

# Package ‘isopam’

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**Type** Package

**Title** Isopam (Clustering)

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**Depends** vegan, cluster

**Suggests** proxy

**Description** Isopam clustering algorithm and utilities.

Isopam optimizes clusters and optionally cluster numbers in a brute force style and aims at an optimum separation by all or some descriptors (typically species).

**License** GPL-2

**LazyLoad** yes

**NeedsCompilation** no

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 andechs

*Fen Meadows*


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### Description

This data set gives the average cover of vascular plant species in subplots nested within 17 whole-plots from mown fen meadows. This is a subset of the data used in Schmidtlein & Sassin (2004).

### Usage

```
data(andechs)
```

### Format

A matrix containing 17 observations and 110 species.

### Source

Schmidtlein, S., Sassin, J. (2004): Mapping of continuous floristic gradients in grasslands using hyperspectral imagery. *Remote Sensing of Environment* **92**, 126–138.

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 isopam

*Isopam (Clustering)*


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### Description

Isopam classification is performed either as a hierarchical, divisive method, or as non-hierarchical partitioning. Optimizes clusters and optionally cluster numbers for maximum performance of group indicators. Developed for matrices representing species abundances in plots.

### Usage

```
isopam (dat, c.fix = FALSE, c.opt = TRUE, c.max = 6,
        l.max = FALSE, stopat = c(1,7), sieve = TRUE,
        Gs = 3.5, ind = NULL, centers = NULL, distance = 'bray',
        k.max = 100, d.max = 7, ..., juice = FALSE)

## S3 method for class 'isopam'
identify(x, ...)
## S3 method for class 'isopam'
plot(x, ...)
```

**Arguments**

<code>dat</code>	data matrix: each row corresponds to an object (typically a plot), each column corresponds to a descriptor (typically a species). All variables must be numeric. Missing values (NAs) are not allowed. At least 3 rows (plots) are required.
<code>c.fix</code>	number of clusters (defaults to FALSE). If a number is given, non-hierarchical partitioning is performed, <code>c.opt</code> and <code>c.max</code> are ignored and <code>l.max</code> is set to one.
<code>c.opt</code>	if TRUE (the default) cluster numbers are optimized in the range between 2 and <code>c.max</code> (slow and thorough). If FALSE groups are divided into two subgroups (quick and dirty).
<code>c.max</code>	maximum number of clusters per partition. Applies to all partitioning steps if <code>c.opt = TRUE</code> .
<code>l.max</code>	maximum number of hierarchy levels. Defaults to FALSE (no maximum number). Note that divisions may stop well before this number is reached (see <code>stopat</code> ). Use <code>l.max = 1</code> for non-hierarchical partitioning (or use <code>c.fix</code> ).
<code>stopat</code>	vector with stopping rules for hierarchical clustering. Two values define if a partition should be retained in hierarchical clustering: the first determines how many indicators must be present per cluster, the second defines the standardized G-value that must be reached by these indicators. <code>stopat</code> is not effective at the first hierarchy level or in non-hierarchical partitioning.
<code>sieve</code>	logical. If TRUE (the default), only descriptors (species) exceeding a threshold defined by <code>Gs</code> are used in the search for a good clustering solution. Their number is multiplied with their mean standardized G-value. The product is used as optimality criterion. If FALSE all descriptors are used for optimization.
<code>Gs</code>	threshold (standardized G value) for descriptors (species) to be considered in the search for a good clustering solution. Effective with <code>sieve = TRUE</code> .
<code>ind</code>	optional vector of column names from <code>dat</code> defining descriptors (species) used as indicators. This turns Isopam in an expert system. Replaces the automated selection of indicators with <code>sieve = TRUE</code> ( <code>ind</code> overrules <code>sieve</code> ).
<code>centers</code>	optional vector with observations used as cluster cores (supervised classification).
<code>distance</code>	distance measure for the distance matrix used as a starting point for Isomap. Any distance measure implemented in packages <b>vegan</b> or <b>proxy</b> can be used (see details).
<code>k.max</code>	maximum Isomap $k$ .
<code>d.max</code>	maximum number of Isomap dimensions.
<code>...</code>	other arguments to S3 functions <code>plot</code> and <code>identify</code> corresponding to <a href="#">hclust</a> .
<code>juice</code>	logical. If TRUE input files for Juice are generated.
<code>x</code>	an <code>isopam</code> result object.

**Details**

Isopam is described in Schmidtlein et al. (2010). It consists of dimensionality reduction (Isomap: Tenenbaum et al. 2000; [isomap](#) in **vegan**) and partitioning of the resulting ordination space (PAM:

Kaufman & Rousseeuw 1990; `pam` in `cluster`). The classification is performed either as a hierarchical, divisive method, or as non-hierarchical partitioning. Compared to other clustering methods, it has the following features: (a) it optimizes partitions for the performance of group indicators (typically species) or for maximum average 'fidelity' of descriptors to groups; (b) it optionally selects the number of clusters per division; (c) the shapes of groups in feature space are not limited to spherical or other regular geometric shapes (thanks to the underlying Isomap algorithm) and (d) the distance measure used for the initial distance matrix can be freely defined.

Currently, the `plot` and `identify` methods for class `isopam` simply link to the `hclust` object `$dendro` resulting from `isopam` in case of hierarchical partitioning. The methods work just like `plot.hclust` and `identify.hclust`.

The preset distance measure is Bray-Curtis (Odum 1950). Distance measures are passed to `vegdist` in `vegan`. If `vegan` does not know the given measure it is passed to `dist` in `proxy`. Measures available in `vegan` are listed in `vegdist`. Measures registered in `proxy` can be listed with `summary(pr_DB)` once `proxy` is loaded. New measures can be defined and registered as described in `?pr_DB`. `isopam` can't deal with distance matrices as a replacement for the original data matrix because it operates on individual descriptors (species).

### Value

<code>call</code>	generating call
<code>distance</code>	distance measure used by Isomap
<code>flat</code>	observations (plots) with group affiliation. Running group numbers for each level of the hierarchy.
<code>hier</code>	observations (plots) with group affiliation. Group identifiers reflect the cluster hierarchy. Not present with only one level of partitioning.
<code>medoids</code>	observations (plots) representing the medoids of the resulting groups.
<code>analytics</code>	table summarizing parameter settings for the final partitioning steps. Name: name of the respective parent cluster (0 in case of the first partition); Subgroups: number of subgroups; <code>Isomap.dim</code> : Isomap dimensions used; <code>Isomap.k.min</code> : minimum possible Isomap $k$ ; <code>Isomap.k</code> : Isomap $k$ used; <code>Isomap.k.max</code> : maximum possible Isomap $k$ ; <code>Ind.N</code> : number of indicators reaching or exceeding $G_s$ ; <code>Ind.Gs</code> : the average standardized $G$ value of these indicators; and <code>Global.Gs</code> : the average standardized $G$ value of all descriptors.
<code>dendro</code>	an object of class <code>hclust</code> representing the clustering. Not present with only one level of partitioning.
<code>dat</code>	data used

### Note

For large datasets, `isopam` may need too much memory or too much computation time. The optimization procedure (selection of Isomap dimensions and  $-k$ , optionally selection of cluster numbers) is based on a brute force approach that takes its time with large data sets. Low speed is inherent to the method, so don't complain. If used with data not representing species in plots make sure that the indicator approach is appropriate.

With very small datasets, the indicator based optimization may fail. In such cases consider using `filtered = FALSE` instead of the default method.

**Author(s)**

Sebastian Schmidlein with contributions from Jason Collison and Lubomir Tichý

**References**

Odum, E.P. (1950): Bird populations in the Highlands (North Carolina) plateau in relation to plant succession and avian invasion. *Ecology* **31**: 587–605.

Kaufman, L., Rousseeuw, P.J. (1990): *Finding groups in data*. Wiley.

Schmidlein, S., Tichý, L., Feilhauer, H., Faude, U. (2010): A brute force approach to vegetation classification. *Journal of Vegetation Science* **21**: 1162–1171.

Tenenbaum, J.B., de Silva, V., Langford, J.C. (2000): A global geometric framework for nonlinear dimensionality reduction. *Science* **290**, 2319–2323.

**See Also**

[isotab](#) for a table of descriptor (species) frequency in clusters.

**Examples**

```
## load data to the current environment
data(andechs)

## call isopam with the standard options
ip<-isopam(andechs)

## examine cluster hierarchy
plot(ip)

## examine grouping
ip$flat

## examine frequency table (second hierarchy level)
isotab(ip, 2)

## non-hierarchical partitioning
ip<-isopam(andechs,c.fix=3)
ip$flat
```

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isotab

*Ordered frequency table for Isopam clusters*

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**Description**

Computes an ordered frequency table based on Isopam clustering results. The upper part of the table lists typical descriptors (usually species) with a significant binding to single clusters (according to customisable thresholds). The lower part of the table is ordered by descending overall frequency.

**Usage**

```
isotab(ip, level = 1, phi.min = 'auto', p.max = .05)
```

**Arguments**

ip	object of class isopam.
level	level in cluster hierarchy starting with 1 = first division.
phi.min	threshold of <i>phi</i> determining which descriptors (species) are listed in the upper part of the table. Applies only to descriptors passing the criterion defined by p.max. If phi.min = 'auto' (the default) isotab suggests a suitable value based on the numbers of clusters, observations, and descriptors.
p.max	threshold of Fisher's <i>p</i> determining which descriptors (species) are listed in the upper part of the table. Applies only to descriptors passing the criterion defined by phi.min.

**Details**

phi.min is based on the standardized *phi* value according to Chitry et al. 2002.

**Value**

tab	dataframe with ordered frequencies and their significance. The latter is derived from Fisher's exact test ( $p \leq 0.05$ : *, $p \leq 0.01$ : **, $p \leq 0.001$ : ***).
n	matrix with cluster sizes.
thresholds	phi.min and p.max used.
typical	dataframe with items (often species) typically found in clusters (according to thresholds).

**Author(s)**

Sebastian Schmidlein

**References**

Chitry, M., Tichy, L., Holt, J., Botta-Dukat, Z. (2002): Determination of diagnostic species with statistical fidelity measures. *Journal of Vegetation Science* **13**, 79–90.

Schmidlein, S., Tichy, L., Feilhauer, H., Faude, U. (2010): A brute force approach to vegetation classification. *Journal of Vegetation Science* (in press).

**See Also**

[isopam](#)

**Examples**

```
## load data to the current environment
data(andechs)

## call isopam with the standard options
ip<-isopam(andechs)

## build table (uppermost hierarchy level)
isotab(ip)

## build table (lower hierarchy level)
isotab(ip,2)
```

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