VSCSE Summer School

Proven Algorithmic Techniques for Many-core Processors

Lecture 9: Directions for Future Studies and Closing Remarks

©Wen-mei W. Hwu and David Kirk/NVIDIA Urbana, Illinois, August 2-5, 2010

Course Objective (Review)

 To master the most commonly used algorithm techniques and computational thinking skills needed for many-core programming

– Especially the simple ones!

- Specifically, to understand
 - Many-core hardware limitations and constraints
 - Desirable and undesirable computation patterns
 - Commonly used algorithm techniques to convert undesirable computation patterns into desirable ones

Agenda (Monday)

- 10:00 11:30 Lecture 1 Introduction to computational thinking for many-core computing
- 12:30 2:00 Lecture 2 Scatter-to-Gather transformation for scalability
- 3:00 4:30 Lecture 3 Loop blocking and register tiling for locality
- Lab 1 2D Blocking and Register tiling

©Wen-mei W. Hwu and David Kirk/NVIDIA Urbana, Illinois, August 2-5, 2010

Agenda Review (Tuesday)

- Lecture 4 Cut-off and Binning for regular data sets
- Lecture 5 Data Layout for Grid Applications
- Keynote 1 Michael Garland (NVIDIA)
 Algorithm Design for GPU Computing
- Lab 1 2D Blocking and Register tiling
- Lab 2 Data Layout Transformation for LBM Methods and Binning for Regular Data Sets

Agenda Review (Wednesday)

- Lecture 6 Dealing with non-uniform data distribution
- Lecture 7 Dealing with dynamic data sets
 - With guest lecture by Jonathan Cohen (NVIDIA) on PDE Solvers and OpenCurrent
- Keynote 2 David Kirk (NVIDIA)
 - Fermi and Future of GPU Computing Technology
- Lab 2 Data Layout Transformation for LBM Methods and Binning for Regular Data Sets

• Lab 3 – Binning for non-uniform data distribution

Agenda Review (Thursday)

- Keynote 3 Lorena Barba (Boston University)
 - Multiplying speedups: fast algorithms on GPUs. A case study on biomolecular electrostatics.
- Lecture 8 Transformations of key computation patterns
 - With guest lecture from Jeremy Meredith (Oak Ridge National Lab) on Accelerating HPC Applications with GPUs – Two Case Studies"
- Hands-on Lab 3 Data binning for non-uniform data distribution
- Hands-on Lab wrap-up discussions

Illinois, August 2-5, 2010

Phillip Colella's "Seven dwarfs"

High-end simulation in the physical sciences = 7 numerical methods:

- . Structured Grids (including locally structured grids, e.g. Adaptive Mesh Refinement)
- . Unstructured Grids
- 3. Fast Fourier Transform
 - L. Dense Linear Algebra
- Sparse Linear Algebra
- 6. Particles
 - 7. Monte Carlo

- If add 4 for embedded, covers all 41 EEMBC benchmarks
 - 8. Search/Sort
 - 9. Filter
 - 10. Combinational logic
 - 11. Finite State Machine
- Note: Data sizes (8 bit to 32 bit) and types (integer, character) differ, but algorithms the same

Slide from "Defining Software Requirements for Scientific Computing", Phillip Colella, 2004

Well-defined targets from algorithmic, software, and architecture standpoint

Seven Techniques in Manycore Programming

- Scatter to Gather transformation
- Granularity coarsening and register tiling
- Tiling for dense matrix data locality
- Data layout and traversal ordering
- Binning and cutoff
- Bin sorting and partitioning for non-uniform data
- Hierarchical queues and kernels for dynamic data

Benefit from other people's experience

- GPU Computing Gems Vol 1
 - Coming December 2010
 - 50 gems in 10 applications areas
 - Scientific simulation, life sciences, statistical modeling, emerging data-intensive applications, electronic design automation, computer vision, ray tracing and rendering, video and imaging processing, signal and audio processing, medical imaging
- GPU Computing Gems Vol 2
 - Coming in April 2011
 - 50+ gems on more application domains, tools,

©Wen-mei W. HELANDBARCHIK Urbana, Illinois, August 2-5, 2010

Phillip Colella's "Seven dwarfs"

High-end simulation in the physical sciences = 7 numerical methods:

- . Structured Grids (including locally structured grids, e.g. Adaptive Mesh Refinement)
- . Unstructured Grids
- 3. Fast Fourier Transform
 - Lense Linear Algebra
- Sparse Linear Algebra
- 6. Particles
 - '. Monte Carlo

- If add 4 for embedded, covers all 41 EEMBC benchmarks
 - 8. Search/Sort
 - 9. Filter
 - 10. Combinational logic
 - 11. Finite State Machine
- Note: Data sizes (8 bit to 32 bit) and types (integer, character) differ, but algorithms the same

Slide from "Defining Software Requirements for Scientific Computing", Phillip Colella, 2004

Well-defined targets from algorithmic, software, and architecture standpoint Courtesy: David Patterson

How about your computation types?

- Are your applications very different from all the seven dwarfs?
- Can you think of any other algorithmic techniques that we should cover?

You can do it.

- Computational thinking is not as hard as you may think it is.
 - The purpose of the course is to make them accessible to domain scientists and engineers.
 - With practice, you can make it look easy!



ANY MORE QUESTIONS?

©Wen-mei W. Hwu and David Kirk/NVIDIA Urbana, Illinois, August 2-5, 2010