CS 240A: Shared Memory & Multicore Programming with Cilk++

- Multicore and NUMA architectures
- Multithreaded Programming
- Cilk++ as a concurrency platform
- Work and Span

Thanks to Charles E. Leiserson for some of these slides

Multicore Architecture



Chip Multiprocessor (CMP)

cc-NUMA Architectures

AMD 8-way Opteron Server (neumann@cs.ucsb.edu)



cc-NUMA Architectures

- No Front Side Bus
- Integrated memory controller
- On-die interconnect among CMPs
- Main memory is <u>physically distributed</u> among CMPs (i.e. each piece of memory has an affinity to a CMP)
- NUMA: Non-uniform memory access.
 - For <u>multi-socket servers only</u>
 - Your desktop is safe (well, for now at least)
 - Triton nodes are not NUMA either

Desktop Multicores Today

This is your AMD Barcelona or Intel Core i7 !



Multithreaded Programming

- POSIX Threads (Pthreads) is a set of threading interfaces developed by the IEEE
- "Assembly language" of shared memory programming
- Programmer has to manually:
 - Create and terminate threads
 - Wait for threads to complete
 - Manage interaction between threads using mutexes, condition variables, etc.

Concurrency Platforms

- Programming directly on PThreads is painful and error-prone.
- With PThreads, you either sacrifice memory usage or load-balance among processors
- A *concurrency platform* provides linguistic support and handles load balancing.
- Examples:
 - Threading Building Blocks (TBB)
 - OpenMP
 - Cilk++

Cilk vs PThreads

How will the following code execute in PThreads? In Cilk?

```
for (i=1; i<100000000; i++) {
    spawn-or-fork foo(i);
}
sync-or-join;</pre>
```

What if foo contains code that waits (e.g., spins) on a variable being set by another instance of foo?

They have different <u>liveness</u> properties:

- Cilk threads are spawned lazily, "may" parallelism
- PThreads are spawned eagerly, "must" parallelism

Cilk vs OpenMP

- Cilk++ guarantees space bounds
 - On P processors, Cilk++ uses no more than P times the stack space of a serial execution.
- Cilk++ has a solution for global variables (called "reducers" / "hyperobjects")
- Cilk++ has nested parallelism that works and provides guaranteed speed-up.
 - Indeed, cilk scheduler is provably optimal.
- Cilk++ has a race detector (cilkscreen) for debugging and software release.
- Keep in mind that platform comparisons are (always will be) subject to debate

Complexity Measures



Series Composition



Work: $T_1(A \cup B) = T_1(A) + T_1(B)$ Span: $T_{\infty}(A \cup B) = T_{\infty}(A) + T_{\infty}(B)$

Parallel Composition



Work: $T_1(A \cup B) = T_1(A) + T_1(B)$ Span: $T_{\infty}(A \cup B) = \max\{T_{\infty}(A), T_{\infty}(B)\}$

Speedup

Def. $T_1/T_P = speedup$ on P processors.

If $T_1/T_P = \Theta(P)$, we have *linear speedup*, = P, we have *perfect linear speedup*, > P, we have *superlinear speedup*, which is not possible in this performance model, because of the Work Law $T_P \ge T_1/P$.

Scheduling

A strand is a sequence of instructions that doesn't contain any parallel constructs

- Cilk++ allows the programmer to e potential parall om ir an application
- The Cilk++ *scheduler* maps strands onto processors dynamically at runtime.
- Since on-line schedulers are complicated, we'll explore the ideas with an off-line scheduler.



Greedy Scheduling

IDEA: Do as much as possible on every step. *Definition:* A strand is *ready* if all its <u>predecessors</u> have executed.

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Greedy Scheduling

IDEA: Do as much as possible on every step.

Definition: A strand is *ready* if all its <u>predecessors</u> have executed.

Complete step

- \geq P strands ready.
- Run any P.

Incomplete step

- < P strands ready.
- Run all of them.



Analysis of Greedy

Theorem: Any greedy scheduler achieves $T_P \le T_1/P + T_\infty$.

Proof.

- # complete steps ≤ T₁/P, since each complete step performs P work.
- # incomplete steps ≤ T_∞, since each incomplete step reduces the span of the <u>unexecuted dag</u> by 1.

P = 3

Optimality of Greedy

Corollary. Any greedy scheduler achieves within a factor of 2 of optimal.

Proof. Let T_P^* be the execution time produced by the optimal scheduler. Since $T_P^* \ge max\{T_1/P, T_\infty\}$ by the Work and Span Laws, we have

 $\begin{array}{l} \mathsf{T}_{\mathsf{P}} &\leq \mathsf{T}_{1}/\mathsf{P} + \mathsf{T}_{\infty} \\ &\leq 2 \cdot \max\{\mathsf{T}_{1}/\mathsf{P}, \, \mathsf{T}_{\infty}\} \\ &\leq 2\mathsf{T}_{\mathsf{P}}^{*} \quad \blacksquare \end{array}$

Linear Speedup

Corollary. Any greedy scheduler achieves near-perfect linear speedup whenever $P \ll T_1/T_{\infty}$. *Proof.* Since $P \ll T_1/T_{\infty}$ is equivalent to $T_{\infty} \ll T_1/P$, the Greedy Scheduling Theorem gives us $T_{P} \leq T_{1}/P + T_{\infty}$ $pprox \mathsf{T}_1/\mathsf{P}$, Thus, the speedup is $T_1/T_P \approx P$. **Definition.** The quantity T_1/PT_{∞} is called the *parallel slackness*.











Each worker (processor) maintains a *work deque* of ready strands, and it manipulates the bottom of the deque like a stack





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Theorem: With sufficient parallelism, workers steal infrequently \Rightarrow *linear speed-up*.

Great, how do we program it?

- Cilk++ is a faithful extension of C++
- Often use divide-and-conquer
- Three (really two) hints to the compiler:
 - cilk_spawn: this function can run in parallel with the caller
 - cilk_sync: all spawned children must return before execution can continue
 - cilk_for: all iterations of this loop can run in parallel
 - Compiler translates cilk_for into cilk_spawn & cilk_sync under the covers

Nested Parallelism

Example: Quicksort The named *child* template <typename T> function may execute void qsort(T begin, T end) in parallel with the if (begin != end) { *parent* caller. T middle = partition(begin, end, bind2nd(less<typename iterator_traits<T>::value_type>(), *begin)); cilk_spawn qsort(begin, middle); gsort(max(begin + 1, middle), end); cilk_sync; Control cannot pass this point until all spawned children have returned.

Cilk++ Loops

Example: Matrix transpose



- A cilk_for loop's iterations execute in parallel.
- The index must be declared in the loop initializer.
- The end condition is evaluated exactly once at the beginning of the loop.
- Loop increments should be a const value

Serial Correctness



Serialization

How to seamlessly switch between serial c++ and parallel cilk++ programs?

Add to the **#ifdef** CILKPAR beginning of #include <cilk.h> your program #else #define cilk for for #define cilk main main #define cilk spawn Compile ! #define cilk sync #endif

cilk++ -DCILKPAR -O2 -o parallel.exe main.cpp
 g++ -O2 -o serial.exe main.cpp

Parallel Correctness



Race Bugs

Definition. A *determinacy race* occurs when two logically parallel instructions access the same memory location and at least one of the instructions performs a write.

Example



Race Bugs

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Types of Races

Suppose that instruction A and instruction B both access a location x, and suppose that $A \parallel B$ (A is parallel to B).

Α	В	Race Type
read	read	none
read	write	read race
write	read	read race
write	write	write race

Two sections of code are *independent* if they have no determinacy races between them.

Avoiding Races

- All the iterations of a cilk_for should be independent.
- Between a **cilk_spawn** and the corresponding **cilk_sync**, the code of the spawned child should be independent of the code of the parent, including code executed by additional spawned or called children.

Ex. cilk_spawn qsort(begin, middle);
qsort(max(begin + 1, middle), end);
cilk_sync;

Note: The arguments to a spawned function are evaluated in the parent before the spawn occurs.

Cilk++ Reducers

- Hyperobjects: reducers, holders, splitters
- Primarily designed as a solution to global variables, but has broader application



Data race !



Hyperobjects under the covers

- A reducer hyperobject<T> includes an associative_binary operator \otimes and an identity element.
- Cilk++ runtime system gives each thread a <u>private view</u> of the global variable
- When threads synchronize, their private views are combined with ⊗

Cilkscreen

- Cilkscreen runs off the binary executable:
 - Compile your program with -fcilkscreen
 - Go to the directory with your executable and say cilkscreen your_program [options]
 - Cilkscreen prints info about any races it detects
- Cilkscreen guarantees to report a race if there exists a parallel execution that could produce results different from the serial execution.
- It runs about 20 times slower than singlethreaded real-time.

Parallelism

Because the Span Law dictates that $T_P \ge T_\infty$, the maximum possible speedup given T_1 and T_∞ is $T_1/T_\infty = parallelism$

= the average amount of work per step along the span.



Three Tips on Parallelism

- Minimize span to maximize parallelism. Try to generate 10 times more parallelism than processors for near-perfect linear speedup.
- 2. If you have plenty of parallelism, try to trade some if it off for *reduced work overheads*.
- 3. Use *divide-and-conquer recursion* or *parallel loops* rather than spawning one small thing off after another.



Three Tips on Overheads

- 1. Make sure that work/#spawns is not too small.
 - Coarsen by using function calls and *inlining* near the leaves of recursion rather than spawning.
- Parallelize *outer loops* if you can, not inner loops (otherwise, you'll have high *burdened parallelism*, which includes runtime and scheduling overhead). If you must parallelize an inner loop, coarsen it, but not too much.
 - 500 iterations should be plenty coarse for even the most meager loop. Fewer iterations should suffice for "fatter" loops.
- 3. Use *reducers* only in sufficiently fat loops.