

spatstat Quick Reference 1.4-3

Type `demo(spatstat)` for an overall demonstration.

Creation, manipulation and plotting of point patterns

An object of class "ppp" describes a point pattern. If the points have marks, these are included as a component vector `marks`.

To create a point pattern:

<code>ppp</code>	create a point pattern from (x, y) and window information <code>ppp(x, y, xlim, ylim)</code> for rectangular window <code>ppp(x, y, poly)</code> for polygonal window <code>ppp(x, y, mask)</code> for binary image window
<code>as.ppp</code>	convert other types of data to a ppp object
<code>setmarks</code>	
<code>%mark%</code>	attach/reassign marks to a point pattern

To simulate a random point pattern:

<code>runifpoint</code>	generate n independent uniform random points
<code>rpoint</code>	generate n independent random points
<code>rmpoint</code>	generate n independent multitype random points
<code>rpoispp</code>	simulate the (in)homogeneous Poisson point process
<code>rmpoispp</code>	simulate the (in)homogeneous multitype Poisson point process
<code>rMaternI</code>	simulate the Matérn Model I inhibition process
<code>rMaternII</code>	simulate the Matérn Model II inhibition process
<code>rSSI</code>	simulate Simple Sequential Inhibition process
<code>rNeymanScott</code>	simulate a general Neyman-Scott process
<code>rMatClust</code>	simulate the Matérn Cluster process
<code>rThomas</code>	simulate the Thomas process
<code>rmh</code>	simulate Gibbs point process using Metropolis-Hastings

Standard point pattern datasets:

Remember to say `data(BrambleCanes)` etc.

<code>amacrine</code>	Austin Hughes' rabbit amacrine cells
<code>bramblecanes</code>	Bramble Canes data
<code>cells</code>	Crick-Ripley biological cells data
<code>ganglia</code>	Wässle et al. cat retinal ganglia data
<code>hamster</code>	Aherne's hamster tumour data
<code>lansing</code>	Lansing Woods data
<code>longleaf</code>	Longleaf Pines data
<code>nztrees</code>	Mark-Esler-Ripley trees data
<code>redwood</code>	Strauss-Ripley redwood saplings data
<code>redwoodfull</code>	Strauss redwood saplings data (full set)
<code>swedishpines</code>	Strand-Ripley swedish pines data

To manipulate a point pattern:

<code>plot.ppp</code>	plot a point pattern <code>plot(X)</code>
<code>"[.ppp"</code>	extract a subset of a point pattern <code>pp[subset]</code> <code>pp[, subwindow]</code>
<code>superimpose</code>	superimpose any number of point patterns
<code>cut.ppp</code>	discretise the marks in a point pattern
<code>unmark</code>	remove marks
<code>setmarks</code>	attach marks or reset marks
<code>rotate</code>	rotate pattern
<code>shift</code>	translate pattern
<code>affine</code>	apply affine transformation
<code>ksmooth.ppp</code>	kernel smoothing
<code>identify.ppp</code>	interactively identify points

See `spatstat.options` to control plotting behaviour.

To create a window:

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

<code>owin</code>	Create a window object <code>owin(xlim, ylim)</code> for rectangular window <code>owin(poly)</code> for polygonal window <code>owin(mask)</code> for binary image window
<code>as.owin</code>	Convert other data to a window object
<code>ripras</code>	Ripley-Rasson estimator of window, given only the points
<code>letterR</code>	polygonal window in the shape of the R logo

To manipulate a window:

<code>plot.owin</code>	plot a window. <code>plot(W)</code>
<code>bounding.box</code>	Find a tight bounding box for the window
<code>erode.owin</code>	erode window by a distance <code>r</code>
<code>complement.owin</code>	invert (inside \leftrightarrow outside)
<code>rotate</code>	rotate window
<code>shift</code>	translate window
<code>affine</code>	apply affine transformation

Digital approximations:

<code>as.mask</code>	Make a discrete pixel approximation of a given window
<code>nearest.raster.point</code>	map continuous coordinates to raster locations
<code>raster.x</code>	raster x coordinates
<code>raster.y</code>	raster y coordinates

See `spatstat.options` to control the approximation

Geometrical computations with windows:

<code>inside.owin</code>	determine whether a point is inside a window
<code>area.owin</code>	compute window's area
<code>diameter</code>	compute window frame's diameter
<code>eroded.areas</code>	compute areas of eroded windows
<code>bdist.points</code>	compute distances from data points to window boundary
<code>bdist.pixels</code>	compute distances from all pixels to window boundary
<code>centroid.owin</code>	compute centroid (centre of mass) of window
<code>is.subset.owin</code>	determine whether one window contains another
<code>trim.owin</code>	intersect a window with a rectangle

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 `ksmooth.ppp`.

<code>im</code>	create a pixel image
<code>as.im</code>	convert other data to a pixel image
<code>plot.im</code>	plot a pixel image on screen as a digital image
<code>contour.im</code>	draw contours of a pixel image
<code>persp.im</code>	draw perspective plot of a pixel image
<code>[.im</code>	extract subset of pixel image
<code>shift.im</code>	apply vector shift to pixel image
<code>X</code>	print very basic information about image <code>X</code>
<code>summary(X)</code>	summary of image <code>X</code>
<code>is.im</code>	test whether an object is a pixel image

Exploratory Data Analysis

Inspection of data

`summary(X)` print useful summary of point pattern **X**
`X` print basic description of point pattern **X**

Summary statistics for a point pattern:

`Fest` empty space function F'
`Gest` nearest neighbour distribution function G
`Kest` Ripley's K -function
`Jest` J -function $J = (1 - G)/(1 - F')$
`allstats` all four functions F' , G , J , K
`pcf` pair correlation function
`Kinhom` K for inhomogeneous point patterns
`Kest.fft` fast K -function using FFT for large datasets
`Kmeasure` reduced second moment measure

Summary statistics for a multitype point pattern:

A multitype point pattern is represented by an object **X** of class "ppp" with a component **X\$marks** which is a factor.

`Gcross, Gdot, Gmulti` multitype nearest neighbour distributions $G_{ij}, G_{i\bullet}$
`Kcross, Kdot, Kmulti` multitype K -functions $K_{ij}, K_{i\bullet}$
`Jcross, Jdot, Jmulti` multitype J -functions $J_{ij}, J_{i\bullet}$
`alltypes` estimates of the above for all i, j pairs

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**.

`markcorr` mark correlation function
`Gmulti` multitype nearest neighbour distribution
`Kmulti` multitype K -function
`Jmulti` 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

Model Fitting

To fit a point process model:

Model fitting in `spatstat` version 1.4 is performed by the function `mpl`. Its result is an object of class `ppm`.

`mpl` Fit a point process model
to a two-dimensional point pattern

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
`fitted.ppm` Compute fitted conditional intensity at quadrature points
`update.ppm` Update the fit
`rmh.ppm` Simulate from fitted model
`print.ppm` Print basic information about a fitted model
`summary.ppm` Summarise a fitted model
See `spatstat.options` to control plotting of fitted model.

To specify a point process model:

The first order “trend” of the model is written as an `S` language formula.

`~1` No trend (stationary)
`~x` First order term $\lambda(x, y) = \exp(\alpha + \beta x)$
where x, y are Cartesian coordinates
`~polynom(x, y, 3)` Log-cubic polynomial trend

The higher order (“interaction”) components are described by an object of class `interact`.

Such objects are created by:

`Poisson()` the Poisson point process
`Strauss()` the Strauss process
`StraussHard()` the Strauss/hard core point process
`Softcore()` pairwise interaction, soft core potential
`PairPiece()` pairwise interaction, piecewise constant
`DiggleGratton()` Diggle-Gratton potential
`LennardJones()` Lennard-Jones potential
`Pairwise()` pairwise interaction, user-supplied potential
`Geyer()` Geyer’s saturation process
`Saturated()` Saturated pair model, user-supplied potential
`OrdThresh()` Ord process, threshold potential
`Ord()` Ord model, user-supplied potential
`MultiStrauss()` multitype Strauss process
`MultiStraussHard()` multitype Strauss/hard core process

Finer control over model fitting:

A quadrature scheme is represented by an object of class "quad".

<code>quadscheme</code>	generate a Berman-Turner quadrature scheme for use by <code>mpl</code>
<code>default.dummy</code>	default pattern of dummy points
<code>gridcentres</code>	dummy points in a rectangular grid
<code>stratrand</code>	stratified random dummy pattern
<code>spokes</code>	radial pattern of dummy points
<code>corners</code>	dummy points at corners of the window
<code>gridweights</code>	quadrature weights by the grid-counting rule
<code>dirichlet.weights</code>	quadrature weights are Dirichlet tile areas
<code>print(Q)</code>	print basic information about quadrature scheme <code>Q</code>
<code>summary(Q)</code>	summary of quadrature scheme <code>Q</code>