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High Performance Comparison-Based Sorting Algorithm on Many-Core GPUs

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Outline

- GPU computation model
- Our sorting algorithm
 - A new **bitonic-based merge sort**, named Warpsort
- Experiment results
- conclusion

GPU computation model

- Massively multi-threaded, data-parallel many-core architecture
- Important features:
 - SIMT execution model
 - Avoid branch divergence
 - Warp-based scheduling
 - implicit hardware synchronization among threads within a warp
 - Access pattern
 - Coalesced vs. non-coalesced

Why merge sort ?

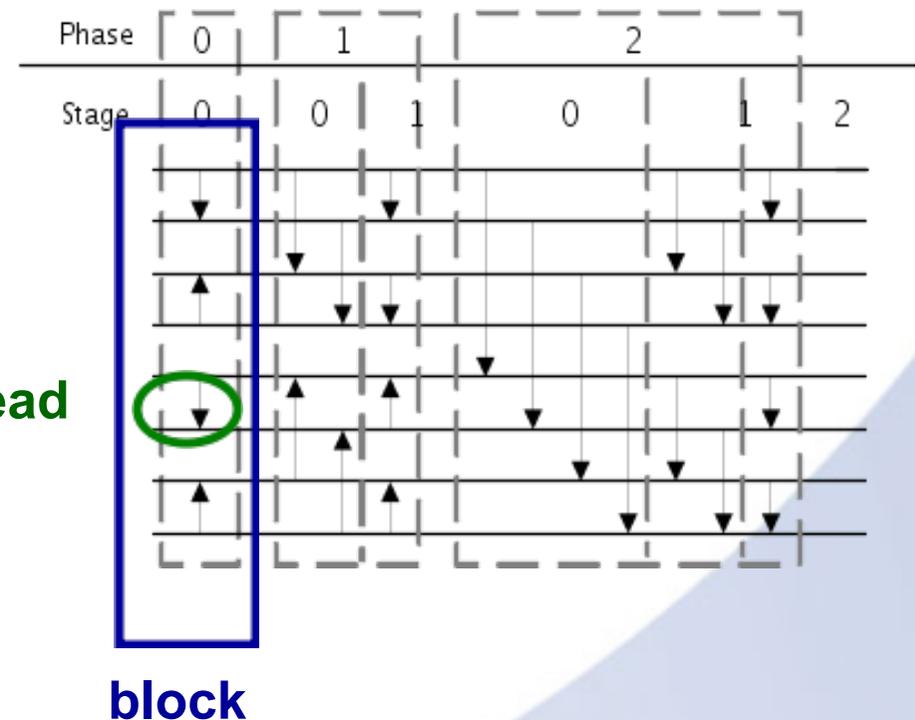
- Similar case with external sorting
 - Limited shared memory on chip vs. limited main memory
- Sequential memory access
 - Easy to meet *coalesced* requirement

Why bitonic-based merge sort ?

- Massively fine-grained parallelism
 - Because of the relatively high complexity, bitonic network is not good at sorting large arrays
 - Only used to sort small subsequences in our implementation
- Again, coalesced memory access requirement

Problems in bitonic network

- naïve implementation
 - Block-based bitonic network
 - One element per thread
- Some problems
 - in each stage
 - n elements produce only $n/2$ compare-and-swap operations
 - Form both ascending pairs and descending pairs
 - Between stages
 - synchronization

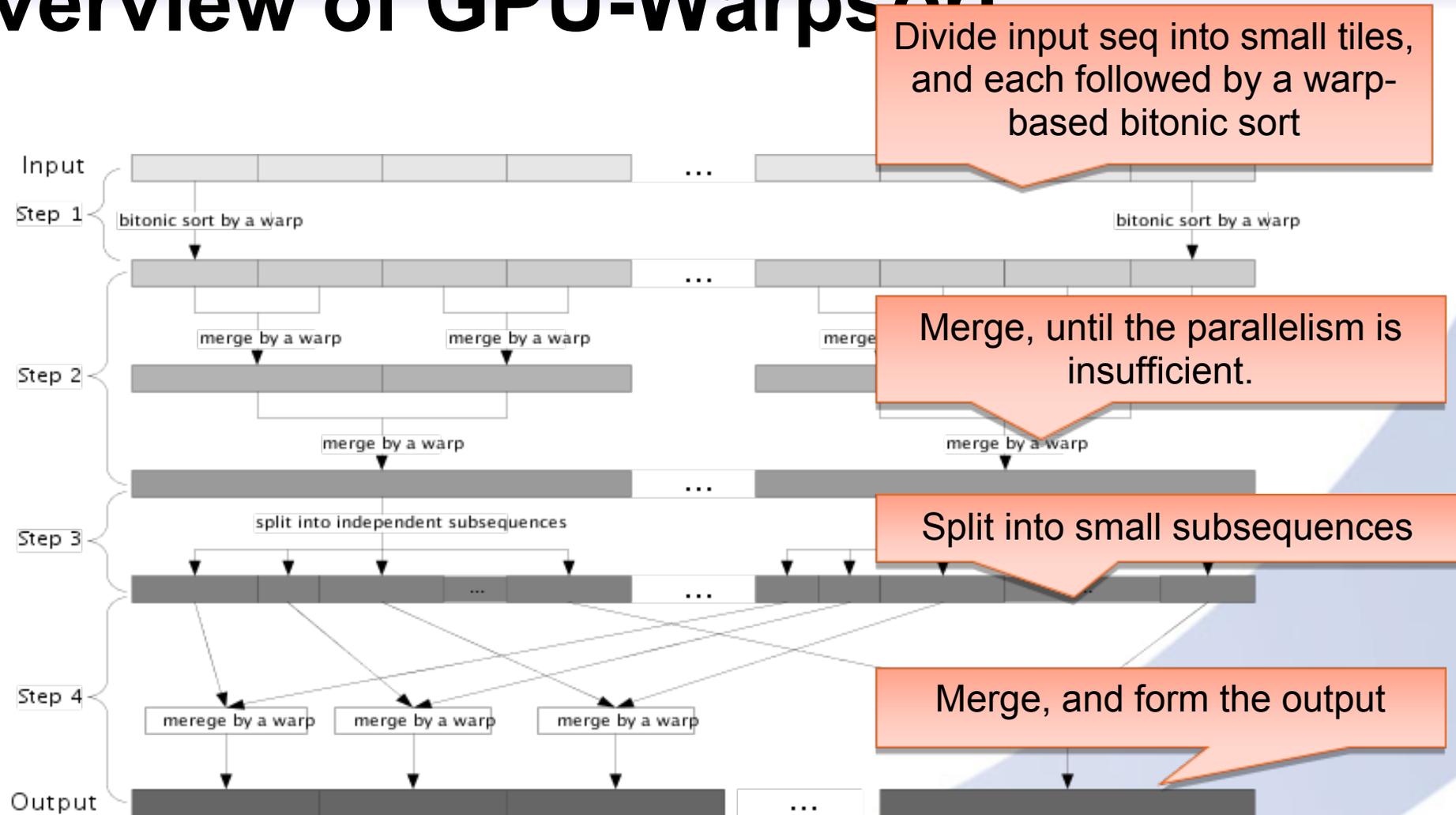


Too many branch divergences and synchronization operations

What we use ?

- Warp-based bitonic network
 - each bitonic network is assigned to an independent **warp**, instead of a **block**
 - Barrier-free, avoid synchronization between stages
 - threads in a warp perform 32 distinct compare-and-swap operations with the same order
 - Avoid branch divergences
 - At least **128** elements per warp
- And further a complete comparison-based sorting algorithm: GPU-Warpsort

Overview of GPU-Warpsort



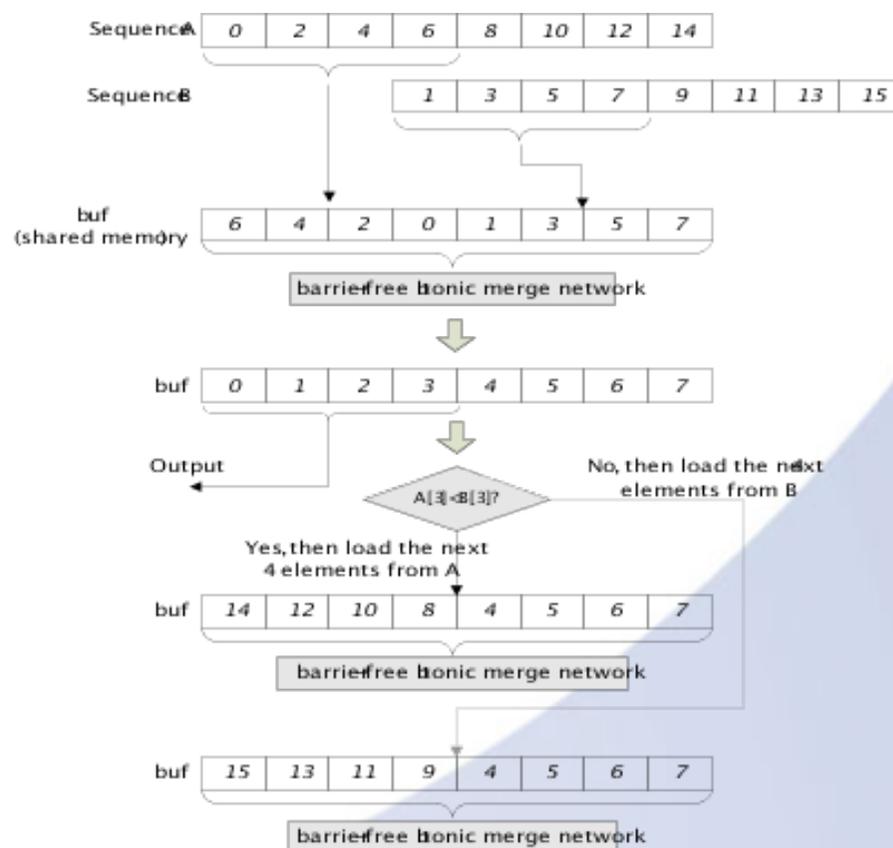
Step1: barrier-free bitonic sort

- divide the input array into equal-sized tiles
- Each tile is sorted by a warp-based bitonic network
 - 128+ elements per tile to avoid branch divergence
 - No need for `__syncthreads()`
 - Ascending pairs + descending pairs
 - Use `max()` and `min()` to replace `if-swap` pairs

```
bitonic_warp_128(key_t *keyin, key_t *keyout){
    /phase 0 to log128-1
    for(i=2; i<128; i*=2){
        for(j=i/2; j>0; j/=2){
            k ← position of preceding element in each pair
                to form ascending order
            (keyin[k0] > keyin[k0+j])
                swap(keyin[k0], keyin[k0+j]);
            l ← position of preceding element in each pair
                to form descending order
            (keyin[k1] < keyin[k1+j])
                swap(keyin[k1], keyin[k1+j]);
        }
    }
    /special case for the last phase
    for(j=128/2; j>0; j/=2){
        k ← position of preceding element in the third
            first pair to form ascending order
        (keyin[k0] > keyin[k0+j])
            swap(keyin[k0], keyin[k0+j]);
        l ← position of preceding element in the third
            second pair to form ascending order
        (keyin[k1] > keyin[k1+j])
            swap(keyin[k1], keyin[k1+j]);
    }
}
```

Step 2: bitonic-based merge sort

- t -element merge sort
 - Allocate a t -element buffer in shared memory
 - Load the $t/2$ smallest elements from seq A and B, respectively
 - Merge
 - Output the lower $t/2$ elements
 - Load the next $t/2$ smallest elements from A or B
- $t = 8$ in this example



Step 3: split into small tiles

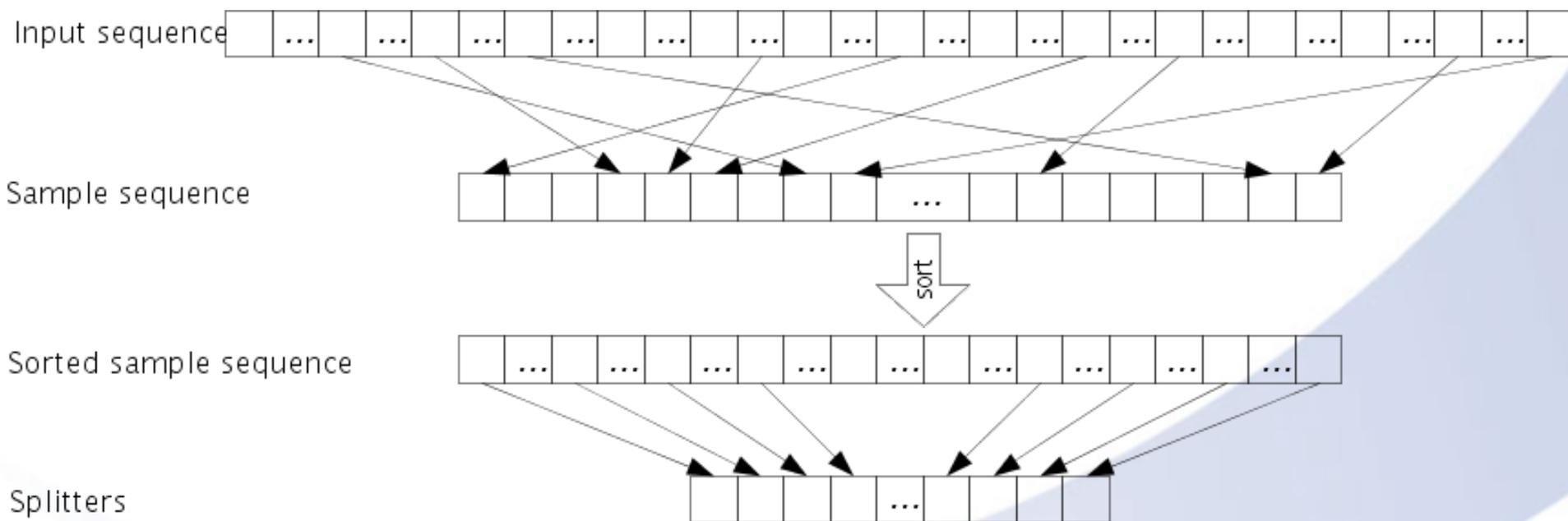
- Problem of merge sort
 - the number of pairs decreases geometrically
 - Can not fit this massively parallel platform
- Method
 - Divide the large seqs into independent small tiles which satisfy:

~~sequence~~ ~~subsequence~~ ~~job~~

~~seqs~~ \ll

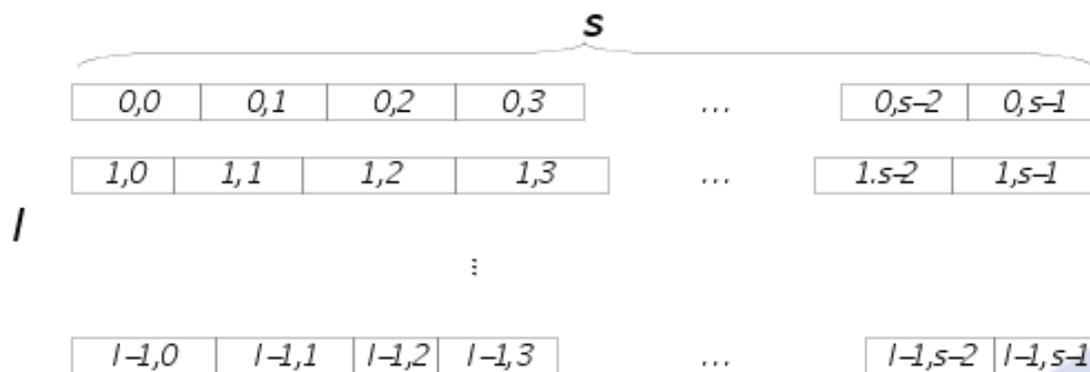
Step 3: split into small tiles (cont.)

- How to get the splitters?
 - Sample the input sequence randomly



Step 4: final merge sort

- Subsequences $(0,i), (1,i), \dots, (l-1,i)$ are merged into S_i
- Then, S_0, S_1, \dots, S_l are assembled into a totally sorted array

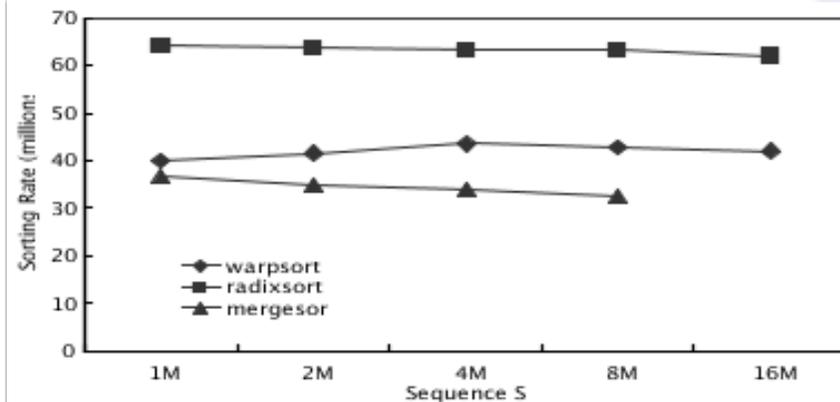
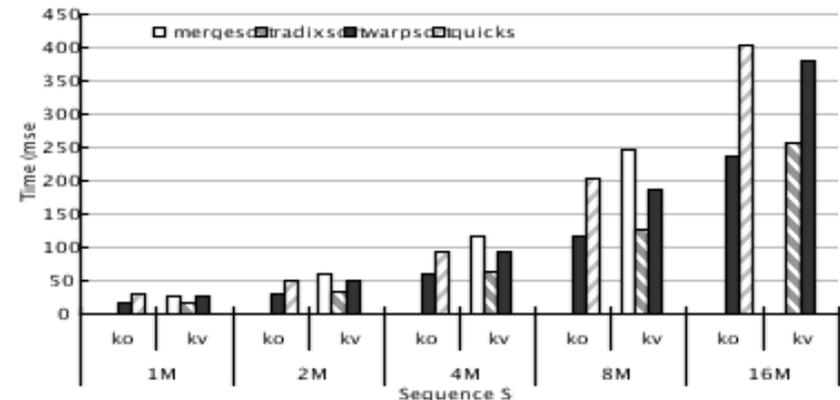


Experimental setup

- Host
 - AMD Opteron880 @ 2.4 GHz, 2GB RAM
- GPU
 - 9800GTX+, 512 MB
- Input sequence
 - Key-only and key-value configurations
 - 32-bit keys and values
 - Sequence size: from 1M to 16M elements
 - Distributions
 - Zero, Sorted, Uniform, Bucket, and Gaussian

Performance comparison

- Mergesort
 - Fastest comparison-based sorting algorithm on GPU (Satish, IPDPS'09)
 - Implementations already compared by Satish are not included
- Quicksort
 - Cederman, ESA'08
- Radixsort
 - Fastest sorting algorithm on GPU (Satish, IPDPS'09)
- Warpsort
 - Our implementation

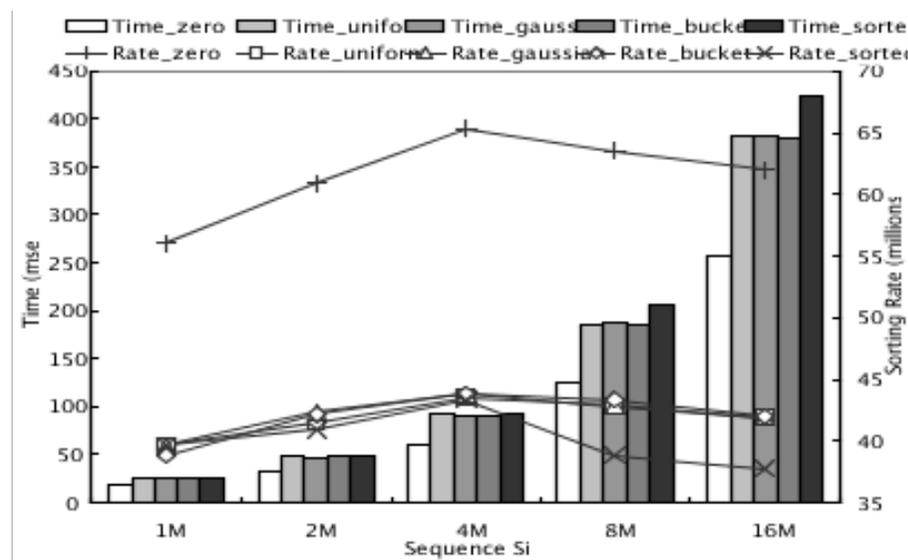


Performance results

- Key-only
 - 70% higher performance than quicksort
- Key-value
 - 20%+ higher performance than mergesort
 - 30%+ for large sequences (>4M)

Results under different distributions

- Uniform, Bucket, and Gaussian distribution almost get the same performance
- Zero distribution is the fastest
- Not excel on Sorted distribution
 - Load imbalance



Conclusion

- We present an efficient comparison-based sorting algorithm for many-core GPUs
 - carefully map the tasks to GPU architecture
 - Use warp-based bitonic network to eliminate barriers
 - provide sufficient homogeneous parallel operations for each thread
 - avoid thread idling or thread divergence
 - totally coalesced global memory accesses when fetching and storing the sequence elements
- The results demonstrate up to 30% higher performance
 - Compared with previous optimized comparison-based algorithms

Thanks