# Package 'saeSim'

October 14, 2025

Type Package
Title Simulation Tools for Small Area Estimation
<b>Version</b> 0.12.0
<pre>URL https://wahani.github.io/saeSim/</pre>
BugReports https://github.com/wahani/saeSim/issues
<b>Depends</b> $R(>=3.1)$ , methods
<b>Imports</b> dplyr (>= 0.2), functional, ggplot2, grDevices, MASS, utils, spdep, stats, parallelMap, tibble
Suggests testthat, knitr, rmarkdown, covr
<b>Description</b> Tools for the simulation of data in the context of small area estimation. Combine all steps of your simulation - from data generation over drawing samples to model fitting - in one object. This enables easy modification and combination of different scenarios. You can store your results in a folder or start the simulation in parallel.
License MIT + file LICENSE
VignetteBuilder knitr
RoxygenNote 7.3.2
Encoding UTF-8
NeedsCompilation no
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Repository CRAN
<b>Date/Publication</b> 2025-10-14 20:20:02 UTC
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agg\_all

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Aggregation function

## Description

This function is intended to be used with sim\_agg and not interactively. This is one implementation for aggregating data in a simulation set-up.

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## Usage

```
agg_all(groupVars = "idD")
```

# Arguments

 ${\sf groupVars}$ 

variable names as character identifying groups to be aggregated.

## **Details**

This function follows the split-apply-combine idiom. Each data set is split by the defined variables. Then the variables within each subset are aggregated (reduced to one row). Logical variables are reduced by any; for characters and factors dummy variables are created and the aggregate is the mean of each dummy; and for numerics the mean (removing NAs).

as.data.frame.sim\_setup

#### See Also

```
sim_agg
```

## **Examples**

```
sim_base() %>% sim_gen_x() %>% sim_gen_e() %>% sim_agg(agg_all())
```

```
as.data.frame.sim_setup

as.data.frame method for sim_setup
```

## Description

Use this method to get a single simulated data.frame out of a sim\_setup object.

#### Usage

```
## S3 method for class 'sim_setup'
as.data.frame(x, row.names = NULL, optional = FALSE, ...)
```

#### **Arguments**

```
x a sim_setup
row.names will have no effect
optional will have no effect
... will have no effect
```

## Description

Use this function to produce plots for an object of class sim\_setup and you like to have plots based on ggplot2. At this time it is a ggplot2 implementation which mimics the behavior of smoothScatter without all the options.

## Usage

```
## S3 method for class 'sim_setup'
autoplot(object, x = "x", y = "y", ...)
```

base\_id

#### **Arguments**

object a sim\_setup

x character of variable name in the data on the x-axis y character of variable name in the data on the y-axis

... is not used

## **Examples**

```
## Not run:
autoplot(sim_base_lm())
## End(Not run)
```

base\_add\_id

Add id-variables to data

## **Description**

Use this function to add id-variables to your data.

## Usage

```
base_add_id(data, domainId)
```

## Arguments

data a data.frame.

domainId variable names in data as character which will identify the areas/domains/groups/cluster

in the data.

base\_id

Construct data with id-variables

## **Description**

This function constructs a data frame with grouping/id variables.

## Usage

```
base_id(nDomains = 10, nUnits = 10)
base_id_temporal(nDomains = 10, nUnits = 10, nTime = 10)
```

comp\_var 5

## **Arguments**

nDomains The number of domains.

nUnits The number of units in each domain. Can have length(nUnits) > 1.

nTime The number of time points for each units.

## Value

Return a data.frame with variables idD as ID-variable for domains, and idU as ID-variable for units.

## **Examples**

```
base_id(2, 2)
base_id(2, c(2, 3))
```

comp\_var

Compute variables in data

## Description

This function is intended to be used with sim\_comp\_pop, sim\_comp\_sample or sim\_comp\_agg and not interactively. This is a wrapper around mutate

## Usage

```
comp_var(...)
```

## **Arguments**

... variables interpreted in the context of that data frame.

## See Also

```
sim_comp_pop, sim_comp_sample, sim_comp_agg
```

```
sim_base_lm() %>% sim_comp_pop(comp_var(yExp = exp(y)))
```

gen\_norm

gen_norm	Generator functions	

## **Description**

These functions are intended to be used with sim\_gen and not interactively. They are designed to draw random numbers according to the setting of grouping variables.

#### **Usage**

```
gen_norm(mean = 0, sd = 1, name = "e")
gen_v_norm(mean = 0, sd = 1, name = "v")
gen_v_sar(mean = 0, sd = 1, rho = 0.5, type = "rook", name)
gen_v_ar1(mean = 0, sd = 1, rho = 0.5, groupVar = "idD", timeVar = "idT", name)
gen_generic(generator, ..., groupVars = NULL, name)
```

#### **Arguments**

mean	the mean passed to the random number generator, for example rnorm.
sd	the standard deviation passed to the random number generator, for example rnorm.
name	name of variable as character in which random numbers are stored.
rho	the correlation used to create the variance covariance matrix for a SAR process - see cell2nb.
type	either "rook" or "queen". See cell2nb for details.
groupVar	a variable name identifying groups.
timeVar	a variable name identifying repeated measurements.
generator	a function producing random numbers.
	arguments passed to generator.
groupVars	names of variables as character. Identify groups within random numbers are constant.

#### **Details**

gen\_norm is used to draw random numbers from a normal distribution where all generated numbers are independent.

gen\_v\_norm and gen\_v\_sar will create an area-level random component. In the case of v\_norm, the error component will be from a normal distribution and i.i.d. from an area-level perspective (all units in an area will have the same value, all areas are independent). v\_sar will also be from a normal distribution, but the errors are correlated. The variance covariance matrix is constructed

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for a SAR(1) - spatial/simultanous autoregressive process. myrnorm is used for the random number generation. gen\_v\_norm and gen\_v\_sar expect a variable idD in the data identifying the areas.

gen\_generic can be used if your world is not normal. You can specify 'any' function as generator, like rnorm. Arguments in . . . are matched by name or position. The first argument of generator is expected to be the number of random numbers (not necessarily named n) and need not to be specified.

#### See Also

```
sim_gen, sim_gen_x, sim_gen_e, sim_gen_ec, sim_gen_v, sim_gen_vc, cell2nb
```

#### **Examples**

```
sim_base() %>% sim_gen_x() %>% sim_gen_e() %>% sim_gen_v() %>% sim_gen(gen_v_sar(name = "vSP"))

# Generic interface
set.seed(1)
dat1 <- sim(base_id() %>%
    sim_gen(gen_generic(rnorm, mean = 0, sd = 4, name = "e")))
set.seed(1)
dat2 <- sim(base_id() %>% sim_gen_e())
all.equal(dat1, dat2)
```

plot.sim\_setup

Plotting methods

#### **Description**

Use this function to produce plots for an object of class sim\_setup.

## Usage

```
## S3 method for class 'sim_setup'
plot(x, y, ...)
```

## **Arguments**

```
x a sim_setupy will be ignored
```

... Arguments to be passed to plot.

#### See Also

```
autoplot
```

8 sample\_fraction

mple_fraction Sampling functions
----------------------------------

## **Description**

These functions are intended to be used with sim\_sample and not interactively. They are wrappers around sample\_frac and sample\_n.

#### Usage

```
sample_fraction(size, replace = FALSE, weight = NULL, groupVars = NULL)
sample_number(size, replace = FALSE, weight = NULL, groupVars = NULL)
sample_numbers(size, replace = FALSE, groupVars = NULL)
sample_cluster_number(size, replace = FALSE, weight = NULL, groupVars)
sample_cluster_fraction(size, replace = FALSE, weight = NULL, groupVars)
```

## **Arguments**

size	<pre><tidy-select> For sample_n(), the number of rows to select. For sample_frac(), the fraction of rows to select. If tbl is grouped, size applies to each group.</tidy-select></pre>	
replace	Sample with or without replacement?	
weight	<tidy-select> Sampling weights. This must evaluate to a vector of non-negative numbers the same length as the input. Weights are automatically standardised to sum to 1.</tidy-select>	
groupVars	character with names of variables to be used for grouping.	

#### **Details**

```
sample_numbers is a vectorized version of sample_number. sample_cluster_number and sample_cluster_fraction will sample clusters (all units in a cluster).
```

```
sim_base_lm() %>% sim_sample(sample_number(5))
sim_base_lm() %>% sim_sample(sample_fraction(0.5))
sim_base_lm() %>% sim_sample(sample_cluster_number(5, groupVars = "idD"))
sim_base_lm() %>% sim_sample(sample_cluster_fraction(0.5, groupVars = "idD"))
```

```
show, sim_setup\_method Show for sim_setup
```

#### **Description**

This is the documentation for the show methods in the package saeSim. In case you don't know, show is for S4-classes like print for S3. If you don't know what that means, don't bother, there is no reason to call show directly, however there is the need to document it.

## Usage

```
## S4 method for signature 'sim_setup'
show(object)
## S4 method for signature 'summary.sim_setup'
show(object)
```

## Arguments

object

Any R object

#### **Details**

Will print the head of a sim\_setup to the console, after converting it to a data. frame.

sim

Start simulation

## **Description**

This function will start the simulation. Use the printing method as long as you are testing the scenario.

# Usage

```
sim(
    x,
    R = 1,
    path = NULL,
    overwrite = TRUE,
    ...,
    suffix = NULL,
    fileExt = ".csv",
    libs = NULL,
    exports = NULL
)
```

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#### **Arguments**

X	a sim_setup
R	number of repetitions.
path	optional path in which the simulation results can be saved. They will we coerced to a data. frame and then saved as 'csv'.
overwrite	TRUE/FALSE. If TRUE files in path are replaced. If FALSE files in path are not replaced and simulation will not be recomputed.
	arguments passed to parallelStart.
suffix	an optional suffix of file names.
fileExt	the file extension. Default is ".csv" - alternative it can be ".RData".
libs	arguments passed to parallelLibrary. Will be used in a call to do.call after coersion with as.list.
exports	arguments passed to parallelExport. Will be used in a call to do.call after coersion with as.list.

#### **Details**

The package parallelMap is utilized as back-end for parallel computations.

Use the argument path to store the simulation results in a directory. This may be a good idea for long running simulations and for those using large data. frames. You can use sim\_read\_data to read them in. The return value will change to NULL in each run.

#### Value

The return value is a list. The elements are the results of each simulation run, typically of class data. frame. In case you specified path, each element is NULL.

```
setup <- sim_base_lm()
resultList <- sim(setup, R = 1)

# For parallel computations you may need to export objects
localFun <- function() cat("Hello World!")
comp_fun <- function(dat) {
   localFun()
   dat
}

res <- sim_base_lm() %>%
   sim_comp_pop(comp_fun) %>%
   sim(
    R = 2,
    mode = "socket", cpus = 2,
    exports = "localFun"
)
```

sim\_agg

sim\_agg

Aggregation component

## **Description**

One of the components which can be added to a simulation set-up. Aggregating the data is a simulation component which can be used to aggregate the population or sample. The aggregation will simply be done after the sampling, if you haven't specified any sampling component, the population is aggregated (makes sense if you draw samples directly from the model).

## Usage

```
sim_agg(simSetup, aggFun = agg_all())
```

## **Arguments**

simSetup a sim\_setup.

aggFun function which controls the aggregation process. At the moment only agg\_all

is defined.

#### **Details**

Potentially you can define an aggFun yourself. Take care that it only has one argument, named dat, and returns the aggregated data as data. frame.

## See Also

```
agg_all, sim_gen, sim_comp_pop, sim_sample, , sim_comp_sample
```

```
# Aggregating the population:
sim_base_lm() %>% sim_agg()

# Aggregating after sampling:
sim_base_lm() %>% sim_sample() %>% sim_agg()

# User aggFun:
sim_base_lm() %>% sim_agg(function(dat) dat[1, ])
```

sim\_base\_lm

sim\_base

Base component

# Description

Use the 'sim\_base' functions to start a new sim\_setup.

## Usage

```
sim_base(data = base_id(100, 100))
```

## Arguments

data

a data.frame.

## **Examples**

```
# Example for a linear model:
sim_base() %>%
  sim_gen_x() %>%
  sim_gen_e()
```

 $sim\_base\_lm$ 

Preconfigured set-ups

## **Description**

sim\_base\_lm() will start a linear model: One regressor, one error component. sim\_base\_lmm() will start a linear mixed model: One regressor, one error component and one random effect for the domain. sim\_base\_lmc() and sim\_base\_lmmc() add outlier contamination to the scenarios. Use these as a quick start, then you probably want to configure your own scenario.

#### Usage

```
sim_base_lm()
sim_base_lmm()
sim_base_lmc()
sim_base_lmmc()
```

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#### **Details**

Additional information on the generated variables:

```
nDomains: 100 domains
nUnits: 100 in each domain
x: is normally distributed with mean of 0 and sd of 4
e: is normally distributed with mean of 0 and sd of 4
v: is normally distributed with mean of 0 and sd of 1, it is a constant within domains
e-cont: as e; probability of unit to be contaminated is 0.05; sd is then 150
v-cont: as v; probability of area to be contaminated is 0.05; sd is then 40
y = 100 + x + v + e
```

#### **Examples**

```
# The preconfigured set-ups:
sim_base_lm()
sim_base_lmm()
sim_base_lmc()
sim_base_lmmc()
```

 $sim\_comp\_n$ 

Preconfigured computation components

## **Description**

sim\_comp\_n and sim\_comp\_N will add the sample and population size in each domain respectively. sim\_comp\_popMean and sim\_comp\_popVar the population mean and variance of the variable y. The data is expected to have a variable idD identifying domains.

## Usage

```
sim_comp_n(simSetup)
sim_comp_N(simSetup)
sim_comp_popMean(simSetup)
sim_comp_popVar(simSetup)
```

# Arguments

```
simSetup a sim_setup.
```

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sim\_comp\_pop

Calculation component

#### **Description**

One of the components which can be added to a sim\_setup. These functions can be used for adding new variables to the data.

## Usage

```
sim_comp_pop(simSetup, fun = comp_var(), by = "")
sim_comp_sample(simSetup, fun = comp_var(), by = "")
sim_comp_agg(simSetup, fun = comp_var(), by = "")
```

## **Arguments**

simSetup a sim\_setup.

fun a function, see details.

by names of variables as character; identifying groups for which fun is applied.

### **Details**

Potentially you can define a function for computation yourself. Take care that it only has one argument, named dat, and returns a data.frame. Use comp\_var for simple data manipulation. Functions added with sim\_comp\_pop are applied before sampling; sim\_comp\_sample after sampling. Functions added with sim\_comp\_agg after aggregation.

#### See Also

comp\_var, sim\_gen, sim\_agg, sim\_sample, sim\_comp\_N, sim\_comp\_n, sim\_comp\_popMean, sim\_comp\_popVar

```
# Standard behavior
sim_base() %>% sim_gen_x() %>% sim_comp_N()

# Custom data modifications
## Add predicted values of a linear model
library(saeSim)

comp_lm <- function(dat) {
   dat$linearPredictor <- predict(lm(y ~ x, data = dat))
   dat
}

sim_base_lm() %>% sim_comp_pop(comp_lm)
```

sim\_gen 15

```
# or if applied after sampling
sim_base_lm() %>% sim_sample() %>% sim_comp_pop(comp_lm)
```

sim\_gen

Generation component

#### **Description**

One of the components which can be added to a sim\_setup.

#### Usage

```
sim_gen(simSetup, generator)
sim_gen_generic(simSetup, ...)
```

## Arguments

```
simSetup a sim_setup.
generator generator function used to generate random numbers.
... arguments passed to gen_generic.
```

#### **Details**

Potentially you can define a generator yourself. Take care that it has one argument, named dat, and returns a data.frame. sim\_gen\_generic is a shortcut to gen\_generic.

### See Also

```
gen_norm, gen_v_norm, gen_v_sar, sim_agg, , sim_comp_pop, sim_sample, sim_gen_x, sim_gen_e,
sim_gen_v, sim_gen_vc, sim_gen_ec
```

```
# Data setup for a mixed model
sim_base() %>% sim_gen_x() %>% sim_gen_v() %>% sim_gen_e()
# Adding contamination in the model error
sim_base() %>% sim_gen_x() %>% sim_gen_v() %>% sim_gen_e() %>% sim_gen_ec()
# Simple user defined generator:
gen_myVar <- function(dat) {
   dat["myVar"] <- rnorm(nrow(dat))
   dat
}
sim_base() %>% sim_gen_x() %>% sim_gen(gen_myVar)
# And a chi-sq(5) distributed 'random-effect':
sim_base() %>% sim_gen_generic(rchisq, df = 5, groupVars = "idD", name = "re")
```

sim\_gen\_cont

oim con cont	Comonation
sim_gen_cont	Generation

Generation Component for contamination

## **Description**

One of the components which can be added to a sim\_setup. It is applied after functions added with sim\_gen.

## Usage

```
sim_gen_cont(simSetup, generator, nCont, type, areaVar = NULL, fixed = TRUE)
```

#### **Arguments**

simSetup a sim\_setup.

generator generator function used to generate random numbers.

nCont gives the number of contaminated observations. Values between 0 and 1 will be

treated as probability. If type is 'unit' and length is larger than 1, the expected length is the number of areas. If type is 'area' and length is larger than 1 the values are interpreted as area positions; i.e. c(1, 3) is interpreted as the first

and 3rd area in the data is contaminated.

type "unit" or "area" - unit- or area-level contamination.

areaVar character with variable name(s) identifying areas.

fixed TRUE fixes the observations which will be contaminated. FALSE will result in

a random selection of observations or areas.

#### See Also

```
sim_gen
```

```
sim_base_lm() %>%
  sim_gen_cont(gen_norm(name = "e"), nCont = 0.05, type = "unit", areaVar = "idD") %>%
  as.data.frame
```

sim\_gen\_x

sim\_gen\_x

Preconfigured generation components

## Description

These are some preconfigured generation components and all wrappers around sim\_gen and sim\_gen\_cont.

## Usage

```
sim_gen_x(simSetup, mean = 0, sd = 4, name = "x")
sim_gen_e(simSetup, mean = 0, sd = 4, name = "e")
sim_gen_ec(
  simSetup,
 mean = 0,
 sd = 150,
 name = "e",
 nCont = 0.05,
  type = "unit",
 areaVar = "idD",
  fixed = TRUE
)
sim_gen_v(simSetup, mean = 0, sd = 1, name = "v")
sim_gen_vc(
 simSetup,
 mean = 0,
 sd = 40,
 name = "v"
 nCont = 0.05,
  type = "area",
 areaVar = "idD",
  fixed = TRUE
)
```

## Arguments

simSetup	a sim_setup.
mean	the mean passed to the random number generator, for example rnorm.
sd	the standard deviation passed to the random number generator, for example rnorm.
name	name of variable as character in which random numbers are stored.

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nCont gives the number of contaminated observations. Values between 0 and 1 will be

treated as probability. If type is 'unit' and length is larger than 1, the expected length is the number of areas. If type is 'area' and length is larger than 1 the values are interpreted as area positions; i.e. c(1, 3) is interpreted as the first

and 3rd area in the data is contaminated.

type "unit" or "area" - unit- or area-level contamination.

areaVar character with variable name(s) identifying areas.

fixed TRUE fixes the observations which will be contaminated. FALSE will result in

a random selection of observations or areas.

#### **Details**

x: fixed-effect component; e: model-error; ec: contaminated model error; v: random-effect (error constant for each domain); vc contaminated random-effect. Note that for contamination you are expected to add both, a non-contaminated component and a contaminated component.

sim\_read\_data

Read in simulated data

#### **Description**

Functions to read in simulation data from folder. Can be csv or RData files.

## Usage

```
sim_read_data(path, ..., returnList = FALSE)
sim_clear_data(path, ...)
sim_read_list(path)
sim_clear_list(path)
```

#### Arguments

path path to the files you want to read in.

... arguments passed to read.csv

returnList if TRUE a list containing the data.frames. Very much like the output of sim. If

FALSE a single data.frame is returned, using bind\_rows

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sim\_resp

Response component

## **Description**

One of the components which can be added to a sim\_setup.

## Usage

```
sim_resp(simSetup, respFun)
sim_resp_eq(simSetup, ...)
```

#### **Arguments**

 $\verb|simSetup| a \verb|sim\_setup|.$ 

respFun a function constructing the response variable

... <data-masking> Name-value pairs. The name gives the name of the column in

the output.

The value can be:

- A vector of length 1, which will be recycled to the correct length.
- A vector the same length as the current group (or the whole data frame if ungrouped).
- NULL, to remove the column.
- A data frame or tibble, to create multiple columns in the output.

#### **Details**

Potentially you can define an respFun yourself. Take care that it only has one argument, named dat, and returns the a data.frame.

#### See Also

```
agg_all, sim_gen, sim_comp_pop, sim_sample, , sim_comp_sample
```

```
base_id() %>% sim_gen_x() %>% sim_gen_e() %>% sim_resp_eq(y = 100 + 2 * x + e)
```

20 sim\_sample

sim\_sample

Sampling component

#### **Description**

One of the components which can be added to a sim\_setup. This component can be used to add a sampling mechanism to the simulation set-up. A sample will be drawn after the population is generated (sim\_gen) and variables on the population are computed (sim\_comp\_pop).

## Usage

```
sim_sample(simSetup, smplFun = sample_number(size = 5L, groupVars = "idD"))
```

#### **Arguments**

```
simSetup a sim_setup.

smplFun function which controls the sampling process.
```

#### **Details**

Potentially you can define a smplFun yourself. Take care that it has one argument, named dat being the data as data.frame, and returns the sample as data.frame.

#### See Also

```
sample_number, sample_fraction
```

```
# Simple random sample - 5% sample:
sim_base_lm() %>% sim_sample(sample_fraction(0.05))
# Simple random sampling proportional to size - 5% in each domain:
sim_base_lm() %>% sim_sample(sample_fraction(0.05, groupVars = "idD"))
# User defined sampling function:
sample_mySampleFun <- function(dat) {
   dat[sample.int(nrow(dat), 10), ]
}
sim_base_lm() %>% sim_sample(sample_mySampleFun)
```

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sim\_simName

Add a name to a sim\_setup

## Description

Use this function to add a name to a sim\_setup in case you are simulating different scenarios. This name will be added if you use the function sim for simulation

## Usage

```
sim_simName(simSetup, name)
```

## **Arguments**

```
simSetup a sim_setup.
name a character
```

## **Examples**

```
sim_base_lm() %>% sim_simName("newName")
```

```
summary,sim_setup-method
```

Summary for a sim\_setup

## Description

Reports a summary of the simulation setup.

## Usage

```
## S4 method for signature 'sim_setup'
summary(object, ...)
```

## **Arguments**

```
object a sim_setup.
... has no effect.
```

```
summary(sim_base_lm())
```

22 %>%

%>% Piping operator

# Description

This is the 'pipe operator' from the package 'magrittr'. Use it to chain all operations for the simulation together. See the original documentation for details: %>%.

# Usage

1hs %>% rhs

# Arguments

The value to be pipedA function or expression

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