# Package 'BRACoD.R'

July 21, 2025

Version 0.0.2.0

Description The goal of this method is to identify associations between bacteria and an environmental variable in 16S or other compositional data. The environmental variable is any variable which is measure for each microbiome sample, for example, a butyrate measurement paired with every sample in the data. Microbiome data is compositional, mean-

ing that the total abundance of each sample sums to 1, and this introduces severe statistical distortions. This method takes a Bayesian approach to correcting for these statistical distortions, in which the total abundance is treated as an unknown variable. This package runs the python implementation using reticulate

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Title BRACoD: Bayesian Regression Analysis of Compositional Data

Imports reticulate
<b>Config/reticulate</b> list( packages = list( list(package = ``BRACoD") ) )
License MIT + file LICENSE
Encoding UTF-8
LazyData true
RoxygenNote 7.1.1.9001
Suggests testthat (>= 3.0.0)
Config/testthat/edition 3
NeedsCompilation no
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<b>Depends</b> R (>= 3.5.0)
Repository CRAN
<b>Date/Publication</b> 2022-03-24 15:10:07 UTC

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convergence\_tests

Perform convergence tests on the p and beta variables

## **Description**

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You may get errors are divergence of some variables after pymc3 samples the posterior. We are not overly concerned about some of the variables, such as the variance, rather we are really interested in the inclusion probabilities (p) and contribution coefficients (beta). The convergence tests that are included here focus on evaluating those two variables.

## Usage

```
convergence_tests(trace, df_relab)
```

# **Arguments**

trace the output of run\_bracod()

df\_relab the microbiome relative abundance

## Value

no return value

install\_bracod

Install BRACoD in python

# **Description**

Uses pip to install the latest BRACoD release in python. You might need to specify a python environment with either reticulate::use\_virtualenv or reticulate::use\_condaenv.

# Usage

```
install_bracod(method = "auto", conda = "auto")
```

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## **Arguments**

method passed to reticulate::py\_install conda passed to reticulate::py\_install

#### Value

no return value

obesity

Example microbiome data

# **Description**

This data is mouse stool microbiome data from a study of obesity.

# Usage

```
data(obesity)
df_scfa
```

## **Format**

a DataFrame of 16S microbiome counts, and a dataframe with corresponding butyrate measurements

An object of class data. frame with 119 rows and 1 columns.

remove\_null

Remove NULL values in your OTU and environmental variable

# Description

This will remove samples that are NULL in the environmental variable, as well as the corresponding samples in your relative abundance data.

## Usage

```
remove_null(df_relab, Y)
```

## **Arguments**

df\_relab microbiome relative abundance data in a dataframe
Y values of the environmental variable

## Value

a list containing 1) the relative abundance data and 2) the Y values

run\_bracod

run_bracod	Run the main BRACoD algorithm

# Description

Uses pymc3 to sample the posterior of the model to determine bacteria that are associated with your environmental variable.

# Usage

```
run_bracod(df_relab, env_var, n_sample = 1000, n_burn = 1000, njobs = 4)
```

# **Arguments**

df_relab	A dataframe of relative microbiome abundances. Samples are rows and bacteria are columns.
env_var	the environmental variable you are evaluating. You need 1 measurement associated with each sample.
n_sample	number of posterior samples.
n_burn	number of burn-in steps before actual sampling stops.
njobs	number of parallel MCMC chains to run.

# Value

the pymc trace object which holds the samples of the posterior distribution

# **Examples**

```
## Not run:
data(obesity)
r <- simulate_microbiome_counts(obesity)
sim_counts <- r[[1]]
sim_y <- r[[2]]
contributions <- r[[3]]
sim_relab <- scale_counts(sim_counts)
trace <- run_bracod(sim_relab, sim_y, n_sample = 1000, n_burn=1000, njobs=4)
## End(Not run)</pre>
```

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scale_counts Normalize OTU counts and add a pseudo count
--

# Description

BRACoD requires relative abundance and cannot handle zeros, so this function adds a small pseudo count (1/10th the smallest non-zero value).

# Usage

```
scale_counts(df_counts)
```

# **Arguments**

df\_counts

A dataframe of OTU counts. Samples are rows and bacteria are columns.

#### Value

a dataframe of relative abundance data

score

Score the results of BRACoD

# Description

This calculate the precision, recall and F1 of your BRACoD results if you know the ground truth, ie. if this is simulated data.

# Usage

```
score(taxon_identified, taxon_actual)
```

## **Arguments**

taxon\_identified

a list of integers corresponding to the indicies of the taxon you identified with BRACoD

taxon\_actual

a list of integers corresponding to the indicies of the taxon that truely contribute to butyrate levels

#### Value

a list containing 1) the precision 2) the recall 3) the f1 metric

## **Examples**

```
## Not run:
df_summary <- summarize_trace(trace, colnames(sim_relab))
taxon_identified <- df_summary$taxon
taxon_actual <- which(contributions != 0)

r <- score(taxon_identified, taxon_actual)

precision <- r[[1]]
recall <- r[[2]]
f1 <- r[[3]]

print(sprintf("Precision: %.2f, Recall: %.2f, F1: %.2f",precision, recall, f1))
## End(Not run)</pre>
```

simulate\_microbiome\_counts

Simulate microbiome counts

# **Description**

Each bacteria's absolute abundance is simulated from a lognormal distribution. Then, convert each sample to relative abundance, and simulate sequencing counts using a multinomial distribution, based on the desired number of reads and the simulated relative abundances. This also simulates an environmental variable that is produced by some of the bacteria.

# Usage

```
simulate_microbiome_counts(
    df,
    n_contributors = 20,
    coeff_contributor = 0,
    min_ab_contributor = -9,
    sd_Y = 1,
    n_reads = 1e+05,
    var_contributor = 5,
    use_uniform = TRUE,
    n_samples_use = NULL,
    corr_value = NULL,
    return_absolute = FALSE,
    seed = NULL
)
```

#### **Arguments**

df

A dataframe of OTU counts that is a model for data simulation. Samples are rows and bacteria are columns.

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n\_contributors the number of bacteria that are to contribute to your environmental variable. coeff\_contributor

the average of the distribution used to simulate the contribution coefficient.

min\_ab\_contributor

The minimum log relative abundance, averaged across samples, to include a

acteria

sd\_Y the standard deviation of the simulated environmental variable

n\_reads the number of reads to be simulated per sample

var\_contributor

If you use a uniform distribution, this is the range of the distribution, with a normal distribution it is the variance used to simulate the contribution coefficient.

use\_uniform use a uniform distribution to simulate the contribution coefficient. Alternative is

the normal distribution.

n\_samples\_use number of microbiome samples to simulate. If NULL, uses the same number of

samples as in your dataframe

corr\_value the bacteria-bacteria correlation value you want to include in the simulation

return\_absolute

returns the abosulte abundance values instead of the simulated microbiome counts

seed random seed for reproducibility

#### Value

a list containing 1) the simulated count data 2) the simulated environmental variable and 3) the simulated contribution coefficients

summarize\_trace

Summarize the results of BRACoD

# **Description**

This summarizes the trace object that run\_bracod() returns. It returns a dataframe that contains two parameters of interest, the average inclusion (p) and the average coefficient (beta), telling you the association between that bacteria and the environmental variable

# Usage

```
summarize_trace(trace, taxon_names = NULL, cutoff = 0.3)
```

#### **Arguments**

trace the pymc3 object that is the output of run\_bracod()

taxon\_names optional, a list of names of the bacteria to include in the results

cutoff this is the cutoff on the average inclusion for inclusion. We recommend a value

of 0.3, but you can lower the value to include less confident taxon or raise the

cutoff to exclude them.

threshold\_count\_data

## Value

a dataframe with information about the bacteria that BRACoD identified

# **Examples**

```
## Not run:
trace <- run_bracod(sim_relab, sim_y, n_sample = 1000, n_burn=1000, njobs=4)
df_summary <- summarize_trace(trace, colnames(sim_relab))
## End(Not run)</pre>
```

## **Description**

This function removes samples below a minimum counts and bacteria below a minimum log abundance. Run this before running BRACoD because the algorithm does not perform well when there are many low abundance bacteria that are only present in a few samples.

## Usage

```
threshold_count_data(df_counts, min_counts = 1000, min_ab = 1e-04)
```

## **Arguments**

df\_counts A dataframe of OTU counts. Samples are rows and bacteria are columns.

min\_counts threshold samples with fewer than this many counts

min\_ab threshold bacteria whose average log abundance is below this

## Value

a dataframe of microbiome counts

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