

# Designing multithreaded code for scalability

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# Designing multithreaded code for scalability

- Scalability
- Limitations
- Designing for Scalability

# Scalability

# Scalability

Modern C++ code runs across a wide variety of platforms:

- Embedded single-core microcontrollers
- Embedded multi-core systems
- Multi-core desktop computers
- Multi-core / multi-socket servers
- Many-core / many-socket HPC systems

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- 1 CPU / 1 core
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Modern C++ code runs across a wide variety of platforms:

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- 1 CPU / 4 cores
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# Scalability

Modern C++ code runs across a wide variety of platforms:

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- 1 CPU / 4 cores
- 1 CPU / 8 cores
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- 1 CPU / 4 cores
- 1 CPU / 8 cores
- 4 CPUs / 32 cores per CPU
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- 1 CPU / 4 cores
- 1 CPU / 8 cores
- 4 CPUs / 32 cores per CPU
- 10000 CPUs / 12 cores per CPU

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- 1 CPU / 8 cores
- 4 CPUs / 32 cores per CPU
- 10000 CPUs / 12 cores per CPU
- Plus GPUs — up to 65536 cores

# Scalability

Communicating between threads has different constraints across these systems.

Your code is **scalable** if it can run on any of these systems without penalty.

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Our software needs to be scalable

# Limitations

# Limitations: Mutex contention

Mutex: **M**utual **E**xclusion

A mutex is a means of **preventing** concurrent execution.

*instead of picking up Dijkstra's cute acronym we should have called the basic synchronization object "the bottleneck" (David Butenhof)*

# Limitations: Mutex contention

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*instead of picking up Dijkstra's cute acronym we should have called the basic synchronization object "the bottleneck" (David Butenhof)*

⇒ For scalable solutions, we need to **avoid** mutex contention.

# Limitations: Atomic contention

Atomic operations can suffer from contention too:

**Read-Modify-Write** operations always affect the **latest** values

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⇒ RMW operations on a single location need to be serialized by the CPUs

⇒ For scalable solutions, we need to be sparing with RMW operations

# Limitations: False Sharing

CPUs synchronize memory at the granularity of a cache line.

Cache lines are typically 16-128 bytes

⇒ objects that are on the same cache line are essentially the same object for contention purposes

# Limitations: Cache Ping-Pong





# Limitations: Cache Ping-Pong

**Cache Ping-Pong** is where a cacheline is continuously shuttled back and forth between two processors. This occurs when two threads are accessing either:

- **the same** atomic variable
- **different** variables on **the same cache line**

This can have a **big** performance impact, because transferring cache lines is **slow**.

# Limitations: Speed of Light

The speed of light is  $3 \times 10^8 \text{ m/s}$

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CPU clocks are around 3GHz

⇒ The speed of light is around 10cm/tick

⇒ There is a hard upper limit on communication speed for multi-socket systems

# Limitations: Memory bandwidth

Intel Xeon Phi 7295:

- 115.2Gb/s Memory bandwidth
- 1.5Ghz Clock speed
- 72 Cores

⇒ 76.8 bytes per clock

⇒ 1.1 bytes per clock per core

# Designing for Scalability

# Strategies: Batch Communications

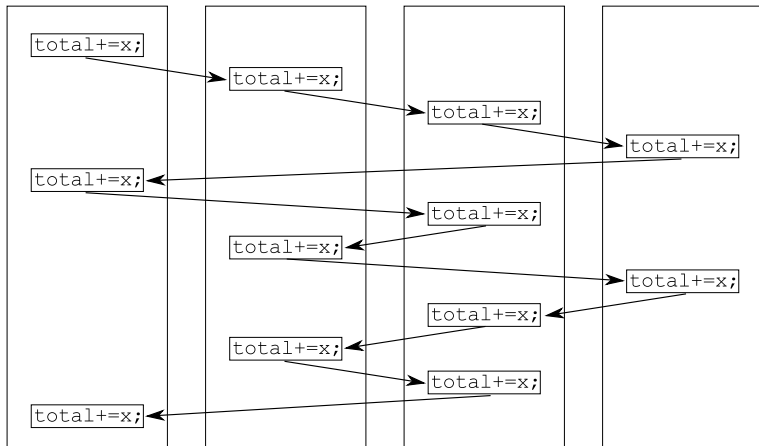
Can you avoid intermediate synchronization?

Each thread works on its own data, and only modifies shared data at the end

# Batch Communication Example

```
std::vector<unsigned> const values=get_values();
std::atomic<unsigned long long> total{0};
unsigned const num_threads=...;
std::vector<joining_thread> threads(num_threads);
for(unsigned t=0;t<num_threads;++t){
    threads[t]=joining_thread([&,t]{
        auto start=...;
        auto end=...;
        std::for_each(start,end,[&](auto x){
            total+=x;
        });
    });
}
```

# Batch Communication Costs

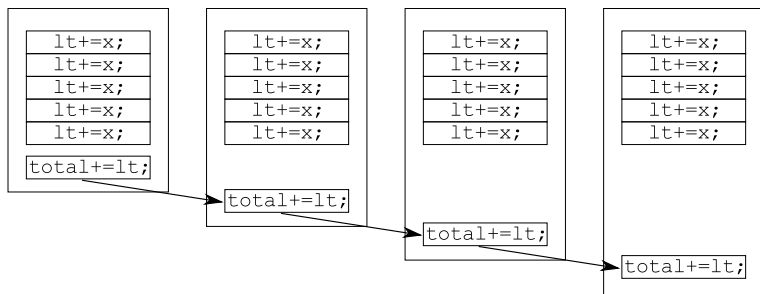




# Batch Communication Example

```
std::vector<unsigned> const values=get_values();
std::atomic<unsigned long long> total{0};
unsigned const num_threads=...;
std::vector<joining_thread> threads(num_threads);
for(unsigned t=0;t<num_threads;++t){
    threads[t]=joining_thread([&,t]{
        auto start=...;
        auto end=...;
        auto local_total=std::accumulate(start,end,0ull);
        total+=local_total;
    });
}
```

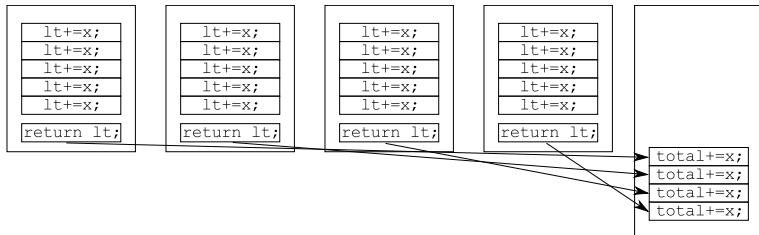
# Batch Communication Costs



# Batch Communication Example

```
std::vector<unsigned> const values=get_values();
unsigned const num_threads=...;
std::vector<std::future<unsigned long long>> futures(
    num_threads);
for(unsigned t=0;t<num_threads;++t){
    futures[t]=std::async(std::launch::async, [&,t]{
        auto start=...;
        auto end=...;
        return std::accumulate(start,end,0ull);
    });
}
unsigned long long total=0;
for(auto& f:futures) total+=f.get();
```

# Batch Communication Costs



# Batch Communication Costs

Sum of 100000000 elements on 4 threads:

Run	Time	Ratio to serial
Serial	0.081s	1
All atomic	9.74s	<b>120x slower!</b>
End atomic	0.052s	<b>1.6x faster</b>
Futures	0.052s	<b>1.6x faster</b>

# Contended Lists

Suppose we have a linked list, accessible by multiple threads, and we might need to add or remove elements. What can we do?

- Use a mutex for the whole list
- Use a mutex for each link in the list
- Use `std::atomic<std::shared_ptr<Node> >` for the node links
- Use `std::atomic<Node*>` for the node links, and a **Safe Reclamation** scheme to ensure Nodes can be removed safely

# Contended Lists: Costs

- Whole list mutex  $\implies$  **big bottleneck**
- Node mutex  $\implies$  lots of small bottlenecks
- `std::atomic<std::shared_ptr<Node> >`  $\implies$  spin-locks, or RMW operations
- `std::atomic<Node>`  $\implies$  low-cost for readers, **big cost** for writers

# Safe Reclamation Options

- Garbage Collection
- RCU
- Hazard Pointers

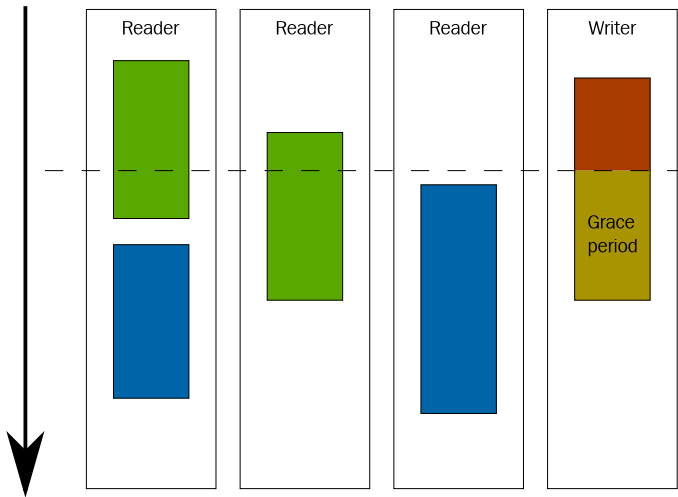


# Safe Reclamation: RCU

Readers just record entry/exit to the read function.

Writers make atomic changes, then wait for a **grace period** before deleting removed objects.

# Safe Reclamation: RCU



# RCU costs

In user space:

- Read side:
  - Atomic read of global marker
  - Two atomic writes to a per-thread location
- Write side:
  - Atomic write of global marker
  - Multiple atomic reads of **all** per-thread locations for readers
  - Mutex locks, delays and spin-loops until all readers ready

In kernel space:

- Read side: **no overhead!**
- Write side:
  - Blocking wait until all processors have cycled a time slice

# Safe Reclamation: Hazard Pointers

Readers store **hazard pointers** referring to objects being accessed

Writers make atomic changes, then check the **hazard pointers** to see if it is safe to delete an object.

# Hazard Pointers Costs

- Read side:
  - Two (or more) atomic writes to a per-thread hazard pointer
  - Spin-loop ensuring value hasn't changed while updating hazard pointer
- Write side:
  - Atomic RMW operation adding to reclamation list
  - Objects not immediately destroyed
  - Period reclamation checks: when **N** objects are queued for reclamation
    - N depends on configuration parameters and number of threads
    - Each reclamation does atomic reads of **all** per-thread hazard pointers for readers
    - Cost of retiring objects varies by orders of magnitude

# Standard Support for Safe Reclamation

There is a proposal under discussion for both RCU and Hazard Pointers, with a sample implementation:

P0566R4: Proposed Wording for Concurrent Data Structures:  
Hazard Pointer and Read-Copy-Update (RCU)

<http://wg21.link/p0566>

RCU implementation:

<https://github.com/paulmckrcu/RCUCPPbindings>

Hazard Pointer implementation:

<https://github.com/facebook/folly/tree/master/folly/experimental/hazptr>

# Sequential Consistency vs Eventual Consistency

## Sequential Consistency:

- All threads see the same view of shared state
- Single Total Order of operations
- This requires serialization, or extensive communication

## Eventual Consistency:

- Threads may see different views of shared state  
Provided each thread has a **self-consistent** view all is well
- All changes propagate to all threads **eventually**
- Cannot write a Single Total Order of operations
- Much less communication required

# Sequential Consistency vs Eventual Consistency

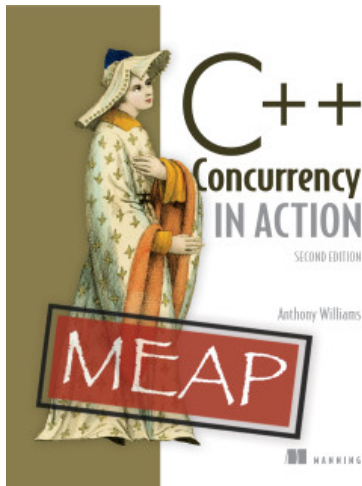
**Sequential Consistency** is easier to reason about.  
**Eventual Consistency** is more scalable.



# Summary

- Multithreaded code needs to be scalable
- Avoid contention
- Avoid cache ping-pong
- Use Safe Reclamation schemes
- Use Eventual Consistency

# My Book



C++ Concurrency in Action:  
Practical Multithreading,  
**Second Edition**

Covers C++17 and the  
Concurrency TS

Early Access Edition now  
available

<http://stdthread.com/book>

# Just::Thread Pro



`just::thread` Pro provides an actor framework, a concurrent hash map, a concurrent queue, synchronized values and a complete implementation of the C++ Concurrency TS, including a lock-free implementation of `atomic_shared_ptr` and RCU.

<http://stdthread.co.uk>

Questions?