General R (and Writing Images)

Reading in the image

Again we will read in the training01_01_mprage.nii.gz file, and assign it to an object called t1:

```
library(neurobase)
t1 = readnii("training01_01_mprage.nii.gz")
```

Data Classes

- Numeric numbers (e.g. 1, 3.673)
- ► Character strings or words ("hey", "I'm a string") in either single or double quotes
- ► Logicals TRUE or FALSE all capital letters and are **not** in quotes.

Data Types

- vector 1-dimensional object of one class (all numeric or all character)
- matrix 2-dimensional object of one class
- data.frame 2-dimensional object, can be multiple classes (like Excel spreadsheet)
- array object of dimensions > 2 of one class. The data in a nifti object is one of these (usually 3-D)
- nifti an array with header information
- list a general holder of things (discuss when necessary)

Vectors

Create a vector of numeric values and assign to variable v

```
v = c(1, 4, 3, 7, 8)
```

► Subsetting (first index is 1, not zero):

```
print(v[4])
```

```
print(v[1:3])
```

[1] 7

```
[1] 1 4 3
```

print(v[c(1,3,5)])

```
[1] 1 3 8
```

· creates a sequence of numbers

Matrices

Create a 3 x 4 numeric matrix and assign to variable m

```
m = matrix(1:12, nrow = 3)
```

► Subsetting - [row,column] format, if row or column missing then all values:

```
print(m[,4])

[1] 10 11 12

print(m[2,])

[1] 2 5 8 11

print(m[1,3])
```

Subsetting with logicals

[1] 7 8

You can either do subsetting with indices or logical vectors:

```
v[v > 5]
[1] 7 8
the which command takes a logical and gets the indices:
which(v > 5)
[1] 4 5
v[which(v > 5)]
```

The subsetting here is similar to that of arrays, so we will use the t1. Since it's 3-dimensions 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
```

Again, we can use a logical operation. Let's get the values of the t1 greater than 400 (head only prints the first 6 values):

```
head(t1[t1 > 400])
```

[1] 421 617 617 479 456 404

The operation results in a nifti object, and if we look at the values, they are logical:

```
class(t1 > 400)
```

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

```
head(t1 > 400)
```

[1] FALSE FALSE FALSE FALSE FALSE

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,] 98 129 1
[2,] 99 129 1
[3,] 100 129 1
[4,] 163 129 1
[5,] 164 129 1
[6,] 190 129 1
```

Again, we can use a logical operation. Let's get the values of the t1 greater than 400 (head only prints the first 6 values):

```
head(t1[t1 > 400])
```

[1] 421 617 617 479 456 404

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 # changed these values!
max(t1_copy) # should be 400
```

[1] 400

```
max(t1)
```

[1] 1505

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:

[1] TRUE

Vectorizing a nifti

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

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

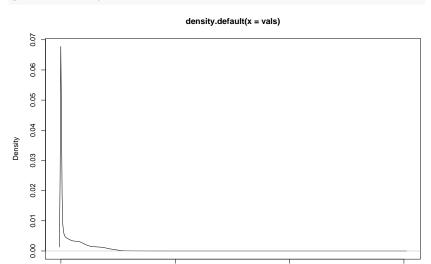
[1] "numeric"

Note an array can be reconstructed by using array(vals, dim = dim(t1)) and will be in the correct order as the way R creates vectors and arrays.

Histogram

From these values we can do all the standard plotting/manipulations of data. For example, let's do a marginal density of the values:

plot(density(vals))

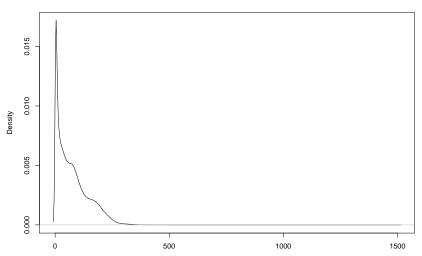


Histogram

You can also pass in a mask to most standard functions:

```
plot(density(t1, mask = t1 > 0))
```

density.default(x = x)



File helpers

Use paste if you want to put strings together with spaces:

Use paste0 if you want no spaces by default:

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

file.path(directory, filename) will paste directory and filename with file separators (e.g. \prime)

```
paste("img", ".nii.gz")
[1] "img .nii.gz"
paste0("img", ".nii.gz")
[1] "img.nii.gz"
file.path("output_directory", paste0("img", ".nii.gz"))
```

Operations with nifti objects

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

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

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
- ► 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 Packages we will use

- ms.lesion contains training/testing data of patients with multiple sclerosis (MS)
 - ▶ from the MS lesion challenge 2016 (http://iacl.ece.jhu.edu/index.php/MSChallenge)
- ▶ kirby21.t1 scan-rescan data for 3 subjects from Landman et al. (2011)
 - https://www.nitrc.org/projects/multimodal

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
- ▶ We have briefly covered subsetting and image manipulation
 - more on that later

Lists

Initialize an empty list and add two elements to it

```
l = list()
1[[1]] = v
1[[2]] = m
print(1)
[[1]]
[1] 1 4 3 7 8
[[2]]
    [,1] [,2] [,3] [,4]
[1,]
    1 4 7 10
[2,] 2 5 8 11
[3,] 3 6
                   12
```

Subsetting: