Package 'spatstat'

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

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Depends R (>= 3.5.0), spatstat.data (>= 2.1-2), spatstat.geom (>= 2.4-0), spatstat.random (>= 2.2-0), spatstat.core (>= 2.4-1), spatstat.linnet (>= 2.3-2), utils

Imports spatstat.utils (>= 2.3-0)

Suggests

Additional_repositories https://spatstat.r-universe.dev

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 2000 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 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 de-

viance, Monte Carlo tests) along with basic tools for model selection (step-

wise(), 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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The Spatstat Package

Description

spatstat-package

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 1m and g1m. 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/.

For information about handling data in **shapefiles**, see Chapter 3, or the Vignette *Handling shapefiles in the spatstat package*, installed with **spatstat**, accessible as vignette('shapefiles').

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.geom** containing functionality for geometrical operations, and defining the main classes of spatial objects
- spatstat.core containing the main functions for statistical analysis and modelling of 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 diagnostics
- IX. 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:

as.ppp

ppp create a point pattern from (x, y) and window information

ppp(x, y, xlim, ylim) for rectangular window

ppp(x, y, poly) for polygonal window ppp(x, y, mask) for binary image window convert other types of data to a ppp object

clickppp interactively add points to a plot 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

apply random displacements to points in a pattern rjitter simulate the Matérn Model I inhibition process rMaternI simulate the Matérn Model II inhibition process rMaternII simulate Simple Sequential Inhibition process rSSI rStrauss simulate Strauss process (perfect simulation) rHardcore simulate Hard Core process (perfect simulation) simulate Strauss-hard core process (perfect simulation) rStraussHard 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 rCauchy 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

rpoispp0nLines 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 Austin Hughes' rabbit amacrine cells
anemones Upton-Fingleton sea anemones data
ants 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 patternfinpinesFinnish Pines datafluInfluenza 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

japanesepines Japanese Pines data
lansing Lansing Woods data
longleaf Longleaf Pines data
mucosa Cells in gastric mucosa
murchison Murchison gold deposits
nbfires New Brunswick fires data
nztrees Mark-Esler-Ripley trees data

Osteocyte lacunae (3D, replicated) osteo Kimboto trees in Paracou, French Guiana paracou ponderosa Getis-Franklin ponderosa pine trees data Pyramidal neurons from 31 brains pyramidal redwood Strauss-Ripley redwood saplings data redwoodfull Strauss redwood saplings data (full set) residualspaper Data from Baddeley et al (2005) Galaxies in an astronomical survey shapley

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:

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

 $\begin{array}{ll} {\sf flipxy} & {\sf swap} \ x \ {\sf and} \ y \ {\sf coordinates} \\ {\sf reflect} & {\sf reflect} \ {\sf in the origin} \end{array}$

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 only 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
as.data.frame.owin	convert window to data frame

Digital approximations:

as.mask Make a discrete pixel approximation of a given window convert window to pixel image as.im.owin pixellate.owin convert window to pixel image commonGrid find common pixel grid for windows map continuous coordinates to raster locations nearest.raster.point raster x coordinates raster.x raster y coordinates raster.y raster.xy raster x and y coordinates convert pixel mask to polygonal window as.polygonal

See spatstat.options to control the approximation

Geometrical computations with windows:

edges extract boundary edges intersection of two windows intersect.owin union.owin union of two windows set subtraction of two windows setminus.owin inside.owin determine whether a point is inside a window area.owin compute area compute perimeter length perimeter diameter.owin compute diameter find largest circle inside a window incircle inradius radius of incircle find connected components of window connected.owin eroded.areas compute areas of eroded windows dilated.areas compute areas of dilated windows bdist.points compute distances from data points to window boundary compute distances from all pixels to window boundary bdist.pixels bdist.tiles boundary distance for each tile in tessellation distmap.owin distance transform image distance transform distfun.owin centroid.owin compute centroid (centre of mass) of window determine whether one window contains another is.subset.owin determine whether a window is convex is.convex compute convex hull convexhull triangulate.owin decompose into triangles as.mask pixel approximation of window 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 setcov spatial covariance function of window extract centres of pixels in mask pixelcentres clickdist measure distance between two points clicked by user

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
pixellate	convert other data to a pixel image
as.matrix.im	convert pixel image to matrix
as.data.frame.im	convert pixel image to data frame
as.function.im	convert pixel image to function
plot.im	plot a pixel image on screen as a digital image
contour.im	draw contours of a pixel image
	ž – – – – – – – – – – – – – – – – – – –
persp.im	draw perspective plot of a pixel image
rgbim	create colour-valued pixel image
hsvim	create colour-valued pixel image
[.im	extract a subset of a pixel image
[<im< td=""><td>replace a subset of a pixel image</td></im<>	replace a subset of a pixel image
rotate.im	rotate pixel image
shift.im	apply vector shift to pixel image
affine.im	apply affine transformation to image
Χ	print very basic information about image X
<pre>summary(X)</pre>	summary of image X
hist.im	histogram of image
mean.im	mean pixel value of image
integral.im	integral of pixel values
quantile.im	quantiles of image
cut.im	convert numeric image to factor image
is.im	test whether an object is a pixel image
interp.im	interpolate a pixel image
blur	apply Gaussian blur to image
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
eval.im	evaluate any expression involving images
im.apply	evaluate a function of several images
scaletointerval	rescale pixel values
zapsmall.im	set very small pixel values to zero
levelset	level set of an image
solutionset	region where an expression is true
imcov	spatial covariance function of image
convolve.im	spatial convolution of images
transect.im	line transect of image
pixelcentres	extract centres of pixels
transmat	convert matrix of pixel values
Ci difolia c	to a different indexing convention
rnoise	random pixel noise
1110120	random pixer noise

Line segment patterns

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

psp	create a line segment pattern
as.psp	convert other data into a line segment pattern
edges	extract edges of a window
is.psp	determine whether a dataset has class "psp"
plot.psp	plot a line segment pattern
print.psp	print basic information
summary.psp	print summary information
[.psp	extract a subset of a line segment pattern
subset.psp	extract subset of line segment pattern
as.data.frame.psp	convert line segment pattern to data frame
marks.psp	extract marks of line segments
marks <psp< td=""><td>assign new marks to line segments</td></psp<>	assign new marks to line segments
unmark.psp	delete marks from line segments
midpoints.psp	compute the midpoints of line segments
endpoints.psp	extract the endpoints of line segments
lengths_psp	compute the lengths of line segments
angles.psp	compute the orientation angles of line segments
superimpose	combine several line segment patterns
flipxy	swap x and y coordinates
rotate.psp	rotate a line segment pattern
shift.psp	shift a line segment pattern
periodify	make several shifted copies
affine.psp	apply an affine transformation
pixellate.psp	approximate line segment pattern by pixel image
as.mask.psp	approximate line segment pattern by binary mask
distmap.psp	compute the distance map of a line segment pattern
distfun.psp	compute the distance map of a line segment pattern
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< td=""><td>change some tiles of a tessellation</td></tess<>	change some tiles of a tessellation
<pre>intersect.tess</pre>	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".

```
create a 3-D point pattern
pp3
plot.pp3
                   plot a 3-D point pattern
                   extract coordinates
coords
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
runifpoint3
                   generate uniform random points in 3-D
                   generate Poisson random points in 3-D
rpoispp3
                   generate simulation envelopes for 3-D pattern
envelope.pp3
box3
                   create a 3-D rectangular box
as.box3
                   convert data to 3-D rectangular box
                   name of unit of length
unitname.box3
                   diameter of box
diameter.box3
                   volume of box
volume.box3
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.

extract coordinates coords as.hyperframe extract coordinates subset.ppx extract subset unitname.ppx name of unit of length npoints count the number of points runifpointx generate uniform random points generate Poisson random points rpoisppx boxx define multidimensional box diameter.boxx diameter of box volume.boxx volume of box

volume.boxx volume of box shortside.boxx shortest side of box 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).

linnet create a linear network

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 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 lpp 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 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
interp.colourmap beachcolourmap one special colour map

II. EXPLORATORY DATA ANALYSIS

Inspection of data:

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 bw.ppl likelihood cross-validated bandwidth selection for density.ppp 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 Allard-Fraley feature detection
nnclean Byers-Raftery feature detection
sharpen.ppp
rhohat Kernel estimate of covariate effect

rho2hat Kernel estimate of effect of two covariates spatialcdf 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 quadratcount Quadrat counts

 $\begin{array}{ll} {\rm intensity.quadratcount} & {\rm Mean\ intensity\ in\ quadrats} \\ {\rm Fest} & {\rm empty\ space\ function\ } F \end{array}$

Gest nearest neighbour distribution function G

Jest J-function J = (1 - G)/(1 - F)

KestRipley's K-functionLestBesag L-functionTstatThird order T-functionall statsall four functions F, G, J, Kpcfpair correlation function

Kinhom K for inhomogeneous point patterns L inhom L for inhomogeneous point patterns per pair correlation for inhomogeneous patterns E for inhomogeneous point patterns E for inhomogeneous point patterns

Finhom F for inhomogeneous point patterns G inhom G for inhomogeneous point patterns G for inhomogeneous point patterns G for inhomogeneous point patterns

localL Getis-Franklin neighbourhood density function

localK neighbourhood K-function localpcf local pair correlation function

 $\begin{array}{ll} {\it localKinhom} & {\it local} \ K \ {\it for inhomogeneous point patterns} \\ {\it localLinhom} & {\it local} \ L \ {\it for inhomogeneous point patterns} \\ \end{array}$

localpcfinhom local pair correlation for inhomogeneous patterns Directional K-function Ksector locally scaled K-function Kscaled Kest.fft fast K-function using FFT for large datasets reduced second moment measure Kmeasure simulation envelopes for a summary function envelope varblock variances and confidence intervals for a summary function lohboot bootstrap for a summary function

Related facilities:

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
nnmap	nearest point image
nnfun	nearest point function
density.ppp	kernel smoothed density
densityHeat.ppp	diffusion kernel smoothed density
Smooth.ppp	spatial interpolation of marks
relrisk	kernel estimate of relative risk
sharpen.ppp	data sharpening
rknn	theoretical distribution of nearest neighbour distance

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.

relrisk	kernel estimation of relative risk
scan.test	spatial scan test of elevated risk
Gcross,Gdot,Gmulti	multitype nearest neighbour distributions $G_{ij}, G_{i\bullet}$
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} estimates of the above for all i, j pairs alltypes multitype *I*-function Iest Kcross.inhom,Kdot.inhom inhomogeneous counterparts of Kcross, Kdot Lcross.inhom,Ldot.inhom inhomogeneous counterparts of Lcross, Ldot pcfcross.inhom,pcfdot.inhom inhomogeneous counterparts of pcfcross, pcfdot localKcross,localKdot local counterparts of Kcross, Kdot

local counterparts of Kcross, Kdot local Lcross, 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:

markmean smoothed local average of marks markvar smoothed local variance of marks

markcorr mark correlation function markcrosscorr mark cross-correlation function

 $\begin{array}{ll} {\rm mark \, variog \, ram} \\ {\rm mark \, mark \, scatter} \\ {\rm Kmark} \end{array} \quad \begin{array}{ll} {\rm mark \, variog \, ram} \\ {\rm mark \, -mark \, scatterplot} \\ {\rm mark \, -weighted} \, K \, {\rm function} \end{array}$

 $\begin{array}{ll} {\sf Emark} & {\sf mark independence\ diagnostic}\ E(r) \\ {\sf Vmark} & {\sf mark\ independence\ diagnostic}\ V(r) \\ {\sf nnmean} & {\sf nearest\ neighbour\ mean\ index} \end{array}$

nnvario nearest neighbour mark variance index

For marks of any type, there are the following:

Gmulti multitype nearest neighbour distribution

Kmulti multitype K-function multitype J-function

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 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:

linear 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	distances between pairs
crossdist.lpp	distances between pairs
nndist.lpp	nearest neighbour distances
nncross.lpp	nearest neighbour distances
nnwhich.lpp	find nearest neighbours
nnfun.lpp	find nearest data point
density.lpp	kernel smoothing estimator of intensity
densityHeat.lpp	diffusion kernel estimate
distfun.lpp	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 FG3est nearest neighbour function GK3est 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).

```
Hest spherical contact distribution H Gfox Foxall G-function Jfox Foxall J-function
```

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
<pre>improve.kppm</pre>	Refine the estimate of trend
simulate.kppm	Generate simulated realisations
vcov.kppm	Variance-covariance matrix of coefficients
coef.kppm	Extract trend coefficients
formula.kppm	Extract trend formula
parameters	Extract all model parameters
clusterfield	Compute offspring density
clusterradius	Radius of support of offspring density
Kmodel.kppm	K function of fitted model
pcfmodel.kppm	Pair correlation of fitted model

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
```

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

command	model
ppm(X)	Complete Spatial Randomness
ppm(X ~ 1)	Complete Spatial Randomness
ppm(X ~ x)	Poisson process with
	intensity loglinear in x coordinate
$ppm(X \sim 1, Strauss(0.1))$	Stationary Strauss process
$ppm(X \sim x, Strauss(0.1))$	Strauss process with
	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

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

Ensure the model is a valid point process

See spatstat.options to control plotting of fitted model.

To specify a point process model:

project.ppm

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.

X ~ 1	No trend (stationary)
X ~ x	Loglinear trend $\lambda(x,y) = \exp(\alpha + \beta x)$
	where x, y are Cartesian coordinates
$X \sim polynom(x,y,3)$	Log-cubic polynomial trend
$X \sim harmonic(x,y,2)$	Log-harmonic polynomial trend
X ~ Z	Loglinear function of covariate Z
	$\lambda(x,y) = \exp(\alpha + \beta Z(x,y))$

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

Hard core process

HierHard() Hierarchical multiype 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 processOrdThresh()Ord process, threshold potentialOrd()Ord model, user-supplied potentialPairPiece()pairwise interaction, piecewise constantPairwise()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(0) plot quadrature scheme 0

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 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 compute simulation envelopes for a fitted model

Point process models on a linear network:

An object of class "lpp" 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 simulation envelopes for envelope.lppm point process model on linear network fitted.lppm fitted intensity values predict.lppm model prediction on linear network linim pixel image on linear network plot.linim plot a pixel image on linear network evaluate expression involving images eval.linim linfun function defined on linear network 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 ~ 1) Complete Spatial Randomness

slrm(X ~ x) Poisson process with

intensity loglinear in x coordinate

slrm(X ~ 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
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 (grid) of points
rMaternI	simulate the Matérn Model I inhibition process
rMaternII	simulate the Matérn Model II inhibition process
rSSI	simulate Simple Sequential Inhibition process
rHardcore	simulate hard core process (perfect simulation)
rStrauss	simulate Strauss 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
rMatClust simulate the Matérn Cluster process

rThomas simulate the Thomas process

rLGCP simulate the log-Gaussian Cox process rGaussPoisson simulate the Gauss-Poisson cluster process

rCauchy simulate Neyman-Scott process with Cauchy clusters

rVarGamma simulate Neyman-Scott process with Variance Gamma clusters

rcell simulate the Baddeley-Silverman cell process

runifpoint0nLines generate n random points along specified line segments generate Poisson random points along specified line segments

Resampling a point pattern:

quadratresample block resampling

rjitter apply random displacements to points in a pattern

rshift random shifting of (subsets of) points

rthin random thinning

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 bits.envelope critical envelope for balanced two-stage Monte Carlo test

qqplot.ppm diagnostic plot for interpoint interaction

scan.test spatial scan statistic/test studpermu.test studentised permutation test segregation.test test of segregation of types

VIII. TESTS AND DIAGNOSTICS

Hypothesis tests:

 χ^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's goodness-of-fit tests berman.test 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 Analysis of Deviance for point process models anova.ppm

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 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 examples from Baddeley et al (2005)

Kcom	model compensator of K function
Gcom	model compensator of G function
Kres	score residual of K function
Gres	score residual of G function
psst	pseudoscore residual of summary function
psstA	pseudoscore residual of empty space function
psstG	pseudoscore residual of G function
compareFit	compare compensators of several fitted model

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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References

Baddeley, A. (2010) *Analysing spatial point patterns in R*. Workshop notes, Version 4.1. Online technical publication, CSIRO. https://research.csiro.au/software/wp-content/uploads/sites/6/2015/02/Rspatialcourse_CMIS_PDF-Standard.pdf

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

Baddeley, A. and Turner, R. (2005a) Spatstat: an R package for analyzing spatial point patterns. *Journal of Statistical Software* **12**:6, 1–42. DOI: 10.18637/jss.v012.i06.

Baddeley, A. and Turner, R. (2005b) Modelling spatial point patterns in R. In: A. Baddeley, P. Gregori, J. Mateu, R. Stoica, and D. Stoyan, editors, *Case Studies in Spatial Point Pattern Modelling*, Lecture Notes in Statistics number 185. Pages 23–74. Springer-Verlag, New York, 2006. ISBN: 0-387-28311-0.

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Baddeley, A., Turner, R., Møller, J. and Hazelton, M. (2005) Residual analysis for spatial point processes. *Journal of the Royal Statistical Society, Series B* **67**, 617–666.

Baddeley, A., Rubak, E. and Møller, J. (2011) Score, pseudo-score and residual diagnostics for spatial point process models. *Statistical Science* **26**, 613–646.

Baddeley, A., Turner, R., Mateu, J. and Bevan, A. (2013) Hybrids of Gibbs point process models and their implementation. *Journal of Statistical Software* **55**:11, 1–43. https://www.jstatsoft.org/v55/i11/

Diggle, P.J. (2003) Statistical analysis of spatial point patterns, Second edition. Arnold.

Diggle, P.J. (2014) *Statistical Analysis of Spatial and Spatio-Temporal Point Patterns*, Third edition. Chapman and Hall/CRC.

Gelfand, A.E., Diggle, P.J., Fuentes, M. and Guttorp, P., editors (2010) *Handbook of Spatial Statistics*. CRC Press.

Huang, F. and Ogata, Y. (1999) Improvements of the maximum pseudo-likelihood estimators in various spatial statistical models. *Journal of Computational and Graphical Statistics* **8**, 510–530.

Illian, J., Penttinen, A., Stoyan, H. and Stoyan, D. (2008) Statistical Analysis and Modelling of Spatial Point Patterns. Wiley.

Waagepetersen, R. An estimating function approach to inference for inhomogeneous Neyman-Scott processes. *Biometrics* **63** (2007) 252–258.

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.

bugfixes 31

Author(s)

```
Adrian Baddeley <Adrian.Baddeley@curtin.edu.au> and Rolf Turner <r.turner@auckland.ac.nz>
```

See Also

latest.news

Examples

beginner

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.

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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.

Value

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

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, ...)
```

latest.news 33

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.

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 <r.turner@auckland.ac.nz> 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)
```

34 latest.news

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".

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 <r.turner@auckland.ac.nz> 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")
```

spatstat.family 35

}

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.sparse
- spatstat.geom
- spatstat.random
- spatstat.core
- 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.

36 spatstat.family

Author(s)

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See Also

latest.news

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