Package: spatstat (via r-universe)

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

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

Depends R (>= 3.5.0), spatstat.data (>= 3.1-4), spatstat.univar (>= 3.1-1), spatstat.geom (>= 3.3-4), spatstat.random (>= 3.3-2), spatstat.explore (>= 3.3-3), spatstat.model (>= 3.3-3), spatstat.linnet (>= 3.2-2), utils

Imports spatstat.utils (>= 3.1-1)

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.

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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:

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 pp(x, y, mask) 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 Woods data lansing 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 display a 2D point pattern:

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

persp.ppp perspective plot of marked point pattern

To manipulate a 2D 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

count the number of points npoints coords extract coordinates, change coordinates extract marks, change marks or attach marks marks rotate pattern rotate shift translate pattern flipxy swap x and y coordinates reflect reflect in the origin make several translated copies periodify apply affine transformation affine apply scalar dilation scalardilate 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 mark value of nearest data point nnmark data sharpening sharpen.ppp identify.ppp interactively identify points unique.ppp remove duplicate points duplicated.ppp determine which points are duplicates map duplicated points to unique points uniquemap.ppp find clumps of points connected.ppp compute Dirichlet-Voronoi tessellation dirichlet 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 approximate point pattern by pixel image as.im.ppp

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 open window by a distance r opening

border difference between window and its erosion/dilation

invert (swap inside and outside) complement.owin

simplify.owin approximate a window by a simple polygon

rotate rotate window

swap x and y coordinates flipxy

translate window shift

make several translated copies periodify 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

as.im.owin convert window to pixel image 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 raster y coordinates raster x and y coordinates raster.xy

convert pixel mask to polygonal window as.polygonal

See spatstat.options to control the approximation

Geometrical computations with windows:

extract boundary edges edges intersection of two windows intersect.owin union of two windows union.owin setminus.owin set subtraction of two windows

determine whether a point is inside a window inside.owin

area.owin compute area

perimeter compute perimeter length compute diameter diameter.owin

find largest circle inside a window incircle

inradius radius of incircle

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

```
compute distances from data points to window boundary
bdist.points
bdist.pixels
                     compute distances from all pixels to window boundary
                     boundary distance for each tile in tessellation
bdist.tiles
                     distance transform image
distmap.owin
distfun.owin
                     distance transform
centroid.owin
                     compute centroid (centre of mass) of window
is.subset.owin
                     determine whether one window contains another
                     determine whether a window is convex
is.convex
convexhull
                     compute convex hull
triangulate.owin
                     decompose into triangles
as.mask
                     pixel approximation of window
                     polygonal approximation of window
as.polygonal
                     test whether window is a rectangle
is.rectangle
                     test whether window is polygonal
is.polygonal
                     test whether window is a mask
is.mask
setcov
                     spatial covariance function of window
pixelcentres
                     extract centres of pixels in mask
                     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
                      convert other data to a pixel image
as.im
                      convert other data to a pixel image
pixellate
as.matrix.im
                      convert pixel image to matrix
                      convert pixel image to data frame
as.data.frame.im
                      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
                      create colour-valued pixel image
rgbim
hsvim
                      create colour-valued pixel image
[.im
                      extract a subset of a pixel image
Γ<-.im
                      replace a subset of a pixel image
                      rotate pixel image
rotate.im
shift.im
                      apply vector shift to pixel image
                      apply affine transformation to image
affine.im
                      print very basic information about image X
Χ
summary(X)
                      summary of image X
hist.im
                      histogram of image
mean.im
                      mean pixel value of image
                      integral of pixel values
integral.im
                      quantiles of image
quantile.im
                      convert numeric image to factor image
cut.im
                      test whether an object is a pixel image
is.im
                      interpolate a pixel image
interp.im
blur
                      apply Gaussian blur to image
```

apply diffusion blur to image blurHeat Smooth.im apply Gaussian blur to image apply diffusion blur to image SmoothHeat.im find connected components connected.im 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.applv scaletointerval rescale pixel values zapsmall.im set very small pixel values to zero level set of an image levelset solutionset region where an expression is true spatial covariance function of image imcov spatial convolution of images convolve.im transect.im line transect of image pixelcentres extract centres of pixels convert matrix of pixel values transmat to a different indexing convention rnoise random pixel 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 extract edges of a window edges determine whether a dataset has class "psp" is.psp plot a line segment pattern plot.psp print basic information print.psp 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 extract marks of line segments marks.psp marks<-.psp assign new marks to line segments unmark.psp delete marks from line segments compute the midpoints of line segments midpoints.psp endpoints.psp extract the endpoints of line segments 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 a line segment pattern rotate.psp 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

psp2mask	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
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".

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

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

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 network of Delaunay triangulation delaunayNetwork dirichletNetwork network of Dirichlet edges methods for linnet objects methods.linnet vertices.linnet nodes of network

join Vertices 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 "1pp" represents a point pattern on a linear network (for example, road accidents on a road network).

create a point pattern on a linear network 1pp methods for 1pp objects methods.lpp subset.lpp method for subset rpoislpp simulate Poisson points on linear network simulate random points on a linear network runiflpp chicago Chicago crime data Dendritic spines data dendrite 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 last few rows of hyperframe tail.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 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

miplot Morisita Index plot

Smoothing:

density.ppp kernel smoothed density/intensity
relrisk kernel estimate of relative risk
relriskHeat diffusion estimate of relative risk
Smooth.ppp spatial interpolation of marks
SmoothHeat.ppp spatial interpolation of marks

bw.diggle cross-validated bandwidth selection for density.ppp

bw.ppl likelihood cross-validated bandwidth selection for density.pppbw.CvL Cronie-Van Lieshout bandwidth selection for density estimation

bw. scottbw. abrambw. relriskScott's rule of thumb for density estimationAbramson's rule for adaptive bandwidthscross-validated bandwidth selection for relrisk

bw.relriskHeatppp cross-validated bandwidth selection for relriskHeat.ppp bw.smoothppp cross-validated bandwidth selection for Smooth.ppp

bw. frac bandwidth selection using window geometry bw. stoyan Stoyan's rule of thumb for bandwidth for pcf

Modern exploratory tools:

clusterset Allard-Fraley feature detection
nnclean Byers-Raftery feature detection
sharpen.ppp Choi-Hall data sharpening

rhohat Kernel estimate of covariate effect

rho2hat Kernel estimate of effect of two covariates

spatialcdf Spatial cumulative distribution function roc 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
intensity.quadratcount	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:

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 find nearest neighbours nnwhich

pairdist distances between all pairs of points distances between points in two patterns crossdist nncross nearest neighbours between two point patterns distance from any location to nearest data point exactdt

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

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

sharpen.ppp data sharpening

Kcross, Kdot, Kmulti

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} , $G_{i\bullet}$

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} multitype pair correlation function $g_{i\bullet}$ pcfdot pcfmulti general pair correlation function

markconnect marked connection function p_{ij} alltypes estimates of the above for all i, j pairs

multitype I-function Iest

Kcross.inhom, Kdot.inhom inhomogeneous counterparts of Kcross, Kdot 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 markvar smoothed local variance of marks markcorr mark correlation function

markcrosscorr mark cross-correlation function

 $\begin{array}{lll} {\tt mark vario} & {\tt mark variogram} \\ {\tt markmark scatter} & {\tt mark-mark scatterplot} \\ {\tt Kmark} & {\tt mark-weighted} \ K \ {\tt function} \\ {\tt Emark} & {\tt mark independence diagnostic} \ E(r) \\ {\tt Vmark} & {\tt mark independence diagnostic} \ V(r) \\ {\tt nnmean} & {\tt nearest neighbour mean index} \\ {\tt nnvario} & {\tt nearest neighbour mark variance index} \\ \end{array}$

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:

 $\begin{array}{lll} \mbox{linearK} & K \mbox{ function on linear network} \\ \mbox{linearpcf} & \mbox{inhomogeneous } K \mbox{ function on linear network} \\ \mbox{linearpcfinhom} & \mbox{inhomogeneous pair correlation on linear network} \\ \mbox{linearpcfinhom} & \mbox{inhomogeneous pair correlation on linear network} \\ \end{array}$

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 Mark connection function on linear network linearmarkconnect linearmarkequal Mark equality function on linear network Pair correlation between two types of points linearpcfcross 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 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
nndist.ppx nnwhich.ppx 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\mbox{-function} \\ {\sf Jfox} & {\sf Foxall} \ J\mbox{-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

 ${\sf Kmodel.kppm}$ 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 lgcp.estpcf fit a log-Gaussian Cox process model fit the Thomas process model thomas.estK fit the Thomas process model thomas.estpcf fit the Matérn Cluster process model matclust.estK fit the Matérn Cluster process model matclust.estpcf cauchy.estK fit a Neyman-Scott Cauchy cluster process fit a Neyman-Scott Cauchy cluster process cauchy.estpcf 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"):

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
predict.ppm	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
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.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
model.frame.ppm	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
<pre>DiggleGratton()</pre>	Diggle-Gratton potential
<pre>DiggleGatesStibbard()</pre>	Diggle-Gates-Stibbard potential
Fiksel()	Fiksel pairwise interaction process
Geyer()	Geyer's saturation process
Hardcore()	Hard core process
HierHard()	Hierarchical multiype hard core process
HierStrauss()	Hierarchical multiype Strauss process
<pre>HierStraussHard()</pre>	Hierarchical multiype 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
0rd()	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 pixelquad quadrature scheme based on image pixels create an object of class "quad"

To inspect a quadrature scheme:

```
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
gridcentres 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 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.

1ppm point process model on linear network 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 pixel image on linear network linim 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"):

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:

generate n independent uniform random points runifpoint 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 generate n independent uniform random points in disc runifdisc rstrat stratified random sample of points systematic random sample (grid) of points rsyst simulate the Matérn Model I inhibition process rMaternI simulate the Matérn Model II inhibition process rMaternII rSSI simulate Simple Sequential Inhibition process simulate hard core process (perfect simulation) rHardcore rStrauss simulate Strauss process (perfect simulation) simulate Strauss-hard core process (perfect simulation) rStraussHard simulate Diggle-Gratton process (perfect simulation) rDiggleGratton rDGS simulate Diggle-Gates-Stibbard process (perfect simulation) simulate Penttinen process (perfect simulation) rPenttinen rNeymanScott simulate a general Neyman-Scott process simulate the Matérn Cluster process rMatClust rThomas simulate the Thomas process rLGCP simulate the log-Gaussian Cox process rGaussPoisson simulate the Gauss-Poisson cluster process simulate Neyman-Scott process with Cauchy clusters rCauchy rVarGamma simulate Neyman-Scott process with Variance Gamma clusters simulate the Baddeley-Silverman cell process rcell runifpointOnLines generate n random points along specified line segments rpoisppOnLines 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
rpoisline	simulate the Poisson line process within a window
rpoislinetess	generate random tessellation using Poisson line process
rMosaicSet	generate random set by selecting some tiles of a tessellation
rMosaicField	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
Clark and Evans test
Spatial distribution goodness-of-fit test
Berman's goodness-of-fit tests
critical envelope for Monte Carlo test of goodness-of-fit
spatial scan statistic/test
Diggle-Cressie-Loosmore-Ford test
Mean Absolute Deviation test
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 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
qqplot.ppm	diagnostic Q-Q plot for interpoint interaction
residualspaper	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 models

Resampling and randomisation procedures

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

```
quadratresample block resampling
rjitter apply random displacements to points in a pattern
```

rshift random shifting of (subsets of) points rthin random 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

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Author(s)

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

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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	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. A character string. 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.

foo 33

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.changes, 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.

34 latest.changes

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)

 $Adrian\ Baddeley\ \verb|\Adrian.Baddeley@curtin.edu.au|>,\ Rolf\ Turner\ \verb|\Colored| Furner\ \verb|\Colored| Furner\ \verb|\Colored| Baddeley\ \verb|\Colored| Rubak\ \verb|\Colored| Furner\ \verb|\Colored| Baddeley\ \verb|\Colored| Rubak\ \verb|\Colored| Furner\ \verb|\Colored| Baddeley\ \verb|\Colored| Rubak\ \verb|\Colored| Baddeley\ \verb|\Colored| Baddeley\ \verb|\Colored| Rubak\ \verb|\Colored| Baddeley\ \verb|\Colored| Rubak\ \verb|\Colored| Baddeley\ \verb|\Colored| Baddeley\ \verb|\Colored| Baddeley\ \verb|\Colored| Rubak\ \verb|\Colored| Rubak\ \verb|\Colored| Baddeley\ Ba$

See Also

beginner

Examples

foo

latest.changes

List Recent Significant Changes to a Function

Description

List all the changes to a particular function in a package, starting from a certain date or version of the package. The default is to list changes in the latest version of the **spatstat** family of packages.

Usage

Arguments

X	Character string giving the name of the function of interest, or a search pattern to be matched. A character vector is permitted.
sinceversion	Earliest version of package for which changes should be listed. A character string. Default is the most recent version.
sincedate	Earliest release date of package for which changes should be listed. A character string or a date-time object. Default is the date of the most recent version.
package	The name of the package (or packages) for which changes are to be listed. A character string or character vector.
show	Logical value indicating whether to display the table of changes on the terminal.

latest.news 35

Details

Details of changes are extracted from the NEWS file of the specified package under the heading 'Significant User-Visible Changes'. All entries for which the first line matches x are selected. The table of changes is displayed (if show=TRUE) and returned invisibly.

The argument sinceversion should be a character string like "1.2-3". The default is the version string of the most recent version.

The argument sincedate should be a character string like "2015-05-27", or a date-time object. The default is the date of the most recent version.

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

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

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

Value

(Invisibly) a data frame, belonging to the class "changetable", 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

bugfixes, latest.news, news.

Examples

latest.changes("plot.symbolmap")

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)
```

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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)

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

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 37

}

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

38 spatstat.family

Value

Character vector of package names.

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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