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

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- Motivation and challenges
- Memory footprint
- RleArray and HDF5Array objects
- Mands-on
- Delayed Array/HDF5Array: Future developments

- Motivation and challenges
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- Mands-or
- DelayedArray/HDF5Array: Future developments

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

- 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

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 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] 63677 8

typeof(m)
## [1] "integer"
```

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"))
## 2.38 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
- \bullet 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:

```
M <- as(m, "HDF6Matrix")

## <63677 x 8> HDF6Matrix object of type "integer":

## [,1] [,2] [,3] [,4] [,6] [,6] [,7] [,8]

## [1,] 679 448 873 408 1138 1047 770 572

## [2,] 0 0 0 0 0 0 0 0 0

## [3,] 467 515 621 365 587 799 417 508

## [4,] 260 211 263 104 245 331 233 229

## [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

## [63676,] 0 0 1 0 0 0 0 0

## [63676,] 0 0 1 0 0 0 0 0

## [63677,] 0 0 0 1 0 0 0 0

## [63677,] 0 0 0 1 0 0 0 0

## [63677,] 0 0 0 1 0 0 0 0

## [63677,] 0 0 0 1 0 0 0 0 0

## [63677,] 0 0 0 1 0 0 0 0 0

## [63677,] 0 0 0 1 0 0 0 0 0

## [63677,] 0 0 0 1 0 0 0 0 0

## [63677,] 0 0 0 0 0 0 0 0 0

## [63677,] 0 0 0 0 0 0 0 0 0

## [63677,] 0 0 0 0 0 0 0 0 0 0

## [63677,] 0 0 0 0 0 0 0 0 0 0

## [63677,] 0 0 0 0 0 0 0 0 0 0

## [63677,] 0 0 0 0 0 0 0 0 0 0
```

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/RtmpnkzHDU/HDF5Array_dump/auto00002.h5"
## Slot "name":
## [1] "/HDF5ArrayAUT000002"
##
## Slot "as_sparse":
## [1] FALSE
## Slot "type":
## [1] NA
## Slot "dim":
## [1] 63677
## Slot "chunkdim":
## [1] 63677 8
## Slot "first_val":
## [1] 679
```

Transposition is delayed:

```
M3 <- t(M2)
M3

## <5 x 3> DelayedMatrix object of type "integer":
## [, 1] [, 2] [, 3]
## [1], 394 172 2112
## [2], 286 168 1867
## [3], 464 264 5137
## [4], 175 118 2667
## [6], 658 241 2735
```

```
seed(M3)
## An object of class "HDF5ArraySeed"
## Slot "filepath":
## [1] "/tmp/RtmpnkzHDU/HDF5Array_dump/auto00002.h5"
## Slot "name":
## [1] "/HDF5ArrayAUT000002"
##
## Slot "as_sparse":
## [1] FALSE
## Slot "type":
## [1] NA
## Slot "dim":
## [1] 63677
## Slot "chunkdim":
## [1] 63677
## Slot "first_val":
## [1] 679
```

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 1667 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 (==, <, ...)</li>
Logic (&, |)
Math (log, sqrt)
and more ...
```

```
M5 <- M == 0
М5
## <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
                TRUE FALSE
## [63676.] TRUE
                                TRUE TRUE
## [63677.] TRUE TRUE TRUE
                                TRUE TRUE
```

```
seed(M5)
## An object of class "HDF5ArraySeed"
## Slot "filepath":
## [1] "/tmp/RtmpnkzHDU/HDF5Array_dump/auto00002.h5"
## Slot "name":
## [1] "/HDF5ArrayAUT000002"
##
## Slot "as_sparse":
## [1] FALSE
## Slot "type":
## [1] NA
##
## Slot "dim":
## [1] 63677
## Slot "chunkdim":
## [1] 63677
## Slot "first_val":
## [1] 679
```

```
M6 <- round(M[i1:14, ] / M[i: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)
```

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 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/RtmpnkzHDU/HDF5Array_dump/auto00003.h5"
## Slot "name":
## [1] "/HDF5ArravAUT000003"
##
## 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
```

or by coercing it to RleArray:

```
seed(M6b)
## An object of class "ChunkedRleArraySeed"
## Slot "breakpoints":
## [1] 32
## Slot "type":
## [1] "double"
## Slot "chunks":
## <environment: 0x563fb25e37f0>
## 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.

```
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/RtmpnkzHDU/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
h51s(hdf5_dumpfile)
                     otype dclass dim
## group name
## 0 / M6c H5T DATASET FLOAT 4 x 8
```

showHDF5DumpLog()

showHDF5DumpLog()

```
## [2023-10-24 16:56:04.152725] #1 In file '/tmp/itmpnkz#DU/HDF51rray_dump/auto00001.h5': creation of dataset
'/HDF51rrayUUT000001' (6367728:integer, chuskdims-6367726, level-6)
## [2023-10-24 16:56:04, 152194] #2 In file '/tmp/itmpnkz#DU/HDF51rray_dump/auto00002.h5': creation of dataset
'/HDF51rrayUUT000002' (6367728:integer, chuskdims-6367728, level-6)
## [2023-10-24 16:56:05.006517] #3 In file '/tmp/itmpnkz#DU/HDF51rray_dump/auto00003.h5': creation of dataset
'/HDF51rrayUUT00003' (488:double, chuskdims-428, level-6)
## [2023-10-24 16:56:05.232826] #4 In file '/tmp/itmpnkz#DU/mydata/H6c.h5': creation of dataset 'H6c' (4x8:double, chuskdims-428, 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 ...

```
DelayedArray:::set_verbose_block_processing(TRUE)

## [1] FALSE

colSums(M)

## Processing block [[1/1, 1/1]] ... OK

## [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, 1/1]] ...
## 0X
## [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 <code>?setHDF5DumpFile</code> 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().

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)