

# Package ‘apcf’

July 22, 2025

**Title** Adapted Pair Correlation Function

**Version** 0.3.2

**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). The pair correlation function describes the spatial distribution of objects, e.g. random, aggregated or regularly spaced. This is a reimplementaion of the method suggested by Nuske et al. (2009) <[doi:10.1016/j.foreco.2009.09.050](https://doi.org/10.1016/j.foreco.2009.09.050)> using the library 'GEOS'.

**License** GPL (>= 3)

**URL** <https://rnuske.github.io/apcf/>, <https://github.com/rnuske/apcf>

**BugReports** <https://github.com/rnuske/apcf/issues>

**Encoding** UTF-8

**Depends** R (>= 3.3.0)

**Imports** Rcpp (>= 0.12), wk (>= 0.6.0)

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**LinkingTo** Rcpp

**SystemRequirements** GEOS (>= 3.4.0)

**VignetteBuilder** knitr

**LazyData** true

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| apcf-package | <i>Adapted Pair Correlation Function</i> |
|--------------|--|

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### Description

A faster implementation of the Adapted Pair Correlation Function presented in Nuske et al. (2009) in C++ using the library GEOS 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 objects of finite size and irregular shape (eg. lakes within a country). The main tasks are (i) the construction of null models by randomizing the objects 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 objects.

This package mainly provides three functions:

- `pat2dists()` creates null models and calculates distances and ratios,
- `dists2pcf()` turns distances and ratios into an edge corrected PCF, and
- `plot()` plots 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. 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.

### Author(s)

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### References

Nuske, R.S., Sprauer, S. and Saborowski, J. (2009): Adapting the pair-correlation function for analysing the spatial distribution of canopy gaps. *Forest Ecology and Management*, 259(1): 107–116. <https://doi.org/10.1016/j.foreco.2009.09.050>

### See Also

Useful links:

- <https://rnuske.github.io/apcf/>
- <https://github.com/rnuske/apcf>
- Report bugs at <https://github.com/rnuske/apcf/issues>

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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

```
dists(df, area, n_obj, max_dist)
```

```
is.dists(obj)
```

**Arguments**

|          |  |
|----------|--|
| df       | A data frame with the columns <code>sim</code> , <code>dist</code> , and <code>ratio</code> 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 <code>dist</code> 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 <code>dists</code>  |

**Value**

An object of class `dists`.

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dists2pcf

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*Convert Distances & Ratios to PCF*


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**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 <code>dists</code> . Usually created by <code>pat2dists()</code>   |
| r      | A step size or a vector of values for the argument <code>r</code> at which <code>g(r)</code> should be evaluated.   |
| r_max  | maximum value for the argument <code>r</code> .   |
| 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 <code>n_sim</code> simulated values used to construct the envelope. A rank of 1 means that the minimum and maximum simulated values will be used. Must be $\geq 1$ and $< n\_sim/2$ . Determines together with <code>n_sim</code> in <code>pat2dists()</code> the alpha level of the envelope. If <code>alpha</code> and <code>n_sim</code> are fix, <code>n_rank</code> 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  $\delta$ , 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  $\alpha = \frac{n\_rank*2}{n\_sim+1}$  according to (Besag and Diggle 1977, Buckland 1984, Stoyan and Stoyan 1994).

## Value

An object of class `fv_pcf` containing the function values of the PCF and the envelope.

## References

- Besag, J. and Diggle, P.J. (1977): Simple Monte Carlo tests for spatial pattern. *Journal of the Royal Statistical Society. Series C (Applied Statistics)*, 26(3): 327–333. <https://doi.org/10.2307/2346974>
- Buckland, S.T. (1984). Monte Carlo Confidence Intervals. *Biometrics*, 40(3): 811-817. <https://doi.org/10.2307/2530926>
- Nuske, R.S., Sprauer, S. and Saborowski, J. (2009): Adapting the pair-correlation function for analysing the spatial distribution of canopy gaps. *Forest Ecology and Management*, 259(1): 107–116. <https://doi.org/10.1016/j.foreco.2009.09.050>
- Penttinen A., Stoyan D., Henttonen H. M. (1992): Marked point processes in forest statistics. *Forest Science*, 38(4): 806–824. <https://doi.org/10.1093/forestscience/38.4.806>
- Ripley, B.D. (1981): *Spatial Statistics*. John Wiley & Sons, New York. <https://doi.org/10.1002/0471725218>
- Silverman, B.W. (1986): *Density Estimation for Statistics and Data Analysis*. Chapman and Hall, London.
- Stoyan, D. and Stoyan, H. (1994) *Fractals, random shapes and point fields: Methods of geometrical statistics*. John Wiley & Sons, Chichester.

## See Also

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

## Examples

```
# it's advised against setting n_sim < 199
ds <- pat2dists(area=sim_area, pattern=sim_pat_reg, 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)
```

fv\_pcf

*Class fv\_pcf: Function Value Table for PCFs***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.

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pat2dists                      *Convert a Pattern to Distances & Ratios*

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### 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_pattern = FALSE,
  verbose = FALSE
)
```

### Arguments

|                            |  |
|----------------------------|--|
| <code>area, pattern</code> | Geodata (polygons) of study area and pattern in the formats WKB (well known binary, list of raw vectors), WKT (well known text) or sf-objects if package sf is available. Via sf all file formats supported by GDAL/OGR are possible.  |
| <code>max_dist</code>      | Maximum distance measured in the pattern. Usually smaller than half the diameter of the study area.  |
| <code>n_sim</code>         | Number of simulated patterns (randomizations) to be generated for computing the envelope and correcting the biased empirical pcf. Determines together with <code>n_rank</code> in <code>dists2pcf()</code> the alpha level of the envelope. If alpha and <code>n_rank</code> are fix, <code>n_sim</code> can be calculated by $(n\_rank * 2 / \alpha) - 1$ for instance $(5 * 2 / 0.05) - 1 = 199$ . |
| <code>max_tries</code>     | How often shall a relocation of an object be tried while randomizing the pattern.  |
| <code>save_pattern</code>  | Shall one simulated pattern be returned in the attributes for debugging/later inspections. The pattern is provided as WKB (list of raw vectors) in the attribute <code>randPattern</code> .  |
| <code>verbose</code>       | Provide progress information in the console. Roman numerals (x: 10, C: 100, D: 500, M: 1000) indicate the progress of the simulation and 'e' denotes 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 using [GEOS](#).

**Value**

An object of class `dists`. If `save_pattern = TRUE` an additional attribute `randPattern` is returned containing a WKB (list of raw vectors).

**References**

Nuske, R.S., Sprauer, S. and Saborowski, J. (2009): Adapting the pair-correlation function for analysing the spatial distribution of canopy gaps. *Forest Ecology and Management*, 259(1): 107–116. <https://doi.org/10.1016/j.foreco.2009.09.050>

**See Also**

`dists2pcf()`, `plot.fv_pcf()`

**Examples**

```
# it's advised against setting n_sim < 199
ds <- pat2dists(area=sim_area, pattern=sim_pat_reg, max_dist=25, n_sim=3)

# verbose and returns one randomized pattern for debugging
ds_plus <- pat2dists(area=sim_area, pattern=sim_pat_reg, max_dist=5, n_sim=3,
                    verbose=TRUE, save_pattern=TRUE)

## Not run:
# wk's plot function needs additional package 'vctrs'
plot(attr(ds_plus, "randPattern"))

## End(Not run)
```

---

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 <code>fv_pcf</code> .  |
| 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 <code>title()</code> .  |
| 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 <code>par()</code> .  |
| ...            | additional parameter, currently without effect  |

### Value

An object of class `fv_pcf` invisibly.

### See Also

`pat2dists()`, `dists2pcf()`

### Examples

```

# it's advised against setting n_sim < 199
ds <- pat2dists(area=sim_area, pattern=sim_pat_reg, 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)

```

---

sim\_patterns                      *Simulated Patterns (sample data)*

---

### Description

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

### Usage

sim\_area

sim\_pat\_clust

sim\_pat\_rand

sim\_pat\_reg

### Format

A set of **WKBs** of class `wk_wkb` containing the study area and three simulated patterns.

| <b>Dataset name</b> | <b>Description</b>          |
|---------------------|-----------------------------|
| sim_area            | study area                  |
| sim_pat_reg         | simulated regular pattern   |
| sim_pat_rand        | simulated random pattern    |
| sim_pat_clust       | simulated clustered pattern |

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 m<sup>2</sup> to 57.7 m<sup>2</sup> with an arithmetic mean of 9.7 m<sup>2</sup> and a median of 5.5 m<sup>2</sup>. The total area of all objects is 969.7 m<sup>2</sup>, meaning 9.7% of the study area is covered by objects.

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  $0.01 \text{ m}^{-2}$ , meaning one point per  $100 \text{ m}^2$ . 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  $\bar{y} = 16.6$  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

Baddeley A., Rubak E. and Turner, R. (2015): *Spatial Point Patterns: Methodology and Applications with R*. Chapman and Hall/CRC, London. <https://doi.org/10.1201/b19708>

Nuske, R.S., Sprauer, S. and Saborowski, J. (2009): Adapting the pair-correlation function for analysing the spatial distribution of canopy gaps. *Forest Ecology and Management*, 259(1): 107–116. <https://doi.org/10.1016/j.foreco.2009.09.050>

### Examples

```
ds <- pat2dists(area=sim_area, pattern=sim_pat_reg, max_dist=25, n_sim=3)
```

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