

5: EDA, Big Jobs, Automation

bit.ly/SISBID3

Exploratory data analysis

- ▶ what were you trying to do?
- ▶ what you're thinking about?
- ▶ what did you observe?
- ▶ what did you conclude, and why?

Avoid

- ▶ "How did I create this plot?"
- ▶ "Why did I decide to omit those six samples?"
- ▶ "Where (on the web) did I find these data?"
- ▶ "What was that interesting gene?"

Basic principles

Step 1: slow down and document.

Step 2: have sympathy for your future self.

Step 3: have a system.

Capturing EDA

- ▶ copy-and-paste from an R file
- ▶ grab code from the `.Rhistory` file
- ▶ Write an informal R Markdown file; make use of the “R Notebook” features.
- ▶ Write code for use with the KnitR function `spin()`
 - Comments like `#' This will become text`
 - Chunk options like so: `#+ chunk_label, echo=FALSE`

A file to spin()

```
#' This is a simple example of an R file for use with spin().  
  
#' We'll start by setting the seed for the RNG.  
set.seed(53079239)  
  
#' We'll first simulate some data with  $x \sim N(\mu=10, \sigma=5)$  and  
#'  $y = 2x + e$ , where  $e \sim N(\mu=0, \sigma=2)$   
x <- rnorm(100, 10, 5)  
y <- 2*x + rnorm(100, 0, 2)  
  
#' Here's a scatterplot of the data.  
plot(x, y, pch=21, bg="slateblue", las=1)
```

Activity

Try out `knitr::spin()`:

- ▶ Create an R script using
 - `#'` for text
 - `#+` for chunk options
- ▶ Compile the script to HTML using `knitr::spin()`
- ▶ Open the resulting HTML file.

Big jobs

- ▶ You don't want `knitr` running for a year.
- ▶ You don't want to re-run things if you don't have to.

Biggish jobs in knitr

- ▶ Manual caching
- ▶ Built-in `cache=TRUE`
- ▶ Split the work into a separate subdirectory.

Manual caching

```
```{r a_code_chunk}
file <- "cache/myfile.RData"

if(file.exists(file)) {
 load(file)
} else{

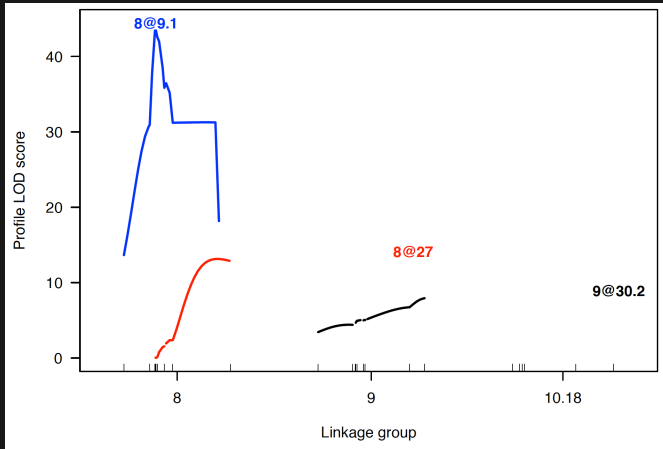
 save(object1, object2, object3, file=file)
}
```
```

Chunk references

```
```{r not_shown, eval=FALSE}  
code_here <- 0
```  
  
```{r a_code_chunk, echo=FALSE}  
file <- "cache/myfile.RData"

if(file.exists(file)) {
 load(file)
} else{
<<not_shown>>
 save(code_here, file=file)
}
```
```

A cache gone bad



Knitr's cache system

```
```{r chunk_name, cache=TRUE}  
load("a_big_file.RData")
med <- apply(object, 2, median, na.rm=TRUE)
```
```

- ▶ Chunk is re-run if edited.
- ▶ Otherwise, objects from previous run are loaded.
- ▶ Don't cache things with side effects
e.g., `options()`, `par()`

Cache dependencies

Manual dependencies

```
```{r chunkA, cache=TRUE}  
Sys.sleep(2)
x <- 5
```\n\n```{r chunkB, cache=TRUE, dependson="chunkA"}  
Sys.sleep(2)  
y <- x + 1  
```\n\n```{r chunkC, cache=TRUE, dependson="chunkB"}  
Sys.sleep(2)
z <- y + 1
```
```

Cache dependencies

Automatic dependencies

```
```{r setup, include=FALSE}  
opts_chunk$set(autodep = TRUE)
dep_auto()
```
```

Parallel computing

If your computer has multiple processors, use `library(parallel)` to make use of them.

- ▶ `detectCores()`
- ▶ `RNGkind("L'Ecuyer-CMRG")` and `mclapply` (Unix/Mac)
- ▶ `makeCluster`, `clusterSetRNGStream`, `clusterApply`, and `stopCluster` (Windows)

Systems for distributed computing

- ▶ At UW-Madison, we use HTCondor
- ▶ There are oddles of similar systems
- ▶ "By hand"

Simulations

- ▶ Computer simulations require RNG seeds (`.Random.seed` in R).
- ▶ Multiple parallel jobs need different seeds.
- ▶ Don't rely on the current seed, or on having it generated from the clock.
- ▶ Use something like `set.seed(91820205 + i)`
- ▶ An alternative is create a big batch of simulated data sets in advance.

Save everything

- ▶ RNG seeds
- ▶ input
- ▶ output
- ▶ version numbers, with `sessionInfo()`
- ▶ raw results
- ▶ script to combine results
- ▶ combined results
- ▶ ReadMe describing the point

Compartmentalize big jobs

- ▶ Separate directory for each batch of big computations.
- ▶ Collect the results in `‘.RData’` or `‘.rds’` files.
- ▶ KnitR-based documents for the analysis/use of those results.

Potential problems

- ▶ Forgetting `save()` in your distributed jobs
- ▶ A bug in the `save()` command
- ▶ Keeping things synchronized
 - Have you re-run the big jobs when upstream data were revised?

Big jobs summary

- ▶ Careful organization and modularization.
- ▶ Save everything.
- ▶ Document everything.
- ▶ Learn the basic skills for distributed computing.

Activity

Try out the knitr cache system.

- ▶ Create an RMarkdown document with multiple dependent chunks.
- ▶ Maybe add `Sys.sleep(5)` to slow things down, as if the jobs were taking a while.
- ▶ Compile the document.
- ▶ Edit a chunk.
- ▶ Re-compile the document and see what was actually run and what was taken from cache.

Automate the process (GNU Make)

```
R/analysis.html: R/analysis.Rmd Data/cleandata.csv
```

```
cd R;R -e "rmarkdown::render('analysis.Rmd')"
```

```
Data/cleandata.csv: R/prepData.R RawData/rawdata.csv
```

```
cd R;R CMD BATCH prepData.R
```

```
RawData/rawdata.csv: Python/xls2csv.py RawData/rawdata.xls
```

```
Python/xls2csv.py RawData/rawdata.xls > RawData/rawdata.csv
```


Automate the process (GNU Make)

```
R/analysis.html: R/analysis.Rmd Data/cleandata.csv  
    cd R;R -e "rmarkdown::render('analysis.Rmd')"  
  
Data/cleandata.csv: R/prepData.R RawData/rawdata.csv  
    cd R;R CMD BATCH prepData.R  
  
RawData/rawdata.csv: Python/xls2csv.py RawData/rawdata.xls  
    Python/xls2csv.py RawData/rawdata.xls > RawData/rawdata.csv
```

Automate the process (GNU Make)

```
R/analysis.html: R/analysis.Rmd Data/cleandata.csv  
    cd R;R -e "rmarkdown::render('analysis.Rmd')"  
  
Data/cleandata.csv: R/prepData.R RawData/rawdata.csv  
    cd R;R CMD BATCH prepData.R  
  
RawData/rawdata.csv: Python/xls2csv.py RawData/rawdata.xls  
    Python/xls2csv.py RawData/rawdata.xls > RawData/rawdata.csv
```

Automate the process (GNU Make)

```
R/analysis.html: R/analysis.Rmd Data/cleandata.csv  
    cd R;R -e "rmarkdown::render('analysis.Rmd')"  
  
Data/cleandata.csv: R/prepData.R RawData/rawdata.csv  
    cd R;R CMD BATCH prepData.R  
  
RawData/rawdata.csv: Python/xls2csv.py RawData/rawdata.xls  
    Python/xls2csv.py RawData/rawdata.xls > RawData/rawdata.csv
```

Automation with GNU Make

- ▶ Make is for more than just compiling software
- ▶ The **essence** of what we're trying to do
- ▶ Automates a workflow
- ▶ Documents the workflow
- ▶ Documents the dependencies among data files, code
- ▶ Re-runs only the necessary code, based on what has changed

Fancier example

```
FIG_DIR = Figs

mypaper.pdf: mypaper.tex $(FIG_DIR)/fig1.pdf $(FIG_DIR)/fig2.pdf
    pdflatex mypaper

# One line for both figures
$(FIG_DIR)/%.pdf: R/%.R
    cd R;R CMD BATCH $(<F)

# Use "make clean" to remove the PDFs
clean:
    rm *.pdf Figs/*.pdf
```

Installing Make

- ▶ On Macs, Make should be installed. Type `make --version` to check.
- ▶ On Windows, probably the easiest is to install **Rtools**, which includes Make.
`cran.r-project.org/bin/windows/Rtools`

How do you use Make?

- ▶ If you name your make file `Makefile`, then just go into the directory containing that file and type `make`
- ▶ If you name your make file `something.else`, then type `make -f something.else`
- ▶ Actually, the commands above will build the **first** target listed in the make file. So I'll often include something like the following.

```
all: target1 target2 target3
```

Then typing `make all` (or just `make`, if `all` is listed first in the file) will build all of those things.

- ▶ To be build a specific target, type `make target`. For example, `make Figs/fig1.pdf`