

Package ‘coda4microbiome’

March 31, 2022

Title Compositional Data Analysis for Microbiome Studies

Version 0.1.1

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.

License MIT + file LICENSE

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

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

Encoding UTF-8

RoxygenNote 7.1.2

LazyData false

Imports corrplot, glmnet, plyr, pROC, ggpubr, ggplot2, qpdf

Suggests knitr, rmarkdown

VignetteBuilder knitr

NeedsCompilation no

Author Malu Calle [aut] (<<https://orcid.org/0000-0001-9334-415X>>),
Toni Susin [aut, cre] (<<https://orcid.org/0000-0002-0874-2784>>)

Maintainer Toni Susin <toni.susin@upc.edu>

Depends R (>= 3.5.0)

Repository CRAN

Date/Publication 2022-03-31 15:30:02 UTC

R topics documented:

coda4microbiome	2
coda_glmnet	3
coda_glmnet0	4
coda_glmnet_longitudinal	5
coda_glmnet_longitudinal0	6

coda_glmnet_longitudinal_null	7
coda_glmnet_null	9
Crohn	10
ecam_filtered	10
explore_logratios	11
explore_lr_longitudinal	12
explore_zeros	14
filter_longitudinal	15
HIV	16
impute_zeros	17
integralFun	17
logratios_matrix	18
metadata	19
MSM_HIV	20
plotMedianCurve	20
plot_prediction	21
plot_signature	22
plot_signature_curves	22
sCD14	24
shannon	24
shannon_effnum	25
shannon_sim	26
taxanames	26
x_Crohn	27
x_ecam	27
x_HIV	28
x_sCD14	28
y_Crohn	29
y_HIV	29
y_sCD14	30
Index	31

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
)
```

Arguments

x	abundance table (rows are samples, columns are variables (taxa))
y	outcome (binary or continuous)
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
showPlots	if TRUE, shows the plots (default = TRUE)

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  
)
```

Arguments

x	abundance table in long format (several rows per individual)
y	outcome (binary)
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
showPlots	if TRUE, shows the plots (default = FALSE)

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 table 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)
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",  
  nfolds = 10,  
  sig = 0.05  
)
```

Arguments

x	abundance table in long format (several rows per individual)
y	outcome (binary)
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 table (rows are samples, columns are variables (taxa))
y	outcome (binary or continuous)
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 table (rows are samples, columns are variables (taxa))
y	outcome (binary or continuous)
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 table in long format (several rows per individual)
y	outcome (binary)
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 table (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,  
  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 table in long format (several rows per individual)
taxanames	names of different taxa
x_time	observation times
subject_id	subject id
metadata	metadata
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
y= metadata$diet # diet ("bd"= breast diet, "fd"=formula diet)
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	<i>impute_zeros</i>
--------------	---------------------

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 table (rows are samples, columns are variables (taxa))

Value

x abundance table with zeros substituted by imputed positive values

Author(s)

M. Calle - T. Susin

Examples

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

integralFun	<i>integralFun</i>
-------------	--------------------

Description

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

Usage

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

Arguments

x	abundance table in long format (several rows per individual)
y	outcome (binary)
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 table (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

ecam_filtered

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	<i>plotMedianCurve</i>
-----------------	------------------------

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	<i>plot_prediction</i>
-----------------	------------------------

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)
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)
```

Arguments

vars	variables selected
coeff	associated coefficients
showPlots	if TRUE, shows the plots (default = TRUE)

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  
)
```

Arguments

varNum	column number of the variables (taxa)
coeff	coefficients (coefficients must sum-up zero)
X	microbiome abundance matrix in long format
Y	binary outcome
x_time	observation times
subject_id	subject id
ini_time	initial time to be analyzed
end_time	end time to be analyzed

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

shannon

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[,])
```

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[1,],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

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)

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)

Index

* data

- Crohn, [10](#)
 - ecam_filtered, [10](#)
 - HIV, [16](#)
 - metadata, [19](#)
 - MSM_HIV, [20](#)
 - sCD14, [24](#)
 - taxanames, [26](#)
 - x_Crohn, [27](#)
 - x_ecam, [27](#)
 - x_HIV, [28](#)
 - x_sCD14, [28](#)
 - y_Crohn, [29](#)
 - y_HIV, [29](#)
 - y_sCD14, [30](#)
- [coda4microbiome, 2](#)
- [coda_glmnet, 3](#)
- [coda_glmnet0, 4](#)
- [coda_glmnet_longitudinal, 5](#)
- [coda_glmnet_longitudinal0, 6](#)
- [coda_glmnet_longitudinal_null, 7](#)
- [coda_glmnet_null, 9](#)
- [Crohn, 10](#)
-
- [ecam_filtered, 10](#)
- [explore_logratios, 11](#)
- [explore_lr_longitudinal, 12](#)
- [explore_zeros, 14](#)
-
- [filter_longitudinal, 15](#)
-
- [HIV, 16](#)
-
- [impute_zeros, 17](#)
- [integralFun, 17](#)
-
- [logratios_matrix, 18](#)
-
- [metadata, 19](#)
- [MSM_HIV, 20](#)
-
- [plot_prediction, 21](#)
- [plot_signature, 22](#)
- [plot_signature_curves, 22](#)
- [plotMedianCurve, 20](#)
-
- [sCD14, 24](#)
- [shannon, 24](#)
- [shannon_effnum, 25](#)
- [shannon_sim, 26](#)
-
- [taxanames, 26](#)
-
- [x_Crohn, 27](#)
- [x_ecam, 27](#)
- [x_HIV, 28](#)
- [x_sCD14, 28](#)
-
- [y_Crohn, 29](#)
- [y_HIV, 29](#)
- [y_sCD14, 30](#)