





- Choice of environments for book was based on how things were when it was written — MPI fairly dominant for distributed memory and OpenMP for shared memory, with Java not so widely used for parallel programming but more familiar/available.
- Slide 3
- All three include more than we had time to cover in class, and have continued to evolve, and then there's a whole new hardware platform (GPUs) ...





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Java 1.5 brought into the standard library a lot of classes previously available as third-party additions — thread pools, locks, various shared-data classes, etc. Java memory model also cleaned up a bit. (Curiously enough, though, the need for explicit multithreading in GUIs seems to have declined, with the notion of the EDT and new classes such as SwingWorker and timers.)



- Graphics processors emerging as a new platform for parallel computing hardware is becoming sophisticated enough to support computation beyond the cards' original purpose, so why not put it to use?
- No consensus yet about programming environments, but OpenCL might emerge as a semi-standard, as MPI and OpenMP did.

• Some interesting challenges, though ...

A Little About GPU Hardware

- Processing hardware seems to typically include many processors working more or less in lockstep, each able to do pipelined/vector operations — i.e., SIMD, making a comeback!
- Typical hardware seems to also include a possibly-complex memory hierarchy separate from the memory hierarchy of the "host computer".



 SIMD hardware makes a data-parallel style of programming a good fit. Not something we really address in our pattern language (yet!), but conceptually similar to *Geometric Decomposition* but more closely synchronized.
 A.k.a. "stream processing"?

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 So, you might express computations as a sequence of whole-array operations, or in terms of applying a "computational kernel" in parallel to many data elements. Whole-array operations included in some programming environments (e.g., Fortran). Current programming environments for GPUs (NVIDIA's CUDA, e.g., and OpenCL) seem to use the computational-kernel idea.

A Little About Programming for GPU Hardware, Continued

• Currently moving data back and forth between host's memory and GPU's memory must be done explicitly. Actually maybe not a bad idea given that it does take time?

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• (Short example?)



Review of Homeworks

- Homeworks 1 and 2 estimating π with Monte Carlo methods. Basic structure is *Task Parallelism*. Complication is that you need a thread-safe RNG.
- Homework 3 Conway's game of life. Basic structure is *Geometric* Decomposition. Basic idea easy, details a bit messy (particularly for MPI).
- Homework 4 quicksort. Basic structure is *Divide and Conquer*.
- For all programs, probably need large problem sizes to get any benefit from multiple UEs. Even then performance may not be amazingly good, but the primary goal is pedagogical rather than practical.



Minute Essay • How did the course compare with your expectations/goals? Did you learn what you hoped to learn? Slide 14