## Package: spatstat (via r-universe)

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Title Spatial Point Pattern Analysis, Model-Fitting, Simulation, Tests

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**Depends** R (>= 3.5.0), spatstat.data (>= 3.1-0), spatstat.univar (>= 3.0-0), spatstat.geom (>= 3.3-0), spatstat.random (>= 3.3-0), spatstat.explore (>= 3.3-0), spatstat.model (>= 3.3-0), spatstat.linnet (>= 3.2-0), utils

**Imports** spatstat.utils (>= 3.0-5)

**Description** Comprehensive open-source toolbox for analysing Spatial Point Patterns. Focused mainly on two-dimensional point patterns, including multitype/marked points, in any spatial region. Also supports three-dimensional point patterns, space-time point patterns in any number of dimensions, point patterns on a linear network, and patterns of other geometrical objects. Supports spatial covariate data such as pixel images. Contains over 3000 functions for plotting spatial data, exploratory data analysis, model-fitting, simulation, spatial sampling, model diagnostics, and formal inference. Data types include point patterns, line segment patterns, spatial windows, pixel images, tessellations, and linear networks. Exploratory methods include quadrat counts, K-functions and their simulation envelopes, nearest neighbour distance and empty space statistics, Fry plots, pair correlation function, kernel smoothed intensity, relative risk estimation with cross-validated bandwidth selection, mark correlation functions, segregation indices, mark dependence diagnostics, and kernel estimates of covariate effects. Formal hypothesis tests of random pattern (chi-squared, Kolmogorov-Smirnov, Monte Carlo, Diggle-Cressie-Loosmore-Ford, Dao-Genton, two-stage Monte Carlo) and tests for covariate effects (Cox-Berman-Waller-Lawson, Kolmogorov-Smirnov, ANOVA) are also supported. Parametric models can be fitted to point pattern data using the functions ppm(), kppm(), slrm(), dppm() similar

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to glm(). Types of models include Poisson, Gibbs and Cox point processes, Neyman-Scott cluster processes, and determinantal point processes. Models may involve dependence on covariates, inter-point interaction, cluster formation and dependence on marks. Models are fitted by maximum likelihood, logistic regression, minimum contrast, and composite likelihood methods. A model can be fitted to a list of point patterns (replicated point pattern data) using the function mppm(). The model can include random effects and fixed effects depending on the experimental design, in addition to all the features listed above. Fitted point process models can be simulated, automatically. Formal hypothesis tests of a fitted model are supported (likelihood ratio test, analysis of deviance, Monte Carlo tests) along with basic tools for model selection (stepwise(), AIC()) and variable selection (sdr). Tools for validating the fitted model include simulation envelopes, residuals, residual plots and Q-Q plots, leverage and influence diagnostics, partial residuals, and added variable plots.

**License** GPL (>= 2)

URL http://spatstat.org/

NeedsCompilation yes

ByteCompile true

BugReports https://github.com/spatstat/spatstat/issues

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spatstat-package

The Spatstat Package

#### Description

This is a summary of the features of **spatstat**, a family of R packages for the statistical analysis of spatial point patterns.

#### **Details**

**spatstat** is a family of R packages for the statistical analysis of spatial data. Its main focus is the analysis of spatial patterns of points in two-dimensional space.

spatstat is designed to support a complete statistical analysis of spatial data. It supports

- creation, manipulation and plotting of point patterns;
- exploratory data analysis;
- · spatial random sampling;
- simulation of point process models;
- parametric model-fitting;
- non-parametric smoothing and regression;
- formal inference (hypothesis tests, confidence intervals);
- · model diagnostics.

Apart from two-dimensional point patterns and point processes, **spatstat** also supports point patterns in three dimensions, point patterns in multidimensional space-time, point patterns on a linear network, patterns of line segments in two dimensions, and spatial tessellations and random sets in two dimensions.

The package can fit several types of point process models to a point pattern dataset:

- Poisson point process models (by Berman-Turner approximate maximum likelihood or by spatial logistic regression)
- Gibbs/Markov point process models (by Baddeley-Turner approximate maximum pseudolikelihood, Coeurjolly-Rubak logistic likelihood, or Huang-Ogata approximate maximum likelihood)
- Cox/cluster point process models (by Waagepetersen's two-step fitting procedure and minimum contrast, composite likelihood, or Palm likelihood)
- determinantal point process models (by Waagepetersen's two-step fitting procedure and minimum contrast, composite likelihood, or Palm likelihood)

The models may include spatial trend, dependence on covariates, and complicated interpoint interactions. Models are specified by a formula in the R language, and are fitted using a function analogous to lm and glm. Fitted models can be printed, plotted, predicted, simulated and so on.

#### **Getting Started**

For a quick introduction to **spatstat**, read the package vignette *Getting started with spatstat* installed with **spatstat**. To read that document, you can either

- visit https://cran.r-project.org/package=spatstat and click on Getting Started with Spatstat
- start R, type library(spatstat) and vignette('getstart')
- start R, type help.start() to open the help browser, and navigate to Packages > spatstat > Vignettes.

Once you have installed **spatstat**, start R and type library(spatstat). Then type beginner for a beginner's introduction, or demo(spatstat) for a demonstration of the package's capabilities.

For a complete course on **spatstat**, and on statistical analysis of spatial point patterns, read the book by Baddeley, Rubak and Turner (2015). Other recommended books on spatial point process methods are Diggle (2014), Gelfand et al (2010) and Illian et al (2008).

The **spatstat** package includes over 50 datasets, which can be useful when learning the package. Type demo(data) to see plots of all datasets available in the package. Type vignette('datasets') for detailed background information on these datasets, and plots of each dataset.

For information on converting your data into **spatstat** format, read Chapter 3 of Baddeley, Rubak and Turner (2015). This chapter is available free online, as one of the sample chapters at the book companion website, https://book.spatstat.org/.

#### Structure of the spatstat family

The original **spatstat** package grew to be very large. It has now been divided into several **sub-packages**:

- spatstat.utils containing basic utilities
- spatstat.sparse containing linear algebra utilities
- spatstat.data containing datasets
- spatstat.univar containing functions for estimating probability distributions of random variables
- spatstat.geom containing functionality for geometrical operations, and defining the main classes of spatial objects
- spatstat.explore containing the main functions for exploratory analysis of spatial data
- **spatstat.model** containing the main functions for parametric statistical modelling and analysis, and formal inference, for spatial data
- spatstat.linnet containing functions for spatial data on a linear network
- spatstat, which simply loads the other sub-packages listed above, and provides documentation.

The breakup has been done in such a way that the user should not notice any difference. Source code that worked with the old **spatstat** package should work with the new **spatstat** family. Code that is documented in our books, journal articles and vignettes should still work.

When you install **spatstat**, the sub-packages listed above are also installed. Then if you load the **spatstat** package by typing library(spatstat), the other sub-packages listed above will automatically be loaded or imported.

This help file covers all the functionality and datasets that are provided in the sub-packages listed above.

#### **Extension packages**

Additionally there are several **extension packages:** 

- spatstat.gui for interactive graphics
- spatstat.local for local likelihood (including geographically weighted regression)
- spatstat.Knet for additional, computationally efficient code for linear networks
- **spatstat.sphere** (under development) for spatial data on a sphere, including spatial data on the earth's surface

The extension packages must be installed separately and loaded explicitly if needed. They also have separate documentation.

## **Updates**

New versions of **spatstat** are released every 8 weeks. Users are advised to update their installation of **spatstat** regularly.

Type latest.news to read the news documentation about changes to the current installed version of **spatstat**.

See the Vignette *Summary of recent updates*, installed with **spatstat**, which describes the main changes to **spatstat** since the book (Baddeley, Rubak and Turner, 2015) was published. It is accessible as vignette('updates').

Type news(package="spatstat") to read news documentation about all previous versions of the package.

#### **FUNCTIONS AND DATASETS**

Following is a summary of the main functions and datasets in the **spatstat** package. Alternatively an alphabetical list of all functions and datasets is available by typing library(help=spatstat).

For further information on any of these, type help(name) or ?name where name is the name of the function or dataset.

## **CONTENTS:**

- I. Creating and manipulating data
- II. Exploratory Data Analysis
- III. Model fitting (Cox and cluster models)
- IV. Model fitting (Poisson and Gibbs models)
- V. Model fitting (determinantal point processes)
- VI. Model fitting (spatial logistic regression)
- VII. Simulation

VIII. Tests and diagnosticsIX. Documentation

#### I. CREATING AND MANIPULATING DATA

## Types of spatial data:

The main types of spatial data supported by **spatstat** are:

ppp point pattern
owin window (spatial region)
im pixel image
psp line segment pattern
tess tessellation
pp3 three-dimensional point pattern
ppx point pattern in any number of dimensions
lpp point pattern on a linear network

## To create a point pattern:

create a point pattern from (x,y) and window information  $\operatorname{ppp}(x, y, x \operatorname{lim}, y \operatorname{lim})$  for rectangular window  $\operatorname{ppp}(x, y, \operatorname{poly})$  for polygonal window  $\operatorname{ppp}(x, y, \operatorname{mask})$  for binary image window  $\operatorname{convert}$  other types of data to a ppp object  $\operatorname{clickppp}$  interactively add points to a plot  $\operatorname{marks}<-$ , %mark% attach/reassign marks to a point pattern

## To simulate a random point pattern:

runifpoint	generate $n$ independent uniform random points
rpoint	generate $n$ independent random points
rmpoint	generate $n$ independent multitype random points
rpoispp	simulate the (in)homogeneous Poisson point process
rmpoispp	simulate the (in)homogeneous multitype Poisson point process
runifdisc	generate $n$ independent uniform random points in disc
rstrat	stratified random sample of points
rsyst	systematic random sample of points
rjitter	apply random displacements to points in a pattern
rMaternI	simulate the Matérn Model I inhibition process
rMaternII	simulate the Matérn Model II inhibition process
rSSI	simulate Simple Sequential Inhibition process
rStrauss	simulate Strauss process (perfect simulation)
rHardcore	simulate Hard Core process (perfect simulation)
rStraussHard	simulate Strauss-hard core process (perfect simulation)
rDiggleGratton	simulate Diggle-Gratton process (perfect simulation)

rDGS simulate Diggle-Gates-Stibbard process (perfect simulation)

rPenttinen simulate Penttinen process (perfect simulation)
rNeymanScott simulate a general Neyman-Scott process
rPoissonCluster simulate a general Poisson cluster process
rMatClust simulate the Matérn Cluster process

rThomas simulate the Thomas process

rGaussPoisson simulate the Gauss-Poisson cluster process simulate Neyman-Scott Cauchy cluster process

rVarGamma simulate Neyman-Scott Variance Gamma cluster process

rthin random thinning

rcell simulate the Baddeley-Silverman cell process

rmh simulate Gibbs point process using Metropolis-Hastings simulate.ppm simulate Gibbs point process using Metropolis-Hastings runifpointOnLines generate n random points along specified line segments generate Poisson random points along specified line segments

## To randomly change an existing point pattern:

rshift random shifting of points

rjitter apply random displacements to points in a pattern

rthin random thinning

rlabel random (re)labelling of a multitype point pattern

quadratresample block resampling

## Standard point pattern datasets:

Datasets in **spatstat** are lazy-loaded, so you can simply type the name of the dataset to use it; there is no need to type data(amacrine) etc.

Type demo(data) to see a display of all the datasets installed with the package.

Type vignette('datasets') for a document giving an overview of all datasets, including background information, and plots.

amacrine
anemones
ants
Upton-Fingleton sea anemones data
Harkness-Isham ant nests data
bdspots
Breakdown spots in microelectrodes

bei Tropical rainforest trees

betacells Waessle et al. cat retinal ganglia data

bramblecanes Bramble Canes data
bronzefilter Bronze Filter Section data
cells Crick-Ripley biological cells data

chicago Chicago crimes

chorley Chorley-Ribble cancer data
clmfires Castilla-La Mancha forest fires

copper Berman-Huntington copper deposits data

dendriteDendritic spinesdemohyperSynthetic point patternsdemopatSynthetic point pattern

finpines Finnish Pines data
flu Influenza virus proteins

gordon People in Gordon Square, London

gorillas Gorilla nest sites

hamster Aherne's hamster tumour data

humberside North Humberside childhood leukaemia data

hyytiala Mixed forest in Hyytiälä, Finland

Japanese Pines data japanesepines lansing Lansing Woods data Longleaf Pines data longleaf mucosa Cells in gastric mucosa Murchison gold deposits murchison New Brunswick fires data nbfires Mark-Esler-Ripley trees data nztrees Osteocyte lacunae (3D, replicated) osteo paracou Kimboto trees in Paracou, French Guiana ponderosa Getis-Franklin ponderosa pine trees data Pyramidal neurons from 31 brains pyramidal redwood Strauss-Ripley redwood saplings data Strauss redwood saplings data (full set) redwoodfull residualspaper Data from Baddeley et al (2005) shapley Galaxies in an astronomical survey

simulated point pattern (inhomogeneous, with interaction)

spiders Spider webs on mortar lines of brick wall

sporophores Mycorrhizal fungi around a tree

spruces Spruce trees in Saxonia

swedishpines Strand-Ripley Swedish pines data

urkiola Urkiola Woods data

waka Trees in Waka national park waterstriders Insects on water surface

## To manipulate a point pattern:

plot.ppp plot a point pattern (e.g. plot(X)) spatstat.gui::iplot plot a point pattern interactively

edit.ppp interactive text editor

[.ppp extract or replace a subset of a point pattern

pp[subset] or pp[subwindow]

subset.ppp extract subset of point pattern satisfying a condition

superimpose combine several point patterns

by .ppp apply a function to sub-patterns of a point pattern

cut.ppp classify the points in a point pattern split.ppp divide pattern into sub-patterns

unmark remove marks

npoints count the number of points

coords extract coordinates, change coordinates
marks extract marks, change marks or attach marks

rotate rotate pattern

shift	translate pattern
flipxy	swap $x$ and $y$ coordinates
reflect	reflect in the origin
periodify	make several translated copies
affine	apply affine transformation
scalardilate	apply scalar dilation
density.ppp	kernel estimation of point pattern intensity
densityHeat.ppp	diffusion kernel estimation of point pattern intensity
Smooth.ppp	kernel smoothing of marks of point pattern
nnmark	mark value of nearest data point
sharpen.ppp	data sharpening
identify.ppp	interactively identify points
unique.ppp	remove duplicate points
duplicated.ppp	determine which points are duplicates
uniquemap.ppp	map duplicated points to unique points
connected.ppp	find clumps of points
dirichlet	compute Dirichlet-Voronoi tessellation
delaunay	compute Delaunay triangulation
delaunayDistance	graph distance in Delaunay triangulation
convexhull	compute convex hull
discretise	discretise coordinates
pixellate.ppp	approximate point pattern by pixel image
as.im.ppp	approximate point pattern by pixel image

See spatstat.options to control plotting behaviour.

## To create a window:

An object of class "owin" describes a spatial region (a window of observation).

owin  Create a window object  owin(xlim, ylim) for rectangular window  owin(poly) for polygonal window  owin(mask) for binary image window	
Window Extract window of another object	
Frame Extract the containing rectangle ('frame') of	another object
as.owin Convert other data to a window object	
square make a square window	
disc make a circular window	
ellipse make an elliptical window	
ripras Ripley-Rasson estimator of window, given or	nly the points
convexhull compute convex hull of something	
letterR polygonal window in the shape of the R logo	•
clickpoly interactively draw a polygonal window	
clickbox interactively draw a rectangle	

## To manipulate a window:

```
plot.owin plot a window.
```

plot(W)

boundingbox Find a tight bounding box for the window

erosion erode window by a distance r dilation dilate window by a distance r closing close window by a distance r opening open window by a distance r

border difference between window and its erosion/dilation

complement.owin invert (swap inside and outside)

simplify.owin approximate a window by a simple polygon

rotate rotate window

flipxy swap x and y coordinates

shift translate window

periodify make several translated copies affine apply affine transformation convert window to data frame

## **Digital approximations:**

as.mask

Make a discrete pixel approximation of a given window

as.im.owin convert window to pixel image pixellate.owin convert window to pixel image commonGrid find common pixel grid for windows

nearest.raster.point map continuous coordinates to raster locations

raster.x raster x coordinates
raster.y raster y coordinates
raster.xy raster x and y coordinates

as.polygonal convert pixel mask to polygonal window

See spatstat.options to control the approximation

#### Geometrical computations with windows:

edges extract boundary edges
intersect.owin intersection of two windows
union.owin union of two windows

setminus.owin set subtraction of two windows

inside.owin determine whether a point is inside a window

area.owin compute area

perimeter compute perimeter length diameter.owin compute diameter

incircle find largest circle inside a window

inradius radius of incircle

connected.owin find connected components of window compute areas of eroded windows dilated.areas compute areas of dilated windows

bdist.points compute distances from data points to window boundary bdist.pixels compute distances from all pixels to window boundary

bdist.tiles boundary distance for each tile in tessellation

```
distance transform image
distmap.owin
distfun.owin
                     distance transform
                     compute centroid (centre of mass) of window
centroid.owin
                     determine whether one window contains another
is.subset.owin
is.convex
                     determine whether a window is convex
convexhul1
                     compute convex hull
triangulate.owin
                     decompose into triangles
                     pixel approximation of window
as.mask
as.polygonal
                     polygonal approximation of window
                     test whether window is a rectangle
is.rectangle
is.polygonal
                     test whether window is polygonal
                     test whether window is a mask
is.mask
                     spatial covariance function of window
setcov
                     extract centres of pixels in mask
pixelcentres
                     measure distance between two points clicked by user
clickdist
```

**Pixel images:** An object of class "im" represents a pixel image. Such objects are returned by some of the functions in **spatstat** including Kmeasure, setcov and density.ppp.

```
im
                      create a pixel image
as.im
                      convert other data to a pixel image
                      convert other data to a pixel image
pixellate
as.matrix.im
                      convert pixel image to matrix
as.data.frame.im
                      convert pixel image to data frame
                      convert pixel image to function
as.function.im
plot.im
                      plot a pixel image on screen as a digital image
                      draw contours of a pixel image
contour.im
                      draw perspective plot of a pixel image
persp.im
rgbim
                      create colour-valued pixel image
                      create colour-valued pixel image
hsvim
                      extract a subset of a pixel image
Γ.im
[<-.im
                      replace a subset of a pixel image
                      rotate pixel image
rotate.im
                      apply vector shift to pixel image
shift.im
                      apply affine transformation to image
affine.im
                      print very basic information about image X
                      summary of image X
summary(X)
hist.im
                      histogram of image
                      mean pixel value of image
mean.im
integral.im
                      integral of pixel values
quantile.im
                      quantiles of image
cut.im
                      convert numeric image to factor image
                      test whether an object is a pixel image
is.im
                      interpolate a pixel image
interp.im
                      apply Gaussian blur to image
blur
Smooth.im
                      apply Gaussian blur to image
connected.im
                      find connected components
compatible.im
                      test whether two images have compatible dimensions
```

harmonise.im make images compatible commonGrid find a common pixel grid for images evaluate any expression involving images eval.im evaluate a function of several images im.apply scaletointerval rescale pixel values zapsmall.im set very small pixel values to zero levelset level set of an image region where an expression is true solutionset spatial covariance function of image imcov convolve.im spatial convolution of images transect.im line transect of image extract centres of pixels pixelcentres convert matrix of pixel values transmat to a different indexing convention random pixel noise rnoise

#### Line segment patterns

An object of class "psp" represents a pattern of straight line segments.

psp create a line segment pattern convert other data into a line segment pattern as.psp edges extract edges of a window determine whether a dataset has class "psp" is.psp plot.psp plot a line segment pattern print basic information print.psp summary.psp print summary information extract a subset of a line segment pattern [.psp extract subset of line segment pattern subset.psp convert line segment pattern to data frame as.data.frame.psp extract marks of line segments marks.psp marks<-.psp assign new marks to line segments unmark.psp delete marks from line segments midpoints.psp compute the midpoints of line segments extract the endpoints of line segments endpoints.psp compute the lengths of line segments lengths\_psp angles.psp compute the orientation angles of line segments superimpose combine several line segment patterns swap x and y coordinates flipxy rotate.psp rotate a line segment pattern shift a line segment pattern shift.psp make several shifted copies periodify affine.psp apply an affine transformation pixellate.psp approximate line segment pattern by pixel image approximate line segment pattern by binary mask psp2mask compute the distance map of a line segment pattern distmap.psp compute the distance map of a line segment pattern distfun.psp density.psp kernel smoothing of line segments selfcrossing.psp find crossing points between line segments

selfcut.psp	cut segments where they cross
crossing.psp	find crossing points between two line segment patterns
extrapolate.psp	extrapolate line segments to infinite lines
nncross	find distance to nearest line segment from a given point
nearestsegment	find line segment closest to a given point
project2segment	find location along a line segment closest to a given point
pointsOnLines	generate points evenly spaced along line segment
rpoisline	generate a realisation of the Poisson line process inside a window
rlinegrid	generate a random array of parallel lines through a window

## **Tessellations**

An object of class "tess" represents a tessellation.

tess	create a tessellation
quadrats	create a tessellation of rectangles
hextess	create a tessellation of hexagons
polartess	tessellation using polar coordinates
quantess	quantile tessellation
venn.tess	Venn diagram tessellation
dirichlet	compute Dirichlet-Voronoi tessellation of points
delaunay	compute Delaunay triangulation of points
as.tess	convert other data to a tessellation
plot.tess	plot a tessellation
tiles	extract all the tiles of a tessellation
[.tess	extract some tiles of a tessellation
[ <tess< th=""><td>change some tiles of a tessellation</td></tess<>	change some tiles of a tessellation
intersect.tess	intersect two tessellations
	or restrict a tessellation to a window
chop.tess	subdivide a tessellation by a line
rpoislinetess	generate tessellation using Poisson line process
tile.areas	area of each tile in tessellation
bdist.tiles	boundary distance for each tile in tessellation
connected.tess	find connected components of tiles
shift.tess	shift a tessellation
rotate.tess	rotate a tessellation
reflect.tess	reflect about the origin
flipxy.tess	reflect about the diagonal
affine.tess	apply affine transformation

## Three-dimensional point patterns

An object of class "pp3" represents a three-dimensional point pattern in a rectangular box. The box is represented by an object of class "box3".

pp3	create a 3-D point pattern
plot.pp3	plot a 3-D point pattern
coords	extract coordinates
as.hyperframe	extract coordinates

extract subset of 3-D point pattern subset.pp3 unitname.pp3 name of unit of length count the number of points npoints generate uniform random points in 3-D runifpoint3 rpoispp3 generate Poisson random points in 3-D envelope.pp3 generate simulation envelopes for 3-D pattern box3 create a 3-D rectangular box convert data to 3-D rectangular box as.box3 unitname.box3 name of unit of length diameter.box3 diameter of box volume.box3 volume of box shortside.box3 shortest side of box eroded.volumes volumes of erosions of box

#### Multi-dimensional space-time point patterns

An object of class "ppx" represents a point pattern in multi-dimensional space and/or time.

create a multidimensional space-time point pattern ррх coords extract coordinates as.hyperframe extract coordinates subset.ppx extract subset unitname.ppx name of unit of length count the number of points npoints runifpointx generate uniform random points generate Poisson random points rpoisppx define multidimensional box boxx diameter of box diameter.boxx volume of box volume.boxx shortest side of box shortside.boxx eroded.volumes.boxx volumes of erosions of box

#### Point patterns on a linear network

An object of class "linnet" represents a linear network (for example, a road network).

create a linear network linnet clickjoin interactively join vertices in network spatstat.gui::iplot.linnet interactively plot network simplenet simple example of network lineardisc disc in a linear network delaunayNetwork network of Delaunay triangulation dirichletNetwork network of Dirichlet edges methods.linnet methods for linnet objects vertices.linnet nodes of network joinVertices join existing vertices in a network insertVertices insert new vertices at positions along a network addVertices add new vertices, extending a network thinNetwork remove vertices or lines from a network

repairNetwork	repair internal format
pixellate.linnet	approximate by pixel image

An object of class "lpp" represents a point pattern on a linear network (for example, road accidents on a road network).

lpp	create a point pattern on a linear network
methods.lpp	methods for 1pp objects
subset.lpp	method for subset
rpoislpp	simulate Poisson points on linear network
runiflpp	simulate random points on a linear network
chicago	Chicago crime data
dendrite	Dendritic spines data
spiders	Spider webs on mortar lines of brick wall

## **Hyperframes**

A hyperframe is like a data frame, except that the entries may be objects of any kind.

hyperframe	araata a himarframa
31	create a hyperframe
as.hyperframe	convert data to hyperframe
plot.hyperframe	plot hyperframe
with.hyperframe	evaluate expression using each row of hyperframe
cbind.hyperframe	combine hyperframes by columns
rbind.hyperframe	combine hyperframes by rows
as.data.frame.hyperframe	convert hyperframe to data frame
subset.hyperframe	method for subset
head.hyperframe	first few rows of hyperframe
tail.hyperframe	last few rows of hyperframe

## Layered objects

A layered object represents data that should be plotted in successive layers, for example, a background and a foreground.

layered	create layered object
plot.layered	plot layered object
[.layered	extract subset of layered object

## Colour maps

A colour map is a mechanism for associating colours with data. It can be regarded as a function, mapping data to colours. Using a colourmap object in a plot command ensures that the mapping from numbers to colours is the same in different plots.

colourmap	create a colour map
plot.colourmap	plot the colour map only
tweak.colourmap	alter individual colour values
<pre>interp.colourmap</pre>	make a smooth transition between colours

beachcolourmap one special colour map

#### II. EXPLORATORY DATA ANALYSIS

#### Inspection of data:

summary(X) print useful summary of point pattern X
X print basic description of point pattern X
any(duplicated(X)) check for duplicated points in pattern X
spatstat.gui::istat(X) Interactive exploratory analysis
spatstat.gui::View.ppp(X) spreadsheet-style viewer

#### Classical exploratory tools:

clarkevans Clark and Evans aggregation index fryplot Fry plot Morisita Index plot

#### **Smoothing:**

density.ppp kernel smoothed density/intensity relrisk kernel estimate of relative risk Smooth.ppp spatial interpolation of marks cross-validated bandwidth selection for density.ppp bw.diggle likelihood cross-validated bandwidth selection for density.ppp lgg.wd bw.CvL Cronie-Van Lieshout bandwidth selection for density estimation bw.scott Scott's rule of thumb for density estimation Abramson's rule for adaptive bandwidths bw.abram bw.relrisk cross-validated bandwidth selection for relrisk bw.smoothppp cross-validated bandwidth selection for Smooth.ppp bw.frac bandwidth selection using window geometry Stoyan's rule of thumb for bandwidth for pcf bw.stoyan

#### Modern exploratory tools:

clusterset
nnclean
Sharpen.ppp
rhohat
rho2hat
spatialcdf
roc

Allard-Fraley feature detection
Byers-Raftery feature detection
Choi-Hall data sharpening
Kernel estimate of covariate effect
Kernel estimate of effect of two covariates
Spatial cumulative distribution function
Receiver operating characteristic curve

**Summary statistics for a point pattern:** Type demo(sumfun) for a demonstration of many of the summary statistics.

intensity Mean intensity

Quadrat counts

quadi account	Quadrat counts
<pre>intensity.quadratcount</pre>	Mean intensity in quadrats
Fest	empty space function $F$
Gest	nearest neighbour distribution function $G$
Jest	<i>J</i> -function $J = (1 - G)/(1 - F)$
Kest	Ripley's K-function
Lest	Besag L-function
Tstat	Third order $T$ -function
allstats	all four functions $F, G, J, K$
pcf	pair correlation function
Kinhom	K for inhomogeneous point patterns
Linhom	L for inhomogeneous point patterns
pcfinhom	pair correlation for inhomogeneous patterns
Finhom	F for inhomogeneous point patterns
Ginhom	G for inhomogeneous point patterns
Jinhom	J for inhomogeneous point patterns
localL	Getis-Franklin neighbourhood density function
localK	neighbourhood K-function
localpcf	local pair correlation function
localKinhom	local K for inhomogeneous point patterns
localLinhom	local L for inhomogeneous point patterns
localpcfinhom	local pair correlation for inhomogeneous patterns
Ksector	Directional K-function
Kscaled	locally scaled $K$ -function
Kest.fft	fast $K$ -function using FFT for large datasets
Kmeasure	reduced second moment measure
envelope	simulation envelopes for a summary function
varblock	variances and confidence intervals
	for a summary function
lohboot	bootstrap for a summary function

## Related facilities:

quadratcount

plot.fv	plot a summary function
eval.fv	evaluate any expression involving summary functions
harmonise.fv	make functions compatible
eval.fasp	evaluate any expression involving an array of functions
with.fv	evaluate an expression for a summary function
Smooth.fv	apply smoothing to a summary function
deriv.fv	calculate derivative of a summary function
pool.fv	pool several estimates of a summary function
nndist	nearest neighbour distances
nnwhich	find nearest neighbours
pairdist	distances between all pairs of points
crossdist	distances between points in two patterns
nncross	nearest neighbours between two point patterns
exactdt	distance from any location to nearest data point
distmap	distance map image

> distfun distance map function nearest point image nnmap nnfun nearest point function kernel smoothed density density.ppp

densityHeat.ppp diffusion kernel smoothed density Smooth.ppp spatial interpolation of marks relrisk kernel estimate of relative risk

data sharpening sharpen.ppp

relrisk

theoretical distribution of nearest neighbour distance rknn

Summary statistics for a multitype point pattern: A multitype point pattern is represented by an object X of class "ppp" such that marks(X) is a factor.

kernel estimation of relative risk scan.test spatial scan test of elevated risk multitype nearest neighbour distributions  $G_{ij}$ ,  $G_{i\bullet}$ Gcross, Gdot, Gmulti Kcross, Kdot, Kmulti multitype K-functions  $K_{ij}, K_{i\bullet}$ Lcross, Ldot multitype L-functions  $L_{ij}, L_{i\bullet}$ 

Jcross, Jdot, Jmulti multitype J-functions  $J_{ij}, J_{i\bullet}$ pcfcross multitype pair correlation function  $g_{ij}$ pcfdot multitype pair correlation function  $g_{i\bullet}$ pcfmulti general pair correlation function markconnect marked connection function  $p_{ij}$ alltypes estimates of the above for all i, j pairs

**Iest** multitype I-function

inhomogeneous counterparts of Kcross, Kdot Kcross.inhom, Kdot.inhom Lcross.inhom,Ldot.inhom inhomogeneous counterparts of Lcross, Ldot inhomogeneous counterparts of pcfcross, pcfdot pcfcross.inhom,pcfdot.inhom

localKcross,localKdot local counterparts of Kcross, Kdot localLcross, localLdot local counterparts of Lcross, Ldot

localKcross.inhom,localLcross.inhom local counterparts of Kcross.inhom, Lcross.inhom

Summary statistics for a marked point pattern: A marked point pattern is represented by an object X of class "ppp" with a component X\$marks. The entries in the vector X\$marks may be numeric, complex, string or any other atomic type. For numeric marks, there are the following functions:

> smoothed local average of marks markmean smoothed local variance of marks markvar markcorr mark correlation function

mark cross-correlation function markcrosscorr

mark variogram markvario markmarkscatter mark-mark scatterplot mark-weighted K function Kmark

mark independence diagnostic E(r)**Emark Vmark** mark independence diagnostic V(r)nearest neighbour mean index nnmean

nearest neighbour mark variance index nnvario

For marks of any type, there are the following:

Alternatively use cut.ppp to convert a marked point pattern to a multitype point pattern.

## **Programming tools:**

```
applynbd apply function to every neighbourhood in a point pattern
markstat apply function to the marks of neighbours in a point pattern
tabulate the marks of neighbours in a point pattern
pppdist find the optimal match between two point patterns
```

## Summary statistics for a point pattern on a linear network:

These are for point patterns on a linear network (class 1pp). For unmarked patterns:

linearK	K function on linear network
linearKinhom	inhomogeneous $K$ function on linear network
linearpcf	pair correlation function on linear network
linearpcfinhom	inhomogeneous pair correlation on linear network

## For multitype patterns:

linearKcross	K function between two types of points
linearKdot	K function from one type to any type
linearKcross.inhom	Inhomogeneous version of linearKcross
linearKdot.inhom	Inhomogeneous version of linearKdot
linearmarkconnect	Mark connection function on linear network
linearmarkequal	Mark equality function on linear network
linearpcfcross	Pair correlation between two types of points
linearpcfdot	Pair correlation from one type to any type
linearpcfcross.inhom	Inhomogeneous version of linearpcfcross
linearpcfdot.inhom	Inhomogeneous version of linearpcfdot

## Related facilities:

pairdist.lpp crossdist.lpp nndist.lpp nncross.lpp nnwhich.lpp nnfun.lpp density.lpp densityHeat.lpp distfun.lpp	distances between pairs distances between pairs nearest neighbour distances nearest neighbour distances find nearest neighbours find nearest data point kernel smoothing estimator of intensity diffusion kernel estimate distance transform
envelope.lpp	simulation envelopes

rpoislpp	simulate Poisson points on linear network
runiflpp	simulate random points on a linear network

It is also possible to fit point process models to 1pp objects. See Section IV.

## Summary statistics for a three-dimensional point pattern:

These are for 3-dimensional point pattern objects (class pp3).

```
F3est empty space function F
G3est nearest neighbour function G
K3est K-function
pcf3est pair correlation function
```

#### Related facilities:

```
envelope.pp3 simulation envelopes
pairdist.pp3 distances between all pairs of points
crossdist.pp3 distances between points in two patterns
nndist.pp3 nearest neighbour distances
nnwhich.pp3 find nearest neighbours
nncross.pp3 find nearest neighbours in another pattern
```

#### Computations for multi-dimensional point pattern:

These are for multi-dimensional space-time point pattern objects (class ppx).

```
pairdist.ppx distances between all pairs of points crossdist.ppx distances between points in two patterns nearest neighbour distances find nearest neighbours
```

#### **Summary statistics for random sets:**

These work for point patterns (class ppp), line segment patterns (class psp) or windows (class owin).

```
 \begin{array}{ll} {\sf Hest} & {\sf spherical\ contact\ distribution\ } H \\ {\sf Gfox} & {\sf Foxall\ } G\text{-function} \\ {\sf Jfox} & {\sf Foxall\ } J\text{-function} \\ \end{array}
```

## III. MODEL FITTING (COX AND CLUSTER MODELS)

Cluster process models (with homogeneous or inhomogeneous intensity) and Cox processes can be fitted by the function kppm. Its result is an object of class "kppm". The fitted model can be printed, plotted, predicted, simulated and updated.

kppm	Fit model
plot.kppm	Plot the fitted model
summary.kppm	Summarise the fitted model
fitted.kppm	Compute fitted intensity

predict.kppm Compute fitted intensity update.kppm Update the model

improve.kppm Refine the estimate of trend simulate.kppm Generate simulated realisations

vcov.kppm Variance-covariance matrix of coefficients

coef.kppmExtract trend coefficientsformula.kppmExtract trend formulaparametersExtract all model parametersclusterfield.kppmCompute offspring density

clusterradius.kppm Radius of support of offspring density

 $\begin{array}{ll} {\sf Kmodel.kppm} & {\it K} \ {\sf function} \ {\sf of} \ {\sf fitted} \ {\sf model} \\ {\sf pcfmodel.kppm} & {\sf Pair} \ {\sf correlation} \ {\sf of} \ {\sf fitted} \ {\sf model} \end{array}$ 

For model selection, you can also use the generic functions step, drop1 and AIC on fitted point process models. For variable selection, see sdr.

The theoretical models can also be simulated, for any choice of parameter values, using rThomas, rMatClust, rCauchy, rVarGamma, and rLGCP.

Lower-level fitting functions include:

lgcp.estK fit a log-Gaussian Cox process model lgcp.estpcf fit a log-Gaussian Cox process model thomas.estK fit the Thomas process model thomas.estpcf fit the Thomas process model matclust.estK fit the Matérn Cluster process model matclust.estpcf fit the Matérn Cluster process model fit a Neyman-Scott Cauchy cluster process cauchy.estK cauchy.estpcf fit a Neyman-Scott Cauchy cluster process vargamma.estK fit a Neyman-Scott Variance Gamma process fit a Neyman-Scott Variance Gamma process vargamma.estpcf low-level algorithm for fitting models mincontrast by the method of minimum contrast

## IV. MODEL FITTING (POISSON AND GIBBS MODELS)

#### Types of models

Poisson point processes are the simplest models for point patterns. A Poisson model assumes that the points are stochastically independent. It may allow the points to have a non-uniform spatial density. The special case of a Poisson process with a uniform spatial density is often called Complete Spatial Randomness.

Poisson point processes are included in the more general class of Gibbs point process models. In a Gibbs model, there is *interaction* or dependence between points. Many different types of interaction can be specified.

For a detailed explanation of how to fit Poisson or Gibbs point process models to point pattern data using **spatstat**, see Baddeley and Turner (2005b) or Baddeley (2008).

#### To fit a Poisson or Gibbs point process model:

Model fitting in **spatstat** is performed mainly by the function ppm. Its result is an object of class "ppm".

Here are some examples, where X is a point pattern (class "ppp"):

```
commandmodelppm(X)Complete Spatial Randomnessppm(X \sim 1)Complete Spatial Randomnessppm(X \sim x)Poisson process with<br/>intensity loglinear in x coordinateppm(X \sim 1, Strauss(0.1))Stationary Strauss processppm(X \sim x, Strauss(0.1))Strauss process with<br/>conditional intensity loglinear in x
```

It is also possible to fit models that depend on other covariates.

## Manipulating the fitted model:

plot.ppm	Plot the fitted model
<pre>predict.ppm</pre>	Compute the spatial trend and conditional intensity
	of the fitted point process model
coef.ppm	Extract the fitted model coefficients
parameters	Extract all model parameters
formula.ppm	Extract the trend formula
<pre>intensity.ppm</pre>	Compute fitted intensity
Kmodel.ppm	K function of fitted model
<pre>pcfmodel.ppm</pre>	pair correlation of fitted model
fitted.ppm	Compute fitted conditional intensity at quadrature points
residuals.ppm	Compute point process residuals at quadrature points
update.ppm	Update the fit
vcov.ppm	Variance-covariance matrix of estimates
rmh.ppm	Simulate from fitted model
simulate.ppm	Simulate from fitted model
print.ppm	Print basic information about a fitted model
summary.ppm	Summarise a fitted model
effectfun	Compute the fitted effect of one covariate
<pre>logLik.ppm</pre>	log-likelihood or log-pseudolikelihood
anova.ppm	Analysis of deviance
<pre>model.frame.ppm</pre>	Extract data frame used to fit model
model.images	Extract spatial data used to fit model
model.depends	Identify variables in the model
as.interact	Interpoint interaction component of model
fitin	Extract fitted interpoint interaction
is.hybrid	Determine whether the model is a hybrid
valid.ppm	Check the model is a valid point process
<pre>project.ppm</pre>	Ensure the model is a valid point process

For model selection, you can also use the generic functions step, drop1 and AIC on fitted point process models. For variable selection, see sdr.

See spatstat.options to control plotting of fitted model.

## To specify a point process model:

The first order "trend" of the model is determined by an R language formula. The formula specifies the form of the *logarithm* of the trend.

```
\begin{array}{lll} {\sf X} \, {\scriptstyle \sim} \, {\sf 1} & & {\sf No} \ {\sf trend} \ ({\sf stationary}) \\ {\sf X} \, {\scriptstyle \sim} \, {\sf x} & & {\sf Loglinear} \ {\sf trend} \ \lambda(x,y) = \exp(\alpha+\beta x) \\ {\sf where} \ x,y \ {\sf are} \ {\sf Cartesian} \ {\sf coordinates} \\ {\sf X} \, {\scriptstyle \sim} \ {\sf polynom}({\sf x},{\sf y},{\sf 3}) & {\sf Log-cubic} \ {\sf polynomial} \ {\sf trend} \\ {\sf X} \, {\scriptstyle \sim} \ {\sf harmonic}({\sf x},{\sf y},{\sf 2}) & {\sf Log-harmonic} \ {\sf polynomial} \ {\sf trend} \\ {\sf X} \, {\scriptstyle \sim} \, {\sf Z} & {\sf Loglinear} \ {\sf function} \ {\sf of} \ {\sf covariate} \ {\sf Z} \\ \lambda(x,y) = \exp(\alpha+\beta Z(x,y)) \end{array}
```

The higher order ("interaction") components are described by an object of class "interact". Such objects are created by:

Poisson()	the Poisson point process
AreaInter()	Area-interaction process
BadGey()	multiscale Geyer process
Concom()	connected component interaction
DiggleGratton()	Diggle-Gratton potential
DiggleGatesStibbard()	Diggle-Gates-Stibbard potential
Fiksel()	Fiksel pairwise interaction process
Geyer()	Geyer's saturation process
Hardcore()	Hard core process
HierHard()	Hierarchical multippe hard core process
HierStrauss()	Hierarchical multiype Strauss process
HierStraussHard()	Hierarchical multippe Strauss-hard core process
Hybrid()	Hybrid of several interactions
LennardJones()	Lennard-Jones potential
MultiHard()	multitype hard core process
MultiStrauss()	multitype Strauss process
MultiStraussHard()	multitype Strauss/hard core process
OrdThresh()	Ord process, threshold potential
Ord()	Ord model, user-supplied potential
PairPiece()	pairwise interaction, piecewise constant
Pairwise()	pairwise interaction, user-supplied potential
Penttinen()	Penttinen pairwise interaction
SatPiece()	Saturated pair model, piecewise constant potential
Saturated()	Saturated pair model, user-supplied potential
Softcore()	pairwise interaction, soft core potential
Strauss()	Strauss process
StraussHard()	Strauss/hard core point process
Triplets()	Geyer triplets process
• • • • • • • • • • • • • • • • • • • •	* * *

Note that it is also possible to combine several such interactions using Hybrid.

## Finer control over model fitting:

A quadrature scheme is represented by an object of class "quad". To create a quadrature scheme, typically use quadscheme.

quadscheme default quadrature scheme

using rectangular cells or Dirichlet cells quadrature scheme based on image pixels

quad create an object of class "quad"

To inspect a quadrature scheme:

pixelquad

plot(Q) plot quadrature scheme Q print(Q) print basic information about quadrature scheme Q

summary(Q) summary of quadrature scheme Q

A quadrature scheme consists of data points, dummy points, and weights. To generate dummy points:

default.dummy default pattern of dummy points dummy points in a rectangular grid stratified random dummy pattern spokes radial pattern of dummy points

corners dummy points at corners of the window

To compute weights:

gridweights quadrature weights by the grid-counting rule dirichletWeights quadrature weights are Dirichlet tile areas

#### Simulation and goodness-of-fit for fitted models:

rmh.ppm simulate realisations of a fitted model simulate.ppm simulate realisations of a fitted model

envelope compute simulation envelopes for a fitted model

## Point process models on a linear network:

An object of class "1pp" represents a pattern of points on a linear network. Point process models can also be fitted to these objects. Currently only Poisson models can be fitted.

point process model on linear network 1ppm anova.lppm analysis of deviance for point process model on linear network envelope.lppm simulation envelopes for point process model on linear network fitted.lppm fitted intensity values model prediction on linear network predict.lppm linim pixel image on linear network plot a pixel image on linear network plot.linim eval.linim evaluate expression involving images function defined on linear network linfun

#### methods.linfun conversion facilities

#### V. MODEL FITTING (DETERMINANTAL POINT PROCESS MODELS)

Code for fitting *determinantal point process models* has recently been added to **spatstat**.

For information, see the help file for dppm.

## VI. MODEL FITTING (SPATIAL LOGISTIC REGRESSION)

#### Logistic regression

Pixel-based spatial logistic regression is an alternative technique for analysing spatial point patterns that is widely used in Geographical Information Systems. It is approximately equivalent to fitting a Poisson point process model.

In pixel-based logistic regression, the spatial domain is divided into small pixels, the presence or absence of a data point in each pixel is recorded, and logistic regression is used to model the presence/absence indicators as a function of any covariates.

Facilities for performing spatial logistic regression are provided in **spatstat** for comparison purposes.

#### Fitting a spatial logistic regression

Spatial logistic regression is performed by the function slrm. Its result is an object of class "slrm". There are many methods for this class, including methods for print, fitted, predict, simulate, anova, coef, logLik, terms, update, formula and vcov.

For example, if X is a point pattern (class "ppp"):

command	model
$slrm(X \sim 1)$	Complete Spatial Randomness
$slrm(X \sim x)$	Poisson process with
	intensity loglinear in $x$ coordinate
$slrm(X \sim Z)$	Poisson process with
	intensity loglinear in covariate Z

## Manipulating a fitted spatial logistic regression

anova.slrm	Analysis of deviance
coef.slrm	Extract fitted coefficients
vcov.slrm	Variance-covariance matrix of fitted coefficients
fitted.slrm	Compute fitted probabilities or intensity
logLik.slrm	Evaluate loglikelihood of fitted model
plot.slrm	Plot fitted probabilities or intensity
predict.slrm	Compute predicted probabilities or intensity with new data
simulate.slrm	Simulate model

There are many other undocumented methods for this class, including methods for print, update, formula and terms. Stepwise model selection is possible using step or stepAIC. For variable selection, see sdr.

#### VII. SIMULATION

There are many ways to generate a random point pattern, line segment pattern, pixel image or tessellation in **spatstat**.

#### Random point patterns:

runifpoint generate n independent uniform random points rpoint generate n independent random points generate n independent multitype random points rmpoint simulate the (in)homogeneous Poisson point process rpoispp simulate the (in)homogeneous multitype Poisson point process rmpoispp runifdisc generate n independent uniform random points in disc stratified random sample of points rstrat systematic random sample (grid) of points rsyst simulate the Matérn Model I inhibition process rMaternI simulate the Matérn Model II inhibition process rMaternII simulate Simple Sequential Inhibition process rSSI rHardcore simulate hard core process (perfect simulation) simulate Strauss process (perfect simulation) rStrauss simulate Strauss-hard core process (perfect simulation) rStraussHard simulate Diggle-Gratton process (perfect simulation) rDiggleGratton simulate Diggle-Gates-Stibbard process (perfect simulation) rDGS rPenttinen simulate Penttinen process (perfect simulation) rNeymanScott simulate a general Neyman-Scott process simulate the Matérn Cluster process rMatClust simulate the Thomas process rThomas simulate the log-Gaussian Cox process rLGCP simulate the Gauss-Poisson cluster process rGaussPoisson rCauchy simulate Neyman-Scott process with Cauchy clusters rVarGamma simulate Neyman-Scott process with Variance Gamma clusters rcell simulate the Baddeley-Silverman cell process generate n random points along specified line segments runifpointOnLines

#### Resampling a point pattern:

rpoisppOnLines

quadratresampleblock resamplingrjitterapply random displacements to points in a patternrshiftrandom shifting of (subsets of) pointsrthinrandom thinning

generate Poisson random points along specified line segments

See also varblock for estimating the variance of a summary statistic by block resampling, and lohboot for another bootstrap technique.

## Fitted point process models:

If you have fitted a point process model to a point pattern dataset, the fitted model can be simulated.

Cluster process models are fitted by the function kppm yielding an object of class "kppm". To generate one or more simulated realisations of this fitted model, use simulate.kppm.

Gibbs point process models are fitted by the function ppm yielding an object of class "ppm". To generate a simulated realisation of this fitted model, use rmh. To generate one or more simulated realisations of the fitted model, use simulate.ppm.

#### Other random patterns:

rlinegrid generate a random array of parallel lines through a window simulate the Poisson line process within a window generate random tessellation using Poisson line process generate random set by selecting some tiles of a tessellation generate random pixel image by assigning random values in each tile of a tessellation

#### Simulation-based inference

envelope critical envelope for Monte Carlo test of goodness-of-fit critical envelope for balanced two-stage Monte Carlo test diagnostic plot for interpoint interaction spatial scan statistic/test studpermu.test studentised permutation test test of segregation of types

#### VIII. TESTS AND DIAGNOSTICS

#### **Hypothesis tests:**

 $\chi^2$  goodness-of-fit test on quadrat counts quadrat.test clarkevans.test Clark and Evans test cdf.test Spatial distribution goodness-of-fit test berman.test Berman's goodness-of-fit tests critical envelope for Monte Carlo test of goodness-of-fit envelope scan.test spatial scan statistic/test dclf.test Diggle-Cressie-Loosmore-Ford test Mean Absolute Deviation test mad.test anova.ppm Analysis of Deviance for point process models

## More recently-developed tests:

dg.test Dao-Genton test
bits.test Balanced independent two-stage test
dclf.progress Progress plot for DCLF test
mad.progress Progress plot for MAD test

## Sensitivity diagnostics:

Classical measures of model sensitivity such as leverage and influence have been adapted to point process models.

leverage.ppm Leverage for point process model

```
influence.ppm Influence for point process model dfbetas.ppm Parameter influence dffit.ppm Effect change diagnostic
```

#### **Diagnostics for covariate effect:**

Classical diagnostics for covariate effects have been adapted to point process models.

parres Partial residual plot
addvar Added variable plot
rhohat Kernel estimate of covariate effect
rho2hat Kernel estimate of covariate effect (bivariate)

#### **Residual diagnostics:**

Residuals for a fitted point process model, and diagnostic plots based on the residuals, were introduced in Baddeley et al (2005) and Baddeley, Rubak and Møller (2011).

Type demo(diagnose) for a demonstration of the diagnostics features.

diagnose.ppm diagnostic plots for spatial trend diagnostic Q-Q plot for interpoint interaction qqplot.ppm residualspaper examples from Baddeley et al (2005) model compensator of K function **Kcom** model compensator of G function Gcom score residual of K function Kres score residual of G function Gres pseudoscore residual of summary function psst pseudoscore residual of empty space function psstA pseudoscore residual of G function psstG compareFit compare compensators of several fitted models

## Resampling and randomisation procedures

You can build your own tests based on randomisation and resampling using the following capabilities:

quadratresampleblock resamplingrjitterapply random displacements to points in a patternrshiftrandom shifting of (subsets of) pointsrthinrandom thinning

#### IX. DOCUMENTATION

The online manual entries are quite detailed and should be consulted first for information about a particular function.

The book Baddeley, Rubak and Turner (2015) is a complete course on analysing spatial point patterns, with full details about **spatstat**.

Older material (which is now out-of-date but is freely available) includes Baddeley and Turner (2005a), a brief overview of the package in its early development; Baddeley and Turner (2005b),

a more detailed explanation of how to fit point process models to data; and Baddeley (2010), a complete set of notes from a 2-day workshop on the use of **spatstat**.

Type citation("spatstat") to get a list of these references.

#### Licence

This library and its documentation are usable under the terms of the "GNU General Public License", a copy of which is distributed with the package.

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beginner 31

beginner

Print Introduction For Beginners

## Description

Prints an introduction for beginners to the spatstat package, or another specified package.

## Usage

```
beginner(package = "spatstat")
```

## **Arguments**

package

Name of package.

#### **Details**

This function prints an introduction for beginners to the **spatstat** package.

The function can be executed simply by typing beginner without parentheses.

If the argument package is given, then the function prints the beginner's help file BEGINNER.txt from the specified package (if it has one).

## Value

Null.

#### Author(s)

```
Adrian Baddeley <Adrian.Baddeley@curtin.edu.au> and Rolf Turner <rolfturner@posteo.net>
```

#### See Also

latest.news

## **Examples**

beginner

32 bugfixes

bugfixes	List Recent Bug Fixes

#### **Description**

List all bug fixes in a package, starting from a certain date or version of the package. Fixes are sorted alphabetically by the name of the affected function. The default is to list bug fixes in the latest version of the **spatstat** family of packages.

#### Usage

#### **Arguments**

sinceversion	Earliest version of package for which bugs should be listed. The default is the current installed version.
sincedate	Earliest release date of package for which bugs should be listed. A character string or a date-time object.
package	The name of the package (or packages) for which bugs are to be listed. A character string or character vector.
show	Logical value indicating whether to display the bug table on the terminal.

#### **Details**

Bug reports are extracted from the NEWS file of the specified package. Only those after a specified date, or after a specified version of the package, are retained. The bug reports are then sorted alphabetically, so that all bugs affecting a particular function are listed consecutively. Finally the table of bug reports is displayed (if show=TRUE) and returned invisibly.

The argument sinceversion should be a character string like "1.2-3". The default is the current installed version of the package.

The argument sincedate should be a character string like "2015-05-27", or a date-time object.

If sinceversion="all" or sincedate="all" then all recorded bugs will be listed.

The special options sinceversion="book" and sincedate="book" are interpreted to mean sincedate="2015-06-05", which gives all bugs reported after publication of the book by Baddeley, Rubak and Turner (2015).

Typing bugfixes without parentheses will display a table of all bugs that were fixed in the current installed version of **spatstat** and its sub-packages.

By default, bugs in the *extension* packages **spatstat.local**, **spatstat.Knet**, **spatstat.gui** are *not* reported. To include these bugs as well, set package=spatstat.family(TRUE, TRUE).

#### Value

(Invisibly) a data frame, belonging to the class "bugtable", which has a print method.

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#### Author(s)

Adrian Baddeley <Adrian.Baddeley@curtin.edu.au>.

#### References

Baddeley, A., Rubak, E. and Turner, R. (2015) *Spatial Point Patterns: Methodology and Applications with R*. Chapman and Hall/CRC Press.

#### See Also

```
latest.news, news.
```

## **Examples**

```
bugfixes
## show all bugs reported after publication of the spatstat book
if(interactive()) bugfixes(sinceversion="book")
```

foo

Foo is Not a Real Name

#### **Description**

The name foo is not a real name: it is a place holder, used to represent the name of any desired thing.

The functions defined here simply print an explanation of the placeholder name foo.

## Usage

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

#### **Arguments**

```
x Ignored.
... Ignored.
```

#### **Details**

The name foo is used by computer scientists as a *place holder*, to represent the name of any desired object or function. It is not the name of an actual object or function; it serves only as an example, to explain a concept.

However, many users misinterpret this convention, and actually type the command foo or foo(). Then they email the package author to inform them that foo is not defined.

To avoid this correspondence, we have now defined an object called foo.

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The function foo() prints a message explaining that foo is not really the name of a variable.

The function can be executed simply by typing foo without parentheses.

#### Value

Null.

## Author(s)

Adrian Baddeley <Adrian.Baddeley@curtin.edu.au>, Rolf Turner <rolfturner@posteo.net> and Ege Rubak <rubak@math.aau.dk>.

## See Also

beginner

## **Examples**

foo

latest.news

Print News About Latest Version of Package

## Description

Prints the news documentation for the current version of spatstat or another specified package.

## Usage

```
latest.news(package = spatstat.family(), doBrowse=FALSE, major=TRUE)
```

## Arguments

package	Name of package for which the latest news should be printed. A character string, or vector of character strings.
doBrowse	Logical value indicating whether to display the results in a browser window instead of printing them.
major	Logical value. If TRUE (the default), print all information for the current major version "x.y". If FALSE, print only the information for the current minor version "x.y-z".

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#### **Details**

This function prints the news documentation about changes in the current installed version of a package.

By default, it prints the latest news about all the sub-packages in the **spatstat** family.

The function can be called simply by typing its name without parentheses (see the Examples).

If major=FALSE, only information for the current minor version "x.y-z" will be printed. If major=TRUE (the default), all information for the current major version "x.y" will be printed, encompassing versions "x.y-0", "x.y-1", up to "x.y-z".

If package is given, then the function reads the news for the specified package from its NEWS file (if it has one) and prints only the entries that refer to the current version of the package.

To see the news for all previous versions as well as the current version, use the R utility news. See the Examples.

#### Value

Null.

#### Author(s)

Adrian Baddeley <Adrian.Baddeley@curtin.edu.au>, Rolf Turner <rolfturner@posteo.net> and Ege Rubak <rubak@math.aau.dk>.

## See Also

```
spatstat.family lists the packages in the spatstat family.
bugfixes lists bug fixes.
news
```

## **Examples**

```
if(interactive()) {
    # current news
    latest.news

# all news
    # news(package="spatstat")
}
```

36 spatstat.family

spatstat.family Names of All Packages in the Spatstat Family

Description

Provides the names of all the packages belonging to the **spatstat** family of packages.

#### Usage

```
spatstat.family(subpackages=TRUE, extensions=FALSE)
```

## **Arguments**

subpackages Logical value specifying whether to include sub-packages.

extensions Logical value specifying whether to include extension packages.

#### **Details**

This function returns a character vector containing the names of the packages that belong to the **spatstat** family.

By default, only the sub-packages are listed, and not the extension packages.

A "sub-package" is a package which is implicitly loaded or imported when the command library(spatstat) is issued. Currently the sub-packages are:

- spatstat.utils
- spatstat.data
- spatstat.univar
- spatstat.sparse
- spatstat.geom
- spatstat.random
- spatstat.explore
- spatstat.model
- spatstat.linnet
- spatstat

An "extension package" is a package which must be loaded explicitly. The extension packages are:

- spatstat.gui
- spatstat.local
- spatstat.Knet

#### Value

Character vector of package names.

spatstat.family 37

## Author(s)

 $Adrian\ Baddeley\ < Adrian.\ Baddeley\ @curtin.\ edu.\ au>,\ Rolf\ Turner\ < rolfturner\ @posteo.\ net> and\ Ege\ Rubak\ < rubak\ @math.\ aau.\ dk>.$ 

## See Also

latest.news

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