# Package 'dbmss'

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Type Package

```
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Description Simple computation of spatial statistic functions of distance to characterize the spa-
      tial structures of mapped objects, following Marcon, Trais-
      sac, Puech, and Lang (2015) <doi:10.18637/jss.v067.c03>.
      Includes classical functions (Ripley's K and others) and more recent ones used by spa-
      tial economists (Duranton and Overman's Kd, Marcon and Puech's M).
      Relies on 'spatstat' for some core calculation.
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dbmss-package

Distance Based Measures of Spatial Structures

#### Description

Simple computation of spatial statistic functions of distance to characterize the spatial structures of mapped objects, including classical ones (Ripley's *K* and others) and more recent ones used by spatial economists (Duranton and Overman's *Kd*, Marcon and Puech's *M*). Relies on spatstat for some core calculation.

#### Author(s)

Eric Marcon, Gabriel Lang, Stephane Traissac, Florence Puech

Maintainer: Eric Marcon < Eric. Marcon@agroparistech.fr>

## References

Marcon, E., and Puech, F. (2003). Evaluating the Geographic Concentration of Industries Using Distance-Based Methods. *Journal of Economic Geography*, 3(4), 409-428.

Marcon, E. and Puech, F. (2010). Measures of the Geographic Concentration of Industries: Improving Distance-Based Methods. *Journal of Economic Geography* 10(5): 745-762.

Marcon, E., Puech F. and Traissac, S. (2012). Characterizing the relative spatial structure of point patterns. *International Journal of Ecology* 2012(Article ID 619281): 11.

Lang G., Marcon E. and Puech F. (2014) Distance-Based Measures of Spatial Concentration: Introducing a Relative Density Function. *HAL* 01082178, 1-18.

Marcon, E., Traissac, S., Puech, F. and Lang, G. (2015). Tools to Characterize Point Patterns: dbmss for R. *Journal of Statistical Software*. 67(3): 1-15.

Marcon, E. and Puech, F. (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

as.Dtable

Converts data to class Dtable

#### **Description**

Creates an object of class "Dtable" representing a set of points with weights and labels and the distances between them.. This is a generic method.

# Usage

```
as.Dtable(X, ...)
  ## S3 method for class 'ppp'
as.Dtable(X, ...)
  ## S3 method for class 'data.frame'
as.Dtable(X, ...)
```

as.wmppp

# **Arguments**

X Data to be converted into a "Dtable".

... Extra arguments.

#### **Details**

This is a generic method, implemented for ppp and data.frame.

Data is first converted to a (wmppp.object). Then, the distance matrix between points is calculated and the marks are kept.

## Value

An object of class "Dtable".

#### See Also

```
as.wmppp
```

as.wmppp

Converts data to class wmppp

# Description

Creates a Weighted, Marked, Planar Point Pattern, *i.e.* an object of class "wmppp" representing a two-dimensional point pattern with weights and labels. This is a generic method.

# Usage

```
as.wmppp(X, ...)
  ## S3 method for class 'ppp'
as.wmppp(X, ...)
  ## S3 method for class 'data.frame'
as.wmppp(X, window = NULL, unitname = NULL, ...)
```

# **Arguments**

X Data to be converted into a weighted, marked, planar point pattern (wmppp.object)

window An object of calls "owin" (owin.object).

unitname Name of unit of length. Either a single character string, or a vector of two

character strings giving the singular and plural forms, respectively.

... Extra arguments.

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## **Details**

This is a generic method, implemented for ppp and data.frame:

• If the dataset X is an object of class "ppp" (ppp.object), the marks are converted to point weights if they are numeric or to point types if they are factors. Default weights are set to 1, default types to "All". If marks are a dataframe with column names equal to PointType and PointWeight, they are not modified. Row names of the dataframe are preserved as row names of the marks, to identify points.

• If the dataset X is a dataframe, see wmppp.

#### Value

An object of class "wmppp".

#### See Also

```
wmppp.object
```

autoplot

ggplot methods to plot dbmss objects

## **Description**

S3 methods for the autoplot generic.

## Usage

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#### **Arguments**

object An object to be plotted.

fmla An R language formula determining which variables or expressions are plotted.

Either a formula object, or a string that can be parsed as a formula. See plot. fv.

... Extra arguments, currently unused.

ObsColor The color of the line representing observed values of the function.

H0Color The color of the line representing the null hypothesis values of the function.

ShadeColor The color of the confidence envelope.

alpha The opacity of the confidence envelope (in function values) or the points (in

maps), between 0 and 1.

main The title of the plot. xlab, ylab The axes labels.

LegendLabels A vector of characters. The first two items describe the observed and null-

hypothesis curves, the third and last item the confidence interval. To be used only in plots with two curves (typically observed and expected values). The

default is 'NULL' to display the full description of functions.

show.window if 'TRUE', the borders of the window containing the points are shown on the

point map.

MaxPointTypes The maximum number of different point types to show. If the point set contains

more of them, the less frequent ones are gathered as "Other". This number must be limited for readability and not to exceed the number of colors offered by the

palette.

Other The name of the point types gathered as "Other".

labelSize The guide of the point size legend in point maps, i.e. what the 'PointSize' mark

represents.

labelColor The guide of the point color legend in point maps, i.e. what the 'PointType'

mark represents.

palette The color palette used to display point types in maps. See scale\_colour\_brewer

windowColor The color used to draw the limits of the windows in point maps.

windowFill The color used to fill the windows in point maps.

# Details

Plots of 'wmppp' objects are a single representation of both point types and point weights. Rectangular and polygonal windows (see owin.object) are supported but mask windows are ignored (use the 'plot' method if necessary).

#### Value

A ggplot object.

# Author(s)

Eric Marcon < Eric.Marcon@agroparistech.fr>, parts of the code from spatstat.explore::plot.fv.

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# **Examples**

dbmssEnvelope.object Class of envelope of function values (fv)

# **Description**

A class "dbmssEnvelope", *i.e.* a particular type of see envelope to represent several estimates of the same function and its confidence envelope.

# **Details**

"dbmssEnvelope" objects are similar to envelope objects. The differences are that the risk level is chosen (instead of the simulation rank to use as the envelope), so the rank is calculated (interpolation is used if necessary), and a global envelope can be calculated following Duranton and Overman (2005).

# References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106

# See Also

summary.dbmssEnvelope, KdEnvelope, MEnvelope

DEnvelope	Estimation of the confidence envelope of the D function under its null hypothesis

# **Description**

Simulates point patterns according to the null hypothesis and returns the envelope of D according to the confidence level.

# Usage

### **Arguments**

X A point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following spatstat.

NumberOfSimulations

The number of simulations to run, 100 by default.

Alpha The risk level, 5% by default.

Cases One of the point types
Controls One of the point types.

Intertype Logical; if TRUE, D is computed as Di in Marcon and Puech (2012).

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

## **Details**

The only null hypothesis is random labeling: marks are distributed randomly across points.

This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

# Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

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## References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. *Ecology* 69(4): 1017-1024.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

## See Also

Dhat

# **Examples**

```
data(paracou16)
# Keep only 20% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.2))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Calculate confidence envelope (should be 1000 simulations, reduced to 20 to save time)
r <- 0:30
NumberOfSimulations <- 20
Alpha <- .05
# Plot the envelope (after normalization by pi.r^2)
autoplot(DEnvelope(X, r, NumberOfSimulations, Alpha,
    "V. Americana", "Q. Rosea", Intertype = TRUE), ./(pi*r^2) ~ r)</pre>
```

Dhat

Estimation of the D function

#### **Description**

Estimates the D function

#### **Usage**

```
Dhat(X, r = NULL, Cases, Controls = NULL, Intertype = FALSE, CheckArguments = TRUE)
```

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#### **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

Cases One of the point types.

Controls One of the point types. If NULL, controls are all types except for cases. Intertype Logical; if TRUE, *D* is computed as *Di* in Marcon and Puech (2012).

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

#### **Details**

The *Di* function allows comparing the structure of the cases to that of the controls around cases, that is to say the comparison is made around the same points. This has been advocated by Arbia et al. (2008) and formalized by Marcon and Puech (2012).

#### Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

#### Note

The computation of Dhat relies on spatstat functions Kest and Kcross.

#### References

Arbia, G., Espa, G. and Quah, D. (2008). A class of spatial econometric methods in the empirical analysis of clusters of firms in the space. *Empirical Economics* 34(1): 81-103.

Diggle, P. J. and Chetwynd, A. G. (1991). Second-Order Analysis of Spatial Clustering for Inhomogeneous Populations. *Biometrics* 47(3): 1155-1163.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

#### See Also

```
Khat, DEnvelope, Kest, Kcross
```

# **Examples**

```
data(paracou16)
autoplot(paracou16)

# Calculate D
r <- 0:30
(Paracou <- Dhat(paracou16, r, "V. Americana", "Q. Rosea", Intertype = TRUE))

# Plot (after normalization by pi.r^2)
autoplot(Paracou, ./(pi*r^2) ~ r)</pre>
```

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Create a Distance table object.	Dtable	Create a Distance table object.	
	Diable	Credie d Distance table object.	

# **Description**

Creates an object of class "Dtable" representing a set of points with weights and labels and the distances between them.

# Usage

```
Dtable(Dmatrix, PointType = NULL, PointWeight = NULL)
```

# **Arguments**

Dmatrix A square matrix containing distances.

PointType A vector describing the point types. Its length must correspond to the number

of points.

PointWeight A vector describing the point weights. Its length must correspond to the number

of points.

## **Details**

The distance matrix is not necessarily symmetric, so distances are understood in the common sense, not in the mathematical sense. Asymmetric distances are appropriate when paths between points are one-way only.

The points of origin are in lines, the targets in columns. The diagonal of the matrix must contain zeros (the distance between a point and itself is 0), and all other distances must be positive (they can be 0).

## Value

An object of class "Dtable". It is a list:

Dmatrix The distance matrix.

n The number of points.

marks A list of two items: PointType, a vector of factors containing the point types

and PointWeight, the numeric vector of weights.

## See Also

as.Dtable

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# **Examples**

```
# A Dtable containing two points
Dmatrix <- matrix(c(0,1,1,0), nrow=2)
PointType <- c("Type1", "Type2")
PointWeight <- c(2,3)
Dtable(Dmatrix, PointType, PointWeight)</pre>
```

envelope.Dtable

Computes simulation envelopes of a summary function.

# **Description**

Prints a useful summary of a confidence envelope of class "dbmssEnvelope"

# Usage

# **Arguments**

Υ	An object of class Dtable.
fun	Function that computes the desired summary statistic for Y.
nsim	Number of simulated point patterns to be generated when computing the envelopes.
nrank	Integer. Rank of the envelope value amongst the nsim simulated values. A rank of 1 means that the minimum and maximum simulated values will be used.
	Extra arguments passed to fun.
funargs	A list, containing extra arguments to be passed to fun.
funYargs	Optional. A list, containing extra arguments to be passed to fun when applied to the original data Y only.
simulate	Optional. Specifies how to generate the simulated point patterns.
verbose	Logical flag indicating whether to print progress reports during the simulations.
savefuns	Logical flag indicating whether to save all the simulated function values.
Yname	Character string that should be used as the name of the data Y when printing or plotting the results.
envir.simul	Environment in which to evaluate the expression simulate, if not the current environment.

# **Details**

This is the S3 method envelope for Dtable objects.

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#### Author(s)

Eric Marcon <Eric.Marcon@agroparistech.fr>. Relies on the envelope engine of spatstat.

gEnvelope	Estimation of the confidence envelope of the g function under its null hypothesis
genverope	

# **Description**

Simulates point patterns according to the null hypothesis and returns the envelope of *g* according to the confidence level.

## Usage

## **Arguments**

X A point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

NumberOfSimulations

The number of simulations to run, 100 by default.

Alpha The risk level, 5% by default.

ReferenceType One of the point types. Default is all point types.

NeighborType One of the point types. Default is all point types.

SimulationType A string describing the null hypothesis to simulate. The null hypothesis may

be "RandomPosition": points are drawn in a Poisson process (default); "RandomLabeling": randomizes point types, keeping locations unchanged; "PopulationIndependence": keeps reference points unchanged, shifts other point locations.

tions.

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

#### **Details**

This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

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#### Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

## References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. *Ecology* 69(4): 1017-1024.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

## See Also

 $\verb|ghat|, \verb|rR| and \verb|omPositionK|, \verb|rR| and \verb|omLocation|, \verb|rPopulationIndependenceK||$ 

# Examples

```
data(paracou16)
# Keep only 20% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.2))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Calculate confidence envelope (should be 1000 simulations, reduced to 10 to save time)
r <- 0:40
NumberOfSimulations <- 10
# Plot the envelope
autoplot(gEnvelope(X, r, NumberOfSimulations))</pre>
```

ghat

Estimation of the g function

# Description

Estimates the g function

# Usage

```
ghat(X, r = NULL, ReferenceType = "", NeighborType = "", CheckArguments = TRUE)
```

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# **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

ReferenceType One of the point types. Default is all point types.

NeighborType One of the point types. Default is all point types.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

## **Details**

The computation of ghat relies on spatstat function sewpcf.

#### Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

## References

Stoyan, D. and Stoyan, H. (1994) Fractals, random shapes and point fields: methods of geometrical statistics. John Wiley and Sons.

#### See Also

gEnvelope

# **Examples**

```
data(paracou16)
autoplot(paracou16)

# Calculate g
r <- 0:30
(Paracou <- ghat(paracou16, r, "Q. Rosea", "V. Americana"))

# Plot
autoplot(Paracou)</pre>
```

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GoFtest	Goodness of Fit test between a distance based measure of spatial
	structure and simulations of its null hypothesis

# **Description**

Calculates the risk to reject the null hypothesis erroneously, based on the distribution of the simulations.

## Usage

GoFtest(Envelope)

# Arguments

Envelope

An envelope object (envelope) containing simulations in its simfuns attribute. It may be the result of any estimation function of the dbmss package or obtained by the envelope function with argument savefuns=TRUE.

#### **Details**

This test was introduced by Diggle(1983) and extensively developed by Loosmore and Ford (2006) for *K*, and applied to *M* by Marcon et al. (2012).

#### Value

A p-value.

## Note

No support exists in the literature to apply the GoF test to non-cumulative functions (g, Kd...).

Ktest is a much better test (it does not rely on simulations) but it is limited to the *K* function against complete spatial randomness (CSR) in a rectangle window.

#### References

Diggle, P. J. (1983). *Statistical analysis of spatial point patterns*. Academic Press, London. 148 p. Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

Marcon, E., F. Puech and S. Traissac (2012). Characterizing the relative spatial structure of point patterns. International *Journal of Ecology* 2012(Article ID 619281): 11.

#### See Also

Ktest

is.wmppp

# **Examples**

```
# Simulate a Matern (Neyman Scott) point pattern
nclust <- function(x0, y0, radius, n) {
   return(runifdisc(n, radius, centre=c(x0, y0)))
}
X <- rNeymanScott(20, 0.2, nclust, radius=0.3, n=10)
autoplot(as.wmppp(X))

# Calculate confidence envelope (should be 1000 simulations, reduced to 50 to save time)
r <- seq(0, 0.3, 0.01)
NumberOfSimulations <- 50
Alpha <- .10
Envelope <- KEnvelope(as.wmppp(X), r, NumberOfSimulations, Alpha)
autoplot(Envelope, ./(pi*r^2) ~ r)

# GoF test. Power is correct if enough simulations are run (say >1000).
paste("p-value =", GoFtest(Envelope))
```

is.wmppp

Test whether an object is a weighted, marked, planar point pattern

# **Description**

Check whether its argument is an object of class "wmppp" (wmppp.object).

# Usage

```
is.wmppp(X)
```

# **Arguments**

Χ

Any object

#### Value

TRUE if X is a weighted, marked, planar point pattern, otherwise FALSE.

# See Also

```
wmppp.object
```

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KdEnvelope	Estimation of the confidence envelope of the Kd function under its null hypothesis
KdEnvelope	

## **Description**

Simulates point patterns according to the null hypothesis and returns the envelope of *Kd* according to the confidence level.

# Usage

### Arguments

X A point pattern (wmppp.object) or a Dtable object.

A vector of distances. If NULL, a default value is set: 512 equally spaced values are used, and the first 256 are returned, corresponding to half the maximum

distance between points (following Duranton and Overman, 2005).

**NumberOfSimulations** 

The number of simulations to run, 100 by default.

Alpha The risk level, 5% by default.

ReferenceType One of the point types.

NeighborType One of the point types. By default, the same as reference type.

Weighted Logical; if TRUE, estimates the *Kemp* function.

Original Logical; if TRUE (by default), the original bandwidth selection by Duranton and

Overman (2005) following Silverman (2006: eq 3.31) is used. If FALSE, it is calculated following Sheather and Jones (1991), *i.e.* the state of the art. See

bw. SJ for more details.

Approximate if not 0 (1 is a good choice), exact distances between pairs of points are rounded

to 1024 times Approximate single values equally spaced between 0 and the largest distance. This technique (Scholl and Brenner, 2015) allows saving a lot of memory when addressing large point sets (the default value is 1 over 10000 points). Increasing Approximate allows better precision at the cost of propor-

tional memory use. Ignored if X is a Dtable object.

Adjust Force the automatically selected bandwidth (following Silverman, 1986) to be

multiplied by Adjust. Setting it to values lower than one (1/2 for example) will

sharpen the estimation. If not 1, Original is ignored.

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MaxRange The maximum value of r to consider, ignored if r is not NULL. Default is "ThirdW",

one third of the diameter of the window. Other choices are "HalfW", and "QuarterW" and "D02005". "HalfW", and "QuarterW" are for half or the quarter of the diameter of the window. "D02005" is for the median distance observed between points, following Duranton and Overman (2005). "ThirdW" should be close to "D02005" but has the advantage to be independent of the point types chosen as ReferenceType and NeighborType, to simplify comparisons between different types. "D02005" is approximated by "ThirdW" if Approximate is not 0. if X is a Dtable object, the diameter of the window is taken as the max distance between

points.

StartFromMinR Logical; if TRUE, points are assumed to be further from each other than the

minimum observed distance, So Kd will not be estimated below it: it is assumed to be 0. If FALSE, by default, distances are smoothed down to r=0. Ignored if

Approximate is not 0: then, estimation always starts from r = 0.

SimulationType A string describing the null hypothesis to simulate. The null hypothesis may

be "RandomLocation": points are redistributed on the actual locations (default); "RandomLabeling": randomizes point types, keeping locations and weights unchanged; "PopulationIndependence": keeps reference points unchanged, ran-

domizes other point locations.

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

## **Details**

This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

# Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. *Ecology* 69(4): 1017-1024.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

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Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

Scholl, T. and Brenner, T. (2015) Optimizing distance-based methods for large data sets, *Journal of Geographical Systems* 17(4): 333-351.

Silverman, B. W. (1986). *Density estimation for statistics and data analysis*. Chapman and Hall, London.

## See Also

Kdhat

# **Examples**

```
data(paracou16)
autoplot(paracou16[paracou16$marks$PointType=="Q. Rosea"])

# Calculate confidence envelope
plot(KdEnvelope(paracou16, , ReferenceType="Q. Rosea", Global=TRUE))

# Center of the confidence interval
Kdhat(paracou16, ReferenceType="") -> kd
lines(kd$Kd ~ kd$r, lty=3, col="green")
```

Kdhat

Estimation of the Kd function

### **Description**

Estimates the Kd function

#### Usage

```
Kdhat(X, r = NULL, ReferenceType, NeighborType = ReferenceType, Weighted = FALSE,
    Original = TRUE, Approximate = ifelse(X$n < 10000, 0, 1), Adjust = 1,
    MaxRange = "ThirdW", StartFromMinR = FALSE, CheckArguments = TRUE)</pre>
```

#### **Arguments**

X A weighted, marked planar point pattern (wmppp.object) or a Dtable object.

r A vector of distances. If NULL, a default value is set: 512 equally spaced values are used, from the smallest distance between points to half the diameter of the

window.

ReferenceType One of the point types. If "", all points are considered (this is not the default

value; NeighborType is ignored then) to estimate the average value of simulated *Kd* values under the null hypothesis of *RandomLocation* (Marcon and Puech,

2012).

NeighborType One of the point types. By default, the same as reference type.

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Weighted Logical; if TRUE, estimates the *Kemp* function.

Original Logical; if TRUE (by default), the original bandwidth selection by Duranton and

Overman (2005) following Silverman (1986: eq 3.31) is used. If FALSE, it is calculated following Sheather and Jones (1991), *i.e.* the state of the art. See

bw. SJ for more details.

Approximate if not 0 (1 is a good choice), exact distances between pairs of points are rounded

to 1024 times Approximate single values equally spaced between 0 and the largest distance. This technique (Scholl and Brenner, 2015) allows saving a lot of memory when addressing large point sets (the default value is 1 over 10000 points). Increasing Approximate allows better precision at the cost of propor-

tional memory use. Ignored if X is a Dtable object.

Adjust Force the automatically selected bandwidth (following Original) to be mul-

tiplied by Adjust. Setting it to values lower than one (1/2 for example) will

sharpen the estimation.

MaxRange The maximum value of r to consider, ignored if r is not NULL. Default is "ThirdW",

one third of the diameter of the window. Other choices are "HalfW", and "QuarterW" and "D02005". "HalfW", and "QuarterW" are for half or the quarter of the diameter of the window. "D02005" is for the median distance observed between points, following Duranton and Overman (2005). "ThirdW" should be close to "DO2005" but has the advantage to be independent of the point types chosen as ReferenceType and NeighborType, to simplify comparisons between different types. "D02005" is approximated by "ThirdW" if Approximate is not 0. if X is a Dtable object, the diameter of the window is taken as the max distance between

points.

StartFromMinR Logical; if TRUE, points are assumed to be further from each other than the min-

imum observed distance, So Kd will not be estimated below it: it is assumed to be 0. If FALSE, distances are smoothed down to r=0. Ignored if Approximate

is not 0: then, estimation always starts from r = 0.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

## **Details**

*Kd* is a density, absolute measure of a point pattern structure. *Kd* is computed efficiently by building a matrix of distances between point pairs and calculating the density of their distribution (the default values of r are those of the density function). The kernel estimator is Gaussian.

The weighted *Kd* function has been named *Kemp* (*emp* is for employees) by Duranton and Overman (2005).

If X is not a Dtable object, the maximum value of r is obtained from the geometry of the window rather than caculating the median distance between points as suggested by Duranton and Overman (2005) to save (a lot of) calculation time.

#### Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

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#### Note

Estimating Kd relies on calculating distances, exactly or approximately (if Approximate is not 0). Then distances are smoothed by estimating their probability density. Reflection is used to estimate density close to the lowest distance, that is the minimum observed distance (if StartFromMinR is TRUE) or 0: all distances below 4 times the estimation kernel bandwith apart from the lowest distance are duplicated (symmetrically with respect to the lowest distance) to avoid edge effects (underestimation of the density close to the lowest distance).

Density estimation heavily relies on the bandwith. Starting from version 2.7, the optimal bandwith is computed from the distribution of distances between pairs of points up to twice the maximum distance considered. The consequence is that choosing a smaller range of distances in argument r results in less smoothed Kd values. The default values (r = NULL, MaxRange = "ThirdW") are such that almost all the pairs of points (except those more than 2/3 of the window diameter apart) are taken into account to determine the bandwith.

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

Scholl, T. and Brenner, T. (2015) Optimizing distance-based methods for large data sets, *Journal of Geographical Systems* 17(4): 333-351.

Sheather, S. J. and Jones, M. C. (1991) A reliable data-based bandwidth selection method for kernel density estimation. *Journal of the Royal Statistical Society series B*, 53, 683-690.

Silverman, B. W. (1986). *Density estimation for statistics and data analysis*. Chapman and Hall, London.

#### See Also

KdEnvelope, Mhat

# **Examples**

```
data(paracou16)
autoplot(paracou16)

# Calculate Kd
(Paracou <- Kdhat(paracou16, , "Q. Rosea", "V. Americana"))
# Plot
autoplot(Paracou)</pre>
```

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KEnvelope	Estimation of the confidence envelope of the K function under its null hypothesis

# Description

Simulates point patterns according to the null hypothesis and returns the envelope of K according to the confidence level.

## Usage

```
KEnvelope(X, r = NULL, NumberOfSimulations = 100, Alpha = 0.05,
    ReferenceType = "", NeighborType = ReferenceType,
    SimulationType = "RandomPosition", Precision = 0, Global = FALSE,
    verbose = interactive())
```

## **Arguments**

X A point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

NumberOfSimulations

The number of simulations to run, 100 by default.

Alpha The risk level, 5% by default.

ReferenceType One of the point types. Default is all point types.

NeighborType One of the point types. By default, the same as reference type.

SimulationType A string describing the null hypothesis to simulate. The null hypothesis may

be "RandomPosition": points are drawn in a Poisson process (default); "RandomLabeling": randomizes point types, keeping locations unchanged; "PopulationIndependence": keeps reference points unchanged, shifts other point locations.

tions.

Precision Accuracy of point coordinates, measured as a part of distance unit. See rRandomPositionK.

Default is 0 for no approximation.

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

#### **Details**

This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

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#### Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. *Ecology* 69(4): 1017-1024.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

Silverman, B. W. (1986). *Density estimation for statistics and data analysis*. Chapman and Hall, London.

## See Also

Khat, rRandomPositionK, rRandomLocation, rPopulationIndependenceK

# **Examples**

```
data(paracou16)
# Keep only 20% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.2))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Calculate confidence envelope (should be 1000 simulations, reduced to 20 to save time)
r <- 0:30
NumberOfSimulations <- 20
# Plot the envelope
autoplot(KEnvelope(X, r, NumberOfSimulations), ./(pi*r^2) ~ r)</pre>
```

Khat

Estimation of the K function

## **Description**

Estimates the *K* function

## Usage

```
Khat(X, r = NULL, ReferenceType = "", NeighborType = ReferenceType, CheckArguments = TRUE)
```

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# **Arguments**

X A	weighted, marked,	planar point	nattern (wmppp	.obiect).
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r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

ReferenceType One of the point types. Default is all point types.

NeighborType One of the point types. By default, the same as reference type.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

## **Details**

*K* is a cumulative, topographic measure of a point pattern structure.

## Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

# Note

The computation of Khat relies on spatstat functions Kest and Kcross.

#### References

Ripley, B. D. (1976). The Foundations of Stochastic Geometry. *Annals of Probability* 4(6): 995-998.

Ripley, B. D. (1977). Modelling Spatial Patterns. *Journal of the Royal Statistical Society B* 39(2): 172-212.

# See Also

```
Lhat, KEnvelope, Ktest
```

# **Examples**

```
data(paracou16)
autoplot(paracou16)

# Calculate K
r <- 0:30
(Paracou <- Khat(paracou16, r))

# Plot (after normalization by pi.r^2)
autoplot(Paracou, ./(pi*r^2) ~ r)</pre>
```

26 KinhomEnvelope

KinhomEnvelope	Estimation of the confidence envelope of the Kinhom function under its null hypothesis

# Description

Simulates point patterns according to the null hypothesis and returns the envelope of *Kinhom* according to the confidence level.

### Usage

## Arguments

X A point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

NumberOfSimulations

The number of simulations to run.

Alpha The risk level.

ReferenceType One of the point types. Default is all point types.

lambda An estimation of the point pattern density, obtained by the density.ppp func-

tion.

SimulationType A string describing the null hypothesis to simulate. The null hypothesis, may

be "RandomPosition": points are drawn in an inhomogenous Poisson process (intensity is either lambda or estimated from X); "RandomLocation": points are redistributed across actual locations; "RandomLabeling": randomizes point types, keeping locations unchanged; "PopulationIndependence": keeps refer-

ence points unchanged, redistributes others across actual locations.

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

#### **Details**

The random location null hypothesis is that of Duranton and Overman (2005). It is appropriate to test the univariate *Kinhom* function of a single point type, redistributing it over all point locations. It allows fixing lambda along simulations so the warning message can be ignored.

The random labeling hypothesis is appropriate for the bivariate Kinhom function.

The population independence hypothesis is that of Marcon and Puech (2010).

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This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

#### Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. *Ecology* 69(4): 1017-1024.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

Marcon, E. and Puech, F. (2010). Measures of the Geographic Concentration of Industries: Improving Distance-Based Methods. *Journal of Economic Geography* 10(5): 745-762.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

## See Also

Kinhomhat

## **Examples**

```
data(paracou16)
# Keep only 20% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.2))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Density of all trees
lambda <- density.ppp(X, bw.diggle(X))
plot(lambda)
V.americana <- X[X$marks$PointType=="V. Americana"]
plot(V.americana, add=TRUE)

# Calculate Kinhom according to the density of all trees
# and confidence envelope (should be 1000 simulations, reduced to 4 to save time)
r <- 0:30
NumberOfSimulations <- 4</pre>
```

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Kinhomhat

Estimation of the inhomogenous K function

## **Description**

Estimates the Kinhom function

#### Usage

```
Kinhomhat(X, r = NULL, ReferenceType = "", lambda = NULL, CheckArguments = TRUE)
```

## **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

ReferenceType One of the point types. Default is all point types.

lambda An estimation of the point pattern density, obtained by the density.ppp func-

tion.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

## **Details**

Kinhom is a cumulative, topographic measure of an inhomogenous point pattern structure.

By default, density estimation is performed at points by density.ppp using the optimal bandwith (bw.diggle). It can be calculated separately (see example), including at pixels if the point pattern is too large for the default estimation to succeed, and provided as the argument lambda: Arbia et al. (2012) for example use another point pattern as a reference to estimate density.

Bivariate *Kinhom* is not currently supported.

## Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

#### Note

The computation of Kinhomhat relies on spatstat functions Kinhom, density.ppp and bw.diggle.

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## References

Baddeley, A. J., J. Moller, et al. (2000). Non- and semi-parametric estimation of interaction in inhomogeneous point patterns. *Statistica Neerlandica* 54(3): 329-350.

Arbia, G., G. Espa, et al. (2012). Clusters of firms in an inhomogeneous space: The high-tech industries in Milan. *Economic Modelling* 29(1): 3-11.

#### See Also

KinhomEnvelope, Kinhom

## **Examples**

```
data(paracou16)

# Density of all trees
lambda <- density.ppp(paracou16, bw.diggle(paracou16))
plot(lambda)

# Reduce the point pattern to one type of trees
V.americana <- paracou16[paracou16$marks$PointType=="V. Americana"]
plot(V.americana, add=TRUE)

# Calculate Kinhom according to the density of all trees
r <- 0:30
autoplot(Kinhomhat(paracou16, r, "V. Americana", lambda), ./(pi*r^2) ~ r)</pre>
```

KmmEnvelope

Estimation of the confidence envelope of the Lmm function under its null hypothesis

# Description

Simulates point patterns according to the null hypothesis and returns the envelope of *Lmm* according to the confidence level.

# Usage

## **Arguments**

X A point pattern (wmppp.object).

A vector of distances. If NULL, a sensible default value is chosen (512 intervals, from 0 to half the diameter of the window) following **spatstat**.

 ${\tt NumberOfSimulations}$ 

The number of simulations to run, 100 by default.

Alpha The risk level, 5% by default.

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ReferenceType One of the point types. Others are ignored. Default is all point types.

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

#### **Details**

This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

#### Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. *Ecology* 69(4): 1017-1024.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

## See Also

Kmmhat

## **Examples**

```
data(paracou16)
# Keep only 20% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.2))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Calculate confidence envelope (should be 1000 simulations, reduced to 4 to save time)
r <- seq(0, 30, 2)
NumberOfSimulations <- 4
Alpha <- .10
autoplot(KmmEnvelope(X, r, NumberOfSimulations, Alpha), ./(pi*r^2) ~ r)</pre>
```

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Kmm	hat	
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Estimation of the Kmm function

# **Description**

Estimates of the Kmm function

#### **Usage**

```
Kmmhat(X, r = NULL, ReferenceType = "", CheckArguments = TRUE)
```

## **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

ReferenceType One of the point types. Others are ignored. Default is all point types.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

## **Details**

The *Kmm* function is used to test the independence of marks.

# Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

# Note

The function is computed using markcorrint in spatstat.

# References

Penttinen, A., Stoyan, D. and Henttonen, H. M. (1992). Marked Point Processes in Forest Statistics. *Forest Science* 38(4): 806-824.

Penttinen, A. (2006). Statistics for Marked Point Patterns. in *The Yearbook of the Finnish Statistical Society*. The Finnish Statistical Society, Helsinki: 70-91.

#### See Also

Lmmhat, LmmEnvelope, markcorrint

32 Ktest

# **Examples**

```
data(paracou16)
# Keep only 50% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.5))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Calculate Kmm
r <- seq(0, 30, 2)
(Paracou <- Kmmhat(X, r))
# Plot
autoplot(Paracou, ./(pi*r^2) ~ r)</pre>
```

Ktest

Test of a point pattern against Complete Spatial Randomness

# Description

Tests the point pattern against CSR using values of the K function

## Usage

```
Ktest(X, r)
```

# Arguments

X A point pattern (ppp.object). Marks are ignored. The window must be a rectangle sensu spatstat (tested by is.rectangle).

r A vector of distances.

#### **Details**

The test returns the risk to reject CSR erroneously, i.e. the p-value of the test, based on the distribution of the *K* function.

If r includes 0, it will be silently removed because no neighbor point can be found at distance 0. The longer r, the more accurate the test is in theory but at the cost of computation time first, and of computation accuracy then because a matrix of size the length of r must be inverted. 10 values in r seems to be a reasonable choice.

#### Value

A p-value.

## Author(s)

Gabriel Lang @agroparistech.fr>, Eric Marcon @Eric.Marcon@agroparistech.fr>

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#### References

Lang, G. and Marcon, E. (2013). Testing randomness of spatial point patterns with the Ripley statistic. *ESAIM: Probability and Statistics*. 17: 767-788.

Marcon, E., S. Traissac, and Lang, G. (2013). A Statistical Test for Ripley's Function Rejection of Poisson Null Hypothesis. *ISRN Ecology* 2013(Article ID 753475): 9.

#### See Also

```
Khat, GoFtest
```

## **Examples**

```
# Simulate a Matern (Neyman Scott) point pattern
nclust <- function(x0, y0, radius, n) {
   return(runifdisc(n, radius, centre=c(x0, y0)))
}
X <- rNeymanScott(20, 0.1, nclust, radius=0.2, n=5)
autoplot(as.wmppp(X))
# Test it
Ktest(X, r=seq(0.1, .5, .1))</pre>
```

kwmppp

Kriged, Weighted, Marked, Planar point pattern

# **Description**

Kriged, Weighted, Marked, Planar point pattern are maps of the individual values of dbmss's.

# Usage

### **Arguments**

X A point pattern (wmppp.object).

fvind An object of class fv, see fv.object, obtained a distance-based method, such as Mhat with individual values (argument Individual = TRUE).

ReferenceType The point type used to calculate the function values. The default value is "", i.e. all point types, but it will generate an error if the actual reference type is

different.

distance The distance at which the function value must be considered. The default value

is the median distance used to calculate the function values.

AllowJitter If TRUE, duplicated points are jittered to avoid their elimination by the kriging

procedure.

Nbx, Nby The number of columns and rows (pixels) of the resulting map, 128 by default.

Increase it for quality, paid by increasing computing time.

Further parameters passed to image.An object of class "kwmppp" to plot.ContourIf TRUE, contours are added to the map.

Palette The color palette of the map.

SuppressMargins

If TRUE, the margins around the map are suppressed.

Contournlevels The number of levels of contours.

Contourcol The color of the contour lines.

CheckArguments If TRUE (default), the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

#### Value

kmppp() produces an object of class "kwmppp" that is also an autoKrige object that can be used to produce alternative maps. It can be plotted.

#### **Examples**

```
ReferenceType <- "V. Americana"
NeighborType <- "Q. Rosea"
# Calculate individual intertype M(distance) values
fvind <- Mhat(paracou16, , ReferenceType, NeighborType, Individual=TRUE)
# Plot the point pattern with values of M(30 meters)
plot(kwmppp(paracou16, fvind, ReferenceType, distance=30))
# Add the reference points to the plot
is.ReferenceType <- paracou16$marks$PointType == ReferenceType
points(x=paracou16$x[is.ReferenceType], y=paracou16$y[is.ReferenceType], pch=20)</pre>
```

LEnvelope Estimation of the confidence envelope of the L function under its null hypothesis

# Description

Simulates point patterns according to the null hypothesis and returns the envelope of L according to the confidence level.

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## Usage

#### **Arguments**

X A point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

NumberOfSimulations

The number of simulations to run, 100 by default.

Alpha The risk level, 5% by default.

ReferenceType One of the point types. Default is all point types.

NeighborType One of the point types. Default is all point types.

SimulationType A string describing the null hypothesis to simulate. The null hypothesis may

be "RandomPosition": points are drawn in a Poisson process (default); "RandomLabeling": randomizes point types, keeping locations unchanged; "PopulationIndependence": keeps reference points unchanged, randomizes other point

locations.

Precision Accuracy of point coordinates, measured as a part of distance unit. See rRandomPositionK.

Default is 0 for no approximation.

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

#### **Details**

This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

#### Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

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## References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. Review of Economic Studies 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. Ecology 69(4): 1017-1024.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. Regional Science and Urban Economics. 62:56-67.

#### See Also

Khat

#### **Examples**

```
data(paracou16)
# Keep only 20% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.2))</pre>
autoplot(X,
  labelSize = expression("Basal area (" ~cm^2~ ")"),
  labelColor = "Species")
# Calculate confidence envelope (should be 1000 simulations, reduced to 20 to save time)
r < -0:30
NumberOfSimulations <- 20
# Plot the envelope
autoplot(LEnvelope(X, r, NumberOfSimulations))
```

Lhat

*Estimation of the L function* 

# **Description**

Estimates the *L* function

# **Usage**

```
Lhat(X, r = NULL, ReferenceType = "", NeighborType = "", CheckArguments = TRUE)
```

# **Arguments**

Χ A weighted, marked, planar point pattern (wmppp.object).

A vector of distances. If NULL, a sensible default value is chosen (512 intervals, from 0 to half the diameter of the window) following **spatstat**.

ReferenceType One of the point types. Default is all point types. LmmEnvelope 37

NeighborType One of the point types. Default is all point types.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

## **Details**

L is the normalized version of K:  $L(r) = \sqrt{\frac{K}{\pi}} - r$ .

#### Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

## Note

L was originally defined as  $L(r)=\sqrt{rac{K}{\pi}}$ . It has been used as  $L(r)=\sqrt{rac{K}{\pi}}-r$  in a part of the literature because this normalization is easier to plot.

## References

Besag, J. E. (1977). Comments on Ripley's paper. *Journal of the Royal Statistical Society B* 39(2): 193-195.

#### See Also

Khat, LEnvelope

## **Examples**

```
data(paracou16)
autoplot(paracou16)

# Calculate L
r <- 0:30
(Paracou <- Lhat(paracou16, r))

# Plot
autoplot(Paracou)</pre>
```

LmmEnvelope

Estimation of the confidence envelope of the Lmm function under its null hypothesis

## **Description**

Simulates point patterns according to the null hypothesis and returns the envelope of *Lmm* according to the confidence level.

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#### Usage

#### **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

NumberOfSimulations

The number of simulations to run, 100 by default.

Alpha The risk level, 5% by default.

ReferenceType One of the point types. Others are ignored. Default is all point types.

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

#### **Details**

This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

# Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. *Ecology* 69(4): 1017-1024.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

## See Also

Lmmhat

Lmmhat 39

## **Examples**

```
data(paracou16)
# Keep only 20% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.2))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Calculate confidence envelope (should be 1000 simulations, reduced to 4 to save time)
r <- seq(0, 30, 2)
NumberOfSimulations <- 4
Alpha <- .10
autoplot(LmmEnvelope(X, r, NumberOfSimulations, Alpha))</pre>
```

Lmmhat

Estimation of the Lmm function

# **Description**

Estimates the *Lmm* function

## Usage

```
Lmmhat(X, r = NULL, ReferenceType = "", CheckArguments = TRUE)
```

## **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

r A vector of distances. If NULL, a sensible default value is chosen (512 intervals,

from 0 to half the diameter of the window) following **spatstat**.

ReferenceType One of the point types. Others are ignored. Default is all point types.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

#### **Details**

 $\mathit{Lmm}$  is the normalized version of  $\mathit{Kmm}$ :  $\mathit{Lmm}(r) = \sqrt{\frac{\mathit{Kmm}}{\pi}} - r$ .

#### Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

## References

Penttinen, A., Stoyan, D. and Henttonen, H. M. (1992). Marked Point Processes in Forest Statistics. *Forest Science* 38(4): 806-824.

Espa, G., Giuliani, D. and Arbia, G. (2010). Weighting Ripley's K-function to account for the firm dimension in the analysis of spatial concentration. *Discussion Papers*, 12/2010. Universita di Trento, Trento: 26.

#### See Also

Kmmhat, LmmEnvelope

# **Examples**

```
data(paracou16)
# Keep only 50% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.5))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Calculate Lmm
r <- seq(0, 30, 2)
(Paracou <- Lmmhat(X, r))
# Plot
autoplot(Paracou)</pre>
```

MEnvelope

Estimation of the confidence envelope of the M function under its null hypothesis

## **Description**

Simulates point patterns according to the null hypothesis and returns the envelope of M according to the confidence level.

## Usage

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### **Arguments**

X A point pattern (wmppp.object) or a Dtable object.

r A vector of distances. If NULL, a default value is set: 32 unequally spaced values

are used up to half the maximum distance between points  $d_m$ . The first value is 0, first steps are small  $(d_m/200)$  then incresase progressively up to  $d_m/20$ .

NumberOfSimulations

The number of simulations to run, 100 by default.

Alpha The risk level, 5% by default.

ReferenceType One of the point types.

NeighborType One of the point types, equal to the reference type by default to caculate univari-

ate M.

CaseControl Logical; if TRUE, the case-control version of M is computed. ReferenceType

points are cases, NeighborType points are controls.

SimulationType A string describing the null hypothesis to simulate. The null hypothesis may

be "RandomLocation": points are redistributed on the actual locations (default); "RandomLabeling": randomizes point types, keeping locations and weights unchanged; "PopulationIndependence": keeps reference points unchanged, ran-

domizes other point locations.

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

#### **Details**

This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

#### Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. *Ecology* 69(4): 1017-1024.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

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Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

## See Also

Mhat

## **Examples**

```
data(paracou16)
# Keep only 50% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.5))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Calculate confidence envelope (should be 1000 simulations, reduced to 4 to save time)
NumberOfSimulations <- 4
Alpha <- .10
autoplot(MEnvelope(X, , NumberOfSimulations, Alpha,
    "V. Americana", "Q. Rosea", FALSE, "RandomLabeling"))</pre>
```

mEnvelope

Estimation of the confidence envelope of the m function under its null hypothesis

## **Description**

Simulates point patterns according to the null hypothesis and returns the envelope of m according to the confidence level.

## Usage

```
mEnvelope(X, r = NULL, NumberOfSimulations = 100, Alpha = 0.05,
    ReferenceType, NeighborType = ReferenceType, CaseControl = FALSE,
    Original = TRUE, Approximate = ifelse(X$n < 10000, 0, 1), Adjust = 1,
    MaxRange = "ThirdW", SimulationType = "RandomLocation", Global = FALSE,
    verbose = interactive())</pre>
```

# **Arguments**

X A point pattern (wmppp.object).

A vector of distances. If NULL, a default value is set: 512 equally spaced values are used up to the median distance between points (following Duranton and Overman, 2005).

NumberOfSimulations

The number of simulations to run, 100 by default.

Alpha The risk level, 5% by default.

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ReferenceType One of the point types.

NeighborType One of the point types, equal to the reference type by default to caculate univari-

ate M.

CaseControl Logical; if TRUE, the case-control version of M is computed. ReferenceType

points are cases, NeighborType points are controls.

Original Logical; if TRUE (by default), the original bandwidth selection by Duranton and

Overman (2005) following Silverman (1986: eq 3.31) is used. If FALSE, it is calculated following Sheather and Jones (1991), *i.e.* the state of the art. See

bw. SJ for more details.

Approximate if not 0 (1 is a good choice), exact distances between pairs of points are rounded

to 1024 times Approximate single values equally spaced between 0 and the largest distance. This technique (Scholl and Brenner, 2015) allows saving a lot of memory when addressing large point sets (the default value is 1 over 10000 points). Increasing Approximate allows better precision at the cost of propor-

tional memory use.

Adjust Force the automatically selected bandwidth (following Original) to be mul-

tiplied by Adjust. Setting it to values lower than one (1/2 for example) will

sharpen the estimation.

MaxRange The maximum value of r to consider, ignored if r is not NULL. Default is "ThirdW",

one third of the diameter of the window. Other choices are "HalfW", and "QuarterW" and "D02005". "HalfW", and "QuarterW" are for half or the quarter of the diameter of the window. "D02005" is for the median distance observed between points, following Duranton and Overman (2005). "ThirdW" should be close to "D02005" but has the advantage to be independent of the point types chosen as ReferenceType and NeighborType, to simplify comparisons between different

types. "D02005" is approximated by "ThirdW" if Approximate is not 0.

SimulationType A string describing the null hypothesis to simulate. The null hypothesis may

be "RandomLocation": points are redistributed on the actual locations (default); "RandomLabeling": randomizes point types, keeping locations and weights unchanged; "PopulationIndependence": keeps reference points unchanged, ran-

domizes other point locations.

Global Logical; if TRUE, a global envelope sensu Duranton and Overman (2005) is cal-

culated.

verbose Logical; if TRUE, print progress reports during the simulations.

# **Details**

This envelope is local by default, that is to say it is computed separately at each distance. See Loosmore and Ford (2006) for a discussion.

The global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

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#### Value

An envelope object (envelope). There are methods for print and plot for this class.

The fv contains the observed value of the function, its average simulated value and the confidence envelope.

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Kenkel, N. C. (1988). Pattern of Self-Thinning in Jack Pine: Testing the Random Mortality Hypothesis. *Ecology* 69(4): 1017-1024.

Lang G., Marcon E. and Puech F. (2014) Distance-Based Measures of Spatial Concentration: Introducing a Relative Density Function. *HAL* 01082178, 1-18.

Loosmore, N. B. and Ford, E. D. (2006). Statistical inference using the G or K point pattern spatial statistics. *Ecology* 87(8): 1925-1931.

Marcon, E. and F. Puech (2017). A typology of distance-based measures of spatial concentration. *Regional Science and Urban Economics*. 62:56-67.

Scholl, T. and Brenner, T. (2015) Optimizing distance-based methods for large data sets, *Journal of Geographical Systems* 17(4): 333-351.

Silverman, B. W. (1986). *Density estimation for statistics and data analysis*. Chapman and Hall, London.

# See Also

mhat

## **Examples**

```
data(paracou16)
# Keep only 50% of points to run this example
X <- as.wmppp(rthin(paracou16, 0.5))
autoplot(X,
    labelSize = expression("Basal area (" ~cm^2~ ")"),
    labelColor = "Species")

# Calculate confidence envelope (should be 1000 simulations, reduced to 4 to save time)
NumberOfSimulations <- 4
Alpha <- .10
autoplot(mEnvelope(X, , NumberOfSimulations, Alpha,
    "V. Americana", "Q. Rosea", Original = FALSE, SimulationType = "RandomLabeling"))</pre>
```

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Mhat	Estimation of the M function	

### **Description**

Estimates the M function

elsewhere.

## Usage

## **Arguments**

Χ A weighted, marked planar point pattern (wmppp.object) or a Dtable object. r A vector of distances. If NULL, a default value is set: 64 unequally spaced values are used up to half the maximum distance between points  $d_m$ . The first value is 0, first steps are small  $(d_m/800)$  then increase progressively up to  $d_m/40$ . ReferenceType One of the point types. NeighborType One of the point types. By default, the same as reference type. CaseControl Logical; if TRUE, the case-control version of M is computed. ReferenceType points are cases, *NeighborType* points are controls. Individual Logical; if TRUE, values of the function around each individual point are returned. CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to save time in simulations for example, when the arguments have been checked

## Details

*M* is a weighted, cumulative, relative measure of a point pattern structure. Its value at any distance is the ratio of neighbors of the *NeighborType* to all points around *ReferenceType* points, normalized by its value over the windows.

If *CaseControl* is TRUE, then *ReferenceType* points are cases and *NeighborType* points are controls. The univariate concentration of cases is calculated as if *NeighborType* was equal to *ReferenceType*, but only controls are considered when counting all points around cases (Marcon et al., 2012). This makes sense when the sampling design is such that all points of *ReferenceType* (the cases) but only a sample of the other points (the controls) are recorded. Then, the whole distribution of points is better represented by the controls alone.

## Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

If Individual is set to TRUE, the object also contains the value of the function around each individual *ReferenceType* point taken as the only reference point. The column names of the fv are "M\_" followed by the point names, i.e. the row names of the marks of the point pattern.

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#### References

Marcon, E. and Puech, F. (2010). Measures of the Geographic Concentration of Industries: Improving Distance-Based Methods. *Journal of Economic Geography* 10(5): 745-762.

Marcon, E., F. Puech and S. Traissac (2012). Characterizing the relative spatial structure of point patterns. *International Journal of Ecology* 2012(Article ID 619281): 11.

Marcon, E., and Puech, F. (2017). A Typology of Distance-Based Measures of Spatial Concentration. *Regional Science and Urban Economics* 62:56-67

#### See Also

MEnvelope, Kdhat

#### **Examples**

```
data(paracou16)
autoplot(paracou16)
# Calculate M
autoplot(Mhat(paracou16, , "V. Americana", "Q. Rosea"))
```

mhat

Estimation of the m function

## **Description**

Estimates the m function

#### Usage

```
mhat(X, r = NULL, ReferenceType, NeighborType = ReferenceType,
    CaseControl = FALSE, Original = TRUE, Approximate = ifelse(X$n < 10000, 0, 1),
    Adjust = 1, MaxRange = "ThirdW", Individual = FALSE, CheckArguments = TRUE)</pre>
```

## **Arguments**

X A weighted, marked planar point pattern (wmppp.object) or a Dtable object.

r A vector of distances. If NULL, a default value is set: 512 equally spaced values are used, from the smallest distance to the range defined by MaxRange. the

between points to half the diameter of the window.

ReferenceType One of the point types.

NeighborType One of the point types. By default, the same as reference type.

CaseControl Logical; if TRUE, the case-control version of M is computed. ReferenceType

points are cases, NeighborType points are controls.

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Original Logical; if TRUE (by default), the original bandwidth selection by Duranton and

Overman (2005) following Silverman (1986: eq 3.31) is used. If FALSE, it is calculated following Sheather and Jones (1991), *i.e.* the state of the art. See

bw. SJ for more details.

Approximate if not 0 (1 is a good choice), exact distances between pairs of points are rounded

to 1024 times Approximate single values equally spaced between 0 and the largest distance. This technique (Scholl and Brenner, 2015) allows saving a lot of memory when addressing large point sets (the default value is 1 over 10000 points). Increasing Approximate allows better precision at the cost of propor-

tional memory use. Ignored if X is a Dtable object.

Adjust Force the automatically selected bandwidth (following Original) to be mul-

tiplied by Adjust. Setting it to values lower than one (1/2 for example) will

sharpen the estimation.

MaxRange The maximum value of r to consider, ignored if r is not NULL. Default is "ThirdW",

one third of the diameter of the window. Other choices are "HalfW", and "QuarterW" and "D02005". "HalfW", and "QuarterW" are for half or the quarter of the diameter of the window. "D02005" is for the median distance observed between points, following Duranton and Overman (2005). "ThirdW" should be close to "D02005" but has the advantage to be independent of the point types chosen as ReferenceType and NeighborType, to simplify comparisons between different types. "D02005" is approximated by "ThirdW" if Approximate is not 0. If X is a Dtable object, the diameter of the window is taken as the max distance between

points.

Individual Logical; if TRUE, values of the function around each individual point are re-

turned.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

#### **Details**

m is a weighted, density, relative measure of a point pattern structure (Lang et al., 2014). Its value at any distance is the ratio of neighbors of the NeighborType to all points around ReferenceType points, normalized by its value over the windows.

The number of neighbors at each distance is estimated by a Gaussian kernel whose bandwith is chosen optimally according to Silverman (1986: eq 3.31). It can be sharpened or smoothed by multiplying it by Adjust. The bandwidth of Sheather and Jones (1991) would be better but it is very slow to calculate for large point patterns and it sometimes fails. It is often sharper than that of Silverman.

If X is not a Dtable object, the maximum value of r is obtained from the geometry of the window rather than caculating the median distance between points as suggested by Duranton and Overman (2005) to save (a lot of) calculation time.

If *CaseControl* is TRUE, then *ReferenceType* points are cases and *NeighborType* points are controls. The univariate concentration of cases is calculated as if *NeighborType* was equal to *ReferenceType*, but only controls are considered when counting all points around cases (Marcon et al., 2012). This makes sense when the sampling design is such that all points of *ReferenceType* (the cases) but only

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a sample of the other points (the controls) are recorded. Then, the whole distribution of points is better represented by the controls alone.

### Value

An object of class fv, see fv. object, which can be plotted directly using plot. fv.

If Individual is set to TRUE, the object also contains the value of the function around each individual *ReferenceType* point taken as the only reference point. The column names of the fv are "m\_" followed by the point names, i.e. the row names of the marks of the point pattern.

#### Note

Estimating m relies on calculating distances, exactly or approximately (if Approximate is not 0). Then distances are smoothed by estimating their probability density. In contrast with Kdhat, reflection is not used to estimate density close to the lowest distance. The same kernel estimation is applied to the distances from reference points of neighbor points and of all points. Since m is a relative function, a ratio of densities is calculated, that makes the features of the estimation vanish.

Density estimation heavily relies on the bandwith. Starting from version 2.7, the optimal bandwith is computed from the distribution of distances between pairs of points up to twice the maximum distance considered. The consequence is that choosing a smaller range of distances in argument r results in less smoothed m values. The default values (r = NULL, MaxRange = "ThirdW") are such that almost all the pairs of points (except those more than 2/3 of the window diameter apart) are taken into account to determine the bandwith.

## References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Lang G., Marcon E. and Puech F. (2014) Distance-Based Measures of Spatial Concentration: Introducing a Relative Density Function. *HAL* 01082178, 1-18.

Marcon, E., F. Puech and S. Traissac (2012). Characterizing the relative spatial structure of point patterns. *International Journal of Ecology* 2012(Article ID 619281): 11.

Scholl, T. and Brenner, T. (2015) Optimizing distance-based methods for large data sets, *Journal of Geographical Systems* 17(4): 333-351.

Sheather, S. J. and Jones, M. C. (1991) A reliable data-based bandwidth selection method for kernel density estimation. *Journal of the Royal Statistical Society series B*, 53, 683-690.

Silverman, B. W. (1986). Density estimation for statistics and data analysis. Chapman and Hall, London.

#### See Also

mEnvelope, Kdhat

## **Examples**

data(paracou16)
autoplot(paracou16)

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```
# Calculate M
autoplot(mhat(paracou16, , "V. Americana", "Q. Rosea"))
```

paracou16

Paracou field station plot 16, partial map

## **Description**

This point pattern is from Paracou field station, French Guiana, managed by Cirad.

## Usage

```
data(paracou16)
```

#### **Format**

An object of class ppp.object representing the point pattern of tree locations in a 250 x 300 meter sampling region. Each tree is marked with its species ("Q. Rosea", "V. Americana" or "Other"), and basal area (square centimeters).

#### **Source**

Permanent data census of Paracou and Marcon et al. (2012).

#### References

Gourlet-Fleury, S., Guehl, J. M. and Laroussinie, O., Eds. (2004). *Ecology & management of a neotropical rainforest. Lessons drawn from Paracou, a long-term experimental research site in French Guiana*. Paris, Elsevier.

Marcon, E., F. Puech and S. Traissac (2012). Characterizing the relative spatial structure of point patterns. *International Journal of Ecology* 2012(Article ID 619281): 11.

# **Examples**

```
data(paracou16)
# Plot (second column of marks is Point Types)
autoplot(paracou16, which.marks=2, leg.side="right")
```

print.dbmssEnvelope

print.dbmssEnvelope
Print a confidence envelope

# **Description**

Prints useful information of a confidence envelope of class "dbmssEnvelope"

## Usage

```
## S3 method for class 'dbmssEnvelope'
print(x, ...)
```

Ignored.

# **Arguments**

x An object of class "dbmssEnvelope".

#### **Details**

"dbmssEnvelope" objects are similar to envelope objects. The way they are printed is different to take into account the possibility of building global envelope following Duranton and Overman (2005): the global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

## References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106

# Examples

```
data(paracou16)
autoplot(paracou16)

# Calculate intertype K envelope
Envelope <- KEnvelope(paracou16, NumberOfSimulations = 20, Global = TRUE,
    ReferenceType = "V. Americana", NeighborType = "Q. Rosea")
autoplot(Envelope)
# print
print(Envelope)</pre>
```

rPopulationIndependenceK

Simulations of a point pattern according to the null hypothesis of population independence defined for K

# **Description**

Simulates of a point pattern according to the null hypothesis of population independence defined for K

# Usage

rPopulationIndependenceK(X, ReferenceType, NeighborType, CheckArguments = TRUE)

# Arguments

X A weighted, marked, planar point pattern (wmppp.object).

ReferenceType One of the point types.

NeighborType One of the point types.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

# **Details**

Reference points are kept unchanged, neighbor type point positions are shifted by rshift. Other points are lost and point weights are not kept (they are set to 1) since the K function ignores them.

## Value

A new weighted, marked, planar point pattern (an object of class wmppp, see wmppp.object).

## References

Goreaud, F. et Pelissier, R. (2003). Avoiding misinterpretation of biotic interactions with the intertype K12 fonction: population independence vs random labelling hypotheses. *Journal of Vegetation Science* 14(5): 681-692.

#### See Also

rPopulationIndependenceM, rRandomLabeling

## **Examples**

```
# Simulate a point pattern with three types
X <- rpoispp(50)
PointType <- sample(c("A", "B", "C"), X$n, replace=TRUE)
PointWeight <- runif(X$n, min=1, max=10)
X$marks <- data.frame(PointType, PointWeight)
X <- as.wmppp(X)

# Plot the point pattern, using PointType as marks
autoplot(X, main="Original pattern")

# Randomize it
Y <- rPopulationIndependenceK(X, "A", "B")
# Points of type "A" are unchanged, points of type "B" have been moved altogether
# Other points are lost and point weights are set to 1
autoplot(Y, main="Randomized pattern")</pre>
```

rPopulationIndependenceM

Simulations of a point pattern according to the null hypothesis of population independence defined for M

# **Description**

Simulates of a point pattern according to the null hypothesis of population independence defined for M

#### Usage

rPopulationIndependenceM(X, ReferenceType, CheckArguments = TRUE)

## **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

ReferenceType One of the point types.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

#### **Details**

Reference points are kept unchanged, other points are redistributed randomly across locations.

## Value

A new weighted, marked, planar point pattern (an object of class wmppp, see wmppp.object).

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## References

Marcon, E. and Puech, F. (2010). Measures of the Geographic Concentration of Industries: Improving Distance-Based Methods. *Journal of Economic Geography* 10(5): 745-762.

Marcon, E., F. Puech and S. Traissac (2012). Characterizing the relative spatial structure of point patterns. *International Journal of Ecology* 2012(Article ID 619281): 11.

#### See Also

rPopulationIndependenceK, rRandomLabelingM

# **Examples**

```
# Simulate a point pattern with five types
X <- rpoispp(50)
PointType <- sample(c("A", "B", "C", "D", "E"), X$n, replace=TRUE)
PointWeight <- runif(X$n, min=1, max=10)
X$marks <- data.frame(PointType, PointWeight)
X <- as.wmppp(X)

# Randomize it
Y <- rPopulationIndependenceM(X, "A")
# Points of type "A" are unchanged,
# all other points have been redistributed randomly across locations
autoplot(Y, main="Randomized pattern")</pre>
```

rRandomLabeling

Simulations of a point pattern according to the null hypothesis of random labeling

# Description

Simulates of a point pattern according to the null hypothesis of random labeling.

#### Usage

```
rRandomLabeling(X, CheckArguments = TRUE)
```

## **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to save time in simulations for example, when the arguments have been checked elsewhere.

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## **Details**

Marks are redistributed randomly across the original point pattern.

#### Value

A new weighted, marked, planar point pattern (an object of class wmppp, see wmppp.object).

## References

Goreaud, F. et Pelissier, R. (2003). Avoiding misinterpretation of biotic interactions with the intertype K12 fonction: population independence vs random labelling hypotheses. *Journal of Vegetation Science* 14(5): 681-692.

#### See Also

rRandomLabelingM, rPopulationIndependenceK

## **Examples**

```
# Simulate a point pattern with five types
X <- rpoispp(50)
PointType <- sample(c("A", "B", "C", "D", "E"), X$n, replace=TRUE)
PointWeight <- runif(X$n, min=1, max=10)
X$marks <- data.frame(PointType, PointWeight)
X <- as.wmppp(X)
autoplot(X, main="Original pattern")
# Randomize it
Y <- rRandomLabeling(X)
# Types and weights have been redistributed randomly across locations
autoplot(Y, main="Randomized pattern")</pre>
```

rRandomLabelingM

Simulations of a point pattern according to the null hypothesis of random labelling defined for M

# **Description**

Simulates of a point pattern according to the null hypothesis of random labelling defined for M

#### Usage

```
rRandomLabelingM(X, CheckArguments = TRUE)
```

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## Arguments

X A weighted, marked, planar point pattern (wmppp.object) or a Dtable object.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to save time in simulations for example, when the arguments have been checked elsewhere.

#### **Details**

Point types are randomized. Locations and weights are kept unchanged. If both types and weights must be randomized together (Duranton and Overman, 2005; Marcon and Puech, 2010), use rRandomLocation.

#### Value

A new weighted, marked, planar point pattern (an object of class wmppp, see wmppp.object).

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Marcon, E. and Puech, F. (2010). Measures of the Geographic Concentration of Industries: Improving Distance-Based Methods. *Journal of Economic Geography* 10(5): 745-762.

Marcon, E., F. Puech and S. Traissac (2012). Characterizing the relative spatial structure of point patterns. *International Journal of Ecology* 2012(Article ID 619281): 11.

#### See Also

rRandomLabeling, rPopulationIndependenceM

## **Examples**

```
# Simulate a point pattern with five types
X <- rpoispp(50)
PointType <- sample(c("A", "B", "C", "D", "E"), X$n, replace=TRUE)
PointWeight <- runif(X$n, min=1, max=10)
X$marks <- data.frame(PointType, PointWeight)
X <- as.wmppp(X)
autoplot(X, main="Original pattern")
# Randomize it
Y <- rRandomLabelingM(X)
# Labels have been redistributed randomly across locations
# But weights are unchanged
autoplot(Y, main="Randomized pattern")</pre>
```

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rRandomLocation	Simulations of a point pattern according to the null hypothesis of random location

## **Description**

Simulates of a point pattern according to the null hypothesis of random location.

## Usage

```
rRandomLocation(X, ReferenceType = "", CheckArguments = TRUE)
```

# **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

ReferenceType One of the point types.

 ${\tt CheckArguments} \quad Logical; if {\tt TRUE}, the function arguments are verified. Should be set to {\tt FALSE} \ to {\tt CheckArguments} \ .$ 

save time in simulations for example, when the arguments have been checked

elsewhere.

# **Details**

Points are redistributed randomly across the locations of the original point pattern. This randomization is equivalent to random labeling, considering the label is both point type and point weight. If ReferenceType is specified, then only reference type points are kept in the original point pattern before randomization.

## Value

A new weighted, marked, planar point pattern (an object of class wmppp, see wmppp.object).

## References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106.

Marcon, E. and Puech, F. (2010). Measures of the Geographic Concentration of Industries: Improving Distance-Based Methods. *Journal of Economic Geography* 10(5): 745-762.

#### See Also

rRandomPositionK

rRandomPositionK 57

## **Examples**

```
# Simulate a point pattern with five types
X <- rpoispp(50)
PointType <- sample(c("A", "B", "C", "D", "E"), X$n, replace=TRUE)
PointWeight <- runif(X$n, min=1, max=10)
X$marks <- data.frame(PointType, PointWeight)
X <- as.wmppp(X)
autoplot(X, main="Original pattern")
# Randomize it
Y <- rRandomLocation(X)
# Points have been redistributed randomly across locations
autoplot(Y, main="Randomized pattern")</pre>
```

rRandomPositionK

Simulations of a point pattern according to the null hypothesis of random position defined for K

# **Description**

Simulations of a point pattern according to the null hypothesis of random position defined for K.

## Usage

```
rRandomPositionK(X, Precision = 0, CheckArguments = TRUE)
```

#### **Arguments**

X A weighted, marked, planar point pattern (wmppp.object).

Precision Accuracy of point coordinates, measured as a part of distance unit. See notes.

Default is 0 for no approximation.

CheckArguments Logical; if TRUE, the function arguments are verified. Should be set to FALSE to

save time in simulations for example, when the arguments have been checked

elsewhere.

#### **Details**

Points marks are kept unchanged and their position is drawn in a binomial process by runifpoint.

#### Value

A new weighted, marked, planar point pattern (an object of class wmppp, see wmppp.object).

#### Note

Simulations in a binomial process keeps the same number of points, so that marks can be redistributed. If a real CSR simulation is needed and marks are useless, use rpoispp.

Actual data coordinates are often rounded. Use the Precision argument to simulate point patterns with the same rounding procedure. For example, if point coordinates are in meters and rounded to the nearest half meter, use Precision = 0.5 so that the same approximation is applied to the simulated point patterns.

#### See Also

rRandomLocation

## **Examples**

```
# Simulate a point pattern with two types
X <- rpoispp(5)
PointType <- sample(c("A", "B"), X$n, replace=TRUE)
PointWeight <- runif(X$n, min=1, max=10)
X$marks <- data.frame(PointType, PointWeight)
X <- as.wmppp(X)
autoplot(X, main="Original pattern")
# Randomize it
Y <- rRandomPositionK(X)
# Points are randomly distributed
autoplot(Y, main="Randomized pattern")</pre>
```

```
spatstat generic functions
```

Methods for weighted, marked planar point patterns (of class wmppp) from spatstat

# Description

**spatstat** methods for a ppp.object applied to a wmppp.object.

## Usage

```
## $3 method for class 'wmppp'
sharpen(X, ...)
## $3 method for class 'wmppp'
split(...)
## $3 method for class 'wmppp'
superimpose(...)
## $3 method for class 'wmppp'
unique(x, ...)
## $3 method for class 'wmppp'
i[j, drop=FALSE, ..., clip=FALSE]
```

## **Arguments**

X, x	A two-dimensional point pattern. An object of class "wmppp".
	Arguments passed to the ppp.object method.
i	Subset index. Either a valid subset index in the usual R sense, indicating which points should be retained, or a window (an object of class "owin") delineating a subset of the original observation window, or a pixel image with logical values defining a subset of the original observation window.
j	Redundant. Included for backward compatibility.
drop	Logical value indicating whether to remove unused levels of the marks, if the marks are a factor.
clip	Logical value indicating how to form the window of the resulting point pattern, when i is a window. If clip=FALSE (the default), the result has window equal to i. If clip=TRUE, the resulting window is the intersection between the window of x and the window i.

#### **Details**

**spatstat** methods for ppp objects returning a ppp object can be applied to a wmppp and return a wpppp with these methods which just call the ppp.object method and change the class of the result for convenience.

Some **spatstat** functions such as **rthin** are not generic so they always return a ppp.object when applied to a wmppp.object. Their result may be converted by as.wmppp.

## Value

An object of class "wmppp".

# See Also

```
sharpen.ppp, split.ppp, superimpose.ppp, unique.ppp
```

summary.dbmssEnvelope Summary of a confidence envelope

# Description

Prints a useful summary of a confidence envelope of class "dbmssEnvelope"

# Usage

```
## S3 method for class 'dbmssEnvelope'
summary(object, ...)
```

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## Arguments

```
object An object of class "dbmssEnvelope". . . . Ignored.
```

#### **Details**

"dbmssEnvelope" objects are similar to envelope objects. Their summary is different to take into account the possibility of building global envelope following Duranton and Overman (2005): the global envelope is calculated by iteration: the simulations reaching one of the upper or lower values at any distance are eliminated at each step. The process is repeated until *Alpha / Number of simulations* simulations are dropped. The remaining upper and lower bounds at all distances constitute the global envelope. Interpolation is used if the exact ratio cannot be reached.

#### References

Duranton, G. and Overman, H. G. (2005). Testing for Localisation Using Micro-Geographic Data. *Review of Economic Studies* 72(4): 1077-1106

# **Examples**

```
data(paracou16)
autoplot(paracou16)

# Calculate intertype K envelope
Envelope <- KEnvelope(paracou16, NumberOfSimulations = 20, Global = TRUE,
   ReferenceType = "V. Americana", NeighborType = "Q. Rosea")
autoplot(Envelope)
summary(Envelope)</pre>
```

wmppp

Create a Weighted, Marked, Planar Point Pattern

# Description

Creates an object of class "wmppp" representing a two-dimensional point pattern with weights and labels.

## Usage

```
wmppp(df, window = NULL, unitname = NULL)
```

#### **Arguments**

df A dataframe with at least two columns containing point coordinates.

window An object of calls "owin" (owin.object).

unitname Name of unit of length. Either a single character string, or a vector of two

character strings giving the singular and plural forms, respectively. Ignored if

window is not NULL.

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## **Details**

Columns named "X", "Y", "PointType", "PointWeight" (capitalization is ignored) are searched to build the "wmppp" object and set the point coordinates, type and weight. If they are not found, columns are used in this order. If columns are missing, PointType is set to "All" and PointWeight to 1. If a "PointName" column is found, it is used to set the row names of the marks, else the original row names are used.

If the window is not specified, a rectangle containing all points is used, and unitname is used.

#### Value

An object of class "wmppp".

#### See Also

```
wmppp.object,
```

## **Examples**

```
# Draw the coordinates of 10 points
X <- runif(10)
Y <- runif(10)
# Draw the point types.
PointType <- sample(c("A", "B"), 10, replace=TRUE)
# Plot the point pattern. Weights are set to 1 ant the window is adjusted.
plot(wmppp(data.frame(X, Y, PointType)), , which.marks=2)</pre>
```

wmppp.object

Class of Weighted, Marked, Planar Point Patterns

# Description

A class "wmppp" to represent a two-dimensional point pattern of class ppp whose marks are a dataframe with two columns:

• PointType: labels, as factors

• PointWeight: weights.

## **Details**

This class represents a two-dimensional point pattern dataset. wmppp objects are also of class ppp.

Objects of class wmppp may be created by the function wmppp and converted from other types of data by the function as.wmppp.

#### See Also

```
ppp.object, wmppp, as.wmppp autoplot.wmppp
```

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# **Examples**

```
# Draw the coordinates of 10 points
X <- runif(10)
Y <- runif(10)
# Draw the point types and weights
PointType <- sample(c("A", "B"), 10, replace=TRUE)
PointWeight <- runif(10)
# Build the point pattern
X <- wmppp(data.frame(X, Y, PointType, PointWeight), owin())
# Plot the point pattern. which.marks=1 for point weights, 2 for point types
par(mfrow=c(1,2))
plot(X, which.marks=1, main="Point weights")
plot(X, which.marks=2, main="Point types")
# Or use autoplot for a ggplot
autoplot(X)</pre>
```

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```