

GPU Teaching Kit

Accelerated Computing



Lecture 1.1 – Course Introduction

Course Introduction and Overview

Course Goals

Learn how to program heterogeneous (异构) parallel computing systems and achieve

- High performance and energy-efficiency
- Functionality(功能性) and maintainability(可维护性)
- Scalability(可扩展性) across future generations
- Portability(可移植性) across vendor devices

Technical subjects

- Parallel programming API, tools and techniques
- Principles and patterns of parallel algorithms
- Processor architecture features and constraints

Course Content

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Module 1 Course Introduction	 Course Introduction and Overview Introduction to Heterogeneous Parallel Computing Portability and Scalability in Heterogeneous Parallel Computing
Module 2 Introduction to CUDA C	 CUDA C vs. CUDA Libs vs. OpenACC Memory Allocation and Data Movement API Functions Data Parallelism and Threads Introduction to CUDA Toolkit
Module 3 CUDA Parallelism Model	 Kernel-Based SPMD Parallel Programming Multidimensional Kernel Configuration Color-to-Greyscale Image Processing Example Blur Image Processing Example
Module 4 Memory Model and Locality	 CUDA Memories Tiled Matrix Multiplication Tiled Matrix Multiplication Kernel Handling Boundary Conditions in Tiling Tiled Kernel for Arbitrary Matrix Dimensions
Module 5 Kernel-based Parallel Programming	 Histogram (Sort) Example Basic Matrix-Matrix Multiplication Example Thread Scheduling Control Divergence

Course Content

Module 6 Performance Considerations: Memory	DRAM Bandwidth Memory Coalescing in CUDA
Module 7 Atomic Operations	Atomic Operations
Module 8 Parallel Computation Patterns (Part 1)	ConvolutionTiled Convolution2D Tiled Convolution Kernel
Module 9 Parallel Computation Patterns (Part 2)	Tiled Convolution AnalysisData Reuse in Tiled Convolution
Module 10 Performance Considerations: Parallel Computation Patterns	ReductionBasic Reduction KernelImproved Reduction Kernel
Module 11 Parallel Computation Patterns (Part 3)	 Scan (Parallel Prefix Sum) Work-Inefficient Parallel Scan Kernel Work-Efficient Parallel Scan Kernel More on Parallel Scan

Course Content

Module 12 Performance Considerations: Scan Applications	 Scan Applications: Per-thread Output Variable Allocation Scan Applications: Radix Sort Performance Considerations (Histogram (Atomics) Example) Performance Considerations (Histogram (Scan) Example)
Module 13 Advanced CUDA Memory Model	Advanced CUDA Memory ModelConstant MemoryTexture Memory
Module 14 Sparse Matrix Computation	 Parallel SpMV Using CSR Padding and Transposition Using a Hybrid Approach to Regulate Padding Sorting and Partitioning for Regularization
Module 15 Merge Sort	 Sequential Merge Algorithm Parallelization Approach Co-Rank Function Implementation Basic Parallel Merge Kernel Tiled Merge Kernel Circular-Buffer Merge Kernel
Module 16 Graph Search	Breadth-First SearchSequential BFS FunctionParallel BFS FunctionOptimizations
Module 17 CUDA Python Using Numba	CUDA Python using Numba

Teaching Arrangement

- 32 Lecture Hours
- 16 Laboratory Hours (Important)

Grade Components

- In-class performance (70%)
 - Final written exam (70%)
 - Regular performance (30%)
 - Attendance (20%)
 - In-class quizzes (80%)
- Lab performance (30%)
 - Nvidia Certifications (50%)
 - 2 NVIDIA certifications
 - 2 corresponding quizzes.
 - 4 assignments (50%)

Students who miss more than three classes will be disqualified from participating in the final exam!!!

Textbooks, Reference Books, and Resources

Textbooks

- 大卫·B. 柯克(David B. Kirk),胡文美(Wen-mei W. Hwu)著
- 大规模并行处理器程序设计(英文版·原书第3版)
- 机械工业出版社, 2020年

Reference Books

- 高性能计算系列丛书·CUDA并行程序设计: GPU编程指南 [CUDA Programming: A Developer's Guide to Parallel Computing with GPUs], 机械工业出版社, 2014年
- Professional Cuda C Programming, Wiley, 2014年

Resources

https://docs.nvidia.com/cuda/doc/index.html



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Lecture 1.2 – Course Introduction

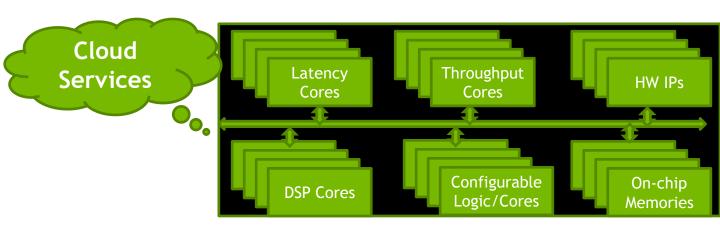
Introduction to Heterogeneous Parallel Computing

Objectives

- To learn the major differences between latency devices (CPU cores) and throughput devices (GPU cores)
- To understand why winning applications increasingly use both types of devices

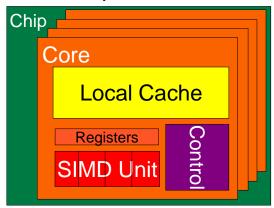
Heterogeneous Parallel Computing

 Use the best match for the job (heterogeneity in mobile SOC(System on a Chip))

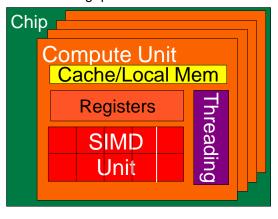


CPU and GPU are designed very differently

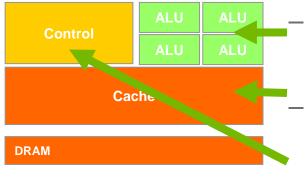
CPULatency Oriented Cores



GPU
Throughput Oriented Cores



CPUs: Latency Oriented Design



Powerful ALU

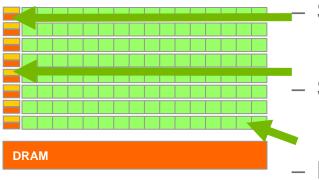
 Reduced operation latency

Large caches

 Convert long latency memory accesses to short latency cache accesses

- Sophisticated control
 - Branch prediction for reduced branch latency
 - Data forwarding for reduced data latency

GPUs: Throughput Oriented Design



Small caches

- To boost memory throughput
- Simple control
 - No branch prediction
 - No data forwarding
- Energy efficient ALUs
 - Many, long latency but heavily pipelined for high throughput
- Require massive number of threads to tolerate latencies
 - Threading logic
 - Thread state

Winning Applications Use Both CPU and GPU

- CPUs for sequential parts where latency matters
 - CPUs can be 10X+ faster than GPUs for sequential code

- GPUs for parallel parts where throughput wins
 - GPUs can be 10X+ faster than CPUs for parallel code

Heterogeneous Parallel Computing in Many Disciplines

Financial Analysis Scientific Simulation Engineering Simulation

Data Intensive Analytics

Medical Imaging

Digital Audio Processing

Digital Video Processing

Computer Vision

Biomedical Informatics

Electronic Design Automation

Statistical Modeling

Numerical Methods

Ray Tracing Rendering Interactive Physics



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Lecture 1.3 – Course Introduction

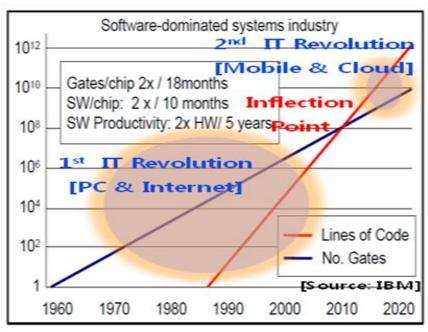
Portability and Scalability in Heterogeneous Parallel Computing

Objectives

 To understand the importance and nature of scalability and portability in parallel programming

Software Dominates System Cost

- SW lines per chip increases at 2x/10 months
- HW gates per chip increases at 2x/18 months
- Future systems must <u>minimize software</u> redevelopment





Scalability



- Scalability
 - The same application runs efficiently on new generations of cores



Scalability

- The same application runs efficiently on new generations of cores
- The same application runs efficiently on more of the same cores

More on Scalability

- Performance growth with HW generations
 - Increasing number of compute units (cores)
 - Increasing number of threads
 - Increasing vector length
 - Increasing pipeline depth
 - Increasing DRAM burst size
 - Increasing number of DRAM channels
 - Increasing data movement latency



- Scalability
- Portability
 - The same application runs efficiently on different types of cores







- Scalability
- Portability
 - The same application runs efficiently on different types of cores
 - The same application runs efficiently on systems with different organizations and interfaces

More on Portability

- Portability across many different HW types
 - Across ISAs (Instruction Set Architectures) X86 vs. ARM, etc.
 - Latency oriented CPUs vs. throughput oriented GPUs
 - Across parallelism models SIMD vs. threading
 - Across memory models Shared memory vs. distributed memory



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