Performance Evaluation of Lock-free Data Structures on GPUs

http://www.cse.iitk.ac.in/~mainakc/lockfree.html

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- Result highlights
- Related work
- Lock-free data structures
- CUDA implementation
- Evaluation methodology
- Empirical results
- Summary



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Talk in One Slide

- Locks are expensive in GPUs

 Thousands of threads cause high contention
- Lock-free data structures offer a possible way to implement irregular computations on GPUs
 - Support for dynamically changing pointer-linked data structures is important in many applications
- Large body of existing research on lock-free data structures for traditional multiprocessors
- This is the first detailed study to explore lock-free linear lists, hash tables, skip lists, and priority queues on CUDA-enabled GPUs

Result highlights

- Significant speedup on Tesla C2070 (Fermi GF100) over 24-core server execution
- Maximum speedup
 - 7.4x for linear lists
 - 11.3x for hash tables
 - 30.7x for skip list
 - 30.8x for priority queue
- Lock-free hash table shows best scalability for a wide range of operation mixes and key ranges
 - Throughput ranges from 20.8 MOPS to 98.9 MOPS



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Related work

- Our lock-free linear list implementation follows a variation of the Harris-Michael construction
- Our hash table implementation leverages the linear list implementation
- Our lock-free skip list construction is due to Herlihy, Lev, and Shavit
- We follow the construction due to Lotan and Shavit for our lock-free priority queue implementation
- More related works are discussed in the paper



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Lock-free data structures

- Linear lists, hash tables, skip lists, priority queues
 - Important computation building blocks
 - We implement a set using these data structures
 - Lock-free and wait-free operations on the set
 - Lock-free operation: infinitely often some instance of this operation finishes in finite number of steps
 - Wait-free operation: every instance of this operation finishes in finite number of steps
 - Correctness criteria: linearizable (except priority queue, which is quiescently consistent)
 - See paper for definition

Lock-free linear list

- Implemented using a sorted singly linked list
- Supported ops: add, delete, search
 - Add(x) returns 0 if x is already in the set;
 otherwise adds x at sorted position and returns 1
 - Delete(x) returns 0 if x is not in the set;
 otherwise removes x from the set and returns 1
 - Search(x) returns 0 or 1 if x is not found or found in the set
 - Add and delete are lock-free
 - Search is wait-free (just walks the list)
 - Delete only logically deletes a node by marking it
 - Subsequent add and delete operations physically remove the logically deleted nodes



The Mark bit is the least significant bit of the aligned 32-bit next field

Needed for logical deletion



Delete(x) logically marks a node. Subsequent add or delete physically deletes it when walking the list.

Lock-free hash table

- Leverages lock-free linear list construction
 - Implemented as a single linear list
 - An array of pointers stores the starting point (head node) of each bucket
 - The head node of each bucket stores a special key
 - Add, delete, and search operations on a bucket start at the head node of that bucket
 - Number of buckets is constant and fixed at the time of CUDA kernel launch
- Supports the same three operations as lockfree linear list

Lock-free hash table

Delete(22): 22 mod 4 = 2





- A skip list is a hierarchy of linear lists
 - Keys present in level n+1 form a subset of the keys present in level n
 - Given that a key is present in level n, there is a probability p of finding the key in level n+1
 - When a new key is inserted, the maximum level up to which this key can be present is decided by a random number r with expected value 1/(1-p)
- A skip list offers expected logarithmic search complexity

Skip list



- Keys are kept sorted at the lowest-level list

 Head and tail nodes maintain the smallest and
 largest keys
- Upper-level lists provide probabilistic shortcuts into the lower-level lists leading to an expected logarithmic search time

Lock-free skip list

- Leverages lock-free linear list implementation
- Additional complications in linking up or removing multiple nodes in different lists
 - Not possible to make multiple Mark+next field modifications atomic using single-word CAS
 - Depending on the traversal path of Add and Delete some middle level node of a marked key may get physically removed while leaving the other levels unchanged: violates the subset property

Lock-free skip list

- Add is made linearizable by adding a key bottom-up
- Delete is made linearizable by logically marking the levels of the key to be deleted top-down
- A key is defined to be present in the set if it is found unmarked in the lowest-level list
- Two major performance bottlenecks
 - Large number of CAS operations
 - Complex code structure leading to significant volume of control flow divergences

Lock-free priority queue

- Supports two operations on the underlying set: Add and DeleteMin
- Leverages lock-free skip list due to its logarithmic search complexity guarantee
 - Makes the Add operation to have expected logarithmic time
 - DeleteMin walks the lowest-level list until an unmarked key is found, which it marks logically using CAS and calls Delete of skip list on that key
- New performance bottleneck
 - Heavy contention near the head due to concurrent DeleteMin operations



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CUDA implementation

- Extensive use of atomicCAS
- All data structures use a generic node class
 All of them build on the basic linear list
- Large number of nodes are pre-allocated
 - Pointers to these are stored in an array
 - A global index points to the next free node
 - An Add operation executes an atomicInc on this index and uses the node pointed to by the pointer at the returned index
- Deleted nodes are not reused
 - Requires an implementation of an elaborate solution to the ABA problem
 - Left to future research



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Evaluation methodology

- Experiments are done on two platforms
 - Tesla C2070 card featuring one GF100 Fermi GPU
 - 14 streaming multiprocessors (SM), each having 32 CUDA cores; thread blocks map to SMs
 - 1.15 GHz core frequency and 1.49 GHz memory frequency
 - 48 KB shared memory and 16 KB L1 cache per thread block; 768 KB globally shared L2 cache
 - Quad processor SMP, each processor having six cores (Intel X7460 CPU) running at 2.66 GHz
 - 16 MB L3 cache shared by six cores in each processor
 - Lock-free implementations use POSIX threads and rely on x86 cmpxchg instruction for realizing the atomicCAS primitive

Evaluation methodology

- Each data structure is evaluated on
 - A range of integer keys [0, 100), [0, 1000), [0, 10000), and [0, 100000)
 - Keys are generated uniformly at random from the range; these are input arguments to the operations
 - Two different mixes of supported operations
 - Different number of operations ranging from 10000 to 100000 in steps of 10000
- Number of thread blocks and threads per block for the CUDA kernel are optimized
 - In most cases, the number of thread blocks is such that each thread carries out one operation
 - 64 threads per block for linear list and 512 threads per block for the rest

Evaluation methodology

 For evaluation on CPU, thread count that offers the best performance is picked

– 24 threads do not always offer the best

- In summary, each experiment shows results using the best performance on the GPU as well as on the CPU
- Lock-free hash table uses ten thousand buckets
- Lock-free skip list uses p=0.5 and 32 levels
 - Lock-free priority queue leverages the lock-free skip list that uses the same parameters



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Lock-free linear list



No major difference between search-heavy and add/delete-heavy op strings



Lock-free hash table



Consistent speedup across all key Baster enduppentixes



Lock-free skip list



Lock-free priority queue



Trends are similar to skip list: speedup increases with Add %





The data are shown for the largest key range

On GPU, the hash table is 36x to 538x faster than linear list
On CPU, the hash table is only 8x to 54x faster than linear list

GPU exposes more concurrency in the lock-free hash table



- ✤ On GPU, the skip list is 2x to 20x faster than linear list
- ✤ GPU still exposes more concurrency than CPU for skip list
- ✤ Hash table shows far better scalability than skip list

Throughput of hash table

- Hash table is the best performing data structure among the four we have evaluated
 - For the largest key range, on a search-heavy op mix [20, 20, 60], the throughput ranges from 28.6 MOPS to 98.9 MOPS on the GPU
 - For an add/delete-heavy op mix [40, 40, 20], the throughput range is 20.8 MOPS to 72.0 MOPS
- Nearly 100 MOPS on a search-heavy op mix



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- First detailed evaluation of four lock-free data structures on CUDA-enabled GPU
- All four data structures offer moderate to high speedup on small to medium key ranges compared to CPU implementations
- Benefits are low for large key ranges in linear lists, skip lists, and priority queues
 - Primarily due to CAS overhead and complex control flow in skip lists and priority queues
- Hash tables offer consistently good speedup on arbitrary key ranges and op mixes
 - Nearly 100 MOPS throughput for search-heavy op mixes and more than 11x speedup over CPU



- Further improvement requires two key architectural innovations in GPUs
 - Fast atomics and high synchronization throughput
 - Helpful for all kinds of scalable implementations
 - Reduction in control flow divergence overhead
 - Helpful for complex lock-free constructions such as skip lists and priority queues

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