

Memory Systems

Fundamentals, Recent Research, Challenges, Opportunities

Lecture 8: Asymmetric Multi-Core

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11 October 2018

Technion Fast Course 2018

Some Readings

- Suleman et al., “Accelerating Critical Section Execution with Asymmetric Multi-Core Architectures,” ASPLOS 2009.
- Joao et al., “Bottleneck Identification and Scheduling in Multithreaded Applications,” ASPLOS 2012.
- Joao et al., “Bottleneck Identification and Scheduling for Multithreaded Applications,” ASPLOS 2012.
- Joao et al., “Utility-Based Acceleration of Multithreaded Applications on Asymmetric CMPs,” ISCA 2013.
- Grochowski et al., “Best of Both Latency and Throughput,” ICCD 2004.

Heterogeneity (Asymmetry)

Heterogeneity (Asymmetry) → Specialization

- Heterogeneity and asymmetry have the same meaning
 - Contrast with homogeneity and symmetry
- Heterogeneity is a very general system design concept (and *life* concept, as well)
- Idea: Instead of having multiple instances of a “resource” to be the same (i.e., homogeneous or symmetric), design some instances to be different (i.e., heterogeneous or asymmetric)
- Different instances can be optimized to be more efficient in executing different types of workloads or satisfying different requirements/goals
 - Heterogeneity enables specialization/customization

Why Asymmetry in Design? (I)

- Different workloads executing in a system can have different behavior
 - Different applications can have different behavior
 - Different execution phases of an application can have different behavior
 - The same application executing at different times can have different behavior (due to input set changes and dynamic events)
 - E.g., locality, predictability of branches, instruction-level parallelism, data dependencies, serial fraction in a parallel program, bottlenecks in parallel portion of a program, interference characteristics, ...
- Systems are designed to satisfy different metrics at the same time
 - There is almost never a single goal in design, depending on design point
 - E.g., Performance, energy efficiency, fairness, predictability, reliability, availability, cost, memory capacity, latency, bandwidth, ...

Why Asymmetry in Design? (II)

- Problem: Symmetric design is one-size-fits-all
- It tries to fit a single-size design to all workloads and metrics
- It is very difficult to come up with a single design
 - that satisfies all workloads even for a single metric
 - that satisfies all design metrics at the same time
- This holds true for different system components, or resources
 - Cores, caches, memory, controllers, interconnect, disks, servers, ...
 - Algorithms, policies, ...

Asymmetry Enables Customization

c	c	c	c
c	c	c	c
c	c	c	c
c	c	c	c

Symmetric

C1		C2	
		C3	
C4	C4	C4	C4
C5	C5	C5	C5

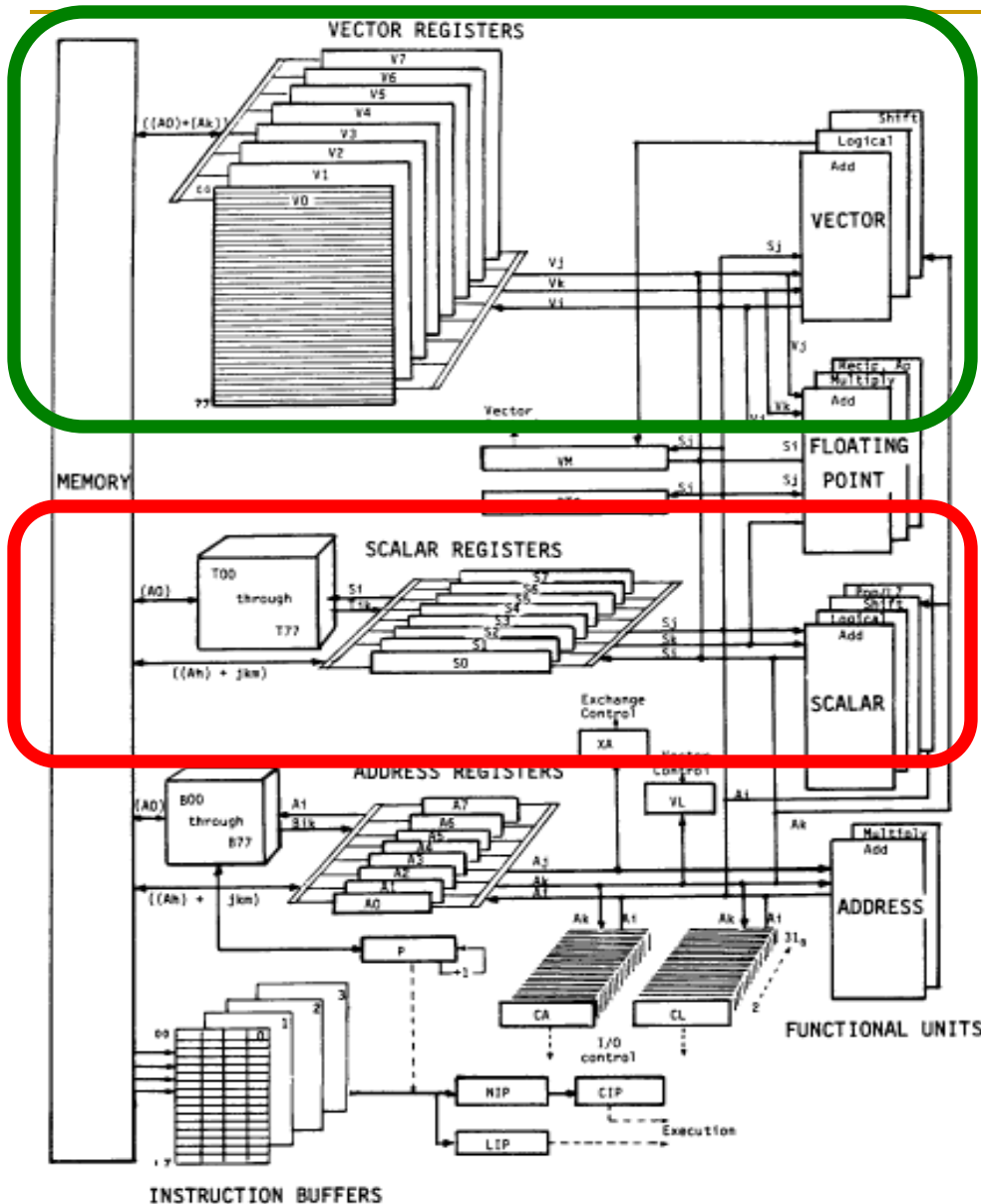
Asymmetric

- Symmetric: One size fits all
 - Energy and performance suboptimal for different “workload” behaviors
- Asymmetric: Enables customization and adaptation
 - Processing requirements vary across workloads (applications and phases)
 - Execute code on best-fit resources (minimal energy, adequate perf.)

We Have Already Seen Examples (Before)

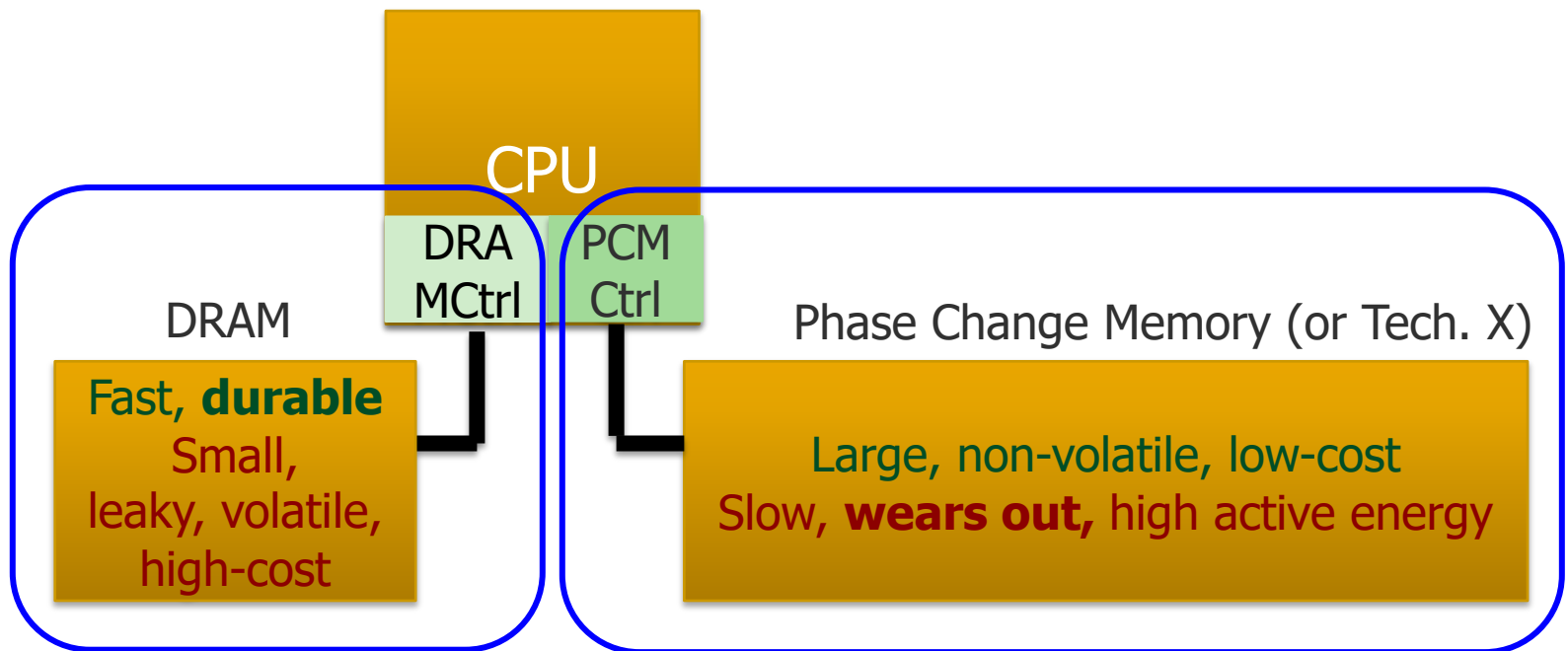
- CRAY-1 design: scalar + vector pipelines
 - Modern processors: scalar instructions + SIMD extensions
 - Decoupled Access Execute: access + execute processors
 - Thread Cluster Memory Scheduling: different memory scheduling policies for different thread clusters
 - RAIDR: Heterogeneous refresh rates in DRAM
 - Heterogeneous-Latency DRAM (Tiered Latency DRAM)
 - Hybrid memory systems
 - DRAM + Phase Change Memory
 - Fast, Costly DRAM + Slow, Cheap DRAM
 - Reliable, Costly DRAM + Unreliable, Cheap DRAM
 - Heterogeneous cache replacement policies
-

An Example Asymmetric Design: CRAY-1



- CRAY-1
- Russell, “The CRAY-1 computer system,” CACM 1978.
- Scalar and vector modes
- 8 64-element vector registers
- 64 bits per element
- 16 memory banks
- 8 64-bit scalar registers
- 8 24-bit address registers

Remember: Hybrid Memory Systems



Hardware/software manage data allocation and movement
to achieve the best of multiple technologies

Meza+, "Enabling Efficient and Scalable Hybrid Memories," IEEE Comp. Arch. Letters, 2012.
Yoon, Meza et al., "Row Buffer Locality Aware Caching Policies for Hybrid Memories," ICCD 2012 Best Paper Award.

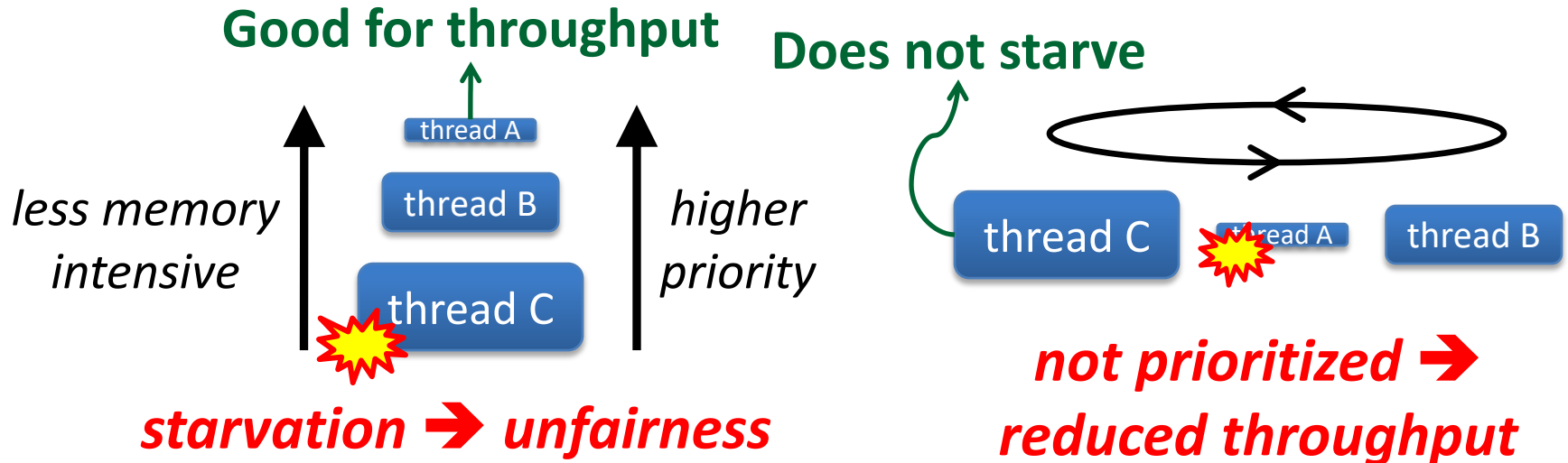
Remember: Throughput vs. Fairness

Throughput biased approach

Prioritize less memory-intensive threads

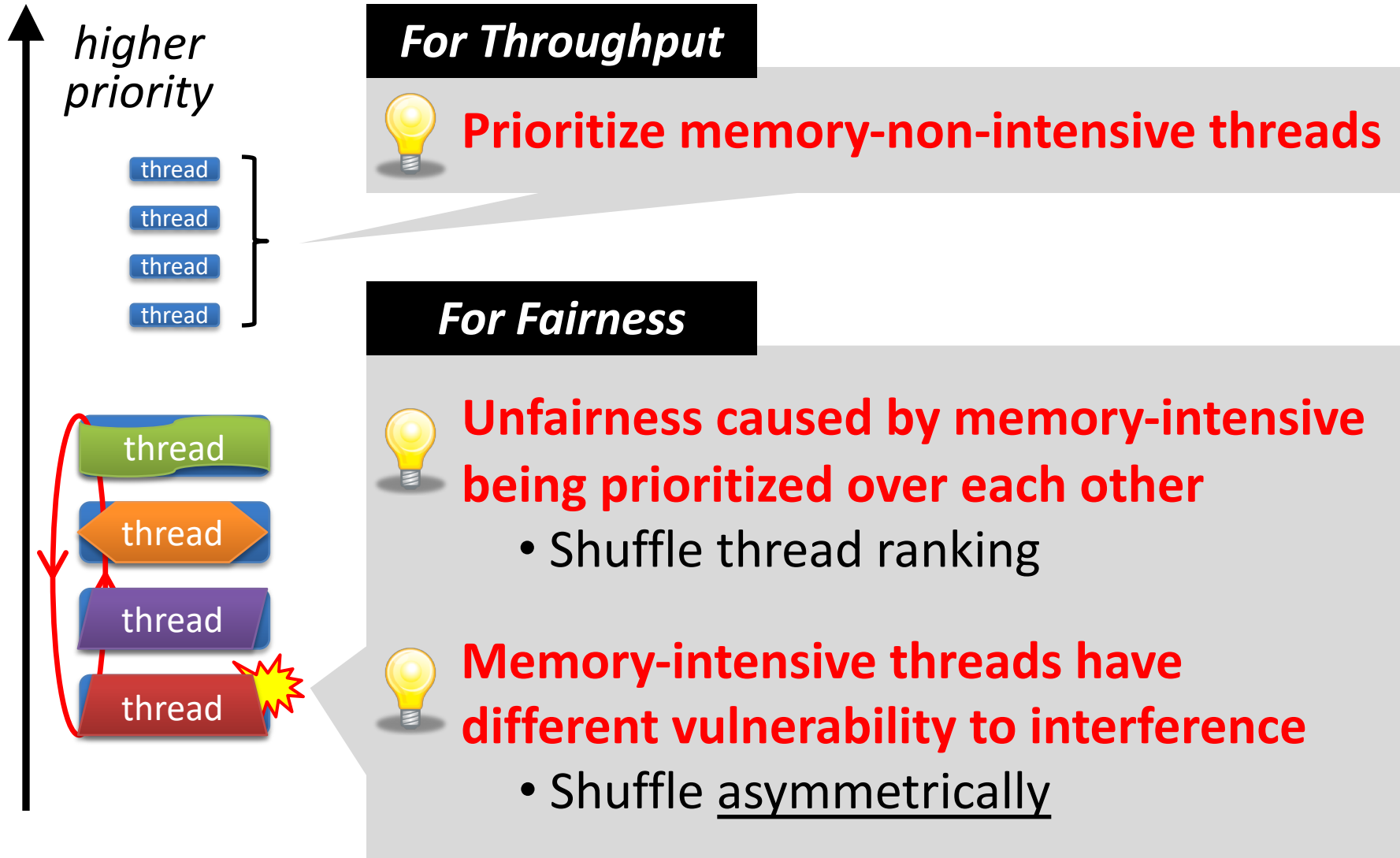
Fairness biased approach

Take turns accessing memory



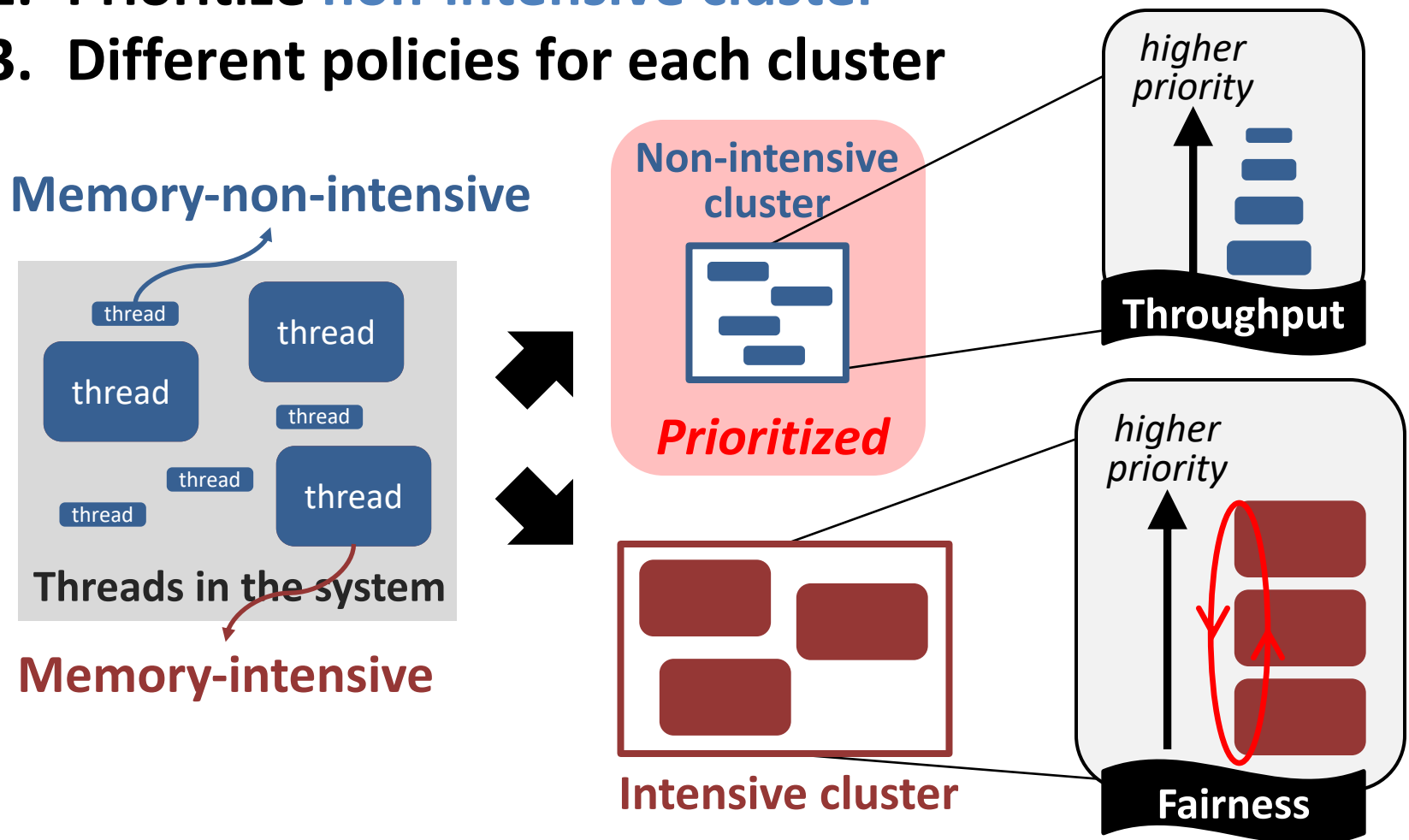
Single policy for all threads is insufficient

Remember: Achieving the Best of Both Worlds

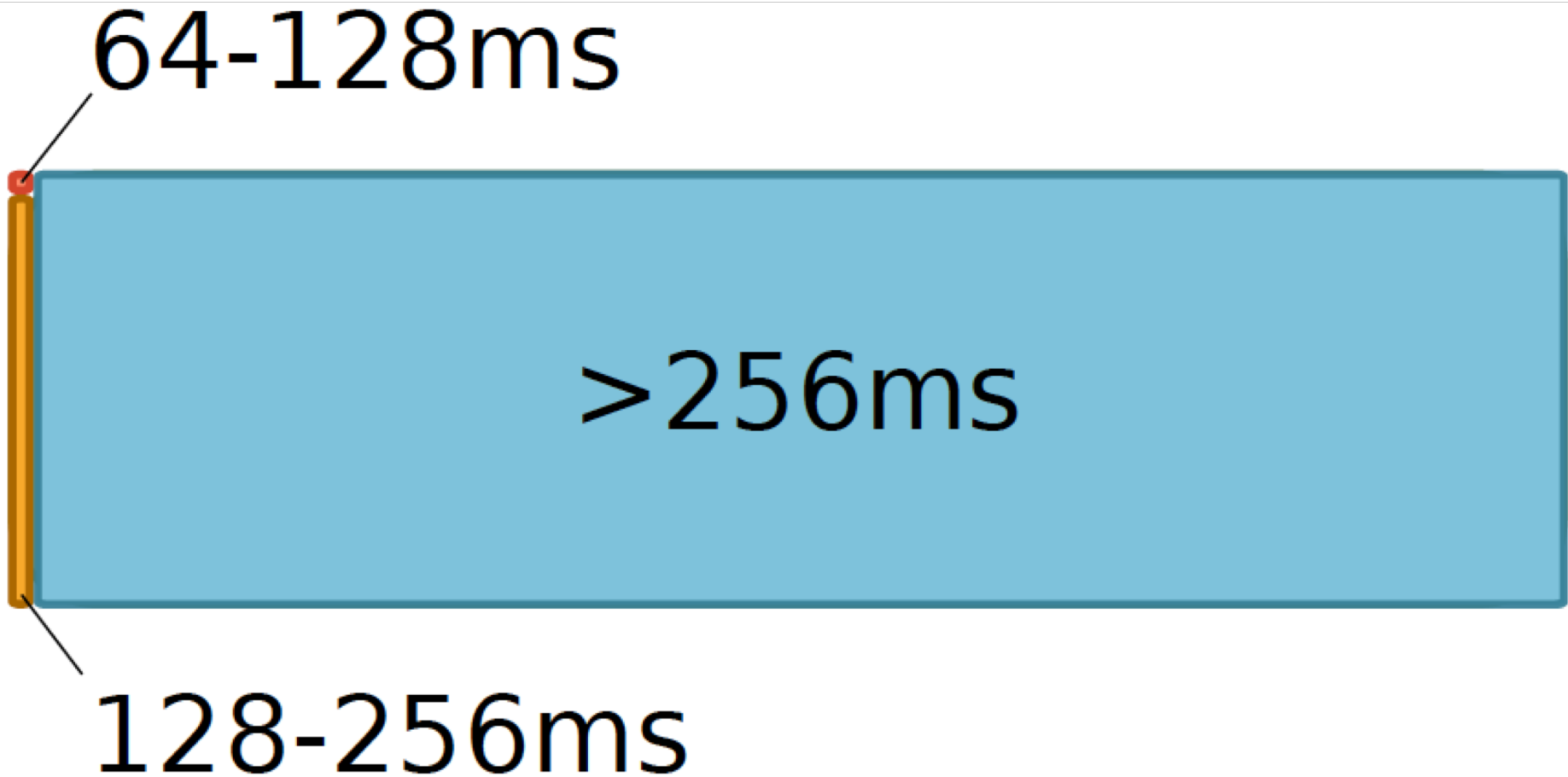


Thread Cluster Memory Scheduling [Kim+ MICRO'10]

1. Group threads into two **clusters**
2. Prioritize **non-intensive cluster**
3. Different policies for each cluster



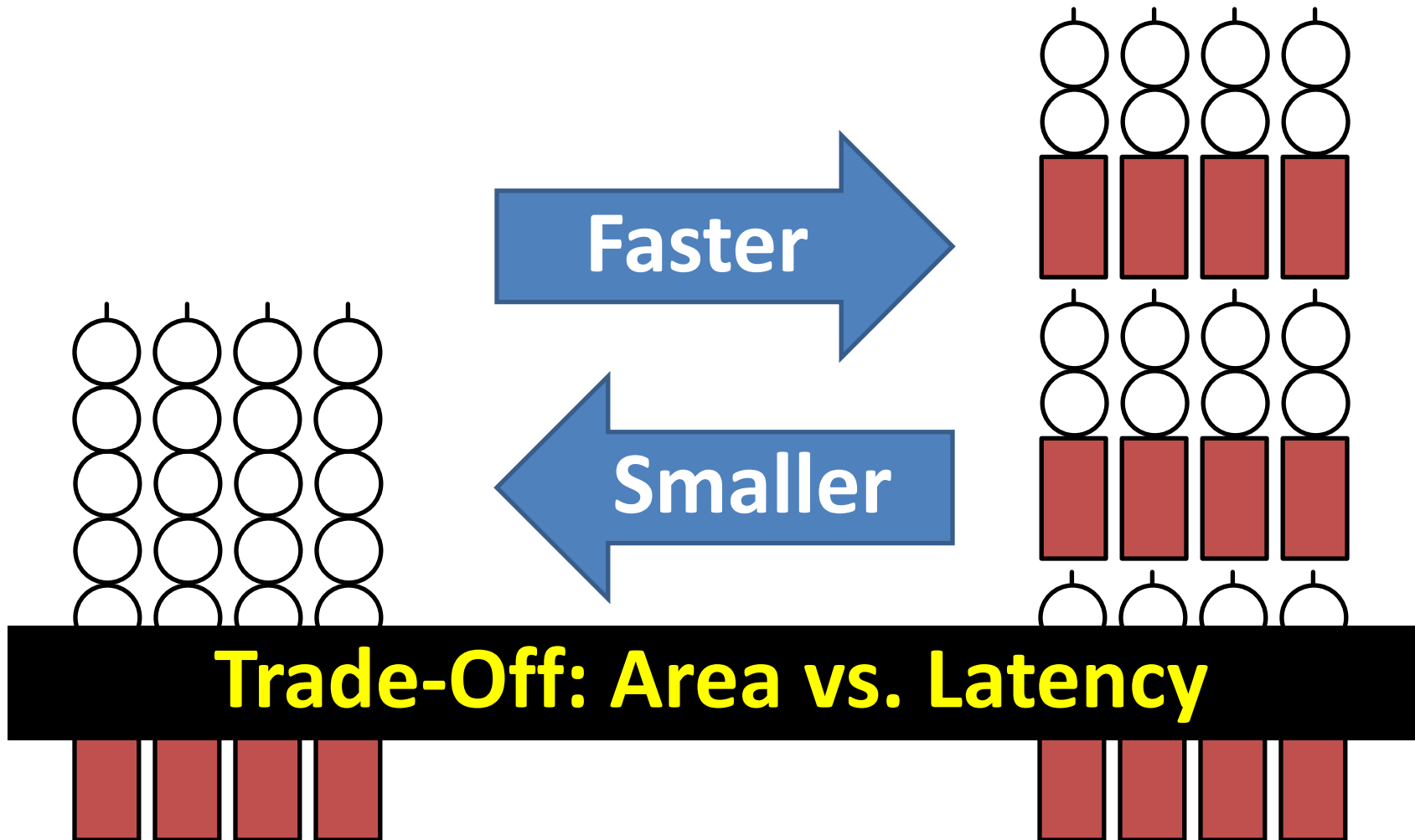
Remember: Heterogeneous Retention Times in DRAM



Trade-Off: Area (Die Size) vs. Latency

Long Bitline

Short Bitline

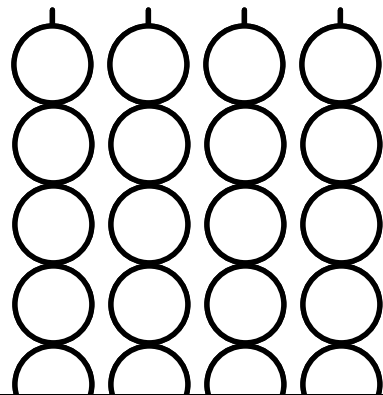


Approximating the Best of Both Worlds

Long Bitline

Small Area

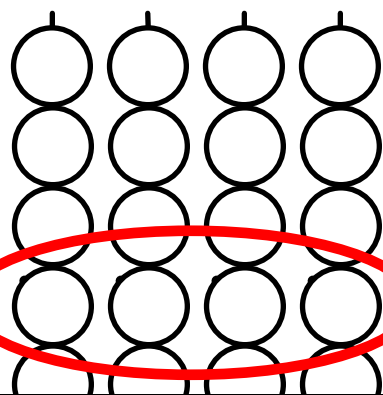
~~High Latency~~



Need Isolation

Our Proposal

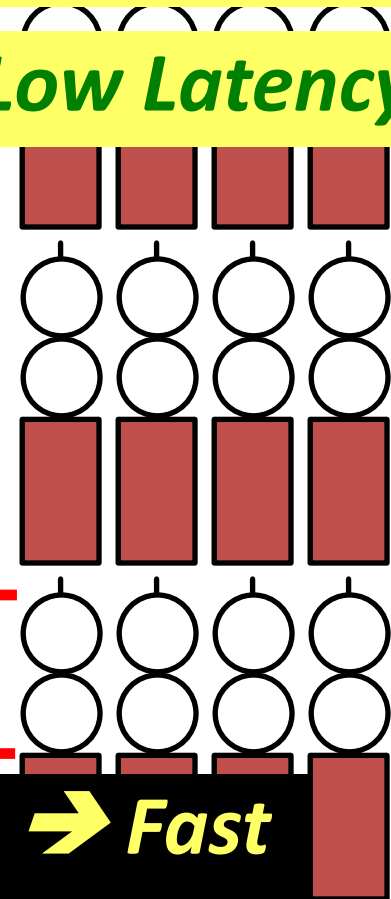
Add Isolation Transistors



Short Bitline

~~Large Area~~

Low Latency



Fast

Approximating the Best of Both Worlds

Long Bitline Tiered-Latency DRAM **Short Bitline**

Small Area

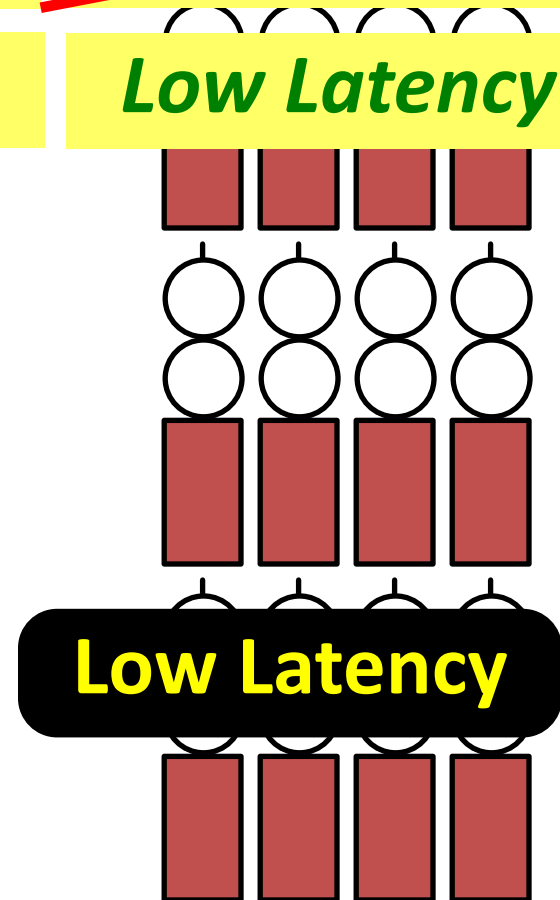
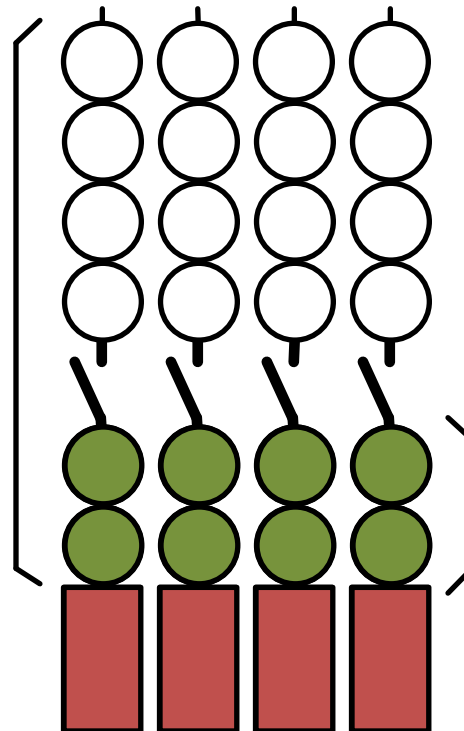
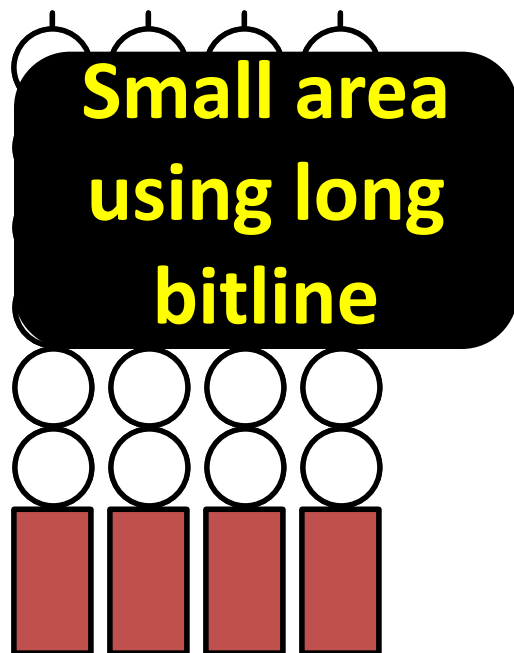
Small Area

~~*Large Area*~~

~~*High Latency*~~

Low Latency

Low Latency



Heterogeneous Interconnects (in Tiler)

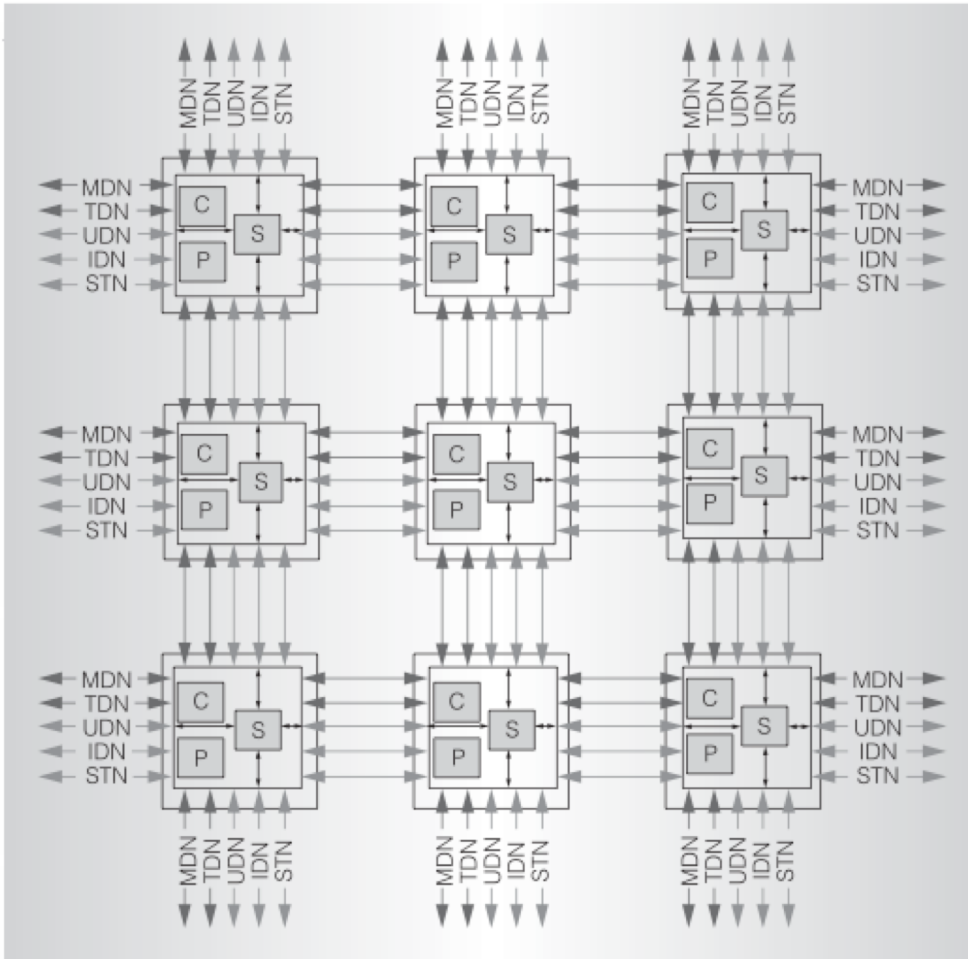


Figure 3. A 3×3 array of tiles connected by networks. (MDN: memory dynamic network; TDN: tile dynamic network; UDN: user dynamic network; IDN: I/O dynamic network; STN: static network.)

- 2D Mesh
- Five networks
- Four packet switched
 - Dimension order routing, wormhole flow control
 - TDN: Cache request packets
 - MDN: Response packets
 - IDN: I/O packets
 - UDN: Core to core messaging
- One circuit switched
 - STN: Low-latency, high-bandwidth static network
 - Streaming data

Aside: Examples from Life

- Heterogeneity is abundant in life
 - both in nature and human-made components
- Humans are heterogeneous
- Cells are heterogeneous → specialized for different tasks
- Organs are heterogeneous
- Cars are heterogeneous
- Buildings are heterogeneous
- Rooms are heterogeneous
- ...

General-Purpose vs. Special-Purpose

- Asymmetry is a way of enabling specialization
- It bridges the gap between purely general purpose and purely special purpose
 - Purely general purpose: Single design for every workload or metric
 - Purely special purpose: Single design per workload or metric
 - Asymmetric: Multiple sub-designs optimized for sets of workloads/metrics and glued together
- The goal of a good asymmetric design is to get the best of both general purpose and special purpose

Asymmetry Advantages and Disadvantages

- Advantages over Symmetric Design
 - + Can enable optimization of multiple metrics
 - + Can enable better adaptation to workload behavior
 - + Can provide special-purpose benefits with general-purpose usability/flexibility

- Disadvantages over Symmetric Design
 - Higher overhead and more complexity in design, verification
 - Higher overhead in management: scheduling onto asymmetric components
 - Overhead in switching between multiple components can lead to degradation

Yet Another Example

- Modern processors integrate general purpose cores and GPUs
 - CPU-GPU systems
 - Heterogeneity in execution models

Three Key Problems in Future Systems

■ Memory system

- ❑ Applications are increasingly data intensive
- ❑ Data storage and movement limits performance & efficiency

■ Efficiency (performance and energy) → scalability

- ❑ Enables scalable systems → new applications
- ❑ Enables better user experience → new usage models

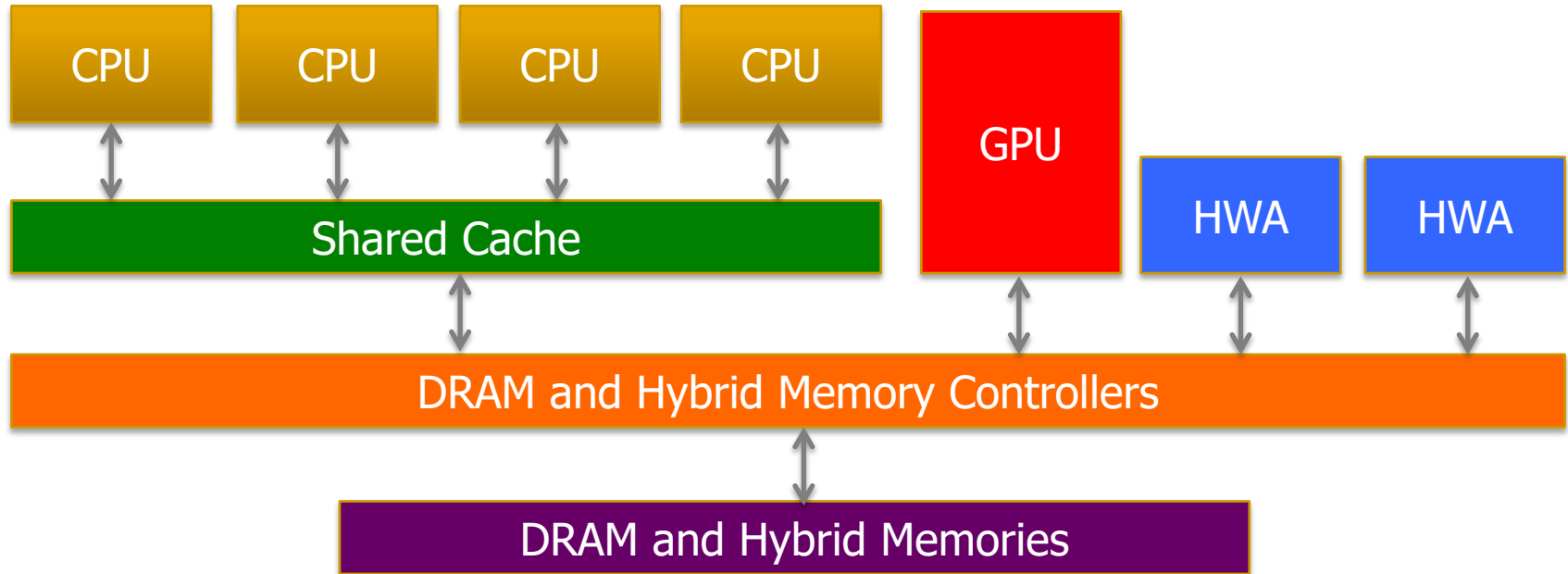
■ Predictability and robustness

**Asymmetric Designs
Can Help Solve These Problems**

Commercial Asymmetric Design Examples

- Integrated CPU-GPU systems (e.g., Intel SandyBridge)
- CPU + Hardware Accelerators (e.g., your cell phone)
- ARM big.LITTLE processor
- IBM Cell processor

Increasing Asymmetry in Modern Systems

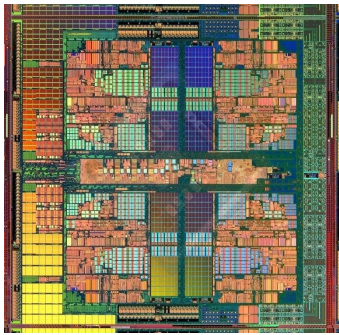


- Heterogeneous agents: CPUs, GPUs, and HWAs
- Heterogeneous memories: Fast vs. Slow DRAM
- Heterogeneous interconnects: Control, Data, Synchronization

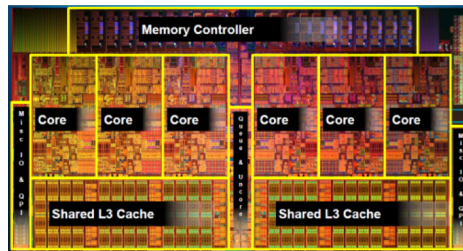
Multi-Core Design: An Asymmetric Perspective

Many Cores on Chip

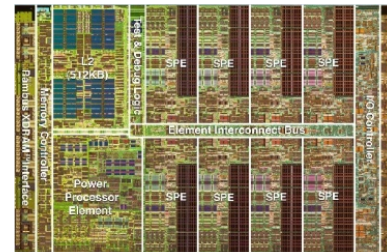
- Simpler and lower power than a single large core
- Large scale parallelism on chip



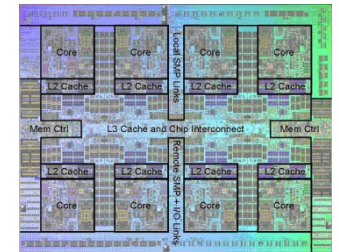
AMD Barcelona
4 cores



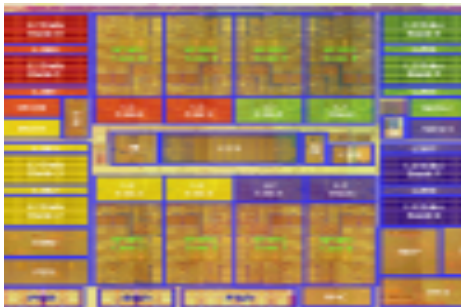
Intel Core i7
8 cores



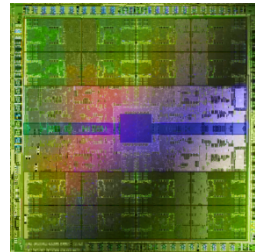
IBM Cell BE
8+1 cores



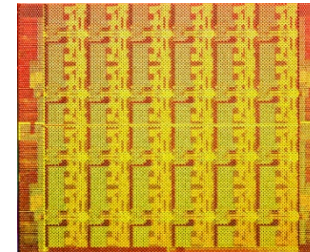
IBM POWER7
8 cores



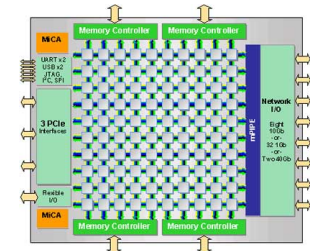
Sun Niagara II
8 cores



Nvidia Fermi
448 "cores"



Intel SCC
48 cores, networked



Tiler TILE Gx
100 cores, networked

With Many Cores on Chip

- What we want:
 - N times the performance with N times the cores when we parallelize an application on N cores
- What we get:
 - Amdahl's Law (serial bottleneck)
 - Bottlenecks in the parallel portion

Caveats of Parallelism

■ Amdahl's Law

- f : Parallelizable fraction of a program
- N : Number of processors

$$\text{Speedup} = \frac{1}{1 - f + \frac{f}{N}}$$

- Amdahl, “Validity of the single processor approach to achieving large scale computing capabilities,” AFIPS 1967.
- **Maximum speedup limited by serial portion: Serial bottleneck**
- **Parallel portion is usually not perfectly parallel**
 - **Synchronization** overhead (e.g., updates to shared data)
 - **Load imbalance** overhead (imperfect parallelization)
 - **Resource sharing** overhead (contention among N processors)

The Problem: Serialized Code Sections

- Many parallel programs cannot be parallelized completely
- Causes of serialized code sections
 - Sequential portions (Amdahl's "serial part")
 - Critical sections
 - Barriers
 - Limiter stages in pipelined programs
- Serialized code sections
 - Reduce performance
 - Limit scalability
 - Waste energy

Example from MySQL

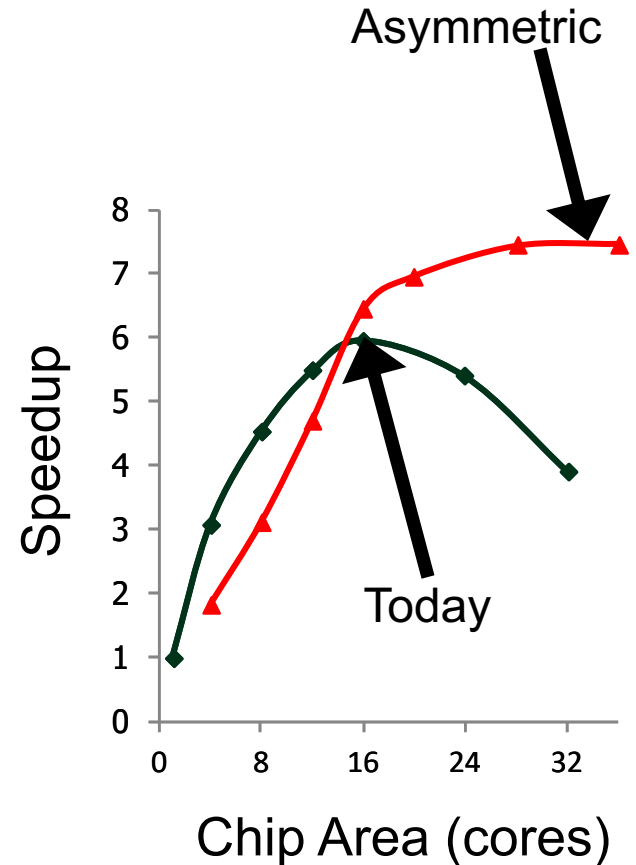
**Critical
Section**

Access Open Tables Cache

Open database tables

Perform the operations
....

Parallel



Demands in Different Code Sections

- What we want:
- In a serialized code section → one powerful “large” core
- In a parallel code section → many wimpy “small” cores
- These two conflict with each other:
 - If you have a single powerful core, you cannot have many cores
 - A small core is much more energy and area efficient than a large core

“Large” vs. “Small” Cores

Large Core

- *Out-of-order*
- *Wide fetch e.g. 4-wide*
- *Deeper pipeline*
- *Aggressive branch predictor (e.g. hybrid)*
- *Multiple functional units*
- *Trace cache*
- *Memory dependence speculation*

Small Core

- *In-order*
- *Narrow Fetch e.g. 2-wide*
- *Shallow pipeline*
- *Simple branch predictor (e.g. Gshare)*
- *Few functional units*

**Large Cores are power inefficient:
e.g., 2x performance for 4x area (power)**

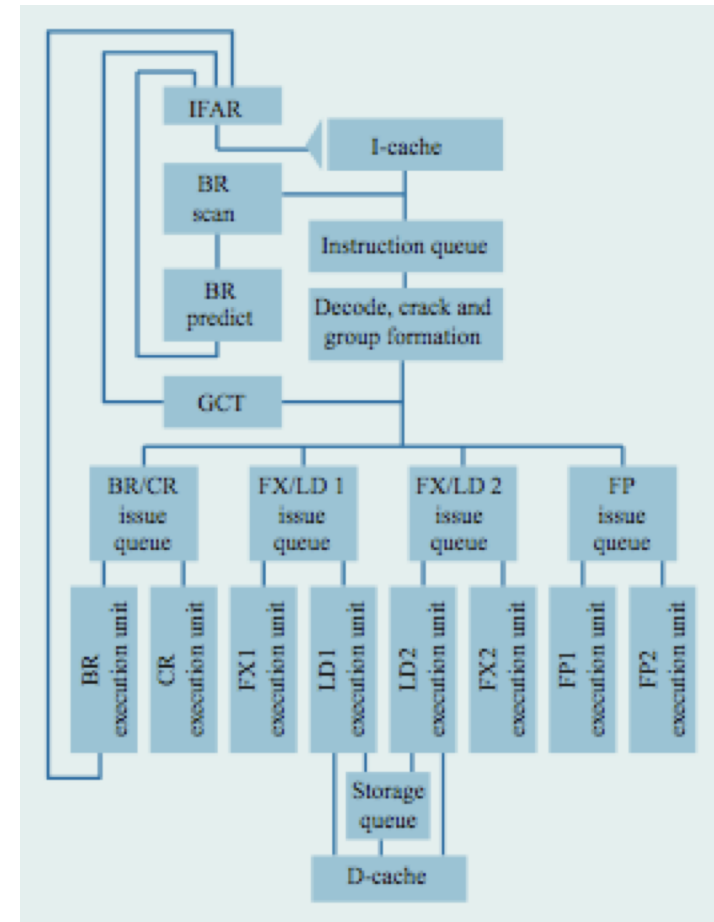
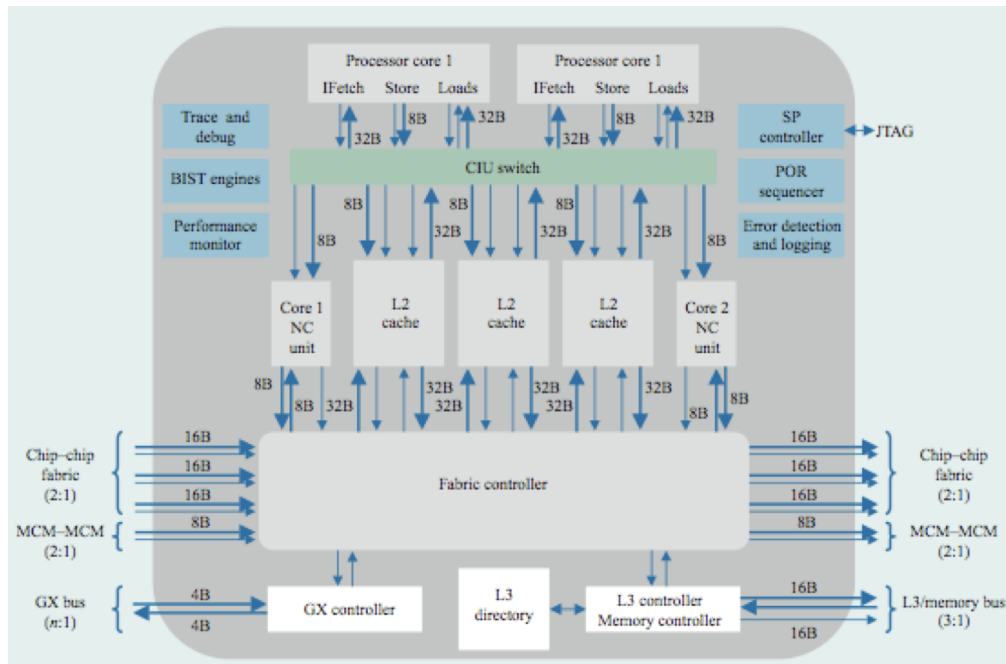
Large vs. Small Cores

- Grochowski et al., “Best of both Latency and Throughput,” ICCD 2004.

	Large core	Small core
Microarchitecture	Out-of-order, 128-256 entry ROB	In-order
Width	3-4	1
Pipeline depth	20-30	5
Normalized performance	5-8x	1x
Normalized power	20-50x	1x
Normalized energy/instruction	4-6x	1x

Meet Large: IBM POWER4

- Tendler et al., “POWER4 system microarchitecture,” IBM J R&D, 2002.
- A symmetric multi-core chip...
- Two powerful cores



IBM POWER4

- 2 cores, out-of-order execution
- 100-entry instruction window in each core
- 8-wide instruction fetch, issue, execute
- Large, local+global hybrid branch predictor
- 1.5MB, 8-way L2 cache
- Aggressive stream based prefetching

IBM POWER5

- Kalla et al., “IBM Power5 Chip: A Dual-Core Multithreaded Processor,” IEEE Micro 2004.

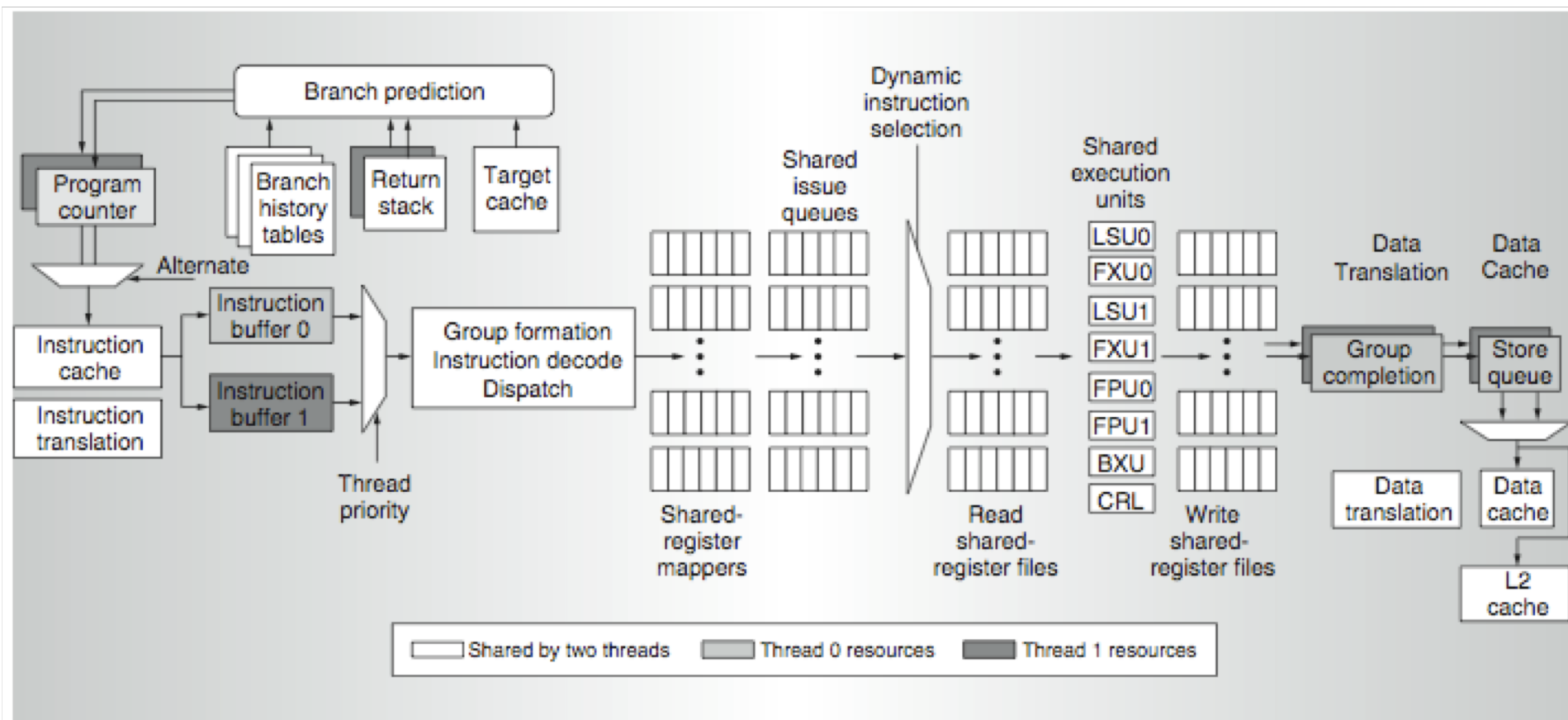
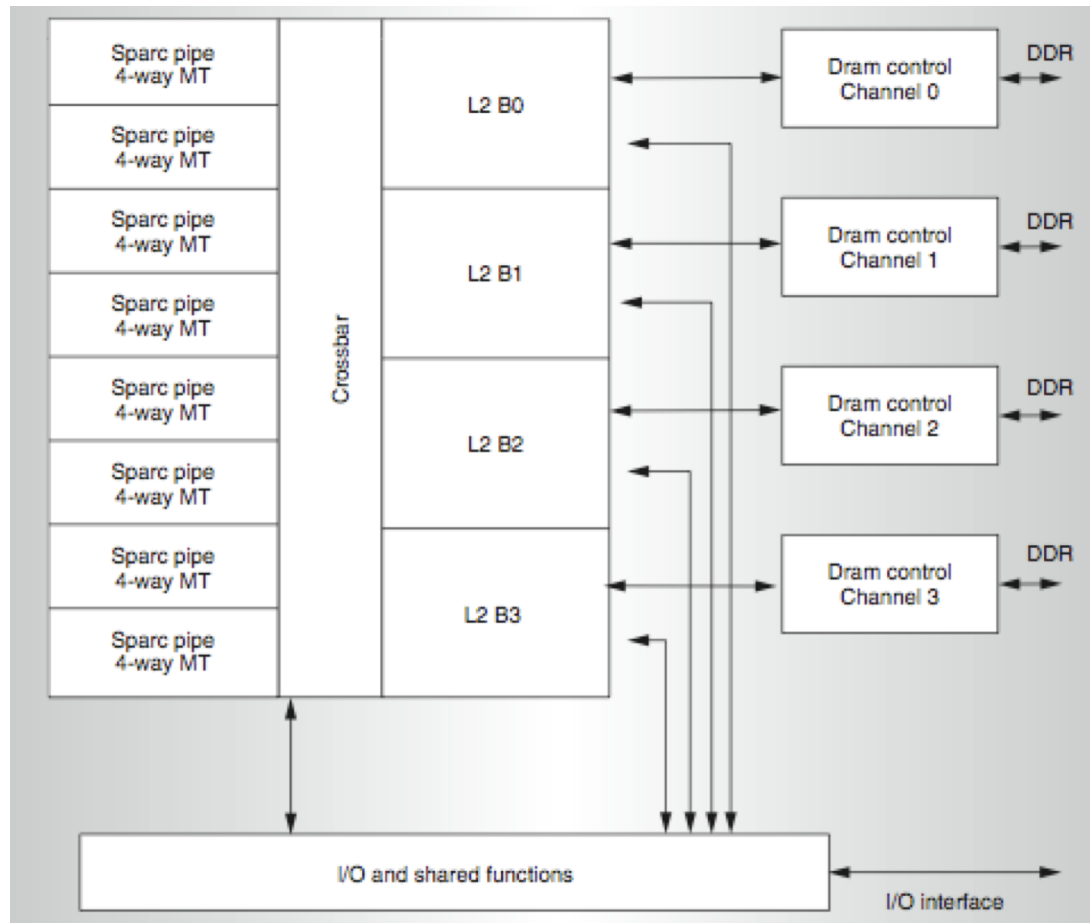


Figure 4. Power5 instruction data flow (BXU = branch execution unit and CRL = condition register logical execution unit).

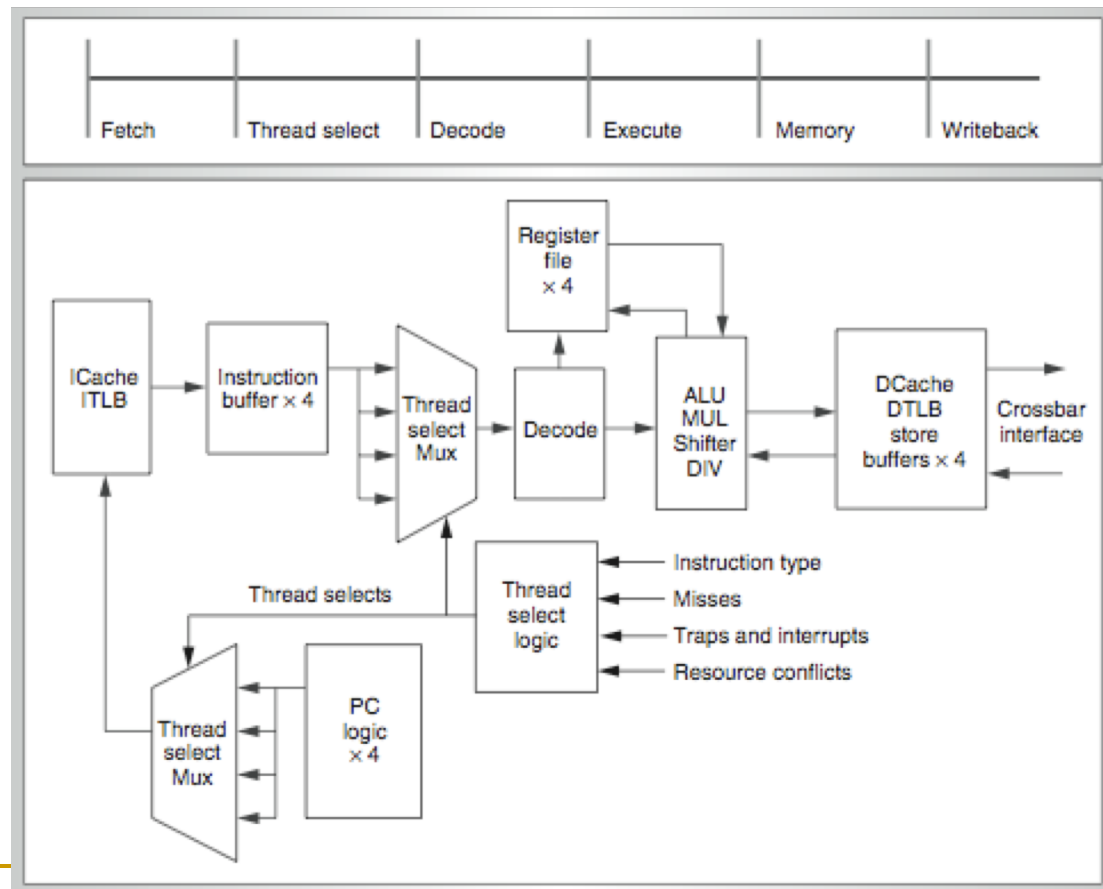
Meet Small: Sun Niagara (UltraSPARC T1)

- Kongetira et al., “Niagara: A 32-Way Multithreaded SPARC Processor,” IEEE Micro 2005.



Niagara Core

- 4-way fine-grain multithreaded, 6-stage, dual-issue in-order
- Round robin thread selection (unless cache miss)
- Shared FP unit among cores



Remember the Demands

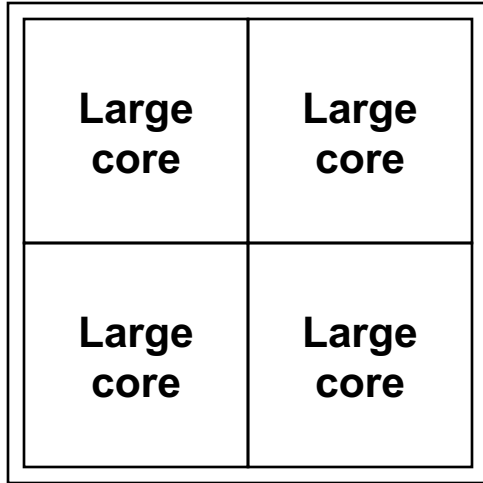
- What we want:
- In a serialized code section → one powerful “large” core
- In a parallel code section → many wimpy “small” cores
- These two conflict with each other:
 - If you have a single powerful core, you cannot have many cores
 - A small core is much more energy and area efficient than a large core
- Can we get the best of both worlds?

Performance vs. Parallelism

Assumptions:

- 1. Small cores takes an area budget of 1 and has performance of 1*
- 2. Large core takes an area budget of 4 and has performance of 2*

Tile-Large Approach



“Tile-Large”

- Tile a few large cores
- IBM Power 5, AMD Barcelona, Intel Core2Quad, Intel Nehalem
- + High performance on single thread, serial code sections (2 units)
- Low throughput on parallel program portions (8 units)

Tile-Small Approach

Small core	Small core	Small core	Small core
Small core	Small core	Small core	Small core
Small core	Small core	Small core	Small core
Small core	Small core	Small core	Small core

“Tile-Small”

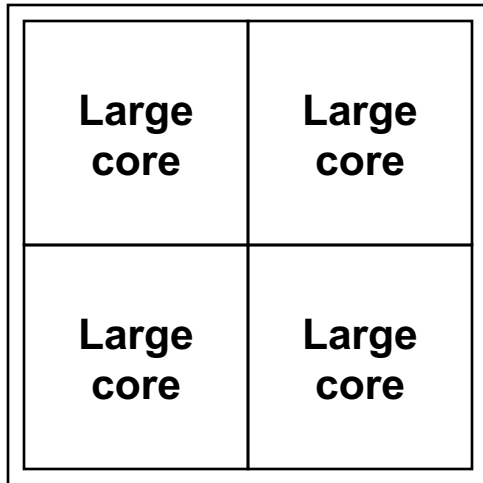
- Tile many small cores
- Sun Niagara, Intel Larrabee, Tiler TILE (tile ultra-small)
 - + High throughput on the parallel part (16 units)
 - Low performance on the serial part, single thread (1 unit)

Can we get the best of both worlds?

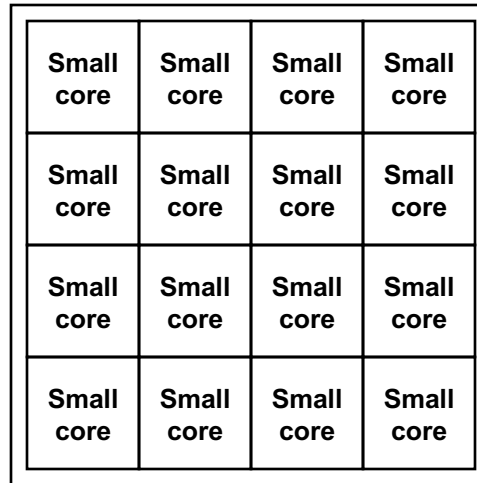
- Tile Large
 - + High performance on single thread, serial code sections (2 units)
 - Low throughput on parallel program portions (8 units)
- Tile Small
 - + High throughput on the parallel part (16 units)
 - Low performance on the serial part, single thread (1 unit),
reduced single-thread performance compared to existing single thread processors
- Idea: Have both large and small on the same chip →
Performance asymmetry

Asymmetric Multi-Core

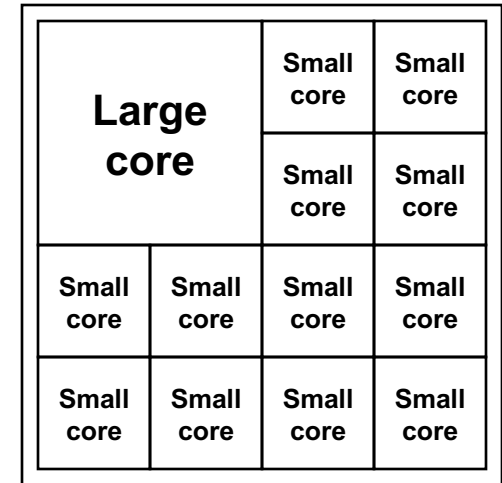
Asymmetric Chip Multiprocessor (ACMP)



“Tile-Large”



“Tile-Small”

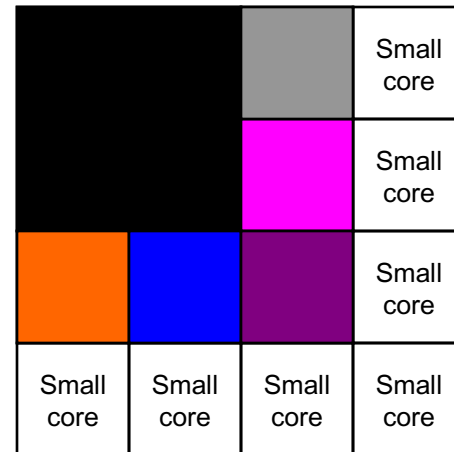
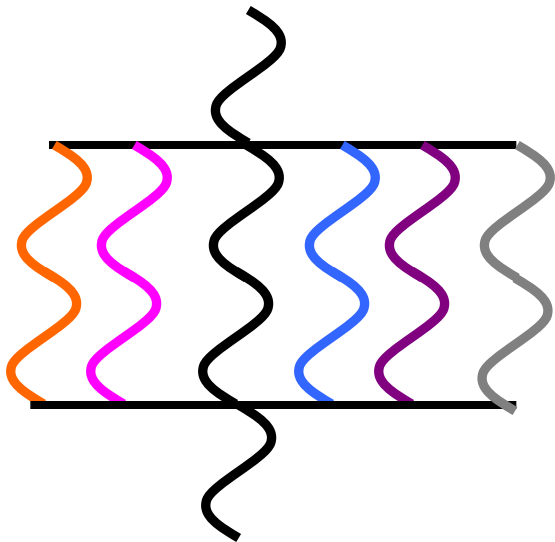


ACMP

- Provide one large core and many small cores
- + Accelerate serial part using the large core (2 units)
- + Execute parallel part on small cores and large core for high throughput (12+2 units)

Accelerating Serial Bottlenecks

Single thread \rightarrow Large core



ACMP Approach

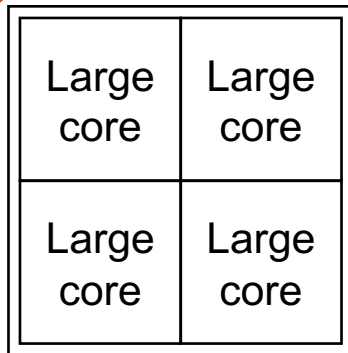
Performance vs. Parallelism

Assumptions:

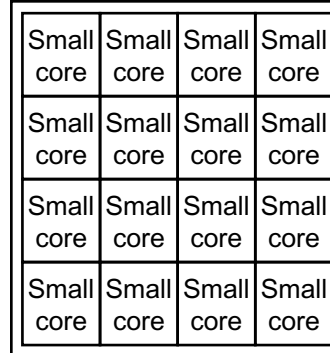
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ACMP Performance vs. Parallelism

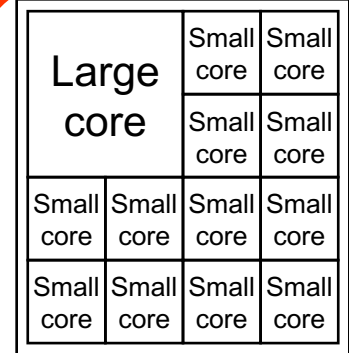
Area-budget = 16 small cores



“Tile-Large”



“Tile-Small”



ACMP

Large Cores	4	0	1
Small Cores	0	16	12
Serial Performance	2	1	2
Parallel Throughput	$2 \times 4 = 8$	$1 \times 16 = 16$	$1 \times 2 + 1 \times 12 = 14$

Amdahl's Law Modified

- Simplified Amdahl's Law for an Asymmetric Multiprocessor
- Assumptions:
 - Serial portion executed on the large core
 - Parallel portion executed on both small cores and large cores
 - f : Parallelizable fraction of a program
 - L : Number of large processors
 - S : Number of small processors
 - X : Speedup of a large processor over a small one

$$\text{Speedup} = \frac{1}{\frac{1-f}{X} + \frac{f}{S + X*L}}$$

Caveats of Parallelism, Revisited

■ Amdahl's Law

- f : Parallelizable fraction of a program
- N : Number of processors

$$\text{Speedup} = \frac{1}{1 - f + \frac{f}{N}}$$

- Amdahl, “Validity of the single processor approach to achieving large scale computing capabilities,” AFIPS 1967.
- **Maximum speedup limited by serial portion: Serial bottleneck**
- **Parallel portion is usually not perfectly parallel**
 - **Synchronization** overhead (e.g., updates to shared data)
 - **Load imbalance** overhead (imperfect parallelization)
 - **Resource sharing** overhead (contention among N processors)

Accelerating Parallel Bottlenecks

- Serialized or imbalanced execution in the parallel portion can also benefit from a large core
- Examples:
 - Critical sections that are contended
 - Parallel stages that take longer than others to execute
- Idea: Dynamically identify these code portions that cause serialization and execute them on a large core

Accelerated Critical Sections

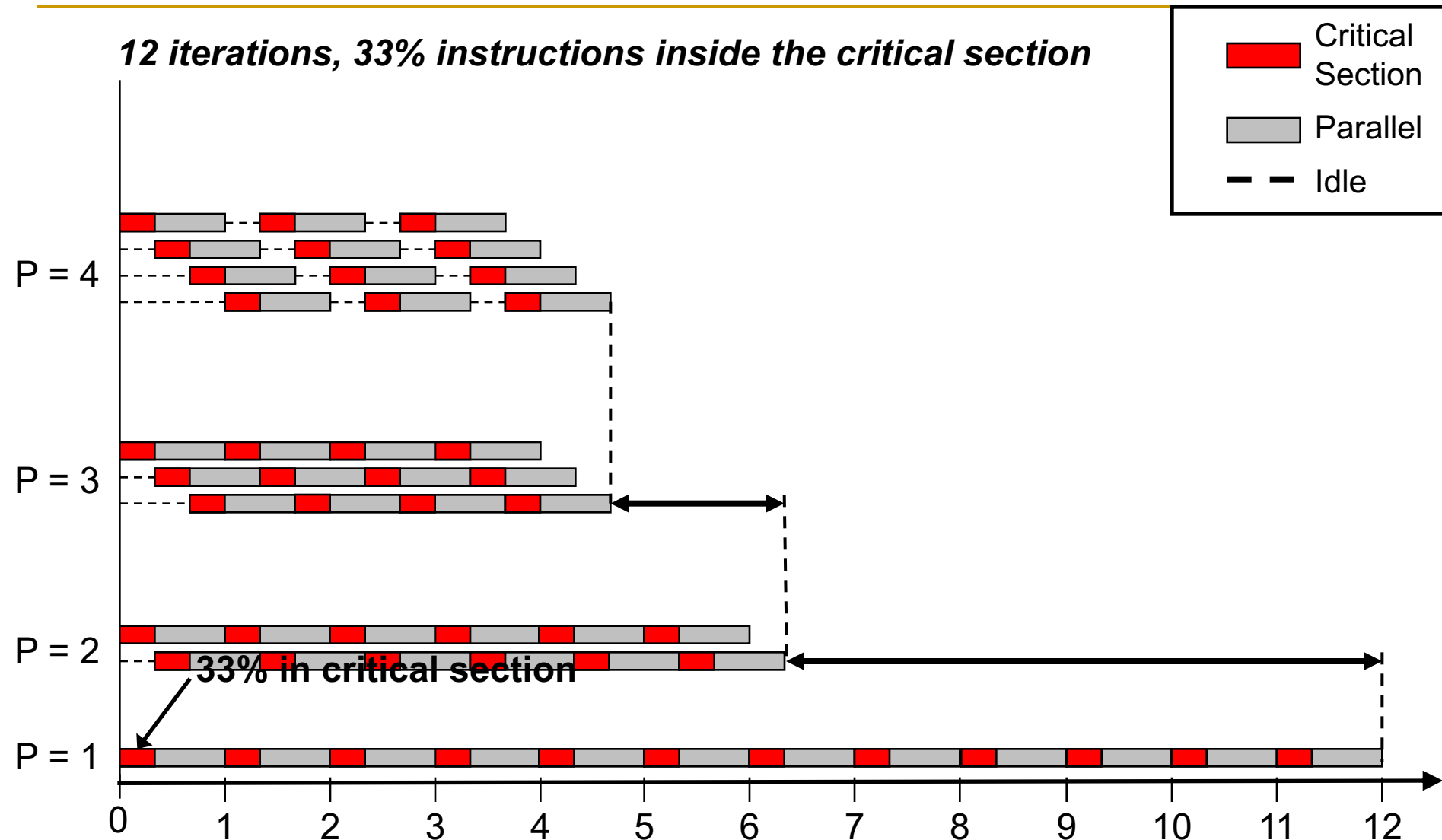
M. Aater Suleman, Onur Mutlu, Moinuddin K. Qureshi, and Yale N. Patt,

"Accelerating Critical Section Execution with Asymmetric Multi-Core Architectures"

*Proceedings of the 14th International Conference on Architectural Support for Programming Languages and Operating Systems (**ASPLOS**), 2009*

Contention for Critical Sections

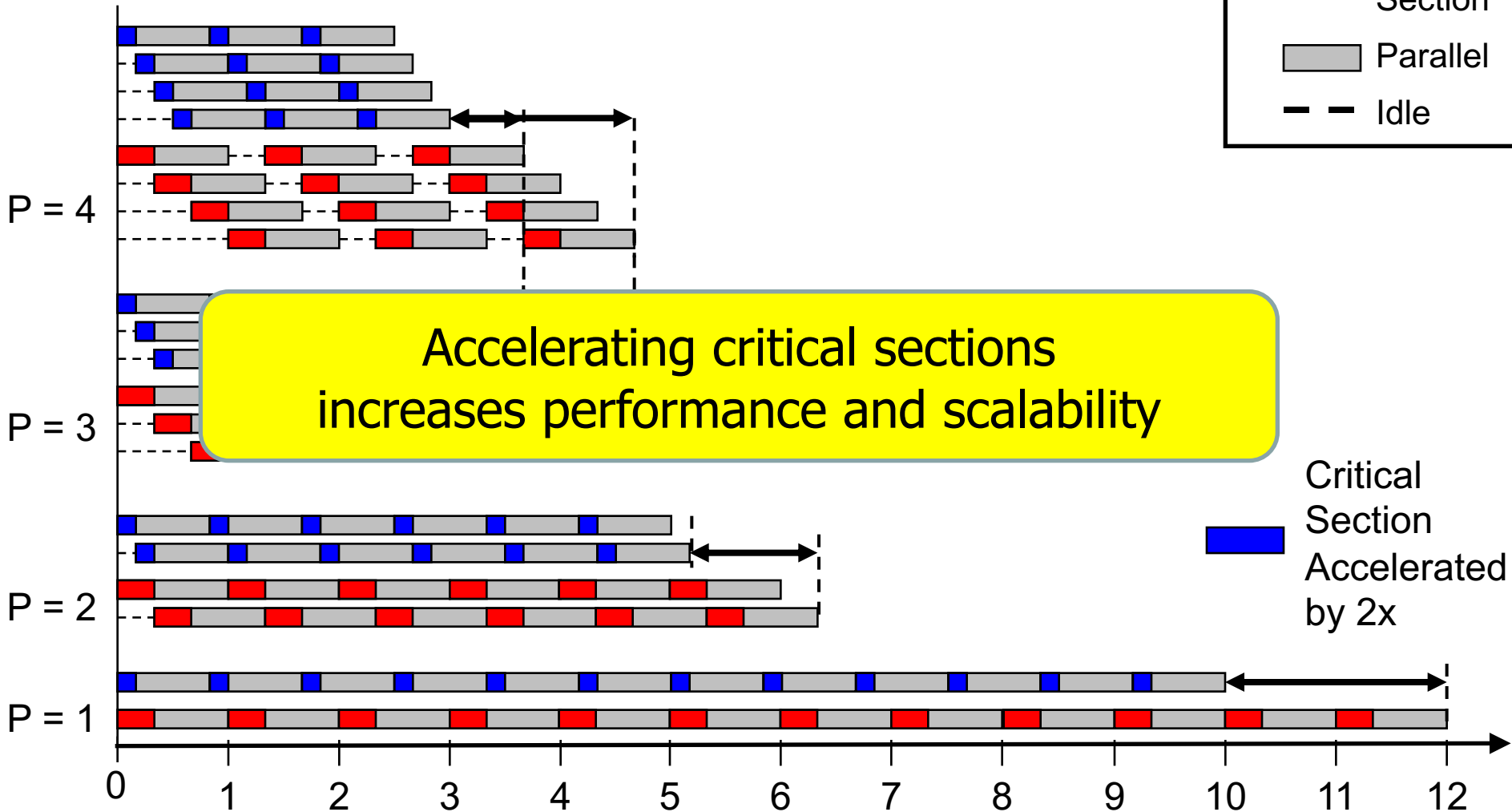
12 iterations, 33% instructions inside the critical section



Contention for Critical Sections

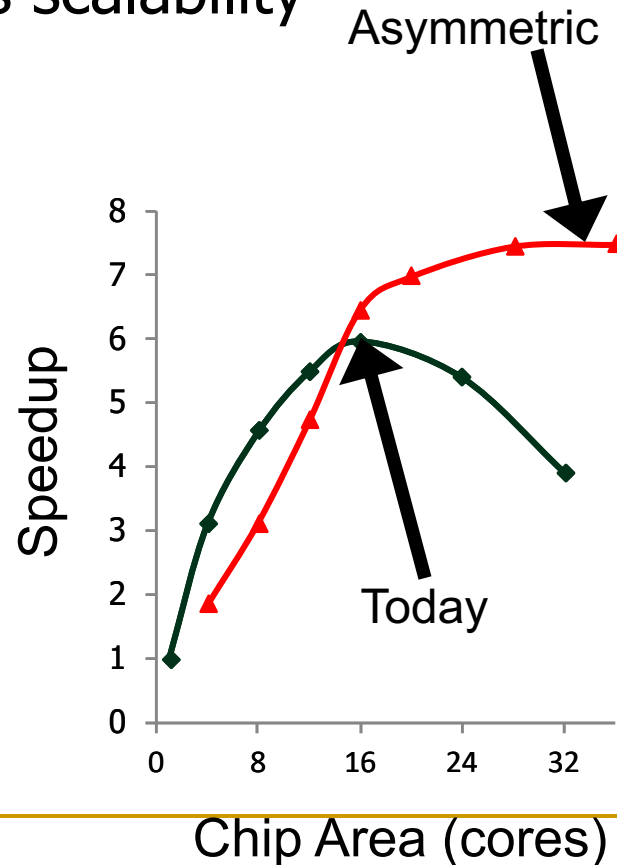
12 iterations, 33% instructions inside the critical section

Critical Section
Parallel
Idle



Impact of Critical Sections on Scalability

- Contention for critical sections leads to serial execution (serialization) of threads in the parallel program portion
- Contention for critical sections increases with the number of threads and limits scalability



MySQL (oltp-1)

A Case for Asymmetry

- Execution time of sequential kernels, critical sections, and limiter stages must be short
 - It is difficult for the programmer to shorten these serialized sections
 - Insufficient domain-specific knowledge
 - Variation in hardware platforms
 - Limited resources
 - Performance-debugging tradeoff
 - Goal: A mechanism to shorten serial bottlenecks without requiring programmer effort
 - Idea: Accelerate serialized code sections by shipping them to powerful cores in an asymmetric multi-core (ACMP)
-

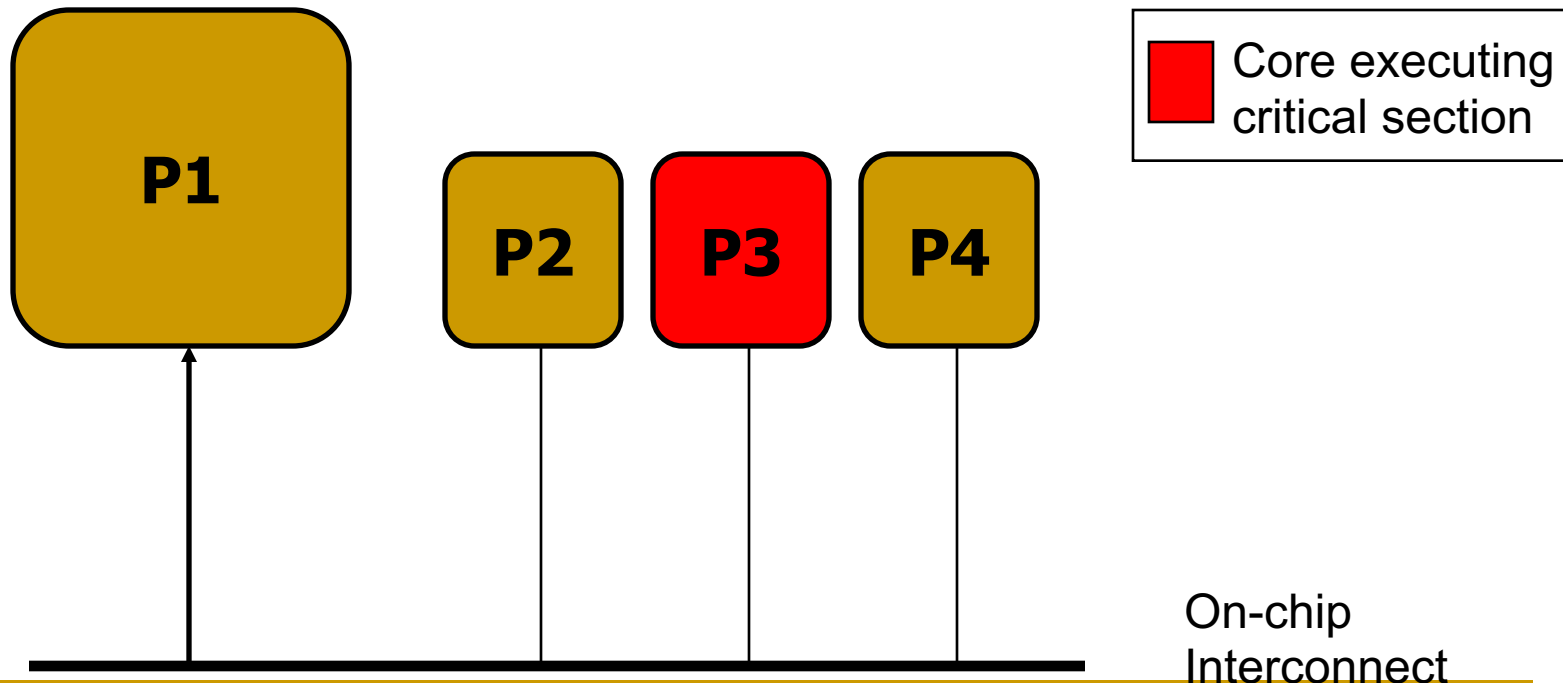
An Example: Accelerated Critical Sections

- Idea: HW/SW ships critical sections to a large, powerful core in an asymmetric multi-core architecture
- Benefit:
 - Reduces serialization due to contended locks
 - Reduces the performance impact of hard-to-parallelize sections
 - Programmer does not need to (heavily) optimize parallel code → fewer bugs, improved productivity
- Suleman et al., “Accelerating Critical Section Execution with Asymmetric Multi-Core Architectures,” ASPLOS 2009, IEEE Micro Top Picks 2010.
- Suleman et al., “Data Marshaling for Multi-Core Architectures,” ISCA 2010, IEEE Micro Top Picks 2011.

Conventional ACMP

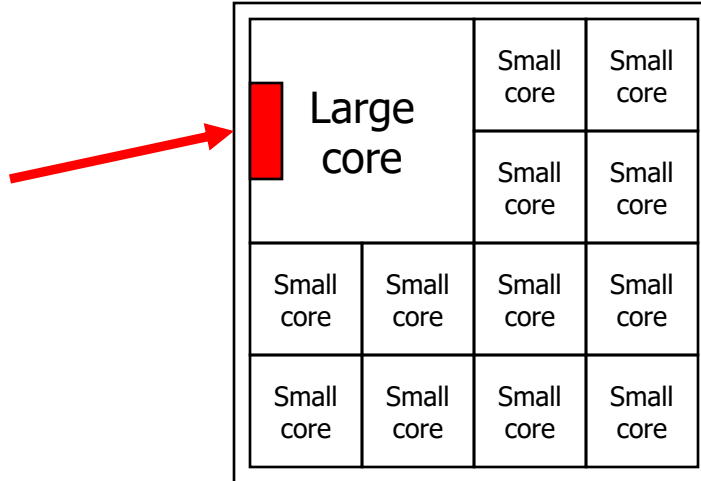
```
EnterCS()  
    PriorityQ.insert(...)  
LeaveCS()
```

1. **P2 encounters a Critical Section**
2. **Sends a request for the lock**
3. **Acquires the lock**
4. **Executes Critical Section**
5. **Releases the lock**



Accelerated Critical Sections (ACS)

**Critical Section
Request Buffer
(CSRB)**



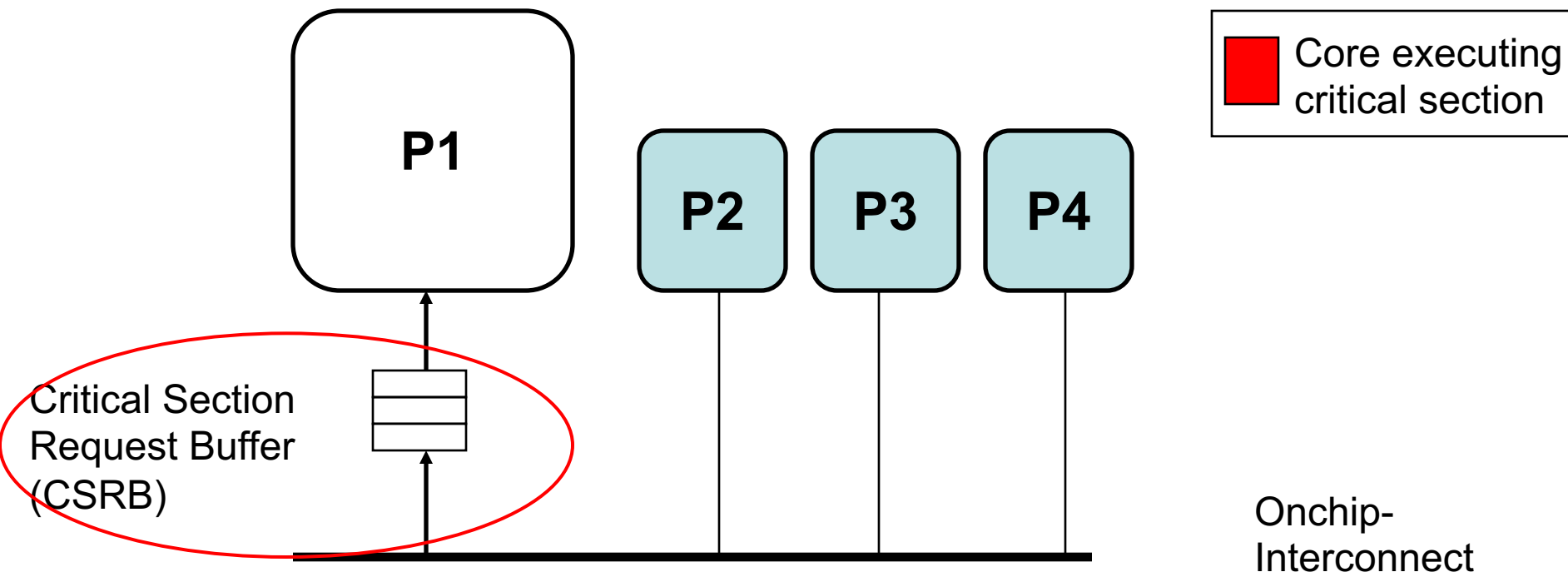
ACMP

- Accelerate Amdahl's serial part **and critical sections** using the large core
 - Suleman et al., "Accelerating Critical Section Execution with Asymmetric Multi-Core Architectures," ASPLOS 2009, IEEE Micro Top Picks 2010.

Accelerated Critical Sections

EnterCS()
 PriorityQ.insert(...)
LeaveCS()

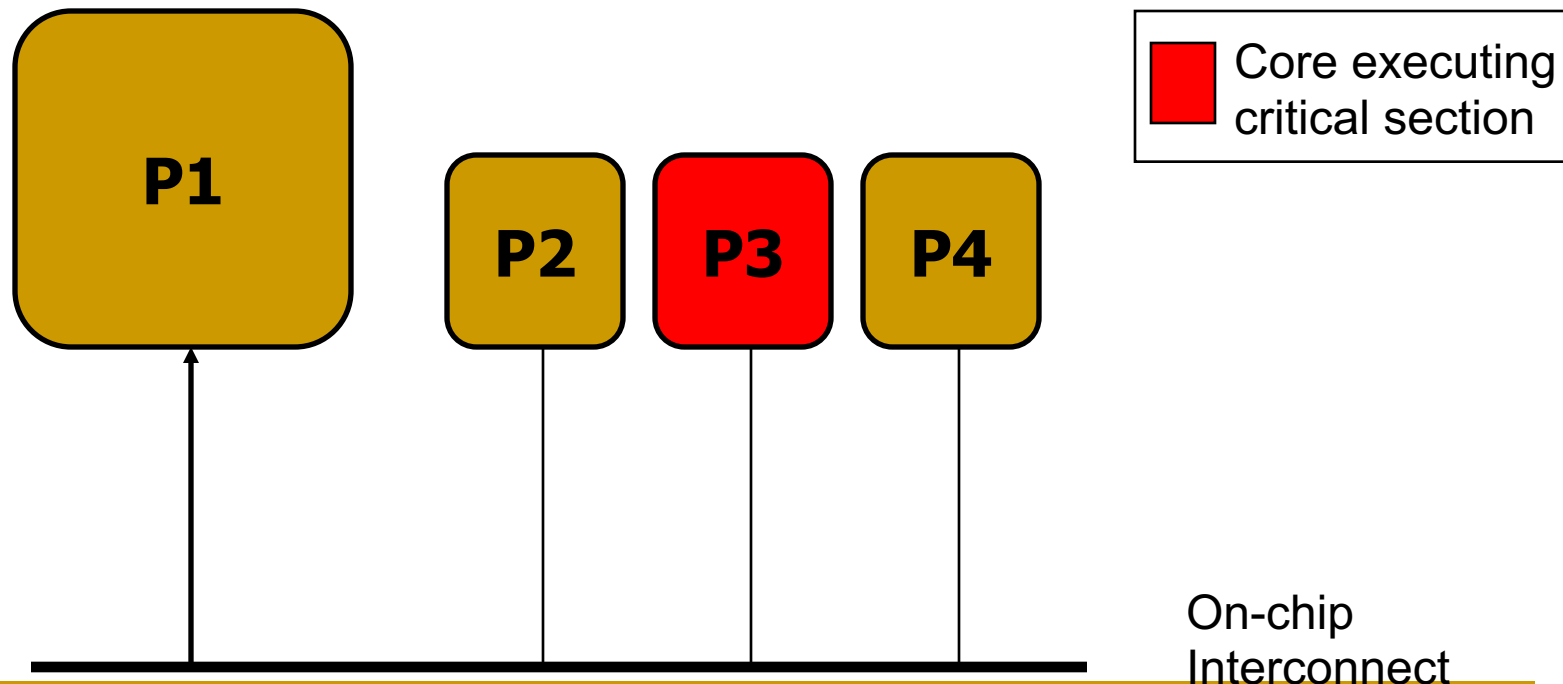
1. P2 encounters a critical section (CSCALL)
2. P2 sends CSCALL Request to CSRB
3. P1 executes Critical Section
4. P1 sends CSDONE signal



Conventional ACMP

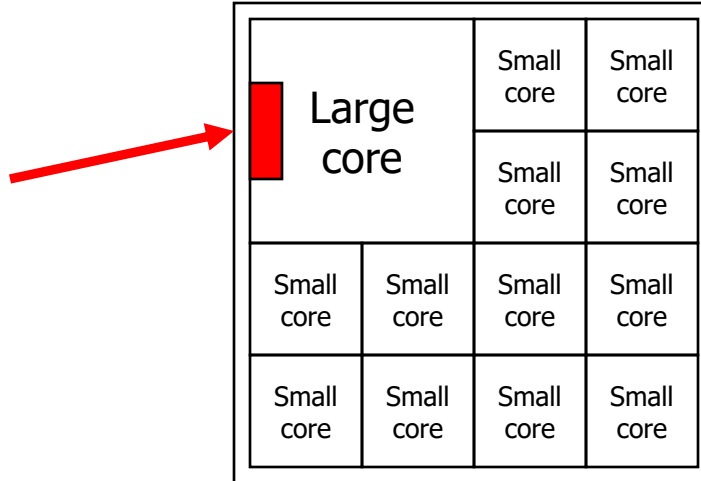
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EnterCS()  
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Accelerated Critical Sections (ACS)

**Critical Section
Request Buffer
(CSRB)**



ACMP

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 - Suleman et al., "Accelerating Critical Section Execution with Asymmetric Multi-Core Architectures," ASPLOS 2009, IEEE Micro Top Picks 2010.

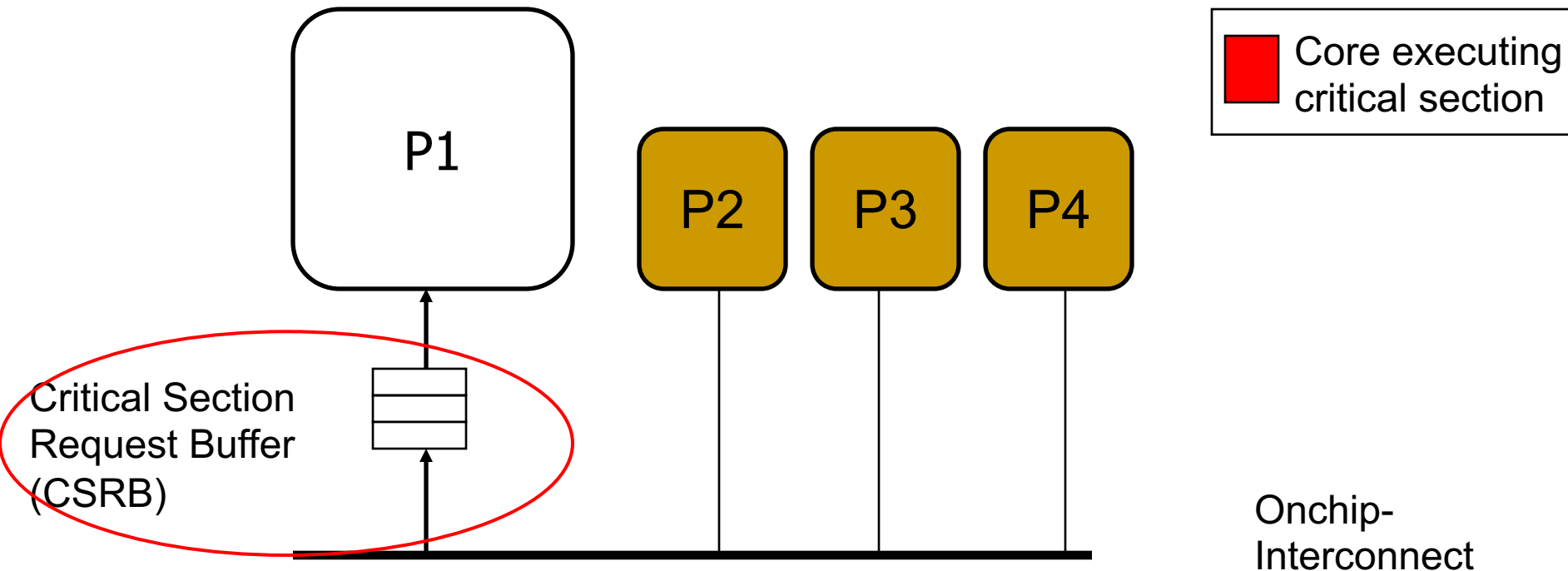
Accelerated Critical Sections (ACS)

EnterCS()

PriorityQ.insert(...)

LeaveCS()

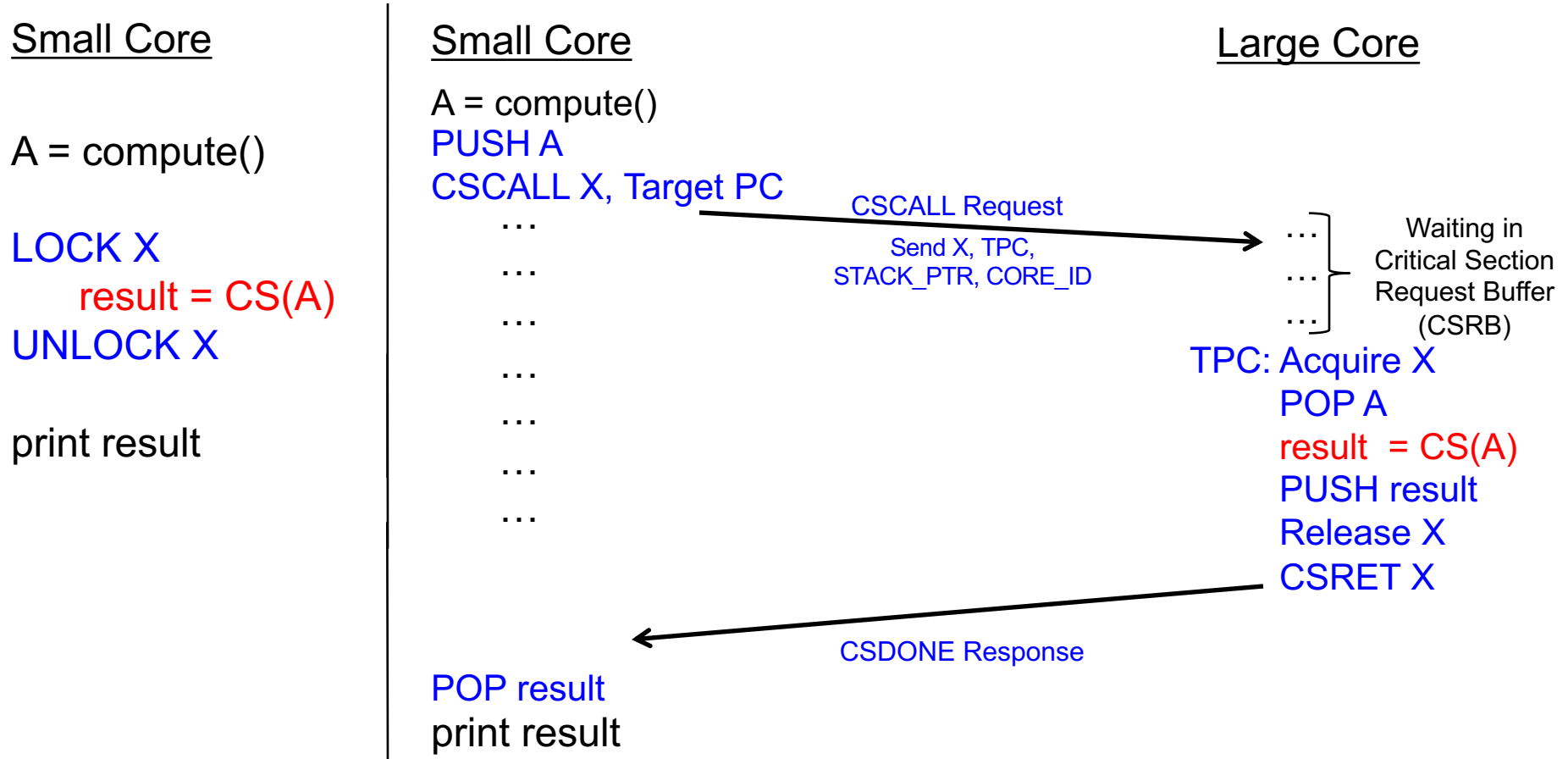
1. P2 encounters a critical section (CSCALL)
2. P2 sends CSCALL Request to CSRB
3. P1 executes Critical Section
4. P1 sends CSDONE signal



ACS Architecture Overview

- ISA extensions
 - `CSCALL LOCK_ADDR, TARGET_PC`
 - `CSRET LOCK_ADDR`
- Compiler/Library inserts `CSCALL/CSRET`
- On a `CSCALL`, the small core:
 - Sends a `CSCALL` request to the large core
 - Arguments: Lock address, Target PC, Stack Pointer, Core ID
 - Stalls and waits for `CSDONE`
- Large Core
 - Critical Section Request Buffer (CSRB)
 - Executes the critical section and sends `CSDONE` to the requesting core

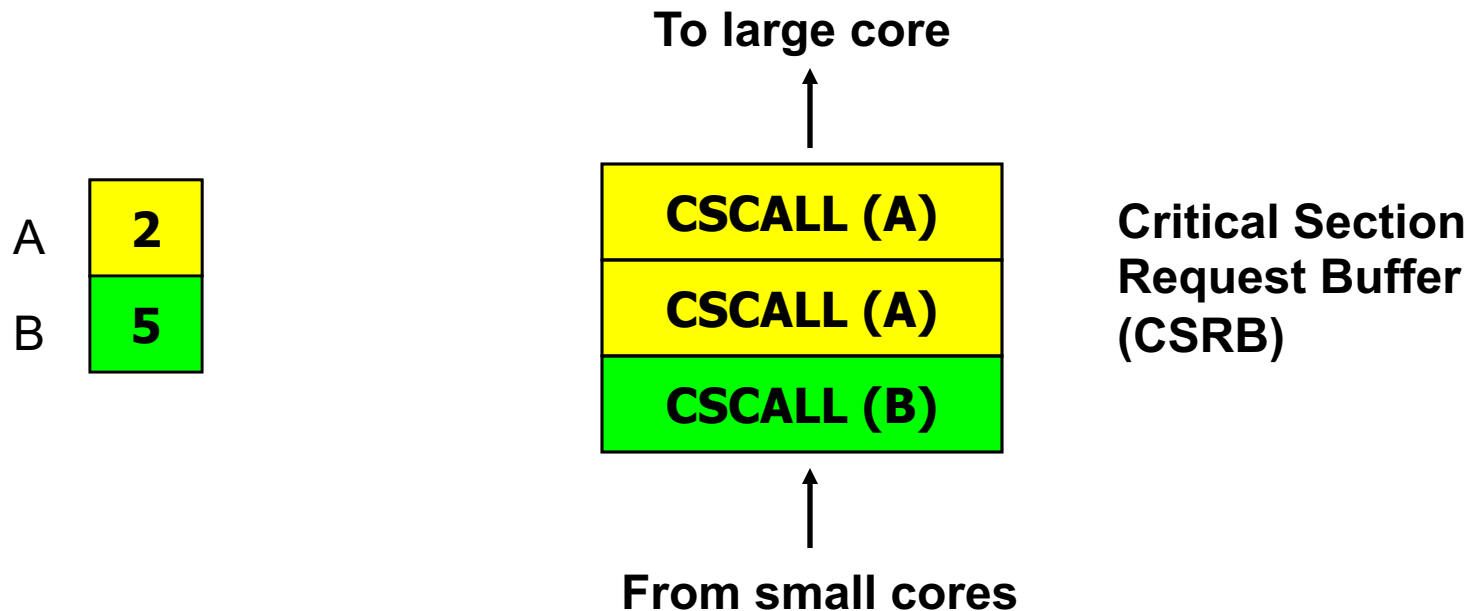
Accelerated Critical Sections (ACS)



- Suleman et al., “Accelerating Critical Section Execution with Asymmetric Multi-Core Architectures,” ASPLOS 2009.

False Serialization

- ACS can serialize independent critical sections
- Selective Acceleration of Critical Sections (SEL)
 - Saturating counters to track false serialization



ACS Performance Tradeoffs

■ Pluses

- + Faster critical section execution
- + Shared locks stay in one place: better lock locality
- + Shared data stays in large core's (large) caches: better shared data locality, less ping-ponging

■ Minuses

- Large core dedicated for critical sections: reduced parallel throughput
- CSCALL and CSDONE control transfer overhead
- Thread-private data needs to be transferred to large core: worse private data locality

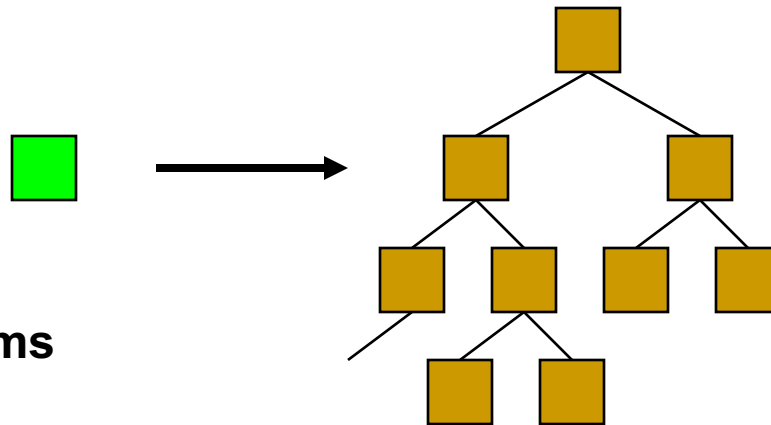
ACS Performance Tradeoffs

- ***Fewer parallel threads vs. accelerated critical sections***
 - Accelerating critical sections offsets loss in throughput
 - As the number of cores (threads) on chip increase:
 - Fractional loss in parallel performance decreases
 - Increased contention for critical sections makes acceleration more beneficial
- ***Overhead of CSCALL/CSDONE vs. better lock locality***
 - ACS avoids “ping-ponging” of locks among caches by keeping them at the large core
- ***More cache misses for private data vs. fewer misses for shared data***

Cache Misses for Private Data

PriorityHeap.insert(NewSubProblems)

Private Data:
NewSubProblems



Shared Data:
The priority heap

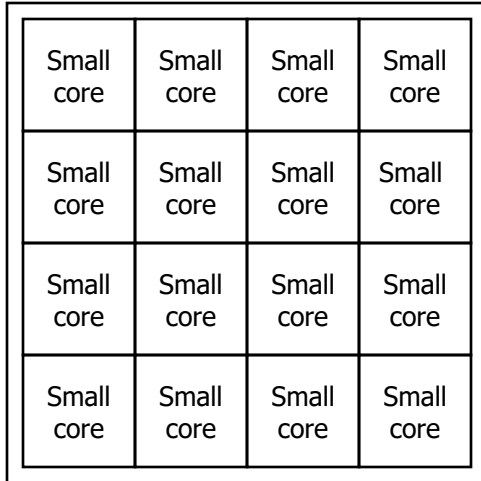
Puzzle Benchmark

ACS Performance Tradeoffs

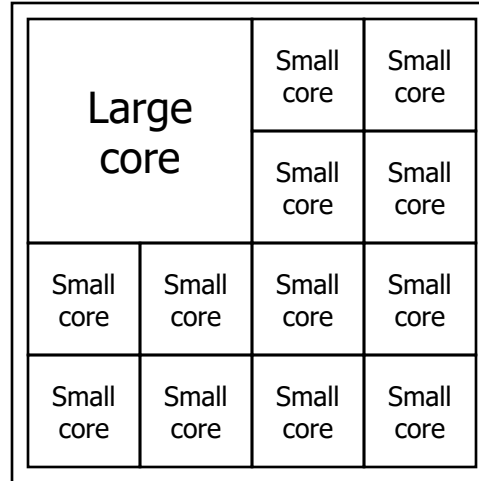
- ***Fewer parallel threads vs. accelerated critical sections***
 - Accelerating critical sections offsets loss in throughput
 - As the number of cores (threads) on chip increase:
 - Fractional loss in parallel performance decreases
 - Increased contention for critical sections makes acceleration more beneficial
- ***Overhead of CSCALL/CSDONE vs. better lock locality***
 - ACS avoids “ping-ponging” of locks among caches by keeping them at the large core
- ***More cache misses for private data vs. fewer misses for shared data***
 - Cache misses reduce if shared data > private data

This problem can be solved

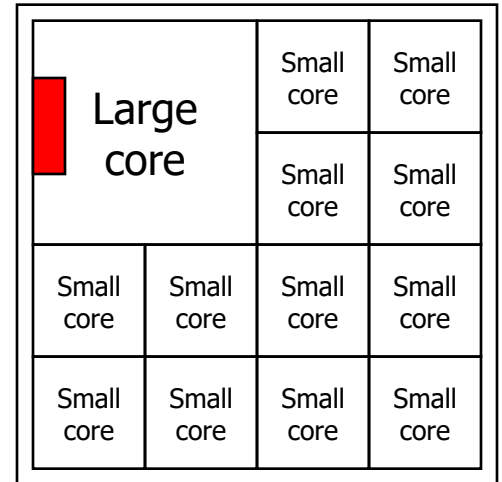
ACS Comparison Points



SCMP



ACMP



ACS

- Conventional locking
 - Large core executes Amdahl's serial part
- Conventional locking
 - Large core executes Amdahl's serial part
- Large core executes Amdahl's serial part and critical sections

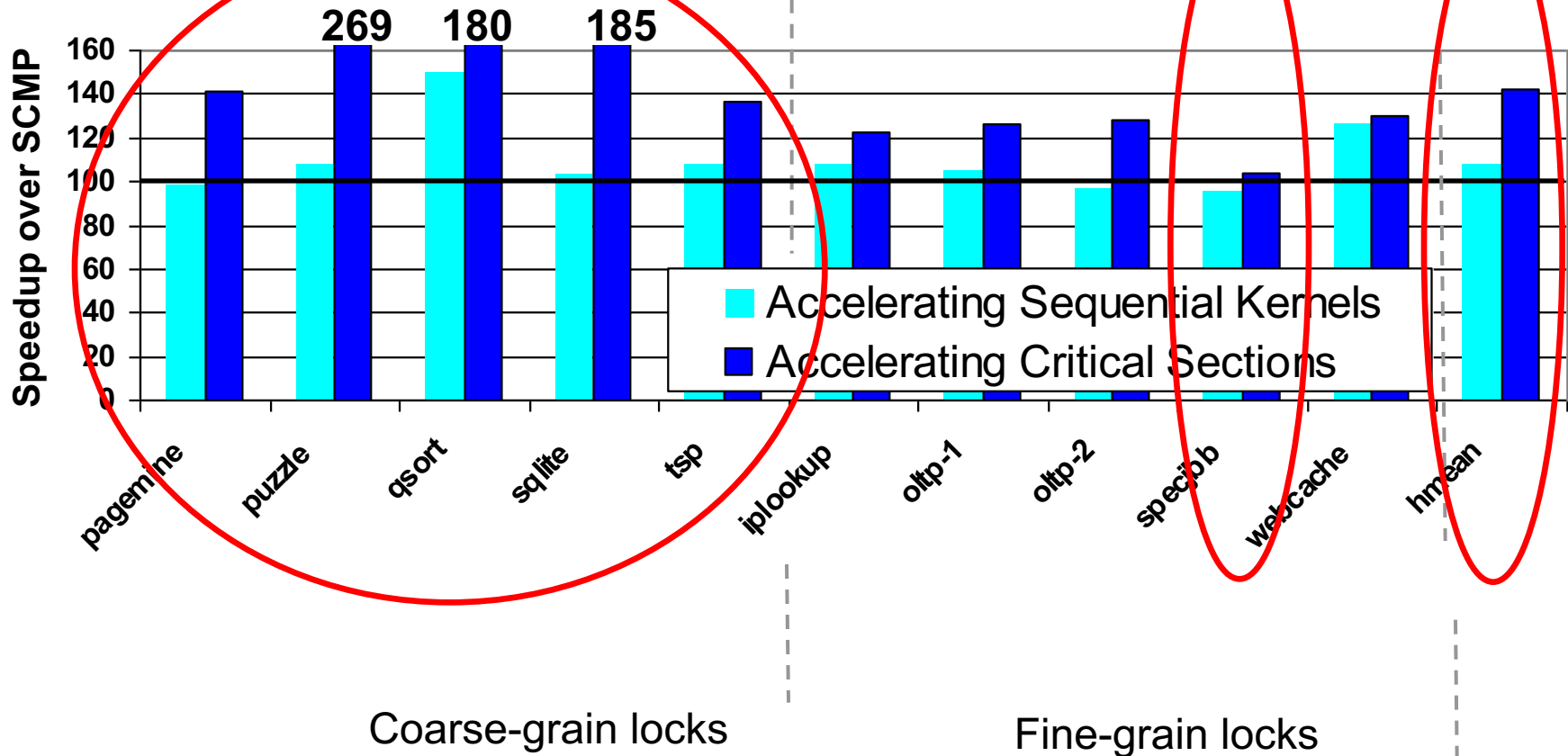
Accelerated Critical Sections: Methodology

- Workloads: 12 critical section intensive applications
 - Data mining kernels, sorting, database, web, networking
- Multi-core x86 simulator
 - 1 large and 28 small cores
 - Aggressive stream prefetcher employed at each core
- Details:
 - Large core: 2GHz, out-of-order, 128-entry ROB, 4-wide, 12-stage
 - Small core: 2GHz, in-order, 2-wide, 5-stage
 - Private 32 KB L1, private 256KB L2, 8MB shared L3
 - On-chip interconnect: Bi-directional ring, 5-cycle hop latency

ACS Performance

Chip Area = 32 small cores
SCMP = 32 small cores
ACMP = 1 large and 28 small cores

Equal-area comparison
Number of threads = *Best threads*

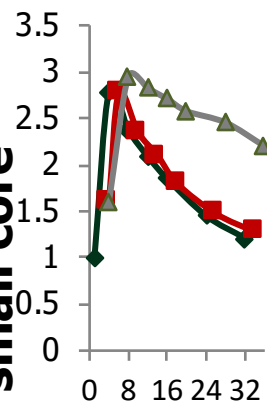


Equal-Area Comparisons

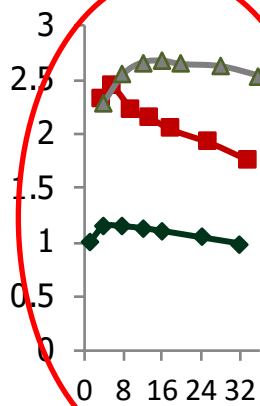
----- **SCMP**
----- **ACMP**
----- **ACS**

Number of threads = No. of cores

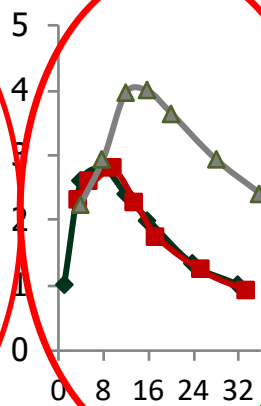
Speedup over a small core



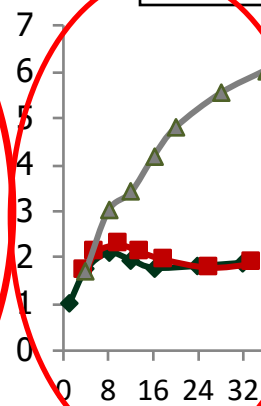
(a) ep



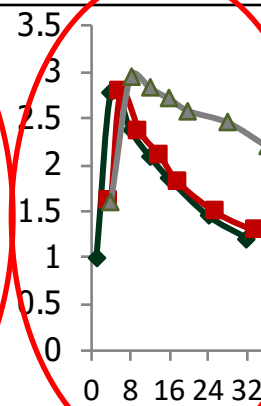
(b) is



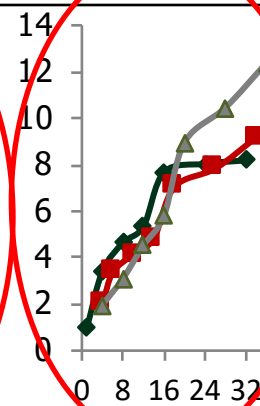
(c) pagemine



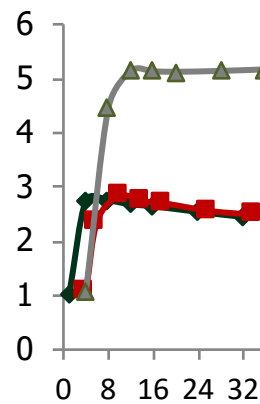
(d) puzzle



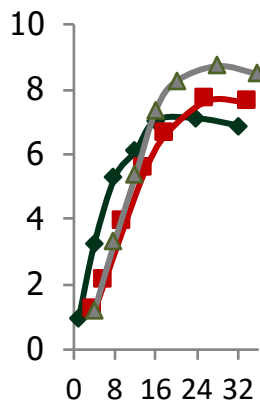
(e) qsort



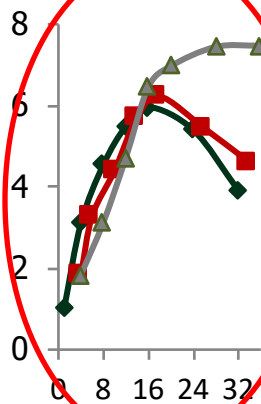
(f) tsp



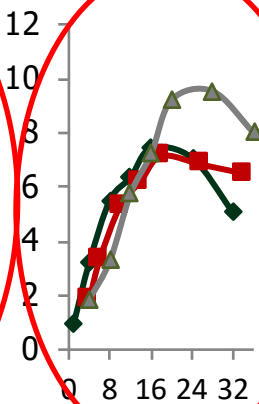
(g) sqlite



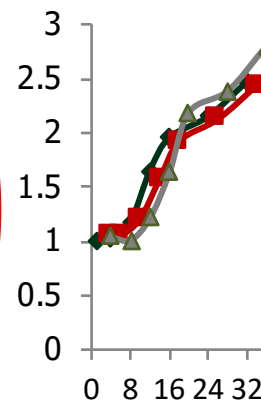
(h) iplookup



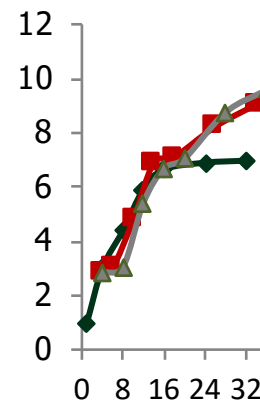
(i) oltp-1



(j) oltp-2



(k) specjbb



(l) webcache

Chip Area (small cores)

ACS Summary

- Critical sections reduce performance and limit scalability
 - Accelerate critical sections by executing them on a powerful core
 - ACS reduces average execution time by:
 - 34% compared to an equal-area SCMP
 - 23% compared to an equal-area ACMP
 - ACS improves scalability of 7 of the 12 workloads
 - Generalizing the idea: Accelerate all bottlenecks (“critical paths”) by executing them on a powerful core
-

More on Accelerated Critical Sections

- M. Aater Suleman, Onur Mutlu, Moinuddin K. Qureshi, and Yale N. Patt, **"Accelerating Critical Section Execution with Asymmetric Multi-Core Architectures"**

Proceedings of the 14th International Conference on Architectural Support for Programming Languages and Operating Systems (ASPLOS), pages 253-264, Washington, DC, March 2009. [Slides \(ppt\)](#)

Accelerating Critical Section Execution with Asymmetric Multi-Core Architectures

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Bottleneck Identification and Scheduling

Jose A. Joao, M. Aater Suleman, Onur Mutlu, and Yale N. Patt,

"Bottleneck Identification and Scheduling in Multithreaded Applications"

Proceedings of the 17th International Conference on Architectural Support for Programming Languages and Operating Systems (ASPLOS), London, UK, March 2012.

BIS Summary

- **Problem:** Performance and scalability of multithreaded applications are limited by serializing bottlenecks
 - ❑ different types: critical sections, barriers, slow pipeline stages
 - ❑ importance (criticality) of a bottleneck can change over time
- **Our Goal:** Dynamically identify the most important bottlenecks and accelerate them
 - ❑ How to identify the most critical bottlenecks
 - ❑ How to efficiently accelerate them
- **Solution:** Bottleneck Identification and Scheduling (BIS)
 - ❑ Software: annotate bottlenecks (BottleneckCall, BottleneckReturn) and implement waiting for bottlenecks with a special instruction (BottleneckWait)
 - ❑ Hardware: identify bottlenecks that cause the most thread waiting and accelerate those bottlenecks on large cores of an asymmetric multi-core system
- Improves multithreaded application performance and scalability, outperforms previous work, and performance improves with more cores

Bottlenecks in Multithreaded Applications

Definition: any code segment for which threads contend (i.e. wait)

Examples:

- **Amdahl's serial portions**
 - Only one thread exists → on the critical path
- **Critical sections**
 - Ensure mutual exclusion → likely to be on the critical path if contended
- **Barriers**
 - Ensure all threads reach a point before continuing → the latest thread arriving is on the critical path
- **Pipeline stages**
 - Different stages of a loop iteration may execute on different threads, slowest stage makes other stages wait → on the critical path

Observation: Limiting Bottlenecks Change Over Time

A=full linked list; B=empty linked list

repeat

Lock A

Traverse list A

Remove X from A

Unlock A

Compute on X

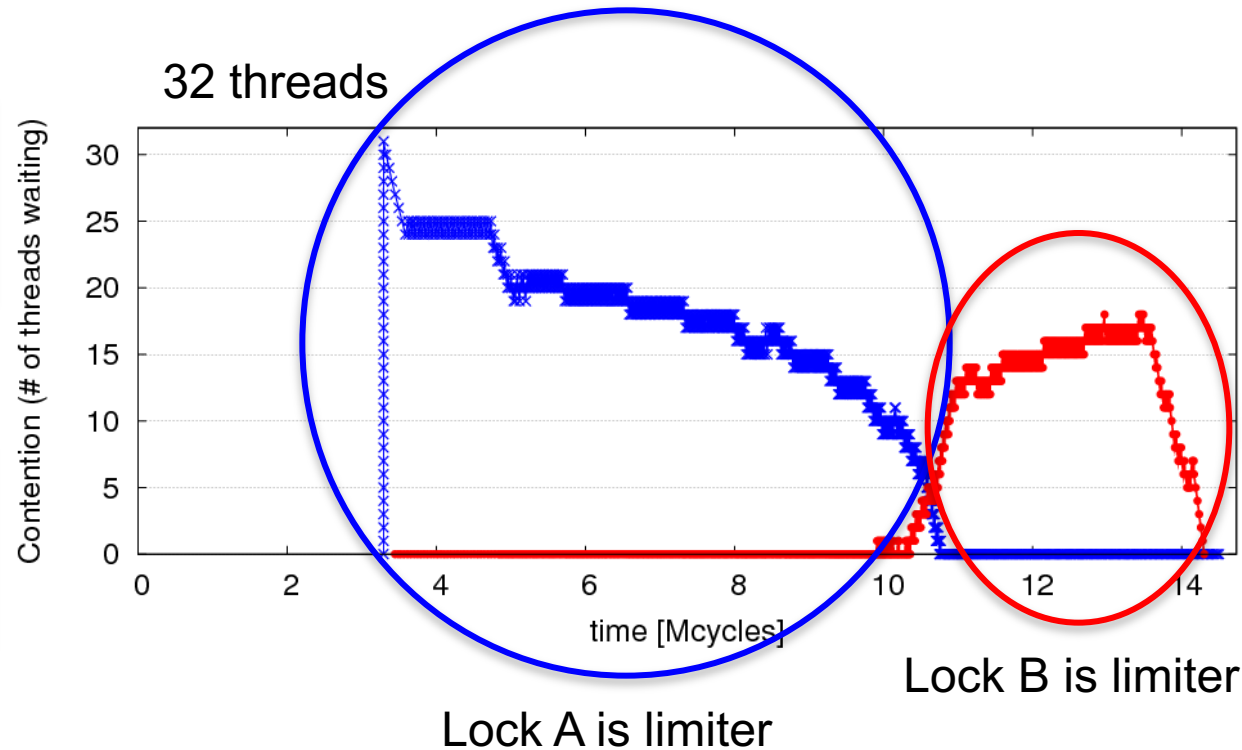
Lock B

Traverse list B

Insert X into B

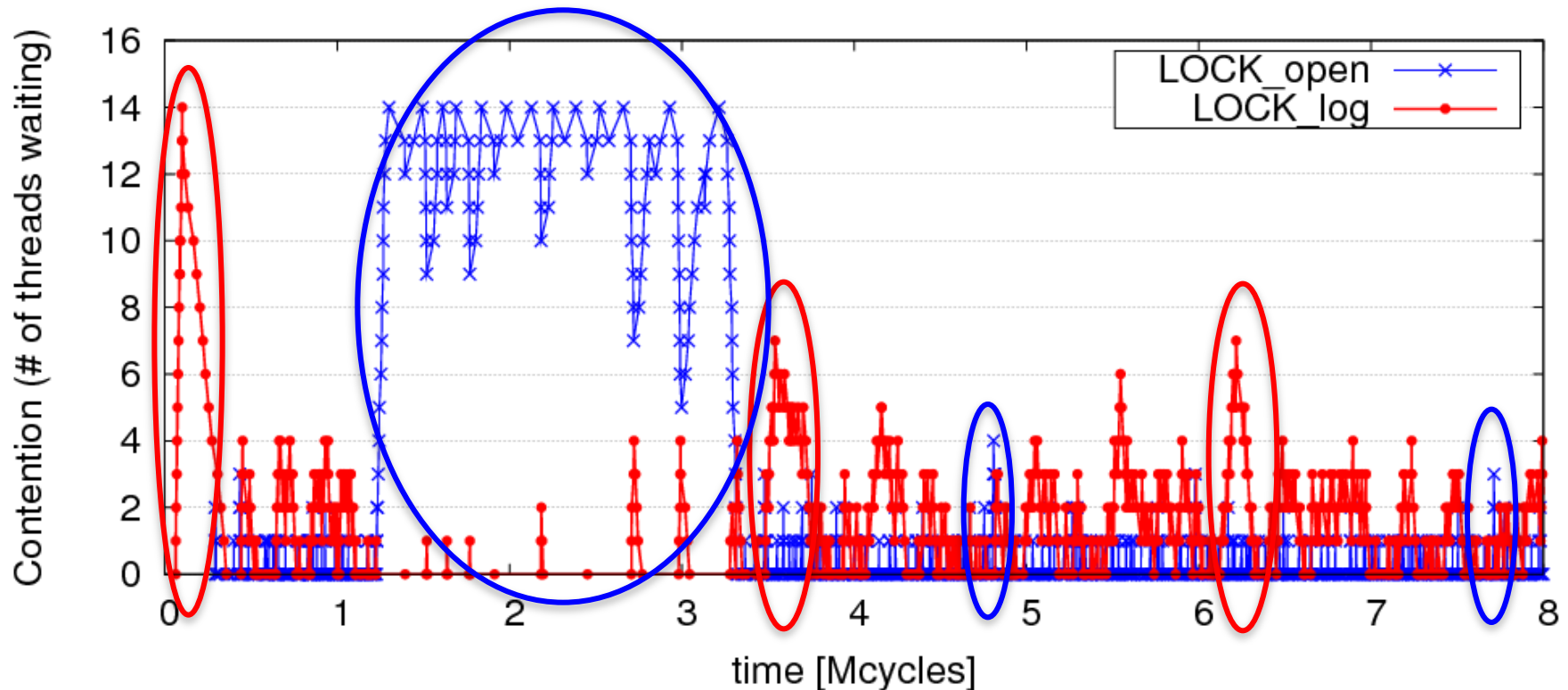
Unlock B

until A is empty



Limiting Bottlenecks Do Change on Real Applications

MySQL running Sysbench queries, 16 threads



Previous Work on Bottleneck Acceleration

- Asymmetric CMP (ACMP) proposals [Annavaram+, ISCA'05]
[Morad+, Comp. Arch. Letters'06] [Suleman+, Tech. Report'07]
 - Accelerate **only the Amdahl's bottleneck**
- Accelerated Critical Sections (ACS) [Suleman+, ASPLOS'09]
 - Accelerate **only critical sections**
 - **Does not take into account importance** of critical sections
- Feedback-Directed Pipelining (FDP) [Suleman+, PACT'10 and PhD thesis'11]
 - Accelerate **only stages with lowest throughput**
 - **Slow to adapt** to phase changes (software based library)

No previous work can accelerate all three types of bottlenecks or quickly adapts to fine-grain changes in the *importance* of bottlenecks

Our goal: general mechanism to identify performance-limiting bottlenecks of any type and accelerate them on an ACMP

Bottleneck Identification and Scheduling (BIS)

- Key insight:
 - Thread waiting reduces parallelism and is likely to reduce performance
 - Code causing the most thread waiting
→ likely critical path

- Key idea:
 - Dynamically identify bottlenecks that cause the most thread waiting
 - Accelerate them (using powerful cores in an ACMP)

Bottleneck Identification and Scheduling (BIS)

Compiler/Library/Programmer

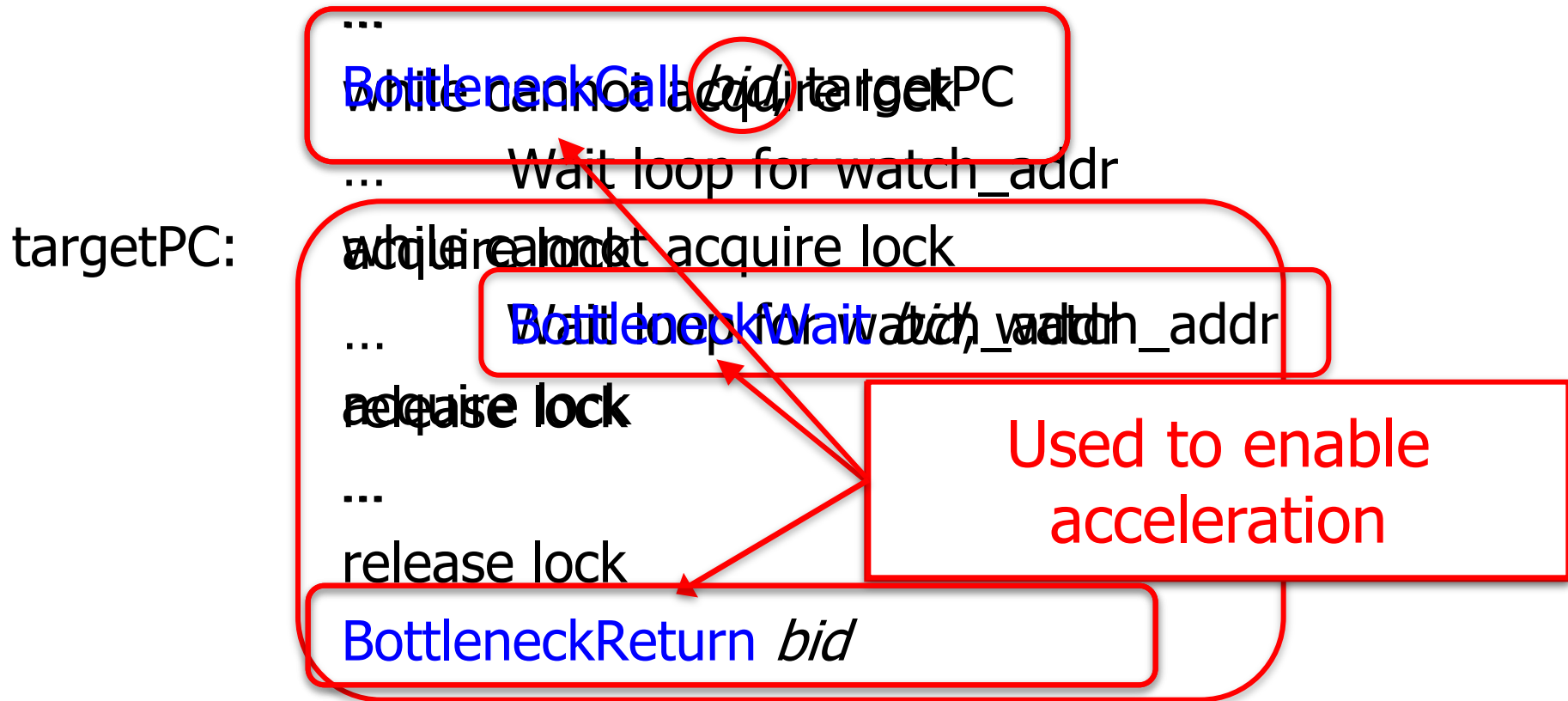
1. Annotate *bottleneck* code
2. Implement *waiting* for bottlenecks

Binary containing
BIS instructions

Hardware

1. Measure *thread waiting cycles (TWC)* for each bottleneck
2. Accelerate bottleneck(s) with the highest TWC

Critical Sections: Code Modifications



Barriers: Code Modifications

...

BottleneckCall *bid*, targetPC

enter barrier

while not all threads in barrier

BottleneckWait *bid*, watch_addr

exit barrier

...

targetPC:

code running for the barrier

...

BottleneckReturn *bid*

Pipeline Stages: Code Modifications

BottleneckCall *bid*, targetPC

...

targetPC:

while not done

while empty queue

BottleneckWait prev_bid

dequeue work

do the work ...

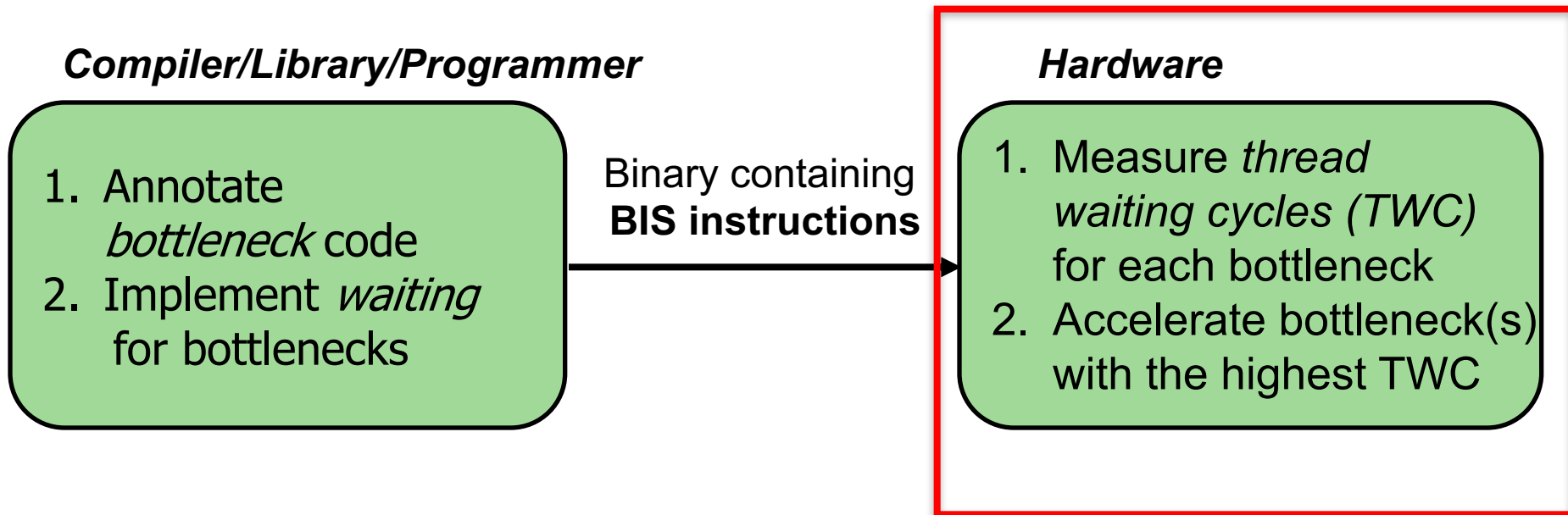
while full queue

BottleneckWait next_bid

enqueue next work

BottleneckReturn *bid*

Bottleneck Identification and Scheduling (BIS)

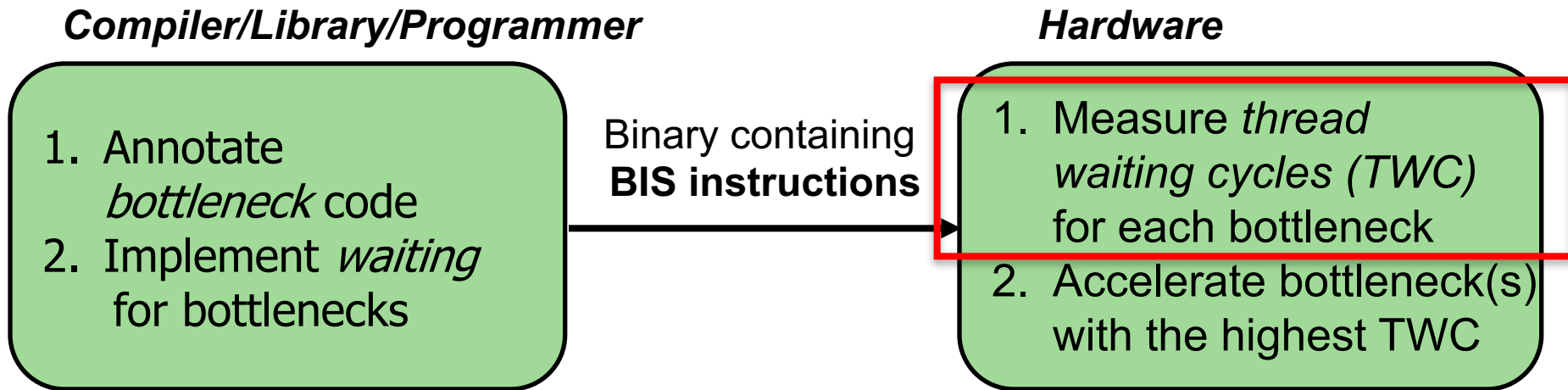


BIS: Hardware Overview

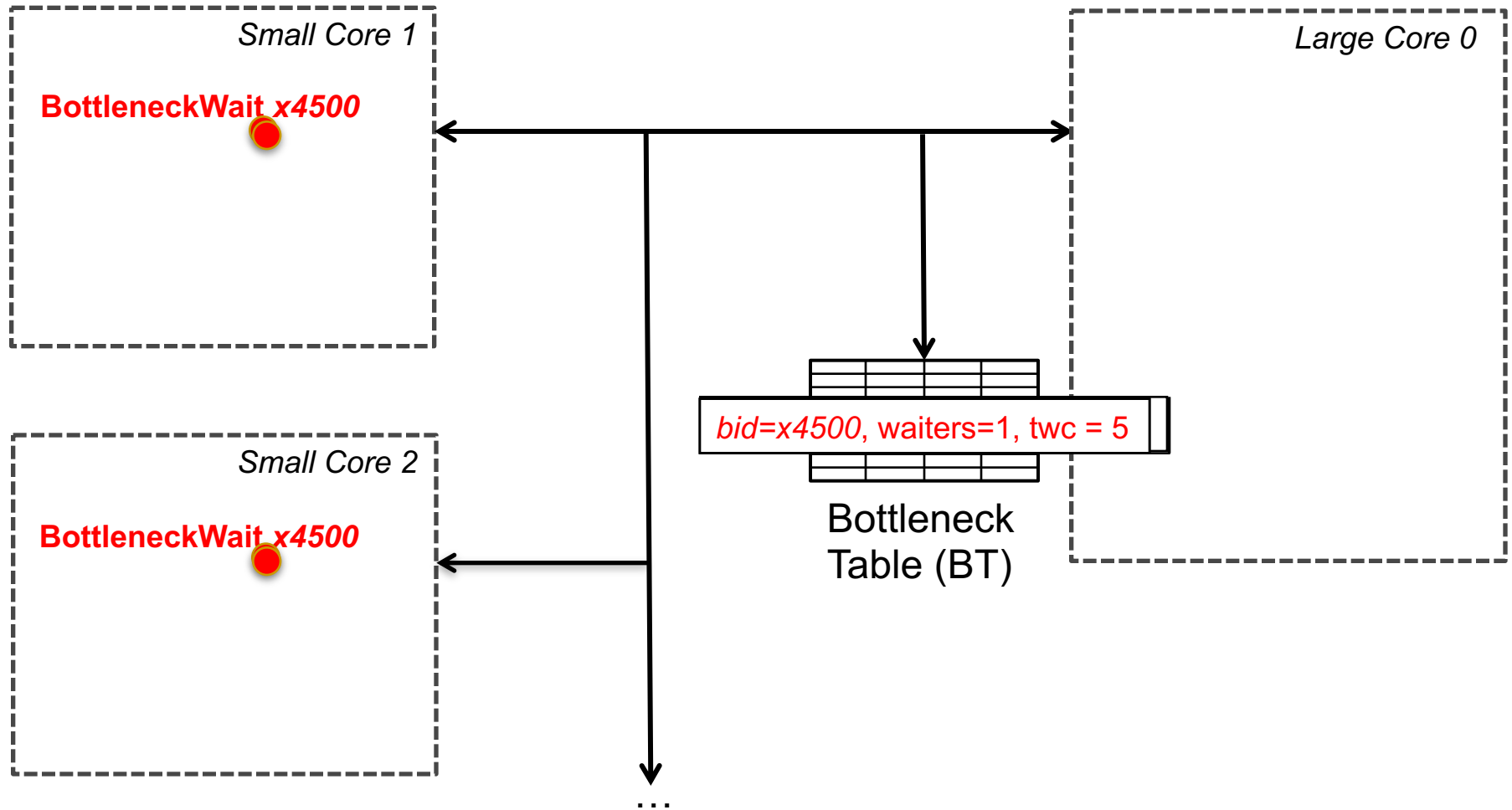
- Performance-limiting bottleneck **identification and acceleration are independent tasks**
- Acceleration can be accomplished in multiple ways
 - ❑ Increasing core frequency/voltage
 - ❑ Prioritization in shared resources [Ebrahimi+, MICRO'11]
 - ❑ **Migration to faster cores in an Asymmetric CMP**

Small core	Small core	Large core	
Small core	Small core		
Small core	Small core	Small core	Small core
Small core	Small core	Small core	Small core

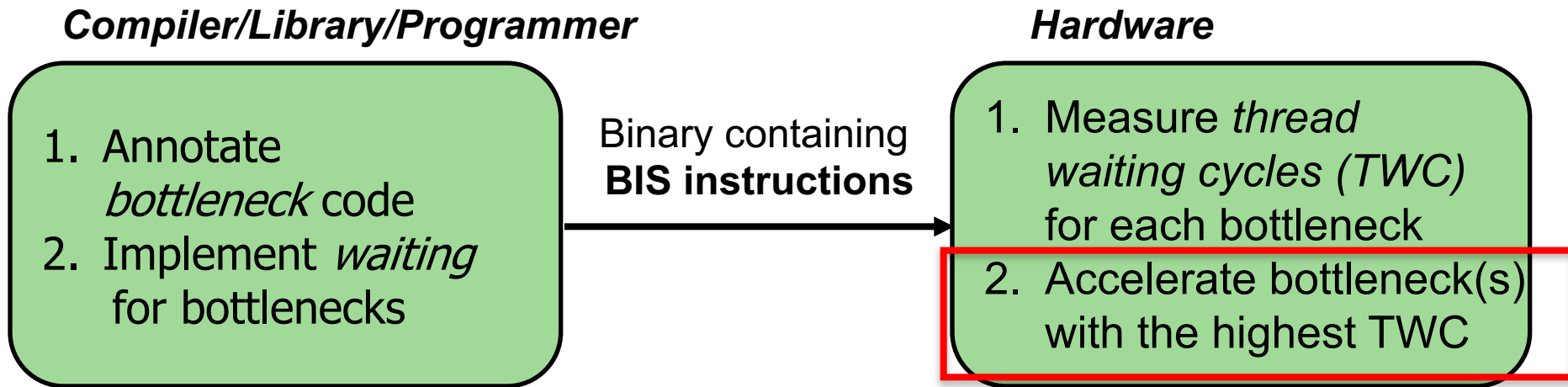
Bottleneck Identification and Scheduling (BIS)



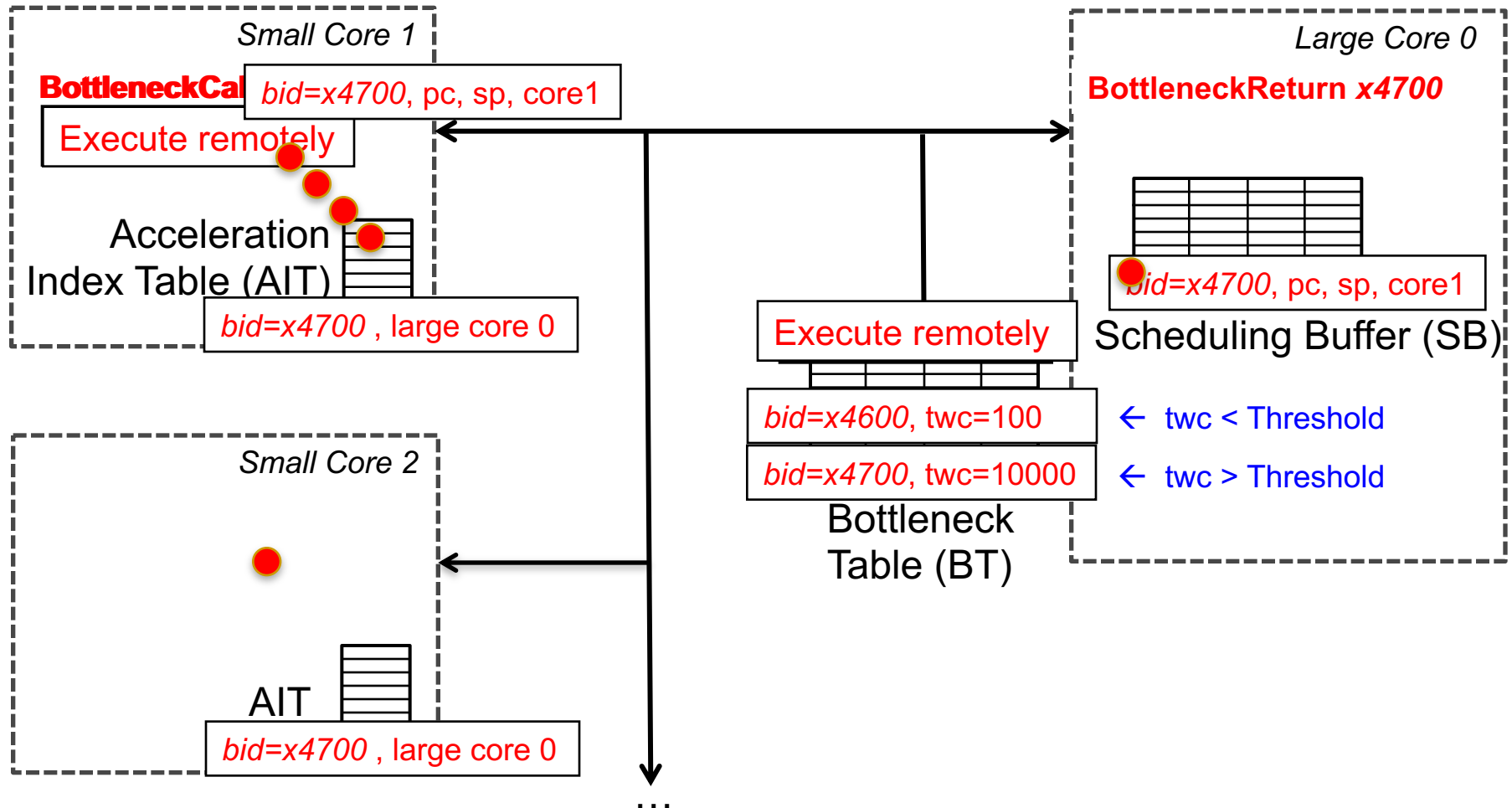
Determining Thread Waiting Cycles for Each Bottleneck



Bottleneck Identification and Scheduling (BIS)



Bottleneck Acceleration



BIS Mechanisms

- Basic mechanisms for BIS:
 - Determining Thread Waiting Cycles ✓
 - Accelerating Bottlenecks ✓

- Mechanisms to improve performance and generality of BIS:
 - Dealing with false serialization
 - Preemptive acceleration
 - Support for multiple large cores

False Serialization and Starvation

- **Observation:** Bottlenecks are picked from Scheduling Buffer in Thread Waiting Cycles order
- **Problem:** An independent bottleneck that is ready to execute has to wait for another bottleneck that has higher thread waiting cycles → **False serialization**
- **Starvation:** Extreme false serialization
- **Solution:** Large core detects when a bottleneck is ready to execute in the Scheduling Buffer but it cannot → sends the bottleneck back to the small core

Preemptive Acceleration

- **Observation:** A bottleneck executing on a small core can become the bottleneck with the highest thread waiting cycles
- **Problem:** This bottleneck should really be accelerated (i.e., executed on the large core)
- **Solution:** The Bottleneck Table detects the situation and sends a preemption signal to the small core. Small core:
 - saves register state on stack, ships the bottleneck to the large core
- Main acceleration mechanism for barriers and pipeline stages

Support for Multiple Large Cores

- **Objective:** to accelerate independent bottlenecks
- Each large core has its own Scheduling Buffer (shared by all of its SMT threads)
- Bottleneck Table assigns each bottleneck to a fixed large core context to
 - preserve cache locality
 - avoid busy waiting
- Preemptive acceleration extended to send multiple instances of a bottleneck to different large core contexts

Hardware Cost

- Main structures:
 - Bottleneck Table (BT): global 32-entry associative cache, minimum-Thread-Waiting-Cycle replacement
 - Scheduling Buffers (SB): one table per large core, as many entries as small cores
 - Acceleration Index Tables (AIT): one 32-entry table per small core
- Off the critical path
- Total storage cost for 56-small-cores, 2-large-cores < 19 KB

BIS Performance Trade-offs

- **Faster bottleneck execution** vs. **fewer parallel threads**
 - Acceleration offsets loss of parallel throughput with large core counts
- **Better shared data locality** vs. **worse private data locality**
 - Shared data stays on large core (good)
 - Private data migrates to large core (bad, but latency hidden with Data Marshaling [Suleman+, ISCA' 10])
- **Benefit of acceleration** vs. **migration latency**
 - Migration latency usually hidden by waiting (good)
 - Unless bottleneck not contended (bad, but likely not on critical path)

Evaluation Methodology

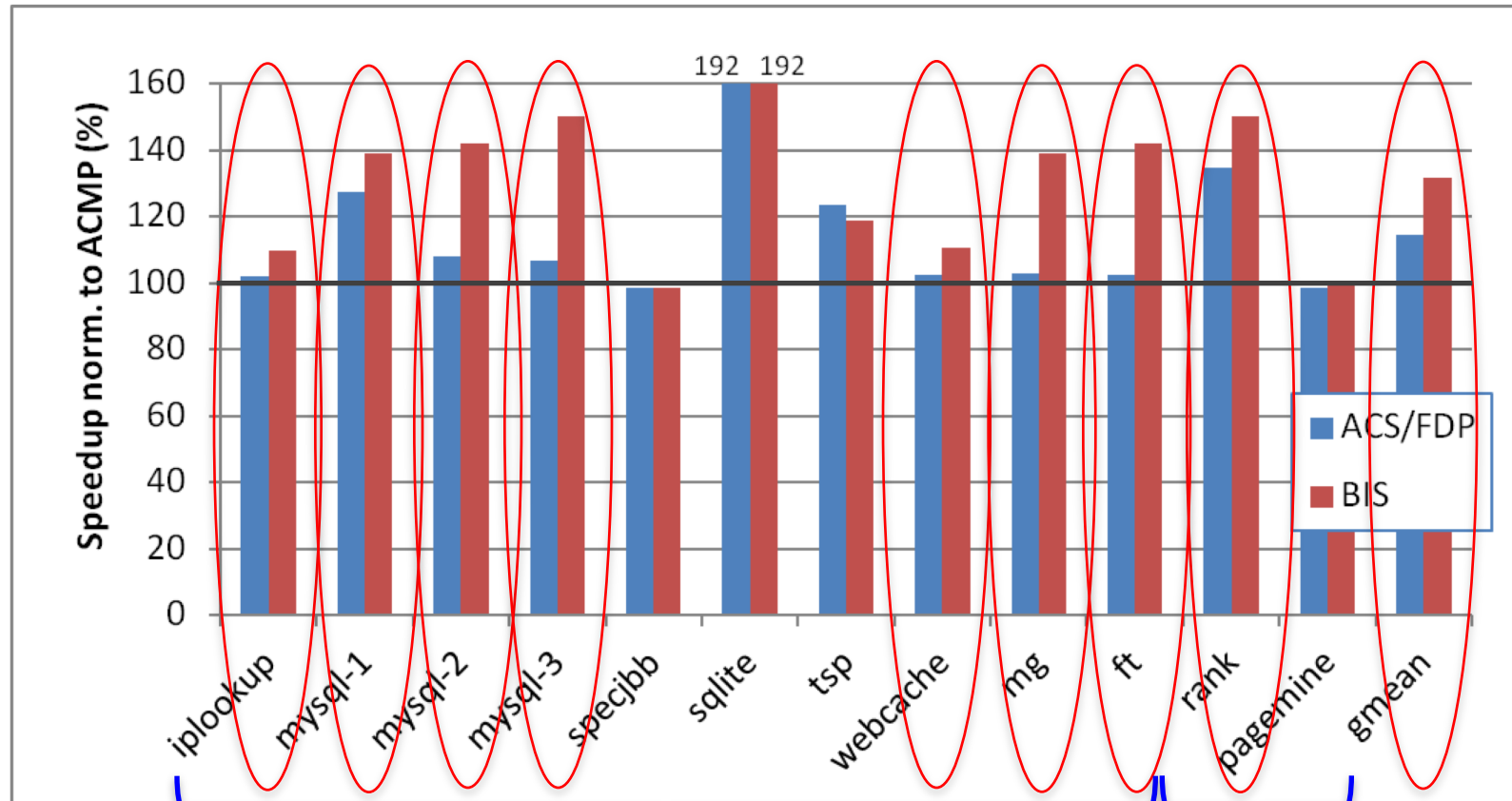
- Workloads: 8 critical section intensive, 2 barrier intensive and 2 pipeline-parallel applications
 - Data mining kernels, scientific, database, web, networking, specjbb
- Cycle-level multi-core x86 simulator
 - 8 to 64 small-core-equivalent area, 0 to 3 large cores, SMT
 - 1 large core is area-equivalent to 4 small cores
- Details:
 - Large core: 4GHz, out-of-order, 128-entry ROB, 4-wide, 12-stage
 - Small core: 4GHz, in-order, 2-wide, 5-stage
 - Private 32KB L1, private 256KB L2, shared 8MB L3
 - On-chip interconnect: Bi-directional ring, 2-cycle hop latency

BIS Comparison Points (Area-Equivalent)

- SCMP (Symmetric CMP)
 - All small cores
- **ACMP** (Asymmetric CMP)
 - Accelerates only Amdahl's serial portions
 - **Our baseline**
- **ACS** (Accelerated Critical Sections)
 - Accelerates only critical sections and Amdahl's serial portions
 - Applicable to multithreaded workloads
(**iplookup, mysql, specjbb, sqlite, tsp, webcache, mg, ft**)
- **FDP** (Feedback-Directed Pipelining)
 - Accelerates only slowest pipeline stages
 - Applicable to pipeline-parallel workloads (**rank, pagemine**)

BIS Performance Improvement

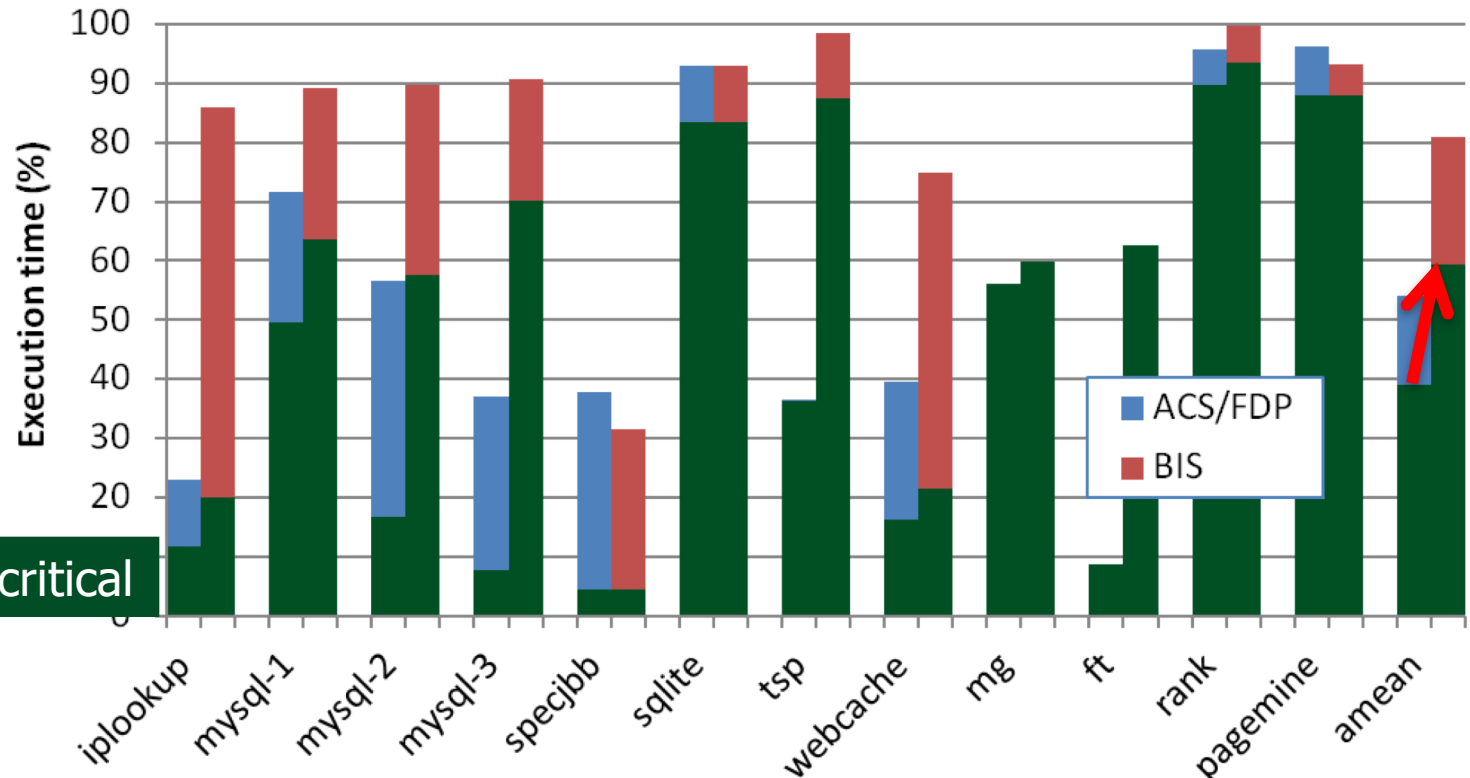
Optimal number of threads, 28 small cores, 1 large core



- BIS outperforms ACS/FDP by 15% and ACMP by 32%
limiting bottlenecks change over barriers, which ACS cannot accelerate
- BIS improves scalability on 4 of the benchmarks

Why Does BIS Work?

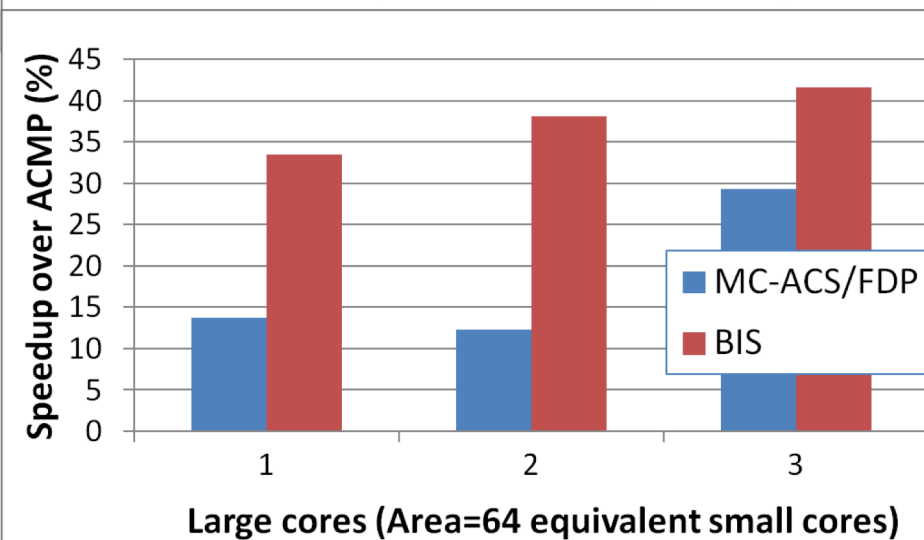
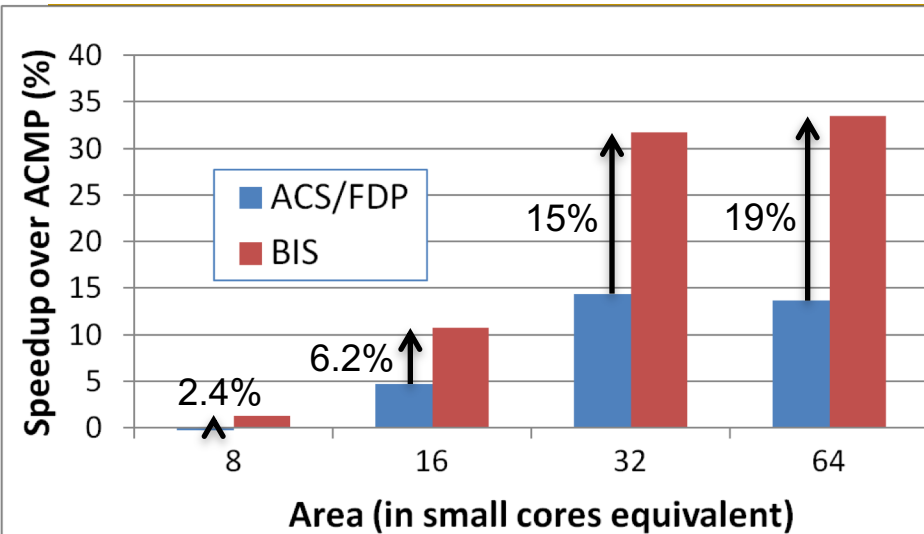
Fraction of execution time spent on predicted-important bottlenecks



Actually critical

- Coverage: fraction of program critical path that is actually identified as bottlenecks
 - 39% (ACS/FDP) to 59% (BIS)
- Accuracy: identified bottlenecks on the critical path over total identified bottlenecks
 - 72% (ACS/FDP) to 73.5% (BIS)

BIS Scaling Results



Performance increases with:

1) More small cores

- Contention due to bottlenecks increases
- Loss of parallel throughput due to large core reduces

2) More large cores

- Can accelerate independent bottlenecks
- *Without reducing parallel throughput (enough cores)*

BIS Summary

- Serializing bottlenecks of different types limit performance of multithreaded applications: Importance changes over time
- BIS is a hardware/software cooperative solution:
 - Dynamically identifies bottlenecks that cause the most thread waiting and accelerates them on large cores of an ACMP
 - Applicable to critical sections, barriers, pipeline stages
- BIS improves application performance and scalability:
 - Performance benefits increase with more cores
- Provides comprehensive fine-grained bottleneck acceleration with no programmer effort

More on Bottleneck Identification & Scheduling

- Jose A. Joao, M. Aater Suleman, Onur Mutlu, and Yale N. Patt,
"Bottleneck Identification and Scheduling in Multithreaded Applications"

Proceedings of the 17th International Conference on Architectural Support for Programming Languages and Operating Systems (ASPLOS), London, UK, March 2012. [Slides \(ppt\)](#) ([pdf](#))

Bottleneck Identification and Scheduling in Multithreaded Applications

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Handling Private Data Locality: Data Marshaling

M. Aater Suleman, Onur Mutlu, Jose A. Joao, Khubaib, and Yale N. Patt,
"Data Marshaling for Multi-core Architectures"
*Proceedings of the 37th International Symposium on Computer Architecture (ISCA),
pages 441-450, Saint-Malo, France, June 2010.*

Staged Execution Model (I)

- Goal: speed up a program by dividing it up into pieces
 - Idea
 - Split program code into *segments*
 - Run each segment on the core best-suited to run it
 - Each core assigned a work-queue, storing segments to be run
 - Benefits
 - Accelerates segments/critical-paths using specialized/heterogeneous cores
 - Exploits inter-segment parallelism
 - Improves locality of within-segment data
 - Examples
 - Accelerated critical sections, Bottleneck identification and scheduling
 - Producer-consumer pipeline parallelism
 - Task parallelism (Cilk, Intel TBB, Apple Grand Central Dispatch)
 - Special-purpose cores and functional units
-

Staged Execution Model (II)



Staged Execution Model (III)

Split code into segments

Segment S0

LOAD X
STORE Y
STORE Y

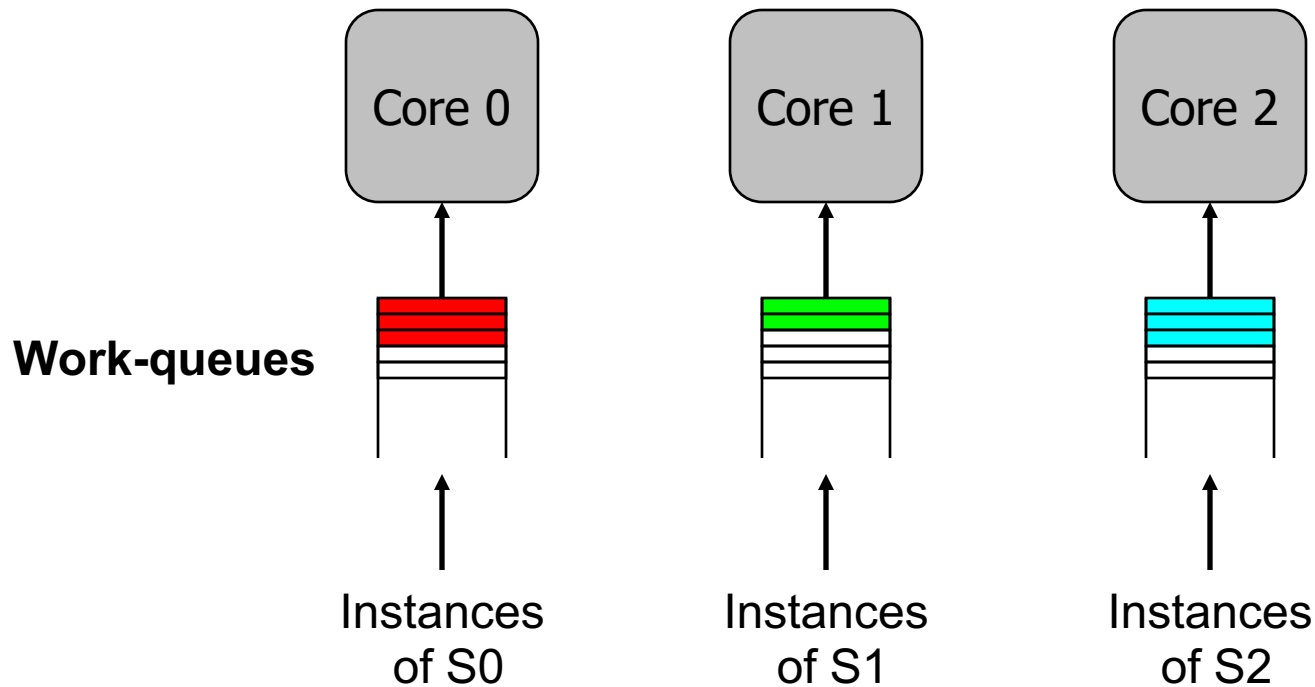
Segment S1

LOAD Y
....
STORE Z

Segment S2

LOAD Z
....

Staged Execution Model (IV)



Staged Execution Model: Segment Spawning

Core 0

Core 1

Core 2

S0

LOAD X
STORE Y
STORE Y

S1

LOAD Y
....
STORE Z

S2

LOAD Z
....

Staged Execution Model: Two Examples

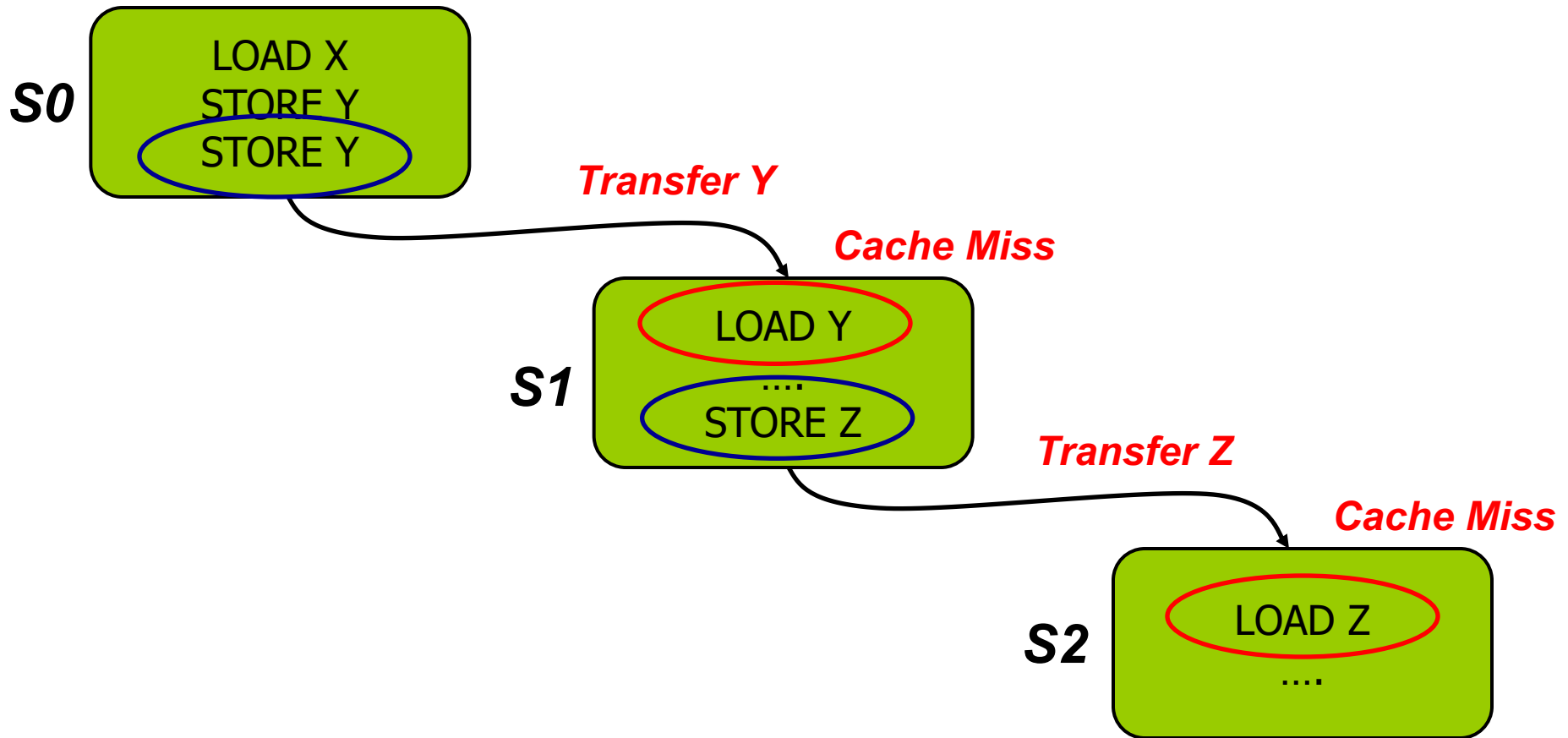
- **Accelerated Critical Sections** [Suleman et al., ASPLOS 2009]
 - Idea: Ship critical sections to a large core in an asymmetric CMP
 - Segment 0: Non-critical section
 - Segment 1: Critical section
 - Benefit: Faster execution of critical section, reduced serialization, improved lock and shared data locality
- **Producer-Consumer Pipeline Parallelism**
 - Idea: Split a loop iteration into multiple “pipeline stages” where one stage consumes data produced by the previous stage → each stage runs on a different core
 - Segment N: Stage N
 - Benefit: Stage-level parallelism, better locality → faster execution

Problem: Locality of Inter-segment Data

Core 0

Core 1

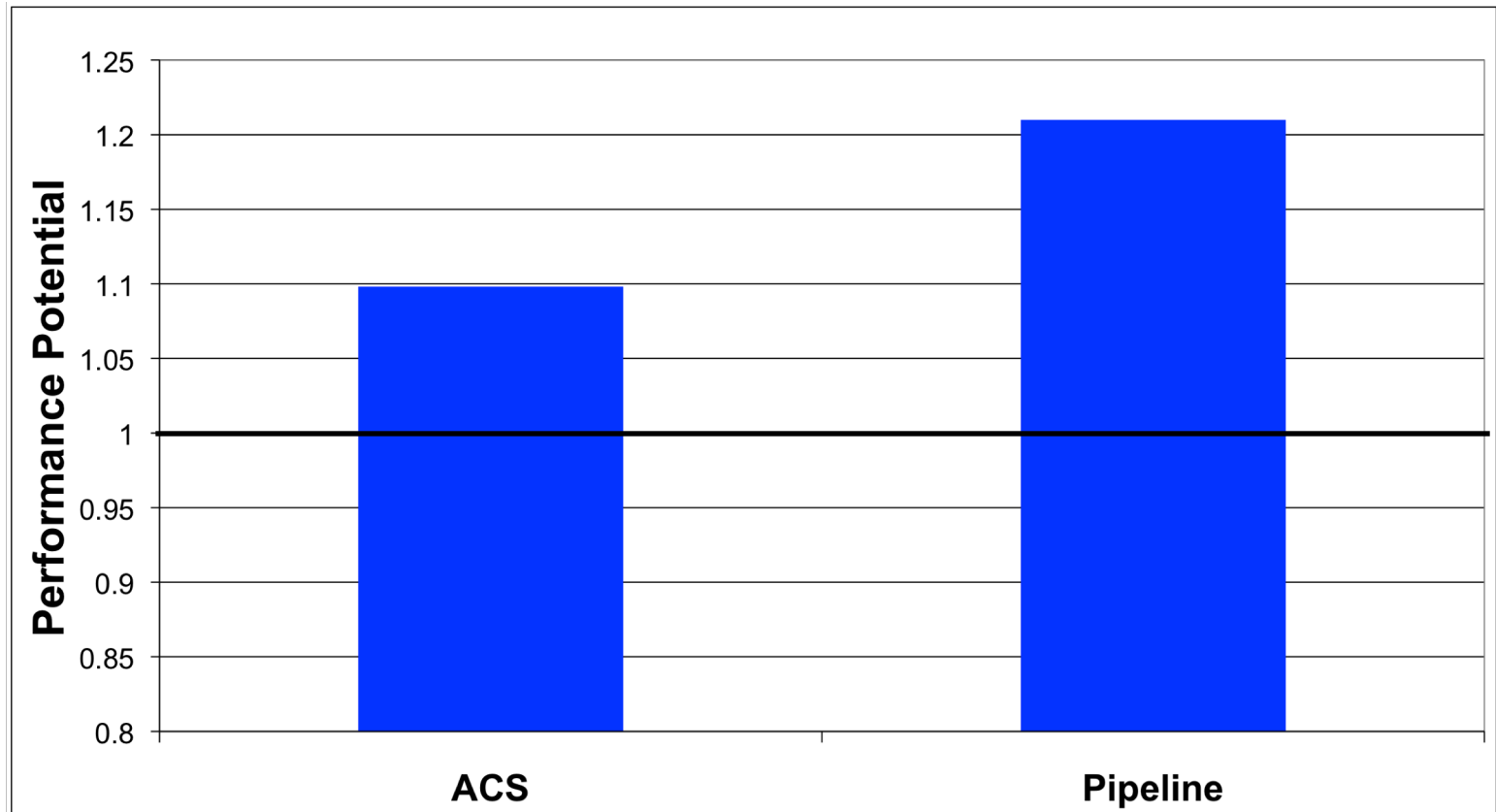
Core 2



Problem: Locality of Inter-segment Data

- Accelerated Critical Sections [Suleman et al., ASPLOS 2010]
 - Idea: Ship critical sections to a large core in an ACMP
 - Problem: Critical section incurs a cache miss when it touches data produced in the non-critical section (i.e., thread private data)
- Producer-Consumer Pipeline Parallelism
 - Idea: Split a loop iteration into multiple “pipeline stages” → each stage runs on a different core
 - Problem: A stage incurs a cache miss when it touches data produced by the previous stage
- Performance of Staged Execution limited by inter-segment cache misses

What if We Eliminated All Inter-segment Misses?

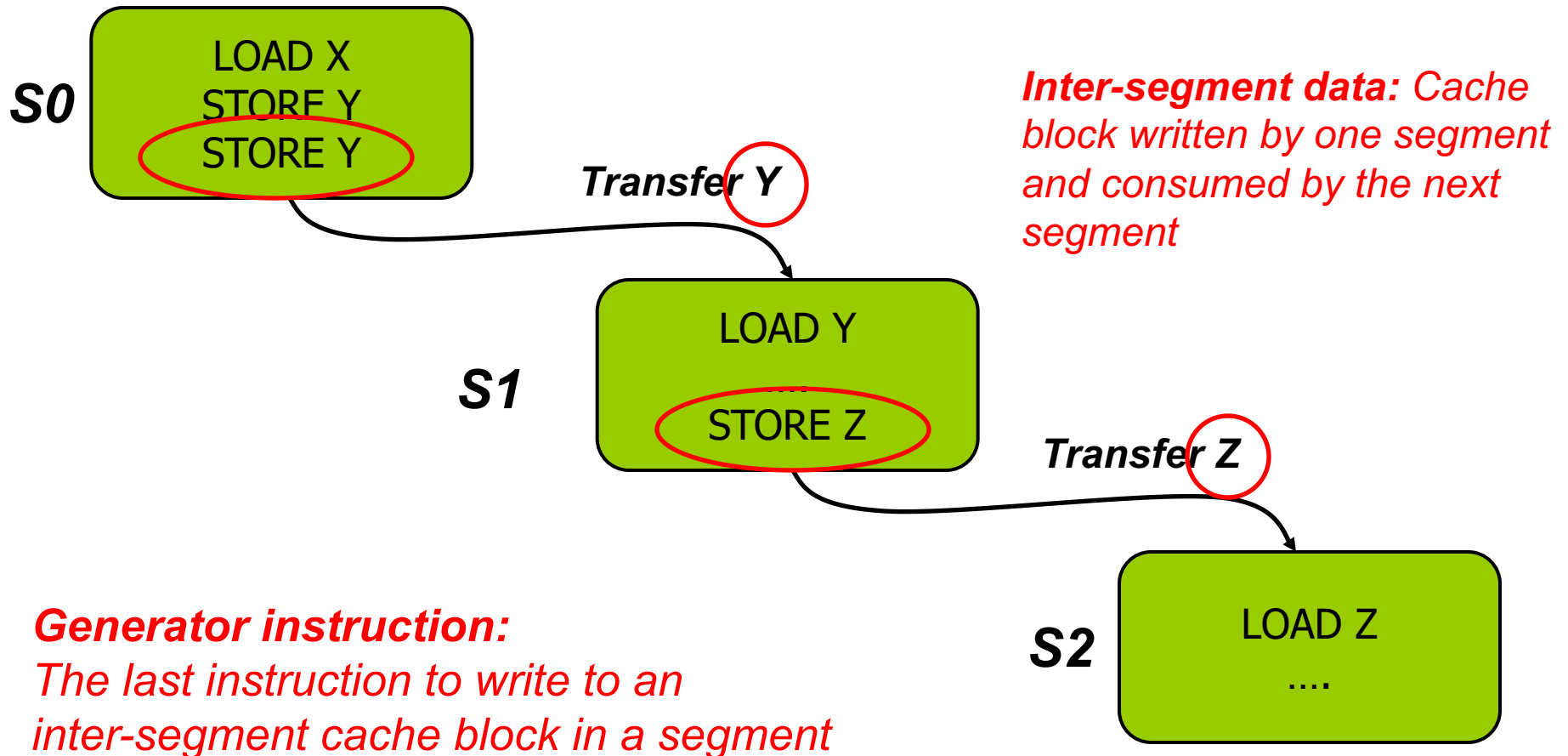


Terminology

Core 0

Core 1

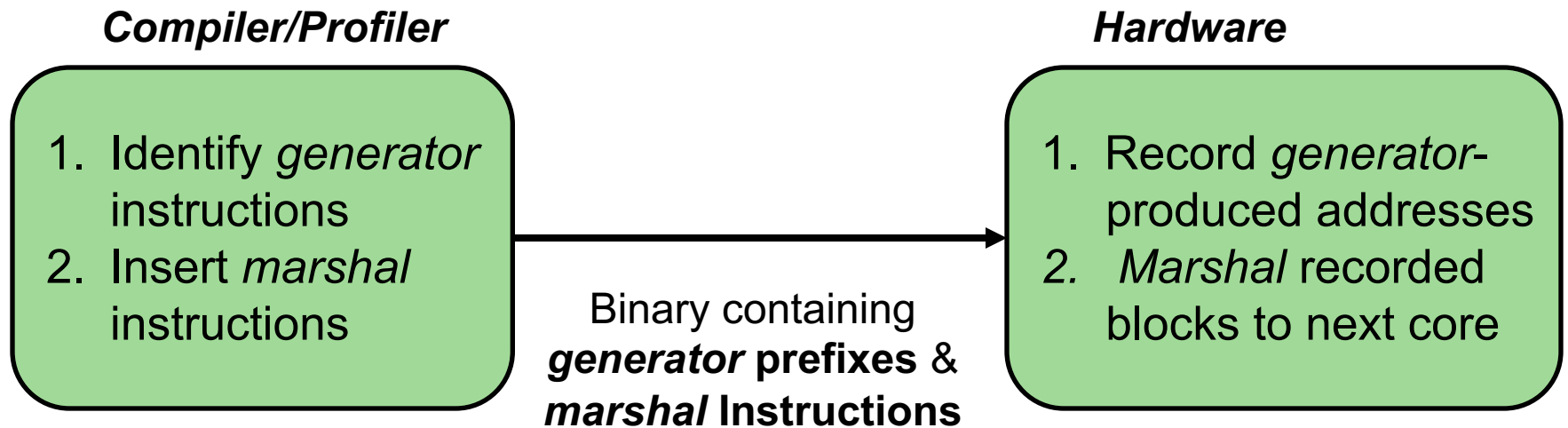
Core 2



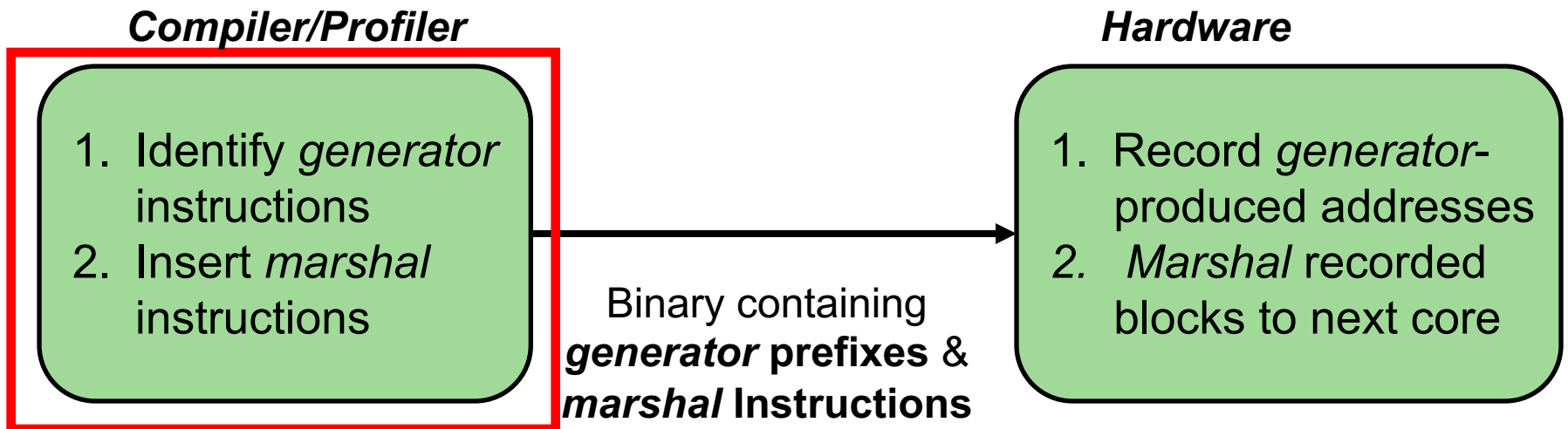
Key Observation and Idea

- Observation: Set of generator instructions is stable over execution time and across input sets
- Idea:
 - Identify the generator instructions
 - Record cache blocks produced by generator instructions
 - Proactively send such cache blocks to the next segment's core before initiating the next segment
- Suleman et al., “Data Marshaling for Multi-Core Architectures,” ISCA 2010, IEEE Micro Top Picks 2011.

Data Marshaling



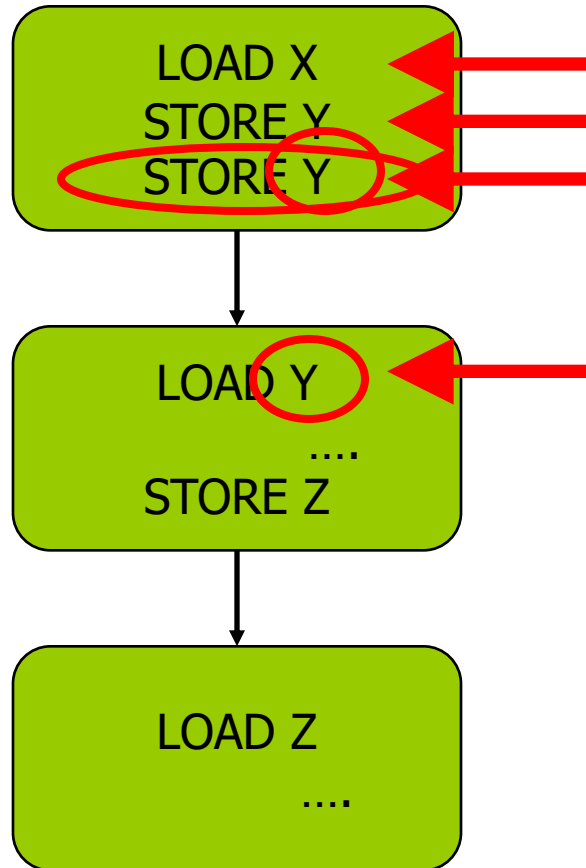
Data Marshaling



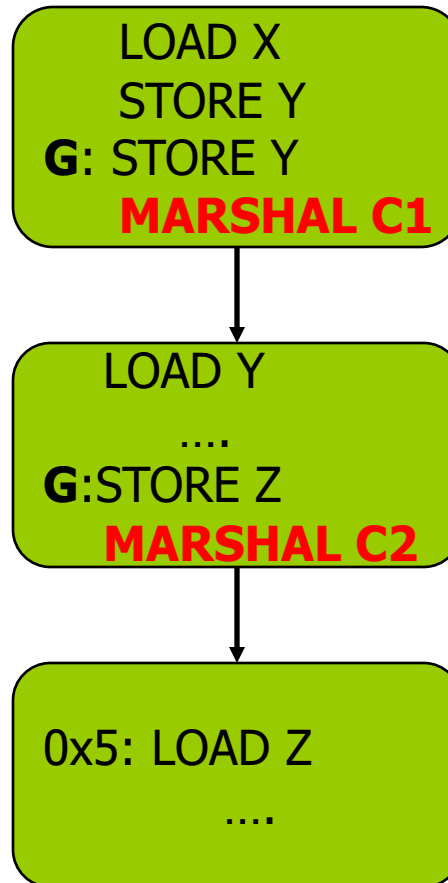
Profiling Algorithm

Inter-segment data

*Mark as Generator
Instruction*



Marshal Instructions



When to send (Marshal)

Where to send (C1)

DM Support/Cost

- Profiler/Compiler: Generators, marshal instructions
- ISA: Generator prefix, marshal instructions
- Library/Hardware: Bind next segment ID to a physical core

- Hardware
 - Marshal Buffer
 - Stores physical addresses of cache blocks to be marshaled
 - 16 entries enough for almost all workloads → 96 bytes per core
 - Ability to execute generator prefixes and marshal instructions
 - Ability to push data to another cache

DM: Advantages, Disadvantages

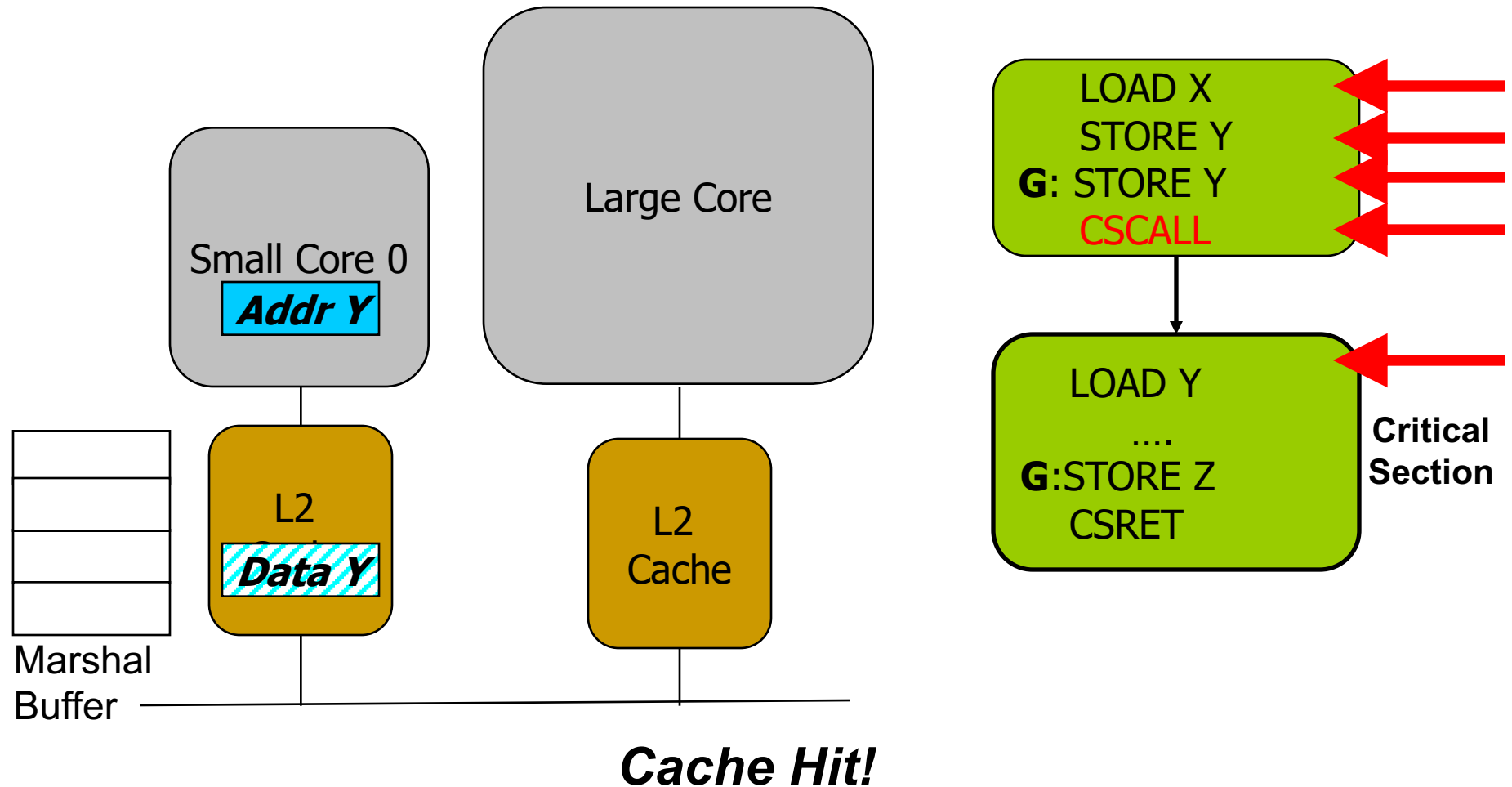
■ Advantages

- ❑ **Timely data transfer**: Push data to core before needed
- ❑ **Can marshal any arbitrary sequence of lines**: Identifies generators, not patterns
- ❑ **Low hardware cost**: Profiler marks generators, no need for hardware to find them

■ Disadvantages

- ❑ **Requires profiler and ISA support**
- ❑ **Not always accurate (generator set is conservative)**: Pollution at remote core, wasted bandwidth on interconnect
 - Not a large problem as number of inter-segment blocks is small

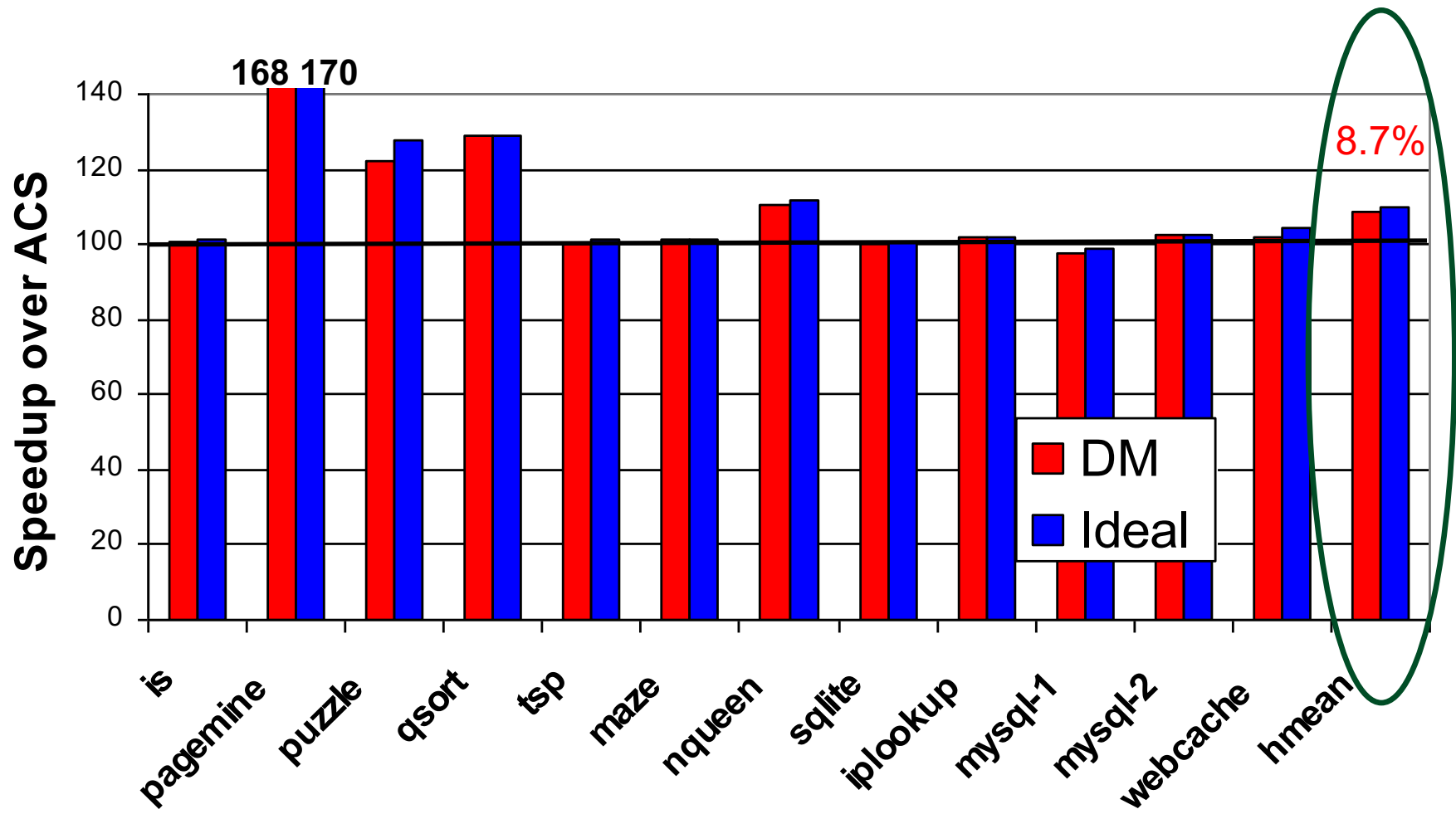
Accelerated Critical Sections with DM



Accelerated Critical Sections: Methodology

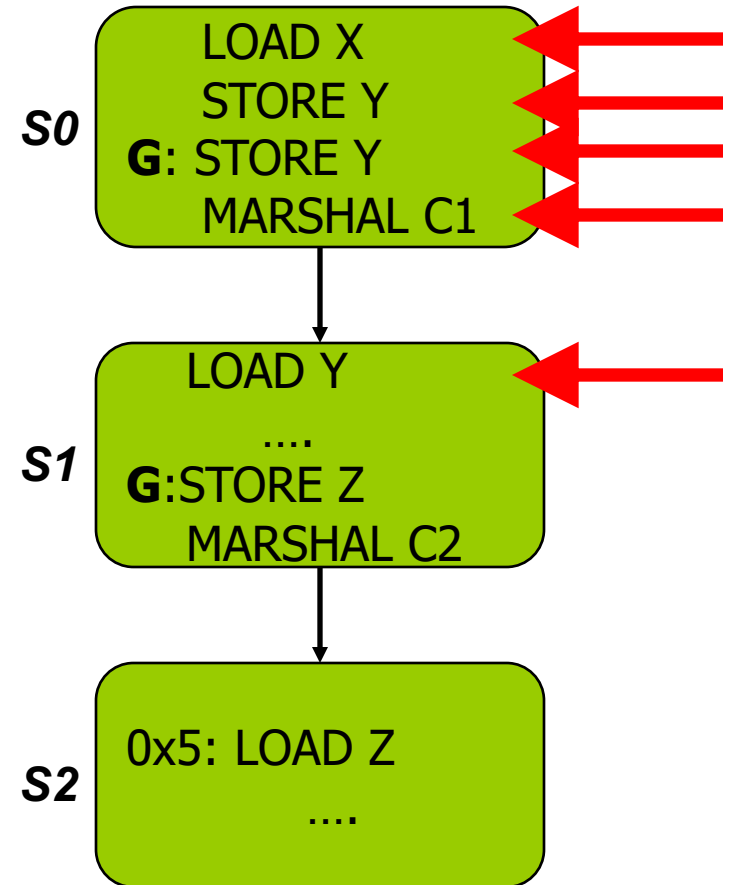
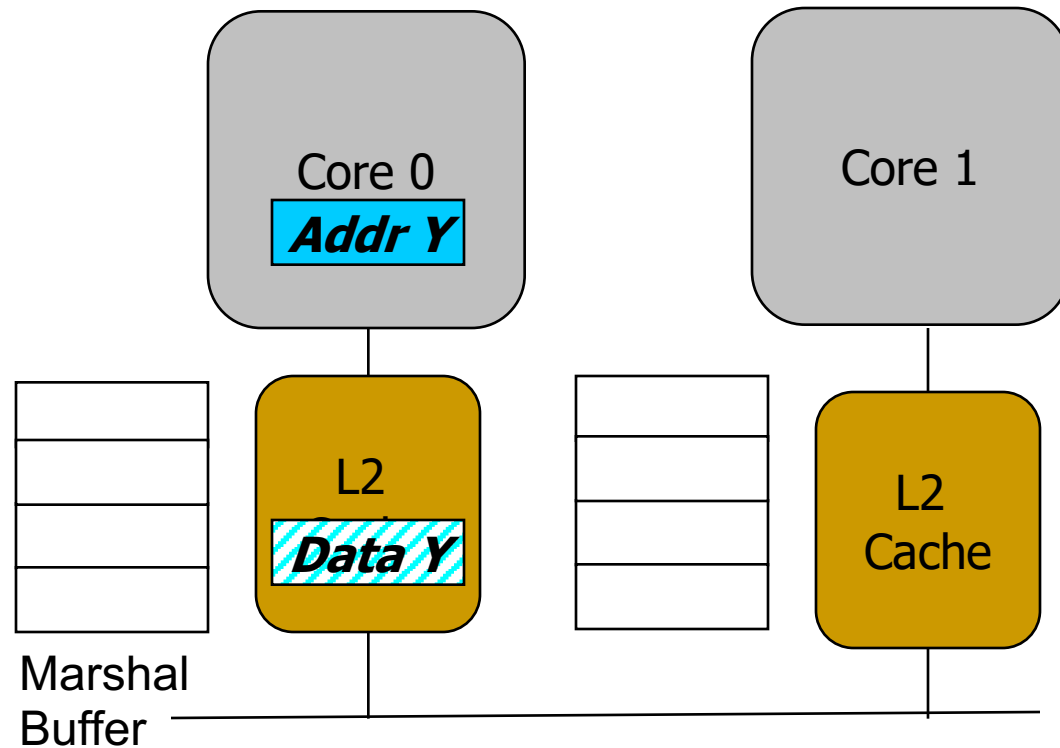
- Workloads: 12 critical section intensive applications
 - Data mining kernels, sorting, database, web, networking
 - Different training and simulation input sets
- Multi-core x86 simulator
 - 1 large and 28 small cores
 - Aggressive stream prefetcher employed at each core
- Details:
 - Large core: 2GHz, out-of-order, 128-entry ROB, 4-wide, 12-stage
 - Small core: 2GHz, in-order, 2-wide, 5-stage
 - Private 32 KB L1, private 256KB L2, 8MB shared L3
 - On-chip interconnect: Bi-directional ring, 5-cycle hop latency

DM on Accelerated Critical Sections: Results



Pipeline Parallelism

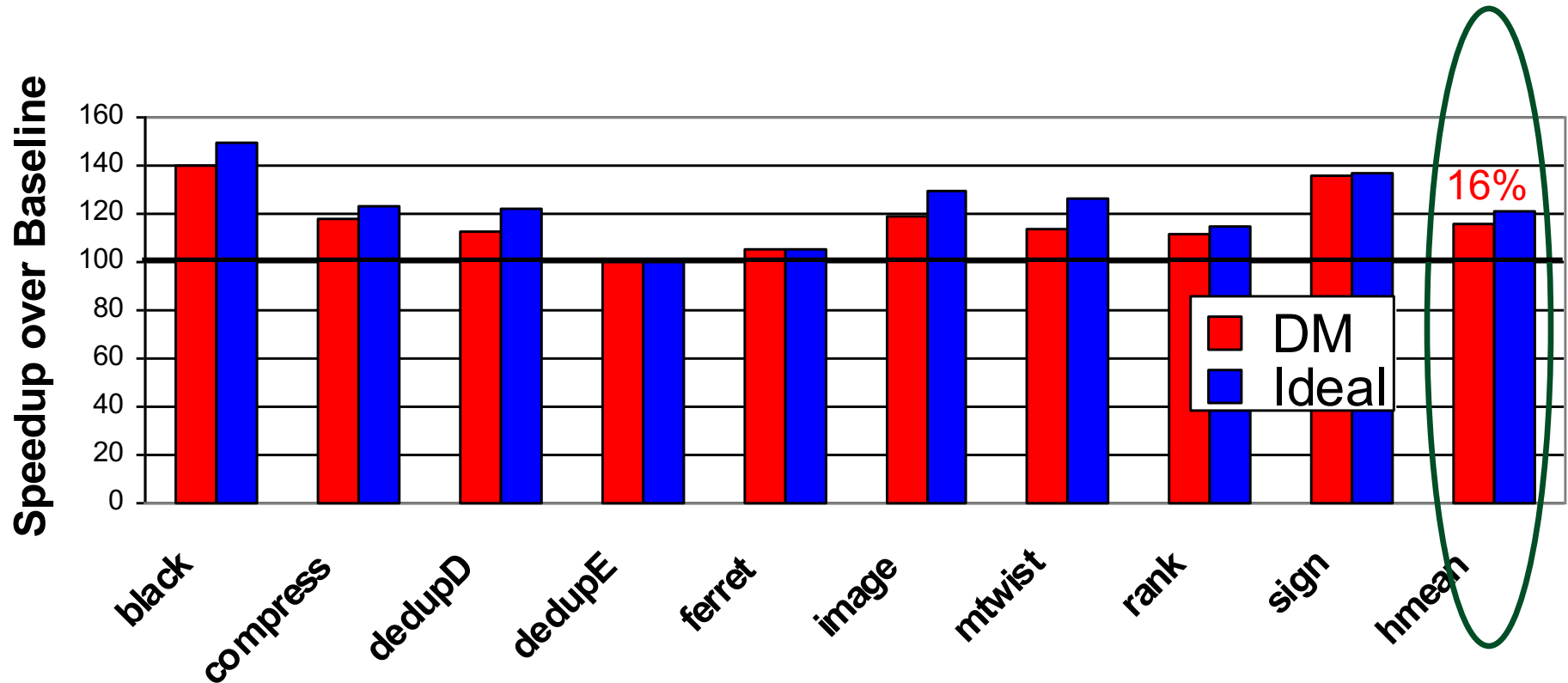
Cache Hit!



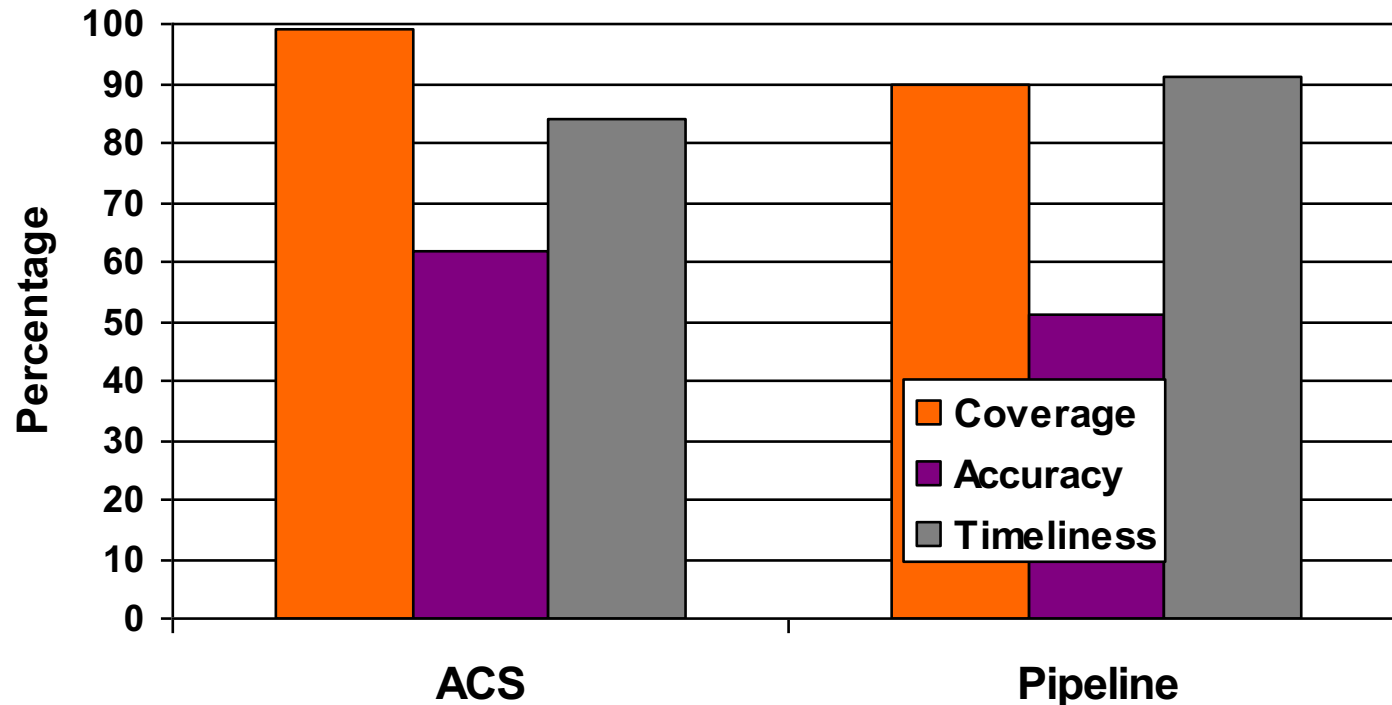
Pipeline Parallelism: Methodology

- Workloads: 9 applications with pipeline parallelism
 - Financial, compression, multimedia, encoding/decoding
 - Different training and simulation input sets
- Multi-core x86 simulator
 - 32-core CMP: 2GHz, in-order, 2-wide, 5-stage
 - Aggressive stream prefetcher employed at each core
 - Private 32 KB L1, private 256KB L2, 8MB shared L3
 - On-chip interconnect: Bi-directional ring, 5-cycle hop latency

DM on Pipeline Parallelism: Results



DM Coverage, Accuracy, Timeliness



- High coverage of inter-segment misses in a timely manner
- Medium accuracy does not impact performance
 - Only 5.0 and 6.8 cache blocks marshaled for average segment

Scaling Results

- DM performance improvement increases with
 - More cores
 - Higher interconnect latency
 - Larger private L2 caches
- Why? Inter-segment data misses become a larger bottleneck
 - More cores → More communication
 - Higher latency → Longer stalls due to communication
 - Larger L2 cache → Communication misses remain

Other Applications of Data Marshaling

- Can be applied to other Staged Execution models
 - Task parallelism models
 - Cilk, Intel TBB, Apple Grand Central Dispatch
 - Special-purpose remote functional units
 - Computation spreading [Chakraborty et al., ASPLOS' 06]
 - Thread motion/migration [e.g., Rangan et al., ISCA' 09]
- Can be an enabler for more aggressive SE models
 - Lowers the cost of data migration
 - an important overhead in remote execution of code segments
 - Remote execution of finer-grained tasks can become more feasible → finer-grained parallelization in multi-cores

Data Marshaling Summary

- **Inter-segment data transfers between cores** limit the benefit of promising Staged Execution (SE) models
- Data Marshaling is a hardware/software cooperative solution: **detect inter-segment data generator instructions and push their data to next segment's core**
 - Significantly reduces cache misses for inter-segment data
 - Low cost, high-coverage, timely for arbitrary address sequences
 - Achieves most of the potential of eliminating such misses
- Applicable to several existing Staged Execution models
 - Accelerated Critical Sections: 9% performance benefit
 - Pipeline Parallelism: 16% performance benefit
- Can enable new models → **very fine-grained remote execution**

More on Bottleneck Identification & Scheduling

- M. Aater Suleman, Onur Mutlu, Jose A. Joao, Khubaib, and Yale N. Patt, **"Data Marshaling for Multi-core Architectures"**
Proceedings of the 37th International Symposium on Computer Architecture (ISCA), pages 441-450, Saint-Malo, France, June 2010. [Slides \(ppt\)](#)

Data Marshaling for Multi-core Architectures

M. Aater Suleman[†] Onur Mutlu[§] José A. Joao[†] Khubaib[†] Yale N. Patt[†]

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onur@cmu.edu

Summary

- Applications and phases have varying performance requirements
- Designs evaluated on multiple metrics/constraints: energy, performance, reliability, fairness, ...
- **One-size-fits-all** design cannot satisfy all requirements and metrics: **cannot get the best of all worlds**
- **Asymmetry** in design enables tradeoffs: **can get the best of all worlds**
 - Asymmetry in core microarch. → **Accelerated Critical Sections, BIS, DM**
→ Good parallel performance + Good serialized performance
 - Asymmetry in memory scheduling → **Thread Cluster Memory Scheduling**
→ Good throughput + good fairness
- Simple asymmetric designs can be effective and low-cost

Readings: Heterogeneous Cores

- M. Aater Suleman, Onur Mutlu, Moinuddin K. Qureshi, and Yale N. Patt, **"Accelerating Critical Section Execution with Asymmetric Multi-Core Architectures"**
Proceedings of the 14th International Conference on Architectural Support for Programming Languages and Operating Systems (ASPLOS), pages 253-264, Washington, DC, March 2009. [Slides \(ppt\)](#)
- M. Aater Suleman, Onur Mutlu, Jose A. Joao, Khubaib, and Yale N. Patt, **"Data Marshaling for Multi-core Architectures"**
Proceedings of the 37th International Symposium on Computer Architecture (ISCA), pages 441-450, Saint-Malo, France, June 2010. [Slides \(ppt\)](#)
- Jose A. Joao, M. Aater Suleman, Onur Mutlu, and Yale N. Patt, **"Bottleneck Identification and Scheduling in Multithreaded Applications"**
Proceedings of the 17th International Conference on Architectural Support for Programming Languages and Operating Systems (ASPLOS), London, UK, March 2012. [Slides \(ppt\)](#) [\(pdf\)](#)
- Jose A. Joao, M. Aater Suleman, Onur Mutlu, and Yale N. Patt, **"Utility-Based Acceleration of Multithreaded Applications on Asymmetric CMPs"**
Proceedings of the 40th International Symposium on Computer Architecture (ISCA), Tel-Aviv, Israel, June 2013. [Slides \(ppt\)](#) [Slides \(pdf\)](#)

A Case for Asymmetry Everywhere

Onur Mutlu,

"Asymmetry Everywhere (with Automatic Resource Management)"

*CRA Workshop on Advancing Computer Architecture Research: Popular
Parallel Programming*, San Diego, CA, February 2010.

Position paper

Asymmetry Enables Customization

c	c	c	c
c	c	c	c
c	c	c	c
c	c	c	c

Symmetric

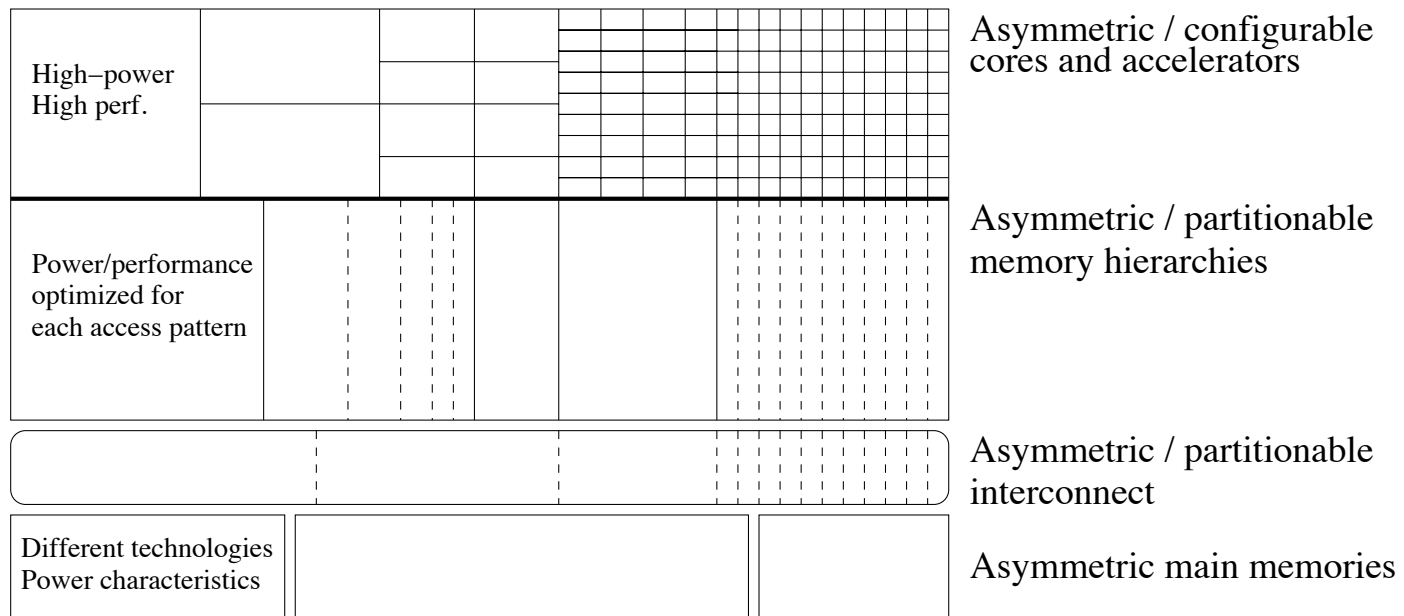
C1		C2	
		C3	
C4	C4	C4	C4
C5	C5	C5	C5

Asymmetric

- Symmetric: One size fits all
 - Energy and performance suboptimal for different phase behaviors
- Asymmetric: Enables tradeoffs and customization
 - Processing requirements vary across applications and phases
 - Execute code on best-fit resources (minimal energy, adequate perf.)

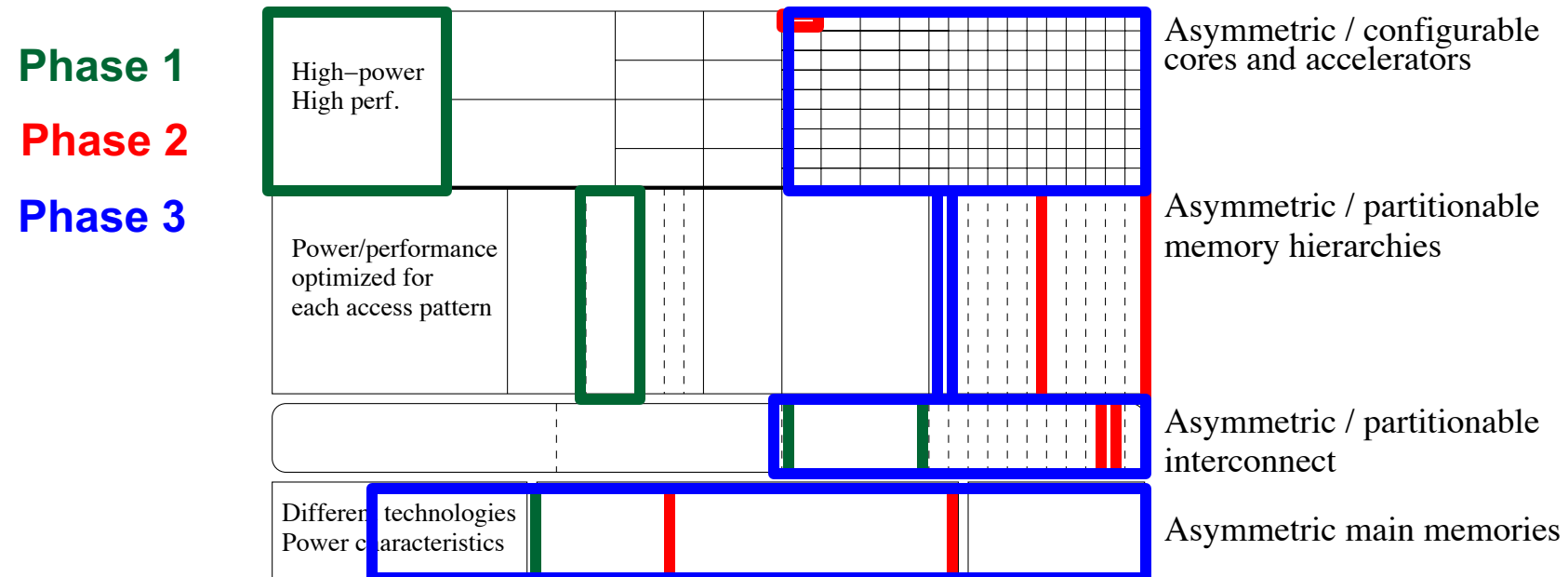
Thought Experiment: Asymmetry Everywhere

- Design each hardware resource with **asymmetric, (re-)configurable, partitionable components**
 - ❑ Different power/performance/reliability characteristics
 - ❑ To fit different computation/access/communication patterns



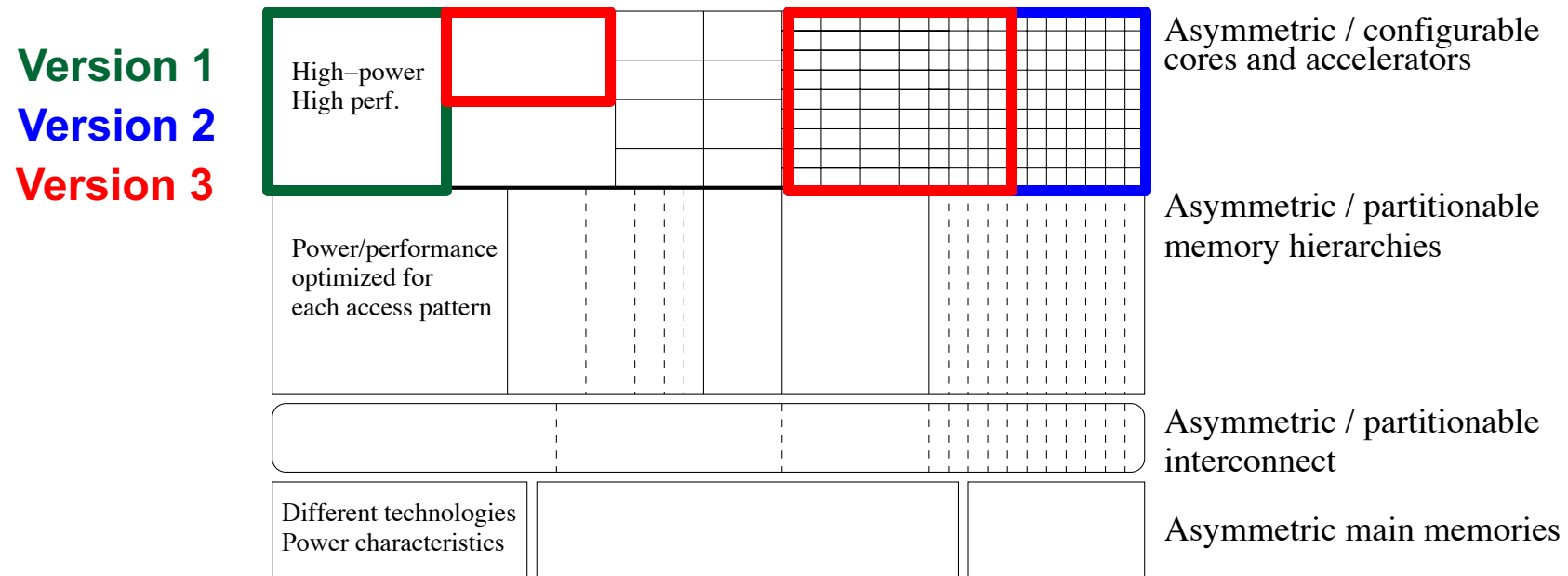
Thought Experiment: Asymmetry Everywhere

- Design the **runtime system (HW & SW)** to **automatically choose** the best-fit components for each phase
 - Satisfy performance/SLA with minimal energy
 - Dynamically stitch together the “best-fit” chip for each phase



Thought Experiment: Asymmetry Everywhere

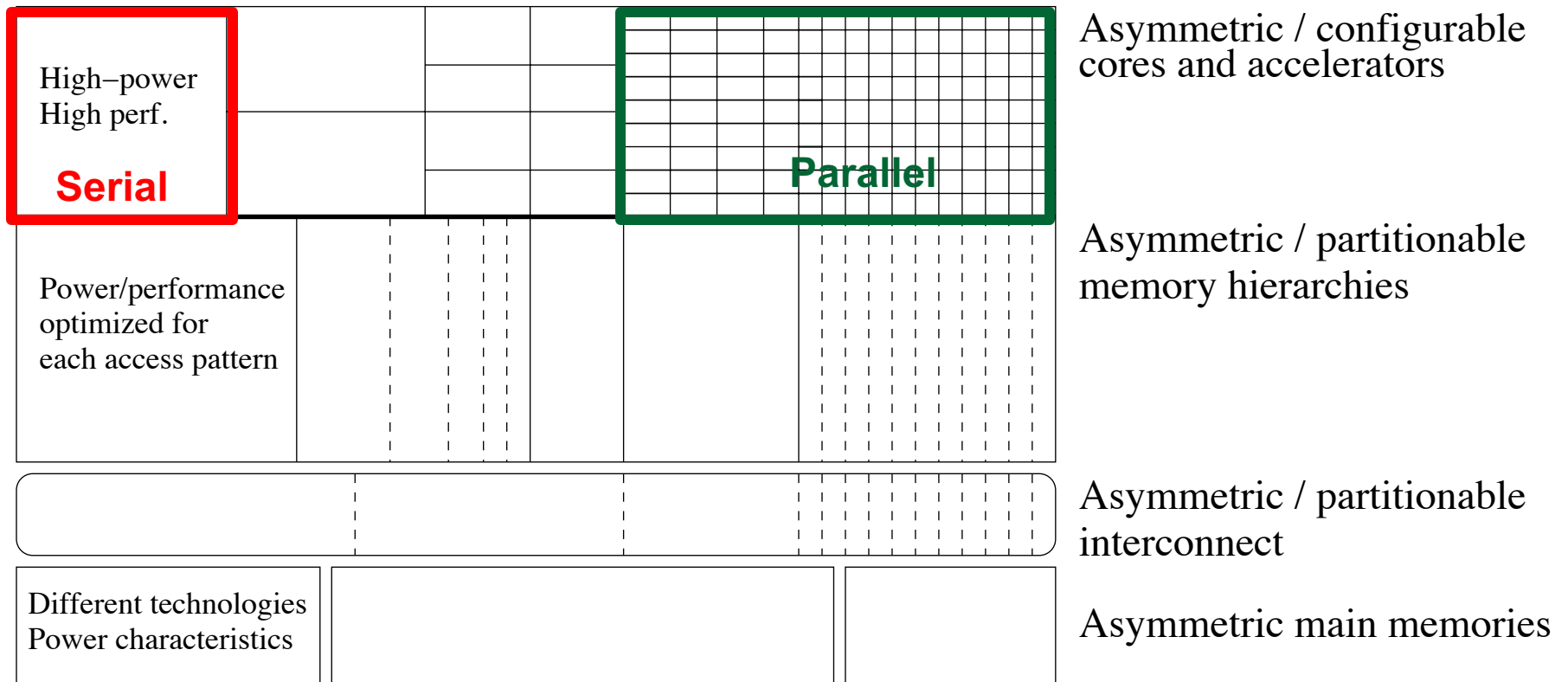
- **Morph software components** to match asymmetric HW components
 - Multiple versions for different resource characteristics



Many Research and Design Questions

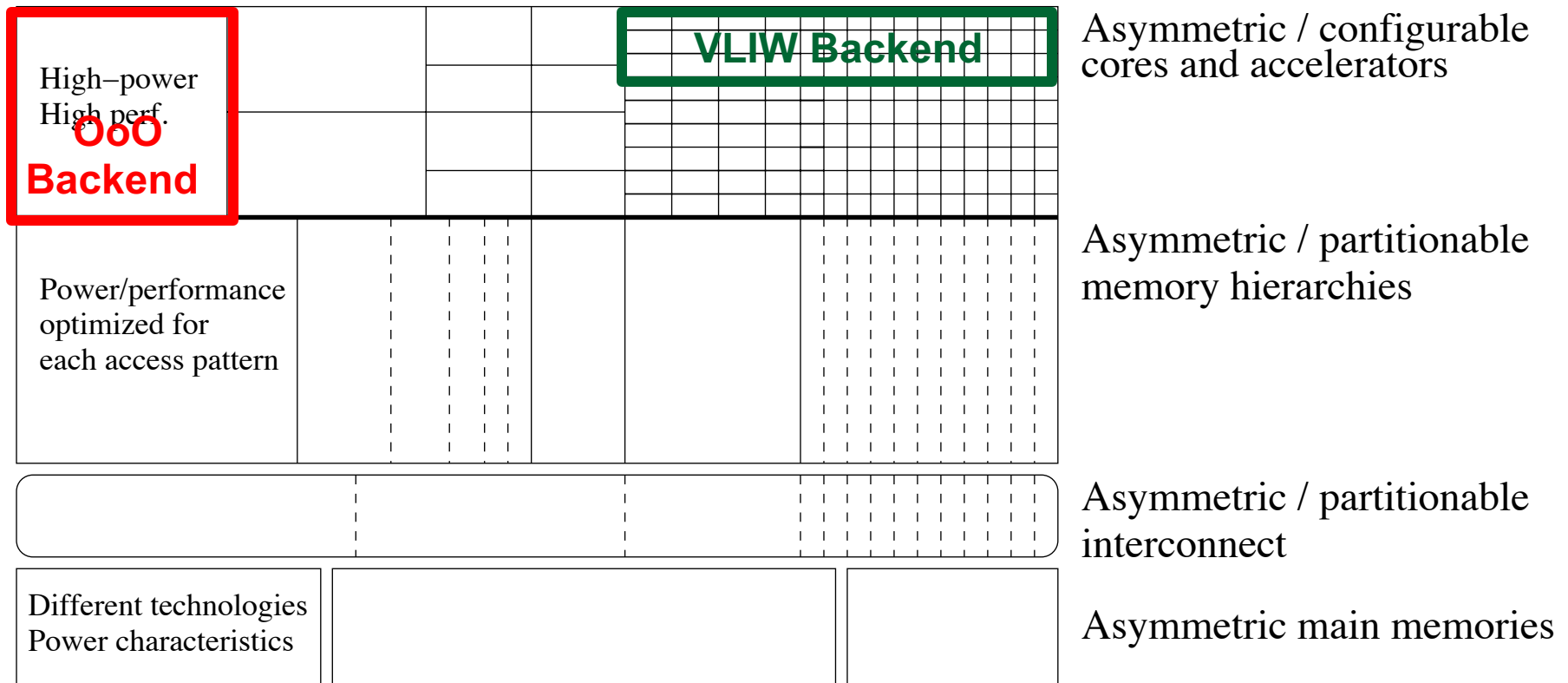
- How to design asymmetric components?
 - Fixed, partitionable, reconfigurable components?
 - What types of asymmetry? Access patterns, technologies?
- What monitoring to perform cooperatively in HW/SW?
 - Automatically discover phase/task requirements
- How to design feedback/control loop between components and runtime system software?
- How to design the runtime to automatically manage resources?
 - Track task behavior, pick “best-fit” components for the entire workload

Exploiting Asymmetry: Simple Examples



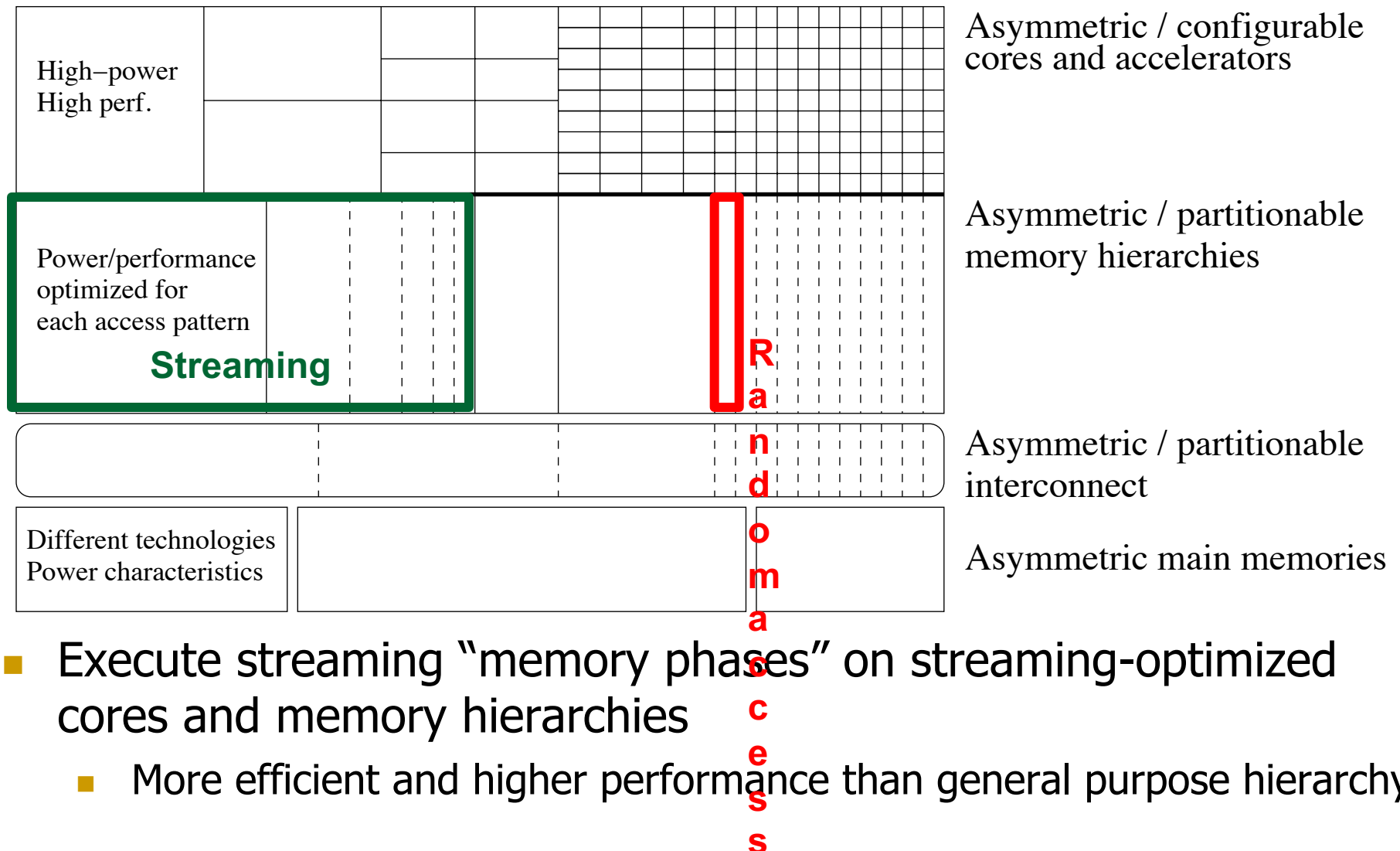
- Execute critical/serial sections on high-power, high-performance cores/resources [Suleman+ ASPLOS'09, ISCA'10, Top Picks'10'11, Joao+ ASPLOS'12, ISCA'13]
 - Programmer can write less optimized, but more likely correct programs

Exploiting Asymmetry: Simple Examples

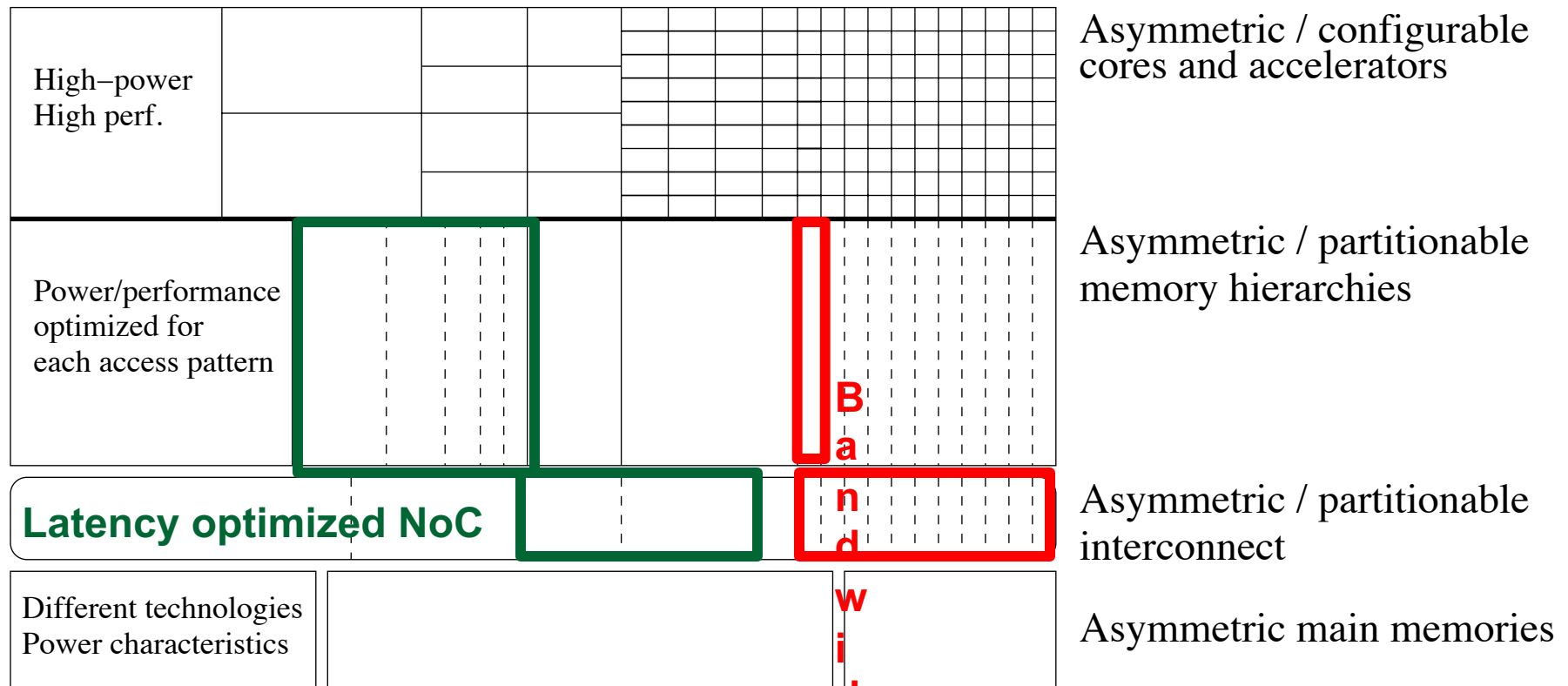


- Execute each code block on the most efficient execution backend for that block [Fallin+ ICCD'14]
 - Enables a much more efficient and still high performance core design

Exploiting Asymmetry: Simple Examples

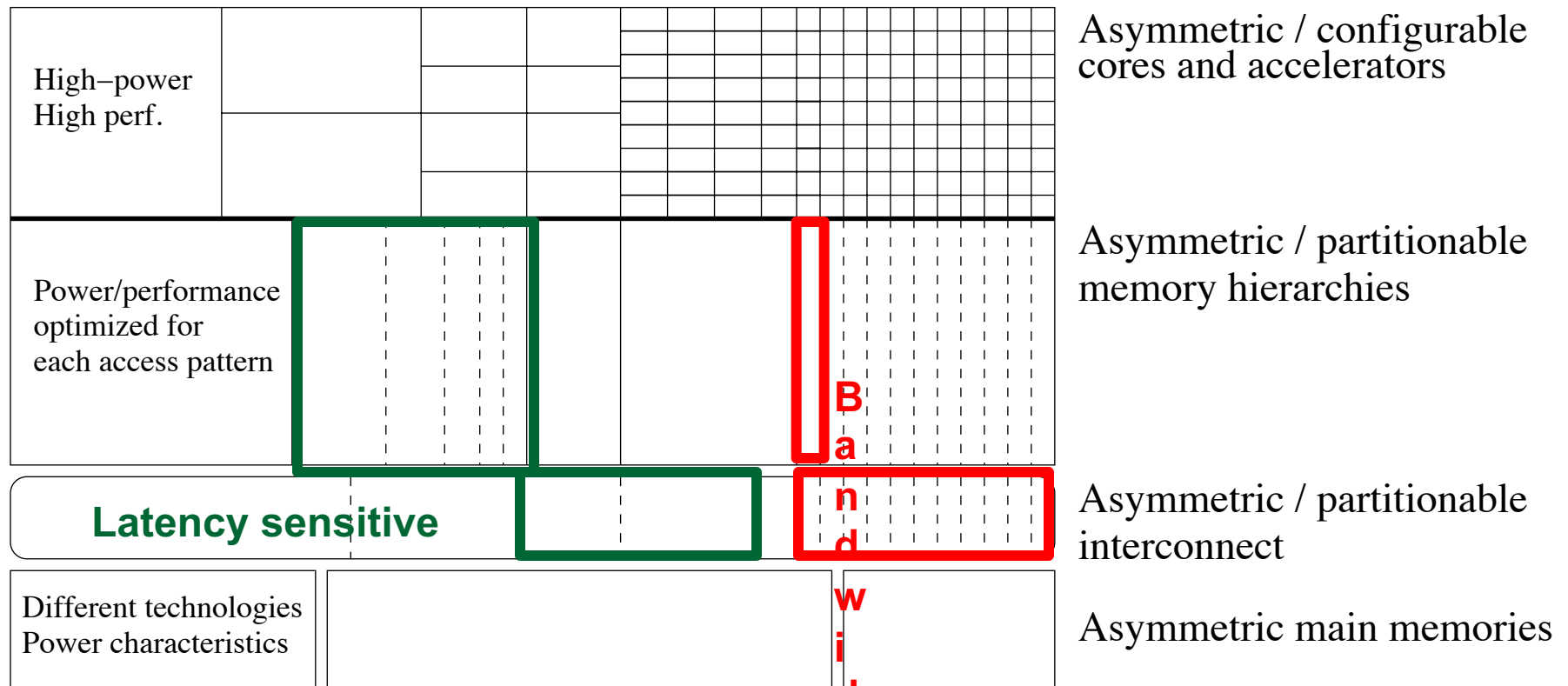


Exploiting Asymmetry: Simple Examples



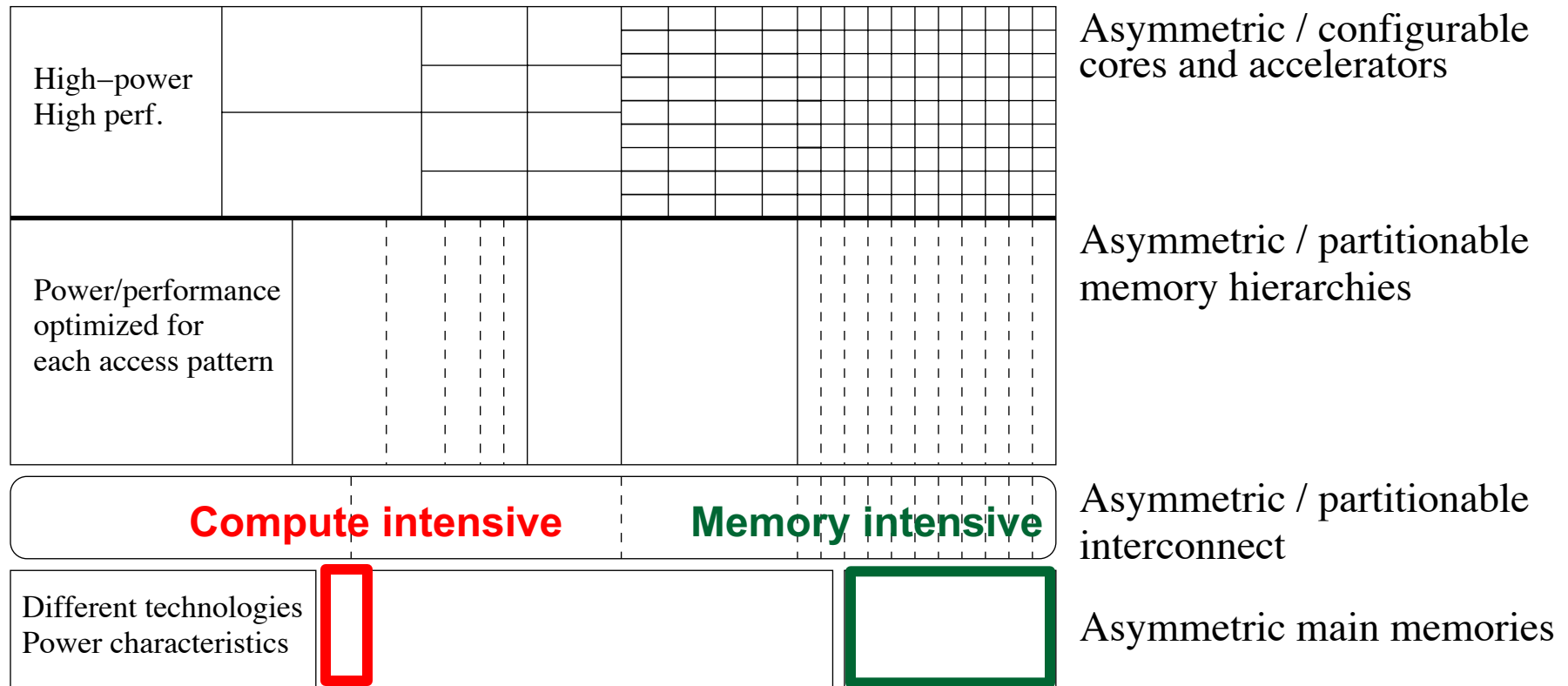
- Execute bandwidth-sensitive threads on a bandwidth-optimized network, latency-sensitive ones on a latency-optimized network [Das+ DAC'13]
 - Higher performance and energy-efficiency than a single network

Exploiting Asymmetry: Simple Examples



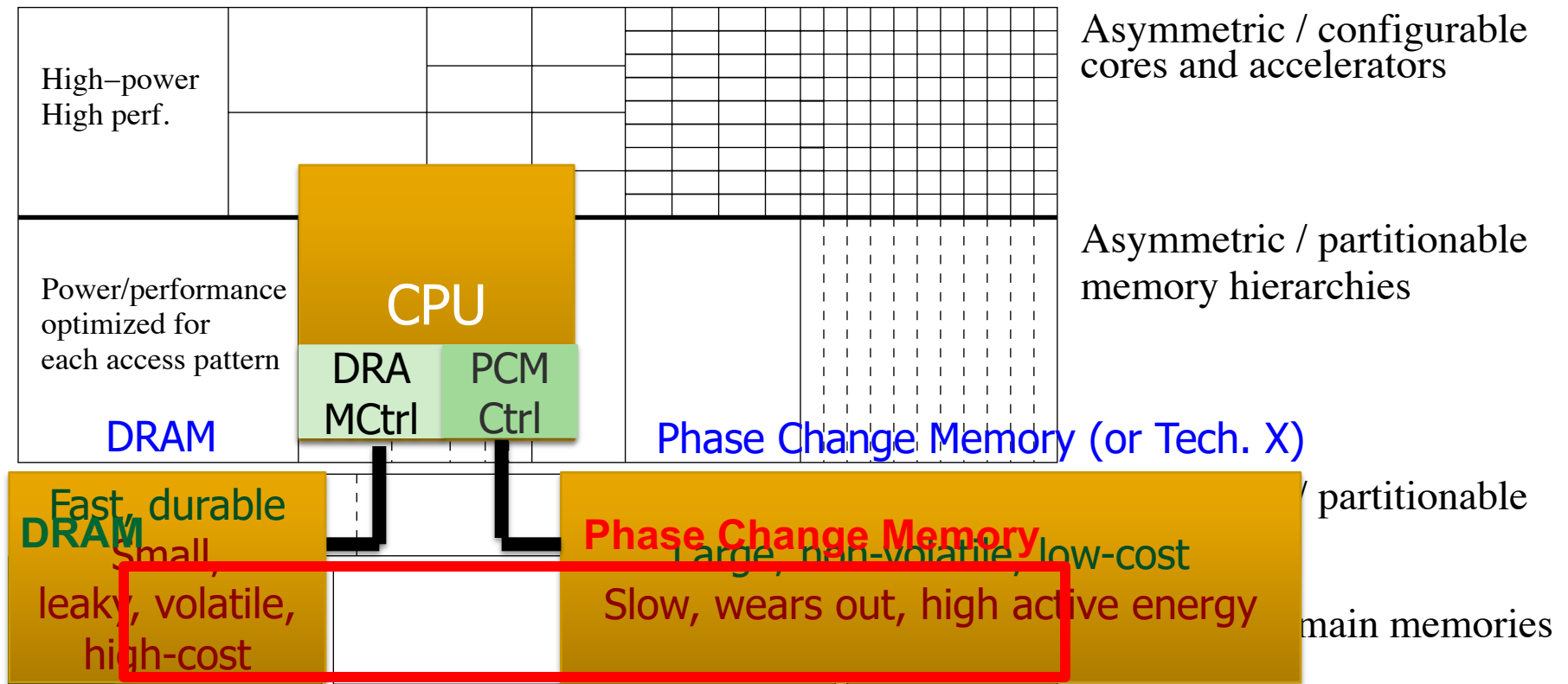
- Partition memory controller and on-chip network bandwidth asymmetrically among threads [Kim+ HPCA 2010, MICRO 2010, Top Picks 2011] [Nychis+ HotNets 2010] [Das+ MICRO 2009, ISCA 2010, Top Picks 2011]
 - Higher performance and energy-efficiency than symmetric/free-for-all

Exploiting Asymmetry: Simple Examples



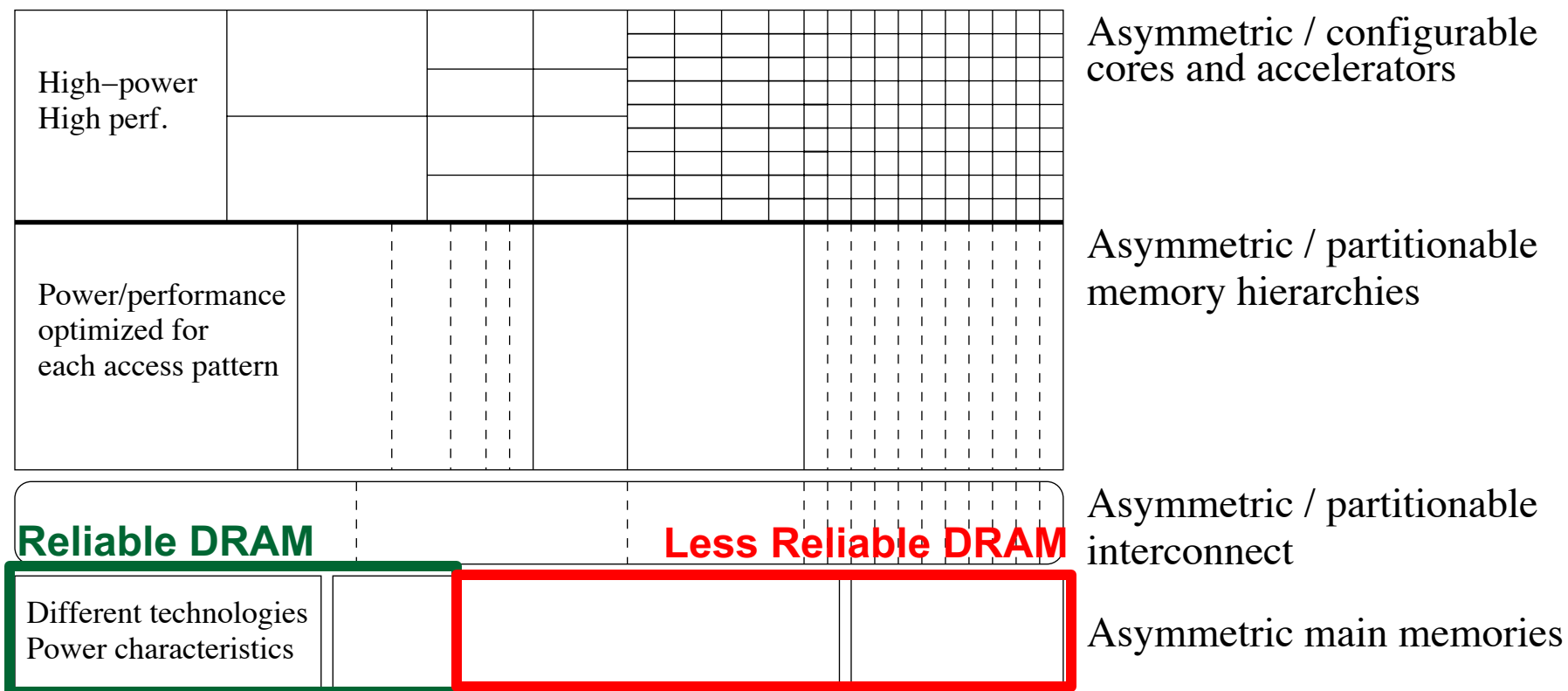
- Have multiple different memory scheduling policies apply them to different sets of threads based on thread behavior [Kim+ MICRO 2010, Top Picks 2011] [Ausavarungnirun+ ISCA 2012]
 - Higher performance and fairness than a homogeneous policy

Exploiting Asymmetry: Simple Examples



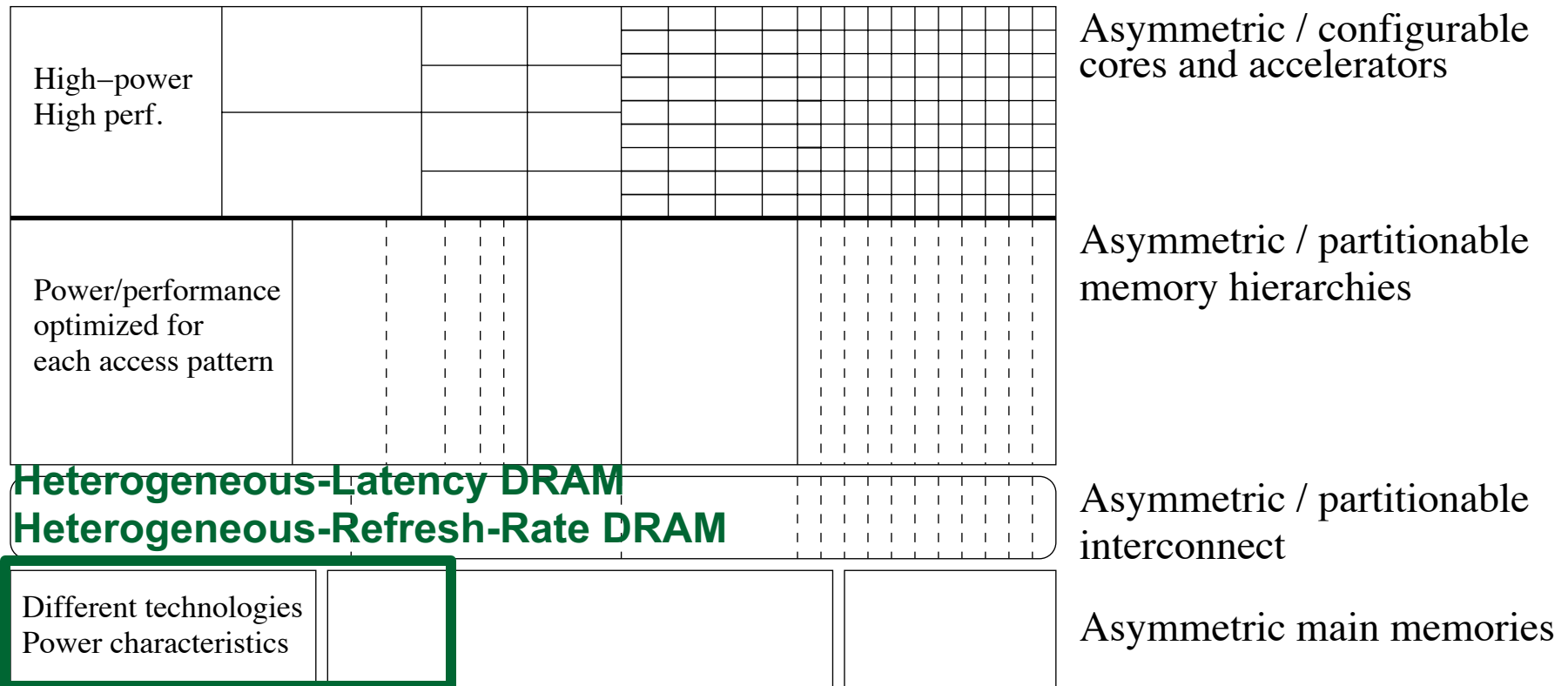
- Build main memory with different technologies with different characteristics (e.g., latency, bandwidth, cost, energy, reliability)
[Meza+ IEEE CAL'12, Yoon+ ICCD'12, Luo+ DSN'14]
 - Higher performance and energy-efficiency than homogeneous memory

Exploiting Asymmetry: Simple Examples



- Build main memory with different technologies with different characteristics (e.g., latency, bandwidth, cost, energy, reliability)
[Meza+ IEEE CAL'12, Yoon+ ICCD'12, Luo+ DSN'14]
 - Lower-cost than homogeneous-reliability memory at same availability

Exploiting Asymmetry: Simple Examples



- Design each memory chip to be heterogeneous to achieve low latency and low energy at reasonably low cost [Lee+ HPCA'13, Liu+ ISCA'12]
 - Higher performance and energy-efficiency than single-level memory

Memory Systems

Fundamentals, Recent Research, Challenges, Opportunities

Lecture 8: Asymmetric Multi-Core

Prof. Onur Mutlu

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<https://people.inf.ethz.ch/omutlu>

11 October 2018

Technion Fast Course 2018

Utility-Based Acceleration of Multithreaded Applications on Asymmetric CMPs

José A. Joao^{*}

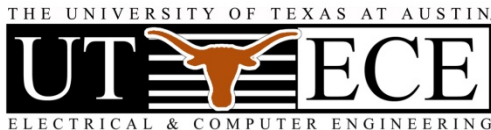
M. Aater Suleman^{*}

Onur Mutlu[‡]

Yale N. Patt^{*}

*** HPS Research Group
University of Texas at Austin**

**‡ Computer Architecture Laboratory
Carnegie Mellon University**



Carnegie Mellon

Asymmetric CMP (ACMP)

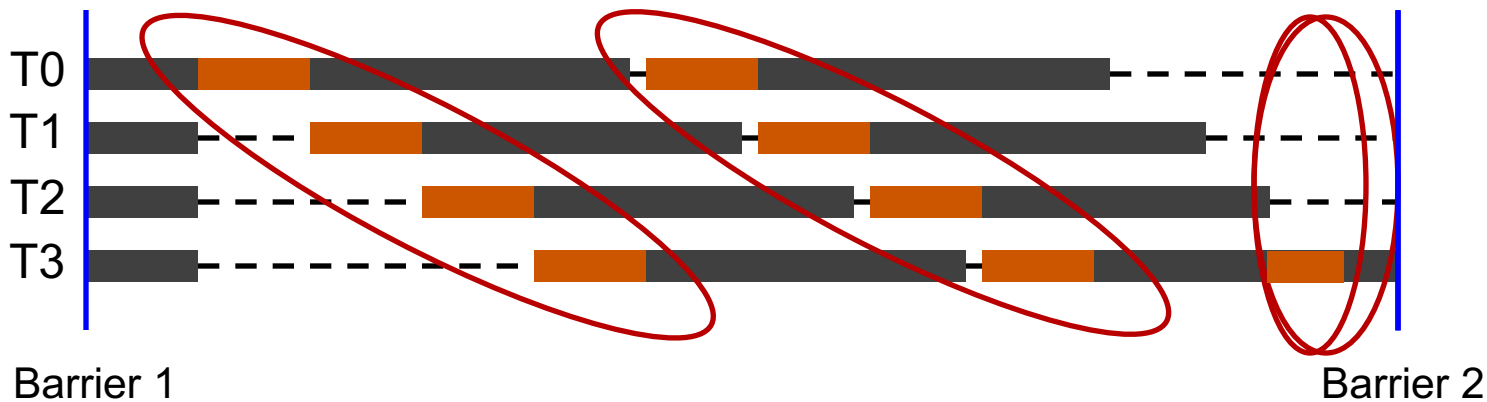
Small core	Small core	Large core	
Small core	Small core		
Small core	Small core	Small core	Small core
Small core	Small core	Small core	Small core

- One or a few large, out-of-order cores, fast
- Many small, in-order cores, power-efficient
- Critical code segments ^{???} run on large cores
- The rest of the code runs on small cores

Bottlenecks

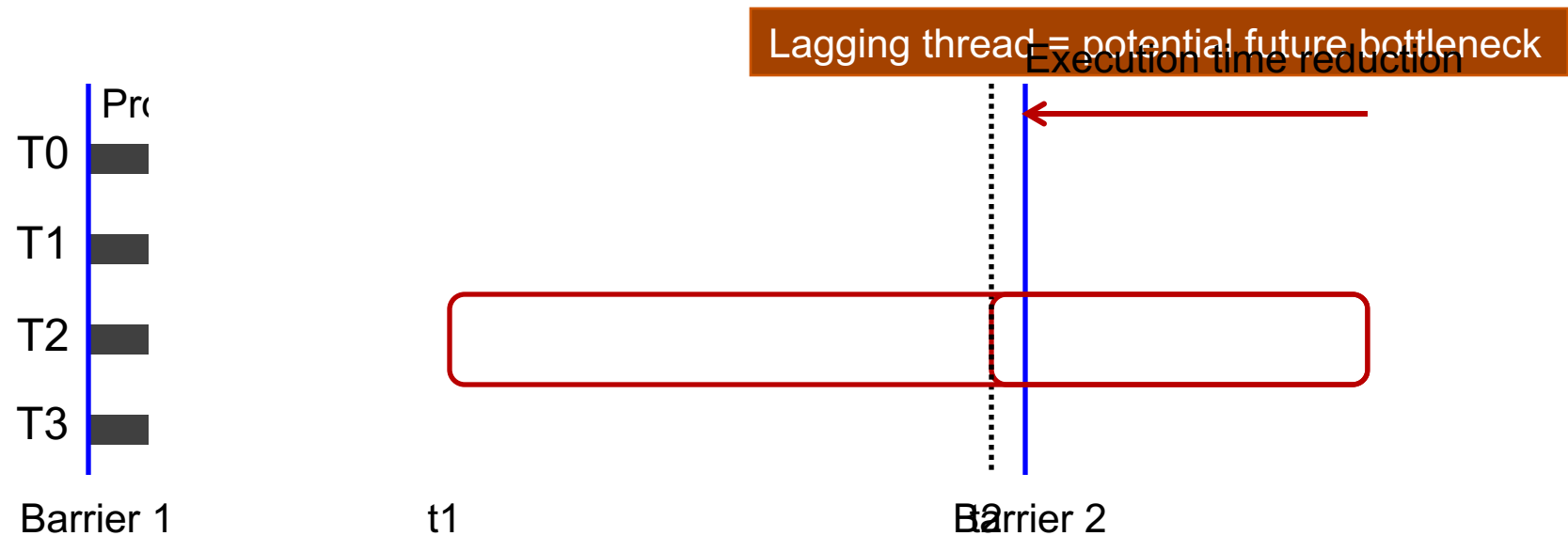


Accelerating Critical Sections (ACS), Suleman et al., ASPLOS' 09



Bottleneck Identification and Scheduling (BIS), Joao et al., ASPLOS' 12

Lagging Threads



T2: Lagging thread

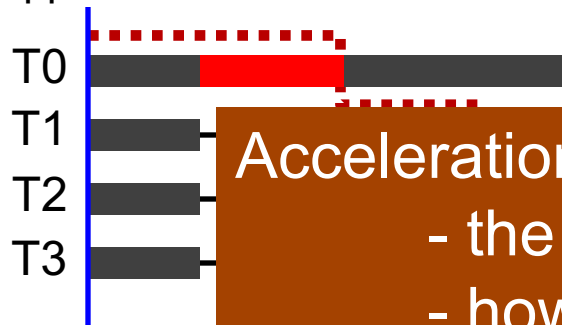
Previous work about progress of multithreaded applications:

- ❑ Meeting points, Cai et al., PACT' 08
- ❑ Thread criticality predictors, Bhattacharjee and Martonosi, ISCA' 09
- ❑ Age-based scheduling (AGETS), Lakshminarayana et al., SC' 09

Two problems

- 1) Do we accelerate bottlenecks or lagging threads?
- 2) Multiple applications: which application do we accelerate?

Application 1



Acceleration decisions need to consider both:

- the criticality of code segments
- how much speedup they get

Application



Utility-Based Acceleration (UBA)

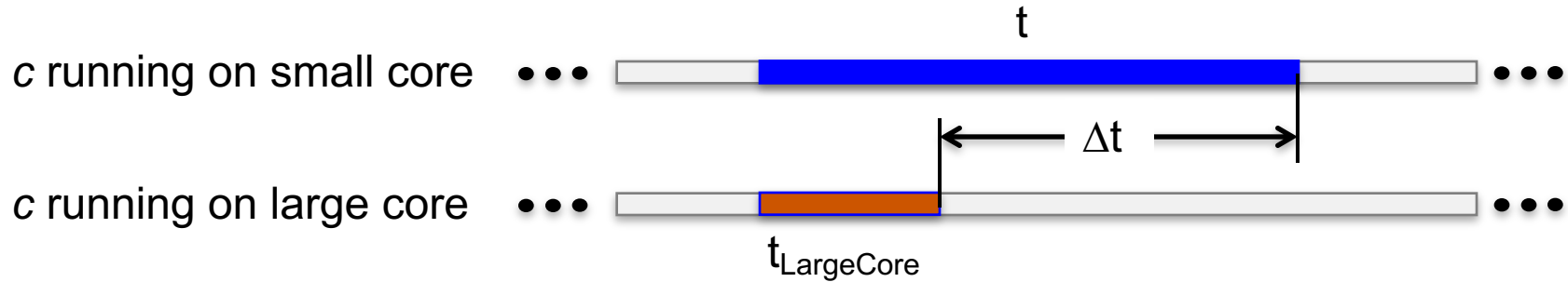
- *Goal*: identify performance-limiting bottlenecks or lagging threads from any running application and accelerate them on large cores of an ACMP
- *Key insight*: a **Utility of Acceleration** metric that combines speedup and criticality of each code segment
- Utility of accelerating code segment c of length t on an application of length T :

$$U_c = \frac{\Delta T}{T} = \underbrace{\left(\frac{\Delta t}{t} \right)}_L \times \underbrace{\left(\frac{t}{T} \right)}_R \times \underbrace{\left(\frac{\Delta T}{\Delta t} \right)}_G$$

L: Local acceleration of c

$$U_c = \textcircled{L} \times R \times G$$

How much code segment c is accelerated



$$L = \frac{\Delta t}{t} = \frac{t - t_{\text{LargeCore}}}{t} = 1 - \frac{1}{S}$$

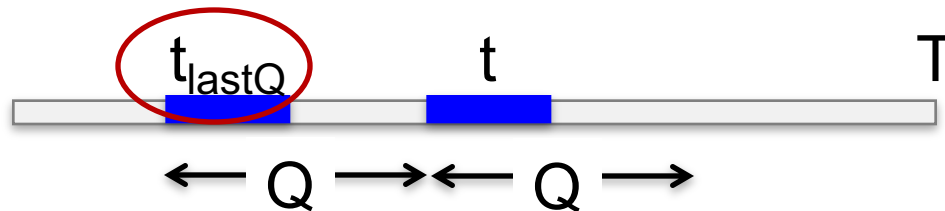
S: speedup of c

- Estimate S = estimate performance on a large core while running on a small core
- Performance Impact Estimation (PIE, Van Craeynest et al., ISCA' 12) : considers both instruction-level parallelism (ILP) and memory-level parallelism (MLP) to estimate CPI

R: Relevance of code segment c

$$U_c = L \times R \times G$$

How relevant code segment c is for the application



Q : scheduling quantum

$$R = \frac{t}{T}$$

$$R_{\text{estimated}} = \frac{t_{\text{last}Q}}{Q}$$

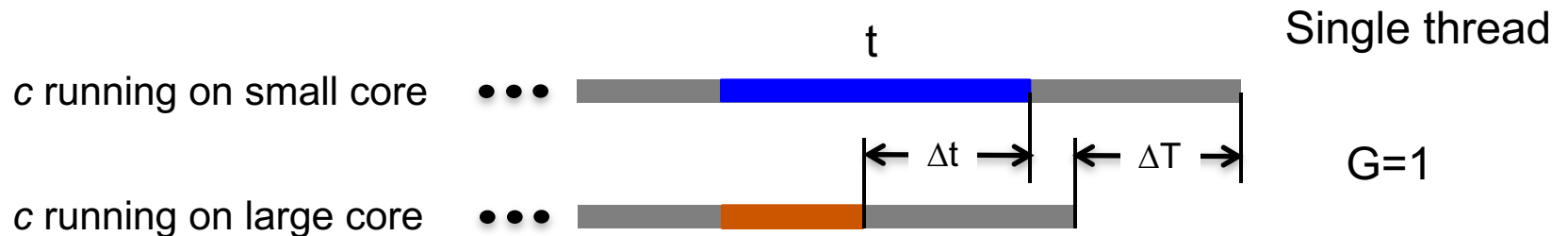
G: Global effect of accelerating c

How much accelerating c reduces total execution time

$$U_c = L \times R \times G$$

$$G = \frac{\Delta T}{\Delta t} \quad \begin{array}{l} \rightarrow \text{Acceleration of } \textit{application} \\ \rightarrow \text{Acceleration of } c \end{array}$$

Criticality of c



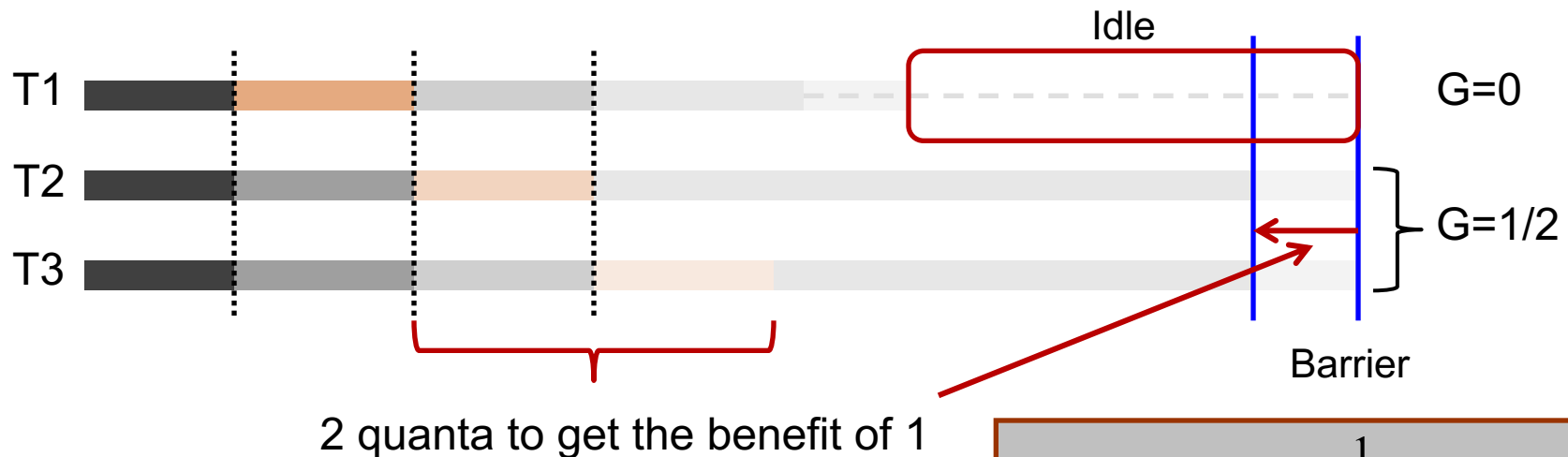
G: Global effect of accelerating c

How much accelerating c reduces total execution time

$$U_c = L \times R \times G$$

$$G = \frac{\Delta T}{\Delta t} \quad \begin{array}{l} \rightarrow \text{Acceleration of } application \\ \rightarrow \text{Acceleration of } c \end{array}$$

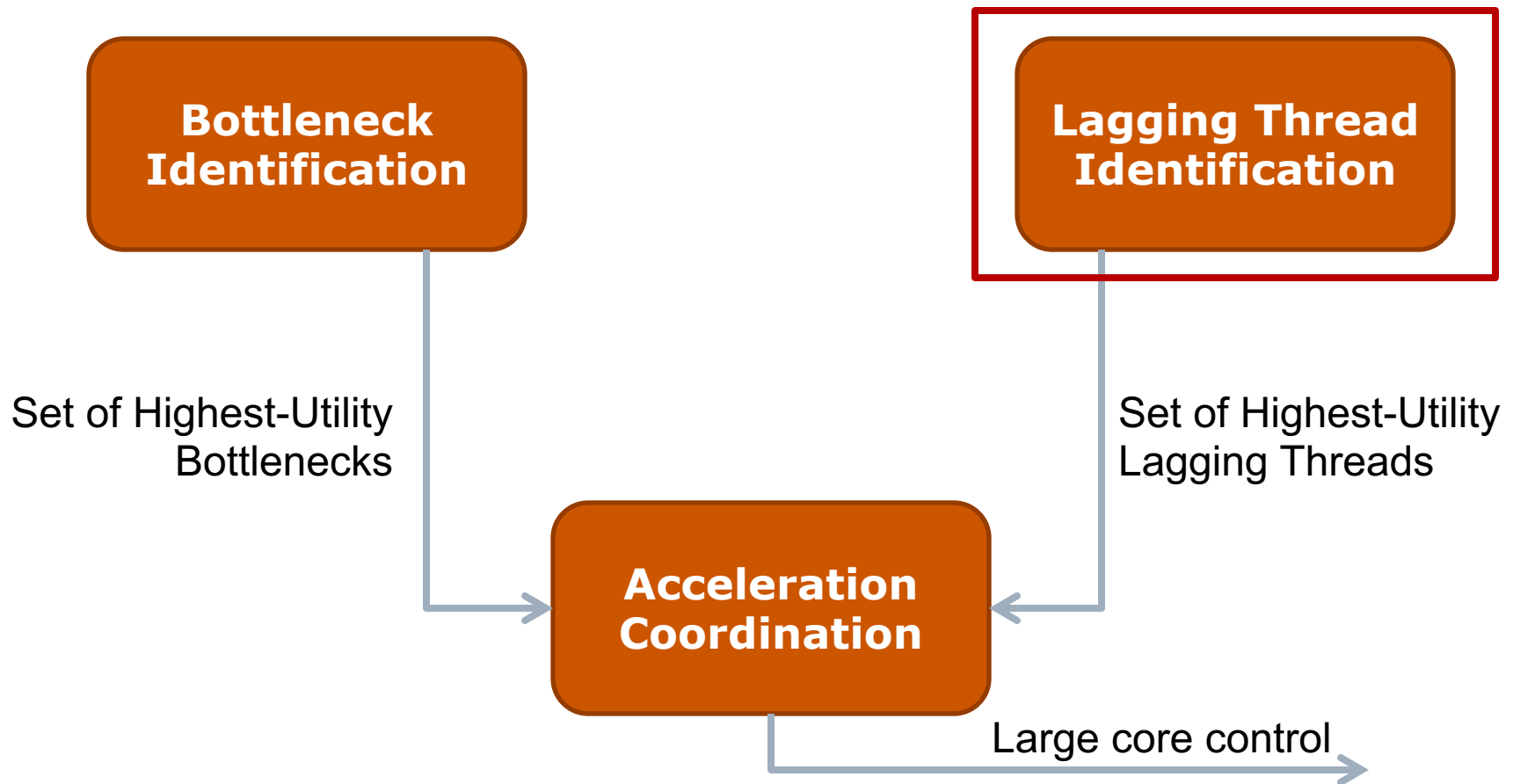
Criticality of c



$$G = \frac{1}{\text{Number of Lagging Threads}}$$

Critical sections: classify into *strongly-contended* and *weakly-contended* and estimate G differently (in the paper)

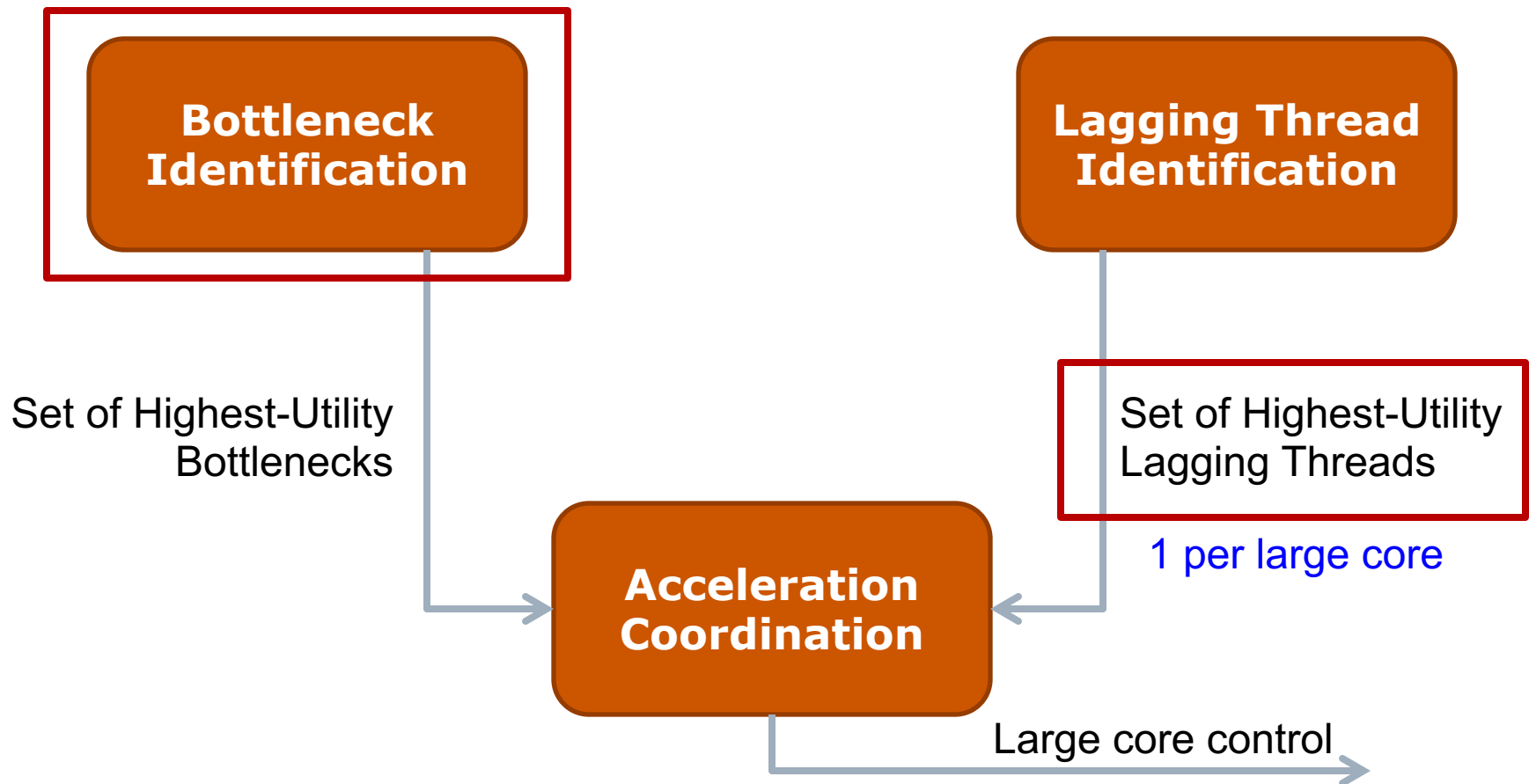
Utility-Based Acceleration (UBA)



Lagging thread identification

- Lagging threads are those that are making the least progress
- How to define and measure progress? → Application-specific problem
 - We borrow from Age-Based Scheduling (SC' 09)
 - Progress metric (committed instructions)
 - Assumption: same number of committed instructions between barriers
 - But we could easily use any other progress metric...
- Minimum progress = $\min P$
- Set of lagging threads = $\{ \text{any thread with progress} < \min P + \Delta P \}$
- Compute Utility for each lagging thread

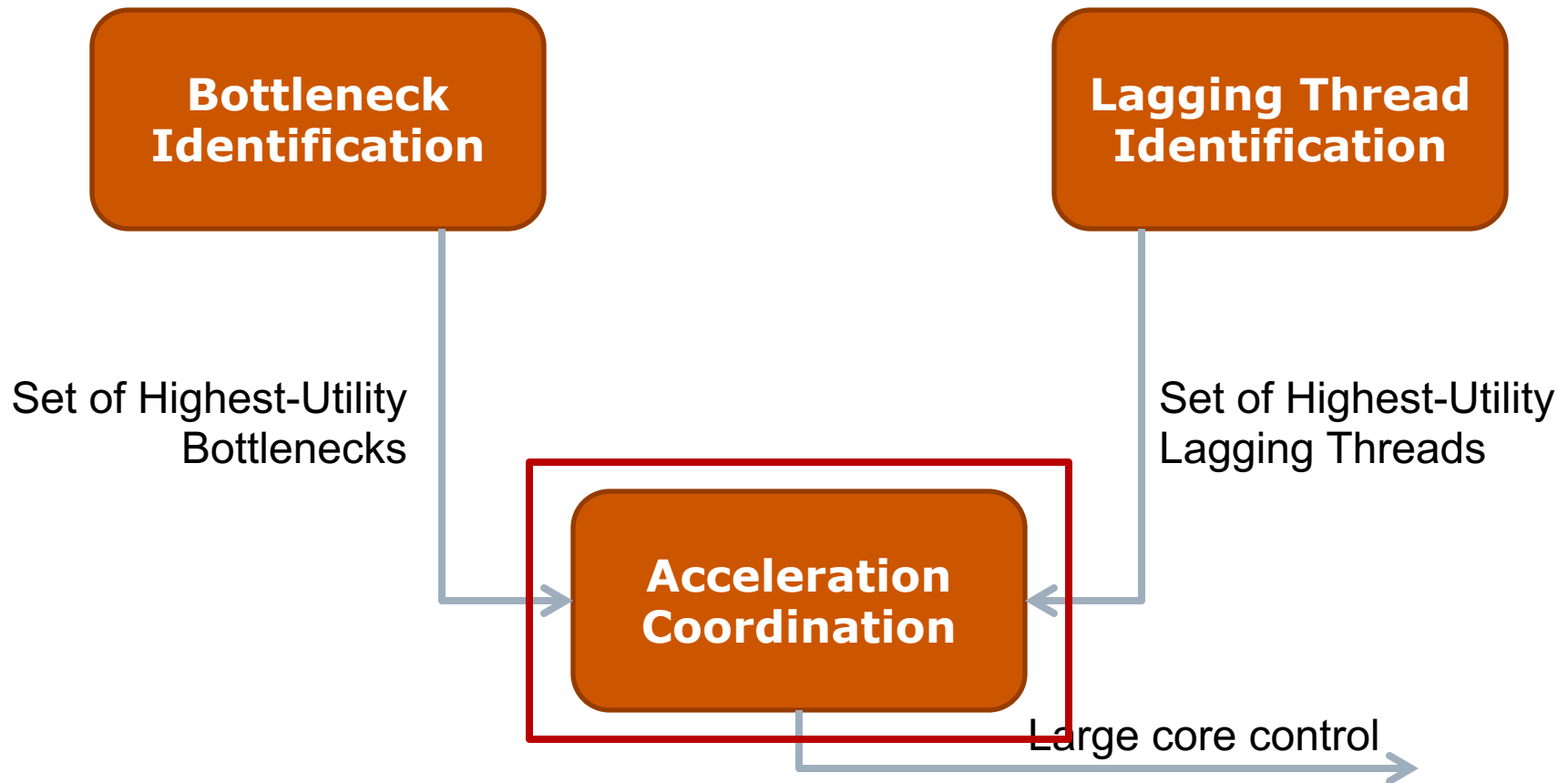
Utility-Based Acceleration (UBA)



Bottleneck identification

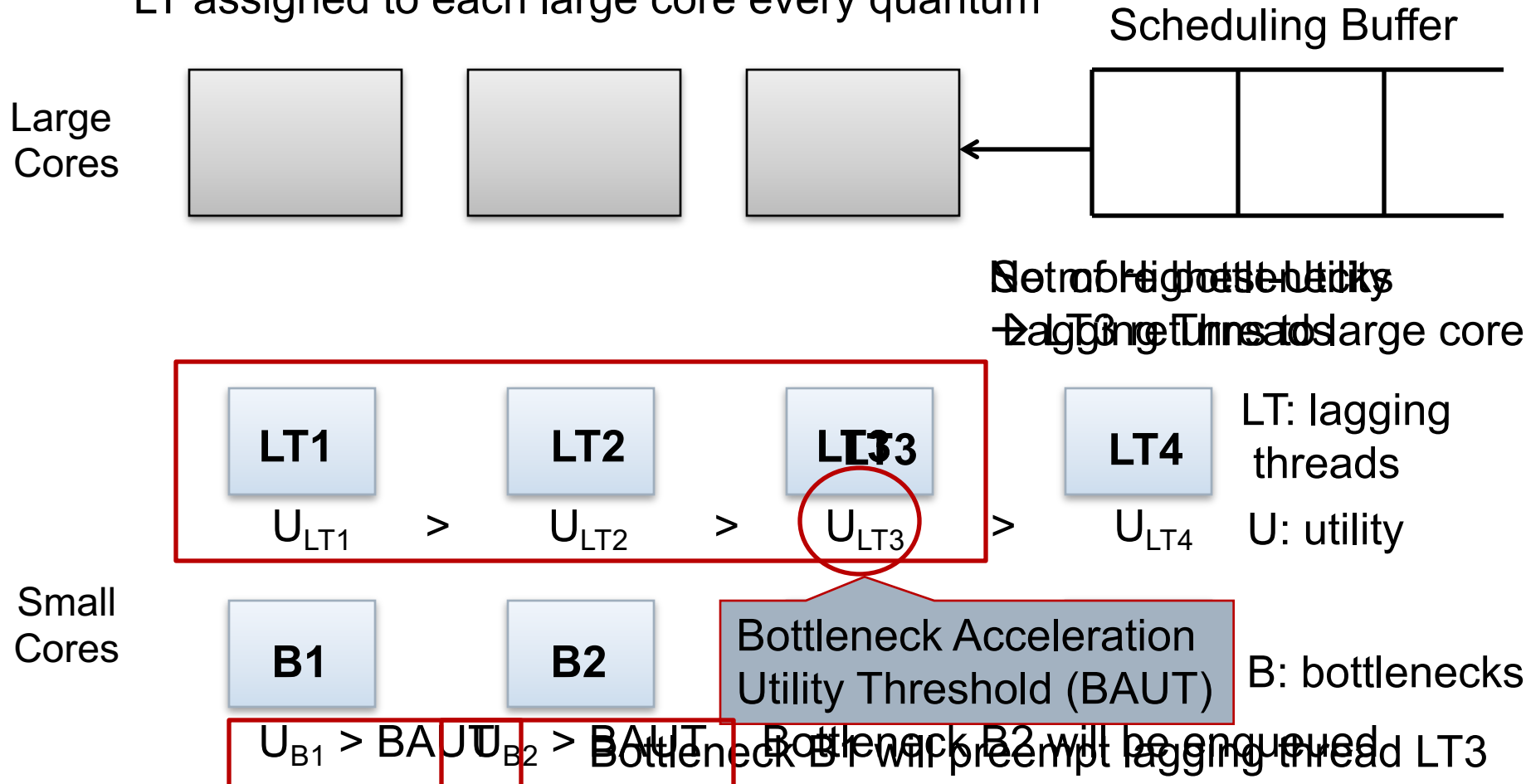
- Software: programmer, compiler or library
 - Delimit potential bottlenecks with **BottleneckCall** and **BottleneckReturn** instructions
 - Replace code that waits with a **BottleneckWait** instruction
- Hardware: Bottleneck Table
 - Keep track of threads executing or waiting for bottlenecks
 - Compute Utility for each bottleneck
 - Determine set of Highest-Utility Bottlenecks
- Similar to our previous work BIS, ASPLOS' 12
 - BIS uses *thread waiting cycles* instead of Utility

Utility-Based Acceleration (UBA)



Acceleration coordination

LT assigned to each large core every quantum



Methodology

■ Workloads

- single-application: 9 multithreaded applications with different impact from bottlenecks
- 2-application: all 55 combinations of (9 MT + 1 ST)
- 4-application: 50 random combinations of (9 MT + 1 ST)

■ Processor configuration

- x86 ISA
- Area of large core = 4 x Area of small core
- Large core: 4GHz, out-of-order, 128-entry ROB, 4-wide, 12-stage
- Small core: 4GHz, in-order, 2-wide, 5-stage
- Private 32KB L1, private 256KB L2, shared 8MB L3
- On-chip interconnect: Bi-directional ring, 2-cycle hop latency

Comparison points

■ Single application

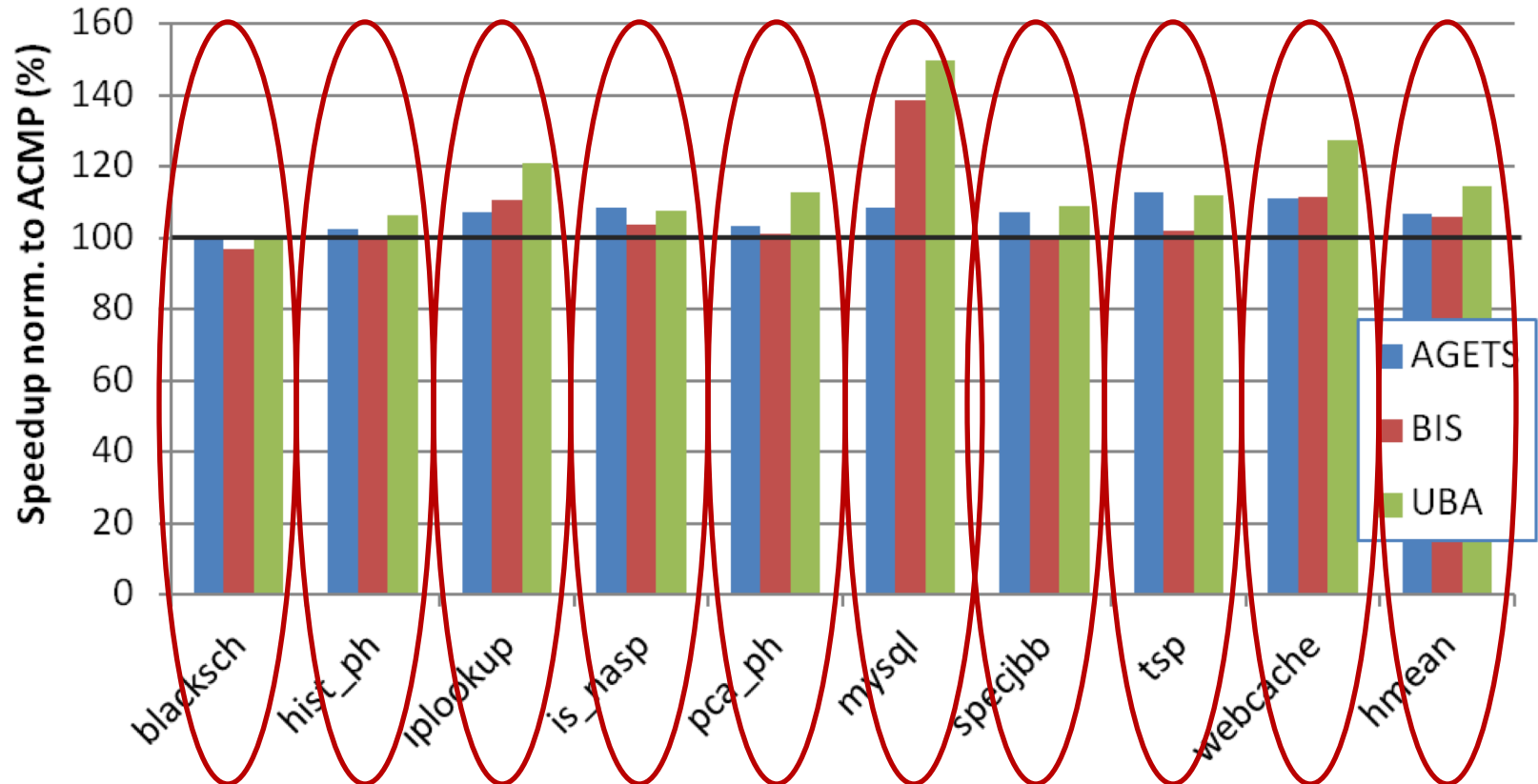
- **ACMP** (Morad et al., Comp. Arch. Letters'06)
 - only accelerates Amdahl's serial bottleneck
- Age-based scheduling (**AGETS**, Lakshminarayana et al., SC'09)
 - only accelerates lagging threads
- Bottleneck Identification and Scheduling (**BIS**, Joao et al., ASPLOS'12)
 - only accelerates bottlenecks

■ Multiple applications

- **AGETS+PIE**: select most lagging thread with AGETS and use PIE across applications
 - only accelerates lagging threads
- **MA-BIS**: BIS with shared large cores across applications
 - only accelerates bottlenecks

Single application, 1 large core

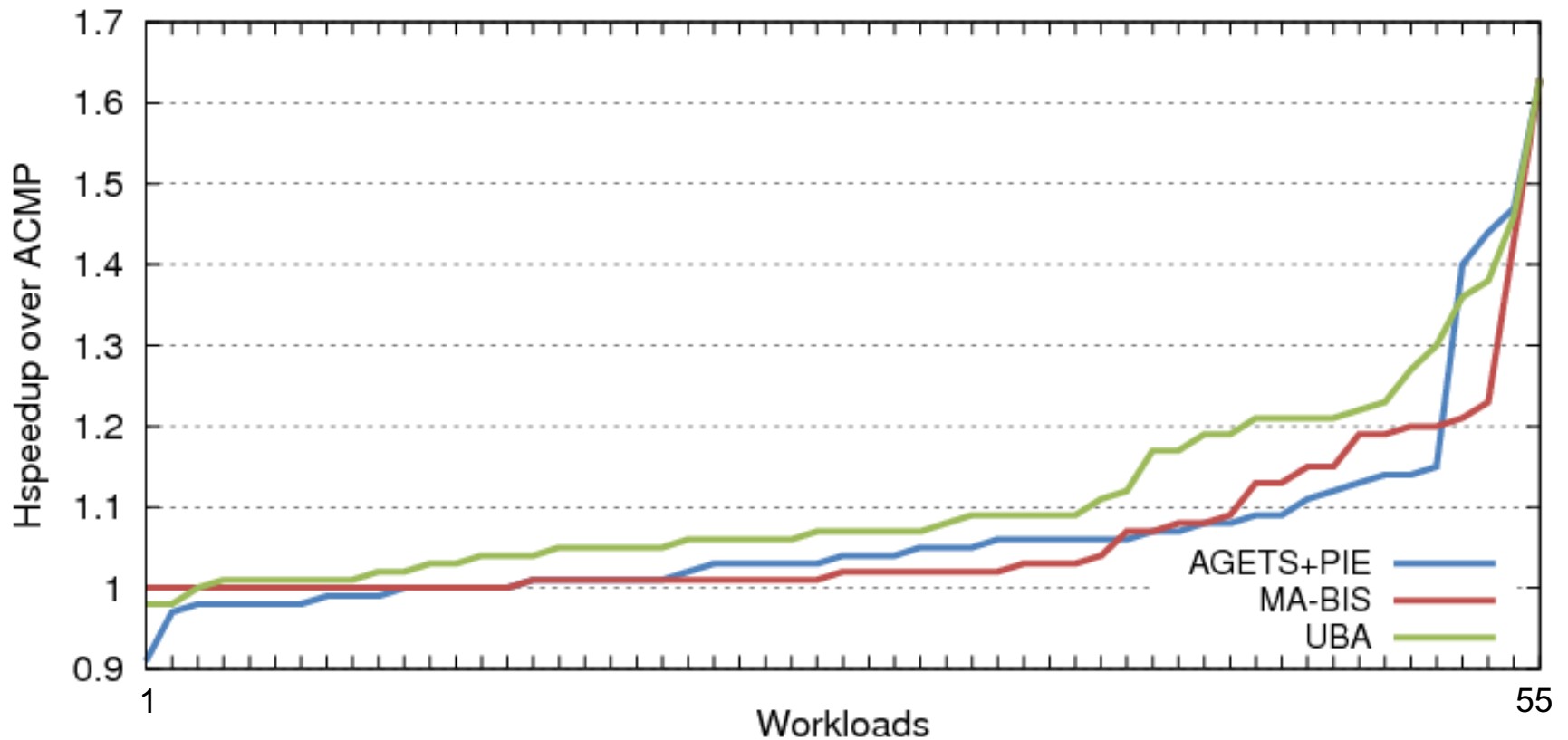
Optimal number of threads, 28 small cores, 1 large core



UBA outperforms both AGETS and BIS by 8%
Neither bottlenecked by small sections. Both AGETS and UBA
UBA's benefit increases with area budget and number of large cores
no lagging threads

Multiple applications

2-application workloads, 60 small cores, 1 large core



UBA improves Hspeedup over AGETS+PIE and MA-BIS by 2 to 9%

Summary

- To effectively use ACMPs:
 - Accelerate both fine-grained bottlenecks and lagging threads
 - Accelerate single and multiple applications
- Utility-Based Acceleration (UBA) is a cooperative software-hardware solution to both problems
- Our [Utility of Acceleration metric](#) combines a measure of acceleration and a measure of criticality to allow meaningful comparisons between code segments
- Utility is implemented for an ACMP but is general enough to be extended to other acceleration mechanisms
- UBA outperforms previous proposals for single applications and their aggressive extensions for multiple-application workloads
- UBA is a comprehensive fine-grained acceleration proposal for parallel applications without programmer effort

Thank You!

Questions?

Utility-Based Acceleration of Multithreaded Applications on Asymmetric CMPs

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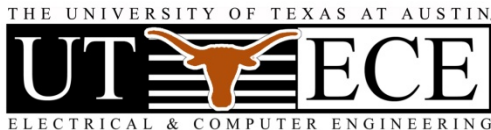
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