Refactoring the MPS/University of Chicago Radiative MHD (MURaM) Model for GPU/CPU Performance Portability Using OpenACC Directives

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2

Outline

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- Description of MURaM
- Challenges porting MURaM to large scale systems
- Results



DESCRIPTION OF MURAM

MURaM (Max Planck University of Chicago Radiative MHD)

- Primary solar model for simulations of the upper convection zone, photosphere and corona
- Jointly developed by HAO, the Max Planck Institute for Solar System Research (MPS) and the Lockheed Martin Solar and Astrophysics Laboratory (LMSAL)



MURaM simulation of solar granulation

MURaM Future Goals

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https://nso.edu/telescopes/dki-solar-telescope/

- Short-term: Solar models capable of running at higher resolutions
 - Higher resolution simulations requires 10-100x more computing time
 - Difficult to accomplish as current CPU runs are hitting a scaling limit
- Long-term: achieve real-time simulation to predict solar events
 - Current models are 10-100x slower than real-time



MURaM simulation: Synthetic coronal emission (SDO/AIA) during M2 Flare and resulting CME

GPU Programming with OpenACC

- Our team has experience working on several other large OpenACC projects
- Some MURaM users will prefer to run the CPU version
- Parallelize the code incrementally, without large rewrites at a time



https://www.gfdl.noaa.gov/ocean-model/

7



https://mpas-dev.github.io/



CHALLENGES PORTING MURAM TO LARGE SCALE SYSTEMS



- RTS accounts for the majority of the runtime
- Within RTS, the *Integrate()* function is the most time consuming
- Integrate() must be run at least once for each of the 24 rays
- Integrate() introduces a data dependency that must be considered carefully when parallelizing



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Optimizing RTS: Problems with *Integrate()*

- The outermost loop over the grid-cells must be sequential
 - This causes many (hundreds) of GPU kernels to be launched for each call to Integrate
 - These GPU kernels only parallelize 2 dimensions of our grid-space, resulting in small, inefficient kernels (with low GPU occupancy)



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Code Performance GPU Occupancy

GPU Occupancy By Kernel

Achieved Theoretical





RESULTS

Experimental Setup: Cobra

Max Planck Computing and Data Facility (MPCDF) in Garching, Germany

- 3,424 compute nodes
 - two Intel Xeon Gold 6148 Skylake (SKL) processors (20 cores at 2.4GHz)
 - 100 Gb/s Omni Path interconnect
- 64 GPU nodes
 - two NVIDIA Tesla V100-PCIe with 32GB HBM2
- CPU compilers/libs: Intel 19.1.3, Intel MPI 2019.9, MKL 2020.2 and FFTW-MPI 3.3.8
- GPU compilers/lib: NVHPC 20.9, CUDA 11, OpenMPI 4.0.5 with UCX 1.8.0 and FFTW-MPI 3.3.8 with multithreading

Strong Scaling: 288³ dataset on 8 GPUs



Strong Scaling: 288x576x576 dataset on up to 96 GPUs



Weak Scaling: 288³ datapoints per GPU



Summary

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- 1.73x speedup using a single NVIDIA V100 GPU over a fully subscribed 40-core Intel Skylake CPU node
- Maintained single source code across multicore CPUs and GPUs
- Multi-band radiation transport will advance understanding of the solar chromosphere

Future Work

- Implementation of ray-merge version of the code
- Explore GPU-enabled FFT libraries (heFFTe)
- Implement ensemble capabilities to get closer to real-time prediction

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