

Advance Your Research: Scale Out Your Computing

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Why Scale Out Your Computing?





Richard Ayoade in the IT Crowd, Series 1, Episode 2 https://giphy.com/gifs/richard-ayoade-it-crowd-maurice-moss-dbtDDSvWErdf2





Photo credit Steve Goldstein





Dr. Timothée Poisot @tpoi

Just use this simple heuristic: if it's slow and the computer goes WOOOOOOOSH, the bottleneck is the CPU; if it's slow and the mouse stop moving, the bottleneck is the memory.

https://twitter.com/tpoi/status/1169692855201402885



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Big Problems

How do you solve a big computational problem?

For example:

- Lots of data
- Thousands of parameters
- Many grid points

Big problem





Our Goal: Scaling Out

We want to be able to tackle big problems.

We want to go from using a small number of resources to a LOT of resources.

There is one strategy to do this with computers: divide and conquer

Big problem





Big Solutions

To solve a big problem, break it into many subproblems and use more/specialized computers.

There are different ways to do this.



Large Scale Computing Approaches



A bigger computer

Sometimes, getting a larger computer can solve your problem more CPUs, more memory, more disk space.

- Workstation
- Server that requires remote log in





Option 1: More CPUs

Run sub-problems on individual CPUs.

Requires programming this capability into the code.



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Specialized computer components (like GPUs)



GPUs are a common way to accelerate certain kinds of computing.

Like before, can be used from a:

- Workstation
- Server



Option 2: Use a GPU

Run sub-problems on "cores" inside a GPU card.

Like before, requires programming this capability into the code.

Examples:

- Neural networks and other deep learning algorithms
- Certain molecular dynamic codes



Use more computers at once.

This is the idea behind the large-scale computing systems like the Open Science Pool.





Option 3: High Performance Computing (HPC)



Solve sub-problems by distributing across multiple CPUs on multiple computers.

time Also requires special coding (MPI) to coordinate calculations across multiple computers.



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HPC Examples:

- Complex physical simulations with many inter-connected components.
- Optimization problems that use many local calculations to drive the global optimization.



Option 4: High Throughput Computing (HTC)



Sub-problems are independent and self-contained, can be run on many computers.

time Most scalable.

No special programming needed.



HTC Examples:

- Test many parameter combinations
- Analyze multiple images or datasets
- Do a replicate/randomized analysis
- Align genome/RNA sequence data



One of our favorite HTC examples: baking the world's largest/longest cake



In computational terms: solving a big problem (the world's longest cake) by executing many small, self-contained tasks (individual cakes) and joining them.

Photos: Arun Sankar via https://www.theguardian.com/world/2020/an/16/indian-bakery rise-to-task-of-making-worlds-longest-cake

Use Case 1: Random Simulations

- Typical task: running a simulation
- One computation needs:
 - 1 core, 512Mb of memory
 - almost no data
 - 10 seconds
- How many times? Millions
- Are there multiple steps involved? No.
- **Bottleneck**: total number of simulations





Use Case 2: Analyzing Multiple Files

- **Typical task**: analyzing a data sample
- One computation needs:
 - 1-16 cores, 8 32GB of memory
 - Input of 5GB, output of 10GB
 - 8-10 hours for whole pipeline
- How many times? Dozens (number of samples)
- Are there multiple steps involved? Yes.
- **Bottleneck**: Size of data, number of samples, certain steps





Discussion

See handout.



OSG School = HTC School

- This week is all about learning more about the practice of high throughput computing:
 - Job submission + troubleshooting (Mon/Tues)
 - Software portability (Tues)
 - Data handling (Wed)
- On two different high throughput computing resources
 - The PATh Facility (Mon)
 - the Open Science Pool (Tues onward)



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