

So You Might Want to Work at a National Lab?

Cynthia A. Phillips (Sandia National Laboratories)



Sandia is a multiprogram laboratory operated by Sandia Corporation, a Lockheed Martin Company, for the United States Department of Energy's National Nuclear Security Administration under contract DE-AC04-94AL85000.

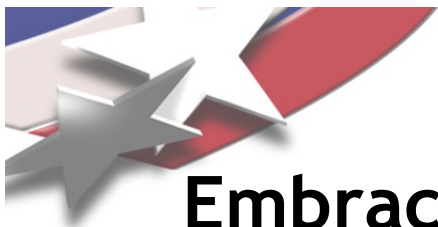




Performance

Why would we want to use a parallel algorithm for an application?

- When we **have** to:
 - Too Slow
 - Too Big
 - Too Inaccurate
- Application evolution
 - More constraints
 - Finer discretization
 - Larger instances
- Parallel computing is a **tool for performance**
 - We'd really like a faster, bigger serial machine
 - Make sure you beat the serial implementation
 - Better algorithms can be a better way to performance



Embrace “Embarrassing” Parallelism

- Embarrassing parallelism increases the maximum feasible problem size
- Buys time to do the harder parallelization if necessary
- Using it can present other interesting algorithmic questions
- Some uses
 - Integer programming pseudocost initialization
 - Randomized rounding
 - Independent problems with different parameters for uncertainty quantification
 - Independent simulations to generate data for stochastic programs (sensor placement in water networks)



Parallel computing jobs at Sandia

- Scientific computing
- Data Sciences including streaming
- Operations Research
- Production computing
- Computer architecture and system softwares
 - Light-weight runtime systems, simulators,
 - Work with vendors
 - Neuromorphic computing

I interviewed other Sandia researchers to reflect more of this breadth



Perspective

Sandians solve **national-scale problems**

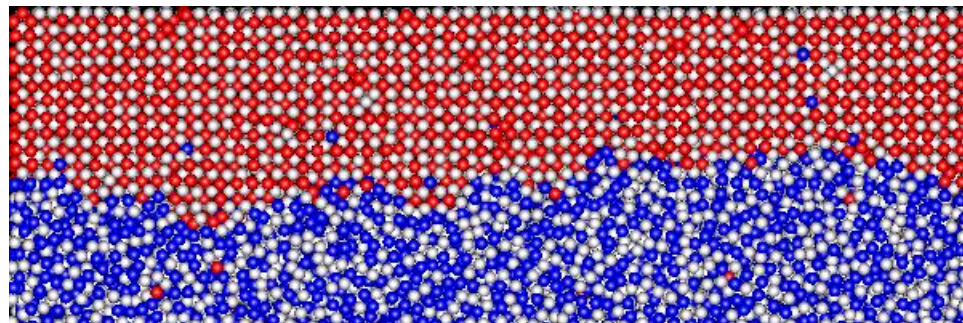
Many people who enter workforce with a bachelor's degree will program cell phones, Hadoop, single multicore

- Cost efficient
 - Energy efficient
- Undergraduate programs have to serve them too



Scientific Computing

- Physical modeling and simulation (e.g. finite element)
- Sample of applications:
 - Shock physics
 - Molecular dynamics
 - Fluid flow
 - Quantum systems
 - Climate
 - More efficient nuclear energy reactors
- Enable the nuclear test ban
- Science, math, physics very important
 - Many Sandia computational scientists not CS

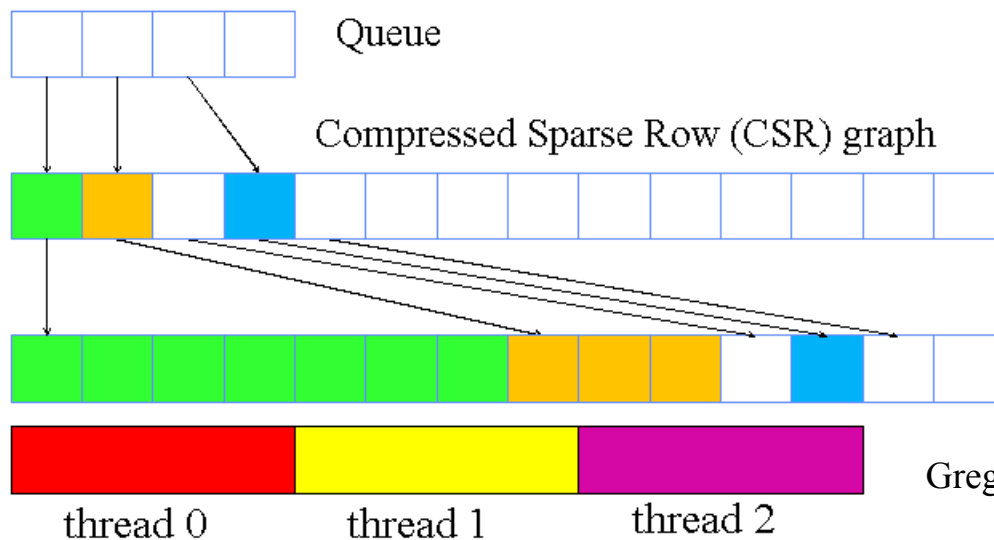


LAAMPS news release 3/3/11

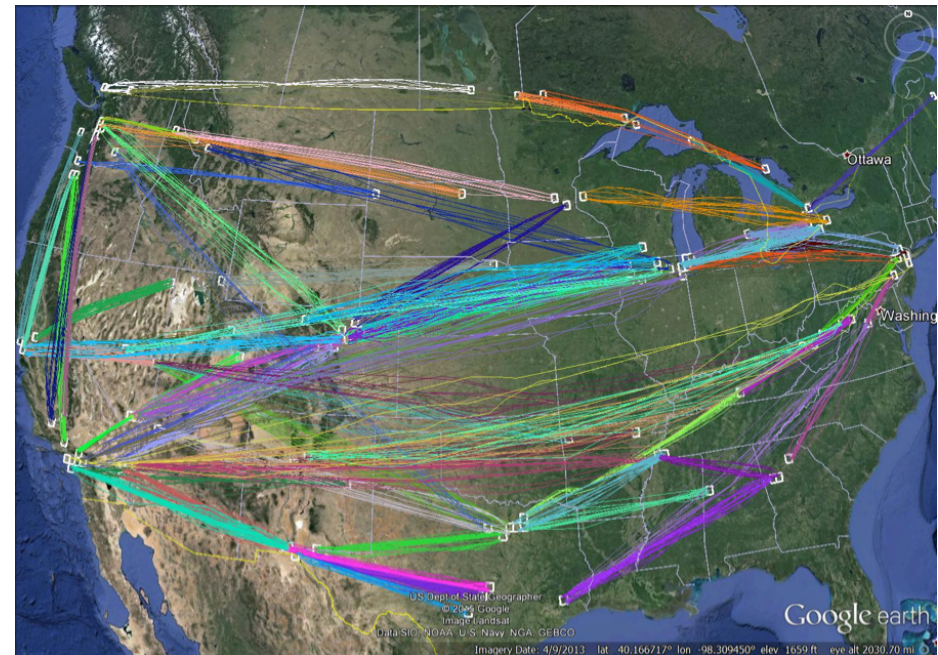


Data Science

- Streaming data (e.g. cybersecurity)
- Streaming images (e.g. non-proliferation)
- Network science
 - Large-scale graph algorithms
- Scientific computing data
 - e.g. visualization



Greg Mackey, 2009

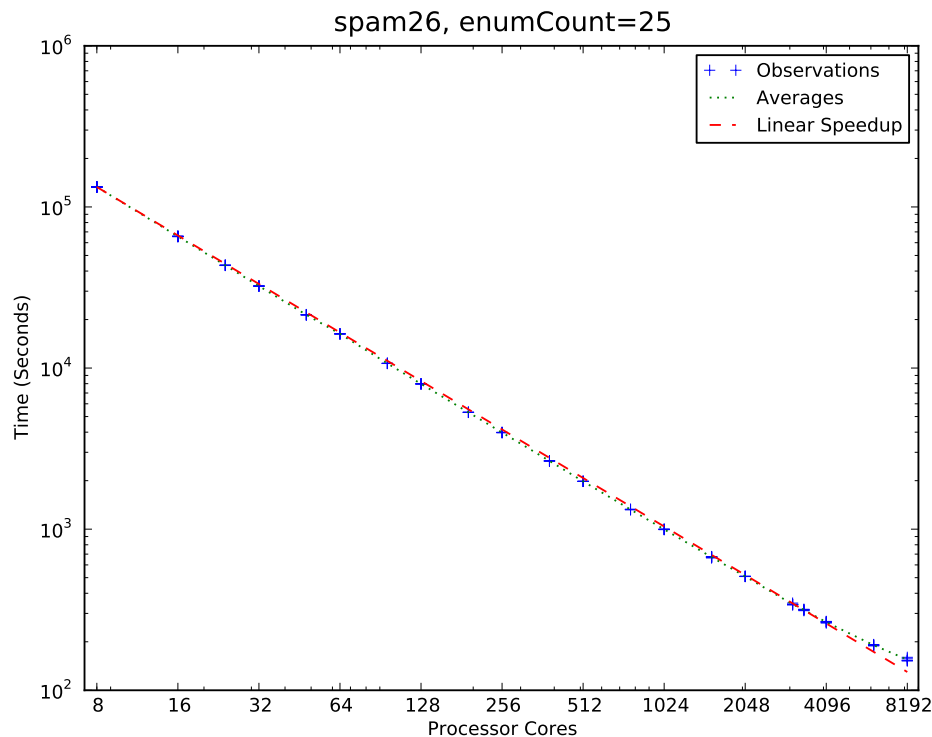


Danny Rintoul, 2015

EduHPC 2015

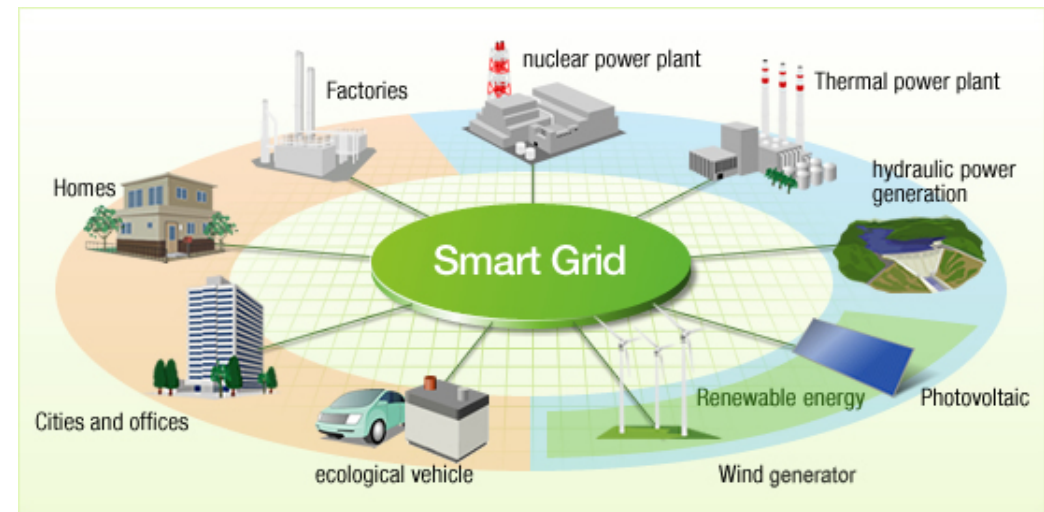
Operations Research (Optimization)

- Branch and bound for general combinatorial optimization
- Stochastic programming



Spam detection, machine learning

Smart grid planning



Hitachi.com, 2015



Production Computing

- Stand up big clusters and administer
 - Create file systems
- Trying new hardware and software
 - Early prototypes, testbeds
 - No user guide. Become local expert
- User support
 - Teaching
 - Workflow, tool chain
 - code certification, testing, monitoring
- Need a deep understanding of hardware and software layers, and math/cs science for application support. Problems happen at scale that you will not learn about in school



Hardware (sample)

- Scientific computing researchers have access to much larger machines at other labs
 - Jaguar (Oak Ridge), > 200k opteron
 - Titan (Oak Ridge), 18k CPU, 18k GPU
 - Cielo (Los Alamos), 143k cores
 - Sequoia, BlueGene/Q (Lawrence Livermore), ~ 1.6M cores



- A Sandia researcher has run on 1.6M cores.

Jaguar



Hardware (sample)

- Capacity Clusters with thousands to 10,000s of cores
 - Capacity machines are for throughput of moderate jobs
 - Capability machines are for jobs that cannot run anywhere else
- Special data science clusters with more memory and better I/O
- GPUs, etc
- Cloudlike, etc

Sandia was an early
Adopter of the Cray XMT



www.jp.cray.com



Competitors

- Many industries want people who know HPC, big data, or can learn
 - Google, eBay, financial companies, Walmart, AmEx, Amazon
 - HPC on a multicore
 - Microsoft
 - Medical: real-time functional MRI
 - Even things like Shazam have a big search real-time problem
- Many **undergraduates are employed early**
 - Great for CS department stats
 - Still want to make their on-the-job learning easier



Parallel Thinking

- The details change frequently still
 - Architectures, languages, tools
- Shared memory processing
 - Learn the basics on something as small as a PC
- Distributed memory computing
 - Problem partitioning, message passing, load balancing
- Concurrency
 - Race conditions
 - One advanced topic: lock-free, wait-free algorithms
- Algorithms

Learn things that will stand the test of time, provide basis for career-long learning



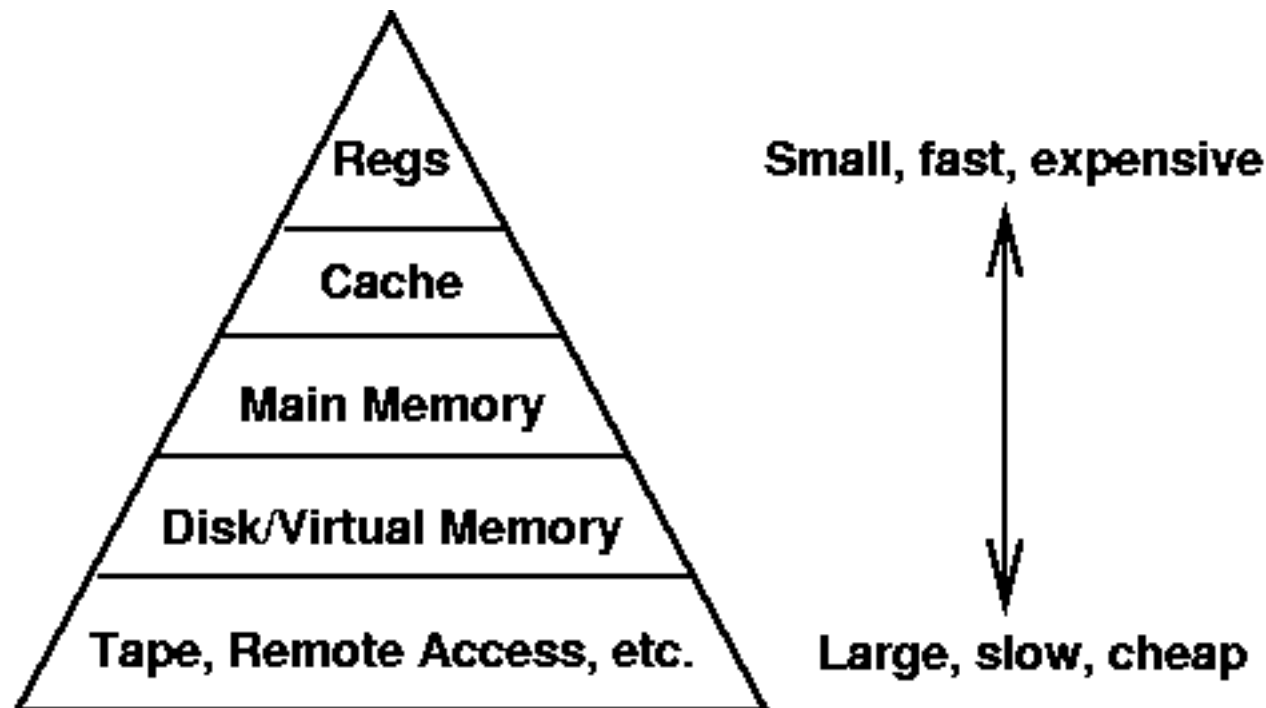
Personal Characteristics

- Curious, willing to learn, learns for fun all the time
- Likes to fix things, make them better
- Attention to detail
- Smart, creative, motivated
- Willing to stay for a while
- Nice. Works well with others
- Flexibility in tool choice, etc



A Basic Skill: Memory Hierarchy

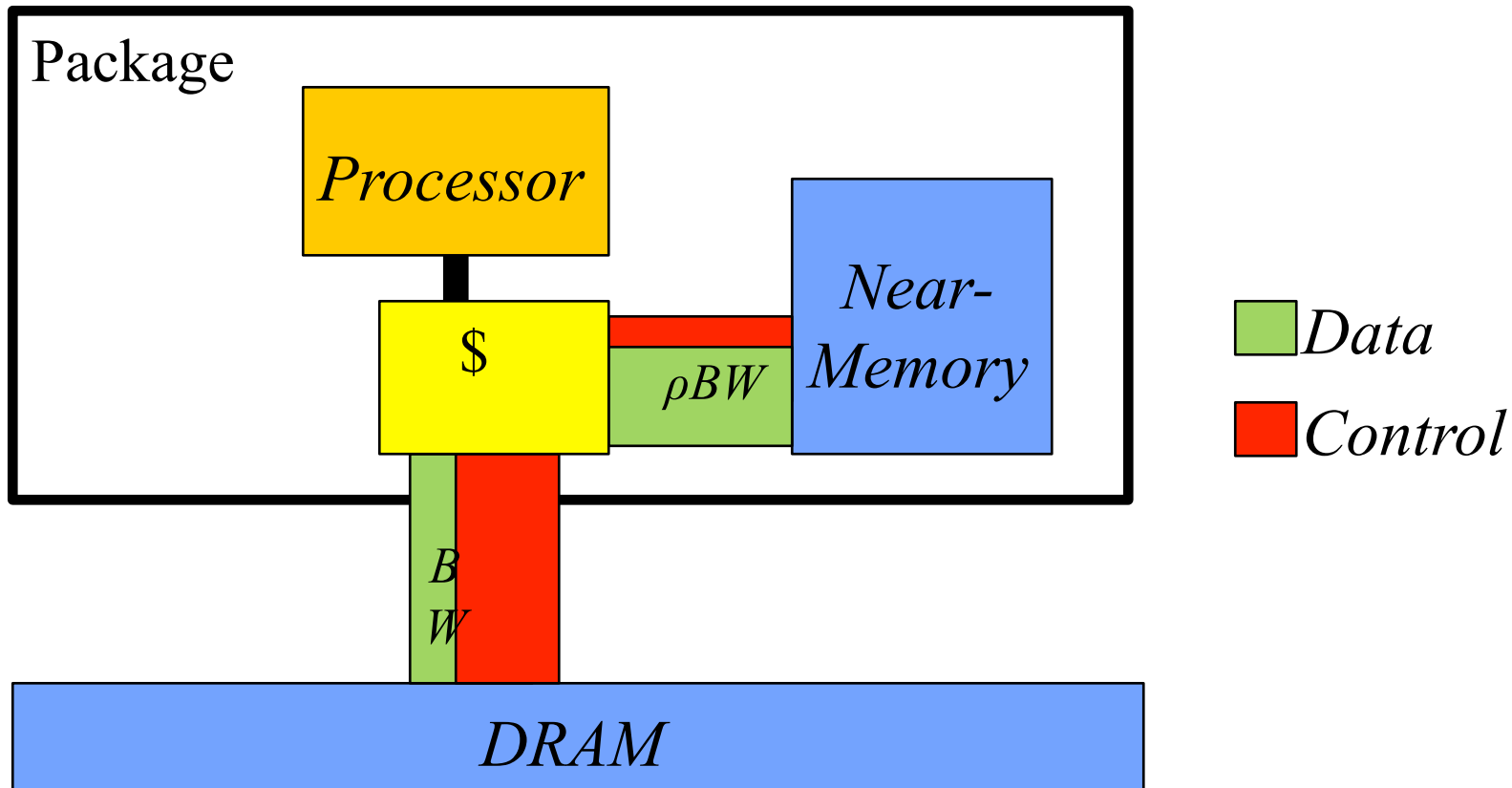
- Network is like another layer of the memory hierarchy
- Optimizing cache performance for serial problems directly helpful





New kind of memory for Trinity

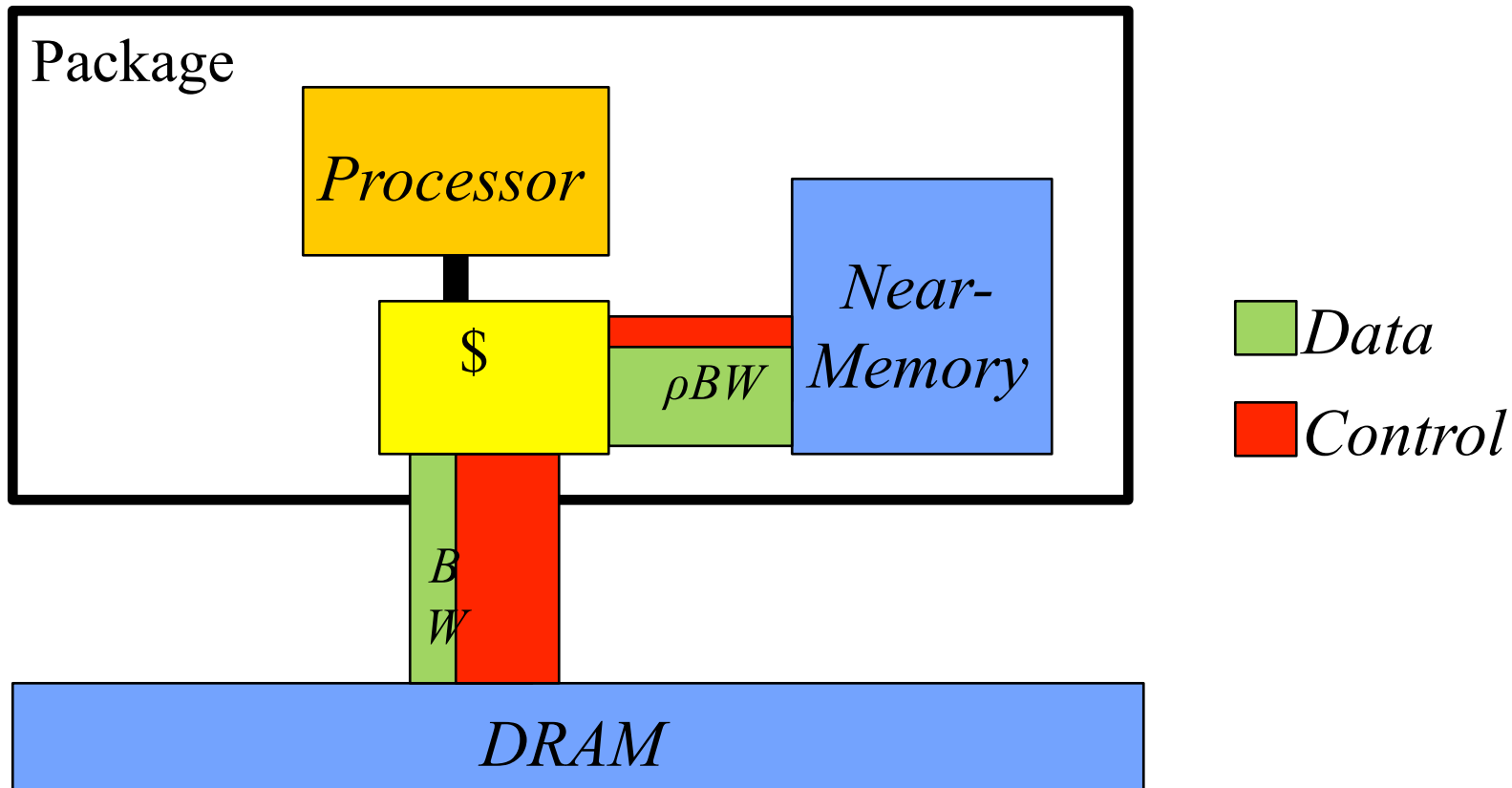
- Next capability machine for NNSA = National Nuclear Security Agency
- Will use Knight's Landing chip:





New kind of memory for Trinity

- Use as a cache?
- User control: justify the cost of reprogramming





Necessary and Sufficient conditions

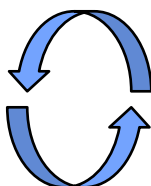
- Divide & conquer, memory boundedness analysis (IPDPS 2015, MemSys 2015)

(1) Memory-boundedness

Aggregate DRAM transfer takes more time than aggregate computation

(4) *Final check: new algorithm isn't cache chunkable*

If the new algorithm works efficiently with cache-sized chunks, scratchpad not needed



(2) Scratchpad chunkable

The computation can be divided into scratchpad-sized chunks that generate manageable partial results (a new algorithm)

(3) Chunk reuse

Each chunk loaded into near-memory must be reused enough to amortize the cost of the copy



Teams

- Especially for scientific computing problems, it takes a team to make a new code
 - Multiple specialties/disciplines required
 - One person cannot know everything
 - Individuals own different parts of the code/project
 - Shared credit



Software Engineering

- Use repositories, document, unit tests, good coding practice
- The performance expert has to speed up a code without deeply understanding the physics
 - Physics “owner” must set up tests to protect all the behavior he wishes to protect
 - Optimization person can commit anything that passes all tests
- The **alternative is a cultural MPI barrier**

Issue: Should use use the code written by the best physicist if it is impossible to test/verify?



Frameworks/packages

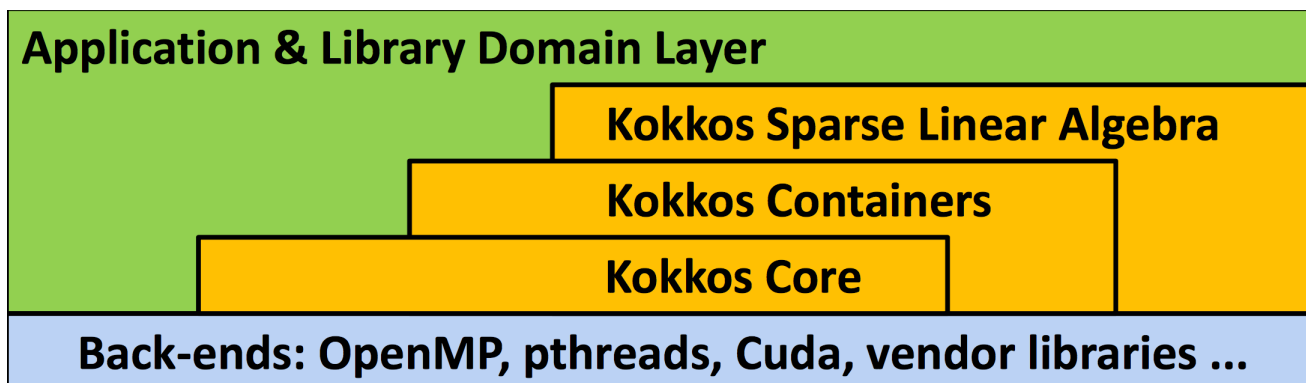
- Trilinos
 - solution of large-scale, complex multi-physics engineering and scientific problems
 - Many of the pieces one needs
 - Meshes, discretizations, load balancing, linear algebra, etc
- Multi-threaded graph library
 - Developed for the Cray XMT
 - Still used locally

Appropriate performance-optimized frameworks can help weather the changes in hardware.



Frameworks/packages

- Kokkos
 - Performance portability to many architectures
 - C++ templates
 - Abstractions for general performance (for, reduce, scan)
- Developers are part of ISO C++ committee
 - C++ wants to make parallelism “native” by 2020
 - Kokkos wants to meet C++ at that point





Scientific Computing Skills

- Good numerical math, numerical PDEs, grasp of discretization
- CS distributed memory computing, MPI background
 - Now moving to MPI + X
- Uncertainty quantification, verification (validation harder)
- Exemplar programs: U. Illinois and U. Texas Austin Institute for Computational Engineering and Sciences
- Predictive Science Academic Alliance program (PSAAP)
 - Department of Energy + National Science Foundation
 - Few universities, multi-million \$, 4-5 years
 - Multidisciplinary. Each university a different application
- Gaming systems also need physics, visualization, and realism



Computing for Scientists

I think the science is harder

What might a CS department put in a graduate computing class for scientists (chem E, mech E, physics, computational materials)?

- Perspective on algorithms
 - “Parallel and Distributed Computation” by Bertsekas and Tsitsiklis. Largely numerical, but high-level, abstract
 - Gives resilience as parallel computing changes
- Basic software engineering
 - Suggested class exercise: use someone matrix-vector multiply in your conjugate-gradient code



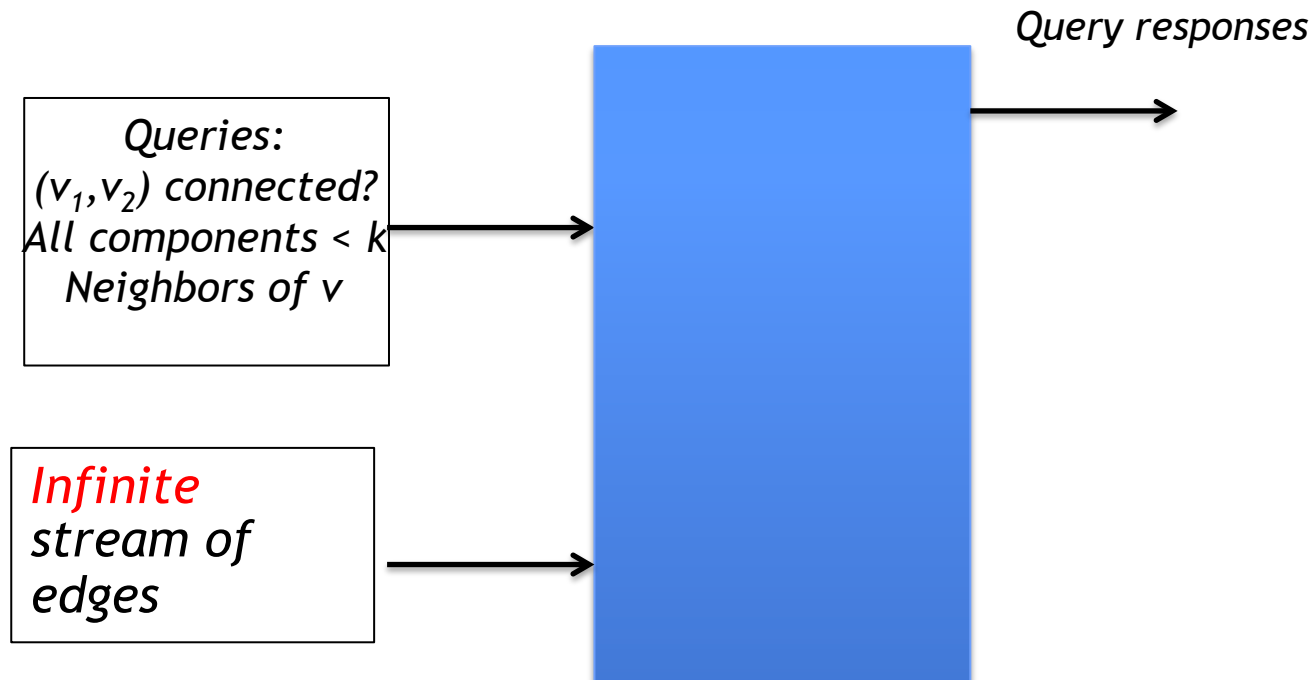
Performance Modeling

- Hardware counters, performance tools, vampir, etc
- Strong scalability
- Weak scalability
- Work/span
 - T_p = time to run an instance on p processors
 - T_1 = serial. T_∞ = minimum time (serial part)
 - Work law: $pT_p \geq T_1$ (Except for cache effects)
 - Span law: $T_p \geq T_\infty$
- A basic understanding of general analysis



Streaming

- Perfect answer vs timely answer
 - Ads, bond trading, cybersecurity
- Write-optimized data structures (databases)
- Streaming benchmarking for acceptance testing
- Cyber (monitoring) model: parallel models to approximate





Paid Internships

- 'Tis the season
 - Labs usually handle summer internships November-December
 - Later is possible, but suboptimal
- A good way to write a program for a larger data set