mpr— title: "General R (Reading and Writing Images)" output: ioslides_presentation: widescreen: yes css: ../styles.css beamer_presentation: default bibliography: ../refs.bib

R Basics

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 > 2 dimensions of one class. The data in a nifti object is one of these (usually 3-D)

Initializing: vectors

• Use c() to create a vector of numeric values:

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

```
[1] 1 4 3 7 8
```

Shortcut for sequential numeric vector:

```
w = 1:5
print(w)
```

```
[1] 1 2 3 4 5
```

The gray boxes denote code, and the lines after denote the output.

Assignment

In R, you can assign using the equals = or arrow <- (aka assignment operator).

The above commands are equivalent to:

```
w = 1:5
w <- 1:5
```

There are no differences in these 2 commands, but just preference (we use =).

Variable/object names:

- · must start with a letter
- cannot contain spaces, \$, quotes, or other special characters
 - o generally just alpha-numeric
- can contain periods (.) and underscores

Help

- To see the documentation for a function, use the ? symbol before the name of the function. This is a shortcut for the help command:
- $\bullet~$ For example, to see documentation for ${\,}_{\text{\tiny C}}$:

```
?c
help(topic = "c")
```

• To search for help files, use a double ?? or help.search:

```
??c
help.search(pattern = "c")
```

Some Details

- R is case sensitive (e.g. y and y are different)
- Commands separated by new line or by a colon (;)
- Use # to comment

You can also be explicit about which package you are using with the :: operator, where the syntax is package::function():

```
Processing math: 100% Processing math: 100%
```

Initializing: matrices and arrays

- Create a 3 x 4 numeric matrix and assign to variable m
 - o note items are added column-wise

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

```
[,1] [,2] [,3] [,4]
[1,] 1 4 7 10
[2,] 2 5 8 11
[3,] 3 6 9 12
```

• Create a 3 x 4 x 3 numeric array and assign to variable a

```
a = array(1:36, dim = c(3, 4, 3))
```

• the dim() function returns the dimensions of the array

```
dim(a)
```

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

Subsetting: vectors

• Subsetting a vector (first index is 1, not zero):

```
print(v)

[1] 1 4 3 7 8

print(v[4])
```

```
[1] 7
```

Processing math: 100% | rint (v[1:3])

```
[1] 1 4 3

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

[1] 1 3 8
```

Subsetting: matrices

• Subsetting a matrix - [row, column] format,

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

[1] 7

print(m[1:2,3:4])

[,1] [,2]
[1,] 7 10
[2,] 8 11
```

• if row or column missing then all values printed:

```
print(m[,4])

[1] 10 11 12

print(m[2,])

[1] 2 5 8 11
```

Subsetting: arrays

• Subsetting - [x,y,z] format:

```
print(a[1,1,1])
 [1] 1
 dim(a[,4,])
 [1] 3 3
This will return an error - need to specify all dims:
 a[,4]
Operators in R: return numeric
   • Arithmetic: +, -, *, /, ^ - exponents
   • Standard math functions: log, abs, sqrt
 print(v); print(w)
 [1] 1 4 3 7 8
 [1] 1 2 3 4 5
 print(v + 4)
 [1] 5 8 7 11 12
 print(v + w)
 [1] 2 6 6 11 13
 print(sqrt(w^2))
```

```
[1] 1 2 3 4 5
```

Operators in R: return logical

• Comparison: >, >=, <, <=, == (equals), != (not equal)

```
• Logical: ! - not, & - and, | - or (a "pipe")
 • all(): function to test all values TRUE and any(): (are any)
print(!FALSE)
[1] TRUE
print(TRUE | FALSE)
[1] TRUE
print(FALSE & FALSE)
[1] FALSE
c(all(c(TRUE, FALSE)), any(c(TRUE, FALSE)))
[1] FALSE TRUE
```

Subsetting with logicals

The ${\tt which}$ command takes a logical and gets the indices of ${\tt TRUE}$:

```
which(v > 5)

[1] 4 5
```

Processing math: 100% which (v > 5)

```
[1] 7 8
```

Or directly pass in a vector of logicals to subset:

```
v[v > 5]
```

```
[1] 7 8
```

This method will be useful later when we are working with images.

Imaging Packages in R

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).

```
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

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

t1

```
NIfTI-1 format

Type : nifti
Data Type : 4 (INT16)
Bits per Pixel : 16
Slice Code : 0 (Unknown)
Intent Code : 0 (None)
Qform Code : 2 (Aligned_Anat)
Sform Code : 1 (Scanner_Anat)
Dimension : 408 x 512 x 152
Pixel Dimension : 0.43 x 0.43 x 0.82
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
```

```
NIfTI-1 format

Type : nifti
Data Type : 4 (INT16)
Bits per Pixel : 16
Slice Code : 0 (Unknown)
Intent Code : 0 (None)
Qform Code : 2 (Aligned_Anat)
Sform Code : 1 (Scanner_Anat)
Dimension : 408 x 512 x 152
Pixel Dimension : 0.43 x 0.43 x 0.82
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

```
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] 402 412 435 448 453 430
```

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,] 180 258 1

[2,] 175 259 1

[3,] 176 259 1

[4,] 177 259 1

[5,] 178 259 1

[6,] 179 259 1
```

But can get the "vector" indices as well:

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

[1] 105036 105439 105440 105441 105442 105443
```

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] 1691
```

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
```

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.frame s (covered later) when you want to do modeling at a voxel level.

File helpers - for constructing filenames

Use paste if you want to put strings together with spaces, paste0 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

Processing math: 100% | fslr - wraps FSL commands to use in R

- · registration, image manipulation, skull stripping
- ANTSR wrapper for Advanced normalization tools (ANTs) code
 - o 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 the MS lesion challenge 2016 (http://iacl.ece.jhu.edu/index.php/MSChallenge (http://iacl.ece.jhu.edu/index.php/MSChallenge))

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 (../index.html)