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September 25th & 26th, 2019, National Center for Atmospheric Research (NCAR) Mesa Lab in Boulder, Colorado



- MURaM Introduction
- OpenACC Introduction
- Development Tools
- Development Roadblocks
- Results





MURaM (Max Planck University of Chicago Radiative MHD)

- The primary solar model used for simulations of the upper convection zone, photosphere and corona
- Jointly developed and used by HAO, the Max Planck Institute for Solar System Research (MPS) and the Lockheed Martin Solar and Astrophysics Laboratory (LMSAL)
- The Daniel K. Inouye Solar Telescope (DKIST), a ~\$300M NSF investment, is expected to advance the resolution of ground based observational solar physics by an order of magnitude
- Requires at least 10-100x increase in computing power compared to current baseline



MURaM simulation of solar granulation



Physics of the MURaM Code

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• Science target

- Realistic simulations of the coupled solar atmosphere
- Detailed comparison with available observations through forward modeling of synthetic observables
- Implemented Physics
 - Single fluid MHD
 - 3D radiative transfer, multi-band + scattering
 - Partial ionization equation of state
 - Heat conduction
 - Optically thin radiative loss
 - Ambipolar diffusion
- Under development
 - Non-equilibrium ionization of hydrogen



Comprehensive model of entire life cycle of a solar prominence (Cheung et al 2018)







3 Ways to program CPU-GPU Architectures

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ANSYS FLUENT COSMO **GAUSSIAN 16** VASP Using OpenACC allowed us to continue development of our fundamental We've effectively used For VASP, OpenACC is the way OpenACC made it practical to develop for GPU-based hardware OpenACC for heterogeneous algorithms and software capabilities simultaneously with the GPU-related computing in ANSYS Fluent Performance is similar and in some cases better than CUDA C, and while retaining a single source for work. In the end, we could use the almost all the COSMO physics with impressive performance. OpenACC dramatically decreases same code base for SMP, cluster/ network and GPU parallelism. PGI's We're now applying this work code GPU development and maintenance efforts. We're excited to collaborate to more of our models and compilers were essential to the success of our efforts. new platforms. with NVIDIA and PGI as an early Image grantery AND/22 adopter of CUDA Unified Memory MPAS-A E3SM NUMECA FINE/Open SYNOPSYS 0 Our team has been evaluating OpenACC as a pathway to performance pathality for the Model for Prediction IMPAS) atmospheric model, Using this approach on the MMS dynamical core, we have active-rol performance on a single Problem of expression on a single Using OpenACC, we've GPU- Porting our unstructured C++ CFD solver FINE/Open to GPUs early access to Summit hardware and accelerated the Synopsys TCAD access to PGI compiler experts. Both using OpenACC would have Sentaurus Device EMW simulator of these were critical to our success. PGI's OpenACC support remains the been impossible two or three to speed up optical simulations of years ago, but OpenACC has image sensors. GPUs are key to much more intrusive programming improving simulation throughput developed enough that we're model approaches. in the design of advanced image " sensors. Chevenne supercomputer. Amage country: Oak Akdyo Klalionai Laborativy results. " GTC VMD GAMERA -**OpenACC** Using OpenACC our scientists Due to Amdahi's law, we need to port more parts of our code to the GPU if we're With OpenACC and a compute were able to achieve the node based on NVIDIA's Tesla going to speed it up. But the sheer number of routines poses a challenge. acceleration needed for P100 GPU, we achieved more integrated fusion simulation with OpenACC directives give us a low-cost approach to getting at least some spee than a 14X speed up over a K a minimum investment of time More Science, Less Programming Computer node running our approach to getuing at teast some speed-up out of these second-tier routines. In many cases it's completely sufficient because with the current algorithms, GPU performance is bandwidth-bound. and effort in learning to program earthquake disaster simulation code -PWscf (Quantum MAS SANJEEVINI **IBM-CFD** ESPRESSO) Adding OpenACC into MAS has given us CUDA Fortran gives us the full performance potential of the CUDA. the ability to migrate medium-sized simulations from a multi-node CPU maintenance and speedup of existing programming model and NVIDIA GPUs. White leveraging the potential of explicit data movement, ISCUF KERNELS codes is a tedious task. OpenACC 6 cluster to a single multi-GPU server. The implementation yielded a portable rsed boundary incompressible provides a great platform for ED, we have obtained order of magnitude The implementation yielded a portable single-source code for both CPU and GPU runs. Future work will add OpenACC to the remaining model features, enabling GPU accelerated realistic solar storm modeling. computational scientists to accomplish directives give us productivity and source code maintainability. It's the best both tasks without involving a lot of efforts or manpower in speeding up the ur legacy codes to GPU os involving search algorithm of both worlds entire computational task. Ì "

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Development Cycle

Name	Routine Summary:	Broadwell (v4) core: (sec)
TVD Diffusion	Update diffusion scheme - using TVD slope + flux limiting.	7.36812
Magnetohydrodyna mics	Calculate right hand side of MHD equations.	6.26662
Radiation Transport	Calculate radiation field and determine heating term (Qtot) required in MHD.	5.55416
Equation of State	Calculate primitive variables from conservative variables. Interpolate the equation of state tables.	2.26398
Time Integration	Performs one time integration.	1.47858
DivB Cleaner	Clean any errors due to non-zero div(B).	0.279718
Boundary Conditions	Update vertical boundary conditions.	0.0855162
Grid Exchange	Grid exchanges (only those in Solver)	0.0667914
Alfven Speed Limiter	Limit Maximum Alfven Velocity	0.0394724
Synchronize timestep	Pick minimum of the radiation, MHD and diffusive timesteps.	4.48E-05



NVPROF: NVIDIA GPU Profiler

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- Profilers give detailed information/feedback about code execution
- For this work, we used NVIDIA's GPU enabled profiler too: NVPROF

https://devblogs.nvidia.com/cuda-pro-tip-nvprof-your-handy-universal-gpu-profiler/

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CUPTI (CUDA Profiling Tools Interface)

ELAWARE.



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give additional

profiler feedback

CUDA Occupancy Calculator

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https://devblogs.nvidia.com/nvidia-turing-architecture-in-depth/

PCAST (PGI Compiler Assisted Software Testing)

- Automated testing features for PGI compiler
- Able to do autocompare (sometimes) to make kernel debugging much easier
- In our case, we used API calls to do some checking manually, but allowed for easy code testing after

\$ pgcc -ta=tesla:autocompare -o a.out example.c

```
$ PGI_COMPARE=summary,compare,abs=1 ./a.out
PCAST a1 comparison-label:0 Float
    idx: 0 FAIL ABS act: 8.40187728e-01 exp: 1.00000000e+00 tol: 1.00000001e-01
    idx: 1 FAIL ABS act: 3.94382924e-01 exp: 1.00000000e+00 tol: 1.00000001e-01
    idx: 2 FAIL ABS act: 7.83099234e-01 exp: 1.00000000e+00 tol: 1.00000001e-01
    idx: 3 FAIL ABS act: 7.98440039e-01 exp: 1.00000000e+00 tol: 1.00000001e-01
```





CUDA Occupancy Report

240x160x160 Dataset

Kernel Name	Theoretical Occupancy	Achieved Occupancy	
MHD	25%	24.9%	
TVD	31%	31.2%	
CONS	25%	24.9%	
Source_Tcheck	25%	24.9%	
	Radiation Transport		
Driver	100%	10.2%	
Interpol	56%	59.9%	
Flux	100%	79%	





RTS Data Dependency Along Rays



Figure 4.1: The intensity at gridpoint F is obtained by solving the transfer equation along the short characteristic \overline{EF} . The intensity at the upwind point E is interpolated from the (already known) intensity values at the surrounding gridpoints, A to D.



Figure 4.2: The walking order of the Short Characteristics method in a 2D grid for a ray direction pointing into the upper right quadrant. Black circles represent gridpoints on the upwind boundaries, where the intensity values are assumed to be known.

- Data dependency is along a plane for each octant, angle combo.
- Depends on resolution ratio, not known until run-time.
- Number of rays per plane can vary.

Vögler, Alexander, et al. "Simulations of magneto-convection in the solar photosphere-Equations, methods, and results of the MURaM code." Astronomy & Astrophysics 429.1 (2005): 335-351.

Solving RTS Data Dependency

- We can deconstruct the 3D grid into a series of 2D slices
- The direction of the slices is dependent on the X,Y,Z direction of the ray
- Parallelize within the slice, but run the slices themselves serially in predetermined order







Profiler driven optimizations

*driver.nvvp	\$ *gpu_i_n2.nvvp %						
9016.995 ms	9016.9975 ms	9017 ms	9017.0 <mark>4.73 µs</mark>	9017.005 ms	9017.0075 ms	9017.01 ms	
	acc_compute	construct@rt.cc:124	42	acc_comp	ute_construct@rt.cc:124	2	
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Experimental Setup

- NCAR Casper system
 - 28 Supermicro nodes featuring Intel Skylake processors
 - 36 cores/node
 - 384GB memory/node
 - 4/8 NVIDIA V100 GPUs/node
 - PGI 19.4, CUDA 10

Results: CPU vs GPU

Routine	GPU time	CPU time	Speedup
RTS	0.361	0.230	0.637
MHD	0.108	0.160	1.48x
TVD	0.056	0.066	1.17x
EOS	0.031	0.071	2.29x
BND	0.004	0.007	1.75x
INT	0.050	0.071	1.42x
DST	0.163	0.031	0.19x
DIVB	0.076	0.029	0.38x
TOTAL	0.853	0.701	0.82x

- Single NVIDIA V100 GPU
- Dual Socket Intel Skylake CPU (36 core)
- Measuring time taken for average timestep with no file I/O
- 192x128x128 sized dataset

MURaM Scaling Seconds Number of V100 GPUs or Dual Socket Broadwell Nodes

Strong Scaling

ELAWARE.



NCAR

Weak Scaling

ELAWARE.



NCAR

Summary

- MURaM
 - Single fluid MHD
 - 3D radiative transfer, multi-band + scattering
 - Partial ionization equation of state
 - Heat conduction
 - Optically thin radiative loss
 - Ambipolar diffusion
- Use OpenACC to port to GPU with directives
 - Incremental changes
 - Maintain single C++ source code
- Tools: NVPROF, CUPTI, CUDA Occupancy Calculator, PGI PCAST