

(A bit of) Advanced R

Efficient R programming

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<http://github/jchiquet/CourseAdvancedR>

Resources

- Gillespie & Lovelace (2016): efficient R programming
- Wickham (2014)

Part 0: Prerequesties

- xpply family, do.call, Reduce

Outline

- 1 Part 1: Benchmark your code
- 2 Part 2: Use multiple cores in your simulations
- 3 Part 3: Be aware of what R is good at
- 4 Part 4: Remember that R is object oriented
- 5 Part 4: Interface with lower-level languages

Quick (and dirty) benchmarking with `system.time()`

One usually relies on the command `system.time(expr)` to evaluate the timings:

```
func.one <- function(n) {return(rnorm(n,0,1))}  
func.two <- function(n) {return(rpois(n,1))}  
  
n <- 1000  
system.time(replicate(100, func.one(n)))
```

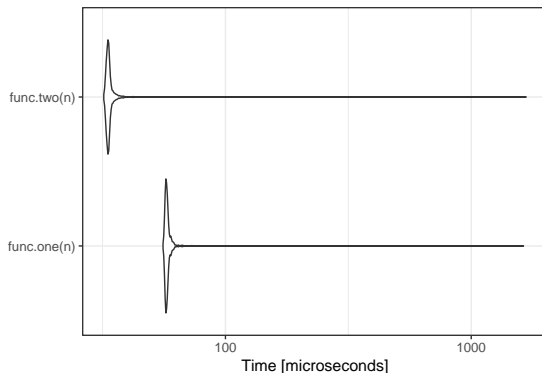
```
##      user  system elapsed  
## 0.008   0.000   0.010
```

```
system.time(replicate(100, func.two(n)))
```

```
##      user  system elapsed  
## 0.008   0.000   0.007
```

Quick benchmarking with microbenchmark

```
func.one <- function(n) {return(rnorm(n,0,1))}  
func.two <- function(n) {return(rpois(n,1))}  
  
library(microbenchmark)  
  
n <- 1000  
res <- microbenchmark(func.one(n), func.two(n), times=1000)  
ggplot2::autoplot(res)
```



Profile your code with base Rprof |

Suppose you want to evaluate which part of the following function is hot:

```
## generate data, center/scale and perform ridge regression
my_func <- function(n,p) {

  require(MASS)

  ## draw data
  x <- matrix(rnorm(n*p),n,p)
  y <- rnorm(n)

  ## center/scale
  xs <- scale(x)
  ys <- y - mean(y)

  ## return ridge's coefficients
  ridge <- lm.ridge(ys~xs+0,lambda=1)

  return(ridge$coef)
}
```

One can rely on the default Rprof function, with somewhat technical outputs

Profile your code with base Rprof II

```
Rprof(file="profiling.out", interval=0.05)
res <- my_func(1000,500)
Rprof(NULL)
```

```
summaryRprof("profiling.out")$by.self
```

##	self.time	self.pct	total.time	total.pct
## "La.svd"	0.90	75.00	0.95	79.17
## "matrix"	0.15	12.50	0.15	12.50
## "FUN"	0.05	4.17	1.20	100.00
## "aperm.default"	0.05	4.17	0.05	4.17
## "na.omit.data.frame"	0.05	4.17	0.05	4.17

```
summaryRprof("profiling.out")$by.total
```


Profile your code with base Rprof III

##	total.time	total.pct	self.time	self.pct
## "FUN"	1.20	100.00	0.05	4.17
## "block_exec"	1.20	100.00	0.00	0.00
## "call_block"	1.20	100.00	0.00	0.00
## "eval"	1.20	100.00	0.00	0.00
## "evaluate"	1.20	100.00	0.00	0.00
## "evaluate_call"	1.20	100.00	0.00	0.00
## "evaluate::evaluate"	1.20	100.00	0.00	0.00
## "handle"	1.20	100.00	0.00	0.00
## "in_dir"	1.20	100.00	0.00	0.00
## "knit"	1.20	100.00	0.00	0.00
## "knitr::knit"	1.20	100.00	0.00	0.00
## "lapply"	1.20	100.00	0.00	0.00
## "my_func"	1.20	100.00	0.00	0.00
## "process_file"	1.20	100.00	0.00	0.00
## "process_group"	1.20	100.00	0.00	0.00
## "process_group.block"	1.20	100.00	0.00	0.00
## "rmarkdown::render"	1.20	100.00	0.00	0.00
## "timing_fn"	1.20	100.00	0.00	0.00
## "withCallingHandlers"	1.20	100.00	0.00	0.00
## "withVisible"	1.20	100.00	0.00	0.00
## "lm.ridge"	1.00	83.33	0.00	0.00
## "La.svd"	0.95	79.17	0.90	75.00
## "svd"	0.95	79.17	0.00	0.00
## "matrix"	0.15	12.50	0.15	12.50
## "scale"	0.10	8.33	0.00	0.00
## "scale.default"	0.10	8.33	0.00	0.00

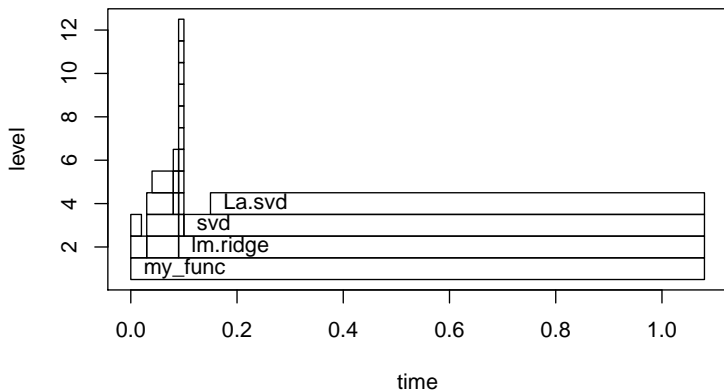
Profile your code with base Rprof IV

## "aperm.default"	0.05	4.17	0.05	4.17
## "na.omit.data.frame"	0.05	4.17	0.05	4.17
## "aperm"	0.05	4.17	0.00	0.00
## "apply"	0.05	4.17	0.00	0.00
## "eval.parent"	0.05	4.17	0.00	0.00
## ".External2"	0.05	4.17	0.00	0.00
## "model.frame.default"	0.05	4.17	0.00	0.00
## "na.omit"	0.05	4.17	0.00	0.00
## "stats::model.frame"	0.05	4.17	0.00	0.00
## "sweep"	0.05	4.17	0.00	0.00

Profile your code with profr

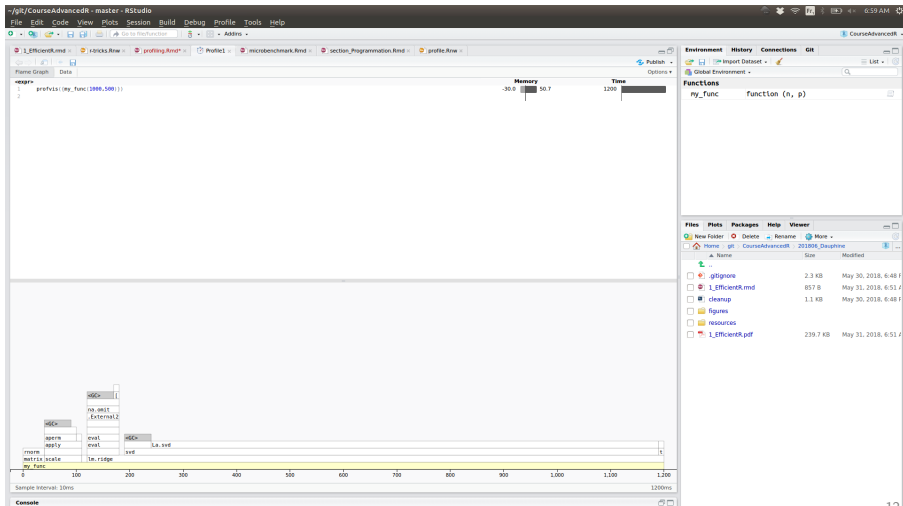
The *profr* package is maybe a little easier to understand...

```
library(profr)
profiling <- profr({my_func(1000,500)}), interval = 0.01)
plot(profiling)
```



Profvis integrates the profiling to the Rstudio API

```
library(profvis)
profvis({my_func(1000,500)})
```



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References

Gillespie, C., & Lovelace, R. (2016). *Efficient r programming*. " O'Reilly Media, Inc." Retrieved from <https://bookdown.org/csgillespie/efficientR/>

Wickham, H. (2014). *Advanced r*. CRC Press. Retrieved from <http://adv-r.had.co.nz/>