PM566 Lab 9

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Problem 1

Three problems that can be solved by parallel computing are: 1. Cross-validation for testing machine learning models (i.e. structure predictions for macromolecules) 2. Resampling-based testing for very large datasets for hypothesis testing 3. Text mining for large datasets

Problem 2

Part 1

```
library(parallel)
library(microbenchmark)
fun1 \leftarrow function(n = 100, k = 4, lambda = 4) {
  x <- NULL
  for (i in 1:n)
    x <- rbind(x, rpois(k, lambda))</pre>
  return(x)
}
fun1alt \leftarrow function(n = 100, k = 4, lambda = 4) {
  x = matrix(rpois(n*k, lambda), ncol = k)
  return(x)
}
microbenchmark::microbenchmark(
  fun1(),
  fun1alt()
## Unit: microseconds
##
                   min
                                              median
         expr
                              lq
                                      mean
                                                           uq
                                                                     max neval
##
       fun1() 296.876 327.3135 550.25273 342.9175 355.355 21313.709
                                                                            100
```

100

fun1alt() 18.667 19.8545 39.16522 20.9380 22.501 1765.959

Part 2

```
set.seed(1234)
M = matrix(runif(12), ncol = 4)
# Find each column's max value
fun2 <- function(x) {</pre>
 apply(x, 2, max)
fun2(M)
## [1] 0.6222994 0.8609154 0.6660838 0.6935913
fun2alt <- function(x) {</pre>
  idx = max.col(t(x))
 x[cbind(idx,1:4)]
fun2alt(M)
## [1] 0.6222994 0.8609154 0.6660838 0.6935913
# Benchmarking
microbenchmark::microbenchmark(
 fun2(M),
 fun2alt(M)
## Unit: microseconds
##
                 min
                                  mean median
         expr
                           lq
                                                    uq
                                                            max neval
       fun2(M) 16.292 17.3550 27.77399 17.938 20.2510 936.125
##
                                                                   100
## fun2alt(M) 12.500 12.8965 36.80780 13.584 14.7715 2298.834
                                                                   100
```

Problem 3

Part 1

```
my_boot <- function(dat, stat, R, ncpus = 1L) {

n <- nrow(dat)
  idx <- matrix(sample.int(n, n*R, TRUE), nrow=n, ncol=R)

# Making the cluster using `ncpus`
  cl = makePSOCKcluster(ncpus)
  clusterSetRNGStream(cl, 123)
  clusterExport(cl, c("stat", "dat", "idx"), envir = environment())

ans <- parLapply(cl, seq_len(R), function(i) {
    stat(dat[idx[,i], , drop=FALSE])
})</pre>
```

```
# Coercing the list into a matrix
ans <- do.call(rbind, ans)

# STEP 4: GOES HERE
ans
}</pre>
```

Part 2

```
my_stat <- function(d) coef(lm(y ~ x, data=d))</pre>
set.seed(1)
n <- 500; R <- 1e4
x \leftarrow cbind(rnorm(n)); y \leftarrow x*5 + rnorm(n)
ans0 <- confint(lm(y~x))</pre>
ans1 <- my_boot(dat = data.frame(x, y), my_stat, R = R, ncpus = 2L)</pre>
t(apply(ans1, 2, quantile, c(.025,.975)))
##
                      2.5%
                                97.5%
## (Intercept) -0.1386903 0.04856752
                4.8685162 5.04351239
ans0
                     2.5 %
                               97.5 %
##
## (Intercept) -0.1379033 0.04797344
                4.8650100 5.04883353
## x
system.time(my_boot(dat = data.frame(x, y), my_stat, R = 4000, ncpus = 1L))
##
      user system elapsed
##
     0.067
            0.013 2.399
system.time(my_boot(dat = data.frame(x, y), my_stat, R = 4000, ncpus = 2L))
##
      user system elapsed
##
     0.089 0.017 1.437
```