

# Package ‘SimInf’

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**Title** A Framework for Data-Driven Stochastic Disease Spread Simulations

**Version** 6.2.0

**Description** Provides an efficient and very flexible framework to conduct data-driven epidemiological modeling in realistic large scale disease spread simulations. The framework integrates infection dynamics in subpopulations as continuous-time Markov chains using the Gillespie stochastic simulation algorithm and incorporates available data such as births, deaths and movements as scheduled events at predefined time-points. Using C code for the numerical solvers and 'OpenMP' (if available) to divide work over multiple processors ensures high performance when simulating a sample outcome. One of our design goals was to make the package extendable and enable usage of the numerical solvers from other R extension packages in order to facilitate complex epidemiological research. The package contains template models and can be extended with user-defined models.

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**License** GPL-3

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**NeedsCompilation** yes

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'SEIR.R' 'SIR.R' 'SISe.R' 'SISe3.R' 'SISe3\_sp.R' 'SISe\_sp.R'  
'SimInf.R' 'SimInf\_mparse.R' 'distance.R' 'package\_skeleton.R'  
'run\_outer.R'

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boxplot,SimInf\_model-method

*Box plot of number of individuals in each compartment*

**Description**

Produce box-and-whisker plot(s) of the number of individuals in each model compartment.

**Usage**

```
## S4 method for signature 'SimInf_model'
boxplot(x, ...)
```

**Arguments**

x                    The model to plot  
 ...                    Additional arguments affecting the plot produced.

**Examples**

```
## Create an 'SIR' model with 10 nodes and initialise
## it with 99 susceptible individuals and one infected
## individual. Let the model run over 100 days.
model <- SIR(u0 = data.frame(S = rep(99, 10),
                             I = rep(1, 10),
                             R = rep(0, 10)),
            tspan = 1:100,
            beta = 0.16,
            gamma = 0.077)

## Run the model and save the result.
result <- run(model, threads = 1)

## Create a boxplot
boxplot(result)
```

---

C\_code

*Extract the C code from a SimInf\_model object*


---

**Description**

Extract the C code from a SimInf\_model object

**Usage**

```
C_code(model, pkg = NULL)
```

**Arguments**

model                The SimInf\_model object to extract the C code from.  
 pkg                    Character vector. If the C code should be used in a package named pkg, the function modifies the C code to facilitate adding the code to the package. Default is to not use this argument and return the C code unmodified.

**Value**

Character vector with C code for the model.

**Examples**

```
## Use the model parser to create a 'SimInf_model' object that
## expresses an SIR model, where 'b' is the transmission rate and
## 'g' is the recovery rate.
model <- mparse(transitions = c("S -> b*S*I/(S+I+R) -> I", "I -> g*I -> R"),
               compartments = c("S", "I", "R"),
               gdata = c(b = 0.16, g = 0.077),
               u0 = data.frame(S = 99, I = 1, R = 0),
               tspan = 1:10)

## View the C code.
C_code(model)

## Modify the C code for a package named "XYZ"
C_code(model, "XYZ")
```

---

distance_matrix	<i>Create a distance matrix between nodes for spatial models</i>
-----------------	--

---

**Description**

Calculate the euclidian distances between coordinates for all coordinates within the cutoff.

**Usage**

```
distance_matrix(x, y, cutoff, min_dist = NULL)
```

**Arguments**

x	Projected x coordinate
y	Projected y coordinate
cutoff	The distance cutoff
min_dist	The minimum distance to separate two nodes. If the coordinates for two nodes are identical, the min_dist must be assigned or an error is raised. Default is NULL i.e. to raise an error.

**Value**

dgCMatrix

**Examples**

```
## Generate a grid 10 x 10 and place one node in each cell
## separated by 100m.
nodes <- expand.grid(x = (0:9) * 100, y = (0:9) * 100)
plot(y ~ x, nodes)

## Define the cutoff to only include neighbors within 300m.
```

```
d <- distance_matrix(x = nodes$x, y = nodes$y, cutoff = 301)

## View the first 10 rows and columns in the distance matrix
d[1:10, 1:10]
```

---

events	<i>Extract the events from a SimInf_model object</i>
--------	--

---

## Description

Extract the scheduled events from a SimInf\_model object.

## Usage

```
events(model)
```

## Arguments

model            The model to extract the events from.

## Value

[SimInf\\_events](#) object.

## Examples

```
## Create an SIR model that includes scheduled events.
model <- SIR(u0 = u0_SIR(),
            tspan = 1:(4 * 365),
            events = events_SIR(),
            beta = 0.16,
            gamma = 0.077)

## Extract the scheduled events from the model and display summary
summary(events(model))

## Extract the scheduled events from the model and plot them
plot(events(model))
```

---

`events_SEIR`*Example data to initialize events for the 'SEIR' model*

---

## Description

Example data to initialize scheduled events for a population of 1600 nodes and demonstrate the [SEIR](#) model.

## Usage

```
events_SEIR()
```

## Details

Example data to initialize scheduled events (see [SimInf\\_events](#)) for a population of 1600 nodes and demonstrate the [SEIR](#) model. The dataset contains 466692 events for 1600 nodes distributed over  $4 * 365$  days. The events are divided into three types: 'Exit' events remove individuals from the population ( $n = 182535$ ), 'Enter' events add individuals to the population ( $n = 182685$ ), and 'External transfer' events move individuals between nodes in the population ( $n = 101472$ ). The vignette contains a detailed description of how scheduled events operate on a model.

## Value

A data.frame

## Examples

```
## Create an 'SEIR' model with 1600 nodes and initialize
## it to run over 4*365 days. Add one infected individual
## to the first node.
u0 <- u0_SEIR()
u0$I[1] <- 1
tspan <- seq(from = 1, to = 4*365, by = 1)
model <- SEIR(u0      = u0,
              tspan   = tspan,
              events  = events_SEIR(),
              beta    = 0.16,
              epsilon = 0.25,
              gamma   = 0.01)

## Display the number of individuals affected by each event type
## per day.
plot(events(model))

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)
plot(result)

## Summarize the trajectory. The summary includes the number of
```

```
## events by event type.
summary(result)
```

---

events\_SIR

*Example data to initialize events for the 'SIR' model*


---

## Description

Example data to initialize scheduled events for a population of 1600 nodes and demonstrate the [SIR](#) model.

## Usage

```
events_SIR()
```

## Details

Example data to initialize scheduled events (see [SimInf\\_events](#)) for a population of 1600 nodes and demonstrate the [SIR](#) model. The dataset contains 466692 events for 1600 nodes distributed over  $4 * 365$  days. The events are divided into three types: 'Exit' events remove individuals from the population ( $n = 182535$ ), 'Enter' events add individuals to the population ( $n = 182685$ ), and 'External transfer' events move individuals between nodes in the population ( $n = 101472$ ). The vignette contains a detailed description of how scheduled events operate on a model.

## Value

A data.frame

## Examples

```
## Create an 'SIR' model with 1600 nodes and initialize
## it to run over 4*365 days. Add one infected individual
## to the first node.
u0 <- u0_SIR()
u0$I[1] <- 1
tspan <- seq(from = 1, to = 4*365, by = 1)
model <- SIR(u0      = u0,
             tspan   = tspan,
             events  = events_SIR(),
             beta    = 0.16,
             gamma   = 0.01)

## Display the number of individuals affected by each event type
## per day.
plot(events(model))

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)
plot(result)
```



```
## Summarize the trajectory. The summary includes the number of
## events by event type.
summary(result)
```

---

events\_SISe

*Example data to initialize events for the 'SISe' model*


---

## Description

Example data to initialize scheduled events for a population of 1600 nodes and demonstrate the [SISe](#) model.

## Usage

```
events_SISe()
```

## Details

Example data to initialize scheduled events (see [SimInf\\_events](#)) for a population of 1600 nodes and demonstrate the [SISe](#) model. The dataset contains 466692 events for 1600 nodes distributed over  $4 * 365$  days. The events are divided into three types: 'Exit' events remove individuals from the population ( $n = 182535$ ), 'Enter' events add individuals to the population ( $n = 182685$ ), and 'External transfer' events move individuals between nodes in the population ( $n = 101472$ ). The vignette contains a detailed description of how scheduled events operate on a model.

## Value

A data.frame

## Examples

```
## Create an 'SISe' model with 1600 nodes and initialize
## it to run over 4*365 days. Add one infected individual
## to the first node.
u0 <- u0_SISe()
u0$I[1] <- 1
tspan <- seq(from = 1, to = 4*365, by = 1)
model <- SISe(u0 = u0, tspan = tspan, events = events_SISe(),
             phi = 0, upsilon = 1.8e-2, gamma = 0.1, alpha = 1,
             beta_t1 = 1.0e-1, beta_t2 = 1.0e-1, beta_t3 = 1.25e-1,
             beta_t4 = 1.25e-1, end_t1 = 91, end_t2 = 182,
             end_t3 = 273, end_t4 = 365, epsilon = 0)

## Display the number of individuals affected by each event type
## per day.
plot(events(model))

## Run the model to generate a single stochastic trajectory.
```

```

result <- run(model, threads = 1)

## Summarize the trajectory. The summary includes the number of
## events by event type.
summary(result)

```

---

events\_SISe3

*Example data to initialize events for the 'SISe3' model*


---

## Description

Example data to initialize scheduled events for a population of 1600 nodes and demonstrate the [SISe3](#) model.

## Usage

```
data(events_SISe3)
```

## Format

A data.frame

## Details

Example data to initialize scheduled events (see [SimInf\\_events](#)) for a population of 1600 nodes and demonstrate the [SISe3](#) model. The dataset contains 783773 events for 1600 nodes distributed over  $4 * 365$  days. The events are divided into three types: 'Exit' events remove individuals from the population ( $n = 182535$ ), 'Enter' events add individuals to the population ( $n = 182685$ ), 'Internal transfer' events move individuals between compartments within one node e.g. ageing ( $n = 317081$ ), and 'External transfer' events move individuals between nodes in the population ( $n = 101472$ ). The vignette contains a detailed description of how scheduled events operate on a model.

## Examples

```

## Create an 'SISe3' model with 1600 nodes and initialize
## it to run over 4*365 days. Add one infected individual
## to the first node.
data("u0_SISe3", package = "SimInf")
data("events_SISe3", package = "SimInf")
u0_SISe3$I_1[1] <- 1
tspan <- seq(from = 1, to = 4*365, by = 1)
model <- SISe3(u0 = u0_SISe3, tspan = tspan, events = events_SISe3,
  phi = rep(0, nrow(u0_SISe3)), epsilon_1 = 1.8e-2,
  epsilon_2 = 1.8e-2, epsilon_3 = 1.8e-2,
  gamma_1 = 0.1, gamma_2 = 0.1, gamma_3 = 0.1,
  alpha = 1, beta_t1 = 1.0e-1, beta_t2 = 1.0e-1,
  beta_t3 = 1.25e-1, beta_t4 = 1.25e-1, end_t1 = 91,
  end_t2 = 182, end_t3 = 273, end_t4 = 365, epsilon = 0)

```

```
## Display the number of individuals affected by each event type
## per day.
plot(events(model))

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)

## Summarize the trajectory. The summary includes the number of
## events by event type.
summary(result)
```

---

gdata

*Extract global data from a SimInf\_model object*

---

## Description

The global data is a numeric vector that is common to all nodes. The global data vector is passed as an argument to the transition rate functions and the post time step function.

## Usage

```
gdata(model)
```

## Arguments

model            The model to get global data from.

## Value

a numeric vector

## Examples

```
## Create an SIR model
model <- SIR(u0 = data.frame(S = 99, I = 1, R = 0),
            tspan = 1:5, beta = 0.16, gamma = 0.077)

## Set 'beta' to a new value
gdata(model, "beta") <- 2

## Extract the global data vector that is common to all nodes
gdata(model)
```

---

`gdata<-`                                    *Set a global data parameter for a SimInf\_model object*

---

### Description

The global data is a numeric vector that is common to all nodes. The global data vector is passed as an argument to the transition rate functions and the post time step function.

### Usage

```
gdata(model, parameter) <- value
```

### Arguments

<code>model</code>	The model to set a global model parameter for.
<code>parameter</code>	The name of the parameter to set.
<code>value</code>	A numeric value.

### Value

a SimInf\_model object

### Examples

```
## Create an SIR model
model <- SIR(u0 = data.frame(S = 99, I = 1, R = 0),
            tspan = 1:5, beta = 0.16, gamma = 0.077)

## Set 'beta' to a new value
gdata(model, "beta") <- 2

## Extract the global data vector that is common to all nodes
gdata(model)
```

---

`indegree`                                    *Determine in-degree for each node in a model*

---

### Description

The number of nodes with inward *external transfer* events to each node.

### Usage

```
indegree(model)
```

**Arguments**

model                    determine in-degree for each node in the model.

**Value**

vector with in-degree for each node.

**Examples**

```
## Create an 'SIR' model with 1600 nodes and initialize
## it with example data.
model <- SIR(u0 = u0_SIR(), tspan = 1:1460, events = events_SIR(),
            beta = 0.16, gamma = 0.077)

## Display indegree for each node in the model.
plot(indegree(model))
```

---

ldata                    *Extract local data from a node*

---

**Description**

The local data is a numeric vector that is specific to a node. The local data vector is passed as an argument to the transition rate functions and the post time step function.

**Usage**

```
ldata(model, node)
```

**Arguments**

model                    The model to get local data from.  
node                      index to node to extract local data from.

**Value**

a numeric vector

**Examples**

```
## Create an 'SISe' model with 1600 nodes.
model <- SISe(u0 = u0_SISe(), tspan = 1:100, events = events_SISe(),
            phi = 0, upsilon = 1.8e-2, gamma = 0.1, alpha = 1,
            beta_t1 = 1.0e-1, beta_t2 = 1.0e-1, beta_t3 = 1.25e-1,
            beta_t4 = 1.25e-1, end_t1 = c(91, 101), end_t2 = c(182, 185),
            end_t3 = c(273, 275), end_t4 = c(365, 360), epsilon = 0)

## Display local data from the first two nodes.
ldata(model, node = 1)
ldata(model, node = 2)
```

mparse

*Model parser to define new models to run in SimInf***Description**

Describe your model in a logical way in R. `mparse` creates a `SimInf_model` object with your model definition that is ready to `run`.

**Usage**

```
mparse(transitions = NULL, compartments = NULL, ldata = NULL,
       gdata = NULL, u0 = NULL, v0 = NULL, tspan = NULL,
       events = NULL, E = NULL, N = NULL, pts_fun = NULL)
```

**Arguments**

<code>transitions</code>	character vector containing transitions on the form "X -> ... -> Y". The left (right) side is the initial (final) state and the propensity is written in between the ->-signs. The special symbol @ is reserved for the empty set. For example, <code>transitions = c("S -&gt; k1*S*I -&gt; I", "I -&gt; k2*I -&gt; R")</code> expresses a SIR model.
<code>compartments</code>	contains the names of the involved compartments, for example, <code>compartments = c("S", "I", "R")</code> .
<code>ldata</code>	optional data for the nodes. Can be specified either as a numeric matrix where column <code>ldata[, j]</code> contains the local data vector for the node <code>j</code> or as a <code>data.frame</code> with one row per node. If it's specified as a matrix, it must have row names to identify the parameters in the transitions. If it's specified as a <code>data.frame</code> , each column is one parameter. The local data vector is passed as an argument to the transition rate functions and the post time step function.
<code>gdata</code>	optional data that are common to all nodes in the model. Can be specified either as a named numeric vector or as a one-row <code>data.frame</code> . The names are used to identify the parameters in the transitions. The global data vector is passed as an argument to the transition rate functions and the post time step function.
<code>u0</code>	A <code>data.frame</code> (or an object that can be coerced to a <code>data.frame</code> with <code>as.data.frame</code> ) with the initial state i.e. the number of individuals in each compartment in each node when the simulation starts..
<code>v0</code>	optional data with the initial continuous state in each node. Can be specified either as a <code>data.frame</code> with one row per node or as a numeric matrix where column <code>v0[, j]</code> contains the initial state vector for the node <code>j</code> . If <code>v0</code> is specified as a <code>data.frame</code> , each column is one parameter. If <code>v0</code> is specified as a matrix, the row names identify the parameters. The 'v' vector is passed as an argument to the transition rate functions and the post time step function. The continuous state can be updated in the post time step function.
<code>tspan</code>	A vector (length $\geq 2$ ) of increasing time points where the state of each node is to be returned. Can be either an integer or a Date vector. A Date vector is coerced to a numeric vector as days, where <code>tspan[1]</code> becomes the day of the

	year of the first year of tspan. The dates are added as names to the numeric vector.
events	A data.frame with the scheduled events. Default is NULL i.e. no scheduled events in the model.
E	matrix to handle scheduled events, see <a href="#">SimInf_events</a> . Default is NULL i.e. no scheduled events in the model.
N	matrix to handle scheduled events, see <a href="#">SimInf_events</a> . Default is NULL i.e. no scheduled events in the model.
pts_fun	optional character vector with C code for the post time step function. The C code should contain only the body of the function i.e. the code between the opening and closing curly brackets.

### Value

a [SimInf\\_model](#) object

### Examples

```
## Not run:
## Use the model parser to create a 'SimInf_model' object that
## expresses a SIR model, where 'b' is the transmission rate and
## 'g' is the recovery rate.
model <- mparse(transitions = c("S -> b*S*I/(S+I+R) -> I",
                               "I -> g*I -> R"),
               compartments = c("S", "I", "R"),
               gdata = c(b = 0.16, g = 0.077),
               u0 = data.frame(S = 100, I = 1, R = 0),
               tspan = 1:100)

## Run and plot the result
set.seed(22)
result <- run(model, threads = 1)
plot(result)

## End(Not run)
```

---

Nn

*Extract number of nodes in a model*

---

### Description

Extract number of nodes in a model.

### Usage

```
Nn(model)
```

**Arguments**

`model` the model object to extract the number of nodes from.

**Value**

the number of nodes in the model.

**Examples**

```
## Create an 'SIR' model with 100 nodes, with 99 susceptible,
## 1 infected and 0 recovered in each node.
u0 <- data.frame(S = rep(99, 100), I = rep(1, 100), R = rep(0, 100))
model <- SIR(u0 = u0, tspan = 1:10, beta = 0.16, gamma = 0.077)

## Display the number of nodes in the model.
Nn(model)
```

---

nodes

*Example data with spatial distribution of nodes*

---

**Description**

Example data to initialize a population of 1600 nodes and demonstrate various models.

**Usage**

```
data(nodes)
```

**Format**

A data.frame

**Examples**

```
## Create an 'SIR' model with 1600 nodes and initialize
## it to run over 4*365 days. Add one infected individual
## to the first node.
u0 <- u0_SIR()
u0$I[1] <- 1
tspan <- seq(from = 1, to = 4*365, by = 1)
model <- SIR(u0      = u0,
             tspan  = tspan,
             events = events_SIR(),
             beta   = 0.16,
             gamma  = 0.077)

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)
```



```

## Determine nodes with one or more infected individuals in the
## trajectory. Extract the 'I' compartment and check for any
## infected individuals in each node.
infected <- colSums(trajectory(result, ~ I, as.is = TRUE)) > 0

## Display infected nodes in 'blue' and non-infected nodes in 'yellow'.
data("nodes", package = "SimInf")
col <- ifelse(infected, "blue", "yellow")
plot(y ~ x, nodes, col = col, pch = 20, cex = 2)

```

---

outdegree	<i>Determine out-degree for each node in a model</i>
-----------	--

---

### Description

The number nodes that are connected with *external transfer* events from each node.

### Usage

```
outdegree(model)
```

### Arguments

model            determine out-degree for each node in the model.

### Value

vector with out-degree for each node.

### Examples

```

## Create an 'SIR' model with 1600 nodes and initialize
## it with example data.
model <- SIR(u0 = u0_SIR(), tspan = 1:1460, events = events_SIR(),
            beta = 0.16, gamma = 0.077)

## Display outdegree for each node in the model.
plot(outdegree(model))

```

---

package_skeleton	<i>Create a package skeleton from a SimInf_model</i>
------------------	--

---

### Description

Describe your model in a logical way in R, then `mparse` creates a `SimInf_model` object with your model definition that can be installed as an add-on R package.

### Usage

```
package_skeleton(model, name = NULL, path = ".", author = NULL,  
  email = NULL, maintainer = NULL, license = "GPL-3")
```

### Arguments

<code>model</code>	The model <code>SimInf_model</code> object with your model to create the package skeleton from.
<code>name</code>	Character string: the package name and directory name for your package.
<code>path</code>	Path to put the package directory in. Default is '.' i.e. the current directory.
<code>author</code>	Author of the package.
<code>email</code>	Email of the package maintainer.
<code>maintainer</code>	Maintainer of the package.
<code>license</code>	License of the package. Default is 'GPL-3'.

### Value

invisible NULL.

### References

Read the *Writing R Extensions* manual for more details.

Once you have created a *source* package you need to install it: see the *R Installation and Administration* manual, [INSTALL](#) and [install.packages](#).

---

pairs, SimInf\_model-method

*Scatterplot of number of individuals in each compartment*

---

### Description

A matrix of scatterplots with the number of individuals in each compartment is produced. The  $ij$ th scatterplot contains  $x[, i]$  plotted against  $x[, j]$ .

### Usage

```
## S4 method for signature 'SimInf_model'
pairs(x, ...)
```

### Arguments

`x`                    The model to plot  
`...`                 Additional arguments affecting the plot produced.

### Examples

```
## Create an 'SIR' model with 10 nodes and initialise
## it with 99 susceptible individuals and one infected
## individual. Let the model run over 100 days.
model <- SIR(u0 = data.frame(S = rep(99, 10),
                             I = rep(1, 10),
                             R = rep(0, 10)),
            tspan = 1:100,
            beta = 0.16,
            gamma = 0.077)

## Run the model and save the result.
result <- run(model, threads = 1)

## Create a scatter plot
pairs(result)
```

---

plot, SimInf\_events-method

*Display the distribution of scheduled events over time*

---

### Description

Display the distribution of scheduled events over time

**Usage**

```
## S4 method for signature 'SimInf_events'
plot(x, frame.plot = FALSE, ...)
```

**Arguments**

x	The events data to plot.
frame.plot	Draw a frame around each plot. Default is FALSE.
...	Additional arguments affecting the plot

---

plot,SimInf\_model-method

*Display the outcome from a simulated trajectory*

---

**Description**

Plot either the median and the quantile range of the counts in all nodes, or plot the counts in specified nodes.

**Usage**

```
## S4 method for signature 'SimInf_model'
plot(x, legend = NULL, col = NULL,
     lty = NULL, lwd = 2, compartments = NULL, node = NULL,
     range = 0.5, ...)
```

**Arguments**

x	The model to plot
legend	The character vector to appear in the legend. Default is to use the names of the compartments.
col	The plotting color for each compartment. Default is black.
lty	The line type for each compartment. Default is the sequence: 1=solid, 2=dashed, 3=dotted, 4=dotdash, 5=longdash, 6=twodash.
lwd	The line width for each compartment. Default is 2.
compartments	Character vector with the compartments in the model to include in the plot. Default is NULL i.e. include all compartments in the model.
node	indices specifying the nodes to include when plotting data. Plot one line for each node. Default (node = NULL) is to extract data from all nodes and plot the median count for the specified compartments.
range	show the quantile range of the count in each compartment. Default is to show the interquartile range i.e. the middle 50% of the count in transparent color. The median value is shown in the same color. Use range = 0.95 to show the middle 95% of the count. To display individual lines for each node, specify range = FALSE.
...	Additional arguments affecting the plot produced.

**Examples**

```

## Not run:
## Create an 'SIR' model with 100 nodes and initialise
## it with 990 susceptible individuals and 10 infected
## individuals in each node. Run the model over 100 days.
model <- SIR(u0 = data.frame(S = rep(990, 100),
                             I = rep(10, 100),
                             R = rep(0, 100)),
            tspan = 1:100,
            beta = 0.16,
            gamma = 0.077)

## Run the model and save the result.
result <- run(model)

## Plot the median and interquartile range of the number
## of susceptible, infected and recovered individuals.
plot(result)

## Plot the median and the middle 95% quantile range of the
## number of susceptible, infected and recovered individuals.
plot(result, range = 0.95)

## Plot the median and interquartile range of the number
## of infected individuals.
plot(result, compartments = "I")

## Plot the number of susceptible, infected
## and recovered individuals in the first
## three nodes.
plot(result, node = 1:3, range = FALSE)

## Plot the number of infected individuals in the first node.
plot(result, compartments = "I", node = 1, range = FALSE)

## End(Not run)

```

---

prevalence

*Calculate prevalence from a model object with trajectory data*


---

**Description**

Calculate the proportion of individuals with disease in the population, or the proportion of nodes with at least one diseased individual, or the proportion of individuals with disease in each node.

**Usage**

```

prevalence(model, formula, type = c("pop", "nop", "wnp"), node = NULL,
           as.is = FALSE)

```

**Arguments**

model	The model with trajectory data to calculate the prevalence from.
formula	A formula that specify the compartments that define the cases with a disease or a condition (numerator), and the compartments that define the entire population of interest (denominator). The left hand side of the formula defines the cases, and the right hand side defines the population, for example, $I \sim S+I+R$ in a 'SIR' model (see 'Examples'). The $.$ (dot) is expanded to all compartments, for example, $I \sim .$ is expanded to $I \sim S+I+R$ in a 'SIR' model (see 'Examples').
type	The type of prevalence measure to calculate at each time point in <code>tspan</code> : <code>pop</code> (population prevalence) calculates the proportion of the individuals (cases) in the population, <code>nop</code> (node prevalence) calculates the proportion of nodes with at least one case, and <code>wnp</code> (within-node prevalence) calculates the proportion of cases within each node. Default is <code>pop</code> .
node	Indices specifying the subset nodes to include in the calculation of the prevalence. Default is <code>NULL</code> , which includes all nodes.
as.is	The default ( <code>as.is = FALSE</code> ) is to generate a <code>data.frame</code> with one row per time-step with the prevalence. Using <code>as.is = TRUE</code> returns the result as a matrix, which is the internal format.

**Value**

A `data.frame` if `as.is = FALSE`, else a matrix.

**Examples**

```
## Create an 'SIR' model with 6 nodes and initialize
## it to run over 10 days.
u0 <- data.frame(S = 100:105, I = c(0, 1, 0, 2, 0, 3), R = rep(0, 6))
model <- SIR(u0 = u0, tspan = 1:10, beta = 0.16, gamma = 0.077)

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)

## Determine the proportion of infected individuals (cases)
## in the population at the time-points in 'tspan'.
prevalence(result, I~S+I+R)

## Identical result is obtained with the shorthand 'I~.'
prevalence(result, I~.)

## Determine the proportion of nodes with infected individuals at
## the time-points in 'tspan'.
prevalence(result, I~S+I+R, type = "nop")

## Determine the proportion of infected individuals in each node
## at the time-points in 'tspan'.
prevalence(result, I~S+I+R, type = "wnp")
```

---

run *Run the SimInf stochastic simulation algorithm*

---

## Description

Run the SimInf stochastic simulation algorithm

## Usage

```
run(model, threads = NULL, solver = c("ssm", "aem"))

## S4 method for signature 'SimInf_model'
run(model, threads = NULL, solver = c("ssm",
  "aem"))
```

## Arguments

model	The siminf model to run.
threads	Number of threads. Default is NULL, i.e. to use all available processors.
solver	Which numerical solver to utilize. Default is 'ssm'.

## Value

[SimInf\\_model](#) object with result from simulation.

## References

- Bauer P, Engblom S, Widgren S (2016) "Fast Event-Based Epidemiological Simulations on National Scales" International Journal of High Performance Computing Applications, 30(4), 438-453. doi:10.1177/1094342016635723
- Bauer P., Engblom S. (2015) Sensitivity Estimation and Inverse Problems in Spatial Stochastic Models of Chemical Kinetics. In: Abdulle A., Deparis S., Kressner D., Nobile F., Picasso M. (eds) Numerical Mathematics and Advanced Applications - ENUMATH 2013. Lecture Notes in Computational Science and Engineering, vol 103. Springer, Cham. Doi: 10.1007/978-3-319-10705-9\_51

## Examples

```
## Create an 'SIR' model with 10 nodes and initialise
## it to run over 100 days.
model <- SIR(u0 = data.frame(S = rep(99, 10),
  I = rep(1, 10),
  R = rep(0, 10)),
  tspan = 1:100,
  beta = 0.16,
  gamma = 0.077)
```

```
## Run the model and save the result.
result <- run(model, threads = 1)

## Plot the proportion of susceptible, infected and recovered
## individuals.
plot(result)
```

---

run\_outer

*Run SimInf\_model on scaled parameters*


---

## Description

Run SimInf\_model on scaled parameters

## Usage

```
run_outer(x, y, model, formula = NULL, FUN = NULL, ...)
```

## Arguments

x	Scale the model gdata parameter values on the right hand side of the formula with x before calling FUN with the scaled model as argument.
y	Scale the model gdata parameter values on the left hand side of the formula with y before calling FUN with the scaled model as argument.
model	The siminf model to scale parameters on and run.
formula	The parameters in the gdata vector matching the left hand side of the formula $a + b \sim c$ will be scaled by y. The parameters in the gdata vector matching the right hand side of the formula $a + b \sim c$ will be scaled by x.
FUN	A function to use on the scaled model 'gdata' parameters.
...	Optional arguments to be passed to FUN.

## Value

Array with dimension  $c(\dim(x), \dim(y))$ .

## Examples

```
## Not run:
## Create an SIR-model with 500 nodes of 99 susceptible individuals
## and one infected individuals.
u0 <- data.frame(S = rep(99, 500), I = rep(1, 500), R = rep(0, 500))
model <- SIR(u0, 1:75, beta = 0.16, gamma = 0.077)

## Define scaling parameters
x <- seq(from = 0.2, to = 1.8, by = 0.05)
y <- seq(from = 0.2, to = 1.1, by = 0.05)
```



```

## Utility function to run the model and estimate the population
## prevalence on day 75.
pop_prev <- function(model) {
  result <- run(model)
  prevalence(result, I~., type = "pop", as.is = TRUE)[75]
}

## Scale 'gamma' with 'y' and 'beta' with 'x' and
## run the model and determine the population prevalence on day
## 500. For each combination of 'x' and 'y', the model parameters
## are scaled and the function 'pop_prev' called with the
## perturbed model.
pop <- run_outer(x, y, model, gamma ~ beta, pop_prev)

## Plot result
contour(x * model@gdata["beta"], y * model@gdata["gamma"],
        pop, method = "edge", bty = "l")

## Utility function to run the model and estimate the node
## prevalence on day 75.
node_prev <- function(model) {
  result <- run(model)
  prevalence(result, I~., type = "nop", as.is = TRUE)[75]
}

## Scale 'gamma' with 'y' and 'beta' with 'x' and
## run the model and determine the node prevalence on day
## 500. For each combination of 'x' and 'y', the model parameters
## are scaled and the function 'node_prev' called with the
## perturbed model.
nop <- run_outer(x, y, model, gamma ~ beta, node_prev)

## Plot result
contour(x * model@gdata["beta"], y * model@gdata["gamma"],
        nop, method = "edge", bty = "l")

## End(Not run)

```

---

SEIR

*Create an SEIR model*


---

### Description

Create an SEIR model to be used by the simulation framework.

### Usage

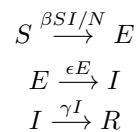
```
SEIR(u0, tspan, events = NULL, beta = NULL, epsilon = NULL,
     gamma = NULL)
```

**Arguments**

<code>u0</code>	A <code>data.frame</code> with the initial state in each node (see ‘Details’).
<code>tspan</code>	A vector (length $\geq 2$ ) of increasing time points where the state of each node is to be returned. Can be either an integer or a Date vector. A Date vector is coerced to a numeric vector as days, where <code>tspan[1]</code> becomes the day of the year of the first year of <code>tspan</code> . The dates are added as names to the numeric vector.
<code>events</code>	a <code>data.frame</code> with the scheduled events, see <a href="#">SimInf_model</a> .
<code>beta</code>	The transmission rate from susceptible to exposed.
<code>epsilon</code>	The incubation rate from exposed to infected.
<code>gamma</code>	The recovery rate from infected to recovered.

**Details**

The SEIR model contains four compartments; number of susceptible (S), number of exposed (E) (those who have been infected but are not yet infectious), number of infectious (I), and number of recovered (R). Moreover, it has three state transitions,



where  $\beta$  is the transmission rate,  $\epsilon$  is the incubation rate,  $\gamma$  is the recovery rate, and  $N = S + E + I + R$ .

The argument `u0` must be a `data.frame` with one row for each node with the following columns:

- S** The number of susceptible in each node
- E** The number of exposed in each node
- I** The number of infected in each node
- R** The number of recovered in each node

**Value**

SEIR

**Examples**

```
## Create a SEIR model object.
model <- SEIR(u0 = data.frame(S = 99, E = 0, I = 1, R = 0),
             tspan = 1:100,
             beta = 0.16,
             epsilon = 0.25,
             gamma = 0.077)

## Run the SEIR model and plot the result.
set.seed(3)
result <- run(model)
plot(result)
```

---

SEIR-class	<i>Definition of the 'SEIR' model</i>
------------	---------------------------------------

---

**Description**

Class to handle the SEIR [SimInf\\_model](#).

---

select_matrix	<i>Extract the select matrix from a SimInf_model object</i>
---------------	---

---

**Description**

Utility function to extract events@E from a SimInf\_model object, see [SimInf\\_events](#)

**Usage**

```
select_matrix(model)
```

**Arguments**

model            The model to extract the select matrix E from.

**Value**

[dgMatrix](#) object.

**Examples**

```
## Create an SIR model
model <- SIR(u0 = data.frame(S = 99, I = 1, R = 0),
            tspan = 1:5, beta = 0.16, gamma = 0.077)

## Extract the select matrix from the model
select_matrix(model)
```

---

`select_matrix<-`            *Set the select matrix for a SimInf\_model object*

---

### Description

Utility function to set `events@E` in a `SimInf_model` object, see [SimInf\\_events](#)

### Usage

```
select_matrix(model) <- value
```

### Arguments

<code>model</code>	The model to set the select matrix for.
<code>value</code>	A matrix.

### Examples

```
## Create an SIR model
model <- SIR(u0 = data.frame(S = 99, I = 1, R = 0),
            tspan = 1:5, beta = 0.16, gamma = 0.077)

## Set the select matrix
select_matrix(model) <- matrix(c(1, 0, 0, 1, 1, 1, 0, 0, 1), nrow = 3)

## Extract the select matrix from the model
select_matrix(model)
```

---

`shift_matrix`            *Extract the shift matrix from a SimInf\_model object*

---

### Description

Utility function to extract the shift matrix `events@N` from a `SimInf_model` object, see [SimInf\\_events](#)

### Usage

```
shift_matrix(model)
```

### Arguments

<code>model</code>	The model to extract the shift matrix <code>events@N</code> from.
--------------------	---

### Value

A matrix.

## Examples

```
## Create an SIR model
model <- SIR(u0 = data.frame(S = 99, I = 1, R = 0),
            tspan = 1:5, beta = 0.16, gamma = 0.077)

## Extract the shift matrix from the model
shift_matrix(model)
```

---

`shift_matrix<-`                    *Set the shift matrix for a SimInf\_model object*

---

## Description

Utility function to set `events@N` in a `SimInf_model` object, see [SimInf\\_events](#)

## Usage

```
shift_matrix(model) <- value
```

## Arguments

<code>model</code>	The model to set the shift matrix <code>events@N</code> .
<code>value</code>	A matrix.

## Value

`SimInf_model` object

## Examples

```
## Create an SIR model
model <- SIR(u0 = data.frame(S = 99, I = 1, R = 0),
            tspan = 1:5, beta = 0.16, gamma = 0.077)

## Set the shift matrix
shift_matrix(model) <- matrix(c(2, 1, 0), nrow = 3)

## Extract the shift matrix from the model
shift_matrix(model)
```

---

show,SimInf\_events-method

*Brief summary of SimInf\_events*

---

### **Description**

Shows the number of scheduled events.

### **Usage**

```
## S4 method for signature 'SimInf_events'  
show(object)
```

### **Arguments**

object            The SimInf\_events object

### **Value**

None (invisible 'NULL').

---

show,SimInf\_model-method

*Brief summary of SimInf\_model*

---

### **Description**

Brief summary of SimInf\_model

### **Usage**

```
## S4 method for signature 'SimInf_model'  
show(object)
```

### **Arguments**

object            The SimInf\_model object

### **Value**

None (invisible 'NULL').

## Examples

```
## Create an 'SIR' model with 10 nodes and initialise
## it to run over 100 days.
model <- SIR(u0 = data.frame(S = rep(99, 10),
                             I = rep(1, 10),
                             R = rep(0, 10)),
            tspan = 1:100,
            beta = 0.16,
            gamma = 0.077)

## Brief summary of the model
model

## Run the model and save the result
result <- run(model, threads = 1)

## Brief summary of the result. Note that 'U' and 'V' are
## non-empty after running the model.
result
```

## Description

The `SimInf` package provides a flexible framework for data-driven spatio-temporal disease spread modeling, designed to efficiently handle population demographics and network data. The framework integrates infection dynamics in each subpopulation as continuous-time Markov chains (CTMC) using the Gillespie stochastic simulation algorithm (SSA) and incorporates available data such as births, deaths or movements as scheduled events. A scheduled event is used to modify the state of a subpopulation at a predefined time-point.

## Details

The `SimInf_model` is central and provides the basis for the framework. A `SimInf_model` object supplies the state-change matrix, the dependency graph, the scheduled events, and the initial state of the system.

All predefined models in `SimInf` have a generating function, with the same name as the model, for example `SIR`.

A model can also be created from a model specification using the `mparse` method.

After a model is created, a simulation is started with a call to the `run` method and if execution is successful, it returns a modified `SimInf_model` object with a single stochastic solution trajectory attached to it.

`SimInf` provides several utility functions to inspect simulated data, for example, `show`, `summary` and `plot`. To facilitate custom analysis, it provides the `trajectory` and `prevalence` methods.

One of our design goal was to make `SimInf` extendable and enable usage of the numerical solvers from other R extension packages in order to facilitate complex epidemiological research. To support

this, SimInf has functionality to generate the required C and R code from a model specification, see [package\\_skeleton](#)

---

 SimInf\_events

 Create a SimInf\_events object
 

---

## Description

The argument events must be a data.frame with the following columns:

**event** Four event types are supported by the current solvers: *exit*, *enter*, *internal transfer*, and *external transfer*. When assigning the events, they can either be coded as a numerical value or a character string: *exit*; 0 or 'exit', *enter*; 1 or 'enter', *internal transfer*; 2 or 'intTrans', and *external transfer*; 3 or 'extTrans'. Internally in **SimInf**, the event type is coded as a numerical value.

**time** When the event occurs i.e., the event is processed when time is reached in the simulation. Can be either an integer or a Date vector. A Date vector is coerced to a numeric vector as days, where t0 determines the offset to match the time of the events to the model tspan vector.

**node** The node that the event operates on. Also the source node for an *external transfer* event.  $1 \leq \text{node}[i] \leq \text{Number of nodes}$ .

**dest** The destination node for an *external transfer* event i.e., individuals are moved from node to dest, where  $1 \leq \text{dest}[i] \leq \text{Number of nodes}$ . Set event = 0 for the other event types. dest is an integer vector.

**n** The number of individuals affected by the event.  $n[i] \geq 0$ .

**proportion** If  $n[i]$  equals zero, the number of individuals affected by event[i] is calculated by summing the number of individuals with proportion[i].  $0 \leq \text{proportion}[i] \leq 1$ .

**select** To process event[i], the compartments affected by the event are specified with select[i] together with the matrix E, where select[i] determines which column in E to use. The specific individuals affected by the event are proportionally sampled from the compartments corresponding to the non-zero entries in the specified column in E[, select[i]], where select is an integer vector.

**shift** Determines how individuals in *internal transfer* and *external transfer* events are shifted to enter another compartment. The sampled individuals are shifted according to column shift[i] in matrix N i.e., N[, shift[i]], where shift is an integer vector. See above for a description of N. Unsuad for the other event types.

## Usage

```
SimInf_events(E = NULL, N = NULL, events = NULL, t0 = NULL)
```



**Arguments**

<b>E</b>	Each row corresponds to one compartment in the model. The non-zero entries in a column indicates the compartments to include in an event. For the <i>exit</i> , <i>internal transfer</i> and <i>external transfer</i> events, a non-zero entry indicate the compartments to sample individuals from. For the <i>enter</i> event, all individuals enter first non-zero compartment. E is sparse matrix of class <code>dgCMatrix</code> .
<b>N</b>	Determines how individuals in <i>internal transfer</i> and <i>external transfer</i> events are shifted to enter another compartment. Each row corresponds to one compartment in the model. The values in a column are added to the current compartment of sampled individuals to specify the destination compartment, for example, a value of 1 in an entry means that sampled individuals in this compartment are moved to the next compartment. Which column to use for each event is specified by the <code>shift</code> vector (see below). N is an integer matrix.
<b>events</b>	A <code>data.frame</code> with events.
<b>t0</b>	If <code>events\$time</code> is a <code>Date</code> vector, then <code>t0</code> determines the offset to match the time of the events to the model <code>tspan</code> vector, see details. If <code>events\$time</code> is a numeric vector, then <code>t0</code> must be <code>NULL</code> .

**Value**

S4 class `SimInf_events`

---

`SimInf_events-class`    *Class "SimInf\_events"*

---

**Description**

Class to hold data for scheduled events to modify the discrete state of individuals in a node at a pre-defined time `t`.

**Slots**

- E** Each row corresponds to one compartment in the model. The non-zero entries in a column indicates the compartments to include in an event. For the *exit*, *internal transfer* and *external transfer* events, a non-zero entry indicate the compartments to sample individuals from. For the *enter* event, all individuals enter first non-zero compartment. E is sparse matrix of class `dgCMatrix`.
- N** Determines how individuals in *internal transfer* and *external transfer* events are shifted to enter another compartment. Each row corresponds to one compartment in the model. The values in a column are added to the current compartment of sampled individuals to specify the destination compartment, for example, a value of 1 in an entry means that sampled individuals in this compartment are moved to the next compartment. Which column to use for each event is specified by the `shift` vector (see below). N is an integer matrix.
- event** Type of event: 0) *exit*, 1) *enter*, 2) *internal transfer*, and 3) *external transfer*. Other values are reserved for future event types and not supported by the current solvers. Integer vector.

- time** Time of when the event occurs i.e., the event is processed when time is reached in the simulation. `time` is an integer vector.
- node** The node that the event operates on. Also the source node for an *external transfer* event. Integer vector.  $1 \leq \text{node}[i] \leq \text{Number of nodes}$ .
- dest** The destination node for an *external transfer* event i.e., individuals are moved from `node` to `dest`, where  $1 \leq \text{dest}[i] \leq \text{Number of nodes}$ . Set `event` = 0 for the other event types. `dest` is an integer vector.
- n** The number of individuals affected by the event. Integer vector.  $n[i] \geq 0$ .
- proportion** If  $n[i]$  equals zero, the number of individuals affected by `event[i]` is calculated by summing the number of individuals with `proportion[i]`. Numeric vector.  $0 \leq \text{proportion}[i] \leq 1$ .
- select** To process `event[i]`, the compartments affected by the event are specified with `select[i]` together with the matrix `E`, where `select[i]` determines which column in `E` to use. The specific individuals affected by the event are proportionally sampled from the compartments corresponding to the non-zero entries in the specified column in `E[, select[i]]`, where `select` is an integer vector.
- shift** Determines how individuals in *internal transfer* and *external transfer* events are shifted to enter another compartment. The sampled individuals are shifted according to column `shift[i]` in matrix `N` i.e., `N[, shift[i]]`, where `shift` is an integer vector. See above for a description of `N`. Unused for the other event types.

---

 SimInf\_model

 Create a SimInf\_model
 

---

## Description

Create a SimInf\_model

## Usage

```
SimInf_model(G, S, tspan, events = NULL, ldata = NULL, gdata = NULL,
             U = NULL, u0 = NULL, v0 = NULL, V = NULL, E = NULL, N = NULL,
             C_code = NULL)
```

## Arguments

- G** Dependency graph that indicates the transition rates that need to be updated after a given state transition has occurred. A non-zero entry in element `G[i, j]` indicates that transition rate `i` needs to be recalculated if the state transition `j` occurs. Sparse matrix ( $Nt \times Nt$ ) of object class `dgMatrix`.
- S** Each column corresponds to a transition, and execution of state transition `j` amounts to adding the `S[, j]` to the state vector of the node where the state transition occurred. Sparse matrix ( $Nc \times Nt$ ) of object class `dgMatrix`.

tspan	A vector (length $\geq 2$ ) of increasing time points where the state of each node is to be returned. Can be either an integer or a Date vector. A Date vector is coerced to a numeric vector as days, where tspan[1] becomes the day of the year of the first year of tspan. The dates are added as names to the numeric vector.
events	A data.frame with the scheduled events.
ldata	A matrix with local data for the nodes. The column ldata[, j] contains the local data vector for the node j. The local data vector is passed as an argument to the transition rate functions and the post time step function.
gdata	A numeric vector with global data that is common to all nodes. The global data vector is passed as an argument to the transition rate functions and the post time step function.
U	The result matrix with the number of individuals in each disease state in every node ( $N_n N_c \times \text{length}(\text{tspan})$ ). U[, j] contains the number of individuals in each disease state at tspan[j]. U[1:Nc, j] contains the state of node 1 at tspan[j]. U[(Nc + 1):(2 * Nc), j] contains the state of node 2 at tspan[j] etc.
u0	The initial state vector. Either a matrix ( $N_c \times N_n$ ) or a data.frame with the number of individuals in each compartment in every node.
v0	The initial continuous state vector in every node. ( $\text{dim}(\text{ldata})[1] \times N_n$ ). The continuous state vector is updated by the specific model during the simulation in the post time step function.
V	The result matrix for the real-valued continuous compartment state ( $N_n \text{dim}(\text{ldata})[1] \times \text{length}(\text{tspan})$ ). V[, j] contains the real-valued state of the system at tspan[j].
E	Sparse matrix to handle scheduled events, see <a href="#">SimInf_events</a> .
N	Sparse matrix to handle scheduled events, see <a href="#">SimInf_events</a> .
C_code	Character vector with optional model C code. If non-empty, the C code is written to a temporary C-file when the run method is called. The temporary C-file is compiled and the resulting DLL is dynamically loaded. The DLL is unloaded and the temporary files are removed after running the model.

**Value**

[SimInf\\_model](#)

---

SimInf\_model-class      *Class "SimInf\_model"*

---

**Description**

Class to handle the siminf data model

**Slots**

- G** Dependency graph that indicates the transition rates that need to be updated after a given state transition has occurred. A non-zero entry in element  $G[i, i]$  indicates that transition rate  $i$  needs to be recalculated if the state transition  $j$  occurs. Sparse matrix ( $Nt \times Nt$ ) of object class `dgMatrix`.
- S** Each column corresponds to a state transition, and execution of state transition  $j$  amounts to adding the  $S[, j]$  column to the state vector  $u[, i]$  of node  $i$  where the transition occurred. Sparse matrix ( $Nc \times Nt$ ) of object class `dgMatrix`.
- U** The result matrix with the number of individuals in each compartment in every node.  $U[, j]$  contains the number of individuals in each compartment at  $tspan[j]$ .  $U[1:Nc, j]$  contains the number of individuals in node 1 at  $tspan[j]$ .  $U[(Nc + 1):(2 * Nc), j]$  contains the number of individuals in node 2 at  $tspan[j]$  etc. Integer matrix ( $N_n N_c \times \text{length}(tspan)$ ).
- U\_sparse** If the model was run to write the solution to a sparse matrix (`dgMatrix`) the `U_sparse` contains the data and `U` is empty. The layout of the data in `U_sparse` is identical to `U`. Please note that `U_sparse` is numeric and `U` is integer.
- V** The result matrix for the real-valued continuous state.  $V[, j]$  contains the real-valued state of the system at  $tspan[j]$ . Numeric matrix ( $N_n \text{dim}(ldata)[1] \times \text{length}(tspan)$ ).
- V\_sparse** If the model was run to write the solution to a sparse matrix (`dgMatrix`) the `V_sparse` contains the data and `V` is empty. The layout of the data in `V_sparse` is identical to `V`.
- ldata** A matrix with local data for the nodes. The column  $ldata[, j]$  contains the local data vector for the node  $j$ . The local data vector is passed as an argument to the transition rate functions and the post time step function.
- gdata** A numeric vector with global data that is common to all nodes. The global data vector is passed as an argument to the transition rate functions and the post time step function.
- tspan** A vector of increasing time points where the state of each node is to be returned.
- u0** The initial state vector ( $N_c \times N_n$ ) with the number of individuals in each compartment in every node.
- v0** The initial value for the real-valued continuous state. Numeric matrix ( $\text{dim}(ldata)[1] \times N_n$ ).
- events** Scheduled events `SimInf_events`
- C\_code** Character vector with optional model C code. If non-empty, the C code is written to a temporary C-file when the run method is called. The temporary C-file is compiled and the resulting DLL is dynamically loaded. The DLL is unloaded and the temporary files are removed after running the model.

SIR

*Create an SIR model***Description**

Create an SIR model to be used by the simulation framework.

**Usage**

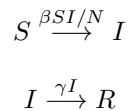
```
SIR(u0, tspan, events = NULL, beta = NULL, gamma = NULL)
```

**Arguments**

<code>u0</code>	A <code>data.frame</code> with the initial state in each node (see ‘Details’).
<code>tspan</code>	A vector (length $\geq 2$ ) of increasing time points where the state of each node is to be returned. Can be either an integer or a Date vector. A Date vector is coerced to a numeric vector as days, where <code>tspan[1]</code> becomes the day of the year of the first year of <code>tspan</code> . The dates are added as names to the numeric vector.
<code>events</code>	a <code>data.frame</code> with the scheduled events, see <a href="#">SimInf_model</a> .
<code>beta</code>	The transmission rate from susceptible to infected.
<code>gamma</code>	The recovery rate from infected to recovered.

**Details**

The SIR model contains three compartments; number of susceptible (S), number of infectious (I), and number of recovered (R). Moreover, it has two state transitions,



where  $\beta$  is the transmission rate,  $\gamma$  is the recovery rate, and  $N = S + I + R$ .

The argument `u0` must be a `data.frame` with one row for each node with the following columns:

**S** The number of susceptible in each node

**I** The number of infected in each node

**R** The number of recovered in each node

**Value**

A [SimInf\\_model](#) of class SIR

**Examples**

```
## Create an SIR model object.
model <- SIR(u0 = data.frame(S = 99, I = 1, R = 0),
            tspan = 1:100,
            beta = 0.16,
            gamma = 0.077)

## Run the SIR model and plot the result.
set.seed(22)
result <- run(model)
plot(result)
```

---

 SIR-class

*Definition of the SIR model*

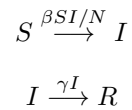

---

### Description

Class to handle the SIR [SimInf\\_model](#).

### Details

The SIR model contains three compartments; number of susceptible (S), number of infectious (I), and number of recovered (R). Moreover, it has two state transitions,



where  $\beta$  is the transmission rate,  $\gamma$  is the recovery rate, and  $N = S + I$ .

### Examples

```
## Create an SIR model object.
model <- SIR(u0 = data.frame(S = 99, I = 1, R = 0),
            tspan = 1:100,
            beta = 0.16,
            gamma = 0.077)

## Run the SIR model and plot the result.
set.seed(22)
result <- run(model)
plot(result)
```

---

 SISe

*Create a SISe model*


---

### Description

Create a SISe model to be used by the simulation framework.

### Usage

```
SISe(u0, tspan, events = NULL, phi = NULL, upsilon = NULL,
     gamma = NULL, alpha = NULL, beta_t1 = NULL, beta_t2 = NULL,
     beta_t3 = NULL, beta_t4 = NULL, end_t1 = NULL, end_t2 = NULL,
     end_t3 = NULL, end_t4 = NULL, epsilon = NULL)
```

**Arguments**

<code>u0</code>	A <code>data.frame</code> with the initial state in each node (see ‘Details’).
<code>tspan</code>	A vector (length $\geq 2$ ) of increasing time points where the state of each node is to be returned. Can be either an <code>integer</code> or a <code>Date</code> vector. A <code>Date</code> vector is coerced to a numeric vector as days, where <code>tspan[1]</code> becomes the day of the year of the first year of <code>tspan</code> . The dates are added as names to the numeric vector.
<code>events</code>	a <code>data.frame</code> with the scheduled events, see <a href="#">SimInf_model</a> .
<code>phi</code>	A numeric vector with the initial environmental infectious pressure in each node. Will be repeated to the length of <code>nrow(u0)</code> . Default is <code>NULL</code> which gives 0 in each node.
<code>upsilon</code>	Indirect transmission rate of the environmental infectious pressure
<code>gamma</code>	The recovery rate from infected to susceptible
<code>alpha</code>	Shed rate from infected individuals
<code>beta_t1</code>	The decay of the environmental infectious pressure in interval 1.
<code>beta_t2</code>	The decay of the environmental infectious pressure in interval 2.
<code>beta_t3</code>	The decay of the environmental infectious pressure in interval 3.
<code>beta_t4</code>	The decay of the environmental infectious pressure in interval 4.
<code>end_t1</code>	vector with the non-inclusive day of the year that ends interval 1 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>end_t2</code>	vector with the non-inclusive day of the year that ends interval 2 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>end_t3</code>	vector with the non-inclusive day of the year that ends interval 3 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>end_t4</code>	vector with the non-inclusive day of the year that ends interval 4 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>epsilon</code>	The background environmental infectious pressure

**Details**

The argument `u0` must be a `data.frame` with one row for each node with the following columns:

**S** The number of susceptible in each node

**I** The number of infected in each node

**Value**

SISe

**Beta**

The time dependent beta is divided into four intervals of the year

where  $0 \leq \text{day} < 365$

Case 1:  $\text{END}_1 < \text{END}_2 < \text{END}_3 < \text{END}_4$

INTERVAL_1	INTERVAL_2	INTERVAL_3	INTERVAL_4	INTERVAL_1
[0, END_1)	[END_1, END_2)	[END_2, END_3)	[END_3, END_4)	[END_4, 365)

Case 2:  $\text{END}_3 < \text{END}_4 < \text{END}_1 < \text{END}_2$

INTERVAL_3	INTERVAL_4	INTERVAL_1	INTERVAL_2	INTERVAL_3
[0, END_3)	[END_3, END_4)	[END_4, END_1)	[END_1, END_2)	[END_2, 365)

Case 3:  $\text{END}_4 < \text{END}_1 < \text{END}_2 < \text{END}_3$

INTERVAL_4	INTERVAL_1	INTERVAL_2	INTERVAL_3	INTERVAL_4
[0, END_4)	[END_4, END_1)	[END_1, END_2)	[END_2, END_3)	[END_3, 365)

---

SISe-class

*Definition of the SISe model*

---

**Description**

Class to handle the SISe [SimInf\\_model](#).

---

SISe3

*Create a 'SISe3' model*

---

**Description**

Create a 'SISe3' model to be used by the simulation framework.

**Usage**

```
SISe3(u0, tspan, events = NULL, phi = NULL,
      epsilon_1 = NULL, epsilon_2 = NULL, epsilon_3 = NULL,
      gamma_1 = NULL, gamma_2 = NULL, gamma_3 = NULL,
      alpha = NULL, beta_t1 = NULL, beta_t2 = NULL,
      beta_t3 = NULL, beta_t4 = NULL, end_t1 = NULL,
      end_t2 = NULL, end_t3 = NULL, end_t4 = NULL,
      epsilon = NULL)
```



**Arguments**

<code>u0</code>	A <code>data.frame</code> with the initial state in each node (see ‘Details’).
<code>tspan</code>	A vector (length $\geq 2$ ) of increasing time points where the state of each node is to be returned. Can be either an integer or a Date vector. A Date vector is coerced to a numeric vector as days, where <code>tspan[1]</code> becomes the day of the year of the first year of <code>tspan</code> . The dates are added as names to the numeric vector.
<code>events</code>	a <code>data.frame</code> with the scheduled events, see <a href="#">SimInf_model</a> .
<code>phi</code>	A numeric vector with the initial environmental infectious pressure in each node. Will be repeated to the length of <code>nrow(u0)</code> . Default is NULL which gives 0 in each node.
<code>upsilon_1</code>	Indirect transmission rate of the environmental infectious pressure in age category 1
<code>upsilon_2</code>	Indirect transmission rate of the environmental infectious pressure in age category 2
<code>upsilon_3</code>	Indirect transmission rate of the environmental infectious pressure in age category 3
<code>gamma_1</code>	The recovery rate from infected to susceptible for age category 1
<code>gamma_2</code>	The recovery rate from infected to susceptible for age category 2
<code>gamma_3</code>	The recovery rate from infected to susceptible for age category 3
<code>alpha</code>	Shed rate from infected individuals
<code>beta_t1</code>	The decay of the environmental infectious pressure in interval 1.
<code>beta_t2</code>	The decay of the environmental infectious pressure in interval 2.
<code>beta_t3</code>	The decay of the environmental infectious pressure in interval 3.
<code>beta_t4</code>	The decay of the environmental infectious pressure in interval 4.
<code>end_t1</code>	vector with the non-inclusive day of the year that ends interval 1 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>end_t2</code>	vector with the non-inclusive day of the year that ends interval 2 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>end_t3</code>	vector with the non-inclusive day of the year that ends interval 3 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>end_t4</code>	vector with the non-inclusive day of the year that ends interval 4 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>epsilon</code>	The background environmental infectious pressure

**Details**

The argument `u0` must be a `data.frame` with one row for each node with the following columns:

**S\_1** The number of susceptible in age category 1

**I\_1** The number of infected in age category 1

**S\_2** The number of susceptible in age category 2

**I\_2** The number of infected in age category 2

**S\_3** The number of susceptible in age category 3

**I\_3** The number of infected in age category 3

### Value

SISe3

### Beta

The time dependent beta is divided into four intervals of the year

where  $0 \leq \text{day} < 365$

Case 1:  $\text{END}_1 < \text{END}_2 < \text{END}_3 < \text{END}_4$

INTERVAL\_1 INTERVAL\_2 INTERVAL\_3 INTERVAL\_4 INTERVAL\_1  
 $[\emptyset, \text{END}_1)$   $[\text{END}_1, \text{END}_2)$   $[\text{END}_2, \text{END}_3)$   $[\text{END}_3, \text{END}_4)$   $[\text{END}_4, 365)$

Case 2:  $\text{END}_3 < \text{END}_4 < \text{END}_1 < \text{END}_2$

INTERVAL\_3 INTERVAL\_4 INTERVAL\_1 INTERVAL\_2 INTERVAL\_3  
 $[\emptyset, \text{END}_3)$   $[\text{END}_3, \text{END}_4)$   $[\text{END}_4, \text{END}_1)$   $[\text{END}_1, \text{END}_2)$   $[\text{END}_2, 365)$

Case 3:  $\text{END}_4 < \text{END}_1 < \text{END}_2 < \text{END}_3$

INTERVAL\_4 INTERVAL\_1 INTERVAL\_2 INTERVAL\_3 INTERVAL\_4  
 $[\emptyset, \text{END}_4)$   $[\text{END}_4, \text{END}_1)$   $[\text{END}_1, \text{END}_2)$   $[\text{END}_2, \text{END}_3)$   $[\text{END}_3, 365)$

---

SISe3-class

*Definition of the 'SISe3' model*

---

### Description

Class to handle the SISe3 [SimInf\\_model](#) model.

---

SISe3\_sp

*Create a SISe3\_sp model*

---

### Description

Create a SISe3\_sp model to be used by the simulation framework.

**Usage**

```
SISe3_sp(u0, tspan, events = NULL, phi = NULL, upsilon_1 = NULL,
         upsilon_2 = NULL, upsilon_3 = NULL, gamma_1 = NULL,
         gamma_2 = NULL, gamma_3 = NULL, alpha = NULL, beta_t1 = NULL,
         beta_t2 = NULL, beta_t3 = NULL, beta_t4 = NULL, end_t1 = NULL,
         end_t2 = NULL, end_t3 = NULL, end_t4 = NULL, distance = NULL,
         coupling = NULL)
```

**Arguments**

<code>u0</code>	A data.frame with the initial state in each node (see ‘Details’).
<code>tspan</code>	A vector (length $\geq 2$ ) of increasing time points where the state of each node is to be returned. Can be either an integer or a Date vector. A Date vector is coerced to a numeric vector as days, where <code>tspan[1]</code> becomes the day of the year of the first year of <code>tspan</code> . The dates are added as names to the numeric vector.
<code>events</code>	a data.frame with the scheduled events, see <a href="#">SimInf_model</a> .
<code>phi</code>	A numeric vector with the initial environmental infectious pressure in each node. Will be repeated to the length of <code>nrow(u0)</code> . Default is NULL which gives 0 in each node.
<code>upsilon_1</code>	Indirect transmission rate of the environmental infectious pressure in age category 1
<code>upsilon_2</code>	Indirect transmission rate of the environmental infectious pressure in age category 2
<code>upsilon_3</code>	Indirect transmission rate of the environmental infectious pressure in age category 3
<code>gamma_1</code>	The recovery rate from infected to susceptible for age category 1
<code>gamma_2</code>	The recovery rate from infected to susceptible for age category 2
<code>gamma_3</code>	The recovery rate from infected to susceptible for age category 3
<code>alpha</code>	Shed rate from infected individuals
<code>beta_t1</code>	The decay of the environmental infectious pressure in interval 1.
<code>beta_t2</code>	The decay of the environmental infectious pressure in interval 2.
<code>beta_t3</code>	The decay of the environmental infectious pressure in interval 3.
<code>beta_t4</code>	The decay of the environmental infectious pressure in interval 4.
<code>end_t1</code>	vector with the non-inclusive day of the year that ends interval 1 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>end_t2</code>	vector with the non-inclusive day of the year that ends interval 2 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>end_t3</code>	vector with the non-inclusive day of the year that ends interval 3 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>end_t4</code>	vector with the non-inclusive day of the year that ends interval 4 in each node. Will be repeated to the length of <code>nrow(u0)</code> .
<code>distance</code>	The distance matrix between neighboring nodes
<code>coupling</code>	The coupling between neighboring nodes

**Details**

The argument `u0` must be a `data.frame` with one row for each node with the following columns:

**S\_1** The number of susceptible in age category 1

**I\_1** The number of infected in age category 1

**S\_2** The number of susceptible in age category 2

**I\_2** The number of infected in age category 2

**S\_3** The number of susceptible in age category 3

**I\_3** The number of infected in age category 3

**Value**

SISe3\_sp

**Beta**

The time dependent beta is divided into four intervals of the year

where  $0 \leq \text{day} < 365$

Case 1:  $\text{END}_1 < \text{END}_2 < \text{END}_3 < \text{END}_4$

INTERVAL_1	INTERVAL_2	INTERVAL_3	INTERVAL_4	INTERVAL_1
$[\text{0}, \text{END}_1)$	$[\text{END}_1, \text{END}_2)$	$[\text{END}_2, \text{END}_3)$	$[\text{END}_3, \text{END}_4)$	$[\text{END}_4, 365)$

Case 2:  $\text{END}_3 < \text{END}_4 < \text{END}_1 < \text{END}_2$

INTERVAL_3	INTERVAL_4	INTERVAL_1	INTERVAL_2	INTERVAL_3
$[\text{0}, \text{END}_3)$	$[\text{END}_3, \text{END}_4)$	$[\text{END}_4, \text{END}_1)$	$[\text{END}_1, \text{END}_2)$	$[\text{END}_2, 365)$

Case 3:  $\text{END}_4 < \text{END}_1 < \text{END}_2 < \text{END}_3$

INTERVAL_4	INTERVAL_1	INTERVAL_2	INTERVAL_3	INTERVAL_4
$[\text{0}, \text{END}_4)$	$[\text{END}_4, \text{END}_1)$	$[\text{END}_1, \text{END}_2)$	$[\text{END}_2, \text{END}_3)$	$[\text{END}_3, 365)$

---

SISe3\_sp-class

*Definition of the 'SISe3\_sp' model*

---

**Description**

Class to handle the SISe3\_sp [SimInf\\_model](#) model.

---

SISe\_sp                      *Create a SISe\_sp model*

---

### Description

Create a SISe\_sp model to be used by the simulation framework.

### Usage

```
SISe_sp(u0, tspan, events = NULL, phi = NULL, upsilon = NULL,
        gamma = NULL, alpha = NULL, beta_t1 = NULL, beta_t2 = NULL,
        beta_t3 = NULL, beta_t4 = NULL, end_t1 = NULL, end_t2 = NULL,
        end_t3 = NULL, end_t4 = NULL, coupling = NULL, distance = NULL)
```

### Arguments

u0	A data.frame with the initial state in each node (see ‘Details’).
tspan	A vector (length >= 2) of increasing time points where the state of each node is to be returned. Can be either an integer or a Date vector. A Date vector is coerced to a numeric vector as days, where tspan[1] becomes the day of the year of the first year of tspan. The dates are added as names to the numeric vector.
events	a data.frame with the scheduled events, see <a href="#">SimInf_model</a> .
phi	A numeric vector with the initial environmental infectious pressure in each node. Will be repeated to the length of nrow(u0). Default is NULL which gives 0 in each node.
upsilon	Indirect transmission rate of the environmental infectious pressure
gamma	The recovery rate from infected to susceptible
alpha	Shed rate from infected individuals
beta_t1	The decay of the environmental infectious pressure in interval 1.
beta_t2	The decay of the environmental infectious pressure in interval 2.
beta_t3	The decay of the environmental infectious pressure in interval 3.
beta_t4	The decay of the environmental infectious pressure in interval 4.
end_t1	vector with the non-inclusive day of the year that ends interval 1 in each node. Will be repeated to the length of nrow(u0).
end_t2	vector with the non-inclusive day of the year that ends interval 2 in each node. Will be repeated to the length of nrow(u0).
end_t3	vector with the non-inclusive day of the year that ends interval 3 in each node. Will be repeated to the length of nrow(u0).
end_t4	vector with the non-inclusive day of the year that ends interval 4 in each node. Will be repeated to the length of nrow(u0).
coupling	The coupling between neighboring nodes
distance	The distance matrix between neighboring nodes

**Details**

The argument `u0` must be a `data.frame` with one row for each node with the following columns:

**S** The number of susceptible

**I** The number of infected

**Value**

SISe\_sp

**Beta**

The time dependent beta is divided into four intervals of the year

where  $0 \leq \text{day} < 365$

Case 1:  $\text{END}_1 < \text{END}_2 < \text{END}_3 < \text{END}_4$

INTERVAL\_1 INTERVAL\_2 INTERVAL\_3 INTERVAL\_4 INTERVAL\_1  
 $[\emptyset, \text{END}_1)$   $[\text{END}_1, \text{END}_2)$   $[\text{END}_2, \text{END}_3)$   $[\text{END}_3, \text{END}_4)$   $[\text{END}_4, 365)$

Case 2:  $\text{END}_3 < \text{END}_4 < \text{END}_1 < \text{END}_2$

INTERVAL\_3 INTERVAL\_4 INTERVAL\_1 INTERVAL\_2 INTERVAL\_3  
 $[\emptyset, \text{END}_3)$   $[\text{END}_3, \text{END}_4)$   $[\text{END}_4, \text{END}_1)$   $[\text{END}_1, \text{END}_2)$   $[\text{END}_2, 365)$

Case 3:  $\text{END}_4 < \text{END}_1 < \text{END}_2 < \text{END}_3$

INTERVAL\_4 INTERVAL\_1 INTERVAL\_2 INTERVAL\_3 INTERVAL\_4  
 $[\emptyset, \text{END}_4)$   $[\text{END}_4, \text{END}_1)$   $[\text{END}_1, \text{END}_2)$   $[\text{END}_2, \text{END}_3)$   $[\text{END}_3, 365)$

---

SISe\_sp-class

*Definition of the SISe\_sp model*

---

**Description**

Class to handle the SISe\_sp [SimInf\\_model](#).

---

summary,SimInf\_events-method

*Detailed summary of a SimInf\_events object*

---

**Description**

Shows the number of scheduled events and the number of scheduled events per event type.

### Usage

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

### Arguments

object	The SimInf_events object
...	Additional arguments affecting the summary produced.

### Value

None (invisible 'NULL').

---

summary, SimInf\_model-method

*Detailed summary of a SimInf\_model object*

---

### Description

Detailed summary of a SimInf\_model object

### Usage

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

### Arguments

object	The SimInf_model object
...	Additional arguments affecting the summary produced.

### Value

None (invisible 'NULL').

---

trajectory	<i>Extract data from a simulated trajectory</i>
------------	---

---

### Description

Extract the number of individuals in each compartment in every node after generating a single stochastic trajectory with `run`.

### Usage

```
trajectory(model, compartments = NULL, node = NULL, as.is = FALSE)
```

### Arguments

model	the model to extract the result from.
compartments	specify the names of the compartments to extract data from. The compartments can be specified as a character vector e.g. <code>compartments = c('S', 'I', 'R')</code> , or as a formula e.g. <code>compartments = ~S+I+R</code> (see ‘Examples’). Default ( <code>compartments=NULL</code> ) is to extract the number of individuals in each compartment i.e. the data from all discrete state compartments in the model. In models that also have continuous state variables e.g. the SISE model, use <code>~.</code> instead of <code>NULL</code> to also include these.
node	indices specifying the subset of nodes to include when extracting data. Default ( <code>node = NULL</code> ) is to extract data from all nodes.
as.is	the default ( <code>as.is = FALSE</code> ) is to generate a data.frame with one row per node and time-step with the number of individuals in each compartment. Using <code>as.is = TRUE</code> returns the result as a matrix, which is the internal format (see ‘Details’).

### Value

A data.frame if `as.is = FALSE`, else a matrix.

### Internal format of the discrete state variables

Description of the layout of the internal matrix (`U`) that is returned if `as.is = TRUE`. `U[, j]` contains the number of individuals in each compartment at `tspan[j]`. `U[1:Nc, j]` contains the number of individuals in node 1 at `tspan[j]`. `U[(Nc + 1):(2 * Nc), j]` contains the number of individuals in node 2 at `tspan[j]` etc, where `Nc` is the number of compartments in the model. The dimension of the matrix is  $N_n N_c \times \text{length}(\text{tspan})$  where  $N_n$  is the number of nodes.

### Internal format of the continuous state variables

Description of the layout of the matrix that is returned if `as.is = TRUE`. The result matrix for the real-valued continuous state. `V[, j]` contains the real-valued state of the system at `tspan[j]`. The dimension of the matrix is  $N_n \dim(\text{ldata})[1] \times \text{length}(\text{tspan})$ .



**Examples**

```

## Create an 'SIR' model with 6 nodes and initialize
## it to run over 10 days.
u0 <- data.frame(S = 100:105, I = 1:6, R = rep(0, 6))
model <- SIR(u0 = u0, tspan = 1:10, beta = 0.16, gamma = 0.077)

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)

## Extract the number of individuals in each compartment at the
## time-points in 'tspan'.
trajectory(result)

## Extract the number of recovered individuals in the first node
## at the time-points in 'tspan'.
trajectory(result, compartments = "R", node = 1)

## Extract the number of recovered individuals in the first and
## third node at the time-points in 'tspan'.
trajectory(result, compartments = "R", node = c(1, 3))

## Create an 'SISe' model with 6 nodes and initialize
## it to run over 10 days.
u0 <- data.frame(S = 100:105, I = 1:6)
model <- SISe(u0 = u0, tspan = 1:10, phi = rep(0, 6),
  epsilon = 0.02, gamma = 0.1, alpha = 1, epsilon = 1.1e-5,
  beta_t1 = 0.15, beta_t2 = 0.15, beta_t3 = 0.15, beta_t4 = 0.15,
  end_t1 = 91, end_t2 = 182, end_t3 = 273, end_t4 = 365)

## Run the model
result <- run(model, threads = 1)

## Extract the continuous state variable 'phi' which represents
## the environmental infectious pressure.
trajectory(result, "phi")

```

---

u0\_SEIR

*Example data to initialize the 'SEIR' model*


---

**Description**

Example data to initialize a population of 1600 nodes and demonstrate the [SEIR](#) model.

**Usage**

```
u0_SEIR()
```

**Details**

A data.frame with the number of individuals in the 'S', 'E', 'I' and 'R' compartments in 1600 nodes. Note that the 'E', 'I' and 'R' compartments are zero.

**Value**

A data.frame

**Examples**

```
## Create an 'SEIR' model with 1600 nodes and initialize it to
## run over 4*365 days and record data at weekly time-points.
## Add ten infected individuals to the first node.
u0 <- u0_SEIR()
u0$I[1] <- 10
tspan <- seq(from = 1, to = 4*365, by = 7)
model <- SEIR(u0      = u0,
              tspan   = tspan,
              events  = events_SEIR(),
              beta    = 0.16,
              epsilon = 0.25,
              gamma   = 0.01)

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)
plot(result)

## Summarize trajectory
summary(result)
```

---

u0\_SIR

*Example data to initialize the 'SIR' model*

---

**Description**

Example data to initialize a population of 1600 nodes and demonstrate the [SIR](#) model.

**Usage**

```
u0_SIR()
```

**Details**

A data.frame with the number of individuals in the 'S', 'I' and 'R' compartments in 1600 nodes. Note that the 'I' and 'R' compartments are zero.

**Value**

A data.frame

**Examples**

```
## Create an 'SIR' model with 1600 nodes and initialize
## it to run over 4*365 days. Add one infected individual
## to the first node.
u0 <- u0_SIR()
u0$I[1] <- 1
tspan <- seq(from = 1, to = 4*365, by = 1)
model <- SIR(u0      = u0,
             tspan   = tspan,
             events  = events_SIR(),
             beta    = 0.16,
             gamma   = 0.01)

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)
plot(result)

## Summarize trajectory
summary(result)
```

---

u0\_SISe

*Example data to initialize the 'SISe' model*


---

**Description**

Example data to initialize a population of 1600 nodes and demonstrate the [SISe](#) model.

**Usage**

```
u0_SISe()
```

**Details**

A data.frame with the number of individuals in the 'S' and 'I' compartments in 1600 nodes. Note that the 'I' compartment is zero.

**Value**

A data.frame

**Examples**

```
## Create an 'SISe' model with 1600 nodes and initialize it to
## run over 4*365 days and record data at weekly time-points.

## Load the initial population and add ten infected individuals to
## the first node.
u0 <- u0_SISe()
u0$I[1] <- 10
```

```

## Define 'tspan' to run the simulation over 4*365 and record the
## state of the system at weekly time-points.
tspan <- seq(from = 1, to = 4*365, by = 7)

## Load scheduled events for the population of nodes with births,
## deaths and between-node movements of individuals.
events <- events_SISe()

## Create an 'SISe' model
model <- SISe(u0 = u0, tspan = tspan, events = events_SISe(),
             phi = 0, epsilon = 1.8e-2, gamma = 0.1, alpha = 1,
             beta_t1 = 1.0e-1, beta_t2 = 1.0e-1, beta_t3 = 1.25e-1,
             beta_t4 = 1.25e-1, end_t1 = 91, end_t2 = 182,
             end_t3 = 273, end_t4 = 365, epsilon = 0)

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)

## Summarize trajectory
summary(result)

## Plot the proportion of nodes with at least one infected
## individual.
plot(prevalence(result, I~S+I, "nop"), type = "l")

```

---

u0\_SISe3

*Example data to initialize the 'SISe3' model*


---

## Description

Example data to initialize a population of 1600 nodes and demonstrate the [SISe3](#) model.

## Usage

```
data(u0_SISe3)
```

## Format

A data.frame

## Details

A data.frame with the number of individuals in the 'S\_1', 'S\_2', 'S\_3', 'I\_1', 'I\_2' and 'I\_3' compartments in 1600 nodes. Note that the 'I\_1', 'I\_2' and 'I\_3' compartments are zero.

**Examples**

```

## Create an 'SISe3' model with 1600 nodes and initialize it to
## run over 4*365 days and record data at weekly time-points.

## Load the initial population and add ten infected individuals to
## I_1 in the first node.
u0 <- u0_SISe3
u0$I_1[1] <- 10

## Define 'tspan' to run the simulation over 4*365 and record the
## state of the system at weekly time-points.
tspan <- seq(from = 1, to = 4*365, by = 7)

## Load scheduled events for the population of nodes with births,
## deaths and between-node movements of individuals.
events <- events_SISe3

## Create a 'SISe3' model
model <- SISe3(u0 = u0, tspan = tspan, events = events,
              phi = rep(0, nrow(u0)), epsilon_1 = 1.8e-2,
              epsilon_2 = 1.8e-2, epsilon_3 = 1.8e-2,
              gamma_1 = 0.1, gamma_2 = 0.1, gamma_3 = 0.1,
              alpha = 1, beta_t1 = 1.0e-1, beta_t2 = 1.0e-1,
              beta_t3 = 1.25e-1, beta_t4 = 1.25e-1, end_t1 = 91,
              end_t2 = 182, end_t3 = 273, end_t4 = 365, epsilon = 0)

## Run the model to generate a single stochastic trajectory.
result <- run(model, threads = 1)

## Summarize trajectory
summary(result)

## Plot the proportion of nodes with at least one infected
## individual.
plot(prevalence(result, I_1 + I_2 + I_3 ~ ., "nop"), type = "l")

```

U&lt;-

*Set a template for where to write the U result matrix***Description**

Using a sparse U result matrix can save a lot of memory if the model contains many nodes and time-points, but where only a few of the data points are of interest for post-processing.

**Usage**

```
U(model) <- value
```

## Arguments

model	The model to set a template for the result matrix U.
value	A data.frame that specify the nodes, time-points and compartments to record the number of individuals at tspan. Use NULL to reset the model to record the number of individuals in each compartment in every node at each time-point in tspan.

## Details

Using a sparse U result matrix can save a lot of memory if the model contains many nodes and time-points, but where only a few of the data points are of interest for post-processing. To use this feature, a template has to be defined for which data points to record. This is done using a data.frame that specifies the time-points (column 'time') and nodes (column 'node') to record the state of the compartments, see 'Examples'. The specified time-points, nodes and compartments must exist in the model, or an error is raised. Note that specifying a template only affects which data-points are recorded for post-processing, it does not affect how the solver simulates the trajectory.

## Examples

```
## Create an 'SIR' model with 6 nodes and initialize it to run over 10 days.
u0 <- data.frame(S = 100:105, I = 1:6, R = rep(0, 6))
model <- SIR(u0 = u0, tspan = 1:10, beta = 0.16, gamma = 0.077)

## Run the model.
result <- run(model, threads = 1)

## Display the trajectory with data for every node at each
## time-point in tspan.
trajectory(result)

## Assume we are only interested in nodes '2' and '4' at the
## time-points '3' and '5'
df <- data.frame(time = c(3, 5, 3, 5),
                 node = c(2, 2, 4, 4),
                 S = c(TRUE, TRUE, TRUE, TRUE),
                 I = c(TRUE, TRUE, TRUE, TRUE),
                 R = c(TRUE, TRUE, TRUE, TRUE))
U(model) <- df
result <- run(model, threads = 1)
trajectory(result)

## We can also specify to record only some of the compartments in
## each time-step.
df <- data.frame(time = c(3, 5, 3, 5),
                 node = c(2, 2, 4, 4),
                 S = c(FALSE, TRUE, TRUE, TRUE),
                 I = c(TRUE, FALSE, TRUE, FALSE),
                 R = c(TRUE, FALSE, TRUE, TRUE))
U(model) <- df
result <- run(model, threads = 1)
trajectory(result)
```

```
## It is possible to use an empty 'data.frame' to specify
## that no data-points should be recorded for the trajectory.
U(model) <- data.frame()
result <- run(model, threads = 1)
trajectory(result)

## Use 'NULL' to reset the model to record data for every node at
## each time-point in tspan.
U(model) <- NULL
result <- run(model, threads = 1)
trajectory(result)
```

---

V<-

*Set a template for where to write the V result matrix*

---

## Description

Using a sparse V result matrix can save a lot of memory if the model contains many nodes and time-points, but where only a few of the data points are of interest for post-processing.

## Usage

```
V(model) <- value
```

## Arguments

model	The model to set a template for the result matrix V.
value	A data.frame that specify the nodes, time-points and compartments of when to record the real-valued continuous state at tspan. Use NULL to reset the model to record the real-valued continuous state in every node at each time-point in tspan.

## Details

Using a sparse V result matrix can save a lot of memory if the model contains many nodes and time-points, but where only a few of the data points are of interest for post-processing. To use this feature, a template has to be defined for which data points to record. This is done using a data.frame that specifies the time-points (column 'time') and nodes (column 'node') to record the state of the continuous state compartments, see 'Examples'. The specified time-points, nodes and compartments must exist in the model, or an error is raised. Note that specifying a template only affects which data-points are recorded for post-processing, it does not affect how the solver simulates the trajectory.

## Examples

```

## Create an 'SISe' model with 6 nodes and initialize
## it to run over 10 days.
u0 <- data.frame(S = 100:105, I = 1:6)
model <- SISe(u0 = u0, tspan = 1:10, phi = rep(0, 6),
  epsilon = 0.02, gamma = 0.1, alpha = 1, epsilon = 1.1e-5,
  beta_t1 = 0.15, beta_t2 = 0.15, beta_t3 = 0.15, beta_t4 = 0.15,
  end_t1 = 91, end_t2 = 182, end_t3 = 273, end_t4 = 365)

## Run the model
result <- run(model, threads = 1)

## Display the continuous state variable 'phi' for every node at
## each time-point in tspan.
trajectory(result, compartments = "phi")

## Assume we are only interested in nodes '2' and '4' at the
## time-points '3' and '5'
df <- data.frame(time = c(3, 5, 3, 5),
  node = c(2, 2, 4, 4),
  phi = c(TRUE, TRUE, TRUE, TRUE))
V(model) <- df
result <- run(model, threads = 1)
trajectory(result, compartments = "phi")

## It is possible to use an empty 'data.frame' to specify
## that no data-points should be recorded for the trajectory.
V(model) <- data.frame()
result <- run(model, threads = 1)
trajectory(result, compartments = "phi")

## Use 'NULL' to reset the model to record data for every node at
## each time-point in tspan.
V(model) <- NULL
result <- run(model, threads = 1)
trajectory(result, compartments = "phi")

```



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