

Package ‘densityClust’

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Type Package
Title Clustering by Fast Search and Find of Density Peaks
Version 0.3.3
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Description An improved implementation (based on k-nearest neighbors) of the density peak clustering algorithm, originally described by Alex Rodriguez and Alessandro Laio (Science, 2014 vol. 344). It can handle large datasets (> 100,000 samples) very efficiently. It was initially implemented by Thomas Lin Pedersen, with inputs from Sean Hughes and later improved by Xiaojie Qiu to handle large datasets with kNNs.
License GPL (>= 2)
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BugReports <https://github.com/thomasp85/densityClust/issues>
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clustered	<i>Check whether a densityCluster object have been clustered</i>
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Description

This function checks whether `findClusters()` has been performed on the given object and returns a boolean depending on the outcome

Usage

```
clustered(x)

## S3 method for class 'densityCluster'
clustered(x)
```

Arguments

x A densityCluster object

Value

TRUE if `findClusters()` have been performed, otherwise FALSE

clusters	<i>Extract cluster membership from a densityCluster object</i>
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Description

This function allows the user to extract the cluster membership of all the observations in the given densityCluster object. The output can be formatted in two ways as described below. Halo observations can be chosen to be removed from the output.

Usage

```
clusters(x, ...)
```

```
## S3 method for class 'densityCluster'
clusters(x, as.list = FALSE, halo.rm = TRUE, ...)
```

Arguments

x	The densityCluster object. <code>findClusters()</code> must have been performed prior to this call to avoid throwing an error.
...	Currently ignored
as.list	Should the output be in the list format. Defaults to FALSE
halo.rm	Logical. should halo observations be removed. Defaults to TRUE

Details

Two formats for the output are available. Either a vector of integers denoting for each observation, which cluster the observation belongs to. If halo observations are removed, these are set to NA. The second format is a list with a vector for each group containing the index for the member observations in the group. If halo observations are removed their indexes are omitted. The list format correspond to the following transform of the vector format `split(1:length(clusters), clusters)`, where `clusters` are the cluster information in vector format.

Value

A vector or list with cluster memberships for the observations in the initial distance matrix

densityClust	<i>Calculate clustering attributes based on the densityClust algorithm</i>
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Description

This function takes a distance matrix and optionally a distance cutoff and calculates the values necessary for clustering based on the algorithm proposed by Alex Rodrigues and Alessandro Laio (see references). The actual assignment to clusters are done in a later step, based on user defined threshold values. If a distance matrix is passed into `distance` the original algorithm described in the paper is used. If a matrix or data.frame is passed instead it is interpreted as point coordinates and `rho` will be estimated based on k-nearest neighbors of each point (`rho` is estimated as $\exp(-\text{mean}(x))$ where x is the distance to the nearest neighbors). This can be useful when data is so large that calculating the full distance matrix can be prohibitive.

Usage

```
densityClust(distance, dc, gaussian = FALSE, verbose = FALSE, ...)
```

Arguments

distance	A distance matrix or a matrix (or data.frame) for the coordinates of the data. If a matrix or data.frame is used the distances and local density will be estimated using a fast k-nearest neighbor approach.
dc	A distance cutoff for calculating the local density. If missing it will be estimated with <code>estimateDc(distance)</code>

gaussian	Logical. Should a gaussian kernel be used to estimate the density (defaults to FALSE)
verbose	Logical. Should the running details be reported
...	Additional parameters passed on to get.knn

Details

The function calculates rho and delta for the observations in the provided distance matrix. If a distance cutoff is not provided this is first estimated using [estimateDc\(\)](#) with default values.

The information kept in the densityCluster object is:

rho A vector of local density values

delta A vector of minimum distances to observations of higher density

distance The initial distance matrix

dc The distance cutoff used to calculate rho

threshold A named vector specifying the threshold values for rho and delta used for cluster detection

peaks A vector of indexes specifying the cluster center for each cluster

clusters A vector of cluster affiliations for each observation. The clusters are referenced as indexes in the peaks vector

halo A logical vector specifying for each observation if it is considered part of the halo

knn_graph kNN graph constructed. It is only applicable to the case where coordinates are used as input. Currently it is set as NA.

nearest_higher_density_neighbor index for the nearest sample with higher density. It is only applicable to the case where coordinates are used as input.

nn.index indices for each cell's k-nearest neighbors. It is only applicable for the case where coordinates are used as input.

nn.dist distance to each cell's k-nearest neighbors. It is only applicable for the case where coordinates are used as input.

Before running findClusters the threshold, peaks, clusters and halo data is NA.

Value

A densityCluster object. See details for a description.

References

Rodriguez, A., & Laio, A. (2014). *Clustering by fast search and find of density peaks*. Science, **344**(6191), 1492-1496. doi:10.1126/science.1242072

See Also

[estimateDc\(\)](#), [findClusters\(\)](#)

Examples

```
irisDist <- dist(iris[,1:4])
irisClust <- densityClust(irisDist, gaussian=TRUE)
plot(irisClust) # Inspect clustering attributes to define thresholds

irisClust <- findClusters(irisClust, rho=2, delta=2)
plotMDS(irisClust)
split(iris[,5], irisClust$clusters)
```

estimateDc

*Estimate the distance cutoff for a specified neighbor rate***Description**

This function calculates a distance cutoff value for a specific distance matrix that makes the average neighbor rate (number of points within the distance cutoff value) fall between the provided range. The authors of the algorithm suggests aiming for a neighbor rate between 1 and 2 percent, but also states that the algorithm is quite robust with regards to more extreme cases.

Usage

```
estimateDc(distance, neighborRateLow = 0.01, neighborRateHigh = 0.02)
```

Arguments

distance	A distance matrix
neighborRateLow	The lower bound of the neighbor rate
neighborRateHigh	The upper bound of the neighbor rate

Value

A numeric value giving the estimated distance cutoff value

Note

If the number of points is larger than 448 (resulting in 100,128 pairwise distances), 100,128 distance pairs will be randomly selected to speed up computation time. Use `set.seed()` prior to calling `estimateDc` in order to ensure reproducible results.

References

Rodriguez, A., & Laio, A. (2014). *Clustering by fast search and find of density peaks*. Science, **344**(6191), 1492-1496. doi:10.1126/science.1242072

Examples

```
irisDist <- dist(iris[,1:4])
estimateDc(irisDist)
```

findClusters	<i>Detect clusters in a densityCluster object</i>
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Description

This function uses the supplied rho and delta thresholds to detect cluster peaks and assign the rest of the observations to one of these clusters. Furthermore core/halo status is calculated. If either rho or delta threshold is missing the user is presented with a decision plot where they are able to click on the plot area to set the threshold. If either rho or delta is set, this takes precedence over the value found by clicking.

Usage

```
findClusters(x, ...)

## S3 method for class 'densityCluster'
findClusters(x, rho, delta, plot = FALSE, peaks = NULL, verbose = FALSE, ...)
```

Arguments

x	A densityCluster object as produced by <code>densityClust()</code>
...	Additional parameters passed on
rho	The threshold for local density when detecting cluster peaks
delta	The threshold for minimum distance to higher density when detecting cluster peaks
plot	Logical. Should a decision plot be shown after cluster detection
peaks	A numeric vector indicates the index of density peaks used for clustering. This vector should be retrieved from the decision plot with caution. No checking involved.
verbose	Logical. Should the running details be reported

Value

A densityCluster object with clusters assigned to all observations

References

Rodriguez, A., & Laio, A. (2014). *Clustering by fast search and find of density peaks*. Science, **344**(6191), 1492-1496. doi:10.1126/science.1242072

Examples

```
irisDist <- dist(iris[,1:4])
irisClust <- densityClust(irisDist, gaussian=TRUE)
plot(irisClust) # Inspect clustering attributes to define thresholds

irisClust <- findClusters(irisClust, rho=2, delta=2)
plotMDS(irisClust)
split(iris[,5], irisClust$clusters)
```

plotDensityClust	<i>Plot densityCluster results</i>
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Description

Generate a single panel of up to three diagnostic plots for a densityClust object.

Usage

```
plotDensityClust(
  x,
  type = "all",
  n = 20,
  mds = NULL,
  dim.x = 1,
  dim.y = 2,
  col = NULL,
  alpha = 0.8
)
```

Arguments

x	A densityCluster object as produced by densityClust
type	A character vector designating which figures to produce. Valid options include "dg" for a decision graph of δ vs. ρ , "gg" for a gamma graph depicting the decrease of γ ($= \delta * \rho$) across samples, and "mds", for a Multi-Dimensional Scaling (MDS) plot of observations. Any combination of these three can be included in the vector, or to produce all plots, specify type = "all".
n	Number of observations to plot in the gamma graph.
mds	A matrix of scores for observations from a Principal Components Analysis or MDS. If omitted, and a MDS plot has been requested, one will be calculated.
dim.x, dim.y	The numbers of the dimensions to plot on the x and y axes of the MDS plot.
col	Vector of colors for clusters.
alpha	Value in 0:1 controlling transparency of points in the decision graph and MDS plot.

Value

A panel of the figures specified in `type` are produced. If designated, clusters are color-coded and labelled. If present in `x`, the ρ and δ thresholds are designated in the decision graph by a set of solid black lines.

Author(s)

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Examples

```
data(iris)
data.dist <- dist(iris[, 1:4])
pca <- princomp(iris[, 1:4])

# Run initial density clustering
dens.clust <- densityClust(data.dist)
op <- par(ask = TRUE)

# Show the decision graph
plotDensityClust(dens.clust, type = "dg")

# Show the decision graph and the gamma graph
plotDensityClust(dens.clust, type = c("dg", "gg"))

# Cluster based on rho and delta
new.clust <- findClusters(dens.clust, rho = 4, delta = 2)

# Show all graphs with clustering
plotDensityClust(new.clust, mds = pca$scores)

par(op)
```

plotMDS

Plot observations using multidimensional scaling and colour by cluster

Description

This function produces an MDS scatterplot based on the distance matrix of the `densityCluster` object (if there is only the coordinates information, a distance matrix will be calculate first), and, if clusters are defined, colours each observation according to cluster affiliation. Observations belonging to a cluster core is plotted with filled circles and observations belonging to the halo with hollow circles. This plotting is not suitable for running large datasets (for example datasets with > 1000 samples). Users are suggested to use other methods, for example tSNE, etc. to visualize their clustering results too.

Usage

```
plotMDS(x, ...)
```

Arguments

`x` A densityCluster object as produced by `densityClust()`
`...` Additional parameters. Currently ignored

See Also

`densityClust()` for creating densityCluster objects, and `plotTSNE()` for an alternative plotting approach.

Examples

```
irisDist <- dist(iris[,1:4])
irisClust <- densityClust(irisDist, gaussian=TRUE)
plot(irisClust) # Inspect clustering attributes to define thresholds

irisClust <- findClusters(irisClust, rho=2, delta=2)
plotMDS(irisClust)
split(iris[,5], irisClust$clusters)
```

plotTSNE	<i>Plot observations using t-distributed neighbor embedding and colour by cluster</i>
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Description

This function produces an t-SNE scatterplot based on the distance matrix of the densityCluster object (if there is only the coordinates information, a distance matrix will be calculate first), and, if clusters are defined, colours each observation according to cluster affiliation. Observations belonging to a cluster core is plotted with filled circles and observations belonging to the halo with hollow circles.

Usage

```
plotTSNE(x, ...)
```

Arguments

`x` A densityCluster object as produced by `densityClust()`
`...` Additional parameters. Currently ignored

See Also

[densityClust\(\)](#) for creating densityCluster objects, and [plotMDS\(\)](#) for an alternative plotting approach.

Examples

```
irisDist <- dist(iris[,1:4])
irisClust <- densityClust(irisDist, gaussian=TRUE)
plot(irisClust) # Inspect clustering attributes to define thresholds

irisClust <- findClusters(irisClust, rho=2, delta=2)
plotTSNE(irisClust)
split(iris[,5], irisClust$clusters)
```

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