

Package ‘gLVInterNetworks’

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Type Package

Title Inference of interaction networks based on generalised Lotka Volterra dynamics

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Description Inference of interaction networks based on the parameterization of generalized Lotka Volterra models on timeseries data

Imports deSolve, MASS, glmnet, FME, igraph, minpack.lm, coda, minqa, rootSolve

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gLVInterNetworks-package

Inference of interaction networks based on generalised Lotka Volterra dynamics

Description

Inference of interaction networks based on the parameterization of generalized Lotka Volterra models on timeseries data

Details

The DESCRIPTION file: This package was not yet installed at build time.

Index: This package was not yet installed at build time.

~~ An overview of how to use the package, including the most important functions ~~

Author(s)

Lukas Hirsch, Florian Centler Maintainer: Lukas Hirsch <lukashirsch@gmail.com>

References

~~ Literature or other references for background information ~~

Examples

```
library(gLVInterNetworks)
data <- gLVgenerateData(species = 2, number_of_interactions = 2, timepoints = 100, noise = 0.01, testData = 20)
## Not run: plot(data, type = "l")
lr <- gLVlinearRegression(data, regularization = TRUE, alpha = 0)
## Not run: summary(lr)
## Not run: plot(lr, type = "l")
## Not run: points(data)
nlr <- gLVnonlinearRegression(data, parms0 = lr$Parms)
## Not run: summary(nlr)
## Not run: plot(nlr, type = "l")
## Not run: points(data)
## Not run: par(mfrow = c(1,2))
## Not run: plotGraph(data, vsize = 0.2, main = "Original interaction network", verbose = TRUE )
## Not run: plotGraph(nlr, vsize = 0.2, main = "Inferred interaction network", verbose = TRUE)
ident <- sensitivityAnalysis(nlr$Parms)
## Print summary of sensitivity matrix
summary(ident$sens)
## Print collinearity index for all parameters together
ident$coll[ident$coll[, "N"]==length(data$Parms),]
```

exampleData2

dataset with 6 species

Description

example random dataset with 6 species

Usage

```
data(exampleData2)
```

Format

A list containing 7 objects.

gLVgenerateData	<i>Generate random data for gLV fitting</i>
-----------------	---

Description

Generates random data simulating time series of cell abundances governed by Lotka Volterra dynamics

Usage

```
gLVgenerateData(species, number_of_interactions, timepoints, noise, testData)
```

Arguments

species	Integer describing the number of independent cellular subcommunities
number_of_interactions	Integer describing the number of non-zero interactions present in the simulated system. These are assigned randomly between the nodes or subcommunities
timepoints	Numeric vector containing the timepoints for which to compute the solutions of the model
noise	The standard deviation of the normally distributed stochastic factor added to the solution of the model at each time step
testData	Number of observations used as test dataset for validation on untrained data, taken from the last measurements of the time series.

Value

Returns a matrix. The first column displays the time points, and the remaining columns correspond each to a independent variable in the system.

Author(s)

Lukas Hirsch

Examples

```
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (species, number_of_interactions, timepoints,
          noise)
{
  "requires inSilico_bio, discrete, addNoise_res"
  raw_data <- inSilico_bio(species, number_of_interactions)
  Pams <- raw_data[[2]]
  res <- raw_data[[1]]
  threshold = 3
  if (any(which(round(diff(abs(rowMeans(res[, -1])), 2), threshold) ==
    0))) {
    end <- min(which(round(diff(abs(rowMeans(res[, -1])),
```

```

        2), threshold) == 0))
    }
    else {
        end <- nrow(res)
    }
    res <- solveLV_bio(Parms, seq(0, end, by = 0.01), res[1,
        -1])
    discretization <- nrow(res)/timepoints
    res <- discrete(res, discretization)
    test_data <- res[(nrow(res) - 19):nrow(res), ]
    obs <- res[1:(nrow(res) - 20), ]
    obs <- addNoise_res(obs, noise)
    timepoints <- nrow(obs)
    dimensions <- ncol(obs[, -1])
    k <- sum(any(Parms == 0))/length(Parms)
    data <- list(species = species, timepoints = timepoints,
        Params = Params, noise = noise, sparsity = k, obs = obs,
        testData = test_data)
    class(data) <- "Sim_data"
    return(data)
}

```

gLVlinearRegression *Parameter estimation of algebraic linear discrete gLV model*

Description

Given multivariatic time series data, this function fits a linear and discrete generalized Lotka Volterra model of the form $\Delta x_i = \alpha_i + \sum \beta_{ij} * x_j$

Usage

```
gLVlinearRegression(data, regularization = FALSE, alpha = 0)
```

Arguments

data	Matrix or table containing time series of measurements in longitudinal form where first column corresponds to the time points and subsequent columns correspond to each model variable
regularization	Boolean flag if regularization of the parameter matrix should be forced
alpha	Regularization parameter for the elastic net. It ranges from 0 (= Ridge regression) to 1 (= LASSO regression) with values in between corresponding to both L1 and L2 penalties weighted by alpha

Details

Some theory and formulas on elastic net

gLVnonlinearRegression

Parameter estimation through gradient search of continuous nonlinear gLV model

Usage

```
gLVnonlinearRegression(data, parms0 = NULL, ftol = 1e-8 , ptol = 1e-8, maxiter = 100, lowerbound =
```

Arguments

data	Data input containing a time series of observations in longitudinal matrix form
parms0	Optional. Starting parameter vector. Default = Zero vector
ftol	Objective function output tolerance before stopping iterative optimization
ptol	Parameter change tolerance in output of objective function
maxiter	Maximal number of iterations allowed before breaking the gradient search algorithm
lowerbound	Numerical vector of equal length as parameter vector describing lower bound for constrained parameter search
upperbound	Numerical vector of equal length as parameter vector describing upper bound for constrained parameter search
method	Method used for optimization of the objective function. Default is "Marq" for Leverberg-Marquandt

Examples

```
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.
```

nlrData2

Result of nonlinear regression on Data2

Description

Result of nonlinear regression on dataset with 6 sp

Usage

```
data(nlrData2)
```

Format

A list containing 12 objects.

plot.Sim_data	<i>Plot function for objects returned by in silico data generation function</i>
---------------	---

Usage

```
plot.Sim_data(x, legend = FALSE, ...)
```

Arguments

x	Object of class Sim_data as returned by function gLVgenerateData()
legend	Set to TRUE to place a basic legend in the topright of the plot
...	

Examples

```
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (x, ...)
{
  matplot(x$obs[, -1])
}
```

plotGraph	<i>Plot interaction network</i>
-----------	---------------------------------

Description

Plots a graph representing the interaction network described by the data

Usage

```
plotGraph(x, vsize = 0.1, main = NULL, verbose = FALSE, keepNames = FALSE, ...)
```

Arguments

x	Object containing parameter matrix in form of x\$Parms
vsize	Integer inversely proportional to the size of the nodes
main	Title of the plot
verbose	Include edge values in output network
keepNames	Set to TRUE if plotted Network should keep the names of the variables as given in the observations table. Default = FALSE
...	

Value

igraph object

Author(s)

Lukas Hirsch

Examples

```
data <- gLVgenerateData(2,2,100,0.1, 20)
#plotGraph(data)
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.
```

sensitivityAnalysis	<i>Compute parameter correlations and multicollinearity for the model output</i>
---------------------	--

Description

This function is a wrapper function for sensFun and Collin from package FME. It calculates both a sensitivity matrix S_{ij} and multicollinearity index for all parameter combinations.

Usage

```
sensitivityAnalysis(Parms)
```

Arguments

Parms	Numeric vector or matrix with the parameter coefficients to test
-------	--

Value

List containing:

sens	Matrix containing sensitivity output values for each parameter and each model variable. The sensitivity matrix S_{ij} contains elements $dy_i/dpar_j * parscale_j / varscale_i$. The scale used to change the value of each parameter can be seen using the summary function on the sens table, and it is set to be the same value of the parameter itself
coll	Table with collinearity index for each possible parameter subset.

Warning

The function needs the original data as a global variable called "data". Please make sure that when using this function to perform identifiability analysis on a set of parameters, that the observation matrix or table used to estimate the parameter set is accessible under the name "data"

Note

For details in output interpretation see package FME

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