# MultipleAlignment Objects

# Marc Carlson Bioconductor Core Team Fred Hutchinson Cancer Research Center Seattle, WA

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### 1 Introduction

The DNAMultipleAlignment RNAMultipleAlignment and AAMultipleAlignment classes allow users to represent groups of aligned DNA, RNA or amino acid sequences as a single object. The frame of reference for aligned sequences is static, so manipulation of these objects is confined to be non-destructive. In practice, this means that these objects contain slots to mask ranges of rows and columns on the original sequence. These masks are then respected by methods that manipulate and display the objects, allowing the user to remove or expose columns and rows without invalidating the original alignment.

# 2 Creation and masking

To create a *MultipleAlignment*, call the appropriate read function to read in and parse the original alignment. There are functions to read clustaW, Phylip and Stolkholm data formats.

```
"Phylip.txt",
                            package="Biostrings"),
                    format="phylip")
 Rows can be renamed with rownames.
> rownames(origMAlign)
[1] "gi|84452153|ref|NM_002449.4|"
                          "gi|208431713|ref|NM_001135625."
[3] "gi|118601823|ref|NM_001079614." "gi|114326503|ref|NM_013601.2|"
[5] "gi|119220589|ref|NM_012982.3|" "gi|148540149|ref|NM_001003098."
[7] "gi|45383056|ref|NM_204559.1|"
                          "qi|213515133|ref|NM 001141603."
> rownames(origMAlign) <- c("Human", "Chimp", "Cow", "Mouse", "Rat",
                    "Dog", "Chicken", "Salmon")
> origMAlign
DNAMultipleAlignment with 8 rows and 2343 columns
[2] ----- Chimp
[3] ----- Cow
[4] ----- Mouse
[5] ----- Rat
[6] ----- Dog
[7] ----- CGGCTCCGCAGC...- Chicken
[8] GGGGGAGACTTCAGAAGTTGTTGTCC...---- Salmon
 To see a more detailed version of your MultipleAlignment object, you can use the detail method, which will show
the details of the alignment interleaved and without the rows and columns that you have masked out.
> detail(origMAlign)
 Applying masks is a simple matter of specifying which ranges to hide.
> maskTest <- origMAlign
> rowmask(maskTest) <- IRanges(start=1,end=3)</pre>
> rowmask(maskTest)
NormalIRanges object with 1 range and 0 metadata columns:
      start end width
    <integer> <integer> <integer>
 [1]
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
[2] ###################### Chimp
[3] #################### Cow
[4] ----- Mouse
[5] ----- Rat
[6] ----- Dog
[7] ----- CGGCTCCGCAGC...- Chicken
[8] GGGGGAGACTTCAGAAGTTGTTGTCC...---- Salmon
```

```
> colmask(maskTest) <- IRanges(start=c(1,1000),end=c(500,2343))
> colmask(maskTest)
NormalIRanges object with 2 ranges and 0 metadata columns:
      start
                  width
             end
   <integer> <integer> <integer>
             500
                   500
 [1]
       1
 [2]
      1000
             2343
                   1344
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
   aln
                                     names
[2] ####################### Chimp
[3] ##################### Cow
[4] ############### Mouse
[5] ##################### Rat
[6] ################# Doq
[7] ########################### Chicken
[8] ################### Salmon
 Remove row and column masks by assigning NULL:
> rowmask(maskTest) <- NULL
> rowmask(maskTest)
NormalIRanges object with 0 ranges and 0 metadata columns:
          end
                width
 <integer> <integer> <integer>
> colmask(maskTest) <- NULL
> colmask(maskTest)
NormalIRanges object with 0 ranges and 0 metadata columns:
    start end
               width
 <integer> <integer> <integer>
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
[2] ----- Chimp
[3] ----- Cow
[4] ----- Mouse
[5] ----- Rat
[6] ----- Dog
[7] -----CGGCTCCGCAGC...---- Chicken
[8] GGGGGAGACTTCAGAAGTTGTTCC...-Salmon
```

When setting a mask, you might want to specify the rows or columns to keep, rather than to hide. To do that, use the *invert* argument. Taking the above example, we can set the exact same masks as before by specifying their inverse and using invert=TRUE.

```
> rowmask(maskTest, invert=TRUE) <- IRanges(start=4,end=8)</pre>
> rowmask(maskTest)
NormalIRanges object with 1 range and 0 metadata columns:
            end
                  width
   <integer> <integer> <integer>
 [1]
       1
           3
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
                                    names
[1] ########################### Human
[2] ######################### Chimp
[3] ################### Cow
[4] ----- Mouse
[5] -----
[6] ----- Dog
[7] ----- CGGCTCCGCAGC...- Chicken
[8] GGGGGAGACTTCAGAAGTTGTTCC...- Salmon
> colmask(maskTest, invert=TRUE) <- IRanges(start=501,end=999)</pre>
> colmask(maskTest)
NormalIRanges object with 2 ranges and 0 metadata columns:
             end
                  width
     start
   <integer> <integer> <integer>
 [1]
        1
            500
                   500
            2343
                  1344
 [2]
      1000
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
[4] ##################### Mouse
[5] ###################### Rat
[6] #################### Doa
[7] ########################## Chicken
[8] ##################### Salmon
```

In addition to being able to invert these masks, you can also choose the way in which the ranges you provide will be merged with any existing masks. The *append* argument allows you to specify the way in which new mask ranges will interact with any existing masks. By default, these masks will be the "union" of the new mask and any existing masks, but you can also specify that these masks be the mask that results from when you "intersect" the current mask and the new mask, or that the new mask simply "replace" the current mask. The *append* argument can be used in combination with the *invert* argument to make things even more interesting. In this case, the inversion of the mask will happen before it is combined with the existing mask. For simplicity, I will only demonstrate this on *rowmask*, but it also works for *colmask*. Before we begin, lets set the masks back to being NULL again.

```
> ## 1st lets null out the masks so we can have a fresh start.
> colmask(maskTest) <- NULL
> rowmask(maskTest) <- NULL</pre>
```

Then we can do a series of examples, starting with the default which uses the "union" value for the *append* argument.

```
> ## Then we can demonstrate how the append argument works
> rowmask(maskTest) <- IRanges(start=1,end=3)</pre>
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
[2] ########################### Chimp
[3] #################### Cow
[4] ----- Mouse
[5] ----- Rat
[6] ----- Dog
[7] ----- CGGCTCCGCAGC...- Chicken
[8] GGGGGAGACTTCAGAAGTTGTTGTCC...----- Salmon
> rowmask(maskTest,append="intersect") <- IRanges(start=2,end=5)</pre>
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
                               names
[2] ###################### Chimp
[3] ################### Cow
[4] ----- Mouse
[5] ----- Rat
[6] ----- Dog
[7] ----- CGGCTCCGCAGC...- Chicken
[8] GGGGGAGACTTCAGAAGTTGTTCC...- Salmon
> rowmask(maskTest,append="replace") <- IRanges(start=5,end=8)</pre>
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
  aln
                               names
[2] ----- Chimp
[3] ----- Cow
[4] ----- Mouse
[5] #################### Rat
[6] ################### Doq
[7] ##################### Chicken
[8] #################### Salmon
> rowmask(maskTest,append="replace",invert=TRUE) <- IRanges(start=5,end=8)</pre>
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
  aln
                               names
[1] ######################## Human
```

```
[3] ################ Cow
[4] ############## Mouse
[5] ----- Rat
[6] ----- Dog
[7] ----- CGGCTCCGCAGC...--- Chicken
[8] GGGGGAGACTTCAGAAGTTGTTCC...-Salmon
> rowmask(maskTest,append="union") <- IRanges(start=7,end=8)</pre>
> maskTest
DNAMultipleAlignment with 8 rows and 2343 columns
                            names
[1] ###################### Human
[3] ################### Cow
[4] ############### Mouse
[5] ----- Rat
[7] ##################### Chicken
[8] ###################### Salmon
```

The function maskMotif works on *MultipleAlignment* objects too, and takes the same arguments that it does elsewhere. maskMotif is useful for masking occurances of a string from columns where it is present in the consensus sequence.

- > tataMasked <- maskMotif(origMAlign, "TATA")</pre>
- > colmask(tataMasked)

NormalIRanges object with 5 ranges and 0 metadata columns:

	start	end	width
	<integer></integer>	<integer></integer>	<integer></integer>
[1]	811	814	4
[2]	1180	1183	4
[3]	1186	1191	6
[4]	1204	1207	4
[5]	1218	1221	4

maskGaps also operates on columns and will mask collumns based on the fraction of each column that contains gaps *min.fraction* along with the width of columns that contain this fraction of gaps *min.block.width*.

```
> autoMasked <- maskGaps(origMAlign, min.fraction=0.5, min.block.width=4)
> autoMasked
```

DNAMultipleAlignment with 8 rows and 2343 columns

Sometimes you may want to cast your *MultipleAlignment* to be a matrix for usage eslewhere. *as.matrix* is supported for these circumstances. The ability to convert one object into another is not very unusual so why mention it? Because when you cast your object, the masks WILL be considered so that the masked rows and columns will be left out of the matrix object.

```
> full = as.matrix(origMAlign)
> dim(full)
[1]    8 2343
> partial = as.matrix(autoMasked)
> dim(partial)
[1]    8 1143
```

One example of where you might want to use *as.matrix* is when using the ape package. For example if you needed to use the *dist.dna* function you would want to use *as.matrix* followed by as .alignment and then the as .DNAbin to create a DNAbin object for the dist.dna.

## 3 Analytic utilities

Once you have masked the sequence, you can then ask questions about the properties of that sequence. For example, you can look at the alphabet frequency of that sequence. The alphabet frequency will only be for the masked sequence.

> alphabetFrequency(autoMasked)

```
G
                  TMRWSYKVHDBN
          C
[1,] 260 351 296 218 0 0 0 0 0 0 0 0 0 0
                           0 0 0 0 0 0 3 339 0 0
[2,] 171 271 231 128 0
                     0
                       0 0
[3,] 277 360 275 209 0 0 0 0
                           0
                             0
                               0 0 0 0 0
[4,] 265 343 277 226 0 0 0 0 0 0 0 0 0 0
                                          32 0 0
[5,] 251 345 287 229 0 0 0 0 0 0 0 0 0 0
[6,] 160 285 241 118 0 0 0 0 0 0 0 0 0 0 339 0 0
[7,] 224 342 273 190 0 0 0 0 0 0 0 0 0 0 114 0 0
[8,] 268 289 273 262 0 0 0 0 0 0 0 0 0 0 51 0 0
```

You can also calculate a consensus matrix, extract the consensus string or look at the consensus views. These methods too will all consider the masking when you run them.

> consensusMatrix(autoMasked, baseOnly=TRUE)[, 84:90]

```
[,1] [,2] [,3] [,4] [,5] [,6] [,7]
Α
           Λ
                  3
                        Λ
                               3
                                      \cap
           3
С
                  0
                        0
                                      1
                                            1
G
           1
                  2
                         4
                               1
                                      2
                                            0
                                                   0
                                                   0
           1
                  0
                        1
                               0
                                     1
                                            1
           3
                  3
                               3
                                            3
                                                   3
other
```

- > substr(consensusString(autoMasked),80,130)
- [1] "####CRGABAMGTCA-YRGCTTCTCYGTSCAWAGGCRRTGRCYTGTTYTCG"
- > consensusViews(autoMasked)

#### Cluster Dendrogram

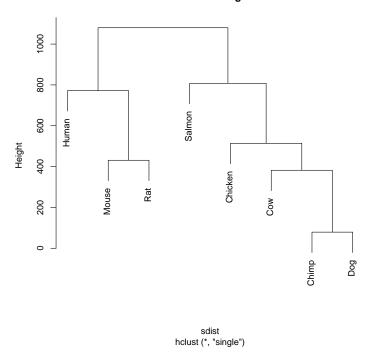


Figure 1: Funky tree produced by using unmasked strings.

```
Views on a 2343-letter BString subject
subject: ------VWVMKYYBSRST...-----
views:
           end width
     start
           325
                 242 [CRGABAMGTCA-YRGCTTCTCYGTSC...GCGSCTCSGCYGGSGCYRYCCTGSGG]
  [1]
        84
       330 332
                   3 [CCR]
  [3]
       338 1191
                 854 [CTGCTGCTGYCGGGVCACGGCGYYCG...TGMTAGTTTTTATGTATAAATATATA]
  [4]
      1198 1241
                  44 [ATAAAATATAAKAC--TTTTTATAYRSCARATGTAAAAATTCAA]
```

You can also cluster the alignments based on their distance to each other. Because you must pass in a DNAS-tringSet, the clustering will also take into account the masking. So for example, you can see how clustering the unmasked DNAMultipleAlignment will draw a funky looking tree.

```
> sdist <- stringDist(as(origMAlign,"DNAStringSet"), method="hamming")
> clust <- hclust(sdist, method = "single")
> pdf(file="badTree.pdf")
> plot(clust)
> dev.off()
null device
```

But, if we use the gap-masked DNAMultipleAlignment, to remove the long uninformative regions, and then make our plot, we can see the real relationships.

#### Cluster Dendrogram

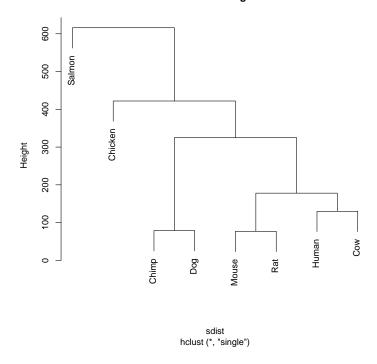


Figure 2: A tree produced by using strings with masked gaps.

```
> sdist <- stringDist(as(autoMasked,"DNAStringSet"), method="hamming")
> clust <- hclust(sdist, method = "single")</pre>
> pdf(file="goodTree.pdf")
> plot(clust)
> dev.off()
null device
          1
> fourgroups <- cutree(clust, 4)
> fourgroups
  Human
          Chimp
                     Cow
                           Mouse
                                      Rat
                                              Dog Chicken
                                                           Salmon
                                                         3
      1
                       1
                               1
                                        1
```

In the "good" plot, the Salmon sequence is once again the most distant which is what we expect to see. A closer examination of the sequence reveals that the similarity between the mouse, rat and human sequences was being inflated by virtue of the fact that those sequences were simply much longer (had more information than) the other species represented. This is what caused the "funky" result. The relationship between the sequences in the funky tree was being driven by extra "length" in the rodent/mouse/human sequences, instead of by the similarity of the conserved regions.

## 4 Exporting to file

One possible export option is to write to fasta files If you need to write your *MultipleAlignment* object out as a fasta file, you can cast it to a *DNAStringSet* and then write it out as a fasta file like so:

```
> DNAStr = as(origMAlign, "DNAStringSet")
> writeXStringSet(DNAStr, file="myFile.fa")
```

One other format that is of interest is the Phylip format. The Phylip format stores the column masking of your object as well as the sequence that you are exporting. So if you have masked the sequence and you write out a Phylip file, this mask will be recorded into the file you export. As with the fasta example above, any rows that you have masked out will be removed from the exported file.

```
> write.phylip(phylipMAlign, filepath="myFile.txt")
```

#### 5 Session Information

All of the output in this vignette was produced under the following conditions:

```
> sessionInfo()
R version 4.2.1 (2022-06-23)
Platform: x86_64-pc-linux-gnu (64-bit)
Running under: Ubuntu 20.04.5 LTS
Matrix products: default
       /home/biocbuild/bbs-3.16-bioc/R/lib/libRblas.so
LAPACK: /home/biocbuild/bbs-3.16-bioc/R/lib/libRlapack.so
locale:
 [1] LC_CTYPE=en_US.UTF-8
                              LC_NUMERIC=C
 [3] LC_TIME=en_GB
                              LC_COLLATE=C
 [5] LC_MONETARY=en_US.UTF-8 LC_MESSAGES=en_US.UTF-8
 [7] LC_PAPER=en_US.UTF-8
                              LC_NAME=C
 [9] LC_ADDRESS=C
                               LC_TELEPHONE=C
[11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C
attached base packages:
[1] stats4
             stats graphics grDevices utils
                                                   datasets methods
[8] base
other attached packages:
[1] Biostrings_2.66.0 GenomeInfoDb_1.34.0 XVector_0.38.0
[4] IRanges_2.32.0
                     S4Vectors_0.36.0 BiocGenerics_0.44.0
loaded via a namespace (and not attached):
[1] zlibbioc_1.44.0
                        compiler_4.2.1
                                                tools_4.2.1
[4] GenomeInfoDbData_1.2.9 RCurl_1.98-1.9
                                                crayon_1.5.2
[7] bitops_1.0-7
```