Package 'rDNAse'

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Index

```
dnacheck
```

Check if the DNA sequence are in the 4 default types

Description

Check if the DNA sequence are in the 4 default types

Usage

dnacheck(x)

Arguments

х

A character vector, as the input DNA sequence.

Details

This function checks if the DNA sequence types are in the 4.

Value

Logical. TRUE if all of the DNA types of the sequence are within the 4 default types. The result character vector

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

Examples

```
x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
dnacheck(x) # TRUE
dnacheck(paste(x, 'Z', sep = '')) # FALSE
```

extrDAC

Description

The Dinucleotide-based Auto Covariance Descriptor

Usage

```
extrDAC(x, index = c("Twist", "Tilt"), nlag = 2, normalization = FALSE,
customprops = NULL, allprop = FALSE)
```

Arguments

х	the input data, which should be a list or file type.
index	the physicochemical indices, it should be a list and there are 38 different physic- ochemical indices (Table 1), which the users can choose.
nlag	an integer larger than or equal to 0 and less than or equal to L-2 (L means the length of the shortest DNA sequence in the dataset). It represents the distance between two dinucleotides.
normaliztion	with this option, the final feature vector will be normalized based on the total occurrences of all kmers. Therefore, the elements in the feature vectors represent the frequencies of kmers. The default value of this parameter is False.
customprops	the users can use their own indices to generate the feature vector. It should be a dict, the key is dinucleotide (string), and its corresponding value is a list type.
allprop	all the 38 physicochemical indices will be employed to generate the feature vec- tor. Its default value is False.

Details

This function calculates the dinucleotide-based auto covariance descriptor

Value

A vector

Note

if the user defined physicochemical indices have not been normalized, it should be normalized.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

References

Dong Q, Zhou S, Guan J. A new taxonomy-based protein fold recognition approach based on autocross-covariance transformation. *Bioinformatics*, 2009, 25(20): 2655-2662.

See Also

See extrDCC and extrDACC

Examples

```
x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
extrDAC(x)
```

extrDACC

The Dinucleotide-based Auto-cross Covariance Descriptor

Description

The Dinucleotide-based Auto-cross Covariance Descriptor

Usage

```
extrDACC(x, index = c("Twist", "Tilt"), nlag = 2, normalization = FALSE,
customprops = NULL, allprop = FALSE)
```

Arguments

x	the input data, which should be a list or file type.
index	the physicochemical indices, it should be a list and there are 38 different physic- ochemical indices (Table 1), which the users can choose.
nlag	an integer larger than or equal to 0 and less than or equal to L-2 (L means the length of the shortest DNA sequence in the dataset). It represents the distance between two dinucleotides.
normaliztion	with this option, the final feature vector will be normalized based on the total occurrences of all kmers. Therefore, the elements in the feature vectors represent the frequencies of kmers. The default value of this parameter is False.
customprops	the users can use their own indices to generate the feature vector. It should be a dict, the key is dinucleotide (string), and its corresponding value is a list type.
allprop	all the 38 physicochemical indices will be employed to generate the feature vec- tor. Its default value is False.

Details

This function calculates the dinucleotide-based auto-cross covariance descriptor

extrDCC

Value

A vector

Note

if the user defined physicochemical indices have not been normalized, it should be normalized.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

References

Dong Q, Zhou S, Guan J. A new taxonomy-based protein fold recognition approach based on autocross-covariance transformation. *Bioinformatics*, 2009, 25(20): 2655-2662.

See Also

See extrDAC and extrDCC

Examples

x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
extrDACC(x)

extrDCC

The Dinucleotide-based Cross Covariance Descriptor

Description

The Dinucleotide-based Cross Covariance Descriptor

Usage

```
extrDCC(x, index = c("Twist", "Tilt"), nlag = 2, normalization = FALSE,
customprops = NULL, allprop = FALSE)
```

Arguments

х	the input data, which should be a list or file type.
index	the physicochemical indices, it should be a list and there are 38 different physic- ochemical indices (Table 1), which the users can choose.
nlag	an integer larger than or equal to 0 and less than or equal to L-2 (L means the length of the shortest DNA sequence in the dataset). It represents the distance between two dinucleotides.

normaliztion	with this option, the final feature vector will be normalized based on the total occurrences of all kmers. Therefore, the elements in the feature vectors represent the frequencies of kmers. The default value of this parameter is False.
customprops	the users can use their own indices to generate the feature vector. It should be a dict, the key is dinucleotide (string), and its corresponding value is a list type.
allprop	all the 38 physicochemical indices will be employed to generate the feature vec- tor. Its default value is False.

Details

This function calculates the dinucleotide-based cross covariance descriptor

Value

A vector

Note

if the user defined physicochemical indices have not been normalized, it should be normalized.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

References

Dong Q, Zhou S, Guan J. A new taxonomy-based protein fold recognition approach based on autocross-covariance transformation. *Bioinformatics*, 2009, 25(20): 2655-2662.

See Also

See extrDAC and extrDACC

Examples

```
x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
extrDCC(x)
```

extrPseDNC

```
The Pseudo Dinucleotide Composition Descriptor
```

Description

The Pseudo Dinucleotide Composition Descriptor

Usage

```
extrPseDNC(x, lambda = 3, w = 0.05, normalize = FALSE,
    customprops = NULL)
```

extrPseDNC

Arguments

x	the input data, which should be a list or file type.
lambda	an integer larger than or equal to 0 and less than or equal to L-2 (L means the length of the shortest sequence in the dataset). It represents the highest counted rank (or tier) of the correlation along a DNA sequence. Its default value is 3.
W	the weight factor ranged from 0 to 1. Its default value is 0.05.
normalize	with this option, the final feature vector will be normalized based on the total occurrences of all kmers. Therefore, the elements in the feature vectors represent the frequencies of kmers. The default value of this parameter is False.
customprops	the users can use their own indices to generate the feature vector. It should be a dict, the key is dinucleotide (string), and its corresponding value is a list type.

Details

This function calculates the pseudo dinucleotide composition Descriptor

Value

A vector

Note

if the user defined physicochemical indices have not been normalized, it should be normalized.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

References

Chen W, Feng P M, Lin H, et al. iRSpot-PseDNC: identify recombination spots with pseudo dinucleotide composition. *Nucleic acids research*, 2013: gks1450.

See Also

See extrPseKNC

Examples

```
x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
extrPseDNC(x)
```

extrPseKNC

Description

The Pseudo K-tupler Composition Descriptor

Usage

```
extrPseKNC(x, lambda = 1, k = 3, normalize = FALSE, w = 0.5,
customprops = NULL)
```

Arguments

х	the input data, which should be a list or file type.
lambda	an integer larger than or equal to 0 and less than or equal to L-2 (L means the length of the shortest sequence in the dataset). It represents the highest counted rank (or tier) of the correlation along a DNA sequence. Its default value is 3.
k	an integer larger than 0 represents the k-tuple. Its default value is 3.
normalize	with this option, the final feature vector will be normalized based on the total occurrences of all kmers. Therefore, the elements in the feature vectors represent the frequencies of kmers. The default value of this parameter is False.
W	the weight factor ranged from 0 to 1. Its default value is 0.05.
customprops	the users can use their own indices to generate the feature vector. It should be a dict, the key is dinucleotide (string), and its corresponding value is a list type.

Details

This function calculates the pseudo k-tupler composition Descriptor

Value

A vector

Note

if the user defined physicochemical indices have not been normalized, it should be normalized.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

References

Guo S H, Deng E Z, Xu L Q, et al. iNuc-PseKNC: a sequence-based predictor for predicting nucleosome positioning in genomes with pseudo k-tuple nucleotide composition. *Bioinformatics*, 2014: btu083.

extrTAC

See Also

See extrPseDNC

Examples

```
x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
extrPseKNC(x)
```

extrTAC

The Trinucleotide-based Auto Covariance Descriptor

Description

The Trinucleotide-based Auto Covariance Descriptor

Usage

```
extrTAC(x, index = c("Dnase I", "Nucleosome"), nlag = 2,
normaliztion = FALSE, customprops = NULL, allprop = FALSE)
```

Arguments

х	the input data, which should be a list or file type.
index	the physicochemical indices, it should be a list and there are 12 different physic- ochemical indices (Table 2), which the users can choose.
nlag	an integer larger than or equal to 0 and less than or equal to L-2 (L means the length of the shortest DNA sequence in the dataset). It represents the distance between two dinucleotides.
normaliztion	with this option, the final feature vector will be normalized based on the total occurrences of all kmers. Therefore, the elements in the feature vectors represent the frequencies of kmers. The default value of this parameter is False.
customprops	the users can use their own indices to generate the feature vector. It should be a dict, the key is dinucleotide (string), and its corresponding value is a list type.
allprop	all the 12 physicochemical indices will be employed to generate the feature vec- tor. Its default value is False.

Details

This function calculates the trinucleotide-based auto covariance Descriptor

Value

A vector

Note

if the user defined physicochemical indices have not been normalized, it should be normalized.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

See Also

See extrTCC and extrTACC

Examples

```
x = x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
extrTAC(x)
```

extrTACC

The Trinucleotide-based Auto-cross Covariance Descriptor

Description

The Trinucleotide-based Auto-cross Covariance Descriptor

Usage

```
extrTACC(x, index = c("Dnase I", "Nucleosome"), nlag = 2,
normaliztion = FALSE, customprops = NULL, allprop = FALSE)
```

Arguments

х	the input data, which should be a list or file type.
index	the physicochemical indices, it should be a list and there are 12 different physic- ochemical indices (Table 2), which the users can choose.
nlag	an integer larger than or equal to 0 and less than or equal to L-2 (L means the length of the shortest DNA sequence in the dataset). It represents the distance between two dinucleotides.
normaliztion	with this option, the final feature vector will be normalized based on the total occurrences of all kmers. Therefore, the elements in the feature vectors represent the frequencies of kmers. The default value of this parameter is False.
customprops	the users can use their own indices to generate the feature vector. It should be a dict, the key is dinucleotide (string), and its corresponding value is a list type.
allprop	all the 12 physicochemical indices will be employed to generate the feature vec- tor. Its default value is False.

Details

This function calculates the trinucleotide-based auto-cross covariance descriptor

Value

A vector

extrTCC

Note

if the user defined physicochemical indices have not been normalized, it should be normalized.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

See Also

See extrTAC and extrTCC

Examples

```
x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
extrTACC(x)
```

```
extrTCC
```

The Trinucleotide-based Cross Covariance Descriptor

Description

The Trinucleotide-based Cross Covariance Descriptor

Usage

extrTCC(x, index = c("Dnase I", "Nucleosome"), nlag = 2, customprops = NULL, normaliztion = FALSE)

Arguments

х	the input data, which should be a list or file type.
index	the physicochemical indices, it should be a list and there are 12 different physic- ochemical indices (Table 2), which the users can choose.
nlag	an integer larger than or equal to 0 and less than or equal to L-2 (L means the length of the shortest DNA sequence in the dataset). It represents the distance between two dinucleotides.
customprops	the users can use their own indices to generate the feature vector. It should be a dict, the key is dinucleotide (string), and its corresponding value is a list type.
normaliztion	with this option, the final feature vector will be normalized based on the total occurrences of all kmers. Therefore, the elements in the feature vectors represent the frequencies of kmers. The default value of this parameter is False.

Details

This function calculates the trinucleotide-based cross covariance Descriptor

Value

A vector

Note

if the user defined physicochemical indices have not been normalized, it should be normalized.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

See Also

See extrTAC and extrTACC

Examples

```
x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
extrTCC(x)
```

getGenbank

Get DNA/RNA Sequences from Genbank by GI ID

Description

Get DNA/RNA Sequences from Genbank by GI ID

Usage

```
getGenbank(id)
```

Arguments

id A character vector, as the GI ID(s).

Details

This function get DNA/RNA sequences from Genbank by GI ID(s).

Value

A list, each component contains one of the DNA/RNA sequences.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

kmer

See Also

See readFASTA for reading FASTA format files.

Examples

Network latency may slow down this example # Only test this when your connection is fast enough require(RCurl)

ids = c(2, 11)
getGenbank(ids)

kmer

The Basic Kmer Descriptor

Description

The Basic Kmer Descriptor

Usage

```
kmer(x, k = 2, upto = FALSE, normalize = FALSE, reverse = FALSE)
```

Arguments

х	the input data, which should be a list or file type.
k	the k value of kmer, it should be an integer larger than 0.
upto	generate all the kmers: 1mer, 2mer,, kmer. The output feature vector is the combination of all these kmers. The default value of this parameter is False.
normalize	with this option, the final feature vector will be normalized based on the total occurrences of all kmers. Therefore, the elements in the feature vectors represent the frequencies of kmers. The default value of this parameter is False.
reverse	make reverse complements into a single feature, The default value of this parameter is False. if reverse is True, this method returns the reverse complement kmer feature vector.

Details

This function calculates the basic kmer descriptor

Value

A vector

if the parameters normalize and upto are both True, and then the feature vector is the combination of all these normalized kmers, e.g. the combination of normalized 1-kmer and normalized 2-kmer when k=2, normalize=True, upto=True.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

References

Noble W S, Kuehn S, Thurman R, et al. Predicting the in vivo signature of human gene regulatory sequences. *Bioinformatics*, 2005, 21 Suppl 1, i338-343. Lee D, Karchin R, Beer M A. Discriminative prediction of mammalian enhancers from DNA sequence. *Genome research*. 2005, 21, 2167-2180.

See Also

See make_kmer_index

Examples

x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
kmer(x)

make_idkmer_vec The Increment Of Diversity Descriptors

Description

The Increment Of Diversity Descriptors

Usage

```
make_idkmer_vec(k = 6, x, pos, neg, upto = TRUE)
```

Arguments

k	the k value of kmer, it should be an integer larger than 0,the default value is 6.
х	the input data, which should be a list or file type.
pos	the positive source data, which should be a or type.
neg	the negative source data, which should be or type.
upto	generate all the kmers: 1mer, 2mer,, kmer. The output feature vector is the combination of all these kmers. The default value of this parameter is True

Details

This function calculates the The Basic Kmer Descriptor

Value

if upto is True, A length k * 2 named vector, k is the k value of kmer; if upto is False, A length 2 named vector

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

References

Chen W, Luo L, Zhang L. The organization of nucleosomes around splice sites. *Nucleic acids research*, 2010, 38(9): 2788-2798. Liu G, Liu J, Cui X, et al. Sequence-dependent prediction of recombination hotspots in Saccharomyces cerevisiae. *Journal of theoretical biology*, 2012, 293: 49-54.

See Also

See kmer

Examples

```
pos = readFASTA(system.file('dnaseq/pos.fasta', package = 'rDNAse'))
neg = readFASTA(system.file('dnaseq/neg.fasta', package = 'rDNAse'))
x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
make_idkmer_vec(k = 6, x, pos, neg)
```

make_kmer_index Calculate The Basic Kmer Feature Vector

Description

Calculate The Basic Kmer Feature Vector

Usage

```
make_kmer_index(k, alphabet = "ACGT")
```

Arguments

k	the k value of kmer, it should be an integer larger than 0.
alphabet	the

Details

This function calculate the basic kmer feature vector.

Value

The result character vector

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

See Also

See kmer

Examples

make_kmer_index(2, alphabet = "ACGT")

parGOSim

DNA Sequence Similarity Calculation based on Gene Ontology (GO) Similarity

Description

DNA Sequence Similarity Calculation based on Gene Ontology (GO) Similarity

Usage

```
parGOSim(golist, type = c("go", "gene"), ont = "MF", organism = "human",
measure = "Resnik", combine = "BMA")
```

Arguments

golist	A character vector, each component contains a character vector of GO terms or one Entrez Gene ID.
type	Input type of golist, 'go' for GO Terms, 'gene' for gene ID.
ont	Default is 'MF', could be one of 'MF', 'BP', or 'CC' subontologies.
organism	Default is 'human', could be one of 'anopheles', 'arabidopsis', 'bovine', 'canine', 'chicken', 'chimp', 'coelicolor', 'ecolik12', 'ecsakai', 'fly', 'human', 'malaria', 'mouse', 'pig', 'rat', 'rhesus', 'worm', 'xenopus', 'yeast' or 'zebrafish'.
measure	Default is 'Resnik', could be one of 'Resnik', 'Lin', 'Rel', 'Jiang' or 'Wang'.
combine	Default is 'BMA', could be one of 'max', 'average', 'rcmax' or 'BMA' for combining semantic similarity scores of multiple GO terms associated with DNA.

Details

This function calculates DNA sequence similarity based on Gene Ontology (GO) similarity.

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parSeqSim

Value

A n x n similarity matrix.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

See Also

See twoGOSim for calculating the GO semantic similarity between two groups of GO terms or two Entrez gene IDs. See parSeqSim for paralleled DNA similarity calculation based on Smith-Waterman local alignment.

Examples

```
## Not run:
# Be careful when testing this since it involves GO similarity computation
# and might produce unpredictable results in some environments
require(GOSemSim)
require(org.Hs.eg.db)
# by GO Terms
go1 = c('G0:0005215', 'G0:0005488', 'G0:0005515', 'G0:0005625', 'G0:0005802', 'G0:0005905') # AP4B1
go2 = c('G0:0005515', 'G0:0005634', 'G0:0005681', 'G0:0008380', 'G0:0031202') # BCAS2
go3 = c('G0:0003735', 'G0:0005622', 'G0:0005840', 'G0:0006412') # PDE4DIP
glist = list(go1, go2, go3)
gsimmat1 = parGOSim(glist, type = 'go', ont = 'CC')
print(gsimmat1)
# by Entrez gene id
genelist = list(c('150', '151', '152', '1814', '1815', '1816'))
gsimmat2 = parGOSim(genelist, type = 'gene')
print(gsimmat2)
## End(Not run)
```

parSeqSim	Parallellized DNA/RNA Sequence Similarity Calculation based on Se-
	quence Alignment

Description

Parallellized DNA/RNA Sequence Similarity Calculation based on Sequence Alignment

Usage

```
parSeqSim(dnalist, cores = 2, type = "local", submat = "BLOSUM62")
```

Arguments

dnalist	A length n list containing n DNA/RNA sequences, each component of the list is a character string, storing one DNA/RNA sequence. Unknown sequences should be represented as ''.
cores	Integer. The number of CPU cores to use for parallel execution, default is 2. Users could use the detectCores() function in the parallel package to see how many cores they could use.
type	Type of alignment, default is 'local', could be 'global' or 'local', where 'global' represents Needleman-Wunsch global alignment; 'local' represents Smith-Waterman local alignment.
submat	Substitution matrix, default is 'BLOSUM62', could be one of 'BLOSUM45', 'BLOSUM50', 'BLOSUM62', 'BLOSUM80', 'BLOSUM100', 'PAM30', 'PAM40', 'PAM70', 'PAM120', 'PAM250'.

Details

This function implemented the parallellized version for calculating DNA/RNA sequence similarity based on sequence alignment.

Value

A n x n similarity matrix.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

See Also

See twoSeqSim for DNA/RNA sequence alignment for two DNA/RNA sequences. See parGOSim for DNA/RNA similarity calculation based on Gene Ontology (GO) semantic similarity.

Examples

Be careful when testing this since it involves parallelisation
and might produce unpredictable results in some environments

```
require(Biostrings)
require(foreach)
require(doParallel)
```

```
s1 = readFASTA(system.file('dnaseq/hs.fasta', package = 'rDNA'))[[1]]
s2 = readFASTA(system.file('dnaseq/hs.fasta', package = 'rDNA'))[[2]]
s3 = readFASTA(system.file('dnaseq/hs.fasta', package = 'rDNA'))[[3]]
s4 = readFASTA(system.file('dnaseq/hs.fasta', package = 'rDNA'))[[4]]
s5 = readFASTA(system.file('dnaseq/hs.fasta', package = 'rDNA'))[[5]]
plist = list(s1, s2, s3, s4, s5)
psimmat = parSeqSim(plist, cores = 2, type = 'local', submat = 'BLOSUM62')
print(psimmat)
```

readFASTA

Description

Read DNA/RNA Sequences in FASTA Format

Usage

readFASTA(file, legacy.mode = TRUE, seqonly = FALSE)

Arguments

file	The name of the file which the sequences in fasta format are to be read from. If it does not contain an absolute or relative path, the file name is relative to the current working directory, getwd. The default here is to read the example.fasta file which is present in the protseq directory of the protr package.
legacy.mode	If set to TRUE, lines starting with a semicolon ';' are ignored. Default value is TRUE.
seqonly	If set to TRUE, only sequences as returned without attempt to modify them or to get their names and annotations (execution time is divided approximately by a factor 3). Default value is FALSE.

Details

This function reads DNA/RNA sequences in FASTA format.

Value

The result character vector

Note

Note

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

References

Pearson, W.R. and Lipman, D.J. (1988) Improved tools for biological sequence comparison. *Proceedings of the National Academy of Sciences of the United States of America*, **85**: 2444-2448

Examples

x = readFASTA(system.file('dnaseq/hs.fasta', package = 'rDNAse'))

revchars

Description

The Reverse chars

Usage

revchars(x)

Arguments

x the input data, which should be a string.

Details

This function calculates Reverse chars

Value

A vector

Note

if the user defined physicochemical indices have not been normalized, it should be normalized.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

Examples

```
x = 'GACTGAACTGCACTTTGGTTTCATATTATTTGCTC'
revchars(x)
```

twoGOSim

Description

DNA Similarity Calculation based on Gene Ontology (GO) Similarity

Usage

```
twoGOSim(id1, id2, type = c("go", "gene"), ont = "MF", organism = "human",
measure = "Resnik", combine = "BMA")
```

Arguments

id1	A character vector. length > 1: each element is a GO term; length = 1: the Entrez Gene ID.
id2	A character vector. length > 1: each element is a GO term; length = 1: the Entrez Gene ID.
type	Input type of id1 and id2, 'go' for GO Terms, 'gene' for gene ID.
ont	Default is 'MF', could be one of 'MF', 'BP', or 'CC' subontologies.
organism	Default is 'human', could be one of 'anopheles', 'arabidopsis', 'bovine', 'canine', 'chicken', 'chimp', 'coelicolor', 'ecolik12', 'ecsakai', 'fly', 'human', 'malaria', 'mouse', 'pig', 'rat', 'rhesus', 'worm', 'xenopus', 'yeast' or 'zebrafish'.
measure	Default is 'Resnik', could be one of 'Resnik', 'Lin', 'Rel', 'Jiang' or 'Wang'.
combine	Default is 'BMA', could be one of 'max', 'average', 'rcmax' or 'BMA' for combining semantic similarity scores of multiple GO terms associated with DNA.

Details

This function calculates the Gene Ontology (GO) similarity between two groups of GO terms or two Entrez gene IDs.

Value

A n x n matrix.

Author(s)

Min-feng Zhu <<wind2zhu@163.com>>

See Also

See parGOSim for DNA similarity calculation based on Gene Ontology (GO) semantic similarity. See parSeqSim for paralleled DNA similarity calculation based on Smith-Waterman local alignment.

Examples

```
## Not run:
# Be careful when testing this since it involves GO similarity computation
# and might produce unpredictable results in some environments
require(GOSemSim)
require(org.Hs.eg.db)
# by GO terms
go1 = c("GO:0004022", "GO:0004024", "GO:0004023")
go2 = c("GO:0009055", "GO:0020037")
gsim1 = twoGOSim(go1, go2, type = 'go', ont = 'MF', measure = 'Wang')
print(gsim1)
# by Entrez gene id
gene1 = '241'
gene2 = '251'
gsim2 = twoGOSim(gene1, gene2, type = 'gene', ont = 'BP', measure = 'Lin')
print(gsim2)
## End(Not run)
```

twoSeqSim

DNA/RNA Sequence Alignment for Two DNA/RNA Sequences

Description

DNA/RNA Sequence Alignment for Two DNA/RNA Sequences

Usage

```
twoSeqSim(seq1, seq2, type = "local", submat = "BLOSUM62")
```

Arguments

seq1	A character string, containing one DNA/RNA sequence.
seq2	A character string, containing another DNA/RNA sequence.
type	Type of alignment, default is 'local', could be 'global' or 'local', where 'global' represents Needleman-Wunsch global alignment; 'local' represents Smith-Waterman local alignment.
submat	Substitution matrix, default is 'BLOSUM62', could be one of 'BLOSUM45', 'BLOSUM50', 'BLOSUM62', 'BLOSUM80', 'BLOSUM100', 'PAM30', 'PAM40', 'PAM70', 'PAM120', 'PAM250'.

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twoSeqSim

Details

This function implements the sequence alignment between two DNA/RNA sequences.

Value

An Biostrings object containing the scores and other alignment information.

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See Also

See parSeqSim for paralleled pairwise DNA/RNA similarity calculation based on sequence alignment. See twoGOSim for calculating the GO semantic similarity between two groups of GO terms or two Entrez gene IDs.

Examples

Be careful when testing this since it involves sequence alignment # and might produce unpredictable results in some environments

```
require(Biostrings)
```

```
s1 = readFASTA(system.file('dnaseq/hs.fasta', package = 'rDNA'))[[1]]
s2 = readFASTA(system.file('dnaseq/hs.fasta', package = 'rDNA'))[[2]]
seqalign = twoSeqSim(s1, s2)
summary(seqalign)
print(seqalign@score)
```

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