

# Package ‘densityClust’

October 24, 2017

**Type** Package

**Title** Clustering by Fast Search and Find of Density Peaks

**Version** 0.3

**Date** 2017-10-24

**Maintainer** Thomas Lin Pedersen <thomasp85@gmail.com>

**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)

**Suggests** testthat

**LinkingTo** Rcpp

**Imports** Rcpp, FNN, Rtsne, ggplot2, ggrepel, grDevices, gridExtra,  
RColorBrewer

**RoxygenNote** 6.0.1

**NeedsCompilation** yes

**Author** Thomas Lin Pedersen [aut, cre],  
Sean Hughes [aut],  
Xiaojie Qiu [aut]

**Repository** CRAN

**Date/Publication** 2017-10-24 12:52:51 UTC

## R topics documented:

densityClust-package . . . . .	2
clustered . . . . .	3
clusters . . . . .	4
densityClust . . . . .	5
estimateDc . . . . .	6

findClusters . . . . .	7
plotDensityClust . . . . .	9
plotMDS . . . . .	10
plotTSNE . . . . .	11
<b>Index</b>	<b>12</b>

---

densityClust-package    *Clustering by fast search and find of density peaks*

---

## Description

This package implement the clustering algorithm described by Alex Rodriguez and Alessandro Laio (2014). It provides the user with tools for generating the initial rho and delta values for each observation as well as using these to assign observations to clusters. This is done in two passes so the user is free to reassign observations to clusters using a new set of rho and delta thresholds, without needing to recalculate everything.

## Plotting

Two types of plots are supported by this package, and both mimics the types of plots used in the publication for the algorithm. The standard plot function produces a decision plot, with optional colouring of cluster peaks if these are assigned. Furthermore `plotMDS()` performs a multidimensional scaling of the distance matrix and plots this as a scatterplot. If clusters are assigned observations are coloured according to their assignment.

## Cluster detection

The two main functions for this package are `densityClust()` and `findClusters()`. The former takes a distance matrix and optionally a distance cutoff and calculates rho and delta for each observation. The latter takes the output of `densityClust()` and make cluster assignment for each observation based on a user defined rho and delta threshold. If the thresholds are not specified the user is able to supply them interactively by clicking on a decision plot.

## Author(s)

**Maintainer:** Thomas Lin Pedersen <thomasp85@gmail.com>

Authors:

- Sean Hughes
- Xiaojie Qiu <xqiu@uw.edu>

## 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**

[densityClust\(\)](#), [findClusters\(\)](#), [plotMDS\(\)](#)

**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)
```

---

clustered

*Check whether a densityCluster object have been clustered*

---

**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`*Extract cluster membership from a densityCluster object*

---

## 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

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

## 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

*Calculate clustering attributes based on the densityClust algorithm***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**

<code>distance</code>	A distance matrix or a matrix (or <code>data.frame</code> ) for the coordinates of the data. If a matrix or <code>data.frame</code> is used the distances and local density will be estimated using a fast k-nearest neighbor approach.
<code>dc</code>	A distance cutoff for calculating the local density. If missing it will be estimated with <code>estimateDc(distance)</code>
<code>gaussian</code>	Logical. Should a gaussian kernel be used to estimate the density (defaults to FALSE)
<code>verbose</code>	Logical. Should the running details be reported
<code>...</code>	Additional parameters passed on to <a href="#">get.knn</a>

**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

*Detect clusters in a densityCluster object*

---

**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      *Plot densityCluster results*

---

### 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 <code>densityClust</code>
type	A character vector designating which figures to produce. Valid options include "dg" for a decision graph of $\delta$ vs. $\rho$ , "gg" for a gamma graph depicting the decrease of $\gamma$ ( $= \delta * \rho$ ) 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 rho and delta thresholds are designated in the decision graph by a set of solid black lines.

### Author(s)

Eric Archer <eric.archer@noaa.gov>

### 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 <code>densityCluster</code> object as produced by <code>densityClust()</code>
...	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>
----------	---------------------------------------------------------------------------------------

---

**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 <a href="#">densityClust()</a>
...	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)
```

# Index

clustered, 3  
clusters, 4

densityClust, 5, 9  
densityClust(), 2, 3, 8, 10, 11  
densityClust-package, 2

estimateDc, 6  
estimateDc(), 5, 6

findClusters, 7  
findClusters(), 2-4, 6

get.knn, 5

plotDensityClust, 9  
plotMDS, 10  
plotMDS(), 2, 3, 11  
plotTSNE, 11  
plotTSNE(), 10

set.seed(), 7