Package ‘apcf’

July 5, 2020

Title  Adapted Pair Correlation Function

Version  0.1.5

Description  The adapted pair correlation function transfers the concept of the pair correlation function from point patterns to patterns of objects of finite size and irregular shape (e.g. lakes within a country). This is a reimplementation of the method suggested by Nuske et al. (2009) <doi:10.1016/j.foreco.2009.09.050> using the libraries ‘GEOS’ and ‘GDAL’ directly instead of through ‘PostGIS’.

License  GPL (>= 3)

URL  https://github.com/rnuske/apcf

BugReports  https://github.com/rnuske/apcf/issues

Encoding  UTF-8

LazyData  true

Depends  R (>= 3.3.0)

Imports  Rcpp (>= 0.12), graphics

Suggests  knitr, rmarkdown

LinkingTo  Rcpp

SystemRequirements  C++11, GDAL (>= 2.0.1), GEOS (>= 3.4.0)

VignetteBuilder  knitr

RoxygenNote  7.1.0

NeedsCompilation  yes

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Description

A faster implementation of the adapted Pair Correlation Function presented in Nuske et al. (2009) in C++ using the libraries GEOS and GDAL directly instead of through PostGIS.

Details

The Adapted Pair Correlation Function transfers the concept of the Pair Correlation Function from point patterns to patterns of patches of finite size and irregular shape (e.g., lakes within a country). The main tasks are (i) the construction of null models by randomizing the patches of the original pattern within the study area, (ii) the edge correction by determining the proportion of a buffer within the study area, and (iii) the calculation of the shortest distances between the patches.

This package mainly provides three functions:

• create null models and calculate distances and ratios,
• turn distances and ratios into an edge corrected PCF, and
• plot Pair Correlation Functions.

Pattern to Distances & Ratios

The task consists of two parts: creating null models / permutations and calculating distances between all objects of a pattern and determining the fraction of the perimeter a buffer inside the study area. Permutations of the original pattern are achieved by randomly rotating and randomly placing all objects within the study area without overlap.

The resulting collection of distances and ratios of each null model and the original pattern are returned as an object of class `dists` (a data.frame with some additional attributes).

The library GEOS (>= 3.4.0) is used for the geometrical analysis of the pattern. Geodata are converted to GEOS Geometries with the help of GDAL/OGR (>= 2.0.1). The GEOS functions are called from C++ Functions which are integrated into R via Rcpp and wrapped in the R function `pat2dists()`.
Create an edge corrected PCF

The `dists` objects are turned into `fv_pcf` objects by the function `dists2pcf()`. A C++ function finds all distances and ratios belonging to a null model or the original pattern (marked with index 0) and calculates a density function using the Epanechnikov kernel and Ripley’s edge correction. Resulting in as many PCFs as null models were created plus a PCF for the original pattern. From the PCF of the null models a pointwise critical envelope is derived. The arithmetic mean of all PCF of the null models is employed for a bias correction of the empirical PCF and the upper and lower bound of the envelope.

Plot a PCF

`plot.fv_pcf()` is an S3 method of the plot function for the class `fv_pcf`. It provides a nice plot of the empirical PCF together with the pointwise critical envelope.

References


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**dists**

*Class dists: Collection of Distances and Ratios*

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**Description**

Advanced Use Only. This low-level function creates an object of class "dists" from raw numerical data.

**Usage**

```r
dists(df, area, n_obj, max_dist)
is.dists(obj)
```

**Arguments**

- `df` A data frame with the columns `sim`, `dist`, and `ratio` containing an indicator of the model run (0:n_sim), distances between the objects of the patterns, and the ratios of a buffer with distance `dist` inside the study area (needed for Ripley’s edge correction).
- `area` size of the study area in square units
- `n_obj` number of objects in the pattern
- `max_dist` maximum distance to be measured in pattern
- `obj` an R object, preferably of class `dists`

**Value**

An object of class `dists`. 
**dists2pcf**

Convert Distances & Ratios to PCF

**Description**

Estimates the adapted Pair Correlation Function (PCF) of a pattern together with a pointwise critical envelope based on distances and ratios calculated by `pat2dists()`.

**Usage**

```
dists2pcf(dists, r, r_max = NULL, kernel = "epanechnikov", stoyan, n_rank)
```

**Arguments**

- **dists**: An object of class `dists`. Usually created by `pat2dists()`.
- **r**: A step size or a vector of values for the argument `r` at which `g(r)` should be evaluated.
- **r_max**: Maximum value for the argument `r`.
- **kernel**: String. Choice of smoothing kernel (only the "epanechnikov" kernel is currently implemented).
- **stoyan**: Bandwidth coefficient (smoothing the Epanechnikov kernel). Penttinen et al. (1992) and Stoyan and Stoyan (1994) suggest values between 0.1 and 0.2.
- **n_rank**: Rank of the value amongst the `n_sim` simulated values used to construct the envelope. A rank of 1 means that the minimum and maximum simulated values will be used. Must be >= 1 and < `n_sim`/2. Determines together with `n_sim` in `pat2dists()` the alpha level of the envelope. If `alpha` and `n_sim` are fix, `n_rank` can be calculated by `(n_sim+1)*alpha/2` eg. `(199+1)*0.05/2 = 5`

**Details**

Since the pair-correlation function is a density function, we employ the frequently used Epanechnikov kernel (Silverman, 1986; Stoyan and Stoyan, 1994; Nuske et al., 2009). The Epanechnikov kernel is a weight function putting maximal weight to pairs with distance exactly equal to `r` but also incorporating pairs only roughly at distance `r` with reduced weight. This weight falls to zero if the actual distance between the points differs from `r` by at least `δ`, the so-called bandwidth parameter, which determines the degree of smoothness of the function. Penttinen et al. (1992) and Stoyan and Stoyan (1994) suggest to set `c` aka stoyan-parameter of `c/\sqrt{\lambda}` between 0.1 and 0.2 with `\lambda` being the intensity of the pattern.

The edge correction is based on suggestions by Ripley (1981). For each pair of objects `i` and `j`, a buffer with buffer distance `r_{ij}` is constructed around the object `i`. The object `j` is then weighted by the inverse of the ratios `p_{ij}` of the buffer perimeter being within the study area. That way we account for the reduced probability of finding objects close to the edge of the study area.

The alpha level of the pointwise critical envelope is `\alpha = \frac{n_{\text{rank}}^2}{2n_{\text{sim}}}` according to (Besag and Diggle, 1977; Buckland, 1984; Stoyan and Stoyan, 1994).
An object of class `fv_pcf` containing the function values of the PCF and the envelope.

**References**


**See Also**

`pat2dists()`, `plot.fv_pcf()`

**Examples**

```r
# it's advised against setting n_sim < 199
ds <- pat2dists(area=system.file("shapes/sim_area.shp", package="apcf"),
               pattern=system.file("shapes/sim_pat_reg.shp", package="apcf"),
               max_dist=25, n_sim=3)

# derive PCF and envelope
pcf <- dists2pcf(ds, r=0.2, r_max=25, stoyan=0.15, n_rank=1)
```

---

**Description**

Advanced Use Only. This low-level function creates an object of class "fv_pcf" from raw numerical data.
Usage

fv_pcf(df, n_sim, n_rank, correc, kernel, stoyan, bw)

is.fv_pcf(obj)

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

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

Arguments

df         A data frame with at least 2 columns named 'r' and 'g' containing the values of the function argument (r) and the corresponding values (g). Usually the upper 'upr' and lower 'lwr' bounds of a pointwise critical envelope are passed along as well.
n_sim      Integer. Number of generated simulated patterns used for computing the envelope
n_rank     Integer. Rank of the envelope value amongst the n_sim simulated values. A rank of 1 means that the minimum and maximum simulated values will be used.
correc     String. Choice of edge correction (eg. "Ripley").
kernel     String. Choice of smoothing kernel (eg. "epanechnikov").
stoyan     Bandwidth coefficient used in smoothing kernel.
bw         Bandwidth used in smoothing kernel.
x, obj, object  an R object, preferably of class fv_pcf
...         additional parameter

Value

An object of class fv_pcf.

pat2dists

Convert a Pattern to Distances & Ratios

Description

Creates n_sim null models by permutation of the original pattern and calculates distances between all object of a pattern closer than max_dist and determines the fractions of the perimeter of buffers with distance dist inside the study area (needed for edge correction).
Usage

pat2dists(
  area,
  pattern,
  max_dist,
  n_sim = 199,
  max_tries = 1e+05,
  save_patterns = NULL,
  verbose = FALSE
)

Arguments

  area, pattern  Data source name of study area and pattern (interpretation varies by driver - for some drivers, dsn is a file name, but may also be a folder, or contain the name and access credentials of a database)

  max_dist  Maximum distance measured in the pattern. Usually smaller than half the diameter of the study area.

  n_sim  Number of simulated patterns (randomizations) to be generated for computing the envelope and correcting the biased empirical pcf. Determines together with n_rank in dists2pcf() the alpha level of the envelope. If alpha and n_rank are fix, n_sim can be calculated by (n_rank*2/alpha)-1 eg. (5*2/0.05)-1 = 199.

  max_tries  How often shall a relocation of an object be tried while randomizing the pattern.

  save_patterns  Shall the simulated patterns be saved as Shapefiles for debugging/later inspections. Might be a large number of files (4 * n_sim). Can be NULL (no export) or a character string providing a basename optionally including a valid/existing path.

  verbose  Provide progress information in the console. Roman numerals (x: 10, C: 100, D: 500, M: 1000) indicate the progress of the simulation and 'e' the empirical PCF.

Details

Null models are created by randomly rotating and randomly placing all objects within the study area without overlap. They are used for correcting the biased pcf and constructing a pointwise critical envelope (cf. Nuske et al. 2009).

Measuring distances between objects and permutation of the pattern is done with GEOS and spatial data are converted to GEOS geometries by GDAL/OGR.

Value

An object of class dists.
plot.fv_pcf

References

j.foreco.2009.09.050

See Also
dists2pcf(), plot.fv_pcf()

Examples

# it's advised against setting n_sim < 199
ds <- pat2dists(area=system.file("shapes/sim_area.shp", package="apcf"),
          pattern=system.file("shapes/sim_pat_reg.shp", package="apcf"),
          max_dist=25, n_sim=3, verbose=TRUE)

plot.fv_pcf

Plot PCF

Description

Plot method for the class "fv_pcf". Draws a pair correlation function and a pointwise critical envelope if available.

Usage

## S3 method for class 'fv_pcf'
plot(
x,
xlim = NULL,
ylim = NULL,
xticks = NULL,
yticks = NULL,
xlab = "r",
ylab = "g(r)",
main = NULL,
sub = NULL,
xaxis = TRUE,
yaxis = TRUE,
ann = graphics::par("ann"),
bty = "l",
...
)
**Arguments**

- **x**
  - an object of class `fv_pcf`.
- **xlim, ylim**
  - the x and y limits of the plot. NULL indicates that the range of the finite values to be plotted should be used.
- **xticks, yticks**
  - points at which tick-marks are to be drawn. By default (when NULL) tickmark locations are computed.
- **xlab, ylab**
  - a label for the x and y axis, respectively.
- **main, sub**
  - a main and sub title for the plot, see also `title()`.
- **xaxis, yaxis**
  - a logical value or a 1-character string giving the desired type of axis. The following values are possible: "n" or FALSE for no axis, "t" for ticks only, "f" or TRUE for full axis, "o" for full axis in the outer margin.
- **ann**
  - a logical value indicating whether the default annotation (title and x and y axis labels) should appear on the plot.
- **bty**
  - a character string which determines the type of box which is drawn about the plotting region, see `par()`.
- **...**
  - additional parameter, currently without effect

**Value**

An object of class `fv_pcf` invisibly.

**See Also**

- `pat2dists()`, `dists2pcf()`

**Examples**

```r
# it's advised against setting n_sim < 199
ds <- pat2dists(area=system.file("shapes/sim_area.shp", package="apcf"),
              pattern=system.file("shapes/sim_pat_reg.shp", package="apcf"),
              max_dist=25, n_sim=3)

# derive PCF and envelope
pcf <- dists2pcf(ds, r=0.2, r_max=25, stoyan=0.15, n_rank=1)

# a simple plot
plot(x=pcf, xlim=c(0, 20), ylim=c(0, 2.2))

# a panel of four plots
op <- par(mfrow=c(2,2), oma=c(3,3,0,0), mar=c(0,0,2,2),
mgp=c(2,0.5,0), tcl=-0.3)
plot(pcf, xaxis='t', yaxis='o', ann=FALSE)
plot(pcf, xaxis='t', yaxis='t', ann=FALSE)
plot(pcf, xaxis='o', yaxis='o', ann=FALSE)
plot(pcf, xaxis='o', yaxis='t')
par(op)
```
Simulated Patterns (sample data)

Description

These simulated patterns were created for testing the Adapted Pair Correlation Function (presented in Nuske et al., 2009).

Format

A set of shapefiles containing the study area and three simulated patterns.

<table>
<thead>
<tr>
<th>Shapefiles</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>sim_area.*</td>
<td>study area</td>
</tr>
<tr>
<td>sim_pat_reg.*</td>
<td>simulated regular pattern</td>
</tr>
<tr>
<td>sim_pat_rand.*</td>
<td>simulated random pattern</td>
</tr>
<tr>
<td>sim_pat_clust.*</td>
<td>simulated clustered pattern</td>
</tr>
</tbody>
</table>

The study area is a square of 100 m x 100 m. A set of n = 100 objects were created and latter placed according to the designated spatial distribution. The size distribution and shapes of the objects are inspired by measurements of canopy gaps. The areas of the objects range from 1.6 m² to 57.7 m² with an arithmetic mean of 9.7 m² and a median of 5.5 m². The total area of all objects is 969.7 m², meaning 9.7\.

For the sim_pat_reg.* dataset, the objects were arranged in a strict regular manner. A centric systematic grid was constructed, and the objects of the set were then randomly rotated and randomly placed by locating the centroids of the objects exactly on the matching randomly numbered grid points, resulting in a regular arrangement of objects with a constant distance of the centroids of 10 m.

For the sim_pat_rand.* dataset with randomly distributed objects, we generated a realisation of the Binomial process with intensity 0.01 m⁻², meaning one point per 100 m². The objects were again randomly rotated and numbered and objects put on matching points with their centroid as close to the point as possible without overlapping other objects.

The sim_pat_clust.* dataset represents a clustered configuration. Again, we first created a point pattern with 100 points and then put the randomly numbered objects on the points. The point pattern was a realisation of Matern’s cluster process with \( w = 0.0006 \text{ m}^{-2} \) or 6 cluster centres per ha, a dispersion radius of \( R = 10 \text{ m} \) and on average \( y = 16.6 \text{ points per cluster} \).

We used the R-package spatstat (Baddeley et al. 2015) for simulating the Binomial process and Matern’s cluster process.

Source

Nuske et al. (2009)
References


Examples

ds <- pat2dists(area=system.file("shapes/sim_area.shp", package="apcf"),
                   pattern=system.file("shapes/sim_pat_reg.shp", package="apcf"),
                   max_dist=25, n_sim=3, verbose=TRUE)
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