# Package 'samovaR'

May 15, 2025

```
Type Package
Title R package for generating model metagenomes with specified properties
Version 0.5
Author@R
     c(person(``Smutin", ``Daniil", email = ``dvsmutin@gmail.com", role = c(``aut", ``cre"),
      comment = c(ORCID = ``0009-0009-1460-4108")),
      person(``Ivanov", ``Artem", role = c(``aut")),
      person(``ctlab", role = c(``fnd")))
Description There is a fundamental problem in modern ***metagenomics***: there are huge differ-
      ences between methodological approaches that strongly influence the results, while remain-
      ing outside the attention of researchers. We propose an approach that utilizes de novo genera-
      tion of the artificial metagenomes - `SamovaR`.
URL https://github.com/dsmutin/samovar/
License MIT + file LICENSE
Encoding UTF-8
LazyData true
Imports tidyverse,
      plotly,
      httr,
     isonlite,
      xml2,
      tsne,
      cluster,
      Matrix,
      shiny,
      here,
      methods,
      scclust
Roxygen list(markdown = TRUE)
RoxygenNote 7.3.2
Suggests ggnewscale,
      knitr,
      rmarkdown,
      testthat (>= 3.0.0)
VignetteBuilder knitr
```

Config/testthat/edition 3

2 build\_samovar

# **Contents**

build_samovar	2
concotion_pour	3
GMrepo_run	4
GMrepo_run2data	4
	5
	6
	7
minmaxscale	8
progress_function	8
read.abundance	8
read.samovar	9
samovar_base	9
samovar_boil	0
samovar_data	1
samovar_run	1
teabag_brew	
tealeaves_pack	2
teatree_trim	3
viz_composition	4
1	6

build\_samovar

Build samovar object

#### **Description**

**Index** 

Samovar network is a 2D-oriented graph with metadata and abundances of species per sample. Oriented graph could be used for network prediction, or for better generation some network could be used as initial (to be implemented) For better understanding of building database and using it in generation, visit github source

# Usage

```
build_samovar(
   samovar_data,
   dist_function = function(x) dist(x),
   network = F,
   k_means = F,
   min_cluster_size = F,
   plot_log = T
)
```

## Arguments

```
samovar_data samovar data after preprocessing stages

network FALSE or graph that can be used for generation. To be implemented

plot_log Logical or path for log plots output

distance_function
```

function used for measuring distances between species based on samples

concotion\_pour 3

```
min_min_cluster_size
FALSE or minimum number of species per cluster
max_min_cluster_size
FALSE or minimum number of species per cluster
```

concotion\_pour

Build samovar object

# Description

Samovar network is a 2D-oriented graph with metadata and abundances of species per sample. Oriented graph could be used for network prediction, or for better generation some network could be used as initial (to be implemented) For better understanding of building database and using it in generation, visit github source

#### Usage

```
concotion_pour(
  samovar_data,
  inner_method = "glm",
  inter_method = "glm",
  inner_model = "gaussian",
  inter_model = "gaussian",
  minimal_cluster = 2,
  probability_calculation = "oriented",
  cluster_connection = "mean"
)
```

#### **Arguments**

samovar_data	samovar data after preprocessing stages	
inner_method	Character, glm, other to be implemented (bootstrap, bsPCA)	
inter_method	Character, glm, other to be implemented (bootstrap, bsPCA)	
inner_model	Character, model processed by glm(). For glm mode only. quasipoisson by default	
inter_model	Character, model processed by glm(). For glm mode only. quasipoisson by default	
cluster_connection		
	Character (mean, median), or function. The way of cluster connection. If function, way of summarize all samples of species cluster	
network	FALSE or graph that can be used for generation. To be implemented	
cooccurrence	Character, co-occurrence calculation.	
	If "simple", calculated as: $P(A B) = sum(A\&B)/sum(A B)$ .	
	If "oriented", calculated as $P(A B) = P(A\&B B)$	
	If "compositional", calculated as $P(A B) = P(A\&B B)$ , and than sampled one of represented conditions of occurence	

4 GMrepo\_run2data

#### **Examples**

```
# download data
teatree <- GMrepo_type2data(number_to_process = 2000)

# filter
tealeaves <- teatree %>%
    teatree_trim(treshhold_species = 3, treshhold_samples = 3, treshhold_amount = 10^(-3))

# normalizing
## if you build teatree by your own, rescaling stage when building via teatree$rescale() or assigning teatree$mi
## good approximation to normal distribution is required for glm generating methods
teabag <- tealeaves %>%
    tealeaves_pack(normalization_function = function(x) log10(x+1))

# clustering
concotion <- teabag %>%
    teabag_brew(min_cluster_size = 4, max_cluster_size = 6)
# remember: if you want to refilter, it is better to re-do welding stage to avoid crashes in future!

# building samovar
samovar <- concotion %>%
    concotion_pour()
```

GMrepo\_run

GMrepo run data class

# Description

GMrepo run data class

#### **Slots**

metadata metadata DataFrame run character

GMrepo\_run2data

Get data from GMrepo\_run object

#### **Description**

Get data from GMrepo\_run object

```
GMrepo_run2data(
  run,
  number_to_out = F,
  at_level = "species",
  QC_filter = "QCStatus"
)
```

GMrepo\_type2data 5

#### **Arguments**

number\_to\_out False by default, maximum number of obtained data

at\_level "species" by default. level to obtain classification from GMrepo

QC\_filter QCStatus by default. Perform auto QC filtering based on metadata column, or

False for no checking.

runs GMrepo\_run object got by GMrepo\_type2run or created by user with new('GMrepo\_run',

metadata = data.frame(), run = run\_list)

#### **Examples**

```
# get data from GMrepo
data_GMrepo <- GMrepo_type2data(mesh_ids = "D006262", number_to_process = 1000)</pre>
# equal to:
run_GMrepo <- GMrepo_type2run(mesh_ids = "D006262", number_to_process = 1000)</pre>
data_GMrepo <- GMrepo_run2data(run_GMrepo)</pre>
# filter runs before obtaining data (OOP updating data!)
run_GMrepo$filter("checking", 1)
# view
data_GMrepo
# access to metadata
data_GMrepo$run
# access to data
data_GMrepo$data
# access to runs
data_GMrepo$run
# access to taxa
data_GMrepo$species
```

GMrepo\_type2data

Get data from GMrepo

#### **Description**

Wrapper around GMrepo\_type2run and GMrepo\_run2data functions

```
GMrepo_type2data(
  mesh_ids = c("D006262"),
  number_to_process = F,
  number_to_out = F,
  at_level = "species",
  QC_filter = "QCStatus"
)
```

6 GMrepo\_type2run

#### **Arguments**

mesh\_ids Character. All types of meshID to use. List of relations between meshID and

phenotype could be obtained using GMrepo\_meshID(). Health meshID by de-

fault

number\_to\_process

False by default, or maximum number of runs per meshID

number\_to\_out False by default, maximum number of obtained data

at\_level "species" by default. level to obtain classification from GMrepo

QC\_filter QCStatus by default. Perform auto QC filtering based on metadata column, or

False for no checking.

#### **Examples**

```
# get data from GMrepo
data_GMrepo <- GMrepo_type2data(mesh_ids = "D006262", number_to_process = 1000)</pre>
# equal to:
run_GMrepo <- GMrepo_type2run(mesh_ids = "D006262", number_to_process = 1000)</pre>
data_GMrepo <- GMrepo_run2data(run_GMrepo)</pre>
# filter runs before obtaining data (OOP updating data!)
run_GMrepo$filter("checking", 1)
# view
data_GMrepo
# access to metadata
data_GMrepo$run
# access to data
data_GMrepo$data
# access to runs
data_GMrepo$run
# access to taxa
data_GMrepo$species
```

GMrepo\_type2run

Get runs from GMrepo by meshID

#### **Description**

Get runs from GMrepo by meshID

```
GMrepo_type2run(mesh_ids = c("D006262"), number_to_process = F)
```

log\_plot 7

## **Arguments**

mesh\_ids Character. All types of meshID to use. List of relations between meshID and phenotype could be obtained using GMrepo\_meshID()

number\_to\_process

False by default, or maximum number of runs per meshID

# **Examples**

```
# get data from GMrepo
data_GMrepo <- GMrepo_type2data(mesh_ids = "D006262", number_to_process = 1000)</pre>
# equal to:
run_GMrepo <- GMrepo_type2run(mesh_ids = "D006262", number_to_process = 1000)</pre>
data_GMrepo <- GMrepo_run2data(run_GMrepo)</pre>
# filter runs before obtaining data (OOP updating data!)
run_GMrepo$filter("checking", 1)
# view
data_GMrepo
# access to metadata
data_GMrepo$run
# access to data
data_GMrepo$data
# access to runs
data_GMrepo$run
# access to taxa
data_GMrepo$species
```

log\_plot

Print a log plot

#### **Description**

Print a log plot

```
log_plot(plot_log, postfix, gg, mode = "ggplot")
```

8 read.abundance

minmaxscale

Misc functions

# Description

Misc functions

# Usage

```
minmaxscale(x)
```

progress\_function

Progress bar

# Description

Progress bar

# Usage

```
progress_function(iters)
```

read.abundance

Build samovar data object from file or environment

# Description

Build samovar data object from file or environment

# Usage

```
read.abundance(data, metadata, ...)
```

# Arguments

data Data.frame or path to abundance file. Row names is using as species list, column

names as sample list. Unique names required

metadata Data.frame or path to metadata file

... Parameters processed by read.table()

read.samovar 9

read.samovar

Read samovar data frame from file

# Description

Prints 'Hello, world!'.

# Usage

hello()

#### **Examples**

hello()

samovar\_base

samovar base class

# Description

samovar base class

#### **Slots**

samovar\_base samovar\_data object

method method to obtain samovar\_base

inner\_cluster\_graph\_method list of graphs in matrix form of inner cluster connections

inter\_cluster\_graph\_method list of graphs in matrix form of inter cluster connections

inner\_cluster\_graph\_prob list of co-occurrence probabilities in matrix form of inner cluster members

inter\_cluster\_graph\_prob list of co-occurrence probabilities in matrix form between clusters properties concotion\_pour() properties

10 samovar\_boil

samovar\_boil

Generate artificial data

#### **Description**

Use pre-built samovar\_data with its parameters

# Usage

```
samovar_boil(
  samovar_base,
  N = 1,
  init_sp = F,
  init_ab = F,
  avoid_zero_generations = T,
  seed = 42
)
```

## Arguments

samovar\_base samovar data after preprocessing and building stages

N number of artificial samples to generate

init\_sp species vector for initializing data generation, or FALSE for usage most com-

mon taxa, auto for choosing random taxa

init\_ab species amount vector (values from 0 to 1) for initializing data generation, or

FALSE for mean initial taxa assignment, or auto for usage from known edf for

each species from init\_sp

 $avoid\_zero\_generations$ 

logical, avoid zero-based generations or not. FALSE might results in underdistributed communities, while TRUE in over-represented with species from

different clusters possibly come from different samples groups

seed initial seed for the seeds generation

# **Examples**

```
# download and prepare data
samovar <- GMrepo_type2data(number_to_process = 2000) %>%
  teatree_trim(treshhold_species = 3, treshhold_samples = 3, treshhold_amount = 10^(-3)) %>%
  tealeaves_pack(normalization_function = function(x) log10(x+1)) %>%
  teabag_brew(min_cluster_size = 4, max_cluster_size = 6) %>%
  concotion_pour()

# generate
new_data <- samovar %>%
  samovar_boil(n = 100)
```

samovar\_data 11

samovar\_data

samovar data class

# Description

samovar data class

#### **Slots**

```
description metadata DataFrame
data DataFrame with species abundances. No NA pass
run character, runs
species character, runs
normalization_function normalization function for samples
reverse_normalization_function reverse normalization function
min_value minimal value after scaling
max_value maximal value after scaling
cluster character vector, enumerated clusters for each species
cluster_size named numeric, cluster sizes per cluster
```

samovar\_run

Samovar run data class

# Description

Samovar run data class

#### **Slots**

```
metadata metadata DataFrame
data data
run character, samle IDs
```

#### Methods

export(Class) Returns the result of coercing the object to Class. No effect on the object itself.

12 tealeaves\_pack

teabag\_brew

Build samovar object

# Description

Samovar network is a 2D-oriented graph with metadata and abundances of species per sample. Oriented graph could be used for network prediction, or for better generation some network could be used as initial (to be implemented) For better understanding of building database and using it in generation, visit github source

# Usage

```
teabag_brew(
  samovar_data,
  dist_function = function(x) dist(x),
  network = F,
  min_cluster_size = 10,
  max_cluster_size = 100,
  plot_log = T
)
```

#### **Arguments**

```
network FALSE or graph that can be used for generation. To be implemented min_cluster_size
FALSE or minimum number of species per cluster
max_cluster_size
FALSE or minimum number of species per cluster
plot_log Logical or path for log plots output
data samovar data after preprocessing stages
distance_function
```

function used for measuring distances between species based on samples

tealeaves\_pack

Scale species abundances

# Description

Scale species abundances

```
tealeaves_pack(
  samovar_data,
  normalization_function = function(x) log10(x + 1),
  plot_log = T
)
```

teatree\_trim 13

#### **Arguments**

```
samovar_data Samovar data object to rescale
normalization_function
Function using for rescaling
plot_log Logical or path for log plots output
```

#### **Examples**

```
# download data
teatree <- GMrepo_type2data(number_to_process = 2000)</pre>
# filter
tealeaves <- teatree %>%
  teatree_trim(treshhold_species = 3, treshhold_samples = 3, treshhold_amount = 10^(-3))
# normalizing
## if you build teatree by your own, rescaling stage when building via teatree$rescale() or assigning teatree$mi
## good approximation to normal distribution is required for glm generating methods
teabag <- tealeaves %>%
  tealeaves_pack(normalization_function = function(x) log10(x+1))
# clustering
concotion <- teabag %>%
  teabag_brew(min_cluster_size = 4, max_cluster_size = 6)
# remember: if you want to refilter, it is better to re-do welding stage to avoid crashes in future!
# building samovar
samovar <- concotion %>%
  concotion_pour()
```

teatree\_trim

Filter species and samples from samovar\_data object

## Description

Filter species and samples from samovar\_data object

```
teatree_trim(
  samovar_data,
  metadata_filter = F,
  treshhold_amount = 10^(-5),
  treshhold_samples = 1,
  treshhold_species = 1,
  drop_species = F,
  drop_unclassified = T
)
```

14 viz\_composition

#### **Arguments**

```
samovar_data Samovar data object to filter

metadata_filter

False, character or data.frame with 2 columns: first contain metadata names for filtering, and second values per column

treshhold_amount

Minimum value to conclude as not the noise.

treshhold_samples

Minimum number of representing samples to keep species.

treshhold_species

Minimum number of representing species to keep samples.

drop_unclassified

Drop unknown and unclassified ranks. True by default
```

#### **Examples**

```
# download data
teatree <- GMrepo_type2data(number_to_process = 2000)</pre>
# filter
tealeaves <- teatree %>%
  teatree_trim(treshhold_species = 3, treshhold_samples = 3, treshhold_amount = 10^(-3))
# normalizing
## if you build teatree by your own, rescaling stage when building via teatree$rescale() or assigning teatree$mi
## good approximation to normal distribution is required for glm generating methods
teabag <- tealeaves %>%
  tealeaves_pack(normalization_function = function(x) log10(x+1))
# clustering
concotion <- teabag %>%
  teabag_brew(min_cluster_size = 4, max_cluster_size = 6)
# remember: if you want to refilter, it is better to re-do welding stage to avoid crashes in future!
# building samovar
samovar <- concotion \%>%
  concotion_pour()
```

 ${\tt viz\_composition}$ 

Visualize composition

#### **Description**

Visualize composition

```
viz_composition(
  data,
  reord_samples = "fpc",
  reord_species = "amount",
```

viz\_composition 15

```
type = "column",
top = 15,
interactive = F,
ggplot_add = F,
bottom_legend = F
```

#### **Arguments**

data data.frame with dimensions of species \* samples, or samovar objects: samovar\_data,

samovar\_base, samovar\_run or GMrepo\_run (with data)

reord\_samples character, fpc, fpc\_scaled, hcl, amount, tsne, or none reorder of samples on plot

reord\_species character, same for reord\_samples

type character, column or tile for composition visualize, or donut (to implement) for

mean composition visualization

top integer, number of top-represented taxa to show, or FALSE to show all

interactive logical. ggplot or plotly object to return ggplot\_add functions to add to ggplot object, or FALSE.

bottom\_legend vector length of samples to show on plot as a color legend, or FALSE

#### **Examples**

```
# Download data
teatree <- GMrepo_type2data(number_to_process = 1000)
#Composition
viz_composition(teatree)</pre>
```

# **Index**

```
build_samovar, 2
concotion_pour, 3
GMrepo_run, 4
GMrepo_run2data, 4
GMrepo_type2data, 5
GMrepo_type2run, 6
log_plot, 7
minmaxscale, 8
progress_function, 8
read.abundance, 8
read.samovar, 9
samovar_base, 9
samovar_boil, 10
samovar\_data, 11
samovar_run, 11
{\tt teabag\_brew},\, {\tt 12}
tealeaves_pack, 12
teatree_trim, 13
\textit{viz\_composition}, \\ 14
```