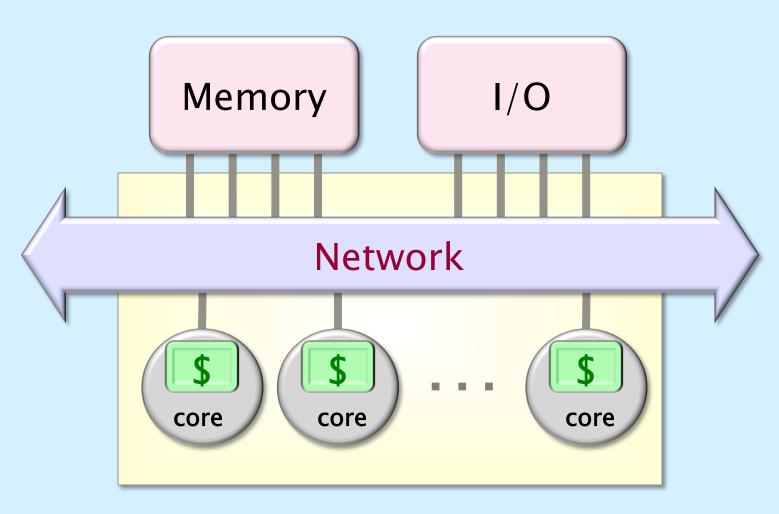
CS 240A: Shared Memory & Multicore Programming with Cilk++

- Multicore and NUMA architectures
- Multithreaded Programming
- Cilk++ as a concurrency platform
- Work, Span, (potential) Parallelism

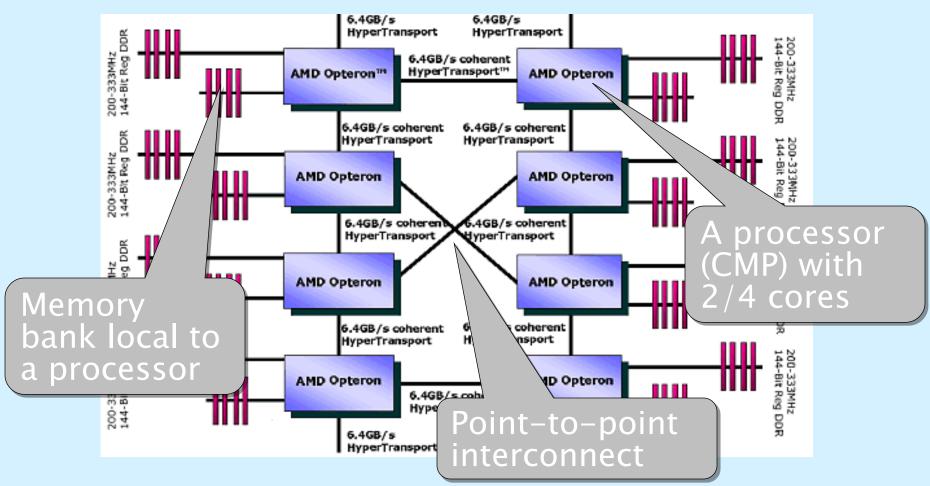
Multicore Architecture



Chip Multiprocessor (CMP)

cc-NUMA Architectures

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

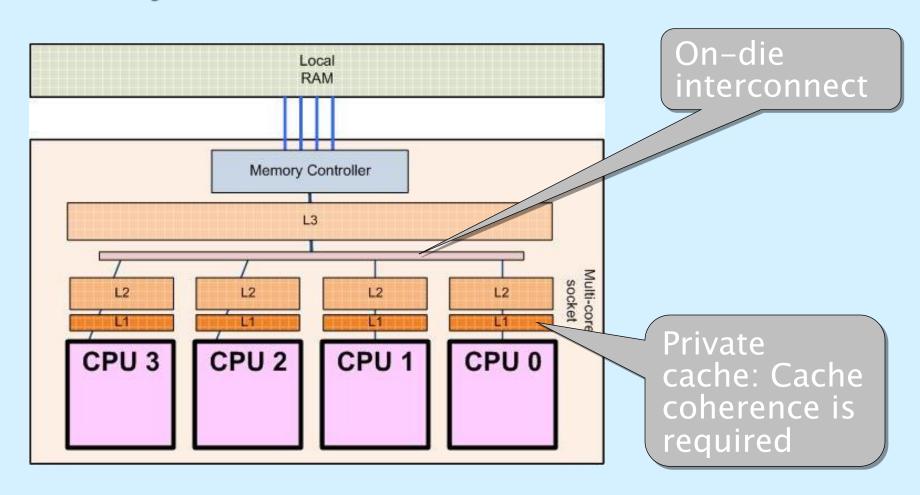


cc-NUMA Architectures

- 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<1000000000; 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:

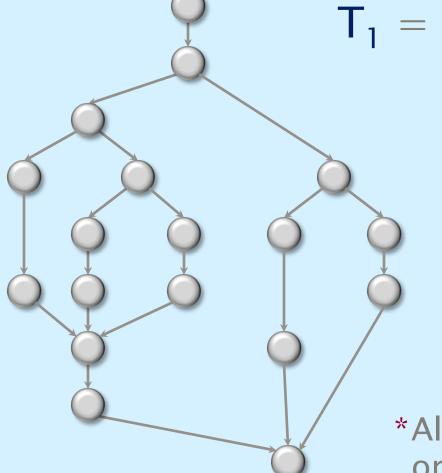
- Cilk threads are spawned lazily, "may" parallelism
- PThreads are spawned eagerly, "must" parallelism

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

 T_P = execution time on P processors



 $T_1 = work$ $T_{\infty} = span^*$

WORK LAW $\bullet T_P \ge T_1/P$

SPAN LAW $T_P \geq T_{\infty}$

*Also called *critical-path length* or *computational depth*.

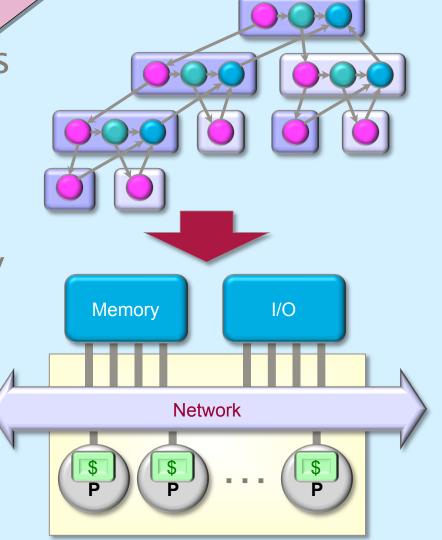
Scheduling

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

• Cilk++ allows the programmer to e ess potential parall sm in an application

The Cilk++ scheduler
maps strands onto
processors dynamically
at runtime.

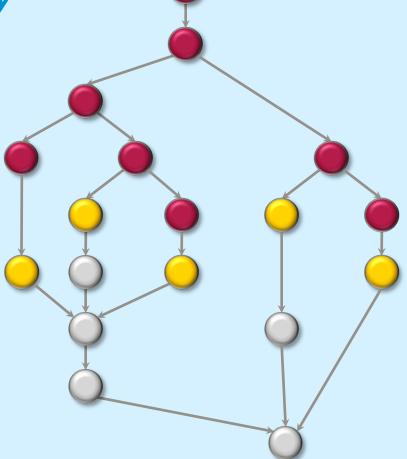
• Since *on-line* schedulers are complicated, we'll explore the ideas with an *off-line* scheduler.



Greedy Scheduling

IDEA: Do as much as possible on every step.

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



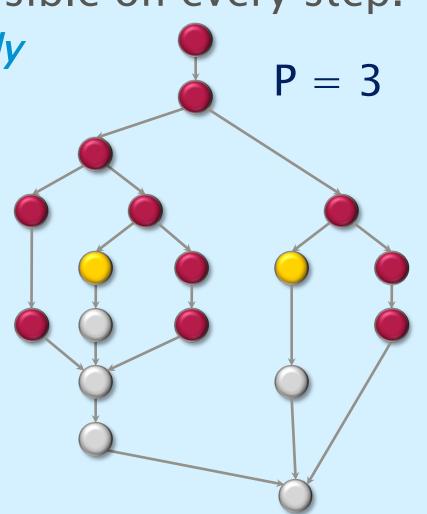
Greedy Scheduling

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Complete step

- ≥ P strands ready.
- Run any P.



Greedy Scheduling

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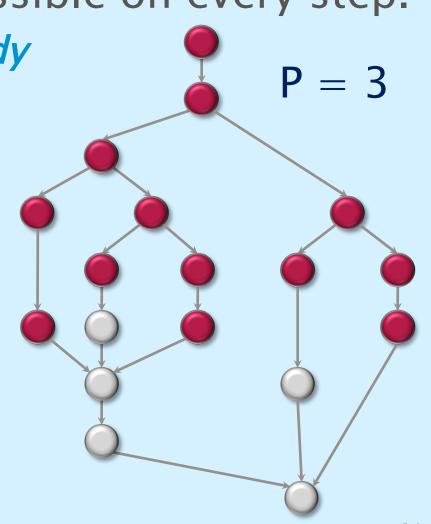
Definition: A strand is **ready** if all its <u>predecessors</u> have executed.

Complete step

- ≥ P strands ready.
- Run any P.

Incomplete step

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



Analysis of Greedy

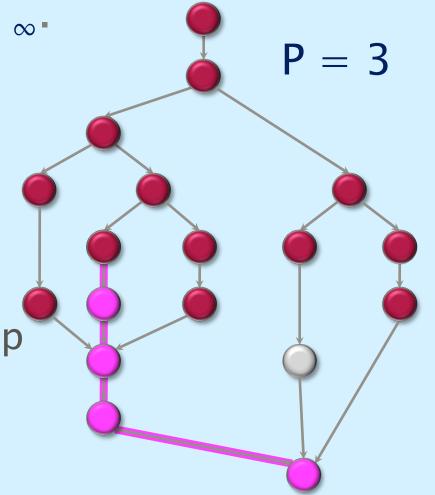
Theorem: Any greedy scheduler achieves

 $T_{P} \leq T_{1}/P + T_{\infty}.$

Proof.

 # complete steps ≤ T₁/P, since each complete step performs P work.

• # incomplete steps $\leq T_{\infty}$, since each incomplete step reduces the span of the unexecuted dag by 1.



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

$$T_P \le T_1/P + T_\infty$$

 $\le 2 \cdot \max\{T_1/P, T_\infty\}$
 $\le 2T_P^*$

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

$$T_P \le T_1/P + T_\infty$$

 $\approx T_1/P$.

Thus, the speedup is $T_1/T_P \approx P$.

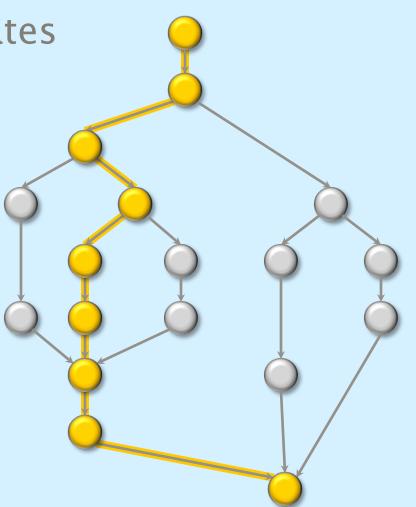
Definition. The quantity T_1/PT_{∞} is called the *parallel slackness*.

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.



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); qsort(max(begin + 1, middle), end); ci k_sync; Control cannot pass this point until all spawned children have returned.

Cilk++ Loops

Example: Matrix transpose

```
cilk_for (int i=1; i<n; ++i) {
    cilk_for (int j=0; j<i; ++j) {
        B[i][j] = A[j][i];
    }
}</pre>
```

- 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

```
int fib (int n) {
   if (n<2) return (n);
   else {
    int x,y;
    x = cilk_spawn fib(n-y = fib(n-2);
    cilk_sync;
   return (x+y);
   }
}</pre>
Cilk++ source
```

The *serialization* is the code with the Cilk++ keywords replaced by null or C++ keywords.

Linker

Binary

int fib (int n) {
 if (n<2) return (n);
 else {
 int x,y;
 x = fib(n-1);
 y = fib(n-2);
 return (x+y);
 }
}</pre>

Cilk++ Runtime Library

Serial correctness can be debugged and verified by running the multithreaded code on a single processor.

Serialization

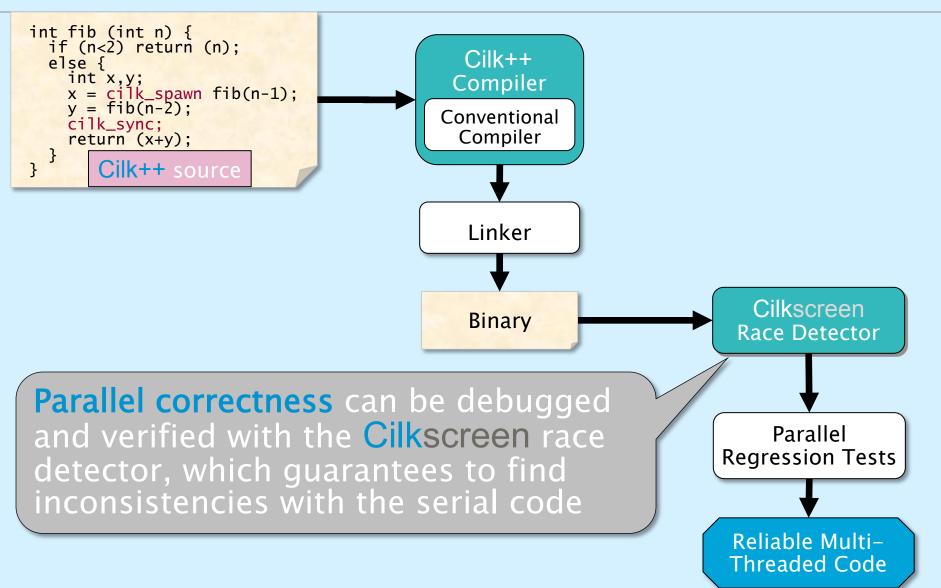
How to seamlessly switch between serial c+ + and parallel cilk++ programs?

```
#ifdef CILKPAR
    #include <cilk.h>
#else
    #define cilk_for for
    #define cilk_main main
    #define cilk_spawn
    #define cilk_sync
#endif
Add to the
beginning of
your program

Compile!
```

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

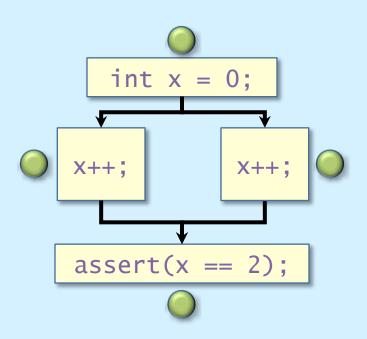
Parallel Correctness



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.

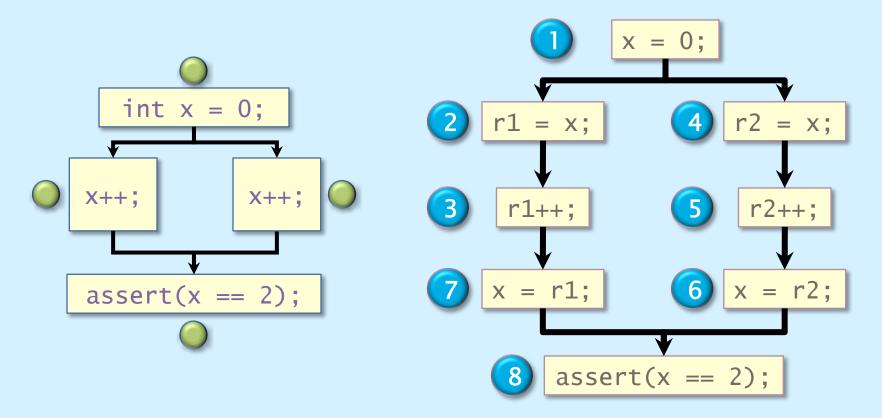
Example



Dependency Graph

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.



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

```
int result = 0;
cilk_for (size_t i = 0; i < N; ++i) {
    result += MyFunc(i);
}</pre>
```

Data race!

```
#include <reducer_opadd.h>
...
cilk::hyperobject < cilk::reducer_opadd < int> > result;
cilk_for (size_t i = 0; i < N; ++i) {
    result() += MyFunc(i);
}
This uses one of the predefined reducers, but you can also write</pre>
```

your own reducer easily

Hyperobjects under the covers

- A reducer hyperobject<T> includes an associative_binary operator ⊗ 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.

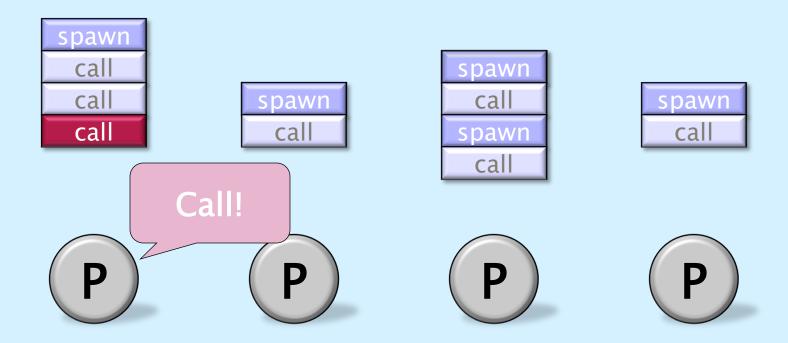
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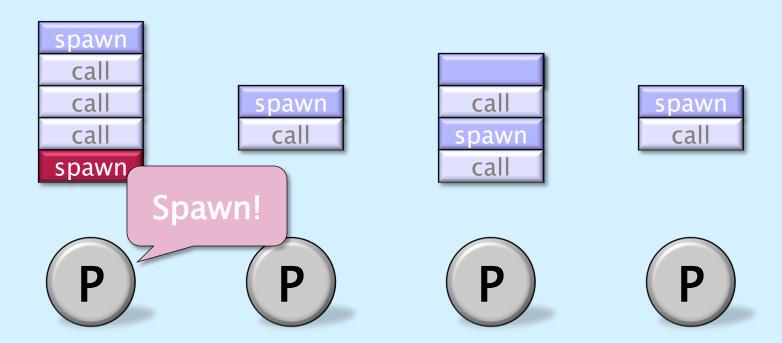
Three Tips on Parallelism

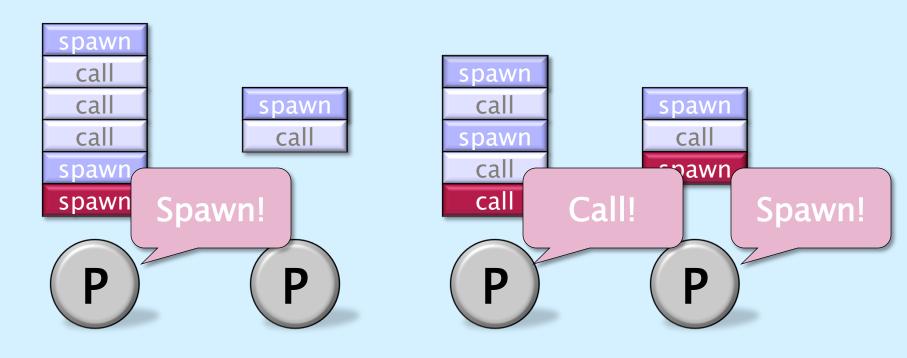
- 1. Minimize span to maximize parallelism. Try to generate 10 times more parallelism than processors for near-perfect linear speedup.
- 2. If you have plenty of parallelism, try to trade some if it off for *reduced work overheads*.
- 3. Use divide-and-conquer recursion or parallel loops rather than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

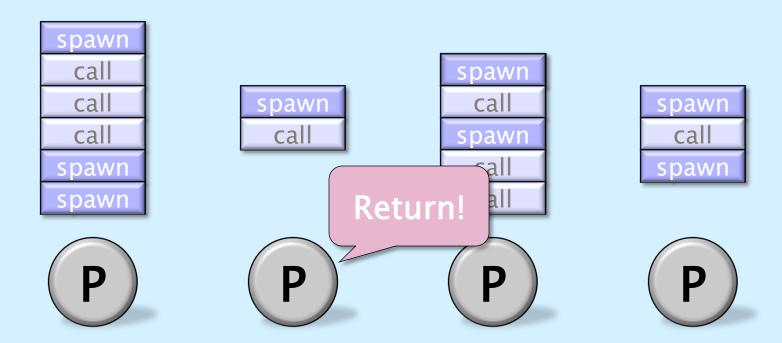
Three Tips on Overheads

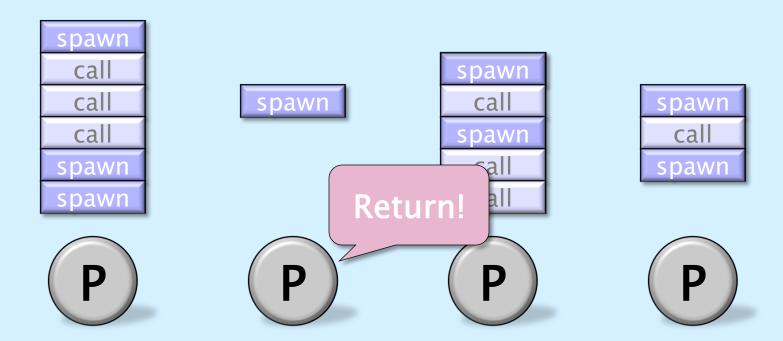
- 1. Make sure that work/#spawns is not too small.
 - Coarsen by using function calls and inlining near the leaves of recursion rather than spawning.
- 2. Parallelize *outer loops* if you can, not inner loops (otherwise, you'll have high *burdened parallelism*, which includes runtime and scheduling overhead). 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.



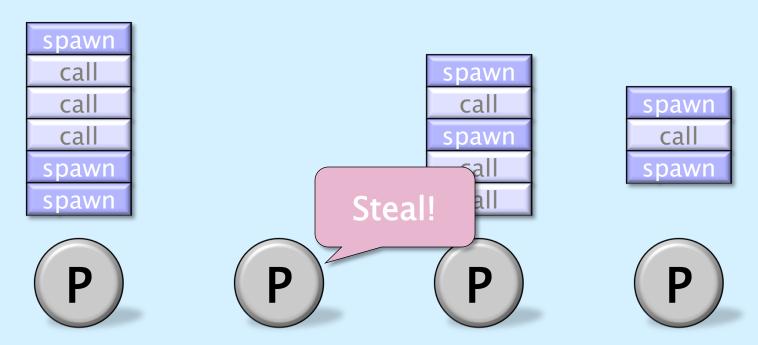






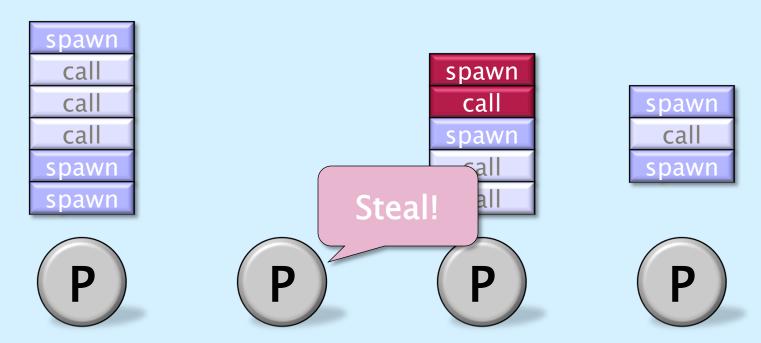


Each worker (processor) maintains a *work deque* of ready strands, and it manipulates the bottom of the deque like a stack





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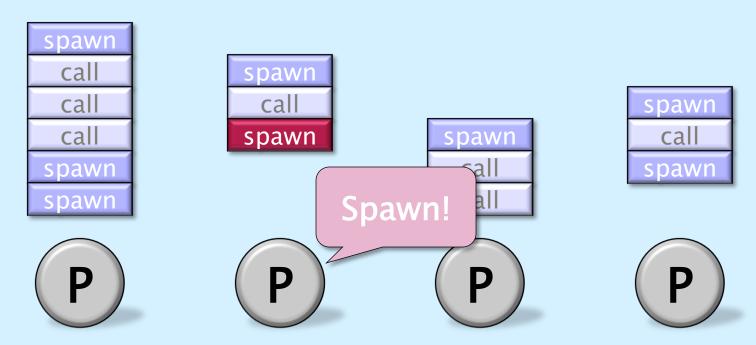






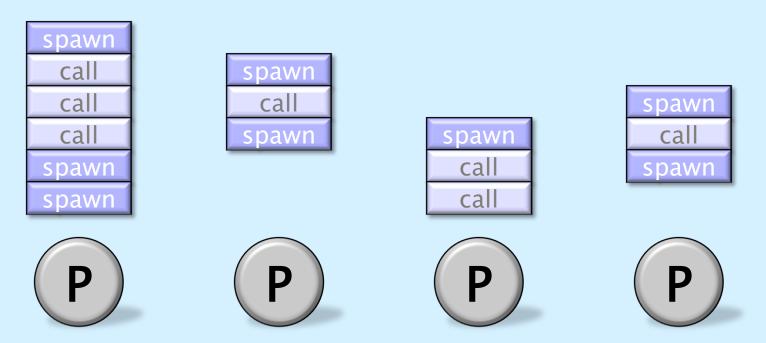


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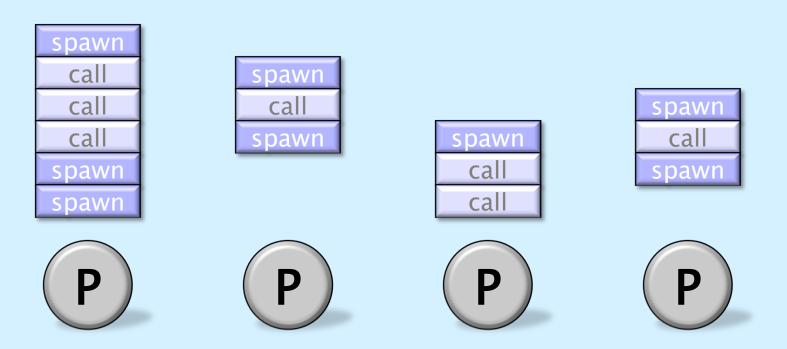


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Theorem: With sufficient parallelism, workers steal infrequently \Rightarrow *linear speed-up*.