

# Working with large arrays in R

## A look at HDF5Array/RleArray/DelayedArray objects

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Bioconductor conference  
Boston

July 2017

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# Motivation and challenges

R ordinary **matrix** or **array** is not suitable for big datasets:

- 10x Genomics dataset (single cell experiment): 30,000 genes  $\times$  1.3 million cells = 36.5 billion values
- in an ordinary integer matrix ==> 136G in memory!

Need for alternative containers:

- but at the same time, the object should be (almost) as easy to manipulate as an ordinary matrix or array
- *standard R matrix/array API*: dim, dimnames, t, is.na, ==, +, log, cbind, max, sum, colSums, etc...
- not limited to 2 dimensions ==> also support arrays of arbitrary number of dimensions

2 approaches: **in-memory data** vs **on-disk data**

## In-memory data

- a 30k x 1.3M matrix might still fit in memory if the data can be efficiently compressed
- example: sparse data (small percentage of nonzero values) ==> *sparse representation* (storage of nonzero values only)
- example: data with long runs of identical values ==> *RLE compression (Run Length Encoding)*
- choose the *smallest type* to store the values: `raw` (1 byte) < `integer` (4 bytes) < `double` (8 bytes)
- if using *RLE compression*:
  - choose the *best orientation* to store the values: *by row* or *by column* (one might give better compression than the other)
  - store the data by chunk ==> opportunity to pick up *best type* and *best orientation* on a chunk basis (instead of for the whole data)
- size of 30k x 1.3M matrix in memory can be reduced from 136G to 16G!

## Examples of in-memory containers

**dgCMatrix** container from the *Matrix* package:

- sparse matrix representation
- nonzero values stored as double

**RleArray** and **RleMatrix** containers from the *DelayedArray* package:

- use RLE compression
- arbitrary number of dimensions
- type of values: any R atomic type (`integer`, `double`, `logical`, `complex`, `character`, and `raw`)

## On-disk data

However...

- if data is too big to fit in memory (even after compression) ==> must use *on-disk representation*
- challenge: should still be (almost) as easy to manipulate as an ordinary matrix! (*standard R matrix/array API*)

## Examples of on-disk containers

Direct manipulation of an **HDF5 dataset** via the *r hdf5* API. Low level API!

**HDF5Array** and **HDF5Matrix** containers from the *HDF5Array* package:

Provide access to the HDF5 dataset via an API that mimics the standard R matrix/array API

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# Memory footprint

## The "airway" dataset

```
library(airway)
data(airway)
m <- unname(assay(airway))
dim(m)

## [1] 63677     8

typeof(m)

## [1] "integer"
```

```
head(m, n=4)

##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,] 679  448  873  408 1138 1047  770  572
## [2,]    0    0    0    0    0    0    0    0
## [3,] 467  515  621  365  587  799  417  508
## [4,] 260  211  263  164  245  331  233  229

tail(m, n=4)

##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [63674,]    0    0    0    0    0    0    0    0
## [63675,]    0    0    0    0    0    0    0    0
## [63676,]    0    0    1    0    0    0    0    0
## [63677,]    0    0    0    0    1    0    0    0

sum(m != 0) / length(m)

## [1] 0.3889591
```

## dgCMatrix vs RleMatrix vs HDF5Matrix

```
library(lobstr) # for obj_size()
obj_size(m)

## 2.04 MB

library(Matrix)
obj_size(as(m, "dgCMatrix"))

## 2.38 MB

library(DelayedArray)
obj_size(as(m, "RleMatrix"))

## 2.22 MB

obj_size(as(t(m), "RleMatrix"))

## 1.74 MB

library(HDF5Array)
obj_size(as(m, "HDF5Matrix"))

## 2.40 kB
```

Some limitations of the sparse matrix implementation in the *Matrix* package:

- nonzero values always stored as `double`, the most memory consuming type
- number of nonzero values must be  $< 2^{31}$
- limited to 2 dimensions: no support for arrays of arbitrary number of dimensions

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## RleArray and HDF5Array objects

RleMatrix/RleArray and HDF5Matrix/HDF5Array provide:

- support all R atomic types
- no limits in size (but each dimension must be  $< 2^{31}$ )
- arbitrary number of dimensions

And also:

- **delayed operations**
- **block processing** (behind the scene)
- TODO: multicore block processing (sequential only at the moment)

## Delayed operations

We start with `HDF5Matrix` object `M`:

```
M <- as(m, "HDF5Matrix")
M

## <63677 x 8> HDF5Matrix object of type "integer":
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,]   679  448  873  408 1138 1047  770  572
## [2,]     0     0     0     0     0     0     0     0
## [3,]   467  515  621  365  587  799  417  508
## [4,]   260  211  263  164  245  331  233  229
## [5,]   60   55   40   35   78   63   76   60
## ...
## [63673,]   0     0     0     1     0     1     0     0
## [63674,]   0     0     0     0     0     0     0     0
## [63675,]   0     0     0     0     0     0     0     0
## [63676,]   0     0     1     0     0     0     0     0
## [63677,]   0     0     0     0     1     0     0     0
```

# RleArray and HDF5Array objects

Subsetting is delayed:

```
M2 <- M[10:12, 1:5]
M2

## <3 x 5> DelayedMatrix object of type "integer":
##      [,1] [,2] [,3] [,4] [,5]
## [1,]  394  236  464  175  658
## [2,]  172  168  264  118  241
## [3,] 2112 1867 5137 2657 2735
```

```
seed(M2)

## An object of class "HDF5ArraySeed"
## Slot "filepath":
## [1] "/tmp/Rtmp9V1F2u/HDF5Array_dump/autoc9fc21be8f84.h5"
##
## Slot "name":
## [1] "/HDF5ArrayAUT000002"
##
## Slot "as_sparse":
## [1] FALSE
##
## Slot "type":
## [1] NA
##
## Slot "dim":
## [1] 63677      8
##
## Slot "chunkdim":
## [1] 63677      8
##
## Slot "first_val":
## [1] 679
```

# RleArray and HDF5Array objects

Transposition is delayed:

```
M3 <- t(M2)
M3

## <5 x 3> DelayedMatrix object of type "integer":
##      [,1] [,2] [,3]
## [1,]  394  172 2112
## [2,]  236  168 1867
## [3,]  464  264 5137
## [4,]  175  118 2657
## [5,]  658  241 2735
```

```
seed(M3)

## An object of class "HDF5ArraySeed"
## Slot "filepath":
## [1] "/tmp/Rtmp9V1F2u/HDF5Array_dump/autoc9fc21be8f84.h5"
##
## Slot "name":
## [1] "/HDF5ArrayAUT000002"
##
## Slot "as_sparse":
## [1] FALSE
##
## Slot "type":
## [1] NA
##
## Slot "dim":
## [1] 63677     8
##
## Slot "chunkdim":
## [1] 63677     8
##
## Slot "first_val":
## [1] 679
```

# RleArray and HDF5Array objects

cbind() / rbind() are delayed:

```
M4 <- cbind(M3, M[1:5, 6:8])  
M4  
  
## <5 x 6> DelayedMatrix object of type "integer":  
##   [,1] [,2] [,3] [,4] [,5] [,6]  
## [1,] 394 172 2112 1047 770 572  
## [2,] 236 168 1867     0     0     0  
## [3,] 464 264 5137  799  417  508  
## [4,] 175 118 2657  331  233  229  
## [5,] 658 241 2735   63    76    60
```

```
seed(M4) # Error! (more than one seed)
```

# RleArray and HDF5Array objects

All the operations in the following groups are delayed:

- Arith (+, -, ...)
- Compare (==, <, ...)
- Logic (&, |)
- Math (log, sqrt)
- and more ...

```
M5 <- M == 0
M5

## <63677 x 8> DelayedMatrix object of type "logical":
##      [,1]  [,2]  [,3] ...  [,7]  [,8]
## [1,] FALSE FALSE FALSE . FALSE FALSE
## [2,] TRUE  TRUE  TRUE . TRUE  TRUE
## [3,] FALSE FALSE FALSE . FALSE FALSE
## [4,] FALSE FALSE FALSE . FALSE FALSE
## [5,] FALSE FALSE FALSE . FALSE FALSE
## ...
## [63673,] TRUE  TRUE  TRUE . TRUE  TRUE
## [63674,] TRUE  TRUE  TRUE . TRUE  TRUE
## [63675,] TRUE  TRUE  TRUE . TRUE  TRUE
## [63676,] TRUE  TRUE  FALSE . TRUE  TRUE
## [63677,] TRUE  TRUE  TRUE . TRUE  TRUE
```

```
seed(M5)

## An object of class "HDF5ArraySeed"
## Slot "filepath":
## [1] "/tmp/Rtmp9V1F2u/HDF5Array_dump/autoc9fc21be8f84.h5"
##
## Slot "name":
## [1] "/HDF5ArrayAUTO000002"
##
## Slot "as_sparse":
## [1] FALSE
##
## Slot "type":
## [1] NA
##
## Slot "dim":
## [1] 63677     8
##
## Slot "chunkdim":
## [1] 63677     8
##
## Slot "first_val":
## [1] 679
```

# RleArray and HDF5Array objects

```
M6 <- round(M[11:14, ] / M[1:4, ], digits=3)
M6

## <4 x 8> DelayedMatrix object of type "double":
##      [,1] [,2] [,3] ... [,7] [,8]
## [1,] 0.253 0.375 0.302   . 0.201 0.309
## [2,]    Inf    Inf    Inf   .    Inf    Inf
## [3,] 1.122 0.948 1.027   . 1.182 0.935
## [4,] 0.273 0.242 0.802   . 0.575 0.751
```

```
seed(M6) # Error! (more than one seed)
```

# RleArray and HDF5Array objects

## Realization

Delayed operations can be **realized** by coercing the DelayedMatrix object to HDF5Array:

```
M6a <- as(M6, "HDF5Array")
M6a

## <4 x 8> HDF5Matrix object of type "double":
##   [,1] [,2] [,3] ... [,7] [,8]
## [1,] 0.253 0.375 0.302 . 0.201 0.309
## [2,] Inf Inf Inf . Inf Inf
## [3,] 1.122 0.948 1.027 . 1.182 0.935
## [4,] 0.273 0.242 0.802 . 0.575 0.751
```

```
seed(M6a)

## An object of class "HDF5ArraySeed"
## Slot "filepath":
## [1] "/tmp/Rtmp9V1F2u/HDF5Array_dump/autoc9fc10f000b2.h5"
##
## Slot "name":
## [1] "/HDF5ArrayAUT000003"
##
## Slot "as_sparse":
## [1] FALSE
##
## Slot "type":
## [1] NA
##
## Slot "dim":
## [1] 4 8
##
## Slot "chunkdim":
## [1] 4 8
##
## Slot "first_val":
## [1] 0.253
```

# RleArray and HDF5Array objects

... or by coercing it to RleArray:

```
M6b <- as(M6, "RleArray")
M6b

## <4 x 8> RleMatrix object of type "double":
##   [,1] [,2] [,3] ... [,7] [,8]
## [1,] 0.253 0.375 0.302 . 0.201 0.309
## [2,] Inf Inf Inf . Inf Inf
## [3,] 1.122 0.948 1.027 . 1.182 0.935
## [4,] 0.273 0.242 0.802 . 0.575 0.751
```

```
seed(M6b)

## An object of class "ChunkedRleArraySeed"
## Slot "breakpoints":
## [1] 32
##
## Slot "type":
## [1] "double"
##
## Slot "chunks":
## <environment: 0x641fa27fb368>
##
## Slot "DIM":
## [1] 4 8
##
## Slot "DIMNAMES":
## [[1]]
## NULL
##
## [[2]]
## NULL
```

# RleArray and HDF5Array objects

## Controlling where HDF5 datasets are realized

*HDF5 dump management utilities:* a set of utilities to control where HDF5 datasets are written to disk.

```
hdf5_dumpfile <- file.path(mydata_dir, "M6c.h5")
setHDF5DumpFile(hdf5_dumpfile)
setHDF5DumpName("M6c")
M6c <- as(M6, "HDF5Array")
```

```
seed(M6c)

## An object of class "HDF5ArraySeed"
## Slot "filepath":
## [1] "/tmp/Rtmp9V1F2u/mydata/M6c.h5"
##
## Slot "name":
## [1] "/M6c"
##
## Slot "as_sparse":
## [1] FALSE
##
## Slot "type":
## [1] NA
##
## Slot "dim":
## [1] 4 8
##
## Slot "chunkdim":
## [1] 4 8
##
## Slot "first_val":
## [1] 0.253

h5ls(hdf5_dumpfile)

##   group name      otype dclass    dim
## 0      / M6c H5I_DATASET FLOAT 4 x 8
```

# RleArray and HDF5Array objects

```
showHDF5DumpLog()
```

```
showHDF5DumpLog()
```

```
## [2025-10-29 23:32:33.005775] #1 In file '/tmp/Rtmp9V1F2u/HDF5Array_dump/autoc9fc7e4b821c.h5': creation of dataset
'/HDF5ArrayAUT00001' (63677x8:integer, chunkdims=63677x8, level=6)
## [2025-10-29 23:32:33.243599] #2 In file '/tmp/Rtmp9V1F2u/HDF5Array_dump/autoc9fc21be8f84.h5': creation of dataset
'/HDF5ArrayAUT00002' (63677x8:integer, chunkdims=63677x8, level=6)
## [2025-10-29 23:32:34.126969] #3 In file '/tmp/Rtmp9V1F2u/HDF5Array_dump/autoc9fc10f000b2.h5': creation of dataset
'/HDF5ArrayAUT00003' (4x8:double, chunkdims=4x8, level=6)
## [2025-10-29 23:32:34.386477] #4 In file '/tmp/Rtmp9V1F2u/mydata/M6c.h5': creation of dataset 'M6c' (4x8:double, chunkdims=4x8,
level=6)
```

## Block processing

The following operations are NOT delayed. They are implemented via a *block processing* mechanism that loads and processes one block at a time:

- operations in the Summary group (`max`, `min`, `sum`, `any`, `all`)
- `mean`
- Matrix row/col summarization operations (`col/rowSums`, `col/rowMeans`, ...)
- `anyNA`, `which`
- `apply`
- and more ...

# RleArray and HDF5Array objects

```
DelayedArray:::set_verbose_block_processing(TRUE)

## [1] FALSE

colSums(M)

## === START walking on vertical strip 1/1 ===
##   / processing <63677 x 8> block from grid position [[1/1, 1/1]] ... ok
## === DONE walking on vertical strip 1/1 ===

## [1] 20637971 18809481 25348649 15163415 24448408 30818215 19126151 21164133
```

## Control the block size:

```
getAutoBlockSize()

## [1] 1e+08

setAutoBlockSize(1e6)

## automatic block size set to 1e+06 bytes (was 1e+08)

colSums(M)

## === START walking on vertical strip 1/1 ===
##   / processing <63677 x 8> block from grid position [[1/1, 1/1]] ... ok
## === DONE walking on vertical strip 1/1 ===

## [1] 20637971 18809481 25348649 15163415 24448408 30818215 19126151 21164133
```

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## Hands-on

1. Load the "airway" dataset.
2. It's wrapped in a `SummarizedExperiment` object. Get the count data as an ordinary matrix.
3. Wrap it in an `HDF5Matrix` object: (1) using `writeHDF5Array()`; then (2) using coercion.
4. When using coercion, where has the data been written on disk?
5. See `?setHDF5DumpFile` for how to control the location of "automatic" HDF5 datasets. Try to control the destination of the data when coercing.

## Hands-on

6. Use `showHDF5DumpLog()` to see all the HDF5 datasets written to disk during the current session.
7. Try some operations on the `HDF5Matrix` object: (1) some delayed ones; (2) some non-delayed ones (block processing).
8. Use `DelayedArray:::set_verbose_block_processing(TRUE)` to see block processing in action.
9. Control the block size with `setAutoBlockSize()`.

10. Stick the `HDF5Matrix` object back in the `SummarizedExperiment` object. The resulting object is an "HDF5-backed `SummarizedExperiment` object".
11. The `HDF5-backed SummarizedExperiment` object can be manipulated (almost) like an in-memory `SummarizedExperiment` object. Try `[`, `cbind`, `rbind` on it.
12. The `SummarizedExperiment` package provides `saveHDF5SummarizedExperiment` to save a `SummarizedExperiment` object (HDF5-backed or not) as an HDF5-backed `SummarizedExperiment` object. Try it.

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## Block processing improvements

Block geometry: (1) better by default, (2) let the user have more control on it

Support multicore

Expose it: `blockApply()`

## HDF5Array improvements

Store the `dimnames` in the HDF5 file (in *HDF5 Dimension Scale datasets* - <https://www.hdfgroup.org/HDF5/Tutor/h5dimscale.html>)

Use better automatic chunk geometry when realizing an HDF5Array object

Block processing should take advantage of the chunk geometry (e.g. `realize()` should use blocks that are clusters of chunks)

Unfortunately: not possible to support multicore realization at the moment (HDF5 does not support concurrent writing to a dataset yet)

## RleArray improvements

Let the user have more control on the chunk geometry when constructing/realizing an RleArray object

Like for HDF5Array objects, block processing should take advantage of the chunk geometry

Support multicore realization

Provide C/C++ low-level API for direct row/column access from C/C++ code (e.g. from the `beachmat` package)