Big jobs/simulations Tools for Reproducible Research

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- Write a test or two
- Commit it to a git repository

So what's the big deal?

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

Unix basics

```
nice +19 R CMD BATCH input.R output.txt &
fg
ctrl-Z
bg
ps ux
top
kill
kill -9
```

Disk thrashing

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Wikipedia

Disk thrashing

In computer science, thrashing occurs when a computer's virtual memory subsystem is in a constant state of paging, rapidly exchanging data in memory for data on disk, to the exclusion of most application-level processing.

Wikipedia

Biggish jobs in knitr

- Manual caching
- ► Built-in cache=TRUE
- ► Split the work and write a Makefile

Manual caching

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

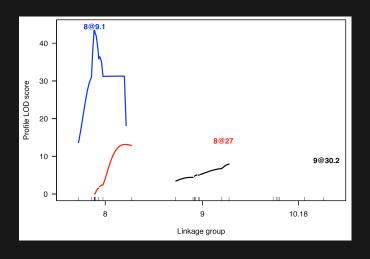
if(file.exists(file)) {
 load(file)
} else{

 save(object1, object2, object3, file=file)
}
...</pre>
```

#### 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
```{r chunkB, cache=TRUE, dependson="chunkA"}
Sys.sleep(2)
v <- x + 1
```{r chunkC, cache=TRUE, dependson="chunkB"}
Sys.sleep(2)
z < -v + 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'Eucyer-CMRG") and mclapply (Unix/Mac)
- ▶ makeCluster, clustersetRNGStream, clusterApply, and stopCluster (Windows)

## Systems for distributed computing

- HTCondor and the UW-Madison CHTC
- Other condor-like systems
- ▶ "By hand"
  - e.g., perl script + template R script

#### 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

#### One Makefile to rule them all

- Separate directory for each batch of big computations.
- Makefile that controls the combination of the results (and everything else).
- KnitR-based documents for the analysis/use of those results.

#### Potential problems

- Forgetting save() in your distributed jobs
- ▶ A bug in the save() command
- make clobbers some important results
  - Scripts should refuse to overwrite output files

## Summary

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