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Start of a Package: Plotting a voxel from 800 Subjects in a Shiny App

Options for reading in NIfTI images

Array like

- `oro.nifti::readNIfTI`/`neurobase::readnii`
- `AnalyzeFMRI::f.read.volume()` - can't read gzipped
- `neuroim::loadVolume("example.nii")`

C++ pointer

- `ANTsCore::antsImageRead`
- `RNifti::readNifti`

From RNifti: https://github.com/jonclayden/RNifti#performance

```
library(microbenchmark)
microbenchmark(AnalyzeFMRI::f.read.volume("example.nii"),
               ANTsRCore::antsImageRead("example.nii"),
               neuroim::loadVolume("example.nii"),
               oro.nifti::readNIfTI("example.nii"),
               RNifti::readNifti("example.nii"),
               tractor.base::readImageFile("example.nii"), unit="ms")
# Unit: milliseconds
#                                     expr       min        lq      mean
#   AnalyzeFMRI::f.read.volume("example.nii") 26.312881 26.769244 29.685981
#     ANTsRCore::antsImageRead("example.nii")  1.150787  1.626918  2.149145
#          neuroim::loadVolume("example.nii") 34.596506 37.732245 54.065227
#        oro.nifti::readNIfTI("example.nii") 57.059828 61.953430 89.386706
#           RNifti::readNifti("example.nii")  0.986773  1.136562  1.683821
#  tractor.base::readImageFile("example.nii") 33.380407 34.096574 34.961812
#     median        uq        max neval
#  27.301693 28.214908 185.441664   100
#   2.210696  2.481675   3.376350   100
#  40.714061 45.425047 192.978225   100
#  65.064212 71.425561 220.709246   100
#   1.501528  1.856601   7.961566   100
#  34.617259 35.257633  42.108584   100
```

What do we want

1. Something fast
2. Has overloaded functions (e.g. `image + image = image`)
3. Can use most plotting functions on it (e.g. `ortho2`)
4. Is array like but has header
5. Keeps memory low
6. Has **random access**

Current issues

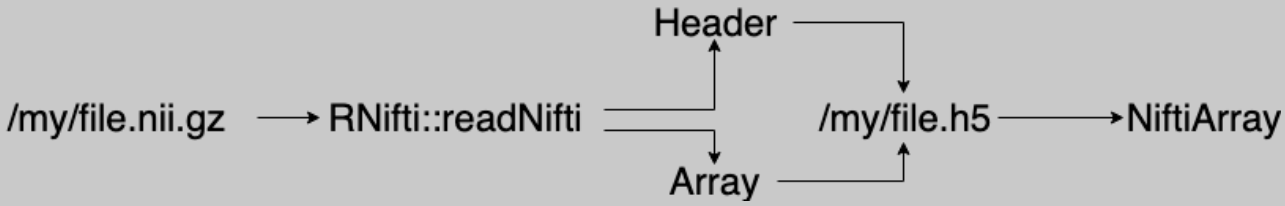
- `oro.nifti::readNIfTI` - slow
- `ANTsCore::antsImageRead` - installation is
- `RNifti::readNifti` - from https://github.com/jonclayden/RNifti/issues/14 :

For your single-precision floating-point image with about 1bn elements, the “native” data size in the C struct is roughly 4 GiB, while the R array uses 8 GiB because it uses double precision. When you use readNifti(…, internal=TRUE) the resulting object has only the C struct, but with readNifti(…, internal=FALSE) the object has both the C struct and the R array, for a total of 12 GiB.

- None have random access (just read in one slice or 1 timepoint)

Solution: NiftiArray

- Stores image and header in HDF5 (Hierarchical Data Format) format
- Based on `HDF5Array` from Bioconductor (https://bioconductor.org/packages/release/bioc/html/HDF5Array.html)



Takes longer than RNifti (the first time)

```
fname = fslr::mni_fname()
system.time({
  rnifti = RNifti::readNifti(fname)
})
```

```
##    user  system elapsed
##   0.125   0.016   0.142
```

```
system.time({
  narray = NiftiArray::writeNiftiArray(fname)
  narray@seed@filepath
  narray
})
```

```
##    user  system elapsed
##   4.512   0.114   4.670
```

Sizes

```
format(object.size(rnifti), units = "Kb")
```

```
## [1] "28208.3 Kb"
```

```
format(object.size(narray), units = "Kb")
```

```
## [1] "8.7 Kb"
```

Header Information

```
print(nifti_header(narray))
```

```
## NIfTI-1 header
##     sizeof_hdr: 348
##       dim_info: 0
##            dim: 3  182  218  182  1  1  1  1
##      intent_p1: 0
##      intent_p2: 0
##      intent_p3: 0
##    intent_code: 0
##       datatype: 8
##         bitpix: 32
##    slice_start: 0
##         pixdim: -1  1  1  1  0  0  0  0
##     vox_offset: 352
##      scl_slope: 0
##      scl_inter: 0
##      slice_end: 0
##     slice_code: 0
##     xyzt_units: 10
##        cal_max: 8000
##        cal_min: 3000
## slice_duration: 0
##        toffset: 0
##        descrip: FSL5.0
##       aux_file:
##     qform_code: 4
##     sform_code: 4
##      quatern_b: 0
##      quatern_c: 1
##      quatern_d: 0
##      qoffset_x: 90
##      qoffset_y: -126
##      qoffset_z: -72
##         srow_x: -1  0  0  90
##         srow_y: 0  1  0  -126
```

Accessing voxels

```
narray[18, 5, 14]
```

```
## [1] 572
```

```
sum(narray > 9990)
```

```
## [1] 20
```

```
narray[ narray > 9990]
```

```
##  [1] 9999 9999 9992 9997 9991 9993 9997 9998 9999 9999 9994 9997 9998 9999
## [15] 9998 9991 9993 9991 9996 9994
```

```
DelayedArray::extract_array(narray, index = list(NULL ,NULL, 1))
```

```
## , , 1
##
##        [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11] [,12]
##   [1,]    0    0    0    0   30   31   29   29   29    29    29    28
##   [2,]    0    0    0    0  266  288  279  279  279   280   281   283
##   [3,]    0    0    0    0  410  455  448  449  444   443   446   452
##   [4,]    0    0    0    0  420  461  453  449  441   443   446   446
##        [,13] [,14] [,15] [,16] [,17] [,18] [,19] [,20] [,21] [,22] [,23]
##   [1,]    28    27    27    27    27    28    28    29    28    27    28
##   [2,]   280   278   277   281   282   279   282   293   289   280   283
##   [3,]   452   453   452   458   462   456   462   477   473   465   469
##   [4,]   450   452   450   458   461   460   464   470   458   459   470
##        [,24] [,25] [,26] [,27] [,28] [,29] [,30] [,31] [,32] [,33] [,34]
##   [1,]    28    27    27    28    30    31    32    32    31    32    32
##   [2,]   288   283   280   284   292   295   299   306   307   303   300
##   [3,]   474   466   460   465   472   473   480   499   507   495   486
##   [4,]   476   471   464   465   469   474   482   498   508   494   485
##        [,35] [,36] [,37] [,38] [,39] [,40] [,41] [,42] [,43] [,44] [,45]
```

What are the benefits?

- Overall most benefits come from saving H5 files on "big" computer
- Random access - the `DelayedArray` package
- Operations on `DelayedArray` objects, the DelayedMatrixStats package https://bioconductor.org/packages/release/bioc/html/DelayedMatrixStats.html

What are the benefits? Reshaping to a Matrix

```
d = dim(narray)
d
```

```
## [1] 182 218 182
```

```
dim(narray) = c(d, 1)
new_dim = c(prod(d), 1)
newarray = writeNiftiArray(narray)
mat = ReshapedNiftiArray(newarray@seed@filepath, dim = new_dim)
dim(mat)
```

```
## [1] 7221032       1
```

```
head(mat)
```

```
## <6 x 1> DelayedMatrix object of type "integer":
##      [,1]
## [1,]    0
## [2,]    0
## [3,]    0
## [4,]    0
## [5,]    0
## [6,]    0
```

Why? Voxel-wise operations in low-memory setting

```
big_mat = DelayedArray::acbind(mat, mat, mat)
dim(big_mat)
```

```
## [1] 7221032       3
```

```
format(object.size(big_mat), units = "Kb")
```

```
## [1] "26.7 Kb"
```

```
rmed = DelayedMatrixStats::rowMedians(big_mat)
```

Why? 4D Example

```
nii_fname = system.file("extdata",
                        "example_4d.nii.gz", package = "RNifti")
from = NiftiArray::NiftiArray(nii_fname)
dim(from)
```

```
## [1] 96 96 60  3
```

```
A <- ReshapedNiftiArray(filepath = from@seed@filepath,
                        name = from@seed@name,
                        dim = c(prod(dim(from)[1:3]), dim(from)[4]))
dim(A)
```

```
## [1] 552960      3
```

```
med = DelayedMatrixStats::rowMedians(A)
```

Back to an image

```
arr = array(med, dim = dim(from)[1:3])
med_array = writeNiftiArray(arr, header = nifti_header(from))
as(med_array, "niftiImage")
```

```
## Image array of mode "double" (4.2 Mb)
## - 96 x 96 x 60 voxels
## - 2.5 x 2.5 x 2.5 mm per voxel
```

Issues

- Need to write H5 file first (do in parallel on cluster)
- Memory tradeoff with speed due to chunking
- Most of this needs things in the same space

Use cases

- Something on Shiny for speed or memory
    - shinyapps.io has very restrictive memory settings for imaging
- Analyzing 1000s of subjects together
- Pass object from R to Matlab: https://www.mathworks.com/help/matlab/ref/hdf5read.html

Extensions

- Anything that works with DelayedArrays should work here
- Check out Bioconductor
- Example: `mbkmeans` - mini-batch k-means https://bioconductor.org/packages/release/bioc/html/mbkmeans.html