Package 'bios2mds'

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Title From BIOlogical Sequences to MultiDimensional Scaling

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Depends R (>= 2.0), gtools, amap, e1071, ggplot2, cluster
Description Bios2mds is primarily dedicated to the analysis of biological sequences by metric Multi- Dimensional Scaling with projection of supplementary data. It contains functions for reading multiple sequence alignment files, calculating distance matrices, performing metric multidimensional scaling and visualizing results.
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Description

The bios2mds package is developed in the Bioinformatics team at Integrated Neurovascular Biology Laboratory, UMR CNRS 6214 / INSERM 771, University of Angers - FRANCE.

This package is dedicated to the analysis of biological sequences by metric MultiDimensional Scaling (MDS) with projection of supplementary data. It contains functions for reading multiple sequence alignment (MSA) files, calculating distance matrices from MSA files, performing MDS analysis and visualizing results. The MDS analysis and visualization tools can be applied to any kind of data.

The main functionalities of bios2mds are summarized below:

(1) BUILDING DISTANCE MATRICES FROM MULTIPLE SEQUENCE ALIGNMENTS:

Several functions allow users to read multiple sequence alignment files and to compute matrices of distances between these sequences:

- read.fasta: reads a multiple sequence alignment in FASTA format.
- read.msf: reads a multiple sequence alignment in MSF format.
- mat.dif: computes a matrix of pairwise distances between sequences based on sequence difference.
- mat.dis: computes a matrix of pairwise distances between sequences based on sequence dissimilarity.
- (2) MULTIDIMENSIONAL SCALING: A function performs metric MDS analysis of a distance matrix between active elements with the option of projecting supplementary elements onto the active space.
 - mmds: performs metric multidimensional scaling.
- (3) GRAPHICAL TOOLS: Several functions are proposed to visualize results of metric MDS analysis:
 - scree.plot: draws the scree plot of eigenvalues.
 - mmds.2D.plot: draws a scatter plot of the MDS coordinates.
 - mmds.plot: wrapper function that draws the scree plot and three scatter plots of MDS coordinates.
 - col.group: colors scatter plots with user provided groupings and colors.
 - write.mmds.pdb: writes MDS coordinates in a PDB formatted file for 3D visualisa-
- (4) CLUSTER ANALYSIS: Several functions allow users to perform data clustering and to assess the clustering robustness:

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- kmeans.run: performs multiple runs of K-means clustering and analyzes clusters
- sil.score: calculates the silhouette score from multiple K-means runs to determine the optimal number of clusters.

(5) DATASETS: Two raw datasets are proposed to test the functionalities of bios2mds. They correspond to the multiple sequence alignments of GPCRs from *H. sapiens* and *D. melanogaster* in .msf and .fa formats (msa/human_gpcr.* and msa/drome_gpcr.*). They are based on the non-redundant sets of non-olfactory class A GPCRs, prepared and analyzed in Deville et al. (2009) and updated with the July 2009 release of Uniprot http://www.uniprot.org. Each MSA file is related to a .csv file that assigns a group and a color to each sequence of the alignment (csv/human_gpcr_group.csv and csv/drome_gpcr_group.csv).

Pre-analyzed data from these two MSA files are in gpcr.

For an index of functions, use library(help = bios2mds).

Details

Package bios2mds
Type Package
Version 1.0
Repository CRAN
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License GPL version 2 or newer

Collate Other useful packages can be found in the CRAN task view.

See http://cran.stat.sfu.ca/web/views/Multivariate.html and http://cran.stat.sfu.ca/web/views/Cluster.html

Author(s)

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Maintainer: Marie Chabbert <marie.chabbert@univ-angers.fr>

References

citation('bios2mds')

Examples

```
# The MSA files provided with the package correspond to the sequence
# alignment of non-olfactory class A G-protein-coupled receptors from
# H. sapiens and D. melanogaster prepared by Deville et al. (2009).
# loading GPCR data
data(gpcr)
# building distance matrices between the aligned GPCR sequences from
# H. sapiens and D. melanogaster
```

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```
human <- read.fasta(system.file("msa/human_qpcr.fa", package = "bios2mds"))</pre>
drome <- read.fasta(system.file("msa/drome_gpcr.fa", package = "bios2mds"))</pre>
#active <- mat.dif(human, human)</pre>
# or
active <- gpcr$dif$sapiens.sapiens
#sup <- mat.dif(drome, human)</pre>
# or
sup <- gpcr$dif$melanogaster.sapiens
# performing MDS analysis of the GPCR sequences from H. sapiens
mmds1 <- mmds(active = active)
mmds1 <-col.group(mmds1, system.file("csv/human_gpcr_group.csv"
,package = "bios2mds"))
# performing MDS analysis of the GPCR sequences from H. sapiens
# with projection of GPCRs from D. melanogaster
# as supplementary elements onto the space of human GPCRs
mmds2 <- mmds(active = active, sup = sup)</pre>
mmds2 <-col.group(mmds2,system.file("csv/human_gpcr_group.csv"</pre>
,package = "bios2mds"))
mmds2 <-col.group(mmds2,system.file("csv/drome_gpcr_group.csv"</pre>
,package = "bios2mds"),data = "sup")
# displaying MDS coordinates
layout (matrix (1:6, 2, 3))
scree.plot(mmds1$eigen.perc, lab = TRUE, title = "Scree plot of metric MDS")
mmds.2D.plot(mmds1, title = "Sequence space of human GPCRs")
mmds.2D.plot(mmds2, title = "Projection of GPCRs from D. melanogaster
onto the space space of human GPCRs ", active.alpha = 0.3)
# writing PDB files for 3D visualization of MDS coordinates
write.mmds.pdb(mmds1)
```

col.group

Links elements in a mmds object to specific groups and colors

Description

Links elements in a mmds object to user-provided groups and colors.

Usage

```
col.group(x, file, data = "active")
```

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Arguments

X	a mmds object obtained from mmds function
file	a string of characters to indicate the file name assigning groups and colors to each active OR each supplementary element of the mmds object.
data	a string of characters to specify whether the assigned elements in the mmds object are active ("active") or supplementary ("sup") data. Default is "active".

Details

col.group assigns each element of the mmds object to user-provided groupings and colors for coloring and labeling mmds scatter plots.

col.group requires a formatted file. See "csv/human_gpcr_group.csv" for an example. Each line corresponds to one element of the mmds object and must contain three parameters separated by ",". The first parameter is the element name, as given in the multiple sequence alignment file.

The second parameter is the group name. Groupings must be provided by the user.

The third parameter is the group color in full letters (example: "black", "green"). Two or more groups can have the same color, but elements within the same group must have the same color. The group is colored by the first color encountered.

Value

Adds data to a mmds object in order to color and label mmds scatter plots with user-provided groupings and colors.

Author(s)

Jean-Michel Becu

See Also

See colors function (default R package).

See getcol in made4 package for special colour palette developed to maximize the contrast between colours.

Examples

```
# performing metric MDS on human GPCRs with projection of
# GPCRs from D. melanogaster as supplementary data:
data(gpcr)
active <- gpcr$dif$sapiens.sapiens
sup <- gpcr$dif$melanogaster.sapiens
mmds1 <- mmds(active = active, sup = sup)
mmds1<-col.group(mmds1, system.file("csv/human_gpcr_group.csv", package = "bios2mds"))
mmds1<-col.group(mmds1, system.file("csv/drome_gpcr_group.csv", package = "bios2mds"), data = "sup")</pre>
```

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dif	Difference score
-----	------------------

Description

Measures the difference score between two aligned amino acid or nucleotide sequences.

Usage

```
dif(seq1, seq2, gap = FALSE, aa.strict = FALSE)
```

Arguments

seq1	a character vector representing a first sequence.
seq2	a character vector representing a second sequence.
gap	a boolean indicating whether the gap character should be taken as a supplementary symbol (TRUE) or not (FALSE). Default is FALSE.
aa.strict	a boolean indicating whether only strict amino acids should be taken into account (TRUE) or not (FALSE). Default is FALSE.

Details

The difference score between two aligned sequences is given by the proportion of sites that differs and is equivalent to 1-PID (percent identity). dif is given by the number of aligned positions (sites) whose symbols differ, divided by the number of aligned positions. dif is equivalent to the p distance defined by Nei and Zhang (2006). In dif, positions with at least one gap can be excluded (gap = FALSE). When gaps are taken as a supplementary symbol (gap = TRUE), sites with gaps in both sequences are excluded.

From Nei and Zhang (2006), the p distance, which is the proportion of sites that differ between two sequences, is estimated by:

$$p = \frac{n_d}{n},$$

where n is the number of sites and n_d is the number of sites with different symbols.

The difference score ranges from 0, for identical sequences, to 1, for completely different sequences.

Value

A single numeric value representing the difference score.

Author(s)

Julien Pele

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References

May AC (2004) Percent sequence identity: the need to be explicit. Structure 12:737-738.

Nei M and Zhang J (2006) Evolutionary Distance: Estimation. Encyclopedia of Life Sciences.

Nei M and Kumar S (2000) Molecular Evolution and Phylogenetics. *Oxford University Press*, New York.

Examples

```
# calculating the difference score between the sequences
# of CLTR1_HUMAN and CLTR2_HUMAN:
aln <- read.fasta(system.file("msa/human_gpcr.fa", package = "bios2mds"))
dif <- dif(aln$CLTR1_HUMAN, aln$CLTR2_HUMAN)
dif</pre>
```

dis

Dissimilarity score

Description

Computes the dissimilarity score between two aligned amino acid sequences, based on substitution matrices.

Usage

```
dis(seq1, seq2, sub.mat.id = "PAM250", gap = NULL)
```

Arguments

a character vector representing a first amino acid sequence.

seq2 a character vector representing a second amino acid sequence.

sub.mat.id a string of characters indicating the amino acid substitution matrix to be taken into account for calculation. This should be one of "PAM40", "PAM120", "PAM250", "BLOSUM45", "BLOSUM62", "BLOSUM80", "PAM250TM" and "PHAT". Default is PAM250. See sub.mat.

a numeric vector of length 2, indicating the penalty for initiating and extending a strand of two gaps (gap ignored, default).

Details

Grishin and Grishin (2002) developped a method to calculate the similarity score with amino acid substitution matrices.

Let s be an amino acid substitution matrix with elements s(a, b), let A be an alignment of n sequences, A_{ik} is a symbol (amino acid or gap: '-') in the site k of the sequence i. For each pair of sequences i and j from A, the following equations(1), (2), (3) and (4) are calculated as follows:

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(1) S_{ij} , called score per site, is obtained as:

$$S_{ij} = \sum_{k \in K_{ij}} s(A_{ik}, A_{jk}) / l(K_{ij})$$

where K_{ik} is the set of sites k such that $A_{ik} ! =$ '-' and $A_{jk} ! =$ '-' and $l(K_{ij})$ is the number of elements in K_{ij} .

(2) T_{ij} , called average upper limit of the score per site, is obtained as:

$$T_{ij} = 0.5 \sum_{k \in K_{ij}} (s(A_{ik}, A_{ik}) + s(A_{jk}, A_{jk})) / l(K_{ij})$$

(3) S_{ij}^{rand} , called score per site expected from random sequences, is obtained as:

$$S_{ij}^{rand} = \sum_{a=1}^{20} \sum_{b=1}^{20} f_j^i(a) f_i^j(b) s(a,b)$$

where $f_j^i(a)$ is the frequency of amino acid 'a' in i^{th} protein sequence of A over all sites in K_{ij} .

(4) V_{ij} , called normalized score (Feng and Doolittle, 1997), is obtained as:

$$V_{ij} = \frac{S_{ij} - S_{ij}^{rand}}{T_{ij} - S_{ij}^{rand}}$$

The normalized score V_{ij} ranges from 0 (for random sequences) to 1 (for identical sequences). However, for very divergent sequences, V_{ij} can become negative due to statistical errors. In this case, dis attributes 0 to negative scores.

The dissimilarity score D_{ij} between sequences i and j is obtained from the similarity score as:

$$D_{ij} = V_{ij} - 1$$

Value

A single numeric value representing the dissimilarity score.

Note

The computation time for dis is higher than for dif because of the increased algorithm complexity.

Author(s)

Julien Pele

References

Grishin VN and Grishin NV (2002) Euclidian space and grouping of biological objects. *Bioinformatics* **18**:1523-1534.

Feng DF and Doolittle RF (1997) Converting amino acid alignment scores into measures of evolutionary time: a simulation study of various relationships. *J Mol Evol* **44**:361-370.

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Examples

```
# calculating dis between the sequences of CLTR1_HUMAN and CLTR2_HUMAN:
aln <- read.fasta(system.file("msa/human_gpcr.fa", package = "bios2mds"))
dis <- dis(aln$CLTR1_HUMAN, aln$CLTR2_HUMAN)
dis</pre>
```

gpcr

Pre-analyzed G-Protein-Coupled Receptor (GPCR) data set

Description

This data set was obtained by the bios2mds analysis of the two multiple sequence alignment files provided in /msa. The MSA files were prepared as previously described by Deville et al. (2009) and updated with the July 2009 release of Uniprot (http://www.uniprot.org). They correspond to non-redundant sets of non-olfactory class A G-protein-coupled receptors (GPCRs) from *H. sapiens* and *D. melanogaster* (283 and 59 sequences, respectively). The data sets from *H. sapiens* and *D. melanogaster* constitute the active and supplementary data sets, respectively.

Usage

```
data(gpcr)
```

Format

gpcr is a named list of three elements:

dif a named list containing two distance matrices calculated from the mat.dif function (distances based on difference scores) with default parameters:

sapiens.sapiens a 283 by 283 matrix of difference scores between the 283 aligned sequences of human GPCRs.

melanogaster.sapiens a 59 by 283 matrix of difference scores between the aligned sequences of GPCRs from *H. sapiens* and *D. melanogaster* (283 and 59 sequences, respectively)..

dis a named list containing sixteen distance matrices calculated from the mat.dis function (distances based on dissimilarity scores) for the eight substitution matrices in sub.mat (other parameters by default):

sapiens.sapiens** are 283 by 283 matrices of dissimilarity scores between the 283 aligned sequences of human GPCRs.

```
sapiens.sapiens$PAM40 is calculated with PAM40.
sapiens.sapiens$PAM120 is calculated with PAM120.
sapiens.sapiens$PAM250 is calculated with PAM250.
sapiens.sapiens$BLOSUM45 is calculated with BLOSUM45.
sapiens.sapiens$BLOSUM62 is calculated with BLOSUM62.
sapiens.sapiens$BLOSUM80 is calculated with BLOSUM80.
sapiens.sapiens$PAM250TM is calculated with PAM250TM.
```

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sapiens.sapiens\$PHAT is calculated with PHAT.

melanogaster.sapiens\$* are 59 by 283 matrices of dissimilarity scores between the 283 aligned sequences from *H. sapiens* and *D. melanogaster*melanogaster.sapiens\$PAM40 is calculated with PAM40.

melanogaster.sapiens\$PAM120 is calculated with PAM120.

melanogaster.sapiens\$PAM250 is calculated with PAM250.

melanogaster.sapiens\$BLOSUM45 is calculated with BLOSUM45.

melanogaster.sapiens\$BLOSUM62 is calculated with BLOSUM62.

melanogaster.sapiens\$BLOSUM80 is calculated with BLOSUM80.

melanogaster.sapiens\$PAM250TM is calculated with PAM250TM.

mmds a typical example of metric MDS analysis with the mmds function of bios2mds using GPCRs from *H. sapiens* as active data and GPCRs from *D. melanogaster* as supplementary data with an example of clustering.

melanogaster.sapiens\$PHAT is calculated with PHAT.

Source

Deville J, Rey J and Chabbert M (2009) An indel in transmembrane helix 2 helps to trace the molecular evolution of class A G-protein-coupled receptors. *J Mol Evol* **68**: 475- 489.

Examples

```
# loading gpcr
data(gpcr)

# displaying the matrix of differences scores between GPCRs sequences
# from H. sapiens
gpcr$dif$sapiens.sapiens

# displaying the matrix of dissimilarity scores between GPCRs sequences
# from H. sapiens
gpcr$dis$sapiens.sapiens$PAM250

# displaying the matrix of dissimilarity scores between GPCRs sequences
# from H. sapiens and D. Melanogaster calculated with the BLOSUM45 matrix
gpcr$dis$melanogaster.sapiens$BLOSUM45

# displaying mmds analysis of the MSA of GPCRs from H. sapiens
# and D. Melanogaster
gpcr$mmds
```

kmeans.run

Multiple runs of K-means analysis

Description

Performs multiple runs of K-means clustering and analyzes data.

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Usage

```
kmeans.run(mat, nb.clus = 2, nb.run = 1000, iter.max = 1000,
method = "euclidean")
```

Arguments

mat	a numeric matrix representing the coordinates of the elements after metric MDS analysis.
nb.clus	a numeric value indicating the number of clusters. Default is 2.
nb.run	a numeric value indicating the number of runs. Default is 1000.
iter.max	a numeric value indicating the maximum number of iterations for K-means. Default is 1000 .
method	a string of characters to determine the distance to be used. This should be one of "euclidean", "maximum", "manhattan", "canberra", "binary", "pearson", "correlation", "spearman" or "kendall". Default is "euclidean".

Details

The aim of K-means clustering is the partition of elements into a user-provided number of clusters. Several runs of K-means analysis on the same data may return different cluster assignments because the K-means procedure attributes random initial centroids for each run. The robustness of an assignment depends on its reproducibility.

The function matchClasses from the e1071 package is used to compare the cluster assignments of the different runs and returns a score of agreement between them. The most frequent clustering solution is selected and used as a reference to assess the reproducibility of the analysis.

kmeans.run returns two lists. In either list, the clusters refer to those observed in the most frequent solution. The first list provides, for each element, the relative ratio of its assignment to each cluster in the different runs. The second list provides, for each cluster, the list of the assigned elements along with the relative assignment to this cluster in the different runs.

Value

A named list of 2 elements:

elements	a named list of elements with the relative assignment of each element to each cluster.
clusters	a named list of clusters with the elements assigned to this cluster in the most frequent solution and their relative assignment to this cluster in multiple runs.

Note

During the K-means procedure, an empty cluster can be obtained if no objects are allocated to the cluster. In kmeans.run, runs with empty clusters are discarded.

kmeans.run requires Kmeans and matchClasses functions from amap and e1071 packages, respectively.

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Author(s)

Julien Pele

Examples

```
# Clustering human GPCRs in 4 groups with 100 runs of K-means
data(gpcr)
coord <- gpcr$mmds$active.coord
kmeans.run1 <- kmeans.run(coord, nb.clus = 4, nb.run = 100)
kmeans.run1$clusters
kmeans.run1$emements</pre>
```

mat.dif

Matrices of difference scores between sequences

Description

Computes a matrix providing the distances based on the difference scores between sequences from two multiple sequence alignments.

Usage

```
mat.dif(align1, align2, gap = FALSE, aa.strict = FALSE)
```

Arguments

align1	a list of character vectors representing a first multiple sequence aligment.
align2	a list of character vectors representing a second multiple sequence aligment.
gap	a logical value indicating whether gap character should be taken as supplementary symbol (TRUE) or not (FALSE). Default is FALSE.
aa.strict	a logical value indicating whether only strict amino acids should be taken into account (TRUE) or not (FALSE). To be used only for amino acid sequences. Default is FALSE.

Details

The difference score between a sequence i from align1 and a sequence j from align2 is calculated by the function dif. To build the matrix of difference-based distances, mat.dif first compares align1 and align2, then build the matrix:

- if align1 and align2 are identical (see identical function from base package), the matrix of difference scores is symmetric (and necessarily square). Thus, mat.dif will only calculate difference scores for each pair (i, j) with i < j. Its main diagonal will contain 0 values. The matrix is equivalent to the matrix of difference scores between sequences from a single multiple sequence alignment file.
- if align1 and align2 are different, the matrix of difference scores is not square. mat.dif will calculate the difference score for every pair (i, j) from align1 and align2.

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Before using mat.dif, users must check the alignment of sequences within align1 and align2 and between align1 and align2.

Value

A named numeric matrix providing the difference-based distances between each pair of sequences from align1 and align2. The number of rows and columns is identical to the number of sequences in align1 and align2, respectively.

Note

mat.dif requires combinations function from gtools package.

Author(s)

Julien Pele

See Also

identity function from bio3d package.

Examples

```
# calculating the matrix of distances based on the difference scores
# between GPCRs sample from H. sapiens and D. melanogaster:
aln_human <- read.fasta(system.file("msa/human_gpcr.fa", package = "bios2mds"))
aln_drome <- read.fasta(system.file("msa/drome_gpcr.fa", package = "bios2mds"))
mat.dif1 <- mat.dif(aln_human[1:5], aln_drome[1:5])
mat.dif1</pre>
```

mat.dis

Matrices of dissimilarity scores between amino acid sequences

Description

Computes a matrix providing the distances based on dissimilarity scores between sequences from two multiple sequence alignments.

Usage

```
mat.dis(align1, align2, sub.mat.id = "PAM250", gap = NULL)
```

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Arguments

align1	a list of character vectors representing a first multiple sequence aligment.
align2	a list of character vectors representing a second multiple sequence aligment.
sub.mat.id	a string of characters indicating the amino acid substitution matrix used for calculation of the dissimilarity score. This should be one of "PAM40", "PAM120", "PAM250", "BLOSUM45", "BLOSUM62", "BLOSUM80", "PAM250TM" and "PHAT". The supported substitution matrices are in <pre>sub.mat</pre> . Default is PAM250.
gap	a numeric vector of length two, indicating the penalty for initiating and extending a gap. Default is NULL (gap ignored).

Details

The dissimilarity score between a sequence *i* from align1 and a sequence *j* from align2 is calculated by the dis function with an amino acid substitution matrix from sub.mat. To build the matrix of dissimilarity-based distances, mat.dis first compares align1 and align2, then build the matrix:

- if align1 and align2 are identical (see identical function from base package), the matrix of distances is symmetric (and necessarily square). Thus, mat.dis will only calculate dissimilarity scores for each pair (i, j) with i < j. Its main diagonal will contain 0 values. The matrix is equivalent to the matrix of dissimilarity between sequences from one multiple alignment file.
- if align1 and align2 are different, the matrix of dissimilarity is not square. mat.dis will calculate the dissimilarity scores for every pair (*i*, *j*) from align1 and align2.

Value

A named numeric matrix providing the dissimilarity-based distances between each pair of sequences from align1 and align2, based on the substitution matrix sub.mat.id. The number of rows and columns is identical to the number of sequences in align1 and align2, respectively.

Note

mat.dis requires combinations function from gtools package.

Author(s)

Julien Pele

Examples

```
# calculating dissimilarity distances between GPCR sequences sample from
#H. sapiens and D. melanogaster, based on the PAM250 matrix:
aln_human <- read.fasta(system.file("msa/human_gpcr.fa", package = "bios2mds"))
aln_drome <- read.fasta(system.file("msa/drome_gpcr.fa", package = "bios2mds"))
mat.dis1 <- mat.dis(aln_human[1:5], aln_drome[1:5])
mat.dis1</pre>
```

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```
# calculating dissimilarity distances between GPCRs sequences sample from
#H. sapiens and D. melanogaster, based on the BLOSUM45 matrix:
aln_human <- read.fasta(system.file("msa/human_gpcr.fa", package = "bios2mds"))
aln_drome <- read.fasta(system.file("msa/drome_gpcr.fa", package = "bios2mds"))
mat.dis1 <- mat.dis(aln_human[1:5], aln_drome[1:5], sub.mat.id = "BLOSUM45")
mat.dis1</pre>
```

mmds

Metric multidimensional scaling

Description

Performs metric MultiDimensional Scaling (MDS) analysis of active elements and projects supplementary elements onto the active space defined by active elements.

Usage

```
mmds(active, sup = NULL, pc = 3)
```

Arguments

active a numeric matrix of distances between active elements.

sup
a numeric matrix of distances between supplementary and active elements. Default is NULL.

pc
a numeric value indicating the number of principal components to be saved.

Default is 3.

Details

Metric multidimensional scaling is a statistical analysis technique aimed at analyzing a matrix of distances between 'active' elements. MDS maps the elements onto a low dimensional space (usually 2D or 3D), In this space, elements are represented by points whose respective distances best approximate the initial distance. In addition, after the metric MDS analysis of active elements, the mmds function allows projecting supplementary elements onto the active space in the context of the R environment. The active space is defined only by the MDS analysis of the active elements. The position of supplementary elements onto the active space depends only on their distances to active elements.

active and sup (if present) must have some characteristics:

- active represents the matrix of pairwise distances between active elements. The active matrix must be symmetric (square and equals to its transpose). The distances on the main diagonal must be equal to 0. The distances do not have to be Euclidean. They can just express a difference or dissimilarity score, with a 0 value between same elements and a positive value between different elements.
- sup represents the matrix of pairwise distances between supplementary (rows) and active (columns) elements and does not have to be symmetric. The number of supplementary elements may be lower or higher than the number of active objects. The names of supplementary elements must be placed in the left column of sup.

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The method for the computation of metric MDS projection of supplementary data is described by Abdi (2007). Briefly, if N is the number of active sequences and D is the N by N matrix of the squared distances computed from the active matrix, the mmds function performs the following steps:

(1) Transforms **D** into a cross-product matrix **S**:

$$S = -0.5[I - (1/N)1] \times D \times [I - (1/N)1],$$

where **I** is the *N* by *N* identity matrix and **1** is an *N* by *N* matrix of ones.

(2) Transforms S into a factor matrix F:

$$oldsymbol{F} = oldsymbol{U}oldsymbol{\Lambda}^{rac{1}{2}},$$

where U is the eigenvector matrix and Λ is the diagonal matrix of the eigenvalues, such as $S = U\Lambda U^T$, where T denotes the transposition operation.

The eigenvectors of S, also called principal components (whose number is smaller or equal to N), form the active space. F gives the coordinates of the active elements in this space.

If a sup matrix is given, the supplementary elements are projected onto the active space as described below. If N_{sup} is the number of supplementary sequences and if D_{sup} is the N_{sup} by N matrix of the squared distances, the mmds function performs the following steps:

(3) Transforms D_{sup} into a cross-product matrix S_{sup} :

$$S_{sup} = -0.5[I - (1/N)1] \times [D_{sup}^{T} - (1/N)D1_{sup}],$$

where $\mathbf{1}_{sup}$ is an N_{sup} by N matrix of ones.

(4) Transforms S_{sup} into a factor matrix F_{sup} :

$$F_{sup} = S_{sup}^{T} F \Lambda^{-1}$$
.

 F_{sup} gives the coordinates of the supplementary elements in the active space.

Value

A object of class 'mmds', which is a named list of four elements:

eigen a numeric vector of the eigenvalues.

eigen.perc a numeric vector of the relative eigenvalues (eigenvalues divided by the sum of

the absolute eigenvalues).

active.coord a named numeric matrix representing the coordinates of active elements.

sup.coord a named numeric matrix representing the coordinates of supplementary elements

on active elements.

Note

If active and/or sup do not contain names:

A tag "A" followed by an incremented number names the rows and the columns of active.

A tag "S" followed by an incremented number names the rows of sup.

The columns of sup are named as the rows of active.

Author(s)

Julien Pele and Jean-Michel Becu

References

Abdi H (2007) Metric multidimensional scaling. In N.J. Salkind (Ed.): *Encyclopedia of Measure-ment and Statistics*. Thousand Oaks (CA): Sage. pp. 598-605.

For further information on multidimentional scaling:

Takane Y, Jung S and Oshima-Takane Y (2009) Multidimensional scaling, in *Handbook of quantitative methods in psychology*, eds Millsap R, Maydeu-Olivares A (Sage Publications, London) pp. 219-242.

For further reading on projection of supplementary elements:

Gower JC (1968) adding a point to vector diagrams in multivariate analysis. *Biometrika* **55**:582-585.

Trosset MW, Pribe CE (2008) The out-of-sample problem for classical multidimensional scaling. *Computational statistics & Data analysis* **52**:4635-4642.

Pele J, Abdi H, Moreau M, Thybert D and Chabbert M (2011) Multidimensional scaling reveals the main evolutionary pathways of class A G-protein-coupled receptors. *PloS ONE* **6**:e19094.

See Also

```
cmdscale function from stats package.
dudi.pco, suprow and supcol functions from ade4 package.
PCA function from FactoMineR package.
```

Examples

```
# performing metric MDS of human GPCRs with projection of
# GPCRs from D. melanogaster as supplementary elements:
data(gpcr)
active <- gpcr$dif$sapiens.sapiens
sup <- gpcr$dif$melanogaster.sapiens
mmds1 <- mmds(active = active, sup = sup)
mmds1$active.coord</pre>
```

mmds.2D.plot

Plots the mmds coordinates onto a 2D space

Description

Displays a scatter plot of the active elements and, if present, of supplementary elements, after a metric MDS analysis.

Usage

```
mmds.2D.plot(x, title = NULL, axis = c(1, 2), xlim = NULL,
ylim = NULL, outfile.type = NULL,
outfile.name = "mmds",new.plot = TRUE, active.col = x$active.col[,3],
active.alpha = 1, sup.col = x$sup.col[,3], active.pch = 20,
sup.pch = 3, active.lab = FALSE, sup.lab = FALSE, active.cex = 2,
sup.cex = 2, active.legend.cex = 2, sup.legend.cex = 2,
active.legend.lwd = 1, sup.legend.lwd = 2, active.lwd = 1, sup.lwd = 4,
legend = TRUE, active.legend.pos = "bottomleft",
sup.legend.pos = "bottomright", active.legend.name = x$active.group[,1],
sup.legend.name = x$sup.group[,1], active.legend.col = x$active.group[,2],
sup.legend.col = x$sup.group[,2], outfile.width = NULL, outfile.height = NULL,
box.lwd = 1, cex.axis = 1, cex.lab = 1, sup.legend.text = 1,
active.legend.text = 1, legend.axis = TRUE, grid = TRUE)
```

Arguments

X	an object of class 'mmds', obtained from mmds function.
title	a string of characters representing the title of the plot. Default is "Metric MDS".
axis	a numeric vector of length two representing the principal components displayed on the plot. Default is $c(1, 2)$.
xlim	a numeric vector representing the range for the x values. Default is full range.
ylim	a numeric vector representing the range for the y values. Default is full range.
outfile.type	a string indicating the extension type of the graph outfile. Default is NULL. If not NULL, this should be one of "pdf", "tiff", "png" or "postscript". In this case, the parameter outfile.name and the forthcoming parameters are activated.
outfile.name	a string ("mmds", default) indicating the name and directory of pdf graph outfile. The extension file is added automaticaly. See outfile.type.
new.plot	a boolean indicating whether a new graphical device should be created (TRUE) or not (FALSE). Default is TRUE.
active.col	a string of characters or character vector representing the color(s) of the active elements. Default is x\$active.col[,3]. It corresponds either to the user-provided colors if the col.group function has been used previously, or to black (filled automaticaly with the mmds function).
active.alpha	a numeric value indicating the alpha value for opacity of active objects. This value must range from 0 (invisible) to 1 (full opacity). Default is 1.
sup.col	a string or character vector representing the color(s) of supplementary elements. Default is x\$sup.col[,3]. It corresponds either to the user-provided colors if the col.group function has been used previously, or to magenta (filled automatically with the mmds function).
active.pch	an integer indicating the symbol of active elements. Default is 20, corresponding to dots.
sup.pch	an integer indicating the symbol of supplementary elements. Default is 3, corresponding to crosses.

active.lab a boolean indicating whether labels of active elements should be displayed (TRUE) or not (FALSE). Default is FALSE. a boolean indicating whether labels of supplementary elements should be dissup.lab played (TRUE) or not (FALSE). Default is FALSE. a numeric value indicating the size of the active symbols. Default is 2. active.cex sup.cex a numeric value indicating the size of the supplementary symbols. Default is 2. active.legend.cex a numeric value indicating the size of active symbols in legend. Default is 2. sup.legend.cex a numeric value indicating the size of supplementary symbols in legend. Default is 2. a numeric value indicating the width of active symbols. Default is 1. active.lwd a numeric value indicating the width of supplementary symbols. Default is 4. sup.lwd active.legend.lwd a numeric value indicating the width of active symbols in legend. Default is 1. sup.legend.lwd a numeric value indicating the width of supplementary objects in legend. Default a boolean indicating whether the legend should be displayed (TRUE) or not legend (FALSE). Default is TRUE. active.legend.pos a string indicating the position of the legend for active elements. Default is "topleft". sup.legend.pos a string indicating the position of the legend for supplementary elements. Default is "topright". active.legend.name a string vector indicating the names of the active groups. Default is x\$active.group[,1]. It corresponds either to the user-provided groups if the col.group function has been used previously, or to "NoGroup" (filled automaticaly with the mmds function). sup.legend.name a string vector indicating the names of the sup groups. Default is x\$sup.group[,1]. It corresponds either to the user-provided groups if the col.group function has been used previously, or to "NoGroup" (filled automaticaly with the mmds function). active.legend.col a string vector indicating the colors of the different active groups. Default is x\$active.group[,2]. It corresponds either to the user-provided colors if the col.group function has been used previously, or to black (filled automaticaly with the mmds function). sup.legend.col a string vector indicating the colors of the different sup groups. Default is x\$sup.group[,2]. It corresponds either to the user-provided colors if the col.group

function has been used previously, or to magenta (filled automaticaly with the

mmds function).

outfile.width

a numeric value in inches indicating the width of graph outfile. Default differs by outfile.type. See pdf, png, postscript. The resolution for tiff and png figures is 150 dpi.

outfile.height

a numeric value in inches indicating the height of graph outfile. Default differs by outfile.type. The resolution for tiff and png figures is 150 dpi. See pdf, codetiff, png, postscript.

box.lwd a numeric value indicating the border width of graph box and legend box. De-

fault is 1.

cex.axis a numeric value indicating the character size for the x and y axes. Default is 1.

cex.lab a numeric value indicating the character size for the labels of the x and y axes. Default is 1.

sup.legend.text

a numeric value indicating the character size of the supplementary tag in legend. Default is 1.

active.legend.text

a numeric value indicating the character size of active tag in legend. Default is

legend.axis a boolean indicating whether axis name should be displayed (TRUE) or not

(FALSE). Default is TRUE.

grid a boolean indicating whether grid should be displayed (TRUE) or not (FALSE).

Default is TRUE.

Details

If mmds.2D.plot is used after the col.group function, the elements are colored by the color scheme provided in the .csv file (see col.group for details). If the col.group function has not been used, the default colors are black and magenta for active and supplementary elements.

mmds.2D.plot helps identify patterns in data and compare active and supplementary elements.

active.alpha argument is helpful for visualization of supplementary elements because it allows the symbols of supplementary elements to be in the foreground as compared to active elements.

Value

Produces a scatter plot on the active graphical device.

Note

mmds.2D.plot requires alpha function from ggplot2 package.

Author(s)

Julien Pele and Jean-Michel Becu

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See Also

```
plot.PCA function from FactoMineR package. png, pdf, postscript functions (default R package).
```

Examples

```
# scatter plot of human GPCRs onto the first two axes from MDS analysis
# with projection of GPCRs from D. melanogaster as supplementary elements:
data(gpcr)
active <- gpcr$dif$sapiens.sapiens
sup <- gpcr$dif$melanogaster.sapiens</pre>
mmds1 <- mmds(active = active, sup = sup)
mmds.2D.plot(mmds1, active.alpha = 0.5, active.lab = TRUE)
# with group information
data(gpcr)
active <- gpcr$dif$sapiens.sapiens
sup <- gpcr$dif$melanogaster.sapiens
mmds1 <- mmds(active = active, sup = sup)
mmds1<-col.group(mmds1, system.file("csv/human_gpcr_group.csv"
,package = "bios2mds"))
mmds1<-col.group(mmds1,system.file("csv/drome_gpcr_group.csv"
,package = "bios2mds"),data = "sup")
mmds.2D.plot(mmds1, active.alpha = 1)
```

mmds.plot

Plots a summary of the mmds results

Description

Displays one scree plot and three scatter plots of mmds results.

Usage

```
mmds.plot(x, new.plot = TRUE, pdf.file = NULL)
```

Arguments

```
an object of class 'mmds', obtained from mmds.

new.plot
a boolean indicating whether a new graphical device should be created (TRUE) or not (FALSE). Default is TRUE.

pdf.file
a string indicating the name and directory of the pdf graph outfile. Default is NULL. If this parameter is not NULL, the parameter new.plot is inactivated.
```

22 random.msa

Details

mmds.plot is a wrapper calling of both scree.plot and mmds.2D.plot. It produces a 2x2 plot with one scree plot of the relative eigenvalues, in the upper left, and three scatter plots. The three scatter plots are generated as follows:

- scatter plot of the elements on the first and second components in the upper right.
- scatter plot of the elements on the first and third components in the lower left.
- scatter plot of the elements on the second and third components in the lower right.

If object x contains supplementary elements, they are also projected onto the three scatter plots. The active and supplementary elements are represented by dots and crosses, respectively. The color.group function may be used before calling mmds.plot to color elements by user-provided groups.

Value

Produces a summary plot of the MDS analysis on the same active graphical device.

Note

The scatter plots can display supplementary objects if their coordinates are present in x input.

Author(s)

Julien Pele and Jean-Michel Becu

See Also

plot.pca function in bio3d package.

Examples

```
# summary plot of the MDS analysis of human GPCRs with projection of GPCRs from
# D. melanogaster as supplementary elements:
data(gpcr)
mmds.plot(gpcr$mmds)
```

random.msa

Random Alignment

Description

Builds a multiple sequence alignment (MSA) of random sequences.

Usage

```
random.msa(nb.seq = 100, id = "SEQ", nb.pos = 100, gap = FALSE,
aa.strict = FALSE)
```

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Arguments

nb.seq	a numeric value indicating the number of sequences in the random MSA. Default is 100.
id	a string of characters used to tag each sequence name. Default is "SEQ". An incremented number is attached to this tag to name each sequence.
nb.pos	a numeric value indicating the length of each sequence in the random MSA. Default is 100.
gap	a logical value indicating whether the gap character should be considered as a supplementary symbol (TRUE) or not (FALSE). Default is FALSE.
aa.strict	a logical value indicating whether only strict amino acids should be taken into account (TRUE) or not (FALSE). Default is FALSE.

Details

random.msa may be used to compare a reference MSA to a random MSA. The random MSA must have the same characteristics as the reference MSA (same number of sequences of same length).

A mmds procedure can be applied to the random MSA to assess the amount of variance due to random mutations in the reference MSA.

Value

A named list whose objects correspond to random sequences.

Note

The subset function is used for random selection of the amino acids. If a truly random procedure is needed, see random package.

Author(s)

Julien Pele

References

For an application of random MSA see:

Pele J, Abdi H, Moreau M, Thybert D and Chabbert M (2011) Multidimensional scaling reveals the main evolutionary pathways of class A G-protein-coupled receptors. *PLoS ONE* **6**: e19094. doi:10.1371.

See Also

permutation and synsequence functions from seginr package.

24 read.fasta

Examples

```
# generating a random sequence alignment with the same characterics
# as human GPCRs:
aln <- read.fasta(system.file("msa/human_gpcr.fa", package = "bios2mds"))
nb.seq <- length(aln)
nb.pos <- length(aln[[1]])
aln.random <- random.msa(nb.seq = nb.seq, nb.pos = nb.pos)</pre>
```

read.fasta

Reads a file in FASTA format

Description

Reads a Multiple Sequence Alignment (MSA) file in FASTA format (.fasta or .fa extension).

Usage

```
read.fasta(file, aa.to.upper = TRUE, gap.to.dash = TRUE)
```

Arguments

file	a string of characters to indicate the name of the MSA file to be read.
aa.to.upper	a logical value indicating whether amino acids should be converted to upper case (TRUE) or not (FALSE). Default is TRUE.
gap.to.dash	a logical value indicating whether the dot (.) and tilde (\sim) gap symbols should be converted to dash (-) character (TRUE) or not (FALSE). Default is TRUE.

Details

Initially, FASTA (for FAST-ALL) was the input format of the FASTA program, used for protein comparison and searching in databases. Presently, FASTA format is a standard format for biological sequences.

The FASTA formatted file of a single sequence displays:

- a single-line description beginning with a greater-than (>) symbol. The following word is the identifier.
- followed by any number of lines, representing biological sequence.

For multiple alignments, the FASTA formatted sequences are concatenated to create a multiple FASTA format.

Value

A named list whose elements correspond to sequences, in the form of character vectors.

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Note

```
For further information about FASTA format, see: http://www.ncbi.nlm.nih.gov/BLAST/fasta.shtml or http://www.ebi.ac.uk/help/formats.html
```

Author(s)

Julien Pele

References

Pearson WR and Lipman DJ (1988) Improved tools for biological sequence comparison. *Proc Natl Acad Sci U S A* 27:2444-2448.

See Also

```
read.fasta function from bio3d package.
read.fasta function from seqinr package.
read.FASTA function from aaMI package (archived).
```

Examples

```
# reading of the multiple sequence alignment of human GPCRS in FASTA format:
aln <- read.fasta(system.file("msa/human_gpcr.fa", package = "bios2mds"))</pre>
```

read.msf

Reads a multiple sequence alignment file in MSF format

Description

Reads a Multiple Sequence Alignment (MSA) file in MSF format (.msf extension).

Usage

```
read.msf(file, aa.to.upper = TRUE, gap.to.dash = TRUE)
```

Arguments

file	a string of characters to indicate the name of the MSA file to be read.
aa.to.upper	a logical value indicating whether amino acids should be converted to upper case (TRUE) or not (FALSE). Default is TRUE.
gap.to.dash	a logical value indicating whether the dot (.) and tilde (\sim) gap symbols should be converted to the dash (-) character (TRUE) or not (FALSE). Default is TRUE.

26 scree.plot

Details

Initially, Multiple Sequence Format (MSF) was the multiple sequence alignment format of the Wisconsin Package (WP) or GCG (Genetic Computer Group). This package is a suite of over 130 sequence analysis programs for database searching, secondary structure prediction or sequence alignment. Presently, numerous multiple sequence alignment editors (Jalview and GeneDoc for example) can read and write MSF files.

MSF file displays several specificities:

- a header containing sequence identifiers and characteristics (length, check and weight).
- a separator symbolized by 2 slashes (//).
- sequences of identifiers, displayed by consecutive blocks.

Value

A named list whose elements correspond to sequences, in the form of character vectors.

Note

read.msf checks the presence of duplicated identifiers in header. Sequences whose identifiers are missing in header are ignored.

Author(s)

Julien Pele

See Also

```
read.alignment function from seqinr package. read.GDoc function from aaMI package (archived).
```

Examples

```
# reading of the multiple sequence alignment of human GPCRs in MSF format:
aln <- read.msf(system.file("msa/human_gpcr.msf", package = "bios2mds"))</pre>
```

scree.plot

Plots the eingenvalues of an MDS analysis

Description

Displays a bar plot of the eigenvalues obtained by MDS

Usage

```
scree.plot(x, lab = FALSE, title = NULL, xlim = NULL,
    ylim = NULL, new.plot = TRUE, pdf.file = NULL)
```

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Arguments

X	a numeric vector representing the raw or the relative eigenvalues of each component.
lab	a boolean indicating whether bar labels should be displayed (TRUE) or not (FALSE). Default is FALSE.
title	a character string representing the title of the plot. Default is "Scree plot".
xlim	a numeric vector representing the range of the x values. Default is full range.
ylim	a numeric vector representing the range of the y values. Default is full range.
new.plot	a boolean indicating whether a new graphical device should be created (TRUE) or not (FALSE). Default is TRUE.
pdf.file	a string indicating the name and directory of the pdf graph outfile. Default is NULL. If this parameter is not NULL, the parameter new plot is inactivated.

Details

A scree plot is a method for determining the optimal number of components useful to describe the data in the context of metric MultiDimensional Scaling (MDS). The scree plot is an histogram showing the eigenvalues of each component. The relative eigenvalues express the ratio of each eigenvalue to the sum of the eigenvalues. The relative eigenvalue of a component gives the proportion of the data variance explained by this component.

The aim is to evaluate the number of components required to capture most information contained in the data. In a scree plot, the relative eigenvalues decrease when the component number increases. The 'elbow' of the plot determines the optimal number of components to describe the data (usually the components before the 'elbow').

Value

Produces a bar plot on the active graphical device.

Note

The scree plot is not an exclusive method to determine the optimal number of components. A shepard plot, which is a scatterplot of the scaled MDS distances against the original distance data, can be another solution. See shepard function from MASS package.

Author(s)

Julien Pele

See Also

```
plot.pca.scree function from bio3d package. goodness.metaMDS function from vegan package.
```

28 sil.score

Examples

```
# displaying the scree plot of the MDS analysis of human GPCRs
data(gpcr)
active <- gpcr$dif$sapiens.sapiens
mmds1 <- mmds(active, pc = 5)
scree.plot(mmds1$eigen.perc, lab = TRUE, title = "Scree plot of metric MDS")</pre>
```

sil.score

Silhouette score

Description

Computes silhouette scores for multiple runs of K-means clustering.

Usage

```
sil.score(mat, nb.clus = c(2:13), nb.run = 100, iter.max = 1000,
method = "euclidean")
```

Arguments

mat	a numeric matrix representing the coordinates of the elements.
nb.clus	a numeric vector indicating the range of the numbers of clusters. Default is $c(2:13)$.
nb.run	a numeric value indicating the number of runs. Default is 100.
iter.max	a numeric value indicating the maximum number of iterations for K-means clustering. Default is 1000 .
method	a string of characters to determine the distance measure. This should be one of "euclidean", "maximum", "manhattan", "canberra" or "binary". Default is "euclidean".

Details

Silhouettes are a general graphical aid for interpretation and validation of cluster analysis. This technique is available through the silhouette function (cluster package). In order to calculate silhouettes, two types of data are needed:

- the collection of all distances between objects. These distances are obtained from application of dist function on the coordinates of the elements in mat with argument method.
- the partition obtained by the application of a clustering technique. In sil.score context, the partition is obtained from the Kmeans function (amap package) with argument method which indicates the cluster to which each element is assigned.

For each element, a silhouette value is calculated and evaluates the degree of confidence in the assignment of the element:

• well-clustered elements have a score near 1,

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• poorly-clustered elements have a score near -1.

Thus, silhouettes indicates the objects that are well or poorly clustered. To summarize the results, for each cluster, the silhouettes values can be displayed as an **average silhouette width**, which is the mean of silhouettes for all the elements assigned to this cluster. Finally, the **overall average silhouette** width is the mean of average silhouette widths of the different clusters.

Silhouette values offer the advantage that they depend only on the partition of the elements. As a consequence, silhouettes can be used to compare the output of the same clustering algorithm applied to the same data but for different numbers of clusters. A range of numbers of clusters can be tested, with the nb.clus argument. The optimal number of clusters is reached for the maximum of the overall silhouette width. This means that the clustering algorithm reaches a strong clustering structure. However, for a given number of clusters, the cluster assignment obtained by different K-means runs can be different because the K-means procedure assigns random initial centroids for each run. It may be necessary to run the K-means procedure several times, with the nb.run argument, to evaluate the uncertainty of the results. In that case, for each number of clusters, the mean of the overall average silhouettes for nb.run K-means runs is calculated. The maximum of this core gives the optimal number of clusters.

Value

A named numeric vector representing the silhouette scores for each number of clusters.

Note

sil.score requires Kmeans and silhouette functions from amap and cluster packages, respectively.

Author(s)

Julien Pele

References

Rousseeuw PJ (1987) Silhouettes: A Graphical Aid to the Interpretation and Validation of Cluster Analysis. *Journal of Computational and Applied Mathematics*, **20**:53-65.

Lovmar L, Ahlford A, Jonsson M and Syvanen AC (2005) Silhouette scores for assessment of SNP genotype clusters. *BMC Genomics*, **6**:35.

Guy B, Vasyl P, Susmita D and Somnath D (2008) clValid: An R Package for Cluster Validation. *Journal of Statistical Software*, **25**.

See Also

connectivity and dunn functions from clValid package. silhouette function from cluster package.

30 sub.mat

Examples

```
# calculating silhouette scores for K-means clustering of human GPCRs:
data(gpcr)
active <- gpcr$dif$sapiens.sapiens
mds <- mmds(active)
sil.score1 <- sil.score(mds$active.coord, nb.clus = c(2:10), nb.run = 100, iter.max = 100)
barplot(sil.score1)</pre>
```

sub.mat

Amino acid substitution matrices

Description

Contains eight amino acid substitution matrices, imported from version 9.1 of the aaindex2 database and PAM matrix calculator of Wageningen Bioinformatics Webportal.

Usage

```
data(sub.mat)
```

Format

A named list with eight elements corresponding to a 20 by 20 named matrix. Rows and columns names correspond to the twenty strict amino acids.

Details

```
PAM40 matrix was produced by "pam" Version 1.0.7
```

PAM120 matrix was produced by "pam" Version 1.0.7

PAM250 log odds matrix for 250 PAMs (Dayhoff et al., 1978)

BLOSUM45 substitution matrix (Henikoff-Henikoff, 1992)

BLOSUM62 substitution matrix (Henikoff-Henikoff, 1992)

BLOSUM80 substitution matrix (Henikoff-Henikoff, 1992)

PAM250TM transmembrane protein exchange matrix (Jones et al., 1994)

PHAT substitution matrix built from hydrophobic and transmembrane regions of the Blocks database (Ng et al., 2000)

Source

The matrices were downloaded from the AAindex database at http://www.genome.jp/aaindex or were calculated on the PAM server at http://www.bioinformatics.nl/tools/pam.html.

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References

Kawashima S and Kanehisa M (2000) AAindex: amino acid index database. *Nucleic Acids Res* **28**:374.

Dayhoff MO, Schwartz R and Orcutt BC (1978) A model of Evolutionary Change in Proteins. Atlas of protein sequence and structure (volume 5, supplement 3 ed.). Nat. Biomed. Res. Found.. pp. 345-358.

Henikoff S and Henikoff JG (1992) Amino acid substitution matrices from protein blocks. *Proc Natl Acad Sci U S A* **89**:10915-9.

Jones DT, Taylor WR and Thornton JM (1994) A mutation data matrix for transmembrane proteins. *FEBS Lett* **339**:269-75.

Ng PC, Henikoff JG and Henikoff S (2000) PHAT: a transmembrane-specific substitution matrix. Predicted hydrophobic and transmembrane. *Bioinformatics* **16**:760-6.

Examples

```
# loading
data(sub.mat)
# displaying PAM40:
sub.mat$PAM40
```

write.mmds.pdb

Writes PDB file

Description

Writes MDS coordinates in the Protein Data Bank format for visualization with a molecular graphics viewer.

Usage

```
write.mmds.pdb(x, axis = c(1, 2, 3), file.pdb = "R.pdb")
```

Arguments

x an object of class 'mmds', obtained from mmds function.

axis a numeric vector of length three the principal components to be displayed. De-

fault is c(1, 2, 3).

file.pdb a string of characters indicating the output PDB file name.

Details

The elements can be visualized in three dimensions (3D) with a molecular viewer as Pymol or Rasmol. If \times contains active and supplementary elements, the active and supplementary elements are numbered from 1 and from 5001, respectively. If group is not NULL, the assignment of an element to a group is indicated by the chain name from A for the first group to Z when the maximum number of groups, 26, is reached.

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Value

Produces a PDB file from the MDS coordinates, with the elements numbered in the order of the MSA file and the groups corresponding to the chain numbers.

Author(s)

Julien Pele and Jean-Michel Becu

References

```
http://www.wwpdb.org/docs.html
```

See Also

write.pdb function from bio3d package.

Examples

```
# writing the first three MDS coordinates of human GPCRs in a PDB file
data(gpcr)
write.mmds.pdb(gpcr$mmds)
```

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