

CUB:

A pattern of "collective" software design, abstraction, and reuse for kernel-level programming

> **DUANE MERRILL, PH.D.** NVIDIA RESEARCH

What is CUB?

1. A design model for *collective* kernel-level primitives

How to make reusable software components for SIMT groups (warps, blocks, etc.)

2. A library of collective primitives

Block-reduce, block-sort, block-histogram, warp-scan, warp-reduce, etc.

3. A library of global primitives (built from collectives)

- Device-reduce, device-sort, device-scan, etc.
- Demonstrate collective composition, performance, *performance-portability*

Outline

1. Software reuse

- 2. SIMT collectives: the "missing" CUDA abstraction layer
- 3. The soul of collective component design
- 4. Using CUB's collective primitives
- 5. Making your own collective primitives
- 6. Other Very Useful Things in CUB
- 7. Final thoughts

Software reuse

Abstraction & composability are fundamental design principles

Reduce redundant programmer effort

- Save time, energy, money
- Reduce buggy software

Encapsulate complexity

- Empower productivity-oriented programmers
- Insulation from underlying hardware

Software reuse empowers a <u>durable</u> programming model

Software reuse

Abstraction & composability are fundamental design principles

Reduce redundant programmer effort

- Save time, energy, money
- Reduce buggy software

Encapsulate complexity

- Empower productivity-oriented programmers
- Insulation from changing capabilities of the underlying hardware
 - NVIDIA has produced nine different CUDA GPU architectures since 2008!

Software reuse empowers a <u>durable</u> programming model

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Parallel programming is hard...

No, <u>cooperative</u> parallel programming is hard...

- Parallel decomposition and grain sizing
- Synchronization
- Deadlock, livelock, and data races
- Plurality of state
- Plurality of flow control (divergence, etc.)

- Bookkeeping control structures
- Memory access conflicts, coalescing, etc.
- Occupancy constraints from SMEM, RF, etc
- Algorithm selection and instruction scheduling
- Special hardware functionality, instructions, etc.

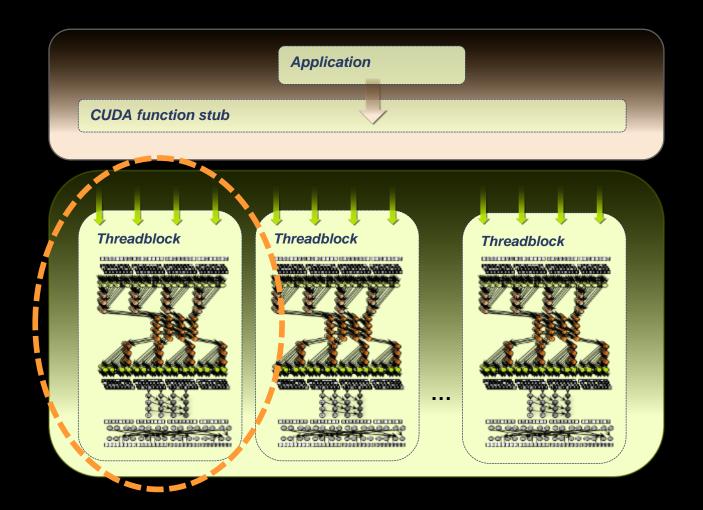
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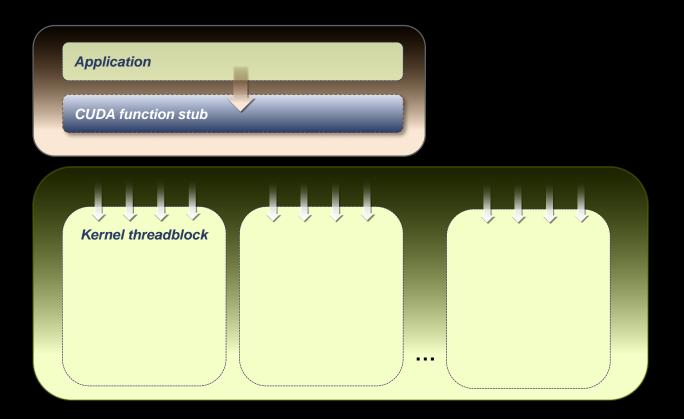
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CUDA today



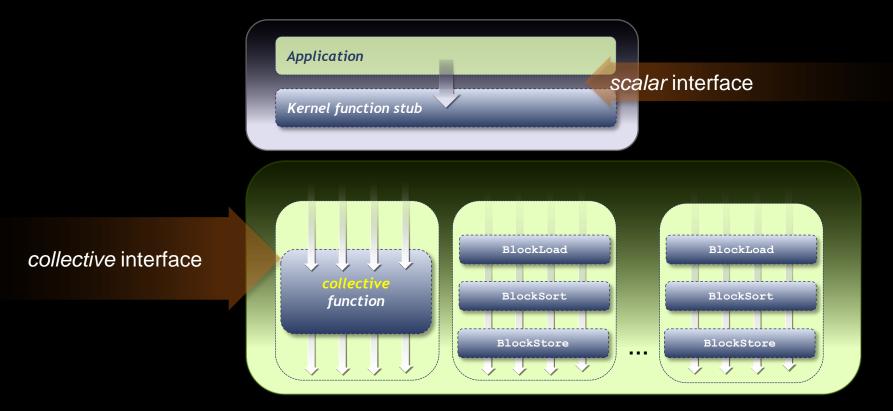
Software abstraction in CUDA



PROBLEM: virtually every CUDA kernel written today is cobbled from scratch

A tunability, portability, and maintenance concern

Software abstraction in CUDA

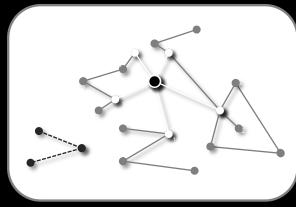


Collective software components

reduce development cost, hide complexity, bugs, etc.



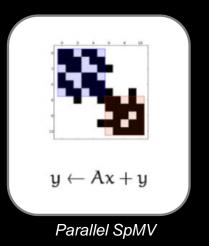
What do these applications have in common?

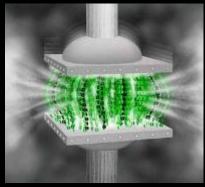


Parallel sparse graph traversal



Parallel radix sort



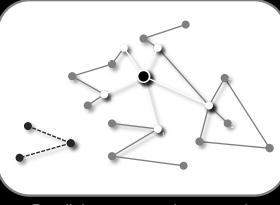


Parallel BWT compression

What do these applications have in common?

Block-wide prefix-scan

Scan for enqueueing



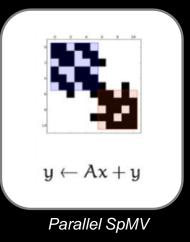
Parallel sparse graph traversal

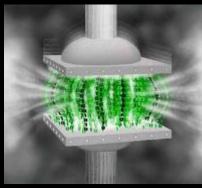


Parallel radix sort

Scan for partitioning

Scan for segmented reduction



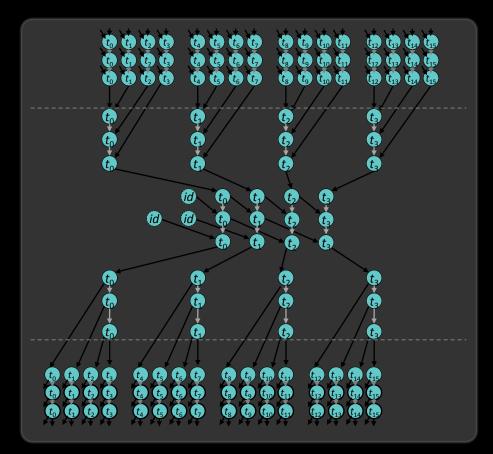


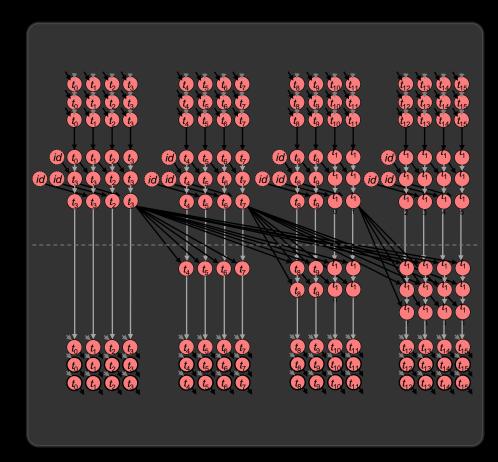
Parallel BWT compression

Scan for solving recurrences (move-to-front)

Examples of parallel scan data flow

16 threads contributing 4 items each





Brent-Kung hybrid

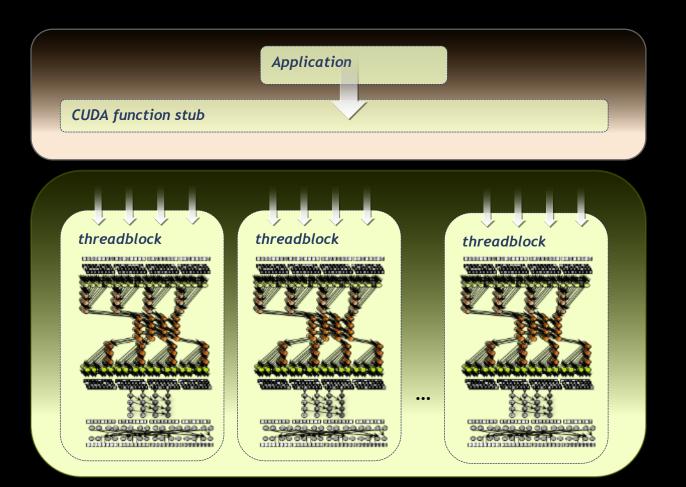
(Work-efficient ~130 binary ops, depth 15)

Kogge-Stone hybrid

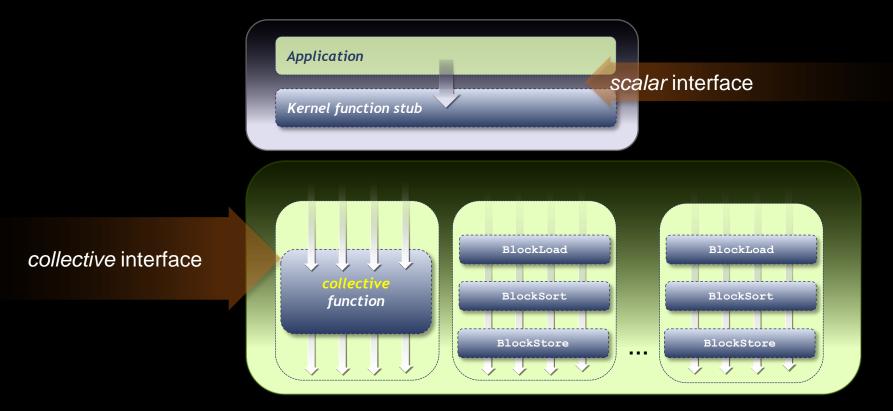
(Depth-efficient ~170 binary ops, depth 12)

CUDA today

Kernel programming is complicating



Software abstraction in CUDA



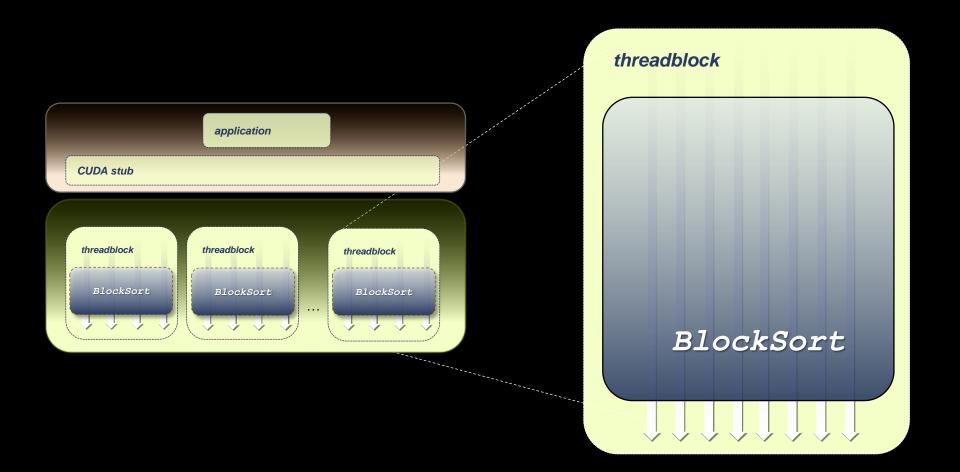
Collective software components

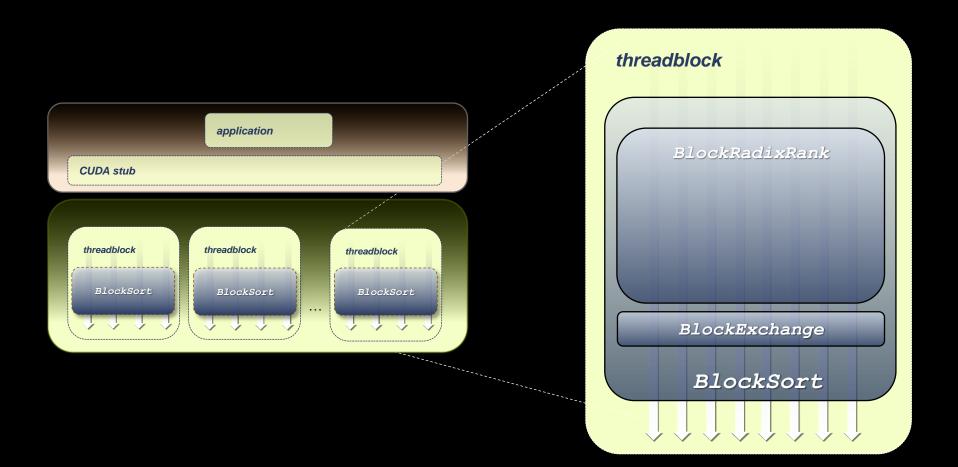
reduce development cost, hide complexity, bugs, etc.

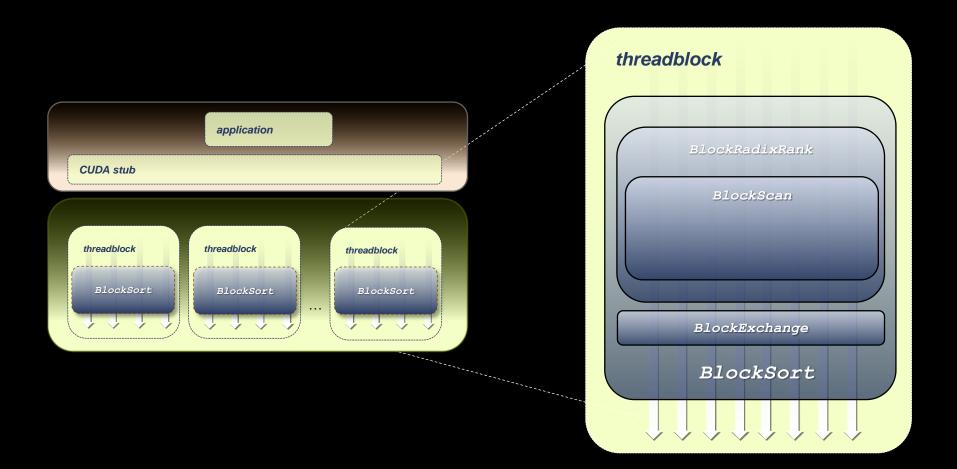


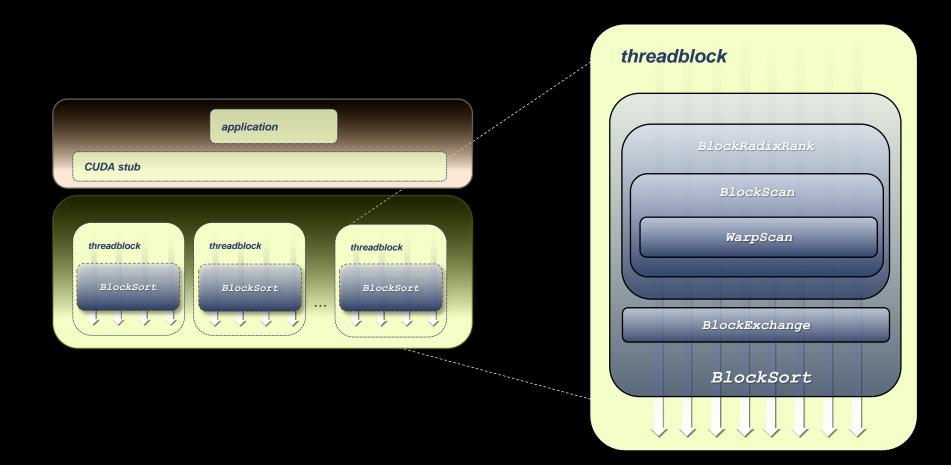
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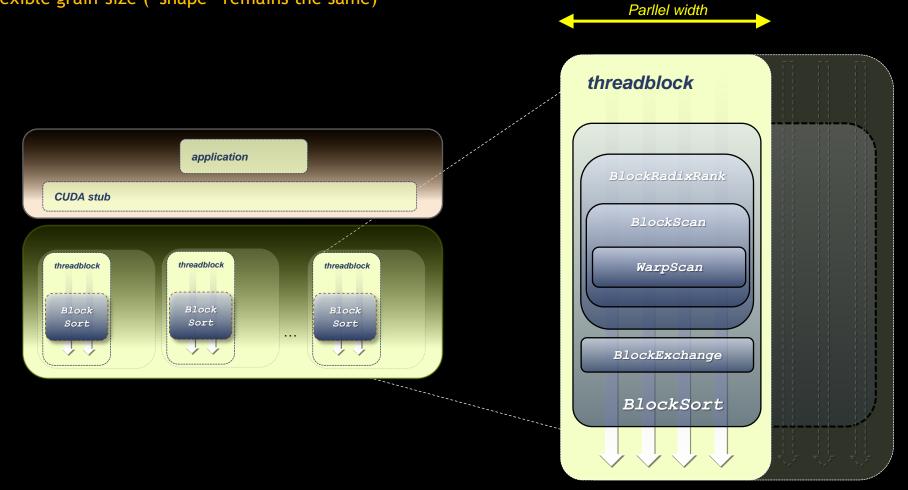




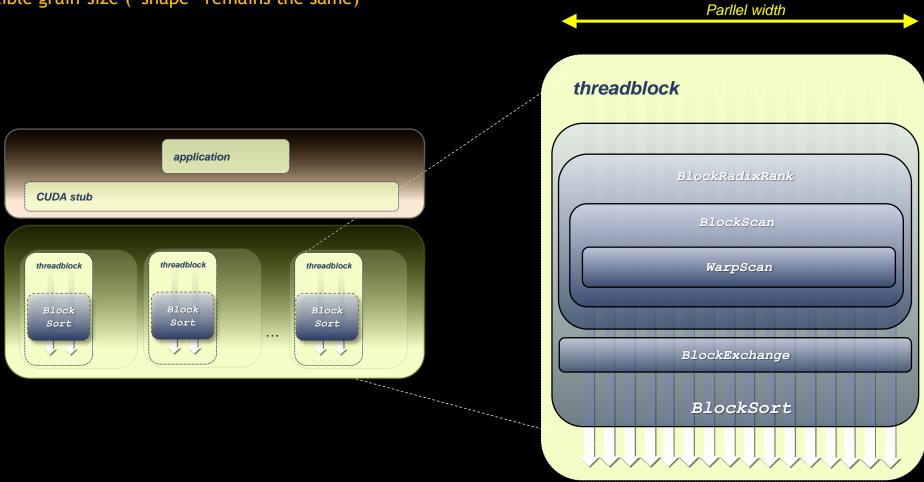


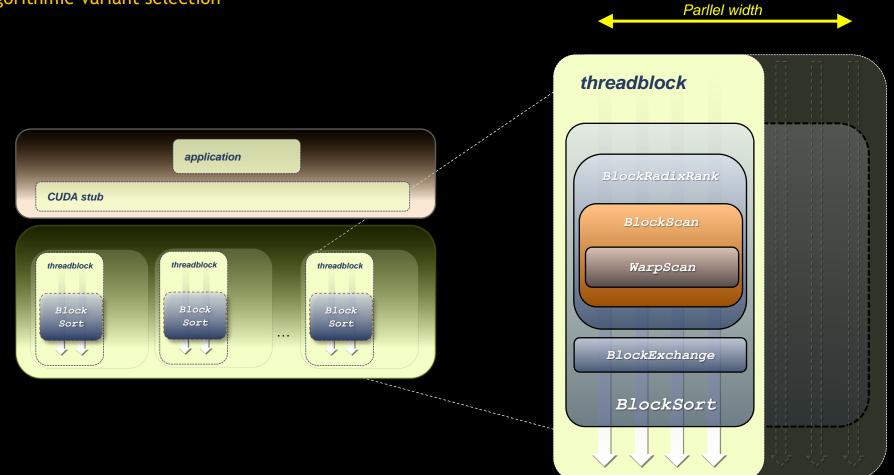


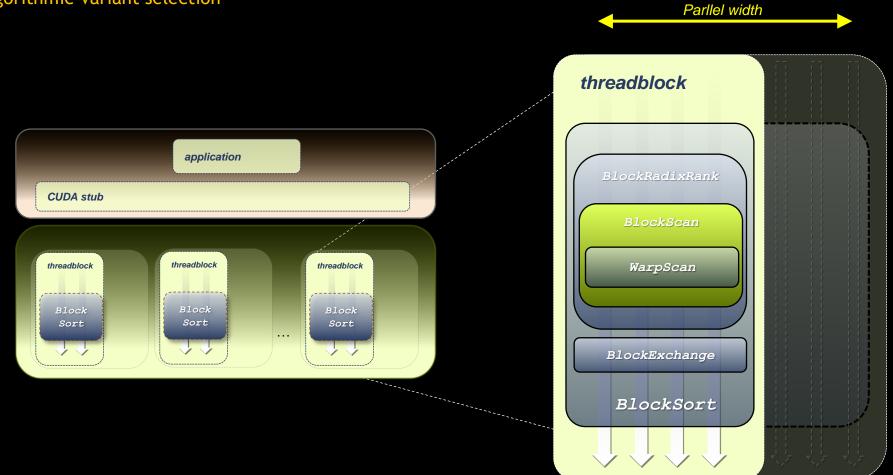
Flexible grain-size ("shape" remains the same)

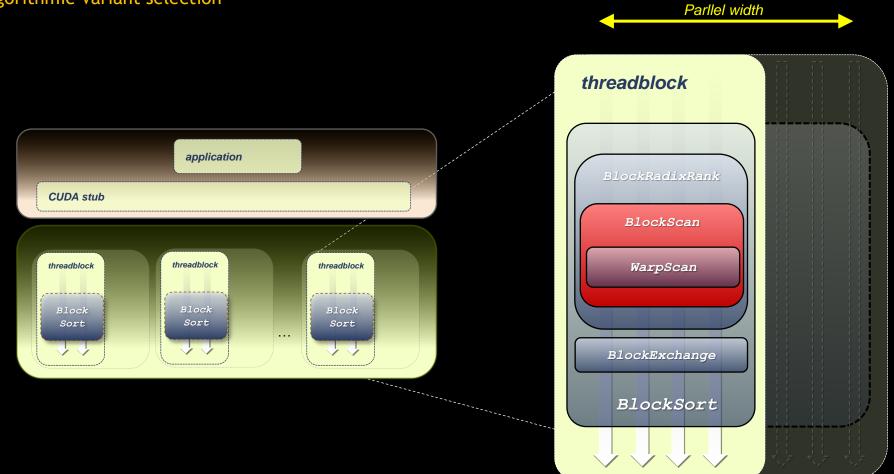


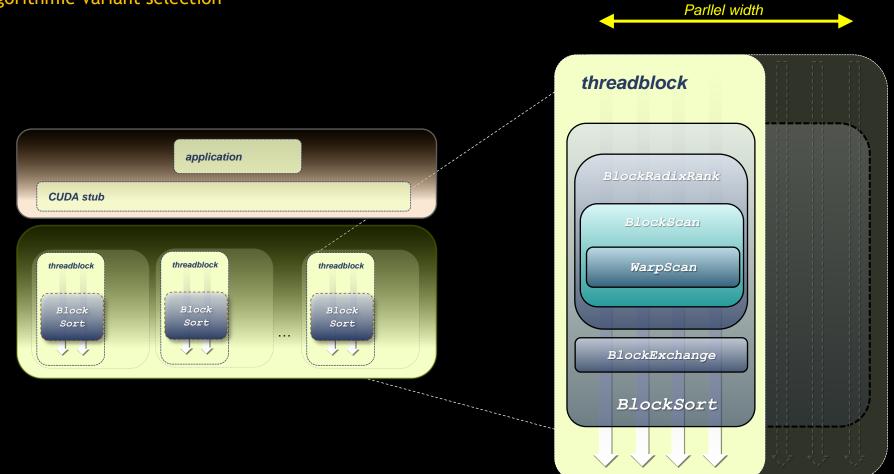
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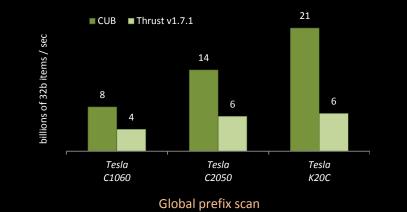
CUB: device-wide performance-portability

vs. Thrust and NPP across the last 4 major NVIDIA arch families (Telsa, Fermi, Kepler, Maxwell)





Global partition-if





Global Histogram

Outline

1. Software reuse

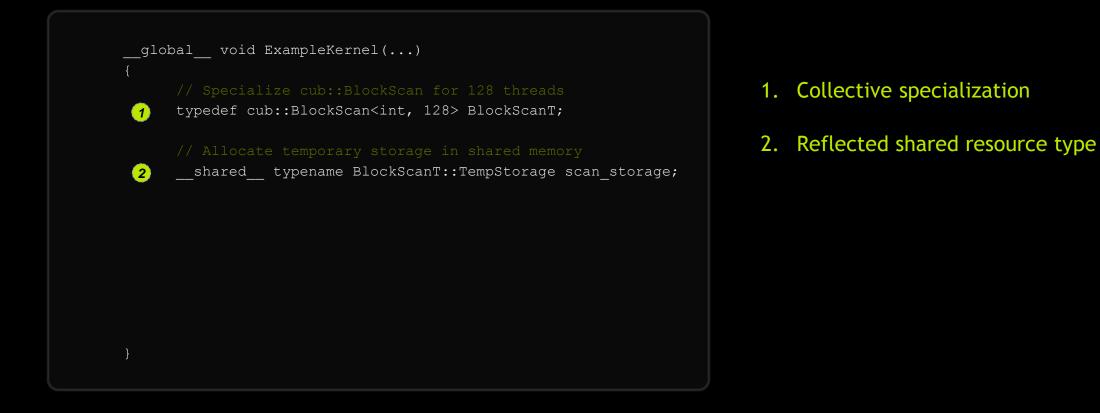
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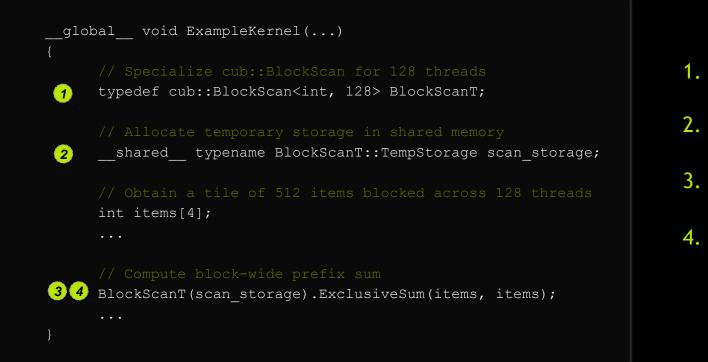
1. Collective specialization

3 parameter fields (specialization, construction, function call) + resource reflection





- 1. Collective specialization
- 2. Reflected shared resource type



- 1. Collective specialization
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- 3. Collective construction
- 4. Collective function call

3 parameter fields (specialization, construction, function call) + resource reflection



- 1. Collective specialization
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// A kernel for computing tiled prefix sums
__global___ void ExampleKernel(int* d_in, int* d_out)

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// Specialize for 128 threads owning 4 integers each

typedef cub::BlockLoad <int*, 128,="" 4=""></int*,>	BlockLoadT;
typedef cub::BlockScan <int, 128=""></int,>	BlockScanT;
<pre>typedef cub::BlockStore<int*, 128,="" 4=""></int*,></pre>	BlockStoreT;

1. Specialize the collective primitive types

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// Allocate temporary storage in shared memory

__shared__ union {

typename BlockLoadT::TempStorage load; typename BlockScanT::TempStorage scan; typename BlockStoreT::TempStorage store; } temp_storage; 1. Specialize the collective primitive types

 Allocate shared memory with a union of TempStorage structuredlayout types

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__shared__ union {

typename BlockLoadT::TempStorage load; typename BlockScanT::TempStorage scan; typename BlockStoreT::TempStorage store;

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// Cooperatively load a tile of 512 items across 128 threads int items[4]; BlockLoadT(temp_storage.load).Load(d_in, items);

- 1. Specialize the collective primitive types
- Allocate shared memory with a union of TempStorage structuredlayout types

3. Block-wide load

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int items[4];
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____syncthreads(); // Barrier for smem reuse

Specialize the collective primitive types

 Allocate shared memory with a union of TempStorage structuredlayout types

- 3. Block-wide load,
- 4. barrier

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// Compute and block-wide exclusive prefix sum
BlockScanT(temp_storage.scan).ExclusiveSum(items, items);

- 1. Specialize the collective primitive types
- Allocate shared memory with a union of TempStorage structuredlayout types

- 3. Block-wide load,
- 4. barrier,
- 5. block-wide scan

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 Allocate shared memory with a union of TempStorage structuredlayout types

- 3. Block-wide load,
- 4. barrier,
- 5. block-wide scan,
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- 1. Specialize the collective primitive types
- Allocate shared memory with a union of TempStorage structuredlayout types

- 3. Block-wide load,
- 4. barrier,
- 5. block-wide scan
- 6. barrier,
- 7. block-wide store

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// Compute and block-wide exclusive prefix sum
BlockScanT(temp_storage.scan).ExclusiveSum(items, items);

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// Cooperatively store a tile of 512 items across 128 threads
BlockStoreT(temp_storage.load).Store(d_in, items);

int* d_in; // = ...
int* d_out; // = ...

// Invoke kernel (GF110 Fermi)

ExampleKernel <<<1, 128>>>(

d_in, d out);

template <typename T>

__global__ void ExampleKernel(T* d_in, T* d_out)

// Specialize for 128 threads owning 4 Ts each

typedef cub::BlockLoad<T*, 128, 4> BlockLoadT; typedef cub::BlockScan<T, 128> BlockScanT; typedef cub::BlockStore<T*, 128, 4> BlockStoreT;

// Allocate temporary storage in shared memory

__shared__ union {

typename BlockLoadT::TempStorage load; typename BlockScanT::TempStorage scan; typename BlockStoreT::TempStorage store; } temp_storage;

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ExampleKernel <<<1, 128>>>(

d_in, d out);

template <int BLOCK_THREADS, typename T>
__global__ void ExampleKernel(T* d_in, T* d_out)

// Specialize for BLOCK_THREADS threads owning 4 integers each
typedef cub::BlockLoad<T*, BLOCK_THREADS, 4> BlockLoadT;
typedef cub::BlockScan<T, BLOCK_THREADS> BlockScanT;
typedef cub::BlockStore<T*, BLOCK_THREADS, 4> BlockStoreT;

// Allocate temporary storage in shared memory

___shared___ union {

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// Allocate temporary storage in shared memory

___shared___ union {

typename BlockLoadT::TempStorage load; typename BlockScanT::TempStorage scan; typename BlockStoreT::TempStorage store; } temp_storage;

// Cooperatively load a tile of items

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ExampleKernel <128, 4> <<<1, 128>>> (

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__global__ void ExampleKernel(T* d_in, T* d_out)

// Specialize for BLOCK_THREADS threads owning ITEMS_PER_THREAD integers each
typedef cub::BlockLoad<T*, BLOCK_THREADS, ITEMS_PER_THREAD, LOAD_ALGO> BlockLoadT;
typedef cub::BlockScan<T, BLOCK_THREADS> BlockScanT;
typedef cub::BlockStore<T*, BLOCK_THREADS, ITEMS_PER_THREAD> BlockStoreT;

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BlockStoreT(temp_storage.load).Store(d_in, items);

int* d_in; // = ...
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template <int BLOCK_THREADS, int ITEMS_PER_THREAD, BlockLoadAlgorithm LOAD_ALGO, BlockScanAlgorithm SCAN_ALGO, typename T> __global__ void ExampleKernel(T* d_in, T* d_out)

// Specialize for BLOCK_THREADS threads owning ITEMS_PER_THREAD integers each
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 d_in,
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// Specialize for BLOCK_THREADS threads owning ITEMS_PER_THREAD integers each
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typedef cub::BlockScan<T, BLOCK_THREADS> BlockScanT;
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// Allocate temporary storage in shared memory

___shared___ union {

typename BlockLoadT::TempStorage load;

typename BlockScanT::TempStorage scan;

typename BlockStoreT::TempStorage store;

} temp_storage;

// Cooperatively load a tile of items

T items[ITEMS_PER_THREAD];

typedef cub::CacheModifiedInputIterator<LOAD_MODIFIER, T> InputItr; BlockLoadT(temp_storage.load).Load(InputItr(d_in), items);

_____syncthreads(); // Barrier for smem reuse

// Compute and block-wide exclusive prefix sum
BlockScanT(temp_storage.scan).ExclusiveSum(items, items);

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int* d_in; // = ...
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ExampleKernel <128, 4, BLOCK_LOAD_WARP_TRANSPOSE, LOAD_DEFAULT, BLOCK_SCAN_RAKING> <<<1, 128>>>(d_in, d_out);

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int* d_in; // = ...
int* d_out; // = ...

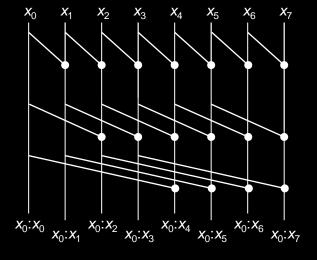
// Invoke kernel (GK110 Kepler)
ExampleKernel <128, 21, BLOCK_LOAD_DIRECT,
 LOAD_LDG, BLOCK_SCAN_WARP_SCANS> <<<1, 128>>>(
 d_in,
 d_out);

Outline

1. Software reuse

- 2. SIMT collectives: the "missing" CUDA abstraction layer
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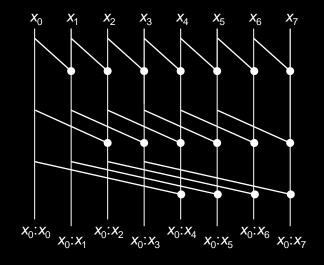
// Simple collective primitive for block-wide prefix sum
template <typename T, int BLOCK_THREADS>
class BlockScan



ł

```
// Simple collective primitive for block-wide prefix sum
template <typename T, int BLOCK_THREADS>
class BlockScan
{
```

// Type of shared memory needed by BlockScan
typedef T TempStorage[BLOCK_THREADS];

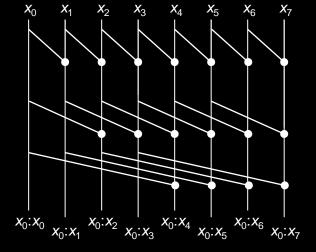


```
// Simple collective primitive for block-wide prefix sum
template <typename T, int BLOCK_THREADS>
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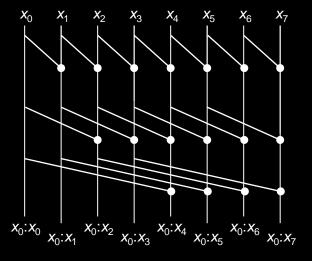
// Type of shared memory needed by BlockScan
typedef T TempStorage[BLOCK THREADS];

// Per-thread data (reference to shared storage)
TempStorage &temp_storage;

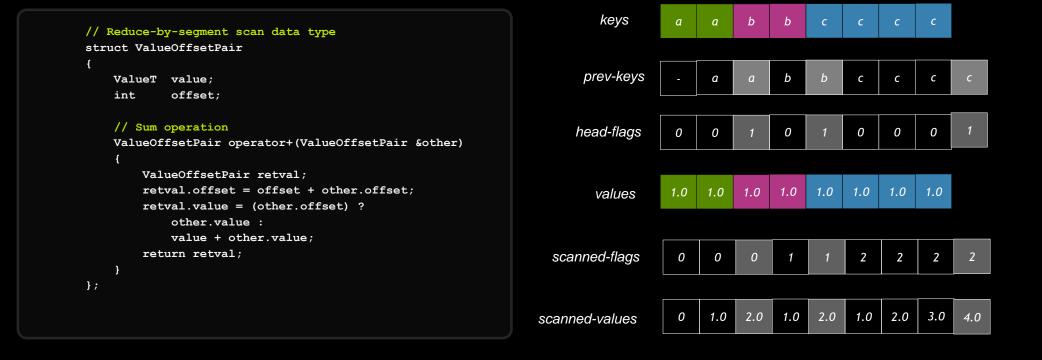
// Constructor
BlockScan (TempStorage &storage) : temp_storage(storage) {}



```
// Simple collective primitive for block-wide prefix sum
template <typename T, int BLOCK THREADS>
class BlockScan
   // Type of shared memory needed by BlockScan
    typedef T TempStorage[BLOCK THREADS];
    // Per-thread data (reference to shared storage)
    TempStorage &temp storage;
    // Constructor
    BlockScan (TempStorage &storage) : temp storage(storage) {}
    // Inclusive prefix sum operation (each thread contributes its own data item)
   T InclusiveSum (T thread data)
        #pragma unroll
        for (int i = 1; i < BLOCK THREADS; i *= 2)</pre>
            temp storage[tid] = thread data;
            syncthreads();
            if (tid - i >= 0) thread data += temp storage[tid];
            ____syncthreads();
        return thread data;
};
```



Block-wide reduce-by-key (simplified)



Block-wide reduce-by-key (simplified)

```
// Block-wide reduce-by-key
template <typename KeyT, typename ValueT, int BLOCK THREADS, int ITEMS PER THREAD>
struct BlockReduceByKey
{
   // Parameterized BlockDiscontinuity type for keys
   typedef BlockDiscontinuity<KeyT, BLOCK THREADS> BlockDiscontinuityT;
   // Parameterized BlockScan type
   typedef BlockScan<ValueOffsetPair, BLOCK THREADS> BlockScanT;
   // Temporary storage type
   union TempStorage
    {
        typename BlockDiscontinuityT::TempStorage discontinuity;
        typename BlockDiscontinuityT::TempStorage scan;
   };
    // Reduce segments using addition operator.
    // Returns the "carry-out" of the last segment
    ValueT Sum(
      TempStorage& temp storage,
                                                       // shared storage reference
                   keys[ITEMS PER THREAD],
                                                       // [in|out] keys
      KeyT
                   values[ITEMS PER THREAD],
                                                       // [in|out] values
      ValueT
                   segment indices [ITEMS PER THREAD]) // [out] segment indices (-1 if invalid)
      int
};
```

Block-wide reduce-by-key (simplified)

// Reduce segments using addition operator.

```
// Returns the "carry-out" of the last segment
ValueT Sum(
```

ValueT Sum(

KeyT keys[ITEMS PER THE	EAD1.
ValueT values[ITEMS_PER_T	HREAD],
int segment_indices[I]	EMS_PER_THREAD])

KeyT prev_keys[ITEMS_PER_THREAD]; ValueOffsetPair scan_items[ITEMS_PER_THREAD];

// Set head segment_flags.

BlockDiscontinuityKeysT(temp_storage.discontinuity).FlagHeads(
 segment_indices, keys, prev_keys);

_____syncthreads();

```
// Unset the flag for the first item
if (threadIdx.x == 0)
    segment indices[0] = 0;
```

```
// Zip values and segment_flags
```

```
for (int ITEM = 0; ITEM < ITEMS_PER_THREAD; ++ITEM)
{</pre>
```

```
scan_items[ITEM].offset = segment_indices[ITEM];
scan_items[ITEM].value = values[ITEM];
```

```
}
```

// Exclusive scan of values and segment_flags
ValueOffsetPair tile_aggregate;
BlockScanT(temp_storage.scan).ExclusiveSum(
 scan_items, scan_items, tile_aggregate);

. . .

// Unzip values and segment indices

```
for (int ITEM = 0; ITEM < ITEMS_PER_THREAD; ++ITEM)
{
    segment_indices[ITEM] = segment_indices[ITEM] ?
        scan_items[ITEM].offset :
        -1;
    keys[ITEM] = prev_keys[ITEM];
    values[ITEM] = scan_items[ITEM].value;
}</pre>
```

// Return "carry-out"
return tile_aggregate.value;

]

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Cache-modified input iterators

```
#include <cub/cub.cuh>
// Standard layout type
struct Foo
    double x;
    char y;
global void Kernel(Foo* d in, Foo* d out)
   // In host or device code: create an LDG wrapper
   cub::CacheModifiedInputIterator<cub::LOAD LDG, Foo> ldg itr(d in);
    cub::CacheModifiedOutputIterator<cub::STORE WT, Foo> volatile itr(d out);
   volatile itr[threadIdx.x] = ldg itr[threadIdx.x];
```

code for sm_35
 Function : _Z6KernelPdS_
MOV R1, c[0x0][0x44];
S2R R0, SR_TID.X;
ISCADD R2, R0, c[0x0][0x140], 0x3;
LDG.64 R4, [R2];
LDG.64 R4, [R2];
ISCADD R0, R0, c[0x0][0x144], 0x4;
TEXDEPBAR 0x1;
ST.WT.64 [R0], R4;
TEXDEPBAR 0x0;
ST.WT.64 [R0+0x8], R2;
EXIT;

LOAD_DEFAULT, ///< Default (no modifier) LOAD_CA, ///< Cache at all levels LOAD_CG, ///< Cache at global level LOAD_CS, ///< Cache streaming (likely to be accessed once) LOAD_CV, ///< Cache as volatile (including cached system line LOAD_LDG, ///< Cache as texture LOAD_VOLATILE, ///< Volatile (any memory space)

Texture obj (and ref) input iterators

```
// Standard layout type
struct Foo
    double x;
template <typename InputIteratorT, typename OutputIteratorT>
global void Kernel(InputIteratorT d in, OutputIteratorT d out)
   d out[threadIdx.x] = d in[threadIdx.x];
// Create a texture object input iterator
Foo* d foo;
cub::TexObjInputIterator<Foo> d foo tex;
d foo tex.BindTexture(d foo);
Kernel<<<1, 32>>>(d foo tex, d foo);
```

d_foo_tex.UnbindTexture();

code for sm_35
Function :
 _Z6KernelIN3cub19TexObjInputIteratorI3FooiEEPS
2 _EvT_T0_

MOV R1, c[0x0][0x44]; S2R R0, SR_TID.X; IADD R2, R0, c[0x0][0x144]; SHF.L R2, RZ, 0x1, R2; IADD R3, R2, 0x1; TLD.LZ.T R2, R2, 0x52, 1D, 0x1; TLD.LZ.P R4, R3, 0x52, 1D, 0x3; ISCADD R0, R0, c[0x0][0x150], 0x4; TEXDEPBAR 0x1; ST [R0], R2; TEXDEPBAR 0x0; ST.64 [R0+0x8], R4; EXIT;

Collective primitives

WarpReduce

- reduction & segmented reduction
- WarpScan

- BlockDiscontinuity
- BlockExchange
- BlockHistogram
- BlockLoad & BlockStore
- BlockRadixSort
- BlockReduce
- BlockScan

Device-wide (global) primitives

(Usable with CDP, streams, and your own memory allocator)

- DeviceHistogram
 - histogram-even
 - histogram-range
- DevicePartition
 - partition-if
 - partition-flagged
- DeviceRadixSort
 - ascending / descending

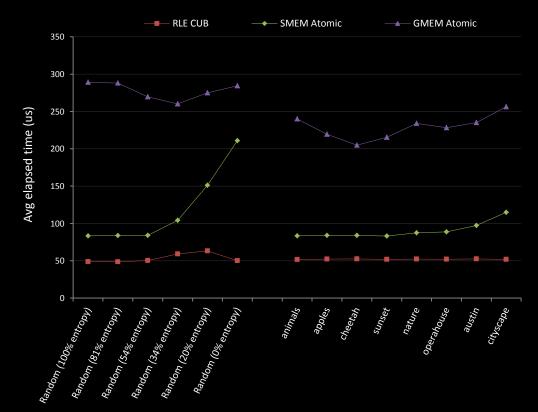
- DeviceReduce
 - reduction
 - arg-min, arg-max
 - reduce-by-key
- DeviceRunLengthEncode
 - RLE
 - Non-trivial segments

- DeviceScan
 - inclusive / exclusive
- DeviceSelect
 - select-flagged
 - select-if
 - keep-unique
- DeviceSpmv

NEW: performance-resilient histogram

Simple intra-thread RLE provides a uniform performance response regardless of input sample distribution

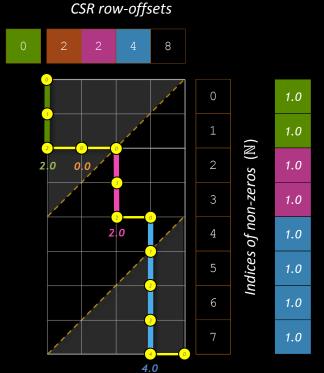
```
// RLE pixel counts within the thread's pixels
int accumulator = 1;
for (int PIXEL = 0;
     PIXEL < PIXELS PER THREAD - 1;
     ++PIXEL)
    if (bins[PIXEL] == bins[PIXEL + 1])
         accumulator++;
    else
         atomicAdd(
             privatized histogram + bins[PIXEL],
             accumulator);
         accumulator = 1;
```



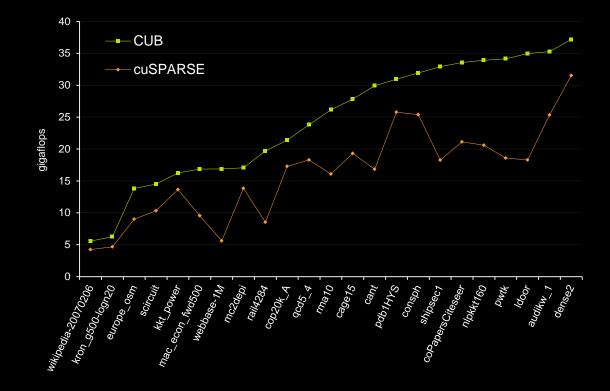
GeForce GTX980: 1-channel (1920x1080 uchar1 pixels)

NEW: CSR SpMV

Merge-based parallel decomposition for load balance







fp32 (Tesla K40)

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Benefits of using CUB primitives

- Simplicity of composition
 - Kernels are simply sequences of primitives
- High performance
 - CUB uses the best known algorithms, abstractions, and strategies, and techniques

Performance portability

• CUB is specialized for the target hardware (e.g., memory conflict rules, special instructions, etc.)

Simplicity of tuning

- CUB adapts to various grain sizes (threads per block, items per thread, etc.)
- CUB provides alterative algorithms

Robustness and durability

CUB supports arbitrary data types and block sizes

Questions?

Please visit the CUB project on GitHub <u>http://nvlabs.github.com/cub</u>

Duane Merrill (dumerrill@nvidia.com)





THANK YOU

JOIN THE CONVERSATION #GTC15 9 f in

