

Parallel Sorting

Michael Garland

NVIDIA Research





Problem Overview

- Given a sequence of n integers, called **keys**

$A = [8 \ 4 \ 3 \ 9 \ 0 \ 9 \ 7]$

- Place keys in output in non-decreasing order

$\text{sorted}(A) = [0 \ 3 \ 4 \ 7 \ 8 \ 9 \ 9]$

- Optionally with equal values in their original order
 - “stable” sorts provide this; “unstable” sorts do not



Why Sorting?

- Put data in order
- Make searching easier
- Build data structures in parallel
- ... and many others



Some assumptions for today

- Keys are integers of fixed length (e.g., 32 bits)
- Keys are not part of larger records
- Sequences reside entirely in main memory
- “Main memory” of the processor we’re using
 - in CPU memory for CPU sorts
 - in GPU memory for GPU sorts



Sorting problems we won't discuss

- **External memory sorting**
 - data doesn't fit in memory all at once
- **Distributed sorting**
 - data resides in physically separate memories
- **Long and/or variable length keys**
 - can significantly change performance trade offs
- **Among others ...**

How do we sort?



Some simple sorts

- **Selection**

- remove the smallest key of the input
- append at the end of the output
- repeat

Sequential
(mostly)

- **Insertion**

- remove the next key of the input
- insert into the output in sorted order
- repeat

- **Transposition**

- find pair where $A[i] > A[i+1]$ and swap them
- repeat until there are none

Parallel
(potentially)

Odd-Even Transposition Sort

- Parallelizing transposition sort:
 - assign 1 thread to each element
 - use odd/even phases to prevent contention

```
while A is not sorted:  
    if is_odd(i) and (A[i+1] < A[i])  
        swap(A[i], A[i+1])  
    barrier  
    if is_even(i) and (A[i+1] < A[i])  
        swap(A[i], A[i+1])  
    barrier
```

requires at most $n/2$ iterations

Counting Sort

- Step 1: Count elements sorting to left of A[i]

```
rank[i] = count( j < i where A[j] ≤ A[i] )  
         + count( j > i where A[j] < A[i] )
```



A[j] ≤ A[i] A[i] A[j] < A[i]

- Step 2: Scatter to position in sorted order

```
permute( A[i] -> A[rank[i]] )
```



... A[i] ...

Counting Sort (alternate)

- Step 1: Count places that $A[i]$ needs to move

```
offset[i] = count( j < i where A[j] > A[i] )  
           - count( j > i where A[j] < A[i] )
```



A[j] > A[i] A[i] A[j] < A[i]

- Step 2: Scatter to position in sorted order

```
permute( A[i] -> A[i-offset[i]] )
```



... A[i] ...

Binary Counting Sort

- If $A[i]$ is 0:

```
offset[i] = count( j < i where A[j] == 1 )
```



count ones before A[i]

- If $A[i]$ is 1:

```
offset[i] = -count( j > i where A[j] == 0 )
```



A[i] count zeros after

- And scatter:

```
permute( A[i] -> A[i-offset[i]] )
```



A Simple Radix Sort

Apply binary counting sort to each bit of the keys, from LSB to MSB

```
def radix_sort(A, msb=32):
    def delta(flag, ones_before, zeros_after):
        if flag==0: return -ones_before
        else:       return +zeros_after

    lsb = 0

    while lsb<msb:
        flags = [(x>>lsb)&1 for x in A]
        ones  = scan(plus, flags)
        zeros = rscan(plus, [f^1 for f in flags])

        offsets = map(delta, flags, ones, zeros)
        A = permute_with_offsets(A, offsets)

        lsb = lsb+1

    return A
```



Is this efficient?

Apply binary counting sort to each bit of the keys, from LSB to MSB

```
def radix_sort(A, msb=32):
    def delta(flag, ones_before, zeros_after):
        if flag==0: return -ones_before
        else:       return +zeros_after

    lsb = 0

    while lsb<msb:
        flags = [(x>>lsb)&1 for x in A]
        ones  = scan(plus, flags)
        zeros = rscan(plus, [f^1 for f in flags])

        offsets = map(delta, flags, ones, zeros)
        A = permute_with_offsets(A, offsets)

        lsb = lsb+1

    return A
```

Radix Sort



- Apply counting sort to successive digits of keys
- Performs d scatter steps for d -digit keys
- Scattering in memory is fundamentally costly

Parallel Radix Sort

- Assign tile of data to each block (1024 elements)
- Build per-block histograms of current digit (4 bit)
- Combine per-block histograms ($P \times 16$)
- Scatter

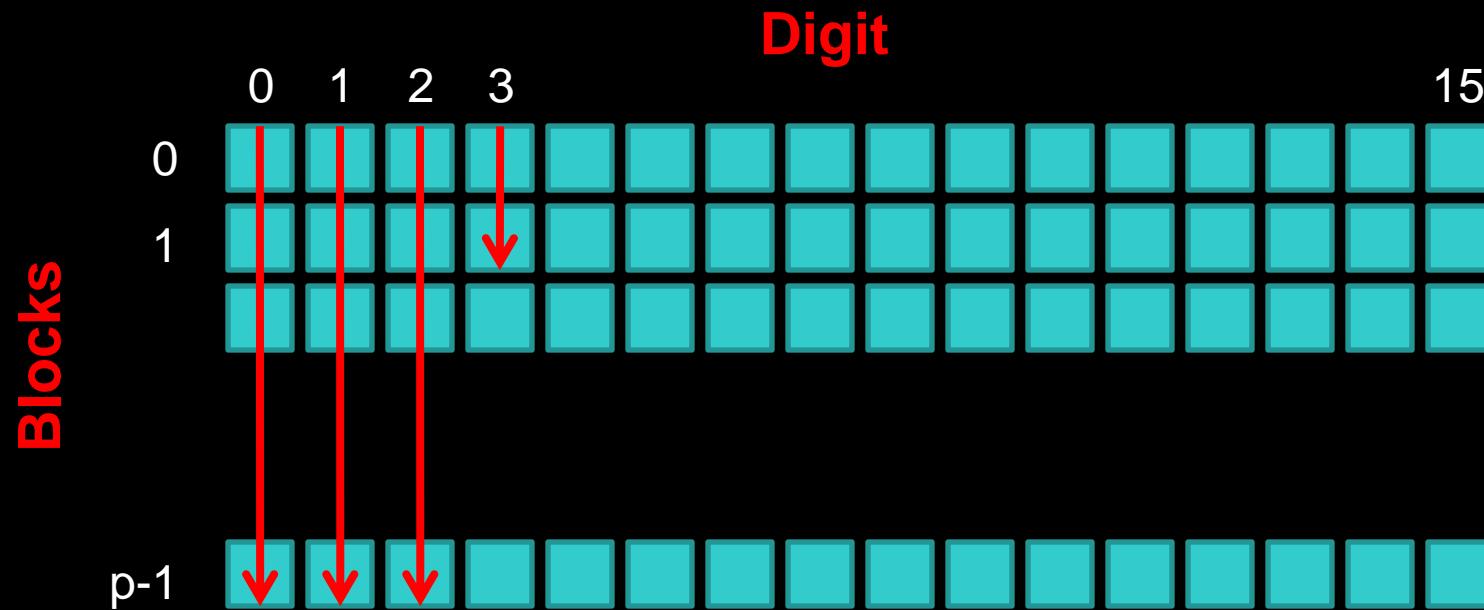


Per-Block Histograms

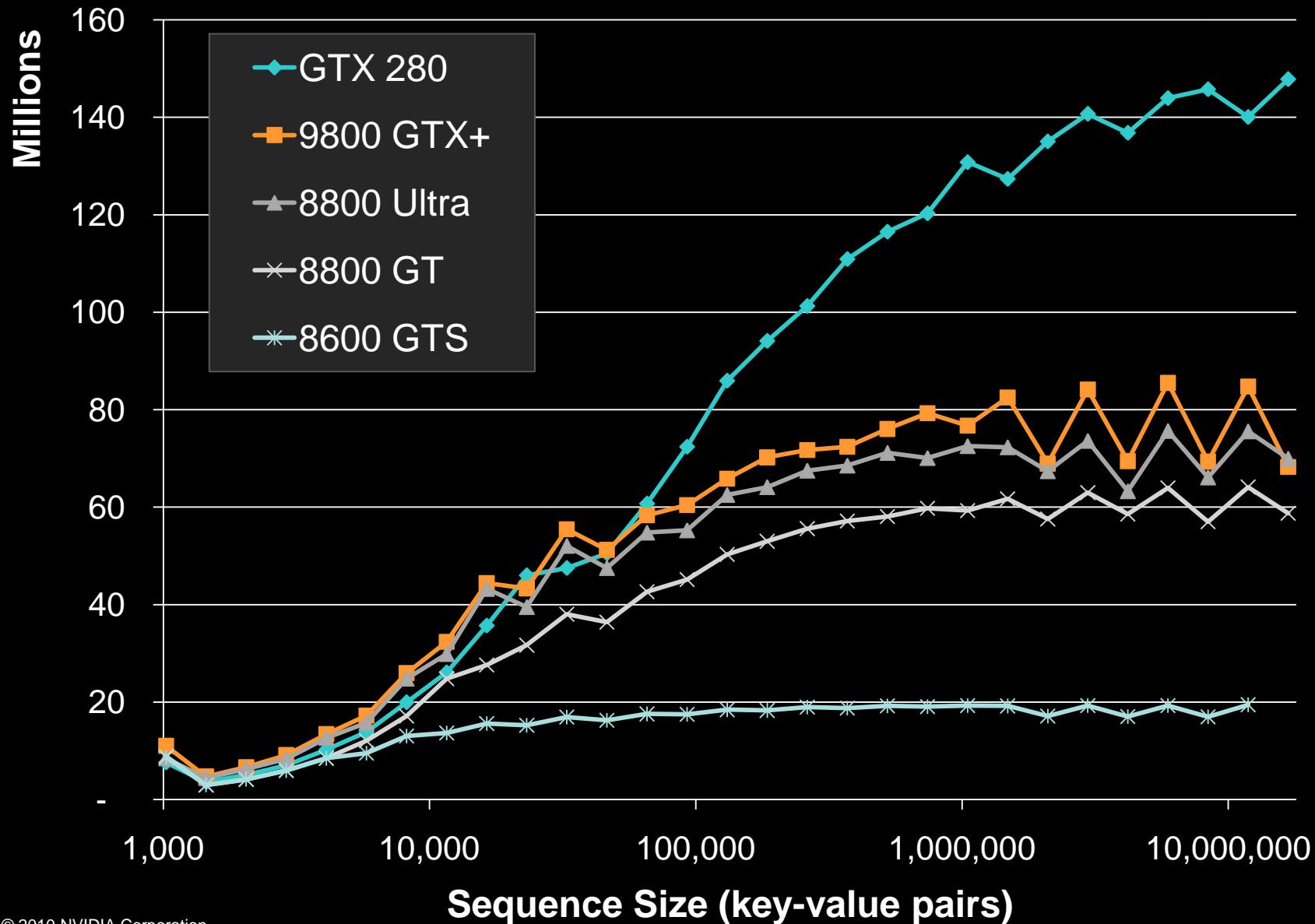
- Perform b parallel splits for b -bit digit
- Each split is just a prefix sum of bits
 - each thread counts 1 bits to its left
- Write bucket counts & partially sorted tile
 - sorting tile improves scatter coherence later

Combining Histograms

- Write per-block counts in column major order & scan

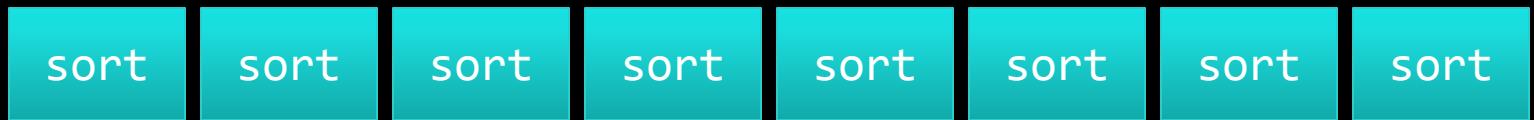


Radix Sorting Rate (pairs/sec)

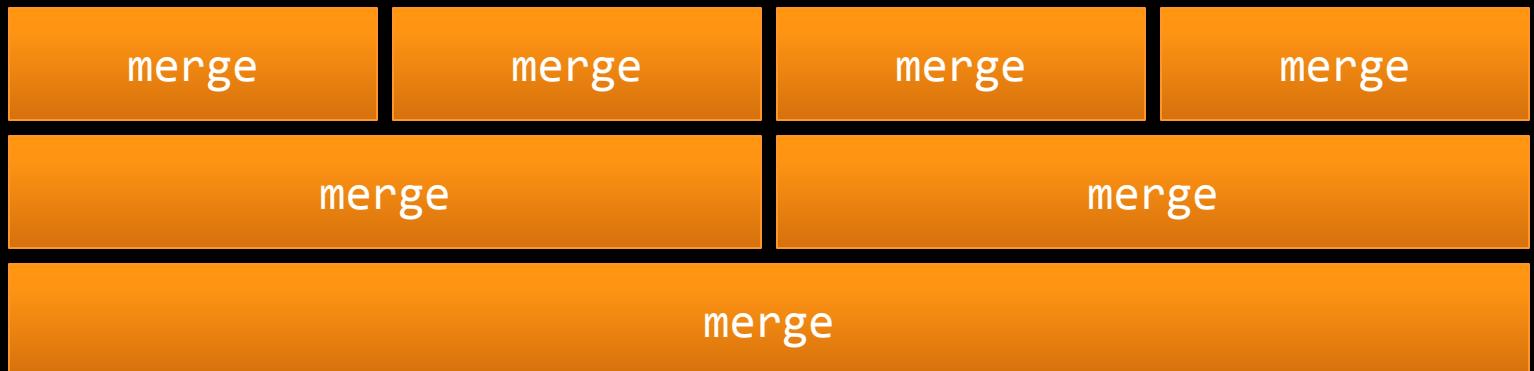


Merge Sort

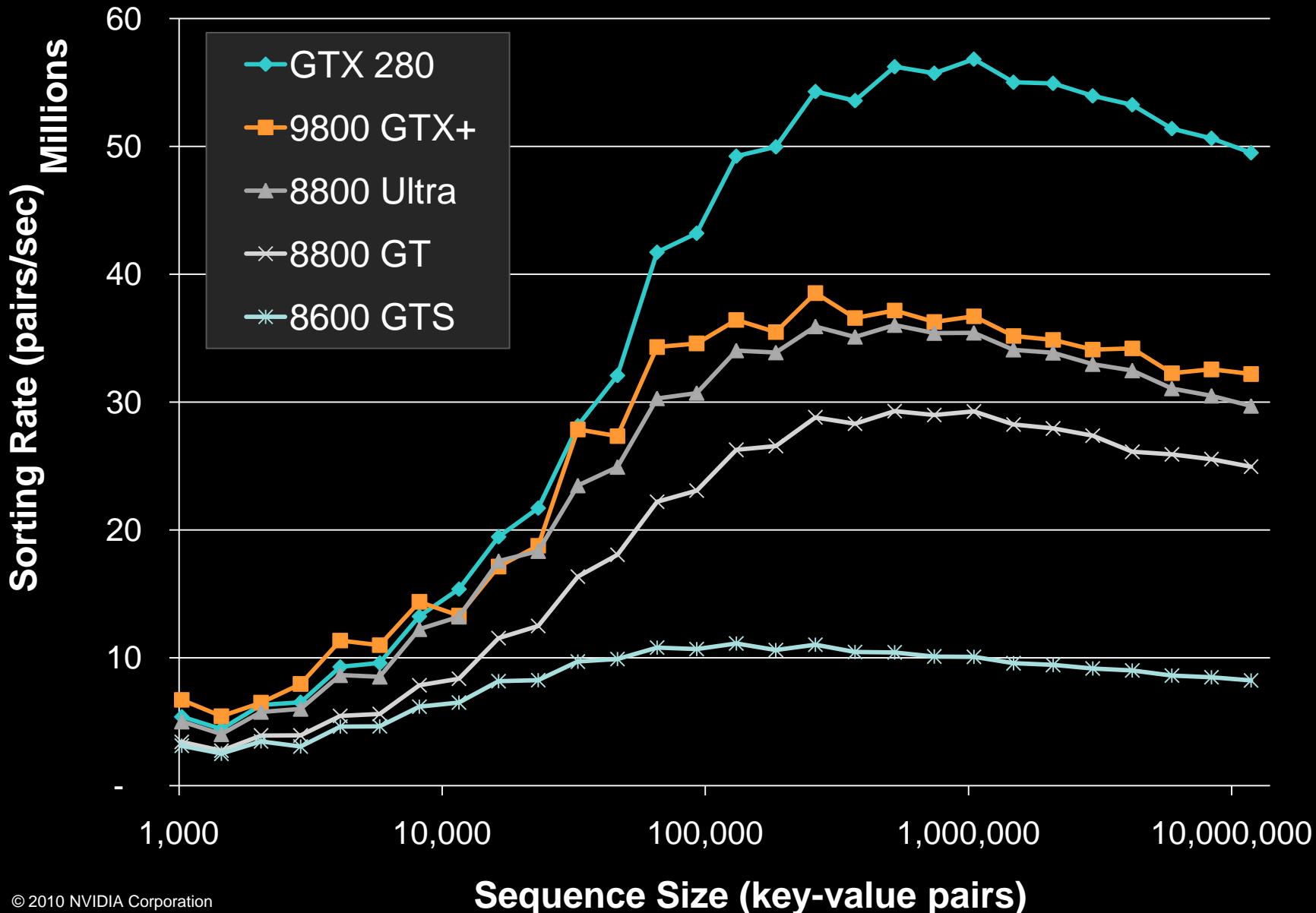
- Divide input array into 256-element tiles
- Sort each tile independently



- Produce sorted output with tree of merges



Merge Sorting Rate





Some other techniques

- **Quicksort / Sample Sort**
 - partition keys into non-overlapping ranges
 - sort each range individually
- **Sorting networks**
 - fixed network of comparison operators
 - e.g., bitonic sort, odd-even merge sort



Questions?

mgarland@nvidia.com



Odd-Even Merge Sort

```
template<typename T, typename Cmp>
__device__ void oddeven_sort(T *keys, int i, int n, Cmp lt)
{
    for(unsigned int p=n/2; p>0; p/=2) {
        unsigned int q=n/2, r=0, d=p;
        while( q>=p ) {
            if( i<(n-d) && (i&p)==r ) {
                unsigned int j = i+d;
                T xi = keys[i], xj = keys[j];

                if( lt(xj,xi) ) {
                    keys[i] = xj;
                    keys[j] = xi;
                }
            }

            d = q-p; q = q/2; r = p;
            __syncthreads();
        }
    }
}
```

Algorithm M, Section 5.2.2
The Art of Computer Programming, Vol 3
D. E. Knuth