## Intro to Parallel Computing in R

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#### 2 Introduction to Parallel Computing in R

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Some of the top most downloaded R packages:

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



### Introduction to Parallel Computing in R

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- 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.

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.



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### 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



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# Key Motivation: Speeding up R



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- 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.

- 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.

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

- 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