# Package 'CoDaLoMic'

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Author Irene Creus Martí [aut, cre] ( <a href="https://orcid.org/0000-0002-7962-4478">https://orcid.org/0000-0002-7962-4478</a> )
Maintainer Irene Creus Martí <ircrmar@mat.upv.es></ircrmar@mat.upv.es>
<b>Description</b> Implementation of models to analyse compositional microbiome time series taking into account the interaction between groups of bacteria. The models implemented are described in Creus-Martí et al (2018, ISBN:978-84-09-07541-6), Creus-Martí et al (2021) <doi:10.1155 2021="" 9951817=""> and Creus-Martí et al (2022) <doi:10.1155 2022="" 4907527="">.</doi:10.1155></doi:10.1155>
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CoDaLoMic

#### **Description**

This package contain functions to model compositional and longitudinal microbiome datasets. This package contains the functions needed to execute the models published in the articles:

- Creus-Martí, I., Moya, A., Santonja, F. J. (2018), A Statistical Model with a Lotka-Volterra Structure for Microbiota Data, Lucas Jodar, Juan Carlos Cortes and Luis Acedo, Modelling for engineering and human behavior 2018, Instituto Universitario de Matematica Multidisciplinar, ISBN: 978-84-09-07541-6.
- Creus-Martí, I., Moya, A., Santonja, F. J. (2021). A Dirichlet autoregressive model for the analysis of microbiota time-series data. Complexity, 2021, 1-16.
- Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

In addition, the package contains one real dataset extracted from:

Marín-Miret, J., Pérez-Cobas, A. E., Domínguez-Santos, R., Pérez-Rocher, B., Latorre, A., & Moya, A. (2024). Adaptability of the gut microbiota of the German cockroach Blattella germanica to a periodic antibiotic treatment. Microbiological Research, 287, 127863.

#### **Details**

We refer to the model described in Creus-Martí (2018) as Dirich-gLV, we refer to the model described in Creus-Martí (2021) as FBM and we refer to the model described in Creus-Martí (2022) as BPBM.

#### Access to Additional Files

This package includes files in the directory 'inst/extdata'. Users can access these files using 'system.file()'. The following files are available:

- README.pdf, README.Rmd, README.R: Basic instructions for using the package.
- 1-s2.0-S0944501324002647-mmc6.xlsx: Original cockroach dataset extracted from Marín-Miret et al (2024).
- Simulated.R: Code use to obtain the Simulated dataset.
- cockroach. R: Code use to obtain the cockroach dataset.

```
To access these files, use: base::system.file("extdata", "filename", package = "CoDaLoMic")
```

For instance: base::system.file("extdata", "README.pdf", package = "CoDaLoMic")

On Windows, the PDF can be opened as follows: base::shell.exec(system.file("extdata", "README.pdf", package = "CoDaLoMic"))

#### Author(s)

Maintainer: Irene Creus Martí <ircrmar@mat.upv.es> (ORCID)

4 B1MODImatrizP

B1MODImatrizP	Ridge regression matrix	
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#### **Description**

Defining the part of the ridge regression matrix that carries the information of the bacteria especieII.

#### Usage

```
B1MODImatrizP(Tt, especieII, especie, E, EspecieMaxima)
```

## **Arguments**

Tt Number of time points available

especieII Number. The number of the row in which the bacteria that we want to use is

placed in the matrix especie.

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

E Number of bacteria available

EspecieMaxima Row in which the bacteria chosen as reference is in especie. This is the bacteria

that is going to be at the denominator of the balance and at the denominator of

the alr transformation.

#### Value

Returns a matrix. The first column contain the number 1 repeated Tt times. The second column contains the alr transformation of the especieII in all time points. The third column contains the balance (whose numerator has all the bacteria except especieII and EspecieMaxima and the denominator contains the EspecieMaxima) in all time points.

```
Tt=2
especie1=cbind(c(0.5,0.3,0.2), c(0.1,0.3,0.6))
especieII=1
E=3
EspecieMaxima=3
B1MODImatrizP(Tt, especieII,especie1, E, EspecieMaxima)
```

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|--|

## **Description**

This function calculates the value of the FBM regression, defined by:

```
\mu_{it} = a_{i1} + a_{i2} \cdot \operatorname{alr}(x_{i,(t-1)}) + a_{i3} \cdot \operatorname{Balance}(x_{i,(t-1)}) \text{ for } i = 1, \dots, D-1 \text{ where } D \text{ is the number of bacterial}
```

#### Usage

```
B1MODImodel(A, especie, E, EspecieMaxima, Tt)
```

## **Arguments**

A	Matrix of dimensions $(E-1)x3$ that contains all the parameters of the model except tau
especie	Matrix that contains at row i the bacterial taxa of bacteria i at all time points.
Е	Number of bacteria available
EspecieMaxima	Row in which the bacteria chosen as reference is in especie. This bacteria is used as reference in the alr tranformation that the model does and it is placed at the denominator of the balance)

Tt Number of time points available.

#### Value

Returns a matrix. The row i contains the regression values of the bacteria i at all time points.

# References

Creus-Martí, I., Moya, A., Santonja, F. J. (2021). A Dirichlet autoregressive model for the analysis of microbiota time-series data. Complexity, 2021, 1-16.

```
df<-data.frame(cbind(c(0.1,0.1,0.8),c(0.2,0.1,0.7)))
E=3
EspecieMaxima=3
set.seed(724)
A=matrix(c(-2:3),2,3)
Tt=2
B1MODImodel(A,df, E, EspecieMaxima,Tt)</pre>
```

BPBM\_Matrix

Balance	Calculating the balance of the FBM model

# Description

Defining a balance where we compare all the bacteria (except the one chosen as reference and the especiel) with the one chosen as reference.

# Usage

```
Balance(A, especieI, especie, E, EspecieMaxima)
```

# Arguments

A	Number of time points for which we calculate the balance
especieI	Number. The bacteria that we do not include in the balance. We must write the number of the row in which the bacteria is placed in the matrix especie.
especie	Matrix that contains at row i the bacterial taxa of bacteria i at all time points.
Е	Number of bacteria available
EspecieMaxima	Row in which the bacteria chosen as reference is in especie. This bacteria is used as reference in the alr tranformation that the model does and it is placed at the denominator of the balance)

## Value

Returns a vector with the value of the balance for all the time points indicated.

## **Examples**

```
Balance(2,2,cbind(c(\emptyset.1,\emptyset.1,\emptyset.8),c(\emptyset.2,\emptyset.1,\emptyset.7)),3,3)
```

BPBM_Matrix	Obtains the matrix of covariates of the BPBM	

# Description

This function writes the matrix of covariates of the BPBM.

# Usage

```
BPBM_Matrix(rows.position, PB, Tt)
```

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#### **Arguments**

PΒ

rows.position Vector. Vector with the number of the rows where the SPBal are in the matrix

PB. We write first the row in which the balance with higher variance is placed,

then the row in which the balance with second higher variance is placed...

Matrix. Each line os the matrix PB contains the values of one principal balance

at all time points.

Tt Number of time points available.

#### **Details**

In an example with two SPBal and three time points, the covariates are written in the following order:

$$\begin{array}{cccc} & & & 1 & & 1 \\ SPBal_{1,t-1} & SPBal_{1,t-2} & SPBal_{1,t-3} \\ SPBal_{2,t-1} & SPBal_{2,t-2} & SPBal_{2,t-3} \end{array}$$

#### Value

Returns a matrix with the covariates of the model.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

## **Examples**

```
matt=matrix(c(1:12),3,4)
rows.position=c(2,3)
BPBM_Matrix(rows.position,matt,4)
```

cockroach

cockroach dataset

# Description

Gut microbiome dataset of a Blatella germanica cockroach treated by kanamycin during three periods of time (days: 1–10, 36–45, 71–80). The data is extracted from Marín-Miret et al (2024), more specifically, the data is the information of the K3 cockroach in the article. The dataset contains 105 time points and 210 genera.

#### Usage

data(cockroach)

#### **Format**

A data frame with 105 rows and 211 columns.

#### References

Marín-Miret, J., Pérez-Cobas, A. E., Domínguez-Santos, R., Pérez-Rocher, B., Latorre, A., & Moya, A. (2024). Adaptability of the gut microbiota of the German cockroach Blattella germanica to a periodic antibiotic treatment. Microbiological Research, 287, 127863.

Estimate\_Param\_EstParmFunc

Estimating Parameters of EstParmFunc

#### **Description**

This function calculates the estimated parameters of the Dirich-gLV model.

#### Usage

Estimate\_Param\_EstParmFunc(Iter.EstParmFunc, paramini, especie, seed = NULL)

# **Arguments**

Iter.EstParmFunc

Number. Number of iterations.

paramini Initial values of the parameters. Vector equal to c(tau.ini, as.vector(pam.ini))

where:

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

The bacteria placed in the last row of the matrix will be used as reference in the

alr transformation.

seed Number. Set a seed. Default seed=NULL.

• pam.ini Matrix. Each row has the parameters of each bacteria, following

the same structure than pam in EstParmFunc

• tau.ini Number. Initial value of the tau parameter in the model

# **Details**

Maximum likelihood estimation is used. This function makes an iterative process, it obtains the value of the parameter that maximize the Dirichlet loglikelihood (defined in EstParmFunc) using the Nelder-Mead method and some initial parameters. Then it uses this value as initial parameters and repeats the process Iter.EstParmFunc times.

#### Value

Returns a list with:

- All.iter: Matrix. Each row has the parameters obtained in each iteration. The parameters are
  in the columns written in the same order that they are written in paramini. In this matrix we
  must observe that in the last iterations the values has really similar or equal values, it not, we
  need to increase the value of Iter.EstParmFunc.
- Param.Estimates: The estimated parameters. The parameters are in the columns written in the same order that they are written in paramini.

#### References

Creus-Martí, I. and Moya, A. and Santonja, F. J. (2018). A Statistical Model with a Lotka-Volterra Structure for Microbiota Data. Lucas Jodar, Juan Carlos Cortes and Luis Acedo, Modelling for engineering and human behavior 2018, Instituto Universitario de Matematica Multidisciplinar. ISBN: 978-84-09-07541-6

#### **Examples**

```
\begin{split} & \texttt{especie=cbind}(\texttt{c}(0.5,0.3,0.2),\texttt{c}(0.1,0.3,0.6)) \\ & \texttt{paramini=c}(100,2,3,4,5,6,7) \\ & \texttt{Estimate\_Param\_EstParmFunc}(5, \texttt{paramini}, \texttt{especie},714) \end{split}
```

Estimate\_Param\_FBM

Estimating Parameters of EstParmFunc\_FBM

## **Description**

This function estimates the parameters of the FBM model.

# Usage

```
Estimate_Param_FBM(
  tau,
  ridge.final,
  Iter.EstParmFunc = 80,
  especie,
  EspecieMaxima,
  Tt,
  E,
  seed = NULL
)
```

#### **Arguments**

tau Number. Value of the tau parameter.

ridge.final Object of class "ridgelm". Values obtained with the ridge regression.

Iter.EstParmFunc

Number. Number of iterations. Default: 80 iterations.

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

The bacteria placed in the last row of the matrix will be used as reference in the

alr transformation and will be at the denominator of the balance.

EspecieMaxima Row in which the bacteria chosen as reference is in especie. This is the bacteria

that is going to be at the denominator of the balance and in the deniminator of the alr transformation. As a result, in this function, EspecieMaxima must be equal to

E.

Tt Number of time points available

E Number. Number of bacteria available.
seed Number. Set a seed. Default seed=NULL.

#### **Details**

Maximum likelihood estimation is used. This function makes an iterative process, for a given value of tau, it obtains the value of the rest of the parameters that maximize the dirichlet loglikelihood (defined in EstParmFunc\_FBM) using the Nelder-Mead method and the values obtained in the ridge regression as initial parameters. Then it uses the values obtained as initial parameters and repeats the process Iter.EstParmFunc times.

The regression of this model is defined by

$$\mu_{it} = a_{i1} + a_{i2} \cdot \operatorname{alr}(x_{i,(t-1)}) + a_{i3} \cdot \operatorname{Balance}(x_{i,(t-1)})$$
 for  $i = 1, \dots, D-1$  where D is the number of bacteria

#### Value

Returns a list with:

- All.iter: Matrix. Each row has the parameters obtained in each iteration. The parameters are in the columns written in following order: a11,a12,a13, a21, a22,a23, ...a(D-1)1,a(D-1)2,a(D-1)3,tau. Where D is the number of bacterial species present in the matrix especie. In this matrix we must observe that in the last iterations the values has really similar or equal values, it not, we need to increase the value of Iter.EstParmFunc.
- Param.Estimates: Vector with the estimated parameters, in the following order: a11,a12,a13, a21, a22,a23, ...a(D-1)1,a(D-1)2,a(D-1)3,tau. Where D is the number of bacterial species present in the matrix especie.
- AIC Number: Value of the AIC.

## References

Creus-Martí, I., Moya, A., Santonja, F. J. (2021). A Dirichlet autoregressive model for the analysis of microbiota time-series data. Complexity, 2021, 1-16.

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#### **Examples**

```
set.seed(123)
especie=t(gtools::rdirichlet(5,c(1,3,1)))
Tt=5
E=3
EspecieMaxima=3
ridge.final=ridgeregression(Tt,especie, E, EspecieMaxima)
tau=20
Iter.EstParmFunc=40
Estimate_Param_FBM(tau,ridge.final,Iter.EstParmFunc, especie,EspecieMaxima,Tt,E, 714)
```

Estimating\_BPBM

Estimating BPBM

# Description

The estimation of the BPBM model is carried out using MCMC. To execute this function it is necessary to have the program Just Another Gibbs Sampler (JAGS) (Plummer, 2003) program installed.

# Usage

```
Estimating_BPBM(
   especie,
   Tt,
   E,
   MatrizPBmodelo,
   nn.chain = 3,
   nn.burnin = 5000,
   nn.sample = 20000,
   nn.thin = 10,
   seed = NULL
)
```

# Arguments

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

Tt Number of time points available.

E Number of bacteria in the dataset.

MatrizPBmodelo Matrix with the covariates of the model. In an example with two SPBal and three time points, the covariates are written in the following order:

```
\begin{array}{cccc} & 1 & & 1 & & 1 \\ SPBal_{1,t-1} & SPBal_{1,t-2} & SPBal_{1,t-3} \\ SPBal_{2,t-1} & SPBal_{2,t-2} & SPBal_{2,t-3} \end{array}
```

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nn.chain	the number of chains to use with the simulation. Default is 3, minimum2.
nn.burnin	the number of burnin iterations. Default is 5000.
nn.sample	the number of iterations to take. Default: 20000. The markov chain will have ("sample"-"burnin")/"thin" iterations.
nn.thin	the thinning interval to be used. Default: 10.
seed	Number. Set a seed. Default seed=NULL.

#### Value

Returns a list with:

- List with:
  - R2jagsOutput: R2jags object with the information of the estimation.
  - Samples All Chains: Matrix. Matrix that has the iterations of all the Markov chains joined.

#### References

- Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.
- Plummer, M. (2003, March). JAGS: A program for analysis of Bayesian graphical models using Gibbs sampling. In Proceedings of the 3rd international workshop on distributed statistical computing (Vol. 124, No. 125.10, pp. 1-10).

EstParmFunc 13

Writting the loglikelihood of the dirichlet

## **Description**

This function calculates the loglikelihood of the dirichlet for the Dirich-gLV model.

# Usage

```
EstParmFunc(parms.vector, especie)
```

# Arguments

parms.vector

Vector equal to c(tau, as.vector(pam)) where:

• pam: Matrix. Each row has the parameters of each bacteria. Following our example, pam has the parameters placed as follows:

• tau: Number. Value of the tau parameter in the model

especie

Matrix that contains at row i the bacterial taxa of bacteria i at all time points . The bacteria placed in the last row of this matrix is the one used as reference in the alr transfromation that the model apply

#### **Details**

In an example with three bacteria, the regression of this model is defined by

$$r_1 \cdot \log(x_1(t)/x_3(t)) + \log(x_1(t)/x_3(t)) \cdot \left[a_{11} \cdot \log(x_1(t)/x_3(t))(t) + a_{12} \cdot \log(x_2(t)/x_3(t))\right]$$

$$r_2 \cdot \log(x_2(t)/x_3(t)) + \log(x_2(t)/x_3(t)) \cdot \left[a_{21} \cdot \log(x_1(t)/x_3(t))(t) + a_{22} \cdot \log(x_2(t)/x_3(t))\right]$$

#### Value

Returns a number with the value of the dirichlet loglikelihood.

# References

Creus-Martí, I. and Moya, A. and Santonja, F. J. (2018). A Statistical Model with a Lotka-Volterra Structure for Microbiota Data. Lucas Jodar, Juan Carlos Cortes and Luis Acedo, Modelling or engineering and human behavior 2018, Instituto Universitario de Matematica Multidisciplinar. ISBN: 978-84-09-07541-6

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#### **Examples**

```
especie1=cbind(c(0.5,0.3,0.2), c(0.1,0.3,0.6))
tau1=0.4
parms1= cbind(c(0.1,0.2),c(-0.2,0.1),c(0.3,0.2))
parms11=c(tau1,as.vector( parms1))
EstParmFunc(parms11,especie1)
```

EstParmFunc\_FBM

Writting the loglikelihood of the dirichlet

#### **Description**

This function calculates the loglikelihood of the dirichlet for the FBM model.

#### Usage

```
EstParmFunc_FBM(param, especie, E, EspecieMaxima, Tt, especiemodi)
```

#### **Arguments**

param	Vector with the parameters in the following order: a11,a12,a13, a21, a22,a23,
	a(D-1)1,a(D-1)2,a(D-1)3,tau. Where D is the number of bacterial species
	present in the matrix especie.
ocnocio	Matrix that contains at row i the heaterial taxe of heateria i at all time points

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

E Number of bacteria available

EspecieMaxima Row in which the bacterial with maximum mean abundance is in especie. This

bacteria is used as reference in the alr transformation that the model does and it

is placed at the denominator of the balance)

Tt Number of bacteria available

especiemodi Matrix that contains at row i the bacterial taxa of bacteria i at time points t=2,...,Tt.

The bacteria are placed in the same order than in especie.

#### **Details**

The regression of this model is defined by

$$\mu_{it} = a_{i1} + a_{i2} \cdot \operatorname{alr}(x_{i,(t-1)}) + a_{i3} \cdot \operatorname{Balance}(x_{i,(t-1)})$$
 for  $i = 1, \dots, D-1$  where  $D$  is the number of bacteria

#### Value

Returns a number with the value of the dirichlet loglikelihood.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2021). A Dirichlet autoregressive model for the analysis of microbiota time-series data. Complexity, 2021, 1-16.

#### **Examples**

```
especie1=cbind(c(0.5,0.3,0.2), c(0.1,0.3,0.6))
especiemodi=especie1[,-1]
tau1=0.4
parms1= cbind(c(0.1,0.2),c(-0.2,0.1),c(0.3,0.2))
parms11=c(as.vector( t(parms1)),tau1)
EstParmFunc_FBM(parms11,especie1,3 ,3 , 2,especiemodi)
```

ExpectedValuess\_BPBM Obtaining the value of the dirichlet parameters, the expected value and the variance.

## **Description**

This function calculates the value of the dirichlet parameters, the expected value and the variance for the BPBM model.

#### Usage

```
ExpectedValuess_BPBM(Estimated.Param, MatrizPBmodelo, E, Tt)
```

#### **Arguments**

Estimated.Param

Vector with the estimate parameters. Column "mean" of the output of "StudyingParam" function.

MatrizPBmodelo Matrix. Output of "Obtaining Value SPBal" called "Matrix SPBal".

E Number of bacteria available.

Tt Number of time points available.

#### **Details**

The regression of this model is defined by:

$$\mu_{it} = a_{i0} + a_{i1} \cdot \text{SPBal}_{1,t-1} + \dots + a_{iM} \cdot \text{SPBal}_{M,t-1}$$

#### Value

Returns a list with:

- Dirichlet.Param: Matrix. Matrix that contains at row i the dirichlet parameter of the bacteria i at all time points.
- Expected. Value: Matrix. Matrix that contains at row i the expected value of the bacteria i at all time points. The bacterias are placed at the same orden than in especies.
- Variance. Value: Matrix. Matrix that contains at row i the variance of the bacteria i at all time points. The bacterias are placed at the same orden than in especies.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

## **Examples**

ExpectedValues\_EstParmFunc\_FBM

Obtaining the value of the dirichlet parameters, the expected value and the variance.

#### Description

This function calculates the value of the dirichlet parameters, the expected value and the variance for the FBM model.

#### Usage

```
ExpectedValues_EstParmFunc_FBM(
  paramEstimadosFinal,
  especie,
  E,
  EspecieMaxima,
  Tt
)
```

#### **Arguments**

paramEstimadosFinal

The estimate parameters, in the following order: a11,a12,a13, a21, a22,a23, ...a(D-1)1,a(D-1)2,a(D-1)3,tau. Where D is the number of bacterial species

present in the matrix especie.

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

E Number of bacteria available.

EspecieMaxima Row in which the bacteria chosen as reference is in especie. This bacteria is

used as reference in the alr tranformation that the model does and it is placed at

the denominator of the balance).

Tt Number of time points available.

#### **Details**

The regression of this model is defined by

```
\mu_{it} = a_{i1} + a_{i2} \cdot \operatorname{alr}(x_{i,(t-1)}) + a_{i3} \cdot \operatorname{Balance}(x_{i,(t-1)}) for i = 1, \dots, D-1 where D is the number of bacteria
```

#### Value

Returns a list with:

- Dirichlet.Param: Matrix. Matrix that contains at row i the dirichlet parameter of the bacteria i at all time points.
- Expected. Value: Matrix. Matrix that contains at row i the expected value of the bacteria i at all time points. The bacterias are placed at the same orden than in especies.
- Variance. Value: Matrix. Matrix that contains at row i the variance of the bacteria i at all time points. The bacterias are placed at the same orden than in especies.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2021). A Dirichlet autoregressive model for the analysis of microbiota time-series data. Complexity, 2021, 1-16.

```
set.seed(123)
especie=t(gtools::rdirichlet(2,c(1,1,3)))
Tt=2
E=3
tau=5
EspecieMaxima=3
Iter.EstParmFunc=5
parms11=c(0.1,0.2,0.3,0.4,0.5,0.6,tau)

ExpectedValues_EstParmFunc_FBM(parms11 , especie,E,EspecieMaxima,Tt)
```

ExpectedValue\_EstParmFunc

Obtaining the value of the Dirichlet parameters, the expected value and the variance.

#### Description

This function calculates the value of the Dirichlet parameters, the expected value and the variance for the Dirich-gLV model.

#### Usage

ExpectedValue\_EstParmFunc(Param.Estimates, especie)

## **Arguments**

Param. Estimates

Vector with the estimated parameters. This value is the output of the Estimate\_Param\_EstParmFunc function.

especie

Matrix that contains at row i the bacterial taxa of bacteria i at all time points. The bacteria placed in the last row of this matrix is the one used as reference in the alr transformation that the model applies.

# Value

Returns a list with:

- Dirichlet.Param: Matrix. Matrix that contains at row i the Dirichlet parameter of the bacteria i at all time points.
- Expected. Value: Matrix. Matrix that contains at row i the expected value of the bacteria i at all time points. The bacterias are placed at the same orden than in especies.
- Variance. Value: Matrix. Matrix that contains at row i the variance of the bacteria i at all time points. The bacterias are placed at the same orden than in especies.

#### References

Creus-Martí, I. and Moya, A. and Santonja, F. J. (2018). A Statistical Model with a Lotka-Volterra Structure for Microbiota Data. Lucas Jodar, Juan Carlos Cortes and Luis Acedo, Modelling or engineering and human behavior 2018, Instituto Universitario de Matematica Multidisciplinar. ISBN: 978-84-09-07541-6

```
especie1=cbind(c(0.5,0.3,0.2), c(0.1,0.3,0.6)) tau1=0.4 parms1= cbind(c(0.1,0.2),c(-0.2,0.1),c(0.3,0.2))
```

```
parms11=c(tau1,as.vector( parms1))
ExpectedValue_EstParmFunc(parms11,especie1)
```

FromVectorToMatrix\_BPBM

Writting the parameters in the matrix form required in BPBM model

# Description

StudyingParam returns a matrix where the value of the parameters is in the column "mean". This function inputs this columns and outputs the parameters in the matrix form required by the BPBM model.

#### Usage

```
FromVectorToMatrix_BPBM(param, MatrizPBmodelo, E)
```

## **Arguments**

param Vector. Column "mean" of the output of "StudyingParam" function.

MatrizPBmodelo Matrix with the covariates of the model. In an example with two SPBal and three time points, the covariates are written in the following order:

$$\begin{array}{cccc} & & & 1 & & 1 \\ SPBal_{1,t-1} & SPBal_{1,t-2} & SPBal_{1,t-3} \\ SPBal_{2,t-1} & SPBal_{2,t-2} & SPBal_{2,t-3} \end{array}$$

E Number of bacteria in the dataset.

#### Value

Returns a matrix with the parameters in the order required by the BPBM model.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

20 FVectorPBmodeloPredi

FVectorPBmodeloPredi Obtaining a vector with the covariates of the prediction

# **Description**

Calculates a vector with the covariates of the BPBM model in one time point.

#### Usage

FVectorPBmodeloPredi(NumSPBal, DemSPBal, v, MatrizPBmodelo)

## **Arguments**

NumSPBal	List. The component i of the list has the number of the row of the matrix especie where the bacteria in the numerator of the principal balance i are placed.
DemSPBal	List. The component i of the list has the number of the row of the matrix especie where the bacteria in the denominator of the principal balance i are placed.
V	Vector. Vector with a coda composition. The bacteria ar in the same orden than the matrix <code>especie</code>
MatrizPBmodelo	the matrix that contains the covariates of the model. The first line es equal to 1 for all columns. The other rows contain the value of one SPBal at all time points. The selected principal balance of the row i+1 has at its numerator the bacteria placed in the rows NumSPBal[[i]] of the "especie". The selected principal balance of the row i+1 has at its denominator the bacteria placed in the rows DemSPBal[[i]] of the "especie".

## Value

Returns a vector where the first component is a 1 and the following components have the values of the SPBal. The SPBal in the component i+1 has at its numerator the bacteria placed in the rows Num[[i]] of the especie. The SPBal of the component i+1 has at its denominator the bacteria placed in the rows Dem[[i]] of the especie.

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## References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

## **Examples**

```
 \begin{array}{l} v \!\!=\!\! c(0.1,0.1,0.2,0.3,0.3) \\ \text{Num2} \!\!<\!\! -\!\! \text{list} (3,c(3,5),1,c(3,5,4)) \\ \text{Dem2} \!\!<\!\! -\!\! \text{list} (5,4,2,c(1,2)) \\ \text{MatrizPBmodelo=rbind} (c(1,1,1,1),\\ c(-0.3,0.2,0.5,0.6),\\ c(-0.4,0.3,0.5,0.6),\\ c(0.5,0.3,0.2,0.7),\\ c(-0.2,0.9,0.2,0.1) \end{array}
```

FVectorPBmodeloPredi(Num2,Dem2,v,MatrizPBmodelo)

Graphics

Plots the time series

# **Description**

Plots the time series

## Usage

```
Graphics(especie, names.especie, esperanza, Variance, Plot.Tipe, Detail)
```

## **Arguments**

especie	Matrix that contains at row i the bacterial taxa of bacteria i at all time points.
names.especie	Vector with the names of the bacteria in the same order that are placed in the especie matrix.
esperanza	Matrix that contains at row i the expected value of the bacterial taxa i at all time points. The bacteria must be placed in the same order than in especie
Variance	Matrix that contains at row i the variance of the bacterial taxa i at all time points. The bacteria must be placed in the same order than in especie
Plot.Tipe	Character. If Plot.Tipe==Data the function displays a graphic of the dataset, if Plot.Tipe==DataExpected the function displays a graphic of the data and the expected values, if Plot.Tipe==All the function displais a graphic woth the data, the expected values and the variance.If Plot.Tipe==Var the function returns the boxplot of the variance at each time point and the variance of each bacteria. If Plot.Tipe==OnlyVar the function returns the boxplots of the variance at each time points.
Detail	Character. If Detail==no the graphic obtained when Plot.Tipe==DataExpected and Plot.Tipe==All will have the same y axis for all the taxa. If Detail==yes these functions will have different y axis.

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#### Value

Returns the indicated plots.

#### **Examples**

```
names.especie=c("Bact1", "Bact2", "Bact3")
especie=cbind(c(0.5,0.3,0.2), c(0.6,0.3,0.1),c(0.4,0.1,0.5),c(0.4,0.1,0.5))
esperanza=especie[,c(1:3)]+0.1
Variance=matrix(c(runif(9,0.001,0.004)), 3,3)

Graphics(especie, names.especie, esperanza, Variance, "Data", "no")
Graphics(especie, names.especie, esperanza, Variance, "DataExpected", "no")
Graphics(especie, names.especie, esperanza, Variance, "All", "no")
```

GraphicsPrediction

Plots the time series

## **Description**

This function takes into account the data used to estimate and the data used to predict.

## Usage

```
GraphicsPrediction(
   especie.All,
   names.especie,
   ExpectedValue.All,
   VarianceValue.All,
   Pred,
   Plot.Tipe,
   Detail
)
```

#### **Arguments**

especie. All Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

names.especie Vector with the names of the bacteria in the same order that are placed in the especie matrix.

ExpectedValue.All

Matrix that contains at row i the expected value of the bacterial taxa i at all time points. The bacteria must be placed in the same order than in especie. This matrix must comply: dim(ExpectedValue.All)[2]=dim(especie.All)[2]-1

VarianceValue.All

Matrix that contains at row i the variance of the bacterial taxa i at all time points. The bacteria must be placed in the same order than in especie. This matrix must comply: dim(VarianceValue.All)[2]=dim(especie.All)[2]-1

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Pred Number. Indicates the time point in which we start predicting.

Plot.Tipe Character. If Plot.Tipe==Data the function displays a graphic of the dataset,

if Plot.Tipe==DataExpected the function displays a graphic of the data and the expected values, if Plot.Tipe==All the function displays a graphic with the data, the expected values and the variance.If Plot.Tipe==Var the function returns the boxplot of the variance at each time point and the variance of each bacteria. If Plot.Tipe==OnlyVar the function returns the boxplots of the variance

ance at each time points.

Detail Character. If Detail==no the graphic obtained when Plot. Tipe==DataExpected

and Plot.Tipe==All will have the same y axis for all the taxa. If Detail==yes

these functions will have different y axis.

#### Value

Returns the indicated plots with a vertical line when the time point is equal to Pred-1, in Pred the prediction has started.

```
names.especie=c("Bact1", "Bact2", "Bact3")
especie.All=cbind(c(0.5,0.3,0.2),
                c(0.6, 0.3, 0.1),
                c(0.4, 0.1, 0.5),
                c(0.4, 0.1, 0.5),
                c(0.4, 0.1, 0.5),
                c(0.4, 0.1, 0.5))
ExpectedValue.All=especie.All[,-1]+0.1
VarianceValue.All=matrix(c(runif(15,0.001,0.004)), 3,5)
Pred=4
GraphicsPrediction(especie.All,
                  names.especie,
                  ExpectedValue.All,
                  VarianceValue.All ,
                  Pred,
                   "Data",
                   "no")
GraphicsPrediction(especie.All,
                 names.especie,
                 ExpectedValue.All,
                 VarianceValue.All ,
                 Pred,
                  "DataExpected",
                  "no")
GraphicsPrediction(especie.All,
                 names.especie,
                 ExpectedValue.All,
                 VarianceValue.All ,
```

```
Pred,
                 "All",
                 "no")
GraphicsPrediction(especie.All,
                 names.especie,
                 ExpectedValue.All,
                 VarianceValue.All ,
                 Pred,
                 "Var",
                 "no")
GraphicsPrediction(especie.All,
                 names.especie,
                 ExpectedValue.All,
                 VarianceValue.All ,
                 Pred,
                 "OnlyVar",
                 "no")
```

GraphicsPredictionBPBM

Plots the time series

# Description

This function takes into account the data used to estimate and the data used to predict. We use this function when we want to observe the results obtained with the BPBM model.

# Usage

```
GraphicsPredictionBPBM(
especie.All,
names.especie,
ExpectedValue.All,
VarianceValue.All,
Pred,
Plot.Tipe,
Varmas,
Varmenos,
Detail
)
```

# **Arguments**

especie. All Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

names.especie Vector with the names of the bacteria in the same order that are placed in the especie matrix.

ExpectedValue.All

Matrix that contains at row i the expected value of the bacterial taxa i at all time points. The bacteria must be placed in the same order than in especie. This matrix must comply: dim(ExpectedValue.All)[2]=dim(especie.All)[2]-1

VarianceValue.All

Plot.Tipe

Varmas

Matrix that contains at row i the variance of the bacterial taxa i at all time points. The bacteria must be placed in the same order than in especie. This matrix must comply: dim(VarianceValue.All)[2]=dim(especie.All)[2]-1

Pred Number. Indicates the time point in which we start predicting.

Character. If Plot.Tipe==Data the function displays a graphic of the dataset, if Plot.Tipe==DataExpected the function displays a graphic of the data and the expected values, if Plot.Tipe==All the function displays a graphic with the data, the expected values and the variance (the varmas and varmenos parameters are introduce and are taken into account to plot the variance of the predicted part).If Plot.Tipe==Var the function returns the boxplot of the variance at each

time point and the variance of each bacteria. If Plot.Tipe==OnlyVar the function returns the boxplots of the variance at each time points.

Matrix. Output of "PredictionBPBM" adding "\$ExpVarmas". Matrix that contains at row i the expected value plus two times the sqrt(variance) of the bacteria i at all time points t=Tt,...,K, the rest of the time points has 0 values. The bacteria

are placed at the same order than in especies.

Varmenos Matrix. Output of "PredictionBPBM" adding "\$ExpVarmenos". Matrix that contains at row i the expected value minus two times the sqrt(variance) of the

bacteria i at all time points t=Tt,...,K, the rest of the time points has 0 values.

The bacteria are placed at the same order than in especies.

Detail Character. If Detail==no the graphic obtained when Plot.Tipe==DataExpected

and Plot.Tipe==All will have the same y axis for all the taxa. If Detail==yes

these functions will have different y axis.

# Value

Returns the indicated plots with a vertical line when the time point is equal to Tt=Pred-1, in Pred=Tt+1 the predicction has started.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

```
c(0.4, 0.1, 0.5),
                 c(0.4, 0.1, 0.5),
                 c(0.4,0.1,0.5))
ExpectedValue.All=especie.All[,-1]+0.1
VarianceValue.All=matrix(c(runif(15,0.001,0.004)), 3,5)
Pred=4
Varmas=cbind(matrix(0,3,2),matrix(c(runif(9,0.001,0.004)) ,3 ,3 ))
Varmenos = cbind(matrix(0,3,2), matrix(c(runif(9,0.001,0.004)), 3,3))
GraphicsPredictionBPBM(especie.All,
                      names.especie,
                      ExpectedValue.All,
                      VarianceValue.All ,
                      Pred ,
                      "Data",
                      Varmas,
                      Varmenos,
                      "no")
GraphicsPredictionBPBM(especie.All,
                      names.especie,
                      ExpectedValue.All,
                      VarianceValue.All ,
                      Pred ,
                      "DataExpected", \\
                      Varmas,
                      Varmenos,
                       "no")
GraphicsPredictionBPBM(especie.All,
                      names.especie,
                      ExpectedValue.All,
                      VarianceValue.All,
                      Pred ,
                      "All",
                      Varmas,
                      Varmenos,
                      "no")
GraphicsPredictionBPBM(especie.All,
                      names.especie,
                      ExpectedValue.All,
                      VarianceValue.All ,
                      Pred ,
                      "Var",
                      Varmas,
                      Varmenos,
                      "no")
GraphicsPredictionBPBM(especie.All,
                      names.especie,
                      ExpectedValue.All,
                      VarianceValue.All ,
```

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```
Pred ,
"OnlyVar",
Varmas,
Varmenos,
"no")
```

GraphicsSPBal

Obtaining the graphic of the SPBal at all time points

## **Description**

The SPBal (of BPBM model) are ordered from highest to lowest variance percentage. The zero is highlight because the closer the value of the balance is to zero, the more similar (in terms of relative abundance) the numerator and denominator groups will be. The farther away from zero, the more different.

## Usage

GraphicsSPBal(MatrizPBmodelo)

## **Arguments**

MatrizPBmodelo Matrix. Output of "ObtainigValueSPBal" function. MatrixSPBal is the matrix that contains the covariates of the model. The first line es equal to 1 for all columns. The other rows contain the value of one SPBal at all time points. The selected principal balance of the row i+1 has at its numerator the bacteria placed in the rows NumSPBal[[i]] of the "especie". The selected principal balance of the row i+1 has at its denominator the bacteria placed in the rows DemSPBal[[i]] of the "especie".

#### Value

Returns a graphic.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

```
MatrizPBmodelo=rbind(c(1,1,1,1,1,1),c(-0.3,0.4,0.3,-0.7,-0.4,-0.6),c(0.3,0.5,-0.3,0.1,0.4,0.1))
GraphicsSPBal(MatrizPBmodelo)
```

LogVeroFuncBUENA

LogVeroFuncBUENA	Writting the loglikelihood of the dirichlet
LOG VCI OI UIICDOLIVA	Willing the logithelihood of the differiel

# Description

This function calculates the loglikelihood of the dirichlet for the BPBM model.

#### Usage

```
LogVeroFuncBUENA(param, MatrizPBmodelo, E, Tt, especiemodi)
```

## **Arguments**

param Vector. Column "mean" of the output of "StudyingParam" function.

MatrizPBmodelo Matrix with the covariates of the model. In an example with two SPBal and three time points, the covariates are written in the following order:

$$\begin{array}{cccc} 1 & 1 & 1 \\ SPBal_{1,t-1} & SPBal_{1,t-2} & SPBal_{1,t-3} \\ SPBal_{2,t-1} & SPBal_{2,t-2} & SPBal_{2,t-3} \end{array}$$

E Number f bacteria in the dataset.

Tt Number of time points available

especiemodi Matrix that contains at row i the bacterial taxa of bacteria i at time points t=2,...,Tt.

#### Value

Returns a number with the value of the dirichlet loglikelihood.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

MaxBacteria 29

```
nn.burnin=1000,
nn.sample=5000,
nn.thin=10)

param=est$SamplesAllChains

especiemodi=especie[,-1]

LogVeroFuncBUENA(param, MatrizPBmodelo, E, Tt, especiemodi)
```

MaxBacteria

Putting the reference bacteria at the last row

#### **Description**

This function calculates the mean abundance of each bacteria. Then, it creates a matrix where each row contains the abundance of one bacteria at all time points but the bacteria with maximum (or minimum or a bacteria indicated by the user) mean abundance is placed at the last row

#### Usage

```
MaxBacteria(nombresOriginal, especieOriginal, E, Tt, which.esp)
```

#### **Arguments**

nombresOriginal

Vector with the bacterial names at the same order than in DaTa. it must be fulfilled that lenght(nombresOriginal)==dim(DaTa)[2]-1

especieOriginal

Matrix that contains at row i the bacterial taxa of bacteria i at all time points

E Number of bacteria available

Tt Number of bacteria available

which.esp If which=="Max" this function puts in

If which=="Max" this function puts in the last position of the matrix the bacteria with maximum mean abundance. If which=="Min" this function puts in the last position of the matrix the bacteria with minimum mean abundance. If which is equal to a number this function puts in the last position of the matrix the bacteria

that is in the "which" row of the especieOriginal matrix.

#### Value

#### Returns a list with

• especie - Matrix that contains at row i the bacterial taxa of bacteria i at all time points but the bacteria with maximum (or minimum) mean abundance (or the bacteria indicated by the user) is placed at the last row.

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• especiemodi - Matrix that contains at row i the bacterial taxa of bacteria i at time points t=2,...,Tt but the bacteria with maximum (or minimum) mean abundance (or the bacteria indicated by the user) is placed at the last row.

- nombres Vector with the bacteria's names placed in the order in which appear in the rows of the matrices especie and especiemodi
- EE Row in which the bacterial with maximum (or minimum) mean abundance was (or the value of "which" if which is numerical).
- EspecieMaxima Row in which the bacterial with maximum (or minimum) mean abundance (or the bacteria indicated by the user) is in especie.) #'

## **Examples**

```
names1=c("Bact1","Bact2","Bact3")
set.seed(314)
esp1=t(gtools::rdirichlet(n=4, c(1,3,1)))
e1=3
t1=4

MaxBacteria(names1,esp1,e1,t1,"Max")

names3=c("Bact1","Bact2","Bact3","Bact4","Bact5")
set.seed(314)
esp3=t(gtools::rdirichlet(n=6, c(6,6,1,6,6)))
e3=5
t3=6
MaxBacteria(names3,esp3,e3,t3,"Min")
```

MaxBacteriaPred

Putting the reference bacteria at the last row

#### **Description**

This function calculates the mean abundance of each bacteria taking into account the time points used to estimate the model (t=1,2,...,Tt). Then, it creates a matrix where each row contains the abundance of one bacteria at all time points but the bacteria with maximum (or minimum) mean abundance (or the bacterial indicated by the user) is placed at the last row

# Usage

```
MaxBacteriaPred(
  nombresOriginal,
  especieOriginal,
  E,
  Tt,
  Pred,
```

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```
K,
  especieOriginal.All,
  which.esp
)
```

#### **Arguments**

nombresOriginal

Vector with the bacterial names at the same order than in DaTa. it must be fulfilled that lenght(nombresOriginal)==dim(DaTa)[2]-1

especieOriginal

Matrix that contains at row i the bacterial taxa of bacteria i at t=1,2,...,Tt, with

Tt=Pred-1.

E Number of bacteria available

Tt Number of time points used to estimate the model (Tt=Pred-1)

Pred Number. The data at t=1,...,Pred-1 will be used to estimate the model. The rest

of the time points will be used to study the capacity of the model to predict. If

Pred==0 all the dataset will be used to estimate the model.

K Number of time points at the data

especieOriginal.All

Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

which.esp

If which=="Max" this function puts in the last position of the matrix the bacteria with maximum mean abundance. If which=="Min" this function puts in the last position of the matrix the bacteria with minimum mean abundance. If which is equal to a number this function puts in the last position of the matrix the bacteria that is in the "which" row of the especieOriginal matrix.

#### Value

#### Returns a list with

- especie Matrix that contains at row i the bacterial taxa of bacteria i at time points t=1,2,...,Tt but the bacteria with with maximum (or minimum) mean abundance (or the bacteria indicated by the user) is placed at the last row.
- especiemodi Matrix that contains at row i the bacterial taxa of bacteria i at time points t=2,...,Tt but the bacteria with with maximum (or minimum) mean abundance (or the bacteria indicated by the user) is placed at the last row.
- nombres Vector with the bacteria's names placed in the order in which appear in the rows of the matrices especie and especiemodi
- EE Row in which the bacterial with maximum (or minimum) mean abundance was (or the value of "which" if which is numerical).
- EspecieMaxima Row in which the bacterial with with maximum (or minimum) mean abundance (or the bacteria indicated by the user) is in especie.) #' #'
- especie.all Matrix that contains at row i the bacterial taxa of bacteria i at all time points (t=1,2,...,K) but the bacteria with with maximum (or minimum) mean abundance (or the bacteria indicated by the user) is placed at the last row.

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• especiemodi.all - Matrix that contains at row i the bacterial taxa of bacteria i at all time points (t=2,...,K) but the bacteria with with maximum (or minimum) mean abundance (or the bacteria indicated by the user) is placed at the last row.

## **Examples**

```
names2=c("Bact1","Bact2","Bact3","Bact4","Bact5")
set.seed(314)
esp2=t(gtools::rdirichlet(n=6, c(1,1,5,1,1)))
e2=5

MaxBacteriaPred(names2,esp2[,-c(4,5,6)],e2,3,Pred=4, 6,esp2, "Max")
```

ObtainigValuePB

Obtaining the principal balances values

## **Description**

Calculates the value of the principal balances (Martín-Fernández et al, 2018) at all time points.

## Usage

```
ObtainigValuePB(Num, Dem, especie, Tt)
```

#### **Arguments**

Num	List. The component i of the list has the number of the row of the matrix especie where the bacteria in the numerator of the principal balance i are placed.
Dem	List. The component i of the list has the number of the row of the matrix especie where the bacteria in the denominator of the principal balance i are placed.
especie	Matrix that contains at row i the bacterial taxa of bacteria i at all time points.
Tt	Number of time points available

# Value

Returns a matrix where the row i has the value of the principal balance at all time points. The principal balance of the row i has at its numerator the bacteria placed in the rows Num[[i]] of the especie. The principal balance of the row i has at its denominator the bacteria placed in the rows Dem[[i]] of the especie.

Obtainig Value SPB al 33

## References

• Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

Martín-Fernández, J. A., Pawlowsky-Glahn, V., Egozcue, J. J., & Tolosona-Delgado, R. (2018).
 Advances in principal balances for compositional data. Mathematical Geosciences, 50, 273-298.

# **Examples**

```
set.seed(314)
esp2=t(gtools::rdirichlet(n=6, c(1,1,5,1,1)))
Num2<-list(3,c(3,5),1,c(3,5,4))
Dem2<-list(5,4,2,c(1,2))
ObtainigValuePB(Num2,Dem2,esp2,6)</pre>
```

ObtainigValueSPBal

Obtaining the selected principal balances values

# Description

Calculates the value of the selected principal balances (SPBal) of the BPBM model at all time points.

#### Usage

```
ObtainigValueSPBal(Num, Dem, especie, Tt)
```

## **Arguments**

Num	List. The component i of the list has the number of the row of the matrix especie where the bacteria in the numerator of the principal balance i are placed.
Dem	List. The component i of the list has the number of the row of the matrix especie where the bacteria in the denominator of the principal balance i are placed.
especie	Matrix that contains at row i the bacterial taxa of bacteria i at all time points.
Tt	Number of time points available

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#### Value

Returns a list with:

• NumSPBal: List. The component i of the list has the number of the row of the matrix especie where the bacteria in the numerator of the selected principal balance i are placed.

- DemSPBal: List. The component i of the list has the number of the row of the matrix especie where the bacteria in the denominator of the selected principal balance i are placed.
- MatrixSPBal: MatrixSPBal is the matrix that contains the covariates of the model. The first line es equal to 1 for all columns. The other rows contain the value of one SPBal at all time points. The selected principal balance of the row i+1 has at its numerator the bacteria placed in the rows NumSPBal[[i]] of the "especie". The selected principal balance of the row i+1 has at its denominator the bacteria placed in the rows DemSPBal[[i]] of the "especie".
- PercenVarianceSPBal: Vector. The component of the vector i contains the percentage of variance of the SPBal with numerator NumSPBal[[i]] and denominator DemSPBal[[i]].

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

#### **Examples**

```
set.seed(314)
esp2=t(gtools::rdirichlet(n=6, c(1,1,5,1,1)))
Num2<-list(3,c(3,5),1,c(3,5,4))
Dem2<-list(5,4,2,c(1,2))
ObtainigValueSPBal(Num2,Dem2,esp2,6)</pre>
```

ObtainingDIC

Writting the loglikelihood of the dirichlet

#### **Description**

This function calculates the loglikelihood of the dirichlet for the BPBM model. Then, it calculates the loglikelihood with the parameters of each iteration of the MCMC chains and introduces the values in a vector called vectorD. DIC=(1/2)\*var(vectorD)+mean(VectorD)

#### Usage

```
ObtainingDIC(cadenas, MatrizPBmodelo, E, Tt, especiemodi)
```

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## **Arguments**

cadenas Matrix with the iterations (in rows) of all the Markov Chains obtained in th

estimation. It is the output of "StudyingParam" adding "\$AllChainsJoined".

MatrizPBmodelo Matrix with the covariates of the model. In an example with two SPBal and

three time points, the covariates are written in the following order:

$$\begin{array}{cccc} & & & 1 & & 1 \\ SPBal_{1,t-1} & SPBal_{1,t-2} & SPBal_{1,t-3} \\ SPBal_{2,t-1} & SPBal_{2,t-2} & SPBal_{2,t-3} \end{array}$$

E Number f bacteria in the dataset.

Tt Number of time points available

especiemodi Matrix that contains at row i the bacterial taxa of bacteria i at time points t=2,...,Tt.

#### Value

Returns a data.frame with the DIC value (using the rule, pD = var/2).

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

```
set.seed(314)
especie=t(gtools::rdirichlet(n=2, c(1,2,3)))
E=3
Tt=2
MatrizPBmodelo=rbind(c(1,1), c(-0.3,0.4), c(0.3,0.5))
set.seed(314)
est=Estimating_BPBM(especie,
                   Τt,
                   MatrizPBmodelo,
                   nn.chain=3,
                   nn.burnin=1000,
                   nn.sample=5000,
                   nn.thin=10)
SumFinal=StudyingParam(est$R2jagsOutput$BUGSoutput$summary
                                                               ,est$SamplesAllChains)
cadenas=SumFinal$AllChainsJoined
especiemodi=especie[,-1]
ObtainingDIC(cadenas, MatrizPBmodelo, E, Tt, especiemodi)
```

36 PBalancePredi

PBalance	Calculating balances
----------	----------------------

# Description

This function calculates the balance that has at the numerator the bacteria placed at Num and has at the denominator the bacteria placed at Dem

# Usage

```
PBalance(A, Num, Dem, especie)
```

## **Arguments**

Α	Number. The balance will be calculated for $t=1,2,,A$ time points.
---	--

Num vector that contains the position in the matrix especies of the families that we

position at the numerator of the balance.

Dem vector that contains the position in the matrix especies of the families that we

position at the denumerator of the balance

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

#### Value

Returns a vector with the value of the balance in each time point.

#### **Examples**

```
especie1=cbind(c(0.5,0.3,0.1,0.1), c(0.1,0.3,0.6,0.1))  \begin{aligned} &\text{Num} = \text{c}(1,2) \\ &\text{Dem} = \text{c}(3,4) \\ &\text{A=2} \\ &\text{PBalance}(\text{A},\text{Num},\text{Dem},\text{especie1}) \end{aligned}
```

PBalancePredi

Calculating balances for a composition

## **Description**

This function calculates the balance that has at the numerator de bacteria placed at Num and has at the denominator the bacteria placed at Dem

#### Usage

```
PBalancePredi(Num, Dem, DatosEsperanzas)
```

PBmodel 37

# **Arguments**

Num vector that contains the position in the matrix especies of the families that we

position at the numerator of the balance.

Dem vector that contains the position in the matrix especies of the families that we

position at the denumerator of the balance

DatosEsperanzas

Vector with a coda composition. The bacteria are in the same orden than the matrix especie

#### Value

Returns the value of the balance

## **Examples**

```
Num=c(1,2)
Dem=c(3,4)
DatosEsperanzas=c(0.1,0.3,0.4,0.2)
PBalancePredi(Num,Dem,DatosEsperanzas)
```

**PBmodel** 

Obtaining the regression value of the BPBM

## Description

This function calculates the value of the BPBM regression, defined by:

$$\mu_{it} = a_{i0} + a_{i1} \cdot \text{SPBal}_{1,t-1} + \dots + a_{iM} \cdot \text{SPBal}_{M,t-1}$$

#### Usage

```
PBmodel(A, MatrizPBmodelo, E, Tt)
```

# **Arguments**

A Matrix that contains all the parameters of the model. The parameters are written in the matrix in the following order (in an example with three bacteria):

```
a10 a11 a12 ... a1M
a20 a21 a22 ... a2M
a30 a31 a32 ... a3M
```

MatrizPBmodelo Matrix. Output of "Obtaining Value SPBal" called "Matrix SPBal".

E Number of bacteria in the dataset.

Tt Number of time points.

#### Value

Returns a matrix. The row i contains the regression values of the bacteria i at all time points.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

#### **Examples**

```
A=rbind(c(1,2,3),c(4,5,6),c(7,8,9))
MatrizPBmodelo=cbind(c(1,2,3),c(4,5,6),c(7,8,9))
E=3
Tt=3
PBmodel(A,MatrizPBmodelo, E,Tt)
```

**PCAbiplot** 

PCA of the estimated parameters

#### **Description**

This function applys a PCA to the estimate parameters (using function "prcomp" with center = TRUE and scale. = TRUE). Then uses the ggbiplot function to plot the biplot.

## Usage

```
PCAbiplot(paramEstimadosFinal, names, E)
```

#### **Arguments**

paramEstimadosFinal

The estimate parameters, in the following order: a11,a12,a13, a21, a22,a23, ...a(D-1)1,a(D-1)2,a(D-1)3,tau. Where D is the number of bacterial especies present in the matrix especie.

names

Vector with the name of the bacteria. The component i has the name of the bacteria i, with i=1,...,D. The bacteria in the las position of the vector is the bacteria used as reference in the alr transformation.

E Number of bacteria available.

#### Value

Returns a list with the PCA biplot, the variance explained of each Principal Component and an object of class "prcomp" with the PCA. In the biplot, "a" denotes the intercept, "b" denotes the parameter that give information about the importance of the bacteria in defining herself in the next time point and "c" denotes denotes the parameter that give information about the importance of the rest of the community in defining the bacteria in the next time point.

Percen\_Variance 39

## **Examples**

```
set.seed(123)
especie=t(gtools::rdirichlet(10,c(1,3,1,2,4)))
names=c("Bact1","Bact2","Bact3","Bact4","Bact5")
tau1=0.4
parms1= cbind(c(0.1,0.2,0.4,0.6),c(-0.2,0.1,0.1,0.3),c(0.3,0.2,0.3,0.5))
paramEstimadosFinal=c(as.vector( t(parms1)),tau1)
PCAbiplot(paramEstimadosFinal,names,5)
```

Percen\_Variance

Percentage of variance

# Description

This function calculates the variance of each row of the matrix PB. Returns the percentage of variance of each row of the matrix PB.

## Usage

```
Percen_Variance(PB)
```

# Arguments

РΒ

Matrix.

# Value

Returns a vector with percentage of variance of each row of the matrix PB.

```
matt=matrix(c(1:4),2,2)
Percen_Variance(matt)
```

40 PlotDendogram

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Plotting a dendogram

## **Description**

Plots the dendogram obtained using the Ward's method for obtaining the principal balances. The process follow in this function is explained in Section 3.1 of (Creus-Martí et al, 2022)

## Usage

```
PlotDendogram(especie, names)
```

# **Arguments**

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

names Vector with the names of the bacteria in the same order that are written in

especie.

#### Value

Returns a list with: the dendogram.

- Num: List. The component i of the list has the number of the row of the matrix especie where the bacteria in the numerator of the principal balance i are placed.
- Dem: List. The component i of the list has the number of the row of the matrix especie where the bacteria in the denominator of the principal balance i are placed.
- dendogram: Plots the dendogram.

## References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

```
names2=c("Bact1","Bact2","Bact3","Bact4","Bact5")
set.seed(314)
esp2=t(gtools::rdirichlet(n=6, c(1,1,5,1,1)))
PlotDendogram(esp2,names2)
```

**PredictionBPBM** 41

PredictionBPBM Predicting using BPBM

#### **Description**

This function calculates the expected value and variance of the bacteria at time point Tt. Then, this function calculates the expected value and variance of the bacteria at time point t=(Tt+1),...,K. It calculates the expected value at each time point for each markov chain iteration. The expected value for each time point is the mean of the expected values of all iterations.mAnalogous with the variance, the dirichlet parameters and the expected valur plus(and minus) two times the sqrt of the variance.

# Usage

```
PredictionBPBM(
 NumSPBal,
 DemSPBal,
 MCMC.CHAINS,
  alpha,
 Κ,
  esperanza,
  Var,
  Ε.
  Tt,
 MatrizPBmodelo
)
```

# **Arguments**

MCMC.CHAINS

alpha

Κ

NumSPBal	List. The component i of the list has the number of the row of the matrix especie where the bacteria in the numerator of the principal balance i are placed.
DemSPBa1	List. The component i of the list has the number of the row of the matrix especie where the bacteria in the denominator of the principal balance i are

placed. Matrix with the iterations of all the chains for all the parameters. Each column

has all the iteration of one parameter. If the cero is in the center of the credible interval of one parameter all its iteration in the Marchov Chain have the value 0. It is output of the "StudyingParam" function adding "\$AllChainsJoined".

Matrix that contains at the row i the Dirichlet parameter of the bacteria i at

t=1,2,3,...,Tt.

Number. The function will calculate the value of the expected value and the variance at Tt and predict for the time points t=Tt+1,..,K. To predict all the time

points available at the data we K=dim(especie.All)-1

Matrix that contains at row i the expected value of the bacterial taxa of bacteria esperanza

i at t=1,2,3,...,Tt-1.

42 **PredictionBPBM** 

Matrix that contains at row i the variance of the bacterial taxa of bacteria i at Var t=1,2,3,...,Tt-1.

Number of bacteria available

Τt Number of bacteria available

MatrizPBmodelo is the matrix that contains the covariates of the model. The first line es equal to 1 for all columns. The other rows contain the value of one SPBal at all time points. The selected principal balance of the row i+1 has at its numerator the bacteria placed in the rows NumSPBal[[i]] of the "especie". The selected principal balance of the row i+1 has at its denominator the bacteria placed in the rows DemSPBal[[i]] of the "especie".

#### Value

Ε

Returns a list with:

- ExpectedValue.All: Matrix. Matrix that contains at row i the expected value of the bacteria i at all time points t=1,2,...,K. The bacterias are placed at the same order than in especies.
- VarianceValue.All: Matrix. Matrix that contains at row i the variance of the bacteria i at all time points t=1,2,...,K. The bacterias are placed at the same order than in especies.
- DirichlerParam.All: Matrix. Matrix that contains at row i the dirichlet parameter of the bacteria i at all time points t=1,2,...,K. The bacterias are placed at the same order than in especies.
- ExpVarmas: Matrix. Matrix that contains at row i the expected value plus two times the sgrt(variance) of the bacteria i at all time points t=Tt,...,K, the rest of the time points has 0 values. The bacterias are placed at the same order than in especies.
- ExpVarmenos: Matrix. Matrix that contains at row i the expected value plus two times the sqrt(variance) of the bacteria i at all time points t=Tt,...,K,the rest of the time points has 0 values. The bacterias are placed at the same order than in especies.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2022). Bayesian hierarchical compositional models for analysing longitudinal abundance data from microbiome studies. Complexity, 2022.

```
NumSPBal=list(1,c(1,2))
DemSPBal=list(2,3)
MCMC.CHAINS=cbind(c(0.1,0.11),
                  c(0.2, 0.21),
                  c(0.3, 0.31),
                  c(-0.1, -0.11),
                  c(0.15, 0.105),
                  c(0.44, 0.41),
                  c(0.3, 0.31),
                  c(0.201,0.221),
                  c(0.13, 0.113))
alpha=cbind(c(0.1,0.2,0.1),c(0.1,0.5,0.3))
K=3
```

PredictionEstParmFunc 43

```
\begin{split} & \operatorname{esperanza=cbind}(c(\emptyset.2,\emptyset.2,\emptyset.6)) \\ & \operatorname{Var=cbind}(c(\emptyset.1,\emptyset.01,\emptyset.11)) \\ & \operatorname{E=3} \\ & \operatorname{Tt=2} \\ & \operatorname{MatrizPBmodelo=cbind}(c(1,\emptyset.3,\emptyset.2)) \\ & \operatorname{PredictionBPBM}(\operatorname{NumSPBal},\operatorname{DemSPBal},\operatorname{MCMC}.\operatorname{CHAINS},\operatorname{alpha},\operatorname{K},\operatorname{esperanza},\operatorname{Var},\operatorname{E},\operatorname{Tt},\operatorname{MatrizPBmodelo}) \\ & \operatorname{MatrizPBmodelo} \\ & \operatorname{CHAINS},\operatorname{alpha},\operatorname{K},\operatorname{esperanza},\operatorname{Var},\operatorname{E},\operatorname{Tt},\operatorname{MatrizPBmodelo}) \\ & \operatorname{MatrizPBmodelo} \\ & \operatorname{MatrixPBmodelo} \\ & \operatorname
```

PredictionEstParmFunc Predicting using dirichl-gLV

# Description

This function calculates the expected value and variance of the bacteria at time point Tt. Then, this function calculates the expected value and variance of the bacteria at time point t=(Tt+1),...,K

## Usage

```
PredictionEstParmFunc(
  paramEstimadosFinal,
  EspecieMaxima,
  alpha,
  K,
  esperanza,
  Var,
  E,
  Tt
)
```

#### Arguments

paramEstimadosFinal

The estimate parameters. Vector equal to c(tau, as.vector(pam)) where:

• pam Matrix. Each row has the parameters of each bacteria. Following our example, pam has the parameters placed as follows:

```
r1 a11 a12
r2 a21 a22
```

• tau Number. Value of the tau parameter in the model

EspecieMaxima

Row in which the bacteria chosen as reference is in especie. This bacteria is used as reference in the alr transformation that the model does.

alpha

Matrix that contains at the row i the dirichlet parameter of the bacteria i at t=1,2,3,...,Tt.

44 PredictionEstParmFunc

K	Number. The function will calculate the value of the expected value and the variance at Tt and predict for the time points t=Tt+1,,K. To predict all the time points available at the data we K=dim(especie.All)-1
esperanza	Matrix that contains at row i the expected value of the bacterial taxa of bacteria i at $t=1,2,3,,Tt-1$ .
Var	Matrix that contains at row i the variance of the bacterial taxa of bacteria i at $t=1,2,3,,Tt-1$ .
E	Number of bacteria available
Tt	Number of time points available

## **Details**

The regression of this model, in an example with three bacteria, is defined by

```
r_1 \cdot log(x_1(t)/x_3(t)) + log(x_1(t)/x_3(t)) \cdot [a_{11} \cdot log(x_1(t)/x_3(t))(t) + a_{12} \cdot log(x_2(t)/x_3(t))]
r_2 \cdot log(x_2(t)/x_3(t)) + log(x_2(t)/x_3(t)) \cdot [a_{21} \cdot log(x_1(t)/x_3(t))(t) + a_{22} \cdot log(x_2(t)/x_3(t))]
```

#### Value

Returns a list with:

- ExpectedValue.All: Matrix. Matrix that contains at row i the expected value of the bacteria i at all time points t=2,...,K. The bacteria are placed at the same order than in especies.
- VarianceValue.All: Matrix. Matrix that contains at row i the variance of the bacteria i at all time points t=2,...,K. The bacteria are placed at the same order than in especies.
- DirichlerParam.All: Matrix. Matrix that contains at row i the dirichlet parameter of the bacteria i at all time points t=2,...,K. The bacteria are placed at the same order than in especies.

#### References

Creus-Martí, I. and Moya, A. and Santonja, F. J. (2018). A Statistical Model with a Lotka-Volterra Structure for Microbiota Data. Lucas Jodar, Juan Carlos Cortes and Luis Acedo, Modelling or engineering and human behavior 2018, Instituto Universitario de Matematica Multidisciplinar. ISBN: 978-84-09-07541-6

```
pam.ini=rbind(c(0.1,0.2,0.3),c(0.4,0.5,0.6))
paramEstimadosFinal=c(5, as.vector(pam.ini))
EspecieMaxima=3
alpha=cbind(c(2,2,3),c(1,1,3))
K=3
esperanza=cbind(c(0.2,0.3,0.5))
Var=cbind(c(0.2,0.3,0.5))
E=3
Tt=2
```

**PredictionFBM** 45

PredictionEstParmFunc(paramEstimadosFinal,EspecieMaxima, alpha,K,esperanza,Var,E,Tt)

PredictionFBM

Predicting using FBM

#### **Description**

This function calculates the expected value and variance of the bacteria at time point Tt. Then, this function calculates the expected value and variance of the bacteria at time point t=(Tt+1),...,K

## Usage

```
PredictionFBM(
  paramEstimadosFinal,
 EspecieMaxima,
  alpha,
  esperanza,
  Var,
 Ε,
  Τt
)
```

#### **Arguments**

paramEstimadosFinal

The estimate parameters, in the following order: a11,a12,a13, a21, a22,a23, ...a(D-1)1,a(D-1)2,a(D-1)3,tau. Where D is the number of bacterial species present in the matrix especie.

EspecieMaxima Row in which the bacteria chosen as reference is in especie. This bacteria is

used as reference in the alr transformation that the model does and it is placed

at the denominator of the balance)

alpha Matrix that contains at the row i the Dirichlet parameter of the bacteria i at

t=1,2,3,...,Tt.

Κ Number. The function will calculate the value of the expected value and the

variance at Tt and predict for the time points t=Tt+1,..,K. To predict all the time

points available at the data we K=dim(especie.All)-1

Matrix that contains at row i the expected value of the bacterial taxa of bacteria esperanza

i at t=1,2,3,...,Tt-1.

Matrix that contains at row i the variance of the bacterial taxa of bacteria i at Var

t=1,2,3,...,Tt-1.

Ε Number of bacteria available Τt Number of bacteria available 46 PredictionFBM

#### **Details**

The regression of this model is defined by

```
\mu_{it} = a_{i1} + a_{i2} \cdot \operatorname{alr}(x_{i,(t-1)}) + a_{i3} \cdot \operatorname{Balance}(x_{i,(t-1)}) for i = 1, \dots, D-1 where D is the number of bacteria
```

#### Value

Returns a list with:

- ExpectedValue.All: Matrix. Matrix that contains at row i the expected value of the bacteria i at all time points t=1,2,...,K. The bacteria are placed at the same order than in especies.
- Variance Value. All: Matrix. Matrix that contains at row i the variance of the bacteria i at all time points t=1,2,...,K. The bacteria are placed at the same order than in especies.
- DirichlerParam.All: Matrix. Matrix that contains at row i the dirichlet parameter of the bacteria i at all time points t=1,2,...,K. The bacteria are placed at the same order than in especies.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2021). A Dirichlet autoregressive model for the analysis of microbiota time-series data. Complexity, 2021, 1-16.

PreparingTheData 47

reparingTheData Preparing dataset

#### **Description**

Preparing the dataset to be introduce in the models' functions. In order to introduce the usage of the package there is a README file. You can find the link to the file using base::system.file("extdata", "README.pdf", package = "CoDaLoMic"). On windows you can open the file with base::shell.exec(system.file("ext" README.pdf", package = "CoDaLoMic")).

### Usage

PreparingTheData(DaTa, Pred)

## **Arguments**

DaTa data.frame. The first column contains the time point information (natural num-

bers 1,2,3...). The rest of the columns contain the relative abundance of each bacteria at the different time points. The values of each column must sum 1.

Pred Number. The data at t=1,...,Pred-1 will be used to estimate the model. The rest

of the time points will be used to study the capacity of the model to predict. If

Pred==0 all the dataset will be used to estimate the model.

### Value

If Pred==0 returns a list with

- Tt The number of time points available.
- E Number of bacteria available
- especieOriginal Matrix that contains at row i the bacterial taxa of bacteria i at all time points.
- especiemodiOriginal Matrix that contains at row i the bacterial taxa of bacteria i at time points t=2,...,Tt.

If Pred!=0 returns a list with

- Tt The number of time points available used to estimate the model (Tt=Pred-1).
- E Number of bacteria available
- especieOriginal Matrix that contains at row i the bacterial taxa of bacteria i at the time points t=1,2,...,Pred-1.
- especiemodiOriginal Matrix that contains at row i the bacterial taxa of bacteria i at time points t=2,...,Pred-1.
- especieOriginal.All Matrix that contains at row i the bacterial taxa of bacteria i at the time points.
- especiemodiOriginal.All Matrix that contains at row i the bacterial taxa of bacteria i at time points.
- K Number of time points available at the dataset.

48 QualityControl

#### **Examples**

QualityControl

Analysing the quality of the estimation

## **Description**

This function calculates the root-mean-square deviation (RMSD), the Nash Sutchiffe Coefficient, the residual sum of squares (RSS) and the mean absolute percentage error (MAPE) for the matrices introduces. This function also calculates the mean of the RMSD, the mean of the Nash Sutchiffe Coefficient and the mean of the RSS.

#### Usage

```
QualityControl(matrixData, matrixExpected, names.especie)
```

#### **Arguments**

matrixData Matrix that contains at row i the bacterial taxa of bacteria i at the time points that we want take into account to calculate the quality control values.

matrixExpected Matrix that contains at row i the expected value of the bacterial taxa i at the time points that we want take into account to calculate the quality control values. The bacteria must be placed in the same order than in matrixData

Vector with the names of the bacteria in the same order that are placed in the

matrixData matrix.

#### Value

Returns a data.frame.

names.especie

ridgeregression 49

#### **Examples**

```
names.especie=c("Bact1", "Bact2", "Bact3")
matrixExpected=matrix(c(1:9),3,3)
matrixData=matrixExpected+0.1
QualityControl(matrixData, matrixExpected,names.especie)
```

ridgeregression

Ridge regression

## **Description**

Ridge regression

## Usage

```
ridgeregression(Tt, especie, E, EspecieMaxima, seed = NULL)
```

#### **Arguments**

Tt Number of time points available

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

The bacteria placed in the last row of the matrix will be used as reference in the

alr transformation and will be at the denominator of the balance.

E Number of bacteria available.

EspecieMaxima Row in which the bacteria used as reference is in especie. This is the bacteria

that is going to be at the denominator of the balance and the denominator of the alr transformation. As a result, in this function, EspecieMaxima must be equal

to E

seed Number. Set a seed. Default seed=NULL.

#### Value

Returns the result of the ridge regression, object of class "ridgelm".

```
set.seed(123)
especie=t(gtools::rdirichlet(10,c(1,3,1,2,4)))
Tt=10
E=5
EspecieMaxima=5
```

50 rxnrate

ridgeregression(Tt,especie, E, EspecieMaxima, 558562316)

rxnrate

Solving the right side of the gLV equations

# **Description**

This function calculates the right side of the gLV equation.

# Usage

rxnrate(State, parms)

#### **Arguments**

State

Vector with a CoDa composition

parms

Matrix. Each row has the parameters of each differential equation. following our example, parms has the parameters placed as follows:

#### **Details**

For instance, if we want to solve the following gLV equations:

$$\frac{dx_1(t)}{dt} = r_1 \cdot x_1(t) + x_1(t) \cdot [a_{11} \cdot x_1(t) + a_{12} \cdot x_2(t) + a_{13} \cdot x_3(t)]$$

$$\frac{dx_2(t)}{dt} = r_2 \cdot x_2(t) + x_2(t) \cdot [a_{21} \cdot x_1(t) + a_{22} \cdot x_2(t) + a_{23} \cdot x_3(t)]$$

$$\frac{dx_3(t)}{dt} = r_3 \cdot x_3(t) + x_3(t) \cdot [a_{31} \cdot x_1(t) + a_{32} \cdot x_2(t) + a_{33} \cdot x_3(t)]$$

This function returns a vector with the value of:

$$r_1 \cdot x_1(t) + x_1(t) \cdot [a_{11} \cdot x_1(t) + a_{12} \cdot x_2(t) + a_{13} \cdot x_3(t)]$$

$$r_2 \cdot x_2(t) + x_2(t) \cdot [a_{21} \cdot x_1(t) + a_{22} \cdot x_2(t) + a_{23} \cdot x_3(t)]$$

$$r_3 \cdot x_3(t) + x_3(t) \cdot [a_{31} \cdot x_1(t) + a_{32} \cdot x_2(t) + a_{33} \cdot x_3(t)]$$

#### Value

Returns a vector with the value of the right side of the gLV equations.

Simulated 51

#### **Examples**

```
cinit1<-c(x1<-0.7,x2<-0.2,x3<-0.1)

parms1= cbind(c(0.1,0.2,-0.1),c(-0.2,0.1,-0.1),c(0.3,0.2,0.3),c(0.1,0.22,0.2))

rxnrate(cinit1,parms1)
```

Simulated

Gut microbiome simulated dataset

#### **Description**

Simulated dataset with 5 microbial taxa and 10 time points. Following the scheme given by Faust et al (2018), we generated the interaction matrix using the algorithm proposed by Klemm and Eguíluz (2002), and we generated the initial abundances using the Poisson distribution. With these two tools, we simulated the data using the generalized Lotka-Volterra structure. We carried out the simulation using the R package seqtime (Faust et al, 2018). Focusing on technical details, to generate the interaction matrix we set the clique size at 4, the diagonal values at -1, the interaction connectance at 0.04, the positive edge percentage at 64

#### Usage

data(Simulated)

#### Format

A data frame with 10 rows and 6 columns.

#### References

- K. Faust, F. Bauchinger, B. Laroche et al., "Signatures ofecological processes in microbial community time series". Microbiome, vol. 6, no. 1, p. 120, 2018
- K. Klemm and V. M. Eguíluz, "Growing scale-free networkswith small-world behavior". Physical Review, vol. 65, no. 5, Article ID 057102, 2002.

StudyingParam

Controlling quality of the convergence in BPBM

## Description

This function controls that the value of the Rhat is between 0.9 and 1.1. In addition, it controls that the effective sample size is bigger than 100 and that the zero is not at the center of the credible interval (the interval between 2.5 and 97.5). We consider that the zero is in the center of the credible interval when the zero is between the 25 and the 75 quantile of the distribution formed by the limits if the credible interval.

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#### Usage

```
StudyingParam(Sum, MCMC.CHAINS)
```

#### **Arguments**

Sum Matrix with the summary of the "Estimating\_BPBM".It is the output of the "Es-

timating\_BPBM" adding "\$R2jagsOutput\$BUGSoutput\$summary".

MCMC.CHAINS Matrix with the values of all the Markov chains for all parameters. It is the

output of the "Estimating\_BPBM" adding "\$SamplesAllChains".

#### Value

Returns a list with:

- Param.Summary: Matrix. The matrix Sum with a zero in the column Sum[, "mean"] when a parameter has the zero in the center of its credible interval.
- AllChainsJoined: Matrix. The matrix MCMC.CHAINS with a zero in all the iterations of the chain when a parameter has the zero in the center of its credible interval.

## **Examples**

Studying Param (est R2jags Output BUGS output summary, est Samples All Chains)

TableBPBM

Obtaining a table with the SPBal information of the BPBM

## **Description**

Returns a table with the percentage of variance that each SPBal has, the bacteria that goes in the numerator and denominator of the balance, the relationship between the group in the numerator and the denominator and the bacteria most influenced by this SPBal.

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#### Usage

```
TableBPBM(
   NumSPBal,
   DemSPBal,
   PerVar,
   MatrizPBmodelo,
   Estimated.Param,
   BB = 0.55,
   names,
   E
)
```

#### **Arguments**

NumSPBal List. Output of "ObtainigValueSPBal" function.List. The component i of the

list has the number of the row of the matrix especie where the bacteria in the

numerator of the selected principal balance i are placed.

DemSPBal List. Output of "ObtainigValueSPBal" function.List. The component i of the

list has the number of the row of the matrix especie where the bacteria in the

denominator of the selected principal balance i are placed.

PerVar Vector. Output of "ObtainigValueSPBal" function. The component of the vector

i contains the percentage of variance of the SPBal with numerator NumSPBal[[i]]

and denominator DemSPBal[[i]].

MatrizPBmodelo Matrix. Output of "ObtainigValueSPBal" function. MatrixSPBal is the matrix

that contains the covariates of the model. The first line es equal to 1 for all columns. The other rows contain the value of one SPBal at all time points. The selected principal balance of the row i+1 has at its numerator the bacteria placed in the rows NumSPBal[[i]] of the "especie". The selected principal balance of the row i+1 has at its denominator the bacteria placed in the rows

DemSPBal[[i]] of the "especie".

Estimated.Param

Vector. Column "mean" of the output of StudyingParam function.

BB The bacteria in the numerator and the denominator of the balance are considered

similar if the mean of the SPBal is between (-BB,BB).Default: 0.55.

names Vector with the bacteria's names placed in the order in which appear in the rows

of the matrices especie and especiemodi.

E number of bacteria in the dataset.

#### Value

Returns the table written in latex

```
NumSPBal=list(c(3,4),3,2)
DemSPBal=list(1,5,4)
PerVar=c(41.37487, 21.08270, 19.16870)
```

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TableFBM

Obtainig a table with the interpretable parameters

#### **Description**

This function returns a table with the interpretable parameters of the FBM model.

#### Usage

```
TableFBM(paramEstimadosFinal, names, E)
```

## **Arguments**

paramEstimadosFinal

The estimate parameters, in the following order: a11,a12,a13, a21, a22,a23, ...a(D-1)1,a(D-1)2,a(D-1)3,tau. Where D is the number of bacterial species present in the matrix especie.

names

Vector of length D. The component i has the name of the bacteria i.

Ε

Number of bacteria available.

# Details

```
\mu_{it} = a_{i1} + a_{i2} \cdot \operatorname{alr}(x_{i,(t-1)}) + a_{i3} \cdot \operatorname{Balance}(x_{i,(t-1)}) for i = 1, \dots, D-1 where D is the number of bacteria
```

#### Value

Returns a table written in latex format.

Table\_alr\_Dirich\_glv 55

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2021). A Dirichlet autoregressive model for the analysis of microbiota time-series data. Complexity, 2021, 1-16.

#### **Examples**

```
paramEstimadosFinal=c(1,2,3,1,2,3,1,2,3)
names=c("Bact1", "Bact2","Bact3")
E=3
TableFBM(paramEstimadosFinal,names,E)
```

## **Description**

This function returns a table with the interpretable parameters of the Dirich-gLV model.

## Usage

```
Table_alr_Dirich_glv(Param.Estimates, especie, names, E)
```

## **Arguments**

Param.Estimates

Vector with the estimates parameters. It is equal to c(tau, as.vector(pam)) where:

• pam Matrix. Each row has the parameters of each bacteria. Following our example, pam has the parameters placed as follows:

• tau Number. Value of the tau parameter in the model

especie	Matrix that contains at row i the bacterial taxa of bacteria i at all time points.
	The bacteria placed in the last row of this matrix is the one used as reference in
	the alr transformation that the model applies.
names	Vector with the name of the bacteria in the same order than are present in the especie matrix.

E Number of bacteria available.

#### **Details**

In an example with three bacteria, the regression of this model is defined by

```
r_1 \cdot log(x_1(t)/x_3(t)) + log(x_1(t)/x_3(t)) \cdot [a_{11} \cdot log(x_1(t)/x_3(t))(t) + a_{12} \cdot log(x_2(t)/x_3(t))]
r_2 \cdot log(x_2(t)/x_3(t)) + log(x_2(t)/x_3(t)) \cdot [a_{21} \cdot log(x_1(t)/x_3(t))(t) + a_{22} \cdot log(x_2(t)/x_3(t))]
```

#### Value

Returns a table written in latex format with the matrix pam.

#### References

Creus-Martí, I. and Moya, A. and Santonja, F. J. (2018). A Statistical Model with a Lotka-Volterra Structure for Microbiota Data. Lucas Jodar, Juan Carlos Cortes and Luis Acedo, Modelling or engineering and human behavior 2018, Instituto Universitario de Matematica Multidisciplinar. ISBN: 978-84-09-07541-6

## **Examples**

```
\label{eq:pam.ini=rbind} \begin{split} &\text{pam.ini=rbind}(c(\emptyset.1,\emptyset.2,\emptyset.3),c(\emptyset.4,\emptyset.5,\emptyset.6)) \\ &\text{paramEstimadosFinal=c(5, as.vector(pam.ini))} \\ &\text{E=3} \\ &\text{especie=cbind}(c(\emptyset.2,\emptyset.4,\emptyset.4),c(\emptyset.1,\emptyset.1,\emptyset.8),c(\emptyset.5,\emptyset.1,\emptyset.4)) \\ &\text{names=c("a","b","c")} \\ &\text{Table\_alr\_Dirich\_glv(paramEstimadosFinal,especie,names,E)} \end{split}
```

TauAndParameters\_EstParmFunc\_FBM

Obtaining the value of tau and the estimate value of the rest of the parameters

#### **Description**

This function estimates the parameters of the FBM model.

# Usage

```
TauAndParameters_EstParmFunc_FBM(
  ttau = 30,
  ridge.final,
  Iter.EstParmFunc = 80,
  especie,
  EspecieMaxima,
```

```
Tt,
E,
seed = NULL
)
```

#### **Arguments**

Number. We estimate de FBM model for the values of tau: 1, 2,..., ttau ridge.final Object of class "ridgelm". Values obtained with the ridge regression.

Iter.EstParmFunc

Number. Number of iterations. Default: 80 iterations.

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

The bacteria placed in the last row of the matrix will be used as reference in the

alr transformation and will be at the denominator of the balance.

EspecieMaxima Row in which the bacteria used as reference is in especie. This is the bacteria

that is going to be at the denominator of the balance and at the denominator of the alr transformartion. As a result, in this function, EspecieMaxima must be

equal to E

Tt Number of time points available

E Number. Number of bacteria available. seed Number. Set a seed. Default seed=NULL.

#### **Details**

We give to the parameter tau the value 1,2,...,ttau. We estimate the FBM model for all this values (using the function "Estimate\_param\_FBM") and we select the value of tau that minimizes the AIC. The regression of this model is defined by

```
\mu_{it} = a_{i1} + a_{i2} \cdot \operatorname{alr}(x_{i,(t-1)}) + a_{i3} \cdot \operatorname{Balance}(x_{i,(t-1)}) for i = 1, \dots, D-1 where D is the number of bacteria
```

#### Value

Returns a list with:

- EstimateParameters: Vector with the estimated parameters, in the following order: a11,a12,a13, a21, a22,a23, ...a(D-1)1,a(D-1)2,a(D-1)3,tau. Where D is the number of bacterial species present in the matrix especie.
- AIC Number: Value of the AIC.
- All.iter: Matrix. Each row has the parameters obtained in each iteration. The parameters are in the columns written in the same order that they are written in Param.Estimates. In this matrix we must observe that in the last iterations the values has really similar or equal values, if not, we need to increase the value of Iter.EstParmFunc.

#### References

Creus-Martí, I., Moya, A., Santonja, F. J. (2021). A Dirichlet autoregressive model for the analysis of microbiota time-series data. Complexity, 2021, 1-16.

58 vecttor

#### **Examples**

```
set.seed(123)
especie=t(gtools::rdirichlet(5,c(1,3,1)))
Tt=5
E=3
EspecieMaxima=3
ridge.final=ridgeregression(Tt,especie, E, EspecieMaxima)
ttau=10
Iter.EstParmFunc=10
TauAndParameters_EstParmFunc_FBM(ttau,ridge.final,Iter.EstParmFunc, especie,EspecieMaxima,Tt,E,714)
```

vecttor

Alr of a bacteria

## **Description**

Writtes a vector with the alr transformation of the bacteria i at time points t=2,...,Tt.

#### Usage

```
vecttor(i, especie, Tt, EspecieMaxima)
```

## **Arguments**

i Number. Position of the bacteria that we make the alr in the matrix especie. i must be different that EspecieMaxima.

especie Matrix that contains at row i the bacterial taxa of bacteria i at all time points.

The bacteria placed in the last row of the matrix will be used as reference in the

alr transformation and will be at the denominator of the balance.

Tt Number of time points available

EspecieMaxima Row in which the bacteria used as reference is in especie. This is the bacteria

that is going to be at the denominator of the balance and the denominator of the alr transformation. As a result, in this function, EspecieMaxima must be equal

to E

#### Value

Returns a vector with the alr transformation of the bacteria i at time points t=2,...,Tt.

```
set.seed(123)
especie=t(gtools::rdirichlet(10,c(1,3,1,2,4)))
Tt=10
EspecieMaxima=5
```

ZeroData 59

i=2

vecttor(i,especie,Tt,EspecieMaxima)

ZeroData

Zero replacement

#### **Description**

In this function the zeros are removed or replaced using functions of "zCompositions" package that can be used with longitudinal data (because they do not use the information of other rows to make the replacement).

## Usage

```
ZeroData(DaTa, method = "multKM", seed = NULL)
```

#### **Arguments**

DaTa

data.frame. The first column contains the time point information (natural numbers 1,2,3...). The rest of the columns contain the relative abundance of each bacteria at the different time points. The values of each column must sum 1.

method

Character.

- If method="multKM" The replacement is carried out with the "multiplicative Kaplan-Meier smoothing spline replacement" (Palarea-Albaladejo and Martín-Fernandez, 2015). Default method. The zeros must be written with a 0.
- If method="multRepl" The replacement is carried out with the "multiplicative simple replacement" (Palarea-Albaladejo and Martín-Fernandez, 2015). The zeros must be written with a 0.
- If method="nozeros" The bacteria that contains zeros are removed. One column is added to the dataset called "Other".

seed

Number. Set a seed. Default seed=NULL.

#### Value

The dataset without zeros.

#### References

Palarea-Albaladejo J. and Martín-Fernandez JA. zCompositions – R package for multivariate imputation of left-censored data under a compositional approach. Chemometrics and Intelligent Laboratory Systems 2015; 143: 85-96.

ZeroData

```
set.seed(2)
dat=gtools::rdirichlet(6,c(1,2,3,1,2,3))
dat2=dat
dat2[2,1]=0
dat2[2,2]=dat[2,1]+dat[2,2]
dat2[4,3]=0
dat2[4,4]=dat[4,3]+dat[4,4]

X <- cbind( c(1:6) ,dat2)

Final=ZeroData(X,"multKM",1)
Final2=ZeroData(X,"multRepl",1)</pre>
```

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