Package foreach

PARALLEL PROGRAMMING IN R



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What is foreach for?

- Developed by Rich Calaway and Steve Weston.
- Provides a new looping construct for repeated execution.
- Supports running loops in parallel.
- Unified interface for sequential and parallel processing.
- Greatly suited for embarrassingly parallel applications.

foreach looping construct

1.5117812 0.3898432 -0.6212406 -2.2146999 1.1249309

```
foreach(...) %do% ...
library(foreach)
foreach(n = rep(\frac{5}{3})) %do% rnorm(n)
[[1]]
[1] -0.6264538 0.1836433 -0.8356286 1.5952808 0.3295078
[[2]]
[1] -0.8204684 0.4874291 0.7383247 0.5757814 -0.3053884
[[3]]
```



Iteration variables

```
foreach(n = rep(\frac{5}{3}), m = \frac{10}{0:2}) %do% rnorm(n, mean = m)
```

```
[[1]]
[1] 0.3735462 1.1836433 0.1643714 2.5952808 1.3295078
[[2]]
[1] 9.179532 10.487429 10.738325 10.575781 9.694612
[[3]]
[1] 101.51178 100.38984 99.37876 97.78530 101.12493
```

Combining results

```
foreach(n = rep(\frac{5}{3}), .combine = rbind) %do% rnorm(n)
```

```
[,1] [,2] [,3] [,4] [,5]
result.1 -0.6264538 0.1836433 -0.8356286 1.5952808 0.3295078
result.2 -0.8204684 0.4874291 0.7383247 0.5757814 -0.3053884
result.3 1.5117812 0.3898432 -0.6212406 -2.2146999 1.1249309
```

```
foreach(n = rep(5, 3), .combine = '+') %do% rnorm(n)
```

```
0.06485897 1.06091561 -0.71854449 -0.04363773 1.14905030
```



List comprehension

```
foreach(x = sample(1:1000, 10), .combine = c) %:% when(x %% 3 == 0 | | x %% 5 == 0) %do% x
```

372 906 201 894 940 657 625

Let's practice!

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foreach & parallel backends

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Popular backends

- doParallel (parallel)
- doFuture (future)
- doSEQ (for consistent sequential interface)

doParallel (Rich Calaway et al.) package

- Interface between foreach and parallel
- Must register via register DoParallel() with cluster info
- Quick registration:

```
library(doParallel)
registerDoParallel(cores = 3)
```

- using multicore functionality for Unix-like systems (fork)
- using snow functionality for Windows systems

doParallel (Rich Calaway et al.) package

• Register by passing a cluster object:

```
library(doParallel)
cl <- makeCluster(3)
registerDoParallel(cl)</pre>
```

• will use *snow* functionality

Using doParallel

Sequential:

```
library(foreach)
foreach(n = rep(5, 3)) %do% rnorm(n)
```

Parallel:

```
library(doParallel)
cl <- makeCluster(3)
registerDoParallel(cl)
foreach(n = rep(5, 3)) %dopar% rnorm(n)</pre>
```

```
[[1]]
[1] -1.1671919 -0.0360007 -0.5972832 1.0380735 -0.0508561

[[2]]
[1] 0.370006 -0.419358 0.131176 0.656627 -0.037162

[[3]]
[1] 0.987222 -1.169738 0.399277 -0.155607 -1.034571
```

doFuture (Henrik Bengtsson) package

- On top of the **future** package
- How to plan the future:
 - sequential
 - cluster
 - multicore
 - multiprocess
- future.batchtools: run processes on HPC clusters (Torque, Slurm, SGE etc.)

Using doFuture

```
library(doFuture)
registerDoFuture()
```

Cluster plan:

```
plan(cluster, workers = 3)
foreach(n = rep(5, 3)) %dopar% rnorm(n)
```

Using doFuture

Multicore plan:

```
plan(multicore)
foreach(n = rep(5, 3)) %dopar% rnorm(n)
```

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Packages future and future.apply

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future package

- Developed by Henrik Bengtsson (now also funded by R Consortium)
- Uniform way to evaluate R expressions asynchronously
- Provides a unified API for sequential and parallel processing of R expressions
- Processing via a construct called **future**
- An abstraction for a value that may be available at some point in the future

Example in plain R:

```
x <- mean(rnorm(n, 0, 1))
y <- mean(rnorm(n, 10, 5))
print(c(x, y))</pre>
```

Via implicit futures:

```
x %<-% mean(rnorm(n, 0, 1))
y %<-% mean(rnorm(n, 10, 5))
print(c(x, y))</pre>
```

Via explicit futures:

```
x <- future(mean(rnorm(n, 0, 1)))
y <- future(mean(rnorm(n, 10, 5)))
print(c(value(x), value(y)))</pre>
```

Sequential and parallel futures

Sequential:

```
plan(sequential)
x %<-% mean(rnorm(n, 0, 1))
y %<-% mean(rnorm(n, 10, 5))
print(c(x, y))</pre>
```

Parallel:

```
plan(multicore)
x %<-% mean(rnorm(n, 0, 1))
y %<-% mean(rnorm(n, 10, 5))
print(c(x, y))</pre>
```

future.apply package

- Developed by Henrik Bengtsson
- Provide parallel API for all the apply functions in base R using futures
- Sibling to foreach
- Functions: future_lapply(), future_sapply(), future_apply(),...

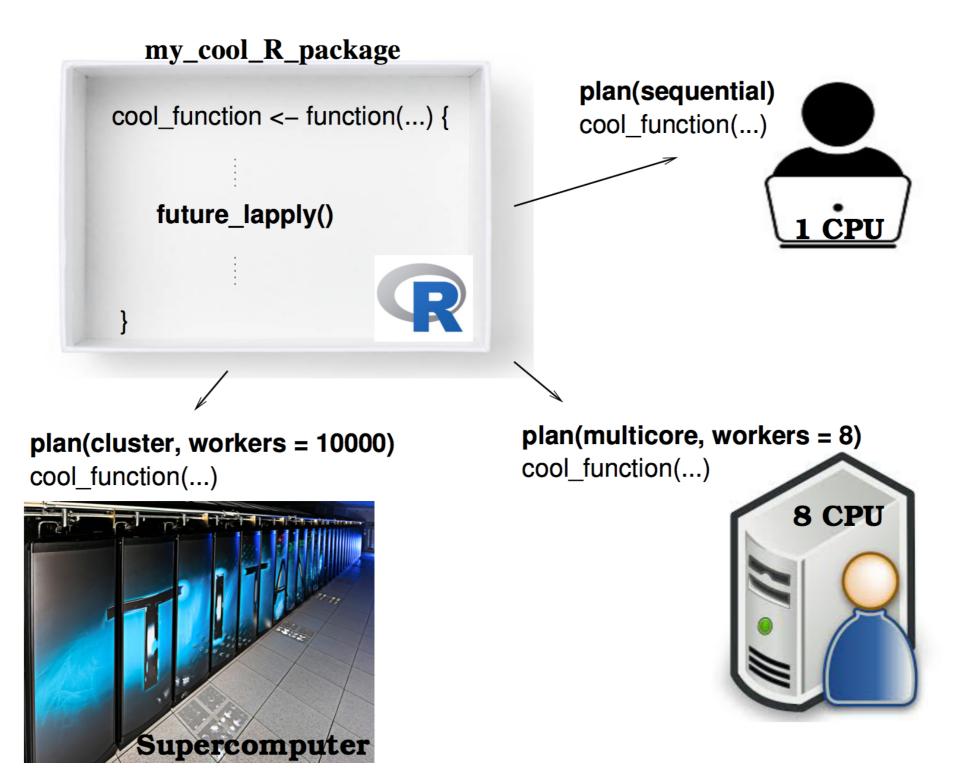
Example of future.apply

```
Using lapply():
```

```
lapply(1:10, rnorm)
Using future_lapply() sequentially:
 plan(sequential)
 future_lapply(1:10, rnorm)
Using future_lapply() on a cluster:
```

```
plan(cluster, workers = 4)
future_lapply(1:10, rnorm)
```





Let's practice!

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Load balancing and scheduling

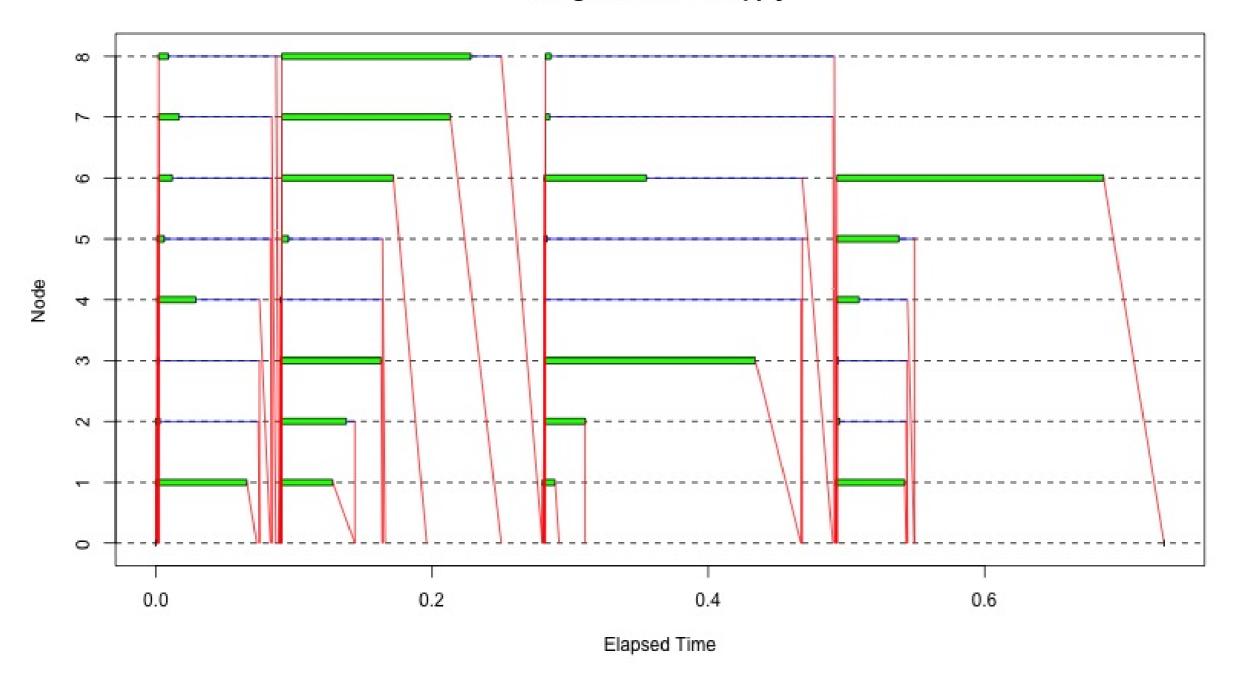
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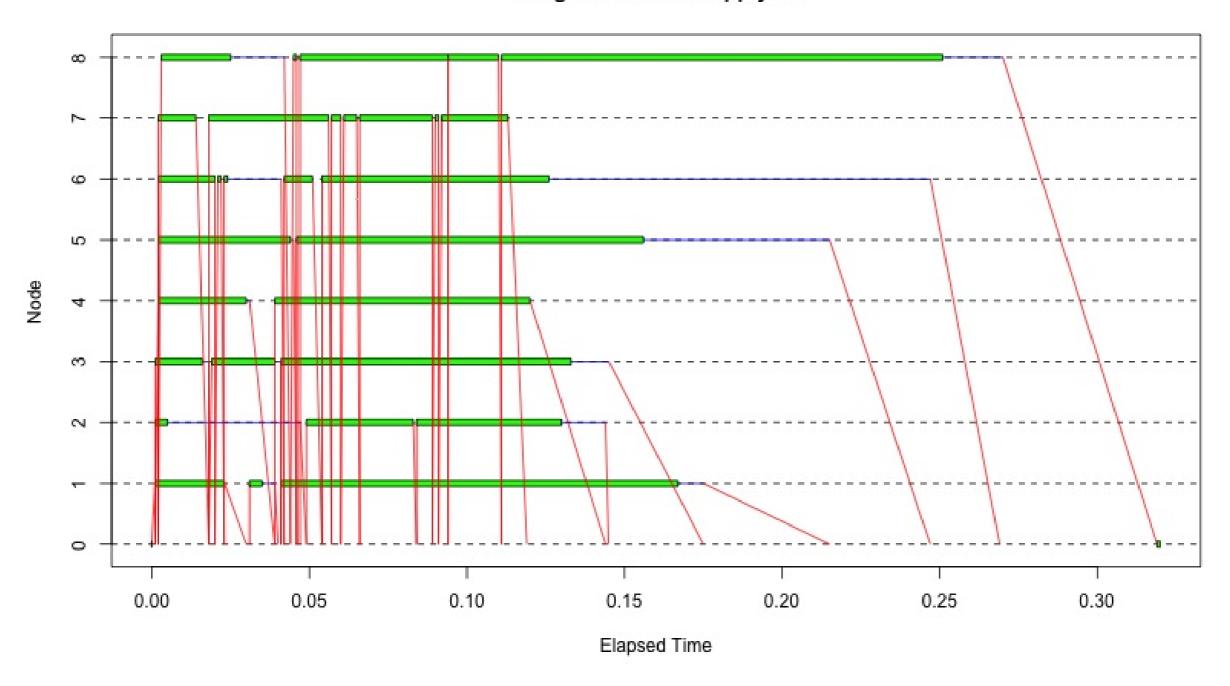


Usage with clusterApply



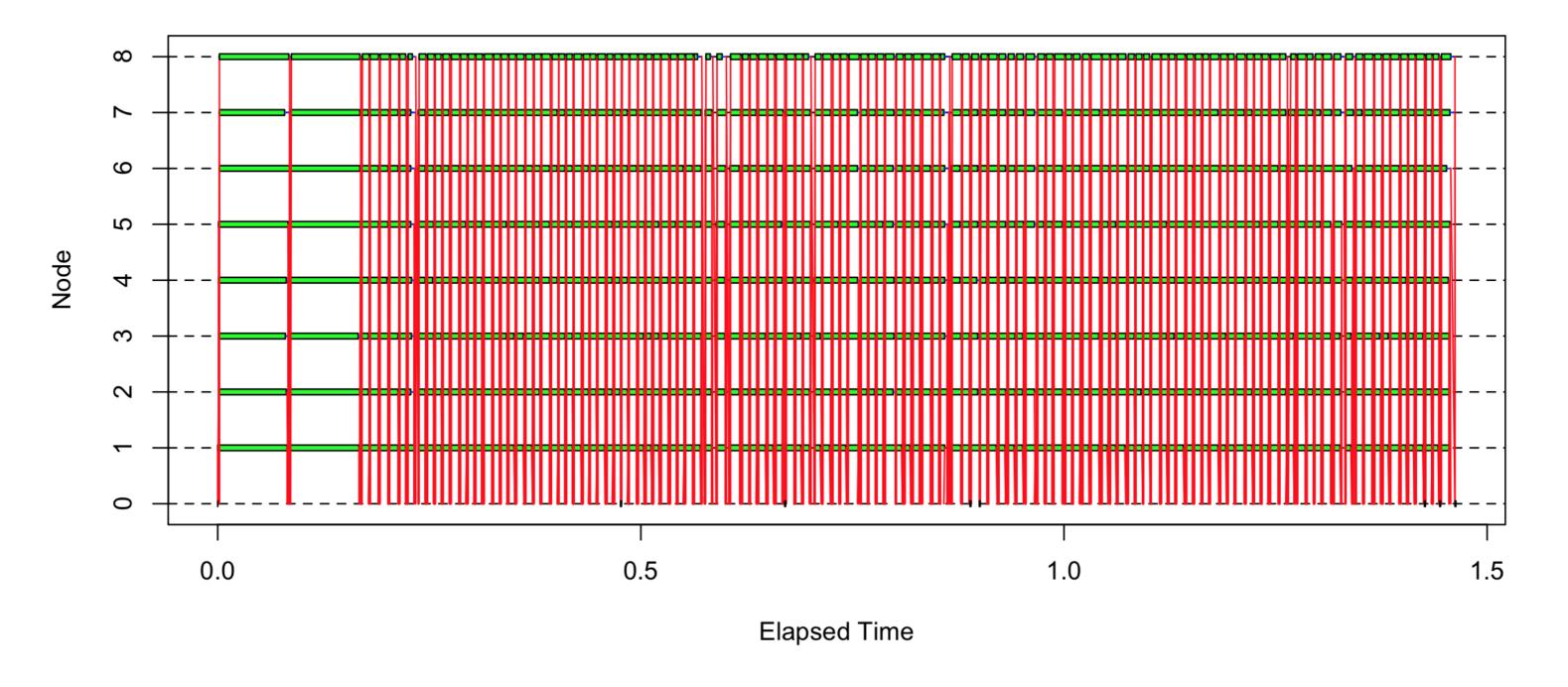


Usage with clusterApplyLB



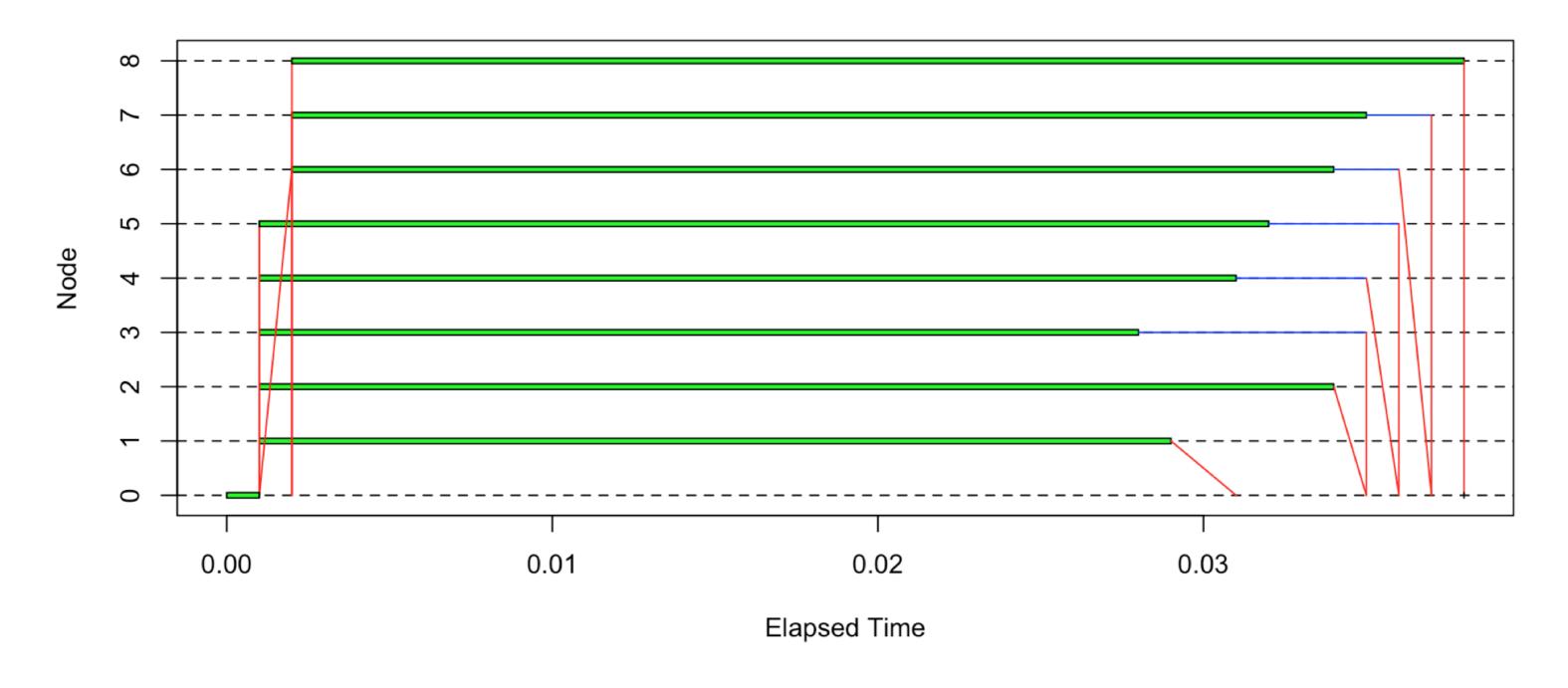


1000 tasks





8 chunks with 1000 tasks





Group 10 tasks into 2 chunks using the parallel package:

```
splitIndices(10, 2)
[[1]]
[1] 1 2 3 4 5
[[2]]
[1] 6 7 8 9 10
clusterApply(cl, x = splitIndices(10, 2), fun = sapply, "*", 100)
[[1]]
[1] 100 200 300 400 500
[[2]]
[1] 600 700 800 900 1000
```

Built into functions parApply() and friends (arg. chunk.size for R >= 3.5)

How to chunk in foreach and future.apply?

For foreach, use functions from the itertools package, e.g.:

```
foreach(s = isplitVector(1:10, chunks = 2)) %dopar% sapply(s, "*", 100)
```

For future.apply, use argument future.scheduling, e.g.

• one chunk per worker (default):

```
future_sapply(1:10, `*`, 100, future.scheduling = 1)
```

• One chunk per task:

```
future_sapply(1:10, `*`, 100, future.scheduling = FALSE)
```

Let's practice!

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