

Intro to Parallel Computing in R

Kevin Lee

Department of Statistics
Western Michigan University

February 22, 2019

- 1 List of Useful R Packages
- 2 Introduction to Parallel Computing in R

List of Useful R Packages

Some of the top most downloaded R packages:

- Check <https://support.rstudio.com/hc/en-us/articles/201057987-Quick-list-of-useful-R-packages>.

- 1 List of Useful R Packages
- 2 Introduction to Parallel Computing in R

Why Parallel Computing?

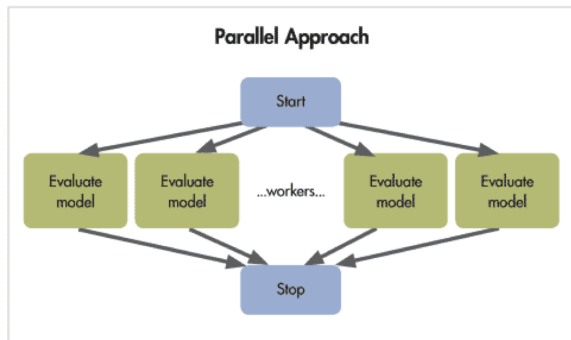
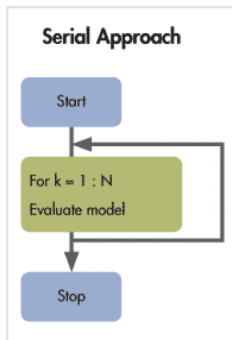
- Many statistical analysis tasks are computationally intensive.
- But at the same time, many problems are “embarrassingly parallel”.
- And often we have multiple cores in our computer!
- However, R only uses a single core.

Embarrassingly Parallel Problems

Easy to speed things up when:

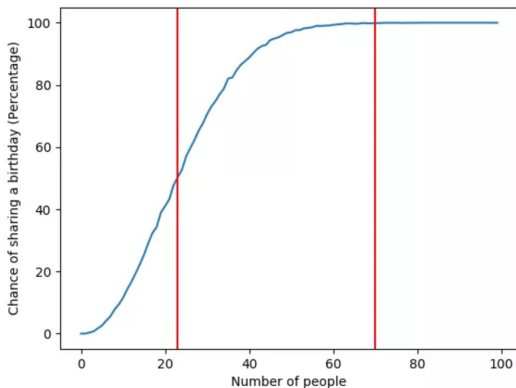
- Calculating similar things many times (e.g. iterations in a loop).
- Calculations are independent of each other.
- Each calculation takes a decent amount of time.

How Does Parallel Computing Works?

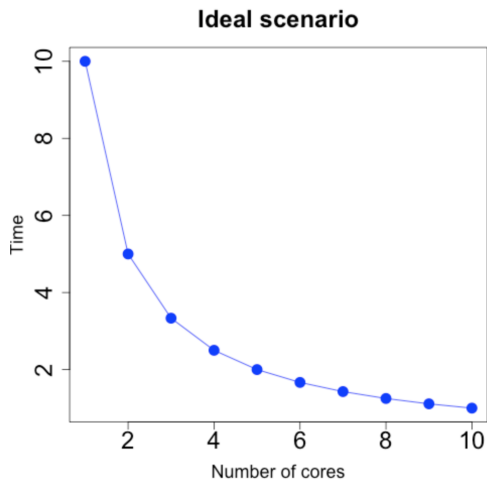


Motivation Example: Birthday Problem

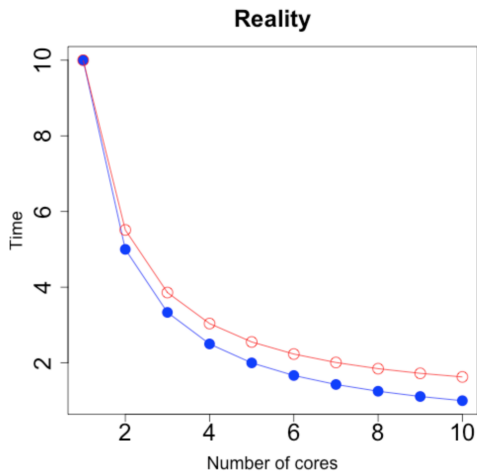
What is the probability that in a group of n people at least two have the same birthday?



Key Motivation: Speeding up R



Key Motivation: Speeding up R



- Repeated executions can be done manually, but it becomes quite tedious to execute repeated operations.
- R comes with various looping constructs that solve this problem. The `for` loop is one of the more common looping constructs, but the `repeat` and `while` statements are also quite useful.
- In addition, there is the family of “apply” functions, which includes `apply`, `lapply`, `sapply`, `tapply`, and others.

foreach package

- The `foreach` package provides a new looping construct for executing R code repeatedly.
- The main reason for using the `foreach` package is that it supports parallel execution.
- It can execute those repeated operations on multiple processors/cores on your computer, or on multiple nodes of a cluster.

- The `doParallel` package is a “parallel backend” for the `foreach` package.
- It provides a mechanism needed to execute `foreach` loops in parallel.
- The `foreach` package must be used in conjunction with a package such as `doParallel` in order to execute code in parallel.

Example: Simulation Study

- We will perform a simulation study to check the performance of bootstrap estimate of standard error of sample mean on different sample sizes.
- We first generate random samples of size 10, 50, and 100 from normal distribution with mean 0 and standard deviation 2. Next we find the bootstrap estimate of standard error of sample mean for each sample size with 2000 bootstrap samples.
- We repeat this 100 times and calculate the root-mean-squared error (RMSE) to compare the performance of bootstrap estimate of standard error of sample mean on different sample sizes.

$$\text{RMSE} = \sqrt{\frac{1}{R} \sum_{r=1}^R \left(\text{SE}(\bar{x}) - \widehat{\text{SE}}_r(\bar{x}^*) \right)^2}$$

- A rule of thumb: If you can wrap your task in an `apply` function or one of its variants then you can also use parallel computing!
- Check how many cores your laptop or desktop has and start using parallel computing in R!

- L. Collado-Torres's website
 - <http://lcolladotor.github.io/>
- Steve Weston's documents.
 - Using the `foreach` Package
 - Getting Started with `doParallel` and `foreach`