Package 'EpiILM'

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Title Spatial and Network Based Individual Level Models for Epidemics

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Imports methods

Author Vineetha Warriyar. K. V., Waleed Almutiry, and Rob Deardon

Maintainer Waleed Almutiry <wkmtierie@qu.edu.sa>

Description Provides tools for simulating from discrete-time individual level models for infectious disease data analysis. This epidemic model class contains spatial and contactnetwork based models with two disease types: Susceptible-Infectious (SI) and Susceptible-Infectious-Removed (SIR).

License GPL $(>= 2)$

URL <https://github.com/vineetha-warriyar/EpiILM>

NeedsCompilation yes

Repository CRAN

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R topics documented:

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EpiILM-package EpiILM*: Spatial and Network Based Individual Level Models for Epidemics*

Description

The R package **EpiILM** is provided for simulating from, and carrying out Bayesian MCMC-based statistical inference for spatial and/or network-based individual-level modelling framework. The package allows for the incorporation of individual-level susceptibility and transmissibility covariates in models, and provides various methods of summarizing epidemic data sets.

Details

The R package EpiILM can be used to carry out simulation of epidemics, estimate the basic reproduction number, plot various epidemic summary graphics, calculate the log-likelihood, carry out Bayesian inference using Metropolis-Hastings MCMC, and implement posterior predictive checks and model selection for a given data set and model. The key functions for this package are detailed in the value section. One of the important functions [epimcmc](#page-11-1) depends heavily on the [MCMC](#page-0-0) from the adaptMCMC package for performing the MCMC analysis. This function implements the robust adaptive Metropolis sampler of Vihola (2012) for tuning the covariance matrix of the (normal) jump distribution adaptively to achieve the desired acceptance rate. The package has other features for making predictions or forecasting for a specific model via the pred.epi function. The main functions, including for epidemic simulation ([epidata](#page-5-1)) and likelihood calculation ([epilike](#page-9-1)) are coded in Fortran in order to achieve the goal of agile implementation.

Value

Key functions for this package:

Author(s)

Vineetha Warriyar. K. V., Waleed Almutiry, and Rob Deardon Maintainer: Waleed Almutiry <wkmtierie@qu.edu.sa>

References

Deardon, R., Brooks, S. P., Grenfell, B. T., Keeling, M. J., Tildesley, M. J., Savill, N. J., Shaw, D. J., and Woolhouse, M. E. (2010). Inference for individual level models of infectious diseases in large populations. *Statistica Sinica*, 20, 239-261.

Vihola, M. (2012) Robust adaptive Metropolis algorithm with coerced acceptance rate. *Statistics and Computing*, 22(5), 997-1008. doi:10.1007/s11222-011-9269-5.

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Examples

```
## Not run:
demo(EpiILM.spatial)
demo(EpiILM.network)
```
End(Not run)

as.epidata *Discrete Time level information of an Epidemic*

Description

This function allows the user to generate objects of class "epidata". The output of this function provides information required to be used in the other functions in the package.

Usage

as.epidata (type, $n, x = NULL, y = NULL, infinite, infperiod = NULL, contact = NULL)$

Arguments

Value

An object of class epidata is returned containing the following:

- type Type of compartment framework, with the choice of "SI" for Susceptible-Infectious diseases and "SIR" for Susceptible-Infectious-Removed.
- XYcoordinates The XY-coordinates of individuals if distance-based ILM is used, otherwise is NULL.

contact Contact network matrix if network-based ILM is used, otherwise is NULL.

inftime The infection times of individuals.

remtime The removal times of individuals when type = "SIR".

See Also

[epidata](#page-5-1), [plot.epidata](#page-14-1).

Examples

```
# generate 100 X-Y coordinates for a distance-based ILM
```

```
x <- runif(100, 0, 10)
y <- runif(100, 0, 10)
```
suppose we know the infection times for a spatial SI model based on above X-Y coordinates

```
inftime \leq-rpois(100, 8)
```
Now we can convert above information to an epidata object with

```
out \leq as.epidata(type = "SI", n = 100, x = x, y = y, inftime = inftime)
out
```
epiBR0 *Basic reproduction number (R0)*

Description

Gives a Monte Carlo estimate of the basic reproduction number for a specified SIR model and data set

Usage

```
epiBR0 (x = NULL, y = NULL, contact = NULL, sus.par, trans.par = NULL, beta,
        spark = NULL, infperiod, Sformula = NULL, Tformula = NULL, tmax,
```
niter)

Arguments

epiBR0 \sim 5

Value

A list is returned with the following components:

Examples

generate 100 X-Y coordinates for a distance-based ILM

```
x <- runif(100, 0, 10)
```

```
y <- runif(100, 0, 10)
```
Suppose we know the length of infectious period for each individual. Also, assume # susceptibility parameter = 1.5 and spatial parameter = 5 for this SIR model

```
infperiod \leq rep(3, 100)
```
For a 1000 iteration with a last observed time point 15, we can estimate the basic # reproduction number using Monte Carlo simulation

out \leq epiBR0(x = x, y = y, sus.par = 1.5, beta = 5, infperiod= infperiod,

tmax = 15, niter = 1000)

out\$BasicR0

Description

This function allows the user to simulate epidemics under different models and scenarios

Usage

```
epidata (type, n, tmin = NULL, tmax, sus.par, trans.par = NULL, beta = NULL, spark = NULL,
          Sformula = NULL, Tformula = NULL, x = NULL, y = NULL,
          inftime = NULL, infperiod = NULL, contact = NULL)
```
Arguments

Details

We consider following two individual level models:

Spatial model:

$$
P(i,t) = 1 - \exp\{-\Omega_S(i)\sum_{j\in I(t)} \Omega_T(j)d_{ij}^{-\beta} - \varepsilon\}
$$

Network model:

$$
P(i,t) = 1 - \exp\{-\Omega_S(i) \sum_{j \in I(t)} \Omega_T(j) (\beta_1 C_{ij}^{(1)} + \dots + \beta_n C_{ij}^{(n)}) - \varepsilon\}
$$

where $P(i, t)$ is the probability that susceptible individual i is infected at time point t, becoming infectious at time t+1; $\Omega_S(i)$ is a susceptibility function which accommodates potential risk factors associated with susceptible individual i contracting the disease; $\Omega_T(j)$ is a transmissibility function which accommodates potential risk factors associated with infectious individual j; ε is a sparks term which represents infections originating from outside the population being observed or some other unobserved infection mechanism.

The susceptibility function can incorporate any individual-level covariates of interest and $\Omega_S(i)$ is treated as a linear function of the covariates, i.e., $\Omega_S(i) = \alpha_0 + \alpha_1 X_1(i) + \alpha_2 X_2(i) + \ldots$ $\alpha_{n_s}X_{n_s}(i)$, where $X_1(i),...,X_{n_s}(i)$ denote n_s covariates associated with susceptible individual \$i\$, along with susceptibility parameters $\alpha_0, \ldots, \alpha_{n_s} > 0$. If the model does not contain any susceptibility covariates then $\Omega_S(i) = \alpha_0$ is used. In a similar way, the transmissibility function can incorporate any individual-level covariates of interest associated with infectious individual. $\Omega_T(j)$ is also treated as a linear function of the covariates, but without the intercept term, i.e., $\Omega_T(j) = \phi_1 X_1(j) + \phi_2 X_2(j) + \ldots + \phi_{n_t} X_{n_t}(j)$, where $X_1(j), \ldots, X_{n_t}(j)$ denote the n_t covariates associated with infectious individual j, along with transmissibility parameters $\phi_1, \dots, \phi_{n_t} > 0$. If the model does not contain any transmissibility covariates then $\Omega_T(j) = 1$ is used.

Value

An object of class epidata is returned containing the following:

type Type of compartment framework, with the choice of "SI" for Susceptible-Infectious diseases and "SIR" for Susceptible-Infectious-Removed

XYcoordinates The XY-coordinates of individuals.

contact Contact network matrix.

inftime The infection times of individuals.

remtime The removal times of individuals when type = "SIR".

References

Deardon, R., Brooks, S. P., Grenfell, B. T., Keeling, M. J., Tildesley, M. J., Savill, N. J., Shaw, D. J., and Woolhouse, M. E. (2010). Inference for individual level models of infectious diseases in large populations. *Statistica Sinica*, 20, 239-261.

Deardon, R., Fang, X., and Kwong, G.P.S. (2014). Statistical modelling of spatio-temporal infectious disease transmission in analyzing and modeling Spatial and temporal dynamics of infectious diseases, *(Ed: D. Chen, B. Moulin, J. Wu), John Wiley & Sons.* Chapter 11.

See Also

[plot.epidata](#page-14-1), [epimcmc](#page-11-1), [epilike](#page-9-1), [pred.epi](#page-18-1).

```
## Example 1: spatial SI model
# generate 100 individuals
x <- runif(100, 0, 10)
y <- runif(100, 0, 10)
covariate \le runif(100, 0, 2)
out1 \leq epidata(type = "SI", n = 100, Sformula = \simcovariate, tmax = 15,
               sus.par = c(0.1, 0.3), beta = 5.0, x = x, y = y)
# Plots of epidemic progression (optional)
plot(out1, plottype = "spatial")
plot(out1, plottype = "curve", curvetype = "newinfect")
## Example 2: spatial SIR model
# generate infectious period(=3) for 100 individuals
lambda <- rep(3, 100)
out2 \leq epidata(type = "SIR", n = 100, tmax = 15, sus.par = 0.3, beta = 5.0, infperiod = lambda,
        x = x, y = yplot(out2, plottype = "spatial")
plot(out2, plottype = "curve", curvetype = "newinfect")
## Example 3: SI network model
contact1 <- matrix(rbinom(10000, 1, 0.1), nrow = 100, ncol = 100)
contact2 <- matrix(rbinom(10000, 1, 0.1), nrow = 100, ncol = 100)
diag(contact1[,] ) <- 0
```
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```
diag(contact2[,] ) <- 0
contact \leq array(c(contact1, contact2), dim = c(100, 100, 2))
out3 <- epidata(type = "SI", n = 100, tmax = 15, sus.par = 0.3, beta = c(3.0, 5.0),
        contact = contact)
plot(out3, plottype = "curve", curvetype = "complete")
plot(out3, plottype = "curve", curvetype = "susceptible")
plot(out3, plottype = "curve", curvetype = "newinfect")
plot(out3, plottype = "curve", curvetype = "totalinfect")
```
epidic *Deviance Information Criterion (DIC)*

Description

Computes the Deviance Information Criterion for individual level models

Usage

epidic (burnin, niter, LLchain, LLpostmean)

Arguments

References

Spiegelhalter, D., Best, N., Carlin, B., Van der Linde, A. (2002). Bayesian Measures of Model Complexity and Fit. *Journal of the Royal Statistical Society. Series B (Statistical Methodology)*, 64(4), 583-639.

```
## Example 1: spatial SI model
# generate 100 individuals
x <- runif(100, 0, 10)
y <- runif(100, 0, 10)
covariate \le- runif(100, 0, 2)
```

```
out1 \le epidata(type = "SI", n = 100, Sformula = \simcovariate, tmax = 15,
              sus.par = c(0.1, 0.3), beta = 5.0, x = x, y = y)
unif_range <- matrix(c(0, 0, 10000, 10000), nrow = 2, ncol = 2)
# estimate parameters
mcmcout \leq epimcmc(out1, tmax = 15, niter = 1500,
                    Sformula = \simcovariate,
                    sus.par.ini = c(0.003, 0.01), beta.ini =0.01,
                    pro.sus.var = c(0.1, 0.1), pro.beta.var = 0.5,
                    prior.sus.par = unif_range,
                   prior.sus.dist = c("uniform","uniform"), prior.beta.dist = "uniform",
                    prior.beta.par = c(0, 10000), adapt = TRUE, acc.rate = 0.5)
# store the estimates
sus.parameters = c(mean(unlist(mcmcout$Estimates[1])), mean(unlist(mcmcout$Estimates[2])))
beta.par = mean(unlist(mcmcout$Estimates[3]))
# likelihood value
loglike <- epilike(out1, tmax = 15, Sformula = ~covariate, sus.par = sus.parameters,
                   beta = beta.par)# deviance information criterion calculation for the above epidemic
dic <- epidic(burnin = 500, niter = 1500, LLchain = mcmcout$Loglikelihood,
              LLpostmean = loglike)
dic
```


Calculates the log likelihood

Description

Calculates the log likelihood for the specified individual level model and data set

Usage

```
epilike (object, tmin = NULL, tmax, sus.par, trans.par = NULL,
        beta = NULL, spark = NULL, Sformula = NULL, Tformula = NULL)
```
Arguments

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Value

Returns the value of the log-likelihood function.

References

Deardon R, Brooks, S. P., Grenfell, B. T., Keeling, M. J., Tildesley, M. J., Savill, N. J., Shaw, D. J., Woolhouse, M. E. (2010). Inference for individual level models of infectious diseases in large populations. *Statistica Sinica*, 20, 239-261.

See Also

[epimcmc](#page-11-1).

```
## Example 1: spatial SI model
# generate 100 individuals
x <- runif(100, 0, 10)
y <- runif(100, 0, 10)
covariate \le- runif(100, 0, 2)
out1 <- epidata(type = "SI", n = 100, Sformula = ~covariate, tmax = 15,
              sus.par = c(0.1, 0.3), beta = 5.0, x = x, y = y)
```

```
epilike(out1, tmax = 15,
        sus.par = c(0.1, 0.3), beta = 5, Sformula = ~covariate)
## Example 2: spatial SIR model
# generate infectious period (=3) for 100 individuals
lambda <- rep(3, 100)
out2 <- epidata(type = "SIR", n = 100, tmax = 15, sus.par =0.3, beta = 5.0,
              infperiod = lambda, x = x, y = y)
epilike(out2,
        tmax = 15, sus.par = 0.3, beta = 5.0)
```
epimcmc *Monte Carlo Simulation*

Description

Runs an MCMC algorithm for the estimation of specified model parameters

Usage

```
epimcmc (object, tmin = NULL, tmax,
```
niter, sus.par.ini, trans.par.ini = NULL, beta.ini = NULL, spark.ini = NULL,

Sformula = NULL, Tformula = NULL,

pro.sus.var, pro.trans.var = NULL, pro.beta.var = NULL, pro.spark.var = NULL,

prior.sus.dist, prior.trans.dist = NULL, prior.beta.dist = NULL,

prior.spark.dist = NULL, prior.sus.par, prior.trans.par, prior.beta.par = NULL,

```
prior.spark.par = NULL, adapt = FALSE, acc.rate = NULL)
```
Arguments

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Details

Independent Gaussian random walks are used as the Metropolis-Hastings MCMC proposal for all parameters. The epimcmc function depends on the [MCMC](#page-0-0) function from the **[adaptMCMC](#page-0-0)** package.

Value

Returns an object of class epimcmc that contains:

type: the compartmental framework model used in the analysis.

kernel.type: the used kernel.type in the function (distance-based or network-based).

Estimates: the MCMC output of the updated model parameters.

Loglikelihood: the loglikelihood of the updated model parameters.

Fullsamples: the MCMC output of all the model parameters (including fixed parameters).

n.sus.par: the number of parameters in the susceptibility function.

n.trans.par: the number of parameters in the transmissibility function.

n.ker.par: the number of parameters in the kernel function.

References

Rob Deardon, Xuan Fang, and Grace P. S. Kwong (2015). Statistical modelling of spatio-temporal infectious disease tranmission in Analyzing and Modeling Spatial and Temporal Dynamics of Infectious Diseases, *(Ed: D. Chen, B. Moulin, J. Wu), John Wiley & Sons.*. Chapter 11.

See Also

[summary.epimcmc](#page-20-1), [plot.epimcmc](#page-16-1), [epidata](#page-5-1), [epilike](#page-9-1), [pred.epi](#page-18-1).

Examples

Example 1: spatial SI model # generate 100 individuals x <- runif(100, 0, 10)

```
y <- runif(100, 0, 10)
covariate \le- runif(100, 0, 2)
out1 <- epidata(type = "SI", n = 100, Sformula = ~covariate, tmax = 15,
               sus.par = c(0.1, 0.3), beta = 5.0, x = x, y = y)
alphapar1 \leq matrix(c(1, 1, 1, 1), ncol = 2, nrow = 2)
betapar1 \leq c(10, 2)
epi \leq epimcmc(object = out1, tmin = 1, tmax = 15,
               niter = 1000, sus.par.ini = c(1, 1), beta.ini = 1,
               Sformula = \simcovariate, pro.sus.var = c(0.5, 0.3), pro.beta.var = 0.1,
               prior.sus.dist = c("gamma", "gamma"), prior.beta.dist = "gamma",
               prior.sus.par = alphapar1, prior.beta.par = betapar1,
               adapt = TRUE, acc.rate = 0.5)
epi
## Example 2: spatial SIR model
lambda <- rep(3, 100)
out2 \le epidata(type = "SIR", n = 100, tmax = 15, sus.par = 0.3, beta = 5.0, infperiod = lambda,
        x = x, y = yalphapar2 \leq c(1, 1)
betapar2 \leq c(1, 1)
epi2 \leq epimcmc(object = out2, tmin = 1, tmax = 15,
               niter = 1000, sus.par.ini = 1, beta.ini = 1,
               Sformula = NULL, pro.sus.var = 0.3, pro.beta.var = 0.1,
               prior.sus.dist = "gamma", prior.beta.dist = "gamma",
               prior.sus.par = alphapar2, prior.beta.par = betapar2,
               adapt = FALSE, acc.rate = NULL)
```

```
epi2
```


Description

Produces various graphs summarizing epidemic of class epidata.

Usage

```
## S3 method for class 'epidata'
plot(x, plottype, curvetype = NULL, time_id = NULL, tmin = NULL, timepoints = NULL, ...)
```
Arguments

Details

The argument plottype has two options. When plottype="spatial" spatial plots are produced for the epidemic progression over time, and when plottype="curve", the argument curvetype has to be specified to one of the four available options: "complete" for plotting the number of susceptible, infected and removed individuals at each time point, "susceptible" for plotting the number of susceptible individuals at each time point, "newinfect" for plotting the number of newly infected individuals at each time point, and "totalinfect" for plotting the cumulative number of infected individuals at each time point.

Value

plot

See Also

[epidata](#page-5-1), [plot.epimcmc](#page-16-1), [plot.pred.epi](#page-17-1).

```
## Example : spatial SI model
# generate 100 individuals
x <- runif(100, 0, 10)
y <- runif(100, 0, 10)
covariate \le- runif(100, 0, 2)
out1 <- epidata(type = "SI", n = 100, Sformula = ~covariate, tmax = 15,
               sus.par = c(0.1, 0.3), beta = 5.0, x = x, y = y)
# Plots of epidemic progression
```
plot.epimcmc 17

```
plot(out1, plottype = "spatial")
plot(out1, plottype = "curve", curvetype = "newinfect")
```
plot.epimcmc *Plot the output of* epimcmc *object*

Description

plot.epimcmc is an S3 method that plots the output of an S3 object of class epimcmc.

Usage

```
## S3 method for class 'epimcmc'
plot(x, partype, start = 1, end = NULL, thin = 1, ...)
```
Arguments

Value

plot.

See Also

[epimcmc](#page-11-1), [summary.epimcmc](#page-20-1), [mcmc](#page-0-0), [plot.mcmc](#page-0-0).

```
## Example : spatial SI model
# generate 100 individuals
set.seed(59991)
x <- runif(100, 0, 10)
y <- runif(100, 0, 10)
covariate \le- runif(100, 0, 2)
```

```
out1 <- epidata(type = "SI", n = 100, Sformula = ~covariate, tmax = 15,
               sus.par = c(0.1, 0.3), beta = 5.0, x = x, y = y)
alphapar1 <- matrix(c(1, 1, 1, 1), ncol = 2, nrow = 2)
betapar1 \leq c(10, 2)
epi \leq epimcmc(object = out1, tmin = 1, tmax = 15,
               niter = 1000, sus.par.ini = c(0.1, 0.1), beta.ini = 5,
               Sformula = \inftycovariate, pro.sus.var = c(0.2, 0.3), pro.beta.var = 0.8,
               prior.sus.dist = c("gamma", "gamma"), prior.beta.dist = "gamma",
               prior.sus.par = alphapar1, prior.beta.par = betapar1,
               adapt = TRUE, acc.rate = 0.5)
# plot estimates
plot(epi, partype = "parameter", start = 100)
```
plot.pred.epi *S3 method to provide plots of posterior predictive check.*

Description

Produces various graphs for the output of the posterior predictive check of class pred.epi.

Usage

```
## S3 method for class 'pred.epi'
plot(x, \ldots)
```
Arguments

Value

plot

See Also

[pred.epi](#page-18-1), [plot.epidata](#page-14-1), [plot.epimcmc](#page-16-1).

Description

Computing the posterior predictive check based on different summary statistics.

Usage

```
pred.epi (object, xx, criterion , n.samples, burnin = NULL, tmin = NULL,
```
Sformula = NULL, Tformula = NULL, showProgressBar = interactive())

Arguments

logical. If TRUE a progress bar is shown.

An object of class pred.epi that contains the following:

type: The compartmental framework model used in the analysis.

criterion: The (multivariate) statistical criteria used in the posterior predictive check.

crit.sim: The output of the evaluated criterion on the simulated epidemics.

crit.obs: The output of the evaluated criterion on the observed epidemics.

tmax: The last time point at which data is observed.

n.samples: The number of simulated epidemics used in the posterior predictive check procedure.

References

Deardon R, Brooks, S. P., Grenfell, B. T., Keeling, M. J., Tildesley, M. J., Savill, N. J., Shaw, D. J., Woolhouse, M. E. (2010). Inference for individual level models of infectious diseases in large populations. *Statistica Sinica*, 20, 239-261.

See Also

[epimcmc](#page-11-1), [epidata](#page-5-1), [epilike](#page-9-1), [plot.pred.epi](#page-17-1).

```
## Example 1: spatial SI model
# generate 100 individuals
set.seed(59991)
x <- runif(100, 0, 10)
y <- runif(100, 0, 10)
covariate <- cbind(runif(100, 0, 2), rbinom(100, 1, 0.5))
out \leq epidata(type = "SI", n = 100, Sformula = \simcovariate, tmax = 15,
               sus.par = c(0.1, 0.3, 0.01), beta = 5.0, x = x, y = y)
alpha alphapar2 <- matrix(c(1, 1, 1, 1, 1, 1), ncol = 2, nrow = 3)
betapar2 \leq c(1, 1)
epi<-epimcmc(object = out, tmin = 1, tmax = 15,
       niter = 500, sus.par.ini = c(1, 1, 1), beta.ini = 1,
       Sformula = \simcovariate,
 pro.sus.var = c(0.5, 0.3, 0.2), pro.beta.var = 0.1,
 prior.sus.dist = c("gamma", "gamma", "gamma"),
 prior.beta.dist = "gamma",
 prior.sus.par = alphapar2, prior.beta.par = betapar2,
       adapt = TRUE, acc.rate = 0.5)
epipred1 <- pred.epi (object = out, xx = epi,
```
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```
criterion = "newly infectious",
  n.samples = 100, burnin = 200, tmin = 1,
  Sformula = ~covariate)
plot(epipred1, col = "red", type = "b", lwd = 2)epipred2 <- pred.epi (object = out, xx = epi,
criterion = "peak time",
 n.samples = 100, burnin = 200, tmin = 1,
  Sformula = \simcovariate)
plot(epipred2, col = "dark gray")
```
summary.epimcmc *Summary method for* epimcmc *objects*

Description

Summarize a [epimcmc](#page-11-1) object and return an object of class summary.epimcmc.

Usage

S3 method for class 'epimcmc' summary(object, ...)

Arguments

See Also

[epimcmc](#page-11-1), [plot.epimcmc](#page-16-1).

```
## Example: spatial SI model
# generate 100 individuals
x <- runif(100, 0, 10)
y <- runif(100, 0, 10)
covariate \le- runif(100, 0, 2)
out1 <- epidata(type = "SI", n = 100, Sformula = ~covariate, tmax = 15,
```

```
sus.par = c(0.1, 0.3), beta = 5.0, x = x, y = y)
alphapar1 <- matrix(c(1, 1, 1, 1), ncol = 2, nrow = 2)betapar1 \leq c(1, 1)epi \leq epimcmc(object = out1, tmin = 1, tmax = 15,
               niter = 1000, sus.par.ini = c(1, 1), beta.ini = 1,
               Sformula = \simcovariate, pro.sus.var = c(0.5, 0.3), pro.beta.var = 0.1,
               prior.sus.dist = c("gamma", "gamma"), prior.beta.dist = "gamma",
               prior.sus.par = alphapar1, prior.beta.par = betapar1,
               adapt = TRUE, acc.rate = 0.5)
# summary of mcmc output
summary(epi)
```
tswv *Tomato Spotted Wilt Virus (TSWV) data*

Description

Data extracted from Hughes et al. (1997). Data obtained from a field experiment as the spatial dynamics of tomato spotted wilt virus (tswv).

Usage

data(tswv)

Format

A data frame with following variables

- x X coordinate
- y Y coordinate

inftime Infection times

removaltime Times at which individuals are removed

References

Hughes, G., McRoberts,N., Madden, L.V., Nelson, S. C. (1997). Validating mathematical models of plant disease progress in space and time. *IMA Journal of Mathematics Applied in Medicine and Biology*, 14, 85-112.

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Examples

data("tswv")

```
x < - tswv$x
y <- tswv$y
inftime <- tswv$inftime
removaltime <- tswv$removaltime
infperiod \leq rep(3, length(x))
# change to epilate object
epidat.tswv <- as.epidata(type = "SIR", n = 520, x = x, y = y,
                            inftime = inftime, infperiod = infperiod)
# plot
plot(epidat.tswv, plottype = "spatial", tmin = 2)
```
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