CUDA Digging further into the programming manual

Schedule

- Application Programming Interface (API)
 - text only part, sorry
- Image utilities (simple CUDA examples)
- Performace considerations
- Matrix multiplication example
- SDK "browse-through"

API

- A minimal set of extensions to the C language that allow the programmer to target portions of the source code for execution on the device
- □ Runtime library
 - A host component that provides functions to control and access one or more compute devices from the host, e.g. memory allocation.
 - A device component that provides device-specific functions, e.g. thread synchronization.
 - A common component that provides built-in vector types and a subset of the C standard library that are supported in both host and device code.

Language extensions

- Roughly speaking only four additions to standard C
 - Function type qualifiers to specify whether a function executes on the host or on the device
 - Variable type qualifiers to specify the memory location on the device
 - A new directive to specify how a kernel is executed on the device
 - Four built-in variables that specify the grid and block dimensions and the block and thread indices

nvcc

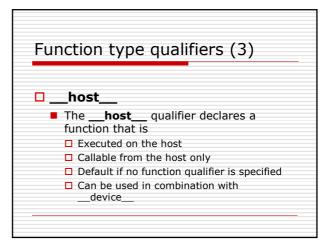
- □ CUDA compiler to handle the standard C extensions
 - Each source file containing these extensions must be compiled with the CUDA compiler nvcc
 - Easily integrated with Visual Studio in Windows or makefiles in Linux

Function type qualifiers (1)

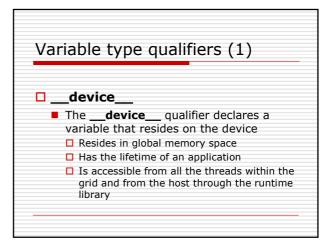
device_

- The __device__ qualifier declares a function that is
 - Executed on the device
 - $\hfill\Box$ Callable from the device only

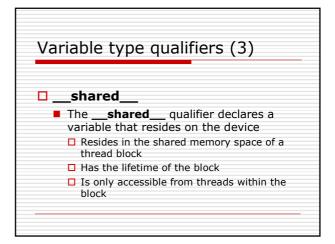
Function type qualifiers (2) __global__ The __global__ qualifier declares a function as being a kernel Executed on the device Callable from the host only



Function type qualifiers restrictions device__ functions are always inlined device__ and __global__ do not support recursion device__ and __global__ cannot declare static variables inside their bodies device__ and __global__ cannot have a variable number of inputs device__ functions cannot have their address taken but function pointers to __global__ functions are supported The __global__ and __host__ qualifiers cannot be used together device__ functions must have void return type Any call to a __global__ function must specify its execution configuration A call to a __global__ function is asynchronous, meaning it returns before the device has completed its execution global__ function parameters are currently passed via shared memory to the device and limited to 256 bytes.



Variable type qualifiers (2) __constant__ The __constant__ qualifier declares a variable that resides on the device Resides in constant memory space Has the lifetime of an application Is accessible from all the threads within the grid and from the host through the runtime library

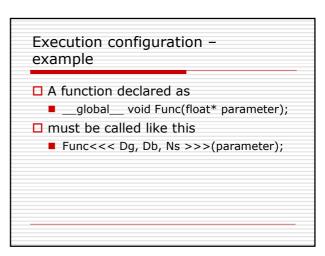


Variable type qualifiers restrictions shared__ and __constant__ cannot be used in combination with each other shared__ and __constant__ variables have implied static storage device__ and __constant__ variables are only allowed at file scope constant__ variables cannot be assigned to from the device, only from the host shared__ variables cannot have an initialization as part of their declaration

Variable type qualifiers – the default case

- An automatic variable declared in device code without any of these qualifiers generally resides in a register
- However in some cases the compiler might choose to place it in local memory
- This is often the case for large structures or arrays that would consume too much register space, and arrays for which the compiler cannot determine that they are indexed with constant quantities

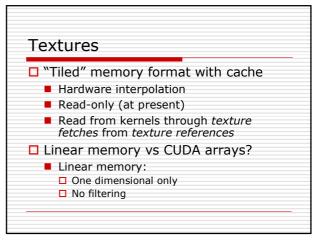
Execution configuration Any call to a _global_ function must specify the execution configuration for that call. The execution configuration defines the dimension of the grid and blocks that will be used to execute the function on the device. It is specified by inserting an expression of the form <<< Dg, Db, Ns >> >> between the function name and the parenthesized argument list, where: ■ Dg is of type dim3 and specifies the dimension and size of the grid, such that 'Dg,x*Dg,y*Dg,y equals the number of blocks being launched ■ Db is of type dim3 and specifies the dimension and size of each block, such that 'Db,x*Db,y*Db,z equals the number of threads per block of the grid is the first of type dim3 and specifies the number of threads per block. ■ Ns is of type size t and specifies the number of bytes in shared method that is dynamically allocated per block for this call in addition the stadio allocated memory. Ns is an optional argument which defaults to 0. The arguments to the execution configuration are evaluated before the actual function arguments.



Built-in variables gridDim This variable is of type dim3 and contains the dimensions of the grid. blockIdx This variable is of type uint3 and contains the block index within the grid. blockDim This variable is of type dim3 and contains the dimensions of the block. threadIdx This variable is of type uint3 and contains the thread index within the block. Restrictions It is not allowed to take the address of any of the built-in variables. It is not allowed to assign values to any of the built-in variables.

Common runtime component Built-in vector types For both host and device char1, char2, char3, char4 uchar1, uchar2, uchar3, uchar4 short1, ushort2, ushort3, ushort4 ushort1, ushort2, ushort3, ushort4 int1, int2, int3, int4 uint1, uint2, uint3, uint4 long1, long2, long3, long4, ulong1, ulong2, ulong3, ulong4, float1, float2, float3, float4 Constructor of the form make_<type_name>(...); For example int2 make_int2(int x, int y);

Mathematical functions <math.mdi>



Device runtime components Synchronization function void syncthreads(); Type conversion unsigned int _float2uint_[rn,rz,ru,rd](float); float _int2float_[rn,rz,ru,rd](int); float _uint2float_[rn,rz,ru,rd](unsigned int); Type casting float _int as_float(int); int _float_as_int(float); Atomic functions Performs a read-modify-write operation on one 32 bit word residing in global memory E.g. unsigned int atomicAdd(unsigned int* address, unsigned int val); At present limited support for integer types only!

Host runtime components Device Multiple devices supported Memory Linear or CUDA arrays OpenGL and DirectX interoperability For visualization of results Asynchronicity global functions and most runtime functions return to the application before the device has completed the requested task

Image utilities <image_utilities.mdi>

Performance Guidelines How to get the most out of the device

Instruction performance

- □ To process an instruction for a warp of threads, a multiprocessor must:
 - Read the instruction operands for each thread of the warp,
 - Execute the instruction,
 - Write the result for each thread of the warp.

Instruction Performance

- ☐ The effective instruction throughput depends on the nominal instruction throughput as well as the memory latency and bandwidth. It is maximized by:
 - Minimizing the use of instructions with low throughput
 - Maximizing the use of the available memory bandwidth for each category of memory
 - Allowing the thread scheduler to overlap memory transactions with mathematical computations as much as possible, which requires that:
 - The program executed by the threads is of high arithmetic intensity, i.e a high number of arithmetic operations per memory operation;

 There are many threads that can be run concurrently

Instruction Throughput

- ☐ To issue one instruction for a warp, a multiprocessor takes 4 clock cycles for
 - floating-point add
 - floating-point multiply
 - floating-point multiply-add
 - integer add
 - bitwise operations
 - compare
 - min, max,
 - type conversion instruction;

Instruction Throughput

- ☐ To issue one instruction for a warp, a multiprocessor takes16 clock cycles for
 - reciprocal
 - reciprocal square root
 - 32-bit integer multiplication;

Instruction Throughput

- □ Integer division and modulo operation are particularly costly and should be avoided if possible
 - Nvidia doesn't tell how costly though...
- □ Other functions are implemented as combinations of several instructions
 - Floating-point square root is implemented as a reciprocal square root followed by a reciprocal, so it takes 32 clock cycles for a warp.
 - Floating-point division takes 36 clock cycles.

Control Flow Instructions

- ☐ Any flow control instruction
 - (if, switch, do, for, while)
- ☐ can significantly impact the effective instruction throughput by causing threads of the same warp to diverge
 - Serialization vs. parallelization

Memory Instructions

- Memory instructions include any instruction that reads from or writes to shared or global memory.
- A multiprocessor takes 4 clock cycles to issue one memory instruction for a warp.
- □ When accessing global memory, there are, in addition, 400 to 600 clock cycles of memory latency!!!

Memory Bandwidth – Global Memory

- □ Uncached
 - One can fetch 32-bit, 64-bit, or 128-bit words into registers in a single instruction
- □ <u>Use coalescence whenever possible</u>

Memory Bandwidth – Constant Memory

- □ Cached
 - Only at cache misses does it cost a read from device memory

Memory Bandwidth – Texture Memory

- □ Cached
 - Only at cache misses does it cost a read from device memory
 - The texture cache is optimized for 2D spatial locality
 - BUT there is a cost associated to copying into the a CUDA array

Memory Bandwidth – Shared Memory

- □ On-chip
 - As fast as registers
 - Avoid bank conflicts
 - shared memory is divided into equally-sized memory modules, called banks, which can be accessed simultaneously
 - any memory read or write request made of n addresses that fall in n distinct memory banks can be serviced simultaneously

Shared Memory — no bank conflicts | these |

2-way and 8-way bank conflicts Theat 1 Theat 2 Theat 3 Theat 3 Theat 4 Theat 5 Theat 5 Theat 5 Theat 5 Theat 6 Theat 7 Theat 8 Theat 7 Theat 8 Theat 9 Theat

Memory Bandwidth − Registers □ On-chip ■ Generally, accessing a register is zero extra clock cycles per instruction

Number of threads per block

- ☐ There should be at least as many blocks as there are multiprocessors in the device.
- Running only one block per multiprocessor can force the multiprocessor to idle during thread synchronization and device memory reads
- ☐ The shared memory available to a block decreases with the number of <u>active</u> blocks
- ☐ The number of threads per block should be chosen as a multiple of the warp size!!!

Number of threads per block

- ☐ Allocating more threads per block is better for efficient time slicing, but the more threads per block, the fewer registers are available per thread.
- □ CUDA occupancy calculator

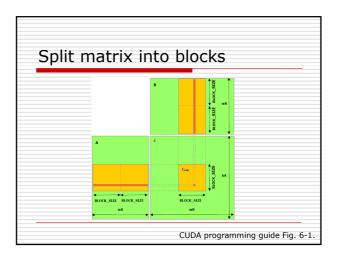
Current generation hardware

<G8x_specs.mdi>

Matrix multiplication

An examle from the CUDA programming guide

Ax=b □ Computing the product C of two matrices A and B of dimensions (wA, hA) and (wB, wA) ■ Each thread block is responsible for computing one square sub-matrix C_{sub} of C; ■ Each thread within the block is responsible for computing one element of C_{sub}.



Matrix multiplication code

<mat_mult.mdi>

Browse-through SDK

As time permits...