

Package ‘miaSim’

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Type Package

Version 1.0.0

Title Microbiome Data Simulation

Description Microbiome time series simulation with generalized Lotka-Volterra model, Self-Organized Instability (SOI), and other models. Hubbell's Neutral model is used to determine the abundance matrix. The resulting abundance matrix is applied to SummarizedExperiment or TreeSummarizedExperiment objects.

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Depends SummarizedExperiment

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Suggests rmarkdown, knitr, BiocStyle, testthat

URL <https://github.com/microbiome/miaSim>

BugReports <https://github.com/microbiome/miaSim/issues>

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powerlawA	<i>Interaction matrix with Power-Law network adjacency matrix</i>
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Description

Where N is the an Interspecific Interaction matrix with values drawn from a normal distribution H the interaction strength heterogeneity drawn from a power-law distribution with the parameter alpha, and G the adjacency matrix of with out-degree that reflects the heterogeneity of the powerlaw. A scaling factor s may be used to constrain the values of the interaction matrix to be within a desired range. Diagonal elements of A are defined by the parameter d.

Usage

```
powerlawA(n.species, alpha = 3, stdev = 1, s = 0.1, d = -1, symmetric = FALSE)
```

Arguments

n.species	integer: the number of species
alpha	numeric: the power-law distribution parameter. Should be > 1. (default: alpha = 3.0) Larger values will give lower interaction strength heterogeneity, whereas values closer to 1 give strong heterogeneity in interaction strengths between the species. In other words, values of alpha close to 1 will give Strongly Interacting Species (SIS).
stdev	numeric: the standard deviation parameter of the normal distribution with mean 0 from which the elements of the nominal interspecific interaction matrix N are drawn. (default: stdev = 1)
s	numeric: scaling parameter with which the final global interaction matrix A is multiplied. (default: s = 0.1)
d	numeric: diagonal values, indicating self-interactions (use negative values for stability). (default: s = 1.0)
symmetric	logical: return a symmetric interaction matrix (default: symmetric = FALSE)

Value

The interaction matrix A with n rows and n columns.

References

Gibson TE, Bashan A, Cao HT, Weiss ST, Liu YY (2016) On the Origins and Control of Community Types in the Human Microbiome. PLOS Computational Biology 12(2): e1004688. <https://doi.org/10.1371/journal.pcbi.1004688>

Examples

```
# Low interaction heterogeneity
A_low <- powerlawA(n.species = 10, alpha = 3)
# Strong interaction heterogeneity
A_strong <- powerlawA(n.species = 10, alpha = 1.01)
```

randomA

Generate random uniform interaction matrix

Description

Generate random simplified interaction matrix from a uniform distribution.

Usage

```
randomA(
  n.species,
  d = -0.5,
  min.strength = -0.5,
  max.strength = 0.5,
  connectance = 0.02,
  symmetric = FALSE
)
```

Arguments

n.species	integer number of species
d	numeric diagonal values (should be negative) (default: d = -0.5)
min.strength	numeric minimal off-diagonal interaction strength (default: min.strength = -0.5)
max.strength	numeric maximal off-diagonal interaction strength (default: max.strength = 0.5)
connectance	numeric (interaction probability) (default: connectance = 0.02)
symmetric	logical return a symmetric interaction matrix (default: symmetric = FALSE)

Value

randomA returns a matrix A with dimensions (n.species x n.species)

Examples

```
high_inter_A <- randomA(n.species = 10, d = -0.4, min.strength = -0.8,
                       max.strength = 0.8, connectance = 0.5)

low_inter_A <- randomA(n.species = 10, connectance = 0.01)
```

simulateGLV

*Generalized Lotka-Volterra (gLV) simulation***Description**

Simulates time series with the generalized Lotka-Volterra model and forms a [SummarizedExperiment](#) object.

Usage

```
simulateGLV(
  n.species,
  A,
  x = runif(n.species),
  b = runif(n.species),
  t.end = 1000,
  norm = FALSE,
  ...
)
```

Arguments

n.species	Integer: number of species
A	interaction matrix
x	Numeric: initial abundances
b	Numeric: growth rates
t.end	Numeric: simulation end time (default: t.end = 1000)
norm	Logical scalar: returns normalised abundances (proportions in each generation) (default: norm = FALSE)
...	additional arguments that can be called from miaSim::tDyn

Details

Simulates a community time series using the generalized Lotka-Volterra model, defined as $dx/dt = x(b+Ax)$, where x is the vector of species abundances, $\text{diag}(x)$ is a diagonal matrix with the diagonal values set to x . A is the interaction matrix and b is the vector of growth rates.

The resulting abundance matrix model is used to construct [SummarizedExperiment](#) object.

Value

simulateGLV returns a [SummarizedExperiment](#) object containing abundance matrix

Examples

```
row_data <- data.frame(Kingdom = "Animalia",
                      Phylum = rep(c("Chordata", "Echinodermata"), c(500, 500)),
                      Class = rep(c("Mammalia", "Asteroidea"), each = 500),
                      ASV = paste0("X", seq_len(1000)),
                      row.names = rownames(paste0("species", seq_len(1000))),
                      stringsAsFactors = FALSE)

row_data <- t(row_data)

col_data <- DataFrame(sampleID = seq_len(1001),
                    time = as.Date(1000, origin = "2000-01-01"),
                    row.names = colnames(paste0("sample", seq_len(1001))))

A <- miaSim::powerlawA(4, alpha = 1.01)

SEobject <- simulateGLV(n.species = 4, A, t.end = 1000)
rowData(SEobject) <- row_data
colData(SEobject) <- col_data
```

simulateHubbell

Hubbell's neutral model simulation

Description

Neutral species abundances simulation according to the Hubbell model.

Usage

```
simulateHubbell(
  n.species,
  M,
  I = 1000,
  d = 10,
  m = 0.02,
  tskip = 0,
  tend,
  norm = FALSE
)
```

Arguments

n.species	Integer: the amount of different species initially in the local community
M	Integer: amount of different species in the metacommunity, including those of the local community
I	Integer: fixed amount of individuals in the local community (default: I = 1000)
d	Integer: fixed amount of deaths of local community individuals in each generation (default: d = 10)
m	Numeric: immigration rate: the probability that a death in the local community is replaced by a migrant of the metacommunity rather than by the birth of a local community member (default: m = 0.02)
tskip	Integer: number of generations that should not be included in the outputted species abundance matrix. (default: tskip = 0)
tend	Integer: number of simulations to be simulated
norm	Logical: whether the time series should be returned with the abundances as proportions (norm = TRUE) or the raw counts (default: norm = FALSE)

Value

simulateHubbell returns a [SummarizedExperiment](#) class object containing matrix with species abundance as rows and time points as columns

References

Rosindell, James et al. "The unified neutral theory of biodiversity and biogeography at age ten." Trends in ecology & evolution vol. 26,7 (2011).

Examples

```
colData <- DataFrame(sampleID = c(seq_len(100)),
                    time = as.Date(100, origin = "2000-01-01"))

rowData <- data.frame(Kingdom = "Animalia",
                    Phylum = rep(c("Platyhelminthes", "Mollusca"), c(50, 50)),
                    Class = rep(c("Turbellaria", "Polyplacophora"), each = 50),
                    ASV1 = paste0("D", seq_len(100)),
                    ASV2 = paste0("E", seq_len(100)),
                    ASV3 = paste0("F", seq_len(100)),
                    ASV4 = paste0("G", seq_len(100)),
                    ASV5 = paste0("H", seq_len(100)),
                    ASV6 = paste0("J", seq_len(100)),
                    ASV7 = paste0("K", seq_len(100)),
                    row.names = rownames(paste0("species", seq_len(10))),
                    stringsAsFactors = FALSE)

rowData <- t(rowData)

ExampleHubbell <- simulateHubbell(n.species = 8, M = 10, I = 1000, d = 50,
                                m = 0.02, tend = 100)
```


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