# Working with large arrays in R A look at HDF5Array/RleArray/DelayedArray objects

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> Bioconductor conference Boston

> > July 2017

- Motivation and challenges
- 2 Memory footprint
- RleArray and HDF5Array objects
- 4 Hands-on
- 5 DelayedArray/HDF5Array: Future developments

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

- ullet 10x Genomics dataset (single cell experiment): 30,000 genes x 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...
- ullet not limited to 2 dimensions ==> also support arrays of arbitrary number of dimensions

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

4/34

#### 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)</li>
- 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)
- $\bullet$  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 rhdf5 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] 64102 8

typeof(m)

## [1] "integer"
```

#### dgCMatrix vs RleMatrix vs HDF5Matrix

```
library(pryr) # for object_size()
object_size(m)
## 2.05 MB
library(Matrix)
object_size(as(m, "dgCMatrix"))
## 2.38 MB
library(DelayedArray)
object_size(as(m, "RleMatrix"))
## 2.22 MB
object_size(as(t(m), "RleMatrix"))
## 1.74 MB
library(HDF5Array)
object_size(as(m, "HDF5Matrix"))
## 1.99 kB
```

# Memory footprint

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

- nonzero values always stored as double, the most memory consuming type
- ullet 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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### $RleMatrix/RleArray\ and\ HDF5Matrix/HDF5Array\ provide:$

- 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:

15 / 34

#### Subsetting is delayed:

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

## <3 x 5> matrix of class DelayedMatrix and 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
```

#### Transposition is delayed:

```
M3 <- t(M2)
M3

## <5 x 3> matrix of class DelayedMatrix and 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"
    ## $lot "filepath":
    ## [1] "/tmp/Rtmp6XozhQ/HDF5Array_dump/auto00002.h5"

## $lot "name":
    ## $lot "file":
    ## $lot "dim":
    ## [1] 64102 8

##
    ## $lot "first_val":
    ## $
```

#### cbind() / rbind() are delayed:

```
M4 <- cbind(M3, M[1:5, 6:8])
M4

## <5 x 6> matrix of class DelayedMatrix and 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)
```

All the operations in the following groups are delayed:

```
• Arith (+, -, ...)
• Compare (==, <, ...)
```

- Logic (&, |)
- Math (log, sqrt)
- and more ...

```
M5 <- M == 0
M5
## <64102 x 8> matrix of class DelayedMatrix and 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
## [64098.] TRUE TRUE TRUE
                             . TRUE TRUE
## [64099.] TRUE TRUE TRUE
                             . TRUE TRUE
## [64100.] TRUE TRUE TRUE
                                TRUE TRUE
## [64101.] TRUE TRUE TRUE
                                TRUE
                                    TRUE
## [64102.] TRUE TRUE TRUE
                                TRUE TRUE
```

```
seed(M5)

## An object of class "HDF5ArraySeed"
: ## Slot "filepath":
## [1] "/tmp/Rtmp6XozhQ/HDF5Array_dump/auto00002.h5"
##
## $$lot "name":
## [1] "/HDF5ArrayAUT000002"
## ## $lot "dim":
## [1] 64102 8
##
## $lot "chunkdim":
## [1] 64102 8
##
## Slot "first_val":
## [1] 679
```

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

## <4 x 8> matrix of class DelayedMatrix and type "double":
## [,1] [,2] [,3] ... [,7] [,8]
## [1,] 0.253 0.375 0.302 . 0.201 0.309
## [2,] Inf 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)
```

#### Realization

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

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

## <4 x 8> matrix of class HDF5Matrix and type "double":
## [,1] [,2] [,3] ... [,7] [,8]
## [1,] 0.253 0.375 0.302 0.0201 0.309
## [2,] Inf Inf 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/Rtmp6XozhQ/HDF5Array_dump/auto00003.h5"
## ## Slot "name":
## [1] "/HDF5ArrayAUT000003"
## ## ## $! "## [1] 4 8
##
## Slot "chunkdim":
## [1] 4 8
##
## Slot "first_val":
## [1] 0.253
```

21 / 34

... or by coercing it to RleArray:

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

## <4 x 8> matrix of class RleMatrix and type "double":
## [,1] [,2] [,3] ... [,7] [,8]
## [1,] 0.253 0.375 0.302 0.201 0.309
## [2,] Inf 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: 0x55ee5bbd3800>
## Slot "DIM":
## [1] 4 8
##
## Slot "DIMNAMES":
## [[1]]
## NULL
##
## [[2]]
## NULL
```

#### Controlling where HDF5 datasets are realized

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

```
setHDF5DumpFile("-/mydata/M6c.h5")
setHDF5DumpMame("M6c")
M6c <- as(M6, "HDF5Array")
```

```
seed (M6c)
## An object of class "HDF5ArraySeed"
## Slot "filepath":
## [1] "/home/biocbuild/mydata/M6c.h5"
## Slot "name":
## [1] "M6c"
## Slot "dim":
## [1] 4 8
## Slot "chunkdim":
## [1] 4 8
## Slot "first val":
## [1] 0.253
h5ls("~/mydata/M6c.h5")
## group name otype dclass dim
## 0 / M6c H5I DATASET FLOAT 4 x 8
```

#### showHDF5DumpLog()

#### showHDF5DumpLog()

- ## [2019-10-29 22:21:58] #1 In file '/tmp/Rtmp6NozhQ/HDF5Array\_dump/auto00001.h5': creation of dataset '/HDF5ArrayAUT000001' (64102x8:integer, chunkdims=64102x8, level=6)
- ## [2019-10-29 22:21:59] #2 In file '/tmp/Rtmp6XozhQ/HDF5Array\_dump/auto00002.h5': creation of dataset '/HDF5ArrayAUT000002' (64102x8:integer, chunkdims=64102x8, level=6)
- ## [2019-10-29 22:21:59] #3 In file '/tmp/Rtmp6XozhQ/HDF5Array\_dump/auto00003.h5': creation of dataset '/HDF5ArrayAUT000003'
- $(4x8:double, chunkdims=4x8, level=6) \\ \#\# \begin{tabular}{l} (2019-10-29 & 22:22:00] \#4 In file '/home/biocbuild/mydata/M6c.h5': creation of dataset 'M6c' (4x8:double, chunkdims=4x8, level=6) \\ \end{tabular}$

### 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 ...

```
DelayedArray:::set_verbose_block_processing(TRUE)

## [1] FALSE

colSums(M)

## Processing block 1/1 ... 0K

## [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)

## Processing block 1/1 ...

## DK

## [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 t	he HDF5	datasets writter	n 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  ${\tt DelayedArray:::set\_verbose\_block\_processing(TRUE)}$  to see block processing in action.

9. Control the block size with setAutoBlockSize().

### Hands-on

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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# Future developments

#### **Block processing improvements**

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

Support multicore

Expose it: blockApply()

### Future developments

#### 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)

### Future developments

#### 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)

34 / 34