

# Package ‘RidgeClust’

January 18, 2011

**Type** Package

**Title** A clustering algorithm based on the package “Mclust” and a merging approach. See `?ridgeline.clust`

**Version** 1.0

**Date** 2010-12-17

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**Description** A 2-step clustering-algorithm. At the 1. step a Mclust- (a Gaussian-Mixture-) clustering of the data is generated. At the 2. step components which lead to unimodal or weakly bimodal mixtures are merged. The measure of the bimodality is the ridgeline ratio. See `?ridgeline.clust`.

**License** GPL-2

**LazyLoad** yes

**Depends** mclust, ellipse

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`plot.RidgeClust` *Plot function.*

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## Description

Redefined generic plot function for the class `RidgeClust`. Plots results of the clustering algorithm `ridgeline.clust` in the case of two-dimensional data. Shows the scatterplot of the original data, the result of `Mclust` and the result after the merging.

**Usage**

```
plot(x, daten)
```

**Arguments**

x                    Result of the function `ridgeline.clust()`.  
daten                Data which was clustered (a matrix or a dataframe with two columns).

**Author(s)**

G. Alexandrovich

**See Also**

[ridgeline.clust](#)

**Examples**

```
#See example of ridgeline.clust
```

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RidgeClustPackage *A clustering algorithm based on Mclust and a merging approach. See ?ridgeline.clust*

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**Description**

A 2-step clustering-algorithm. At the 1. step a Mclust- (a Gaussian-Mixture-) clustering of the data is generated. At the 2. step components which lead to unimodal or weakly bimodal mixtures are merged. The measure of the bimodality is the ridgeline ratio. See `?ridgeline.clust`.

**Details**

Package:    ridgeline.clust  
Type:        Package  
Version:    1.0  
Date:        2010-12-17  
License:    GPL-2  
LazyLoad:   yes

```
ridgeline.clust(data)
```

**Author(s)**

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**References**

Alexandrovich G. (2010): Analytische Eigenschaften von Mischungen elliptischer Verteilungen und deren Anwendung in der Clusteranalyse, diploma thesis, Marburg University

Fraley, C., Raftery, A.E. (2006): Technical Report MCLUST Version 3 for R.

Hennig, C. (2009): Methods for merging Gaussian mixture components, Research Report no. 302, Department of Statistical Science, UCL

Surajit Ray and Bruce G. Lindsay (2005): The topography of multivariate normal mixtures, Ann. Statist. 33, Number 5, pp. 2042-2065

### See Also

[ridgeline.clust](#) [plot.RidgeClust](#)

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ridgeline.clust	<i>Computes a clustering of the data</i>
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### Description

A 2-step clustering-algorithm. At the 1. step a clustering based on a Gaussian mixture model is calculated for the data (the package Mclust is applied here). At the 2. step Gaussian components which lead to weakly separated (i.e. unimodal or weakly bimodal) submixtures are merged, so that the number of clusters is reduced.

### Usage

```
ridgeline.clust(daten, rst = 0.7, anz = 15, G = NULL, modelNames = NULL,
prior = NULL, control = emControl(), initialization = NULL, warn = FALSE, ...)
```

### Arguments

daten	Data to be clustered, a dataframe or a matrix.
rst	Cutoff value for the ridgeline ratio ( $0 \leq \text{rst} \leq 1$ ). $\text{rst} = 1$ implies merging only at a strict unimodality, $\text{rst}$ near 0 implicates merging of strongly separated components.
anz	Parameter for the mode-searching routine. For high-dimensional data must be increased. A suitable choice is $(15/2) * \text{Dimension}$
G	Mclust parameter. Vector with numbers of components for the clustering. See Mclust reference
modelNames	Covariance structure of the components. See Mclust reference
prior	Priors. See Mclust reference
control	Parameter-Object for the EM-Algorithm. See Mclust reference
initialization	See Mclust reference
warn	See Mclust reference
...	See Mclust reference

**Value**

A list of 2 Objects.

<code>Mclust</code>	Result of the Mclust-algorithm. See Mclust reference.
<code>Merged</code>	Result of the merging-algorithm. It is a list with following components:
<code>Merged\$class</code>	Vector of cluster-labels for each datapoint
<code>Merged\$info</code>	A $k \times 3$ matrix, where $k$ is the number of merging iterations. $i$ 'th row of this matrix consists of the numbers of the merged components a iteration $i$ and the ridgeline-ratio of this components. For example <code>info[1,] = c(1,2,0.8)</code> means that in the first iteration components 1 and 2 were merged and their ridgeline-ratio was 0.8. The new component gets the number 1 (the lesser one)
<code>Merged\$diecluster</code>	A list the $i$ -th element of which contains the numbers of old-clusters which were merged and build now the $i$ 'th cluster.
<code>Merged\$new_posteriors</code>	A $N \times M$ Matrix, where $N$ is the number of datapoints and $M$ is the number of clusters, which contains posteriori-probabilities, for each cluster and datapoint.

**Author(s)**

Grigory Alexandrovich

**References**

- Alexandrovich G. (2010): Analytische Eigenschaften von Mischungen elliptischer Verteilungen und deren Anwendung in der Clusteranalyse, diploma thesis, Marburg University
- Fraley, C., Raftery, A.E. (2006): Technical Report MCLUST Version 3 for R.
- Hennig, C. (2009): Methods for merging Gaussian mixture components, Research Report no. 302, Department of Statistical Science, UCL
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**See Also**

[Mclust plot.RidgeClust](#)

**Examples**

```
library(mvtnorm)

#Simulate 1000 points from a gaussian mixture of 5 components:
#Defining the means
mu_vek <- array(dim=c(2,5))
mu_vek[,1] <- c(4,5)
mu_vek[,2] <- c(1.5,5)
mu_vek[,3] <- c(2,4.5)
mu_vek[,4] <- c(4.1,1)
mu_vek[,5] <- c(5,1)

#Defining the covariances
sigma_vek <- array(dim=c(2,2,5))
```

```
sigma_vek[, ,1] <- rbind(c(0.3,0.05),c(0.05,.3))
sigma_vek[, ,2] <- rbind(c(0.1,0.05),c(0.05,.1))
sigma_vek[, ,3] <- rbind(c(0.2,0),c(0,0.2))
sigma_vek[, ,4] <- rbind(c(0.2,.1),c(.1,.2))
sigma_vek[, ,5] <- rbind(c(.2,.1),c(.1,.2))

#Defining the weights
weight <- c(0.2,0.25,0.15,0.2,0.2)

#Simulation loop
ind <- 1
dat <- array(dim=c(1000,2))

while(ind <= 1000)
{
  component <- rmultinom(1,1,weight)
  component <- which(component == 1)
  dat[ind,] <- rmvnorm(1, mean = mu_vek[,component], sigma = sigma_vek[, ,component])
  ind <- ind + 1
}

#Clustering
result <- ridgeline.clust(dat)

#Plot
plot(result,dat)
```

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