# Imaging Packages in R

Processing math: 100%

## Some packages we will use

All packages we will discuss are loaded on the RStudio Server:

- oro.nifti reading/writing NIfTI images
  - made the nifti object/data class: like an array but with header information
    - the main data class we will use
- neurobase extends oro.nifti and provides helpful imaging functions

#### Let's load them:

```
library(oro.nifti)
library(neurobase)
```

## Reading in NIfTI images: assignment

We will use the readnii function (from neurobase) to read in a nifti object (this is an R object).

Here we read in the "training01\_01\_t1.nii.gz" file, and assign it to an object called t1:

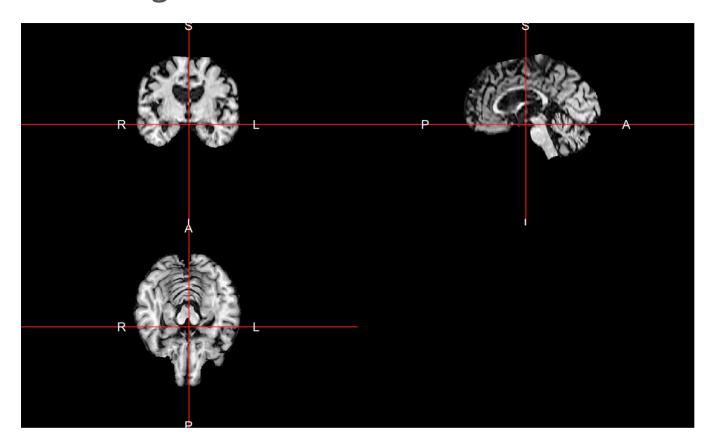
```
t1 = readnii("training01_01_t1.nii.gz")
```

Now, an object t1 is in memory/the workspace.

```
class(t1)

[1] "nifti"
attr(,"package")
[1] "oro.nifti"
```

## nifti images



#### nifti images

By default, if you simply pass the object, it is printed, we can also do print (t1):

t1

```
Type : nifti
Data Type : 16 (FLOAT32)
Bits per Pixel : 32
Slice Code : 0 (Unknown)
Intent Code : 0 (None)
Qform Code : 1 (Scanner_Anat)
Sform Code : 0 (Unknown)
Dimension : 192 x 512 x 512
Pixel Dimension : 0.8 x 0.47 x 0.47
Voxel Units : mm
Time Units : Unknown
```

#### Operations with nifti objects

These work with an image and a number (img + 2) or two images of the same dimensions img1 + img2.

- Comparison: >, >=, <, <=, == (equals), != (not equal)</li>
- Logical: ! not, & and, | or (a "pipe")
- Arithmetic: +, -, \*, /, ^ exponents
- · Standard math functions: log, abs, sqrt

```
t1 + t1 + 2 # still a nifti
```

```
Type : nifti
Data Type : 16 (FLOAT32)
Bits per Pixel : 32
Slice Code : 0 (Unknown)
Intent Code : 0 (None)
Qform Code : 1 (Scanner_Anat)
Sform Code : 0 (Unknown)
Dimension : 192 x 512 x 512
Pixel Dimension : 0.8 x 0.47 x 0.47
Voxel Units : mm
Time Units : Unknown
```

## Working with nifti objects

Again, we can use a logical operation. Let's create an image indicating values over 400:

```
class(t1 > 400) # still a nifti
[1] "nifti"
attr(,"package")
[1] "oro.nifti"
head(t1 > 400) # values are now logical vs. numeric
```

[1] FALSE FALSE FALSE FALSE FALSE

We will refer to images such as t1 > 400 as a "mask", simply binary images with logical values in them (or 0s and 1s)

#### Subsetting with nifti objects: like arrays

The subsetting here is similar to that of arrays. Since t1 is 3-dimensional the subsetting goes to the 3rd dimension:

```
t1[5, 4, 3]

[1] 0

t1[5, 4, ] # returns a vector of numbers (1-d)

t1[, 4, ] # returns a 2-d matrix

t1[1, , ] # returns a 2-d matrix
```

- You can subset with a logical array of the same dimensions!
- We can view values of the t1 greater than 400 (head only prints the first 6 values):

```
head(t1[ t1 > 400 ]) # produces a vector of numbers

[1] 409.0842 414.6199 416.8652 430.9107 413.1289 422.2794
```

## which with nifti objects

The which function works to get indices, but you can pass the arr.ind = TRUE argument to get "array" indices:

```
head(which(t1 > 400, arr.ind = TRUE))

dim1 dim2 dim3
[1,] 129 376 179
[2,] 128 377 179
[3,] 128 378 179
[4,] 134 251 194
[5,] 135 251 194
[6,] 135 252 194
```

But can get the "vector" indices as well:

```
head(which(t1 > 400, arr.ind = FALSE))
[1] 17570241 17570432 17570624 19020806 19020807 19020999
```

## Working with nifti objects: reassignment

Subsetting can work on the left hand side of assignment too:

```
t1_copy = t1
t1_copy[ t1_copy > 400 ] = 400 # replaced these values!
max(t1_copy) # should be 400

[1] 400

max(t1)

[1] 829.8354
```

Note, although t1\_copy was copied from t1, they are not linked - if you change values in t1 copy, values in t1 are unchanged.

#### Writing Images out

We now can write out this modified t1\_copy image:

We have seen that file.exists returns TRUE if a file exists

useful in conjunction with all: all(file.exists(VECTOR\_OF\_FILES))

#### Vectorizing a nifti

To convert a nifti to a vector, you can simply use the c() function:

```
vals = c(t1)
class(vals)

[1] "numeric"
```

Essentially "strings out" the array. If you do array(c(t1), dim = dim(t1)), this will put things back "in order" of the t1.

Vectorizing is useful for making data.frames (covered later) when you want to do modeling at a voxel level.

```
df = data.frame(t1 = c(t1), mask = c(t1 > 400)); head(df)
```

- t1 mask
- 1 0 FALSE
- 2 0 FALSE
- 3 0 FALSE
- 4 0 FALSE
- 5 0 FALSE
- 6 0 FALSE

## File helpers - for constructing filenames

Use paste if you want to put strings together with spaces, paste 0 no spaces by default.

file.path(directory, filename) will paste directory and filename w/file separators (e.g. /)

```
c(paste("img", ".nii.gz"), paste0("img", ".nii.gz"))
[1] "img .nii.gz" "img.nii.gz"

x = file.path("output_directory", paste0("img", ".nii.gz")); print(x)

[1] "output_directory/img.nii.gz"

nii.stub will strip off the nifti extension. If bn = TRUE, it removes the directory as well:

c(nii.stub(x), nii.stub(x, bn = TRUE))

[1] "output_directory/img" "img"
```

#### Main Packages we will use

- oro.nifti reading/writing NIfTI images
- neurobase extends oro.nifti and provides helpful imaging functions
- fslr wraps FSL commands to use in R
  - registration, image manipulation, skull stripping
- ANTSR wrapper for Advanced normalization tools (ANTs) code
  - registration, inhomogeneity correction, lots of tools
- extrantsr allows ANTSR to work with objects from oro.nifti

#### Data Package we will use

- ms.lesion contains training/testing data of patients with multiple sclerosis (MS)
  - from an open MS MRI data set (Lesjak et al. 2017)

#### Conclusions

- · We have (briefly) covered some R data classes and types to get you started
- We will be using nifti objects
  - They are special 3-dimensional arrays
  - Contain numbers or logicals
- readnii and writenii are used for reading/writing nifti objects to NIfTI files
- We have briefly covered subsetting and image manipulation
  - more on that later

#### Website

#### http://johnmuschelli.com/imaging\_in\_r

Lesjak, Žiga, Alfiia Galimzianova, Aleš Koren, Matej Lukin, Franjo Pernuš, Boštjan Likar, and Žiga Špiclin. 2017. "A Novel Public MR Image Dataset of Multiple Sclerosis Patients with Lesion Segmentations Based on Multi-Rater Consensus." . Springer, 1–13.