Programming with Big Data in R

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Affiliations and Support

The pbdR Core Team http://r-pbd.org

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About This Presentation

Downloads

This presentation and supplemental materials are available at:

http://r-pbd.org/user2013



About This Presentation

Speaking Serial R with a Parallel Accent

The content of this presentation is based in part on the **pbdDEMO** vignette *Speaking Serial R with a Parallel Accent*

https://github.com/wrathematics/pbdDEMO/blob/master/inst/doc/pbdDEMO-guide.pdf?raw=true

It contains more examples, and sometimes added detail.



About This Presentation

Installation Instructions

Installation instructions for setting up a pbdR environment are available:

This includes instructions for installing R, MPI, and pbdR.



About This Presentation

Conventions

We use:

- "•" as a decimal mark
- "," as order of magnitude separator

Example	Yes	No	
One million	1,000,000	1.000.000	
One half	0.5	0,5	
One thousand and one half	1,000.5	1.000, 5	



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A Concise Introduction to Parallelism

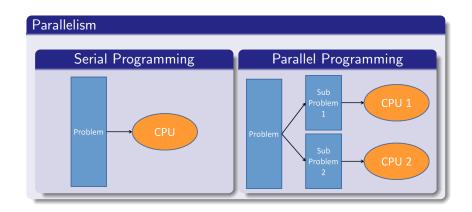
What is Parallelism?

Broadly, doing more than one thing at a time.

The simultaneous use of multiple compute resources to solve a computational problem:

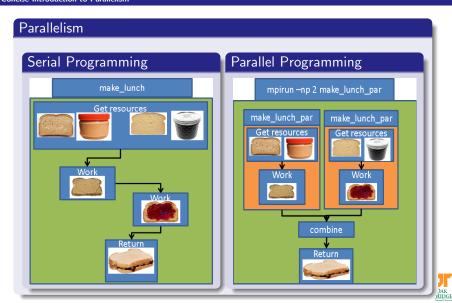


A Concise Introduction to Parallelism





A Concise Introduction to Parallelism



A Concise Introduction to Parallelism

Kinds of Parallelism

- Data Parallelism: Data is distributed
- Task Parallelism: Tasks are distributed



A Concise Introduction to Parallelism

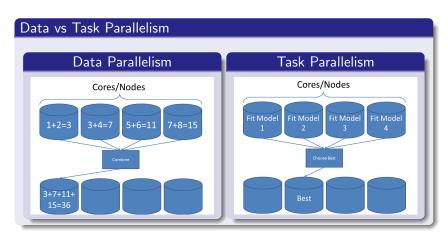
pbdR Paradigms: Data Parallelism

With data parallelism:

- No one processor/node owns all the data.
- Processors own local pieces of a (conceptually) global object



A Concise Introduction to Parallelism





Common Terminology

Difficulty

- Implicit parallelism: Parallel details hidden from user
- 2 Explicit parallelism: Some assembly required...
- 3 Embarrassingly Parallel: Also called loosely coupled. Obvious how to make parallel; lots of independence in computations.
- **1** Tightly Coupled: Opposite of embarrassingly parallel; lots of dependence in computations.



Common Terminology

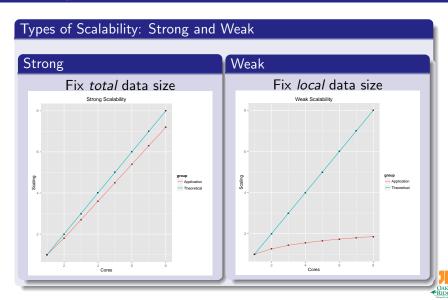
Scalability

Scalability: unitless measure of performance;

$$\frac{\tau_i}{\tau_0}$$



Common Terminology



 Introduction
 pbdR
 pbdMPI
 GBD
 Break
 Stats eg's
 pbdDMAT
 pbdDMAT eg's
 Wrapup

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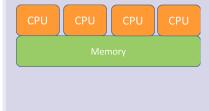
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Common Terminology

Shared and Distributed Memory Machines

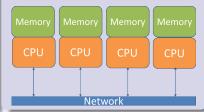
Shared Memory

Different processors can directly access and modify each others' memory. There is only one node.



Distributed

Different processors/nodes can not directly access/modify different processors'/nodes' memory.





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Common Terminology

Shared and Distributed Memory Machines

Shared Memory Machines

Thousands of cores



Nautilus, University of Tennessee

1024 cores

Distributed Memory Machines

Hundreds of thousands of cores



112,896 cores



R and Parallelism

R and Parallelism

What about R?



R and Parallelism

Problems with Serial R

- Slow.
- ② If you don't know what you're doing, it's really slow.
- Open Performance improvements usually for small machines.
- Very ram intensive.
- 6 Chokes on big data.



R and Parallelism

Shared Memory 1 foreach 2 parallel 3 snow 4 multicore Distributed 1 Rmpi 2 R+Hadoop 3 pbdR



R and Parallelism

R and Parallelism

The solution to many of R's problems is parallelism. However . . .

What we have

- Mostly serial.
- 2 Mostly not distributed
- Data parallelism mostly explicit

What we want

- Mostly parallel.
- Mostly distributed.
- Mostly implicit.



R and Parallelism

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Why We Need Parallelism

- Saves time (long term).
- ② Data size is skyrocketing.
- 3 Necessary for many problems.
- Like it or not, it's coming.
- 1t's really cool.

