Languages and Compilers for Multicore Computing Systems

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Topics

- Multicore Systems
- Parallelism
- The Challenge and the Opportunity
The Problem and Why Does It Matter?

- Computers are hitting a performance limit

- “The biggest problem Computer Science has ever faced.” John Hennessy

- “The best opportunity Computer Science has to improve user productivity, application performance, and system integrity.” Fran Allen
The Problem

- Transistors continue to shrink
- More and more transistors fit on a chip
- The chips are faster and faster
- Result: HOT CHIPS!
Real Performance Stops Growing as Fast

Performance (GOPS)


1000
100
10
1
0.1
0.01

Gap

Transistors
Real Performance
Solution: Multicores and Parallelism

- Simpler, slower, cooler processors
- Multiple processors on a chip
- Processors can work on independent parts of the same task
- Software and users organize work to maximize PARALLELISM
Parallelism Solves the Performance Problem! (or does it?)

The Parallel Hammer

We have defined the tool - it is up to you to figure out how to use it!
Moore’s Law predicts that the number of cores will double every 18 - 24 months:
- 2007 - 8 cores on a chip
- 2009 - 16 cores
- 2013 - 64 cores
- 2015 - 128 cores
- 2021 - 1k cores

20?? - LUNATIC LEVELS OF PARALLELISM!!
Parallelism is Moving to the Desktop, the Laptop, Handhelds.... !!

- High performance computing applications and computers have long used parallelism for performance.

- Microprocessors now need parallelism to stay on the performance curve.
Peak Performance Computers

Doubling time = 1.5 yr.
Software Capability is Way Behind Hardware
Multi-core processors are here now

- Can compilers be constructed that abstract thread and data level parallelism from today’s sequential languages?
- Do we need new programming languages?
- Is just more threads the answer?
- Is data parallelism the low hanging fruit?
- Is the answer specialized runtime, middleware, programming models?

Research in these areas is critical
Automatic Parallelization is Hard

- Disambiguating data references is hard:
  - Storage reuse
  - Procedure boundaries
  - Pointers
  - Caches

- Forming useful parallelism is a challenge
  - Data is a problem
  - Multiple models of parallelism
My Wish List of Changes and Opportunities

- Establish what we mean be PARALLELISM
- Establish new clearly defined hierarchical computational models
- Application needs and software capabilities determine the computer architecture not the other way around
- Eliminate caches
- Eliminate pointers
OPPORTUNITIES

- New very high level languages
- New compiler techniques to manage data locality, integrity, ownership, … in the presence of parallelism.
- Influence the architects before it is too late
- Rebuild the software stack
- Establish overall system goals:
  - User Productivity
  - Application Performance
END OF TALK
BEGINNING OF DISCUSSION
*PTRAN (mid-1980s to mid-1990s)*

- Research on automatic parallelization
  - Program Dependence Graphs
  - Constructing Useful Parallelism
  - Static Single Assignment (SSA)
  - Whole Program Analysis Framework
- Compilers for RP3, 3090
  - IBM’s XL Family of Compilers
  - Fortran 90
- Run-time technologies for parallel code
  - Dynamic Process Scheduling
  - Debugging
  - Visualization
Exploiting Multi-Core Multi-Function Chips: Some Options

- Functional threads
- Assist threads
- Speculative threads
- Explicit threads
- Parallel languages

**Programming Intrusiveness**

- Traditional Compiler
- Parallelizing Compiler
- JIT Parallelizing Compiler
- Directives + Compiler
- Parallel Language Compiler

**Compiler Innovations**

- No change to customer code
- Rewrite program

**Hardware Innovations**

- Single-thread program
- Java program
- Annotated program
The Anatomy of a Supercomputer

1 Gigaflop

32 Gigaflops

2 Teraflops

16 Teraflops

1 Petaflop

1,000,000 processors
32,768 chips
512 boards

chip (32 processors)

board (64 chips)

cabinet (8 boards)

(8x8 cabinets)