#### Mergesort with CUDA

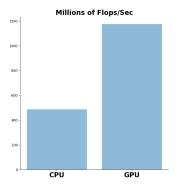
Fabian Schuetze

25.02.2020

Fork on Github: https://github.com/FabianSchuetze/mergesort

#### GPU: fast, cheap, and approachable

Fast:



Cheap: CPU: \$400, GPU: \$200

Approachable: CUDA: C++ Syntax, called from C and C++

Intro I: Simple Program

Intro II: Nvidia's two big Architecture Decisions

Merge

Memory Hirachy of CUDA

Merging with local memory

Intro I: Simple Program

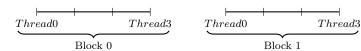
#### A Simple Cuda Program

1: Call Cuda Program from C and C++:

```
dim3 Grid(2)
dim3 Block(4)
square<<<Grid, Block>>>(...)
```

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### A Simple Cuda Program

```
1: Call Cuda Program from C and C++:
```

- dim3 Grid(2)
  dim3 Block(4)
- 3 square<<<Grid, Block>>>(...)
  - 2: Blocks & Grid spawn Threads:



#### 3: Cuda Code follows C++ Syntax with Extension:

```
1  __global__ void square(int *array, int n) {
2     int tid = blockDim.x * blockIdx.x + threadIdx.x;
3     if (tid < n) {
4         array[tid] = array[tid] * array[tid];
5     }
6     }</pre>
```

## How a full CUDA Program is Run:

Intro I: Simple Program

```
__global__ void square(int *array, int n) {
       int tid = blockDim.x * blockIdx.x + threadIdx.x;
       if (tid < n) array[tid] = array[tid] * array[tid];</pre>
   }
5
   int main() {
6
       int a[3]; fill array(a, 3);
7
       // 1. Initialize Memory on GPU and copy from RAM//
8
       unsigned int sz = 3 * sizeof(int);
9
       int* d a;
10
       cudaMalloc((int**)&d_a, sz);
11
       cudaMemcpy(d a, a, sz, cudaMemcpyHostToDevice);
12
       // 2. Define Thread Structure and Launch Kernel //
13
       dim3 block(4), grid(2);
14
       square << grid, block >>> (d a, 3);
15
       // 2. Copy GPU memory back to RAM //
16
       cudaMemcpy(a, d_a, sz, cudaMemcpyDeviceToHost);
17
```

# Intro II: Nvidia's two big Architecture Decisions

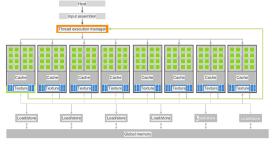
#### Software/Hardware interface

```
dim3 block(4);
dim3 grid(2);
square<<<grid, block>>>(...);
```

1. Software: Specify Blocks and Threads

#### Software/Hardware interface

**GPU:** Rectangles are Sets of Processors



- 1. Software: Specify Blocks and Threads
- 2. Hardware:
  - 1. Blocks to SM: Block  $1 \rightarrow SM1$ , Block  $2 \rightarrow SM2 \dots$

#### Software/Hardware interface

GPU: Group of Threads allocated to Processors

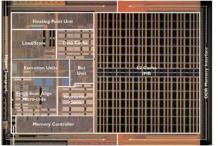


- 1. Software: Specify Blocks and Threads
- 2. Hardware:
  - 1. Blocks to SM: Block  $1 \to SM1$ , Block  $2 \to SM2 \dots$
  - 2. Threads to Processors: Wrapped threads are allocated

#### Reasons for strong GPU performance

Patterson and Hennessy (2009)

AMD Opteron: FP and Int Execution Units are small

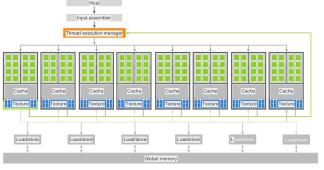


CPU: Chip full with fancy Branch Prediction, Caches, etc.

#### Reasons for strong GPU performance

Patterson and Hennessy (2009)

#### Nvidia GPU: Small Blocks are Execution Units



CPU: Chip full with fancy Branch Prediction, Caches, etc. GPU:

1. Simplicity: Just Cores on chip

#### Reasons for strong GPU performance

Patterson and Hennessy (2009)

Multithreading: Context Switches are Cheap



CPU: Chip full with fancy Branch Prediction, Caches, etc. GPU:

- 1. Simplicity: Just Cores on chip
- 2. Multithreading: Wrap1 in long-latency op: Start Wrap2

# Merge

#### Serial Merge

```
A = \begin{bmatrix} 5 & 7 & 8 & 12 \end{bmatrix}; \quad B = \begin{bmatrix} 3 & 4 & 6 & 10 \end{bmatrix}
                  C = \begin{bmatrix} ? & ? & ? & ? & ? & ? & ? \end{bmatrix}
    void merge(T* a, T* b, T* c, int sz_a, int sz_b) {
          int i = 0, j = 0, k = 0;
2
          while (k < sz a + sz b)
3
                if (i == sz a)
4
                     c[k++] = b[j++];
5
                else if (j == sz b)
6
                     c[k++] = a[i++]:
7
               else if (a[i] <= b[i])
8
                     c[k++] = a[i++]:
9
                else
10
                     c[k++] = b[j++];
11
12
    Problem: complexity, \mathcal{O}(n)
          Merge
```

$$A = \begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}; \quad B = \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}$$
  
 $C = \begin{bmatrix} ? & ? & ? & ? & ? & ? \end{bmatrix}$ 

$$A = \begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}; \quad B = \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}$$
  
 $C = \begin{bmatrix} ? & ? & ? & ? & ? & ? \end{bmatrix}$ 

Naive: 2 Threads, half A and B

$$A = \begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}; \quad B = \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}$$
 
$$C = \begin{bmatrix} ? & ? & ? & ? & ? & ? \end{bmatrix}$$

Naive: 2 Threads, half A and B

Result:

Thread 1: Merge:

$$A_1 = \begin{bmatrix} 0 & 0 \end{bmatrix}; \quad B_1 = \begin{bmatrix} 1 & 1 \end{bmatrix}; \qquad C_1 = \begin{bmatrix} 0 & 0 & 1 & 1 \end{bmatrix}$$

Thread 2: Merge:

$$A_2 = \begin{bmatrix} 0 & 0 \end{bmatrix}; \quad B_2 = \begin{bmatrix} 1 & 1 \end{bmatrix}; \qquad C_2 = \begin{bmatrix} 0 & 0 & 1 & 1 \end{bmatrix}$$

$$A = \begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}; \quad B = \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}$$
 
$$C = \begin{bmatrix} ? & ? & ? & ? & ? & ? \end{bmatrix}$$

Naive: 2 Threads, half A and B

Result:

Thread 1: Merge:

$$A_1 = \begin{bmatrix} 0 & 0 \end{bmatrix}; \quad B_1 = \begin{bmatrix} 1 & 1 \end{bmatrix}; \qquad C_1 = \begin{bmatrix} 0 & 0 & 1 & 1 \end{bmatrix}$$

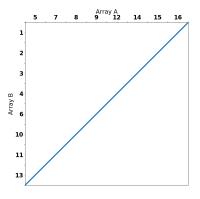
Thread 2: Merge:

$$A_2 = \begin{bmatrix} 0 & 0 \end{bmatrix}; \quad B_2 = \begin{bmatrix} 1 & 1 \end{bmatrix}; \qquad C_2 = \begin{bmatrix} 0 & 0 & 1 & 1 \end{bmatrix}$$

Naive strategy: Doesn't work

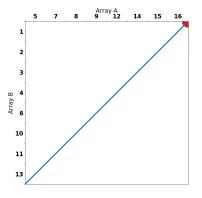
Odeh et al. (2012); Baxter (2016)

How to split two arrays in equal chuncks?



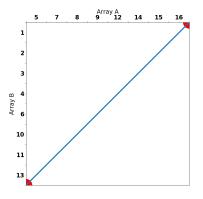
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Feasible split: Array A to Thread 1, B to Thread 2



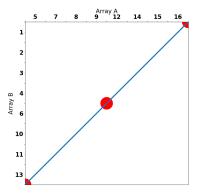
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Another split: Array B to Thread 1, A to Thread 2



Odeh et al. (2012); Baxter (2016)

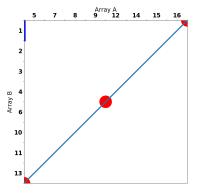
Another (as before): Thread 1 gets half of A and B



Summary: All allocations along vertical line split work equally!

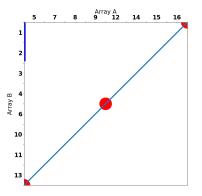
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Mergepath: Optimal split: One Elemet of B



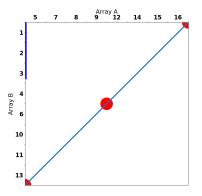
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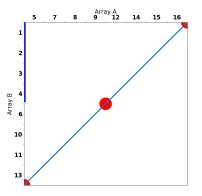
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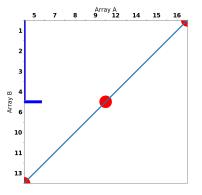
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Mergepath: Optimal split: Another Elemet of B



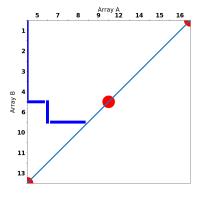
Odeh et al. (2012); Baxter (2016)

#### Mergepath: Optimal split: First elment of A



Odeh et al. (2012); Baxter (2016)

#### Mergepath: Optimal split!



Split index of A: 3 Split index of B: 5

## Merge with Cuda (Divide-and-Conquer)

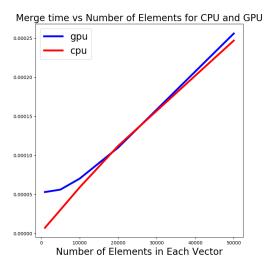
```
1 __global
   void paralleMerge(int* a, int sz_a, int* b, int sz_b,
2
                      int* c, int length)
3
       int diag = threadIdx.x * length;
5
       int a split = mergepath(a, sz_a, b, sz_b, diag);
6
       int b_split = diag - a_split;
7
       merge(a, a_split, sz_a, b, b_split, sz_b, c, diag,
8
             length);
9
   }
10
```

Each identifies split indices

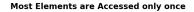
Split indices suffice to merge two sub-arrays into c

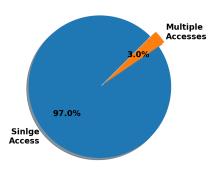
#### Problem: Slow as a Snail

std::merge (with -O3 optimization) is as fast!

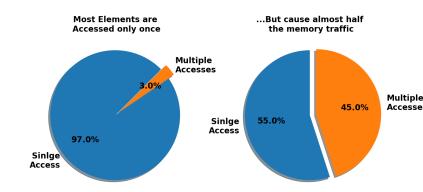


#### Reason: Too much global memory access





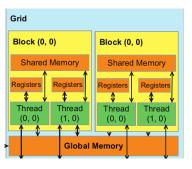
#### Reason: Too much global memory access



## Memory Hirachy of CUDA

#### Shared memory: Tiny but Fast

Kirk and Hwu (2012)



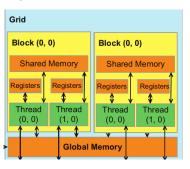
#### Global $\Leftrightarrow$ Shared:

Visibility: All threads  $\Leftrightarrow$  Private for Block Size: 2 GB  $\Leftrightarrow$  Total/Block 192/48 KB

Latency:  $8 \text{ GB/s} \Leftrightarrow 80 \text{ GB/s}$ 

#### Shared memory: Tiny but Fast

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Visibility: All threads  $\Leftrightarrow$  Private for Block Size: 2 GB  $\Leftrightarrow$  Total/Block 192/48 KB

Latency:  $8 \text{ GB/s} \Leftrightarrow 80 \text{ GB/s}$ 

Conclusion: Load Subarrays in Shared  $\Rightarrow$  faster memory access



## Merging with shared memory: Use MergePath twice

- 1. Problem: Shared memory tiny: 48 Kb
- 2. Solution: Split arrays into subarrays by block
- 3. Approach (Divide-and-Conquer twice):

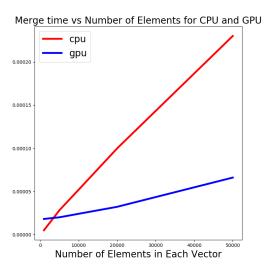
Identify subarrays with mergepath; smaller than 48KB

Load into shared memeory

Threads use mergepath again (now from shared memory)

Threads merge subset (as before)

#### With Shared Memory: GPU faster than std::merge



- Sean Baxter. Intro Modern GPU. 2016 (cit. on pp. 22–31).
- David B. Kirk and Wen-mei W. Hwu. Programming Massively Parallel Processors: A Hands-on Approach. English. 2 edition. Amsterdam: Morgan Kaufmann, Dec. 2012 (cit. on pp. 37, 38).
- Saher Odeh et al. "Merge Path Parallel Merging Made Simple". In: 2012 IEEE 26th International Parallel and Distributed Processing Symposium Workshops PhD Forum. ISSN: null. May 2012, pp. 1611–1618 (cit. on pp. 22–31).
- David A. Patterson and John L. Hennessy. Computer organization and design: the hardware/software interface. 4th ed. Burlington, MA: Morgan Kaufmann Publishers, 2009 (cit. on pp. 13–15).