



### **Examples of reduction algorithms**

**Extracting small data from larger** 

Finding max or min

- Calculating median or average
  - Histograms
  - **Common problems!**



### **Sequentially trivial**

### Loop through data

Add, min/max, accumulate results

Fits badly in massive parallelism!







### **Tree-based approach**

## Each level parallel! Can be split onto large numbers of threads

but

the parallelism is reduced for each level, and the results need to be reorganized to a smaller number of threads!







### Multiple kernel runs for varying size!

### For n = k downto 0 do Launch 2<sup>n</sup> kernels

Multiple levels can be merged into one - but not all of them!





# Important note: You can not synchronize between blocks!

### Why?

## Complex hardware Risk for deadlock between blocks that are not simultaneously active







### Many important optimizations:

- Avoid "if" statements, divergent branches
  - Avoid bank conflicts in shared memory
  - Loop unrolling to avoid loop overhead (classic old-style optimization!)



### Huge speed difference reported by Harris

	Time (2 <sup>22</sup> ints)	Bandwidth	Step Speedup	Cumulati Speedu
Kernel 1: interleaved addressing with divergent branching	8.054 ms	2.083 GB/s		
Kernel 2: interleaved addressing with bank conflicts	3.456 ms	4.854 GB/s	2.33x	2.3
Kernel 3: sequential addressing	1.722 ms	9.741 GB/s	2.01x	4.6
Kernel 4: first add during global load	0.965 ms	17.377 GB/s	1.78x	8.3
Kernel 5: unroll last warp	0.536 ms	31.289 GB/s	1.8x	15.0
Kernel 6: completely unrolled	0.381 ms	43.996 GB/s	1.41x	21.1
Kernel 7: multiple elements per thread	0.268 ms	62.671 GB/s	1.42x	30.0





### **Conclusions:**

- Multiple kernel runs for varying problem size
- Multiple kernel runs for synchronizing blocks
- Optimizing matters! Not only shared memory and coalescing!