameter (default = "lambda.1se") |
| nfolds | number of folds |
| sig | significance value (default = 0.05) |

Value

list with "accuracy" and "confidence interval"

Author(s)

M. Calle - T. Susin

Examples

```
set.seed(123) # to reproduce the results  
data(ecam_filtered, package = "coda4microbiome") # load the data
```

```

x=x_ecam # microbiome abundance
x_time = metadata$day_of_life # observation times
subject_id = metadata$studyid # subject id
y= metadata$diet # diet ("bd"= breast diet, "fd"=formula diet)
ini_time = 0
end_time = 90

coda_glmnet_longitudinal_null (x,y, x_time, subject_id, ini_time, end_time,
                              lambda="lambda.min",nfolds=4, niter=3)

```

coda_glmnet_null *coda_glmnet_null*

Description

Performs a permutational test for the `coda_glmnet()` algorithm: It provides the distribution of results under the null hypothesis by implementing the `coda_glmnet()` on different rearrangements of the response variable.

Usage

```

coda_glmnet_null(
  x,
  y,
  niter = 100,
  covar = NULL,
  lambda = "lambda.1se",
  alpha = 0.9,
  sig = 0.05
)

```

Arguments

| | |
|--------|---|
| x | abundance matrix or data frame (rows are samples, columns are variables (taxa)) |
| y | outcome (binary or continuous); data type: numeric, character or factor vector |
| niter | number of iterations (default = 100) |
| covar | data frame with covariates (default = NULL) |
| lambda | penalization parameter (default = "lambda.1se") |
| alpha | elastic net parameter (default = 0.9) |
| sig | significance level for the confidence interval (default = 0.05) |

Value

a list with "accuracy" and "confidence interval"

Author(s)

M. Calle - T. Susin

Examples

```
data(Crohn, package = "coda4microbiome")

coda_glmnet_null(x=x_Crohn[, (1:10)], y=y_Crohn, niter=2, covar=NULL, lambda="lambda.1se",
                alpha=0.9, sig=0.05)
```

| | |
|-------|--------------|
| Crohn | <i>Crohn</i> |
|-------|--------------|

Description

Microbiome composition at genus level from a Crohn's disease study: 48 taxa and 975 individuals (662 patients with Crohn's disease and 313 controls)

Format

The dataset contains two objects:

x_Crohn microbiome abundance matrix for 975 individuals (rows) and 48 genera (columns)

y_Crohn a factor, indicating if the sample corresponds to a case (*CD*) or a control (*no*).

References

doi: [10.1016/j.chom.2014.02.005](https://doi.org/10.1016/j.chom.2014.02.005)

| | |
|---------------|----------------------|
| ecam_filtered | <i>ecam_filtered</i> |
|---------------|----------------------|

Description

Microbiome composition at genus level from Early childhood and the microbiome (ECAM) study (Bokulich et al. 2016). Metadata and microbiome data were downloaded from <https://github.com/caporaso-lab/longitudinal-notebooks>. Filtered data contains information on 42 children and 37 taxa.

Format

The dataset contains three objects:

x_ecam microbiome abundance matrix in long format (873 rows) and 37 genera (columns)

taxanames vector containing the taxonomy of the 37 taxa

metadata matrix with information on the individuals at the observation time

References

Bokulich et al. (2016). Antibiotics, birth mode, and diet shape microbiome maturation during early life. *Sci Transl Med* 8:343

explore_logratios *explore_logratios*

Description

Explores the association of each log-ratio with the outcome. Summarizes the importance of each variable (taxa) as the aggregation of the association measures of those log-ratios involving the variable. The output includes a plot of the association of the log-ratio with the outcome where the variables (taxa) are ranked by importance

Usage

```
explore_logratios(
  x,
  y,
  decreasing = TRUE,
  measure = "AUC",
  covar = NULL,
  shownames = FALSE,
  maxrow = 15,
  maxcol = 15,
  showtitle = TRUE,
  mar = c(0, 0, 1, 0)
)
```

Arguments

| | |
|------------|---|
| x | abundance matrix or data frame (rows are samples, columns are variables (taxa)) |
| y | outcome (binary or continuous); data type: numeric, character or factor vector |
| decreasing | order of importance (default = TRUE) |
| measure | association measures "AUC", "Pearson", "Spearman", "glm" (default = "AUC") |
| covar | data frame with covariates (default = NULL) |
| shownames | logical, if TRUE, shows the names of the variables in the rows of the plot (default = FALSE) |
| maxrow | maximum number of rows to display in the plot (default = 15) |
| maxcol | maximum number of columns to display in the plot (default = 15) |
| showtitle | logical, if TRUE, shows the title of the plot (default = TRUE) |
| mar | mar numerical vector of the form c(bottom, left, top, right) which gives the number of lines of margin to be specified on the four sides of the plot (default mar=c(0,0,1,0)) |

Value

list with "max log-ratio", "names max log-ratio", "order of importance", "name of most important variables", "association log-ratio with y" and "top log-ratios plot"

Author(s)

M. Calle - T. Susin

Examples

```
data(HIV, package = "coda4microbiome")  
  
explore_logratios(x_HIV,y_HIV)
```

```
explore_lr_longitudinal  
    explore_lr_longitudinal
```

Description

Explores the association of summary (integral) of each log-ratio trajectory with the outcome. Summarizes the importance of each variable (taxa) as the aggregation of the association measures of those log-ratios involving the variable. The output includes a plot of the association of the log-ratio with the outcome where the variables (taxa) are ranked by importance

Usage

```
explore_lr_longitudinal(  
  x,  
  y,  
  x_time,  
  subject_id,  
  ini_time,  
  end_time,  
  showPlots = FALSE,  
  decreasing = TRUE,  
  covar = NULL,  
  shownames = FALSE,  
  maxrow = 15,  
  maxcol = 15,  
  showtitle = TRUE,  
  mar = c(0, 0, 1, 0)  
)
```

Arguments

| | |
|------------|---|
| x | abundance matrix or data frame in long format (several rows per individual) |
| y | outcome (binary); data type: numeric, character or factor vector |
| x_time | observation times |
| subject_id | subject id |
| ini_time | initial time to be analyzed |
| end_time | end time to be analyzed |
| showPlots | if TRUE, shows the plot (default = FALSE) |
| decreasing | order of importance (default = TRUE) |
| covar | data frame with covariates (default = NULL) |
| shownames | if TRUE, shows the names of the variables in the rows of the plot (default = FALSE) |
| maxrow | maximum number of rows to display in the plot (default = 15) |
| maxcol | maximum number of columns to display in the plot (default = 15) |
| showtitle | logical, if TRUE, shows the title of the plot (default = TRUE) |
| mar | mar numerical vector of the form c(bottom, left, top, right) which gives the number of lines of margin to be specified on the four sides of the plot (default mar=c(0,0,1,0)) |

Value

list with "max log-ratio", "names max log-ratio", "order of importance", "name of most important variables", "association log-ratio with y", "top log-ratios plot"

Author(s)

M. Calle - T. Susin

Examples

```
set.seed(123) # to reproduce the results

data(ecam_filtered, package = "coda4microbiome") # load the data

x=x_ecam # microbiome abundance
x_time = metadata$day_of_life # observation times
subject_id = metadata$studyid # subject id
y= metadata$diet # diet ("bd"= breast diet, "fd"=formula diet)
ini_time = 0
end_time = 90

ecam_logratios<-explore_lr_longitudinal(x,y,x_time,subject_id,ini_time,end_time)
```

`explore_zeros`*explore_zeros*

Description

Provides the proportion of zeros for a pair of variables (taxa) in table `x` and the proportion of samples with zero in both variables. A bar plot with this information is also provided. Results can be stratified by a categorical variable.

Usage

```
explore_zeros(x, id1, id2, strata = NULL)
```

Arguments

| | |
|---------------------|---|
| <code>x</code> | abundance matrix or data frame (rows are samples, columns are variables (taxa)) |
| <code>id1</code> | column number in <code>x</code> for the first taxa |
| <code>id2</code> | column number in <code>x</code> for the second taxa |
| <code>strata</code> | stratification variable (default = NULL) |

Value

a list with the frequency table and the associated bar plot

Author(s)

M. Calle - T. Susin

Examples

```
data(HIV, package = "coda4microbiome")  
explore_zeros(x_HIV,5,6)  
explore_zeros(x_HIV,5,6, strata=y_HIV)
```

filter_longitudinal *filter_longitudinal*

Description

Filters those individuals and taxa with enough longitudinal information

Usage

```
filter_longitudinal(  
  x,  
  taxanames = NULL,  
  x_time,  
  subject_id,  
  metadata,  
  ini_time = min(x_time),  
  end_time = max(x_time),  
  percent_indv = 0.7,  
  min_obs = 3  
)
```

Arguments

| | |
|--------------|---|
| x | abundance matrix or data frame in long format (several rows per individual) |
| taxanames | names of different taxa |
| x_time | observation times |
| subject_id | subject id |
| metadata | matrix sample data |
| ini_time | initial time to be analyzed |
| end_time | end time to be analyzed |
| percent_indv | percentage of individuals with more than min_obs observations |
| min_obs | required minimum number of observations per individual |

Value

list with filtered abundance table, taxanames and metadata

Author(s)

M. Calle - T. Susin

Examples

```
data(ecam_filtered, package = "coda4microbiome") # load the data

x=x_ecam # microbiome abundance
x_time = metadata$day_of_life # observation times
subject_id = metadata$studyid # subject id
ini_time = 0
end_time = 360

data_filtered<-filter_longitudinal(x,taxanames,x_time, subject_id, metadata,
                                   ini_time, end_time, min_obs=4)
```

HIV

HIV

Description

Microbiome abundances (60 taxa and 155 individuals) from an HIV study (Noguera-Julian et al. 2016).

Format

The dataset contains three objects:

x_HIV microbiome abundance matrix for 155 individuals (rows) and 60 genera (columns)

y_HIV a factor, specifying if the individual is HIV positive or (Pos) or negative (Neg).

MSM_HIV a factor, indicating sexual preferences: MSM (*Men who have Sex with Men*) or not (nonMSM).

References

doi: [10.1016/j.ebiom.2016.01.032](https://doi.org/10.1016/j.ebiom.2016.01.032)

impute_zeros

impute_zeros

Description

Simple imputation: When the abundance table contains zeros, a positive value is added to all the values in the table. It adds 1 when the minimum of table is larger than 1 (i.e. tables of counts) or it adds half of the minimum value of the table, otherwise.

Usage

```
impute_zeros(x)
```

Arguments

x abundance matrix or data frame (rows are samples, columns are variables (taxa))

Value

x abundance matrix or data frame with zeros substituted by imputed positive values

Author(s)

M. Calle - T. Susin

Examples

```
data(HIV, package = "coda4microbiome")  
x<-impute_zeros(x_HIV)
```

integralFun

integralFun

Description

Integral of the curve trajectory of subject_id in the interval a,b

Usage

```
integralFun(x, y, id, a, b)
```

Arguments

x abundance matrix or data frame in long format (several rows per individual)
y outcome (binary); data type: numeric, character or factor vector
id subjects-ids
a interval initial time
b interval final time

Value

matrix with integrals for each individual (rows) and each taxa (columns)

Author(s)

M. Calle - T. Susin

logratios_matrix *logratios_matrix*

Description

Computes a large matrix with all the log-ratios between pairs of taxa (columns) in the abundance table

Usage

```
logratios_matrix(x)
```

Arguments

x abundance matrix or data frame (rows are samples, columns are variables (taxa))

Value

list with matrix of log-ratios, matrix indicating the pairs of variables involved in each log-ratio, and a matrix indicating the names of the variables involved in each log-ratio.

Author(s)

M. Calle - T. Susin

Examples

```
data(HIV, package = "coda4microbiome")
lrHIV<-logratios_matrix(x_HIV[, (1:4)])
# matrix of log-ratios (first 6 rows and 6 columns):
lrHIV[[1]][1:6,1:6]
# variables involved in each log-ratio
head(lrHIV[[2]])
# names of the variables involved in each log-ratio
head(lrHIV[[3]])
```

| | |
|----------|----------------------|
| metadata | <i>ecam_filtered</i> |
|----------|----------------------|

Description

Microbiome composition at genus level from Early childhood and the microbiome (ECAM) study (Bokulich et al. 2016). Metadata and microbiome data were downloaded from <https://github.com/caporaso-lab/longitudinal-notebooks>. Filtered data contains information on 42 children and 37 taxa.

Format

The dataset contains three objects:

x_ecam microbiome abundance matrix in long format (873 rows) and 37 genera (columns)

taxanames vector containing the taxonomy of the 37 taxa

metadata matrix with information on the individuals at the observation time

References

Bokulich et al. (2016). Antibiotics, birth mode, and diet shape microbiome maturation during early life. *Sci Transl Med* 8:343

| | |
|---------|------------|
| MSM_HIV | <i>HIV</i> |
|---------|------------|

Description

Microbiome abundances (60 taxa and 155 individuals) from an HIV study (Noguera-Julian et al. 2016).

Format

The dataset contains three objects:

x_HIV microbiome abundance matrix for 155 individuals (rows) and 60 genera (columns)

y_HIV a factor, specifying if the individual is HIV positive or (Pos) or negative (Neg).

MSM_HIV a factor, indicating sexual preferences: MSM (*Men who have Sex with Men*) or not (nonMSM).

References

doi: [10.1016/j.ebiom.2016.01.032](https://doi.org/10.1016/j.ebiom.2016.01.032)

plotMedianCurve *plotMedianCurve*

Description

Internal plot function

Usage

```
plotMedianCurve(iNum, iDen, X, Y, x_time, subject_id, ini_time, end_time)
```

Arguments

| | |
|------------|---|
| iNum | . |
| iDen | . |
| X | . |
| Y | . |
| x_time | . |
| subject_id | . |
| ini_time | . |
| end_time | . |

Value

.

Author(s)

M. Calle - T. Susin

plot_prediction *plot_prediction*

Description

Plot of the predictions of a fitted model: Multiple box-plot and density plots for binary outcomes and Regression plot for continuous outcome

Usage

```
plot_prediction(prediction, y, strata = NULL, showPlots = TRUE)
```

Arguments

| | |
|------------|--|
| prediction | the fitted values of predictions for the model |
| y | outcome (binary or continuous); data type: numeric, character or factor vector |
| strata | stratification variable (default = NULL) |
| showPlots | if TRUE, shows the plots (default = TRUE) |

Value

prediction plot

Author(s)

M. Calle - T. Susin

Examples

```
# prediction plot for the log-ratio between columns 3 and 32 on HIV status
data(HIV, package = "coda4microbiome")
x<-impute_zeros(x_HIV)
lr<-log(x[,3])-log(x[,32])
plot_prediction(lr, y_HIV)
```

plot_signature *plot_signature*

Description

Graphical representation of the variables selected and their coefficients

Usage

```
plot_signature(vars, coeff, showPlots = TRUE, varnames = NULL)
```

Arguments

| | |
|-----------|---|
| vars | variables selected |
| coeff | associated coefficients |
| showPlots | if TRUE, shows the plots (default = TRUE) |
| varnames | if TRUE, shows the names of the variables |

Value

bar plot

Author(s)

M. Calle - T. Susin

Examples

```
plot_signature(c(2,10, 3, 15, 4), c(0.8, -0.1, 0.2, -0.6, -0.3))
```

plot_signature_curves *plot_signature_curves*

Description

Graphical representation of the signature trajectories

Usage

```
plot_signature_curves(  
  varNum,  
  coeff,  
  x,  
  y,  
  x_time,  
  subject_id,  
  ini_time,  
  end_time,  
  color = c("chocolate1", "slateblue2"),  
  showLabel = TRUE,  
  location = "bottomright",  
  inset = c(0.01, 0.02),  
  cex = 0.8,  
  y.intersp = 0.7,  
  main_title = NULL  
)
```

Arguments

| | |
|--------|--|
| varNum | column number of the variables (taxa) |
| coeff | coefficients (coefficients must sum-up zero) |
| x | microbiome abundance matrix in long format |

| | |
|-------------------------|--|
| <code>y</code> | binary outcome; data type: numeric, character or factor vector |
| <code>x_time</code> | observation times |
| <code>subject_id</code> | subject id |
| <code>ini_time</code> | initial time to be analyzed |
| <code>end_time</code> | end time to be analyzed |
| <code>color</code> | color graphical parameter |
| <code>showLabel</code> | graphical parameter (see <code>help(label)</code>) |
| <code>location</code> | graphical parameter (see <code>help(label)</code>) |
| <code>inset</code> | graphical parameter (see <code>help(label)</code>) |
| <code>cex</code> | graphical parameter (see <code>help(label)</code>) |
| <code>y.intersp</code> | graphical parameter (see <code>help(label)</code>) |
| <code>main_title</code> | title plot |

Value

trajectories plot

Author(s)

M. Calle - T. Susin

Examples

```
x=matrix(c(2, 3, 4, 1, 2, 5, 10, 20, 15, 30, 40, 12), ncol=2)
x_time = c(0,10,20,1,15, 25)
subject_id = c(1,1,1,2,2,2)
y=c(0,0,0,1,1,1)
plot_signature_curves(varNum=c(1,2), coeff=c(1,-1), x, y,x_time, subject_id,
  ini_time=0, end_time=25)
```

sCD14

sCD14

Description

Microbiome composition (60 taxa and 151 individuals) and inflammatory parameter sCD14 from an HIV study (Noguera-Julian et al. 2016). A dataset containing the number of counts of 60 different genera in a group of 151 samples (including HIV - infected and non - infected patients).

Format

The dataset contains two objects:

x_sCD14 microbiome abundance matrix for 151 individuals (rows) and 60 genera (columns)

y_sCD14 a numeric variable with the value of the inflammation parameter sCD14 for each sample

References

Rivera-Pinto et al. (2018) Balances: a new perspective for microbiome analysis. *mSystems* 3 (4)

| | |
|---------|----------------|
| shannon | <i>shannon</i> |
|---------|----------------|

Description

Shannon information

Usage

shannon(x)

Arguments

x abundance composition (vector)

Value

shannon information

Author(s)

M. Calle - T. Susin

Examples

```
data(HIV, package = "coda4microbiome")
shannon(x_HIV[,1])
```

| | |
|----------------|-----------------------|
| shannon_effnum | <i>shannon_effnum</i> |
|----------------|-----------------------|

Description

Shannon effective number of variables in a composition

Usage

shannon_effnum(x)

Arguments

x abundance composition (vector)

Value

shannon information

Author(s)

M. Calle - T. Susin

Examples

```
data(HIV, package = "coda4microbiome")
shannon_effnum(x_HIV[,])
```

| | |
|-------------|--------------------|
| shannon_sim | <i>shannon_sim</i> |
|-------------|--------------------|

Description

Shannon similarity between two compositions

Usage

```
shannon_sim(x, y)
```

Arguments

x abundance composition (vector)
y abundance composition (vector)

Value

shannon similarity (value between 0 and 1)

Author(s)

M. Calle - T. Susin

Examples

```
data(HIV, package = "coda4microbiome")
shannon_sim(x_HIV[,], x_HIV[2,])
```

| | |
|-----------|----------------------|
| taxanames | <i>ecam_filtered</i> |
|-----------|----------------------|

Description

Microbiome composition at genus level from Early childhood and the microbiome (ECAM) study (Bokulich et al. 2016). Metadata and microbiome data were downloaded from <https://github.com/caporaso-lab/longitudinal-notebooks>. Filtered data contains information on 42 children and 37 taxa.

Format

The dataset contains three objects:

x_ecam microbiome abundance matrix in long format (873 rows) and 37 genera (columns)

taxanames vector containing the taxonomy of the 37 taxa

metadata matrix with information on the individuals at the observation time

References

Bokulich et al. (2016). Antibiotics, birth mode, and diet shape microbiome maturation during early life. *Sci Transl Med* 8:343

| | |
|---------|--------------|
| x_Crohn | <i>Crohn</i> |
|---------|--------------|

Description

Microbiome composition at genus level from a Crohn's disease study: 48 taxa and 975 individuals (662 patients with Crohn's disease and 313 controls)

Format

The dataset contains two objects:

x_Crohn microbiome abundance matrix for 975 individuals (rows) and 48 genera (columns)

y_Crohn a factor, indicating if the sample corresponds to a case (*CD*) or a control (*no*).

References

doi: [10.1016/j.chom.2014.02.005](https://doi.org/10.1016/j.chom.2014.02.005)

| | |
|--------|----------------------|
| x_ecam | <i>ecam_filtered</i> |
|--------|----------------------|

Description

Microbiome composition at genus level from Early childhood and the microbiome (ECAM) study (Bokulich et al. 2016). Metadata and microbiome data were downloaded from <https://github.com/caporaso-lab/longitudinal-notebooks>. Filtered data contains information on 42 children and 37 taxa.

Format

The dataset contains three objects:

x_ecam microbiome abundance matrix in long format (873 rows) and 37 genera (columns)

taxanames vector containing the taxonomy of the 37 taxa

metadata matrix with information on the individuals at the observation time

References

Bokulich et al. (2016). Antibiotics, birth mode, and diet shape microbiome maturation during early life. *Sci Transl Med* 8:343

| | |
|-------|------------|
| x_HIV | <i>HIV</i> |
|-------|------------|

Description

Microbiome abundances (60 taxa and 155 individuals) from an HIV study (Noguera-Julian et al. 2016).

Format

The dataset contains three objects:

x_HIV microbiome abundance matrix for 155 individuals (rows) and 60 genera (columns)

y_HIV a factor, specifying if the individual is HIV positive or (Pos) or negative (Neg).

MSM_HIV a factor, indicating sexual preferences: MSM (*Men who have Sex with Men*) or not (nonMSM).

References

doi: [10.1016/j.ebiom.2016.01.032](https://doi.org/10.1016/j.ebiom.2016.01.032)

x_sCD14

sCD14

Description

Microbiome composition (60 taxa and 151 individuals) and inflammatory parameter sCD14 from an HIV study (Noguera-Julian et al. 2016). A dataset containing the number of counts of 60 different genera in a group of 151 samples (including HIV - infected and non - infected patients).

Format

The dataset contains two objects:

x_sCD14 microbiome abundance matrix for 151 individuals (rows) and 60 genera (columns)

y_sCD14 a numeric variable with the value of the inflammation parameter sCD14 for each sample

References

Rivera-Pinto et al. (2018) Balances: a new perspective for microbiome analysis. *mSystems* 3 (4)

y_Crohn

Crohn

Description

Microbiome composition at genus level from a Crohn's disease study: 48 taxa and 975 individuals (662 patients with Crohn's disease and 313 controls)

Format

The dataset contains two objects:

x_Crohn microbiome abundance matrix for 975 individuals (rows) and 48 genera (columns)

y_Crohn a factor, indicating if the sample corresponds to a case (*CD*) or a control (*no*).

References

doi: [10.1016/j.chom.2014.02.005](https://doi.org/10.1016/j.chom.2014.02.005)

y_HIV

HIV

Description

Microbiome abundances (60 taxa and 155 individuals) from an HIV study (Noguera-Julian et al. 2016).

Format

The dataset contains three objects:

x_HIV microbiome abundance matrix for 155 individuals (rows) and 60 genera (columns)

y_HIV a factor, specifying if the individual is HIV positive or (Pos) or negative (Neg).

MSM_HIV a factor, indicating sexual preferences: MSM (*Men who have Sex with Men*) or not (nonMSM).

References

doi: [10.1016/j.ebiom.2016.01.032](https://doi.org/10.1016/j.ebiom.2016.01.032)

y_sCD14

sCD14

Description

Microbiome composition (60 taxa and 151 individuals) and inflammatory parameter sCD14 from an HIV study (Noguera-Julian et al. 2016). A dataset containing the number of counts of 60 different genera in a group of 151 samples (including HIV - infected and non - infected patients).

Format

The dataset contains two objects:

x_sCD14 microbiome abundance matrix for 151 individuals (rows) and 60 genera (columns)

y_sCD14 a numeric variable with the value of the inflammation parameter sCD14 for each sample

References

Rivera-Pinto et al. (2018) Balances: a new perspective for microbiome analysis. *mSystems* 3 (4)

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