WaveScalar: the Executive Summary

Dataflow machine

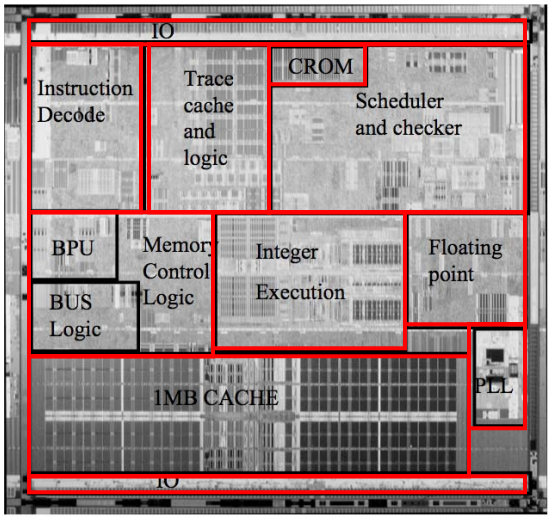
- good at exploiting ILP
- dataflow parallelism + traditional coarser-grain parallelism
  - cheap thread management
- low operand latency because of a hierarchical organization
- memory ordering enforced through **wave-ordered memory**
  - no special languages

WaveScalar

Additional motivation:

- increasing disparity between computation (fast transistors) & communication (long wires)
- increasing circuit complexity
- decreasing fabrication reliability

Monolithic von Neumann Processors



A phenomenal success today.
But in 2016?

☹ Performance
Centralized processing & control
Long wires
e.g., operand broadcast networks

☹ Complexity
40-75% of "design" time is design verification

☹ Defect tolerance
1 flaw -> paperweight

WaveScalar's Microarchitecture

Good performance via distributed microarchitecture ☺

- hundreds of PEs
- dataflow execution – no centralized control
- short point-to-point (producer to consumer) operand communication
- organized hierarchically for fast communication between neighboring PEs
- scalable

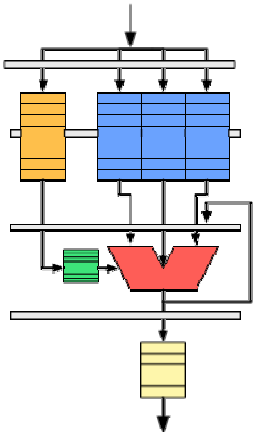
Low design complexity through simple, identical PEs ☺

- design one & stamp out thousands

Defect tolerance ☺

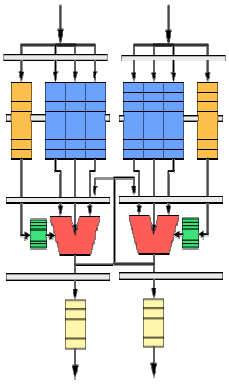
- route around a bad PE

Processing Element



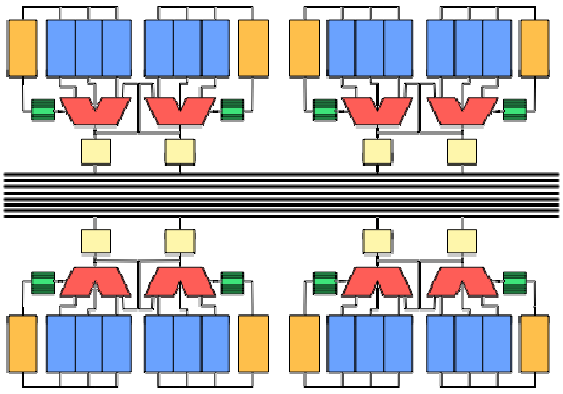
- Simple, small (.5M transistors)
- 5-stage pipeline (receive input operands, match tags, instruction schedule, execute, send output)
- Holds 64 (decoded) instructions
- 128-entry token store
- 4-entry output buffer

PEs in a Pod

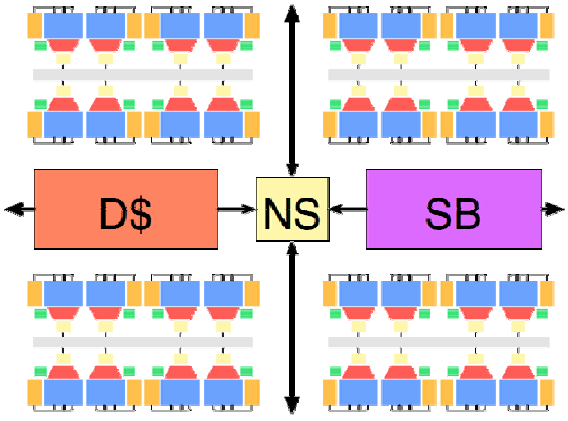


- Share operand bypass network
- Back-to-back producer-consumer execution across PEs
- Relieve congestion on intra-domain bus

Domain



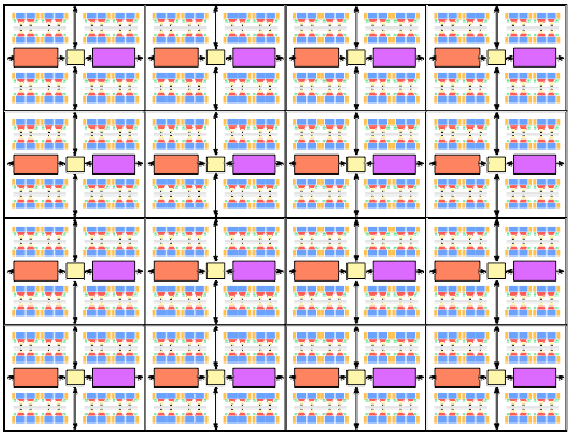
Cluster



WaveScalar Processor

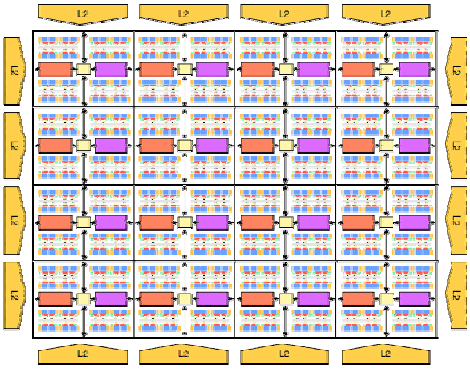
Long distance communication

- dynamic routing
- grid-based network
- 2-cycle hop/cluster



Whole Chip

- Can hold 32K instructions
- Normal memory hierarchy
- Traditional directory-based cache coherence
- ~400 mm^2 in 90 nm technology
- 1GHz.
- ~85 watts



WaveScalar Instruction Placement

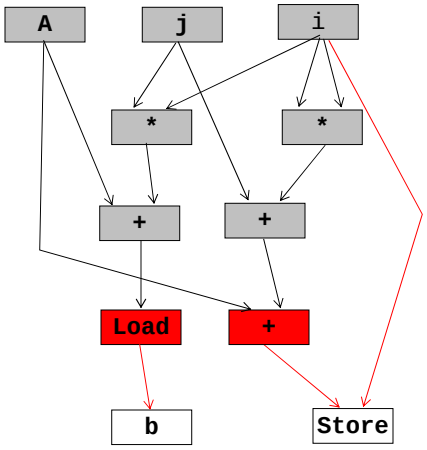
Place instructions in PEs to maximize data locality & instruction-level parallelism.

- Instruction placement algorithm based on a performance model that captures the important performance factors
- Carve the dataflow graph into segments
  - a particular depth to make chains of dependent instructions that will be placed in the same pod
  - a particular width to make multiple independent chains that will be placed in different, but near-by pods
- Snakes segments across PES in the chip on demand
- K-loop bounding to prevent instruction “explosion”

Example to Illustrate the Memory Ordering Problem

```
A[j + i*i] = i;

b = A[i*j];
```



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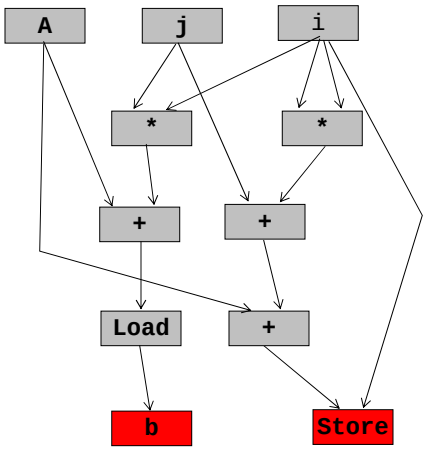

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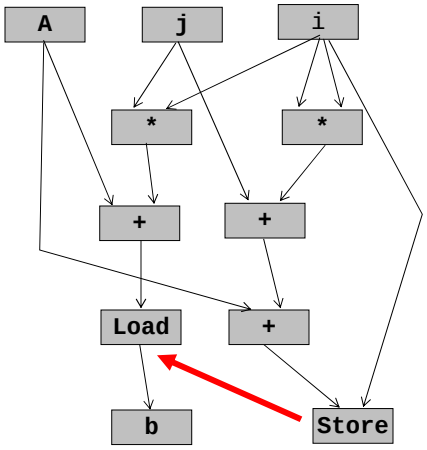
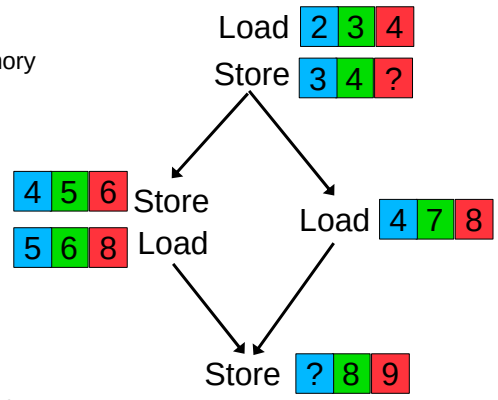
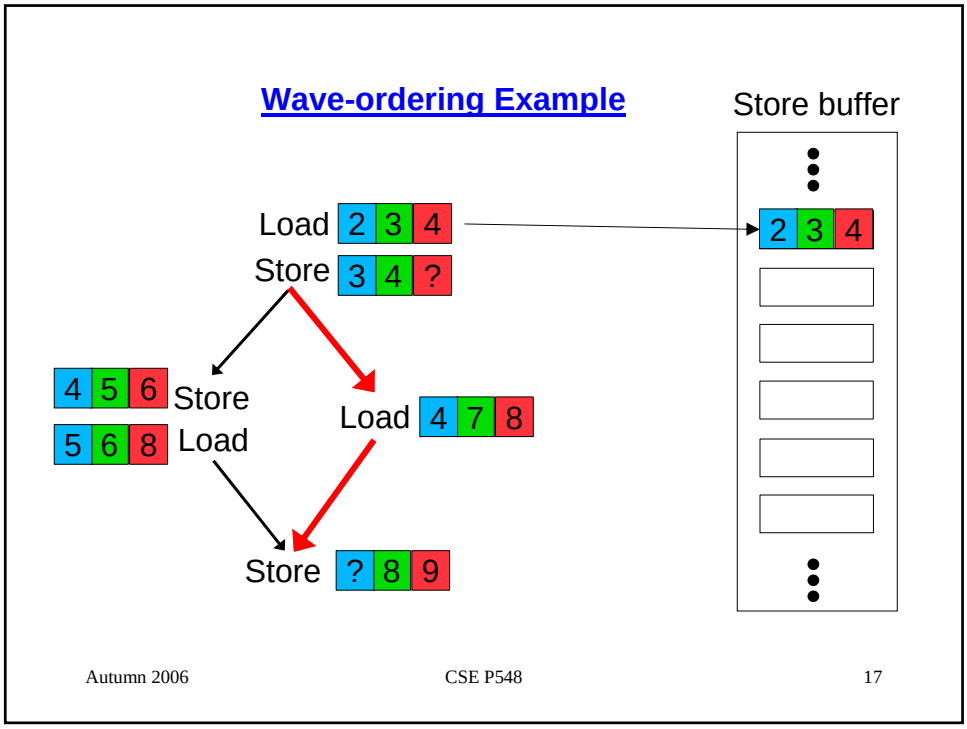


Wave-ordered Memory

- Compiler annotates memory operations
  - Sequence #
  - Successor
  - Predecessor
- Send memory requests in any order
- Hardware reconstructs the correct order




Wave-ordering Example
Store buffer
Load 2 3 4
Store 3 4 ?
4 5 6 Store
5 6 8 Load
Load 4 7 8
Store ? 8 9
2 3 4
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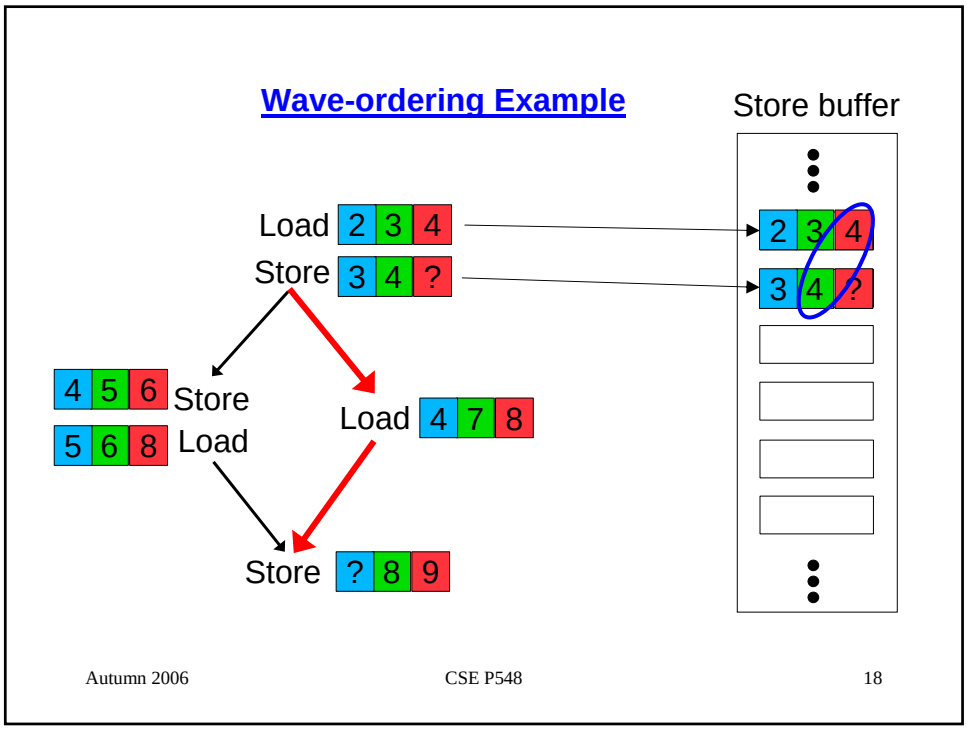
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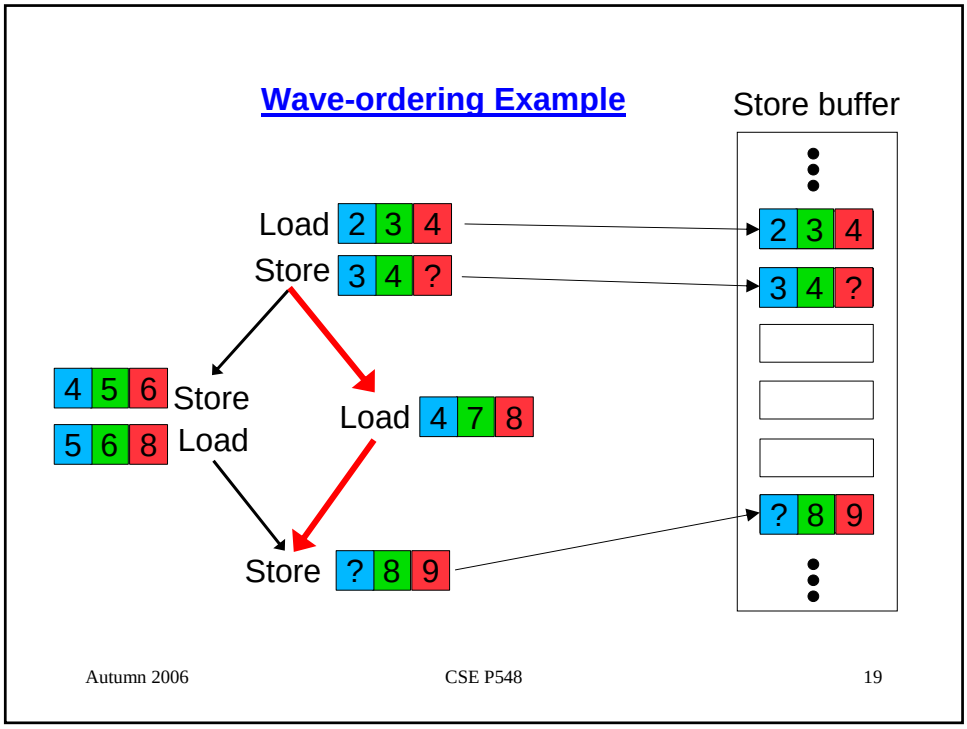
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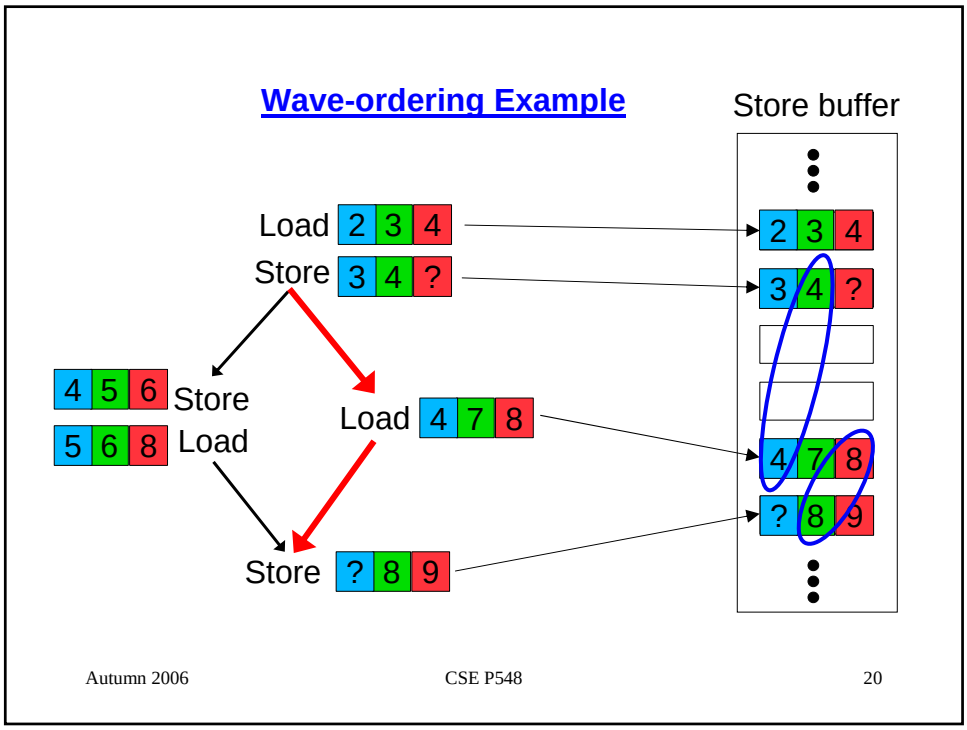
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Wave-ordered Memory

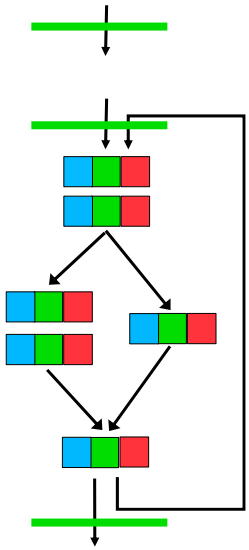
Waves are loop-free sections of the dataflow graph

Each dynamic wave has a **wave number**

Wave number is incremented between waves

Ordering memory:

- wave-numbers
- sequence number within a wave

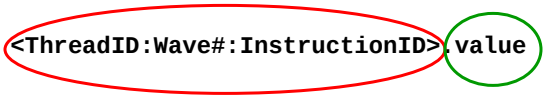


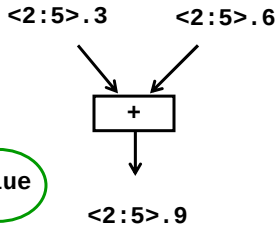
WaveScalar Tag-matching

WaveScalar tag

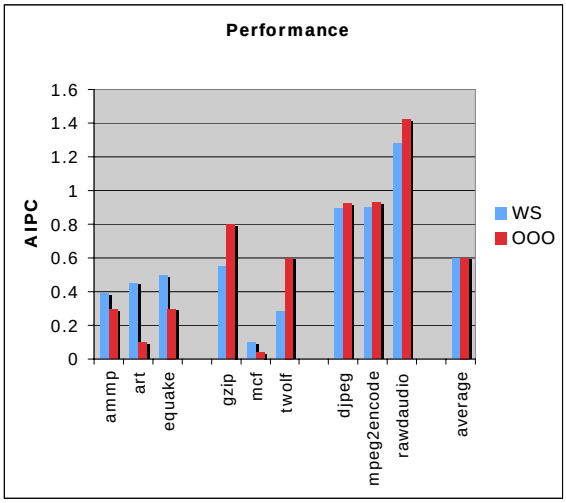
- thread identifier
- wave number

Token: **tag** & **value**

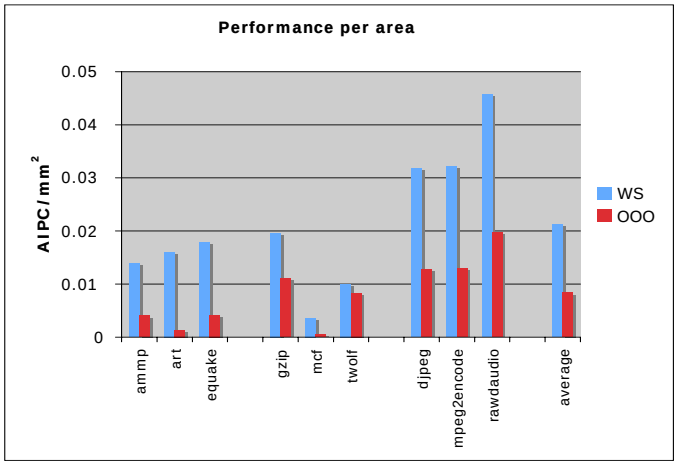




Single-thread Performance



Single-thread Performance per Area



Multithreading the WaveCache

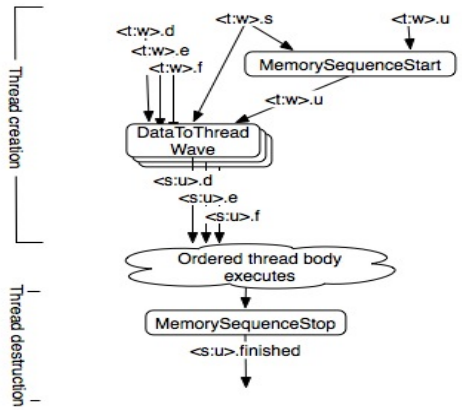
Architectural-support for WaveScalar threads

- instructions to start & stop memory orderings, i.e., threads
- memory-free synchronization to allow exclusive access to data (thread communicate instruction)
- fence instruction to force all previous memory operations to fully execute (to allow other threads to see the results of this one's memory ops)

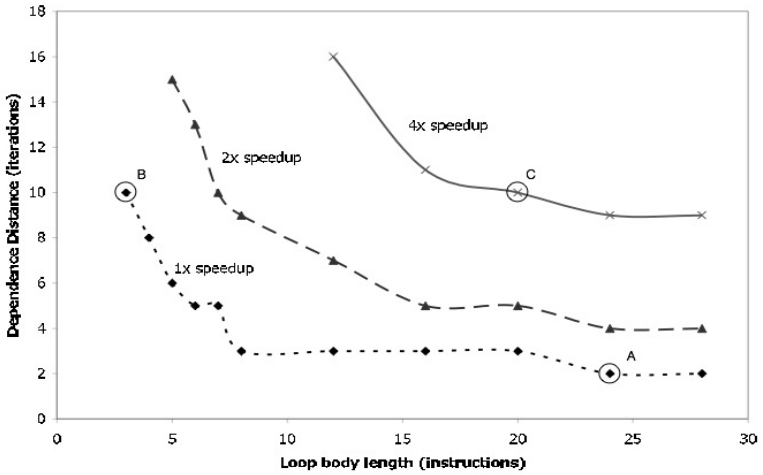
Combine to build threads with multiple granularities

- coarse-grain threads: 25-168X over a single thread; 2-16X over CMP, 5-11X over SMT
- fine-grain, dataflow-style threads: 18-242X over single thread
- combine the two in the same application: 1.6X or 7.9X -> 9X

Creating & Terminating a Thread



Thread Creation Overhead



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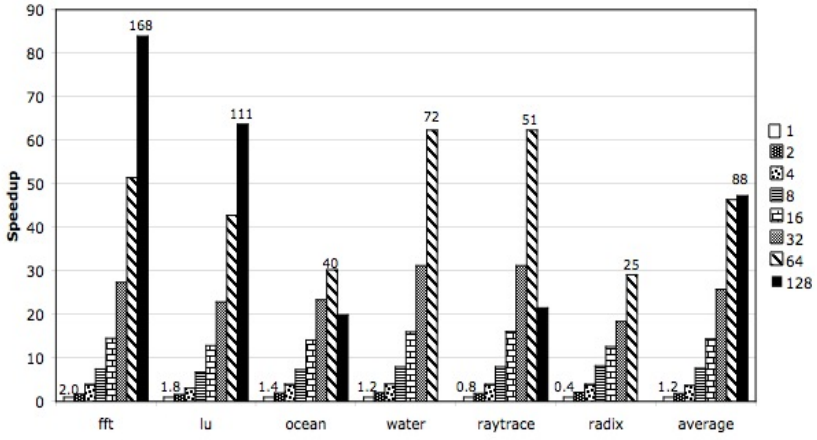
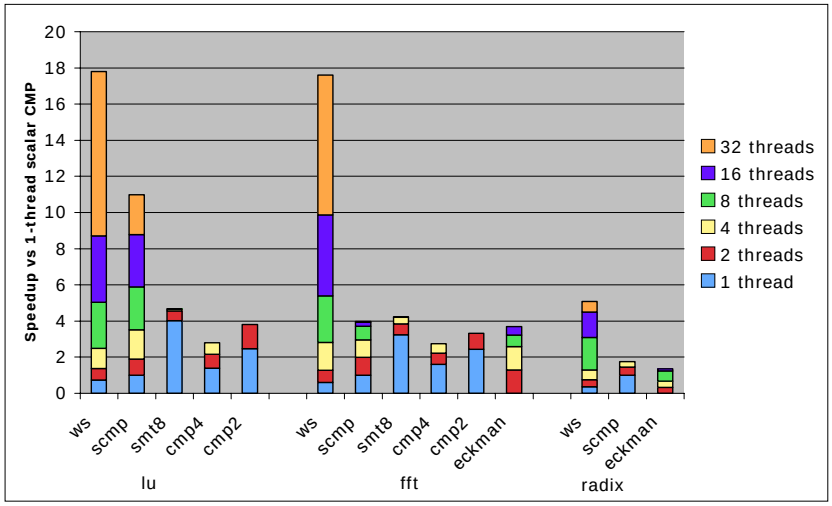
Performance of Coarse-grain Parallelism

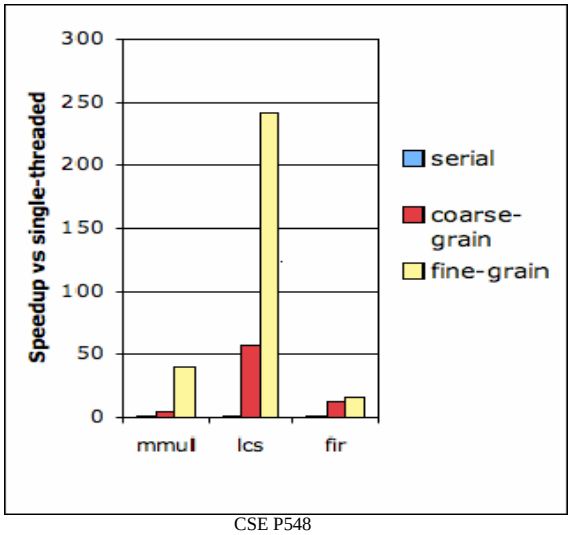


CMP Comparison



Relies on:
Cheap synchronization
Load once, pass data (not load/compute/store)

Performance of Fine-grain Parallelism



Building the WaveCache

RTL-level implementation

- some didn't believe it could be built in a normal-sized chip
- some didn't believe it could achieve a decent cycle time and load-use latencies
- Verilog & Synopsis CAD tools

Different WaveCache's for different applications

- 1 cluster: low-cost, low power, single-thread or embedded
  - 42 mm² in 90 nm process technology, 2.2 AIPC on Splash2
- 16 clusters: multiple threads, higher performance: 378 mm² , 15.8 AIPC

Board-level FPGA implementation

- OS & real application simulations