Introduction to parallel computing with R

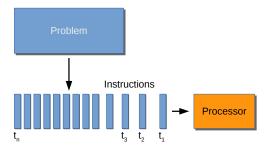
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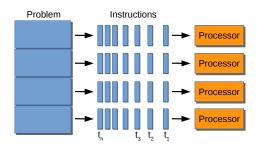
Serial computation

- Traditionally, software has been written for serial computation
 - ▶ A problem is broken into a **discrete series** of instructions
 - Instructions are executed sequentially one after another on a single processor
 - Only one instruction may execute at any moment in time



Parallel computing

- Parallel computing is the simultaneous use of multiple compute resources to solve a computational problem
 - A problem is broken into discrete parts that can be solved concurrently
 - Each part is further broken down to a series of instructions which execute simultaneously on different processors
 - ► An overall control/coordination mechanism is employed



Motivation for parallel computing

- Consider the case when you need to repeat a computation, or a series of computations, many (many) times, or/and when those individual computations are time-consuming, e.g.:
 - Running a simulation model using multiple different parameter sets,
 - Running multiple MCMC chains simultaneously,
 - Carrying out bootstrapping, cross-validation, etc.

It takes time...

- ► Nowadays (almost) all computers have multicore processors
- As long as the computations do not need to communicate, they
 can be spread across multiple cores and executed in
 parallel, thus reducing computation time



Parallel computing in R

- The package parallel
 - ▶ The **foundational package** for parallel computing in R
 - Comes pre-installed with base R in recent R versions (since R 2.14.0)
 - Builds on multicore (works for unix-alikes) & snow (works for Winblows)
 - Provides parallel apply functions
- The package doParallel is a parallel backend for the foreach package which enable the execution of for loops in parallel
- Some task specific packages include an option for parallel computation: e.g. boot, caret, pls, plyr

How many cores do you have on your computer?

```
library(parallel)
detectCores()
```

[1] 8

Create a parallel backend

You need to create a cluster with desired number of cores

cl <- makeCluster(2)</pre>

Execute computation in parallel using parallel apply functions

```
parLapply(cl, X, FUN, ...)
parSapply(cl, X, FUN, ...)
parApply(cl, X, MARGIN, FUN, ...)
```

Execute computation in parallel using foreach

```
Don't forget to:

* load the `doParallel` package
* register your cluster before the computations
library(doParallel)
registerDoParallel(cl)
x <- foreach(..., .combine) %dopar% {
}</pre>
```

Stop your cluster

In any case don't forget to stop your cluster when you're done
stopCluster(cl)

Exercises

Exercise 1 Column means of a large matrix

Consider a matrix of 10 rows \times 1 million columns with normally distributed data of mean 0 and variance 1

- Generate this matrix
- Compute the column means using apply, parApply and colMeans
- Record and compare the computation times

One solution

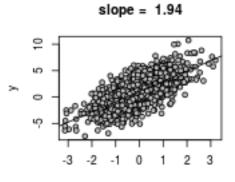
```
mymat <- matrix(rnorm(10*1000000), nrow = 10, ncol = 1000000)</pre>
system.time(out <- apply(mymat, 2, mean))</pre>
##
      user system elapsed
     5.640 0.008 5.648
##
library(parallel)
cl <- makeCluster(4)</pre>
system.time(out <- parApply(cl, mymat, 2, mean))</pre>
##
      user system elapsed
     2.696
             0.128 4.453
##
system.time(out <- colMeans(mymat))</pre>
##
      user system elapsed
     0.008
             0.000 0.009
##
```

Exercise 2 Bootstrap confidence interval of a regression slope

You're interested in computing a 95% confidence interval of the slope of the following regression

```
x <- rnorm(n = 1000, mean = 0, sd = 1)
y <- rnorm(n = 1000, mean = 1 + 2 * x, sd = 2)
mydata <- data.frame(x, y)
myreg <- lm(y ~ x, data = mydata)</pre>
```

Exercise 2 Bootstrap confidence interval of a regression slope



Exercise 2 Bootstrap confidence interval of a regression slope

- Reproduce the dataset, the regression and the graph
- Generate 1000 boostrap samples (function sample())
- ► Compute the regression slope within each bootsrap sample
- ▶ Use the quantile() function to get a 95% confidence interval
- Do the analysis both in serial and parallel
- Record and compare the computation times

for loop solution

```
a <- proc.time()
boot_a <- rep(NA, 1000)
for(i in 1:1000) {
  bootstrap_data <- mydata[sample(nrow(mydata), nrow(mydata),</pre>
                                   replace=TRUE), ]
  boot_a[i] <- unname(lm(y ~ x,bootstrap_data)$coef[2])</pre>
c(quantile(boot a, c(0.025, 0.975)))
## 2.5% 97.5%
## 1.8113 2.0619
proc.time() - a
##
      user system elapsed
     1.316 0.000 1.316
##
```

foreach loop solution

user system elapsed

0 004 0 000 0 005

##

library(doParallel) ## Error in library(doParallel): there is no package called 'doParallel' cl <- makeCluster(4)</pre> registerDoParallel(cl = cl) ## Error in eval(expr, envir, enclos): could not find function "registerDoParal a <- proc.time() boot b <- foreach(i = 1:1000, .combine=c) %dopar% { bootstrap_data <- mydata[sample(nrow(mydata), nrow(mydata),</pre> replace=TRUE),] unname(lm(y ~ x, bootstrap_data)\$coef[2]) ## Error in eval(expr, envir, enclos): could not find function "%dopar%" stopCluster(cl) c(quantile(boot_b, c(0.025, 0.975))) ## Error in quantile(boot_b, c(0.025, 0.975)): object 'boot_b' not found proc.time() - a

boot solution

boot solution

```
boot.ci(boot_c, type = "perc")

## BOOTSTRAP CONFIDENCE INTERVAL CALCULATIONS
## Based on 1000 bootstrap replicates
##
## CALL:
## boot.ci(boot.out = boot_c, type = "perc")
##
## Intervals:
## Level Percentile
## 95% ( 1.810,  2.063 )
## Calculations and Intervals on Original Scale
```

Exercise 3 Predicted Residual Sum of Squares (PRESS)

The PRESS is a statistic used to assess the predictive ability of a model : PRESS = $\sum_{i=1}^{n} (y_i - \hat{y}_{i,-i})^2$ (Allen, 1971)

- \Leftrightarrow the prediction residual sum of squares within a leave-one-out cross-validation.
 - Consider the regression from exercise 2, can you compute it's PRESS?
 - Use both for and foreach loops for the computations and compare the corresponding execution times
 - ► How fast is your code in comparison to the function PRESS of the MPV library?

for loop solution

```
a <- proc.time()
pred <- rep(NA, nrow(mydata))</pre>
for (i in 1:nrow(mydata)){
  dat <- mydata[- i, ]</pre>
  mod \leftarrow lm(y \sim x, data = dat)
  pred[i] <- predict(object = mod, mydata[i, ])</pre>
sum((pred - mydata$y)^2)
## [1] 4262.1
proc.time() - a
     user system elapsed
##
     1.732
              0.012 1.742
##
```

foreach loop solution

```
library(doParallel)
```

user system elapsed

0 000 0 000 0 007

##

```
## Error in library(doParallel): there is no package called 'doParallel'
cl <- makeCluster(4)</pre>
registerDoParallel(cl = cl)
## Error in eval(expr, envir, enclos): could not find function "registerDoParal
a <- proc.time()
pred <- foreach(i = 1:nrow(mydata), .combine = c) %dopar% {</pre>
  dat <- mydata[- i, ]</pre>
  mod \leftarrow lm(y \sim x, data = dat)
  predict(object = mod, mydata[i, ])
## Error in eval(expr, envir, enclos): could not find function "%dopar%"
stopCluster(cl)
sum((pred - mydata$y)^2)
## [1] 4262.1
proc.time() - a
```

PRESS function from MPV package

```
library(MPV)
## Error in library(MPV): there is no package called 'MPV'
a <- proc.time()
myreg <- lm(y ~ x, data = mydata)
PRESS (myreg)
## Error in eval(expr, envir, enclos): could not find func-
proc.time() - a
     user system elapsed
##
    0.004 0.000 0.005
##
```

References

- ▶ Blaise Barney. Introduction to Parallel Computing (2016). https://computing.llnl.gov/tutorials/parallel_comp/
- ▶ Clint Leach. Introduction to parallel computing in R (2014). http://michaeljkoontz.weebly.com/uploads/1/9/9/4/19940979/parallel.pdf
- ▶ Steve Weston & Rich Calaway. Getting Started with doParallel and foreach (2015). https://cran.rproject.org/web/packages/doParallel/vignettes/gettingstartedParallel.pdf
- ▶ Dirk Eddelbuettel. CRAN Task View: High-Performance and
- Parallel Computing with R (2016). https://cran.rproject.org/web/views/HighPerformanceComputing.html
 - ▶ Roger D Peng. R Programming for Data Science (2016). https://bookdown.org/rdpeng/rprogdatascience/ ▶ Colin Gillespie & Robin Lovelace. Efficient R programming
 - (2016). https://bookdown.org/csgillespie/efficientR/