

# Speeding up R code using Rcpp and foreach packages.

Pedro Albuquerque

Universidade de Brasília

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R is a good choice as open source and free software.

But R has some **disadvantages**:

- Limited memory for big datasets.
- Inefficient standard loops.

There are some solutions for both **disadvantages**, especially to speed up:

- Using apply family functions and vectorized operations.
- Using parallel processing.
- Using C++ (Rcpp package).



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# foreach package.

Suppose we are interested in estimate a variance of a complicated function of parameters, for example, the coefficient of variation:

$$\theta = \frac{\mu}{\sigma}$$

where  $\mu$  and  $\sigma$  are the population mean and standard deviation.

We can estimate the variance of  $\theta$  using, for example, the bootstrap estimator.



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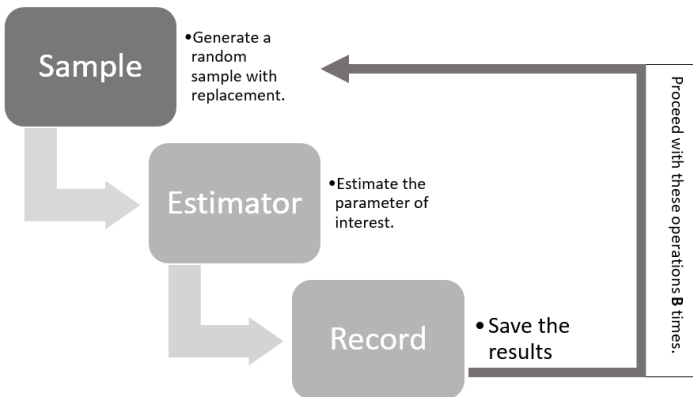
$$\theta = \frac{\mu}{\sigma}$$

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# Bootstrap.



# Bootstrap.

| Bootstrap sample (b=1,...B) | Estimator                                                                           |
|-----------------------------|-------------------------------------------------------------------------------------|
| 1                           | $\hat{\theta}_1 = \hat{\mu}_1 / \hat{\sigma}_1$                                     |
| 2                           | $\hat{\theta}_2 = \hat{\mu}_2 / \hat{\sigma}_2$                                     |
| 3                           | $\hat{\theta}_3 = \hat{\mu}_3 / \hat{\sigma}_3$                                     |
| .                           |                                                                                     |
| .                           |                                                                                     |
| B                           | $\hat{\theta}_B = \hat{\mu}_B / \hat{\sigma}_B$                                     |
| <b>MEAN</b>                 | $\bar{E}(\hat{\theta}) = \sum \hat{\theta}_b / B$                                   |
| <b>VARIANCE</b>             | $\bar{V}(\hat{\theta}) = \sum (\hat{\theta}_b - \bar{E}(\hat{\theta}))^2 / (B - 1)$ |

# Application.

```
1 #Set the seed
2 set.seed(1)
3 #Generate fake data
4 data <- rnorm(n=10000, mean=10, sd=10)
5 #True coefficient of variation
6 CV.true <- 10/10
7 #Estimated coefficient of variation
8 CV.hat <- mean(data)/sd(data)
```

Algorithm 1: Generating fake data.

The estimate is  $\hat{\theta} \approx 0.9813$ .

# Application.

```

1 #Create the bootstrap vector
2 result <- rep(NA, 100000)
3 #Start the clock
4 ptm <- proc.time()
5 for(b in 1:100000){
6   #Generate the random sample with replacement
7   ids <- sample(x=length(data), size=length(data), replace
8     =T)
9   sample <- data(ids)
10  #Calculate the CV
11  result(b) <- mean(sample)/sd(sample)
12 }
13 #Stop the clock – Time elapsed 20.40.
14 proc.time() – ptm

```



Algorithm 2: Time elapsed 20.40



# Application.

```
1 library(foreach)
2 library(doParallel)
3 #See how many cores we have
4 ncl<-detectCores()
5 #Register the cores
6 cl <- makeCluster(ncl)
7 registerDoParallel(cl)
```

Algorithm 3: Using foreach package.



# Aplication.

```

1 #Start the clock
2 ptm <- proc.time()
3 #Create the bootstrap vector
4 result <- rep(NA, 100000)
5 #Bootstrap using foreach
6 result <- foreach(b=1:100000, .combine=c) %dopar% {
7   #Generate the random sample with replacement
8   ids <- sample(x=length(data), size=length(data), replace=T)
9   sample <- data(ids)
10  #Calculate the CV
11  mean(sample)/sd(sample)
12 }
13 #Stop the clock – Time elapsed 5.32.
14 proc.time() - ptm
15 #Release the cores
16 stopCluster(cl)

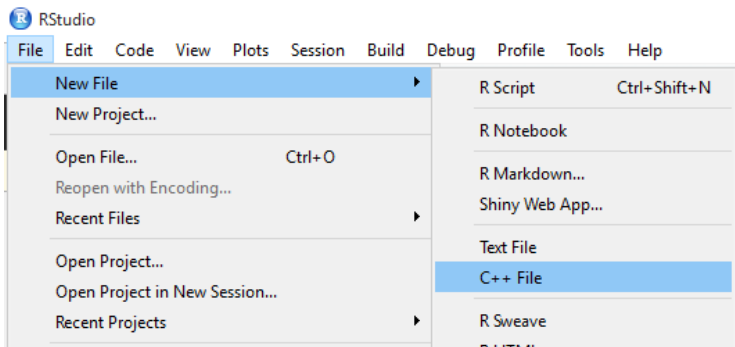
```

Algorithm 4: Time elapsed 5.32.



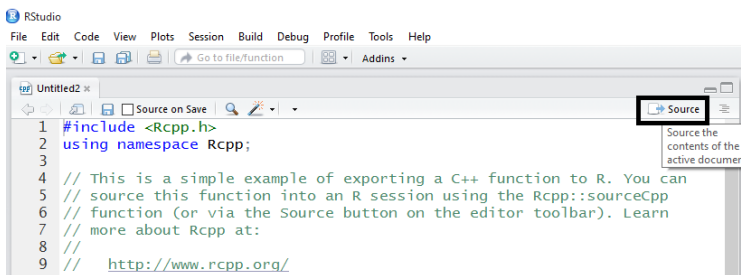
# Rcpp package.

Another solution is to use the Rcpp package:



# Rcpp package.

We need to compile the functions before use in R:



The screenshot shows the RStudio interface with a C++ code editor. The code is as follows:

```
1 #include <Rcpp.h>
2 using namespace Rcpp;
3
4 // This is a simple example of exporting a C++ function to R. You can
5 // source this function into an R session using the Rcpp::sourceCpp
6 // function (or via the Source button on the editor toolbar). Learn
7 // more about Rcpp at:
8 //
9 // http://www.rcpp.org/
```

The 'Source' button in the toolbar is highlighted with a black box, and a tooltip is visible next to it that reads: 'Source the contents of the active document'.

isPrime function.

# isPrime function.

Suppose we are interested in find if a number is prime or not.

```
1 #Create the isPrime function
2 isPrime <- function(num){
3   prime <- TRUE
4   den <- num -1
5   while(prime==TRUE & den >1){
6     #Remainder of the division
7     if (num%%den==0){
8       prime <- FALSE
9     }
10    #Decrease the number
11    den <- (den - 1)
12  }
13  return(prime)
14 }
15 #Example isPrime(12)
```

Algorithm 5: isPrime function.



isPrime function.

# isPrime function.

Using the same idea with the Rcpp:

```
1 #include <Rcpp.h>
2 using namespace Rcpp;
3 // ((Rcpp::export))
4 bool isPrimeCpp(int num) {
5     bool prime = true;
6     int den = num - 1;
7     while(prime==true & den >1){
8         //Test if is divided
9         if(num % den==0){
10            prime = false;
11        }
12        //Decrease the number
13        den = (den - 1);
14    }
15    return(prime);
16 }
```

isPrime function.

# isPrime function.

You can save your functions in a *cpp* file and invoke using:

```
1 #Rcpp library
2 library(Rcpp)
3 #Call the functions
4 sourceCpp("MyFile.cpp")
5 #Example
6 isPrimeCpp(12)
```

Algorithm 6: isPrime function.



- Create a Rcpp function to calculate the sample mean of a vector.
- Create a Rcpp function to calculate the  $k$  Fibonacci numbers.
- Use foreach and the bootstrap technique to estimate an arbitrary parameter function:  $g(\theta) = \sqrt{\theta}/2$ .