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

Reuben D. Budiardja

Scientific Computing Group, Oak Ridge Leadership Computing Facility, Oak Ridge National Laboratory

Christian Cardall

Physics Division, Oak Ridge National Laboratory



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The Application

General Astrophysics Simulation System (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
 → 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)
 → 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 () \leftarrow 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)
 - default copy to-from
- Presence check: if fail, new variable is created on device
 sometime requires explicit association to avoid unintentional data
 - movement

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Offloading Computational Kernel

```
subroutine AddKernel (A, B, D_A, D_B, D_C, C)
 1
\mathbf{2}
3
    real (KDR), dimension (:), intent (in) :: A, B
          (c_ptr), intent (in) :: D_A, D_B, D_C
 4
    type
5
    real (KDR), dimension (:), intent (out) :: C
6
7
    integer (KDI) :: i
                                                          call F % Initialize &
8
9
                                          Tells OpenMP data location on GPU iables])
     call AssociateHost ( D_A, A
10
    call AssociateHost
                         D_B, B
                                          \rightarrowavoid (implicit) allocation & transfer eDevice ()
11
     call AssociateHost ( D_C, C
                                                          call r % upualedevice ( )
12
    !$OMP target teams distribute parallel do schedule (static, 1)
13
                                                          call AddKernel &
14
    do i = 1, size ( C )
                                                             (F % Value (:, 1),
     C(i) = A(i) + B(i)
15
16
                                                               F % Value ( :, 2 ), &
    end do
17
     !$OMP 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),
     call DisassociateHost
21
                             А
                                                               F % Value (:, 3))
22
  end subroutine AddKernel
23
```

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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)
2
 3
    real (KDR), dimension (-1:, -1:, -1:), intent (in) :: V
    type (c_ptr), intent (in) :: D_V, D_dV
 4
    real (KDR), dimension (-1:, -1:, -1:), intent (out) :: dV
5
6
7
    integer (KDI) :: i, j, k
8
9
     call AssociateHost ( D_V, V )
     call AssociateHost ( D_dV, dV )
10
11
     !$OMP target teams distribute parallel do collapse (3) schedule (static, 1)
12
13
    do k = 1, nZ
      do j = 1, nY
14
15
        do i = 0, nX + 2
          dV (i, j, k) = V (i, j, k) - V (i - 1, j, k)
16
17
        end do
18
      end do
    end do
19
20
     !$OMP 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

1:	Call: Initialize ()
2:	Call: GhostExchange ()
3:	Set: Time = StartTime
4:	while Time $<$ FinishTime do
5:	Call: ComputeTimeStep () \rightarrow TimeStep
6:	Set: $FluidOld = FluidCurrent$
7:	
8:	Call: ComputeDifferences ()
9:	Call: Compute Reconstruction $()$
10:	Call: ComputeFluxes ()
11:	Call: ComputeUpdate (TimeStep) \rightarrow FluidUpdate
12:	Set: $FluidCurrent = FluidOld + FluidUpdate$
13:	
14:	Call: GhostExchange ()
15:	
16:	Call: ComputeDifferences ()
17:	Call: Compute Reconstruction $()$
18:	Call: ComputeFluxes ()
19:	Call: ComputeUpdate (TimeStep)
20:	Set: FluidCurrent = $0.5 *$ (FluidOld + FluidCurrent + FluidUpdate)
21:	
22:	Call: GhostExchange ()
23:	end while



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:
7:
     DEVICE: Set: FluidOld = FluidCurrent
8:
9:
     DEVICE: Call: ComputeDifferences ()
     DEVICE: Call: ComputeReconstruction ()
10:
     DEVICE: Call: ComputeFluxes ()
11:
     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)
     DEVICE: Set: FluidCurrent = 0.5 * (FluidOld + FluidCurrent + FluidUpdate)
23:
24:
25:
     TRANSFER: Call: FluidCurrent % UpdateHost ()
26:
      Call: GhostExchange ()
27: end while
```



Performance Results

(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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Implications (Then and Now)

- c. 2010: 1024³ cells with 64000 processors (Jaguar), ~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