COOPERATIVE GROUPS

FLEXIBLE GROUPS OF THREADS

21 June 2023 | Andreas Herten | Forschungszentrum Jülich ***Handout Version***

Member of the Helmholtz Association

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Overview, Outline

At a Glance

- Cooperative Groups: New model to work with thread groups
- Thread groups are entities, intrinsic function as member functions

Contents

Gather Last-Minute Material

Place cursor in box when done:

Now run

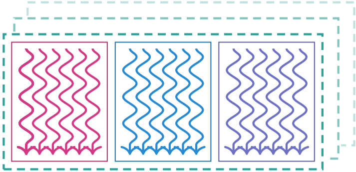
```
jsc-material-reset-03
```

I’m done!

Standard CUDA Threading Model

Before CUDA 9

- Many threads, combined into blocks, on a grid; in 3D
- Operation: Single Instruction, Multiple Threads (SIMT)
- Thread waiting for result of instruction? Use computational resource with other threads in meantime!
- Group of threads execute in lockstep: **Warp** (currently 32 threads)
  - Same instructions
  - Branching possible
  - Predicates (and masks)
- Shared memory: Fast, shared between threads of block
- Synchronization between threads of blocks:
  `__syncthreads()` – barrier for all threads of block



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Cooperative Groups

Introduction

New Model: Cooperative Groups

- Motivation to extend classical model

Algorithmic	Not all algorithms map easily to available synchronization methods; **synchronization** should be more flexible; easier to utilize low-level concepts
Design	Make groups of threads explicit **entities**
Hardware	Access new **hardware features** (*Independent Thread Scheduling*, *Thread Block Clusters*)

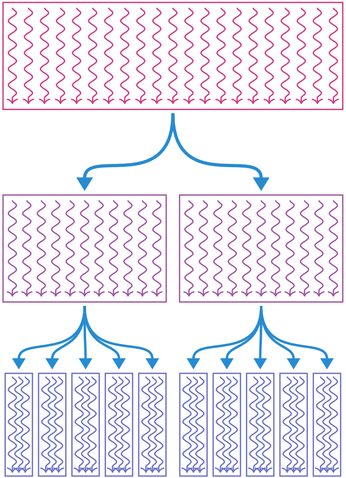
→ **Cooperative Groups** (CG)

A flexible model for synchronization and communication within groups of threads.

- All in `namespace cooperative_groups` (`cooperative_groups.h` header)
- Following in text: `cooperative_groups::func()` ⟶ `cg::func()`

```
namespace cg = cooperative_groups;
```

Division of Thread Blocks



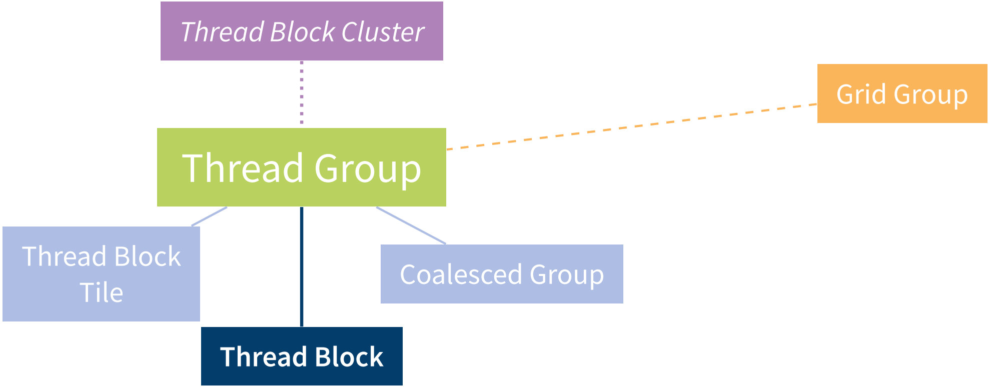
- Start with block of certain size
- Divide into smaller sub-groups
- Continue diving, if algorithm makes it necessity
- Methods for dynamic or static divisions (*tiles*)
- In each level: thread of group has unique ID (local index instead of global index)

→ Use functions and collectives on sub-set of all threads

Cooperative Groups

Thread Groups Overview

Thread Group Landscape



Common Methods of Cooperative Groups

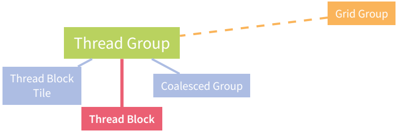
- Fundamental type: `thread_group`
- Every CG has following member functions

`sync()`	Synchronize the threads of this group (alternative `cg::sync(g)`) *Before: `__syncthreads()` for whole block*
`thread_rank()`	Get unique ID of current thread in this group (*local index*) *Before: `threadIdx.x` for index in block*
`size()`	Number of threads in this group *Before: `blockDim.x` for number of threads in block*
`is_valid()`	*Group is technically ok*

Cooperative Groups

Thread Blocks

Cooperative Thread Blocks



- Easiest entry point to thread groups: `cg::this_thread_block()`
- Additional member functions

`thread_index()`	Thread index within block (3D)
`group_index()`	Block index within grid (3D)

- Blocks (and groups) are now concrete entities

→ Design functions to represent this!

Example: Print Rank Function

```
__device__ void printRank(cg::thread_group g) {
    printf("Rank %d\n", g.thread_rank());
}
__global__ void allPrint() {
    cg::thread_block b = cg::this_thread_block();

    printRank(b);
}
int main() {
    allPrint<<<1, 23>>>();
}
```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Task Base Code: Shared Memory Reduction

Inner logic: Function

```
int * array;
cudaMallocManaged(&array, sizeof(int) * N);

for (int i = 0; i < N; i++)
    array[i] = rand() % 1024;

int blocks = 1;
int threads = N;
maxKernel<<<blocks, threads, threads * sizeof(int)>>>(array);
```

Allocate this much shared memory per block

```
__global__ void maxKernel(int * array) {
    extern __shared__ int shmem_temp[];  // threads * sizeof(int)

    int threadIndex = threadIdx.x;
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```

One value for each thread

```
    int maxValue = maxFunction(shmem_temp, myValue);
```

Call function with

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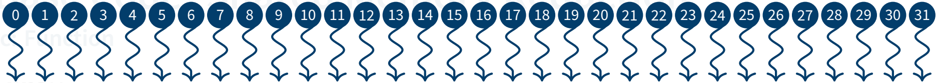


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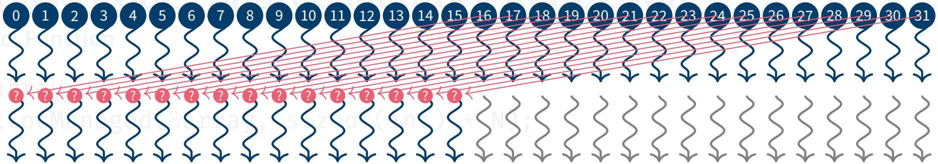


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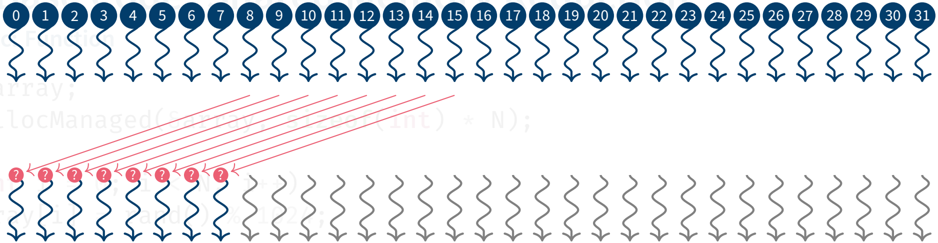


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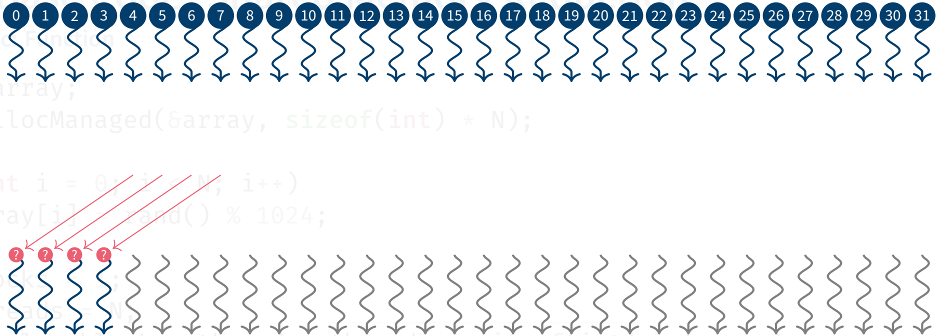


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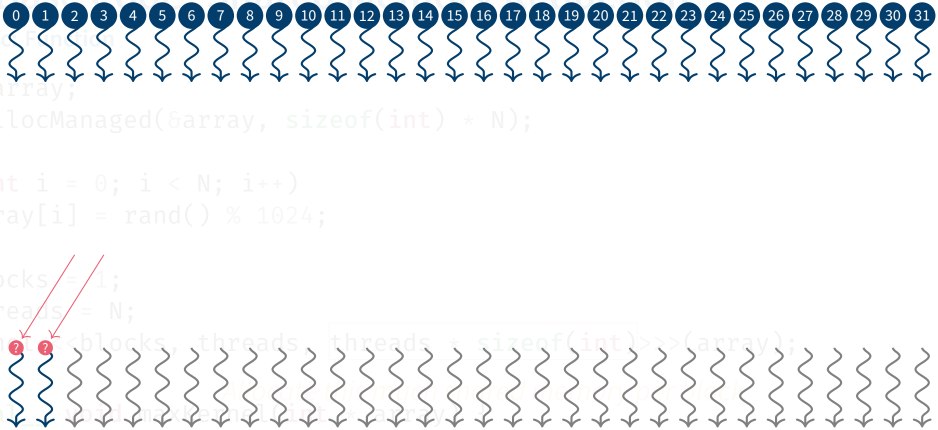


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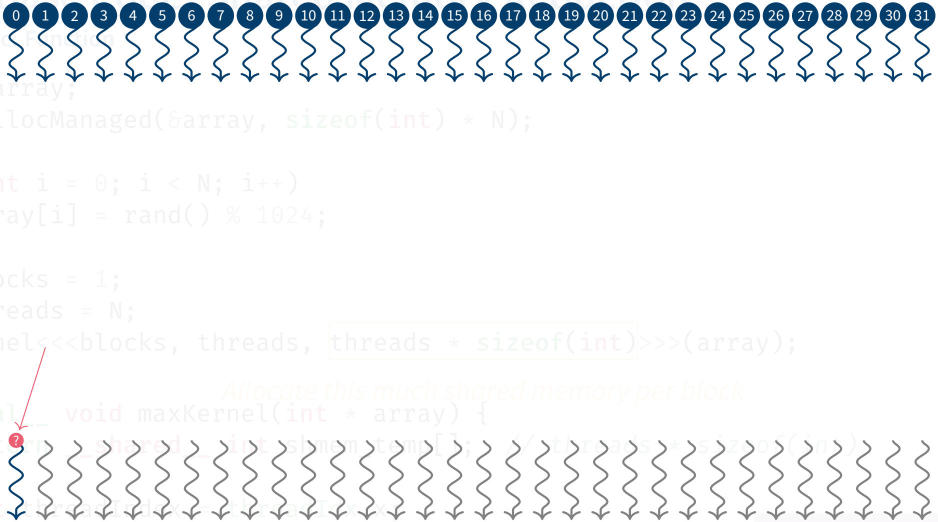


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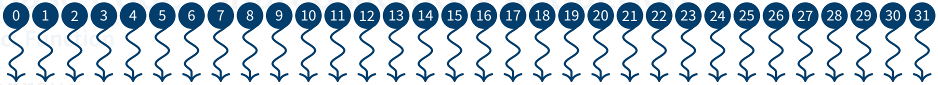


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One value for each thread

Call function with

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Implementing a Cooperative Groups Kernel

TASK 1

From old to new

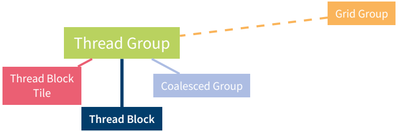
- Location of code: `03-Cooperative_Groups/exercises/tasks/task1`
- See `Instructions.md` for explanations
- Follow TODOs to port kernel/device function from traditional CUDA threading model to new CG model
- Compile with `make`, submit to batch system with `make run`
- See also CUDA C programming guide for details on Cooperative Groups

Cooperative Groups

Tiling Groups

Tiles of Groups

Dynamically-tiled



- Divide into smaller groups with `cg::tiled_partition()`
- Will automatically create smaller groups from parent group
- Examples
  - Create groups of size 32 of current block

    ```
    cg::thread_group tile32 = cg::tiled_partition(cg::this_thread_block(), 32);
    ```
  - Create sub-groups of size 4

    ```
    cg::thread_group tile4  = cg::tiled_partition(tile32, 4);
    ```
- **Note**: Currently, only supported partition sizes are 1, 2, 4, 8, 16, 32

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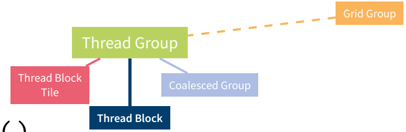
Tiles of Groups

Statically-tiled: `thread_block_tile`



- Second version of function: `cg::tiled_partition<>()`
- Size of tile is template parameter

→ Known at compile time! Optimizations possible!

- Partition size: 1, . . . , 32, **64**, **128**, **256**, **512**! *(<A100: extra work needed)*
- Returns `thread_block_tile` object with additional member functions
  - `.shfl()`, `.shfl_down()`, `.shfl_up()`, `.shfl_xor()`
  - `.any()`, `.all()`, `.ballot()`; `.match_any()`, `.match_all()`

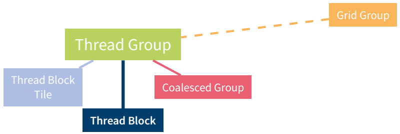
  → Intrinsic functions to work with threads inside a warp *(more later)*
- Example

```
cg::thread_block_tile<32> tile32 = cg::tiled_partition<32>(cg::this_thread_block());
cg::thread_block_tile<4>  tile4  = cg::tiled_partition<4> (tile32);
```

Cooperative Groups

Coalesced Groups

Coalesced Group



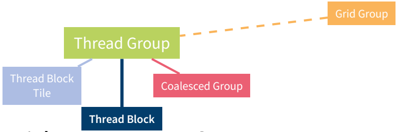
- Get group of threads which is not diverged
- Threads have same state at point of API call
- `cg::coalesced_group active_threads = cg::coalesced_threads();`
- Example

```
cg::coalesced_group active_threads = cg::coalesced_threads();
if (i < 5) {
    cg::coalesced_group if_true_threads = cg::coalesced_threads();
    int rank = if_true_threads.thread_rank();
    cg::thread_group partition = cg::tiled_partition(if_true_threads, 2);
}
```

Cooperative Groups

Binary Partition

Binary Partition



- Get group of coalesced threads for which a condition is either `true` or `false`
- Threads have same state at point of API call and belong to one of two *buckets*
- `cg::coalesced_group partitioned_threads = cg::binary_partition(group, condition);`
- *Beta* feature, details might change
- Example

```
cg::thread_block cta = cg::this_thread_block();
cg::thread_block_tile<32> tile32 = cg::tiled_partition<32>(cta);
auto subTile = cg::binary_partition(tile32, isEven(array[cta.thread_rank()]) );
```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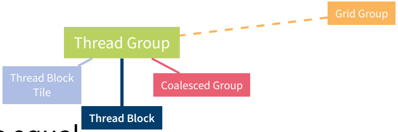
Cooperative Groups

Labeled Partition

Labeled Partition



- Get group of coalesced threads for which a condition is equal
- Threads have same state at point of API call and belong to same *bucket*
- Extension of binary partition to general case
- `cg::coalesced_group partitioned_threads = cg::labeled_partition(group, condition);`
- *Beta* feature, details might change
- Example

```
cg::coalesced_group active = cg::coalesced_threads();
auto labeledGroup = cg::labeled_partition(active, bucket);
```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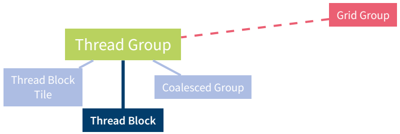
Cooperative Groups

Larger Groups

Grid Group



- Grid of blocks can also be entity now
- Synchronize across all blocks:
```
cg::grid_group grid = cg::this_grid();
grid.sync();
```
- Condition
  1. Blocks must be co-resident on device (Occupancy Calculator)
  2. Kernel must be launched with Cooperative Launch API `cudaLaunchCooperativeKernel()` instead of `<<<,>>>` syntax

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Cooperative Groups with Tiled Partitions

TASK 2

Sub-divisions

- Location of code: `03-Cooperative_Groups/exercises/tasks/task2`
- See `Instructions.md` for explanations
- Follow TODOs to tile a CG and use kernel from Task 1; atomic operations needed

  Aside!
- Compile with `make`, submit to batch system with `make run`
- See also CUDA C programming guide for details on Cooperative Groups

Aside: Atomic Operations

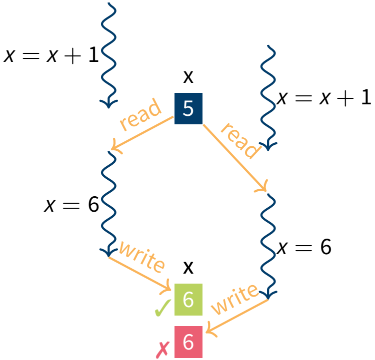
Motivation

- Order execution of CUDA threads non-deterministic
- No problem, if each thread works on distinct data element
- What, if threads collaborate and share data? Read/Write to same element?

→ Atomic operations
  - Safe way to read and write to memory position by different threads
  - Data in global or shared memory
  - Example: `atomicAdd(&array[i], myvalue)`

- See CUDA Documentation

`array[1] = array[1] + myvalue`



Aside: Atomic Operations

Examples

- First argument to function (always): `address` of a value to potentially change
- Old value of `address` usually returned
- `int atomicOp(int * removeVal, int myVal)`
- Examples

`atomicAdd(int* address, int val)` Add `val` to the value at `address`

`atomicExch(int* address, int val)` Store `val` at `address` location; return old value

`atomicMin(int* address, int val)` Store the minimum of `val` and the value at `address` at `address` location; return old value

`atomicCAS(int* address, int compare, int val)` The value at `address` is compared to `compare`. If true, `val` is stored at `address`; if false, the old value at `address` is stored. The old value at `address` is returned. Basic function: Compare And Swap

Cooperative Groups with Tiled Partitions

TASK 2

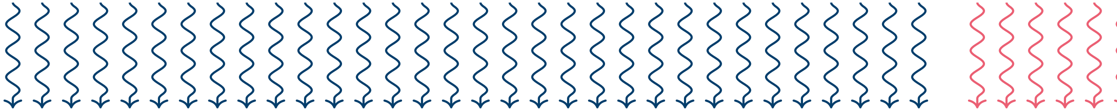
Sub-divisions

- Location of code: `03-Cooperative_Groups/exercises/tasks/task2`
- See `Instructions.md` for explanations
- Follow TODOs to tile a CG and use kernel from Task 1; atomic operations needed
- Compile with `make`, submit to batch system with `make run`
- See also CUDA C programming guide for details on Cooperative Groups

Warp-Synchronous Programming

Warp-Level Intrinsics

- Smallest set of executed threads: Warp
- Warp: 32 threads executed in SIMT/SIMD fashion
- Exchange data between threads of warp
  - Global memory: Slow
  - Shared memory: Faster
  - Directly (registers): Even faster
- Safe method access without race conditions
  - Global/shared memory: Atomic operations
  - Registers: **Warp-aggregated Atomic operations**



Warp Intrinsics Overview

`shfl(int lane)`	Copy data from a target warp lane; also: other flavors (next slide)
`all(int pred)`	If predicate (*comparison, relation*) evaluates to non-zero (*true*) for all threads, return non-zero (*true*)
`any(int pred)`	If predicate evaluates to non-zero for any thread, return non-zero
`ballot(int pred)`	Return a bit mask which has 1s set for all thread for which predicate evaluates to non-zero
`match_any(T value)`	Return a bit mask of threads which have same value of `value` as current thread; also: `match_all(T value)`

- Available as global device functions, with additional selection *mask* as first element (as `__shufl_sync()` etc.)
- Available as **member functions** of a `cg::tiled_partition` group (as `g.shfl()` etc.)
- Intrinsics automatically synchronize after operation – new since CUDA 9
- Value can only be retrieved if targeted lane also invokes intrinsic
- Per clock cycle: 32 shuffle instructions per SM → **very fast!**

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Warp Intrinsic Example

Everyday I'm Shuffeling

- `shfl()`: Copy data from target warp lane
- Different flavors

`shfl()`	Copy data from warp lane with ID directly
`shfl_up()`	Copy data from relative warp lane with lower ID (shuffle *upstream*)
`shfl_down()`	Copy data from relative warp lane with higher ID (shuffle *downstream*)
`shfl_xor()`	Copy data from relative warp lane with ID as calculated by a bitwise XOR

- **Example: `shfl_down(value, N)` with `N` = 16, 8, …**

Kernel → Warp-Level Reduction w/o Shared Memory

TASK 3

Expert level 11

- Location of code: `03-Cooperative_Groups/exercises/tasks/task3`
- See `Instructions.md` for explanations
- Follow TODOs to modify `maxKernel()` such that it uses warp-level atomic operations (and no shared memory)
- Compile with `make`, submit to batch system with `make run`
- See also CUDA C programming guide for details on warp-level functions

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Collective Operations

Collective Operations

- In-group programming (ideally: warp-level programming) can get last bits of performance; but quite advanced
- Help: Collective operations on thread groups (new and slightly less advanced)

`cg::sync()`	Synchronize threads in group
`cg::memcpy_async()`	Copy from global to shared memory in group, non-blocking; also: `cg::wait`
`cg::reduce()`	Reduction operation in group; hardware-accelerated operators: `plus()`, `less()`, `greater()`, `bit_and()`, `bit_xor()`, `bit_or()`
`cg::inclusive_scan()`	Scan operation in group (also: `cg::exclusive_scan()`)

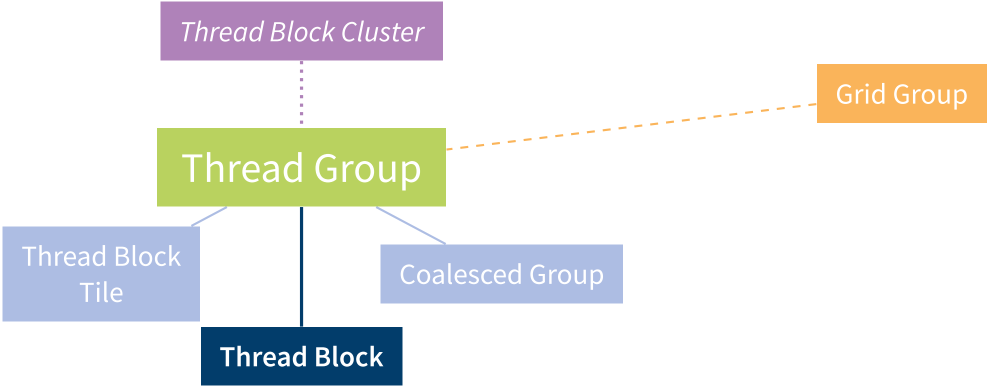
Cooperative Reduce Collective Example

```
__shared__ int reduction_s[BLOCKSIZE];
cg::thread_block cta = cg::this_thread_block();
cg::thread_block_tile<32> tile = cg::tiled_partition<32>(cta);

const int tid = cta.thread_rank();
int value = A[tid];
reduction_s[tid] = cg::reduce(tile, value, cg::plus<int>());
// reduction_s contains tile-sum at all positions associated to tile
cg::sync(cta);
// Still to do: sum partial tile sums
```

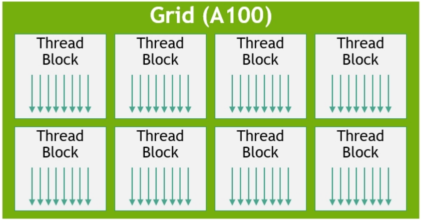
Block Clusters

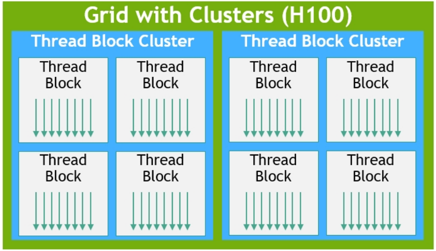
Thread Group Landscape



Moare Hierarchy

- New feature available in next-gen H100 GPU in Compute Capability 9.0
- Extend hierarchy:
  Threads → Thread Blocks → Grids Threads → Thread Blocks → Thread Block Clusters → Grids
- Exposes the GPC (*GPU Processing Cluster*) hardware to software – **only through CG**
- Enables collaboration of some SMs of GPC; access shared memory (incl. atomics, like `sync( )`); max. 16 blocks per cluster





Using Block Clusters

- Two possibilities for usage
  1. Through annotating intrinsic at kernel definition `__cluster_dims__(X,Y,Z)` (compile-time only)
  2. Through special kernel launch call `cudaLaunchKernelEx()` (also run-time)
- Guaranteed to be co-scheduled (running at same time)
- Use `cg::this_cluster` to get cluster
- Member functions (highlights)

`sync()`	Sync in the cluster
`thread_rank()`	Get rank within cluster
`map_shared_rank()`	Get address of shared memory of another block of cluster

- See cluster group documentation and thread block cluster introduction

Conclusions

Conclusions

- **CG** alternative model to create groups
- Groups are **entities**, have member functions
- Synchronizing is important (not mentioned before: `__syncwarps()`)
- **Warp-level functions** easily accessible from groups
- Some new device features only exposed through CG
- See also further literature in Appendix

Thank you for your attention!

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Appendix

Appendix

Further Literature

Glossary

References: Images

Further Literature

- NVIDIA Developer Blog: Cooperative Groups: Flexible CUDA Thread Programming
- NVIDIA Developer Blog: Inside Volta: The World's Most Advanced Data Center GPU
- NVIDIA Developer Blog: Using CUDA Warp-Level Primitives
- Talk at GPU Technology Conference 2018: Cooperative Groups by Kyrylo Perelygin and Yuan Lin
- Talk: Warp-synchronous programming with Cooperative Groups by Sylvain Collange
- Book: CUDA Programming by Shane Cook

Glossary I

API A programmatic interface to software by well-defined functions. Short for application programming interface. 35, 37, 39, 41

CUDA Computing platform for GPUs from NVIDIA. Provides, among others, CUDA C/C++. 4, 30, 42, 43, 45, 48, 50

NVIDIA US technology company creating GPUs. 63

CG Cooperative Groups. 6, 10, 30, 42, 45, 59

GPU Graphics Processing Unit. 63

SIMD Single Instruction, Multiple Data. 47

SIMT Single Instruction, Multiple Threads. 4, 47

SM Streaming Multiprocessor. 48

References: Images, Graphics I

[1] Yuriy Rzhemovskiy. *Teenage Penguins*. Freely available at Unsplash. URL: `https://unsplash.com/photos/qFxS5FkUSAQ`.