Package: lgcp (via r-universe)

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Title Log-Gaussian Cox Process

Type Package

LazyLoad yes

Author Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle. Additional code contributions from Edzer Pebesma, Dominic Schumacher.

Description Spatial and spatio-temporal modelling of point patterns using the log-Gaussian Cox process. Bayesian inference for spatial, spatiotemporal, multivariate and aggregated point processes using Markov chain Monte Carlo. See Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2015) <doi:10.18637/jss.v063.i07>.

Version 2.0

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Imports spatstat.geom, spatstat.explore, spatstat.utils, sp, sf, raster, tcltk, iterators, ncdf4, methods, rpanel (>= 1.1-3), fields, Matrix

RoxygenNote 7.2.3

Encoding UTF-8

NeedsCompilation no

Depends R (>= 2.10)

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Repository https://bentaylor1.r-universe.dev

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Contents

`lgcp-package` *lgcp*

Description

An R package for spatiotemporal prediction and forecasting for log-Gaussian Cox processes.

Usage

```
lgcp
```

Format

An object of class `logical` of length 1.

Details

This package was not yet installed at build time.

Index: This package was not yet installed at build time.

For examples and further details of the package, type vignette("lgcp"), or refer to the paper associated with this package.

The content of `lgcp` can be broken up as follows:

Datasets wpopdata.rda, wtowncoords.rda, wtowns.rda. Giving regional and town poopulations as well as town coordinates,are provided by Wikipedia and The Office for National Statistics under respectively the Creative Commons Attribution-ShareAlike 3.0 Unported License and the Open Government Licence.

Data manipulation

Model fitting and parameter estimation

Unconditional and conditional simulation

Summary statistics, diagnostics and visualisation

Dependencies

The `lgcp` package depends upon some other important contributions to CRAN in order to operate; their uses here are indicated:

spatstat, sp, RandomFields, iterators, ncdf, methods, tcltk, rgl, rpanel, fields, rgdal, maptools, rgeos, raster

Citation

To see how to cite `lgcp`, type `citation("lgcp")` at the console.

Author(s)

Benjamin Taylor, Health and Medicine, Lancaster University, Tilman Davies, Institute of Fundamental Sciences - Statistics, Massey University, New Zealand., Barry Rowlingson, Health and Medicine, Lancaster University Peter Diggle, Health and Medicine, Lancaster University

References

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
3. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
4. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

`.onAttach` *.onAttach function*

Description

A function to print a welcome message on loading package

Usage

```
.onAttach(libname, pkgname)
```

Arguments

`libname`	libname argument
`pkgname`	pkgname argument

Value

...

add.list *add.list function*

Description

This function adds the elements of two list objects together and returns the result in another list object.

Usage

```
add.list(list1, list2)
```

Arguments

`list1`	a list of objects that could be summed using "+"
`list2`	a list of objects that could be summed using "+"

Value

a list with ith entry the sum of list1[[i]] and list2[[i]]

addTemporalCovariates *addTemporalCovariates function*

Description

A function to 'bolt on' temporal data onto a spatial covariate design matrix. The function takes a spatial design matrix, Z(s) and converts it to a spatiotemporal design matrix Z(s,t) when the effects can be separably decomposed i.e.,
Z(s,t)beta = Z_1(s)beta_1 + Z_2(t)beta_2

An example of this function in action is given in the vignette "Bayesian_lgcp", in the section on spatiotemporal data.

Usage

```
addTemporalCovariates(temporal.formula, T, laglength, tdata, Zmat)
```

Arguments

`temporal.formula`	a formula of the form t ~ tvar1 + tvar2 etc. Where the left hand side is a "t". Note there should not be an intercept term in both of the the spatial and temporal components.
`T`	the time point of interest

`laglength`	the number of previous time points to include in the analysis
`tdata`	a data frame with variable t minimally including times (T-laglength):T and var1, var2 etc.
`Zmat`	the spatial covariates Z(s), obtained by using the getZmat function.

Details

The main idea of this function is: having created a spatial Z(s) using getZmat, to create a dummy dataset tdata and temporal formula corresponding to the temporal component of the separable effects. The entries in the model matrix Z(s,t) corresponsing to the time covariates are constant over the observation window in space, but in general vary from time-point to time-point.

Note that if there is an intercept in the spatial part of the model e.g., X ~ var1 + var2, then in the temporal model, the intercept should be removed i.e., t ~ tvar1 + tvar2 - 1

Value

A list of design matrices, one for each time, Z(s,t) for t in (T-laglength):T

See Also

chooseCellwidth, getpolyol, guessinterp, getZmat, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars

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`affine.fromFunction` *affine.fromFunction function*

Description

An affine transformation of an object of class `fromFunction`

Usage

```
## S3 method for class 'fromFunction'
affine(X, mat, ...)
```

Arguments

`X`	an object of class fromFunction
`mat`	matrix of affine transformation
`...`	additional arguments

Value

the object acted on by the transformation matrix

`affine.fromSPDF` *affine.fromSPDF function*

Description

An affine transformation of an object of class `fromSPDF`

Usage

```
## S3 method for class 'fromSPDF'
affine(X, mat, ...)
```

Arguments

`X`	an object of class fromSPDF
`mat`	matrix of affine transformation
`...`	additional arguments

Value

the object acted on by the transformation matrix

`affine.fromXYZ` *affine.fromXYZ function*

Description

An affine transformation of an object of class `fromXYZ`. Nearest Neighbour interpolation

Usage

```
## S3 method for class 'fromXYZ'
affine(X, mat, ...)
```

Arguments

`X`	an object of class fromFunction
`mat`	matrix of affine transformation
`...`	additional arguments

Value

the object acted on by the transformation matrix

`affine.SpatialPolygonsDataFrame` *affine.SpatialPolygonsDataFrame function*

Description

An affine transformation of an object of class `SpatialPolygonsDataFrame`

Usage

```
## S3 method for class 'SpatialPolygonsDataFrame'
affine(X, mat, ...)
```

Arguments

`X`	an object of class fromFunction
`mat`	matrix of affine transformation
`...`	additional arguments

Value

the object acted on by the transformation matrix

`affine.stppp` *affine.stppp function*

Description

An affine transformation of an object of class `stppp`

Usage

```
## S3 method for class 'stppp'
affine(X, mat, ...)
```

Arguments

`X`	an object of class stppp
`mat`	matrix of affine transformation
`...`	additional arguments

Value

the object acted on by the transformation matrix

aggCovInfo — *aggCovInfo function*

Description

Generic function for aggregation of covariate information.

Usage

```
aggCovInfo(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method aggCovInfo

aggCovInfo.ArealWeightedMean — *aggCovInfo.ArealWeightedMean function*

Description

Aggregation via weighted mean.

Usage

```
## S3 method for class 'ArealWeightedMean'
aggCovInfo(obj, regwts, ...)
```

Arguments

`obj`	an ArealWeightedMean object
`regwts`	regional (areal) weighting vector
`...`	additional arguments

Value

Areal weighted mean.

`aggCovInfo.ArealWeightedSum` *aggCovInfo.ArealWeightedSum function*

Description

Aggregation via weighted sum. Use to sum up population counts in regions.

Usage

```
## S3 method for class 'ArealWeightedSum'
aggCovInfo(obj, regwts, ...)
```

Arguments

`obj`	an ArealWeightedSum object
`regwts`	regional (areal) weighting vector
`...`	additional arguments

Value

Areal weighted Sum.

`aggCovInfo.Majority` *aggCovInfo.Majority function*

Description

Aggregation via majority.

Usage

```
## S3 method for class 'Majority'
aggCovInfo(obj, regwts, ...)
```

Arguments

`obj`	an Majority object
`regwts`	regional (areal) weighting vector
`...`	additional arguments

Value

The most popular cell type.

`aggregateCovariateInfo` *aggregateCovariateInfo function*

Description

A function called by cov.interp.fft to allocate and perform interpolation of covariate infomation onto the FFT grid

Usage

```
aggregateCovariateInfo(cellidx, cidx, gidx, df, fftovl, classes, polyareas)
```

Arguments

`cellidx`	the index of the cell
`cidx`	index of covariate, no longer used
`gidx`	grid index
`df`	the data frame containing the covariate information
`fftovl`	an overlay of the fft grid onto the SpatialPolygonsDataFrame or SpatialPixelsDataFrame objects
`classes`	vector of class attributes of the dataframe
`polyareas`	polygon areas of the SpatialPolygonsDataFrame or SpatialPixelsDataFrame objects

Value

the interpolated covariate information onto the FFT grid

`aggregateformulaList` *aggregateformulaList function*

Description

An internal function to collect terms from a formulalist. Not intended for general use.

Usage

```
aggregateformulaList(x, ...)
```

Arguments

`x`	an object of class "formulaList"
`...`	other arguments

Value

a formula of the form X ~ var1 + var2 tec.

`andrieuthomsh` *andrieuthomsh function*

Description

A Robbins-Munro stochastic approximation update is used to adapt the tuning parameter of the proposal kernel. The idea is to update the tuning parameter at each iteration of the sampler:

$$h^{(i+1)} = h^{(i)} + \eta^{(i+1)}(\alpha^{(i)} - \alpha_{opt}),$$

where $h^{(i)}$ and $\alpha^{(i)}$ are the tuning parameter and acceptance probability at iteration i and α_{opt} is a target acceptance probability. For Gaussian targets, and in the limit as the dimension of the problem tends to infinity, an appropriate target acceptance probability for MALA algorithms is 0.574. The sequence $\{\eta^{(i)}\}$ is chosen so that $\sum_{i=0}^{\infty} \eta^{(i)}$ is infinite whilst $\sum_{i=0}^{\infty} \left(\eta^{(i)}\right)^{1+\epsilon}$ is finite for $\epsilon > 0$. These two conditions ensure that any value of h can be reached, but in a way that maintains the ergodic behaviour of the chain. One class of sequences with this property is,

$$\eta^{(i)} = \frac{C}{i^{\alpha}},$$

where $\alpha \in (0, 1]$ and $C > 0$.The scheme is set via the `mcmcpars` function.

Usage

```
andrieuthomsh(inith, alpha, C, targetacceptance = 0.574)
```

Arguments

`inith`	initial h
`alpha`	parameter α
`C`	parameter C
`targetacceptance`	target acceptance probability

Value

an object of class andrieuthomsh

References

1. Andrieu C, Thoms J (2008). A tutorial on adaptive MCMC. Statistics and Computing, 18(4), 343-373.
2. Robbins H, Munro S (1951). A Stochastic Approximation Methods. The Annals of Mathematical Statistics, 22(3), 400-407.
3. Roberts G, Rosenthal J (2001). Optimal Scaling for Various Metropolis-Hastings Algorithms. Statistical Science, 16(4), 351-367.

See Also

mcmcpars, lgcpPredict

Examples

```
andrieuthomsh(inith=1,alpha=0.5,C=1,targetacceptance=0.574)
```

`as.array.lgcpgrid` *as.array.lgcpgrid function*

Description

Method to convert an lgcpgrid object into an array.

Usage

```
## S3 method for class 'lgcpgrid'
as.array(x, ...)
```

Arguments

`x`	an object of class lgcpgrid
`...`	other arguments

Value

conversion from lgcpgrid to array

`as.fromXYZ` *as.fromXYZ function*

Description

Generic function for conversion to a fromXYZ object (eg as would have been produced by spatialAtRisk for example.)

Usage

```
as.fromXYZ(X, ...)
```

Arguments

`X`	an object
`...`	additional arguments

Value

generic function returning method as.fromXYZ

See Also

as.im.fromXYZ, as.im.fromSPDF, as.im.fromFunction, as.fromXYZ

`as.fromXYZ.fromFunction` *as.fromXYZ.fromFunction function*

Description

Method for converting from the fromFunction class of objects to the fromXYZ class of objects. Clearly this requires the user to specify a grid onto which to compute the discretised verion.

Usage

```
## S3 method for class 'fromFunction'
as.fromXYZ(X, xyt, M = 100, N = 100, ...)
```

Arguments

X	an object of class fromFunction
xyt	and objects of class stppp
M	number of cells in x direction
N	number of cells in y direction
...	additional arguments

Value

object of class im containing normalised intensities

See Also

as.im.fromXYZ, as.im.fromSPDF, as.im.fromFunction, as.fromXYZ

as.im.fromFunction *as.im.fromFunction function*

Description

Convert an object of class fromFunction(created by spatialAtRisk for example) into a spatstat im object.

Usage

```
## S3 method for class 'fromFunction'
as.im(X, xyt, M = 100, N = 100, ...)
```

Arguments

X	an object of class fromSPDF
xyt	and objects of class stppp
M	number of cells in x direction
N	number of cells in y direction
...	additional arguments

Value

object of class im containing normalised intensities

See Also

as.im.fromXYZ, as.im.fromSPDF, as.im.fromFunction, as.fromXYZ

as.im.fromSPDF *as.im.fromSPDF function*

Description

Convert an object of class fromSPDF (created by spatialAtRisk for example) into a spatstat im object.

Usage

```
## S3 method for class 'fromSPDF'
as.im(X, ncells = 100, ...)
```

Arguments

`X`	an object of class fromSPDF
`ncells`	number of cells to divide range into; default 100
`...`	additional arguments

Value

object of class im containing normalised intensities

See Also

as.im.fromXYZ, as.im.fromSPDF, as.im.fromFunction, as.fromXYZ

`as.im.fromXYZ` *as.im.fromXYZ function*

Description

Convert an object of class fromXYZ (created by spatialAtRisk for example) into a spatstat im object.

Usage

```
## S3 method for class 'fromXYZ'
as.im(X, ...)
```

Arguments

`X`	object of class fromXYZ
`...`	additional arguments

Value

object of class im containing normalised intensities

See Also

as.im.fromSPDF, as.im.fromFunction, as.fromXYZ `as.list.lgcpgrid` *as.list.lgcpgrid function*

Description

Method to convert an lgcpgrid object into a list of matrices.

Usage

```
## S3 method for class 'lgcpgrid'
as.list(x, ...)
```

Arguments

`x`	an object of class lgcpgrid
`...`	other arguments

Value

conversion from lgcpgrid to list

See Also

lgcpgrid.list, lgcpgrid.array, print.lgcpgrid, summary.lgcpgrid, quantile.lgcpgrid, image.lgcpgrid, plot.lgcpgrid

`as.owin.stapp` *as.owin.stapp function*

Description

A function to extract the SpatialPolygons part of W and return it as an owin object.

Usage

```
## S3 method for class 'stapp'
as.owin(W, ..., fatal = TRUE)
```

Arguments

`W`	see ?as.owin
`...`	see ?as.owin
`fatal`	see ?as.owin

Value

an owin object

`as.owinlist` *as.owinlist function*

Description

Generic function for creating lists of owin objects

Usage

```
as.owinlist(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method as.owinlist

`as.owinlist.SpatialPolygonsDataFrame` *as.owinlist.SpatialPolygonsDataFrame function*

Description

A function to create a list of owin objects from a SpatialPolygonsDataFrame

Usage

```
## S3 method for class 'SpatialPolygonsDataFrame'
as.owinlist(obj, dmin = 0, check = TRUE, subset = rep(TRUE, length(obj)), ...)
```

Arguments

`obj`	a SpatialPolygonsDataFrame object
`dmin`	purpose is to simplify the SpatialPolygons. A numeric value giving the smallest permissible length of an edge. See ? simplify.owin
`check`	whether or not to use spatstat functions to check the validity of SpatialPolygons objects
`subset`	logical vector. Subset of regions to extract and conver to owin objects. By default, all regions are extracted.
`...`	additional arguments

Value

a list of owin objects corresponding to the constituent Polygons objects

as.owinlist.stapp *as.owinlist.stapp function*

Description

A function to create a list of owin objects from a stapp

Usage

```
## S3 method for class 'stapp'
as.owinlist(obj, dmin = 0, check = TRUE, ...)
```

Arguments

obj	an stapp object
dmin	purpose is to simplify the SpatialPolygons. A numeric value giving the smallest permissible length of an edge. See ? simplify.owin
check	whether or not to use spatstat functions to check the validity of SpatialPolygons objects
...	additional arguments

Value

a list of owin objects corresponding to the constituent Polygons objects

as.ppp.mstppp *as.ppp.mstppp function*

Description

Convert from mstppp to ppp. Can be useful for data handling.

Usage

```
## S3 method for class 'mstppp'
as.ppp(X, ..., fatal = TRUE)
```

Arguments

X	an object of class mstppp
...	additional arguments
fatal	logical value, see details in generic ?as.ppp

Value

a ppp object without observation times

as.ppp.stppp *as.ppp.stppp function*

Description

Convert from stppp to ppp. Can be useful for data handling.

Usage

```
## S3 method for class 'stppp'
as.ppp(X, ..., fatal = TRUE)
```

Arguments

X	an object of class stppp
...	additional arguments
fatal	logical value, see details in generic ?as.ppp

Value

a ppp object without observation times

as.SpatialGridDataFrame *as.SpatialGridDataFrame function*

Description

Generic method for convertign to an object of class SpatialGridDataFrame.

Usage

```
as.SpatialGridDataFrame(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

method as.SpatialGridDataFrame

See Also

as.SpatialGridDataFrame.fromXYZ

`as.SpatialGridDataFrame.fromXYZ`
as.SpatialGridDataFrame.fromXYZ function

Description

Method for converting objects of class fromXYZ into those of class SpatialGridDataFrame

Usage

```
## S3 method for class 'fromXYZ'
as.SpatialGridDataFrame(obj, ...)
```

Arguments

`obj`	an object of class spatialAtRisk
`...`	additional arguments

Value

an object of class SpatialGridDataFrame

See Also

as.SpatialGridDataFrame

`as.SpatialPixelsDataFrame`
as.SpatialPixelsDataFrame function

Description

Generic function for conversion to SpatialPixels objects.

Usage

```
as.SpatialPixelsDataFrame(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method as.SpatialPixels

See Also

as.SpatialPixelsDataFrame.lgcpgrid

---

as.SpatialPixelsDataFrame.lgcpgrid
as.SpatialPixelsDataFrame.lgcpgrid function

Description

Method to convert lgcpgrid objects to SpatialPixelsDataFrame objects.

Usage

```
## S3 method for class 'lgcpgrid'
as.SpatialPixelsDataFrame(obj, ...)
```

Arguments

obj	an lgcpgrid object
...	additional arguments to be passed to SpatialPoints, eg a proj4string

Value

Either a SpatialPixelsDataFrame, or a list consisting of SpatialPixelsDataFrame objects.

---

as.stppp *as.stppp function*

Description

Generic function for converting to stppp objects

Usage

```
as.stppp(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

method as.stppp

`as.stppp.stapp` *as.stppp.stapp function*

Description

A function to convert stapp objects to stppp objects for use in lgcpPredict. The regional counts in the stapp object are assigned a random location within each areal region proportional to a population density (if that is available) else the counts are distributed uniformly across the observation windows.

Usage

```
## S3 method for class 'stapp'
as.stppp(obj, popden = NULL, n = 100, dmin = 0, check = TRUE, ...)
```

Arguments

`obj`	an object of class stapp
`popden`	a 'spatialAtRisk' of sub-class 'fromXYZ' object representing the population density, or for better results, lambda(s) can also be used here. Cases are distributed across the spatial region according to popden. NULL by default, which has the effect of assigning counts uniformly.
`n`	if popden is NULL, then this parameter controls the resolution of the uniform. Otherwise if popden is of class 'fromFunction', it controls the size of the imputation grid used for sampling. Default is 100.
`dmin`	If any reginal counts are missing, then a set of polygonal 'holes' in the observation window will be computed for each. dmin is the parameter used to control the simplification of these holes (see ?simplify.owin). default is zero.
`check`	logical. If any reginal counts are missing, then roughly speaking, check specifies whether to check the 'holes'.
`...`	additional arguments

Value

...

`assigninterp` *assigninterp function*

Description

A function to assign an interpolation type to a variable in a data frame.

Usage

```
assigninterp(df, vars, value)
```

Arguments

`df`	a data frame
`vars`	character vector giving name of variables
`value`	an interpolation type, posssible options are given by typing interptypes(), see ?interptypes

Details

The three types of interpolation method employed in the package lgcp are:

1. 'Majority' The interpolated value corresponds to the value of the covariate occupying the largest area of the computational cell.
2. 'ArealWeightedMean' The interpolated value corresponds to the mean of all covariate values contributing to the computational cell weighted by their respective areas.
3. 'ArealWeightedSum' The interpolated value is the sum of all contributing covariates weighed by the proportion of area with respect to the covariate polygons. For example, suppose region A has the same area as a computational grid cell and has 500 inhabitants. If that region occupies half of a computational grid cell, then this interpolation type assigns 250 inhabitants from A to the computational grid cell.

Value

assigns an interpolation type to a variable

See Also

chooseCellwidth, getpolyol, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars

Examples

```
## Not run: spdf a SpatialPolygonsDataFrame
## Not run: spdf@data <- assigninterp(df=spdf@data,vars="pop",value="ArealWeightedSum")
```

at — *at function*

Description

at function

Usage

```
at(t, mu, theta)
```

Arguments

`t`	change in time parameter, see Brix and Diggle (2001)
`mu`	mean
`theta`	parameter beta in Brix and Diggle

Value

...

autocorr — *autocorr function*

Description

This function requires data to have been dumped to disk: see `?dump2dir` and `?setoutput`. The routine `autocorr.lgcpPredict` computes cellwise selected autocorrelations of Y. Since computing the quantiles is an expensive operation, the option to output the quantiles on a subregion of interest is also provided (by setting the argument `inWindow`, which has a sensible default).

Usage

```
autocorr(
  x,
  lags,
  tidx = NULL,
  inWindow = x$xyt$window,
  crop2parentwindow = TRUE,
  ...
)
```

Arguments

`x`	an object of class lgcpPredict
`lags`	a vector of the required lags
`tidx`	the index number of the the time interval of interest, default is the last time point.
`inWindow`	an observation owin window on which to compute the autocorrelations, can speed up calculation. Default is xxytwindow, set to NULL for full grid.
`crop2parentwindow`	logical: whether to only compute autocorrelations for cells inside xxytwindow (the 'parent window')
`...`	additional arguments

Value

an array, the [,,i]th slice being the grid of cell-wise autocorrelations.

See Also

lgcpPredict, dump2dir, setoutput, plot.lgcpAutocorr, ltar, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

`autocorrMultitype` *autocorrMultitype function*

Description

A function to compute cell-wise autocorrelation in the latent field at specifiec lags

Usage

```
autocorrMultitype(
  x,
  lags,
  fieldno,
  inWindow = x$xyt$window,
  crop2parentwindow = TRUE,
  ...
)
```

Arguments

`x`	an object of class lgcpPredictMultitypeSpatialPlusParameters
`lags`	the lags at which to compute the autocorrelation
`fieldno`	the index of the lateyt field, the i in Y_i, see the help file for lgcpPredictMultitypeSpatialPlusParameters. IN diagnostic checking ,this command should be called for each field in the model.

`inWindow`	an observation owin window on which to compute the autocorrelations, can speed up calculation. Default is xxytwindow, set to NULL for full grid.
`crop2parentwindow`	logical: whether to only compute autocorrelations for cells inside xxytwindow (the 'parent window')
`...`	other arguments

Value

an array, the [,,i]th slice being the grid of cell-wise autocorrelations.

`BetaParameters` *BetaParameters function*

Description

An internal function to declare a vector a parameter vector for the main effects.

Usage

```
BetaParameters(beta)
```

Arguments

`beta`	a vector

Value

...

`betavals` *betavals function*

Description

A function to return the sampled beta from a call to the function lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars or lgcpPredictMultitypeSpatialPlusPars

Usage

```
betavals(lg)
```

Arguments

`lg`	an object produced by a call to lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars orlgcpPredictMultitypeSpatialPlusPars

Value

the posterior sampled beta

See Also

ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, etavals

---

`blockcircbase` *blockcircbase function*

Description

Compute the base matrix of a continuous Gaussian field. Computed as a block circulant matrix on a torus where x and y is the x and y centroids (must be equally spaced)

Usage

```
blockcircbase(x, y, sigma, phi, model, additionalparameters, inverse = FALSE)
```

Arguments

`x`	x centroids, an equally spaced vector
`y`	y centroids, an equally spaced vector
`sigma`	spatial variance parameter
`phi`	spatial decay parameter
`model`	covariance model, see ?CovarianceFct
`additionalparameters`	additional parameters for chosen covariance model. See ?CovarianceFct
`inverse`	logical. Whether to return the base matrix of the inverse covariance matrix (ie the base matrix for the precision matrix), default is FALSE

Value

the base matrix of a block circulant matrix representing a stationary covariance function on a toral grid.

`blockcircbaseFunction` *blockcircbaseFunction function*

Description

Compute the base matrix of a continuous Gaussian field. Computed as a block circulant matrix on a torus where x and y is the x and y centroids (must be equally spaced). This is an extension of the function blockcircbase to extend the range of covariance functions that can be fitted to the model.

Usage

```
blockcircbaseFunction(x, y, CovFunction, CovParameters, inverse = FALSE)
```

Arguments

`x`	x centroids, an equally spaced vector
`y`	y centroids, an equally spaced vector
`CovFunction`	a function of distance, returning the covariance between points that distance apart
`CovParameters`	an object of class CovParamters, see ?CovParameters
`inverse`	logical. Whether to return the base matrix of the inverse covariance matrix (ie the base matrix for the precision matrix), default is FALSE

Value

the base matrix of a block circulant matrix representing a stationary covariance function on a toral grid.

See Also

chooseCellwidth, getpolyol, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars

`bt.scalar` *bt.scalar function*

Description

bt.scalar function

Usage

```
bt.scalar(t, theta)
```

Arguments

`t`	change in time, see Brix and Diggle (2001)
`theta`	parameter beta in Brix and Diggle

Value

...

`checkObsWin` *checkObsWin function*

Description

A function to run on an object generated by the "selectObsWindow" function. Plots the observation window with grid, use as a visual aid to check the choice of cell width is correct.

Usage

```
checkObsWin(ow)
```

Arguments

`ow`	an object generated by selectObsWindow, see ?selectObsWindow

Value

a plot of the observation window and grid

See Also

chooseCellwidth

`chooseCellwidth` *chooseCellwidth function*

Description

A function to help choose the cell width (the parameter "cellwidth" in lgcpPredictSpatialPlusPars, for example) prior to setting up the FFT grid, before an MCMC run.

Usage

```
chooseCellwidth(obj, cwinit)
```

Arguments

`obj`	an object of class ppp, stppp, SpatialPolygonsDataFrame, or owin
`cwinit`	the cell width

Details

Ideally this function should be used after having made a preliminary guess at the parameters of the latent field.The idea is to run chooseCellwidth several times, adjusting the parameter "cwinit" so as to balance available computational resources with output grid size.

Value

produces a plot of the observation window and computational grid.

See Also

getpolyol, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars

---

`circulant` *circulant function*

Description

generic function for constructing circulant matrices

Usage

```
circulant(x, ...)
```

Arguments

`x`	an object
`...`	additional arguments

Value

method circulant

circulant.matrix *circulant.matrix function*

Description

If x is a matrix whose columns are the bases of the sub-blocks of a block circulant matrix, then this function returns the block circulant matrix of interest.

Usage

```
## S3 method for class 'matrix'
circulant(x, ...)
```

Arguments

`x`	a matrix object
`...`	additional arguments

Value

If x is a matrix whose columns are the bases of the sub-blocks of a block circulant matrix, then this function returns the block circulant matrix of interest.

circulant.numeric *circulant.numeric function*

Description

returns a circulant matrix with base x

Usage

```
## S3 method for class 'numeric'
circulant(x, ...)
```

Arguments

`x`	an numeric object
`...`	additional arguments

Value

a circulant matrix with base x

clearinterp *clearinterp function*

Description

A function to remove the interpolation methods from a data frame.

Usage

```
clearinterp(df)
```

Arguments

`df`	a data frame

Value

removes the interpolation methods

computeGradtruncSpatial *computeGradtruncSpatial function*

Description

Advanced use only. A function to compute a gradient truncation parameter for 'spatial only' MALA via simulation. The function requires an FFT 'grid' to be pre-computed, see fftgrid.

Usage

```
computeGradtruncSpatial(
  nsims = 100,
  scale = 1,
  nis,
  mu,
  rootQeigs,
  invrootQeigs,
  scaleconst,
  spatial,
  cellarea
)
```

Arguments

`nsims`	The number of simulations to use in computation of gradient truncation.
`scale`	multiplicative scaling constant, returned value is scale (times) max(gradient over simulations). Default scale is 1.
`nis`	cell counts on the extended grid
`mu`	parameter of latent field, mu
`rootQeigs`	root of eigenvalues of precision matrix of latent field
`invrootQeigs`	reciprocal root of eigenvalues of precision matrix of latent field
`scaleconst`	expected number of cases, or ML estimate of this quantity
`spatial`	spatial at risk interpolated onto grid of requisite size
`cellarea`	cell area

Value

gradient truncation parameter

See Also

fftgrid

`computeGradtruncSpatioTemporal`

computeGradtruncSpatioTemporal function

Description

Advanced use only. A function to compute a gradient truncation parameter for 'spatial only' MALA via simulation. The function requires an FFT 'grid' to be pre-computed, see fftgrid.

Usage

```
computeGradtruncSpatioTemporal(
  nsims = 100,
  scale = 1,
  nis,
  mu,
  rootQeigs,
  invrootQeigs,
  spatial,
  temporal,
  bt,
  cellarea
)
```

Arguments

`nsims`	The number of simulations to use in computation of gradient truncation.
`scale`	multiplicative scaling constant, returned value is scale (times) max(gradient over simulations). Default scale is 1.
`nis`	cell counts on the extended grid
`mu`	parameter of latent field, mu
`rootQeigs`	root of eigenvalues of precision matrix of latent field
`invrootQeigs`	reciprocal root of eigenvalues of precision matrix of latent field
`spatial`	spatial at risk interpolated onto grid of requisite size
`temporal`	fitted temporal values
`bt`	vectoer of variances b(delta t) in Brix and Diggle 2001
`cellarea`	cell area

Value

gradient truncation parameter

See Also

fftgrid

`condProbs` *condProbs function*

Description

A function to compute the conditional type-probabilities from a multivariate LGCP. See the vignette "Bayesian_lgcp" for a full explanation of this.

Usage

```
condProbs(obj)
```

Arguments

`obj`	an lgcpPredictMultitypeSpatialPlusParameters object

Details

We suppose there are K point types of interest. The model for point-type k is as follows:

X_k(s) ~ Poisson[R_k(s)]

R_k(s) = C_A lambda_k(s) exp[Z_k(s)beta_k+Y_k(s)]

Here X_k(s) is the number of events of type k in the computational grid cell containing the point s, R_k(s) is the Poisson rate, C_A is the cell area, lambda_k(s) is a known offset, Z_k(s) is a vector of measured covariates and Y_i(s) where i = 1,...,K+1 are latent Gaussian processes on the computational grid. The other parameters in the model are beta_k , the covariate effects for the kth type; and eta_i = [log(sigma_i),log(phi_i)], the parameters of the process Y_i for i = 1,...,K+1 on an appropriately transformed (again, in this case log) scale.

The term 'conditional probability of type k' means the probability that at a particular location there will be an event of type k, which denoted p_k.

Value

an lgcpgrid object containing the consitional type-probabilities for each type

See Also

segProbs, postcov.lgcpPredictSpatialOnlyPlusParameters, postcov.lgcpPredictAggregateSpatialPlusParameters, postcov.lgcpPredictSpatioTemporalPlusParameters, postcov.lgcpPredictMultitypeSpatialPlusParameters, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

constanth — *constanth function*

Description

This function is used to set up a constant acceptance scheme in the argument `mcmc.control` of the function `lgcpPredict`. The scheme is set via the `mcmcpars` function.

Usage

```
constanth(h)
```

Arguments

h	an object

Value

object of class constanth

See Also

mcmcpars, lgcpPredict

Examples

```
constanth(0.01)
```

`constantInTime` *constantInTime function*

Description

Generic function for creating constant-in-time temporalAtRisk objects, that is for models where mu(t) can be assumed to be constant in time. The assumption being that the global at-risk population does not change in size over time.

Usage

```
constantInTime(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Details

For further details of temporalAtRisk objects, see ?temporalAtRisk>

Value

method constantInTime

See Also

temporalAtRisk, spatialAtRisk, temporalAtRisk.numeric, temporalAtRisk.function, constantInTime.numeric, constantInTime.stppp, print.temporalAtRisk, plot.temporalAtRisk `constantInTime.numeric` *constantInTime.numeric function*

Description

Create a constant-in-time temporalAtRisk object from a numeric object of length 1. The returned temporalAtRisk object is assumed to have been scaled correctly by the user so that mu(t) = E(number of cases in a unit time interval).

Usage

```
## S3 method for class 'numeric'
constantInTime(obj, tlim, warn = TRUE, ...)
```

Arguments

`obj`	numeric constant
`tlim`	vector of length 2 giving time limits
`warn`	Issue a warning if the given temporal intensity treated is treated as 'known'?
`...`	additional arguments

Details

For further details of temporalAtRisk objects, see ?temporalAtRisk>

Value

a function f(t) giving the (constant) temporal intensity at time t for integer t in the interval [tlim[1],tlim[2]] of class temporalAtRisk

See Also

temporalAtRisk, spatialAtRisk, temporalAtRisk.numeric, temporalAtRisk.function, constantInTime, constantInTime.stppp, print.temporalAtRisk, plot.temporalAtRisk,

constantInTime.stppp *constantInTime.stppp function*

Description

Create a constant-in-time temporalAtRisk object from an stppp object. The returned temporalAtRisk object is scaled to return mu(t) = E(number of cases in a unit time interval).

Usage

```
## S3 method for class 'stppp'
constantInTime(obj, ...)
```

Arguments

`obj`	an object of class stppp.
`...`	additional arguments

Details

For further details of temporalAtRisk objects, see ?temporalAtRisk>

Value

a function f(t) giving the (constant) temporal intensity at time t for integer t in the interval [tlim[1],tlim[2]] of class temporalAtRisk

See Also

temporalAtRisk, spatialAtRisk, temporalAtRisk.numeric, temporalAtRisk.function, constantInTime, constantInTime.numeric, print.temporalAtRisk, plot.temporalAtRisk,

cov.interp.fft *cov.interp.fft function*

Description

A function to interpolate covariate values onto the fft grid, ready for analysis

Usage

```
cov.interp.fft(
  formula,
  W,
  regionalcovariates = NULL,
  pixelcovariates = NULL,
  mcens,
  ncens,
  cellInside,
  overl = NULL
)
```

Arguments

`formula`	an object of class formula (or one that can be coerced to that class) starting with X ~ (eg X~var1+var2 *NOT for example* Y~var1+var2): a symbolic description of the model to be fitted.
`W`	an owin observation window
`regionalcovariates`	an optional SpatialPolygonsDataFrame
`pixelcovariates`	an optional SpatialPixelsDataFrame
`mcens`	x-coordinates of output grid centroids (not fft grid centroids ie *not* the extended grid)
`ncens`	y-coordinates of output grid centroids (not fft grid centroids ie *not* the extended grid)
`cellInside`	a 0-1 matrix indicating which computational cells are inside the observation window
`overl`	an overlay of the computational grid onto the SpatialPolygonsDataFrame or SpatialPixelsDataFrame.

Value

The interpolated design matrix, ready for analysis

`CovarianceFct` *CovarianceFct function*

Description

A function to

Usage

```
CovarianceFct(u, sigma, phi, model, additionalparameters)
```

Arguments

`u`	distance
`sigma`	parameter sigma
`phi`	parameter phi
`model`	character string, the model
`additionalparameters`	additional parameters for the covariance function that will be fixed.

Value

the covariance function evaluated at the specified distances

`covEffects` *covEffects function*

Description

A function used in conjunction with the function "expectation" to compute the main covariate effects,
lambda(s) exp[Z(s)beta]
in each computational grid cell. Currently only implemented for spatial processes (lgcpPredictSpatialPlusPars and lgcpPredictAggregateSpatialPlusPars).

Usage

```
covEffects(Y, beta, eta, Z, otherargs)
```

Arguments

`Y`	the latent field
`beta`	the main effects
`eta`	the parameters of the latent field
`Z`	the design matrix
`otherargs`	other arguments to the function (see vignette "Bayesian_lgcp" for an explanation)

Value

the main effects

See Also

expectation, lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars

Examples

```
## Not run: ex <- expectation(lg,covEffects)[[1]] # lg is output from spatial LGCP MCMC
```

`CovFunction` *CovFunction function*

Description

A Generic method used to specify the choice of covariance function for use in the MCMC algorithm. For further details and examples, see the vignette "Bayesian_lgcp".

Usage

```
CovFunction(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method CovFunction

See Also

CovFunction.function, exponentialCovFct, RandomFieldsCovFct, SpikedExponentialCovFct

`CovFunction.function` *CovFunction.function function*

Description

A function used to define the covariance function for the latent field prior to running the MCMC algorithm

Usage

```
## S3 method for class '`function`'
CovFunction(obj, ...)
```

Arguments

`obj`	a function object
`...`	additional arguments

Value

the covariance function ready to run the MCMC routine.

See Also

exponentialCovFct, RandomFieldsCovFct, SpikedExponentialCovFct, CovarianceFct

Examples

```
## Not run: cf1 <- CovFunction(exponentialCovFct)
## Not run: cf2 <- CovFunction(RandomFieldsCovFct(model="matern",additionalparameters=1))
```

CovParameters — *CovParameters function*

Description

A function to provide a structure for the parameters of the latent field. Not intended for general use.

Usage

```
CovParameters(list)
```

Arguments

`list`	a list

Value

an object used in the MCMC routine.

Cvb — *Cvb function*

Description

This function is used in `thetaEst` to estimate the temporal correlation parameter, theta.

Usage

```
Cvb(xyt, spatial.intensity, N = 100, spatial.covmodel, covpars)
```

Arguments

`xyt`	object of class stppp
`spatial.intensity`	bivariate density estimate of lambda, an object of class im (produced from density.ppp for example)
`N`	number of integration points
`spatial.covmodel`	spatial covariance model
`covpars`	additional covariance parameters

Value

a function, see below. Computes Monte carlo estimate of function C(v;beta) in Brix and Diggle 2001 pp 829 (... note later corrigendum to paper (2003) corrects the expression given in this paper)

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.

See Also

thetaEst

---

d.func *d.func function*

Description

d.func function

Usage

```
d.func(mat1il, mat2jk, i, j, l, k)
```

Arguments

mat1il	matrix 1
mat2jk	matrix 2
i	index matrix 1 number 1
j	index matrix 2 number 1
l	index matrix 1 number 2
k	index matrix 2 number 2

Value

...

density.stppp	*density.stppp function*

Description

A wrapper function for density.ppp.

Usage

```
## S3 method for class 'stppp'
density(x, bandwidth = NULL, ...)
```

Arguments

x	an stppp object
bandwidth	'bandwidth' parameter, equivanent to parameter sigma in ?density.ppp ie standard deviation of isotropic Gaussian smoothing kernel.
...	additional arguments to be passed to density.ppp

Value

bivariate density estimate of xyt; not this is a wrapper function for density.ppp

See Also

density.ppp

discreteWindow	*discreteWindow function*

Description

Generic function for extracting the FFT discrete window.

Usage

```
discreteWindow(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

method discreteWindow

See Also

discreteWindow.lgcpPredict

`discreteWindow.lgcpPredict`
discreteWindow.lgcpPredict function

Description

A function for extracting the FFT discrete window from an lgcpPredict object.

Usage

```
## S3 method for class 'lgcpPredict'
discreteWindow(obj, inclusion = "touching", ...)
```

Arguments

`obj`	an lgcpPredict object
`inclusion`	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.
`...`	additional arguments

Value

...

`dump2dir` *dump2dir function*

Description

This function, when set by the `gridfunction` argument of setoutput, in turn called by the argument `output.control` of `lgcpPredict` facilitates the dumping of data to disk. Data is dumped to a netCDF file, `simout.nc`, stored in the directory specified by the user. If the directory does not exist, then it will be created. Since the requested data dumped to disk may be very large in a run of `lgcpPredict`, by default, the user is prompted as to whether to proceed with prediction, this can be turned off by setting the option `forceSave=TRUE` detailed here. To save space, or increase the number of simulations that can be stored for a fixed disk space the option to only save the last time point is also available (`lastonly=TRUE`, which is the default setting).

Usage

```
dump2dir(dirname, lastonly = TRUE, forceSave = FALSE)
```

Arguments

`dirname`	character vector of length 1 containing the name of the directory to create
`lastonly`	only save output from time T? (see ?lgcpPredict for definition of T)
`forceSave`	option to override display of menu

Value

object of class dump2dir

See Also

setoutput, \ GFinitialise, GFupdate, GFfinalise, GFreturnvalue

---

`eigenfrombase` *eigenfrombase function*

Description

A function to compute the eigenvalues of an SPD block circulant matrix given the base matrix.

Usage

```
eigenfrombase(x)
```

Arguments

`x`	the base matrix

Value

the eigenvalues

`etavals` *etavals function*

Description

A function to return the sampled eta from a call to the function lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars or lgcpPredictMultitypeSpatialPlusPars

Usage

```
etavals(lg)
```

Arguments

lg	an object produced by a call to lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars orlgcpPredictMultitypeSpatialPlusPars

Value

the posterior sampled eta

See Also

ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals

---

`EvaluatePrior` *EvaluatePrior function*

Description

An internal function used in the MCMC routine to evaluate the prior for a given set of parameters

Usage

```
EvaluatePrior(etaParameters, betaParameters, prior)
```

Arguments

etaParameters	the paramter eta
betaParameters	the parameter beta
prior	the prior

Value

the prior evaluated at the given values.

exceedProbs *exceedProbs function*

Description

This function can be called using `MonteCarloAverage` (see `fun3` the examples in the help file for MonteCarloAverage). It computes exceedance probabilities,

$$P[\exp(Y_{t_1:t_2}) > k],$$

that is the probability that the relative reisk exceeds threshold k. Note that it is possible to pass vectors of tresholds to the function, and the exceedance probabilities will be computed for each of these.

Usage

```
exceedProbs(threshold, direction = "upper")
```

Arguments

`threshold`	vector of threshold levels for the indicator function
`direction`	default 'upper' giving exceedance probabilities, alternative is 'lower', which gives 'subordinate probabilities'

Value

a function of Y that computes the indicator function I(exp(Y)>threshold) evaluated for each cell of a matrix Y If several tresholds are specified an array is returned with the [,,i]th slice equal to I(exp(Y)>threshold[i])

See Also

MonteCarloAverage, setoutput

exceedProbsAggregated *exceedProbsAggregated function*

Description

NOTE THIS FUNCTION IS IN TESTING AT PRESENT

Usage

```
exceedProbsAggregated(threshold, lg = NULL, lastonly = TRUE)
```

Arguments

`threshold`	vector of threshold levels for the indicator function
`lg`	an object of class aggregatedPredict
`lastonly`	logical, whether to only compute the exceedances for the last time point. default is TRUE

Details

This function computes regional exceedance probabilities after MCMC has finished, it requires the information to have been dumped to disk, and to have been computed using the function lgcpPredictAggregated

$$P[\exp(Y_{t_1:t_2}) > k],$$

that is the probability that the relative risk exceeds threshold k. Note that it is possible to pass vectors of tresholds to the function, and the exceedance probabilities will be computed for each of these.

Value

a function of Y that computes the indicator function I(exp(Y)>threshold) evaluated for each cell of a matrix Y, but with values aggregated to regions If several tresholds are specified an array is returned with the [,,i]th slice equal to I(exp(Y)>threshold[i])

See Also

lgcpPredictAggregated

`expectation` *expectation function*

Description

Generic function used in the computation of Monte Carlo expectations.

Usage

```
expectation(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method expectation

expectation.lgcpPredict *expectation.lgcpPredict function*

Description

This function requires data to have been dumped to disk: see ?dump2dir and ?setoutput. This function computes the Monte Carlo Average of a function where data from a run of lgcpPredict has been dumped to disk.

Usage

```
## S3 method for class 'lgcpPredict'
expectation(obj, fun, maxit = NULL, ...)
```

Arguments

obj	an object of class lgcpPredict
fun	a function accepting a single argument that returns a numeric vector, matrix or array object
maxit	Not used in ordinary circumstances. Defines subset of samples over which to compute expectation. Expectation is computed using information from iterations 1:maxit, where 1 is the first non-burn in iteration dumped to disk.
...	additional arguments

Details

A Monte Carlo Average is computed as:

$$E_{\pi(Y_{t_1:t_2}|X_{t_1:t_2})}[g(Y_{t_1:t_2})] \approx \frac{1}{n}\sum_{i=1}^{n} g(Y_{t_1:t_2}^{(i)})$$

where g is a function of interest, $Y_{t_1:t_2}^{(i)}$ is the ith retained sample from the target and n is the total number of retained iterations. For example, to compute the mean of $Y_{t_1:t_2}$ set,

$$g(Y_{t_1:t_2}) = Y_{t_1:t_2},$$

the output from such a Monte Carlo average would be a set of $t_2 - t_1$ grids, each cell of which being equal to the mean over all retained iterations of the algorithm (NOTE: this is just an example computation, in practice, there is no need to compute the mean on line explicitly, as this is already done by default in lgcpPredict).

Value

the expectated value of that function

See Also

lgcpPredict, dump2dir, setoutput

`expectation.lgcpPredictSpatialOnlyPlusParameters` *expectation.lgcpPredictSpatialOnlyPlusParameters function*

Description

This function requires data to have been dumped to disk: see `?dump2dir` and `?setoutput`. This function computes the Monte Carlo Average of a function where data from a run of `lgcpPredict` has been dumped to disk.

Usage

```
"expectation(obj,fun,maxit=NULL,...)"
```

Arguments

`obj`	an object of class lgcpPredictSpatialOnlyPlusParameters
`fun`	a function with arguments 'Y', 'beta', 'eta', 'Z' and 'otherargs'. See vignette("Bayesian_lgcp") for an example
`maxit`	Not used in ordinary circumstances. Defines subset of samples over which to compute expectation. Expectation is computed using information from iterations 1:maxit, where 1 is the first non-burn in iteration dumped to disk.
`...`	additional arguments

Value

the expectated value of that function

`exponentialCovFct` *exponentialCovFct function*

Description

A function to declare and also evaluate an exponential covariance function.

Usage

```
exponentialCovFct(d, CovParameters)
```

Arguments

`d`	toral distance
`CovParameters`	parameters of the latent field, an object of class "CovParamaters".

Value

the exponential covariance function

See Also

CovFunction.function, RandomFieldsCovFct, SpikedExponentialCovFct

---

`extendspatialAtRisk` *extendspatialAtRisk function*

Description

A function to extend a spatialAtRisk object, used in interpolating the fft grid NOTE THIS DOES NOT RETURN A PROPER spatialAtRisk OBJECT SINCE THE NORMALISING CONSTANT IS PUT BACK IN.

Usage

```
extendspatialAtRisk(spatial)
```

Arguments

`spatial`	a spatialAtRisk object inheriting class 'fromXYZ'

Value

the spatialAtRisk object on a slightly larger grid, with zeros appearing outside the original extent.

`extract` *extract function*

Description

Generic function for extracting information dumped to disk. See extract.lgcpPredict for further information.

Usage

```
extract(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method extract

See Also

extract.lgcpPredict

`extract.lgcpPredict` *extract.lgcpPredict function*

Description

This function requires data to have been dumped to disk: see `?dump2dir` and `?setoutput`. `extract.lgcpPredict` extracts chunks of data that have been dumped to disk. The subset of data can either be specified using an (x,y,t,s) box or (window,t,s) region where window is a polygonal subregion of interest.

Usage

```
## S3 method for class 'lgcpPredict'
extract(
  obj,
  x = NULL,
  y = NULL,
  t,
  s = -1,
  inWindow = NULL,
  crop2parentwindow = TRUE,
  ...
)
```

Arguments

`obj`	an object of class lgcpPredict
`x`	range of x-indices: vector (eg c(2,4)) corresponding to desired subset of x coordinates. If equal to -1, then all cells in this dimension are extracted
`y`	range of y-indices as above
`t`	range of t-indices: time indices of interest
`s`	range of s-indices ie the simulation indices of interest
`inWindow`	an observation owin window over which to extract the data (alternative to specifying x and y).
`crop2parentwindow`	logical: whether to only extract cells inside objxytwindow (the 'parent window')
`...`	additional arguments

Value

extracted array

See Also

lgcpPredict, loc2poly, dump2dir, setoutput

`Extract.mstppp` *Extract.mstppp function*

Description

extracting subsets of an mstppp object.

Usage

```
"x[subset]"
```

Arguments

`x`	an object of class mstppp
`subset`	subsetto extract

Value

extracts subset of an mstppp object

`Extract.stppp` *Extract.stppp function*

Description

extracting subsets of an stppp object.

Usage

```
"x[subset]"
```

Arguments

`x`	an object of class stppp
`subset`	the subset to extract

Value

extracts subset of an stppp object

Examples

```
## Not run: xyt <- lgcpSim()
## Not run: xyt
## Not run: xyt[xyt$t>0.5]
```

`fftgrid` *fftgrid function*

Description

! As of lgcp version 0.9-5, this function is no longer used !

Usage

```
fftgrid(xyt, M, N, spatial, sigma, phi, model, covpars, inclusion = "touching")
```

Arguments

`xyt`	object of class stppp
`M`	number of centroids in x-direction
`N`	number of centroids in y-direction
`spatial`	an object of class spatialAtRisk
`sigma`	scaling paramter for spatial covariance function, see Brix and Diggle (2001)
`phi`	scaling paramter for spatial covariance function, see Brix and Diggle (2001)
`model`	correlation type see ?CovarianceFct
`covpars`	vector of additional parameters for certain classes of covariance function (eg Matern), these must be supplied in the order given in ?CovarianceFct
`inclusion`	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Details

Advanced use only. Computes various quantities for use in `lgcpPredict`, `lgcpSim` .

Value

fft objects for use in MALA

`fftinterpolate` *fftinterpolate function*

Description

Generic function used for computing interpolations used in the function fftgrid.

Usage

```
fftinterpolate(spatial, ...)
```

Arguments

`spatial`	an object
`...`	additional arguments

Value

method fftinterpolate

See Also

fftgrid

`fftinterpolate.fromFunction` *fftinterpolate.fromFunction function*

Description

This method performs interpolation within the function `fftgrid` for `fromFunction` objects.

Usage

```
## S3 method for class 'fromFunction'
fftinterpolate(spatial, mcens, ncens, ext, ...)
```

Arguments

`spatial`	objects of class spatialAtRisk
`mcens`	x-coordinates of interpolation grid in extended space
`ncens`	y-coordinates of interpolation grid in extended space
`ext`	integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.
`...`	additional arguments

Value

matrix of interpolated values

See Also

fftgrid, spatialAtRisk.function

`fftinterpolate.fromSPDF`

fftinterpolate.fromSPDF function

Description

This method performs interpolation within the function `fftgrid` for `fromSPDF` objects.

Usage

```
## S3 method for class 'fromSPDF'
fftinterpolate(spatial, mcens, ncens, ext, ...)
```

Arguments

`spatial`	objects of class spatialAtRisk
`mcens`	x-coordinates of interpolation grid in extended space
`ncens`	y-coordinates of interpolation grid in extended space
`ext`	integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.
`...`	additional arguments

Value

matrix of interpolated values

See Also

fftgrid, spatialAtRisk.SpatialPolygonsDataFrame

fftinterpolate.fromXYZ *interpolate.fromXYZ function*

Description

This method performs interpolation within the function `fftgrid` for `fromXYZ` objects.

Usage

```
## S3 method for class 'fromXYZ'
fftinterpolate(spatial, mcens, ncens, ext, ...)
```

Arguments

`spatial`	objects of class spatialAtRisk
`mcens`	x-coordinates of interpolation grid in extended space
`ncens`	y-coordinates of interpolation grid in extended space
`ext`	integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.
`...`	additional arguments

Value

matrix of interpolated values

See Also

fftgrid, spatialAtRisk.fromXYZ

fftmultiply *fftmultiply function*

Description

A function to pre-multiply a vector by a block cirulant matrix

Usage

```
fftmultiply(efb, vector)
```

Arguments

`efb`	eigenvalues of the matrix
`vector`	the vector

Value

a vector: the product of the matrix and the vector.

`formulaList` *formulaList function*

Description

A function to creat an object of class "formulaList" from a list of "formula" objects; use to define the model for the main effects prior to running the multivariate MCMC algorithm.

Usage

```
formulaList(X)
```

Arguments

`X`	a list object, each element of which is a formula

Value

an object of class "formulaList"

`GAfinalise` *GAfinalise function*

Description

Generic function defining the the finalisation step for the `gridAverage` class of functions. The function is called invisibly within `MALAlgcp` and facilitates the computation of Monte Carlo Averages online.

Usage

```
GAfinalise(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

method GAfinalise

See Also

setoutput, GAinitialise, GAupdate, GAreturnvalue

`GAfinalise.MonteCarloAverage`

GAfinalise.MonteCarloAverage function

Description

Finalise a Monte Carlo averaging scheme. Divide the sum by the number of iterations.

Usage

```
## S3 method for class 'MonteCarloAverage'
GAfinalise(F, ...)
```

Arguments

`F`	an object of class MonteCarloAverage
`...`	additional arguments

Value

computes Monte Carlo averages

See Also

MonteCarloAverage, setoutput, GAinitialise, GAupdate, GAfinalise, GAreturnvalue

---

`GAfinalise.nullAverage`

GAfinalise.nullAverage function

Description

This is a null function and performs no action.

Usage

```
## S3 method for class 'nullAverage'
GAfinalise(F, ...)
```

Arguments

`F`	an object of class nullAverage
`...`	additional arguments

Value

nothing

See Also

nullAverage, setoutput, GAinitialise, GAupdate, GAfinalise, GAreturnvalue

---

`GAinitialise` *GAinitialise function*

Description

Generic function defining the the initialisation step for the `gridAverage` class of functions. The function is called invisibly within `MALAlgcp` and facilitates the computation of Monte Carlo Averages online.

Usage

```
GAinitialise(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

method GAinitialise

See Also

setoutput, GAupdate, GAfinalise, GAreturnvalue

GAinitialise.MonteCarloAverage

GAinitialise.MonteCarloAverage function

Description

Initialise a Monte Carlo averaging scheme.

Usage

```
## S3 method for class 'MonteCarloAverage'
GAinitialise(F, ...)
```

Arguments

`F`	an object of class MonteCarloAverage
`...`	additional arguments

Value

nothing

See Also

MonteCarloAverage, setoutput, GAinitialise, GAupdate, GAfinalise, GAreturnvalue

GAinitialise.nullAverage

GAinitialise.nullAverage function

Description

This is a null function and performs no action.

Usage

```
## S3 method for class 'nullAverage'
GAinitialise(F, ...)
```

Arguments

`F`	an object of class nullAverage
`...`	additional arguments

Value

nothing

See Also

nullAverage, setoutput, GAinitialise, GAupdate, GAfinalise, GAreturnvalue

---

GammafromY *GammafromY function*

Description

A function to change Ys (spatially correlated noise) into Gammas (white noise). Used in the MALA algorithm.

Usage

```
GammafromY(Y, rootQeigs, mu)
```

Arguments

`Y`	Y matrix
`rootQeigs`	square root of the eigenvectors of the precision matrix
`mu`	parameter of the latent Gaussian field

Value

Gamma

GAreturnvalue *GAreturnvalue function*

Description

Generic function defining the the returned value for the `gridAverage` class of functions. The function is called invisibly within `MALAlgcp` and facilitates the computation of Monte Carlo Averages online.

Usage

```
GAreturnvalue(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

method GAreturnvalue

See Also

setoutput, GAinitialise, GAupdate, GAfinalise

GAreturnvalue.MonteCarloAverage

GAreturnvalue.MonteCarloAverage function

Description

Returns the required Monte Carlo average.

Usage

```
## S3 method for class 'MonteCarloAverage'
GAreturnvalue(F, ...)
```

Arguments

`F`	an object of class MonteCarloAverage
`...`	additional arguments

Value

results from MonteCarloAverage

See Also

MonteCarloAverage, setoutput, GAinitialise, GAupdate, GAfinalise, GAreturnvalue

---

GAreturnvalue.nullAverage

GAreturnvalue.nullAverage function##'

Description

This is a null function and performs no action.

Usage

```
## S3 method for class 'nullAverage'
GAreturnvalue(F, ...)
```

Arguments

`F`	an object of class nullAverage
`...`	additional arguments

Value

nothing

See Also

nullAverage, setoutput, GAinitialise, GAupdate, GAfinalise, GAreturnvalue

---

`GAupdate` *GAupdate function*

Description

Generic function defining the the update step for the `gridAverage` class of functions. The function is called invisibly within `MALAlgcp` and facilitates the computation of Monte Carlo Averages online.

Usage

```
GAupdate(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

method GAupdate

See Also

setoutput, GAinitialise, GAfinalise, GAreturnvalue

`GAupdate.MonteCarloAverage` *GAupdate.MonteCarloAverage function*

Description

Update a Monte Carlo averaging scheme. This function performs the Monte Carlo sum online.

Usage

```
## S3 method for class 'MonteCarloAverage'
GAupdate(F, ...)
```

Arguments

`F`	an object of class MonteCarloAverage
`...`	additional arguments

Value

updates Monte Carlo sums

See Also

MonteCarloAverage, setoutput, GAinitialise, GAupdate, GAfinalise, GAreturnvalue

`GAupdate.nullAverage` *GAupdate.nullAverage function*

Description

This is a null function and performs no action.

Usage

```
## S3 method for class 'nullAverage'
GAupdate(F, ...)
```

Arguments

`F`	an object of class nullAverage
`...`	additional arguments

Value

nothing

See Also

nullAverage, setoutput, GAinitialise, GAupdate, GAfinalise, GAreturnvalue

---

GaussianPrior *GaussianPrior function*

Description

A function to create a Gaussian prior.

Usage

```
GaussianPrior(mean, variance)
```

Arguments

`mean`	a vector of length 2 representing the mean.
`variance`	a 2x2 matrix representing the variance.

Value

an object of class LogGaussianPrior that can be passed to the function PriorSpec.

See Also

LogGaussianPrior, linkPriorSpec.list

Examples

```
## Not run: GaussianPrior(mean=rep(0,9),variance=diag(10^6,9))
```

---

gDisjoint_wg *gDisjoint_wg function*

Description

A function to

Usage

```
gDisjoint_wg(w, gri)
```

Arguments

`w`	X
`gri`	X

Value

...

`genFFTgrid` *genFFTgrid function*

Description

A function to generate an FFT grid and associated quantities including cell dimensions, size of extended grid, centroids, cell area, cellInside matrix (a 0/1 matrix: is the centroid of the cell inside the observation window?)

Usage

```
genFFTgrid(study.region, M, N, ext, inclusion = "touching")
```

Arguments

`study.region`	an owin object
`M`	number of cells in x direction
`N`	number of cells in y direction
`ext`	multiplying constant: the size of the extended grid: ext*M by ext*N
`inclusion`	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Value

a list

`getCellCounts` *getCellCounts function*

Description

This function is used to count the number of observations falling inside grid cells.

Usage

```
getCellCounts(x, y, xgrid, ygrid)
```

Arguments

x	x-coordinates of events
y	y-coordinates of events
xgrid	x-coordinates of grid centroids
ygrid	y-coordinates of grid centroids

Value

The number of observations in each grid cell.

---

getCounts — *getCounts function*

Description

This function is used to count the number of observations falling inside grid cells, the output is used in the function lgcpPredict.

Usage

```
getCounts(xyt, subset = rep(TRUE, xyt$n), M, N, ext)
```

Arguments

xyt	stppp or ppp data object
subset	Logical vector. Subset of data of interest, by default this is all data.
M	number of centroids in x-direction
N	number of cnetroids in y-direction
ext	how far to extend the grid eg (M,N) to (ext*M,ext*N)

Value

The number of observations in each grid cell returned on a grid suitable for use in the extended FFT space.

See Also

lgcpPredict

Examples

```
require(spatstat.explore)
xyt <- stppp(ppp(runif(100),runif(100)),t=1:100,tlim=c(1,100))
cts <- getCounts(xyt,M=64,N=64,ext=2) # gives an output grid of size 128 by 128
ctssub <- cts[1:64,1:64] # returns the cell counts in the observation
                         # window of interest
```

getCovParameters *getCovParameters function*

Description

Internal function for retrieving covariance parameters. not indended for general use.

Usage

```
getCovParameters(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

method getCovParameters

getCovParameters.GPrealisation *getCovParameters.GPrealisation function*

Description

Internal function for retrieving covariance parameters. not indended for general use.

Usage

```
## S3 method for class 'GPrealisation'
getCovParameters(obj, ...)
```

Arguments

obj	an GPrealisation object
...	additional arguments

Value

...

`getCovParameters.list` *getCovParameters.list function*

Description

Internal function for retrieving covariance parameters. not indended for general use.

Usage

```
## S3 method for class 'list'
getCovParameters(obj, ...)
```

Arguments

`obj`	an list object
`...`	additional arguments

Value

...

`getinterp` *getinterp function*

Description

A function to get the interpolation methods from a data frame

Usage

```
getinterp(df)
```

Arguments

`df`	a data frame

Details

The three types of interpolation method employed in the package lgcp are:

1. 'Majority' The interpolated value corresponds to the value of the covariate occupying the largest area of the computational cell.
2. 'ArealWeightedMean' The interpolated value corresponds to the mean of all covariate values contributing to the computational cell weighted by their respective areas.

3. 'ArealWeightedSum' The interpolated value is the sum of all contributing covariates weighed by the proportion of area with respect to the covariate polygons. For example, suppose region A has the same area as a computational grid cell and has 500 inhabitants. If that region occupies half of a computational grid cell, then this interpolation type assigns 250 inhabitants from A to the computational grid cell.

Value

the interpolation methods

---

getlgcpPredictSpatialINLA *getlgcpPredictSpatialINLA function*

Description

A function to download and 'install' lgcpPredictSpatialINLA into the lgcp namespace.

Usage

```
getlgcpPredictSpatialINLA()
```

Value

Does not return anything

---

getLHSformulaList *getLHSformulaList function*

Description

A function to retrieve the dependent variables from a formulaList object. Not intended for general use.

Usage

```
getLHSformulaList(fl)
```

Arguments

`fl`	an object of class "formulaList"

Value

the indepentdent variables

getpolyol *getpolyol function*

Description

A function to perform polygon/polygon overlay operations and form the computational grid, on which inference will eventually take place. For details and examples of using this fucntion, please see the package vignette "Bayesian_lgcp"

Usage

```
getpolyol(
  data,
  regionalcovariates = NULL,
  pixelcovariates = NULL,
  cellwidth,
  ext = 2,
  inclusion = "touching"
)
```

Arguments

- `data` an object of class ppp or SpatialPolygonsDataFrame, containing the event counts, i.e. the dataset that will eventually be analysed
- `regionalcovariates` an object of class SpatialPolygonsDataFrame containg regionally measured covariate information
- `pixelcovariates` X an object of class SpatialPixelsDataFrame containg regionally measured covariate information
- `cellwidth` the chosen cell width
- `ext` the amount by which to extend the observation window in forming the FFT grid, default is 2. In the case that the point pattern has long range spatial correlation, this may need to be increased.
- `inclusion` criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former, the default, includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Value

an object of class lgcppolyol, which can then be fed into the function getZmat.

See Also

chooseCellwidth, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars

---

`getRotation` *getRotation function*

Description

Generic function for the computation of rotation matrices.

Usage

```
getRotation(xyt, ...)
```

Arguments

`xyt`	an object
`...`	additional arguments

Value

method getRotation

See Also

getRotation.stppp

`getRotation.default` *getRotation.default function*

Description

Presently there is no default method, see ?getRotation.stppp

Usage

```
## Default S3 method:
getRotation(xyt, ...)
```

Arguments

`xyt`	an object
`...`	additional arguments

Value

currently no default implementation

See Also

getRotation.stppp

`getRotation.stppp` *getRotation.stppp function*

Description

Compute rotation matrix if observation window is a polygonal boundary

Usage

```
## S3 method for class 'stppp'
getRotation(xyt, ...)
```

Arguments

`xyt`	an object of class stppp
`...`	additional arguments

Value

the optimal rotation matrix and rotated data and observation window. Note it may or may not be advantageous to rotate the window, this information is displayed prior to the MALA routine when using lgcpPredict

---

`getup` *getup function*

Description

A function to get an object from a parent frame.

Usage

```
getup(n, lev = 1)
```

Arguments

`n`	a character string, the name of the object
`lev`	how many levels up the hierarchy to go (see the argument "envir" from the function "get"), default is 1.

Value

...

getZmat *getZmat function*

Description

A function to construct a design matrix for use with the Bayesian MCMC routines in lgcp. See the vignette "Bayesian_lgcp" for further details on how to use this function.

Usage

```
getZmat(
  formula,
  data,
  regionalcovariates = NULL,
  pixelcovariates = NULL,
  cellwidth,
  ext = 2,
  inclusion = "touching",
  overl = NULL
)
```

Arguments

`formula` a formula object of the form X ~ var1 + var2 etc. The name of the dependent variable must be "X". Only accepts 'simple' formulae, such as the example given.

`data` the data to be analysed (using, for example lgcpPredictSpatialPlusPars). Either an object of class ppp, or an object of class SpatialPolygonsDataFrame

`regionalcovariates` an optional SpatialPolygonsDataFrame object containing covariate information, if applicable

`pixelcovariates` an optional SpatialPixelsDataFrame object containing covariate information, if applicable

`cellwidth` the width of computational cells

`ext` integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.

`inclusion` criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former, the default, includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

`overl`	an object of class "lgcppolyol", created by the function getpolyol. Such an object contains the FFT grid and a polygon/polygon overlay and speeds up computation massively.

Details

For example, a spatial LGCP model for the would have the form:

X(s) ~ Poisson[R(s)]

R(s) = C_A lambda(s) exp[Z(s)beta+Y(s)]

The function getZmat helps create the matrix Z. The returned object is passed onto an MCMC function, for example lgcpPredictSpatialPlusPars or lgcpPredictAggregateSpatialPlusPars. This function can also be used to help construct Z for use with lgcpPredictSpatioTemporalPlusPars and lgcpPredictMultitypeSpatialPlusPars, but these functions require a list of such objects: see the vignette "Bayesian_lgcp" for examples.

Value

a design matrix for passing on to the Bayesian MCMC functions

See Also

chooseCellwidth, getpolyol, guessinterp, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars

---

`getZmats` *getZmats function*

Description

An internal function to create Z_k from an lgcpZmat object, for use in the multivariate MCMC algorithm. Not intended for general use.

Usage

```
getZmats(Zmat, formulaList)
```

Arguments

`Zmat`	an objecty of class "lgcpZmat"
`formulaList`	an object of class "formulaList"

Value

design matrices for each of the point types

GFfinalise *GFfinalise function*

Description

Generic function defining the the finalisation step for the `gridFunction` class of objects. The function is called invisibly within `MALAlgcp` and facilitates the dumping of data to disk

Usage

```
GFfinalise(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

method GFfinalise

See Also

setoutput, GFinitialise, GFupdate, GFreturnvalue

GFfinalise.dump2dir *GFfinalise.dump2dir function*

Description

This function finalises the dumping of data to a netCDF file.

Usage

```
## S3 method for class 'dump2dir'
GFfinalise(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

nothing

See Also

dump2dir, setoutput, GFinitialise, GFupdate, GFfinalise, GFreturnvalue

---

`GFfinalise.nullFunction` *GFfinalise.nullFunction function*

Description

This is a null function and performs no action.

Usage

```
## S3 method for class 'nullFunction'
GFfinalise(F, ...)
```

Arguments

`F`	an object of class dump2dir
`...`	additional arguments

Value

nothing

See Also

nullFunction, setoutput, GFinitialise, GFupdate, GFfinalise, GFreturnvalue

---

`GFinitialise` *GFinitialise function*

Description

Generic function defining the the initialisation step for the `gridFunction` class of objects. The function is called invisibly within `MALAlgcp` and facilitates the dumping of data to disk

Usage

```
GFinitialise(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

method GFinitialise

See Also

setoutput, GFupdate, GFfinalise, GFreturnvalue

GFinitialise.dump2dir *GFinitialise.dump2dir function*

Description

Creates a directory (if necessary) and allocates space for a netCDF dump.

Usage

```
## S3 method for class 'dump2dir'
GFinitialise(F, ...)
```

Arguments

F	an object of class dump2dir
...	additional arguments

Value

creates initialisation file and folder

See Also

dump2dir, setoutput, GFinitialise, GFupdate, GFfinalise, GFreturnvalue

---

GFinitialise.nullFunction *GFinitialise.nullFunction function*

Description

This is a null function and performs no action.

Usage

```
## S3 method for class 'nullFunction'
GFinitialise(F, ...)
```

Arguments

`F`	an object of class dump2dir
`...`	additional arguments

Value

nothing

See Also

nullFunction, setoutput, GFinitialise, GFupdate, GFfinalise, GFreturnvalue

---

`GFreturnvalue` *GFreturnvalue function*

Description

Generic function defining the the returned value for the `gridFunction` class of objects. The function is called invisibly within `MALAlgcp` and facilitates the dumping of data to disk

Usage

```
GFreturnvalue(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

method GFreturnvalue

See Also

setoutput, GFinitialise, GFupdate, GFfinalise

GFreturnvalue.dump2dir

GFreturnvalue.dump2dir function

Description

This function returns the name of the directory the netCDF file was written to.

Usage

```
## S3 method for class 'dump2dir'
GFreturnvalue(F, ...)
```

Arguments

F	an object
...	additional arguments

Value

display where files have been written to

See Also

dump2dir, setoutput, GFinitialise, GFupdate, GFfinalise, GFreturnvalue

GFreturnvalue.nullFunction

GFreturnvalue.nullFunction function

Description

This is a null function and performs no action.

Usage

```
## S3 method for class 'nullFunction'
GFreturnvalue(F, ...)
```

Arguments

F	an object of class dump2dir
...	additional arguments

Value

nothing

See Also

nullFunction, setoutput, GFinitialise, GFupdate, GFfinalise, GFreturnvalue

---

GFupdate *GFupdate function*

Description

Generic function defining the the update step for the `gridFunction` class of objects. The function is called invisibly within `MALAlgcp` and facilitates the dumping of data to disk

Usage

```
GFupdate(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

method GFupdate

See Also

setoutput, GFinitialise, GFfinalise, GFreturnvalue

GFupdate.dump2dir *GFupdate.dump2dir function*

Description

This function gets the required information from `MALAlgcp` and writes the data to the netCDF file.

Usage

```
## S3 method for class 'dump2dir'
GFupdate(F, ...)
```

Arguments

`F`	an object
`...`	additional arguments

Value

saves latent field

See Also

dump2dir, setoutput, GFinitialise, GFupdate, GFfinalise, GFreturnvalue

---

`GFupdate.nullFunction` *GFupdate.nullFunction function*

Description

This is a null function and performs no action.

Usage

```
## S3 method for class 'nullFunction'
GFupdate(F, ...)
```

Arguments

`F`	an object of class dump2dir
`...`	additional arguments

Value

nothing

See Also

nullFunction, setoutput, GFinitialise, GFupdate, GFfinalise, GFreturnvalue ginhomAverage *ginhomAverage function*

Description

A function to estimate the inhomogeneous pair correlation function for a spatiotemporal point process. See equation (8) of Diggle P, Rowlingson B, Su T (2005).

Usage

```
ginhomAverage(
  xyt,
  spatial.intensity,
  temporal.intensity,
  time.window = xyt$tlim,
  rvals = NULL,
  correction = "iso",
  suppresswarnings = FALSE,
  ...
)
```

Arguments

xyt	an object of class stppp
spatial.intensity	A spatialAtRisk object
temporal.intensity	A temporalAtRisk object
time.window	time interval contained in the interval xyt$tlim over which to compute average. Useful if there is a lot of data over a lot of time points.
rvals	Vector of values for the argument r at which g(r) should be evaluated (see ?pcfinhom). There is a sensible default.
correction	choice of edge correction to use, see ?pcfinhom, default is Ripley isotropic correction
suppresswarnings	Whether or not to suppress warnings generated by pcfinhom
...	other parameters to be passed to pcfinhom, see ?pcfinhom

Value

time average of inhomogenous pcf, equation (13) of Brix and Diggle 2001.

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Baddeley AJ, Moller J, Waagepetersen R (2000). Non-and semi-parametric estimation of interaction in inhomogeneous point patterns. Statistica Neerlandica, 54, 329-350.
3. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
4. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

KinhomAverage, spatialparsEst, thetaEst, lambdaEst, muEst

`gIntersects_pg` *gIntersects_pg function*

Description

A function to

Usage

```
gIntersects_pg(spdf, grid)
```

Arguments

`spdf`	X
`grid`	X

Value

...

gOverlay — *gOverlay function*

Description

A function to overlay the FFT grid, a SpatialPolygons object, onto a SpatialPolygonsDataFrame object.

Usage

```
gOverlay(grid, spdf)
```

Arguments

`grid`	the FFT grid, a SpatialPolygons object
`spdf`	a SpatialPolygonsDataFrame object

Details

this code was adapted from Roger Bivand:
https://stat.ethz.ch/pipermail/r-sig-geo/2011-June/012099.html

Value

a matrix describing the features of the overlay: the originating indices of grid and spdf (all non-trivial intersections) and the area of each intersection.

GPdrv — *GPdrv function*

Description

A function to compute the first derivatives of the log target with respect to the paramters of the latent field. Not intended for general purpose use.

Usage

```
GPdrv(
  GP,
  prior,
  Z,
  Zt,
  eta,
  beta,
  nis,
```

```
    cellarea,
    spatial,
    gradtrunc,
    fftgrid,
    covfunction,
    d,
    eps = 1e-06
)
```

Arguments

GP	an object of class GPrealisation
prior	priors for the model
Z	design matirix on the FFT grid
Zt	transpose of the design matrix
eta	vector of parameters, eta
beta	vector of parameters, beta
nis	cell counts on the extended grid
cellarea	the cell area
spatial	the poisson offset
gradtrunc	gradient truncation parameter
fftgrid	an object of class FFTgrid
covfunction	the choice of covariance function, see ?CovFunction
d	matrix of toral distances
eps	the finite difference step size

Value

first derivatives of the log target at the specified paramters Y, eta and beta

---

GPdrv2 *GPdrv2 function*

Description

A function to compute the second derivative of the log target with respect to the paramters of the latent field. Not intended for general purpose use.

Usage

```
GPdrv2(
  GP,
  prior,
  Z,
  Zt,
  eta,
  beta,
  nis,
  cellarea,
  spatial,
  gradtrunc,
  fftgrid,
  covfunction,
  d,
  eps = 1e-06
)
```

Arguments

`GP`	an object of class GPrealisation
`prior`	priors for the model
`Z`	design matirix on the FFT grid
`Zt`	transpose of the design matrix
`eta`	vector of parameters, eta
`beta`	vector of parameters, beta
`nis`	cell counts on the extended grid
`cellarea`	the cell area
`spatial`	the poisson offset
`gradtrunc`	gradient truncation parameter
`fftgrid`	an object of class FFTgrid
`covfunction`	the choice of covariance function, see ?CovFunction
`d`	matrix of toral distances
`eps`	the finite difference step size

Value

first and second derivatives of the log target at the specified paramters Y, eta and beta

GPdrv2_Multitype *GPdrv2_Multitype function*

Description

A function to compute the second derivatives of the log target for the multivariate model with respect to the paramters of the latent field. Not intended for general use.

Usage

```
GPdrv2_Multitype(
  GPlist,
  priorlist,
  Zlist,
  Ztlist,
  etalist,
  betalist,
  nis,
  cellarea,
  spatial,
  gradtrunc,
  fftgrid,
  covfunction,
  d,
  eps = 1e-06,
  k
)
```

Arguments

GPlist	a list of objects of class GPrealisation
priorlist	list of priors for the model
Zlist	list of design matirices on the FFT grid
Ztlist	list of transpose design matrices
etalist	list of parameters, eta, for each realisation
betalist	clist of parameters, beta, for each realisation
nis	cell counts of each type the extended grid
cellarea	the cell area
spatial	list of poisson offsets for each type
gradtrunc	gradient truncation parameter
fftgrid	an object of class FFTgrid
covfunction	list giving the choice of covariance function for each type, see ?CovFunction
d	matrix of toral distances
eps	the finite difference step size
k	index of type for which to compute the gradient and hessian

Value

first and second derivatives of the log target for tyupe k at the specified paramters Y, eta and beta

---

`GPlist2array` *GPlist2array function*

Description

An internal function for turning a list of GPrealisation objects into an an array by a particular common element of the GPrealisation object

Usage

```
GPlist2array(GPlist, element)
```

Arguments

`GPlist`	an object of class GPrealisation
`element`	the name of the element of GPlist[[1]] (for example) to extract, e.g. "Y"

Value

an array

`GPrealisation` *GPrealisation function*

Description

A function to store a realisation of a spatial gaussian process for use in MCMC algorithms that include Bayesian parameter estimation. Stores not only the realisation, but also computational quantities.

Usage

```
GPrealisation(gamma, fftgrid, covFunction, covParameters, d)
```

Arguments

`gamma`	the transformed (white noise) realisation of the process
`fftgrid`	an object of class FFTgrid, see ?genFFTgrid
`covFunction`	an object of class function returning the spatial covariance
`covParameters`	an object of class CovParamaters, see ?CovParamaters
`d`	matrix of grid distances

Value

a realisation of a spatial Gaussian process on a regular grid

---

`grid2spdf` *grid2spdf function*

Description

A function to convert a regular (x,y) grid of centroids into a SpatialPoints object

Usage

```
grid2spdf(xgrid, ygrid, proj4string = CRS(as.character(NA)))
```

Arguments

`xgrid`	vector of x centroids (equally spaced)
`ygrid`	vector of x centroids (equally spaced)
`proj4string`	an optional proj4string, projection string for the grid, set using the function CRS

Value

a SpatialPolygonsDataFrame

`grid2spix` *grid2spix function*

Description

A function to convert a regular (x,y) grid of centroids into a SpatialPixels object

Usage

```
grid2spix(xgrid, ygrid, proj4string = CRS(as.character(NA)))
```

Arguments

`xgrid`	vector of x centroids (equally spaced)
`ygrid`	vector of x centroids (equally spaced)
`proj4string`	an optional proj4string, projection string for the grid, set using the function CRS

Value

a SpatialPixels object

`grid2spoly` *grid2spoly function*

Description

A function to convert a regular (x,y) grid of centroids into a SpatialPolygons object

Usage

```
grid2spoly(xgrid, ygrid, proj4string = CRS(as.character(NA)))
```

Arguments

`xgrid`	vector of x centroids (equally spaced)
`ygrid`	vector of x centroids (equally spaced)
`proj4string`	proj 4 string: specify in the usual way

Value

a SpatialPolygons object

`grid2spts` *grid2spts function*

Description

A function to convert a regular (x,y) grid of centroids into a SpatialPoints object

Usage

```
grid2spts(xgrid, ygrid, proj4string = CRS(as.character(NA)))
```

Arguments

`xgrid`	vector of x centroids (equally spaced)
`ygrid`	vector of x centroids (equally spaced)
`proj4string`	an optional proj4string, projection string for the grid, set using the function CRS

Value

a SpatialPoints object

gridav — *gridav function*

Description

A generic function for returning `gridmeans` objects.

Usage

```
gridav(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method gridav

See Also

setoutput, lgcpgrid

gridav.lgcpPredict — *gridav.lgcpPredict function*

Description

Accessor function for `lgcpPredict objects`: returns the `gridmeans` argument set in the `output.control` argument of the function `lgcpPredict`.

Usage

```
## S3 method for class 'lgcpPredict'
gridav(obj, fun = NULL, ...)
```

Arguments

`obj`	an object of class lgcpPredict
`fun`	an optional character vector of length 1 giving the name of a function to return Monte Carlo average of
`...`	additional arguments

Value

returns the output from the gridmeans option of the setoutput argument of lgcpPredict

See Also

setoutput, lgcpgrid

`gridfun` *gridfun function*

Description

A generic function for returning `gridfunction` objects.

Usage

```
gridfun(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method gridfun

See Also

setoutput, lgcpgrid

`gridfun.lgcpPredict` *gridfun.lgcpPredict function*

Description

Accessor function for `lgcpPredict objects`: returns the `gridfunction` argument set in the `output.control` argument of the function `lgcpPredict`.

Usage

```
## S3 method for class 'lgcpPredict'
gridfun(obj, ...)
```

Arguments

`obj`	an object of class lgcpPredict
`...`	additional arguments

Value

returns the output from the gridfunction option of the setoutput argument of lgcpPredict

See Also

setoutput, lgcpgrid

`gridInWindow` *gridInWindow function*

Description

For the grid defined by x-coordinates, xvals, and y-coordinates, yvals, and an owin object W, this function just returns a logical matrix M, whose [i,j] entry is TRUE if the point(xvals[i], yvals[j]) is inside the observation window.

Usage

```
gridInWindow(xvals, yvals, win, inclusion = "touching")
```

Arguments

`xvals`	x coordinates
`yvals`	y coordinates
`win`	owin object
`inclusion`	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Value

matrix of TRUE/FALSE, which elements of the grid are inside the observation window win

gTouches_wg — *gTouches_wg function*

Description

A function to

Usage

```
gTouches_wg(w, gri)
```

Arguments

`w`	X
`gri`	X

Value

...

gu — *gu function*

Description

gu function

Usage

```
gu(u, sigma, phi, model, additionalparameters)
```

Arguments

`u`	distance
`sigma`	variance parameter, see Brix and Diggle (2001)
`phi`	scale parameter, see Brix and Diggle (2001)
`model`	correlation type, see ?CovarianceFct
`additionalparameters`	vector of additional parameters for certain classes of covariance function (eg Matern), these must be supplied in the order given in ?CovarianceFct

Value

this is just a wrapper for CovarianceFct `guessinterp` *guessinterp function*

Description

A function to guess provisional interpolational methods to variables in a data frame. Numeric variables are assigned interpolation by areal weighted mean (see below); factor, character and other types of variable are assigned interpolation by majority vote (see below). Not that the interpolation type ArealWeightedSum is not assigned automatically.

Usage

```
guessinterp(df)
```

Arguments

`df`	a data frame

Details

The three types of interpolation method employed in the package lgcp are:

1. 'Majority' The interpolated value corresponds to the value of the covariate occupying the largest area of the computational cell.
2. 'ArealWeightedMean' The interpolated value corresponds to the mean of all covariate values contributing to the computational cell weighted by their respective areas.
3. 'ArealWeightedSum' The interpolated value is the sum of all contributing covariates weighed by the proportion of area with respect to the covariate polygons. For example, suppose region A has the same area as a computational grid cell and has 500 inhabitants. If that region occupies half of a computational grid cell, then this interpolation type assigns 250 inhabitants from A to the computational grid cell.

Value

the data frame, but with attributes describing the interpolation method for each variable

See Also

chooseCellwidth, getpolyol, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars

Examples

```
## Not run: spdf a SpatialPolygonsDataFrame
## Not run: spdf@data <- guessinterp(spdf@data)
```

`hasNext` *generic hasNext method*

Description

test if an iterator has any more values to go

Usage

```
hasNext(obj)
```

Arguments

`obj`	an iterator

`hasNext.iter` *hasNext.iter function*

Description

method for iter objects test if an iterator has any more values to go

Usage

```
## S3 method for class 'iter'
hasNext(obj)
```

Arguments

`obj`	an iterator

`hvals` *hvals function*

Description

Generic function to return the values of the proposal scaling h in the MCMC algorithm.

Usage

```
hvals(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method hvals

`hvals.lgcpPredict` *hvals.lgcpPredict function*

Description

Accessor function returning the value of h, the MALA proposal scaling constant over the iterations of the algorithm for objects of class `lgcpPredict`

Usage

```
## S3 method for class 'lgcpPredict'
hvals(obj, ...)
```

Arguments

`obj`	an object of class lgcpPredict
`...`	additional arguments

Value

returns the values of h taken during the progress of the algorithm

See Also

lgcpPredict

identify.lgcpPredict — *identify.lgcpPredict function*

Description

Identifies the indices of grid cells on plots of `lgcpPredict` objects. Can be used to identify a small number of cells for further information eg trace or autocorrelation plots (provided data has been dumped to disk). On calling `identify(lg)` for example (see code below), the user can click multiply with the left mouse button on the graphics device; once the user has selected all points of interest, the right button is pressed, which returns them.

Usage

```
## S3 method for class 'lgcpPredict'
identify(x, ...)
```

Arguments

`x`	an object of class lgcpPredict
`...`	additional arguments

Value

a 2 x n matrix containing the grid indices of the points of interest, where n is the number of points selected via the mouse.

See Also

lgcpPredict, loc2poly

Examples

```
## Not run: plot(lg) # lg an lgcpPredict object
## Not run: pt_indices <- identify(lg)
```

identifygrid — *identifygrid function*

Description

Identifies the indices of grid cells on plots of objects.

Usage

```
identifygrid(x, y)
```

Arguments

`x`	the x grid centroids
`y`	the y grid centroids

Value

a 2 x n matrix containing the grid indices of the points of interest, where n is the number of points selected via the mouse.

See Also

lgcpPredict, loc2poly, identify.lgcpPredict

`image.lgcpgrid` *image.lgcpgrid function*

Description

Produce an image plot of an lgcpgrid object.

Usage

```
## S3 method for class 'lgcpgrid'
image(x, sel = 1:x$len, ask = TRUE, ...)
```

Arguments

`x`	an object of class lgcpgrid
`sel`	vector of integers between 1 and grid$len: which grids to plot. Default NULL, in which case all grids are plotted.
`ask`	logical; if TRUE the user is asked before each plot
`...`	other arguments

Value

grid plotting

See Also

lgcpgrid.list, lgcpgrid.array, as.list.lgcpgrid, print.lgcpgrid, summary.lgcpgrid, quantile.lgcpgrid, plot.lgcpgrid

initialiseAMCMC — *initialiseAMCMC function*

Description

A generic to be used for the purpose of user-defined adaptive MCMC schemes, initialiseAMCMC tells the MALA algorithm which value of h to use first. See lgcp vignette, codevignette("lgcp"), for further details on writing adaptive MCMC schemes.

Usage

```
initialiseAMCMC(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

method intialiseAMCMC

See Also

initialiseAMCMC.constanth, initialiseAMCMC.andrieuthomsh

initialiseAMCMC.andrieuthomsh — *initaliseAMCMC.andrieuthomsh function*

Description

Initialises the andrieuthomsh adaptive scheme.

Usage

```
## S3 method for class 'andrieuthomsh'
initialiseAMCMC(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

initial h for scheme

References

1. Andrieu C, Thoms J (2008). A tutorial on adaptive MCMC. Statistics and Computing, 18(4), 343-373.
2. Robbins H, Munro S (1951). A Stochastic Approximation Methods. The Annals of Mathematical Statistics, 22(3), 400-407.
3. Roberts G, Rosenthal J (2001). Optimal Scaling for Various Metropolis-Hastings Algorithms. Statistical Science, 16(4), 351-367.

See Also

andrieuthomsh

---

initialiseAMCMC.constanth

initaliseAMCMC.constanth function

Description

Initialises the constanth adaptive scheme.

Usage

```
## S3 method for class 'constanth'
initialiseAMCMC(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

initial h for scheme

See Also

constanth

`integerise` *integerise function*

Description

Generic function for converting the time variable of an stppp object.

Usage

```
integerise(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method integerise

See Also

integerise.stppp

`integerise.mstppp` *integerise.mstppp function*

Description

Function for converting the times and time limits of an mstppp object into integer values.

Usage

```
## S3 method for class 'mstppp'
integerise(obj, ...)
```

Arguments

`obj`	an mstppp object
`...`	additional arguments

Value

The mstppp object, but with integerised times.

integerise.stppp *integerise.stppp function*

Description

Function for converting the times and time limits of an stppp object into integer values. Do this before estimating mu(t), and hence before creating the temporalAtRisk object. Not taking this step is possible in lgcp, but can cause minor complications connected with the scaling of mu(t).

Usage

```
## S3 method for class 'stppp'
integerise(obj, ...)
```

Arguments

`obj`	an stppp object
`...`	additional arguments

Value

The stppp object, but with integerised times.

intens *intens function*

Description

Generic function to return the Poisson Intensity.

Usage

```
intens(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method intens

See Also

lgcpPredict, intens.lgcpPredict

`intens.lgcpPredict` *intens.lgcpPredict function*

Description

Accessor function returning the Poisson intensity as an lgcpgrid object.

Usage

```
## S3 method for class 'lgcpPredict'
intens(obj, ...)
```

Arguments

`obj`	an lgcpPredict object
`...`	additional arguments

Value

the cell-wise mean Poisson intensity, as computed by MCMC.

See Also

lgcpPredict

`intens.lgcpSimMultitypeSpatialPlusParameters`
intens.lgcpSimMultitypeSpatialPlusParameters function

Description

A function to return the cellwise Poisson intensity used during in constructing the simulated data.

Usage

```
"intens(obj, ...)"
```

Arguments

`obj`	an object of class lgcpSimMultitypeSpatialPlusParameters
`...`	other parameters

Value

the Poisson intensity

`intens.lgcpSimSpatialPlusParameters` *intens.lgcpSimSpatialPlusParameters function*

Description

A function to return the cellwise Poisson intensity used during in constructing the simulated data.

Usage

```
## S3 method for class 'lgcpSimSpatialPlusParameters'
intens(obj, ...)
```

Arguments

`obj`	an object of class lgcpSimSpatialPlusParameters
`...`	other parameters

Value

the Poisson intensity

`interptypes` *interptypes function*

Description

A function to return the types of covariate interpolation available

Usage

```
interptypes()
```

Details

The three types of interpolation method employed in the package lgcp are:

1. 'Majority' The interpolated value corresponds to the value of the covariate occupying the largest area of the computational cell.
2. 'ArealWeightedMean' The interpolated value corresponds to the mean of all covariate values contributing to the computational cell weighted by their respective areas.
3. 'ArealWeightedSum' The interpolated value is the sum of all contributing covariates weighed by the proportion of area with respect to the covariate polygons. For example, suppose region A has the same area as a computational grid cell and has 500 inhabitants. If that region occupies half of a computational grid cell, then this interpolation type assigns 250 inhabitants from A to the computational grid cell.

Value

character string of available interpolation types

`inversebase` *inversebase function*

Description

A function to compute the base of the inverse os a block circulant matrix, given the base of the matrix

Usage

```
inversebase(x)
```

Arguments

`x`	the base matrix of a block circulant matrix

Value

the base matrix of the inverse of the circulant matrix

`is.burnin` *is this a burn-in iteration?*

Description

if this mcmc iteration is in the burn-in period, return TRUE

Usage

```
is.burnin(obj)
```

Arguments

`obj`	an mcmc iterator

Value

TRUE or FALSE

is.pow2 — *is.pow2 function*

Description

Tests whether a number id

Usage

```
is.pow2(num)
```

Arguments

num	a numeric

Value

logical: is num a power of 2?

Examples

```
is.pow2(128)  # TRUE
is.pow2(64.9) # FALSE
```

is.retain — *do we retain this iteration?*

Description

if this mcmc iteration is one not thinned out, this is true

Usage

```
is.retain(obj)
```

Arguments

obj	an mcmc iterator

Value

TRUE or FALSE

`is.SPD` *is.SPD function*

Description

A function to compute whether a block circulant matrix is symmetric positive definite (SPD), given its base matrix.

Usage

```
is.SPD(base)
```

Arguments

`base`	base matrix of a block circulant matrix

Value

logical, whether the circulant matrix the base represents is SPD

`iteration` *iteration number*

Description

within a loop, this is the iteration number we are currently doing.

Usage

```
iteration(obj)
```

Arguments

`obj`	an mcmc iterator

Details

get the iteration number

Value

integer iteration number, starting from 1.

KinhomAverage *KinhomAverage function*

Description

A function to estimate the inhomogeneous K function for a spatiotemporal point process. The method of computation is similar to ginhomAverage, see eq (8) Diggle P, Rowlingson B, Su T (2005) to see how this is computed.

Usage

```
KinhomAverage(
  xyt,
  spatial.intensity,
  temporal.intensity,
  time.window = xyt$tlim,
  rvals = NULL,
  correction = "iso",
  suppresswarnings = FALSE
)
```

Arguments

`xyt`	an object of class stppp
`spatial.intensity`	A spatialAtRisk object
`temporal.intensity`	A temporalAtRisk object
`time.window`	time interval contained in the interval xyt$tlim over which to compute average. Useful if there is a lot of data over a lot of time points.
`rvals`	Vector of values for the argument r at which the inhmogeneous K function should be evaluated (see ?Kinhom). There is a sensible default.
`correction`	choice of edge correction to use, see ?Kinhom, default is Ripley isotropic correction
`suppresswarnings`	Whether or not to suppress warnings generated by Kinhom

Value

time average of inhomogenous K function.

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/

2. Baddeley AJ, Moller J, Waagepetersen R (2000). Non-and semi-parametric estimation of interaction in inhomogeneous point patterns. Statistica Neerlandica, 54, 329-350.

3. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.

4. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

ginhomAverage, spatialparsEst, thetaEst, lambdaEst, muEst

`lambdaEst` *lambdaEst function*

Description

Generic function for estimating bivariate densities by eye. Specific methods exist for stppp objects and ppp objects.

Usage

```
lambdaEst(xyt, ...)
```

Arguments

`xyt`	an object
`...`	additional arguments

Value

method lambdaEst

See Also

lambdaEst.stppp, lambdaEst.ppp

lambdaEst.ppp *lambdaEst.ppp function*

Description

A tool for the visual estimation of lambda(s) via a 2 dimensional smoothing of the case locations. For parameter estimation, the alternative is to estimate lambda(s) by some other means, convert it into a spatialAtRisk object and then into a pixel image object using the build in coercion methods, this `im` object can then be fed to ginhomAverage, KinhomAverage or thetaEst for instance.

Usage

```
## S3 method for class 'ppp'
lambdaEst(xyt, weights = c(), edge = TRUE, bw = NULL, ...)
```

Arguments

`xyt`	object of class stppp
`weights`	Optional vector of weights to be attached to the points. May include negative values. See ?density.ppp.
`edge`	Logical flag: if TRUE, apply edge correction. See ?density.ppp.
`bw`	optional bandwidth. Set to NULL by default, which calls teh resolve.2D.kernel function for computing an initial value of this
`...`	arguments to be passed to plot

Details

The function lambdaEst is built directly on the density.ppp function and as such, implements a bivariate Gaussian smoothing kernel. The bandwidth is initially that which is automatically chosen by the default method of density.ppp. Since image plots of these kernel density estimates may not have appropriate colour scales, the ability to adjust this is given with the slider 'colour adjustment'. With colour adjustment set to 1, the default image.plot for the equivalent pixel image object is shown and for values less than 1, the colour scheme is more spread out, allowing the user to get a better feel for the density that is being fitted. NOTE: colour adjustment does not affect the returned density and the user should be aware that the returned density will 'look like' that displayed when colour adjustment is set equal to 1.

Value

This is an rpanel function for visual choice of lambda(s), the output is a variable, varname, with the density *per unit time* the variable varname can be fed to the function ginhomAverage or KinhomAverage as the argument density (see for example ?ginhomAverage), or into the function thetaEst as the argument spatial.intensity.

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
3. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

spatialAtRisk, ginhomAverage, KinhomAverage, spatialparsEst, thetaEst, muEst

---

`lambdaEst.stppp` *lambdaEst.stppp function*

Description

A tool for the visual estimation of lambda(s) via a 2 dimensional smoothing of the case locations. For parameter estimation, the alternative is to estimate lambda(s) by some other means, convert it into a spatialAtRisk object and then into a pixel image object using the build in coercion methods, this `im` object can then be fed to ginhomAverage, KinhomAverage or thetaEst for instance.

Usage

```
## S3 method for class 'stppp'
lambdaEst(xyt, weights = c(), edge = TRUE, bw = NULL, ...)
```

Arguments

`xyt`	object of class stppp
`weights`	Optional vector of weights to be attached to the points. May include negative values. See ?density.ppp.
`edge`	Logical flag: if TRUE, apply edge correction. See ?density.ppp.
`bw`	optional bandwidth. Set to NULL by default, which calls teh resolve.2D.kernel function for computing an initial value of this
`...`	arguments to be passed to plot

Details

The function lambdaEst is built directly on the density.ppp function and as such, implements a bivariate Gaussian smoothing kernel. The bandwidth is initially that which is automatically chosen by the default method of density.ppp. Since image plots of these kernel density estimates may not have appropriate colour scales, the ability to adjust this is given with the slider 'colour adjustment'. With colour adjustment set to 1, the default image.plot for the equivalent pixel image object is shown and for values less than 1, the colour scheme is more spread out, allowing the user to get a better feel for the density that is being fitted. NOTE: colour adjustment does not affect the returned density and the user should be aware that the returned density will 'look like' that displayed when colour adjustment is set equal to 1.

Value

This is an rpanel function for visual choice of lambda(s), the output is a variable, varname, with the density *per unit time* the variable varname can be fed to the function ginhomAverage or KinhomAverage as the argument density (see for example ?ginhomAverage), or into the function thetaEst as the argument spatial.intensity.

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
3. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

spatialAtRisk, ginhomAverage, KinhomAverage, spatialparsEst, thetaEst, muEst

---

`lgcpbayes` *lgcpbayes function*

Description

Display the introductory vignette for the lgcp package.

Usage

```
lgcpbayes()
```

Value

displays the vignette by calling browseURL `lgcpForecast` *lgcpForecast function*

Description

Function to produce forecasts for the mean field Y at times beyond the last time point in the analysis (given by the argument `T` in the function `lgcpPredict`).

Usage

```
lgcpForecast(
  lg,
  ptimes,
  spatial.intensity,
  temporal.intensity,
  inclusion = "touching"
)
```

Arguments

`lg`	an object of class lgcpPredict
`ptimes`	vector of time points for prediction. Must start strictly after last inferred time point.
`spatial.intensity`	the fixed spatial component: an object of that can be coerced to one of class spatialAtRisk
`temporal.intensity`	the fixed temporal component: either a numeric vector, or a function that can be coerced into an object of class temporalAtRisk
`inclusion`	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Value

forcasted relative risk, Poisson intensities and Y values over grid, together with approximate variance.

References

Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.

See Also

lgcpPredict

`lgcpgrid` *lgcpgrid function*

Description

Generic function for the hadling of list objects where each element of the list is a matrix. Each matrix is assumed to have the same dimension. Such objects arise from the various routines in the package lgcp.

Usage

```
lgcpgrid(grid, ...)
```

Arguments

`grid`	a list object with each member of the list being a numeric matrix, each matrix having the same dimension
`...`	other arguments

Details

lgcpgrid objects are list objects with names len, nrow, ncol, grid, xvals, yvals, zvals. The first three elements of the list store the dimension of the object, the fourth element, grid, is itself a list object consisting of matrices in which the data is stored. The last three arguments can be used to give what is effectively a 3 dimensional array a physical reference.

For example, the mean of Y from a call to lgcpPredict, obj$y.mean for example, is stored in an lgcpgrid object. If several time points have been stored in the call to lgcpPredict, then the grid element of the lgcpgrid object contains the output for each of the time points in succession. So the first element, obj$y.mean$grid[[1]],contains the output from the first time point and so on.

Value

method lgcpgrid

See Also

lgcpgrid.list, lgcpgrid.array, lgcpgrid.matrix `lgcpgrid.array` *lgcpgrid.array function*

Description

Creates an lgcp grid object from an 3-dimensional array.

Usage

```
## S3 method for class 'array'
lgcpgrid(
  grid,
  xvals = 1:dim(grid)[1],
  yvals = 1:dim(grid)[2],
  zvals = 1:dim(grid)[3],
  ...
)
```

Arguments

`grid`	a three dimensional array object
`xvals`	optional vector of x-coordinates associated to grid. By default, this is the cell index in the x direction.
`yvals`	optional vector of y-coordinates associated to grid. By default, this is the cell index in the y direction.
`zvals`	optional vector of z-coordinates (time) associated to grid. By default, this is the cell index in the z direction.
`...`	other arguments

Value

an object of class lgcpgrid

See Also

lgcpgrid.list, as.list.lgcpgrid, print.lgcpgrid, summary.lgcpgrid, quantile.lgcpgrid, image.lgcpgrid, plot.lgcpgrid

`lgcpgrid.list` *lgcpgrid.list function*

Description

Creates an lgcpgrid object from a list object plus some optional coordinates. Note that each element of the list should be a matrix, and that each matrix should have the same dimension.

Usage

```
## S3 method for class 'list'
lgcpgrid(
  grid,
  xvals = 1:dim(grid[[1]])[1],
  yvals = 1:dim(grid[[1]])[2],
  zvals = 1:length(grid),
  ...
)
```

Arguments

`grid`	a list object with each member of the list being a numeric matrix, each matrix having the same dimension
`xvals`	optional vector of x-coordinates associated to grid. By default, this is the cell index in the x direction.
`yvals`	optional vector of y-coordinates associated to grid. By default, this is the cell index in the y direction.
`zvals`	optional vector of z-coordinates (time) associated to grid. By default, this is the cell index in the z direction.
`...`	other arguments

Value

an object of class lgcpgrid

See Also

lgcpgrid.array, as.list.lgcpgrid, print.lgcpgrid, summary.lgcpgrid, quantile.lgcpgrid, image.lgcpgrid, plot.lgcpgrid

lgcpgrid.matrix — *lgcpgrid.matrix function*

Description

Creates an lgcp grid object from an 2-dimensional matrix.

Usage

```
## S3 method for class 'matrix'
lgcpgrid(grid, xvals = 1:nrow(grid), yvals = 1:ncol(grid), ...)
```

Arguments

`grid`	a three dimensional array object
`xvals`	optional vector of x-coordinates associated to grid. By default, this is the cell index in the x direction.
`yvals`	optional vector of y-coordinates associated to grid. By default, this is the cell index in the y direction.
`...`	other arguments

Value

an object of class lgcpgrid

See Also

lgcpgrid.list, as.list.lgcpgrid, print.lgcpgrid, summary.lgcpgrid, quantile.lgcpgrid, image.lgcpgrid, plot.lgcpgrid

lgcpInits — *lgcpInits function*

Description

A function to declare initial values for a run of the MCMC routine. If specified, the MCMC algorithm will calibrate the proposal density using these as provisional estimates of the parameters.

Usage

```
lgcpInits(etainit = NULL, betainit = NULL)
```

Arguments

`etainit`	a vector, the initial value of eta to use
`betainit`	a vector, the initial value of beta to use, this vector must have names the same as the variable names in the formula in use, and in the same order.

Details

It is not necessary to supply intial values to the MCMC routine, by default the functions lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars and lgcpPredictMultitypeSpatialPlusPars will initialise the MCMC as follows. For eta, if no initial value is specified then the initial value of eta in the MCMC run will be the prior mean. For beta, if no initial value is specified then the initial value of beta in the MCMC run will be estimated from an overdispersed Poisson fit to the cell counts, ignoring spatial correlation. The user cannot specify an initial value of Y (or equivalently Gamma), as a sensible value is chosen by the MCMC function.

A secondary function of specifying initial values is to help design the MCMC proposal matrix, which is based on these initial estimates.

Value

an object of class lgcpInits used in the MCMC routine.

See Also

chooseCellwidth, getpolyol, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, CovFunction, lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars

Examples

```
## Not run: INITS <- lgcpInits(etainit=log(c(sqrt(1.5),275)), betainit=NULL)
```

`lgcppars` *lgcppars function*

Description

A function for setting the parameters sigma, phi and theta for `lgcpPredict`. Note that the returned set of parameters also features mu=-0.5*sigma^2, gives mean(exp(Y)) = 1.

Usage

```
lgcppars(sigma = NULL, phi = NULL, theta = NULL, mu = NULL, beta = NULL)
```

Arguments

`sigma`	sigma parameter
`phi`	phi parameter
`theta`	this is 'beta' parameter in Brix and Diggle (2001)
`mu`	the mean of the latent field, if equal to NULL, this is set to -sigma^2/2
`beta`	ONLY USED IN case where there is covariate information.

See Also

lgcpPredict

`lgcpPredict` *lgcpPredict function*

Description

The function `lgcpPredict` performs spatiotemporal prediction for log-Gaussian Cox Processes

Usage

```
lgcpPredict(
  xyt,
  T,
  laglength,
  model.parameters = lgcppars(),
  spatial.covmodel = "exponential",
  covpars = c(),
  cellwidth = NULL,
  gridsize = NULL,
  spatial.intensity,
  temporal.intensity,
  mcmc.control,
  output.control = setoutput(),
  missing.data.areas = NULL,
  autorotate = FALSE,
  gradtrunc = Inf,
  ext = 2,
  inclusion = "touching"
)
```

Arguments

`xyt`	a spatio-temporal point pattern object, see ?stppp
`T`	time point of interest

`laglength` specifies lag window, so that data from and including time (T-laglength) to time T is used in the MALA algorithm

`model.parameters` values for parameters, see ?lgcppars

`spatial.covmodel` correlation type see ?CovarianceFct

`covpars` vector of additional parameters for certain classes of covariance function (eg Matern), these must be supplied in the order given in ?CovarianceFct

`cellwidth` width of grid cells on which to do MALA (grid cells are square) in same units as observation window. Note EITHER gridsize OR cellwidth must be specified.

`gridsize` size of output grid required. Note EITHER gridsize OR cellwidthe must be specified.

`spatial.intensity` the fixed spatial component: an object of that can be coerced to one of class spatialAtRisk

`temporal.intensity` the fixed temporal component: either a numeric vector, or a function that can be coerced into an object of class temporalAtRisk

`mcmc.control` MCMC paramters, see ?mcmcpars

`output.control` output choice, see ?setoutput

`missing.data.areas` a list of owin objects (of length laglength+1) which has xyt$window as a base window, but with polygonal holes specifying spatial areas where there is missing data.

`autorotate` logical: whether or not to automatically do MCMC on optimised, rotated grid.

`gradtrunc` truncation for gradient vector equal to H parameter Moller et al 1998 pp 473. Default is Inf, which means no gradient truncation. Set to NULL to estimate this automatically (though note that this may not necessarily be a good choice). The default seems to work in most settings.

`ext` integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays very slowly (compared withe the size of hte observation window), increasing 'ext' may be necessary.

`inclusion` criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window. further notes on autorotate argument: If set to TRUE, and the argument spatial is not NULL, then the argument spatial must be computed in the original frame of reference (ie NOT in the rotated frame). Autorotate performs bilinear interpolation (via interp.im) on an inverse transformed grid; if there is no computational advantage in doing this, a warning message will be issued. Note that best accuracy is achieved by manually rotating xyt and then computing spatial on the transformed xyt and finally feeding these in as arguments to the function lgcpPredict. By default autorotate is set to FALSE.

Details

The following is a mathematical description of a log-Gaussian Cox Process, it is best viewed in the pdf version of the manual.

Let $\mathcal{Y}(s,t)$ be a spatiotemporal Gaussian process, $W \subset R^2$ be an observation window in space and $T \subset R_{\geq 0}$ be an interval of time of interest. Cases occur at spatio-temporal positions $(x,t) \in W \times T$ according to an inhomogeneous spatio-temporal Cox process, i.e. a Poisson process with a stochastic intensity $R(x,t)$, The number of cases, $X_{S,[t_1,t_2]}$, arising in any $S \subseteq W$ during the interval $[t_1,t_2] \subseteq T$ is then Poisson distributed conditional on $R(\cdot)$,

$$X_{S,[t_1,t_2]} \sim \text{Poisson}\left\{\int_S \int_{t_1}^{t_2} R(s,t)dsdt\right\}$$

Following Brix and Diggle (2001) and Diggle et al (2005), the intensity is decomposed multiplicatively as

$$R(s,t) = \lambda(s)\mu(t)\exp\{\mathcal{Y}(s,t)\}.$$

In the above, the fixed spatial component, $\lambda : R^2 \mapsto R_{\geq 0}$, is a known function, proportional to the population at risk at each point in space and scaled so that

$$\int_W \lambda(s)ds = 1,$$

whilst the fixed temporal component, $\mu : R_{\geq 0} \mapsto R_{\geq 0}$, is also a known function with

$$\mu(t)\delta t = E[X_{W,\delta t}],$$

for t in a small interval of time, δt, over which the rate of the process over W can be considered constant.

NOTE: the xyt stppp object can be recorded in continuous time, but for the purposes of prediciton, discretisation must take place. For the time dimension, this is achieved invisibly by `as.integer(xyt$t)` and `as.integer(xyt$tlim)`. Therefore, before running an analysis please make sure that this is commensurate with the physical inerpretation and requirements of your output. The spatial discretisation is chosen with the argument cellwidth (or gridsize). If the chosen discretisation in time and space is too coarse for a given set of parameters (sigma, phi and theta) then the proper correlation structures implied by the model will not be captured in the output.

Before calling this function, the user must decide on the time point of interest, the number of intervals of data to use, the parameters, spatial covariance model, spatial discretisation, fixed spatial ($\lambda(s)$) and temporal ($\mu(t)$) components, mcmc parameters, and whether or not any output is required.

Value

the results of fitting the model in an object of class `lgcpPredict`

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/

2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
3. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
4. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
5. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

See Also

KinhomAverage, ginhomAverage, lambdaEst, muEst, spatialparsEst, thetaEst, spatialAtRisk, temporalAtRisk, lgcppars, CovarianceFct, mcmcpars, setoutput print.lgcpPredict, xvals.lgcpPredict, yvals.lgcpPredict, plot.lgcpPredict, meanfield.lgcpPredict, rr.lgcpPredict, serr.lgcpPredict, intens.lgcpPredict, varfield.lgcpPredict, gridfun.lgcpPredict, gridav.lgcpPredict, hvals.lgcpPredict, window.lgcpPredict, mcmctrace.lgcpPredict, plotExceed.lgcpPredict, quantile.lgcpPredict, identify.lgcpPredict, expectation.lgcpPredict, extract.lgcpPredict, showGrid.lgcpPredict

---

`lgcpPredictAggregated` *lgcpPredictAggregated function*

Description

The function `lgcpPredict` performs spatiotemporal prediction for log-Gaussian Cox Processes for point process data where counts have been aggregated to the regional level. This is achieved by imputation of the regional counts onto a spatial continuum; if something is known about the underlying spatial density of cases, then this information can be added to improve the quality of the imputation, without this, the counts are distributed uniformly within regions.

Usage

```
lgcpPredictAggregated(
  app,
  popden = NULL,
  T,
  laglength,
  model.parameters = lgcppars(),
  spatial.covmodel = "exponential",
  covpars = c(),
  cellwidth = NULL,
  gridsize = NULL,
  spatial.intensity,
  temporal.intensity,
  mcmc.control,
  output.control = setoutput(),
  autorotate = FALSE,
```

```
  gradtrunc = NULL,
  n = 100,
  dmin = 0,
  check = TRUE
)
```

Arguments

`app` a spatio-temporal aggregated point pattern object, see ?stapp

`popden` a spatialAtRisk object of class 'fromFunction' describing the population density, if known. Default is NULL, which gives a uniform density on each region.

`T` time point of interest

`laglength` specifies lag window, so that data from and including time (T-laglength) to time T is used in the MALA algorithm

`model.parameters` values for parameters, see ?lgcppars

`spatial.covmodel` correlation type see ?CovarianceFct

`covpars` vector of additional parameters for certain classes of covariance function (eg Matern), these must be supplied in the order given in ?CovarianceFct

`cellwidth` width of grid cells on which to do MALA (grid cells are square). Note EITHER gridsize OR cellwidthe must be specified.

`gridsize` size of output grid required. Note EITHER gridsize OR cellwidthe must be specified.

`spatial.intensity` the fixed spatial component: an object of that can be coerced to one of class spatialAtRisk

`temporal.intensity` the fixed temporal component: either a numeric vector, or a function that can be coerced into an object of class temporalAtRisk

`mcmc.control` MCMC paramters, see ?mcmcpars

`output.control` output choice, see ?setoutput

`autorotate` logical: whether or not to automatically do MCMC on optimised, rotated grid.

`gradtrunc` truncation for gradient vector equal to H parameter Moller et al 1998 pp 473. Set to NULL to estimate this automatically (default). Set to zero for no gradient truncation.

`n` parameter for as.stppp. If popden is NULL, then this parameter controls the resolution of the uniform. Otherwise if popden is of class 'fromFunction', it controls the size of the imputation grid used for sampling. Default is 100.

`dmin` parameter for as.stppp. If any reginal counts are missing, then a set of polygonal 'holes' in the observation window will be computed for each. dmin is the parameter used to control the simplification of these holes (see ?simplify.owin). default is zero.

check — logical parameter for as.stppp. If any reginal counts are missing, then roughly speaking, check specifies whether to check the 'holes'. further notes on autorotate argument: If set to TRUE, and the argument spatial is not NULL, then the argument spatial must be computed in the original frame of reference (ie NOT in the rotated frame). Autorotate performs bilinear interpolation (via interp.im) on an inverse transformed grid; if there is no computational advantage in doing this, a warning message will be issued. Note that best accuracy is achieved by manually rotating xyt and then computing spatial on the transformed xyt and finally feeding these in as arguments to the function lgcpPredict. By default autorotate is set to FALSE.

Details

The following is a mathematical description of a log-Gaussian Cox Process, it is best viewed in the pdf version of the manual.

Let $\mathcal{Y}(s,t)$ be a spatiotemporal Gaussian process, $W \subset R^2$ be an observation window in space and $T \subset R_{\geq 0}$ be an interval of time of interest. Cases occur at spatio-temporal positions $(x,t) \in W \times T$ according to an inhomogeneous spatio-temporal Cox process, i.e. a Poisson process with a stochastic intensity $R(x,t)$, The number of cases, $X_{S,[t_1,t_2]}$, arising in any $S \subseteq W$ during the interval $[t_1,t_2] \subseteq T$ is then Poisson distributed conditional on $R(\cdot)$,

$$X_{S,[t_1,t_2]} \sim \text{Poisson}\left\{\int_S \int_{t_1}^{t_2} R(s,t) ds dt\right\}$$

Following Brix and Diggle (2001) and Diggle et al (2005), the intensity is decomposed multiplicatively as

$$R(s,t) = \lambda(s)\mu(t)\exp\{\mathcal{Y}(s,t)\}.$$

In the above, the fixed spatial component, $\lambda : R^2 \mapsto R_{\geq 0}$, is a known function, proportional to the population at risk at each point in space and scaled so that

$$\int_W \lambda(s) ds = 1,$$

whilst the fixed temporal component, $\mu : R_{\geq 0} \mapsto R_{\geq 0}$, is also a known function with

$$\mu(t)\delta t = E[X_{W,\delta t}],$$

for t in a small interval of time, δt, over which the rate of the process over W can be considered constant.

NOTE: the xyt stppp object can be recorded in continuous time, but for the purposes of prediciton, discretisation must take place. For the time dimension, this is achieved invisibly by `as.integer(xyt$t)` and `as.integer(xyt$tlim)`. Therefore, before running an analysis please make sure that this is commensurate with the physical inerpretation and requirements of your output. The spatial discretisation is chosen with the argument cellwidth (or gridsize). If the chosen discretisation in time and space is too coarse for a given set of parameters (sigma, phi and theta) then the proper correlation structures implied by the model will not be captured in the output.

Before calling this function, the user must decide on the time point of interest, the number of intervals of data to use, the parameters, spatial covariance model, spatial discretisation, fixed spatial ($\lambda(s)$) and temporal ($\mu(t)$) components, mcmc parameters, and whether or not any output is required.

Value

the results of fitting the model in an object of class `lgcpPredict`

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
3. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
4. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
5. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

See Also

KinhomAverage, ginhomAverage, lambdaEst, muEst, spatialparsEst, thetaEst, spatialAtRisk, temporalAtRisk, lgcppars, CovarianceFct, mcmcpars, setoutput print.lgcpPredict, xvals.lgcpPredict, yvals.lgcpPredict, plot.lgcpPredict, meanfield.lgcpPredict, rr.lgcpPredict, serr.lgcpPredict, intens.lgcpPredict, varfield.lgcpPredict, gridfun.lgcpPredict, gridav.lgcpPredict, hvals.lgcpPredict, window.lgcpPredict, mcmctrace.lgcpPredict, plotExceed.lgcpPredict, quantile.lgcpPredict, identify.lgcpPredict, expectation.lgcpPredict, extract.lgcpPredict, showGrid.lgcpPredict

---

`lgcpPredictAggregateSpatialPlusPars`

lgcpPredictAggregateSpatialPlusPars function

Description

A function to deliver fully Bayesian inference for the aggregated spatial log-Gaussian Cox process.

Usage

```
lgcpPredictAggregateSpatialPlusPars(
  formula,
  spdf,
  Zmat = NULL,
  overlayInZmat = FALSE,
  model.priors,
  model.inits = lgcpInits(),
  spatial.covmodel,
  cellwidth = NULL,
```

```
  poisson.offset = NULL,
  mcmc.control,
  output.control = setoutput(),
  gradtrunc = Inf,
  ext = 2,
  Nfreq = 101,
  inclusion = "touching",
  overlapping = FALSE,
  pixwts = NULL
)
```

Arguments

formula	a formula object of the form X ~ var1 + var2 etc. The name of the dependent variable must be "X". Only accepts 'simple' formulae, such as the example given.
spdf	a SpatialPolygonsDataFrame object with variable "X", the event counts per region.
Zmat	design matrix Z (see below) constructed with getZmat
overlayInZmat	if the covariate information in Zmat also comes from spdf, set to TRUE to avoid replicating the overlay operations. Default is FALSE.
model.priors	model priors, set using lgcpPrior
model.inits	model initial values. The default is NULL, in which case lgcp will use the prior mean to initialise eta and beta will be initialised from an oversispersed glm fit to the data. Otherwise use lgcpInits to specify.
spatial.covmodel	choice of spatial covariance function. See ?CovFunction
cellwidth	the width of computational cells
poisson.offset	A SpatialAtRisk object defining lambda (see below)
mcmc.control	MCMC paramters, see ?mcmcpars
output.control	output choice, see ?setoutput
gradtrunc	truncation for gradient vector equal to H parameter Moller et al 1998 pp 473. Default is Inf, which means no gradient truncation, which seems to work in most settings.
ext	integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.
Nfreq	the sampling frequency for the cell counts. Default is every 101 iterations.
inclusion	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former, the default, includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.
overlapping	logical does spdf contain overlapping polygons? Default is FALSE. If set to TRUE, spdf can contain a variable named 'sintens' that gives the sampling intensity for each polygon; the default is to assume that cases are evenly split between overlapping regions.

`pixwts` optional matrix of dimension (NM) x (number of regions in spdf) where M, N are the number of cells in the x and y directions (not the number of cells on the Fourier grid, rather the number of cell on the output grid). The ith row of this matrix are the probabilities that for the ith grid cell (in the same order as expand.grid(mcens,ncens)) a case belongs to each of the regions in spdf. Including this object overrides 'sintens' in the overlapping option above.

Details

See the vignette "Bayesian_lgcp" for examples of this code in use.

In this case, we OBSERVE case counts in the regions of a SpatialPolygonsDataFrame; the counts are stored as a variable, X. The model for the UNOBSERVED data, X(s), is as follows:

X(s) ~ Poisson[R(s)]

R(s) = C_A lambda(s) exp[Z(s)beta+Y(s)]

Here X(s) is the number of events in the cell of the computational grid containing s, R(s) is the Poisson rate, C_A is the cell area, lambda(s) is a known offset, Z(s) is a vector of measured covariates and Y(s) is the latent Gaussian process on the computational grid. The other parameters in the model are beta, the covariate effects; and eta=[log(sigma),log(phi)], the parameters of the process Y on an appropriately transformed (in this case log) scale.

We recommend the user takes the following steps before running this method:

1. Compute approximate values of the parameters, eta, of the process Y using the function minimum.contrast. These approximate values are used for two main reasons: (1) to help inform the size of the computational grid, since we will need to use a cell width that enables us to capture the dependence properties of Y and (2) to help inform the proposal kernel for the MCMC algorithm.
2. Choose an appropriate grid on which to perform inference using the function chooseCellwidth; this will partly be determined by the results of the first stage and partly by the available computational resource available to perform inference.
3. Using the function getpolyol, construct the computational grid and polygon overlays, as required. As this can be an expensive step, we recommend that the user saves this object after it has been constructed and in future reference to the data, reloads this object, rather than having to re-compute it (provided the computational grid has not changed).
4. Decide on which covariates are to play a part in the analysis and use the lgcp function getZmat to interpolate these onto the computational grid. Note that having saved the results from the previous step, this is a relatively quick operation, and allows the user to quickly construct different design matrices, Z, from different candidate models for the data
5. If required, set up the population offset using SpatialAtRisk functions (see the vignette "Bayesian_lgcp"); specify the priors using lgcpPrior; and if desired, the initial values for the MCMC, using the function lgcpInits.

6. Run the MCMC algorithm and save the output to disk. We recommend dumping information to disk using the dump2dir function in the output.control argument because it offers much greater flexibility in terms of MCMC diagnosis and post-processing.
7. Perform post-processing analyses including MCMC diagnostic checks and produce summaries of the posterior expectations we require for presentation. (see the vignette "Bayesian_lgcp" for further details). Functions of use in this step include traceplots, autocorr, parautocorr, ltar, parsummary, priorpost, postcov, textsummary, expectation, exceedProbs and lgcp:::expectation.lgcpPredict

Value

an object of class lgcpPredictAggregateSpatialPlusParameters

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle. Bayesian Inference and Data Augmentation Schemes for Spatial, Spatiotemporal and Multivariate Log-Gaussian Cox Processes in R. Submitted.
2. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
3. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
4. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
5. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
6. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

See Also

linkchooseCellWidth, getpolyol, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

lgcpPredictMultitypeSpatialPlusPars

lgcpPredictMultitypeSpatialPlusPars function

Description

A function to deliver fully Bayesian inference for a multitype spatial log-Gaussian Cox process.

Usage

```
lgcpPredictMultitypeSpatialPlusPars(
  formulaList,
  sd,
  typemark = NULL,
  Zmat = NULL,
  model.priorsList,
  model.initsList = NULL,
  spatial.covmodelList,
  cellwidth = NULL,
  poisson.offset = NULL,
  mcmc.control,
  output.control = setoutput(),
  gradtrunc = Inf,
  ext = 2,
  inclusion = "touching"
)
```

Arguments

`formulaList` an object of class formulaList, see ?formulaList. A list of formulae of the form t1 ~ var1 + var2 etc. The name of the dependent variable must correspond to the name of the point type. Only accepts 'simple' formulae, such as the example given.

`sd` a marked ppp object, the mark of interest must be able to be coerced to a factor variable

`typemark` if there are multiple marks, thrun the MCMC algorithm for spatial point process data. Not for general purpose use.is sets the name of the mark by which

`Zmat` design matrix including all covariate effects from each point type, constructed with getZmat

`model.priorsList` model priors, a list object of length the number of types, each element set using lgcpPrior

`model.initsList` list of model initial values (of length the number of types). The default is NULL, in which case lgcp will use the prior mean to initialise eta and beta will be initialised from an oversispersed glm fit to the data. Otherwise use lgcpInits to specify.

`spatial.covmodelList` list of spatial covariance functions (of length the number of types). See ?CovFunction

`cellwidth` the width of computational cells

`poisson.offset` A list of SpatialAtRisk objects (of length the number of types) defining lambda_k (see below)

`mcmc.control` MCMC paramters, see ?mcmcpars

`output.control` output choice, see ?setoutput `gradtrunc` truncation for gradient vector equal to H parameter Moller et al 1998 pp 473. Default is Inf, which means no gradient truncation, which seems to work in most settings.

`ext` integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.

`inclusion` criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former, the default, includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Details

See the vignette "Bayesian_lgcp" for examples of this code in use.

We suppose there are K point types of interest. The model for point-type k is as follows:

X_k(s) ~ Poisson[R_k(s)]

R_k(s) = C_A lambda_k(s) exp[Z_k(s)beta_k+Y_k(s)]

Here X_k(s) is the number of events of type k in the computational grid cell containing the point s, R_k(s) is the Poisson rate, C_A is the cell area, lambda_k(s) is a known offset, Z_k(s) is a vector of measured covariates and Y_i(s) where i = 1,...,K+1 are latent Gaussian processes on the computational grid. The other parameters in the model are beta_k , the covariate effects for the kth type; and eta_i = [log(sigma_i),log(phi_i)], the parameters of the process Y_i for i = 1,...,K+1 on an appropriately transformed (again, in this case log) scale.

We recommend the user takes the following steps before running this method:

1. Compute approximate values of the parameters, eta, of the process Y using the function minimum.contrast. These approximate values are used for two main reasons: (1) to help inform the size of the computational grid, since we will need to use a cell width that enables us to capture the dependence properties of Y and (2) to help inform the proposal kernel for the MCMC algorithm.
2. Choose an appropriate grid on which to perform inference using the function chooseCellwidth; this will partly be determined by the results of the first stage and partly by the available computational resource available to perform inference.
3. Using the function getpolyol, construct the computational grid and polygon overlays, as required. As this can be an expensive step, we recommend that the user saves this object after it has been constructed and in future reference to the data, reloads this object, rather than having to re-compute it (provided the computational grid has not changed).
4. Decide on which covariates are to play a part in the analysis and use the lgcp function getZmat to interpolate these onto the computational grid. Note that having saved the results from the previous step, this is a relatively quick operation, and allows the user to quickly construct different design matrices, Z, from different candidate models for the data

5. If required, set up the population offset using SpatialAtRisk functions (see the vignette "Bayesian_lgcp"); specify the priors using lgcpPrior; and if desired, the initial values for the MCMC, using the function lgcpInits.
6. Run the MCMC algorithm and save the output to disk. We recommend dumping information to disk using the dump2dir function in the output.control argument because it offers much greater flexibility in terms of MCMC diagnosis and post-processing.
7. Perform post-processing analyses including MCMC diagnostic checks and produce summaries of the posterior expectations we require for presentation. (see the vignette "Bayesian_lgcp" for further details). Functions of use in this step include traceplots, autocorr, parautocorr, ltar, parsummary, priorpost, postcov, textsummary, expectation, exceedProbs and lgcp:::expectation.lgcpPredict

Value

an object of class lgcpPredictMultitypeSpatialPlusParameters

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle. Bayesian Inference and Data Augmentation Schemes for Spatial, Spatiotemporal and Multivariate Log-Gaussian Cox Processes in R. Submitted.
2. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
3. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
4. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
5. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
6. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

See Also

linkchooseCellWidth, getpolyol, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

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`lgcpPredictSpatial` *lgcpPredictSpatial function*

Description

The function `lgcpPredictSpatial` performs spatial prediction for log-Gaussian Cox Processes

Usage

```
lgcpPredictSpatial(
  sd,
  model.parameters = lgcppars(),
  spatial.covmodel = "exponential",
  covpars = c(),
  cellwidth = NULL,
  gridsize = NULL,
  spatial.intensity,
  spatial.offset = NULL,
  mcmc.control,
  output.control = setoutput(),
  gradtrunc = Inf,
  ext = 2,
  inclusion = "touching"
)
```

Arguments

`sd` a spatial point pattern object, see ?ppp

`model.parameters` values for parameters, see ?lgcppars

`spatial.covmodel` correlation type see ?CovarianceFct

`covpars` vector of additional parameters for certain classes of covariance function (eg Matern), these must be supplied in the order given in ?CovarianceFct

`cellwidth` width of grid cells on which to do MALA (grid cells are square) in same units as observation window. Note EITHER gridsize OR cellwidthe must be specified.

`gridsize` size of output grid required. Note EITHER gridsize OR cellwidthe must be specified.

`spatial.intensity` the fixed spatial component: an object of that can be coerced to one of class spatialAtRisk

`spatial.offset` Numeric of length 1. Optional offset parameter, corresponding to the expected number of cases. NULL by default, in which case, this is estimateed from teh data.

`mcmc.control` MCMC paramters, see ?mcmcpars

`output.control` output choice, see ?setoutput

`gradtrunc` truncation for gradient vector equal to H parameter Moller et al 1998 pp 473. Default is Inf, which means no gradient truncation. Set to NULL to estimate this automatically (though note that this may not necessarily be a good choice). The default seems to work in most settings.

`ext` integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.

`inclusion` criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former, the default, includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Details

The following is a mathematical description of a log-Gaussian Cox Process, it is best viewed in the pdf version of the manual.

Let $\mathcal{Y}(s)$ be a spatial Gaussian process and $W \subset R^2$ be an observation window in space. Cases occur at spatial positions $x \in W$ according to an inhomogeneous spatial Cox process, i.e. a Poisson process with a stochastic intensity $R(x)$, The number of cases, X_S, arising in any $S \subseteq W$ is then Poisson distributed conditional on $R(\cdot)$,

$$X_S \sim \text{Poisson}\left\{\int_S R(s)ds\right\}$$

Following Brix and Diggle (2001) and Diggle et al (2005) (but ignoring temporal variation), the intensity is decomposed multiplicatively as

$$R(s,t) = \lambda(s)\exp\{\mathcal{Y}(s,t)\}.$$

In the above, the fixed spatial component, $\lambda : R^2 \mapsto R_{\geq 0}$, is a known function, proportional to the population at risk at each point in space and scaled so that

$$\int_W \lambda(s)ds = 1.$$

Before calling this function, the user must decide on the parameters, spatial covariance model, spatial discretisation, fixed spatial ($\lambda(s)$) component, mcmc parameters, and whether or not any output is required. Note there is no autorotate option for this function.

Value

the results of fitting the model in an object of class `lgcpPredict`

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
3. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
4. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
5. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

See Also

lgcpPredict KinhomAverage, ginhomAverage, lambdaEst, muEst, spatialparsEst, thetaEst, spatialAtRisk, temporalAtRisk, lgcppars, CovarianceFct, mcmcpars, setoutput print.lgcpPredict, xvals.lgcpPredict, yvals.lgcpPredict, plot.lgcpPredict, meanfield.lgcpPredict, rr.lgcpPredict, serr.lgcpPredict, intens.lgcpPredict, varfield.lgcpPredict, gridfun.lgcpPredict, gridav.lgcpPredict, hvals.lgcpPredict, window.lgcpPredict, mcmctrace.lgcpPredict, plotExceed.lgcpPredict, quantile.lgcpPredict, identify.lgcpPredict, expectation.lgcpPredict, extract.lgcpPredict, showGrid.lgcpPredict

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`lgcpPredictSpatialINLA` *lgcpPredictSpatialINLA function*

Description

—————————————————————- !IMPORTANT! after library(lgcp) this will be a dummy function. In order to use, type getlgcpPredictSpatialINLA() at the console. This will download and install the true function. —————————————————————-

Usage

```
lgcpPredictSpatialINLA(
  sd,
  ns,
  model.parameters = lgcppars(),
  spatial.covmodel = "exponential",
  covpars = c(),
  cellwidth = NULL,
  gridsize = NULL,
  spatial.intensity,
  ext = 2,
  optimverbose = FALSE,
  inlaverbose = TRUE,
  generic0hyper = list(theta = list(initial = 0, fixed = TRUE)),
  strategy = "simplified.laplace",
  method = "Nelder-Mead"
)
```

Arguments

`sd` a spatial point pattern object, see ?ppp

`ns` size of neighbourhood to use for GMRF approximation ns=1 corresponds to 3^2-1=8 eight neighbours around each point, ns=2 corresponds to 5^2-1=24 neighbours etc ...

`model.parameters`
values for parameters, see ?lgcppars

`spatial.covmodel`
: correlation type see ?CovarianceFct

`covpars`
: vector of additional parameters for certain classes of covariance function (eg Matern), these must be supplied in the order given in ?CovarianceFct

`cellwidth`
: width of grid cells on which to do MALA (grid cells are square). Note EITHER gridsize OR cellwidthe must be specified.

`gridsize`
: size of output grid required. Note EITHER gridsize OR cellwidthe must be specified.

`spatial.intensity`
: the fixed spatial component: an object of that can be coerced to one of class spatialAtRisk

`ext`
: integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.

`optimverbose`
: logical whether to print optimisation details of covariance matching step

`inlaverbose`
: loogical whether to print the inla fitting procedure to the console

`generic0hyper`
: optional hyperparameter list specification for "generic0" INLA model. default is list(theta=list(initial=0,fixed=TRUE)), which effectively treats the precision matrix as known.

`strategy`
: inla strategy

`method`
: optimisation method to be used in function matchcovariance, default is "Nelder-Mead". See ?matchcovariance

Details

The function `lgcpPredictSpatialINLA` performs spatial prediction for log-Gaussian Cox Processes using the integrated nested Laplace approximation.

The following is a mathematical description of a log-Gaussian Cox Process, it is best viewed in the pdf version of the manual.

Let $\mathcal{Y}(s)$ be a spatial Gaussian process and $W \subset R^2$ be an observation window in space. Cases occur at spatial positions $x \in W$ according to an inhomogeneous spatial Cox process, i.e. a Poisson process with a stochastic intensity $R(x)$, The number of cases, X_S, arising in any $S \subseteq W$ is then Poisson distributed conditional on $R(\cdot)$,

$$X_S \sim \text{Poisson}\left\{\int_S R(s)ds\right\}$$

Following Brix and Diggle (2001) and Diggle et al (2005) (but ignoring temporal variation), the intensity is decomposed multiplicatively as

$$R(s,t) = \lambda(s)\exp\{\mathcal{Y}(s,t)\}.$$

In the above, the fixed spatial component, $\lambda : R^2 \mapsto R_{\geq 0}$, is a known function, proportional to the population at risk at each point in space and scaled so that

$$\int_W \lambda(s)ds = 1.$$

Before calling this function, the user must decide on the parameters, spatial covariance model, spatial discretisation, fixed spatial ($\lambda(s)$) component and whether or not any output is required. Note there is no autorotate option for this function.

Value

the results of fitting the model in an object of class `lgcpPredict`

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
3. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
4. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
5. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

See Also

lgcpPredict KinhomAverage, ginhomAverage, lambdaEst, muEst, spatialparsEst, thetaEst, spatialAtRisk, temporalAtRisk, lgcppars, CovarianceFct, mcmcpars, setoutput print.lgcpPredict, xvals.lgcpPredict, yvals.lgcpPredict, plot.lgcpPredict, meanfield.lgcpPredict, rr.lgcpPredict, serr.lgcpPredict, intens.lgcpPredict, varfield.lgcpPredict, gridfun.lgcpPredict, gridav.lgcpPredict, hvals.lgcpPredict, window.lgcpPredict, mcmctrace.lgcpPredict, plotExceed.lgcpPredict, quantile.lgcpPredict, identify.lgcpPredict, expectation.lgcpPredict, extract.lgcpPredict, showGrid.lgcpPredict,

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`lgcpPredictSpatialPlusPars`

lgcpPredictSpatialPlusPars function

Description

A function to deliver fully Bayesian inference for the spatial log-Gaussian Cox process.

Usage

```
lgcpPredictSpatialPlusPars(
  formula,
  sd,
  Zmat = NULL,
  model.priors,
```

```
    model.inits = lgcpInits(),
    spatial.covmodel,
    cellwidth = NULL,
    poisson.offset = NULL,
    mcmc.control,
    output.control = setoutput(),
    gradtrunc = Inf,
    ext = 2,
    inclusion = "touching"
)
```

Arguments

`formula`	a formula object of the form X ~ var1 + var2 etc. The name of the dependent variable must be "X". Only accepts 'simple' formulae, such as the example given.
`sd`	a spatstat ppp object
`Zmat`	design matrix Z (see below) constructed with getZmat
`model.priors`	model priors, set using lgcpPrior
`model.inits`	model initial values. The default is NULL, in which case lgcp will use the prior mean to initialise eta and beta will be initialised from an oversispersed glm fit to the data. Otherwise use lgcpInits to specify.
`spatial.covmodel`	choice of spatial covariance function. See ?CovFunction
`cellwidth`	the width of computational cells
`poisson.offset`	A SpatialAtRisk object defining lambda (see below)
`mcmc.control`	MCMC paramters, see ?mcmcpars
`output.control`	output choice, see ?setoutput
`gradtrunc`	truncation for gradient vector equal to H parameter Moller et al 1998 pp 473. Default is Inf, which means no gradient truncation, which seems to work in most settings.
`ext`	integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.
`inclusion`	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former, the default, includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Details

See the vignette "Bayesian_lgcp" for examples of this code in use.

The model for the data is as follows:

X(s) ~ Poisson[R(s)]

R(s) = C_A lambda(s) exp[Z(s)beta+Y(s)]

Here X(s) is the number of events in the cell of the computational grid containing s, R(s) is the Poisson rate, C_A is the cell area, lambda(s) is a known offset, Z(s) is a vector of measured covariates and Y(s) is the latent Gaussian process on the computational grid. The other parameters in the model are beta, the covariate effects; and eta=[log(sigma),log(phi)], the parameters of the process Y on an appropriately transformed (in this case log) scale.

We recommend the user takes the following steps before running this method:

1. Compute approximate values of the parameters, eta, of the process Y using the function minimum.contrast. These approximate values are used for two main reasons: (1) to help inform the size of the computational grid, since we will need to use a cell width that enables us to capture the dependence properties of Y and (2) to help inform the proposal kernel for the MCMC algorithm.
2. Choose an appropriate grid on which to perform inference using the function chooseCellwidth; this will partly be determined by the results of the first stage and partly by the available computational resource available to perform inference.
3. Using the function getpolyol, construct the computational grid and polygon overlays, as required. As this can be an expensive step, we recommend that the user saves this object after it has been constructed and in future reference to the data, reloads this object, rather than having to re-compute it (provided the computational grid has not changed).
4. Decide on which covariates are to play a part in the analysis and use the lgcp function getZmat to interpolate these onto the computational grid. Note that having saved the results from the previous step, this is a relatively quick operation, and allows the user to quickly construct different design matrices, Z, from different candidate models for the data
5. If required, set up the population offset using SpatialAtRisk functions (see the vignette "Bayesian_lgcp"); specify the priors using lgcpPrior; and if desired, the initial values for the MCMC, using the function lgcpInits.
6. Run the MCMC algorithm and save the output to disk. We recommend dumping information to disk using the dump2dir function in the output.control argument because it offers much greater flexibility in terms of MCMC diagnosis and post-processing.
7. Perform post-processing analyses including MCMC diagnostic checks and produce summaries of the posterior expectations we require for presentation. (see the vignette "Bayesian_lgcp" for further details). Functions of use in this step include traceplots, autocorr, parautocorr, ltar, parsummary, priorpost, postcov, textsummary, expectation, exceedProbs and lgcp:::expectation.lgcpPredict

Value

an object of class lgcpPredictSpatialOnlyPlusParameters

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle. Bayesian Inference and Data Augmentation Schemes for Spatial, Spatiotemporal and Multivariate Log-Gaussian Cox Processes in R. Submitted.

2. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
3. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
4. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
5. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
6. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

See Also

linkchooseCellWidth, getpolyol, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

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lgcpPredictSpatioTemporalPlusPars

lgcpPredictSpatioTemporalPlusPars function

Description

A function to deliver fully Bayesian inference for the spatiotemporal log-Gaussian Cox process.

Usage

```
lgcpPredictSpatioTemporalPlusPars(
  formula,
  xyt,
  T,
  laglength,
  ZmatList = NULL,
  model.priors,
  model.inits = lgcpInits(),
  spatial.covmodel,
  cellwidth = NULL,
  poisson.offset = NULL,
  mcmc.control,
  output.control = setoutput(),
  gradtrunc = Inf,
  ext = 2,
  inclusion = "touching"
)
```

Arguments

`formula`	a formula object of the form X ~ var1 + var2 etc. The name of the dependent variable must be "X". Only accepts 'simple' formulae, such as the example given.
`xyt`	An object of class stppp
`T`	the time point of interest
`laglength`	the number of previous time points to include in the analysis
`ZmatList`	A list of design matrices Z constructed with getZmat and possibly addTemporal-Covariates see the details below and Bayesian_lgcp vignette for details on how to construct this.
`model.priors`	model priors, set using lgcpPrior
`model.inits`	model initial values. The default is NULL, in which case lgcp will use the prior mean to initialise eta and beta will be initialised from an oversispersed glm fit to the data. Otherwise use lgcpInits to specify.
`spatial.covmodel`	choice of spatial covariance function. See ?CovFunction
`cellwidth`	the width of computational cells
`poisson.offset`	A list of SpatialAtRisk objects (of length the number of types) defining lambda_k (see below)
`mcmc.control`	MCMC paramters, see ?mcmcpars
`output.control`	output choice, see ?setoutput
`gradtrunc`	truncation for gradient vector equal to H parameter Moller et al 1998 pp 473. Default is Inf, which means no gradient truncation, which seems to work in most settings.
`ext`	integer multiple by which grid should be extended, default is 2. Generally this will not need to be altered, but if the spatial correlation decays slowly, increasing 'ext' may be necessary.
`inclusion`	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former, the default, includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Details

See the vignette "Bayesian_lgcp" for examples of this code in use.

The model for the data is as follows:

X(s) ~ Poisson[R(s,t)]

R(s) = C_A lambda(s,t) exp[Z(s,t)beta+Y(s,t)]

Here X(s,t) is the number of events in the cell of the computational grid containing s, R(s,t) is the Poisson rate, C_A is the cell area, lambda(s,t) is a known offset, Z(s,t) is a vector of measured covariates and Y(s,t) is the latent Gaussian process on the computational grid. The other parameters in the model are beta, the covariate effects; and eta=[log(sigma),log(phi),log(theta)], the parameters of the process Y on an appropriately transformed (in this case log) scale.

We recommend the user takes the following steps before running this method:

1. Compute approximate values of the parameters, eta, of the process Y using the function minimum.contrast. These approximate values are used for two main reasons: (1) to help inform the size of the computational grid, since we will need to use a cell width that enables us to capture the dependence properties of Y and (2) to help inform the proposal kernel for the MCMC algorithm.
2. Choose an appropriate grid on which to perform inference using the function chooseCellwidth; this will partly be determined by the results of the first stage and partly by the available computational resource available to perform inference.
3. Using the function getpolyol, construct the computational grid and polygon overlays, as required. As this can be an expensive step, we recommend that the user saves this object after it has been constructed and in future reference to the data, reloads this object, rather than having to re-compute it (provided the computational grid has not changed).
4. Decide on which covariates are to play a part in the analysis and use the lgcp function getZmat to interpolate these onto the computational grid. Note that having saved the results from the previous step, this is a relatively quick operation, and allows the user to quickly construct different design matrices, Z, from different candidate models for the data
5. If required, set up the population offset using SpatialAtRisk functions (see the vignette "Bayesian_lgcp"); specify the priors using lgcpPrior; and if desired, the initial values for the MCMC, using the function lgcpInits.
6. Run the MCMC algorithm and save the output to disk. We recommend dumping information to disk using the dump2dir function in the output.control argument because it offers much greater flexibility in terms of MCMC diagnosis and post-processing.
7. Perform post-processing analyses including MCMC diagnostic checks and produce summaries of the posterior expectations we require for presentation. (see the vignette "Bayesian_lgcp" for further details). Functions of use in this step include traceplots, autocorr, parautocorr, ltar, parsummary, priorpost, postcov, textsummary, expectation, exceedProbs and lgcp:::expectation.lgcpPredict

The user must provide a list of design matrices to use this function. In the interpolation step above, there are three cases to consider

1. where Z(s,t) cannot be decomposed, i.e., Z are true spatiotemporal covariates. In this case, each element of the list must be constructed separately using the function getZmat on the covariates for each time point.
2. Z(s,t)beta = Z_1(s)beta_1 + Z_2(t)beta_2: the spatial and temporal effects are separable; in this case use the function addTemporalCovariates, to aid in the construction of the list.
3. Z(s,t)beta = Z(s)beta, in which case the user only needs to perform the interpolation using getZmat once, each of the elements of the list will then be identical.

4. Z(s,t)beta = Z(t)beta in this case we follow the procedure for the separable case above. For example, if dotw is a temporal covariate we would use formula <- X ~ dotw for the main algorithm, formula.spatial <- X ~ 1 to interpolate the spatial covariates using getZmat, followed by temporal.formula <- t ~ dotw - 1 using addTemporalCovariates to construct the list of design matrices, Zmat.

Value

an object of class lgcpPredictSpatioTemporalPlusParameters

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle. Bayesian Inference and Data Augmentation Schemes for Spatial, Spatiotemporal and Multivariate Log-Gaussian Cox Processes in R. Submitted.
2. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
3. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
4. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
5. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
6. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

See Also

linkchooseCellWidth, getpolyol, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictMultitypeSpatialPlusPars, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

`lgcpPrior` *lgcpPrior function*

Description

A function to create the prior for beta and eta ready for a run of the MCMC algorithm.

Usage

```
lgcpPrior(etaprior = NULL, betaprior = NULL)
```

Arguments

etaprior
: an object of class PriorSpec defining the prior for the parameters of the latent field, eta. See ?PriorSpec.list.

betaprior
: etaprior an object of class PriorSpec defining the prior for the parameters of main effects, beta. See ?PriorSpec.list.

Value

an R structure representing the prior density ready for a run of the MCMC algorithm.

See Also

GaussianPrior, LogGaussianPrior, PriorSpec.list, chooseCellwidth, getpolyol, guessinterp, getZmat, addTemporalCovariates, lgcpPrior, lgcpInits, CovFunction lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars, lgcpPredictMultitypeSpatialPlusPars

Examples

```
lgcpPrior(etaprior=PriorSpec(LogGaussianPrior(mean=log(c(1,500)),
    variance=diag(0.15,2))),betaprior=PriorSpec(GaussianPrior(mean=rep(0,9),
    variance=diag(10^6,9))))
```

---

lgcpSim *lgcpSim function*

Description

Approximate simulation from a spatiotemoporal log-Gaussian Cox Process. Returns an stppp object.

Usage

```
lgcpSim(
  owin = NULL,
  tlim = as.integer(c(0, 10)),
  spatial.intensity = NULL,
  temporal.intensity = NULL,
  cellwidth = 0.05,
  model.parameters = lgcppars(sigma = 2, phi = 0.2, theta = 1),
  spatial.covmodel = "exponential",
  covpars = c(),
  returnintensities = FALSE,
  progressbar = TRUE,
  ext = 2,
  plot = FALSE,
  ratepow = 0.25,
```

```
  sleeptime = 0,
  inclusion = "touching"
)
```

Arguments

owin	polygonal observation window
tlim	time interval on which to simulate data
spatial.intensity	object that can be coerced into a spatialAtRisk object. if NULL then uniform spatial is chosen
temporal.intensity	the fixed temporal component: either a numeric vector, or a function that can be coerced into an object of class temporalAtRisk
cellwidth	width of cells in same units as observation window
model.parameters	parameters of model, see ?lgcppars.
spatial.covmodel	spatial covariance function, default is exponential, see ?CovarianceFct
covpars	vector of additional parameters for spatial covariance function, in order they appear in chosen model in ?CovarianceFct
returnintensities	logigal, whether to return the spatial intensities and true field Y at each time. Default FALSE.
progressbar	logical, whether to print a progress bar. Default TRUE.
ext	how much to extend the parameter space by. Default is 2.
plot	logical, whether to plot intensities.
ratepow	power that intensity is raised to for plotting purposes (makes the plot more pleasign to the eye), defaul 0.25
sleeptime	time in seconds to sleep between plots
inclusion	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Details

The following is a mathematical description of a log-Gaussian Cox Process, it is best viewed in the pdf version of the manual.

Let $\mathcal{Y}(s,t)$ be a spatiotemporal Gaussian process, $W \subset R^2$ be an observation window in space and $T \subset R_{\geq 0}$ be an interval of time of interest. Cases occur at spatio-temporal positions $(x,t) \in W \times T$ according to an inhomogeneous spatio-temporal Cox process, i.e. a Poisson process with a stochastic intensity $R(x,t)$, The number of cases, $X_{S,[t_1,t_2]}$, arising in any $S \subseteq W$ during the interval $[t_1,t_2] \subseteq T$ is then Poisson distributed conditional on $R(\cdot)$,

$$X_{S,[t_1,t_2]} \sim \text{Poisson}\left\{\int_S \int_{t_1}^{t_2} R(s,t)dsdt\right\}$$

Following Brix and Diggle (2001) and Diggle et al (2005), the intensity is decomposed multiplicatively as

$$R(s,t) = \lambda(s)\mu(t)\exp\{\mathcal{Y}(s,t)\}.$$

In the above, the fixed spatial component, $\lambda : R^2 \mapsto R_{\geq 0}$, is a known function, proportional to the population at risk at each point in space and scaled so that

$$\int_W \lambda(s)ds = 1,$$

whilst the fixed temporal component, $\mu : R_{\geq 0} \mapsto R_{\geq 0}$, is also a known function with

$$\mu(t)\delta t = E[X_{W,\delta t}],$$

for t in a small interval of time, δt, over which the rate of the process over W can be considered constant.

Value

an stppp object containing the data

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
3. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.
4. Wood ATA, Chan G (1994). Simulation of Stationary Gaussian Processes in [0,1]d. Journal of Computational and Graphical Statistics, 3(4), 409-432.
5. Moller J, Syversveen AR, Waagepetersen RP (1998). Log Gaussian Cox Processes. Scandinavian Journal of Statistics, 25(3), 451-482.

See Also

lgcpPredict, showGrid.stppp, stppp

Examples

```
## Not run: library(spatstat.explore); library(spatstat.utils); xyt <- lgcpSim()
```

lgcpSimMultitypeSpatialCovariates

lgcpSimMultitypeSpatialCovariates function

Description

A function to Simulate multivariate point process models

Usage

```
lgcpSimMultitypeSpatialCovariates(
  formulaList,
  owin,
  regionalcovariates,
  pixelcovariates,
  betaList,
  spatial.offsetList = NULL,
  cellwidth,
  model.parameters,
  spatial.covmodel = "exponential",
  covpars = c(),
  ext = 2,
  plot = FALSE,
  inclusion = "touching"
)
```

Arguments

- `formulaList` a list of formulae objetcs
- `owin` a spatstat owin object on which to simulate the data
- `regionalcovariates` a SpatialPolygonsDataFrame object
- `pixelcovariates` a SpatialPixelsDataFrame object
- `betaList` list of beta parameters
- `spatial.offsetList` list of poisson offsets
- `cellwidth` cellwidth
- `model.parameters` model parameters, a list eg list(sigma=1,phi=0.2)
- `spatial.covmodel` the choice of spatial covariance model, can be anything from the RandomFields covariance function, CovariacenFct.
- `covpars` additional covariance parameters, for the chosen model, optional.
- `ext` number of times to extend the simulation window

`plot`	whether to plot the results automatically
`inclusion`	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former, the default, includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Value

a marked ppp object, the simulated data

`lgcpSimSpatial` *lgcpSimSpatial function*

Description

A function to simulate from a log gaussian process

Usage

```
lgcpSimSpatial(
  owin = NULL,
  spatial.intensity = NULL,
  expectednumcases = 100,
  cellwidth = 0.05,
  model.parameters = lgcppars(sigma = 2, phi = 0.2),
  spatial.covmodel = "exponential",
  covpars = c(),
  ext = 2,
  plot = FALSE,
  inclusion = "touching"
)
```

Arguments

`owin`	observation window
`spatial.intensity`	an object that can be coerced to one of class spatialAtRisk
`expectednumcases`	the expected number of cases
`cellwidth`	width of cells in same units as observation window
`model.parameters`	parameters of model, see ?lgcppars. Only set sigma and phi for spatial model.
`spatial.covmodel`	spatial covariance function, default is exponential, see ?CovarianceFct
`covpars`	vector of additional parameters for spatial covariance function, in order they appear in chosen model in ?CovarianceFct

`ext`	how much to extend the parameter space by. Default is 2.
`plot`	logical, whether to plot the latent field.
`inclusion`	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Value

a ppp object containing the data

`lgcpSimSpatialCovariates` *lgcpSimSpatialCovariates function*

Description

A function to simulate a spatial LGCP.

Usage

```
lgcpSimSpatialCovariates(
  formula,
  owin,
  regionalcovariates = NULL,
  pixelcovariates = NULL,
  Zmat = NULL,
  beta,
  poisson.offset = NULL,
  cellwidth,
  model.parameters,
  spatial.covmodel = "exponential",
  covpars = c(),
  ext = 2,
  plot = FALSE,
  inclusion = "touching"
)
```

Arguments

`formula`	a formula of the form X ~ var1 + var2 etc.
`owin`	the observation window on which to do the simulation
`regionalcovariates`	an optional object of class SpatialPolygonsDataFrame containing covariates
`pixelcovariates`	an optional object of class SpatialPixelsDataFrame containing covariates

Zmat	optional design matrix, if the polygon/polygon overlays have already been computed
beta	the parameters, beta for the model
poisson.offset	the poisson offet, created using a SpatialAtRisk.fromXYZ class of objects
cellwidth	the with of cells on which to do the simulation
model.parameters	the paramters of the model eg list(sigma=1,phi=0.2)
spatial.covmodel	the choice of spatial covariance model, can be anything from the RandomFields covariance function, CovariacenFct.
covpars	additional covariance parameters, for the chosen model, optional.
ext	the amount by which to extend the observation grid in each direction, default is 2
plot	whether to plot the resulting data
inclusion	criterion for cells being included into observation window. Either 'touching' or 'centroid'. The former, the default, includes all cells that touch the observation window, the latter includes all cells whose centroids are inside the observation window.

Value

a ppp onject containing the simulated data

`lgcpvignette` *lgcpvignette function*

Description

Display the introductory vignette for the lgcp package.

Usage

```
lgcpvignette()
```

Value

displays the vignette by calling browseURL `loc2poly` *loc2poly function*

Description

Converts a polygon selected via the mouse in a graphics window into an polygonal owin object. (Make sure the x and y scales are correct!) Points must be selected traversing the required window in one direction (ie either clockwise, or anticlockwise), points must not be overlapping. Select the sequence of edges via left mouse button clicks and store the polygon with a right click.

Usage

```
loc2poly(n = 512, type = "l", col = "black", ...)
```

Arguments

`n`	the maximum number of points to locate
`type`	same as argument type in function locator. see ?locator. Default draws lines
`col`	colour of lines/points
`...`	other arguments to pass to locate

Value

a polygonal owin object

See Also

lgcpPredict, identify.lgcpPredict

Examples

```
## Not run: plot(lg) # lg an lgcpPredict object
## Not run: subwin <- loc2poly())
```

`LogGaussianPrior` *LogGaussianPrior function*

Description

A function to create a Gaussian prior on the log scale

Usage

```
LogGaussianPrior(mean, variance)
```

Arguments

`mean`	a vector of length 2 representing the mean (on the log scale)
`variance`	a 2x2 matrix representing the variance (on the log scale)

Value

an object of class LogGaussianPrior that can be passed to the function PriorSpec.

See Also

GaussianPrior, linkPriorSpec.list

Examples

```
## Not run: LogGaussianPrior(mean=log(c(1,500)),variance=diag(0.15,2))
```

`loop.mcmc` *loop over an iterator*

Description

useful for testing progress bars

Usage

```
loop.mcmc(object, sleep = 1)
```

Arguments

`object`	an mcmc iterator
`sleep`	pause between iterations in seconds

`ltar` *ltar function*

Description

A function to return the sampled log-target from a call to the function lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars or lgcpPredictMultitype-SpatialPlusPars. This is used as a convergence diagnostic.

Usage

```
ltar(lg)
```

Arguments

`lg` an object produced by a call to lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars orlgcpPredictMultitypeSpatialPlusPars

Value

the log-target from each saved iteration of the MCMC chain.

See Also

autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

MALAlgcp *MALAlgcp function*

Description

ADVANCED USE ONLY A function to perform MALA for the spatial only case

Usage

```
MALAlgcp(
  mcmcloop,
  inits,
  adaptivescheme,
  M,
  N,
  Mext,
  Next,
  sigma,
  phi,
  theta,
  mu,
  nis,
  cellarea,
  spatialvals,
  temporal.fitted,
  tdiff,
  scaleconst,
  rootQeigs,
  invrootQeigs,
  cellInside,
  MCMCdiag,
  gradtrunc,
  gridfun,
```

```
    gridav,
    mcens,
    ncens,
    aggtimes
)
```

Arguments

mcmcloop	an mcmcLoop object
inits	initial values from mcmc.control
adaptivescheme	adaptive scheme from mcmc.control
M	number of cells in x direction on output grid
N	number of cells in y direction on output grid
Mext	number of cells in x direction on extended output grid
Next	number of cells in y direction on extended output grid
sigma	spatial covariance parameter sigma
phi	spatial covariance parameter phi
theta	temporal correlation parameter theta
mu	spatial covariance parameter mu
nis	cell counts matrix
cellarea	area of cells
spatialvals	spatial at risk, function lambda, interpolated onto the requisite grid
temporal.fitted	temporal fitted values representing mu(t)
tdiff	vecto of time differences with convention that the first element is Inf
scaleconst	expected number of observations
rootQeigs	square root of eigenvalues of precision matrix
invrootQeigs	inverse square root of eigenvalues of precision matrix
cellInside	logical matrix dictating whether cells are inside the observation window
MCMCdiag	defunct
gradtrunc	gradient truncation parameter
gridfun	grid functions
gridav	grid average functions
mcens	x-coordinates of cell centroids
ncens	y-coordinates of cell centroids
aggtimes	z-coordinates of cell centroids (ie time)

Value

object passed back to lgcpPredictSpatial

MALAlgcpAggregateSpatial.PlusPars
MALAlgcpAggregateSpatial.PlusPars function

Description

A function to run the MCMC algorithm for aggregated spatial point process data. Not for general purpose use.

Usage

```
MALAlgcpAggregateSpatial.PlusPars(
  mcmcloop,
  inits,
  adaptivescheme,
  M,
  N,
  Mext,
  Next,
  mcens,
  ncens,
  formula,
  Zmat,
  model.priors,
  model.inits,
  fftgrid,
  spatial.covmodel,
  nis,
  cellarea,
  spatialvals,
  cellInside,
  MCMCdiag,
  gradtrunc,
  gridfun,
  gridav,
  d,
  spdf,
  ol,
  Nfreq
)
```

Arguments

mcmcloop	details of the mcmc loop
inits	initial values
adaptivescheme	the adaptive MCMC scheme

M	number of grid cells in x direction
N	number of grid cells in y direction
Mext	number of extended grid cells in x direction
Next	number of extended grid cells in y direction
mcens	centroids in x direction
ncens	centroids in y direction
formula	a formula object of the form X ~ var1 + var2 etc.
Zmat	design matrix constructed using getZmat
model.priors	model priors, constructed using lgcpPrior
model.inits	initial values for the MCMC
fftgrid	an objects of class FFTgrid, see genFFTgrid
spatial.covmodel	spatial covariance model, consructed with CovFunction
nis	cell counts on the etended grid
cellarea	the cell area
spatialvals	inerpolated poisson offset on fft grid
cellInside	0-1 matrix indicating inclusion in the observation window
MCMCdiag	not used
gradtrunc	gradient truncation parameter
gridfun	used to specify other actions to be taken, e.g. dumping MCMC output to disk.
gridav	used for computing Monte Carlo expectations online
d	matrix of toral distances
spdf	the SpatialPolygonsDataFrame containing the aggregate counts as a variable X
ol	overlay of fft grid onto spdf
Nfreq	frequency at which to resample nis

Value

output from the MCMC run

`MALAlgcpMultitypeSpatial.PlusPars`
MALAlgcpMultitypeSpatial.PlusPars function

Description

A function to run the MCMC algorithm for multivariate spatial point process data. Not for general purpose use.

Usage

```
MALAlgcpMultitypeSpatial.PlusPars(
  mcmcloop,
  inits,
  adaptivescheme,
  M,
  N,
  Mext,
  Next,
  mcens,
  ncens,
  formulaList,
  zml,
  Zmat,
  model.priorsList,
  model.initsList,
  fftgrid,
  spatial.covmodelList,
  nis,
  cellarea,
  spatialvals,
  cellInside,
  MCMCdiag,
  gradtrunc,
  gridfun,
  gridav,
  marks,
  ntypes,
  d
)
```

Arguments

mcmcloop	details of the mcmc loop
inits	initial values
adaptivescheme	the adaptive MCMC scheme
M	number of grid cells in x direction
N	number of grid cells in y direction
Mext	number of extended grid cells in x direction
Next	number of extended grid cells in y direction
mcens	centroids in x direction
ncens	centroids in y direction
formulaList	a list of formula objects of the form X ~ var1 + var2 etc.
zml	list of design matrices
Zmat	a design matrix constructed using getZmat

`model.priorsList`	list of model priors, see lgcpPriors
`model.initsList`	list of model initial values, see lgcpInits
`fftgrid`	an objects of class FFTgrid, see genFFTgrid
`spatial.covmodelList`	list of spatial covariance models constructed using CovFunction
`nis`	cell counts on the etended grid
`cellarea`	the cell area
`spatialvals`	inerpolated poisson offset on fft grid
`cellInside`	0-1 matrix indicating inclusion in the observation window
`MCMCdiag`	not used
`gradtrunc`	gradient truncation parameter
`gridfun`	used to specify other actions to be taken, e.g. dumping MCMC output to disk.
`gridav`	used for computing Monte Carlo expectations online
`marks`	the marks from the marked ppp object
`ntypes`	the number of types being analysed
`d`	matrix of toral distances

Value

output from the MCMC run

`MALAlgcpSpatial` *MALAlgcpSpatial function*

Description

ADVANCED USE ONLY A function to perform MALA for the spatial only case

Usage

```
MALAlgcpSpatial(
  mcmcloop,
  inits,
  adaptivescheme,
  M,
  N,
  Mext,
  Next,
  sigma,
  phi,
  mu,
```

```
    nis,
    cellarea,
    spatialvals,
    scaleconst,
    rootQeigs,
    invrootQeigs,
    cellInside,
    MCMCdiag,
    gradtrunc,
    gridfun,
    gridav,
    mcens,
    ncens
)
```

Arguments

mcmcloop	an mcmcLoop object
inits	initial values from mcmc.control
adaptivescheme	adaptive scheme from mcmc.control
M	number of cells in x direction on output grid
N	number of cells in y direction on output grid
Mext	number of cells in x direction on extended output grid
Next	number of cells in y direction on extended output grid
sigma	spatial covariance parameter sigma
phi	spatial covariance parameter phi
mu	spatial covariance parameter mu
nis	cell counts matrix
cellarea	area of cells
spatialvals	spatial at risk, function lambda, interpolated onto the requisite grid
scaleconst	expected number of observations
rootQeigs	square root of eigenvalues of precision matrix
invrootQeigs	inverse square root of eigenvalues of precision matrix
cellInside	logical matrix dictating whether cells are inside the observation window
MCMCdiag	defunct
gradtrunc	gradient truncation parameter
gridfun	grid functions
gridav	grid average functions
mcens	x-coordinates of cell centroids
ncens	y-coordinates of cell centroids

Value

object passed back to lgcpPredictSpatial

MALAlgcpSpatial.PlusPars

MALAlgcpSpatial.PlusPars function

Description

A function to run the MCMC algorithm for spatial point process data. Not for general purpose use.

Usage

```
MALAlgcpSpatial.PlusPars(
  mcmcloop,
  inits,
  adaptivescheme,
  M,
  N,
  Mext,
  Next,
  mcens,
  ncens,
  formula,
  Zmat,
  model.priors,
  model.inits,
  fftgrid,
  spatial.covmodel,
  nis,
  cellarea,
  spatialvals,
  cellInside,
  MCMCdiag,
  gradtrunc,
  gridfun,
  gridav,
  d
)
```

Arguments

`mcmcloop`	details of the mcmc loop
`inits`	initial values
`adaptivescheme`	the adaptive MCMC scheme
`M`	number of grid cells in x direction
`N`	number of grid cells in y direction
`Mext`	number of extended grid cells in x direction

Next	number of extended grid cells in y direction
mcens	centroids in x direction
ncens	centroids in y direction
formula	a formula object of the form X ~ var1 + var2 etc.
Zmat	design matrix constructed using getZmat
model.priors	model priors, constructed using lgcpPrior
model.inits	initial values for the MCMC
fftgrid	an objects of class FFTgrid, see genFFTgrid
spatial.covmodel	spatial covariance model, consructed with CovFunction
nis	cell counts on the etended grid
cellarea	the cell area
spatialvals	inerpolated poisson offset on fft grid
cellInside	0-1 matrix indicating inclusion in the observation window
MCMCdiag	not used
gradtrunc	gradient truncation parameter
gridfun	used to specify other actions to be taken, e.g. dumping MCMC output to disk.
gridav	used for computing Monte Carlo expectations online
d	matrix of toral distances

Value

output from the MCMC run

`MALAlgcpSpatioTemporal.PlusPars`
MALAlgcpSpatioTemporal.PlusPars function

Description

A function to run the MCMC algorithm for spatiotemporal point process data. Not for general purpose use.

Usage

```
MALAlgcpSpatioTemporal.PlusPars(
  mcmcloop,
  inits,
  adaptivescheme,
  M,
  N,
  Mext,
```

```
  Next,
  mcens,
  ncens,
  formula,
  ZmatList,
  model.priors,
  model.inits,
  fftgrid,
  spatial.covmodel,
  nis,
  tdiff,
  cellarea,
  spatialvals,
  cellInside,
  MCMCdiag,
  gradtrunc,
  gridfun,
  gridav,
  d,
  aggtimes,
  spatialOnlyCovariates
)
```

Arguments

`mcmcloop`	details of the mcmc loop
`inits`	initial values
`adaptivescheme`	the adaptive MCMC scheme
`M`	number of grid cells in x direction
`N`	number of grid cells in y direction
`Mext`	number of extended grid cells in x direction
`Next`	number of extended grid cells in y direction
`mcens`	centroids in x direction
`ncens`	centroids in y direction
`formula`	a formula object of the form X ~ var1 + var2 etc.
`ZmatList`	list of design matrices constructed using getZmat
`model.priors`	model priors, constructed using lgcpPrior
`model.inits`	initial values for the MCMC
`fftgrid`	an objects of class FFTgrid, see genFFTgrid
`spatial.covmodel`	spatial covariance model, consructed with CovFunction
`nis`	cell counts on the etended grid
`tdiff`	vector of time differences
`cellarea`	the cell area

spatialvals inerpolated poisson offset on fft grid

cellInside 0-1 matrix indicating inclusion in the observation window

MCMCdiag not used

gradtrunc gradient truncation parameter

gridfun used to specify other actions to be taken, e.g. dumping MCMC output to disk.

gridav used for computing Monte Carlo expectations online

d matrix of toral distances

aggtimes the aggregate times

spatialOnlyCovariates
whether this is a 'spatial' only problem

Value

output from the MCMC run

---

matchcovariance *matchcovariance function*

Description

A function to match the covariance matrix of a Gaussian Field with an approximate GMRF with neighbourhood size ns.

Usage

```
matchcovariance(
  xg,
  yg,
  ns,
  sigma,
  phi,
  model,
  additionalparameters,
  verbose = TRUE,
  r = 1,
  method = "Nelder-Mead"
)
```

Arguments

xg x grid must be equally spaced

yg y grid must be equally spaced

ns neighbourhood size

sigma spatial variability parameter

`phi`	spatial dependence parameter
`model`	covariance model, see ?CovarianceFct
`additionalparameters`	additional parameters for chosen covariance model
`verbose`	whether or not to print stuff generated by the optimiser
`r`	parameter used in optimisation, see Rue and Held (2005) pp 188. default value 1.
`method`	The choice of optimising routine must either be 'Nelder-Mead' or 'BFGS'. see ?optim

Value

...

`maternCovFct15` *maternCovFct15 function*

Description

A function to declare and also evaluate an Matern 1.5 covariance function.

Usage

```
maternCovFct15(d, CovParameters)
```

Arguments

`d`	toral distance
`CovParameters`	parameters of the latent field, an object of class "CovParamaters".

Value

the exponential covariance function

Author(s)

Dominic Schumacher

See Also

CovFunction.function, RandomFieldsCovFct, SpikedExponentialCovFct

---

`maternCovFct25` *maternCovFct25 function*

Description

A function to declare and also evaluate an Matern 2.5 covariance function.

Usage

```
maternCovFct25(d, CovParameters)
```

Arguments

`d`	toral distance
`CovParameters`	parameters of the latent field, an object of class "CovParamaters".

Value

the exponential covariance function

Author(s)

Dominic Schumacher

See Also

CovFunction.function, RandomFieldsCovFct, SpikedExponentialCovFct

---

`mcmcLoop` *iterator for MCMC loops*

Description

control an MCMC loop with this iterator

Usage

```
mcmcLoop(N, burnin, thin, trim = TRUE, progressor = mcmcProgressPrint)
```

Arguments

`N`	number of iterations
`burnin`	length of burn-in
`thin`	frequency of thinning
`trim`	whether to cut off iterations after the last retained iteration
`progressor`	a function that returns a progress object

`mcmcpars` *mcmcpars function*

Description

A function for setting MCMC options in a run of `lgcpPredict` for example.

Usage

```
mcmcpars(mala.length, burnin, retain, inits = NULL, adaptivescheme)
```

Arguments

`mala.length`	default = 100,
`burnin`	default = floor(mala.length/2),
`retain`	thinning parameter eg operated on chain every 'retain' iteration (eg store output or compute some posterior functional)
`inits`	optional initial values for MCMC
`adaptivescheme`	the type of adaptive mcmc to use, see ?constanth (constant h) or ?andrieuthomsh (adaptive MCMC of Andrieu and Thoms (2008))

Value

mcmc parameters

See Also

lgcpPredict

`mcmcProgressNone` *null progress monitor*

Description

a progress monitor that does nothing

Usage

```
mcmcProgressNone(mcmcloop)
```

Arguments

`mcmcloop`	an mcmc loop iterator

Value

a progress monitor

`mcmcProgressPrint` *printing progress monitor*

Description

a progress monitor that prints each iteration

Usage

```
mcmcProgressPrint(mcmcloop)
```

Arguments

`mcmcloop`	an mcmc loop iterator

Value

a progress monitor

`mcmcProgressTextBar` *text bar progress monitor*

Description

a progress monitor that uses a text progress bar

Usage

```
mcmcProgressTextBar(mcmcloop)
```

Arguments

`mcmcloop`	an mcmc loop iterator

Value

a progress monitor

`mcmcProgressTk` *graphical progress monitor*

Description

a progress monitor that uses tcltk dialogs

Usage

```
mcmcProgressTk(mcmcloop)
```

Arguments

`mcmcloop`	an mcmc loop iterator

Value

a progress monitor

`mcmctrace` *mcmctrace function*

Description

Generic function to extract the information required to produce MCMC trace plots.

Usage

```
mcmctrace(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method mcmctrace

mcmctrace.lgcpPredict *mcmctrace.lgcpPredict function*

Description

If MCMCdiag was positive when lgcpPredict was called, then this retrieves information from the chains stored.

Usage

```
## S3 method for class 'lgcpPredict'
mcmctrace(obj, ...)
```

Arguments

obj	an object of class lgcpPredict
...	additional arguments

Value

returns the saved MCMC chains in an object of class mcmcdiag.

See Also

lgcpPredict, plot.mcmcdiag

meanfield *meanfield function*

Description

Generic function to extract the mean of the latent field Y.

Usage

```
meanfield(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

method meanfield

meanfield.lgcpPredict *meanfield.lgcpPredict function*

Description

This is an accessor function for objects of class `lgcpPredict` and returns the mean of the field Y as an lgcpgrid object.

Usage

```
## S3 method for class 'lgcpPredict'
meanfield(obj, ...)
```

Arguments

`obj`	an object of class lgcpPredict
`...`	additional arguments

Value

returns the cell-wise mean of Y computed via Monte Carlo.

See Also

lgcpPredict, lgcpgrid

meanfield.lgcpPredictINLA *meanfield.lgcpPredictINLA function*

Description

A function to return the mean of the latent field from a call to lgcpPredictINLA output.

Usage

```
## S3 method for class 'lgcpPredictINLA'
meanfield(obj, ...)
```

Arguments

`obj`	an object of class lgcpPredictINLA
`...`	other arguments

Value

the mean of the latent field

MonteCarloAverage *MonteCarloAverage function*

Description

This function creates an object of class `MonteCarloAverage`. The purpose of the function is to compute Monte Carlo expectations online in the function `lgcpPredict`, it is set in the argument `gridmeans` of the argument `output.control`.

Usage

```
MonteCarloAverage(funlist, lastonly = TRUE)
```

Arguments

`funlist`	a character vector of names of functions, each accepting single argument Y
`lastonly`	compute average using only time T? (see ?lgcpPredict for definition of T)

Details

A Monte Carlo Average is computed as:

$$E_{\pi(Y_{t_1:t_2}|X_{t_1:t_2})}[g(Y_{t_1:t_2})] \approx \frac{1}{n}\sum_{i=1}^{n} g(Y_{t_1:t_2}^{(i)})$$

where g is a function of interest, $Y_{t_1:t_2}^{(i)}$ is the ith retained sample from the target and n is the total number of retained iterations. For example, to compute the mean of $Y_{t_1:t_2}$ set,

$$g(Y_{t_1:t_2}) = Y_{t_1:t_2},$$

the output from such a Monte Carlo average would be a set of $t_2 - t_1$ grids, each cell of which being equal to the mean over all retained iterations of the algorithm (NOTE: this is just an example computation, in practice, there is no need to compute the mean on line explicitly, as this is already done by defaul in `lgcpPredict`). For further examples, see below. The option `last=TRUE` computes,

$$E_{\pi(Y_{t_1:t_2}|X_{t_1:t_2})}[g(Y_{t_2})],$$

so in this case the expectation over the last time point only is computed. This can save computation time.

Value

object of class MonteCarloAverage

See Also

setoutput, lgcpPredict, GAinitialise, GAupdate, GAfinalise, GAreturnvalue, exceedProbs

Examples

```
fun1 <- function(x){return(x)}   # gives the mean
fun2 <- function(x){return(x^2)} # computes E(X^2). Can be used with the
                                 # mean to compute variances, since
                                 # Var(X) = E(X^2) - E(X)^2
fun3 <- exceedProbs(c(1.5,2,3))  # exceedance probabilities,
                                 #see ?exceedProbs
mca <- MonteCarloAverage(c("fun1","fun2","fun3"))
mca2 <- MonteCarloAverage(c("fun1","fun2","fun3"),lastonly=TRUE)
```

---

mstppp *mstppp function*

Description

Generic function used in the construction of marked space-time planar point patterns. An mstppp object is like an stppp object, but with an extra component containing a data frame (the mark information).

Usage

```
mstppp(P, ...)
```

Arguments

P	an object
...	additional arguments

Details

Observations are assumed to occur in the plane and the observation window is assumed not to change over time.

Value

method mstppp

See Also

mstppp, mstppp.ppp, mstppp.list

mstppp.list *mstppp.list function*

Description

Construct a marked space-time planar point pattern from a list object

Usage

```
## S3 method for class 'list'
mstppp(P, ...)
```

Arguments

`P`	list object containing $xyt, an (n x 3) matrix corresponding to (x,y,t) values; $tlim, a vector of length 2 givign the observation time window, $window giving an owin spatial observation winow, see ?owin for more details, and $data, a data frame containing the collection of marks
`...`	additional arguments

Value

an object of class mstppp

See Also

mstppp, mstppp.ppp,

mstppp.ppp *mstppp.ppp function*

Description

Construct a marked space-time planar point pattern from a ppp object

Usage

```
## S3 method for class 'ppp'
mstppp(P, t, tlim, data, ...)
```

Arguments

`P`	a spatstat ppp object
`t`	a vector of length P$n
`tlim`	a vector of length 2 specifying the observation time window
`data`	a data frame containing the collection of marks
`...`	additional arguments

Value

an object of class mstppp

See Also

mstppp, mstppp.list

`mstppp.stppp` *mstppp.stppp function*

Description

Construct a marked space-time planar point pattern from an stppp object

Usage

```
## S3 method for class 'stppp'
mstppp(P, data, ...)
```

Arguments

`P`	an lgcp stppp object
`data`	a data frame containing the collection of marks
`...`	additional arguments

Value

an object of class mstppp

See Also

mstppp, mstppp.list

`muEst` *muEst function*

Description

Computes a non-parametric estimate of mu(t). For the purposes of performing prediction, the alternatives are: (1) use a parameteric model as in Diggle P, Rowlingson B, Su T (2005), or (2) a constantInTime model.

Usage

```
muEst(xyt, ...)
```

Arguments

`xyt`	an stppp object
`...`	additional arguments to be passed to lowess

Value

object of class temporalAtRisk giving the smoothed mut using the lowess function

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
3. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

temporalAtRisk, constantInTime, ginhomAverage, KinhomAverage, spatialparsEst, thetaEst, lambdaEst

---

`multiply.list` *multiply.list function*

Description

This function multiplies the elements of two list objects together and returns the result in another list object.

Usage

```
multiply.list(list1, list2)
```

Arguments

`list1`	a list of objects that could be summed using "+"
`list2`	a list of objects that could be summed using "+"

Value

a list with ith entry the sum of list1[[i]] and list2[[i]]

neattable *neattable function*

Description

Function to print right-aligned tables to the console.

Usage

```
neattable(mat, indent = 0)
```

Arguments

mat	a numeric or character matrix object
indent	indent

Value

prints to screen with specified indent

Examples

```
mat <- rbind(c("one","two","three"),matrix(round(runif(9),3),3,3))
neattable(mat)
```

neigh2D *neigh2D function*

Description

A function to compute the neighbours of a cell on a toral grid

Usage

```
neigh2D(i, j, ns, M, N)
```

Arguments

i	cell index i
j	cell index j
ns	number of neighbours either side
M	size of grid in x direction
N	size of grid in y direction

Value

the cell indices of the neighbours

nextStep *next step of an MCMC chain*

Description

just a wrapper for nextElem really.

Usage

```
nextStep(object)
```

Arguments

`object`	an mcmc loop object

nullAverage *nullAverage function*

Description

A null scheme, that does not perform any computation in the running of `lgcpPredict`, it is the default value of `gridmeans` in the argument `output.control`.

Usage

```
nullAverage()
```

Value

object of class nullAverage

See Also

setoutput, lgcpPredict, GAinitialise, GAupdate, GAfinalise, GAreturnvalue

`nullFunction` *nullFunction function*

Description

This is a null function and performs no action.

Usage

```
nullFunction()
```

Value

object of class nullFunction

See Also

setoutput, GFinitialise, GFupdate, GFfinalise, GFreturnvalue

`numCases` *numCases function*

Description

A function used in conjunction with the function "expectation" to compute the expected number of cases in each computational grid cell. Currently only implemented for spatial processes (lgcpPredictSpatialPlusPars and lgcpPredictAggregateSpatialPlusPars).

Usage

```
numCases(Y, beta, eta, Z, otherargs)
```

Arguments

`Y`	the latent field
`beta`	the main effects
`eta`	the parameters of the latent field
`Z`	the design matrix
`otherargs`	other arguments to the function (see vignette "Bayesian_lgcp" for an explanation)

Value

the number of cases in each cell

See Also

expectation, lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars

Examples

```
## Not run: ex <- expectation(lg,numCases)[[1]] # lg is output from spatial LGCP MCMC
```

---

osppp2latlon — *osppp2latlon function*

Description

A function to transform a ppp object in the OSGB projection (epsg:27700) to a ppp object in the latitude/longitude (epsg:4326) projection.

Usage

```
osppp2latlon(obj)
```

Arguments

obj	a ppp object in OSGB

Value

a pppobject in Lat/Lon

---

osppp2merc — *osppp2merc function*

Description

A function to transform a ppp object in the OS GB projection (epsg:27700) to a ppp object in the Mercator (epsg:3857) projection.

Usage

```
osppp2merc(obj)
```

Arguments

obj	a ppp object in OSGB

Value

a ppp object in Mercator

paramprec *paramprec function*

Description

A function to compute the precision matrix of a GMRF on an M x N toral grid with neighbourhood size ns. Note that the precision matrix is block circulant. The returned function operates on a parameter vector as in Rue and Held (2005) pp 187.

Usage

```
paramprec(ns, M, N)
```

Arguments

`ns`	neighbourhood size
`M`	number of cells in x direction
`N`	number of cells in y direction

Value

a function that returns the precision matrix given a parameter vector.

paramprecbase *paramprecbase function*

Description

A function to compute the parametrised base matrix of a precision matrix of a GMRF on an M x N toral grid with neighbourhood size ns. Note that the precision matrix is block circulant. The returned function operates on a parameter vector as in Rue and Held (2005) pp 187.

Usage

```
paramprecbase(ns, M, N, inverse = FALSE)
```

Arguments

`ns`	neighbourhood size
`M`	number of x cells
`N`	number of y cells
`inverse`	whether or not to compute the base matrix of the inverse precision matrix (ie the covariance matrix). default is FALSE

Value

a functioin that returns the base matrix of the precision matrix `parautocorr` *parautocorr function*

Description

A function to produce autocorrelation plots for the paramerers beta and eta from a call to the function lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars or lgcpPredictMultitypeSpatialPlusPars

Usage

```
parautocorr(obj, xlab = "Lag", ylab = NULL, main = "", ask = TRUE, ...)
```

Arguments

`obj`	an object produced by a call to lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars orlgcpPredictMultitypeSpatialPlusPars
`xlab`	optional label for x-axis, there is a sensible default.
`ylab`	optional label for y-axis, there is a sensible default.
`main`	optional title of the plot, there is a sensible default.
`ask`	the paramter "ask", see ?par
`...`	other arguments passed to the function "hist"

Value

produces autocorrelation plots of the parameters beta and eta

See Also

ltar, autocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

`parsummary` *parsummary function*

Description

A function to produce a summary table for the parameters beta and eta from a call to the function lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars or lgcpPredictMultitypeSpatialPlusPars

Usage

```
parsummary(obj, expon = TRUE, LaTeX = FALSE, ...)
```

Arguments

`obj`	an object produced by a call to lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars orlgcpPredictMultitypeSpatialPlusPars
`expon`	whether to exponentiate the results, so that the parameters beta haev the interpretation of "relative risk per unit increase in the covariate" default is TRUE
`LaTeX`	whether to print paramter names using LaTeX symbols (if the table is later to be exported to a LaTeX document)
`...`	other arguments

Value

a data frame containing the median, 0.025 and 0.975 quantiles.

See Also

ltar, autocorr, parautocorr, traceplots, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

`plot.fromSPDF` *plot.fromSPDF function*

Description

Plot method for objects of class fromSPDF.

Usage

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

Arguments

`x`	an object of class spatialAtRisk
`...`	additional arguments

Value

prints the object

plot.fromXYZ *plot.fromXYZ function*

Description

Plot method for objects of class fromXYZ.

Usage

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

Arguments

x	object of class spatialAtRisk
...	additional arguments

Value

an image plot

plot.lgcpAutocorr *plot.lgcpAutocorr function*

Description

Plots `lgcpAutocorr` objects: output from `autocorr`

Usage

```
## S3 method for class 'lgcpAutocorr'
plot(x, sel = 1:dim(x)[3], ask = TRUE, crop = TRUE, plotwin = FALSE, ...)
```

Arguments

x	an object of class lgcpAutocorr
sel	vector of integers between 1 and grid$len: which grids to plot. Default NULL, in which case all grids are plotted.
ask	logical; if TRUE the user is asked before each plot
crop	whether or not to crop to bounding box of observation window
plotwin	logical whether to plot the window attr(x,"window"), default is FALSE
...	other arguments passed to image.plot

Value

a plot

See Also

autocorr

Examples

```
## Not run: ac <- autocorr(lg,qt=c(1,2,3))
                          # assumes that lg has class lgcpPredict
## Not run: plot(ac)
```

---

plot.lgcpgrid *plot.lgcpgrid function*

Description

This is a wrapper function for image.lgcpgrid

Usage

```
## S3 method for class 'lgcpgrid'
plot(x, sel = 1:x$len, ask = TRUE, ...)
```

Arguments

x	an object of class lgcpgrid
sel	vector of integers between 1 and grid$len: which grids to plot. Default NULL, in which case all grids are plotted.
ask	logical; if TRUE the user is asked before each plot
...	other arguments

Value

an image-type plot

See Also

lgcpgrid.list, lgcpgrid.array, as.list.lgcpgrid, print.lgcpgrid, summary.lgcpgrid,quantile.lgcpgrid, image.lgcpgrid

plot.lgcpPredict *plot.lgcpPredict function*

Description

Simple plotting function for objects of class lgcpPredict.

Usage

```
## S3 method for class 'lgcpPredict'
plot(
  x,
  type = "relrisk",
  sel = 1:x$EY.mean$len,
  plotdata = TRUE,
  ask = TRUE,
  clipWindow = TRUE,
  ...
)
```

Arguments

x	an object of class lgcpPredict
type	Character string: what type of plot to produce. Choices are "relrisk" (=exp(Y)); "serr" (standard error of relative risk); or "intensity" (=lambda*mu*exp(Y)).
sel	vector of integers between 1 and grid$len: which grids to plot. Default NULL, in which case all grids are plotted.
plotdata	whether or not to overlay the data
ask	logical; if TRUE the user is asked before each plot
clipWindow	whether to plot grid cells outside the observation window
...	additional arguments passed to image.plot

Value

plots the Monte Carlo mean of quantities obtained via simulation. By default the mean relative risk is plotted.

See Also

lgcpPredict

plot.lgcpQuantiles *plot.lgcpQuantiles function*

Description

Plots `lgcpQuantiles` objects: output from `quantiles.lgcpPredict`

Usage

```
## S3 method for class 'lgcpQuantiles'
plot(x, sel = 1:dim(x)[3], ask = TRUE, crop = TRUE, plotwin = FALSE, ...)
```

Arguments

x	an object of class lgcpQuantiles
sel	vector of integers between 1 and grid$len: which grids to plot. Default NULL, in which case all grids are plotted.
ask	logical; if TRUE the user is asked before each plot
crop	whether or not to crop to bounding box of observation window
plotwin	logical whether to plot the window attr(x,"window"), default is FALSE
...	other arguments passed to image.plot

Value

grid plotting This is a wrapper function for image.lgcpgrid

See Also

quantile.lgcpPredict

Examples

```
## Not run: qtiles <- quantile(lg,qt=c(0.5,0.75,0.9),fun=exp)
                          # assumed that lg has class lgcpPredict
## Not run: plot(qtiles)
```

plot.lgcpZmat *plot.lgcpZmat function*

Description

A function to plot lgcpZmat objects

Usage

```
## S3 method for class 'lgcpZmat'
plot(
  x,
  ask = TRUE,
  pow = 1,
  main = NULL,
  misscol = "black",
  obswin = NULL,
  ...
)
```

Arguments

x	an lgcpZmat object, see ?getZmat
ask	graphical parameter ask, see ?par
pow	power parameter, raises the image values to this power (helps with visualisation, default is 1.)
main	title for plot, default is null which gives an automatic title to the plot (the name of the covariate)
misscol	colour to identify imputed grid cells, default is yellow
obswin	optional observation window to add to plot using plot(obswin).
...	other paramters

Value

a sequence of plots of the interpolated covariate values

`plot.mcmcdiag` *plot.mcmcdiag function*

Description

The command `plot(trace(lg))`, where `lg` is an object of class `lgcpPredict` will plot the mcmc traces of a subset of the cells, provided they have been stored, see `mcmpars`.

Usage

```
## S3 method for class 'mcmcdiag'
plot(x, idx = 1:dim(x$trace)[2], ...)
```

Arguments

`x`	an object of class mcmcdiag
`idx`	vector of chain indices to plot, default plots all chains
`...`	additional arguments passed to plot

Value

plots the saved MCMC chains

See Also

mcmctrace.lgcpPredict, mcmcpars,

`plot.mstppp` *plot.mstppp function*

Description

Plot method for mstppp objects

Usage

```
## S3 method for class 'mstppp'
plot(x, cols = "red", ...)
```

Arguments

`x`	an object of class mstppp
`cols`	optional vector of colours to plot points with
`...`	additional arguments passed to plot

Value

plots the mstppp object x

`plot.stppp` *plot.stppp function*

Description

Plot method for stppp objects

Usage

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

Arguments

`x`	an object of class stppp
`...`	additional arguments passed to plot

Value

plots the stppp object x

`plot.temporalAtRisk` *plot.temporalAtRisk function*

Description

Pot a temporalAtRisk object.

Usage

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

Arguments

`x`	an object
`...`	additional arguments

Value

print the object

See Also

temporalAtRisk, spatialAtRisk, temporalAtRisk.numeric, temporalAtRisk.function, constantInTime, constantInTime.numeric, constantInTime.stppp, print.temporalAtRisk,

---

plotExceed *plotExceed function*

Description

A generic function for plotting exceedance probabilities.

Usage

```
plotExceed(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

generic function returning method plotExceed

See Also

plotExceed.lgcpPredict, plotExceed.array

---

plotExceed.array *plotExceed.array function*

Description

Function for plotting exceedance probabilities stored in array objects. Used in `plotExceed.lgcpPredict`.

Usage

```
## S3 method for class 'array'
plotExceed(
  obj,
  fun,
  lgcppredict = NULL,
  xvals = NULL,
  yvals = NULL,
  window = NULL,
```

```
    cases = NULL,
    nlevel = 64,
    ask = TRUE,
    mapunderlay = NULL,
    alpha = 1,
    sub = NULL,
    ...
)
```

Arguments

obj	an object
fun	the name of the function used to compute exceedances (character vector of length 1). Note that the named function must be in memory.
lgcppredict	an object of class lgcpPredict that can be used to supply an observation window and x and y coordinates
xvals	optional vector giving x coords of centroids of cells
yvals	optional vector giving y coords of centroids of cells
window	optional obervation window
cases	optional xy (n x 2) matrix of locations of cases to plot
nlevel	number of colour levels to use in plot, default is 64
ask	whether or not to ask for a new plot between plotting exceedances at different thresholds.
mapunderlay	optional underlay to plot underneath maps of exceedance probabilities. Use in conjunction with rainbow parameter 'alpha' (eg alpha=0.3) to set transparency of exceedance layer.
alpha	graphical parameter takign values in [0,1] controlling transparency of exceedance layer. Default is 1.
sub	optional subtitle for plot
...	additional arguments passed to image.plot

Value

generic function returning method plotExceed

See Also

plotExceed.lgcpPredict

plotExceed.lgcpPredict

plotExceed.lgcpPredict function

Description

Function for plotting exceedance probabilities stored in `lgcpPredict` ojects.

Usage

```
## S3 method for class 'lgcpPredict'
plotExceed(
  obj,
  fun,
  nlevel = 64,
  ask = TRUE,
  plotcases = FALSE,
  mapunderlay = NULL,
  alpha = 1,
  ...
)
```

Arguments

`obj`	an object
`fun`	the name of the function used to compute exceedances (character vector of length 1). Note that the named function must be in memory.
`nlevel`	number of colour levels to use in plot, default is 64
`ask`	whether or not to ask for a new plot between plotting exceedances at different thresholds.
`plotcases`	whether or not to plot the cases on the map
`mapunderlay`	optional underlay to plot underneath maps of exceedance probabilities. Use in conjunction with rainbow parameter 'alpha' (eg alpha=0.3) to set transparency of exceedance layer.
`alpha`	graphical parameter takign values in [0,1] controlling transparency of exceedance layer. Default is 1.
`...`	additional arguments passed to image.plot

Value

plot of exceedances

See Also

lgcpPredict, MonteCarloAverage, setoutput

Examples

```
## Not run: exceedfun <- exceedProbs(c(1.5,2,4))
## Not run:
    plot(lg,"exceedfun") # lg is an object of class lgcpPredict
                         # in which the Monte Carlo mean of
                         # "exceedfun" was computed
                         # see ?MonteCarloAverage and ?setoutput

## End(Not run)
```

`plotit` *plotit function*

Description

A function to plot various objects. A developmental tool: not intended for general use

Usage

```
plotit(x)
```

Arguments

`x`	an a list, matrix, or GPrealisation object.

Value

plots the objects.

`postcov` *postcov function*

Description

Generic function for producing plots of the posterior covariance function from a call to the function lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars or lgcpPredictMultitypeSpatialPlusPars.

Usage

```
postcov(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method postcov

See Also

postcov.lgcpPredictSpatialOnlyPlusParameters,postcov.lgcpPredictAggregateSpatialPlusParameters, postcov.lgcpPredictSpatioTemporalPlusParameters, postcov.lgcpPredictMultitypeSpatialPlusParameters, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, exceedProbs, betavals, etavals

---

postcov.lgcpPredictAggregateSpatialPlusParameters

postcov.lgcpPredictAggregateSpatialPlusParameters function

Description

A function for producing plots of the posterior covariance function.

Usage

```
"postcov(obj,qts=c(0.025,0.5,0.975),covmodel=NULL,ask=TRUE,...)"
```

Arguments

obj	an lgcpPredictAggregateSpatialPlusParameters object
qts	vector of quantiles of length 3, default is 0.025, 0.5, 0.975
covmodel	the assumed covariance model. NULL by default, this information is read in from the object obj, so generally does not need to be set.
ask	parameter "ask", see ?par
...	additional arguments

Value

...

See Also

postcov.lgcpPredictSpatialOnlyPlusParameters, postcov.lgcpPredictAggregateSpatialPlusParameters, postcov.lgcpPredictSpatioTemporalPlusParameters, postcov.lgcpPredictMultitypeSpatialPlusParameters, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

postcov.lgcpPredictMultitypeSpatialPlusParameters *postcov.lgcpPredictMultitypeSpatialPlusParameters function*

Description

A function for producing plots of the posterior covariance function.

Usage

```
"postcov(obj,qts=c(0.025,0.5,0.975),covmodel=NULL,ask=TRUE,...)"
```

Arguments

obj	an lgcpPredictMultitypeSpatialPlusParameters object
qts	vector of quantiles of length 3, default is 0.025, 0.5, 0.975
covmodel	the assumed covariance model. NULL by default, this information is read in from the object obj, so generally does not need to be set.
ask	parameter "ask", see ?par
...	additional arguments

Value

plots of the posterior covariance function for each type.

See Also

postcov.lgcpPredictSpatialOnlyPlusParameters, postcov.lgcpPredictAggregateSpatialPlusParameters, postcov.lgcpPredictSpatioTemporalPlusParameters, postcov.lgcpPredictMultitypeSpatialPlusParameters, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

postcov.lgcpPredictSpatialOnlyPlusParameters *postcov.lgcpPredictSpatialOnlyPlusParameters function*

Description

A function for producing plots of the posterior spatial covariance function.

Usage

```
"postcov(obj,qts=c(0.025,0.5,0.975),covmodel=NULL,ask=TRUE,...)"
```

Arguments

`obj`	an lgcpPredictSpatialOnlyPlusParameters object
`qts`	vector of quantiles of length 3, default is 0.025, 0.5, 0.975
`covmodel`	the assumed covariance model. NULL by default, this information is read in from the object obj, so generally does not need to be set.
`ask`	parameter "ask", see ?par
`...`	additional arguments

Value

a plot of the posterior covariance function.

See Also

postcov.lgcpPredictSpatialOnlyPlusParameters, postcov.lgcpPredictAggregateSpatialPlusParameters, postcov.lgcpPredictSpatioTemporalPlusParameters, postcov.lgcpPredictMultitypeSpatialPlusParameters, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

`postcov.lgcpPredictSpatioTemporalPlusParameters` *postcov.lgcpPredictSpatioTemporalPlusParameters function*

Description

A function for producing plots of the posterior spatiotemporal covariance function.

Usage

```
"postcov(obj,qts=c(0.025,0.5,0.975),covmodel=NULL,ask=TRUE,...)"
```

Arguments

`obj`	an lgcpPredictSpatioTemporalPlusParameters object
`qts`	vector of quantiles of length 3, default is 0.025, 0.5, 0.975
`covmodel`	the assumed covariance model. NULL by default, this information is read in from the object obj, so generally does not need to be set.
`ask`	parameter "ask", see ?par
`...`	additional arguments

Value

a plot of the posterior spatial covariance function and temporal correlation function.

See Also

postcov.lgcpPredictSpatialOnlyPlusParameters, postcov.lgcpPredictAggregateSpatialPlusParameters, postcov.lgcpPredictSpatioTemporalPlusParameters, postcov.lgcpPredictMultitypeSpatialPlusParameters, ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

print.dump2dir *print.dump2dir function*

Description

Display function for dump2dir objects.

Usage

```
## S3 method for class 'dump2dir'
print(x, ...)
```

Arguments

x	an object of class dump2dir
...	additional arguments

Value

nothing

See Also

dump2dir,

---

print.fromFunction *print.fromFunction function*

Description

Print method for objects of class fromFunction.

Usage

```
## S3 method for class 'fromFunction'
print(x, ...)
```

Arguments

`x`	an object of class spatialAtRisk
`...`	additional arguments

Value

prints the object

`print.fromSPDF` *print.fromSPDF function*

Description

Print method for objects of class fromSPDF.

Usage

```
## S3 method for class 'fromSPDF'
print(x, ...)
```

Arguments

`x`	an object of class spatialAtRisk
`...`	additional arguments

Value

prints the object

`print.fromXYZ` *print.fromXYZ function*

Description

Print method for objects of class fromXYZ.

Usage

```
## S3 method for class 'fromXYZ'
print(x, ...)
```

Arguments

`x`	an object of class spatialAtRisk
`...`	additional arguments

Value

prints the object

`print.gridaverage` *print.gridaverage function*

Description

Print method for `gridaverage` objects

Usage

```
## S3 method for class 'gridaverage'
print(x, ...)
```

Arguments

`x`	an object of class gridaverage
`...`	other arguments

Value

just prints out details

`print.lgcpgrid` *print.lgcpgrid function*

Description

Print method for lgcp grid objects.

Usage

```
## S3 method for class 'lgcpgrid'
print(x, ...)
```

Arguments

`x`	an object of class lgcpgrid
`...`	other arguments

Value

just prints out details to the console

See Also

lgcpgrid.list, lgcpgrid.array, as.list.lgcpgrid, summary.lgcpgrid quantile.lgcpgrid image.lgcpgrid plot.lgcpgrid

print.lgcpPredict *print.lgcpPredict function*

Description

Print method for lgcpPredict objects.

Usage

```
## S3 method for class 'lgcpPredict'
print(x, ...)
```

Arguments

x	an object of class lgcpPredict
...	additional arguments

Value

just prints information to the screen

See Also

lgcpPredict

print.mcmc *print.mcmc function*

Description

print method print an mcmc iterator's details

Usage

```
## S3 method for class 'mcmc'
print(x, ...)
```

Arguments

x	a mcmc iterator
...	other args

`print.mstppp` *print.mstppp function*

Description

Print method for mstppp objects

Usage

```
## S3 method for class 'mstppp'
print(x, ...)
```

Arguments

`x`	an object of class mstppp
`...`	additional arguments

Value

prints the mstppp object x

`print.stapp` *print.stapp function*

Description

Print method for stapp objects

Usage

```
## S3 method for class 'stapp'
print(x, printhead = TRUE, ...)
```

Arguments

`x`	an object of class stapp
`printhead`	whether or not to print the head of the counts matrix
`...`	additional arguments

Value

prints the stapp object x

`print.stppp` *print.stppp function*

Description

Print method for stppp objects

Usage

```
## S3 method for class 'stppp'
print(x, ...)
```

Arguments

`x`	an object of class stppp
`...`	additional arguments

Value

prints the stppp object x

`print.temporalAtRisk` *print.temporalAtRisk function*

Description

Printing method for temporalAtRisk objects.

Usage

```
## S3 method for class 'temporalAtRisk'
print(x, ...)
```

Arguments

`x`	an object
`...`	additional arguments

Value

print the object

See Also

temporalAtRisk, spatialAtRisk, temporalAtRisk.numeric, temporalAtRisk.function, constantInTime, constantInTime.numeric, constantInTime.stppp, plot.temporalAtRisk `priorpost` *priorpost function*

Description

A function to plot the prior and posterior densities of the model parameters eta and beta. The prior appears as a red line and the posterior appears as a histogram.

Usage

```
priorpost(
  obj,
  breaks = 30,
  xlab = NULL,
  ylab = "Density",
  main = "",
  ask = TRUE,
  ...
)
```

Arguments

`obj`	an object produced by a call to lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars or lgcpPredictMultitypeSpatialPlusPars
`breaks`	"breaks" paramter from the function "hist"
`xlab`	optional label for x-axis, there is a sensible default.
`ylab`	optional label for y-axis, there is a sensible default.
`main`	optional title of the plot, there is a sensible default.
`ask`	the paramter "ask", see ?par
`...`	other arguments passed to the function "hist"

Value

plots of the prior and posterior of the model parameters eta and beta.

See Also

ltar, autocorr, parautocorr, traceplots, parsummary, textsummary, postcov, exceedProbs, betavals, etavals

PriorSpec *PriorSpec function*

Description

Generic for declaring that an object is of valid type for use as as prior in lgcp. For further details and examples, see the vignette "Bayesian_lgcp".

Usage

```
PriorSpec(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method PriorSpec

See Also

PriorSpec.list

PriorSpec.list *PriorSpec.list function*

Description

Method for declaring a Bayesian prior density in lgcp. Checks to confirm that the object obj has the requisite components for functioning as a prior.

Usage

```
## S3 method for class 'list'
PriorSpec(obj, ...)
```

Arguments

`obj`	a list object defining a prior , see ?GaussianPrior and ?LogGaussianPrior
`...`	additional arguments

Value

an object suitable for use in a call to the MCMC routines

See Also

GaussianPrior, LogGaussianPrior

Examples

```
## Not run: PriorSpec(LogGaussianPrior(mean=log(c(1,500)),variance=diag(0.15,2)))
## Not run: PriorSpec(GaussianPrior(mean=rep(0,9),variance=diag(10^6,9)))
```

---

quantile.lgcpgrid *quantile.lgcpgrid function*

Description

Quantile method for lgcp objects. This just applies the quantile function to each of the elements of x$grid

Usage

```
## S3 method for class 'lgcpgrid'
quantile(x, ...)
```

Arguments

x	an object of class lgcpgrid
...	other arguments

Value

Quantiles per grid, see ?quantile for further options

See Also

lgcpgrid.list, lgcpgrid.array, as.list.lgcpgrid, print.lgcpgrid, summary.lgcpgrid, image.lgcpgrid, plot.lgcpgrid quantile.lgcpPredict *quantile.lgcpPredict function*

Description

This function requires data to have been dumped to disk: see ?dump2dir and ?setoutput. The routine quantile.lgcpPredict computes quantiles of functions of Y. For example, to get cell-wise quantiles of exceedance probabilities, set fun=exp. Since computign the quantiles is an expensive operation, the option to output the quantiles on a subregion of interest is also provided (by setting the argument inWindow, which has a sensible default).

Usage

```
## S3 method for class 'lgcpPredict'
quantile(
  x,
  qt,
  tidx = NULL,
  fun = NULL,
  inWindow = x$xyt$window,
  crop2parentwindow = TRUE,
  startidx = 1,
  sampcount = NULL,
  ...
)
```

Arguments

x	an object of class lgcpPredict
qt	a vector of the required quantiles
tidx	the index number of the the time interval of interest, default is the last time point.
fun	a 1-1 function (default the identity function) to be applied cell-wise to the grid. Must be able to evaluate sapply(vec,fun) for vectors vec.
inWindow	an observation owin window on which to compute the quantiles, can speed up calculation. Default is xxytwindow.
crop2parentwindow	logical: whether to only compute the quantiles for cells inside xxytwindow (the 'parent window')
startidx	optional starting sample index for computing quantiles. Default is 1.
sampcount	number of samples to include in computation of quantiles after startidx. Default is all
...	additional arguments

Value

an array, the [,,i]th slice being the grid of cell-wise quantiles, qt[i], of fun(Y), where Y is the MCMC output dumped to disk.

See Also

lgcpPredict, dump2dir, setoutput, plot.lgcpQuantiles

RandomFieldsCovFct *RandomFieldsCovFct function*

Description

A function to declare and also evaluate an covariance function from the RandomFields Package. See ?CovarianceFct. Note that the present version of lgcp only offers estimation for sigma and phi, any additional paramters are treated as fixed.

Usage

```
RandomFieldsCovFct(model, additionalparameters = c())
```

Arguments

`model`	the choice of model e.g. "matern"
`additionalparameters`	additional parameters for chosen covariance model. See ?CovarianceFct

Value

a covariance function from the RandomFields package

See Also

CovFunction.function, exponentialCovFct, SpikedExponentialCovFct, CovarianceFct

Examples

```
## Not run: RandomFieldsCovFct(model="matern",additionalparameters=1)
```

`raster.lgcpgrid` *raster.lgcpgrid function*

Description

A function to convert lgcpgrid objects into either a raster object, or a RasterBrick object.

Usage

```
## S3 method for class 'lgcpgrid'
raster(x, crs = NA, transpose = FALSE, ...)
```

Arguments

`x`	an lgcpgrid object
`crs`	PROJ4 type description of a map projection (optional). See ?raster
`transpose`	Logical. Transpose the data? See ?brick method for array
`...`	additional arguments

Value

...

`rescale.mstppp` *rescale.mstppp function*

Description

Rescale an mstppp object. Similar to rescale.ppp

Usage

```
## S3 method for class 'mstppp'
rescale(X, s, unitname)
```

Arguments

`X`	an object of class mstppp
`s`	scale as in rescale.ppp: x and y coordinaes are scaled by 1/s
`unitname`	parameter as defined in ?rescale

Value

a ppp object without observation times

rescale.stppp *rescale.stppp function*

Description

Rescale an stppp object. Similar to rescale.ppp

Usage

```
## S3 method for class 'stppp'
rescale(X, s, unitname)
```

Arguments

X	an object of class stppp
s	scale as in rescale.ppp: x and y coordinaes are scaled by 1/s
unitname	parameter as defined in ?rescale

Value

a ppp object without observation times

resetLoop *reset iterator*

Description

call this to reset an iterator's state to the initial

Usage

```
resetLoop(obj)
```

Arguments

obj	an mcmc iterator

`rgauss` *rgauss function*

Description

A function to simulate a Gaussian field on a regular square lattice, the returned object is of class lgcpgrid.

Usage

```
rgauss(
  n = 1,
  range = c(0, 1),
  ncells = 128,
  spatial.covmodel = "exponential",
  model.parameters = lgcppars(sigma = 2, phi = 0.1),
  covpars = c(),
  ext = 2
)
```

Arguments

`n`	the number of realisations to generate. Default is 1.
`range`	a vector of length 2, defining the left-most and right most cell centroids in the x-direction. Note that the centroids in the y-direction are the same as those in the x-direction.
`ncells`	the number of cells, typially a power of 2
`spatial.covmodel`	spatial covariance function, default is exponential, see ?CovarianceFct
`model.parameters`	parameters of model, see ?lgcppars. Only set sigma and phi for spatial model.
`covpars`	vector of additional parameters for spatial covariance function, in order they appear in chosen model in ?CovarianceFct
`ext`	how much to extend the parameter space by. Default is 2.

Value

an lgcp grid object containing the simulated field(s).

roteffgain *roteffgain function*

Description

Compute whether there might be any advantage in rotating the observation window in the object xyt for a proposed cell width.

Usage

```
roteffgain(xyt, cellwidth)
```

Arguments

`xyt`	an object of class stppp
`cellwidth`	size of grid on which to do MALA

Value

whether or not there woud be any efficiency gain in the MALA by rotating window

See Also

getRotation.stppp

rotmat *rotmat function*

Description

This function returns a rotation matrix corresponding to an anticlockwise rotation of theta radians about the origin

Usage

```
rotmat(theta)
```

Arguments

`theta`	an angle in radians

Value

the transformation matrix corresponding to an anticlockwise rotation of theta radians about the origin `rr` *rr function*

Description

Generic function to return relative risk.

Usage

```
rr(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method rr

See Also

lgcpPredict, rr.lgcpPredict

`rr.lgcpPredict` *rr.lgcpPredict function*

Description

Accessor function returning the relative risk = exp(Y) as an lgcpgrid object.

Usage

```
## S3 method for class 'lgcpPredict'
rr(obj, ...)
```

Arguments

`obj`	an lgcpPredict object
`...`	additional arguments

Value

the relative risk as computed my MCMC

See Also

lgcpPredict

samplePosterior *samplePosterior function*

Description

A function to draw a sample from the posterior of a spatial LGCP. Randomly selects an index i, and returns the ith value of eta, the ith value of beta and the ith value of Y as a named list.

Usage

```
samplePosterior(x)
```

Arguments

`x`	an object of class lgcpPredictSpatialOnlyPlusParameters or lgcpPredictAggregateSpatialPlusParameters

Value

a sample from the posterior named list object with names elements "eta", "beta" and "Y".

segProbs *segProbs function*

Description

A function to compute segregation probabilities from a multivariate LGCP. See the vignette "Bayesian_lgcp" for a full explanation of this.

Usage

```
segProbs(obj, domprob)
```

Arguments

`obj`	an lgcpPredictMultitypeSpatialPlusParameters object
`domprob`	the threshold beyond which we declare a type as dominant e.g. a value of 0.8 would mean we would consider each type to be dominant if the conditional probability of an event of a given type at that location exceeded 0.8.

Details

We suppose there are K point types of interest. The model for point-type k is as follows:

X_k(s) ~ Poisson[R_k(s)]

R_k(s) = C_A lambda_k(s) exp[Z_k(s)beta_k+Y_k(s)]

Here X_k(s) is the number of events of type k in the computational grid cell containing the point s, R_k(s) is the Poisson rate, C_A is the cell area, lambda_k(s) is a known offset, Z_k(s) is a vector of measured covariates and Y_i(s) where i = 1,...,K+1 are latent Gaussian processes on the computational grid. The other parameters in the model are beta_k , the covariate effects for the kth type; and eta_i = [log(sigma_i),log(phi_i)], the parameters of the process Y_i for i = 1,...,K+1 on an appropriately transformed (again, in this case log) scale.

The term 'conditional probability of type k' means the probability that at a particular location, x, there will be an event of type k, we denote this p_k(x).

It is also of interest to scientists to be able to illustrate spatial regions where a genotype dominates a posteriori. We say that type k dominates at position x if p_k(x)>c, where c (the parameter domprob) is a threshold is a threshold set by the user. Let A_k(c,q) denote the set of locations x for which P[p_k(x)>c|X] > q.

As the quantities c and q tend to 1 each area A_k(c,p) shrinks towards the empty set; this happens more slowly in a highly segregated pattern compared with a weakly segregated one.

The function segProbs computes P[p_k(x)>c|X] for each type, from which plots of P[p_k(x)>c|X] > q can be produced.

Value

an lgcpgrid object contatining the segregation probabilities.

`seintens` *seintens function*

Description

Generic function to return the standard error of the Poisson Intensity.

Usage

```
seintens(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method seintens

See Also

lgcpPredict, seintens.lgcpPredict

---

seintens.lgcpPredict *seintens.lgcpPredict function*

Description

Accessor function returning the standard error of the Poisson intensity as an lgcpgrid object.

Usage

```
## S3 method for class 'lgcpPredict'
seintens(obj, ...)
```

Arguments

obj	an lgcpPredict object
...	additional arguments

Value

the cell-wise standard error of the Poisson intensity, as computed by MCMC.

See Also

lgcpPredict

---

selectObsWindow *selectObsWindow function*

Description

See ?selectObsWindow.stppp for further details on usage. This is a generic function for the purpose of selecting an observation window (or more precisely a bounding box) to contain the extended FFT grid.

Usage

```
selectObsWindow(xyt, ...)
```

Arguments

`xyt`	an object
`...`	additional arguments

Value

method selectObsWindow

See Also

selectObsWindow.default, selectObsWindow.stppp

`selectObsWindow.default`
selectObsWindow.default function

Description

Default method, note at present, there is only an implementation for stppp objects.

Usage

```
## Default S3 method:
selectObsWindow(xyt, cellwidth, ...)
```

Arguments

`xyt`	an object
`cellwidth`	size of the grid spacing in chosen units (equivalent to the cell width argument in lgcpPredict)
`...`	additional arguments

Details

!!NOTE!! that this function also returns the grid ($xvals and $yvals) on which the FFT (and hence MALA) will be performed. It is useful to define spatialAtRiskobjects on this grid to prevent loss of information from the bilinear interpolation that takes place as part of the fitting algorithm.

Value

this is the same as selectObsWindow.stppp

See Also

spatialAtRisk selectObsWindow.stppp

`selectObsWindow.stppp` *selectObsWindow.stppp function*

Description

This function computes an appropriate observation window on which to perform prediction. Since the FFT grid must have dimension 2^M by 2^N for some M and N, the window `xyt$window`, is extended to allow this to be fit in for a given cell width.

Usage

```
## S3 method for class 'stppp'
selectObsWindow(xyt, cellwidth, ...)
```

Arguments

`xyt`	an object of class stppp
`cellwidth`	size of the grid spacing in chosen units (equivalent to the cell width argument in lgcpPredict)
`...`	additional arguments

Details

!!NOTE!! that this function also returns the grid ($xvals and $yvals) on which the FFT (and hence MALA) will be performed. It is useful to define spatialAtRiskobjects on this grid to prevent loss of information from the bilinear interpolation that takes place as part of the fitting algorithm.

Value

a resized stppp object together with grid sizes M and N ready for FFT, together with the FFT grid locations, can be useful for estimating lambda(s)

See Also

spatialAtRisk

serr — *serr function*

Description

Generic function to return standard error of relative risk.

Usage

```
serr(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

method serr

See Also

lgcpPredict, serr.lgcpPredict

serr.lgcpPredict — *serr.lgcpPredict function*

Description

Accessor function returning the standard error of relative risk as an lgcpgrid object.

Usage

```
## S3 method for class 'lgcpPredict'
serr(obj, ...)
```

Arguments

obj	an lgcpPredict object
...	additional arguments

Value

Standard error of the relative risk as computed by MCMC.

See Also

lgcpPredict

setoutput *setoutput function*

Description

Sets output functionality for lgcpPredict via the main functions dump2dir and MonteCarloAverage. Note that it is possible for the user to create their own `gridfunction` and `gridmeans` schemes.

Usage

```
setoutput(gridfunction = NULL, gridmeans = NULL)
```

Arguments

`gridfunction`	what to do with the latent field, but default this set to nothing, but could save output to a directory, see ?dump2dir
`gridmeans`	list of Monte Carlo averages to compute, see ?MonteCarloAverage

Value

output parameters

See Also

lgcpPredict, dump2dir, MonteCarloAverage

setTxtProgressBar2 *set the progress bar*

Description

update a text progress bar. See help(txtProgressBar) for more info.

Usage

```
setTxtProgressBar2(pb, value, title = NULL, label = NULL)
```

Arguments

`pb`	text progress bar object
`value`	new value
`title`	ignored
`label`	text for end of progress bar

showGrid — *showGrid function*

Description

Generic method for displaying the FFT grid used in computation.

Usage

```
showGrid(x, ...)
```

Arguments

`x`	an object
`...`	additional arguments

Value

generic function returning method showGrid

See Also

showGrid.default, showGrid.lgcpPredict, showGrid.stppp

showGrid.default — *showGrid.default function*

Description

Default method for printing a grid to a screen. Arguments are vectors giving the x any y coordinates of the centroids.

Usage

```
## Default S3 method:
showGrid(x, y, ...)
```

Arguments

`x`	an vector of grid values for the x coordinates
`y`	an vector of grid values for the y coordinates
`...`	additional arguments passed to points

Value

plots grid centroids on the current graphics device

See Also

showGrid.lgcpPredict, showGrid.stppp

---

showGrid.lgcpPredict *showGrid.lgcpPredict function*

Description

This function displays the FFT grid used on a plot of an `lgcpPredict` object. First plot the object using for example `plot(lg)`, where `lg` is an object of class `lgcpPredict`, then for any of the plots produced, a call to `showGrid(lg,pch=="+",cex=0.5)` will display the centroids of the FFT grid.

Usage

```
## S3 method for class 'lgcpPredict'
showGrid(x, ...)
```

Arguments

`x`	an object of class lgcpPredict
`...`	additional arguments passed to points

Value

plots grid centroids on the current graphics device

See Also

lgcpPredict, showGrid.default, showGrid.stppp

---

showGrid.stppp *showGrid.stppp function*

Description

If an stppp object has been created via simulation, ie using the function `lgcpSim`, then this function will display the grid centroids that were used in the simulation

Usage

```
## S3 method for class 'stppp'
showGrid(x, ...)
```

Arguments

`x`	an object of class stppp. Note this function oly applies to SIMULATED data.
`...`	additional arguments passed to points

Value

plots grid centroids on the current graphics device. FOR SIMULATED DATA ONLY.

See Also

lgcpSim, showGrid.default, showGrid.lgcpPredict

Examples

```
## Not run: xyt <- lgcpSim()
## Not run: plot(xyt)
## Not run: showGrid(xyt,pch="+",cex=0.5)
```

`smultiply.list` *smultiply.list function*

Description

This function multiplies each element of a list by a scalar constant.

Usage

```
smultiply.list(list, const)
```

Arguments

`list`	a list of objects that could be summed using "+"
`const`	a numeric constant

Value

a list with ith entry the scalar multiple of const * list[[i]]

sparsebase *sparsebase function*

Description

A function that returns the full precision matrix in sparse format from the base of a block circulant matrix, see ?Matrix::sparseMatrix

Usage

```
sparsebase(base)
```

Arguments

base	base matrix of a block circulant matrix

Value

...

spatialAtRisk *spatialAtRisk function*

Description

The methods for this generic function:spatialAtRisk.default, spatialAtRisk.fromXYZ, spatialAtRisk.im, spatialAtRisk.function, spatialAtRisk.SpatialGridDataFrame, spatialAtRisk.SpatialPolygonsDataFrame and spatialAtRisk.bivden are used to represent the fixed spatial component, lambda(s) in the log-Gaussian Cox process model. Typically lambda(s) would be represented as a spatstat object of class im, that encodes population density information. However, regardless of the physical interpretation of lambda(s), in lgcp we assume that it integrates to 1 over the observation window. The above methods make sure this condition is satisfied (with the exception of the method for objects of class function), as well as providing a framework for manipulating these structures. lgcp uses bilinear interpolation to project a user supplied lambda(s) onto a discrete grid ready for inference via MCMC, this grid can be obtained via the selectObsWindow function.

Usage

```
spatialAtRisk(X, ...)
```

Arguments

X	an object
...	additional arguments

Details

Generic function used in the construction of spatialAtRisk objects. The class of spatialAtRisk objects provide a framework for describing the spatial inhomogeneity of the at-risk population, lambda(s). This is in contrast to the class of temporalAtRisk objects, which describe the global levels of the population at risk, mu(t).

Unless the user has specified lambda(s) directly by an R function (a mapping the from the real plane onto the non-negative real numbers, see ?spatialAtRisk.function), then it is only necessary to describe the population at risk up to a constant of proportionality, as the routines automatically normalise the lambda provided to integrate to 1.

For reference purposes, the following is a mathematical description of a log-Gaussian Cox Process, it is best viewed in the pdf version of the manual.

Let $\mathcal{Y}(s,t)$ be a spatiotemporal Gaussian process, $W \subset R^2$ be an observation window in space and $T \subset R_{\geq 0}$ be an interval of time of interest. Cases occur at spatio-temporal positions $(x,t) \in W \times T$ according to an inhomogeneous spatio-temporal Cox process, i.e. a Poisson process with a stochastic intensity $R(x,t)$, The number of cases, $X_{S,[t_1,t_2]}$, arising in any $S \subseteq W$ during the interval $[t_1,t_2] \subseteq T$ is then Poisson distributed conditional on $R(\cdot)$,

$$X_{S,[t_1,t_2]} \sim \text{Poisson}\left\{\int_S \int_{t_1}^{t_2} R(s,t) ds dt\right\}$$

Following Brix and Diggle (2001) and Diggle et al (2005), the intensity is decomposed multiplicatively as

$$R(s,t) = \lambda(s)\mu(t)\exp\{\mathcal{Y}(s,t)\}.$$

In the above, the fixed spatial component, $\lambda : R^2 \mapsto R_{\geq 0}$, is a known function, proportional to the population at risk at each point in space and scaled so that

$$\int_W \lambda(s) ds = 1,$$

whilst the fixed temporal component, $\mu : R_{\geq 0} \mapsto R_{\geq 0}$, is also a known function with

$$\mu(t)\delta t = E[X_{W,\delta t}],$$

for t in a small interval of time, δt, over which the rate of the process over W can be considered constant.

Value

method spatialAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

selectObsWindow lgcpPredict, linklgcpSim, spatialAtRisk.default, spatialAtRisk.fromXYZ, spatialAtRisk.im, spatialAtRisk.function, spatialAtRisk.SpatialGridDataFrame, spatialAtRisk.SpatialPolygonsDataFrame, spatialAtRisk.bivden

spatialAtRisk.bivden *spatialAtRisk.bivden function*

Description

Creates a spatialAtRisk object from a sparr bivden object

Usage

```
## S3 method for class 'bivden'
spatialAtRisk(X, ...)
```

Arguments

X	a bivden object
...	additional arguments

Value

object of class spatialAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

lgcpPredict, linklgcpSim, spatialAtRisk.default, spatialAtRisk.fromXYZ, spatialAtRisk.im, spatialAtRisk.function, spatialAtRisk.SpatialGridDataFrame, spatialAtRisk.SpatialPolygonsDataFrame

spatialAtRisk.default *spatialAtRisk.default function*

Description

The default method for creating a spatialAtRisk object, which attempts to extract x, y and Zm values from the object using `xvals`, `yvals` and `zvals`.

Usage

```
## Default S3 method:
spatialAtRisk(X, ...)
```

Arguments

`X`	an object
`...`	additional arguments

Value

object of class spatialAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

lgcpPredict, linklgcpSim, spatialAtRisk.fromXYZ, spatialAtRisk.im, spatialAtRisk.function, spatialAtRisk.SpatialGridDataFrame, spatialAtRisk.SpatialPolygonsDataFrame, spatialAtRisk.bivden, `xvals`, `yvals`, `zvals`

`spatialAtRisk.fromXYZ` *spatialAtRisk.fromXYZ function*

Description

Creates a spatialAtRisk object from a list of X, Y, Zm giving respectively the x and y coordinates of the grid and the 'z' values ie so that Zm[i,j] is proportional to the at-risk population at X[i], Y[j].

Usage

```
## S3 method for class 'fromXYZ'
spatialAtRisk(X, Y, Zm, ...)
```

Arguments

`X`	vector of x-coordinates
`Y`	vector of y-coordinates
`Zm`	matrix such that Zm[i,j] = f(x[i],y[j]) for some function f
`...`	additional arguments

Value

object of class spatialAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

lgcpPredict, linklgcpSim, spatialAtRisk.default, spatialAtRisk.im, spatialAtRisk.function, spatialAtRisk.SpatialGridDataFrame, spatialAtRisk.SpatialPolygonsDataFrame, spatialAtRisk.bivden

---

`spatialAtRisk.function` *spatialAtRisk.function function*

Description

Creates a spatialAtRisk object from a function mapping R^2 onto the non negative reals. Note that for spatialAtRisk objects defined in this manner, the user is responsible for ensurng that the integral of the function is 1 over the observation window of interest.

Usage

```
## S3 method for class '`function`'
spatialAtRisk(X, warn = TRUE, ...)
```

Arguments

`X`	a function with accepts arguments x and y that returns the at risk population at coordinate (x,y), which should be a numeric of length 1
`warn`	whether to issue a warning or not
`...`	additional arguments

Value

object of class spatialAtRisk NOTE The function provided is assumed to integrate to 1 over the observation window, the user is responsible for ensuring this is the case.

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

lgcpPredict, linklgcpSim, spatialAtRisk.default, spatialAtRisk.fromXYZ, spatialAtRisk.im, spatialAtRisk.SpatialGridDataFrame, spatialAtRisk.SpatialPolygonsDataFrame, spatialAtRisk.bivden

spatialAtRisk.im *spatialAtRisk.im function*

Description

Creates a spatialAtRisk object from a spatstat pixel image (im) object.

Usage

```
## S3 method for class 'im'
spatialAtRisk(X, ...)
```

Arguments

`X`	object of class im
`...`	additional arguments

Value

object of class spatialAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

lgcpPredict, linklgcpSim, spatialAtRisk.default, spatialAtRisk.fromXYZ, spatialAtRisk.function, spatialAtRisk.SpatialGridDataFrame, spatialAtRisk.SpatialPolygonsDataFrame, spatialAtRisk.bivden

spatialAtRisk.lgcpgrid *spatialAtRisk.lgcpgrid function*

Description

Creates a spatialAtRisk object from an lgcpgrid object

Usage

```
## S3 method for class 'lgcpgrid'
spatialAtRisk(X, idx = length(X$grid), ...)
```

Arguments

X	an lgcpgrid object
idx	in the case that X$grid is a list of length > 1, this argument specifies which element of the list to convert. By default, it is the last.
...	additional arguments

Value

object of class spatialAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

lgcpPredict, linklgcpSim, spatialAtRisk.default, spatialAtRisk.fromXYZ, spatialAtRisk.im, spatialAtRisk.function, spatialAtRisk.SpatialGridDataFrame, spatialAtRisk.SpatialPolygonsDataFrame

---

spatialAtRisk.SpatialGridDataFrame

spatialAtRisk.SpatialGridDataFrame function

Description

Creates a spatialAtRisk object from an sp SpatialGridDataFrame object

Usage

```
## S3 method for class 'SpatialGridDataFrame'
spatialAtRisk(X, ...)
```

Arguments

X	a SpatialGridDataFrame object
...	additional arguments

Value

object of class spatialAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

lgcpPredict, linklgcpSim, spatialAtRisk.default, spatialAtRisk.fromXYZ, spatialAtRisk.im, spatialAtRisk.function, spatialAtRisk.SpatialPolygonsDataFrame, spatialAtRisk.bivden

---

`spatialAtRisk.SpatialPolygonsDataFrame`
spatialAtRisk.SpatialPolygonsDataFrame function

Description

Creates a spatialAtRisk object from a SpatialPolygonsDataFrame object.

Usage

```
## S3 method for class 'SpatialPolygonsDataFrame'
spatialAtRisk(X, ...)
```

Arguments

`X`	a SpatialPolygonsDataFrame object; one column of the data frame should have name "atrisk", containing the aggregate population at risk for that region
`...`	additional arguments

Value

object of class spatialAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

lgcpPredict, linklgcpSim, spatialAtRisk.default, spatialAtRisk.fromXYZ, spatialAtRisk.im, spatialAtRisk.function, spatialAtRisk.SpatialGridDataFrame, spatialAtRisk.bivden

spatialIntensities *spatialIntensities function*

Description

Generic method for extracting spatial intensities.

Usage

```
spatialIntensities(X, ...)
```

Arguments

X	an object
...	additional arguments

Value

method spatialintensities

See Also

spatialIntensities.fromXYZ, spatialIntensities.fromSPDF

spatialIntensities.fromSPDF *spatialIntensities.fromSPDF function*

Description

Extract the spatial intensities from an object of class fromSPDF (as would have been created by spatialAtRisk.SpatialPolygonsDataFrame for example).

Usage

```
## S3 method for class 'fromSPDF'
spatialIntensities(X, xyt, ...)
```

Arguments

X	an object of class fromSPDF
xyt	object of class stppp or a list object of numeric vectors with names $x, $y
...	additional arguments

Value

normalised spatial intensities

See Also

spatialIntensities, spatialIntensities.fromXYZ

`spatialIntensities.fromXYZ`
spatialIntensities.fromXYZ function

Description

Extract the spatial intensities from an object of class fromXYZ (as would have been created by spatialAtRisk for example).

Usage

```
## S3 method for class 'fromXYZ'
spatialIntensities(X, xyt, ...)
```

Arguments

`X`	object of class fromXYZ
`xyt`	object of class stppp or a list object of numeric vectors with names $x, $y
`...`	additional arguments

Value

normalised spatial intensities

See Also

spatialIntensities, spatialIntensities.fromSPDF spatialparsEst *spatialparsEst function*

Description

Having estimated either the pair correlation or K functions using respectively ginhomAverage or KinhomAverage, the spatial parameters sigma and phi can be estimated. This function provides a visual tool for this estimation procedure.

Usage

```
spatialparsEst(
  gk,
  sigma.range,
  phi.range,
  spatial.covmodel,
  covpars = c(),
  guess = FALSE
)
```

Arguments

gk	an R object; output from the function KinhomAverage or ginhomAverage
sigma.range	range of sigma values to consider
phi.range	range of phi values to consider
spatial.covmodel	correlation type see ?CovarianceFct
covpars	vector of additional parameters for certain classes of covariance function (eg Matern), these must be supplied in the order given in ?CovarianceFct
guess	logical. Perform an initial guess at paramters? Alternative (the default) sets initial values in the middle of sigma.range and phi.range. NOTE: automatic parameter estimation can be can be unreliable.

Details

To get a good choice of parameters, it is likely that the routine will have to be called several times in order to refine the choice of sigma.range and phi.range.

Value

rpanel function to help choose sigma nad phi by eye

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Baddeley AJ, Moller J, Waagepetersen R (2000). Non-and semi-parametric estimation of interaction in inhomogeneous point patterns. Statistica Neerlandica, 54, 329-350.
3. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
4. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

ginhomAverage, KinhomAverage, thetaEst, lambdaEst, muEst

`SpatialPolygonsDataFrame.stapp`

SpatialPolygonsDataFrame.stapp function

Description

A function to return the SpatialPolygonsDataFrame part of an stapp object

Usage

```
SpatialPolygonsDataFrame.stapp(from)
```

Arguments

`from`	stapp object

Value

an object of class SpatialPolygonsDataFrame

`SpikedExponentialCovFct` *SpikedExponentialCovFct function*

Description

A function to declare and also evaluate a spiked exponential covariance function. Note that the present version of lgcp only offers estimation for sigma and phi, the additional parameter 'spikevar' is treated as fixed.

Usage

```
SpikedExponentialCovFct(d, CovParameters, spikevar = 1)
```

Arguments

`d`	toral distance
`CovParameters`	parameters of the latent field, an object of class "CovParamaters".
`spikevar`	the additional variance at distance 0

Value

the spiked exponential covariance function; note that the spikevariance is currently not estimated as part of the MCMC routine, and is thus treated as a fixed parameter.

See Also

CovFunction.function, exponentialCovFct, RandomFieldsCovFct

`stapp` *stapp function*

Description

Generic function for space-time aggregated point-process data

Usage

```
stapp(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method stapp

`stapp.list` *stapp.list function*

Description

A wrapper function for stapp.SpatialPolygonsDataFrame

Usage

```
## S3 method for class 'list'
stapp(obj, ...)
```

Arguments

`obj`	an list object as described above, see ?stapp.SpatialPolygonsDataFrame for further details on the requirements of the list
`...`	additional arguments

Details

Construct a space-time aggregated point-process (stapp) object from a list object. The first element of the list should be a SpatialPolygonsDataFrame, the second element of the list a counts matrix, the third element of the list a vector of times, the fourth element a vector giving the bounds of the temporal observation window and the fifth element a spatstat owin object giving the spatial observation window.

Value

an object of class stapp

`stapp.SpatialPolygonsDataFrame`
stapp.SpatialPolygonsDataFrame function

Description

Construct a space-time aggregated point-process (stapp) object from a SpatialPolygonsDataFrame (along with some other info)

Usage

```
## S3 method for class 'SpatialPolygonsDataFrame'
stapp(obj, counts, t, tlim, window, ...)
```

Arguments

obj	an SpatialPolygonsDataFrame object
counts	a (length(t) by N) matrix containing aggregated case counts for each of the geographical regions defined by the SpatialPolygonsDataFrame, where N is the number of regions
t	vector of times, for each element of t there should correspond a column in the matrix 'counts'
tlim	vector giving the upper and lower bounds of the temporal observation window
window	the observation window, of class owin, see ?owin
...	additional arguments

Value

an object of class stapp

---

stGPrealisation *stGPrealisation function*

Description

A function to store a realisation of a spatiotemporal gaussian process for use in MCMC algorithms that include Bayesian parameter estimation. Stores not only the realisation, but also computational quantities.

Usage

```
stGPrealisation(gamma, fftgrid, covFunction, covParameters, d, tdiff)
```

Arguments

gamma	the transformed (white noise) realisation of the process
fftgrid	an object of class FFTgrid, see ?genFFTgrid
covFunction	an object of class function returning the spatial covariance
covParameters	an object of class CovParamaters, see ?CovParamaters
d	matrix of grid distances
tdiff	vector of time differences

Value

a realisation of a spatiotemporal Gaussian process on a regular grid stppp *stppp function*

Description

Generic function used in the construction of space-time planar point patterns. An stppp object is like a ppp object, but with extra components for (1) a vector giving the time at whcih the event occurred and (2) a time-window over which observations occurred. Observations are assumed to occur in the plane and the observation window is assumed not to change over time.

Usage

```
stppp(P, ...)
```

Arguments

P	an object
...	additional arguments

Value

method stppp

See Also

stppp, stppp.ppp, stppp.list

stppp.list *stppp.list function*

Description

Construct a space-time planar point pattern from a list object

Usage

```
## S3 method for class 'list'
stppp(P, ...)
```

Arguments

P	list object containing $data, an (n x 3) matrix corresponding to (x,y,t) values; $tlim, a vector of length 2 givign the observation time window; and $window giving an owin spatial observation winow, see ?owin for more details
...	additional arguments

Value

an object of class stppp

See Also

stppp, stppp.ppp,

---

stppp.ppp *stppp.ppp function*

Description

Construct a space-time planar point pattern from a ppp object

Usage

```
## S3 method for class 'ppp'
stppp(P, t, tlim, ...)
```

Arguments

`P`	a spatstat ppp object
`t`	a vector of length P$n
`tlim`	a vector of length 2 specifying the observation time window
`...`	additional arguments

Value

an object of class stppp

See Also

stppp, stppp.list

summary.lgcpgrid *summary.lgcpgrid function*

Description

Summary method for lgcp objects. This just applies the summary function to each of the elements of object$grid.

Usage

```
## S3 method for class 'lgcpgrid'
summary(object, ...)
```

Arguments

`object`	an object of class lgcpgrid
`...`	other arguments

Value

Summary per grid, see ?summary for further options

See Also

lgcpgrid.list, lgcpgrid.array, as.list.lgcpgrid, print.lgcpgrid, quantile.lgcpgrid, image.lgcpgrid, plot.lgcpgrid

summary.mcmc *summary.mcmc function*

Description

summary of an mcmc iterator print out values of an iterator and reset it. DONT call this in a loop that uses this iterator - it will reset it. And break.

Usage

```
## S3 method for class 'mcmc'
summary(object, ...)
```

Arguments

`object`	an mcmc iterator
`...`	other args

target.and.grad.AggregateSpatialPlusPars

target.and.grad.AggregateSpatialPlusPars function

Description

A function to compute the target and gradient for the Bayesian aggregated point process model. Not for general use.

Usage

```
target.and.grad.AggregateSpatialPlusPars(
  GP,
  prior,
  Z,
  Zt,
  eta,
  beta,
  nis,
  cellarea,
  spatial,
  gradtrunc
)
```

Arguments

GP	an object constructed using GPrealisation
prior	the prior, created using lgcpPrior
Z	the design matrix on the full FFT grid
Zt	the transpose of the design matrix
eta	the model parameter, eta
beta	the model parameters, beta
nis	cell counts on the FFT grid
cellarea	the cell area
spatial	the poisson offset
gradtrunc	the gradient truncation parameter

Value

the target and gradient

target.and.grad.MultitypespatialPlusPars
target.and.grad.MultitypespatialPlusPars function

Description

A function to compute the taget an gradient for the Bayesian multivariate lgcp

Usage

```
target.and.grad.MultitypespatialPlusPars(
  GPlist,
  priorlist,
  Zlist,
  Ztlist,
  eta,
  beta,
  nis,
  cellarea,
  spatial,
  gradtrunc
)
```

Arguments

GPlist	list of Gaussian processes
priorlist	list of priors
Zlist	list of design matrices on the FFT gridd
Ztlist	list of transposed design matrices
eta	LGCP model parameter eta
beta	LGCP model parameter beta
nis	matrix of cell counts on the extended grid
cellarea	the cell area
spatial	the poisson offset interpolated onto the correcy grid
gradtrunc	gradient truncation paramter

Value

the target and gradient

target.and.grad.spatial

target.and.grad.spatial function

Description

A function to compute the target and gradient for 'spatial only' MALA

Usage

```
target.and.grad.spatial(
  Gamma,
  nis,
  cellarea,
  rootQeigs,
  invrootQeigs,
  mu,
  spatial,
  logspat,
  scaleconst,
  gradtrunc
)
```

Arguments

Gamma	current state of the chain, Gamma
nis	matrix of cell counts
cellarea	area of cells, a positive number
rootQeigs	square root of the eigenvectors of the precision matrix
invrootQeigs	inverse square root of the eigenvectors of the precision matrix
mu	parameter of the latent Gaussian field
spatial	spatial at risk function, lambda, interpolated onto correct grid
logspat	log of spatial at risk function, lambda*scaleconst, interpolated onto correct grid
scaleconst	the expected number of cases
gradtrunc	gradient truncation parameter

Value

the back-transformed Y, its exponential, the log-target and gradient for use in MALAlgcpSpatial

---

`target.and.grad.spatialPlusPars`
target.and.grad.spatialPlusPars function

Description

A function to compute the target and gradient for the Bayesian spatial LGCP

Usage

```
target.and.grad.spatialPlusPars(
  GP,
  prior,
  Z,
  Zt,
  eta,
  beta,
  nis,
  cellarea,
  spatial,
  gradtrunc
)
```

Arguments

`GP`	an object created using GPrealisation
`prior`	the model priors, created using lgcpPrior
`Z`	the design matrix on the FFT grid
`Zt`	transpose of the design matrix
`eta`	the paramters, eta
`beta`	the parameters, beta
`nis`	cell counts on the FFT grid
`cellarea`	the cell area
`spatial`	poisson offset
`gradtrunc`	the gradient truncation parameter

Value

the target and graient for this model

`target.and.grad.spatiotemporal` *target.and.grad.spatiotemporal function*

Description

A function to compute the target and gradient for 'spatial only' MALA

Usage

```
target.and.grad.spatiotemporal(
  Gamma,
  nis,
  cellarea,
  rootQeigs,
  invrootQeigs,
  mu,
  spatial,
  logspat,
  temporal,
  bt,
  gt,
  gradtrunc
)
```

Arguments

`Gamma`	current state of the chain, Gamma
`nis`	matrix of cell counts
`cellarea`	area of cells, a positive number
`rootQeigs`	square root of the eigenvectors of the precision matrix
`invrootQeigs`	inverse square root of the eigenvectors of the precision matrix
`mu`	parameter of the latent Gaussian field
`spatial`	spatial at risk function, lambda, interpolated onto correct grid
`logspat`	log of spatial at risk function, lambda*scaleconst, interpolated onto correct grid
`temporal`	fitted temoporal values
`bt`	in Brix and Diggle vector b(delta t)
`gt`	in Brix and Diggle vector g(delta t) (ie the coefficient of R in G(t)), with convention that (deltat[1])=Inf
`gradtrunc`	gradient truncation parameter

Value

the back-transformed Y, its exponential, the log-target and gradient for use in MALAlgcp

---

`target.and.grad.SpatioTemporalPlusPars`
target.and.grad.SpatioTemporalPlusPars function

Description

A function to compute the target and gradient for the Bayesian spatiotemporal LGCP.

Usage

```
target.and.grad.SpatioTemporalPlusPars(
  GP,
  prior,
  Z,
  Zt,
  eta,
  beta,
  nis,
  cellarea,
  spatial,
  gradtrunc,
  ETA0,
  tdiff
)
```

Arguments

`GP`	an object created using the stGPrealisation function
`prior`	the priors for hte model, created using lgcpPrior
`Z`	the design matrix on the FFT grid
`Zt`	the transpose of the design matrix
`eta`	the paramers eta
`beta`	the parameters beta
`nis`	the cell counts on the FFT grid
`cellarea`	the cell area
`spatial`	the poisson offset
`gradtrunc`	the gradient truncation parameter
`ETA0`	the initial value of eta
`tdiff`	vector of time differences between time points

Value

the target and gradient for the spatiotemporal model.

temporalAtRisk *temporalAtRisk function*

Description

Generic function used in the construction of temporalAtRisk objects. A temporalAtRisk object describes the at risk population globally in an observation time window [t_1,t_2]. Therefore, for any t in [t_1,t_2], a temporalAtRisk object should be able to return the global at risk population, mu(t) = E(number of cases in the unit time interval containing t). This is in contrast to the class of spatialAtRisk objects, which describe the spatial inhomogeneity in the population at risk, lambda(s).

Usage

```
temporalAtRisk(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Details

Note that in the prediction routine, lgcpPredict, and the simulation routine, lgcpSim, time discretisation is achieved using `as.integer` on both observation times and time limits t_1 and t_2 (which may be stored as non-integer values). The functions that create temporalAtRisk objects therefore return piecewise cconstant step-functions. that can be evaluated for any real t in [t_1,t_2], but with the restriction that mu(t_i) = mu(t_j) whenever `as.integer(t_i)==as.integer(t_j)`.

A temporalAtRisk object may be (1) 'assumed known', or (2) scaled to a particular dataset. In the latter case, in the routines available (temporalAtRisk.numeric and temporalAtRisk.function), the stppp dataset of interest should be referenced, in which case the scaling of mu(t) will be done automatically. Otherwise, for example for simulation purposes, no scaling of mu(t) occurs, and it is assumed that the mu(t) corresponds to the expected number of cases during the unit time interval containnig t. For reference purposes, the following is a mathematical description of a log-Gaussian Cox Process, it is best viewed in the pdf version of the manual.

Let $\mathcal{Y}(s,t)$ be a spatiotemporal Gaussian process, $W \subset R^2$ be an observation window in space and $T \subset R_{\geq 0}$ be an interval of time of interest. Cases occur at spatio-temporal positions $(x,t) \in W \times T$ according to an inhomogeneous spatio-temporal Cox process, i.e. a Poisson process with a stochastic intensity $R(x,t)$, The number of cases, $X_{S,[t_1,t_2]}$, arising in any $S \subseteq W$ during the interval $[t_1,t_2] \subseteq T$ is then Poisson distributed conditional on $R(\cdot)$,

$$X_{S,[t_1,t_2]} \sim \text{Poisson}\left\{\int_S \int_{t_1}^{t_2} R(s,t)dsdt\right\}$$

Following Brix and Diggle (2001) and Diggle et al (2005), the intensity is decomposed multiplicatively as

$$R(s,t) = \lambda(s)\mu(t)\exp\{\mathcal{Y}(s,t)\}.$$

In the above, the fixed spatial component, $\lambda : R^2 \mapsto R_{\geq 0}$, is a known function, proportional to the population at risk at each point in space and scaled so that

$$\int_W \lambda(s)ds = 1,$$

whilst the fixed temporal component, $\mu : R_{\geq 0} \mapsto R_{\geq 0}$, is also a known function with

$$\mu(t)\delta t = E[X_{W,\delta t}],$$

for t in a small interval of time, δt, over which the rate of the process over W can be considered constant.

Value

method temporalAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

spatialAtRisk, lgcpPredict, lgcpSim, temporalAtRisk.numeric, temporalAtRisk.function, constantInTime, constantInTime.numeric, constantInTime.stppp, print.temporalAtRisk, plot.temporalAtRisk

---

`temporalAtRisk.function`

temporalAtRisk.function function

Description

Create a temporalAtRisk object from a function.

Usage

```
## S3 method for class '`function`'
temporalAtRisk(obj, tlim, xyt = NULL, warn = TRUE, ...)
```

Arguments

`obj`	a function accepting single, scalar, numeric argument, t, that returns the temporal intensity for time t
`tlim`	an integer vector of length 2 giving the time limits of the observation window
`xyt`	an object of class stppp. If NULL (default) then the function returned is not scaled. Otherwise, the function is scaled so that f(t) = expected number of counts at time t.
`warn`	Issue a warning if the given temporal intensity treated is treated as 'known'?
`...`	additional arguments

Details

Note that in the prediction routine, lgcpPredict, and the simulation routine, lgcpSim, time discretisation is achieved using `as.integer` on both observation times and time limits t_1 and t_2 (which may be stored as non-integer values). The functions that create temporalAtRisk objects therefore return piecewise cconstant step-functions. that can be evaluated for any real t in [t_1,t_2], but with the restriction that mu(t_i) = mu(t_j) whenever `as.integer(t_i)==as.integer(t_j)`.

A temporalAtRisk object may be (1) 'assumed known', corresponding to the default argument `xyt=NULL`; or (2) scaled to a particular dataset (argument xyt=[stppp object of interest]). In the latter case, in the routines available (temporalAtRisk.numeric and temporalAtRisk.function), the dataset of interest should be referenced, in which case the scaling of mu(t) will be done automatically. Otherwise, for example for simulation purposes, no scaling of mu(t) occurs, and it is assumed that the mu(t) corresponds to the expected number of cases during the unit time interval containnig t.

Value

a function f(t) giving the temporal intensity at time t for integer t in the interval [tlim[1],tlim[2]] of class temporalAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

temporalAtRisk, spatialAtRisk, temporalAtRisk.numeric, constantInTime, constantInTime.numeric, constantInTime.stppp, print.temporalAtRisk, plot.temporalAtRisk

---

`temporalAtRisk.numeric` *temporalAtRisk.numeric function*

Description

Create a temporalAtRisk object from a numeric vector.

Usage

```
## S3 method for class 'numeric'
temporalAtRisk(obj, tlim, xyt = NULL, warn = TRUE, ...)
```

Arguments

`obj`	a numeric vector of length (tlim[2]-tlim[1] + 1) giving the temporal intensity up to a constant of proportionality at each integer time within the interval defined by tlim
`tlim`	an integer vector of length 2 giving the time limits of the observation window
`xyt`	an object of class stppp. If NULL (default) then the function returned is not scaled. Otherwise, the function is scaled so that f(t) = expected number of counts at time t.
`warn`	Issue a warning if the given temporal intensity treated is treated as 'known'?
`...`	additional arguments

Details

Note that in the prediction routine, lgcpPredict, and the simulation routine, lgcpSim, time discretisation is achieved using `as.integer` on both observation times and time limits t_1 and t_2 (which may be stored as non-integer values). The functions that create temporalAtRisk objects therefore return piecewise constant step-functions that can be evaluated for any real t in [t_1,t_2], but with the restriction that mu(t_i) = mu(t_j) whenever `as.integer(t_i)==as.integer(t_j)`.

A temporalAtRisk object may be (1) 'assumed known', corresponding to the default argument `xyt=NULL`; or (2) scaled to a particular dataset (argument xyt=[stppp object of interest]). In the latter case, in the routines available (temporalAtRisk.numeric and temporalAtRisk.function), the dataset of interest should be referenced, in which case the scaling of mu(t) will be done automatically. Otherwise, for example for simulation purposes, no scaling of mu(t) occurs, and it is assumed that the mu(t) corresponds to the expected number of cases during the unit time interval containing t.

Value

a function f(t) giving the temporal intensity at time t for integer t in the interval as.integer([tlim[1],tlim[2]]) of class temporalAtRisk

1. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
2. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

temporalAtRisk, spatialAtRisk, temporalAtRisk.function, constantInTime, constantInTime.numeric, constantInTime.stppp, print.temporalAtRisk, plot.temporalAtRisk `tempRaster` *tempRaster function*

Description

A function to create a temporary raster object from an x-y regular grid of cell centroids. Useful for projection from one raster to another.

Usage

```
tempRaster(mcens, ncens)
```

Arguments

`mcens`	vector of equally-spaced coordinates of cell centroids in x-direction
`ncens`	vector of equally-spaced coordinates of cell centroids in y-direction

Value

an empty raster object

`textsummary` *textsummary function*

Description

A function to print a text description of the inferred paramerers beta and eta from a call to the function lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars or lgcpPredictMultitypeSpatialPlusPars

Usage

```
textsummary(obj, digits = 3, scientific = -3, inclIntercept = FALSE, ...)
```

Arguments

`obj`	an object produced by a call to lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars orlgcpPredictMultitypeSpatialPlusPars
`digits`	see the option "digits" in ?format
`scientific`	see the option "scientific" in ?format
`inclIntercept`	logical: whether to summarise the intercept term, default is FALSE.
`...`	other arguments passed to the function "format"

Value

A text summary, that can be pasted into a LaTeX document and later edited.

See Also

ltar, autocorr, parautocorr, traceplots, parsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

thetaEst *thetaEst function*

Description

A tool to visually estimate the temporal correlation parameter theta; note that sigma and phi must have first been estiamted.

Usage

```
thetaEst(
  xyt,
  spatial.intensity = NULL,
  temporal.intensity = NULL,
  sigma,
  phi,
  theta.range = c(0, 10),
  N = 100,
  spatial.covmodel = "exponential",
  covpars = c()
)
```

Arguments

`xyt`	object of class stppp
`spatial.intensity`	A spatial at risk object OR a bivariate density estimate of lambda, an object of class im (produced from density.ppp for example),
`temporal.intensity`	either an object of class temporalAtRisk, or one that can be coerced into that form. If NULL (default), this is estimated from the data, seee ?muEst
`sigma`	estimate of parameter sigma
`phi`	estimate of parameter phi
`theta.range`	range of theta values to consider
`N`	number of integration points in computation of C(v,beta) (see Brix and Diggle 2003, corrigendum to Brix and Diggle 2001)
`spatial.covmodel`	spatial covariance model
`covpars`	additional covariance parameters

Value

An r panel tool for visual estimation of temporal parameter theta NOTE if lambdaEst has been invoked to estimate lambda, then the returned density should be passed to thetaEst as the argument spatial.intensity

References

1. Benjamin M. Taylor, Tilman M. Davies, Barry S. Rowlingson, Peter J. Diggle (2013). Journal of Statistical Software, 52(4), 1-40. URL http://www.jstatsoft.org/v52/i04/
2. Brix A, Diggle PJ (2001). Spatiotemporal Prediction for log-Gaussian Cox processes. Journal of the Royal Statistical Society, Series B, 63(4), 823-841.
3. Diggle P, Rowlingson B, Su T (2005). Point Process Methodology for On-line Spatio-temporal Disease Surveillance. Environmetrics, 16(5), 423-434.

See Also

ginhomAverage, KinhomAverage, spatialparsEst, lambdaEst, muEst

---

toral.cov.mat *toral.cov.mat function*

Description

A function to compute the covariance matrix of a stationary process on a torus.

Usage

```
toral.cov.mat(xg, yg, sigma, phi, model, additionalparameters)
```

Arguments

`xg`	x grid
`yg`	y grid
`sigma`	spatial variability parameter
`phi`	spatial decay parameter
`model`	model for covariance, see ?CovarianceFct
`additionalparameters`	additional parameters for covariance structure

Value

circulant covariacne matrix

touchingowin	*touchingowin function*

Description

A function to compute which cells are touching an owin or spatial polygons object

Usage

```
touchingowin(x, y, w)
```

Arguments

`x`	grid centroids in x-direction note this will be expanded into a GRID of (x,y) values in the function
`y`	grid centroids in y-direction note this will be expanded into a GRID of (x,y) values in the function
`w`	an owin or SpatialPolygons object

Value

vector of TRUE or FALSE according to whether the cell

traceplots	*traceplots function*

Description

A function to produce trace plots for the paramerers beta and eta from a call to the function lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars or lgcpPredictMultitypeSpatialPlusPars

Usage

```
traceplots(obj, xlab = "Sample No.", ylab = NULL, main = "", ask = TRUE, ...)
```

Arguments

`obj`	an object produced by a call to lgcpPredictSpatialPlusPars, lgcpPredictAggregateSpatialPlusPars, lgcpPredictSpatioTemporalPlusPars orlgcpPredictMultitypeSpatialPlusPars
`xlab`	optional label for x-axis, there is a sensible default.
`ylab`	optional label for y-axis, there is a sensible default.
`main`	optional title of the plot, there is a sensible default.
`ask`	the paramter "ask", see ?par
`...`	other arguments passed to the function "hist"

Value

produces MCMC trace plots of the parameters beta and eta

See Also

ltar, autocorr, parautocorr, parsummary, textsummary, priorpost, postcov, exceedProbs, betavals, etavals

---

transblack — *transblack function*

Description

A function to return a transparent black colour.

Usage

```
transblack(alpha = 0.1)
```

Arguments

`alpha`	transparency parameter, see ?rgb

Value

character string of colour

---

transblue — *transblue function*

Description

A function to return a transparent blue colour.

Usage

```
transblue(alpha = 0.1)
```

Arguments

`alpha`	transparency parameter, see ?rgb

Value

character string of colour

transgreen — *transgreen function*

Description

A function to return a transparent green colour.

Usage

```
transgreen(alpha = 0.1)
```

Arguments

alpha	transparency parameter, see ?rgb

Value

character string of colour

transred — *transred function*

Description

A function to return a transparent red colour.

Usage

```
transred(alpha = 0.1)
```

Arguments

alpha	transparency parameter, see ?rgb

Value

character string of colour

txtProgressBar2 — *A text progress bar with label*

Description

This is the base txtProgressBar but with a little modification to implement the label parameter for style=3. For full info see txtProgressBar

Usage

```
txtProgressBar2(
  min = 0,
  max = 1,
  initial = 0,
  char = "=",
  width = NA,
  title = "",
  label = "",
  style = 1
)
```

Arguments

min	min value for bar
max	max value for bar
initial	initial value for bar
char	the character (or character string) to form the progress bar.
width	progress bar width
title	ignored
label	text to put at the end of the bar
style	bar style

updateAMCMC — *updateAMCMC function*

Description

A generic to be used for the purpose of user-defined adaptive MCMC schemes, updateAMCMC tells the MALA algorithm how to update the value of h. See lgcp vignette, codevignette("lgcp"), for further details on writing adaptive MCMC schemes.

Usage

```
updateAMCMC(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

method updateAMCMC

See Also

updateAMCMC.constanth, updateAMCMC.andrieuthomsh

`updateAMCMC.andrieuthomsh`

updateAMCMC.andrieuthomsh function

Description

Updates the andrieuthomsh adaptive scheme.

Usage

```
## S3 method for class 'andrieuthomsh'
updateAMCMC(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

update and return current h for scheme

References

1. Andrieu C, Thoms J (2008). A tutorial on adaptive MCMC. Statistics and Computing, 18(4), 343-373.
2. Robbins H, Munro S (1951). A Stochastic Approximation Methods. The Annals of Mathematical Statistics, 22(3), 400-407.
3. Roberts G, Rosenthal J (2001). Optimal Scaling for Various Metropolis-Hastings Algorithms. Statistical Science, 16(4), 351-367.

See Also

andrieuthomsh

updateAMCMC.constanth *updateAMCMC.constanth function*

Description

Updates the constanth adaptive scheme.

Usage

```
## S3 method for class 'constanth'
updateAMCMC(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

update and return current h for scheme

See Also

constanth

varfield *varfield function*

Description

Generic function to extract the variance of the latent field Y.

Usage

```
varfield(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

method meanfield

See Also

lgcpPredict

varfield.lgcpPredict *varfield.lgcpPredict function*

Description

This is an accessor function for objects of class `lgcpPredict` and returns the variance of the field Y as an lgcpgrid object.

Usage

```
## S3 method for class 'lgcpPredict'
varfield(obj, ...)
```

Arguments

`obj`	an object of class lgcpPredict
`...`	additional arguments

Value

returns the cell-wise variance of Y computed via Monte Carlo.

See Also

lgcpPredict

varfield.lgcpPredictINLA *varfield.lgcpPredictINLA function*

Description

A function to return the variance of the latent field from a call to lgcpPredictINLA output.

Usage

```
## S3 method for class 'lgcpPredictINLA'
varfield(obj, ...)
```

Arguments

`obj`	an object of class lgcpPredictINLA
`...`	other arguments

Value

the variance of the latent field

window.lgcpPredict *window.lgcpPredict function*

Description

Accessor function returning the observation window from objects of class `lgcpPredict`. Note that for computational purposes, the window of an `stppp` object will be extended to accommodate the requirement that the dimensions must be powers of 2. The function `window.lgcpPredict` returns the extended window.

Usage

```
## S3 method for class 'lgcpPredict'
window(x, ...)
```

Arguments

`x`	an object of class lgcpPredict
`...`	additional arguments

Value

returns the observation window used durign computation

See Also

lgcpPredict

wpopdata *Population of Welsh counties*

Description

Population of Welsh counties

Usage

```
data(wpopdata)
```

Format

matrix

Source

ONS

References

http://www.statistics.gov.uk/default.asp

`wtowncoords` *Welsh town details: location*

Description

Welsh town details: location

Usage

```
data(wtowncoords)
```

Format

matrix

Source

Wikipedia

References

https://www.wikipedia.org/

`wtowns` *Welsh town details: population*

Description

Welsh town details: population

Usage

```
data(wtowns)
```

Format

matrix

Source

ONS

References

http://www.statistics.gov.uk/default.asp xvals *xvals function*

Description

Generic for extractign the 'x values' from an object.

Usage

```
xvals(obj, ...)
```

Arguments

obj	an object of class spatialAtRisk
...	additional arguments

Value

the xvals method

See Also

yvals, zvals, xvals.default, yvals.default, zvals.default, xvals.fromXYZ, yvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

xvals.default *xvals.default function*

Description

Default method for extracting 'x values' looks for $X, $x in that order.

Usage

```
## Default S3 method:
xvals(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

the x values

See Also

xvals, yvals, zvals, yvals.default, zvals.default, xvals.fromXYZ, yvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

---

xvals.fromXYZ *xvals.fromXYZ function*

Description

Method for extracting 'x values' from an object of class fromXYZ

Usage

```
## S3 method for class 'fromXYZ'
xvals(obj, ...)
```

Arguments

obj	a spatialAtRisk object
...	additional arguments

Value

the x values

See Also

xvals, yvals, zvals, xvals.default, yvals.default, zvals.default, yvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

---

xvals.lgcpPredict *xvals.lgcpPredict function*

Description

Gets the x-coordinates of the centroids of the prediction grid.

Usage

```
## S3 method for class 'lgcpPredict'
xvals(obj, ...)
```

Arguments

obj	an object of class lgcpPredict
...	additional arguments

Value

the x coordinates of the centroids of the grid

See Also

lgcpPredict

`xvals.SpatialGridDataFrame`

xvals.SpatialGridDataFrame function

Description

Method for extracting 'x values' from an object of class spatialGridDataFrame

Usage

```
## S3 method for class 'SpatialGridDataFrame'
xvals(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

the x values

See Also

xvals, yvals, zvals, xvals.default, yvals.default, zvals.default, xvals.fromXYZ, yvals.fromXYZ, zvals.fromXYZ, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

YfromGamma *YfromGamma function*

Description

A function to change Gammas (white noise) into Ys (spatially correlated noise). Used in the MALA algorithm.

Usage

```
YfromGamma(Gamma, invrootQeigs, mu)
```

Arguments

Gamma	Gamma matrix
invrootQeigs	inverse square root of the eigenvectors of the precision matrix
mu	parameter of the latent Gaussian field

Value

Y

yvals *yvals function*

Description

Generic for extractign the 'y values' from an object.

Usage

```
yvals(obj, ...)
```

Arguments

obj	an object of class spatialAtRisk
...	additional arguments

Value

the yvals method

See Also

xvals, zvals, xvals.default, yvals.default, zvals.default, xvals.fromXYZ, yvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame `yvals.default` *yvals.default function*

Description

Default method for extracting 'y values' looks for $Y, $y in that order.

Usage

```
## Default S3 method:
yvals(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

the y values

See Also

xvals, yvals, zvals, xvals.default, zvals.default, xvals.fromXYZ, yvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

`yvals.fromXYZ` *yvals.fromXYZ function*

Description

Method for extracting 'y values' from an object of class fromXYZ

Usage

```
## S3 method for class 'fromXYZ'
yvals(obj, ...)
```

Arguments

`obj`	a spatialAtRisk object
`...`	additional arguments

Value

the y values

See Also

xvals, yvals, zvals, xvals.default, yvals.default, zvals.default, xvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

---

`yvals.lgcpPredict` *yvals.lgcpPredict function*

Description

Gets the y-coordinates of the centroids of the prediction grid.

Usage

```
## S3 method for class 'lgcpPredict'
yvals(obj, ...)
```

Arguments

`obj`	an object of class lgcpPredict
`...`	additional arguments

Value

the y coordinates of the centroids of the grid

See Also

lgcpPredict

`yvals.SpatialGridDataFrame` *yvals.SpatialGridDataFrame function*

Description

Method for extracting 'y values' from an object of class SpatialGridDataFrame

Usage

```
## S3 method for class 'SpatialGridDataFrame'
yvals(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

the y values

See Also

xvals, yvals, zvals, xvals.default, yvals.default, zvals.default, xvals.fromXYZ, yvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

---

`zvals` *zvals function*

Description

Generic for extractign the 'z values' from an object.

Usage

```
zvals(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

the zvals method

See Also

xvals, yvals, xvals.default, yvals.default, zvals.default, xvals.fromXYZ, yvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

---

`zvals.default` *zvals.default function*

Description

Default method for extracting 'z values' looks for $Zm, $Z, $z in that order.

Usage

```
## Default S3 method:
zvals(obj, ...)
```

Arguments

`obj`	an object
`...`	additional arguments

Value

the x values

See Also

xvals, yvals, zvals, xvals.default, yvals.default, xvals.fromXYZ, yvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

---

`zvals.fromXYZ` *zvals.fromXYZ function*

Description

Method for extracting 'z values' from an object of class fromXYZ

Usage

```
## S3 method for class 'fromXYZ'
zvals(obj, ...)
```

Arguments

`obj`	a spatialAtRisk object
`...`	additional arguments

Value

the z values

See Also

xvals, yvals, zvals, xvals.default, yvals.default, zvals.default, xvals.fromXYZ, yvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame, zvals.SpatialGridDataFrame

`zvals.SpatialGridDataFrame`
zvals.SpatialGridDataFrame function

Description

Method for extracting 'z values' from an object of class SpatialGridDataFrame

Usage

```
## S3 method for class 'SpatialGridDataFrame'
zvals(obj, ...)
```

Arguments

obj	an object
...	additional arguments

Value

the z values

See Also

xvals, yvals, zvals, xvals.default, yvals.default, zvals.default, xvals.fromXYZ, yvals.fromXYZ, zvals.fromXYZ, xvals.SpatialGridDataFrame, yvals.SpatialGridDataFrame

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