

Package ‘parafac4microbiome’

May 20, 2025

Title Parallel Factor Analysis Modelling of Longitudinal Microbiome Data

Version 1.2.1

Description Creation and selection of PARAllel FACtor Analysis (PARAFAC) models of longitudinal microbiome data. You can import your own data with our import functions or use one of the example datasets to create your own PARAFAC models. Selection of the optimal number of components can be done using `assessModelQuality()` and `assessModelStability()`. The selected model can then be plotted using `plotPARAFACmodel()`. The Parallel Factor Analysis method was originally described by Carroll and Chang (1970) <[doi:10.1007/BF02310791](https://doi.org/10.1007/BF02310791)> and Harshman (1970) <<https://www.psychology.uwo.ca/faculty/harshman/wpppfac0.pdf>>.

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<https://github.com/GRvanderPloeg/parafac4microbiome/>

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assessModelQuality *Create randomly initialized models to determine the correct number of components by assessing model quality metrics.*

Description

Create randomly initialized models to determine the correct number of components by assessing model quality metrics.

Usage

```
assessModelQuality(
  X,
  minNumComponents = 1,
  maxNumComponents = 5,
  numRepetitions = 100,
  method = "als",
  ctol = 1e-06,
  maxit = 2500,
  max_fn = 10000,
  rel_tol = 1e-08,
  abs_tol = 1e-08,
  grad_tol = 1e-08,
  numCores = 1
)
```

Arguments

- X Input data
- minNumComponents Minimum number of components (default 1).
- maxNumComponents Maximum number of components (default 5).
- numRepetitions Number of randomly initialized models to create (default 100).
- method Use ALS algorithm ("als", default) or use all-at-once optimization ("opt"). The all-at-once optimization is based on a nonlinear conjugate gradient method with Hestenes-Stiefel updates and the More-Thuente line search algorithm.
- ctol Change in SSQ needed for model to be converged (default 1e-6).
- maxit Maximum number of iterations (default 2500).
- max_fn Maximum number of function evaluations allowed without convergence in the OPT case (default 10000).
- rel_tol Relative change in loss tolerated to call the algorithm converged in the OPT case (default 1e-8).

| | |
|----------|---|
| abs_tol | Absolute loss tolerated to call the algorithm converged in the OPT case (default 1e-8). |
| grad_tol | Tolerance on the two-norm of the gradient divided over the number of elements in the gradient in the OPT case (default 1e-8). |
| numCores | Number of cores to use. If set larger than 1, it will run the job in parallel (default 1) |

Value

A list object of the following:

- plots: Plots of all assessed metrics and an overview plot showing a summary of all of them.
- metrics: metrics of every created model (number of iterations, sum of squared errors, CORCONDIA score and variance explained).
- models: all created models.

Examples

```
X = Fujita2023$data

# Run assessModelQuality with less strict convergence parameters as example
assessment = assessModelQuality(X,
                                minNumComponents=1,
                                maxNumComponents=3,
                                numRepetitions=5,
                                ctol=1e-4,
                                maxit=250)

assessment$plots$overview
```

assessModelStability *Bootstrapping procedure to determine PARAFAC model stability for a given number of components.*

Description

Bootstrapping procedure to determine PARAFAC model stability for a given number of components.

Usage

```
assessModelStability(
  dataset,
  minNumComponents = 1,
  maxNumComponents = 5,
  numFolds = dim(dataset$data)[1],
  considerGroups = FALSE,
  groupVariable = "",
```

```

colourCols = NULL,
legendTitles = NULL,
xLabels = NULL,
legendColNums = NULL,
arrangeModes = NULL,
method = "als",
ctol = 1e-06,
maxit = 2500,
max_fn = 10000,
rel_tol = 1e-08,
abs_tol = 1e-08,
grad_tol = 1e-08,
numCores = 1
)

```

Arguments

| | |
|------------------|--|
| dataset | See Fujita2023 , Shao2019 or vanderPloeg2024 . |
| minNumComponents | Minimum number of components (default 1). |
| maxNumComponents | Maximum number of components (default 5). |
| numFolds | Number of bootstrapped models to create. |
| considerGroups | Consider subject groups in calculating sparsity (default FALSE) |
| groupVariable | Column name in dataset\$mode1 that should be used to consider groups (default "") |
| colourCols | Vector of strings stating which column names should be factorized for colours per mode. |
| legendTitles | Vector of strings stating the legend title per mode. |
| xLabels | Vector of strings stating the x-axis labels per mode. |
| legendColNums | Vector of integers stating the desired number of columns for the legends per mode. |
| arrangeModes | Vector of boolean values per mode, stating if the loadings should be arranged according to colourCols (TRUE) or not (FALSE). |
| method | Use ALS algorithm ("als", default) or use all-at-once optimization ("opt"). The all-at-once optimization is based on a nonlinear conjugate gradient method with Hestenes-Stiefel updates and the More-Thuente line search algorithm. |
| ctol | Relative change in loss tolerated to call the algorithm converged in the ALS case (default 1e-4). |
| maxit | Maximum number of iterations allowed without convergence in the ALS case (default 500). |
| max_fn | Maximum number of function evaluations allowed without convergence in the OPT case (default 10000). |
| rel_tol | Relative change in loss tolerated to call the algorithm converged in the OPT case (default 1e-8). |

| | |
|----------|---|
| abs_tol | Absolute loss tolerated to call the algorithm converged in the OPT case (default 1e-8). |
| grad_tol | Tolerance on the two-norm of the gradient divided over the number of elements in the gradient in the OPT case (default 1e-8). |
| numCores | Number of cores to use. If set larger than 1, it will run the job in parallel (default 1) |

Value

A list containing the following:

- **models**: All stabilized sign-flipped bootstrapped PARAFAC models.
- **modelPlots**: A list of plots of the median model with error bars for each number of components.
- **FMSplot**: A bar plot showing the Factor Match Scores per number of components (see Li et al., 2024).
- **FMS**: FMS values that the FMS plot is based on.

Examples

```
processedFujita = processDataCube(Fujita2023, sparsityThreshold=0.99, centerMode=1, scaleMode=2)
modelStability = assessModelStability(processedFujita,
                                     minNumComponents=1,
                                     maxNumComponents=2,
                                     ctol=1e-4,
                                     maxit=250)
```

| | |
|--------------|---|
| calculateFMS | <i>Calculate Factor Match Score for all initialized models.</i> |
|--------------|---|

Description

Calculate Factor Match Score for all initialized models.

Usage

```
calculateFMS(models)
```

Arguments

models Output of `parafac()` using `output="all"`.

Value

Vector containing FMS scores of all comparisons

Examples

```
A = array(rnorm(108*2), c(108, 2))
B = array(rnorm(100*2), c(100, 2))
C = array(rnorm(10*2), c(10, 2))
X = reinflaTensor(A, B, C)
models = parafac(X, 2, initialization="random", nstart=10, maxit=2, output="all")
calculateFMS(models)
```

| | |
|-------------------|---|
| calculateSparsity | <i>Calculate sparsity across the feature mode of a multi-way array.</i> |
|-------------------|---|

Description

Calculate sparsity across the feature mode of a multi-way array.

Usage

```
calculateSparsity(dataset, considerGroups = FALSE, groupVariable = "")
```

Arguments

| | |
|----------------|--|
| dataset | See Fujita2023 , Shao2019 or vanderPloeg2024 . |
| considerGroups | Consider subject groups in calculating sparsity (default FALSE) |
| groupVariable | Column name in dataset\$model that should be used to consider groups (default "") |

Value

Vector of sparsity fractions (N x J) where N is the number of groups and J is the number of features.

Examples

```
# No groups
sparsity = calculateSparsity(Fujita2023)
length(sparsity)
hist(sparsity)

# Consider groups
colnames(Shao2019$model)
sparsity = calculateSparsity(Shao2019, considerGroups=TRUE, groupVariable="Delivery_mode")
dim(sparsity)
hist(sparsity[1,])
hist(sparsity[2,])
```

| | |
|-----------------|--|
| calculateVarExp | <i>Calculate the variation explained by a PARAFAC model.</i> |
|-----------------|--|

Description

Calculate the variation explained by a PARAFAC model.

Usage

```
calculateVarExp(Fac, X)
```

Arguments

| | |
|-----|---|
| Fac | Fac object output from the <code>parafac()</code> function. |
| X | Input data of the PARAFAC model. |

Value

The variation explained by the model, expressed as a fraction (between 0-1).

Examples

```
X = Fujita2023$data
model = parafac(X, nfac=1, nstart=1, verbose=FALSE)
calculateVarExp(model$Fac, X)
```

| | |
|------------------------|---|
| calcVarExpPerComponent | <i>Calculate the variance explained of a PARAFAC model, per component</i> |
|------------------------|---|

Description

Calculate the variance explained of a PARAFAC model, per component

Usage

```
calcVarExpPerComponent(Fac, X)
```

Arguments

| | |
|-----|------------------------------|
| Fac | Fac object output of a model |
| X | Input dataset |

Value

Vector of scalars of the percentage of variation explained per component

Examples

```
X = array(rnorm(108*100*10), c(108,100,10))
model = parafac(X, 2)
calcVarExpPerComponent(model$Fac, X)
```

corcondia

Core Consistency Diagnostic (CORCONDIA) calculation

Description

Core Consistency Diagnostic (CORCONDIA) calculation

Usage

```
corcondia(X, Fac)
```

Arguments

| | |
|-----|--------------------------|
| X | Input data matrix |
| Fac | PARAFAC model Fac object |

Value

Scalar of the CORCONDIA value

Examples

```
X = Fujita2023$data
model = parafac(X, 2)
corcondia(X, model$Fac)
```

fac_to_vect

Vectorize Fac object

Description

Vectorize Fac object

Usage

```
fac_to_vect(Fac)
```

Arguments

| | |
|-----|--|
| Fac | Fac object output of parafac . |
|-----|--|

Value

Vectorized Fac object

Examples

```
set.seed(123)
A = array(rnorm(108*2), c(108, 2))
B = array(rnorm(100*2), c(100, 2))
C = array(rnorm(10*2), c(10, 2))
Fac = list(A, B, C)
v = fac_to_vect(Fac)
```

flipLoadings

Sign flip the loadings of many randomly initialized models to make consistent overview plots.

Description

Sign flip the loadings of many randomly initialized models to make consistent overview plots.

Usage

```
flipLoadings(models, X)
```

Arguments

models Output of [parafac](#).
X Input dataset of parafac modelling procedure.

Value

models with sign flipped components where applicable.

Examples

```
A = array(rnorm(108*2), c(108,2))
B = array(rnorm(100*2), c(100,2))
C = array(rnorm(10*2), c(10,2))
X = reinflateTensor(A, B, C)
models = parafac(X, 2, nstart=10, output="all", sortComponents=TRUE)
flippedModels = flipLoadings(models, X)
```

Fujita2023

Fujita2023 longitudinal microbiome data

Description

The Fujita2023 longitudinal microbiome dataset as a three-dimensional array, with replicates in mode 1, microbial abundances in mode 2 and time in mode 3.

Usage

```
Fujita2023
```

Format

Fujita2023:

A list object with three elements:

data Array object of the data cube

mode1 Dataframe with all the subject metadata, ordered the same as the rows in the data cube.

mode2 Taxonomic classification of the microbiota, ordered the same as the columns in the data cube.

mode3 Dataframe with the time metadata, ordered the same as the third dimension in the array.

...

Source

[doi:10.1186/s40168023014745](https://doi.org/10.1186/s40168023014745)

```
importMicrobiotaProcess
```

Import MicrobiotaProcess object for PARAFAC modelling

Description

Import MicrobiotaProcess object for PARAFAC modelling

Usage

```
importMicrobiotaProcess(MPobject, subjectIDs, thirdMode, taxa_are_rows = TRUE)
```

Arguments

| | |
|---------------|---|
| MPObject | MicrobiotaProcess object containing at least an OTU table and sample information, preferably also taxonomic information. |
| subjectIDs | Column name in the sample information corresponding to the subject IDs. |
| thirdMode | Column name in the sample information corresponding to the study design aspect to put in the third mode of the data cube. |
| taxa_are_rows | Boolean specifying if the taxa are in the rows of the OTU table (TRUE) or not (FALSE). |

Value

List object containing:

- 'data': data cube
- 'mode1': metadata of the subject mode
- 'mode2': taxonomy information
- 'mode3': metadata of the third mode

Examples

```
## Not run:
library(MicrobiotaProcess)
data("mouse.time.mpse")
importMicrobiotaProcess(mouse.time.mpse, subjectIDs = "Sample",
                        thirdMode = "time", taxa_are_rows = TRUE)

## End(Not run)
```

```
importPhyloseq      Import Phyloseq object for PARAFAC modelling
```

Description

Import Phyloseq object for PARAFAC modelling

Usage

```
importPhyloseq(phyloseqObject, subjectIDs, thirdMode)
```

Arguments

| | |
|----------------|---|
| phyloseqObject | Phyloseq object containing at least an otu table and sample data, preferably also taxonomic information. |
| subjectIDs | Column name in sam_data corresponding to the subject IDs. |
| thirdMode | Column name in sam_data corresponding to the study design aspect to put in the third mode of the data cube. |

Value

List object containing:

- 'data': data cube
- 'mode1': metadata of the subject mode
- 'mode2': taxonomy information
- 'mode3': metadata of the third mode

Examples

```
library(phyloseq)
data(GlobalPatterns)
GP = GlobalPatterns

# Add custom subject IDs to the sample data to make this example work
alteredSampleData = sample_data(GP)
alteredSampleData$subjectID = c(1,2,3,1,2,1,2,3,1,2,1,2,1,2,3,1,2,3,1,2,3,4,5,1,2,3)
df = phyloseq(otu_table(GP), tax_table(GP), alteredSampleData)

# Make a data cube with SampleType (soil, feces, etc.) as the third mode.
result = importPhyloseq(df, subjectIDs = "subjectID", thirdMode="SampleType")
```

```
importTreeSummarizedExperiment
  Import TreeSummarizedExperiment object for PARAFAC modelling
```

Description

Import TreeSummarizedExperiment object for PARAFAC modelling

Usage

```
importTreeSummarizedExperiment(
  treeObject,
  subjectIDs,
  thirdMode,
  taxa_are_rows
)
```

Arguments

| | |
|---------------|---|
| treeObject | TreeSummarizedExperiment object containing at least an OTU table and sample information, preferably also taxonomic information. |
| subjectIDs | Column name in the sample information corresponding to the subject IDs. |
| thirdMode | Column name in the sample information corresponding to the study design aspect to put in the third mode of the data cube. |
| taxa_are_rows | Boolean specifying if the taxa are in the rows of the OTU table (TRUE) or not (FALSE). |

Value

List object containing:

- 'data': data cube
- 'model': metadata of the subject mode
- 'mode2': taxonomy information
- 'mode3': metadata of the third mode

Examples

```
library(TreeSummarizedExperiment)

fakeOTU = t(rTensor::k_unfold(rTensor::as.tensor(Fujita2023$data), 2)@data)
fakeTaxa = as.matrix(Fujita2023$mode2)
fakeSam = as.data.frame(cbind(rep(1:8, 110), rep(1:110, each=8)))
colnames(fakeSam) = c("replicate.id", "timepoint")

fakeTreeObj = TreeSummarizedExperiment(assays = list(Count = fakeOTU),
                                       rowData = fakeSam,
                                       colData = fakeTaxa)
dataset = importTreeSummarizedExperiment(fakeTreeObj,
                                       subjectIDs="replicate.id",
                                       thirdMode="timepoint",
                                       taxa_are_rows=FALSE)
```

```
initializePARAFAC      Initialize PARAFAC algorithm input vectors
```

Description

Initialize PARAFAC algorithm input vectors

Usage

```
initializePARAFAC(Tensor, nfac, initialization = "random", output = "Fac")
```

Arguments

| | |
|----------------|---|
| Tensor | Input dataset matrix or tensor |
| nfac | Number of components to initialize. |
| initialization | Either "random" for random initialization or "svd" for svd based. |
| output | Output the initialized components as a Fac object ("Fac", default) or as a vector ("vect"). |

Value

Fac or vector with initialized components.

Examples

```
A = array(rnorm(108,2), c(108,2))
B = array(rnorm(100,2), c(100,2))
C = array(rnorm(10,2), c(10,2))
Tensor = reinflateTensor(A, B, C, returnAsTensor=TRUE)
init = initializePARAFAC(Tensor, 2)
```

| | |
|----------------|---------------------------------|
| multiwayCenter | <i>Center a multi-way array</i> |
|----------------|---------------------------------|

Description

Center a multi-way array

Usage

```
multiwayCenter(X, mode = 1)
```

Arguments

| | |
|------|------------------------------------|
| X | Multi-way array |
| mode | Mode to center across (default 1). |

Value

Centered multi-way array

Examples

```
cube_cnt = multiwayCenter(Fujita2023$data)
```

| | |
|-------------|--|
| multiwayCLR | <i>Perform a centered log-ratio transform over a multi-way array</i> |
|-------------|--|

Description

Note: Propagates NAs corresponding to missing samples.

Usage

```
multiwayCLR(X, pseudocount = 1)
```

Arguments

| | |
|-------------|---------------------------------------|
| X | Multi-way array of counts |
| pseudocount | Pseudocount value to use (default 1). |

Value

CLRed cube

Examples

```
cubeCLR = multiwayCLR(Fujita2023$data)
```

| | |
|---------------|--------------------------------|
| multiwayScale | <i>Scale a multi-way array</i> |
|---------------|--------------------------------|

Description

Scale a multi-way array

Usage

```
multiwayScale(X, mode = 2)
```

Arguments

| | |
|------|---|
| X | Multi-way array |
| mode | Mode to scale within: 1=subjects,2=features,3=time (default 2). |

Value

Scaled multi-way array

Examples

```
cube_scl = multiwayCenter(Fujita2023$data)
```

| | |
|---------|---------------------------------|
| parafac | <i>Parallel Factor Analysis</i> |
|---------|---------------------------------|

Description

Parallel Factor Analysis

Usage

```

parafac(
  Tensor,
  nfac,
  nstart = 1,
  maxit = 500,
  max_fn = 10000,
  ctol = 1e-04,
  rel_tol = 1e-08,
  abs_tol = 1e-08,
  grad_tol = 1e-08,
  initialization = "random",
  method = "als",
  verbose = FALSE,
  output = "best",
  sortComponents = FALSE
)

```

Arguments

| | |
|----------------|--|
| Tensor | 3-way matrix of numeric data |
| nfac | Number of factors (components) to fit. |
| nstart | Number of models to randomly initialize (default 1). |
| maxit | Maximum number of iterations allowed without convergence in the ALS case (default 500). |
| max_fn | Maximum number of function evaluations allowed without convergence in the OPT case (default 10000). |
| ctol | Relative change in loss tolerated to call the algorithm converged in the ALS case (default 1e-4). |
| rel_tol | Relative change in loss tolerated to call the algorithm converged in the OPT case (default 1e-8). |
| abs_tol | Absolute loss tolerated to call the algorithm converged in the OPT case (default 1e-8). |
| grad_tol | Tolerance on the two-norm of the gradient divided over the number of elements in the gradient in the OPT case (default 1e-8). |
| initialization | "Random" for randomly initialized input vectors or "nvec" for svd-based best guess. |
| method | Use ALS algorithm ("als", default) or use all-at-once optimization ("opt"). The all-at-once optimization is based on a nonlinear conjugate gradient method with Hestenes-Stiefel updates and the More-Thuente line search algorithm. |
| verbose | [Deprecated] verbose output |
| output | String ("best"/"all") Return only the best model of the nstart models ("best") or return all of them in a list object ("all"). |
| sortComponents | Boolean to sort the components based on their variance explained (default FALSE) |

Value

List object of the PARAFAC model or models.

Examples

```
X = array(rnorm(108*100*10), c(108,100,10))
model = parafac(X, 2)
```

parafac_core_als *Internal PARAFAC alternating least-squares (ALS) core algorithm*

Description

Internal PARAFAC alternating least-squares (ALS) core algorithm

Usage

```
parafac_core_als(Tensor, nfac, init, maxit = 500, ctol = 1e-04)
```

Arguments

| | |
|--------|---|
| Tensor | Tensor data object |
| nfac | Number of components to compute |
| init | Initialization from initializePARAFAC . |
| maxit | Maximum number of iterations to run (default 500). |
| ctol | Loss function tolerance for convergence (default 1e-4) |

Value

List containing a Fac object and the loss per iteration

Examples

```
A = array(rnorm(108*2), c(108,2))
B = array(rnorm(100*2), c(100,2))
C = array(rnorm(10*2), c(10,2))
Tensor = reinflateTensor(A, B, C)
init = initializePARAFAC(Tensor, 2)
model = parafac_core_als(Tensor, 2, init)
```

| | |
|-------------|--|
| parafac_fun | <i>PARAFAC loss function calculation</i> |
|-------------|--|

Description

PARAFAC loss function calculation

Usage

```
parafac_fun(x, Tensor, lambdas = NULL)
```

Arguments

| | |
|---------|---|
| x | Vector of fitted loadings generated by the PARAFAC algorithm, can also be a Fac object |
| Tensor | input data |
| lambdas | If lambdas (from the kruskal tensor case) are generated to make the Fac norm 1, they can be supplied. |

Value

Scalar value of the loss function

Examples

```
A = array(rnorm(108*2), c(108,2))
B = array(rnorm(100*2), c(100,2))
C = array(rnorm(10*2), c(10,2))
X = reinflateTensor(A, B, C)
model = parafac(X, 2)
f = parafac_fun(model$Fac, X)
```

| | |
|------------------|---|
| parafac_gradient | <i>Calculate gradient of PARAFAC model.</i> |
|------------------|---|

Description

Calculate gradient of PARAFAC model.

Usage

```
parafac_gradient(x, Tensor)
```

Arguments

| | |
|--------|--|
| x | Vector of fitted loadings generated by the PARAFAC algorithm, can also be a Fac object |
| Tensor | input data |

Value

Gradient of the PARAFAC model.

Examples

```
A = array(rnorm(108*2), c(108,2))
B = array(rnorm(100*2), c(100,2))
C = array(rnorm(10*2), c(10,2))
X = reinflateTensor(A, B, C)
init = initializePARAFAC(X, 2)
g = parafac_gradient(init, X)
```

plotModelMetric *Plot diagnostics of many initialized PARAFAC models.*

Description

Plot diagnostics of many initialized PARAFAC models.

Usage

```
plotModelMetric(
  metric,
  plottingMode = "box",
  ylabel = "metric",
  titleString = ""
)
```

Arguments

| | |
|--------------|--|
| metric | Matrix of metrics per initialized model (number of models x number of components). |
| plottingMode | Plot the metrics as a box plot ("box", default) or as a bar plot ("bar"). |
| ylabel | String of the y axis label (default "metric"). |
| titleString | String of the plot title (default ""). |

Value

A plot of the metrics

Examples

```
varExp = array(runif(100*2, min=50, max=100), c(100,2))
plotModelMetric(varExp, plottingMode="box", ylabel="Variation explained (%)")
```

plotModelStability *Plot a summary of the loadings of many initialized parafac models.*

Description

Plot a summary of the loadings of many initialized parafac models.

Usage

```
plotModelStability(
  models,
  dataset,
  colourCols = NULL,
  legendTitles = NULL,
  xLabels = NULL,
  legendColNums = NULL,
  arrangeModes = NULL,
  continuousModes = NULL,
  overallTitle = ""
)
```

Arguments

| | |
|---------------|--|
| models | Models list output from <code>parafac()</code> using <code>output="all"</code> . |
| dataset | A longitudinal microbiome dataset, ideally processed with <code>processDataCube()</code> , formatted as follows: <ul style="list-style-type: none"> data Array object of the data cube mode1 Dataframe with all the subject metadata, ordered the same as the rows in the data cube. mode2 Taxonomic classification of the microbiota, ordered the same as the columns in the data cube. mode3 Dataframe with the time metadata, ordered the same as the third dimension in the array. |
| colourCols | Vector of strings stating which column names should be factorized for colours per mode. |
| legendTitles | Vector of strings stating the legend title per mode. |
| xLabels | Vector of strings stating the x-axis labels per mode. |
| legendColNums | Vector of integers stating the desired number of columns for the legends per mode. |

arrangeModes Vector of boolean values per mode, stating if the loadings should be arranged according to colourCols (TRUE) or not (FALSE).
continuousModes Vector of boolean values per mode, stating if the loadings should be plotted as a line plot (TRUE) or a bar plot (FALSE).
overallTitle Overall title of the plot.

Value

Plot of loadings with error bars

Examples

```

processedFujita = processDataCube(Fujita2023, sparsityThreshold=0.99, centerMode=1, scaleMode=2)
models = parafac(processedFujita$data, 2, nstart=10, output="all")
plotModelStability(models, processedFujita)
  
```

| | |
|----------------------------|---|
| <code>plotModelTCCs</code> | <i>Plots Tucker Congruence Coefficients of randomly initialized models.</i> |
|----------------------------|---|

Description

Plots Tucker Congruence Coefficients of randomly initialized models.

Usage

```
plotModelTCCs(models)
```

Arguments

`models` Models list output of `parafac()` using `output="all"`.

Value

Plot of TCCs

Examples

```

processedFujita = processDataCube(Fujita2023, sparsityThreshold=0.99, centerMode=1, scaleMode=2)
models = parafac(processedFujita$data, 3, nstart=10, output="all")
plotModelTCCs(models)
  
```

plotPARAFACmodel *Plot a PARAFAC model*

Description

Plot a PARAFAC model

Usage

```
plotPARAFACmodel(
  model,
  dataset,
  numComponents,
  colourCols = NULL,
  legendTitles = NULL,
  xLabels = NULL,
  legendColNums = NULL,
  arrangeModes = NULL,
  continuousModes = NULL,
  overallTitle = ""
)
```

Arguments

| | |
|---------------|--|
| model | Model output from parafac() . |
| dataset | A longitudinal microbiome dataset, ideally processed with processDataCube() , formatted as follows: data Array object of the data cube mode1 Dataframe with all the subject metadata, ordered the same as the rows in the data cube. mode2 Taxonomic classification of the microbiota, ordered the same as the columns in the data cube. mode3 Dataframe with the time metadata, ordered the same as the third dimension in the array. |
| numComponents | Number of PARAFAC components in the model. |
| colourCols | Vector of strings stating which column names should be factorized for colours per mode. |
| legendTitles | Vector of strings stating the legend title per mode. |
| xLabels | Vector of strings stating the x-axis labels per mode. |
| legendColNums | Vector of integers stating the desired number of columns for the legends per mode. |
| arrangeModes | Vector of boolean values per mode, stating if the loadings should be arranged according to colourCols (TRUE) or not (FALSE). |

continuousModes Vector of boolean values per mode, stating if the loadings should be plotted as a line plot (TRUE) or a bar plot (FALSE).

overallTitle Overall title of the plot.

Value

Plot object

Examples

```
library(multiway)
library(dplyr)
library(ggplot2)
set.seed(0)

# Process the data
processedFujita = processDataCube(Fujita2023, sparsityThreshold=0.9, centerMode=1, scaleMode=2)

# Make PARAFAC model
model = parafac(processedFujita$data, nfac=2, nstart=10, verbose=FALSE)

# Make plot
plotPARAFACmodel(model, processedFujita,
  numComponents = 2,
  colourCols = c("", "Genus", ""),
  legendTitles = c("", "Genus", ""),
  xLabels = c("Replicate", "Feature index", "Time point"),
  legendColNums = c(0,5,0),
  arrangeModes = c(FALSE, TRUE, FALSE),
  continuousModes = c(FALSE,FALSE,TRUE),
  overallTitle = "Fujita PARAFAC model")
```

processDataCube

Process a multi-way array of count data.

Description

Process a multi-way array of count data.

Usage

```
processDataCube(
  dataset,
  sparsityThreshold = 1,
  considerGroups = FALSE,
  groupVariable = "",
  CLR = TRUE,
```

```

    centerMode = 0,
    scaleMode = 0
  )

```

Arguments

dataset A longitudinal microbiome dataset, formatted as follows:

- data** Array object of the data cube filled with counts
- mode1** Dataframe with all the subject metadata, ordered the same as the rows in the data cube.
- mode2** Taxonomic classification of the microbiota, ordered the same as the columns in the data cube.
- mode3** Dataframe with the time metadata, ordered the same as the third dimension in the array.

See [Fujita2023](#), [Shao2019](#) or [vanderPloeg2024](#) for more information.

sparsityThreshold Maximum sparsity for a feature to be selected (default=1, i.e. do not select features).

considerGroups Consider groups when calculating sparsity (default=FALSE).

groupVariable Column name in dataset\$mode1 that should be used to consider groups (default="").

CLR Perform a centered log-ratio transformation of the count data (default=TRUE).

centerMode Mode to center across: 1=subjects,2=features,3=time (default 0, i.e. do not center). See [multiwayCenter\(\)](#) for more information.

scaleMode Mode to scale within: 1=subjects,2=features,3=time (default 0, i.e. do not scale). See [multiwayScale\(\)](#) for more information.

Value

CLRred, centered and scaled cube

Examples

```
processedCube = processDataCube(Fujita2023)
```

reinflateFac *Calculate Xhat from a model Fac object*

Description

Calculate Xhat from a model Fac object

Usage

```
reinflateFac(Fac, X, returnAsTensor = FALSE)
```

Arguments

Fac Fac object from parafac
 X Input data X
 returnAsTensor Boolean to return Xhat as rTensor tensor (TRUE) or matrix (default, FALSE).

Value

Xhat

Examples

```

processedFujita = processDataCube(Fujita2023, sparsityThreshold=0.99, centerMode=1, scaleMode=2)
model = parafac(processedFujita$data, nfac=1, nstart=1, verbose=FALSE)
Xhat = reinflateFac(model$Fac, processedFujita$data)

```

reinflateTensor *Create a tensor out of a set of matrices similar to a component model.*

Description

Create a tensor out of a set of matrices similar to a component model.

Usage

```
reinflateTensor(A, B, C, returnAsTensor = FALSE)
```

Arguments

A I x N matrix corresponding to loadings in the first mode for N components.
 B J x N matrix corresponding to loadings in the second mode for N components.
 C K x N matrix corresponding to loadings in the third mode for N components.
 returnAsTensor Boolean return as rTensor S4 tensor object (default FALSE).

Value

M, an I x J x K tensor.

Examples

```

A = rnorm(108)
B = rnorm(100)
C = rnorm(10)
M = reinflateTensor(A,B,C)

```

| | |
|-------------|---|
| reshapeData | <i>Reorganize longitudinal microbiome into a data cube ready for PARAFAC modelling.</i> |
|-------------|---|

Description

Reorganize longitudinal microbiome into a data cube ready for PARAFAC modelling.

Usage

```
reshapeData(  
  Xlong,  
  subjectMetadata,  
  featureMetadata,  
  timepointMetadata,  
  timepointOrder = sort(unique(timepointMetadata))  
)
```

Arguments

| | |
|-------------------|--|
| Xlong | Longitudinal microbiome count data in matrix (long) format. |
| subjectMetadata | Vector containing the subjects corresponding to the measurements in Xlong. |
| featureMetadata | Taxonomic classification of the microbiota, ordered the same as the columns in Xlong. |
| timepointMetadata | Vector containing the time points corresponding to the measurements in Xlong. |
| timepointOrder | Vector containing the required order of the timepoints in timepointMetadata (default: sort(unique(timepointMetadata))). |

Value

#' A list object containing

data Array object of the data cube

mode1 Dataframe with the subjects, ordered the same as the rows in the data cube.

mode2 Taxonomic classification of the microbiota, ordered the same as the columns in the data cube.

mode3 Dataframe with the time metadata, ordered the same as the third dimension in the data cube.

...

Examples

```
library(dplyr)

Xlong = array(rnorm(108*5*10), c(108*5, 10))
subjects = rep(1:108, 5)
features = rep(1:10)
timepoints = rep(1:5, each=108)

dataset = reshapeData(Xlong, subjects, features, timepoints)
```

Shao2019

Shao2019 longitudinal microbiome data

Description

The Shao2019 longitudinal microbiome dataset as a three-dimensional array, with subjects in mode 1, microbial abundances in mode 2 and time in mode 3. Note: only time points 4, 7, 21 and Infancy are used. Note: all-zero microbial abundances have been removed to save disk space.

Usage

Shao2019

Format

Shao2019:

A list object with three elements:

data Array object of the data cube

mode1 Dataframe with all the subject metadata, ordered the same as the rows in the data cube.

mode2 Taxonomic classification of the microbiota, ordered the same as the columns in the data cube.

mode3 Dataframe with the time metadata, ordered the same as the third dimension in the array.

...

Source

[doi:10.1038/s4158601915601](https://doi.org/10.1038/s4158601915601)

| | |
|----------------|---|
| sortComponents | <i>Sort PARAFAC components based on variance explained per component.</i> |
|----------------|---|

Description

Sort PARAFAC components based on variance explained per component.

Usage

```
sortComponents(Fac, X)
```

Arguments

| | |
|-----|--|
| Fac | Fac object output of a parafac model |
| X | Input data |

Value

Fac object of sorted components

Examples

```
X = array(rnorm(108*100*10), c(108,100,10))
model = parafac(X, 2)
sortedFac = sortComponents(model$Fac, X)
```

| | |
|--------------------------|--|
| transformPARAFACloadings | |
|--------------------------|--|

Transform PARAFAC loadings to an orthonormal basis. Note: this function only works for 3-way PARAFAC models.

Description

Transform PARAFAC loadings to an orthonormal basis. Note: this function only works for 3-way PARAFAC models.

Usage

```
transformPARAFACloadings(Fac, modeToCorrect, moreOutput = FALSE)
```

Arguments

| | |
|---------------|---|
| Fac | Fac object from a PARAFAC object, see parafac() . |
| modeToCorrect | Correct the subject (1), feature (2) or time mode (3). |
| moreOutput | Give orthonormal basis and transformation matrices as part of output (default FALSE). |

Value

Corrected loadings of the specified mode.

Examples

```
processedFujita = processDataCube(Fujita2023, sparsityThreshold=0.99, centerMode=1, scaleMode=2)
model = parafac(processedFujita$data, nfac=2, nstart=1, verbose=FALSE)
transformedA = transformPARAFACloadings(model$Fac, 1)
```

vanderPloeg2024

vanderPloeg2024 longitudinal dataset

Description

The vanderPloeg2024 longitudinal data containing six oral microbiome niches, as well as salivary metabolomics.

Usage

```
vanderPloeg2024
```

Format

vanderPloeg2024:

Each measured dataset contains three elements:

data Array object of the data cube

mode1 Dataframe with all the subject metadata, ordered the same as the rows in the data cube.

mode2 Taxonomic classification of the microbiota, ordered the same as the columns in the data cube.

mode3 Dataframe with the time metadata, ordered the same as the third dimension in the array.

...

Source

[doi:10.1101/2024.03.18.585469](https://doi.org/10.1101/2024.03.18.585469)

| | |
|-------------|--|
| vect_to_fac | <i>Convert vectorized output of PARAFAC to a Fac list object with all loadings per mode.</i> |
|-------------|--|

Description

Convert vectorized output of PARAFAC to a Fac list object with all loadings per mode.

Usage

```
vect_to_fac(vect, X, sortComponents = FALSE)
```

Arguments

| | |
|----------------|--|
| vect | Vectorized output of PARAFAC modelling |
| X | Input data |
| sortComponents | Sort the order of the components by variation explained (default FALSE). |

Value

Fac: list object with all loadings in all components per mode, ordered the same way as Z\$modes.

Examples

```
set.seed(123)
A = array(rnorm(108*2), c(108, 2))
B = array(rnorm(100*2), c(100, 2))
C = array(rnorm(10*2), c(10, 2))

X = reinflateTensor(A, B, C)
result = initializePARAFAC(X, 2, initialization="random", output="vect")
Fac = vect_to_fac(result, X)
```

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