

BiostringsTools: Interface to Tools for Biostrings (alignment, classification, database)

Michael Hahsler
Southern Methodist University

Anurag Nagar
Southern Methodist University

Abstract

There are many stand-alone tools available for Bioinformatics. This package aims at using R and the Biostrings package as the common interface for several important tools for multiple sequence alignment (*clustalw*, *kalign*), classification (RDP), sequence retrieval (BLAST) as well as database driven sequence management for 16S rRNA.

Keywords: bioinformatics, Bioconductor, biostrings, sequence alignment, sequence classification, sequence management.

1. Introduction

There are many tools available for sequence alignment and classification. Some tools are: BALibase ([Smith and Waterman 1981](#)), BLAST ([Altschul, Gish, Miller, Myers, and Lipman 1990](#)), T-Coffee ([Notredame, Higgins, and Heringa 2000](#)), MAFFT ([Katoh, Misawa, Kuma, and Miyata 2002](#)), MUSCLE ([Edgar 2004b,a](#)), Kalign ([Lassmann and Sonnhammer 2006](#)) and ClustalW2 and ClustalX2 ([Larkin, Blackshields, Brown, Chenna, McGettigan, McWilliam, Valentin, Wallace, Wilm, Lopez, Thompson, Gibson, and Higgins 2007](#)). Typically, these tools have a command-line interface and the input and output data is stored in files using various formats. Also the parameters supplied to the command-line interface are different. All this makes using and comparing several approaches time consuming and error prone. The R-based Bioconductor project ([Gentleman, Carey, Bates, and others 2004](#)) provides important infrastructure to handle and manipulate bioinformatics data. The **Biostrings** package in particular provides infrastructure for DNA, RNA and protein sequences as well as (multiple) alignments. Also algorithms for sequence alignment are included. However, for multiple sequence alignment using BLAST the user needs to export the data into a file and then run the needed tool manually and re-import the results. Also, **Biostrings** stores sets of sequences in memory and does not directly support storing and querying classification information.

In **BiostringsTools** we provide a simple interface to a growing set of popular tools. The tools are called directly from within R and no manual data export or import is needed. Currently we interface *clustalw*, *kalign*, *RDP* and *clustalw*. **BiostringsTools** also provides database backed sequence management where large amounts of sequences and classification information can be stored and used for selective and efficient sequence retrieval.

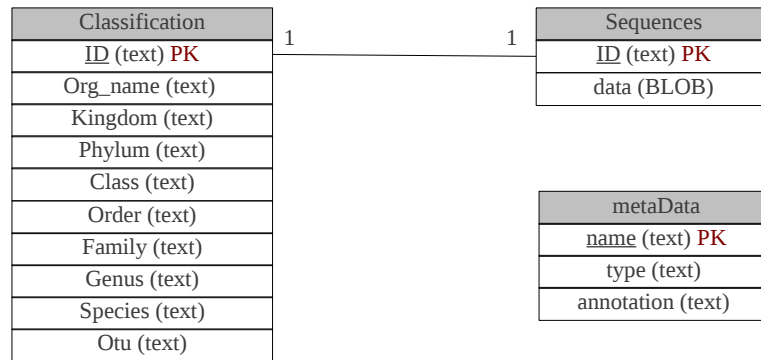


Figure 1: Entity Relationship diagram of GenDB

2. Installing Third-Party Software

BiostringsTools is designed to make installation of third-party software (RDP, clustal, kalign, MAFFT, BLAST and boxshade) easy by providing `BiostringsTools_Software_Wizard()`. With this wizard the needed software can be installed individually. This is shown in the example section.

Additional software is stored in a subdirectory of the home directory called `BiostringsTools`.

3. GenDB: Sequence storage and management

BiostringsTools provides a databases (GenDB) which can be used for efficient storage and retrieval of genetic sequences. By default the light-weight SQLite database is used, but any other compatible database such as MySQL or Oracle can also be used. Figure 1 shows the basic table layout of a GenDB instance with a table containing classification information, a table containing the sequence information and a meta data table. Each sequence we will have an entry in the classification table and an corresponding entry in the sequence table. The tables are connected by a unique sequence ID as the primary key.

GenDB is easy to use. First, we load the library into the R environment.

```
R> library(BiostringsTools)
```

To start we need to create an empty GenDB to store and organize sequences.

```
R> db<-createGenDB("example.sqlite")
```

```
R> db
```

```
Object of class GenDB with 0 sequences
```

```
DB File: example.sqlite
```

```
Tables:
```

The above command creates an empty database with a table structure similar to Figure 1 and stores it in the file `example.sqlite`. If a GenDB already exists, then it can be opened using `openGenDB()`.

The next step is to import sequences into the database by reading FASTA files. This is accomplished by function `addSequences()`. This function automatically extracts the classification information from the FASTA file's description lines. The default is to expect classification in the format used by the Greengenes project, however other meta data readers can be implemented (see manual page for `addSequences`).

The command below uses a FASTA file provided by the package, hence we use `system.file()` instead of just a string with the file name.

```
R> addSequences(db,
+   system.file("examples/Firmicutes.fasta", package="BiostringsTools"))
```

```
Read 100 sequences. Added 100 sequences.
```

After inserting the sequences, various querying and limiting functions can be used to check the data and obtain a subset of the sequences. To get a count of the number of sequences in the database, the function `nSequences()` can be used.

```
R> nSequences(db)
```

```
[1] 100
```

The function `getSequences()` returns the sequences as a vector. In the following example we get all sequences in the database and then show the first 50 bases of the first sequence.

```
R> s <- getSequences(db)
R> s
```

```
A DNAStringSet instance of length 100
```

	width	seq	names
[1]	1521	TTTGATCCTGGCTCAGG...CGGCTGGATCACCTCCT	1250
[2]	1392	ACGGGTGAGTAACGCGT...TTGGGGTGAAGTCGTAA	13651
[3]	1384	TAGTGGCGGACGGGTGA...TCGAATTTGGGTCAAGT	13652
[4]	1672	GGCGTGCCTAACACATG...TGTAACACGACTTCAT	13654
[5]	1386	ATCTCACCTCTCAATAG...CGAAGGTGGGGTTGGTG	13655
...	
[96]	1446	ATGCAAGTCGAACGGGG...GGGGCCGATGATTGGGG	13857
[97]	1511	ATCCTGGCTCAGGACGA...AGTCGTAACAAGGTAGC	13858
[98]	1544	ATCCTGGCTCAGGACGA...GGTGGATCACCTCCTTC	13860
[99]	1482	GGACGAACGCTGGCGGC...GCCGATGATTGGGGTGA	13861
[100]	1485	GACGAACGCTGGCGGCG...GAAGTCGTAACAAGGTA	13862

```
R> length(s)
```

```
[1] 100
```

```
R> s[[1]]
```

```
1521-letter "DNASTring" instance
seq: TTTGATCCTGGCTCAGGACGAACGCTGGCGG...TGTACCGGAAGGTGCGGCTGGATCACCTCCT
```

```
R> substr(s[[1]], 1, 50)
```

```
50-letter "DNASTring" instance
seq: TTTGATCCTGGCTCAGGACGAACGCTGGCGGCGTGCCTAATGCATGCAAG
```

Sequences in the database can also be filtered using classification information. For example, we can get all sequences of the genus name “Desulfosporomusa” by specifying rank and name.

```
R> s <- getSequences(db, rank="Genus", name="Desulfosporomusa")
R> s
```

A DNASTringSet instance of length 0

To obtain a single sequence, getSequences can be used with rank equal to "id" and supplying the sequence's greengenes ID as the name.

```
R> s <- getSequences(db, rank="id", name="1250")
R> s
```

```
A DNASTringSet instance of length 1
      width seq                                     names
[1] 1521 TTTGATCCTGGCTCAGGA...GCGGCTGGATCACCTCCT 1250
```

The database also stores a classification hierarchy. We can obtain the classification hierarchy used in the database with getTaxonomyNames().

```
R> getTaxonomyNames(db)

[1] "Kingdom" "Phylum" "Class" "Order" "Family" "Genus"
[7] "Species" "Otu" "Org_name" "Id"
```

To obtain all unique names stored in the database for a given rank we can use getRank().

```
R> getRank(db, rank="Order")

[1] "Thermoanaerobacterales" "Clostridiales"
```

The 100 sequences in our example data base contain organisms from different orders. We can obtain the rank name for each sequence individually by using all=TRUE or count how many sequences we have for each genus using count=TRUE.

```
R> getRank(db, rank="Genus", all=TRUE)
```

- [1] Coprothermobacter
- [2] Desulfotomaculum; Unclassified
- [3] Desulfotomaculum
- [4] Desulfotomaculum; Unclassified
- [5] Desulfotomaculum
- [6] Desulfotomaculum; Unclassified
- [7] Desulfotomaculum; Unclassified
- [8] Desulfotomaculum; Unclassified
- [9] Desulfotomaculum
- [10] Pelotomaculum; Unclassified
- [11] Desulfotomaculum; Unclassified
- [12] Desulfotomaculum; Unclassified
- [13] Pelotomaculum; Unclassified
- [14] Desulfotomaculum; Unclassified
- [15] Desulfotomaculum; Unclassified
- [16] Desulfotomaculum; Unclassified
- [17] Desulfotomaculum
- [18] Pelotomaculum; Unclassified
- [19] Desulfotomaculum
- [20] Desulfotomaculum
- [21] Desulfotomaculum
- [22] Desulfotomaculum
- [23] Desulfotomaculum; Unclassified
- [24] Desulfotomaculum
- [25] Pelotomaculum; Unclassified
- [26] Syntrophomonas; Unclassified
- [27] Syntrophomonas; Unclassified
- [28] Syntrophomonas
- [29] Syntrophomonas; Unclassified
- [30] Syntrophomonas; Unclassified
- [31] unknown
- [32] Syntrophomonas; Unclassified
- [33] Moorella
- [34] Moorella
- [35] Moorella
- [36] Moorella
- [37] Thermacetogenium
- [38] Thermaerobacter; Unclassified
- [39] Carboxydotherrmus; Unclassified
- [40] Carboxydotherrmus; Unclassified
- [41] Thermoanaerobacterium
- [42] Thermoanaerobacterium
- [43] Thermoanaerobacterium; Unclassified
- [44] Thermoanaerobacterium; Unclassified
- [45] Thermoanaerobacterium
- [46] Thermoanaerobacterium
- [47] Thermoanaerobacterium

[48] Thermoanaerobacterium
[49] Thermoanaerobacter; Unclassified
[50] Thermoanaerobacter; Unclassified
[51] Thermoanaerobacter; Unclassified
[52] Thermoanaerobacter; Unclassified
[53] Thermoanaerobacter; Unclassified
[54] Thermoanaerobacter
[55] Thermoanaerobacter; Unclassified
[56] Thermoanaerobacter; Unclassified
[57] Thermoanaerobacter; Unclassified
[58] Thermoanaerobacter; Unclassified
[59] Selenomonas
[60] Selenomonas
[61] Selenomonas
[62] Selenomonas; Unclassified
[63] Selenomonas; Unclassified
[64] Mitsuokella
[65] Selenomonas
[66] Selenomonas
[67] Selenomonas
[68] unknown
[69] Selenomonas
[70] Veillonella
[71] Veillonella
[72] Veillonella; Unclassified
[73] Veillonella
[74] Veillonella; Unclassified
[75] Dialister
[76] Dialister
[77] Dialister
[78] Desulfosporomusa; Unclassified
[79] Desulfosporomusa; Unclassified
[80] unknown
[81] unknown
[82] Desulfosporomusa; Unclassified
[83] Thermosinus; Unclassified
[84] Thermosinus; Unclassified
[85] unknown
[86] Desulfosporomusa; Unclassified
[87] Desulfosporomusa; Unclassified
[88] Desulfosporomusa; Unclassified
[89] Desulfosporomusa; Unclassified
[90] unknown
[91] unknown
[92] Acidaminococcus
[93] Acidaminococcus
[94] unknown

```

[95] unknown
[96] unknown
[97] Phascolarctobacterium
[98] Phascolarctobacterium
[99] unknown
[100] unknown
25 Levels: Acidaminococcus ... Veillonella; Unclassified

```

```
R> getRank(db, rank="Genus", count=TRUE)
```

```

              unknown
              12
Desulfotomaculum; Unclassified
              11
      Desulfotomaculum
              9
Thermoanaerobacter; Unclassified
              9
Desulfosporomusa; Unclassified
              7
      Selenomonas
              7
      Thermoanaerobacterium
              6
Syntrophomonas; Unclassified
              5
      Moorella
              4
Pelotomaculum; Unclassified
              4
      Dialister
              3
      Veillonella
              3
      Acidaminococcus
              2
Carboxydotherrmus; Unclassified
              2
      Phascolarctobacterium
              2
      Selenomonas; Unclassified
              2
Thermoanaerobacterium; Unclassified
              2
      Thermosinus; Unclassified
              2
      Veillonella; Unclassified

```

```

                2
Coprothermobacter
                1
        Mitsuokella
                1
        Syntrophomonas
                1
        Thermacetogenium
                1
Thermaerobacter; Unclassified
                1
        Thermoanaerobacter
                1

```

This information can be easily turned into a barplot showing the abundance of different orders in the data database (see Figure 3).

```

R> oldpar <- par(mar=c(12,5,5,5)) ### make space for labels
R> barplot(sort(
+   getRank(db, rank="Genus", count=TRUE, removeUnknown=TRUE),
+   decreasing=TRUE), las=2)
R> par(oldpar)

```

Filtering also works for `getRank()`. For example, we can find the genera within the order “Thermoanaerobacterales”.

```

R> getRank(db, rank="Gen",
+   whereRank="Ord", whereName="Thermo%")

[1] "Coprothermobacter"
[2] "Moorella"
[3] "Thermacetogenium"
[4] "Carboxydotherrmus; Unclassified"
[5] "Thermoanaerobacter; Unclassified"
[6] "Thermoanaerobacter"

```

Note that partial matching is performed for the ranks (i.e., from “Gen” to Genus and “Ord” to Order) and also for the name from “Thermo%” to Thermoanaerobacterales. Partial matching is available for ranks and names in most operations in **BiostringsTools**.

We can also get the complete classification hierarchy for different ranks down to individual sequences. In the following we get the classification hierarchy for genus *Thermaerobacter*, then all orders matching *Therm* and then for a list of names.

```

R> getHierarchy(db, rank="Genus", name="Thermaerobacter")

[1] Kingdom  Phylum  Class    Order    Family   Genus    Species
[8] Otu      Org_name Id
<0 rows> (or 0-length row.names)

```

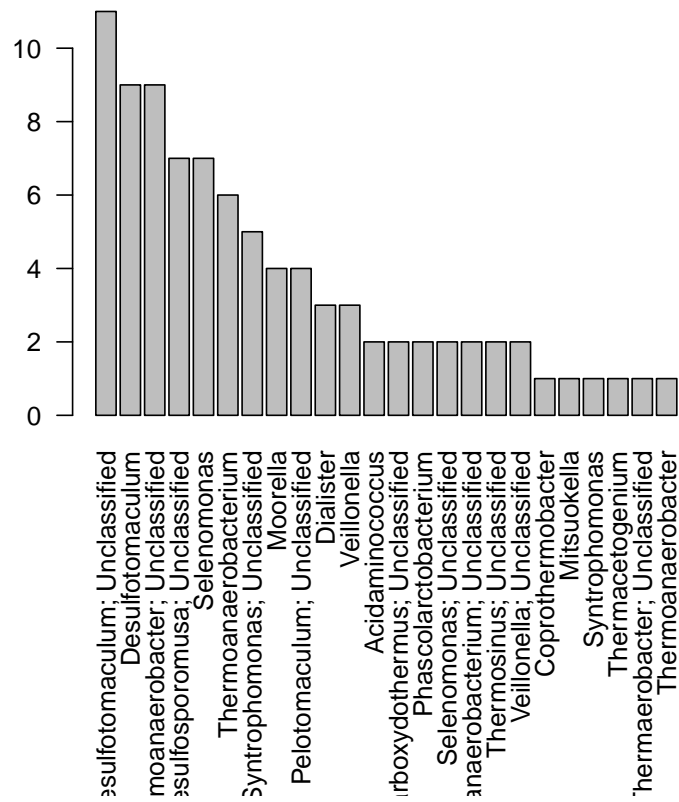



Figure 2: Abundance of different orders in the database.

```
R> getHierarchy(db, rank="Genus", name="Therm%")
```

	Kingdom	Phylum	Class	Order
1	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
2	Bacteria	Firmicutes	Clostridia	Clostridiales
3	Bacteria	Firmicutes	Clostridia	Clostridiales
4	Bacteria	Firmicutes	Clostridia	Clostridiales
5	Bacteria	Firmicutes	Clostridia	Clostridiales
6	Bacteria	Firmicutes	Clostridia	Clostridiales
7	Bacteria	Firmicutes	Clostridia	Clostridiales
8	Bacteria	Firmicutes	Clostridia	Clostridiales
9	Bacteria	Firmicutes	Clostridia	Clostridiales
10	Bacteria	Firmicutes	Clostridia	Clostridiales
11	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
12	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
13	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
14	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
15	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
16	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
17	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
18	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
19	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
20	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
21	Bacteria	Firmicutes	Clostridia	Clostridiales
22	Bacteria	Firmicutes	Clostridia	Clostridiales

	Family
1	Thermoanaerobacteraceae
2	Sulfobacillaceae
3	Thermoanaerobacterales Family III. Incertae Sedis
4	Thermoanaerobacterales Family III. Incertae Sedis
5	Thermoanaerobacterales Family III. Incertae Sedis
6	Thermoanaerobacterales Family III. Incertae Sedis
7	Thermoanaerobacterales Family III. Incertae Sedis
8	Thermoanaerobacterales Family III. Incertae Sedis
9	Thermoanaerobacterales Family III. Incertae Sedis
10	Thermoanaerobacterales Family III. Incertae Sedis
11	Thermoanaerobacteraceae
12	Thermoanaerobacteraceae
13	Thermoanaerobacteraceae
14	Thermoanaerobacteraceae
15	Thermoanaerobacteraceae
16	Thermoanaerobacteraceae
17	Thermoanaerobacteraceae
18	Thermoanaerobacteraceae
19	Thermoanaerobacteraceae
20	Thermoanaerobacteraceae
21	Veillonellaceae

22	Veillonellaceae		
	Genus		
1	Thermacetogenium		
2	Thermaerobacter; Unclassified		
3	Thermoanaerobacterium		
4	Thermoanaerobacterium		
5	Thermoanaerobacterium; Unclassified		
6	Thermoanaerobacterium; Unclassified		
7	Thermoanaerobacterium		
8	Thermoanaerobacterium		
9	Thermoanaerobacterium		
10	Thermoanaerobacterium		
11	Thermoanaerobacter; Unclassified		
12	Thermoanaerobacter; Unclassified		
13	Thermoanaerobacter; Unclassified		
14	Thermoanaerobacter; Unclassified		
15	Thermoanaerobacter; Unclassified		
16	Thermoanaerobacter		
17	Thermoanaerobacter; Unclassified		
18	Thermoanaerobacter; Unclassified		
19	Thermoanaerobacter; Unclassified		
20	Thermoanaerobacter; Unclassified		
21	Thermosinus; Unclassified		
22	Thermosinus; Unclassified		
	Species	Otu	
1	unknown	2273	
2	unknown	2189	
3	Thermoanaerobacterium saccharolyticum	2200	
4	Thermoanaerobacterium saccharolyticum	2200	
5	unknown	2199	
6	unknown	2199	
7	Thermoanaerobacterium saccharolyticum	2200	
8	Thermoanaerobacterium saccharolyticum	2200	
9	Thermoanaerobacterium saccharolyticum	2200	
10	Thermoanaerobacterium saccharolyticum	2200	
11	unknown	2274	
12	unknown	2274	
13	unknown	2274	
14	unknown	2274	
15	unknown	2274	
16	Thermoanaerobacter mathranii	2276	
17	unknown	2274	
18	unknown	2274	
19	unknown	2274	
20	unknown	2274	
21	unknown	2228	
22	unknown	2228	

	Org_name
1	AB020336.1_Thermacetogenium_phaeum_str._PB
2	AB011495.1_Thermaerobacter_marianensis
3	L09172.1_Thermoanaerobacterium_xylanolyticum_str._LXIIT
4	L09171.1_Thermoanaerobacterium_thermosulfurigenes_str._4BT
5	U75993.1_Thermoanaerobacterium_zeae_str._mel2
6	U40229.1_Thermoanaerobacterium_polysaccharolyticum
7	L09170.1_Thermoanerobacter_brockii_subsp._lactiethylicus_str._ZE-1
8	L09169.1_Thermoanaerobacterium_saccharolyticum_str._B6A-RI
9	X76743.1_Clostridium_thermoamylolyticum_str._DSM2335
10	X93359.1_Thermoanaerobacterium_aotearoense_str._JW/SL-NZ613T
11	L09160.1_Thermoanaerobacter_kivui
12	X92513.1_Thermoanaerobacter_wiegelii_str._Rt8.B1
13	U51198.1_Thermoanaerobacter_sp._str._AB11_Ad
14	Y16940.1_Thermoanaerobacter_sulfurophilus_str._L-64
15	L09167.1_Thermoanaerobacter_thermocopriae_str._JT-3T
16	Y11279.1_Thermoanaerobacter_mathranii_str._A3
17	L09164.1_Thermoanaerobacter_ethanolicus_ATCC_33223_str._39E
18	L09165.1_Thermoanaerobacter_brockii
19	L09166.1_Thermoanaerobacter_finii
20	U14330.1_Thermoanerobacter_brockii_subsp._lactiethylicus_str._SEBR_5268
21	AJ009459.1_anaerobic_TCB-transforming_consortium_clone_SJA-29
22	AJ009486.1_TCB-transforming_consortium_clone_SJA-112

	Id
1	13717
2	13721
3	13755
4	13757
5	13758
6	13759
7	13760
8	13761
9	13762
10	13763
11	13765
12	13766
13	13767
14	13768
15	13780
16	13781
17	13789
18	13790
19	13792
20	13793
21	13840
22	13842

```
R> getHierarchy(db, rank="Genus", name=c("Acid%", "Thermo%"))
```

	Kingdom	Phylum	Class	Order
1	Bacteria	Firmicutes	Clostridia	Clostridiales
2	Bacteria	Firmicutes	Clostridia	Clostridiales
3	Bacteria	Firmicutes	Clostridia	Clostridiales
4	Bacteria	Firmicutes	Clostridia	Clostridiales
5	Bacteria	Firmicutes	Clostridia	Clostridiales
6	Bacteria	Firmicutes	Clostridia	Clostridiales
7	Bacteria	Firmicutes	Clostridia	Clostridiales
8	Bacteria	Firmicutes	Clostridia	Clostridiales
9	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
10	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
11	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
12	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
13	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
14	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
15	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
16	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
17	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
18	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales
19	Bacteria	Firmicutes	Clostridia	Clostridiales
20	Bacteria	Firmicutes	Clostridia	Clostridiales
21	Bacteria	Firmicutes	Clostridia	Clostridiales
22	Bacteria	Firmicutes	Clostridia	Clostridiales

	Family
1	Thermoanaerobacterales Family III. Incertae Sedis
2	Thermoanaerobacterales Family III. Incertae Sedis
3	Thermoanaerobacterales Family III. Incertae Sedis
4	Thermoanaerobacterales Family III. Incertae Sedis
5	Thermoanaerobacterales Family III. Incertae Sedis
6	Thermoanaerobacterales Family III. Incertae Sedis
7	Thermoanaerobacterales Family III. Incertae Sedis
8	Thermoanaerobacterales Family III. Incertae Sedis
9	Thermoanaerobacteraceae
10	Thermoanaerobacteraceae
11	Thermoanaerobacteraceae
12	Thermoanaerobacteraceae
13	Thermoanaerobacteraceae
14	Thermoanaerobacteraceae
15	Thermoanaerobacteraceae
16	Thermoanaerobacteraceae
17	Thermoanaerobacteraceae
18	Thermoanaerobacteraceae
19	Veillonellaceae
20	Veillonellaceae
21	Veillonellaceae

22 Veillonellaceae

	Genus
1	Thermoanaerobacterium
2	Thermoanaerobacterium
3	Thermoanaerobacterium; Unclassified
4	Thermoanaerobacterium; Unclassified
5	Thermoanaerobacterium
6	Thermoanaerobacterium
7	Thermoanaerobacterium
8	Thermoanaerobacterium
9	Thermoanaerobacter; Unclassified
10	Thermoanaerobacter; Unclassified
11	Thermoanaerobacter; Unclassified
12	Thermoanaerobacter; Unclassified
13	Thermoanaerobacter; Unclassified
14	Thermoanaerobacter
15	Thermoanaerobacter; Unclassified
16	Thermoanaerobacter; Unclassified
17	Thermoanaerobacter; Unclassified
18	Thermoanaerobacter; Unclassified
19	Thermosinus; Unclassified
20	Thermosinus; Unclassified
21	Acidaminococcus
22	Acidaminococcus

	Species	Otu
1	Thermoanaerobacterium saccharolyticum	2200
2	Thermoanaerobacterium saccharolyticum	2200
3	unknown	2199
4	unknown	2199
5	Thermoanaerobacterium saccharolyticum	2200
6	Thermoanaerobacterium saccharolyticum	2200
7	Thermoanaerobacterium saccharolyticum	2200
8	Thermoanaerobacterium saccharolyticum	2200
9	unknown	2274
10	unknown	2274
11	unknown	2274
12	unknown	2274
13	unknown	2274
14	Thermoanaerobacter mathranii	2276
15	unknown	2274
16	unknown	2274
17	unknown	2274
18	unknown	2274
19	unknown	2228
20	unknown	2228
21	Acidaminococcus fermentans	2203
22	Acidaminococcus fermentans	2203

	Org_name
1	L09172.1_Thermoanaerobacterium_xylanolyticum_str._LXIIT
2	L09171.1_Thermoanaerobacterium_thermosulfurigenes_str._4BT
3	U75993.1_Thermoanaerobacterium_zeae_str._mel2
4	U40229.1_Thermoanaerobacterium_polysaccharolyticum
5	L09170.1_Thermoanaerobacter_brockii_subsp._lactiethylicus_str._ZE-1
6	L09169.1_Thermoanaerobacterium_saccharolyticum_str._B6A-RI
7	X76743.1_Clostridium_thermoamylolyticum_str._DSM2335
8	X93359.1_Thermoanaerobacterium_aotearoense_str._JW/SL-NZ613T
9	L09160.1_Thermoanaerobacter_kivui
10	X92513.1_Thermoanaerobacter_wiegelii_str._Rt8.B1
11	U51198.1_Thermoanaerobacter_sp._str._AB11_Ad
12	Y16940.1_Thermoanaerobacter_sulfurophilus_str._L-64
13	L09167.1_Thermoanaerobacter_thermocopriae_str._JT-3T
14	Y11279.1_Thermoanaerobacter_mathranii_str._A3
15	L09164.1_Thermoanaerobacter_ethanolicus_ATCC_33223_str._39E
16	L09165.1_Thermoanaerobacter_brockii
17	L09166.1_Thermoanaerobacter_finii
18	U14330.1_Thermoanaerobacter_brockii_subsp._lactiethylicus_str._SEBR_5268
19	AJ009459.1_anaerobic_TCB-transforming_consortium_clone_SJA-29
20	AJ009486.1_TCB-transforming_consortium_clone_SJA-112
21	X78017.1_Acidaminococcus_fermentans_str._DSM_20731
22	X77951.1_Acidaminococcus_fermentans_str._AO
Id	
1	13755
2	13757
3	13758
4	13759
5	13760
6	13761
7	13762
8	13763
9	13765
10	13766
11	13767
12	13768
13	13780
14	13781
15	13789
16	13790
17	13792
18	13793
19	13840
20	13842
21	13852
22	13853

To get individual sequences we can use again the unique sequence id.

```
R> getHierarchy(db, rank="id", name="1250")
```

	Kingdom	Phylum	Class	Order	
1	Bacteria	Firmicutes	Clostridia	Thermoanaerobacterales	
		Family	Genus	Species	Otu
1	Thermodesulfobiaceae	Coprothermobacter	unknown	2281	
					Org_name Id
1	X69335.1_Coprothermobacter	proteolyticus_str.	ATCC_35245	1250	

Finally, we can close a GenDB after we are done working with it. The database can later be reopened using `openGenDB()`.

```
R> closeGenDB(db)
```

To permanently remove the database we need to delete the file (for SQLite databases) or remove the database using the administrative tool for the database management system.

```
R> unlink("example.sqlite")
```

4. Multiple Sequence Alignment

Multiple Sequence Alignment (MSA) involves comparing and aligning more than two sequences to each other and also possibly to many others in a sequence database. The aim is to discover regions of high similarity for all the sequences taken together. The sequences are generally related such as those from the same species or same phylum.

Although, computationally complex, MSA is quite often what biologists need to identify and characterize sequences from a given group. Sequences might also share an evolutionary relationship, such as having a common ancestor. Such sequences are said to be homologous. Similarly, biologists might be interested in the similarity of genes from different organisms and want to compare their sequences. Another area of application is to find regions which are conserved for a given species or genus. Such conserved regions can be used for identification and classification of organisms.

MSA is a NP-hard problem ?? and is computationally more complex than pairwise alignment. Various algorithms that are used for pairwise alignment, such as dynamic programming, can also be used for MSA but have much greater run time requirements. To obtain results in reasonable time, various heuristics have been proposed such as Progressive Alignment, Iterative Refinement methods, and Hidden Markov Models ?. Out of these, progressive alignment is the most commonly used in many tools for MSA such as Clustal?.

Current methods for Clustal are through an online interface through the The European Bioinformatics Institute website at <http://www.ebi.ac.uk/Tools/msa/clustalw2/> or through a web-service also at the same website. There is no current tool that can be run through the command line for a batch of sequences. Our package addresses this need by providing an interface that can be used for DNA Sequences.

The **BiostringsTools** provides a rich set of functionality for MSA operations including visualization options. The commands below will illustrate that in detail.

4.1. clustalw

Install the clustal software. This has to be done only once.

```
R> BiostringsTools_Software_Wizard(clustal = TRUE)
```

BiostringsTools Software Installation Wizard for LINUX

clustalw ... installed.

We read an example FASTA file with DNA, take the first 60 nucleotides and run clustal.

```
R> dna <- readDNASTringSet(system.file("examples/DNA_example.fasta",
+   package="BiostringsTools"))
R> dna <- narrow(dna, start=1, end=60)
R> al <- clustal(dna)
R> al
```

DNAMultipleAlignment with 5 rows and 98 columns

	aln	names
[1]	-----...-GTGGCGGACGGGTGAGTAA	4403
[2]	-----...-GTGGCGGACGG-----	4404
[3]	-----...CGTGGCGCA-----	4399
[4]	AGAGTTTGATCCTGGCTCAGA...-----	1675
[5]	AGAGTTTGATTATGGCTCAGA...-----	4411

Using detail the alignment can be inspected.

```
R> detail(al)
```

Plot produces the sequence logo shown in [Figure 3](#).

```
R> plot(al, 1, 40)
```

Boxshade can also be used for producing a pdf of the alignment. [Figure 4](#) shows the result.

```
R> BiostringsTools_Software_Wizard(boxshade = TRUE)
```

BiostringsTools Software Installation Wizard for LINUX

boxshade ... installed.

```
R> boxshade(al, file="alignment.pdf")
```

Clustal can also be used for RNA and protein sequences.

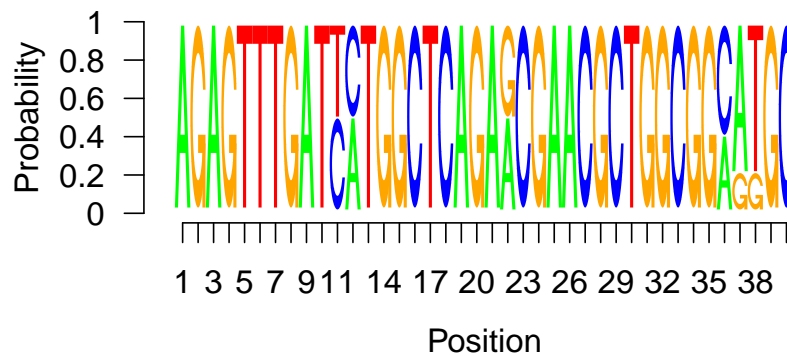


Figure 3: Sequence logo of alignment.

```

4403  ---  -----GGAATGCTNAACACATGCAAGTCGCACGG--
4404  ---  -----GCTGGCGGAATGCTTAACACATGCAAGTCGCACGGGG
4399  ---  -----GCTGGCGGCAAGCCTAACACATGCAAGTCGAACGGGG
1675  AGA  TTTGATCCTGGCTCAGAACGAACGCTGGCGGCCTGCCTAACACATGCAAGTCGAAC----
4411  AGA  TTTGATTATGGCTCAGAGCGAACGCTGGCGGCATGCTTAACACATGCAAGTCGCAC----
consensus  gctggcGGcatGcttAACACATGCAAGTCGcACgg

4403  ---  GCAGC--AATGTCA-GTGGCGGACGGGTGAGTAA
4404  ---  GTTTC--GGCCTTA-GTGGCGGACGG-----
4399  ---  ACCTTCGGGTCTTACGTGGCGCA-----
1675  AGA  -----
4411  AGA  -----
consensus  g      aa  t  a  gtggcg  a

```

Figure 4: Representation of a DNA multiple alignment using boxshade.

```
R> rna <- readRNAStringSet(system.file("examples/RNA_example.fasta",
+   package="BiostringsTools"))
R> rna
```

```
A RNAStringSet instance of length 5
      width seq                                     names
[1]  1481 AGAGUUUGAUCCUGGCUC...AGUCGUAACAAGGUAACC 1675 AB015560.1 d...
[2]  1404 GCUGGCGGCAGGCCUAAC...UAAGGUCAGCGACUGGGG 4399 D14432.1 Rho...
[3]  1426 GGAAUGCUNAACACAUGC...GGUAGCCGUAGGGGAACC 4403 X72908.1 Ros...
[4]  1362 GCUGGCGGAAUGCUUAAAC...UAGGUGUCUAGGCUAACC 4404 AF173825.1 A...
[5]  1458 AGAGUUUGAUUAUGGCUC...UCGUAACAAGGUAACCGU 4411 Y07647.2 Dre...
```

```
R> al <- clustal(rna)
R> al
```

```
RNAMultipleAlignment with 5 rows and 1500 columns
      aln                                     names
[1] -----...AAGGUAGCCGUAGGGGAACC 4403
[2] -----...----- 4404
[3] AGAGUUUGAUUAUGGCUCAGA...AAGGUAACCGU----- 4411
[4] -----...----- 4399
[5] AGAGUUUGAUCCUGGCUCAGA...AAGGUAACC----- 1675
```

```
R> aa <- readAAStringSet(system.file("examples/Protein_example.fasta",
+   package="BiostringsTools"))
R> aa
```

```
A AAStringSet instance of length 5
      width seq                                     names
[1]  170 MKKSWRRIWIFGLLSIW...DVYYLEAPFFQGRKCGGT gi|340754543|ref|...
[2]  233 MYIIWKLLFFKGENVVEH...KEEEVISVDDILKKRRE gi|340754544|ref|...
[3]  326 MKRSLSGIQPSGILHLGN...KKVQEAKEIVGLLGNIYR gi|340754545|ref|...
[4]  317 MKYYSGVDLGGTNTKIGL...VLGNEAGILGAAALFMLS gi|340754546|ref|...
[5]  337 MKKMGIILGALVLAAGLV...IVLVPSIGIDKENVAEYK gi|340754547|ref|...
```

```
R> al <- clustal(aa)
R> al
```

```
AAMultipleAlignment with 5 rows and 358 columns
      aln                                     names
[1] ---MKKSWRRIWIFGLLSIW...----- gi|340754543|ref|...
[2] ---MYIIWKLLFFKGENVVEH...----- gi|340754544|ref|...
[3] MKKMGIILGALVLAAGLVGCG...DKENVAEYK----- gi|340754547|ref|...
[4] ---MKRSLSGIQPSGILHLGN...ASKKVQEAKEIVGLLGNIYR gi|340754545|ref|...
[5] ----MKYYSGVDLGGTNTKIG...----- gi|340754546|ref|...
```

4.2. kalign

Another popular technique for MSA is based on the KAlign algorithm [Lassmann and Sonnhammer \(2005\)](#). It uses a progressive method for sequence alignment by first calculating pairwise distances between sequences and then constructing a guide tree from these pairwise alignments. The guide tree is used to progressively create the multiple sequence alignment profile. KAlign uses the Wu-Manber approximate string matching algorithm [Wu and Manber \(1992\)](#) for distance calculation. KAlign has been evaluated to be faster and more efficient than other methods [Lassmann and Sonnhammer \(2005\)](#) due to the use of the approximate string matching algorithm and efficient guide tree generation.

```
R> BiostringsTools_Software_Wizard(kalign = TRUE)
```

BiostringsTools Software Installation Wizard for LINUX

kalign ... installed.

```
R> dna <- readDNASTringSet(system.file("examples/DNA_example.fasta",
+   package="BiostringsTools"))
R> dna
```

```

A DNASTringSet instance of length 5
      width seq                                     names
[1]  1481 AGAGTTTGATCCTGGCTC...AGTCGTAACAAGGTAACC 1675 AB015560.1 d...
[2]  1404 GCTGGCGGCAGGCCTAAC...TAAGGTCAGCGACTGGGG 4399 D14432.1 Rho...
[3]  1426 GGAATGCTNAACACATGC...GGTAGCCGTAGGGGAACC 4403 X72908.1 Ros...
[4]  1362 GCTGGCGGAATGCTTAAC...TAGGTGTCTAGGCTAACC 4404 AF173825.1 A...
[5]  1458 AGAGTTTGATTATGGCTC...TCGTAACAAGGTAACCGT 4411 Y07647.2 Dre...
```

```
R> ### align the sequences
R> al <- kalign(dna)
R> al
```

DNAMultipleAlignment with 5 rows and 1502 columns

```

      aln                                     names
[1] AGAGTTTGATCCTGGCTCAGA...-----CAAGGTAAC--C 1675 AB015560.1 d...
[2] G-----...-----TGGG-----G 4399 D14432.1 Rho...
[3] G-----...GGTAGCCGTAGGGGAAC--C 4403 X72908.1 Ros...
[4] G-----...-----TAGGCTAAC--C 4404 AF173825.1 A...
[5] AGAGTTTGATTATGGCTCAGA...-----CAAGGTAACCGT 4411 Y07647.2 Dre...
```

4.3. MUSCLE

```
R> BiostringsTools_Software_Wizard(muscle = TRUE)
```

BiostringsTools Software Installation Wizard for LINUX

MUSCLE ... installed.

```
R> dna <- readDNAStringSet(system.file("examples/DNA_example.fasta",
+   package="BiostringsTools"))
R> dna
```

```

A DNAStringSet instance of length 5
      width seq                                     names
[1]  1481 AGAGTTTGGATCCTGGGCTC...AGTCGTAACAAGGTAACC 1675 AB015560.1 d...
[2]  1404 GCTGGCGGCAGGCCTAAC...TAAGGTCAGCGACTGGGG 4399 D14432.1 Rho...
[3]  1426 GGAATGCTNAACACATGC...GGTAGCCGTAGGGGAACC 4403 X72908.1 Ros...
[4]  1362 GCTGGCGGAATGCTTAAC...TAGGTGTCTAGGCTAACC 4404 AF173825.1 A...
[5]  1458 AGAGTTTGGATTATGGGCTC...TCGTAACAAGGTAACCGT 4411 Y07647.2 Dre...
```

```
R> al <- muscle(dna)
R> al
```

```

DNAMultipleAlignment with 5 rows and 1502 columns
      aln                                     names
[1] AGAGTTTGGATCCTGGGCTCAGA...AAGGTAACC----- 1675
[2] -----...----- 4399
[3] AGAGTTTGGATTATGGGCTCAGA...AAGGTAACCGT----- 4411
[4] -----...AAGGTAGCCGTAGGGGAACC 4403
[5] -----...----- 4404
```

4.4. MAFFT

```
R> BiostringsTools_Software_Wizard(mafft = TRUE)
```

BiostringsTools Software Installation Wizard for LINUX

mafft ... installed.

```
R> dna <- readDNAStringSet(system.file("examples/DNA_example.fasta",
+   package="BiostringsTools"))
R> dna
```

```

A DNAStringSet instance of length 5
      width seq                                     names
[1]  1481 AGAGTTTGGATCCTGGGCTC...AGTCGTAACAAGGTAACC 1675 AB015560.1 d...
[2]  1404 GCTGGCGGCAGGCCTAAC...TAAGGTCAGCGACTGGGG 4399 D14432.1 Rho...
[3]  1426 GGAATGCTNAACACATGC...GGTAGCCGTAGGGGAACC 4403 X72908.1 Ros...
[4]  1362 GCTGGCGGAATGCTTAAC...TAGGTGTCTAGGCTAACC 4404 AF173825.1 A...
[5]  1458 AGAGTTTGGATTATGGGCTC...TCGTAACAAGGTAACCGT 4411 Y07647.2 Dre...
```

```
R> al <- mafft(dna)
R> al
```

DNAMultipleAlignment with 5 rows and 1499 columns

	aln	names
[1]	AGAGTTTGATCCTGGCTCAGA...AAGGTAACC-----	1675
[2]	-----...-----	4399
[3]	-----...AAGGTAGCCGTAGGGGAACC	4403
[4]	-----...-----	4404
[5]	AGAGTTTGATTATGGCTCAGA...AAGGTAACCGT-----	4411

5. Classification with RDP

The Ribosomal Database Project (RDP) provides various tools and services to the scientific community for data related to 16S rRNA sequences. Among other tools, it provides a hierarchical browser and a classifier that can be used to assign sequences to taxonomies. The classifier uses a Naive Bayesian approach to quickly and accurately classify sequences. The classifier uses an alignment-free approach and compares the word frequency distribution with word size of 8 [Wang, Garrity, Tiedje, and Cole \(2007\)](#).

The RDP classifier needs to be trained first before it can be used. The default classifier comes trained with sequences from the microbial 16S rRNA gene.

First, we install RDP.

```
R> BiostringsTools_Software_Wizard(rdp = TRUE)
```

BiostringsTools Software Installation Wizard for LINUX

RDP ... installed.

5.1. Using the default RDP classifier

For this example we load some test sequences. we also shorten the names to only the sequence ID.

```
R> seq <- readRNAStringSet(system.file("examples/RNA_example.fasta",
+   package="BiostringsTools"))
R> names(seq) <- sapply(strsplit(names(seq), " "), "[", 1)
R> seq
```

A RNAStringSet instance of length 5

	width	seq	names
[1]	1481	AGAGUUUGAUCCUGGCUC...AGUCGUAACAAGGUAACC	1675
[2]	1404	GCUGGCGGCAGGCCUAAC...UAAGGUCAGCGACUGGGG	4399
[3]	1426	GGAAUGCUNAACAACAU...GGUAGCCGUAGGGGAACC	4403
[4]	1362	GCUGGCGGAAUGCUUAAC...UAGGUGUCUAGGCUAACC	4404
[5]	1458	AGAGUUUGAUUAUGGCUC...UCGUAACAAGGUAACCGU	4411

Next, we apply RDP with the default training set.

```
R> predict(rdp(), seq)
```

	rootrank	domain	phylum	class
1675	Root	Bacteria	Proteobacteria	Deltaproteobacteria
4399	Root	Bacteria	Proteobacteria	Alphaproteobacteria
4403	Root	Bacteria	Proteobacteria	Alphaproteobacteria
4404	Root	Bacteria	Proteobacteria	Alphaproteobacteria
4411	Root	Bacteria	Proteobacteria	Alphaproteobacteria
		order	family	genus
1675		<NA>	<NA>	<NA>
4399	Rhodospirillales	Rhodospirillaceae	Rhodovibrio	
4403	Rhodospirillales	Acetobacteraceae	Roseococcus	
4404	Rhodospirillales	Acetobacteraceae	Roseococcus	
4411	Rhodospirillales	Acetobacteraceae	<NA>	

5.2. Training a custom RDP classifier

RDP can be trained using `trainRDP()`.

```
R> trainingSequences <- readDNAStringSet(
+   system.file("examples/trainingSequences.fasta", package="BiostringsTools"))
R> customRDP <- trainRDP(trainingSequences, dir = "myRDP")
R> customRDP
```

RDPClassifier

Location: /home/hahsler/BiostringsTools/myRDP

```
R> testSequences <- readDNAStringSet(
+   system.file("examples/testSequences.fasta", package="BiostringsTools"))
R> predict(customRDP, testSequences)
```

	rootrank	Kingdom	Phylum	Class	Order
13811	Root	Bacteria	Firmicutes	Clostridia	Clostridiales
13813	Root	Bacteria	Firmicutes	Clostridia	Clostridiales
13678	Root	Bacteria	Firmicutes	Clostridia	Clostridiales
13755	Root	Bacteria	Firmicutes	Clostridia	Clostridiales
13661	Root	Bacteria	Firmicutes	Clostridia	Clostridiales
				Family	
13811				Veillonellaceae	
13813				Veillonellaceae	
13678				Peptococcaceae	
13755	Thermoanaerobacterales		Family III.	Incertae Sedis	
13661				Peptococcaceae	
		Genus			

```

13811      Selenomonas
13813      Selenomonas
13678      Desulfotomaculum
13755 Thermoanaerobacterium
13661      Desulfotomaculum

```

The clustom classifier is stored on disc and can be recalled anytime using `rdp()`.

```
R> customRDP <- rdp(dir = "myRDP")
```

To permanently remove the classifier use `removeRDP()`.

```
R> removeRDP(customRDP)
```

6. Sequence Retrieval with BLAST

First we install BLAST.

```
R> BiostringsTools_Software_Wizard(blast = TRUE)
```

BiostringsTools Software Installation Wizard for LINUX

BLAST ... installed.

Next, we need a BLAST database. The installation wizard can install the default 16S rRNA database into the BiostringsTools folder. Now, we can initialize BLAST with the database.

```
R> BiostringsTools_Software_Wizard(blast16S = TRUE)
```

BiostringsTools Software Installation Wizard for LINUX

16SMicrobialDB ... installed.

```
R> bl <- blast(db=~ /BiostringsTools/16SMicrobialDB/16SMicrobial")
R> bl
```

BLAST Database

Location: /home/hahsler/BiostringsTools/16SMicrobialDB/16SMicrobial

Database: 16S Microbial Sequences

14,868 sequences; 21,718,706 total bases

Date: Jun 2, 2014 12:00 AM

Longest sequence: 2,211 bases

Volumes:

/home/hahsler/BiostringsTools/16SMicrobialDB/16SMicrobial

We load again a few sequences.

```
R> seq <- readRNAStringSet(system.file("examples/RNA_example.fasta",
+   package="BiostringsTools"))
R> ## shorten names
R> names(seq) <- sapply(strsplit(names(seq), " "), "[", 1)
R> seq
```

```
A RNAStringSet instance of length 5
      width seq                                     names
[1]  1481 AGAGUUUGAUCCUGGCUC...AGUCGUAACAAGGUAACC 1675
[2]  1404 GCUGGCGGCAGGCCUAAC...UAAGGUCAGCGACUGGGG 4399
[3]  1426 GGAAUGCUNAAACAUGC...GGUAGCCGUAGGGGAACC 4403
[4]  1362 GCUGGCGGAAUGCUUAAC...UAGGUGUCUAGGCUAACC 4404
[5]  1458 AGAGUUUGAUUAUGGCUC...UCGUAACAAGGUAACCGU 4411
```

Using, predict we can BLAST the sequences.

```
R> cl <- predict(bl, seq[1,])
R> cl[1:5,]
```

	QueryID		SubjectID	Perc.Ident	Alignment.Length
1	1675	gi 559795231 ref NR_104821.1		90.82	1459
2	1675	gi 444304125 ref NR_074549.1		85.99	1249
3	1675	gi 444304125 ref NR_074549.1		94.20	69
4	1675	gi 265678428 ref NR_028730.1		82.53	1494
5	1675	gi 343201138 ref NR_041853.1		82.30	1531

	Mismatches	Gap.Openings	Q.start	Q.end	S.start	S.end	E	Bits
1	124	9	16	1468	5	1459	0e+00	1943
2	158	15	235	1478	247	1483	0e+00	1321
3	4	0	1	69	1	69	3e-22	106
4	206	34	31	1475	1	1488	0e+00	1271
5	210	40	3	1481	1	1522	0e+00	1269

7. Creating Random Sequences

Creating random sequences given letter probabilities.

```
R> seqs <- random_sequences(100, number=10, prob=c(a=.5, c=.3, g=.1, t=.1))
R> seqs
```

```
A DNAStringSet instance of length 10
      width seq                                     names
[1]   100 CCCGCAACCCCATAGAAA...AGAAAGATAAACAACA 1
[2]   100 CAAAAAAAACATAATTAA...TAGCACCTAGGGGCTCC 2
```

```
[3] 100 CACCCAAATCAACCTCCA...CAAACGCATACCCACAA 3
[4] 100 TCATAATCCTCAAAAAA...AACATTCCCATCCAAC 4
[5] 100 ACCCACACACGTAGACCA...AACCCACCTACACACCC 5
[6] 100 GGACGCGACATTCAACCA...AAATTCTGACACCCCAA 6
[7] 100 AACAAGACAAGAATAACC...GAGACAGAACAAACACA 7
[8] 100 CCAAAACACCTTAAAAAT...ACGACACACCCACGAGA 8
[9] 100 GCAACAACACATCAAAGA...CTAAATCCAAACCTGC 9
[10] 100 ATATAAACAAAAAAATT...TAATAAACTACACATAG 10
```

Creating random sequences using dinucleotides transition probabilities

```
R> prob <- matrix(runif(16), nrow=4, ncol=4, dimnames=list(DNA_BASES, DNA_BASES))
R> prob <- prob/rowSums(prob)
R> seqs <- random_sequences(100, number=10, prob=prob)
R> seqs
```

```
A DNAStringSet instance of length 10
      width seq                      names
[1] 100 CCGGGGCCCTTAGGGTCGA...GGGGGGGATTCTGGTT 1
[2] 100 TTCTTCGGGAGTCGAGGA...AGAGGCGTAATCGGTT 2
[3] 100 CGGGCCCCCTCAGGCGA...GTCTACCTATATTCTAT 3
[4] 100 AAGGTAAGGGGGGAGGG...ATTTAAGGGGGAAGCG 4
[5] 100 GGGCGTAGATAGAGTCTA...ATAACGACTATAGAGGG 5
[6] 100 TCTTCCTCGCTAGTCCCT...TAGATCAGGAAGGGGGA 6
[7] 100 TCCGAAGTGGCCCCGGG...GGAGGACCTCTATCTAG 7
[8] 100 GAGGGATCTCCCGATTA...GAGGGGAGGCGAATAGG 8
[9] 100 AGCCCTCGTCCTCTCACT...GATTCCGTACGGGGGAT 9
[10] 100 GGACAGTCCCTTTAGGTA...GGATCGAGTCTTTCTCT 10
```

Creates a set of sequences which are random mutations (with base changes, insertions and deletions) for a given DNA, RNA or AA sequence.

```
R> s <- random_sequences(100, number=1)
R> s
```

```
A DNAStringSet instance of length 1
      width seq                      names
[1] 100 GGCTTTAATCCGAGGCCA...CCTGTGGGGTGGGCACTG 1
```

```
R> ### create 10 sequences with 1 percent base changes, insertions and deletions
R> m <- mutations(s, 10, change=0.01, insertion=0.01, deletion=0.01)
R> m
```

```
A DNAStringSet instance of length 10
      width seq                      names
[1] 100 GGCTTTAATCCGAGGCCA...TGTGGGGTGCGGCACTG 1_mutation_1
[2] 101 GGCTTTAATCCGAGGCCA...TGTGGGGTGCGGCACTG 1_mutation_2
```

```
[3] 100 GGCTTTATCCGAGGCCAC...CTGTGGGGTGGGCACTG 1_mutation_3
[4] 101 GGCTTTAATCCGAGGCCA...CTGTGGGGTGGGCACTG 1_mutation_4
[5] 102 GGCTTTAATCCGAGGCCA...CTGTGGGGTGGGCACTG 1_mutation_5
[6] 100 GGCTTTAATCCGAGGCCA...CTGTGGGGTGGGCACTG 1_mutation_6
[7] 101 GGCTTTAATCCGAGGCCA...CCTGTGGGGTGGGCATG 1_mutation_7
[8] 100 GGCTTTAATCCGAGGCCA...CCTGTGGGGTGGGCACTG 1_mutation_8
[9] 99 GGCTTTAACCGAGGCCAC...CCTGTGGGGTGGGCAAT 1_mutation_9
[10] 100 GGCTTTAATCCGAGGCCA...CTGTGGGGTGGGCACTT 1_mutation_10
```

```
R> clustal(c(s,m))
```

DNAMultipleAlignment with 11 rows and 109 columns

	aln	names
[1]	GGCTTTAATCCGAGGCCACC...ACCTGTGGGGTGCGGCACTG	1_mutation_1
[2]	GGCTTTAATCCGAGGCCACC...ACCTGTGGGGTGCGGCACTG	1_mutation_2
[3]	GGCTTTA-TCCGAGGCCACC...ACCTGTGGGGTG-GGCACTG	1_mutation_3
[4]	GGCTTTAATCCGAGGCCACC...ACCTGTGGGGTG-GGCACTG	1_mutation_5
[5]	GGCTTTAATCCGAGGCCACC...ACCTGTGGGGTG-GGCACTG	1
[6]	GGCTTTAATCCGAGGCCACC...ACCTGTGGGGTG-GGCACTG	1_mutation_4
[7]	GGCTTTAATCCGAGGCCACC...ACCTGTGGGGTG-G-CACTG	1_mutation_8
[8]	GGCTTTAATCCGAGGCCACC...ACCTGTGGGGTG-GGCACTG	1_mutation_6
[9]	GGCTTTAA-CCGAGGCCACC...ACCTGTGGGGTG-GGCAAT-	1_mutation_9
[10]	GGCTTTAATCCGAGGCCACC...ACCTGTGGGGTG-GGCATG-	1_mutation_7
[11]	GGCTTTAATCCGAGGCCACC...ACCTGTGGGGTG-GGCACTT	1_mutation_10

8. Calculating Distances between Sequences

Calculating distances between sequences is important for many bioinformatics applications. The following distance metrics are available in **BiostringsTools**:

- Feature frequency profile (distFFP): A FFP is the normalized (by the number of k-mers in the sequence) count of each possible k-mer in a sequence. The distance is defined as the Jensen-Shannon divergence (JSD) between FFPs (Sims and Kim, 2011).
- Composition Vector (distCV): A CV is a vector with the frequencies of each k-mer in the sequence minus the expected frequency of random background nice obtained from a Markov Model (not implemented yet!). The cosine distance is used between CVs. (Qi et al, 2007).
- Numerical Summarization Vector (distNSV): An NSV is frequency distribution of all possible k-mers in a sequence. The Manhattan distance is used between NSVs (Nagar and Hahsler, 2013).
- Distance between sets of k-mers (distkMer): Each sequence is represented as a set of k-mers. The Jaccard (binary) distance is used between sets (number of unique shared k-mers over the total number of unique k-mers in both sequences).

- Distance based on SimRank (distSimRank): 1-simRank (see simRank).
- Edit (Levenshtein) Distance (distEdit): Edit distance between sequences.
- Distance based on alignment score (distAlignment): see stringDist in Biostrings.
- Evolutionary distances (distApe): see dist.dna in ape.

```
R> s <- mutations(random_sequences(100), 100)
```

```
R> s
```

```
A DNASTringSet instance of length 100
```

	width	seq	names
[1]	103	GCTGTAGTGTGCGCCGAG...GGACTACATTTTAGTGG	1_mutation_1
[2]	99	GCTGTAGGTCGCCAAGT...AGGACTACATTTTAGTGG	1_mutation_2
[3]	101	GCTGTAGGTCGCACAAG...GGACTACATTTTAGTGG	1_mutation_3
[4]	102	GCTGTATGTGCGCCAAGT...GGACTACATTTTAGTGG	1_mutation_4
[5]	99	GCTGTAGGTGCACAAGT...GGACTACATTTTAGTGG	1_mutation_5
...
[96]	102	GCTGTGAGGTCGCCAAG...GACTACATTTTAGTTGG	1_mutation_96
[97]	101	GCTGTAGGTCGCCAAGT...GGACTACATTTTAGTGG	1_mutation_97
[98]	101	GCTGTGGTCGCCAAGTA...GGACTACATTTTAGTGG	1_mutation_98
[99]	101	GCTGTAGGTCGCCAAGT...GGACTACATGTTAGTGG	1_mutation_99
[100]	100	GCATGTAGGTCGCCAGT...GGACTACATTTTAGTGG	1_mutation_100

```
R> ### calculate NSV distance
```

```
R> dNSV <- distNSV(s)
```

```
R> ### relationship with edit distance
```

```
R> dEdit <- distEdit(s)
```

```
R> df <- data.frame(dNSV=as.vector(dNSV), dEdit=as.vector(dEdit))
```

```
R> plot(sapply(df, jitter), cex=.1)
```

```
R> ### add lower bound (2*k, for Manhattan distance)
```

```
R> abline(0,1/(2*3), col="red", lwd=2)
```

```
R> ### add regression line
```

```
R> abline(lm(dEdit~dNSV, data=df), col="blue", lwd=2)
```

```
R> ### check correlation
```

```
R> cor(dNSV,dEdit)
```

```
[1] 0.8336
```

9. Conclusion

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Affiliation:

Michael Hahsler
Engineering Management, Information, and Systems
Lyle School of Engineering
Southern Methodist University
P.O. Box 750123
Dallas, TX 75275-0123
E-mail: mhahsler@lyle.smu.edu
URL: <http://lyle.smu.edu/~mhahsler>

Anurag Nagar
Computer Science and Engineering
Lyle School of Engineering
Southern Methodist University
P.O. Box 750122
Dallas, TX 75275-0122
E-mail: anagar@smu.edu