

Package ‘fastbeta’

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Title Fast Approximation of Time-Varying Infectious Disease
Transmission Rates

Description A fast method for approximating time-varying infectious disease
transmission rates from disease incidence time series and other
data, based on a discrete time approximation of an SEIR model,
as analyzed in Jagan et al. (2020) <[doi:10.1371/journal.pcbi.1008124](https://doi.org/10.1371/journal.pcbi.1008124)>.

License GPL (>= 2)

URL <https://github.com/davidearn/fastbeta>

BugReports <https://github.com/davidearn/fastbeta/issues>

Depends R (>= 4.3.0)

Imports grDevices, graphics, stats

Suggests adaptivetau, deSolve, tools, utils

BuildResaveData no

NeedsCompilation yes

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deconvolve

*Richardson-Lucy Deconvolution***Description**

Performs a modified Richardson-Lucy iteration for the purpose of estimating incidence from reported incidence or mortality, conditional on a reporting probability and on a distribution of the time to reporting.

Usage

```
deconvolve(x, prob = 1, delay = 1,
           start, tol = 1, iter.max = 32L, complete = FALSE)
```

Arguments

<code>x</code>	a numeric vector of length <code>n</code> giving the number of infections or deaths reported during <code>n</code> observation intervals of equal duration.
<code>prob</code>	a numeric vector of length <code>d+n</code> such that <code>prob[d+i]</code> is the probability that an infection during interval <code>i</code> is eventually reported. <code>prob</code> of length 1 is recycled.
<code>delay</code>	a numeric vector of length <code>d+1</code> such that <code>delay[j]</code> is the probability that an infection during interval <code>i</code> is reported during interval <code>i+j-1</code> , given that it is eventually reported. <code>delay</code> need not sum to 1 but must not sum to 0.
<code>start</code>	a numeric vector of length <code>d+n</code> giving a starting value for the iteration. <code>start[d+i]</code> estimates the expected number of infections during interval <code>i</code> that are eventually reported. If missing, then a starting value is generated by padding <code>x</code> on the left and right with <code>d-d0</code> and <code>d0</code> zeros, choosing <code>d0 = which.max(delay)-1</code> .
<code>tol</code>	a tolerance indicating a stopping condition; see the reference.
<code>iter.max</code>	the maximum number of iterations.
<code>complete</code>	a logical flag indicating if the result should preserve successive updates to <code>start</code> .

Value

A list with elements:

<code>value</code>	the result of updating <code>start</code> <code>iter</code> times then dividing by <code>prob</code> . If <code>complete = TRUE</code> , then <code>value</code> is a <code>(d+n)</code> -by- <code>(1+iter)</code> matrix containing <code>start</code> and the <code>iter</code> successive updates, each divided by <code>prob</code> .
<code>chisq</code>	the chi-squared statistics corresponding to <code>value</code> .
<code>iter</code>	the number of iterations performed.

References

Goldstein, E., Dushoff, J., Ma, J., Plotkin, J. B., Earn, D. J. D., & Lipsitch, M. (2020). Reconstructing influenza incidence by deconvolution of daily mortality time series. *Proceedings of the National Academy of Sciences U. S. A.*, *106*(51), 21825-21829. doi:10.1073/pnas.0902958106

Examples

```

set.seed(2L)
n <- 200L
d <- 50L
p <- 0.1
prob <- plogis(rlogis(d + n, location = qlogis(p), scale = 0.1))
delay <- diff(pgamma(0L:(d + 1L), 12, 0.4))

h <- function (x, a = 1, b = 1, c = 0) a * exp(-b * (x - c)^2)
ans <- floor(h(seq(-60, 60, length.out = d + n), a = 1000, b = 0.001))

x0 <- rbinom(d + n, ans, prob)
x <- tabulate(rep.int(1L:(d + n), x0) +
             sample(0L:d, size = sum(x0), replace = TRUE, prob = delay),
             d + n)[-1L:d]

str(D0 <- deconvolve(x, prob, delay, complete = FALSE))
str(D1 <- deconvolve(x, prob, delay, complete = TRUE))

matplot(-(d - 1L):n,
        cbind(x0, c(rep.int(NA, d), x), prob * D0[["value"]], p * ans),
        type = c("p", "p", "p", "l"),
        col = c(1L, 1L, 2L, 4L), pch = c(16L, 1L, 16L, NA),
        lty = c(0L, 0L, 0L, 1L), lwd = c(NA, NA, NA, 3L),
        xlab = "time", ylab = "count")
legend("topleft", NULL,
      c("actual", "actual+delay", "actual+delay+deconvolution", "p*h"),
      col = c(1L, 1L, 2L, 4L), pch = c(16L, 1L, 16L, NA),
      lty = c(0L, 0L, 0L, 1L), lwd = c(NA, NA, NA, 3L),
      bty = "n")

plot(0L:D1[["iter"]], D1[["chisq"]], xlab = "iterations", ylab = quote(chi^2))
abline(h = 1, lty = 2L)

```

fastbeta

*Estimate a Time-Varying Infectious Disease Transmission Rate***Description**

Generates a discrete approximation of a time-varying infectious disease transmission rate from an equally spaced incidence time series and other data.

Usage

```

fastbeta(series, sigma = gamma, gamma = 1, delta = 0,
         init, m = length(init) - n - 2L, n = 1L, ...)

```

Arguments

series	a “multiple time series” object, inheriting from class <code>mts</code> , with three columns storing (“parallel”, equally spaced) time series of incidence, births, and the per capita natural mortality rate, in that order.
sigma, gamma, delta	non-negative numbers. <code>m*sigma</code> , <code>n*gamma</code> , and <code>delta</code> are the rates of removal from each latent, infectious, and recovered compartment.
init	a numeric vector of length $1+m+n+1$ giving an initial state with compartments ordered as (S, E, I, R) .
m	a non-negative integer indicating a number of latent stages.
n	a positive integer indicating a number of infectious stages.
...	optional arguments passed to <code>deconvolve</code> , if the first column of series represents <i>observed</i> incidence rather than actual or estimated incidence.

Details

The algorithm implemented by `fastbeta` is based on an SEIR model with

- m latent stages ($E^i, i = 1, \dots, m$);
- n infectious stages ($I^j, j = 1, \dots, n$);
- time-varying rates β, ν , and μ of transmission, birth, and natural death; and
- constant rates $m\sigma, n\gamma$, and δ of removal from each latent, infectious, and recovered compartment, where removal from the recovered compartment implies return to the susceptible compartment (loss of immunity).

It is derived by linearizing of the system of ordinary differential equations

$$\begin{aligned}
 dS/dt &= \delta R - (\lambda(t) + \mu(t))S + \nu(t) \\
 dE^1/dt &= \lambda(t)S - (m\sigma + \mu(t))E^1 \\
 dE^{i+1}/dt &= m\sigma E^i - (m\sigma + \mu(t))E^{i+1} \\
 dI^1/dt &= m\sigma E^m - (n\gamma + \mu(t))I^1 \\
 dI^{j+1}/dt &= n\gamma I^j - (n\gamma + \mu(t))I^{j+1} \\
 dR/dt &= n\gamma I^n - (\delta + \mu(t))R
 \end{aligned}
 \quad \lambda(t) = \beta(t) \sum_j I^j$$

and substituting actual or estimated incidence and births for definite integrals of λS and ν . This procedure yields a system of linear difference equations from which one recovers a discrete approximation of β :

$$\begin{aligned}
 E_{t+1}^1 &= [(1 - \frac{1}{2}(m\sigma + \mu_t))E_t^1 + Z_{t+1}] / [1 + \frac{1}{2}(m\sigma + \mu_{t+1})] \\
 E_{t+1}^{i+1} &= [(1 - \frac{1}{2}(m\sigma + \mu_t))E_t^{i+1} + \frac{1}{2}m\sigma(E_t^i + E_{t+1}^i)] / [1 + \frac{1}{2}(m\sigma + \mu_{t+1})] \\
 I_{t+1}^1 &= [(1 - \frac{1}{2}(n\gamma + \mu_t))I_t^1 + \frac{1}{2}m\sigma(E_t^m + E_{t+1}^m)] / [1 + \frac{1}{2}(n\gamma + \mu_{t+1})] \\
 I_{t+1}^{j+1} &= [(1 - \frac{1}{2}(n\gamma + \mu_t))I_t^{j+1} + \frac{1}{2}n\gamma(I_t^j + I_{t+1}^j)] / [1 + \frac{1}{2}(n\gamma + \mu_{t+1})] \\
 R_{t+1} &= [(1 - \frac{1}{2}(\delta + \mu_t))R_t + \frac{1}{2}n\gamma(I_t^n + I_{t+1}^n)] / [1 + \frac{1}{2}(\delta + \mu_{t+1})] \\
 S_{t+1} &= [(1 - \frac{1}{2}(\mu_t))S_t + \frac{1}{2}(\delta(R_t + R_{t+1}) - Z_{t+1} + B_{t+1})] / [1 + \frac{1}{2}(\mu_{t+1})]
 \end{aligned}
 \quad \beta_t = (Z_t + Z_{t+1}) / (2S_t \sum_j I_t^j)$$

where we use the notation

$$X_t \sim X(t) : X = S, E^i, I^j, R, Z, B, \mu, \beta$$

$$Z(t) = \int_{t-1}^t \lambda(s) S(s) ds$$

$$B(t) = \int_{t-1}^t \nu(s) ds$$

and it is understood that the independent variable t is a unitless measure of time relative to the spacing of the substituted time series of incidence and births.

Value

A “multiple time series” object, inheriting from class `mts`, with $1+m+n+1+1$ columns (named S, E, I, R, and beta) storing the result of the iteration described in ‘Details’. It is completely parallel to argument series, having the same `tsp` attribute.

References

Jagan, M., deJonge, M. S., Krylova, O., & Earn, D. J. D. (2020). Fast estimation of time-varying infectious disease transmission rates. *PLOS Computational Biology*, 16(9), Article e1008124, 1-39. [doi:10.1371/journal.pcbi.1008124](https://doi.org/10.1371/journal.pcbi.1008124)

Examples

```
if (requireNamespace("adaptivetau")) withAutoprint({

data(seir.ts02, package = "fastbeta")
a <- attributes(seir.ts02)
str(seir.ts02)
plot(seir.ts02)

## We suppose that we have perfect knowledge of incidence,
## births, and the data-generating parameters
series <- cbind(seir.ts02[, c("Z", "B")], mu = a[["mu"]](0))
colnames(series) <- c("Z", "B", "mu") # FIXME: stats::cbind.ts mangles dimnames

args <- c(list(series = series),
          a[c("sigma", "gamma", "delta", "init", "m", "n")])
str(args)

X <- do.call(fastbeta, args)
str(X)
plot(X)

plot(X[, "beta"], ylab = "transmission rate")
lines(a[["beta"]](time(X)), col = "red") # the "truth"

})
```

fastbeta.bootstrap *Parametric Bootstrapping*

Description

A simple wrapper around [fastbeta](#) using it to generate a “primary” estimate of a time-varying transmission rate and r bootstrap estimates. Bootstrap estimates are computed for incidence time series simulated using [seir](#), with transmission rate defined as the linear interpolant of the primary estimate.

Usage

```
fastbeta.bootstrap(r,
                  series, sigma = gamma, gamma = 1, delta = 0,
                  init, m = length(init) - n - 2L, n = 1L, ...)
```

Arguments

<code>r</code>	a non-negative integer indicating a number of replications.
<code>series</code>	a “multiple time series” object, inheriting from class <code>mts</code> , with three columns storing (“parallel”, equally spaced) time series of incidence, births, and the per capita natural mortality rate, in that order.
<code>sigma, gamma, delta</code>	non-negative numbers. $m \times \text{sigma}$, $n \times \text{gamma}$, and <code>delta</code> are the rates of removal from each latent, infectious, and recovered compartment.
<code>init</code>	a numeric vector of length $1+m+n+1$ giving an initial state with compartments ordered as (S, E, I, R) .
<code>m</code>	a non-negative integer indicating a number of latent stages.
<code>n</code>	a positive integer indicating a number of infectious stages.
<code>...</code>	optional arguments passed to seir and/or deconvolve . Both take optional arguments <code>prob</code> and <code>delay</code> . When <code>prob</code> is supplied but not <code>delay</code> , seir and deconvolve receive <code>prob</code> as is. When both are supplied, seir receives <code>prob</code> as is, whereas deconvolve receives <code>prob</code> augmented with $\text{length}(\text{delay})-1$ ones.

Value

A “multiple time series” object, inheriting from class `mts`, with $1+r$ columns storing the one primary and r bootstrap estimates. It is completely parallel to argument `series`, having the same `tsp` attribute.

Examples

```

if (requireNamespace("adaptivetau")) withAutoprint({

  data(seir.ts02, package = "fastbeta")
  a <- attributes(seir.ts02)
  str(seir.ts02)
  plot(seir.ts02)

  ## We suppose that we have perfect knowledge of incidence,
  ## births, and the data-generating parameters
  series <- cbind(seir.ts02[, c("Z", "B")], mu = a[["mu"]](0))
  colnames(series) <- c("Z", "B", "mu") # FIXME: stats::cbind.ts mangles dimnames

  args <- c(list(r = 100L, series = series),
            a[c("sigma", "gamma", "delta", "init", "m", "n")])
  str(args)

  R <- do.call(fastbeta.bootstrap, args)
  str(R)
  plot(R)
  plot(R, level = 0.95)

})

```

ptpi

Peak to Peak Iteration

Description

Approximates the state of an SEIR model at a reference time from an equally spaced, T -periodic incidence time series and other data. The algorithm relies on a strong assumption: that the incidence time series was generated by the asymptotic dynamics of an SEIR model admitting a locally stable, T -periodic attractor. Hence do interpret with care.

Usage

```

ptpi(series, sigma = gamma, gamma = 1, delta = 0,
      init, m = length(init) - n - 2L, n = 1L,
      start = tsp(series)[1L], end = tsp(series)[2L],
      tol = 1e-03, iter.max = 32L,
      backcalc = FALSE, complete = FALSE, ...)

```

Arguments

series a “multiple time series” object, inheriting from class `mts`, with three columns storing (“parallel”, equally spaced) time series of incidence, births, and the per capita natural mortality rate, in that order.

<code>sigma, gamma, delta</code>	non-negative numbers. $m \times \text{sigma}$, $n \times \text{gamma}$, and <code>delta</code> are the rates of removal from each latent, infectious, and recovered compartment.
<code>init</code>	a numeric vector of length $1+m+n+1$ giving an initial guess for the state at time start.
<code>m</code>	a non-negative integer indicating a number of latent stages.
<code>n</code>	a positive integer indicating a number of infectious stages.
<code>start, end</code>	start and end times for the iteration, whose difference should be approximately equal to an integer number of periods. One often chooses the time of the first peak in the incidence time series and the time of the last peak in phase with the first.
<code>tol</code>	a tolerance indicating a stopping condition; see ‘Details’.
<code>iter.max</code>	the maximum number of iterations.
<code>backcalc</code>	a logical indicating if the state at time <code>tsp(series)[1]</code> should be back-calculated from the state at time <code>start</code> if that is later.
<code>complete</code>	a logical indicating if intermediate states should be recorded in an array. Useful mainly for didactic or diagnostic purposes.
<code>...</code>	optional arguments passed to <code>deconvolve</code> , if the first column of <code>series</code> represents <i>observed</i> incidence rather than actual or estimated incidence.

Details

ptpi can be understood as an iterative application of `fastbeta` to a subset of series. The basic algorithm can be expressed in R code as:

```
w <- window(series, start, end); i <- nrow(s); j <- seq_along(init)
diff <- Inf; iter <- 0L
while (diff > tol && iter < iter.max) {
  init. <- init
  init <- fastbeta(w, sigma, gamma, delta, init, m, n)[i, j]
  diff <- sqrt(sum((init - init.)^2) / sum(init.^2))
  iter <- iter + 1L
}
value <- init
```

Back-calculation involves solving a linear system of equations; the back-calculated result can mislead if the system is ill-conditioned.

Value

A list with elements:

<code>value</code>	an approximation of the state at time <code>start</code> or at time <code>tsp(series)[1L]</code> , depending on <code>backcalc</code> .
<code>diff</code>	the relative difference between the last two approximations.
<code>iter</code>	the number of iterations performed.

x if complete = TRUE, then a “multiple time series” object, inheriting from class `mts`, with dimensions `c(nrow(w), length(value), iter)`, where `w = window(series, start, end)`. `x[, , k]` contains the state at each time(`w`) in iteration `k`.

References

Jagan, M., deJonge, M. S., Krylova, O., & Earn, D. J. D. (2020). Fast estimation of time-varying infectious disease transmission rates. *PLOS Computational Biology*, *16*(9), Article e1008124, 1-39. [doi:10.1371/journal.pcbi.1008124](https://doi.org/10.1371/journal.pcbi.1008124)

Examples

```
if (requireNamespace("deSolve")) withAutoprint({

  data(seir.ts01, package = "fastbeta")
  a <- attributes(seir.ts01); p <- length(a[["init"]])
  str(seir.ts01)
  plot(seir.ts01)

  ## We suppose that we have perfect knowledge of incidence,
  ## births, and the data-generating parameters, except for
  ## the initial state, which we "guess"
  series <- cbind(seir.ts01[, c("Z", "B")], mu = a[["mu"]](0))
  colnames(series) <- c("Z", "B", "mu") # FIXME: stats::cbind.ts mangles dimnames

  plot(series[, "Z"])
  start <- 23; end <- 231
  abline(v = c(start, end), lty = 2)

  set.seed(0L)
  args <- c(list(series = series),
           a[c("sigma", "gamma", "delta", "init", "m", "n")],
           list(start = start, end = end, complete = TRUE))
  init <- seir.ts01[which.min(abs(time(seir.ts01) - start)), seq_len(p)]
  args[["init"]] <- init * rlnorm(p, 0, 0.1)
  str(args)

  L <- do.call(ptpi, args)
  str(L)

  S <- L[["x"]][, "S", ]
  plot(S, plot.type = "single")
  lines(seir.ts01[, "S"], col = "red", lwd = 4) # the "truth"
  abline(h = L[["value"]][["S"], v = start, col = "blue", lwd = 4, lty = 2)

  ## Relative error
  L[["value"]] / init - 1

})
```

seir

*Simulate Infectious Disease Time Series***Description**

Simulates incidence time series based on an SEIR model with user-defined forcing and a simple model for observation error.

Note that simulation code depends on availability of suggested packages **adaptivetau** and **deSolve**. If the dependency cannot be loaded then an error is signaled.

Usage

```
seir(length.out = 1L,
      beta, nu, mu, sigma = gamma, gamma = 1, delta = 0,
      init, m = length(init) - n - 2L, n = 1L,
      stochastic = TRUE, prob = 1, delay = 1,
      useCompiled = TRUE, ...)
```

A basic wrapper for the 'm = 0L' case:

```
sir(length.out = 1L,
     beta, nu, mu, gamma = 1, delta = 0,
     init, n = 1L,
     stochastic = TRUE, prob = 1, delay = 1,
     useCompiled = TRUE, ...)
```

Arguments

length.out	a non-negative integer indicating the time series length.
beta, nu, mu	functions of one or more arguments returning transmission, birth, and natural death rates at the time point indicated by the first argument. Arguments after the first must be strictly optional. The functions need not be vectorized.
sigma, gamma, delta	non-negative numbers. $m \times \text{sigma}$, $n \times \text{gamma}$, and delta are the rates of removal from each latent, infectious, and recovered compartment.
init	a numeric vector of length $1+m+n+1$ giving an initial state with compartments ordered as (S, E, I, R) .
m	a non-negative integer indicating a number of latent stages.
n	a positive integer indicating a number of infectious stages.
stochastic	a logical indicating if the simulation should be stochastic; see 'Details'.
prob	a numeric vector of length n such that $\text{prob}[i]$ is the probability that an infection during interval i is eventually observed. prob of length 1 is recycled.
delay	a numeric vector of positive length such that $\text{delay}[i]$ is the probability that an infection during interval j is observed during interval $j+i-1$, given that it is eventually observed. delay need not sum to 1 but must not sum to 0.

useCompiled a logical indicating if derivatives should be computed by compiled C functions rather than by R functions (which *may* be *byte-compiled*). Set to FALSE only if TRUE seems to cause problems, and in that case please report the problems with [bug.report](#)(package = "fastbeta").

... optional arguments passed to [lsoda](#) (directly) or [ssa.adaptivetau](#) (via its list argument `tl.params`), depending on `stochastic`.

Details

Simulations are based on an SEIR model with

- m latent stages ($E^i, i = 1, \dots, m$);
- n infectious stages ($I^j, j = 1, \dots, n$);
- time-varying rates β, ν , and μ of transmission, birth, and natural death; and
- constant rates $m\sigma, n\gamma$, and δ of removal from each latent, infectious, and recovered compartment, where removal from the recovered compartment implies return to the susceptible compartment (loss of immunity).

`seir(stochastic = FALSE)` works by numerically integrating the system of ordinary differential equations

$$\begin{aligned} dS/dt &= \delta R - (\lambda(t) + \mu(t))S + \nu(t) \\ dE^1/dt &= \lambda(t)S - (m\sigma + \mu(t))E^1 \\ dE^{i+1}/dt &= m\sigma E^i - (m\sigma + \mu(t))E^{i+1} \\ dI^1/dt &= m\sigma E^m - (n\gamma + \mu(t))I^1 \\ dI^{j+1}/dt &= n\gamma I^j - (n\gamma + \mu(t))I^{j+1} \\ dR/dt &= n\gamma I^n - (\delta + \mu(t))R \end{aligned} \quad \lambda(t) = \beta(t) \sum_j I^j$$

where it is understood that the independent variable t is a unitless measure of time relative to an observation interval. To get time series of incidence and births, the system is augmented with two equations describing *cumulative* incidence and births

$$dZ/dt = \lambda(t)S \quad dB/dt = \nu(t)$$

and the *augmented* system is numerically integrated. Observed incidence is simulated from incidence by scaling the latter by `prob` and convolving the result with `delay`.

`seir(stochastic = TRUE)` works by simulating a Markov process corresponding to the augmented system, as described in the reference. Observed incidence is simulated from incidence by binning binomial samples taken with probabilities `prob` over future observation intervals according to multinomial samples taken with probabilities `delay`.

Value

A “multiple time series” object, inheriting from class `mts`. Beneath the class, it is a `length.out-by-1+m+n+1+2` numeric matrix with columns S, E, I, R, Z, and B, where Z and B specify incidence and births as the number of infections and births since the previous time point.

If `prob` or `delay` is not missing, then there is an additional column `Z.obs` specifying *observed* incidence as the number of infections observed since the previous time point. The first `length(delay)` elements of this column contain partial counts.

References

Cao, Y., Gillespie, D. T., & Petzold, L. R. (2007). Adaptive explicit-implicit tau-leaping method with automatic tau selection. *Journal of Chemical Physics*, 126(22), Article 224101, 1-9. doi:10.1063/1.2745299

See Also

[seir.library.](#)

Examples

```

if (requireNamespace("adaptivetau")) withAutoprint({

beta <- function (t, a = 1e-01, b = 1e-05) b * (1 + a * sinpi(t / 26))
nu   <- function (t) 1e+03
mu   <- function (t) 1e-03

sigma <- 0.5
gamma <- 0.5
delta <- 0

init <- c(S = 50200, E = 1895, I = 1892, R = 946011)

length.out <- 250L
prob <- 0.1
delay <- diff(pgamma(0:8, 2.5))

set.seed(0L)
X <- seir(length.out, beta, nu, mu, sigma, gamma, delta, init,
          prob = prob, delay = delay, epsilon = 0.002)
##
## default epsilon = 0.05 allows too big leaps => spurious noise
##
str(X)
plot(X)

r <- 10L
Y <- do.call(cbind, replicate(r, simplify = FALSE,
  seir(length.out, beta, nu, mu, sigma, gamma, delta, init,
    prob = prob, delay = delay, epsilon = 0.002)[, "Z.obs"]))
str(Y) # FIXME: stats::cbind.ts mangles dimnames
plot(window(Y, start = tsp(Y)[1L] + length(delay) / tsp(Y)[3L]),
      ##
      ## discards points showing edge effects due to 'delay'
      ##
      plot.type = "single", col = seq_len(r), ylab = "case reports")

})

```

Description

Infectious disease time series simulated using [seir](#), for use primarily in examples, tests, and vignettes. Users should not rely on simulation details, which may change between package versions.

Note that simulation code depends on availability of suggested packages **adaptivetau** and **deSolve**. If the dependency cannot be loaded then the value of the data set is NULL.

Usage

```
## if (requireNamespace("deSolve"))
data(seir.ts01, package = "fastbeta")
## else ...

## if (requireNamespace("adaptivetau"))
data(seir.ts02, package = "fastbeta")
## else ...
```

Format

A “multiple time series” object, inheriting from class [mts](#), always a subset of the result of a call to [seir](#), discarding transient behaviour. Simulation parameters may be preserved as attributes.

Source

Scripts sourced by [data](#) to reproduce the simulations are located in subdirectory ‘data’ of the **fastbeta** installation; see, e.g. `system.file("data", "seir.ts01.R", package = "fastbeta")`.

See Also

[seir](#).

Examples

```
if (requireNamespace("deSolve")) withAutoprint({

data(seir.ts01, package = "fastbeta")
str(seir.ts01)
plot(seir.ts01)

})

if (requireNamespace("adaptivetau")) withAutoprint({

data(seir.ts02, package = "fastbeta")
```

```
str(seir.ts02)
plot(seir.ts02)

})
```

smallpox

Smallpox Mortality in London, England, 1661-1930

Description

Time series of deaths due to smallpox, deaths due to all causes, and births in London, England, from 1661 to 1930, as recorded in the London Bills of Mortality and the Registrar General's Weekly Returns.

Usage

```
data(smallpox, package = "fastbeta")
```

Format

A data frame with 13923 observations of 5 variables:

from start date of the record.

nday length of the record, which is the number of days (typically 7) over which deaths and births were counted.

smallpox count of deaths due to smallpox.

allcauses count of deaths due to all causes.

births count of births.

Source

A precise description of the data set and its correspondence to the original source documents is provided in the reference.

A script generating the smallpox data frame from a CSV file accompanying the reference is available as `system.file("scripts", "smallpox.R", package = "fastbeta")`.

References

Krylova, O. & Earn, D. J. D. (2020). Patterns of smallpox mortality in London, England, over three centuries. *PLOS Biology*, 18(12), Article e3000506, 1-27. [doi:10.1371/journal.pbio.3000506](https://doi.org/10.1371/journal.pbio.3000506)

Examples

```
data(smallpox, package = "fastbeta")
str(smallpox)
table(smallpox[["nday"]]) # not all 7 days, hence:
plot(7 * smallpox / as.double(nday) ~ from, smallpox, type = "l")
```

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