Lecture 22: Profiling and and microbenchmarking

An order of operations for programming

- 1. Make it run
- 2. Make it right
- 3. Make it fast

When speed matters

- You are working with very large data
- You are running a process (a simulation, a data analysis, etc.) many times
- A piece of code will be called many times (e.g., choosing a split in a decision tree)

Goals

- Learn how to identify bottlenecks in code
- Learn approaches for more efficient code in R
- Time permitting: learn how to use C++ to make code faster

Example: timing code

Suppose we want to compute the mean of each column of a data frame:

```
1  n <- 100000
2  cols <- 150
3  data_mat <- matrix(rnorm(n * cols, mean = 5), ncol = cols)
4  data <- as.data.frame(data_mat)
5
6  means <- rep(NA, cols)
7  for(i in 1:cols){
8   means[i] <- mean(data[,i])
9 }</pre>
```

Example: timing code

Suppose we want to compute the mean of each column of a data frame:

```
1  n <- 100000
2  cols <- 150
3  data_mat <- matrix(rnorm(n * cols, mean = 5), ncol = cols)
4  data <- as.data.frame(data_mat)
5
6  system.time({
7    means <- rep(NA, cols)
8   for(i in 1:cols){
9    means[i] <- mean(data[,i])
10  }
11 })</pre>
```

```
user system elapsed 1.930 0.017 1.960
```

Alternatives

```
1 means <- rep(NA, cols)
2 for(i in 1:cols){
3  means[i] <- mean(data[,i])
4 }</pre>
```

What are the alternatives to this for-loop approach?

Alternatives

```
1 # Option 1: for loop
 2 for loop means <- function(data){</pre>
     cols <- ncol(data)</pre>
 4 means <- rep(NA, cols)
 5 for(i in 1:cols){
 6 means[i] <- mean(data[,i])</pre>
    return(means)
10 means <- for_loop_means(data)</pre>
11
12 # Option 2: apply
13 means <- apply(data, 2, mean)
14
15 # Option 3: colMeans
16 means <- colMeans(data)</pre>
```

Comparing performance

Microbenchmarking: Evaluating the performance of a small piece of code

```
1 bench::mark(
      means <- for loop means(data),</pre>
      means <- apply(data, 2, mean),</pre>
     means <- colMeans(data),</pre>
    check = F
 6 )
# A tibble: 3 \times 6
  expression
                                     min
                                           median `itr/sec` mem alloc
`qc/sec`
  <bch:expr>
                                <bch:tm> <bch:tm>
                                                       <dbl> <bch:byt>
<dbl>
1 means <- for loop means(data)</pre>
                                  1.93s
                                            1.93s
                                                       0.519
                                                                1.85KB
0
2 means <- apply(data, 2, mean) 2.03s</pre>
                                            2.03s
                                                       0.493 400.57MB
0.987
3 means <- colMeans(data) 461.4ms 469.34ms
                                                       2.13 114.45MB
1.07
```

Profiling

```
library(profvis)
  2 profvis({
         means <- for_loop_means(data)</pre>
         means <- apply(data, 2, mean)</pre>
         means <- colMeans(data)</pre>
     })
    profvis({
        means <- for loop means(data)</pre>
                                                                                                        1790
                                                                                                       1840
        means <- apply(data, 2, mean)</pre>
                                                                                            400.6
                                                                                                        480
                                                                                            114.5
        means <- colMeans(data)</pre>
5
    })
                                                      mean.default
mean.default
for loop means
                                                   apply
                                                                                                        colMeans
```

Space for efficiency increases?

1 colMeans

```
function (x, na.rm = FALSE, dims = 1L)
{
   if (is.data.frame(x))
        x <- as.matrix(x)
   if (!is.array(x) || length(dn <- dim(x)) < 2L)
        stop("'x' must be an array of at least two dimensions")
   if (dims < 1L || dims > length(dn) - 1L)
        stop("invalid 'dims'")
   n <- prod(dn[id <- seq_len(dims)])
   dn <- dn[-id]
   z <- if (is.complex(x))
        .Internal(colMeans(Re(x), n, prod(dn), na.rm)) + (0+li) *
        .Internal(colMeans(Im(x), n, prod(dn), na.rm))
   else .Internal(colMeans(x, n, prod(dn), na.rm))</pre>
```

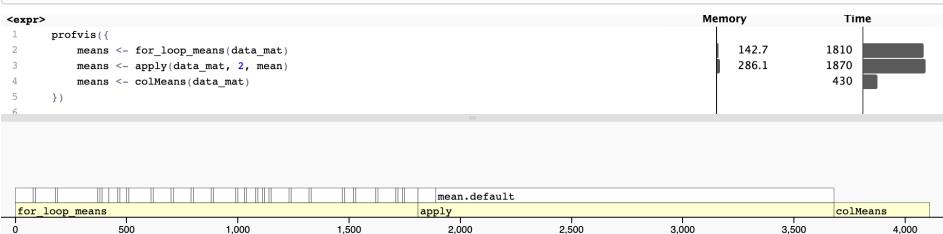
Increase efficiency by avoiding extraneous steps

```
1 n < 100000
 2 cols <- 150
 3 data mat <- matrix(rnorm(n * cols, mean = 5), ncol = cols)</pre>
 4 data <- as.data.frame(data mat)</pre>
 5
   bench::mark(
      means <- colMeans(data mat),</pre>
      means <- colMeans(data),</pre>
    check = F
10)
# A tibble: 2 \times 6
  expression
                                    min
                                           median `itr/sec` mem alloc
`qc/sec`
  <bch:expr>
                               <bch:tm> <bch:tm>
                                                      <dbl> <bch:byt>
<dbl>
1 means <- colMeans(data mat)</pre>
                                 437ms
                                            437ms
                                                       2.29
                                                                1.22KB
0
                                            453ms
                                                       2.21 114.45MB
2 means <- colMeans(data)</pre>
                            453ms
2.21
```

Profiling

Sample Interval: 10ms

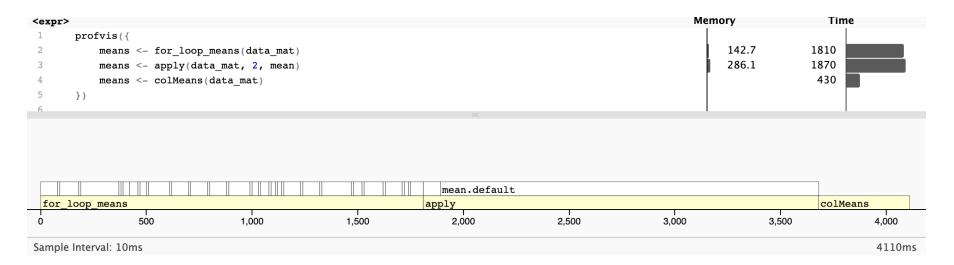
```
1 profvis({
2   means <- for_loop_means(data_mat)
3   means <- apply(data_mat, 2, mean)
4   means <- colMeans(data_mat)
5 })</pre>
```



4110ms

Profiling





Class activity

https://sta279-

f23.github.io/class_activities/ca_lecture_24.html