ECE408/CS483/CSE408 Fall 2022

**Applied Parallel Programming** 

Lecture 18
Atomic Operations and
Histogramming

#### Course Reminders

- MP5.2 is due this week
- Project Milestone 2: Baseline Convolution Kernel
  - Due November 4th

# Objective

- To understand atomic operations
  - Read-modify-write in parallel computation
  - A primitive form of "critical regions" in parallel programs
  - Use of atomic operations in CUDA
  - Why atomic operations reduce memory system throughput
  - How to avoid atomic operations in some parallel algorithms
- To learn practical histogram programming techniques
  - Basic histogram algorithm using atomic operations
  - Atomic operation throughput
  - Privatization

#### A Common Collaboration Pattern

- Multiple bank tellers count the total amount of cash in the safe
- Each grab a pile and count
- Have a central display of the running total
- Whenever someone finishes counting a pile, add the subtotal of the pile to the running total
- A bad outcome
  - Some of the piles were not accounted for

#### A Common Arbitration Pattern

- Multiple customers booking air tickets
- Each
  - Brings up a flight seat map
  - Decides on a seat
  - Update the the seat map, mark the seat as taken
- A bad outcome
  - Multiple passengers ended up booking the same seat

# Read-Modify-Write Operations

thread1: Old  $\leftarrow$  Mem[x]

New  $\leftarrow$  Old + 1

 $Mem[x] \leftarrow New$ 

thread2: Old  $\leftarrow$  Mem[x]

New  $\leftarrow$  Old + 1

 $Mem[x] \leftarrow New$ 

If Mem[x] was **initially 0**, what would the value of Mem[x] be after threads 1 and 2 have completed?

- What does each thread get in their Old variable?

The answer may vary due to data races. To avoid data races, you should use atomic operations

Time	Thread 1	Thread 2
1	$(0) Old \leftarrow Mem[x]$	
2	(1) New $\leftarrow$ Old + 1	
3	(1) $Mem[x] \leftarrow New$	
4		(1) Old $\leftarrow$ Mem[x]
5		(2) New ← Old + 1
6		(2) $Mem[x] \leftarrow New$

- Thread 1 Old = 0
- Thread 2 Old = 1
- Mem[x] = 2 after the sequence

Time	Thread 1	Thread 2
1		$(0)  \text{Old} \leftarrow \text{Mem}[x]$
2		(1) New $\leftarrow$ Old + 1
3		(1) $Mem[x] \leftarrow New$
4	$(1) Old \leftarrow Mem[x]$	
5	(2) New ← Old + 1	
6	(2) $Mem[x] \leftarrow New$	

- Thread 1 Old = 1
- Thread 2 Old = 0
- Mem[x] = 2 after the sequence

Time	Thread 1	Thread 2
1	$(0) Old \leftarrow Mem[x]$	
2	(1) New $\leftarrow$ Old + 1	
3		$(0) Old \leftarrow Mem[x]$
4	(1) $Mem[x] \leftarrow New$	
5		(1) New $\leftarrow$ Old + 1
6		(1) $Mem[x] \leftarrow New$

- Thread 1 Old = 0
- Thread 2 Old = 0
- Mem[x] = 1 after the sequence

Time	Thread 1	Thread 2
1		$(0) Old \leftarrow Mem[x]$
2		(1) New $\leftarrow$ Old + 1
3	$(0) Old \leftarrow Mem[x]$	
4		(1) $Mem[x] \leftarrow New$
5	(1) New ← Old + 1	
6	(1) $Mem[x] \leftarrow New$	

- Thread 1 Old = 0
- Thread 2 Old = 0
- Mem[x] = 1 after the sequence

# Atomic Operations Prevent Interleaving

```
thread1: Old \leftarrow Mem[x]
New \leftarrow Old + 1
Mem[x] \leftarrow New
```

thread2: Old  $\leftarrow$  Mem[x] New  $\leftarrow$  Old + 1 Mem[x]  $\leftarrow$  New

Or

thread2: Old  $\leftarrow$  Mem[x] New  $\leftarrow$  Old + 1 Mem[x]  $\leftarrow$  New

thread1: Old  $\leftarrow$  Mem[x] New  $\leftarrow$  Old + 1 Mem[x]  $\leftarrow$  New

# Without Atomic Operations

Mem[x] initialized to 0

thread1: Old  $\leftarrow$  Mem[x]

New ← Old + 1

 $Mem[x] \leftarrow New$ 

thread2: Old  $\leftarrow$  Mem[x]

New  $\leftarrow$  Old + 1

 $Mem[x] \leftarrow New$ 

- Both threads receive 0
- Mem[x] becomes 1

#### Needed When Threads Write to Same Location

#### When two threads

- may write to the same memory location,
- the program may need atomic operations.

#### Sharing is not always easy to recognize...

- Do two insertions into a hash table share data?
- What about two graph node updates based on all of the nodes' neighbors?
- What if nodes are on same side of bipartite graph?

# What Exactly is "Atomic?"

To a high-energy photon, atoms are not.

Atomicity is ALWAYS with respect to something.

#### Two sections of code

- that execute atomically with respect to one another
- appear to the software as though
- the programs' execution did not interleave at all.

#### What Can Go Wrong?

#### Common failure mode:

- Programmer thinks operations are independent.
- Hasn't considered input data for which they are not.
- Or another programmer reuses code without understanding assumptions that imply independence.

Also: atomicity does not constrain relative order.

### Implementing Atomic Operations

- Many ISAs offer synchronization primitives,
  - instructions with one (or more) address operands
  - that execute atomically with respect to
     one another when used on the same address.
- Mostly read, modify, write operations
  - Bit test and set
  - Compare and swap / exchange
  - Swap / exchange
  - Fetch and increment / add

# Atomicity Enforced by Microarchitecture

When synchronization primitives execute,

- hardware ensures that no other thread
- accesses the location until the operation is complete.

Other threads that access the location

- are typically stalled or held in a queue until their turn.
- Threads perform atomic operations serially.

### Atomic Operations in CUDA

- Function calls that are translated into single ISA instructions (a.k.a. *intrinsics*)
  - Atomic add, sub, inc, dec, min, max, exch (exchange), CAS (compare and swap)
  - Read CUDA C programming Guide for more details

#### Atomic Add

int atomicAdd(int\* address, int val);

reads the 32-bit word **old** pointed to by **address** in global or shared memory, computes **(old + val)**, and stores the result back to memory at the same address. The function returns **old**.

#### More Atomic Adds in CUDA

- Unsigned 32-bit integer atomic add unsigned int atomicAdd(unsigned int\* address, unsigned int val);
- Unsigned 64-bit integer atomic add unsigned long long int atomicAdd(unsigned long long int\* address, unsigned long long int val);
- Single-precision floating-point atomic add (capability > 2.0)
  - float atomicAdd(float\* address, float val);

# Building synchronization with atomics

- How would we build syncthreads () for block?
- How would we create syncthreads () for entire grid?
  - And why would this not be a good idea?
- How would we create a critical section? I.e., one thread per block executing a particular section of code?

- How would we create a critical section per grid?
  - Why doesn't this have the same issue as \_\_syncthreads() for grid?

# Atomic Compare and Swap (CAS)

```
Bool atomicCAS(int *address, int old, int new)
    if (*address != old)
        return false;
    *address = new;
    return true;
```

#### Atomic Add using CAS

```
int atomicAdd(int* address, int value)
   Bool done = false;
   while (!done) {
        old v = *address;
       done = atomicCAS(address, old v, old v + value);
    return old v;
```

# Histogramming

- A method for extracting notable features and patterns from large data sets
  - Feature extraction for object recognition in images
  - Fraud detection in credit card transactions
  - Correlating heavenly object movements in astrophysics

**—** ...

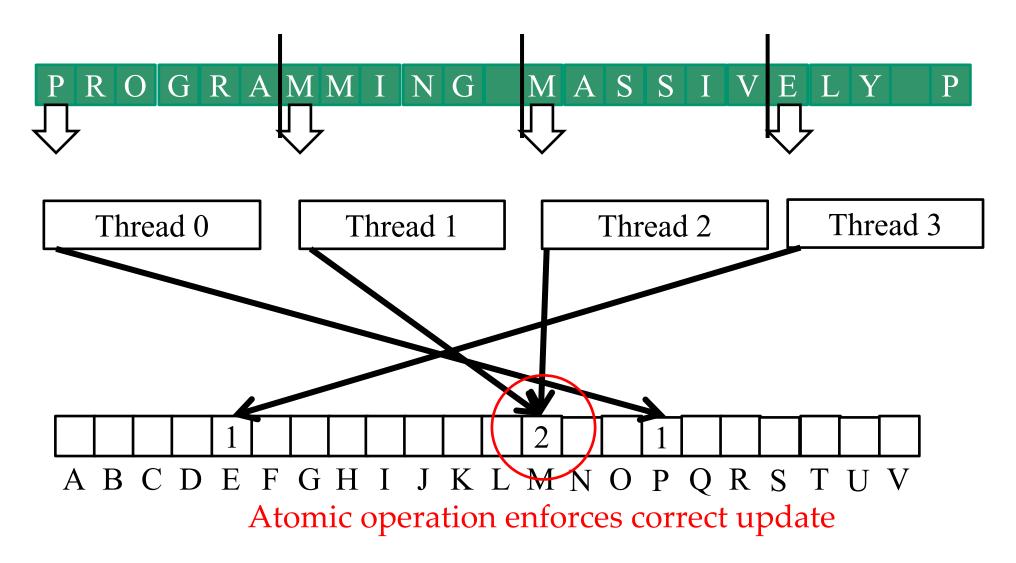
• Basic histograms - for each element in the data set, use the value to identify a "bin" to increment

# A Histogram Example

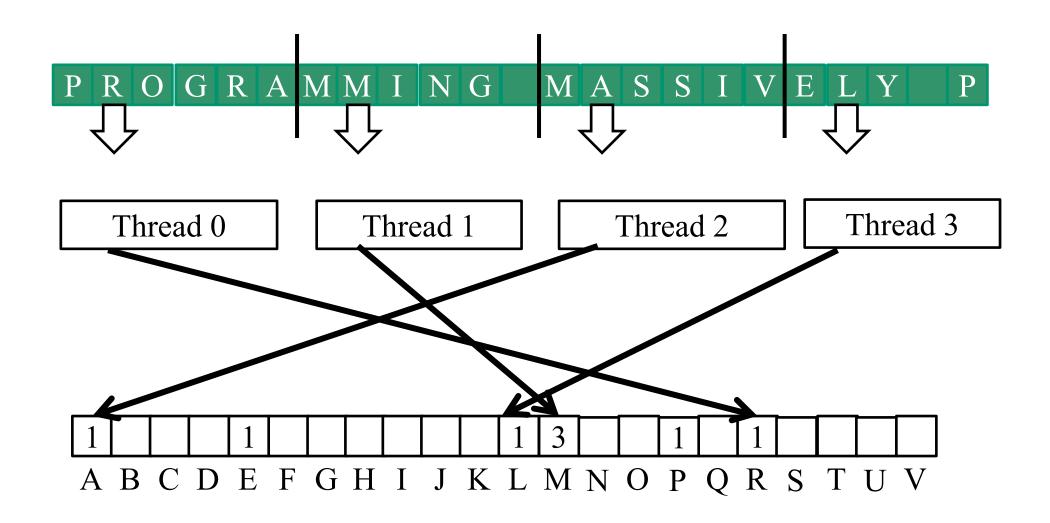
- In sentence "Programming Massively Parallel Processors" build a histogram of frequencies of each letter
- A(4), C(1), E(1), G(1), ...

• How do you do this in parallel?

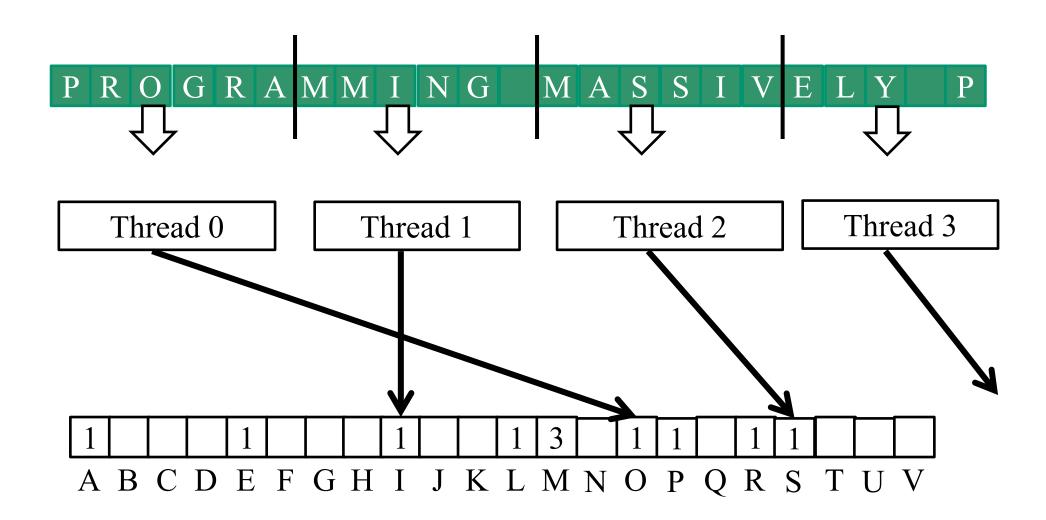
#### Iteration $#1 - 1^{st}$ letter in each section



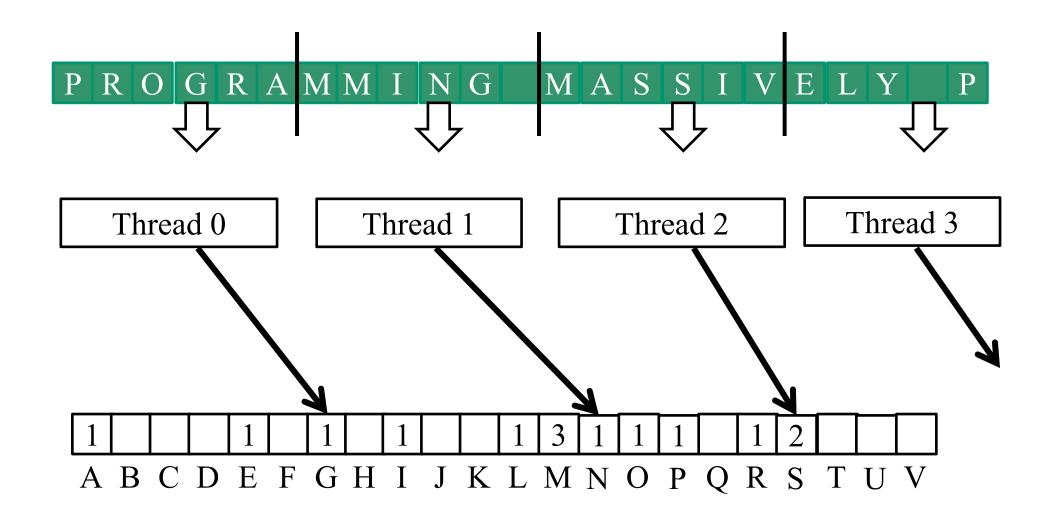
#### Iteration $\#2 - 2^{nd}$ letter in each section



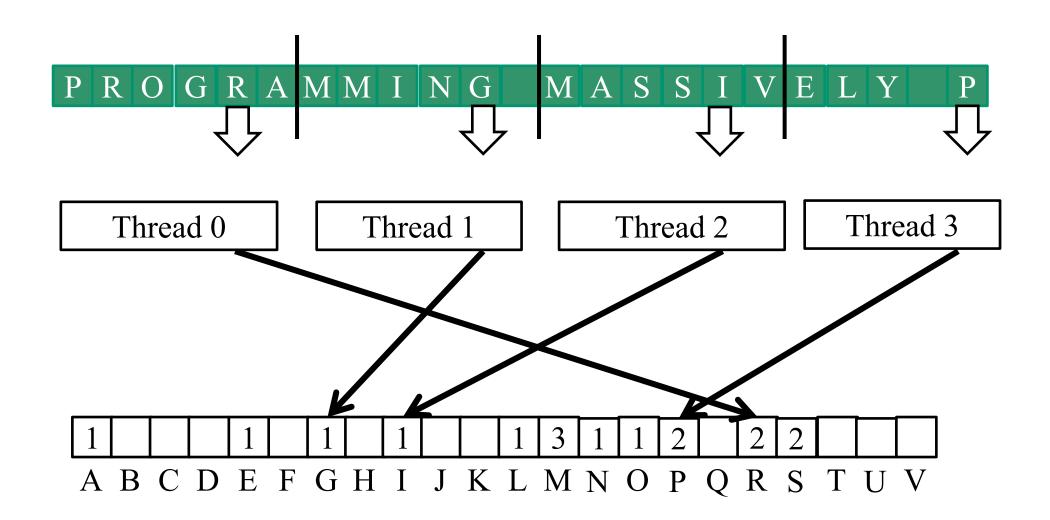
#### Iteration #3



#### Iteration #4

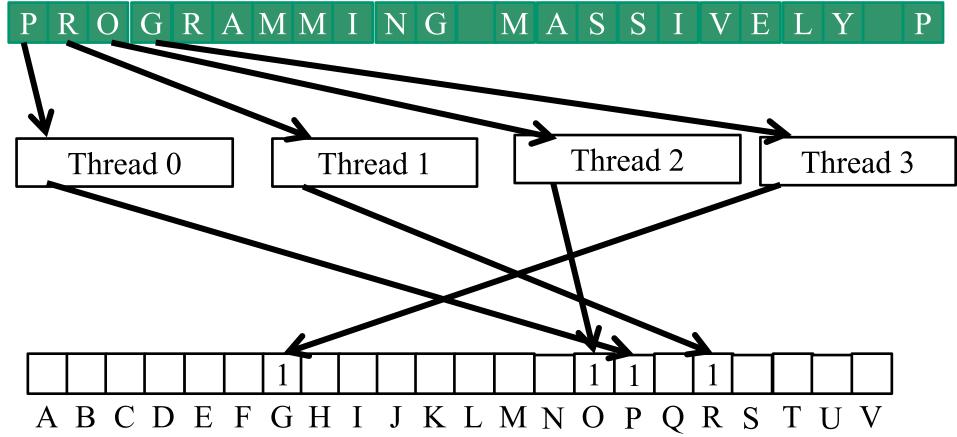


#### Iteration #5



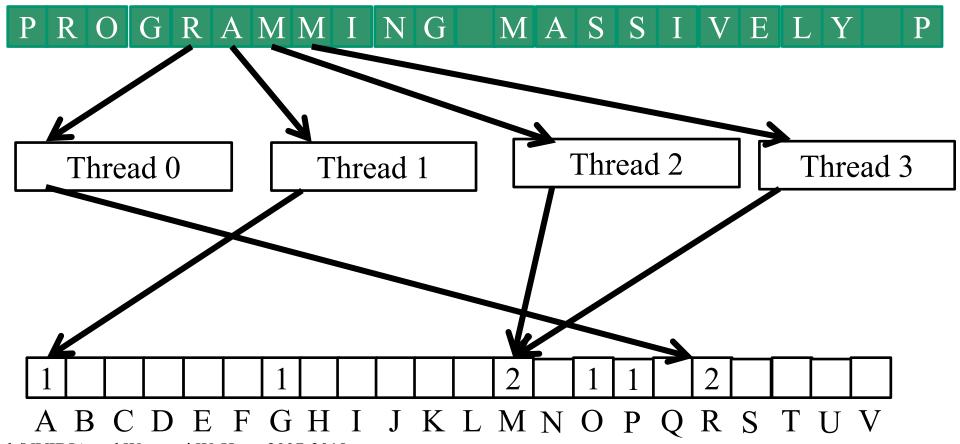
#### A better approach

- Reads from the input array are not coalesced
  - Assign inputs to each thread in a strided pattern
  - Adjacent threads process adjacent input letters



#### Iteration 2

• All threads move to the next section of input



#### A Histogram Kernel

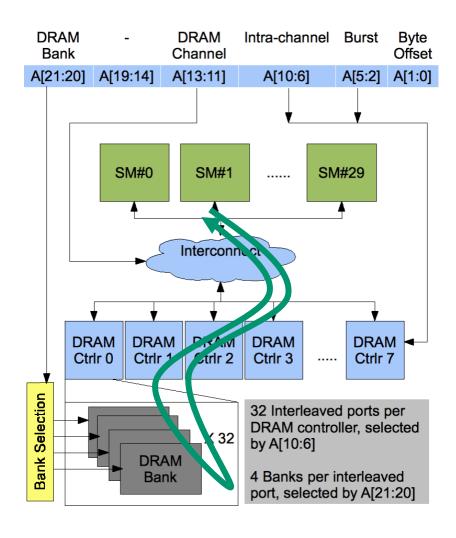
- The kernel receives a pointer to the input buffer
- Each thread process the input in a strided pattern

### More on the Histogram Kernel

```
// All threads in the grid collectively handle
// blockDim.x * gridDim.x consecutive elements

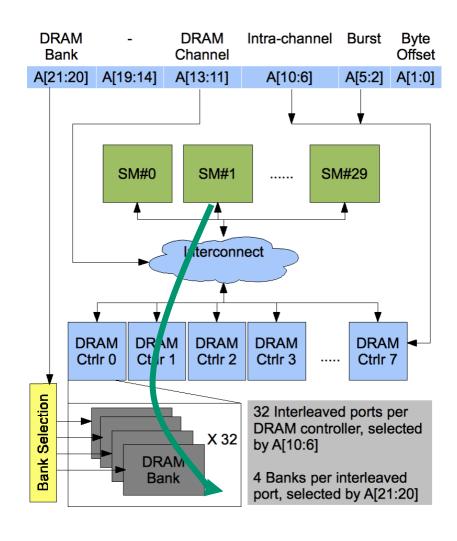
while (i < size) {
   atomicAdd( &(histo[buffer[i]]), 1);
   i += stride;
}</pre>
```

#### Atomic Operations on DRAM



• An atomic operation starts with a read, with a latency of a few hundred cycles

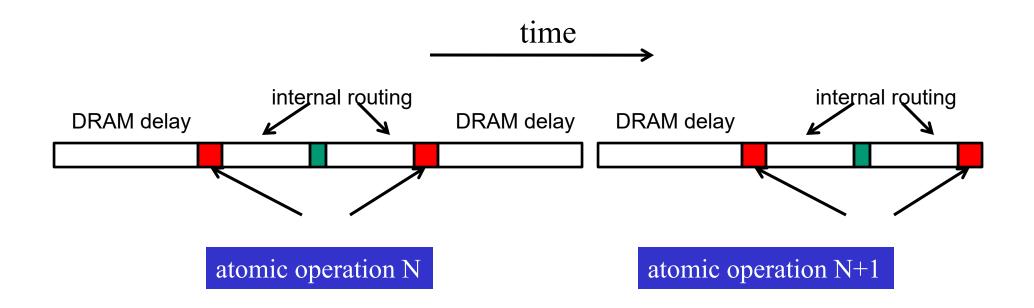
#### Atomic Operations on DRAM



- An atomic operation starts with a read, with a latency of a few hundred cycles
- The atomic operation ends with a write, with a latency of a few hundred cycles
- During this whole time, no one else can access the location

#### Atomic Operations on DRAM

- Each Load-Modify-Store has two full memory access delays
  - All atomic operations on the same variable (RAM location) are serialized



#### Latency determines throughput of atomic operations

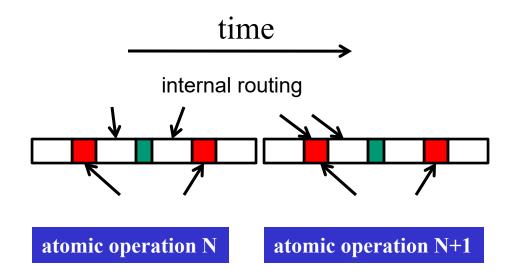
- Throughput of an atomic operation is the rate at which the application can execute an atomic operation on a particular location.
- The rate is limited by the total latency of the read-modify-write sequence, typically more than 1000 cycles for global memory (DRAM) locations.
- This means that if many threads attempt to do atomic operation on the same location (contention), the memory bandwidth is reduced to < 1/1000!

# You may have a similar experience in supermarket checkout

- Some customers realize that they missed an item after they started to check out
- They run to the isle and get the item while the line waits
  - The rate of check is reduced due to the long latency of running to the isle and back.
- Imagine a store where every customer starts the check out before they even fetch any of the items
  - The rate of the checkout will be 1 / (entire shopping time of each customer)

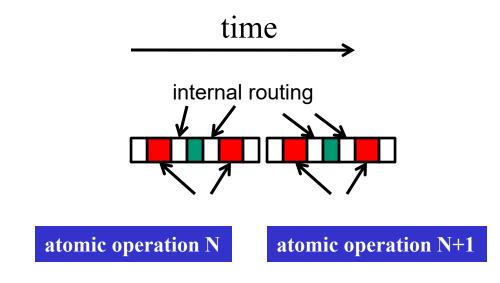
# Hardware Improvements

- Atomic operations on L2 cache
  - medium latency, but still serialized
  - Global to all blocks
  - "Free improvement" on Global Memory atomics



# Hardware Improvements

- Atomic operations on Shared Memory
  - Very short latency, but still serialized
  - Private to each thread block
  - Need algorithm work by programmers (more later)



#### Atomics in Shared Memory Requires Privatization

• Create private copies of the histo[] array for each thread block

```
global
void histo kernel (unsigned char *buffer,
                      long size, unsigned int *histo) {
      shared unsigned int histo private [256];
   // warning: this will not work correctly if there are fewer than 256 threads!
   if (threadIdx.x < 256)
        histo private[threadIdx.x] = 0;
     syncthreads();
```

# Build Private Histogram

```
int i = threadIdx.x + blockIdx.x * blockDim.x;
// stride is total number of threads
int stride = blockDim.x * gridDim.x;
while (i < size) {</pre>
     atomicAdd( &(private histo[buffer[i]), 1);
     i += stride;
```

### Build Final Histogram

#### More on Privatization

- Privatization is a powerful and frequently used techniques for parallelizing applications
- The operation needs to be associative and commutative
  - Histogram add operation is associative and commutative
- The histogram size needs to be small
  - Fits into shared memory
- What if the histogram is too large to privatize?

### ANY MORE QUESTIONS READ CHAPTER 9