Package 'RelativeDistClust'

September 22, 2025

Type Package			
Title Clustering with a Novel Non Euclidean Relative Distance			
Version 0.1.0			
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Description Using the novel Relative Distance to cluster datasets. Implementation of a clustering approach based on the k-means algorithm that can be used with any distance. In addition, implementation of the Hartigan and Wong method to accommodate alternative distance metrics. Both methods can operate with any distance measure, provided a suitable method is available to compute cluster centers under the chosen metric. Additionally, the k-medoids algorithm is implemented, offering a robust alternative for clustering without the need of computing cluster centers under the chosen metric. All three methods are designed to support Relative distances, Euclidean distances, and any user-defined distance functions. The Hartigan and Wong method is described in Hartigan and Wong (1979) <doi:10.2307 2346830=""> and an explanation of the k-medoids algorithm can be found in Reynolds et al (2006) <doi:10.1007 s10852-005-9022-1="">.</doi:10.1007></doi:10.2307>			
License GPL-3			
Encoding UTF-8			
Imports compositions, proxy, utils, ggpubr, factoextra, ggplot2			
Suggests testthat (>= 3.0.0), clusterSim, fpc, gtools, cluster			
Config/testthat/edition 3			
RoxygenNote 7.3.2			
NeedsCompilation no			
Repository CRAN			
Date/Publication 2025-09-22 11:50:06 UTC			
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add_unique_numbers

Add values to a vector if they are not already in it

Description

This function adds two values to a vector if the values are not already in the vector.

Usage

```
add_unique_numbers(vector, num1, num2)
```

Arguments

vector	Vector with values
num1	Number. Value that will be added to the vector it it is no already in it.
num2	Number. Value that will be added to the vector it it is no already in it.

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Value

Returns the vector with the values added if they are not alredy in the vector.

Examples

```
mi_vector <- c(1, 2, 3, 4, 5)
num1 <- 8
num2 <- 10
mi_vector <- add_unique_numbers(mi_vector, num1, num2)</pre>
```

 $add_unique_numbers2$

Add one value to a vector if it is not already there

Description

This function adds one value to a vector if it is not already in the vector.

Usage

```
add_unique_numbers2(vector, num1)
```

Arguments

vector

Vector with values

num1

Number. Value that will be added to the vector it it is no already in it.

Value

Returns the vector with the value added if it is not already in the vector.

```
mi_vector <- c(1, 2, 3, 4, 5)
num1 <- 8
mi_vector <- add_unique_numbers2(mi_vector, num1)</pre>
```

AitchisonDistance

Aitchison distance

Description

This function calculates the Aitchison distance between two vectors.

Usage

```
AitchisonDistance(vect1, vect2)
```

Arguments

vect1 vector vect2 vector

Value

A number with the distance between vect1 and vect2.

Examples

```
AitchisonDistance(c(1,2,3), c(4,5,6))
```

BrayCurtisDissimilarity

Bray-Curtis dissimilarity

Description

This function calculates the Bray-Curtis dissimilarity between two vectors

Usage

```
BrayCurtisDissimilarity(x, y)
```

Arguments

x vectory vector

Value

A number with the Bray-Curtis dissimilarity between x and y.

centers_function_mean 5

Examples

```
BrayCurtisDissimilarity(c(1,2,3), c(4,5,6))
```

Description

This function calculates the center of a group using the mean of its components.

Usage

```
centers_function_mean(data, grouping)
```

Arguments

data Matrix. The points that we want to group are in the rows.

grouping List. List with the number of the rows of the data matrix that are in the group

[[i]].

Value

A matrix. The row i contains the centers of the group in [[i]].

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```
centers_function_RelativeDistance
```

Center of a cluster when the Relative distance is used.

Description

This function calculates the center of a group when the Relative distance is used to group.

Usage

```
centers_function_RelativeDistance(data, grouping)
```

Arguments

data Matrix. The points that we want to group are in the rows.

grouping List. List with the number of the rows of the data matrix that are in the group

[[i]].

Value

A matrix. The row i contains the centers of the group in [[i]].

Examples

ClustPlot

Plotting the clustring results

Description

This function performs a PCA to reduce the dataset to two dimensions. Then, it draws the points, marks the center of the groups, the exact groups and the obtained groups.

Usage

```
ClustPlot(data, grouping, exact_grouping, centers, k)
```

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Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

grouping List with information of the groups obtained using some clustering method.

Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to the group i are.

exact_grouping List with the information of the real groups present in the data. Each component

of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the

matrix data where the points belonging to the group i are.

centers Matrix. Each row contains the center of each group. The groups are obtained

using some clustering methods.

k Number. Number of groups.

Value

Returns a plot where it is possible to visualize the he points, the center of the groups, the exact groups (represented in the type of point used to represent the data) and the obtained groups (observed in the geometric froms that join the points).

Examples

DaviesBouldinIndex

Davies-Bouldin index

Description

This function calculates the Davies-Bouldin index as is defined by Davies and Bouldin (1979) without imposing that the use of the euclidean distance. This function allows calculating the Davies-Bouldin index using different distances.

Usage

```
DaviesBouldinIndex(data, FHW_output, distance)
```

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

FHW_output List. List with:

- centers: the information of the centers updated.
- grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

Function. This function designs how the distance is going to be calculated. It must have as input two vectors and as output the distance of these vectors.

Value

Returns a number, the value of the Davies-Bouldin index.

References

Davies, D. L., & Bouldin, D. W. (1979). A cluster separation measure. IEEE transactions on pattern analysis and machine intelligence, (2), 224-227.

Examples

DistanceBetweenGroups Distance between groups

Description

This function calculates the distance between points in two groups. For each point in the first group, it calculates the distance from that point to all points in the second group. Finally, it takes the minimum distance obtained.

Usage

```
DistanceBetweenGroups(group1, group2, FHW_output, distance, data)
```

Arguments

group2 Number. Number of the first group.

FHW_output List. Output of the Hartigan_and_Wong function. List with:

• centers: the information of the centers updated.

• grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

Value

Returns a number, the value of the minimum distance between pair of points of the two groups.

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DistanceSameGroup

Distance between points in the same group

Description

This function calculates the distance between points in the same group. This function calculates the distance between the pair of points in the group. Then, takes the maximum distance.

Usage

```
DistanceSameGroup(group1, FHW_output, data, distance)
```

Arguments

group1 Number. Number of the group.

FHW_output List. List with:

• centers: the information of the centers updated.

• grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row

of the matrix data where the points belonging to group i are.

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

Value

Returns a number, the value of the maximum distance between pair of points of the group.

Dist_IC1_IC2

Dist_IC1_IC2

Finding IC1 and IC2 from a distance matrix

Description

This function finds the IC1 and IC2 from a distance matrix. IC1 and IC2 are the closets and second closest cluster centers.

Usage

```
Dist_IC1_IC2(Dist_e_cent)
```

Arguments

Dist_e_cent Matrix. The position (i,j) contains the distance between the taxa i and the center j.

Value

Returns a matrix. The first column contain the IC1 and the second column contain the IC2.

Examples

```
dist=rbind(c(1,2,3),c(6,19,2),c(2,4,1),c(2,3,9))
Dist_IC1_IC2(dist)
```

DosMinimos

Finding the two smallest values for each row of a matrix

Description

This function finds the two smallest values for each row of a matrix matriz.

Usage

```
DosMinimos(matriz)
```

Arguments

matriz

Matrix

Value

Returns a matrix. The row i contains the two minimum values of the row i of the matrix matriz. The first column of the matriz contains the smallest value.

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Examples

```
ma=rbind(c(5,4,3,2,1), c(10,9,8,7,6), c(120,119,103,104,105))
DosMinimos(ma)
```

DunnIndex

Dunn's index

Description

This function calculates the Dunn's index as is defined in Bezdek and Pal (1995) without imposing that the use of the euclidean distance. This function allows calculating the Dunn's index using different distances.

Usage

```
DunnIndex(data, FHW_output, distance)
```

Arguments

data

Matrix with dim(data)[1] points of dim(data)[2] dimensions.

FHW_output

List. List with:

- centers: the information of the centers updated.
- grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

distance

Function. This function designs how the distance is going to be calculated. It must have as input two vectors and as output the distance of these vectors.

Value

Returns a number, the value of the Dunn's index.

References

Bezdek, J. C., & Pal, N. R. (1995, November). Cluster validation with generalized Dunn's indices. In Proceedings 1995 second New Zealand international two-stream conference on artificial neural networks and expert systems (pp. 190-193). IEEE.

d_i_other_group

Examples

d_i_other_group

Distance between a point and a group

Description

This function calculates the distance between the point i of the data matrix and all the components in the group num.

Usage

```
d_i_other_group(data, i, distance, FHW_output, num)
```

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

i Number. Number of the row of data where the point is.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

FHW_output List. List with:

- centers: the information of the centers updated.
- grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

num Number. Number of the group from FHW_output\$grouping.

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Value

Returns a vector. The component j contains the distance between the point in the row i of the data matrix and the point j of the group num.

Examples

ECDentroCluster

Sum of squared errors within the cluster

Description

The sum of squared errors within the cluster (also known as inertia) is calculated. We calculate the squared distance between the points that belong to a cluster and the cluster centroid. Then, we sum all the squared distances obtained. In this function the user can choose the distance that want to use to calculate the sum of squared errors within the cluster.

Usage

```
ECDentroCluster(data, FHW_output, distance)
```

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

FHW_output List. List with:

• centers: the information of the centers updated.

ECDentroCluster3

• grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

distance

Function. This function designs how the distance is going to be calculated. It must have as input two vectors and as output the distance of these vectors.

Value

Returns a vector. The component i contains the sum of squared errors value of group i.

Examples

```
set.seed(231)
data1=gtools::rdirichlet(10,c(1,1,1,4,4))
data=t(data1)
grouping=list(c(1,2,3),c(4,5))
centers=centers_function_mean(data, grouping)
FHW_output=list(centers=centers, grouping=grouping)
distance=Euclideandistance
ECDentroCluster(data, FHW_output, distance)
```

ECDentroCluster3

Sum of errors within the cluster

Description

We calculate the distance between the points that belong to a cluster and the cluster centroid. Then, we sum all the distances obtained. In this function the user can choose the distance that want to use to calculate the sum of errors within the cluster.

Usage

```
ECDentroCluster3(data, FHW_output, distance)
```

Arguments

data

Matrix with dim(data)[1] points of dim(data)[2] dimensions.

FHW_output

List. List with:

- centers: the information of the centers updated.
- grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

distance

Function. This function designs how the distance is going to be calculated. It must have as input two vectors and as output the distance of these vectors.

Value

Returns a vector. The component i contains the sum of squared errors value of group i.

Examples

```
#'set.seed(231)
data1=gtools::rdirichlet(10,c(1,1,1,4,4))
data=t(data1)
grouping=list(c(1,2,3),c(4,5))
centers=centers_function_mean(data, grouping)
FHW_output=list(centers=centers, grouping=grouping)
distance=Euclideandistance
ECDentroCluster3(data, FHW_output, distance)
```

encontrar_componente

Finding the component in the list that contains a value

Description

This function finds in which component of the list lista the number valor is.

Usage

```
encontrar_componente(lista, valor)
```

Arguments

lista List. Each component of the list has a vector. The different vector can not

contain the same number.

valor Number. We want to know in which component of the list lista the number

valor is.

Value

Returns a number. Return the number of the component of list that contains the number valor.

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Examples

```
mi_lista <- list(
a = c(1, 2, 3),
b = c(6,7,8,9),
c = c(4,5)
)
valor=7
encontrar_componente(mi_lista, valor)</pre>
```

Euclideandistance

Euclidean distance

Description

This function calculates the euclidean distance between two vectors

Usage

```
Euclideandistance(vect1, vect2)
```

Arguments

```
vect1 vector
vect2 vector
```

Value

A number with the distance between vect1 and vect2.

```
Euclideandistance(c(1,2,3), c(4,5,6))
```

Hartigan_and_Wong

Hartigan_and_Wong

Flexibilization of the Hartigan and Wong algorithm

Description

This function implements the Hartigan and Wong algorithm (Hartigan and Wong, 1979) without imposing the use of the euclidean distance and without imposing that the centers of the groups are calculated by averaging the points. This function allow the use of other distances and different ways to calculate the centers of the groups.

Usage

```
Hartigan_and_Wong(
  data,
  distance,
  k,
  centers_function,
  init_centers,
  seed = NULL,
  ITER
)
```

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

k Number. Number of groups into which we are going to group the different

points.

centers_function

Function. This function designs how the centers of the groups will be calculated. It must have as input data and grouping and as output a matrix that has the centers. This matrix will have as many rows as centers. With grouping we mean a list. The list component i has a vector with the numbers of the row of

the matrix data where the points belonging to group i are.

init_centers Function. This function designs how we are going to calculate the initial centers.

The input must be the data, distance and k and the output must be a matrix

where each row has the center of one group.

seed Number. Number to fix a seed and be able to reproduce your results.

ITER Number. Maximum number of iterations.

Value

Returns a list with:

- centers: the information of the centers updated. Matrix with dim(centers)[1] centers of dim(centers)[2] dimensions.
- grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

References

Hartigan, J. A., & Wong, M. A. (1979). Algorithm AS 136: A k-means clustering algorithm. Journal of the royal statistical society. series c (applied statistics), 28(1), 100-108.

Examples

Hartigan_and_Wong_total

Hartigan and Wong algorithm

Description

This function apply the Hartigan_and_Wong to different number of groups and calculates quality metrics as Silhouette.

Usage

```
Hartigan_and_Wong_total(
  data,
  distance,
  centers_function,
  init_centers,
  seed = NULL,
```

```
ITER,
KK = 10,
index = "DaviesBouldin",
k = NULL
)
```

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

centers_function

Function. This function designs how the centers of the groups will be calculated. It must have as input data and grouping and as output a matrix that has the centers. This matrix will have as many rows as centers. With grouping we mean a list. The list component i has a vector with the numbers of the row of

the matrix data where the points belonging to group i are.

init_centers Function. This function designs how we are going to calculate the initial centers.

The input must be the data, distance and k and the output must be a matrix

where each row has the center of one group.

seed Number. Number to fix a seed and be able to reproduce your results.

ITER Number. Maximum number of iterations.

KK Number. Calculates the algorithm for the number of groups 2,3,...,KK. Default

KK=10.

index Character. If index="Silhouette" the function returns the results obtained

with the number of groups (between 2 and KK) that maximize the Silhouette index. If index="DaviesBouldin" the function returns the results obtained with the number of groups (between 2 and KK) that minimize the Davies Bouldin index. If index="Dunn" the function returns the results obtained with the number of groups (between 2 and KK) that maximize the Dunn index. Default:

"DaviesBouldin".

Number. If k is not NULL the function returns the results obtained with k

groups.

Value

k

Returns a list with:

- Number of groups: Number of groups took into account to cluster.
- Output_of_grouping: list with the centers and the clusters.
- Quality: vector with the Silhouette index, Davies Bouldin Index, the Dunn index, the Within Cluster Sum (WCS) and the time (in seconds) that the function Hartigan_and_Wong needs to be executed. The WCS is equal to the sum of the distance of each point to the center of its group.

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References

Hartigan, J. A., & Wong, M. A. (1979). Algorithm AS 136: A k-means clustering algorithm. Journal of the royal statistical society. series c (applied statistics), 28(1), 100-108.

Examples

init_centers_hw

Initializing the centers

Description

This function initializes the cluster centers following the procedure described in the 'Additional Comments' section of Hartigan and Wong (1979), without restricting the method to the use of Euclidean distance.

Usage

```
init_centers_hw(data, distance, k, centers_function)
```

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

k Number. Number of groups into which we are going to group the different

points.

centers_function

Function. This function designs how the centers of the groups will be calculated. It must have as input data and grouping and as output a matrix that has the

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centers. This matrix will have as many rows as centers. With grouping we mean a list. The list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

Value

Returns a matrix where each row is the center of a group.

References

Hartigan, J. A., & Wong, M. A. (1979). Algorithm AS 136: A k-means clustering algorithm. Journal of the royal statistical society. series c (applied statistics), 28(1), 100-108.

Examples

init_centers_random In

Initializing the centers

Description

This function initializes the centers of the groups randomly.

Usage

```
init_centers_random(data, distance, k, centers_function)
```

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

k Number. Number of groups into which we are going to group the different

points.

centers_function

Function. This function designs how the centers of the groups will be calculated. It must have as input data and grouping and as output a matrix that has the centers. This matrix will have as many rows as centers. With grouping we mean a list. The list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

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Value

Returns a matrix where each row is the center of a group.

Examples

kmedois_distance

K-Medoids

Description

This function apply the K-Medoids with any distance to different number of groups and calculates quality metrics as Silhouette.

Usage

```
kmedois_distance(data, distance, KK = 10, index = "DaviesBouldin", k = NULL)
```

Arguments

data	Matrix with dim(data)[1] points of dim(data)[2] dimensions.
distance	Function. This function designs how the distance is going to be calculated. It must have as input two vectors and as output the distance of these vectors.
KK	Number. Calculates the K-Medoids for the number of groups $2,3,,KK$. Default $KK=10$.
index	Character. If index="Silhouette" the function returns the results obtained with the number of groups (between 2 and KK) that maximize the Silhouette index. If index="DaviesBouldin" the function returns the results obtained with the number of groups (between 2 and KK) that minimize the Davies Bouldin index. If index="Dunn" the function returns the results obtained with the number of groups (between 2 and KK) that maximize the Dunn index. Default: "DaviesBouldin".
k	Number. If k is not NULL the function returns the results obtained with the K-Medoids for k groups.

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Value

Returns a list with:

- Number_of_groups: Number of groups took into account to cluster.
- Output_of_grouping: list with the centers and the clusters.
- Quality: vector with the Silhouette index, Davies Bouldin Index, the Dunn index, the Within Cluster Sum (WCS) and the time (in seconds) that the algorithm needs to be executed. The WCS is equal to the sum of the distance of each point to the center of its group.

Examples

ManhattanDistance

Manhattan distance

Description

This function calculates the Manhattan distance between two vectors

Usage

```
ManhattanDistance(x, y)
```

Arguments

```
x vector
y vector
```

Value

A number with the distance between x and y.

```
ManhattanDistance(c(1,2,3), c(4,5,6))
```

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NEC Non Euclidean Algorithm to Cluster

Description

We give initial centers, calculate the distance between each point and each center and assign each point to the center with minimum distance. Calculate the center of the group and repeat the process. The process is stopped when the distance between a center and the previous one is small than COTA or the maximum number of iterations is reached.

Usage

NEC(data, distance, k, centers_function, init_centers, seed = NULL, ITER, COTA)

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

k Number. Number of groups into which we are going to group the different

points.

centers_function

Function. This function designs how the centers of the groups will be calculated. It must have as input data and grouping and as output a matrix that has the centers. This matrix will have as many rows as centers. With grouping we mean a list. The list component i has a vector with the numbers of the row of the matrix data where the resists belowing to group is group.

the matrix data where the points belonging to group i are.

init_centers Function. This function designs how we are going to calculate the initial centers.

The input must be the data, distance and k and the output must be a matrix

where each row has the center of one group.

seed Number. Number to fix a seed and be able to reproduce your results.

ITER Number. Maximum number of iterations.

COTA Number. The process is stopped when the distance between a center and the

previous one is smaller than COTA.

Value

Returns a list with:

- FHW_output; is a list with
 - centers: the information of the centers updated. Matrix with dim(centers)[1] centers of dim(centers)[2] dimensions.
 - grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

NEC_total

Stop_Criteria: returns the distance between one center and the previous one for all the iterations

- Chanche_yes_no: matrix, in the position [i,j] returns "yes" if the point i have changed its group in the iteration j and return "no" if the point have not changed.
- all_output: is a list with the information of the center and the groups of each iteration of the process

Examples

NEC_total

NEC algorithm

Description

This function apply the NEC to different number of groups and calculates quality metrics as Silhouette.

Usage

```
NEC_total(
  data,
  distance,
  centers_function,
  init_centers,
  seed = NULL,
  ITER,
  COTA,
  KK = 10,
  index = "DaviesBouldinIndex",
  k = NULL
)
```

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Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

centers_function

Function. This function designs how the centers of the groups will be calculated. It must have as input data and grouping and as output a matrix that has the centers. This matrix will have as many rows as centers. With grouping we mean a list. The list component i has a vector with the numbers of the row of

the matrix data where the points belonging to group i are.

init_centers Function. This function designs how we are going to calculate the initial centers.

The input must be the data, distance and k and the output must be a matrix

where each row has the center of one group.

seed Number. Number to fix a seed and be able to reproduce your results.

ITER Number. Maximum number of iterations.

COTA Number. The process is stopped when the distance between a center and the

previous one is smaller than COTA.

KK Number. Calculates the algorithm for the number of groups 2,3,...,KK. Default

KK=10.

index Character. If index="Silhouette" the function returns the results obtained

with the number of groups (between 2 and KK) that maximize the Silhouette index. If index="DaviesBouldin" the function returns the results obtained with the number of groups (between 2 and KK) that minimize the Davies Bouldin index. If index="Dunn" the function returns the results obtained with the number of groups (between 2 and KK) that maximize the Dunn index. Default:

"DaviesBouldin".

k Number. If k is not NULL the function returns the results obtained with k

groups.

Value

Returns a list with:

- Number_of_groups: Number of groups took into account to cluster.
- Output_of_grouping: list with the centers and the clusters.
- Quality: vector with the Silhouette index, Davies Bouldin Index, the Dunn index, the Within Cluster Sum (WCS) and the time (in seconds) that algorithm needs to be executed. The WCS is equal to the sum of the distance of each point to the center of its group.

Number_of_failes

Number_of_failes

Comparison of groupings

Description

This function compares the real clustering with a clustering obtained with some mathematical method. For each group, this function calculates the number of components that are in the expected grouping that are not in the real grouping. This function adds this value for all groups. It calculates it for all possible combinations of groups and returns the minimum value.

Usage

```
Number_of_failes(grouping_exact, grouping_obtained)
```

Arguments

grouping_exact List. Each component of the list contains a vector with the components of one group. This list represents the actual grouping of the data.

grouping_obtained

List. Each component of the list contains a vector with the components of one group. This list represents the grouping obtained by some mathematical method.

Value

Returns a number with the quantity of points that are misclassified in the grouping_obtained.

```
grouping_exact=list(c(1,2,3,4,5),c(6,7),c(8,9))
grouping_obtained=list(c(1,3,7),c(2,4,6),c(8,9,5))
Number_of_failes(grouping_exact, grouping_obtained)
```

RelativeDistance 29

RelativeDistance

Relative Distance

Description

This function calculates the Relative Distance between two vectors.

Usage

```
RelativeDistance(vect1, vect2)
```

Arguments

vect1 vector vect2 vector

Value

A number with the distance between vect1 and vect2.

Examples

```
RelativeDistance(c(1,2,3), c(4,5,6))
```

 $\\ {\tt Silhouette}\\$

Silhouette

Description

This function calculates the Silhouette as is defined in Rousseeuw (1987) without imposing that the use of the euclidean distance. This allows calculating the Silhouette using different distances. Note that the Silhouette must be calculated using a distance that is a a ratio scale (Rousseeuw, 1987).

Usage

```
Silhouette(data, FHW_output, distance)
```

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Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

FHW_output List. List with:

- centers: the information of the centers updated.
- grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

distance

Function. This function designs how the distance is going to be calculated. It must have as input two vectors and as output the distance of these vectors.

Value

Returns a vector. The component i contains the Silhouette value of the point in the row i of the data matrix.

References

Rousseeuw, P.J. (1987) Silhouettes: A graphical aid to the interpretation and validation of cluster analysis. J. Comput. Appl. Math., 20, 53–65.

Step4 31

Description

This function implements the Step 4 of the Hartigan and Wong (Hartigan and Wong, 1979) algorithm without imposing that the use of the euclidean distance and without imposing that the centers of the groups are calculated by averaging the points. This function allows other distances to be used and allows the centers of the groups to be calculated in different ways.

Usage

```
Step4(
  data,
  centers,
  grouping,
  LIVE_SET_original,
  distance,
  centers_function,
  Ic12_change,
  index
)
```

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

centers Matrix with dim(centers)[1] centers of dim(centers)[2] dimensions.

grouping List. Each component of the list contains a vector with the points that belong to

that group. More specifically, the list component i has a vector with the numbers

of the row of the matrix data where the points belonging to group i are.

LIVE_SET_original

Vector that contains the groups that have been modified in the previous Step 6.

The Step 6 is described in Hartigan and Wong (1979).

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

centers_function

Function. This function designs how the centers of the groups will be calculated.

It must have as input data and grouping and as output a matrix that has the

centers. This matrix will have as many rows as centers.

Ic12_change Matrix. The first row contains the IC1 of each point. The second column con-

tains the IC2 of each point. IC1 and IC2 are the closets and second closest

cluster centers.

index Number. When a point is reallocated, index becomes zero.

Value

Returns a list with:

• centers: the information of the centers updated. Matrix with dim(centers)[1] centers of dim(centers)[2] dimensions.

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• IC1andIC2: the information of the IC1 and IC2 updated. Matrix. The first row contains the IC1 of each point. The second column contains the IC2 of each point. IC1 and IC2 are the closets and second closest cluster centers.

- grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.
- Live_set: Vector. Contains the groups that have been modified during the Step 4.
- no_Change: vector with the points that do not change its group. More specifically, contains the row of the matrix data where these points are.

References

Hartigan, J. A., & Wong, M. A. (1979). Algorithm AS 136: A k-means clustering algorithm. Journal of the royal statistical society. series c (applied statistics), 28(1), 100-108.

```
set.seed(231)
data1=gtools::rdirichlet(10,c(1,1,4,4,20,20))
data=t(data1)
k=3
seed=5
if(!is.null(seed)){
 set.seed(seed)
centers <- data[sample(1:nrow(data), k), ]</pre>
#We calculate the distance between each row of the data matrix and the centers
Dist_e_cent=matrix(0,dim(data)[1],dim(centers)[1])
for (i in 1:(dim(data)[1])){
for (j in 1:(dim(centers)[1])){
Dist_e_cent[i,j]=Euclideandistance(data[i,],centers[j,])
}
}
Ic12=Dist_IC1_IC2(Dist_e_cent)
Ic12_change=Ic12
Group=Ic12[,1]
grouping<-list()</pre>
for(i in 1:(max(Group))){
grouping[[i]]=which(Group==i)
#Update the clusters centers.
centers=centers_function_mean(data, grouping)
```

Step6 33

Step6

Step 6 of the Hartigan and Wong algorithm

Description

This function implements the Step 6 of the Hartigan and Wong (Hartigan and Wong, 1979) algorithm without imposing that the use of the euclidean distance and without imposing that the centers of the groups are calculated by averaging the points. This function allows other distances to be used and allows the centers of the groups to be calculated in different ways.

Usage

```
Step6(
  data,
  centers,
  grouping,
  distance,
  centers_function,
  Ic12_change,
  Ic12,
  index
)
```

Arguments

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

centers Matrix with dim(centers)[1] centers of dim(centers)[2] dimensions.

grouping List. Each component of the list contains a vector with the points that belong to

that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i are.

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distance Function. This function designs how the distance is going to be calculated. It must have as input two vectors and as output the distance of these vectors.

centers_function

Function. This function designs how the centers of the groups will be calculated. It must have as input data and grouping and as output a matrix that has the

centers. This matrix will have as many rows as centers.

Matrix. Contains IC1 and IC2 after the Step 4 is carried out. The first row Ic12_change

contains the IC1 of each point. The second column contains the IC2 of each

point. IC1 and IC2 are the closets and second closest cluster centers.

Ic12 Matrix. Contains IC1 and IC2 before the Step 4 is carried out. The first row

contains the IC1 of each point. The second column contains the IC2 of each

point. IC1 and IC2 are the closets and second closest cluster centers.

index Number. When a point is reallocated, index becomes zero.

Value

Returns a list with:

- centers: the information of the centers updated. Matrix with dim(centers)[1] centers of dim(centers)[2] dimensions.
- IC1 and IC2: the information of the IC1 and IC2 updated. Matrix. The first row contains the IC1 of each point. The second column contains the IC2 of each point. IC1 and IC2 are the closets and second closest cluster centers.
- grouping: the information of the groups updated. List. Each component of the list contains a vector with the points that belong to that group. More specifically, the list component i has a vector with the numbers of the row of the matrix data where the points belonging to group i
- Live_set: Vector. Contains the groups that have been modified during the Step 6.
- index: number. The information of index updated.

References

Hartigan, J. A., & Wong, M. A. (1979). Algorithm AS 136: A k-means clustering algorithm. Journal of the royal statistical society. series c (applied statistics), 28(1), 100-108.

```
set.seed(231)
data12=gtools::rdirichlet(10,c(1,1,4,4,20,20))
data1=t(data12)
k=3
seed=5
distance<- function(vect1, vect2){</pre>
sqrt(sum((vect1-vect2)^2))
}
centers_function<-function(data, grouping){</pre>
```

to_minimize 35

```
center=matrix(0,length(grouping), dim(data)[2])
 for (i in 1:(length(grouping))){
   if(length(grouping[[i]])==1){
     center[i,]=data[grouping[[i]],]
     center[i,]=apply(data[grouping[[i]],],2,mean)
   }
 }
 return(center)
}
if(!is.null(seed)){
 set.seed(seed)
}
centers <- data1[sample(1:nrow(data1), k), ]</pre>
#We calculate the distance between each row of the data matrix and the centers
Dist_e_cent=matrix(0,dim(data1)[1],dim(centers)[1])
for (i in 1:(dim(data1)[1])){
 for (j in 1:(dim(centers)[1])){
   Dist_e_cent[i,j]=distance(data1[i,],centers[j,])
 }
}
#We obtain the IC1 and IC2 for each taxa
Ic12_change=Dist_IC1_IC2(Dist_e_cent)
Group=Ic12_change[,1]
grouping<-list()</pre>
for(i in 1:(max(Group))){
grouping[[i]]=which(Group==i)
}
#Update the clusters centers.
centers=centers_function(data1, grouping)
Ic12=cbind(c(1,1,3,3,2,2),c(1,2,1,2,3,3))
P1=Step6(data1, centers, grouping, distance, centers_function, Ic12_change,Ic12, 0)
```

to_minimize

Sum of the distance between the points in a group and a given center.

Description

This function calculates the sum of the distance between the points in a group and a given center of the group. The function calculates these values for all groups and then adds them together. The user can choose which distance to choose.

36 vector_a_lista

Usage

```
to_minimize(inicenters_v, data, grouping, distance)
```

Arguments

inicenters_v Vector. Vector with the centers of the groups that has more than one point.

The centres are arranged by the number of the group. If a group has only one component, this center is not included in the vector. The vector contain all the components of the center of the first group (if this group has more than one point, otherwise the vector will start with the components of the center of the second group), then all the components of the second group (if this group has more than one point), then all the components of the third group (if this group has more than one point), and so on until the center of all groups with

more than one point are introduced.

data Matrix with dim(data)[1] points of dim(data)[2] dimensions.

grouping List. Each component of the list contains a vector with the points that belong to

that group. More specifically, the list component i has a vector with the numbers

of the row of the matrix data where the points belonging to group i are.

distance Function. This function designs how the distance is going to be calculated. It

must have as input two vectors and as output the distance of these vectors.

Value

Returns a number. First this function calculates the distance between each point of a group and its given center and sum these values. Then, the function sum the values obtained for each group. This is the output.

Examples

```
grouping=list(c(1,2,3),c(4,5),c(6,7))
set.seed(451)
data=t(gtools::rdirichlet(10, c(1,1,1,4,4,9,9)))
inicenters=runif(dim(data)[2]*length(grouping), 0.1, 0.9)
inicenters_v=as.vector(inicenters)
to_minimize(inicenters_v, data, grouping, Euclideandistance)
```

vector_a_lista

Vector to list

Description

This function returns a list. The component of the list i contains the positions of the vector that are equal to i.

Usage

```
vector_a_lista(clustering_vector)
```

vector_a_lista 37

Arguments

Value

Returns a list. The component of the list i contains the positions of the vector that are equal to i.

```
vect=c(1,1,1,1,1,2,2,2,2,2,3,3,3,3,3)
vector_a_lista(vect)
```

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