Targeting GPUs using OpenMP Directives on Summit with GenASiS: A Simple and Effective Fortran Experience

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The Application

*Gen*eral *A*strophysics *Si*mulation *S*ystem (GenASiS)

- Designed for parallel, large scale simulations
	- weak-scale to ~100 thousands MPI processes
- Written entirely in modern Fortran (2003, 2008)
- Modular, object-oriented design, and extensible
- Multi-physics solvers:
	- (Magneto)-hydrodynamics (HLL, HLLC solvers)
	- Explicit 2nd order time-integration
	- Self-gravity, polytropic & nuclear EoS
	- Grey and spectral neutrino transport
- CPU only code with OpenMP for threading (prior to this work)

The Application

- Studied the role fluid instabilities --convection and Standing Accretion Shock Instability (SASI) --- in supernova dynamics
- Discovered exponential magnetic field amplification by SASI in progenitor star \rightarrow origin of neutron star magnetic fields
- Refactored to three major subdivisions: *Basics*, *Mathematics*, *Physics* → allowing unit testing, ad-hoc/standalone tests, mini-apps

Paths to Targeting GPU

- CUDA
	- requires rewrite of all computational kernels
	- loss of Fortran semantics (multi-d arrays, pointer/array remapping)
	- requires interfacing with the rest of the (Fortran) code
- CUDA Fortran
	- non standard extension to Fortran (XL, PGI)
	- cannot easily fall back to standard Fortran
- Directives (OpenMP)
	- retain Fortran semantics
	- OpenMP 4.5 has excellent support from IBM XL (Summit), CCE (Titan)
		- with excellent support for modern Fortran

Lower-Level GenASiS Functionality

- Fortran wrappers to OpenMP APIs
	- call AllocateDevice(Value, D_Value)
		- \rightarrow omp_target_alloc()

call AssociateHost(D_Value, Value) \rightarrow omp target associate ptr()

call UpdateDevice(Value, D_Value), call UpdateHost(Value, D_Value)

 \rightarrow omp_target_memcpy()

- Affirmative control of data movement
- Persistent memory allocation on the device

Value : Fortran array D Value : type(c_ptr), GPU address

Higher-level GenASiS Functionality

- StorageForm :
	- a class for data and metadata; the 'heart' of data storage facility in GenASiS
	- metadata includes units, variable names (for I/O, visualization)
	- used to group together a set of related physical variables (e.g. Fluid)
	- render more generic and simplified code for I/O, ghost exchange, prolongation & restriction (AMR mesh)
- Data: StorageForm % Value (nCells, nVariables)
- Methods:
	- call StorageForm % Initialize () **← allocate data on host**
	- call StorageForm % AllocateDevice () **← allocate data on GPU**
	- call StorageForm % Update{Device,Host} () **← transfer data**

Sidebar: OpenMP Memory for Offload

• OpenMP maps host (CPU) variables to device (GPU) (explicitly or implicitly)

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- default copy to-from
- *• Presence check*: if fail, new variable is created on device
	- *–* sometime requires explicit association to avoid unintentional data movement

Offloading Computational Kernel

```
subroutine AddKernel (A, B, D_A, D_B, D_C, C)
2
3
     real (KDR), dimension (: ), intent (in) :: A, B
           (c<sub>-</sub>ptr), intent (m): D_A, D_B, D_C\overline{4}true\overline{5}real (KDR), dimension (: ), intent (out) :: C\,6\sqrt{ }integer ( KDI ) :: icall F % Initialize &
8
9
     call AssociateHost (D.A. A
                                            Tells OpenMP data location on GPU 1ables])
10
     call AssociateHost (D_B, B
                                            →avoid (implicit) allocation & transfer eDevice ()
11
     call AssociateHost (D_C, C
                                                             call F % UpdateDevice ( )
12
13
     \ell : SOMP target teams distribute parallel do schedule (static, 1)
                                                             call AddKernel &
14
     do i = 1, size (C) ( F % Value ( :, 1 ),
15
     C(i) = A(i) + B(i)16
     end do
                                                                   F % Value ( :, 2 ), &
17
     SOMP end target teams distribute parallel do
                                                                   F % D_Value ( 1 ), 
18
                                                                   F % D_Value ( 2 ), &
19
     call DisassociateHost
20
     call DisassociateHost
                             ( B
                                                                   F % D_Value ( 3 ), 
21
     call DisassociateHost
                              \mathbf{A} F % Value ( :, 3 ) ) 
22
23 end subroutine AddKernel
```
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Example of Kernel with Pointer Remapping

real (KDR), dimension (:, :, :), pointer :: V, dV

V (−1:nX+2, −1:nY+2, −1:nZ+2) => F % Value (: , iV) dV (−1:nX+2 , −1:nY+2 , −1:nZ+2) => dF % Value (: , iV)

call ComputeDifferences_X (V, F % D_Value (iV), ...)

Example of Kernel with Pointer Remapping

```
1 subroutine ComputeDifference_X (V, D_V, D_dV, dV)
 \overline{2}3
     real (KDR), dimension (-1:, -1:, -1:), intent (in) :: V
     type (c_ptr), intent (in) :: D<sub>V</sub>, D<sub>dV</sub>
 \overline{4}real (KDR), dimension (-1; -1; -1; ), intent (out) :: dV
 5
 \,6\,7integer ( KDI ) :: i, j, k\,8\,9
     call AssociateHost (D_V, V)
     call AssociateHost (D_dV, dV)
10
11
     \ell /somp target teams distribute parallel do collapse (3) schedule (static, 1)
12
     do k = 1, nZ
13
       \bf{do} j = 1, nY14
15
         do i = 0, nX + 2dV (i, j, k) = V (i, j, k) - V (i - 1, j, k)
16
17
         end do
18
       end do
     end do
19
20
     SOMP end target teams distribute parallel do
21
22
     call DisassociateHost (dV)
23
     call DisassociateHost (V)
24
25 end subroutine ComputeDifferences_X
```
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Porting a Fluid Dynamics Application: RiemannProblem

Initial (left) and final (right) density of 1D and 3D RiemannProblem

Fluid Evolution on CPU

Fluid Evolution on GPU

```
1: HOST: Call: Initialize ()
2: Call: GhostExchange()3: HOST: Set: Time = StartTime
4: while Time < FinishTime do
     HOST: Call: ComputeTimeStep () \rightarrow TimeStep
5:
     TRANSFER: Call: FluidCurrent % UpdateDevice ()
6:DEVICE: Set: FluidOld = FluidCurrent
7:8:
     DEVICE: Call: ComputeDifferences ()
9:
     DEVICE: Call: ComputeReconstruction ()
10:11:DEVICE: Call: ComputeFluxes ()
     DEVICE: Call: ComputeUpdate (TimeStep) \rightarrow FluidUpdate
12:DEVICE: Set: FluidCurrent = FluidOld + FluidUpdate
13:14:TRANSFER: Call: FluidCurrent % UpdateHost ()
15:HOST: Call: GhostExchange()16:
     TRANSFER: Call: FluidCurrent % UpdateDevice ()
17:18:
     DEVICE: Call: ComputeDifferences ()
19:
     DEVICE: Call: ComputeReconstruction ()
20:DEVICE: Call: ComputeFluxes ()
21:22:DEVICE: Call: ComputeUpdate (TimeStep)
23:
     DEVICE: Set: FluidCurrent = 0.5 * (FluidOld + FluidCurrent + FluidUpdate)
24:25:TRANSFER: Call: FluidCurrent % UpdateHost ()
26:Call: GhostExchange()27: end while
```


Performance Results

Summit Node (2) IBM Power9 + (6) NVIDIA Volta V100

Performance Results: Weak-Scaling 3D RiemannProblem

Performance Results: Kernel Timings

■ 2 Threads ■ 4 Threads ■ 7 Threads ■ GPU

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Performance Results: Kernel Speedups

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Performance Results: Timing Distribution

Beyond OpenMP: Using Pinned Memory

To optimize data transfers:

- pinned memory: page-locked host memory allocated using cudaMallocHost() or cudaHostAlloc()
- created another Fortran wrapper in GenASiS, used in StorageForm initialization method as an option StorageForm % Initialize (…, PinnedOption = .true.)
- No mechanism to do this with OpenMP 4.5 (but perhaps in 5.x)

Performance Results: Using Pinned Memory

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Number of MPI Ranks

Implications (Then and Now)

- c. 2010: 1024³ cells with 64000 processors (Jaguar), \sim 3s per timestep Now: 64 GPUs (11 nodes) on Summit, ~1.2s per timestep
- Enable us to do higher-fidelity simulations, ensemble studies for trends in observables
	- plan to perform ~200 2D grey transport supernova simulations, tens of 3D grey transport, and a handful of 3D spectral transport simulations
- First step towards full Boltzmann radiation transport (6-D problem + time) with exascale computing

Remaining Issues and Future Work

- A single code-base with OpenMP for multi-threading and offload
	- Fall back to multi-threading with target if-clause is problematic
	- team distribute directive introduce deleterious effects for multi-threading
- Kernel-launch parallelism and CUDA streams
	- no mechanism within OpenMP to affect and select stream
- Better compilers support for OpenMP 4.5 5.x
- Using CUDA-aware MPI for GPU Direct
	- benefit (vs. manual staging on host) depends on message sizes

Conclusion

- Using OpenMP allows for a simple and effective porting of Fortran code to target GPU
	- 2 3 months "walltime" efforts for this project
	- $-$ ~1 day / week "person time" efforts
- OpenMP 4.5 (and later) is a path to port code to GPU
	- more compilers are supporting OpenMP offload (XL, GCC, CCE, Intel, PGI, LLVM-based)
- Code available on <http://github.com/GenASiS>
	- paper describing this work in preparation