Fortran on GPUs

STAR Working Session: GPGPUs

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Fortran on GPUs

Answers to the most important questions in 15 minutes!

• Fortran for HPC, not Vis!
• Who provides and supports it?
• What is it?
• How does it work?
• Why is it important?
• What are the capabilities and benefits?

• Why is it so exciting?
Who provides and supports Fortran on GPUs?

- Available on NVIDIA's GPUs
- Developed by “The Portland Group” (PGI)
  - Partnership with NVIDIA
  - Strong interest by NVIDIA to keep the compiler up-to-date as the evolution of CUDA continues
- Name: PGI CUDA Fortran Compiler
- Two usage models
  - CUDA Fortran, allows also for automatic kernel generation
  - CUDA Fortran Accelerator Directives (similar to OpenMP)
What is CUDA Fortran?

- Extension to Fortran that enables access to NVIDIA GPUs through “native” Fortran
- No C code necessary
  - Host code written in Fortran
  - Device code (Kernels) written in Fortran
  - No need to call C-kernels from Fortran
- Complete CUDA C API available in Fortran
How does CUDA Fortran work?

- PGI CUDA Fortran leverages:
  - Existing Fortran constructs
    - Variable/function/subroutine attributes, allocate/deallocate
    - Array syntax for data transfer
  - CUDA C language elements (translated to native Fortran!) and naming conventions
    - Host, global, device, kernel, etc.
    - thread grids, thread blocks
- Compile with flag: `pgf90 -Mcuda` (or `-Mcuda=emu`)
- Generally: Kernels are simple by nature
  - Number of instructions, number of arguments in calls, etc.
  - Limited complexity of code-logic (IF-THEN-ELSE constructs)
  - Limited complexity of language constructs (OO, Polymorph.)
- Source-to-source translation of the kernel
  - Details hidden under the covers
Why is CUDA Fortran important?

- One half of the users of our large clusters at TACC (Ranger, Lonestar, Longhorn) code in Fortran
  - Half of all source code is written in Fortran
  - Half of the service units (SUs) are consumed by Fortran executables
  - The other half is C (including some C++)
  - … and there are users that need to move on to one of these languages

- The same applies to other HPC communities
  - The fraction of Fortran users and Fortran code in HPC is large

- Vendors that embrace Fortran users have a huge advantage in the “Battle of the Accelerators”
What are the capabilities and benefits?

• CUDA Fortran provides full* access to the GPU
  – Kernel launch
  – GPU memory
    • Main memory (GDDR5)
    • Shared and constant memory
    • Registers
    • *no easy access to texture memory yet
  – “Pinned” memory on CPU
  – Synchronous data transfer
  – Asynchronous data transfer and streams
  – Grids and blocks of threads
  – Thread synchronization
What are the capabilities and benefits?

- CUDA Fortran looks and feels like Fortran
- Existing Fortran language elements are expanded
  - Attributes for variables and subroutines/functions
    - Declare where a variable/array is stored: host or device
    - Declare where a subroutine/function is executed: host or device
  - Array syntax
    - Transfer of data between host and device
  - Predefined structures with thread grid & block information
- No “clunky” API calls to declare/allocate variables and to transfer data
  - Full API is exposed,
  - all major API calls are integrated into the Fortran language
Why is CUDA Fortran so exciting?

- CUDA C exposes the API to programmers
- CUDA Fortran does the same, but goes much farther ...
- CUDA Fortran is an extension to the Fortran language

Why is access to the API not enough?

- Compilers are here to help!
  - Integration allows compiler to optimize for performance
  - HPC languages (Fortran, C++) evolve to fit our needs
  - Code becomes more readable if API is integrated
  - Automatic kernel generation
  - Better code can be written faster ...
- We are already “stuck” with one “bare-metal” API
  - All attempts to integrate MPI have given us bad performance
  - We can’t afford to pile on API after API
program cuda_fortran

real, allocatable, dimension(:,:,:), &
  device :: h               ! h on Host
real, allocatable, dimension(:,:,:), device :: d               ! d on Device
real :: a = 2.1 ! variable

allocate (h(512,5,20), d(512,5,20)) ! Allocate on Host and Device

h = ... ! Preset h

d = h    ! Transfer 3-dim array from host to device (Array syntax)

dimblock = dim3(512,1,1)  ! Block of Threads (dim3: derived type)
dimgrid  = dim3(5, 20,1)  ! Grid of Threads

call sub_gpu<<<dimgrid,dimblock>>>(d,a)

h = d    ! Transfer 3-dim array back

end
program cuda_fortran

real, allocatable, dimension(:,:,:,:)            :: h          ! h on Host
real, allocatable, dimension(:,:,:,:), device  :: d          ! d on Device
real                                         :: a = 2.1   ! variable

allocate (h(512,5,20), d(512,5,20))       ! Allocate on Host and Device

h = ...    ! Preset h

! Transfer 3-dim array from host to device (Array syntax)
d = h

dimblock = dim3(512,1,1)  ! Block of Threads (dim3: derived type)
dimgrid  = dim3(5, 20,1)  ! Grid of Threads

call sub_gpu<<<dimgrid,dimblock>>>(d,a)

! Transfer 3-dim array back
h = d

end
module madd_device_module
  use cudafor
contains
  subroutine madd_dev(a,b,c,n1,n2)
    real, dimension(:,:), device :: a,b,c
    integer                      :: n1,n2
    real                         :: sum = 0.
!$cuf kernel do(2) <<<(*,*),(32,4)>>>
    do j=1, n2
      do i=1, n1
        a(i,j) = b(i,j) + c(i,j)
        sum    = sum    + a(i,j)
      enddo
    enddo
  end subroutine
end module

No Synchronization beyond a thread block!
A global reduction requires:
• Reduction on block level (Step I) within kernel
• Exiting kernel for global synchronization
• Second kernel for reduction on grid level (Step II)
• Both reductions are operations on a tree
CUDA code becomes very convoluted very quickly!

New attributes    Thread parameters
Data/Subroutine on device
Compiler directive creating kernel
Chevron syntax for Kernel launch

"Kernelize" the two (2) loops
!$cuf similar to OpenMP !$omp
Use 2D blocks: x=32 y=4
Grid shape automatic (*,*)
Compiler can handle a
"global" reduction

CUDA Fortran "Kernel Loop Directives"
CUDA Fortran is forward-looking …

• CUDA Fortran empowers 50% of the HPC community
  – Exposure of the API
  – Integration into language, way ahead of C and C++
  – Expect to see integration of CUDA into C++
  – GPU-language extension is easier to learn than “bare” API

• Other vendors? AMD, Intel

• OpenMP accelerator directives: pushed by Cray, etc.

• Allow the compiler to:
  – help us code more efficiently
  – optimize for execution speed
  – minimize and group data transfer

• APIs can be the way to go (e.g. MPI), but …

Getting to Exascale means developing new languages/language paradigms, not using more bare APIs