

The background features several decorative elements: a large orange ring in the top-left, a large dashed blue circle in the top-center, a large teal ring in the bottom-right, and various smaller solid and dashed circles in yellow, pink, green, and cyan scattered throughout.

Parallel Computing

What is it and when to use it

Types of Computations

Input



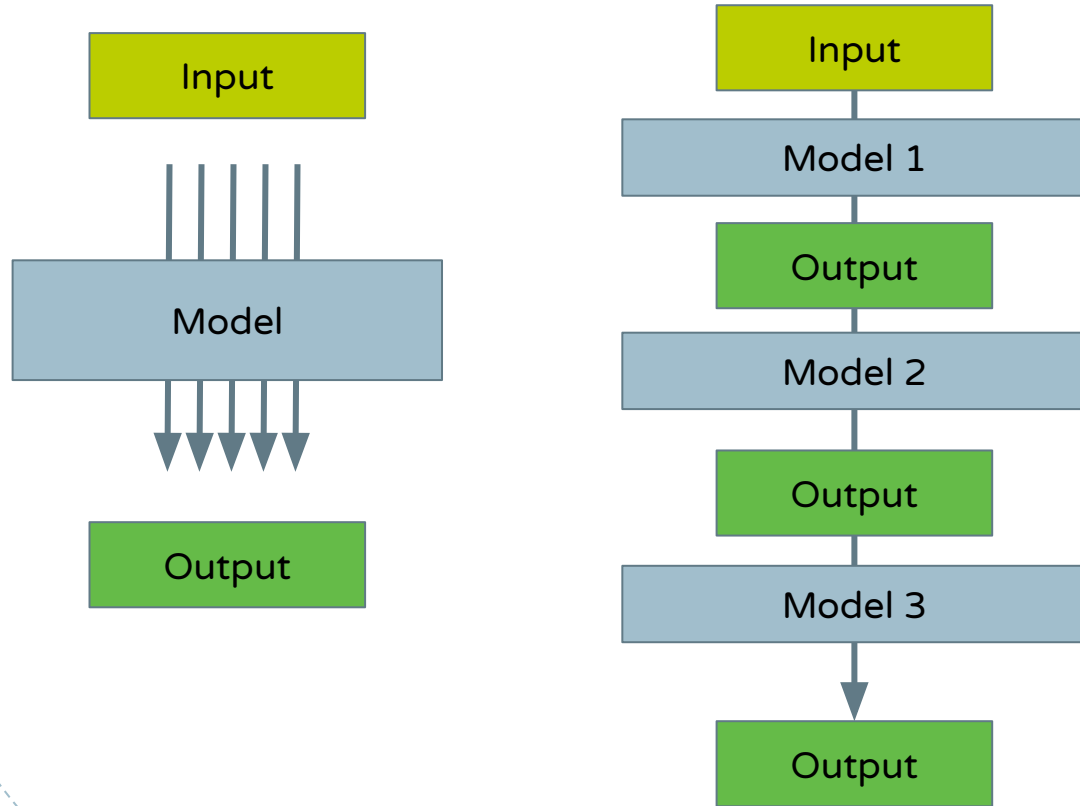
Model



Output

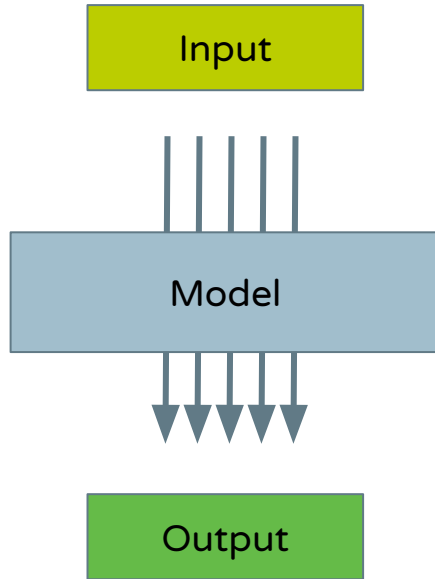
Outputs are independent of each other

Types of Computations

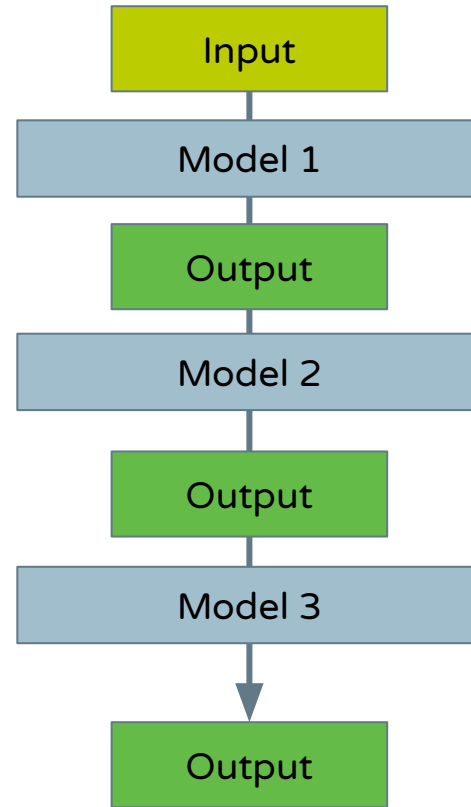


Types of Computations

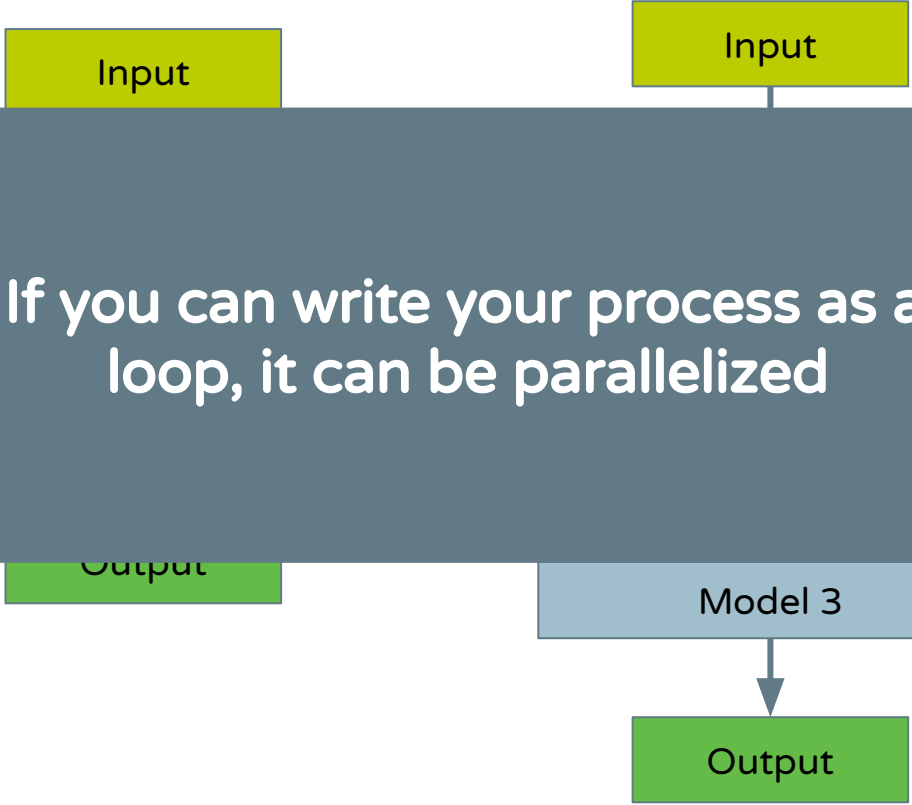
Parallel Problem:
Outputs are independent of each other



Serial Problem:
Output becomes the input for next computation



Types of Computations



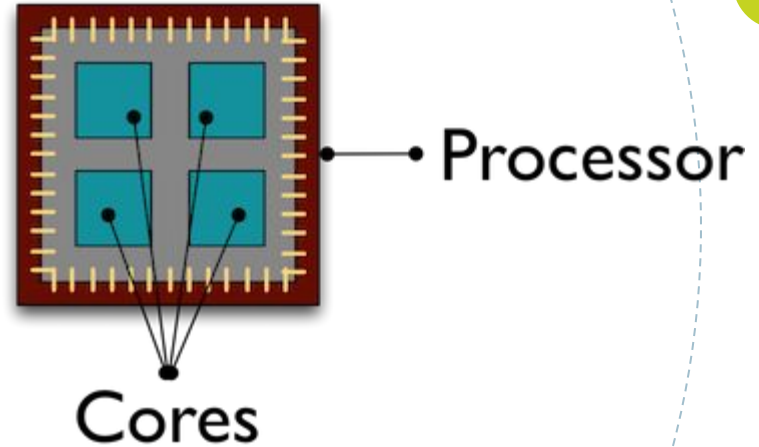
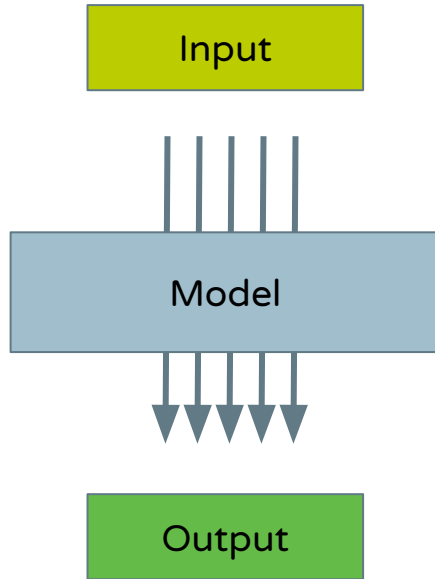
Parallel Problem:

Outputs are independent of each other

Serial Problem:

Output becomes the input for next computation

Each core can run a separate calculation



Standard computers often have 4-8 cores

Sockets

Launches R on new core and manually exports functions, objects, packages, etc.

- Works on all OS
- More flexible/complicated to implement

Forking

Creates a copy of the current R instance on new core

- Does not work on Windows
- Faster
- Can be unstable/odd behavior

Two Common Packages

parallel

- feels like an apply statement
- some functionality built into base R
- `mclapply()`, but is treated as `lapply` on Windows

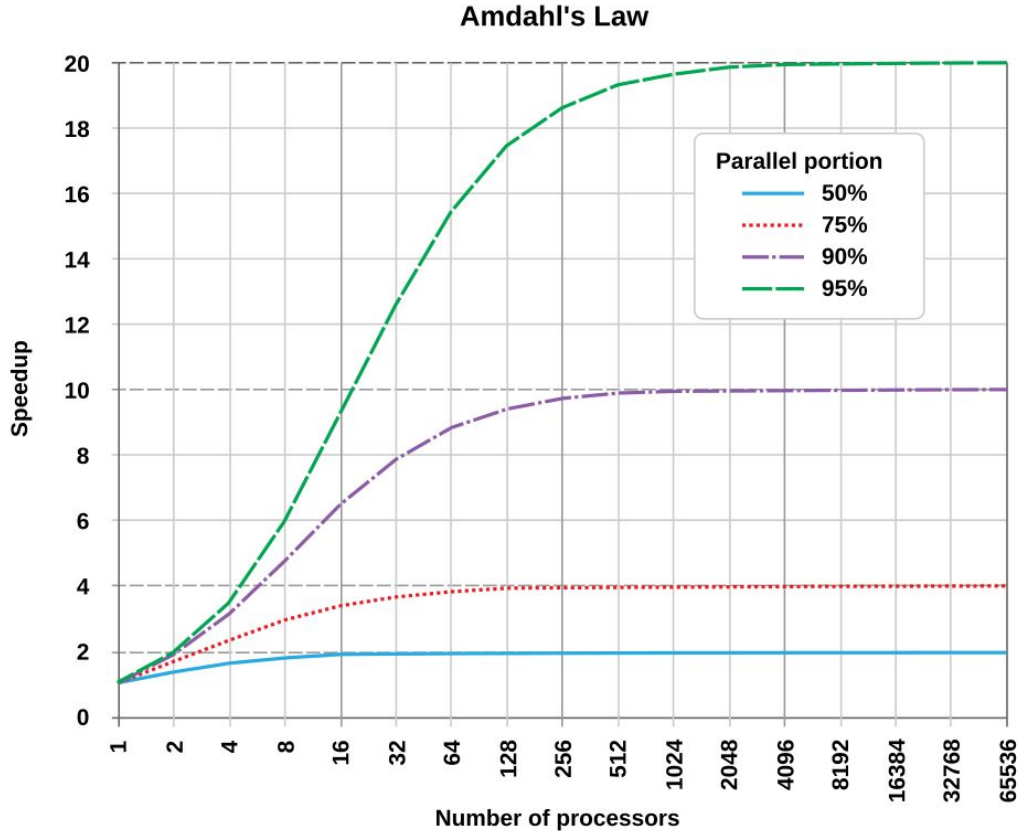
doParallel

- feels like loops
- uses the `foreach` package
- `%dopar%`

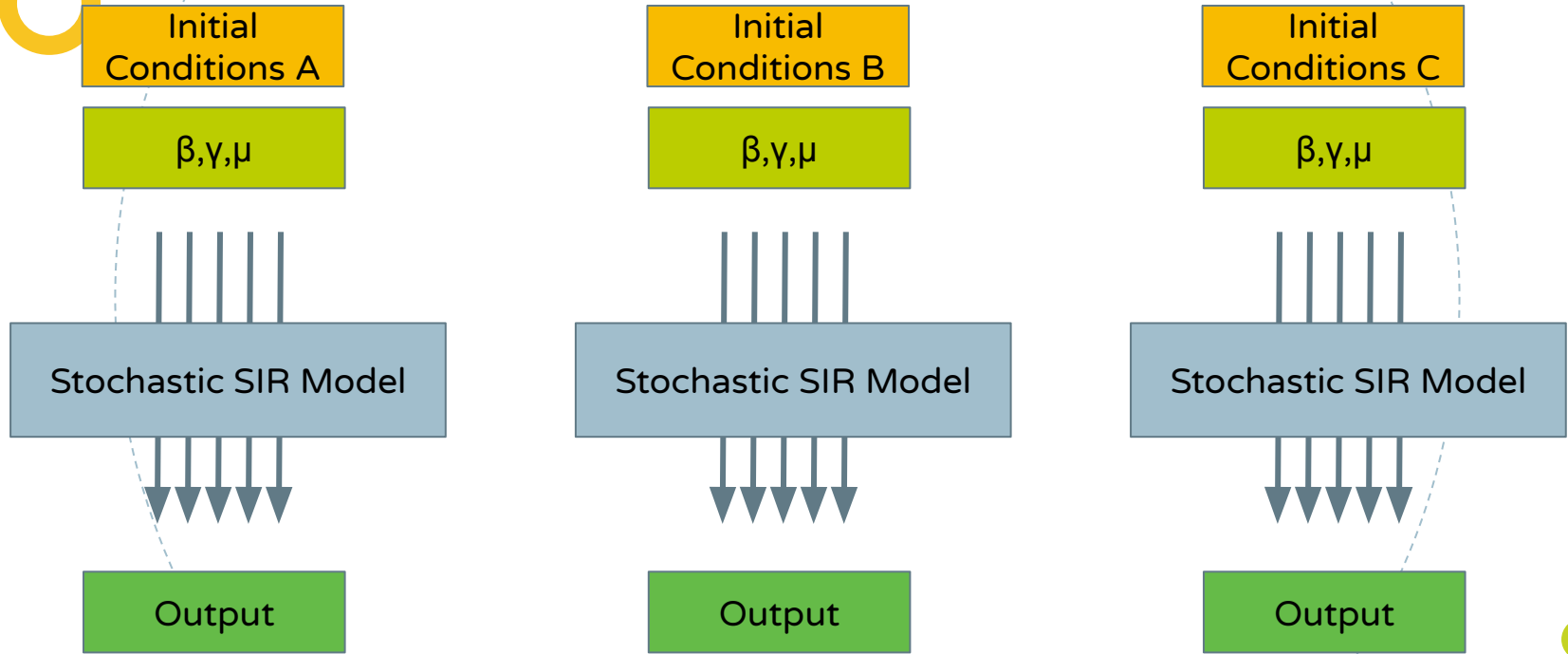
* Choosing approach is dependent on operating system and coding preference

2x Cores \neq 2x Speed

There is a computational cost to manage additional cores



My common use case





Further readings

Vignettes:

[Getting Started with doParallel and foreach
Package `parallel`](#)

Applied example of parallel package: [Stats
506: Parallel Processing](#)

Comparison of parallel and doParallel: [23
Additional Resources: Parallel Computing in R |
Reproducible Research Techniques for
Synthesis](#)

