

NON-BLOCKING DATA STRUCTURES AND TRANSACTIONAL MEMORY

Tim Harris, 3 Nov 2017

Lecture 1/3

- Introduction
- Basic spin-locks
- Queue-based locks
- Hierarchical locks
- Reader-writer locks
- Reading without locking
- Flat combining
- Recent research: parallel work distribution

Overview

- Building shared memory data structures
 - Lists, queues, hashtables, ...
- Why?
 - Used directly by applications (e.g., in C/C++, Java, C#, ...)
 - Used in the language runtime system (e.g., management of work, implementations of message passing, ...)
 - Used in traditional operating systems (e.g., synchronization between top/bottom-half code)
- Why not?
 - Don't think of "threads + shared data structures" as a default/good/complete/desirable programming model
 - It's better to have shared memory and not need it...

Different techniques for different problems

Ease to
write

Correctness

When can it
be used?

How fast is it?

How well
does it scale?

Suggested reading

- “The art of multiprocessor programming”, Herlihy & Shavit
– excellent coverage of shared memory data structures,
from both practical and theoretical perspectives

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- Recent research papers on in-memory databases built from concurrent data structures – search for SILO, and work from Haibo Chen and colleagues
- “Transactional memory, 2nd edition”, Harris, Larus, Rajwar – survey of transactional memory work, with 350+ references
- “NOrec: streamlining STM by abolishing ownership records”, Dalessandro, Spear, Scott, PPOPP 2010

Basic spin-locks

Test and set (pseudo-code)

```
bool testAndSet(bool *b) {  
    bool result;  
    atomic {  
        result = *b;  
        *b = TRUE;  
    }  
    return result;  
}
```

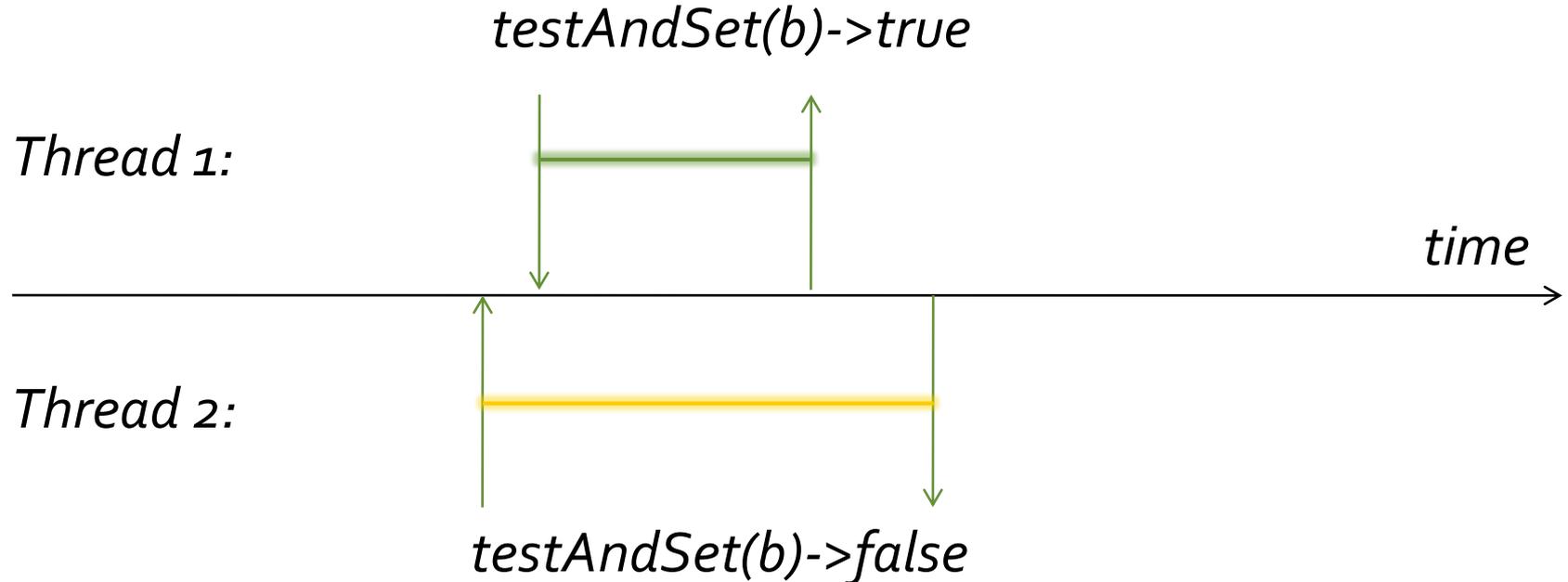
Pointer to a location holding a boolean value (TRUE/FALSE)

Read the current contents of the location b points to...

...set the contents of *b to TRUE

Test and set

- Suppose two threads use it at once



Test and set lock

lock:

FALSE

FALSE => lock available
TRUE => lock held

```
void acquireLock(bool *lock) {  
    while (testAndSet(lock)) {  
        /* Nothing */  
    }  
}
```

Each call tries to acquire the lock, returning TRUE if it is already held

```
void releaseLock(bool *lock) {  
    *lock = FALSE;  
}
```

NB: all this is pseudo-code, assuming SC memory

Test and set lock

lock:

TRUE

```
void acquireLock(bool *lock) {  
    while (testAndSet(lock)) {  
        /* Nothing */  
    }  
}
```

```
void releaseLock(bool *lock) {  
    *lock = FALSE;  
}
```

Thread 1



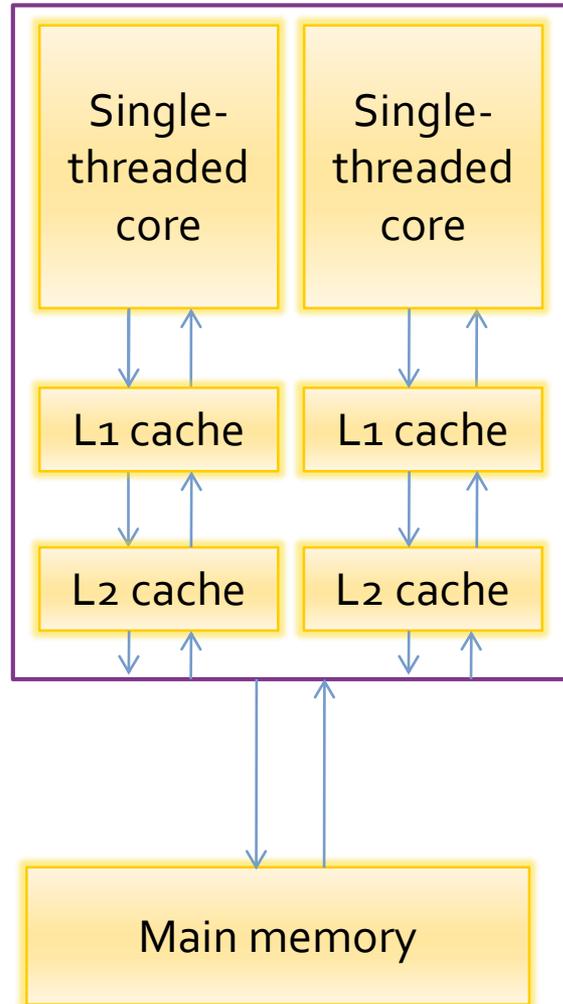
Thread 2



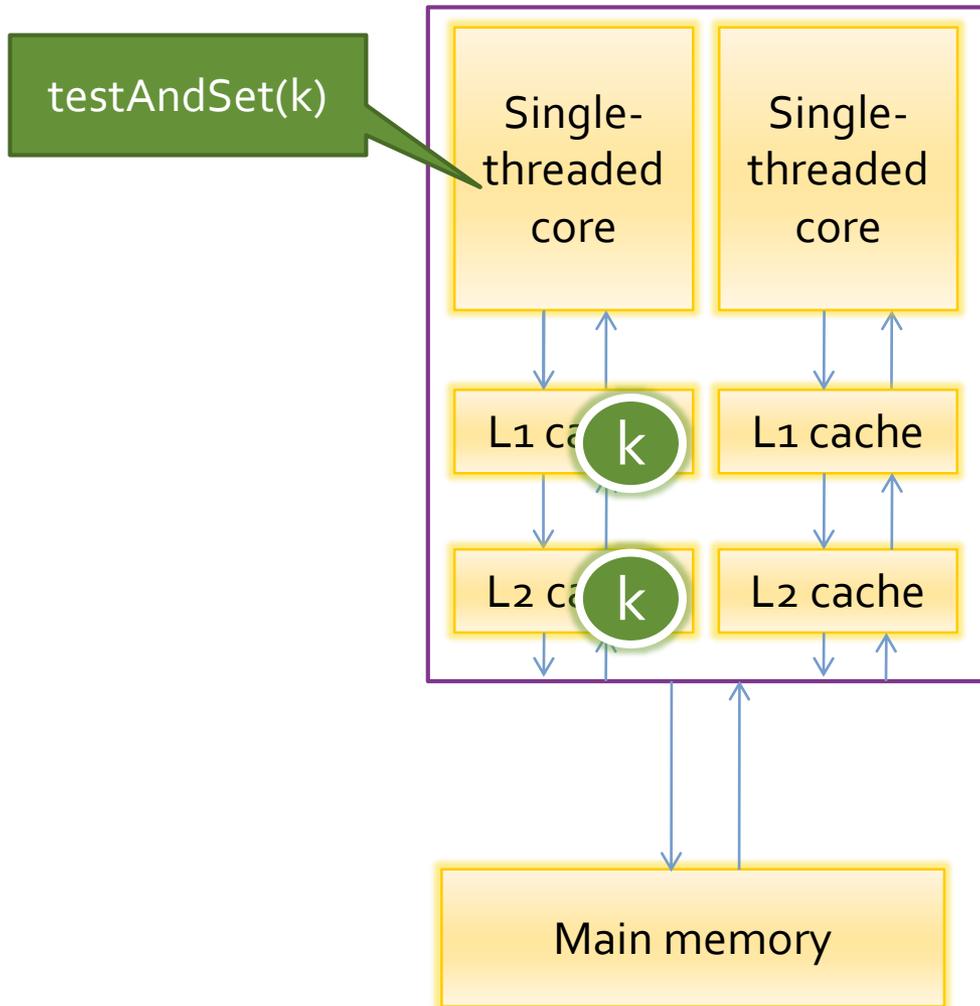
What are the problems here?

testAndSet
implementation
causes contention

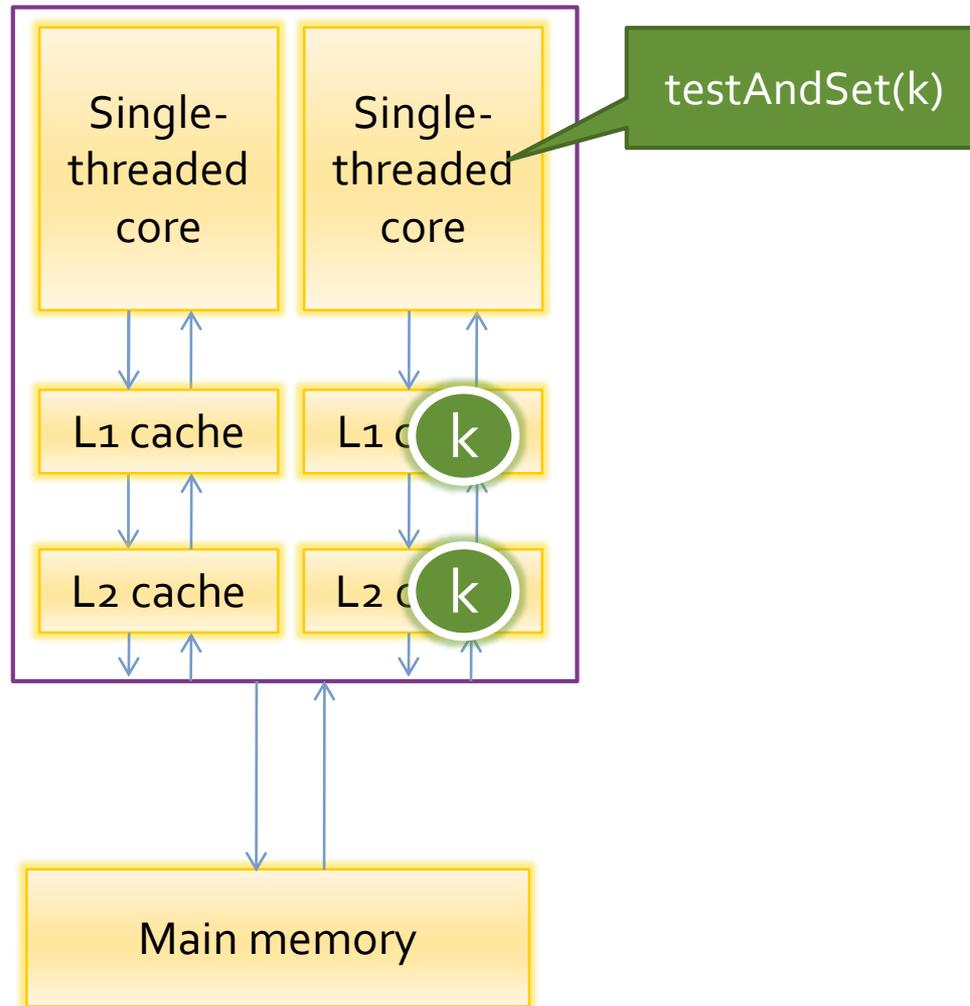
Contention from testAndSet



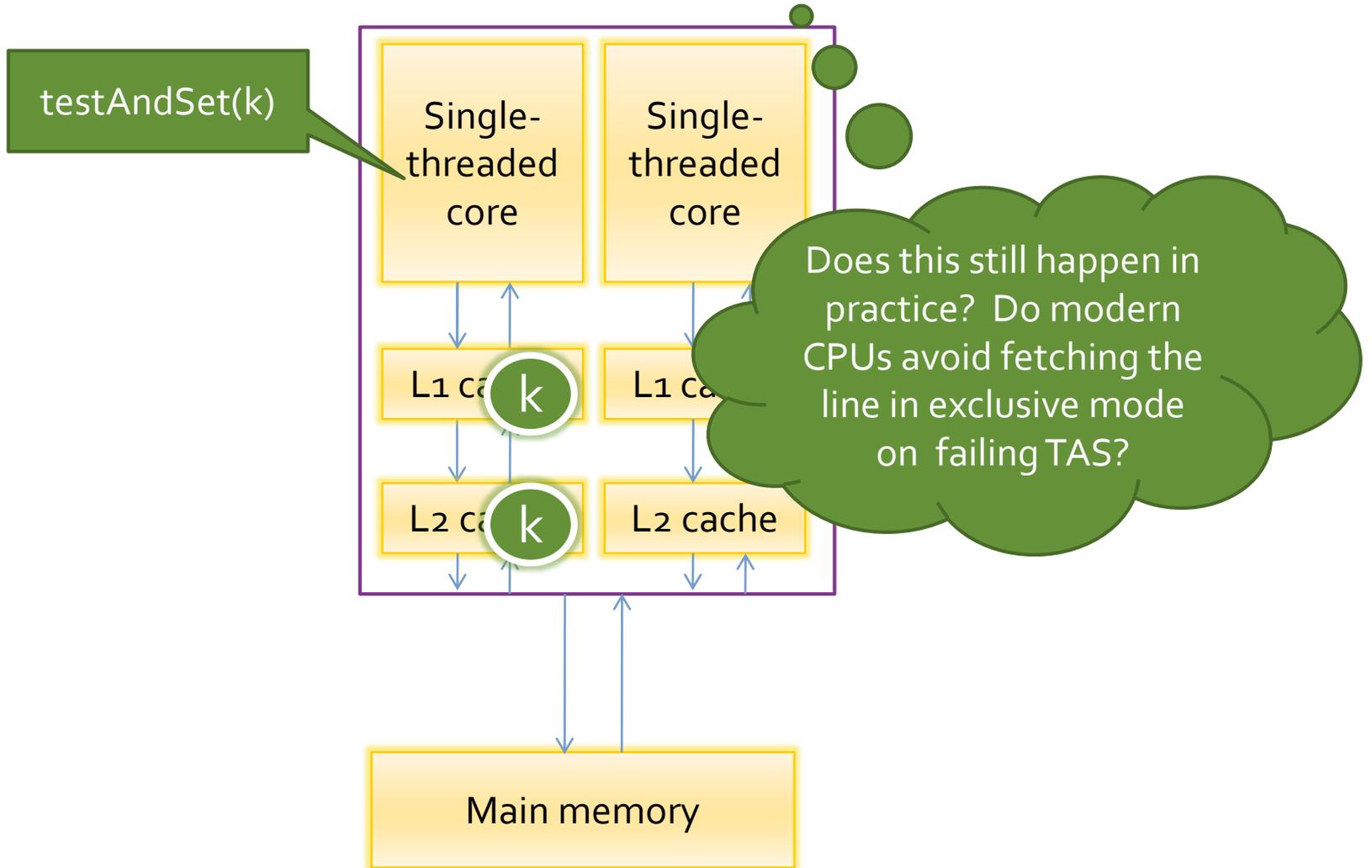
Multi-core h/w – separate L2



Multi-core h/w – separate L2



Multi-core h/w – separate L2



What are the problems here?

testAndSet
implementation
causes contention

No control over
locking policy

Only supports mutual
exclusion: not reader-
writer locking

Spinning may waste
resources while
waiting

General problem

- No logical conflict between two failed lock acquires
- Cache protocol introduces a physical conflict
- For a good algorithm: only introduce physical conflicts if a logical conflict occurs
 - In a lock: successful lock-acquire & failed lock-acquire
 - In a set: successful insert(10) & failed insert(10)
- But not:
 - In a lock: two failed lock acquires
 - In a set: successful insert(10) & successful insert(20)
 - In a non-empty queue: enqueue on the left and remove on the right

Test and test and set lock

lock:

FALSE

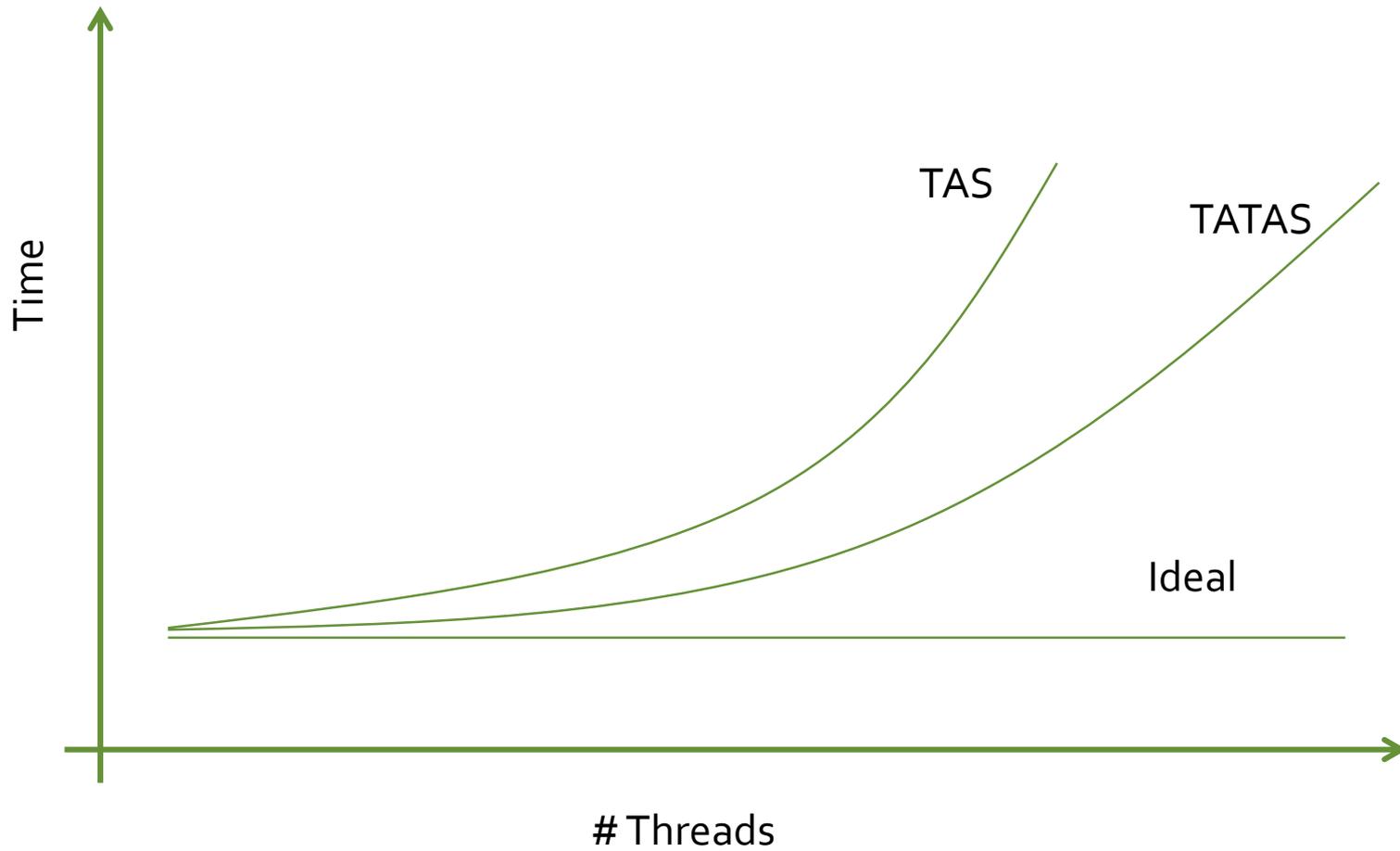
FALSE => lock available
TRUE => lock held

```
void acquireLock(bool *lock) {  
  do {  
    while (*lock) { }  
  } while (testAndSet(lock));  
}
```

Spin while the lock is
held... only do
testAndSet when it is
clear

```
void releaseLock(bool *lock) {  
  *lock = FALSE;  
}
```

Performance



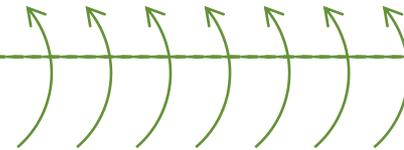
Stampedes

lock:

TRUE

```
void acquireLock(bool *lock) {  
  do {  
    while (*lock) {}  
  } while (testAndSet(lock));  
}
```

```
void releaseLock(bool *lock) {  
  *lock = FALSE;  
}
```



Back-off algorithms

1. Start by spinning, watching the lock for “s” iterations
2. If the lock does not become free, wait locally for “w” (*without watching the lock*)

What should “s” be?
What should “w” be?

Time spent spinning on the lock “s”

- Lower values:
 - Less time to build up a set of threads that will stampede
 - Less contention in the memory system, if remote reads incur a cost
 - Risk of a delay in noticing when the lock becomes free if we are not watching
- Higher values:
 - Less likelihood of a delay between a lock being released and a waiting thread noticing

Local waiting time “w”

- Lower values:
 - More responsive to the lock becoming available
- Higher values:
 - If the lock doesn't become available then the thread makes fewer accesses to the shared variable

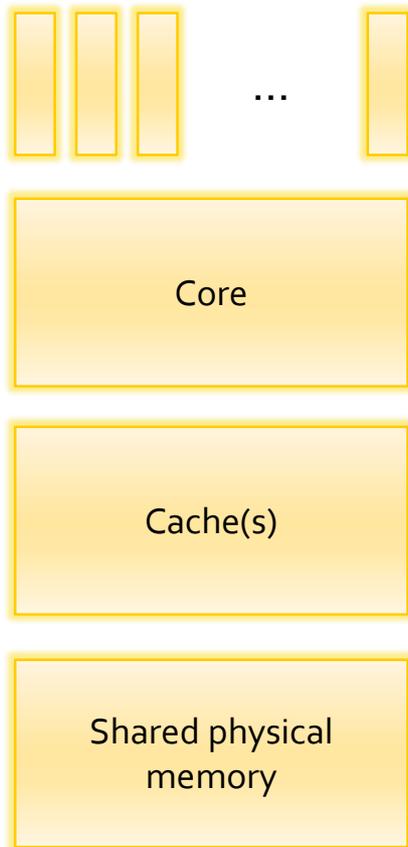
Methodical approach

- For a given workload and performance model:
 - What is the best that could be done (i.e. given an “oracle” with perfect knowledge of when the lock becomes free)?
 - How does a practical algorithm compare with this?
- Look for an algorithm with a bound between its performance and that of the oracle
- “Competitive spinning”

Rule of thumb

- Spin on the lock for a duration that's comparable with the shortest back-off interval
- Exponentially increase the per-thread back-off interval (resetting it when the lock is acquired)
- Use a maximum back-off interval that is large enough that waiting threads don't interfere with the other threads' performance

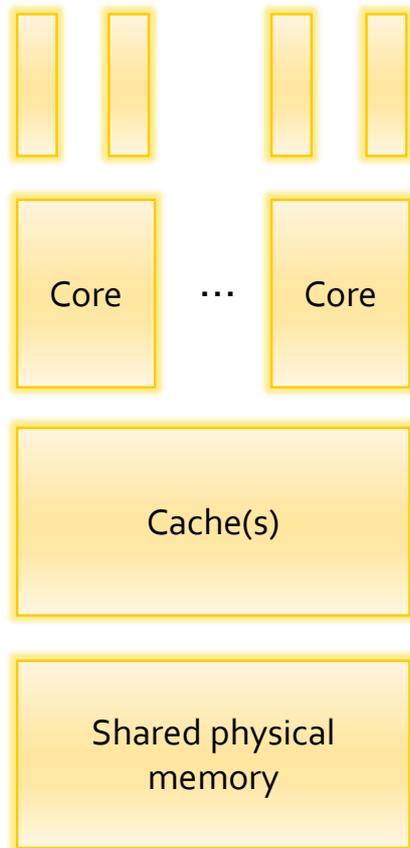
Systems problems



Lots of h/w threads
multiplexed over a core

- The threads need to “wait efficiently”
- Not consuming processing resources (contending with lock holder) & not consuming power
- “monitor” / “mwait” operations – e.g., SPARC M7

Systems problems



S/W threads multiplexed
on cores

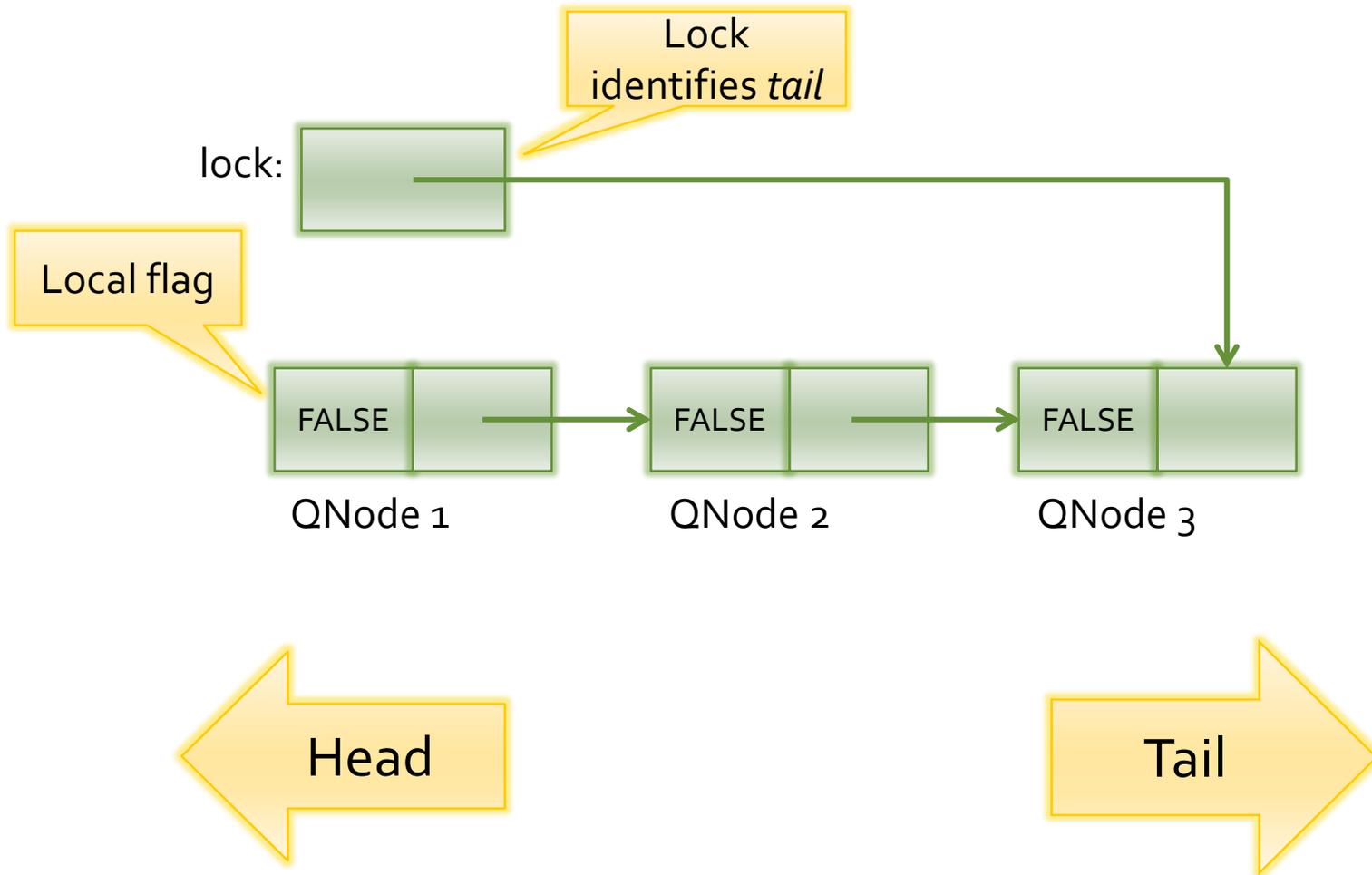
- Spinning gets in the way of other s/w threads, even if done efficiently
- For long delays, may need to actually block and unblock
- ...as with back-off, how long to spin for before blocking?

Queue-based locks

Queue-based locks

- Lock holders queue up: immediately provides FCFS behavior
- Each spins *locally* on a flag in their queue entry: no remote memory accesses while waiting
- A lock release wakes the next thread directly: no stampede

MCS locks



MCS lock acquire

lock:



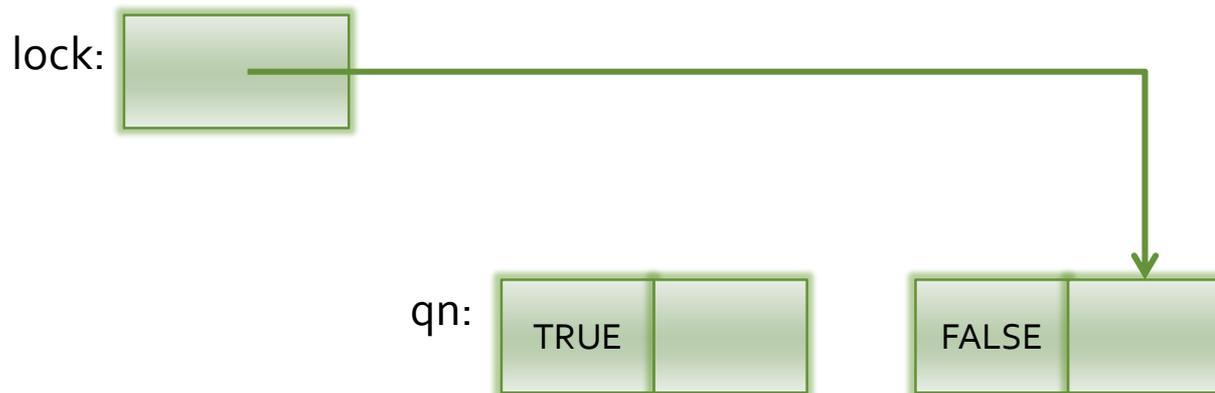
```
void acquireMCS(mcs *lock, QNode *qn) {  
    QNode *prev;  
    qn->flag = false;  
    qn->next = NULL;  
    while (true) {  
        prev = lock->tail;  
        /* Label 1 */  
        if (CAS(&lock->tail, prev, qn)) break;  
    }  
    if (prev != NULL) {  
        prev->next = qn; /* Label 2 */  
        while (!qn->flag) { } // Spin  
    }  
}
```

Find previous
tail node

Atomically replace
"prev" with "qn" in
the lock itself

Add link within
the queue

MCS lock release



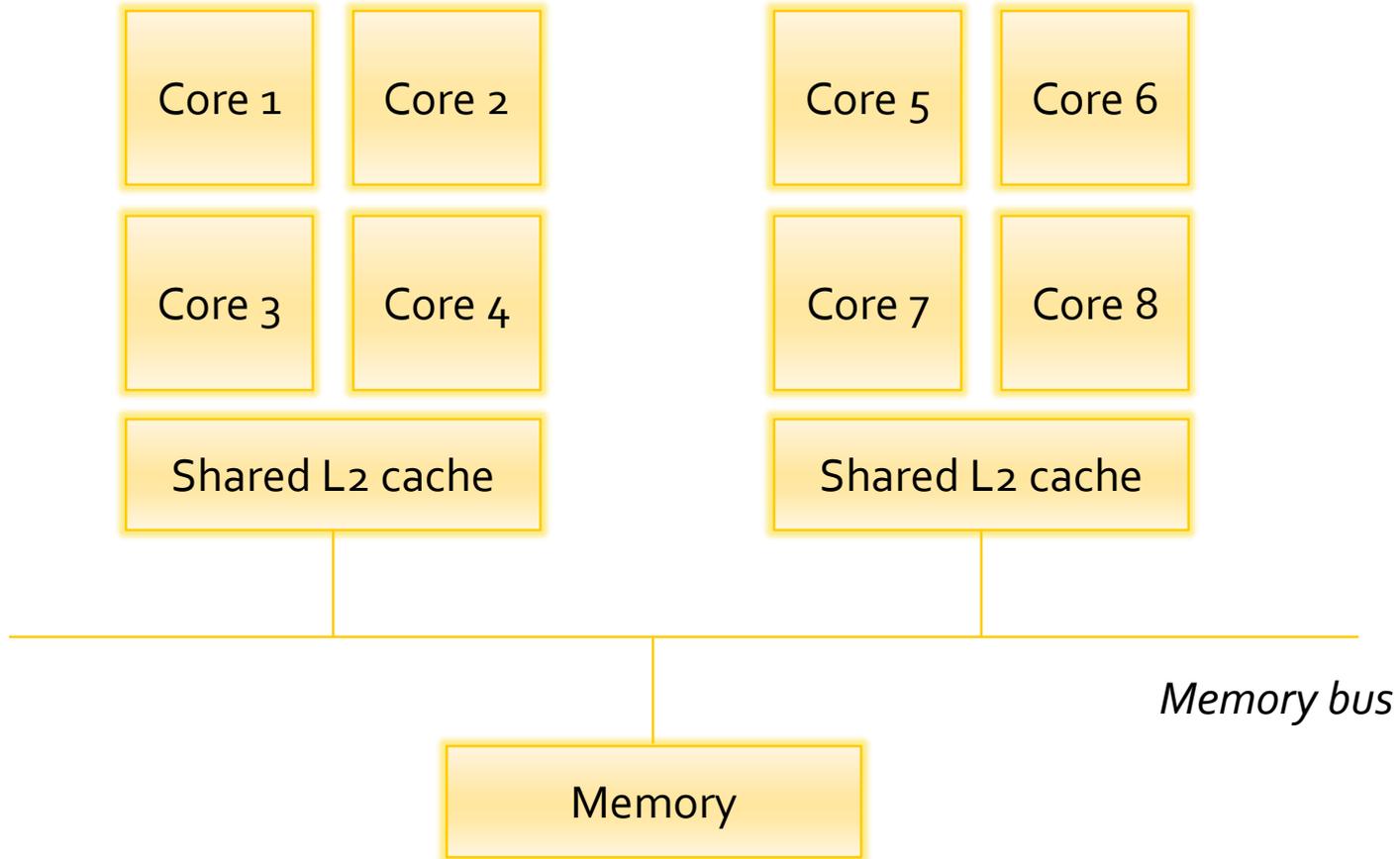
```
void releaseMCS(mcs *lock, QNode *qn) {  
    if (lock->tail = qn) {  
        if (CAS(&lock->tail, qn, NULL)) return;  
    }  
    while (qn->next == NULL) { }  
    qn->next->flag = TRUE;  
}
```

If we were at the tail
then remove us

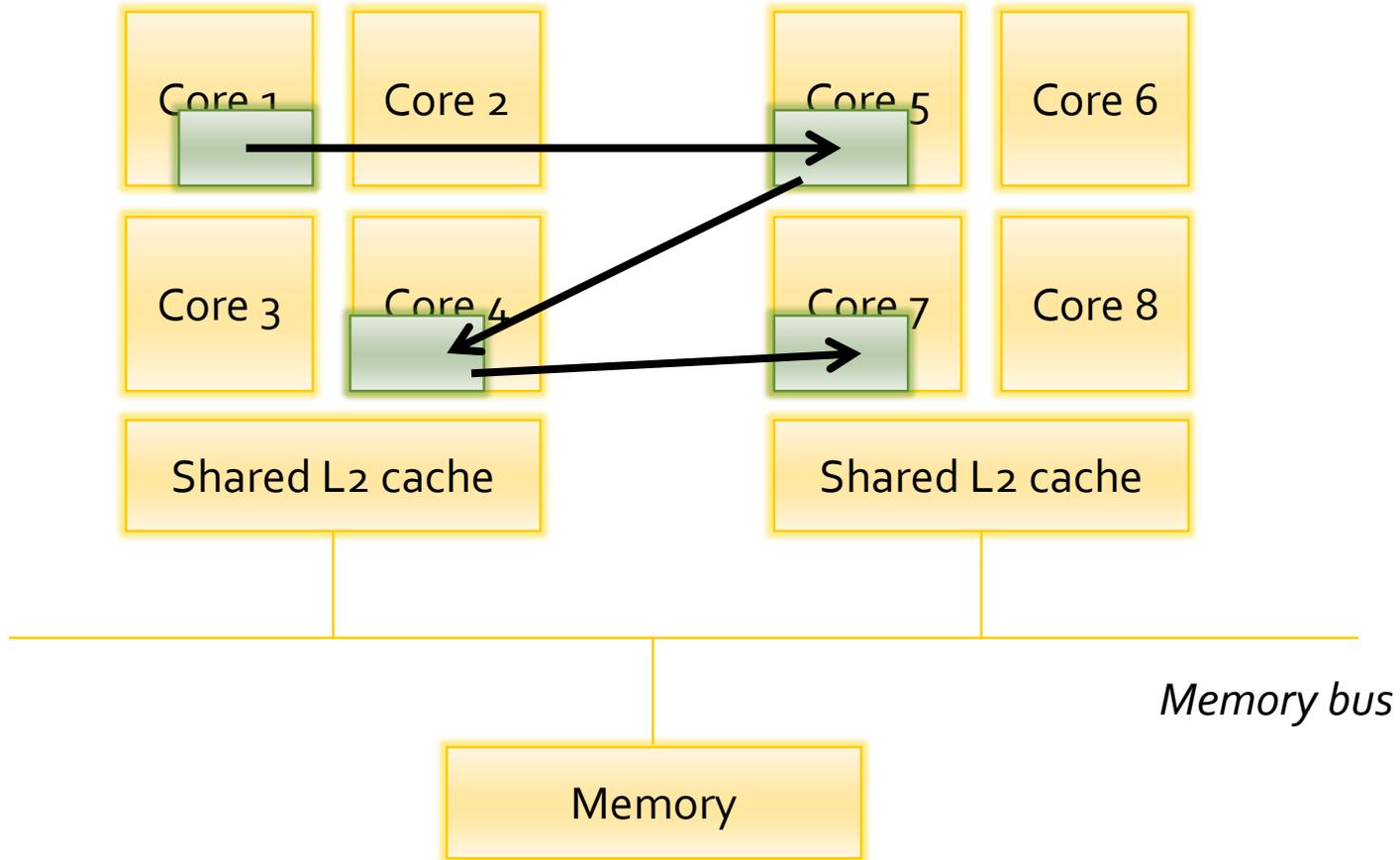
Wait for next lock holder
to announce themselves;
signal them

Hierarchical locks

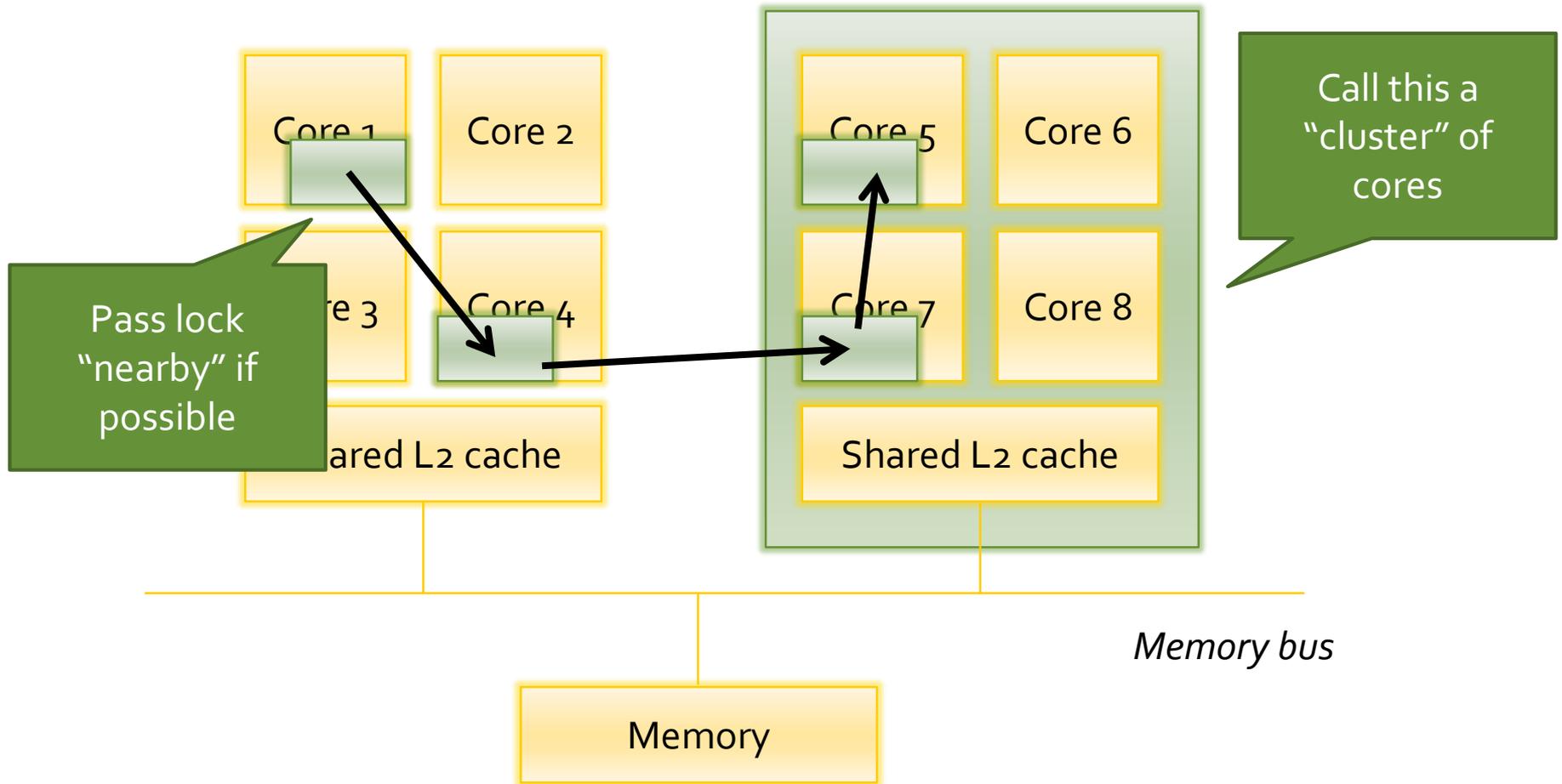
Hierarchical locks



Hierarchical locks



Hierarchical locks



Hierarchical TATAS with backoff

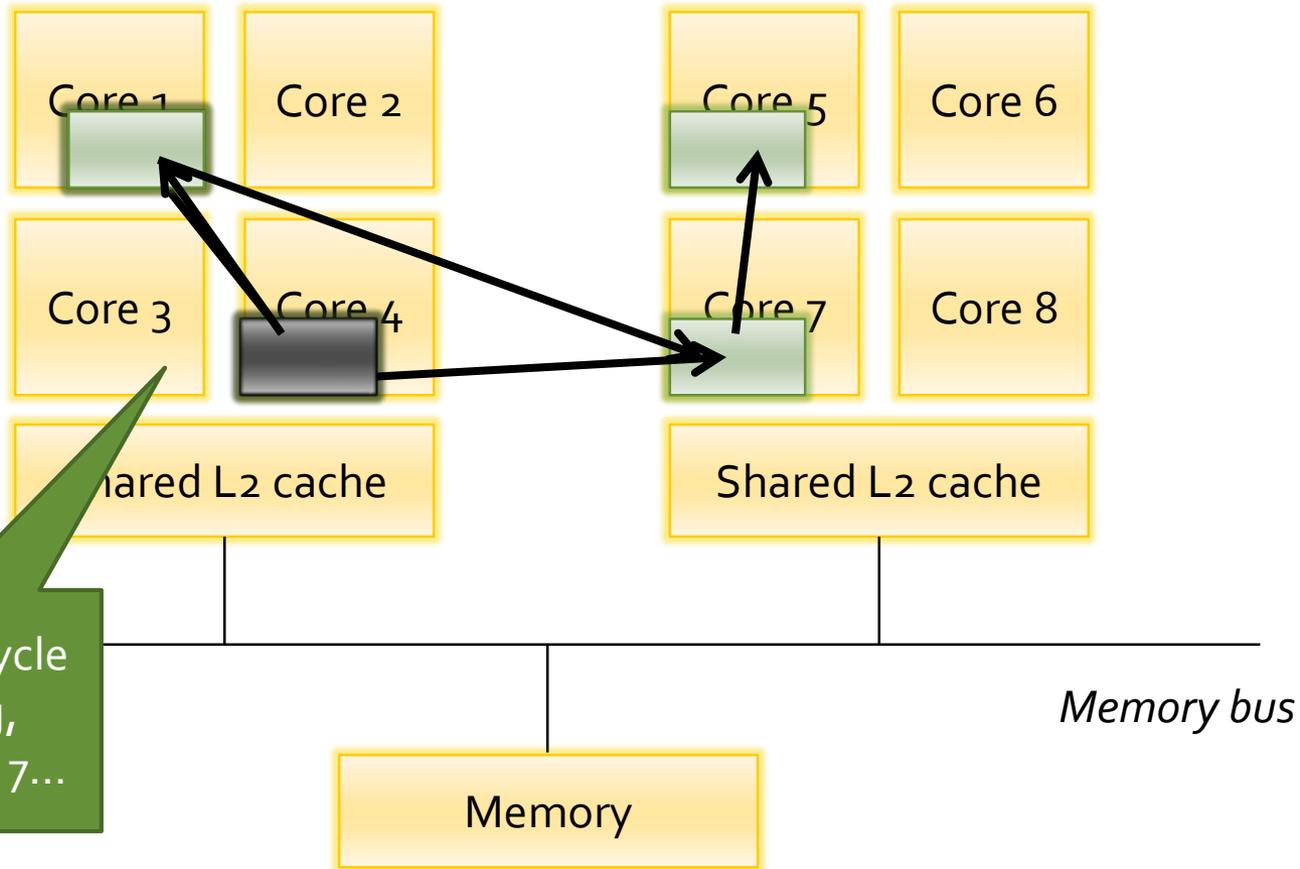
lock:

-1

-1 => lock available
n => lock held by cluster n

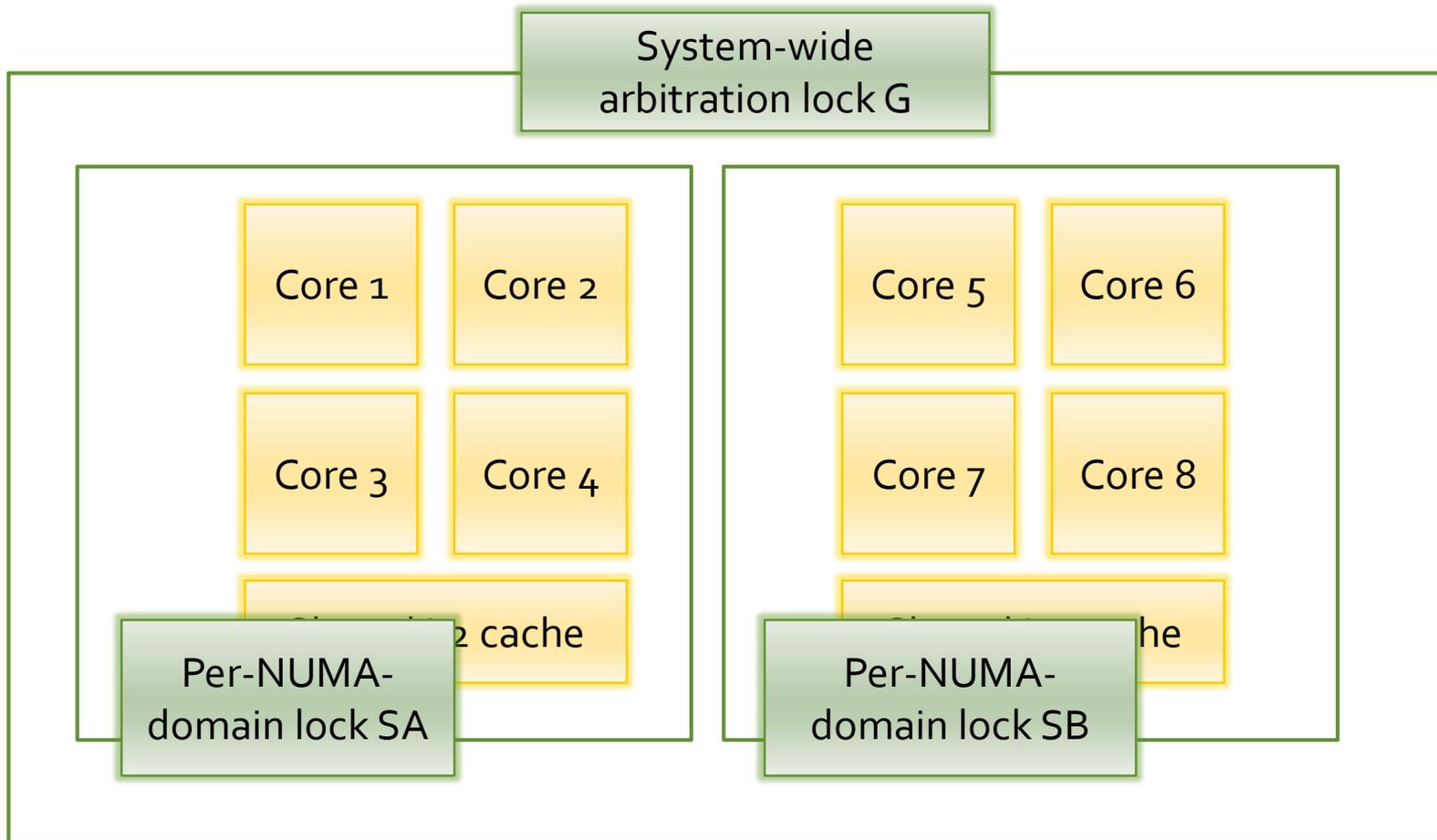
```
void acquireLock(bool *lock) {  
  do {  
    holder = *lock;  
    if (holder != -1) {  
      if (holder == MY_CLUSTER) {  
        BackOff(SHORT);  
      } else {  
        BackOff(LONG);  
      }  
    }  
  } while (!CAS(lock, -1, MY_CLUSTER));  
}
```

Hierarchical locks: unfairness v throughput



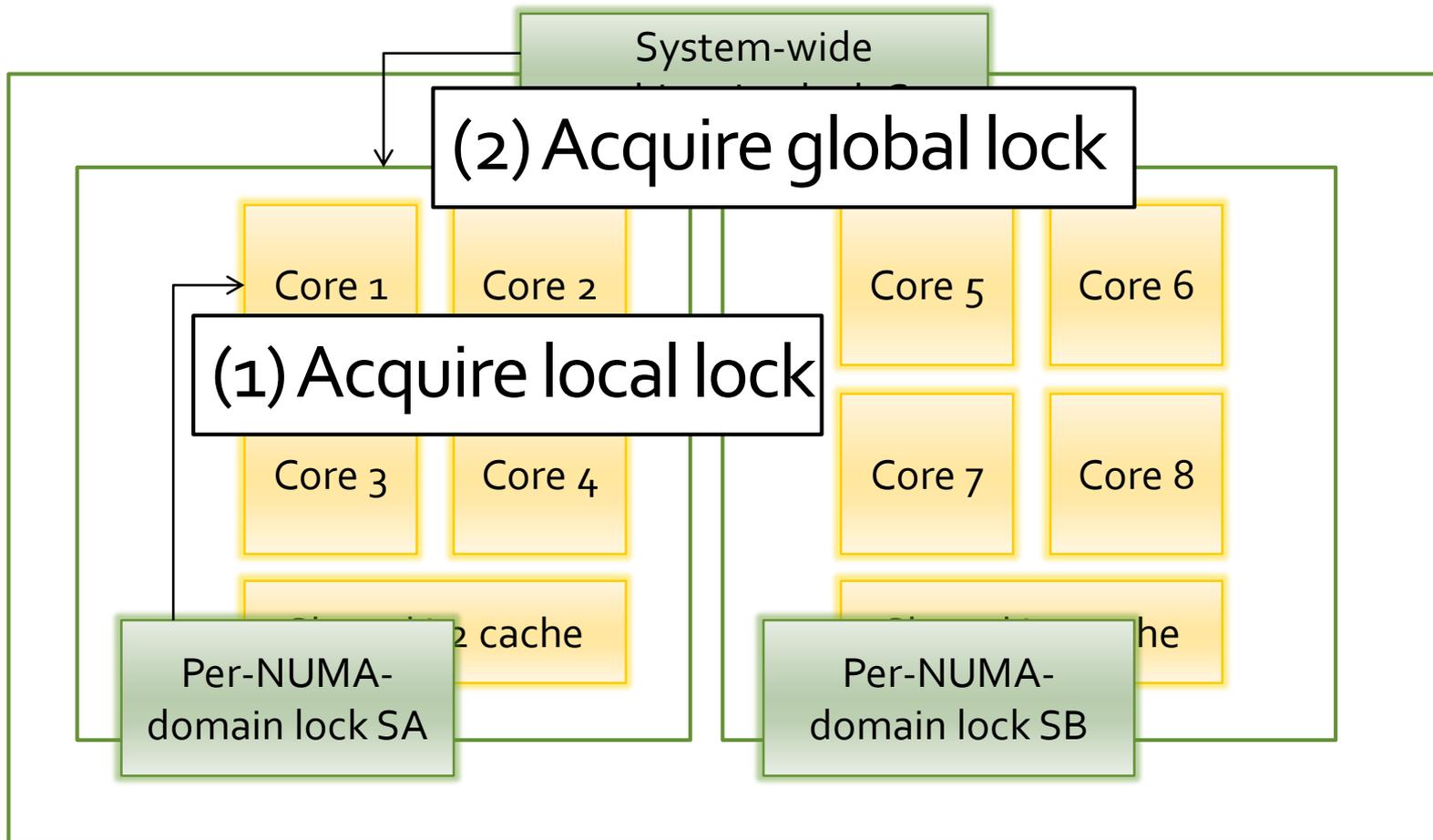
Lock cohorting

- “Lock Cohorting: A General Technique for Designing NUMA Locks”, Dice *et al* PPOPP 2012



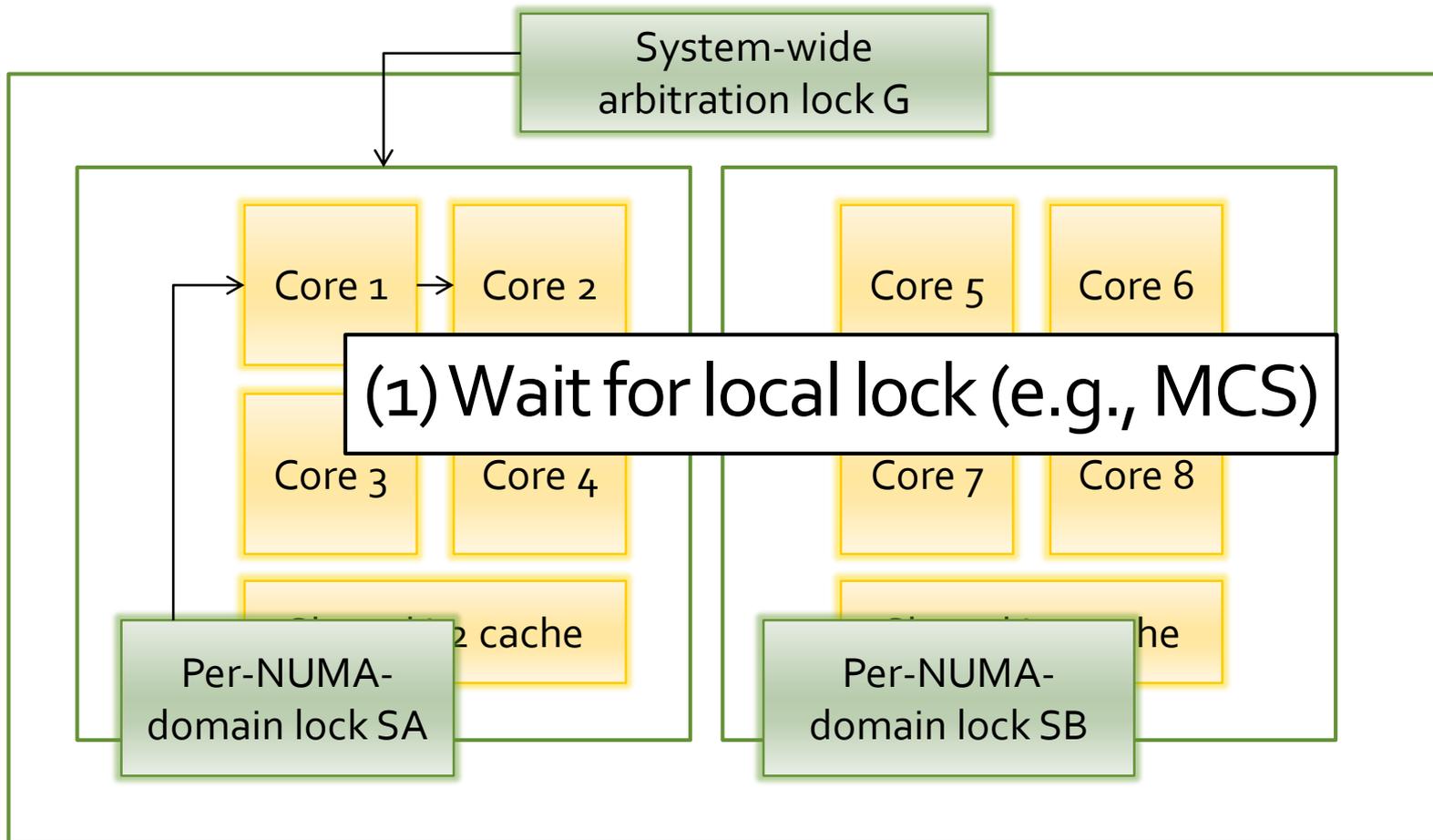
Lock cohorting

- Lock acquire, uncontended



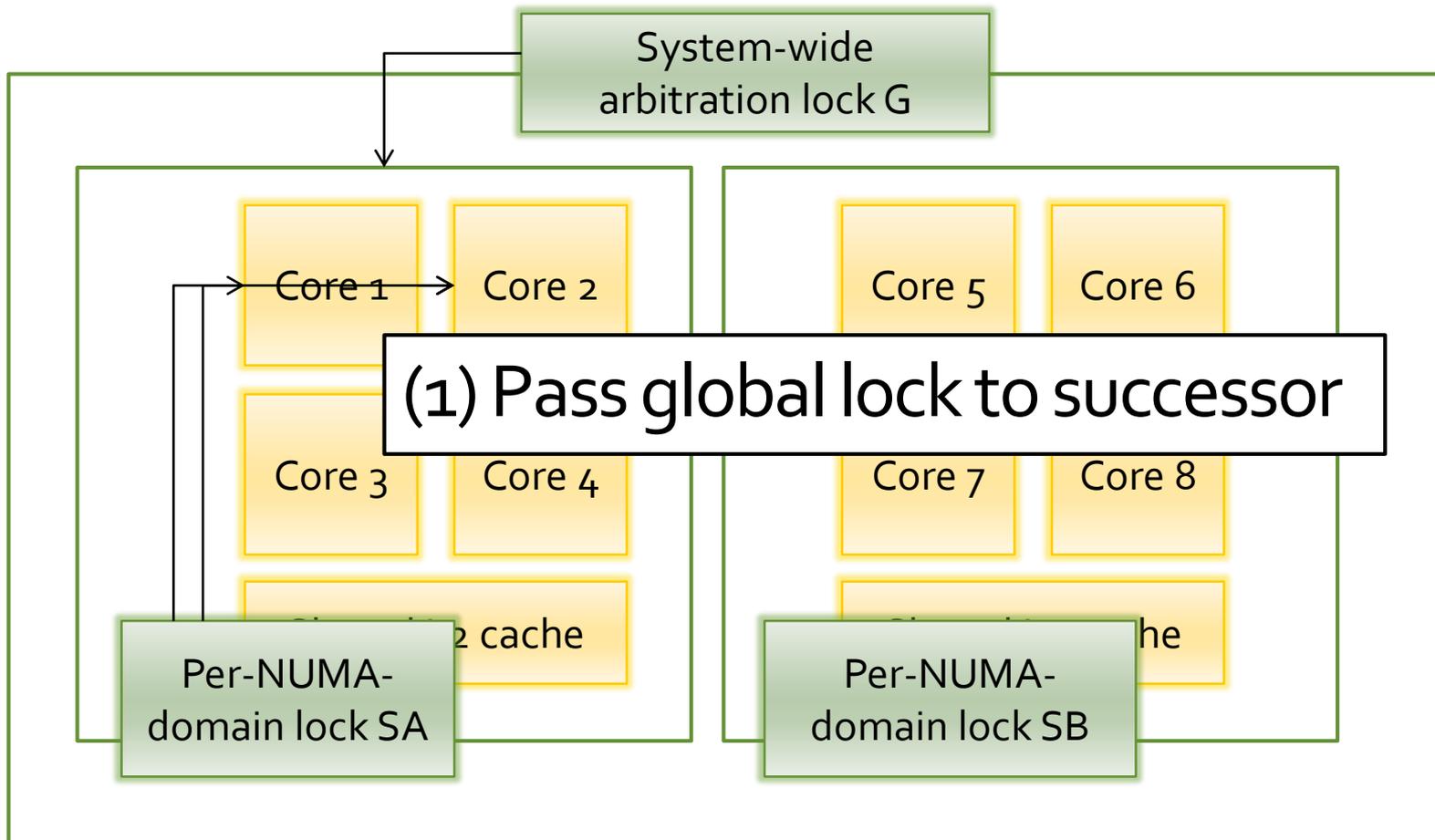
Lock cohorting

- Lock acquire, contended



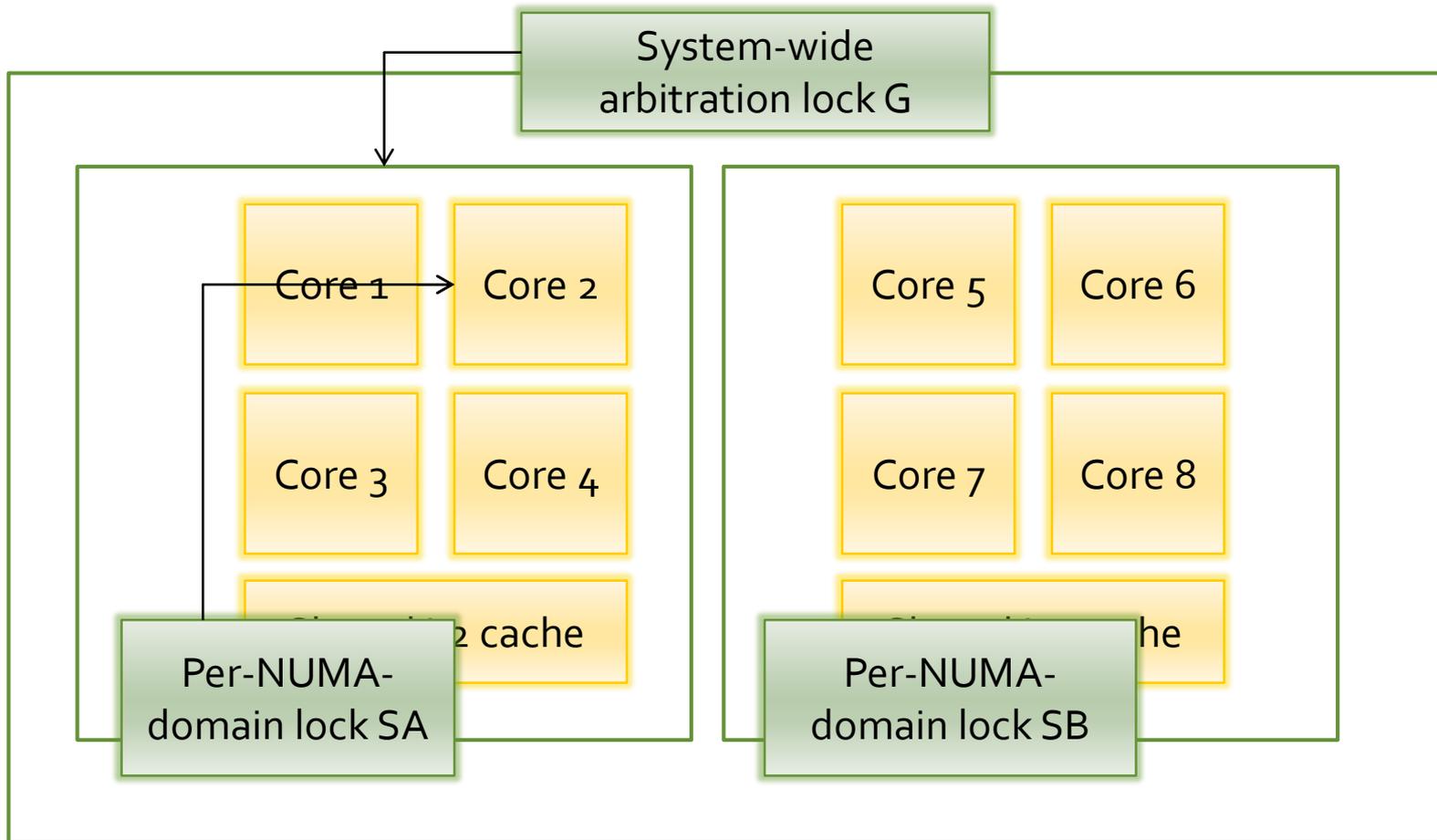
Lock cohorting

- Lock release, with successor



Lock cohorting, requirements

- Global: “thread oblivious” (acq one thread, release another)
- Local lock: “cohort detection” (can test for successors)



Reader-writer locks

Reader-writer locks (TATAS-like)

lock:

0

-1 => Locked for write
0 => Lock available
+n => Locked by n readers

```
void acquireWrite(int *lock) {  
    do {  
        if ((*lock == 0) &&  
            (CAS(lock, 0, -1))) {  
            break;  
        } while (1);  
    }  
}
```

```
void acquireRead(int *lock) {  
    do {  
        int oldVal = *lock;  
        if ((oldVal >= 0) &&  
            (CAS(lock, oldVal, oldVal+1))) {  
            break;  
        } } while (1);  
}
```

```
void releaseWrite(int *lock) {  
    *lock = 0;  
}
```

```
void releaseRead(int *lock) {  
    FADD(lock, -1); // Atomic fetch-and-add  
}
```

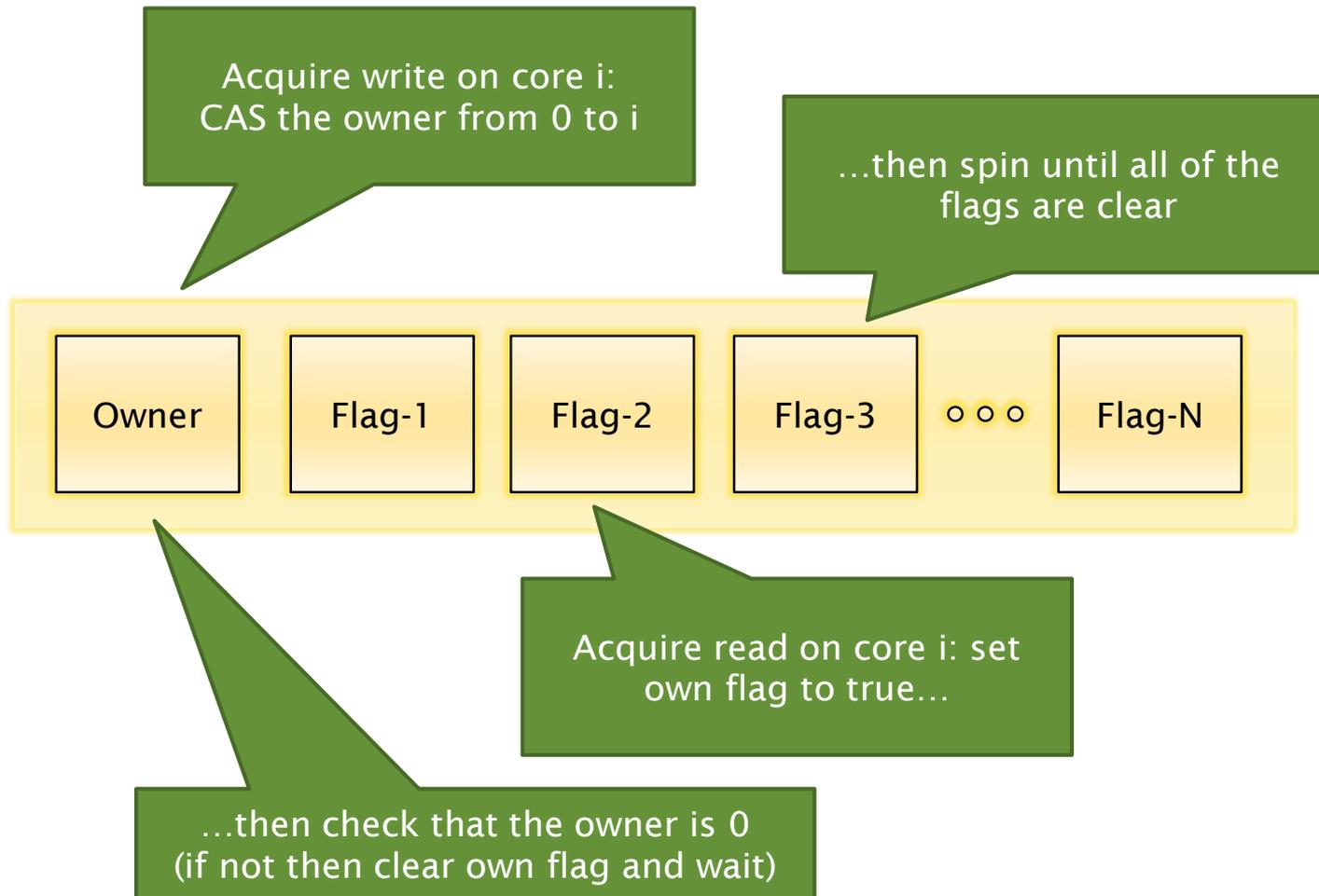
The problem with readers

```
int readCount() {  
    acquireRead(lock);  
    int result = count;  
    releaseRead(lock);  
    return result;  
}
```

```
void incrementCount() {  
    acquireWrite(lock);  
    count++;  
    releaseWrite(lock);  
}
```

- Each acquireRead fetches the cache line holding the lock in exclusive mode
 - Again: acquireRead are not logically conflicting, but this introduces a physical conflict
- The time spent managing the lock is likely to vastly dominate the actual time looking at the counter
- Many workloads are read-mostly...

Keeping readers separate



Keeping readers separate

- With care, readers do not need to synchronize with other readers
 - Extend the flags to be whole cache lines
 - Pack multiple locks flags for the same thread onto the same line
 - Exploit the cache structure in the machine: Dice & Shavit's TLRW byte-lock on SPARC Niagara
- If "N" threads is very large..
 - Dedicate the flags to specific important threads
 - Replace the flags with ordinary multi-reader locks
 - Replace the flags with per-NUMA-domain multi-reader locks

Other locking techniques

- Affinity
 - Allow one thread fast access to the lock
 - “One thread” – e.g., previous lock holder
 - “Fast access” – e.g., with fewer / no atomic CAS operations
 - Mike Burrows “Implementing unnecessary mutexes”
(Do the assumptions hold? How slow is an uncontended CAS on a modern machine? Are these techniques still useful?)

Other locking techniques

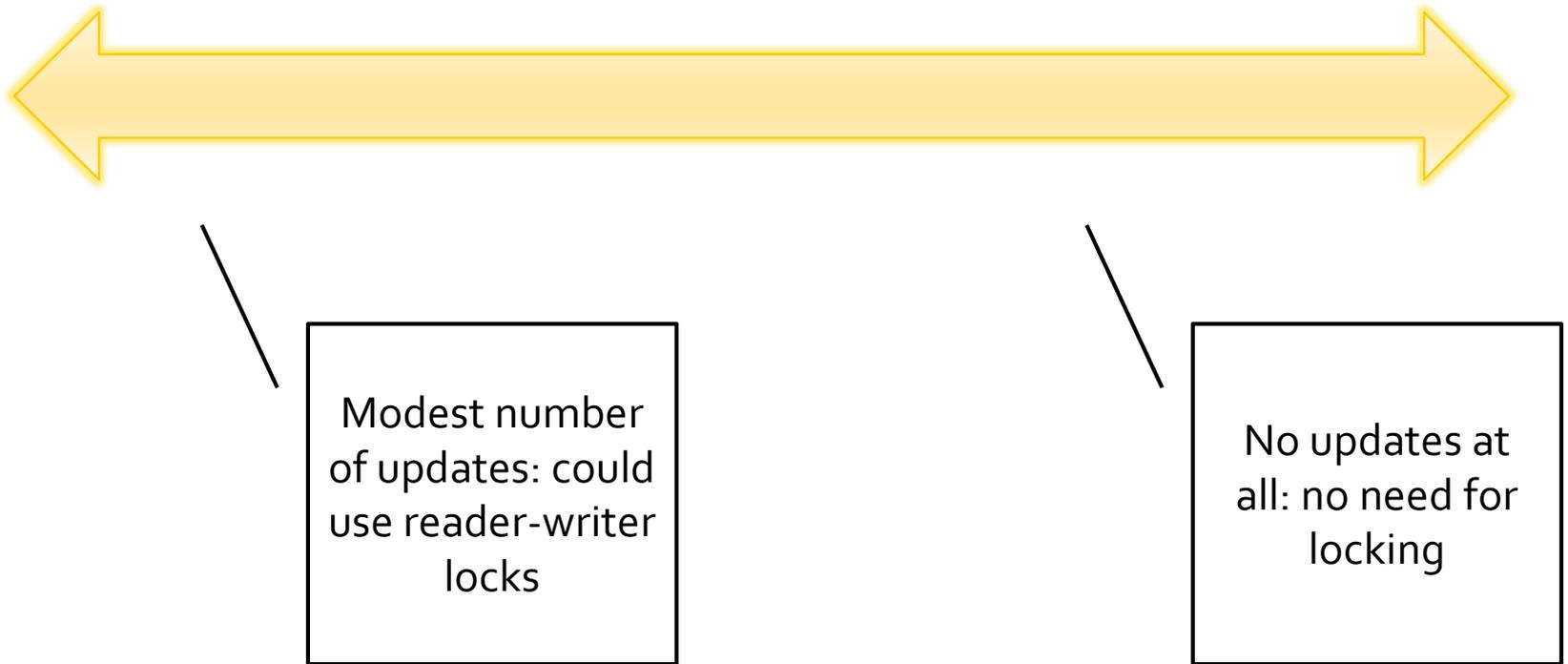
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(Do the assumptions hold? How slow is an uncontended CAS on a modern machine? Are these techniques still useful?)
- Inflation
 - Start out with a simple lock for likely-to-be-uncontended use
 - Replace with a “proper” lock if contended
 - David Bacon (thin locks), Agesen *et al* (meta-locks)
 - Motivating example: standard libraries in Java

Where are we

- Amdahl's law: to scale to large numbers of cores, we need critical sections to be rare and/or short
- A lock implementation may involve updating a few memory locations
- Accessing a data structure may involve only a few memory locations too
- If we try to shrink critical sections then the time in the lock implementation becomes proportionately greater
- So:
 - try to make the cost of the operations in the critical section lower, or
 - try to write critical sections correctly without locking

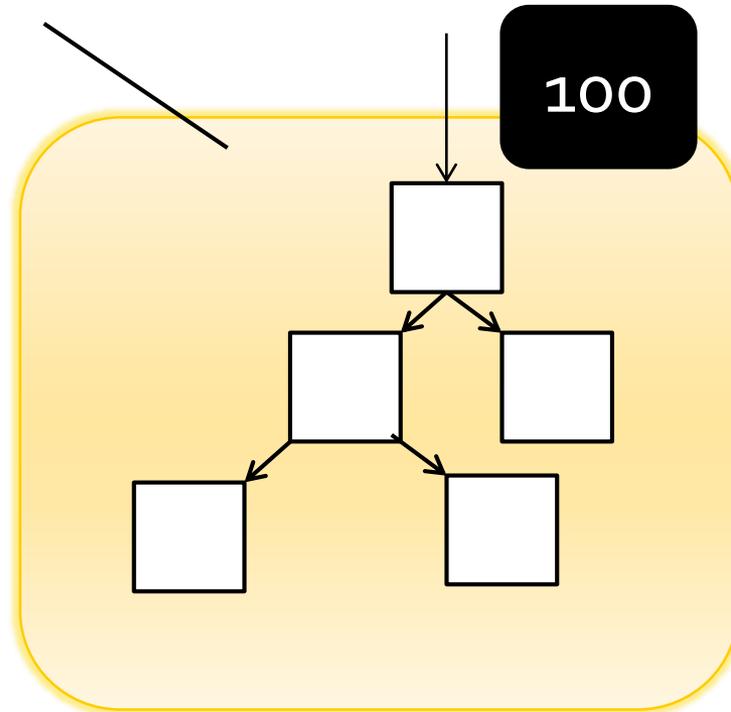
Reading without locking

What if updates are very rare



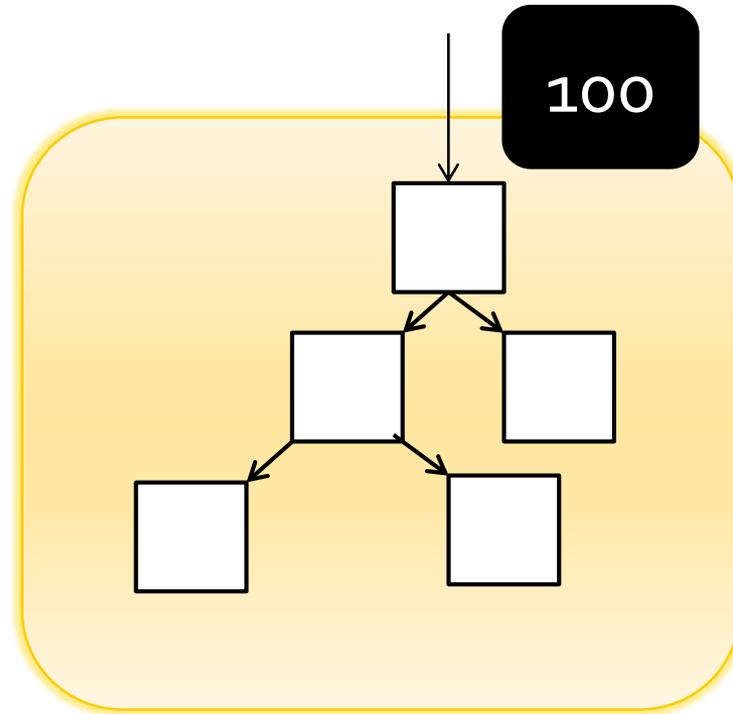
Version numbers

Sequential
data structure
with write
lock



Per-data-
structure
version
number

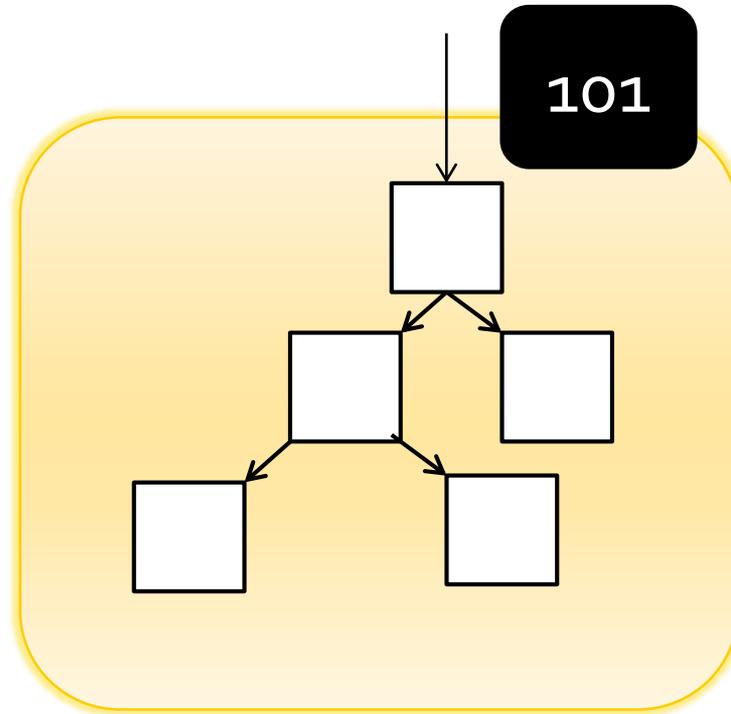
Version numbers: writers



Version numbers: writers

Writers:

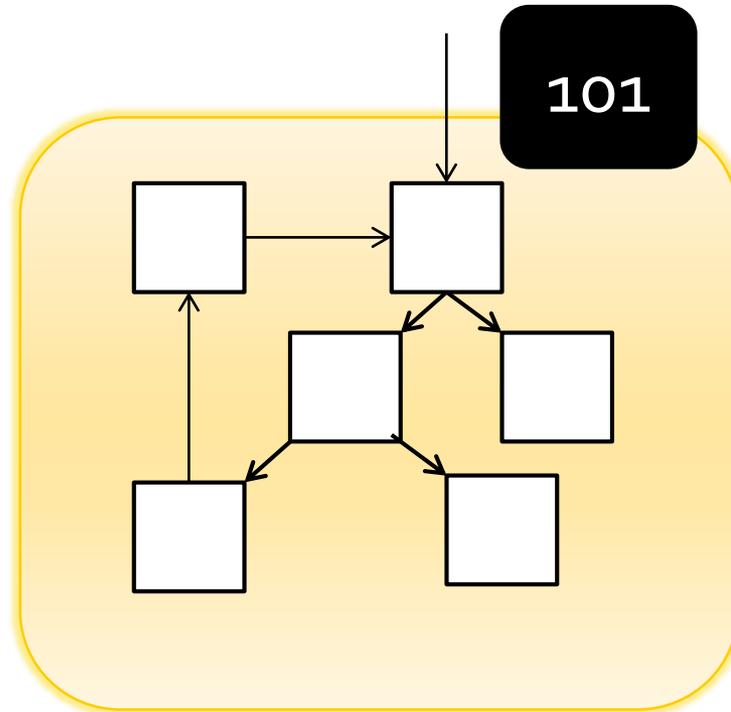
1. Take write lock
2. Increment version number



Version numbers: writers

Writers:

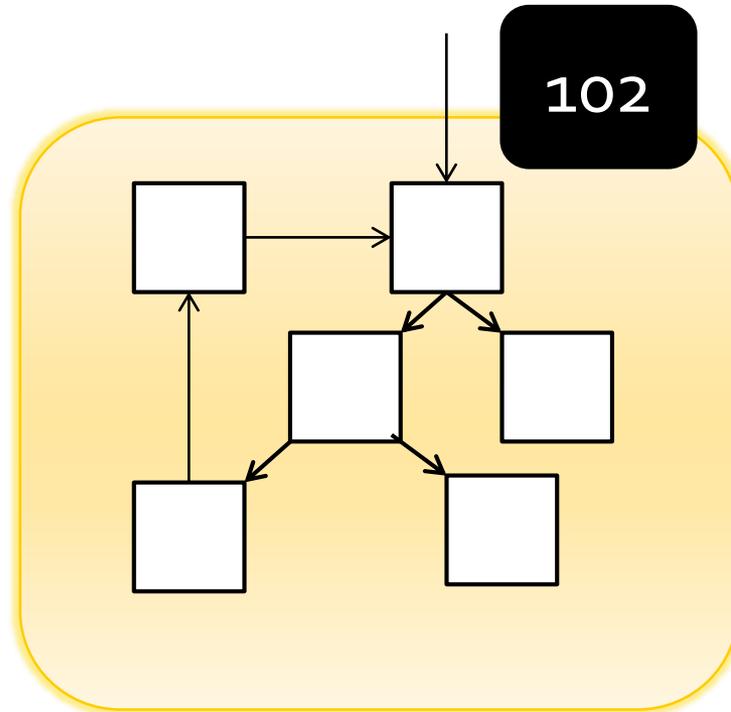
1. Take write lock
2. Increment version number
3. Make update



Version numbers: writers

Writers:

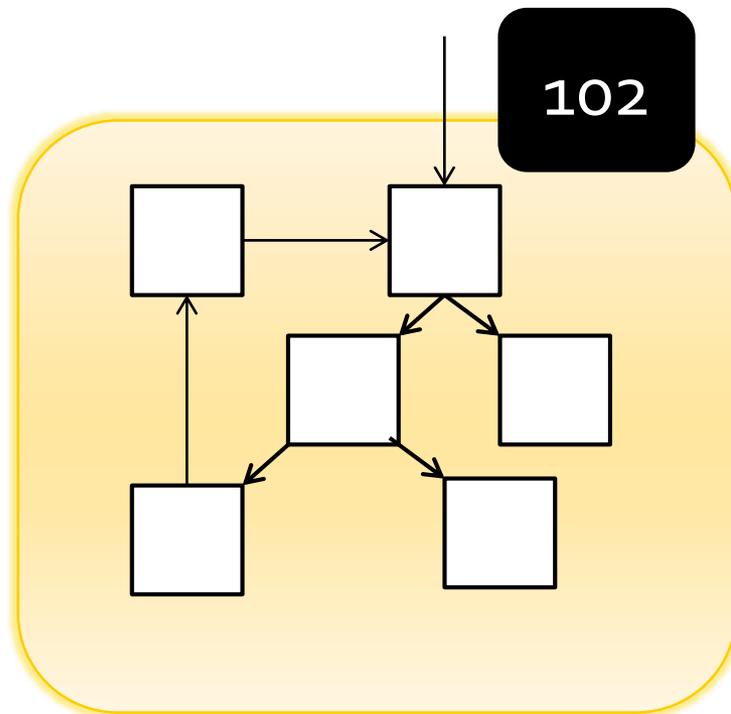
1. Take write lock
2. Increment version number
3. Make update
4. Increment version number
5. Release write lock



Version numbers: readers

Writers:

1. Take write lock
2. Increment version number
3. Make update
4. Increment version number
5. Release write lock



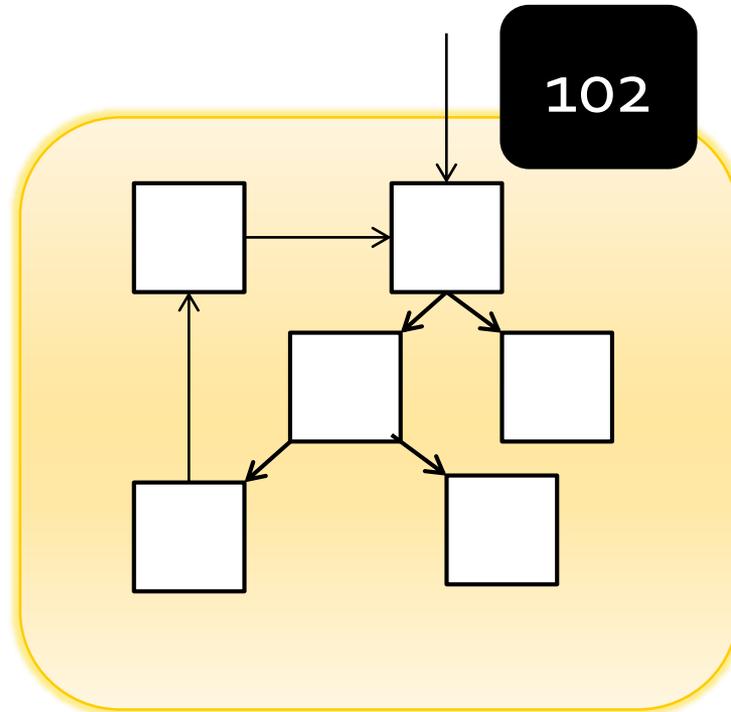
Readers:

1. Wait for version number to be even

Version numbers: readers

Writers:

1. Take write lock
2. Increment version number
3. Make update
4. Increment version number
5. Release write lock



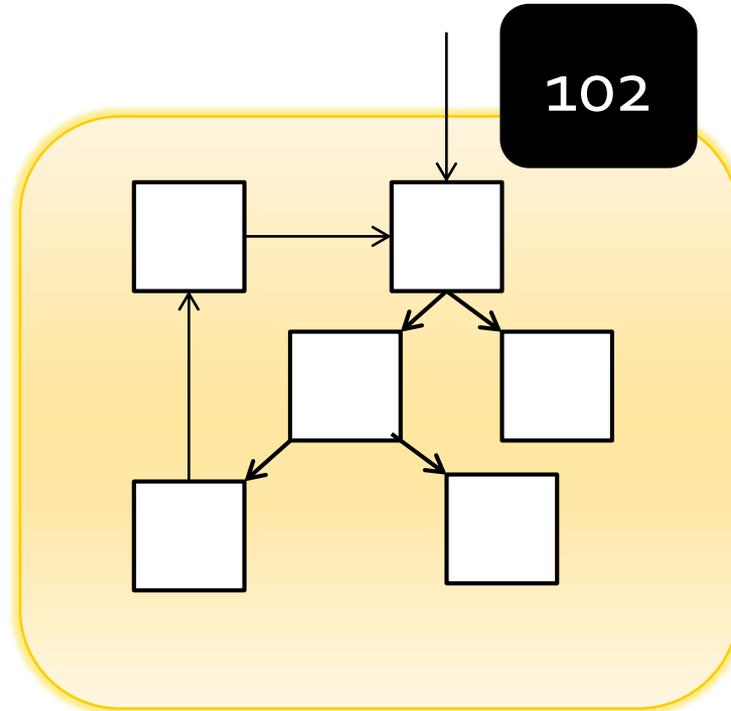
Readers:

1. Wait for version number to be even
2. Do operation

Version numbers: readers

Writers:

1. Take write lock
2. Increment version number
3. Make update
4. Increment version number
5. Release write lock



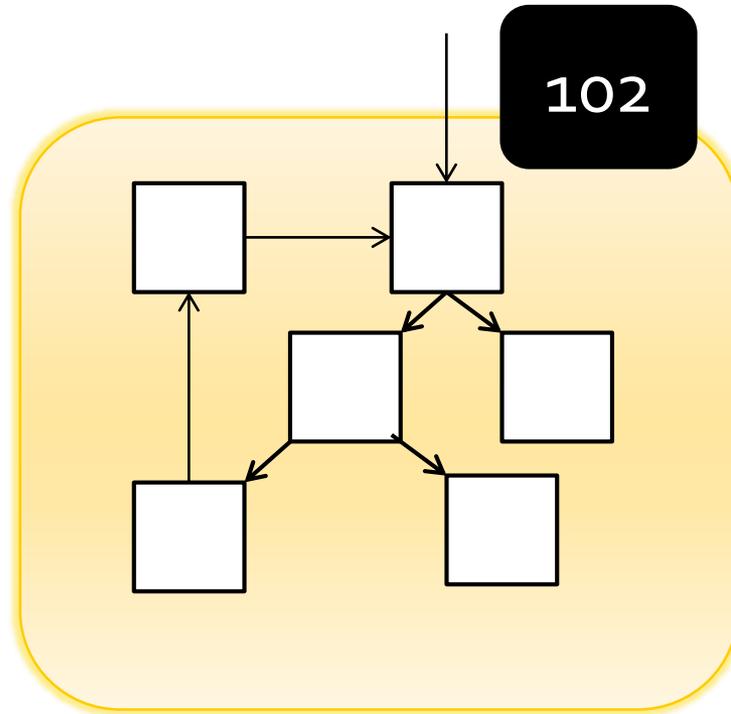
Readers:

1. Wait for version number to be even
2. Do operation
3. Has the version number changed?
4. Yes? Go to 1

Why do we need the two steps?

Writers:

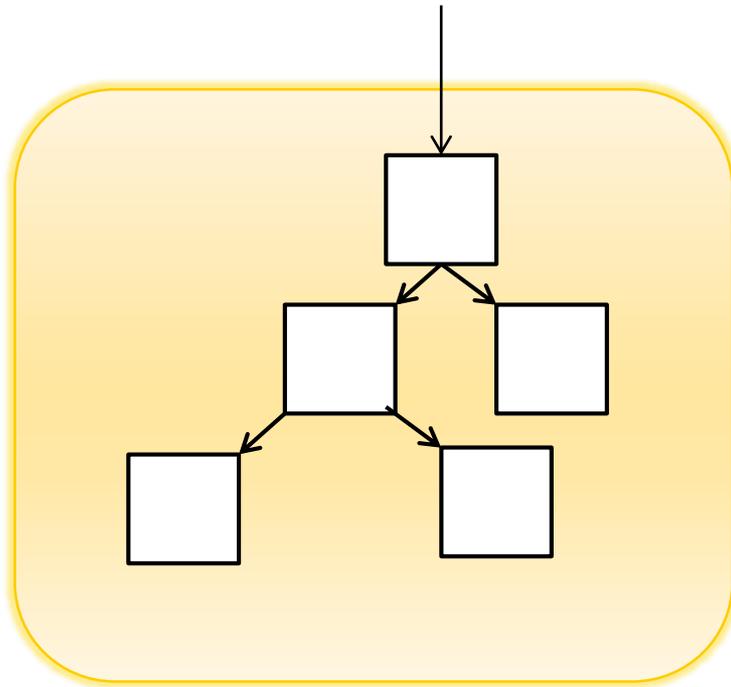
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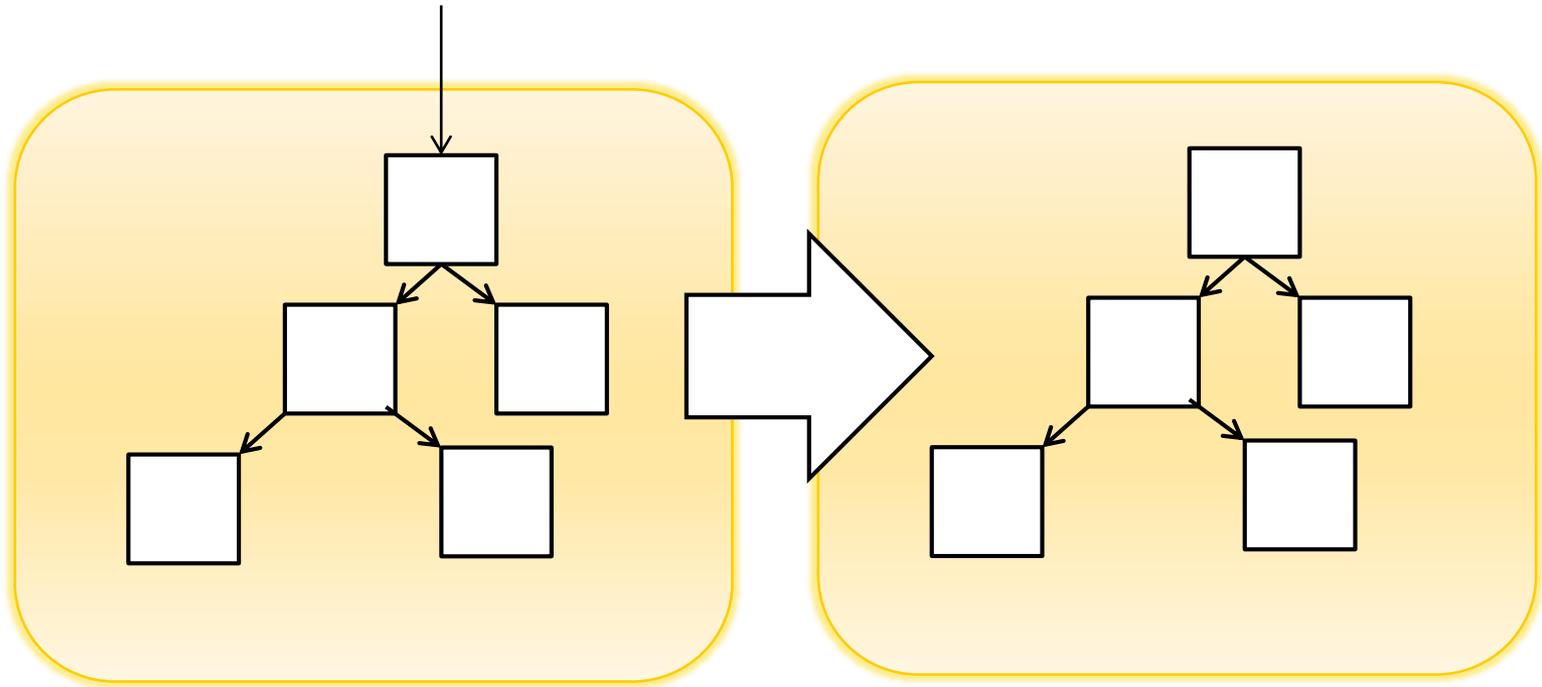
Readers:

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2. Do operation
3. Has the version number changed?
4. Yes? Go to 1

Read-Copy-Update (RCU)

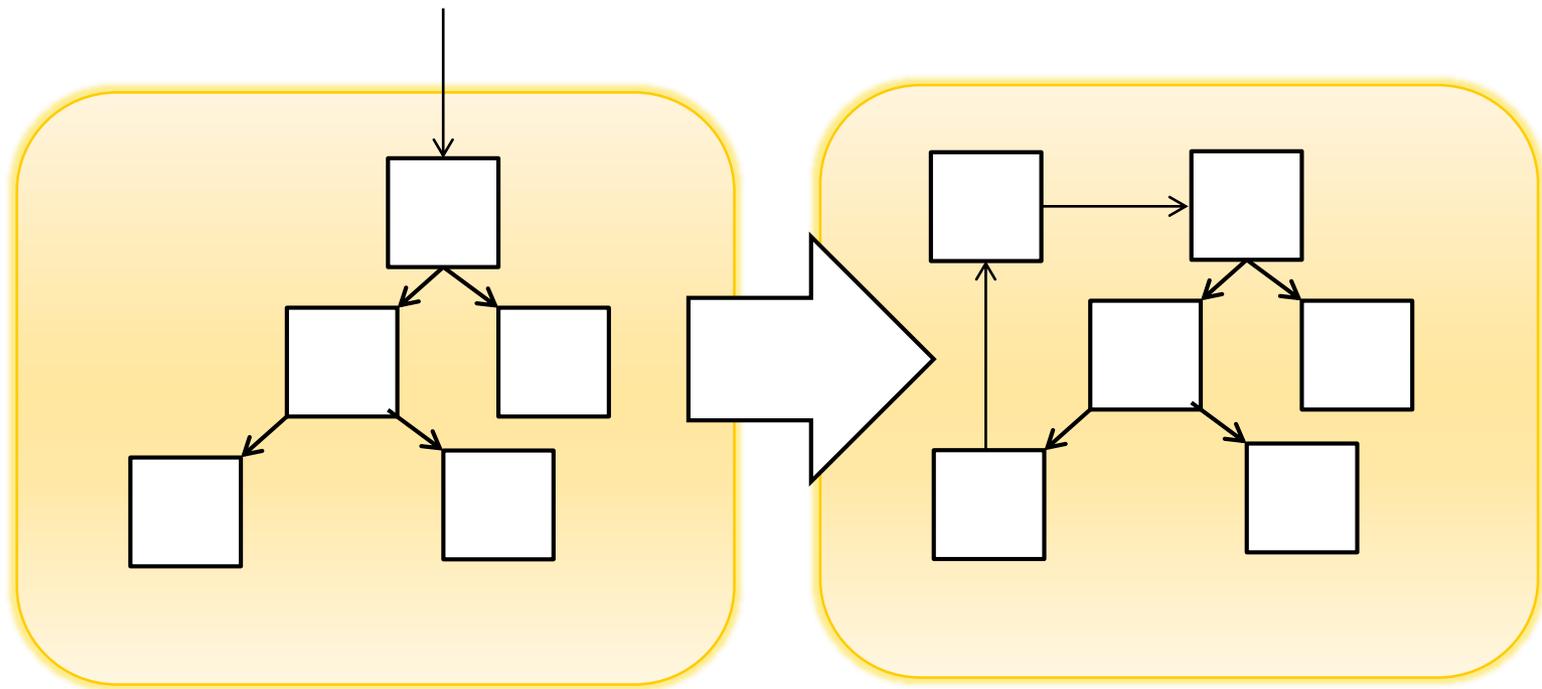


Read-Copy-Update (RCU)



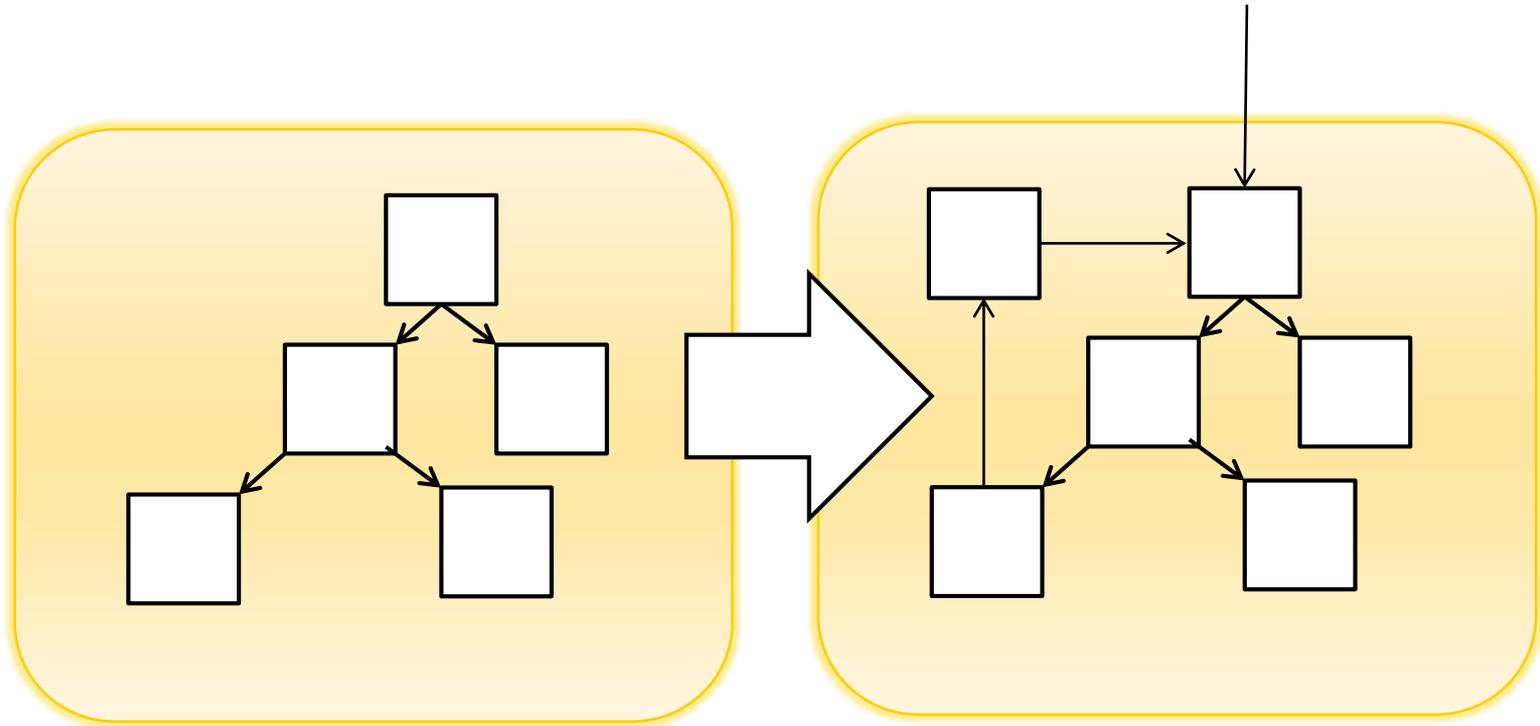
1. Copy existing structure

Read-Copy-Update (RCU)



1. Copy existing structure
2. Update copy

Read-Copy-Update (RCU)



1. Copy existing structure
2. Update copy
3. Install copy with CAS on root pointer

Read-Copy-Update (RCU)

- Use locking to serialize updates (typically)
 - ...but allow readers to operate concurrently with updates
- Ensure that readers don't go wrong if they access data mid-update
 - Have data structures reachable via a single root pointer: update the root pointer rather than updating the data structure in-place
 - Ensure that updates don't affect readers – e.g., initializing nodes before splicing them into a list, and retaining “next” pointers in deleted nodes
 - Exact semantics offered can be subtle (ongoing research direction)
- Memory management problems common with lock-free data structures

When will these techniques be effective?

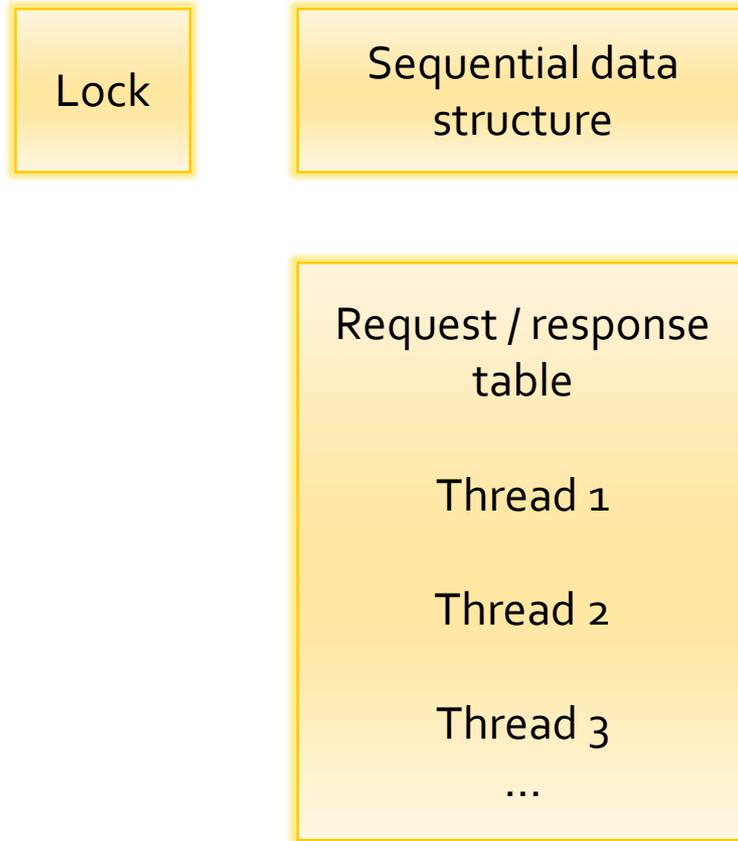
- Update rate low
 - So the need to serialize updates is OK
- Readers behaviour is OK mid-update
 - E.g., structure small enough to clone, rather than update in place
 - Readers will be OK until a version number check (not enter endless loops / crash / etc.)
- Deallocation or re-use of memory can be controlled

Flat combining

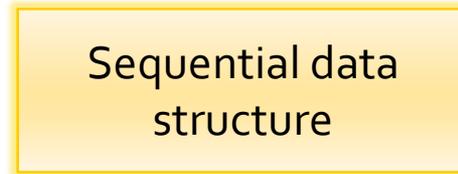
Flat combining

- “Flat Combining and the Synchronization-Parallelism Tradeoff”, Hendler *et al*
- Intuition:
 - Acquiring and releasing a lock involves numerous cache line transfers on the interconnect
 - These may take hundreds of cycles (e.g., between cores in different NUMA nodes)
 - The work protected by the lock may involve only a few memory accesses...
 - ...and these accesses may be likely to hit in the cache of the previous lock holder (but miss in your own)
 - So: if a lock is not available, request that the current lock holder does the work on your behalf

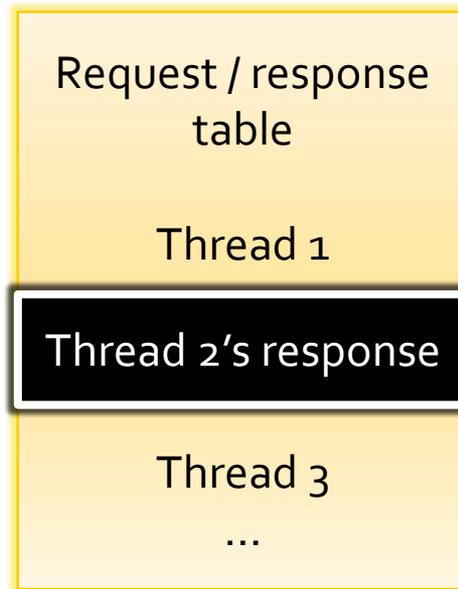
Flat combining



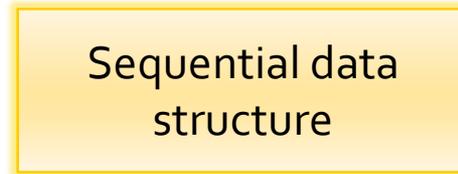
Flat combining: uncontended acquire



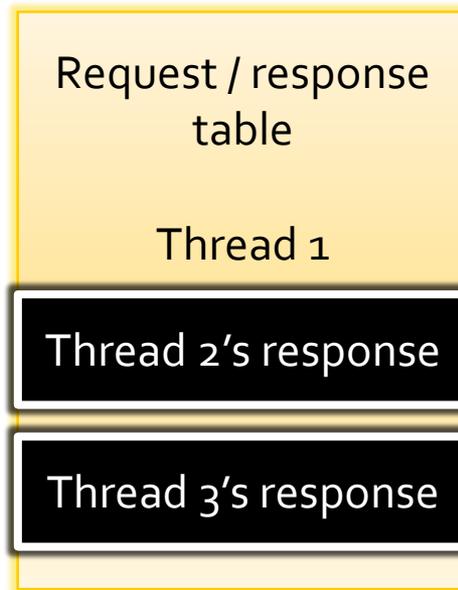
1. Write proposed op to req/resp table
2. Acquire lock if it is free
3. Process requests
4. Release lock
5. Pick up response



Flat combining: contended acquire



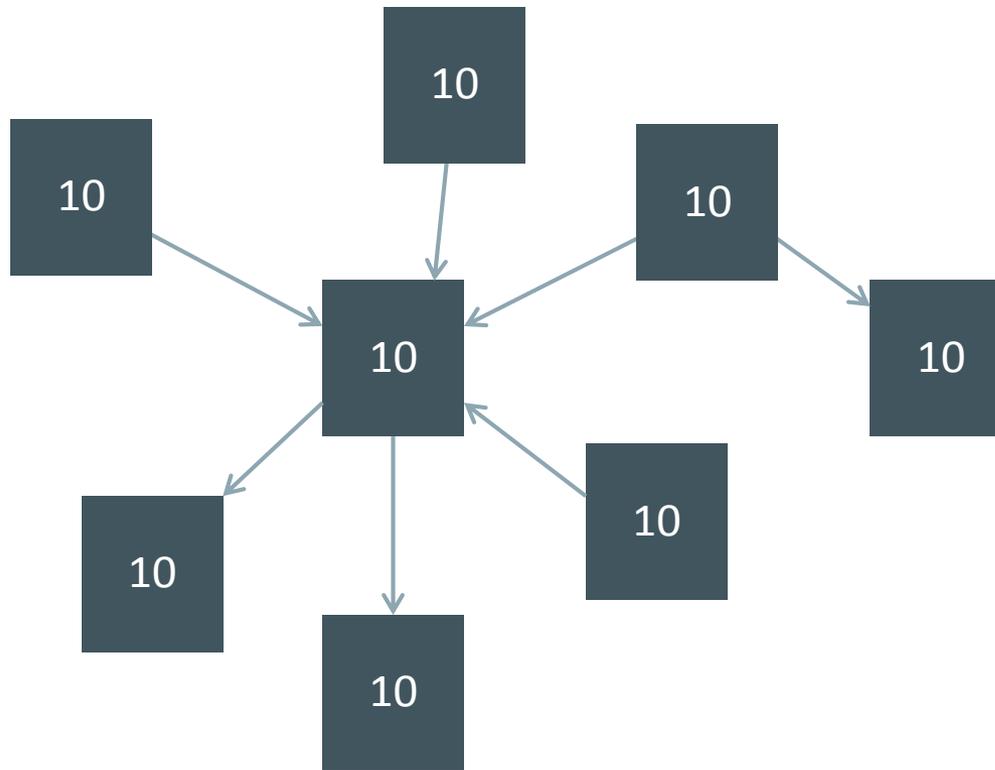
1. Write proposed op to req/resp table
2. See lock is not free
3. Wait for response
4. Pick up response



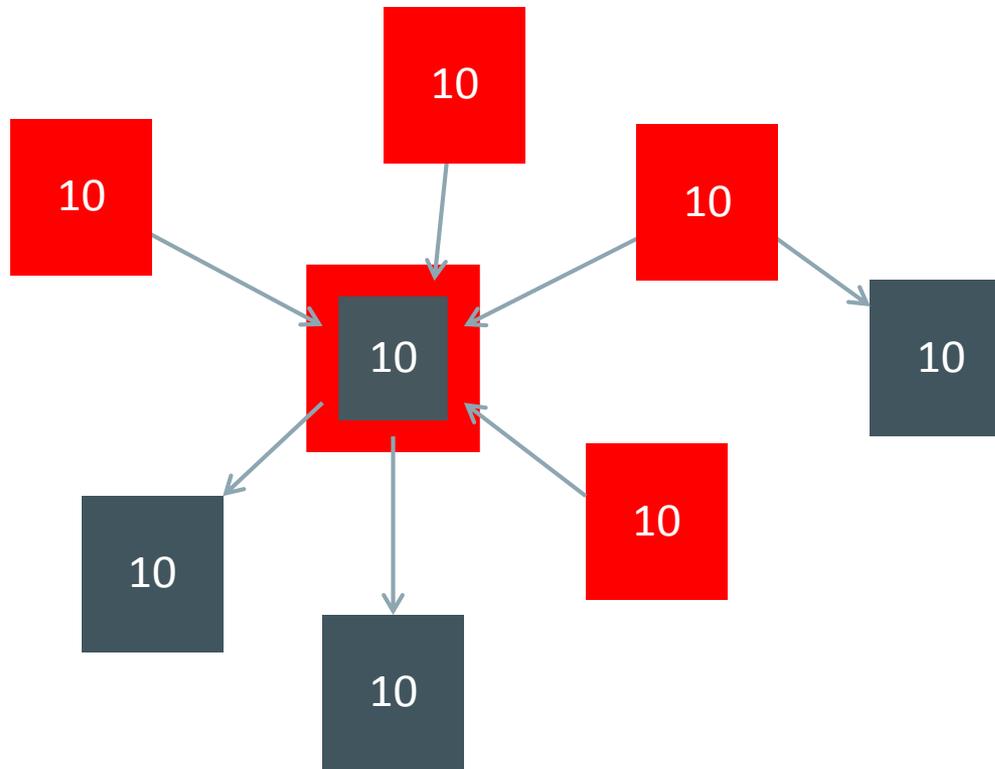
Recent research: Parallel work distribution

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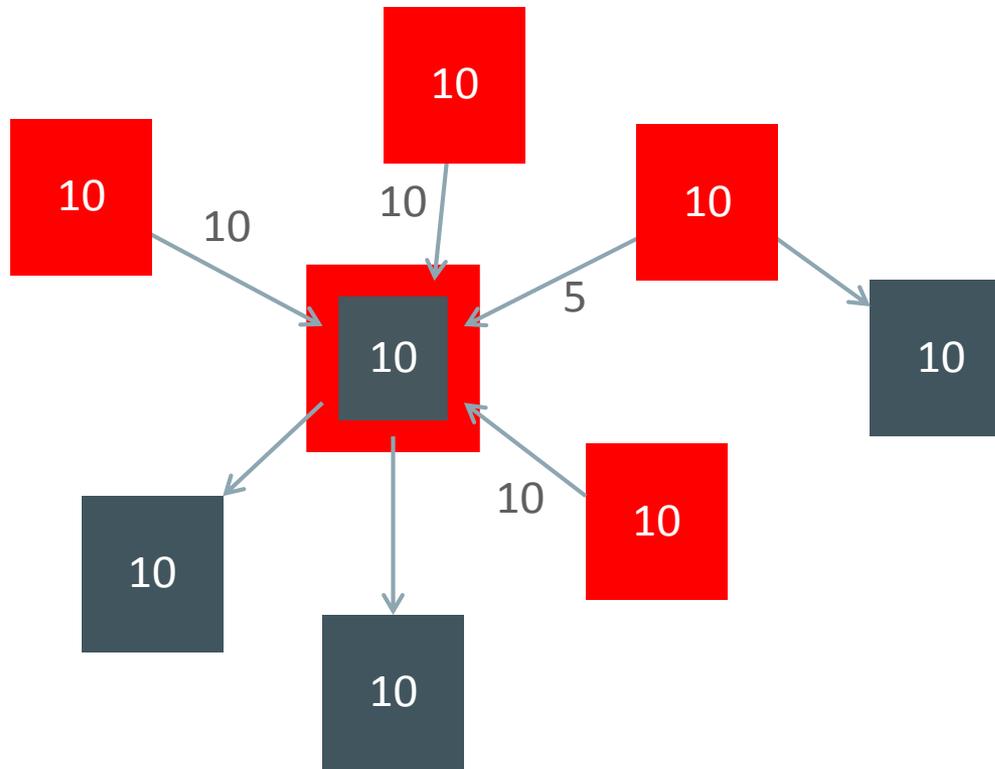
PageRank inner loop



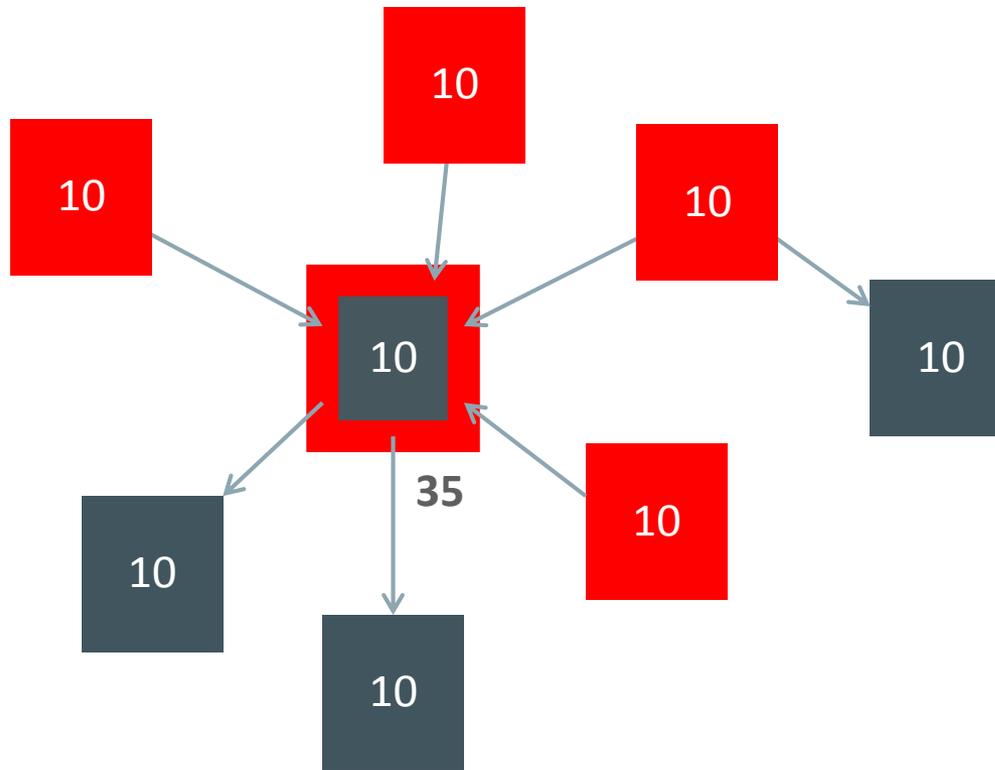
PageRank inner loop



PageRank inner loop



PageRank inner loop



Batch size / load imbalance trade-off



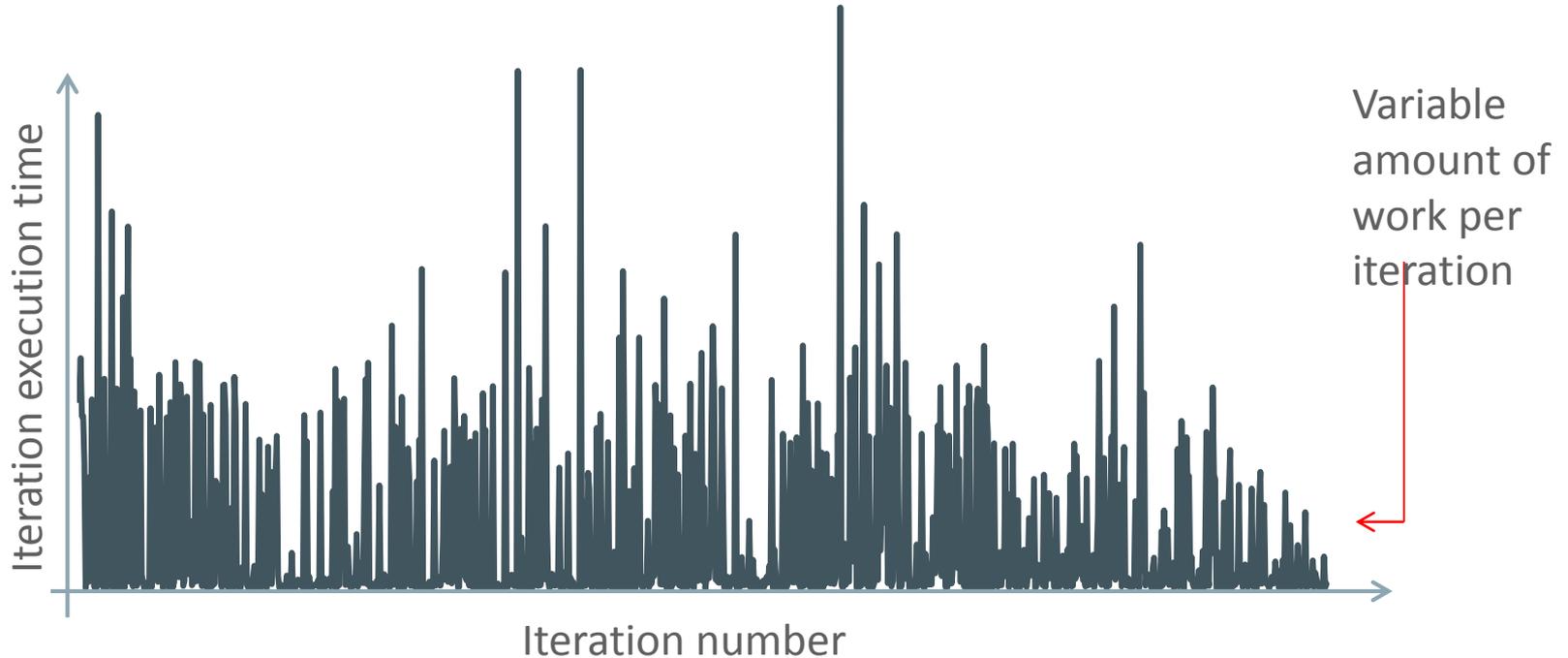
Divide into large batches of vertices

Reduce overheads
Risk load imbalance

Divide into small batches of vertices

Increase overheads distributing work
Achieve better load balance

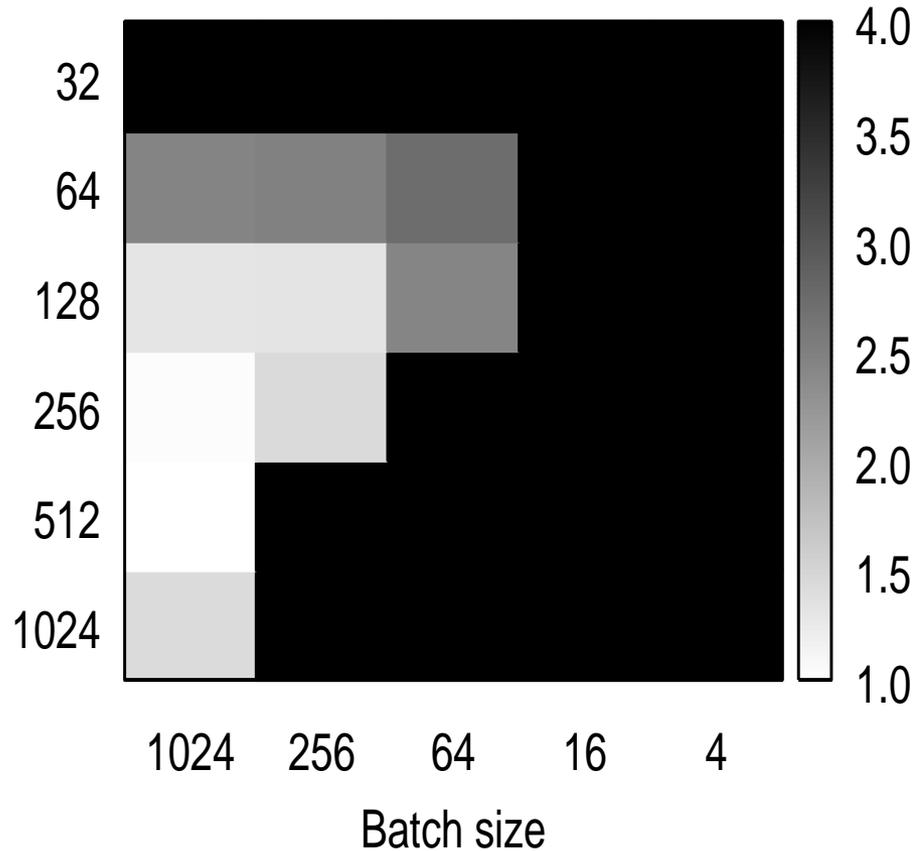
Batch size / load imbalance trade-off



(Actual data – #out-edges of the top 1000 nodes in the SNAP Twitter dataset)

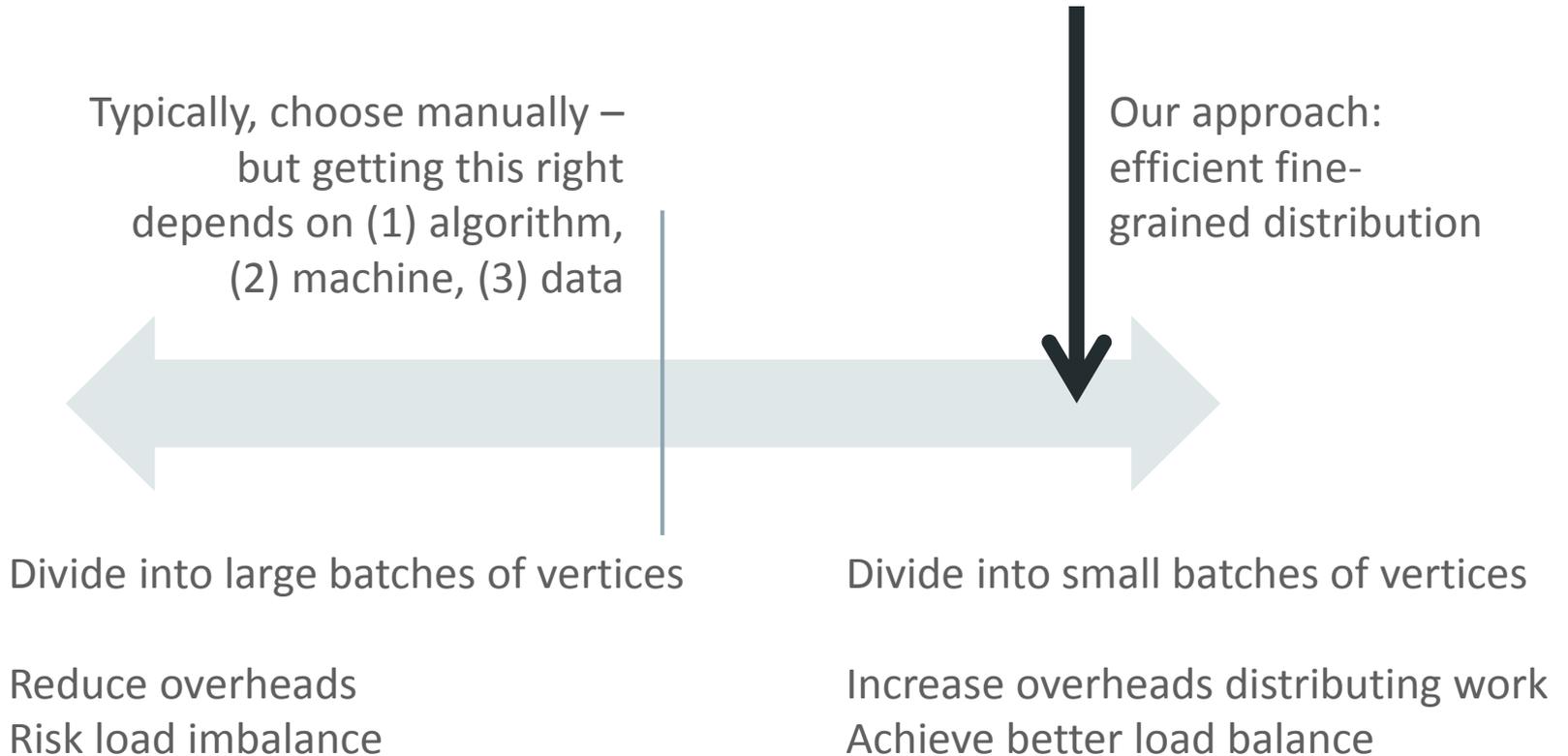
Example performance

Complete PageRank execution, SNAP LiveJournal data set



8-socket SPARC T5
16 cores per socket
8 h/w threads per core

Batch size / load imbalance trade-off

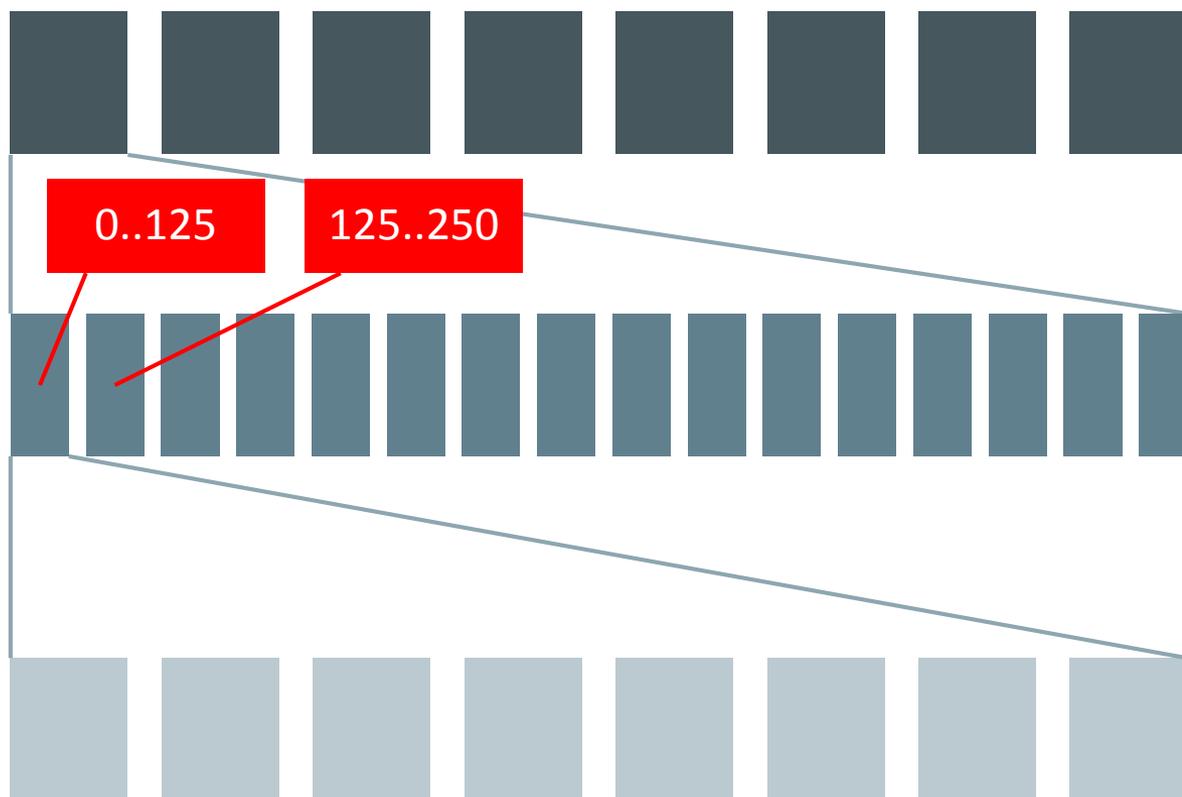


Consider distributing 0..16000 vertices, batch size 10



8 sockets

Consider distributing 0..16000 vertices, batch size 10



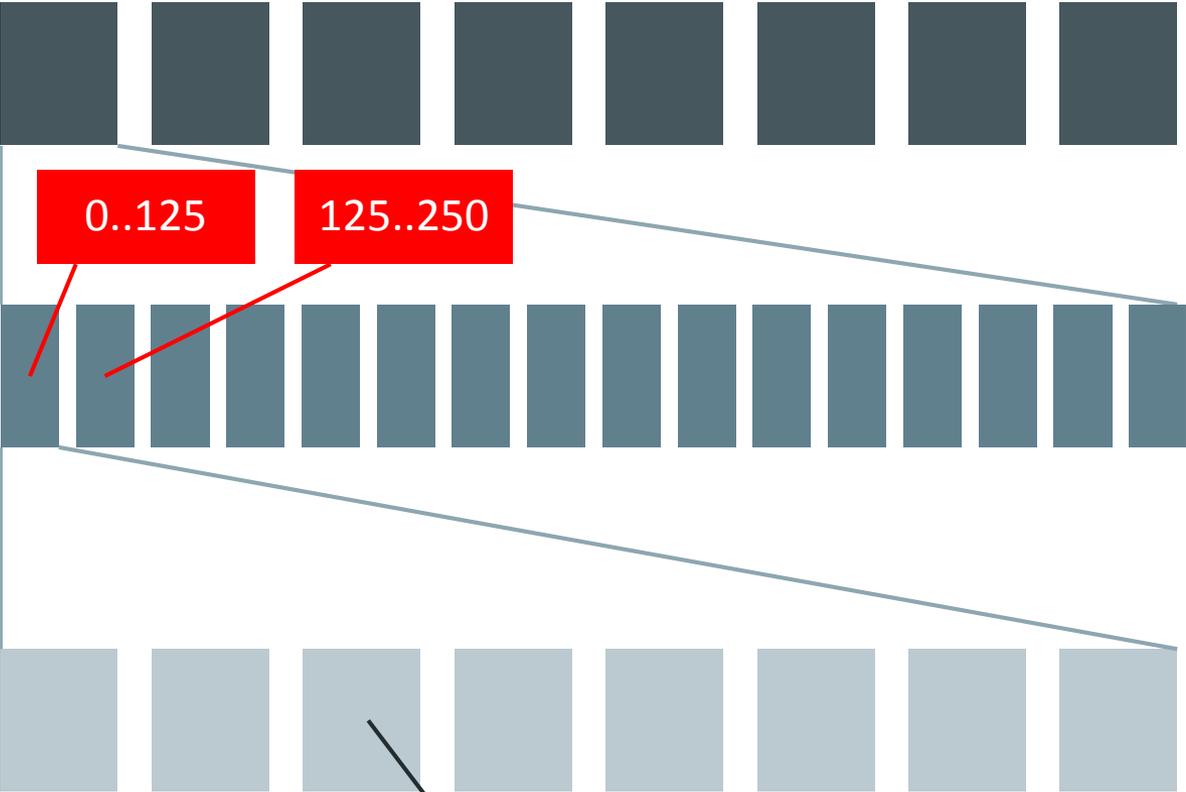
8 sockets

Distribute vertices at
start of loop down to
per-core counters

16 cores per socket

8 h/w threads per core

Consider distributing 0..16000 vertices, batch size 10



8 sockets

Distribute vertices at
start of loop down to
per-core counters

16 cores per socket

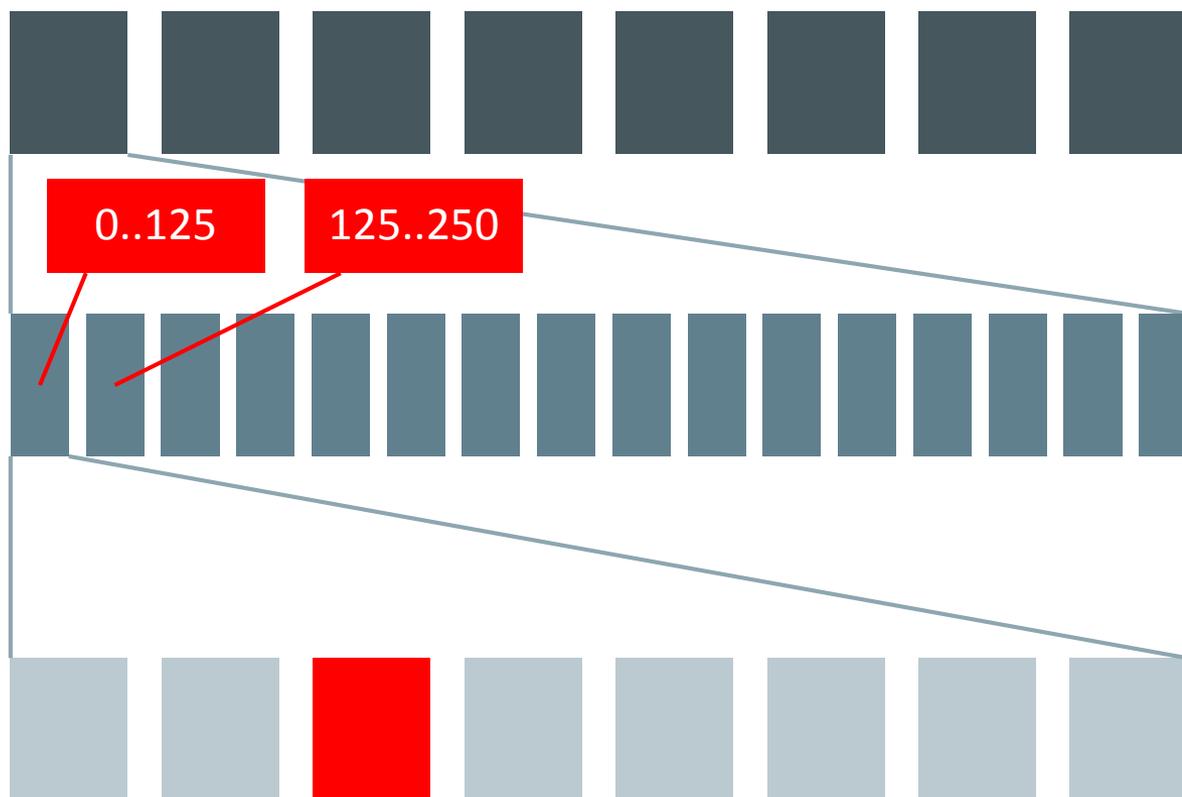
Aggregate requests
upwards within a core

8 h/w threads per core

Per-thread request flags



Consider distributing 0..16000 vertices, batch size 10



8 sockets

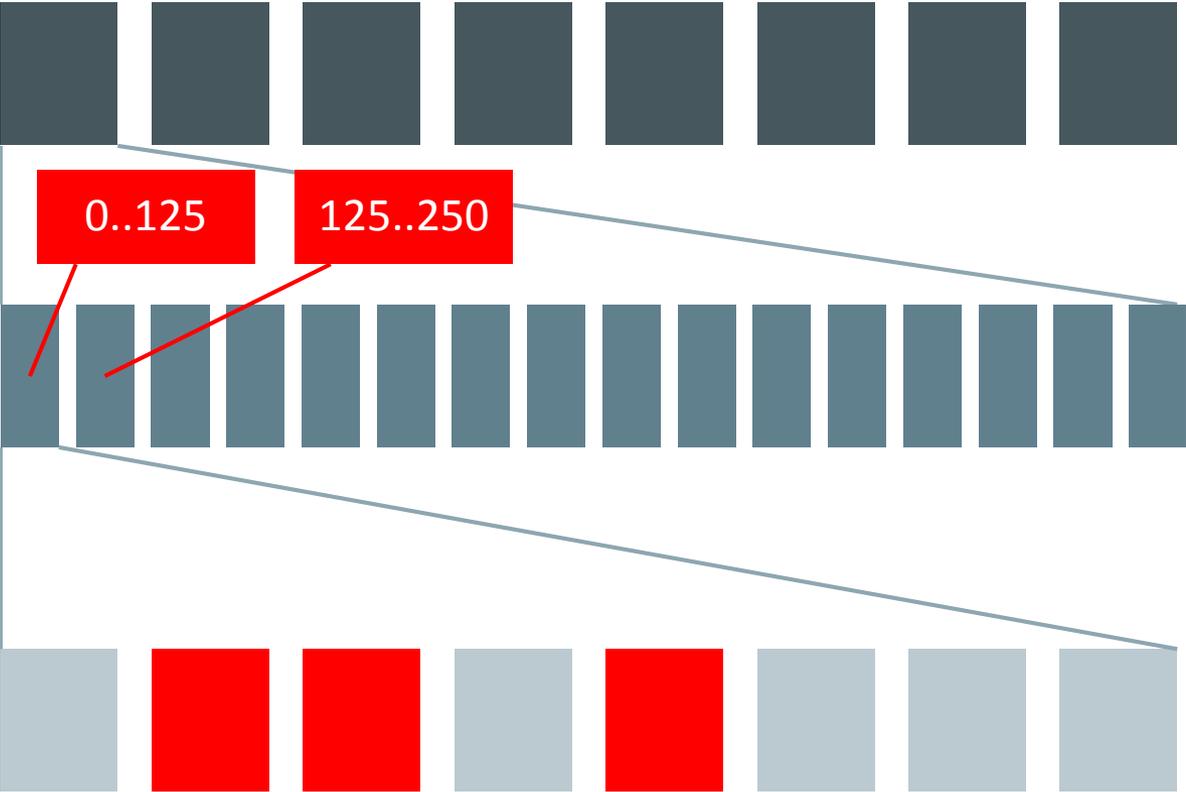
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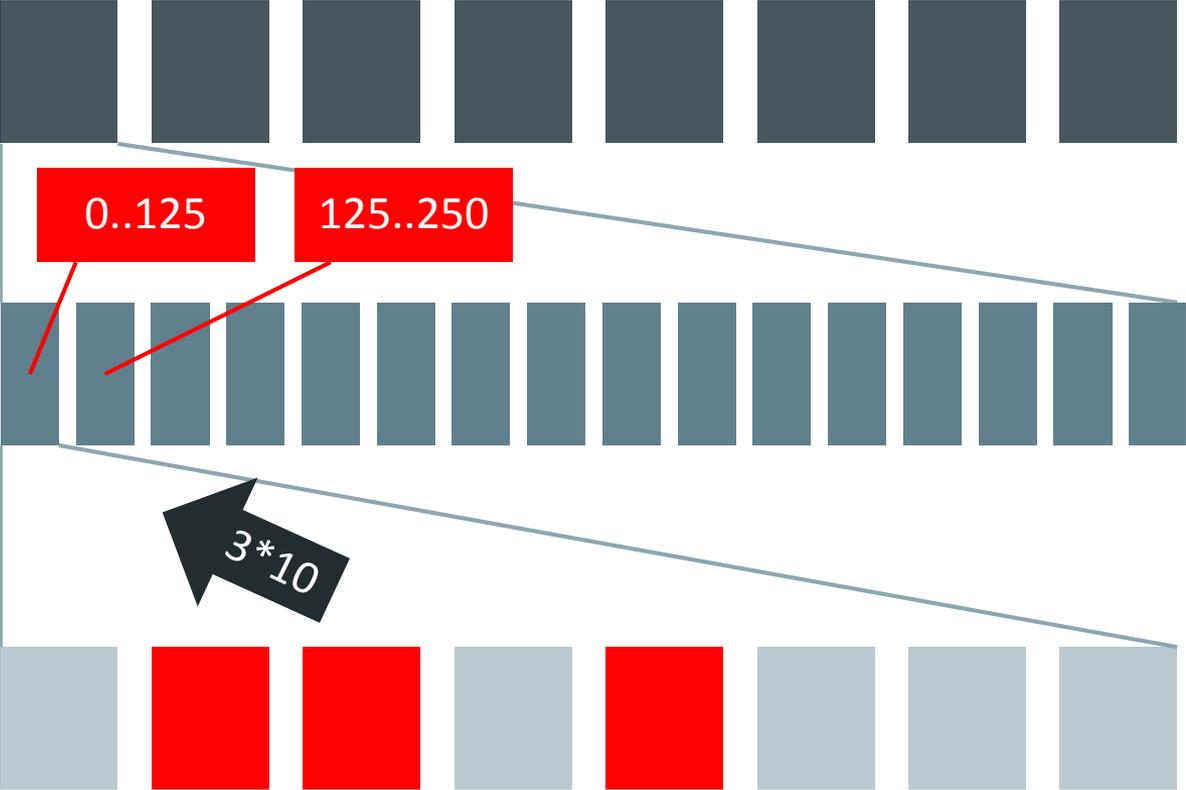
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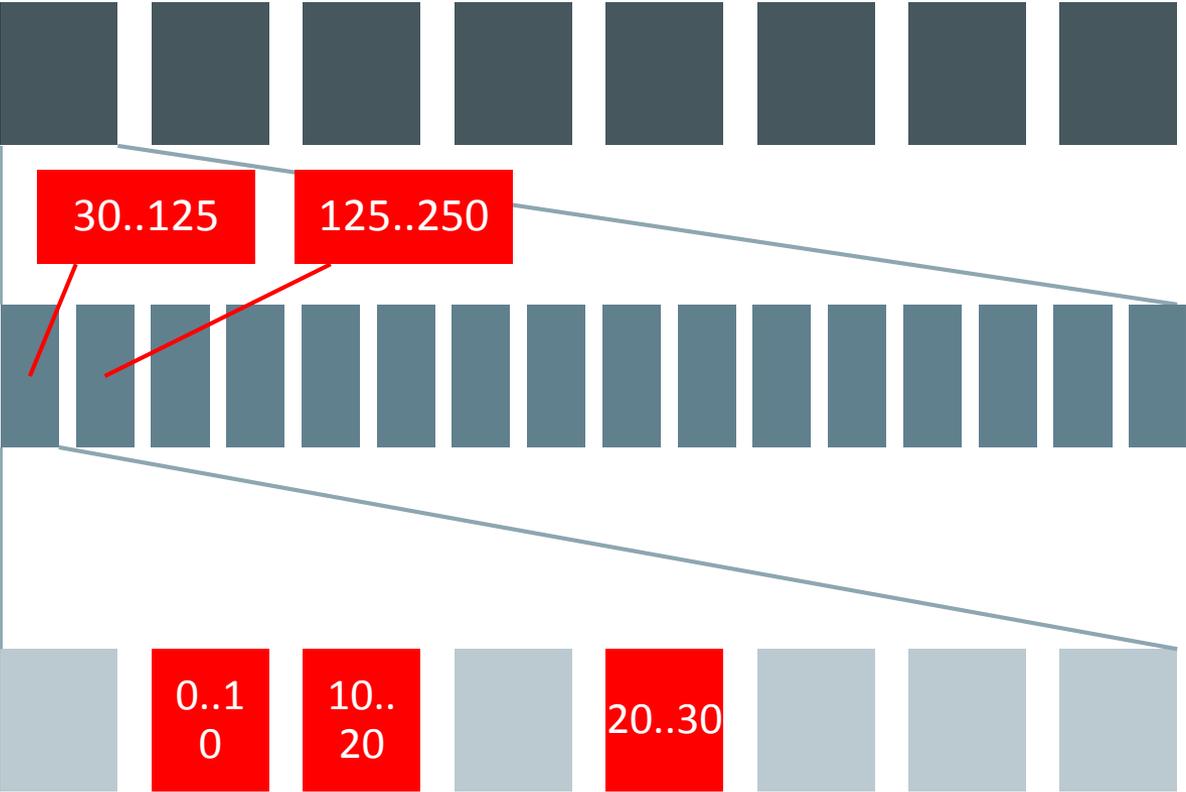
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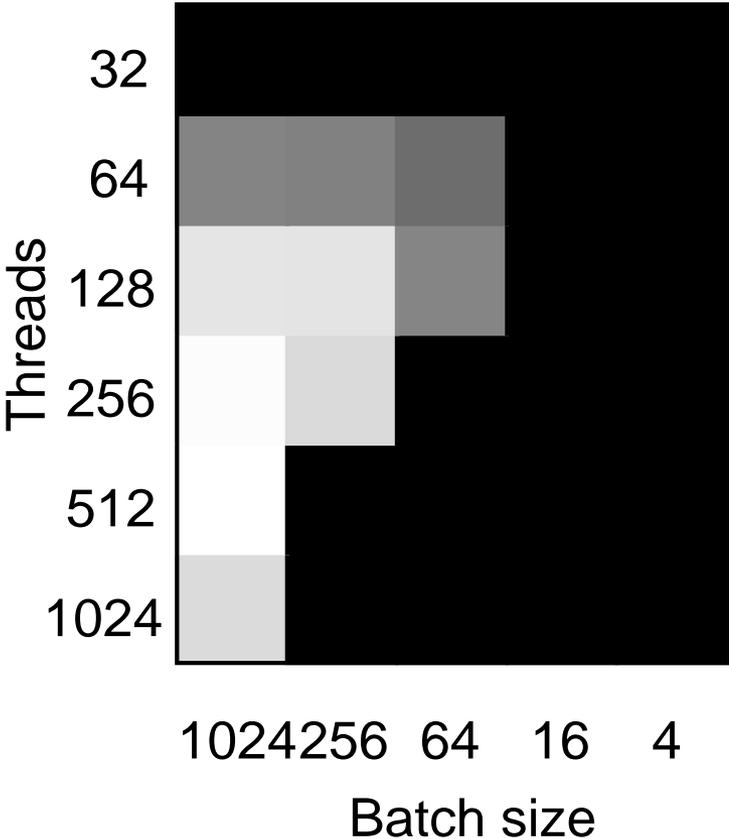
16 cores per socket

Aggregate requests
upwards within a core

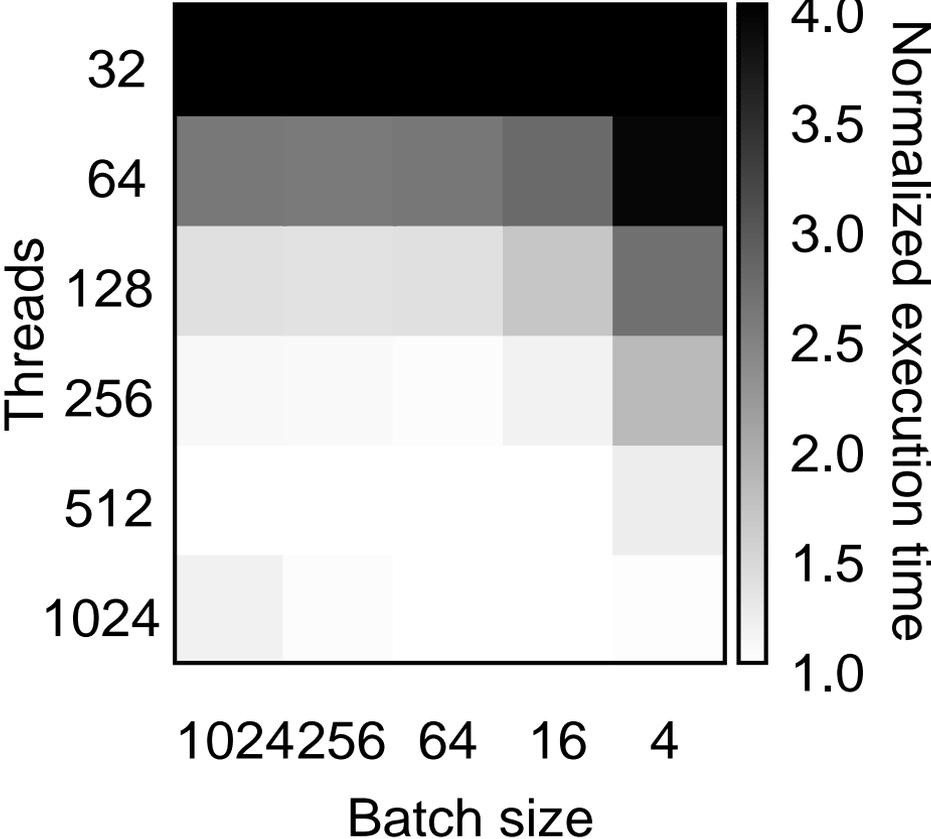
8 h/w threads per core

PageRank – SNAP LiveJournal (4.8M vertices, 69M edges)

Before



After



More details

- [Callisto-RTS: Fine-Grain Parallel Loops](#)
- Tim Harris, Stefan Kaestle, *USENIX ATC 2015*
- <https://timharris.uk/papers/2015-atc-callisto.pdf>