

# Lecture 2: Basic Performance Models For Extreme Scale Systems

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# Performance is Key

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- Parallelism is (usually) used to get more performance
  - ◆ How do you know if you are making good (not even best) use of a parallel system?
- Even measurement-based approaches can be (and all too often are) performed without any real basis of comparison
  - ◆ The key questions are
    - Where is most of the time spent?
    - What is the achievable performance, and how do I get there?
  - ◆ This latter is often overlooked, leading to erroneous conclusions based on the (immature) state of compiler / runtime / code implementations



# Tuning A (Parallel) Code

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- Typical Approach
  - ◆ Profile code. Determine where most time is being spent
  - ◆ Study code. Measure absolute performance, look at performance counters, compare FLOP rates
  - ◆ Improve code that takes a long time, reduce time spent in “unproductive” operations
- Why this isn’t the right Approach:
  - ◆ How do you know when you are done?
  - ◆ How do you know how much performance improvement you can obtain?
- Why is it hard to know?
  - ◆ Many problems are too hard to solve without extreme scale computing
  - ◆ Its getting harder and harder to provide performance without specialized hardware



# Blue Waters Computing System



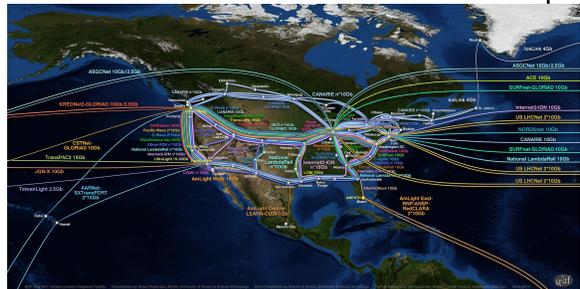
10/40/100 Gb Ethernet Switch

IB Switch

> 1 TB/sec

120+ Gb/sec

100 GB/sec



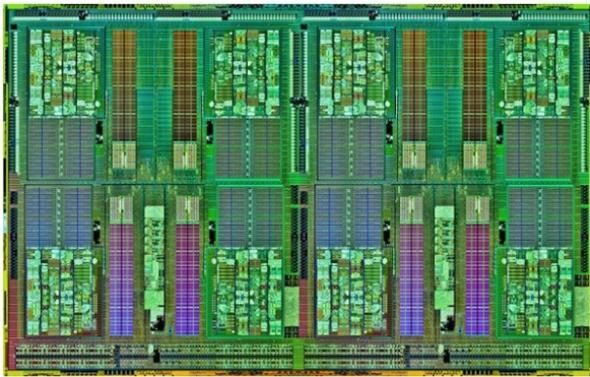
WAN

Spectra Logic: 300 PBs

Sonexion: 26 PBs



# Heart of Blue Waters: Two Chips

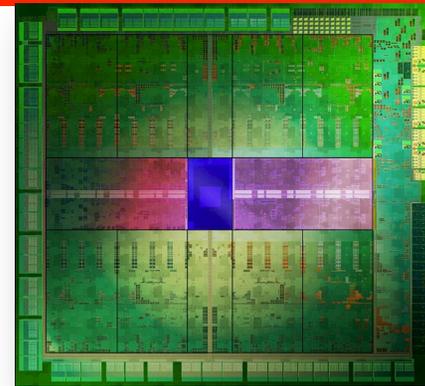


## AMD Interlagos

*157 GF peak performance*

### Features:

- 2.3-2.6 GHz
- 8 core modules, 16 threads
- On-chip Caches
  - L1 (I:8x64KB; D:16x16KB)
  - L2 (8x2MB)
- Memory Subsystem
  - Four memory channels
  - 51.2 GB/s bandwidth



## NVIDIA Kepler

*1,400 GF peak performance*

### Features:

- 15 Streaming multiprocessors (SMX)
  - SMX: 192 sp CUDA cores, 64 dp units, 32 special function units
  - L1 caches/shared mem (64KB, 48KB)
  - L2 cache (1536KB)
- Memory subsystem
  - Six memory channels
  - 180 GB/s bandwidth



# What is an Extreme Scale System Today?

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- Tianhe 2 (China):
  - ◆ 16,000 nodes, each with 2 Intel Ivy Bridge Xeon processors and 3 Xeon Phi coprocessors
  - ◆ 3,120,000 cores
  - ◆ Interconnect is a “fat tree” of 13 switches, each with 576 ports
- Sequoia (USA):
  - ◆ IBM Blue Gene/Q. 98,304 nodes, each with 16 (+1) cores
  - ◆ Interconnect is 5 dimensional torus



# Likely Directions for Extreme Scale Systems

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- 5 Years (2020)
  - ◆ Peak performance over 1 ExaFLOPs ( $10^{18}$  ops/sec)
  - ◆ 100k “nodes”
  - ◆ Heterogeneous nodes
- 10 Years (2025)
  - ◆ Peak performance over 30 ExaFLOPs
  - ◆ Computing distributed throughout node and memory
- 15 Years (2030)
  - ◆ Peak performance over 100 ExaFLOPs
  - ◆ Radically different systems emerging
    - New digital logic, e.g., nanotubes
    - New computing models, e.g., quantum or molecular



# Why Performance Modeling?

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- What is the goal?
  - ◆ It is *not* precise predictions
  - ◆ It *is* insight into whether a code is achieving the performance it could, and if not, how to fix it
- Performance modeling can be used
  - ◆ To estimate the baseline performance
  - ◆ To estimate the potential benefit of a nontrivial change to the code
  - ◆ To identify the critical resource



# What do I mean by Performance Modeling?

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- Actually two different models
  - ◆ First, an analytic expression based on the application code
  - ◆ Second, an analytic expression based on the application's *algorithm* and data structures
- Note that a series of measurements from benchmarks are *not* a performance model
- Why this sort of modeling
  - ◆ The obvious: extrapolation to other systems, such as scalability in nodes or different interconnect
  - ◆ Also: comparison of the two models with observed performance can identify
    - Inefficiencies in compilation/runtime
    - Mismatch in developer expectations



# Different Philosophies for Performance Models

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- Simulation:
  - ◆ Very accurate prediction, little insight beyond specifics of the simulation itself
- Traditional Performance Modeling (PM):
  - ◆ Focuses on accurate predictions
  - ◆ Tool for computer scientists, not application developers
- PM as part of the software engineering process
  - ◆ PM for design, tuning and optimization
  - ◆ PMs are developed with algorithms and used in each step of the development cycle
  - Performance Engineering



# Example

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- Lets look at a simple example
- Matrix-matrix multiply
  - ◆ Classic example, often used in discussion of compiler optimizations
  - ◆ Core of the “HPLinpack” benchmark
  - ◆ Simple to express: In Fortran,  
do i=1, n  
  do j=1,n  
    c(i,j) = 0  
    do k=1,n  
      c(i,j) = c(i,j) + a(i,k) \* b(k,j)



# Performance Estimate

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- How fast should this run?
  - ◆ Standard complexity analysis in numerical analysis counts floating point operations
  - ◆ Our matrix-matrix multiply algorithm has  $2n^3$  floating point operations
    - 3 nested loops, each with  $n$  iterations
    - 1 multiply, 1 add in each inner iteration
  - ◆ For  $n=100$ ,  $2 \times 10^6$  operations, or about 1 msec on a 2GHz processor :)
  - ◆ For  $n=1000$ ,  $2 \times 10^9$  operations, or about 1 sec



# The Reality

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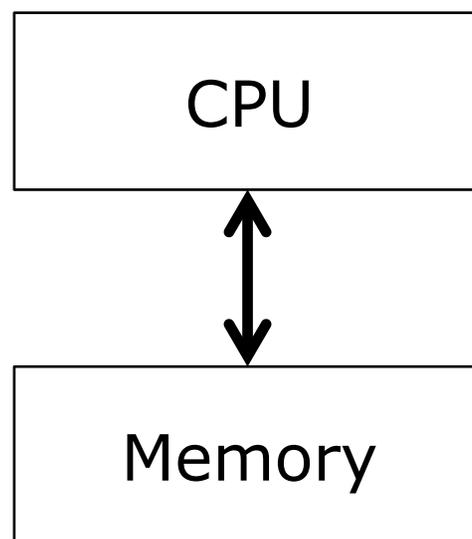
- $N=100$ 
  - ◆ 1818 MF (1.1ms)
- $N=1000$ 
  - ◆ 335 MF (6s)
- What this tells us:
  - ◆ Obvious expression of algorithms are not transformed into leading performance.



# Thinking about Performance

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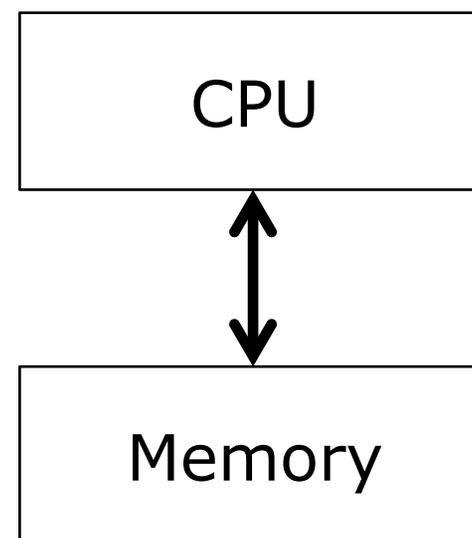
- The performance model assumes the computer looks like the figure on the right
  - ◆ Memory is infinitely large
  - ◆ Memory is infinitely fast



# Thinking about Performance

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- We will incrementally improve our performance models by adding features to our model of the computer hardware
  - ◆ That model of the computer hardware is a major part of what is often called an *execution model*
- In the first enhancement, lets make memory not infinitely fast



# A Simple Performance Model

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- Use the following:
  - ◆ Number of operations (e.g., floating point multiply)
  - ◆ Number of loads from memory
  - ◆ Number of stores to memory
- We are ignoring for now the many features of an architecture that are used to optimize performance
  - ◆ We will cover many of them during the class



# A Simple Example

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- Consider this code:  
Do  $i=1,n$   
     $y(i) = a*x(i) + y(i)$   
enddo
- $2n$  operations (floating add, floating multiply)
- $2n$  Loads ( $x(i)$  and  $y(i)$  for  $i=1$  to  $n$ )
- $N$  Stores ( $y(i)$ )



# Performance Model

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- Assume that
  - $c$  = time for operation
  - $r$  = time to read an element
  - $w$  = time to write an element
- Then a very crude estimate of the time for this operation is
$$T = n(2c + 2r + w)$$
- Call this a *model* because it is too crude to be an estimate



# Some Comments on This Model

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- Many analysis of algorithms set  $r$  and  $w$  to zero
- We will spend much of our time considering different ways to model *communication* time
  - ◆ Load and Store to memory
  - ◆ Sharing of data between threads
  - ◆ Communication between nodes in a parallel computer
  - ◆ Load and Store to a file system



# Discussion Topics for Matrix- Matrix Multiply

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- Why do you think the algorithm runs slowly at large sizes?
- Why do you think the compiler doesn't do a better job?
- What about other algorithms such as Strassen's algorithm?
  - ◆ How would that algorithm change this analysis?

