Development of a Portable, GPU-Accelerated High-Order Discontinuous Galerkin CFD Code for Compressible Flows on Hybrid Grids

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February 7, 2014



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Outline

- ① A brief overview
- ② Motivation and objective
- ③ Governing equations of fluid dynamics
- Discontinuous Galerkin formulation
- ⑤ OpenACC-based parallelism
- 6 Numerical examples
- ⑦ Concluding remarks
- 8 Future work

GPU Computing in CFD – Overview

GPGPU

General-purpose computing on graphics processing units

• The utilization of GPU, which typically handles computation only for computer graphics, to perform computation in applications traditionally handled by CPU.

Why?

The great potential and scalability of GPGPU for CFD applications!

How?

Offload the computing-intensive portion from the host CPUs to the GPU devices.

Programming models

- **OpenCL**: the currently dominant open GPGPU programming language
- **NVIDIA's CUDA**: the dominant proprietary framework
- **OpenACC**: a collection of directives designed to simplify parallel programming

GPU Computing in CFD – Overview

A variety of applications (selected)

- For the finite difference method (FDM)
 - E. Elsen, P. LeGresley, and E. Darve.
 <u>Large Calculation of the Flow over a Hypersonic Vehicle Using a GPU.</u> J. Comput. Phys., 227(24):10148–10161, 2008.

For the finite volume method (FVM)

 A. Corrigan, F. Camelli, R. Löhner, and J. Wallin.
 <u>Running Unstructured Grid-based CFD Solvers on Modern Graphics Hardware.</u> *Int. J. Numer. Methods Fluids*, 66(2):221–229, 2011.

For the spectral difference method (SDM)

- B. Zimmerman, Z. Wang, and M. Visbal.
 <u>High-Order Spectral Difference: Verification and Acceleration Using GPU Computing.</u> AIAA Paper, 2013-2491, 2013.
- For the discontinuous Galerkin method (DGM)
 - A. Klöckner, T. Warburton, J. Bridge, and J. S. Hesthaven
 <u>Nodal Discontinuous Galerkin Methods on Graphics Processors.</u> J. Comput. Phys., 228(21):7863–7882, 2009.

Observation: most GPU-related CFD solvers are based on CUDA

Motivation & Objectives

Motivation

Tap the power of GPU parallel computing for the aerodynamic design of Unmanned Aerial Vehicle (UAV) with CFD toolkits



MQ-1 Predator UAV (Source: grabcad.com CAD by Chao Wey)

Motivation & Objectives

Objectives

A portable, efficient and ultimately competitive GPU parallelization strategy for extensible and sustainable high-fidelity CFD code development

An unavoidable debate – CUDA, OpenCL or OpenACC?

It really depends on your needs, e.g., for us, cross-platform portability

- Support from a wide range of compiler and accelerator vendors
- C/C++ and Fortran
- Best performance
- · Less fine-tuning effort, especially for a legacy code package
- Available computing resource



GPU-Computing Framework

Option 1. – based on NVIDIA's CUDA

- Why CUDA?
 - □ Mature and ever-updating GPU parallelization standards for HPC
 - Currently wide user community support
 - □ Achievable optimal performance with fine-tuning
 - □ Strong GPU-accelerated library support, e.g., CULA tools
 -
- Why not CUDA?
 - □ Complex and explicit layout of threads on GPU for each kernel function
 - Excessive workload to upgrade an existing CFD package
 - □ Uncertainty in the vendor's long-term development strategy
 - □ Constrained **portability** of the developed code on non-CUDA devices

CUDA alone can not satisfy not only our primary design goals, but also many others who hesitate to adopt GPU computing!

GPU-Computing Framework

Option 2. – based on OpenACC

Directives for accelerators http://www.openacc-standard.org/

Why OpenACC?

□ Simple directive-based GPU parallelization strategy, similar to OpenMP

□ Multi-compiler / multi-platform support

Growing supporting community

Why not OpenACC?

□ If a fine-tuned, best-performance code is what you pursue

□ A number of limitations compared with CUDA



Our choice: OpenACC meets most of our design requirements!

GPU-Computing Framework

Desirable features of the resulting CFD code

- Multi-compiler compatibility
 - GNU Fortran compiler
 - Intel Fortran compiler
 - PGI Accelerator Fortran compiler (with OpenACC support)
 - □ CAPS Fortran compiler (with **OpenACC** support)



(intel) Intel[®] Software **Development Products**



- Cross-platform portability
 - □ Intel CPUs
 - □ AMD CPUs / APUs (with potential **OpenACC** support in 2014)
 - □ NVIDIA CUDA-enabled GPUs (with **OpenACC** support)





- Extensible and sustainable programming schemes
- Competitive performance

Legacy CFD Package

RDGFLO – a baseline code for OpenACC-based GPU parallelization

- A <u>Reconstructed Discontinuous Galerkin finite element FLOw solver</u>
 - □ High-order solution of compressible flows on 3-D hybrid grids
 - Explicit / implicit solution schemes
 - Domain-partition based MPI parallel computing

Gallery



ONERA M6 wing

wing/pylon/finnedstore configuration **Boeing 747 aircraft**

Governing Equations of Fluid Dynamics

The Navier-Stokes equations for unsteady compressible flows

$$\frac{\partial \mathbf{U}}{\partial t} + \frac{\partial \mathbf{F}_k(\mathbf{U}, t)}{\partial \mathbf{X}_k} = \frac{\partial \mathbf{G}_k(\mathbf{U}, \nabla \mathbf{U}, t)}{\partial \mathbf{X}_k}$$

where the summation convention is used. The conservative variable vector \mathbf{U} , advective flux vector \mathbf{F} , and viscous flux \mathbf{G} are defined by

$$\mathbf{U} = \begin{pmatrix} \rho \\ \rho u_i \\ \rho e \end{pmatrix} \qquad \mathbf{F}_j = \begin{pmatrix} \rho u_j \\ \rho u_i u_j + p \delta_{ij} \\ u_j (\rho e + p) \end{pmatrix} \qquad \mathbf{G}_j = \begin{pmatrix} 0 \\ \tau_{ij} \\ u_l \tau_{ij} + q_j \end{pmatrix}$$

Governing Equations of Fluid Dynamics

The Navier-Stokes equations for unsteady compressible flows (cont.)

The pressure *p* can be computed from the equation of state (EOS)

$$p = (\gamma - 1)(\rho e - \frac{1}{2}\rho u_k u_k)$$

which is valid for perfect gas. The ratio of specific heats γ is assumed to be constant and equal to 1.4.

The viscous stress tensor τ_{ii} and heat flux vector q_i are given by

$$\tau_{ij} = \mu \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right) - \frac{2}{3} \mu \frac{\partial u_k}{\partial x_k} \delta_{ij} \qquad q_j = \frac{1}{\gamma - 1} \frac{\mu}{Pr} \frac{\partial T}{\partial x_j}$$

where *T* is the temperature of the fluid, *Pr* the laminar Prandtl number, which is 0.7 for air. μ represents the molecular viscosity, which can be determined through the Sutherlands law

$$\frac{\mu}{\mu_0} = \left(\frac{T}{T_0}\right)^{3/2} \frac{T_0 + S}{T + S}$$

where μ_0 is the viscosity at the reference temperature T_0 and S = 110K.

Discontinuous Galerkin Method

Weak formulation of the governing equations

$$\frac{d}{dt} \int_{\Omega_e} \mathbf{U}_h B_i \, d\Omega + \int_{\Gamma_e} \mathbf{F}_k \mathbf{n}_k B_i \, d\Gamma - \int_{\Omega_e} \mathbf{F}_k \frac{\partial B_i}{\partial \mathbf{x}_k} \, d\Omega =$$

$$\int_{\Gamma_e} \mathbf{G}_k \mathbf{n}_k B_i \, d\Gamma - \int_{\Omega_e} \mathbf{G}_k \frac{\partial B_i}{\partial \mathbf{X}_k} \, d\Omega, \ 1 \le i \le N$$

where $B_i(\mathbf{x})$ is the basis function of polynomials of degree p, and N is the dimension of the polynomial space p.

The Taylor-basis discontinuous Galerkin (DG(p)) solution in each element

$$\mathbf{U} = \sum_{i=1}^{N} \mathbf{U}_{i} B_{i}$$

For example, the underlying piecewise linear polynomial DG(P1) solution

$$\mathbf{U} = \overline{\mathbf{U}} + \left(\frac{\partial \mathbf{U}}{\partial x}\Delta x\right)\frac{x - x_c}{\Delta x} + \left(\frac{\partial \mathbf{U}}{\partial y}\Delta y\right)\frac{y - y_c}{\Delta y} + \left(\frac{\partial \mathbf{U}}{\partial z}\Delta z\right)\frac{z - z_c}{\Delta z}$$

where (x_c, y_c, z_c) is the coordinate of the element center.

Discontinuous Galerkin Method

Hierarchical WENO reconstruction

- A quadratic polynomial DG(P2) solution is obtained via a hierarchical WENO reconstruction approach in each element
 - 1) Least-squares reconstruction to obtain an initial quadratic solution
 - 2) WENO reconstruction for the second derivatives to maintain linear stability
 - 3) WENO reconstruction for the first derivatives to maintain nonlinear stability

Reference articles

- H. Luo, Y. Xia, S. Li, and R. Nourgaliev. <u>A Hermite WENO Reconstruction-Based Discontinuous Galerkin Method for the Euler</u> <u>Equations on Tetrahedral grids</u>. *J. Comput. Phys.* 231(16):5489–5503, 2012.
- H. Luo, Y. Xia, S. Spiegel, R. Nourgaliev, and Z. Jiang. <u>A Reconstructed Discontinuous Galerkin Method Based on a Hierarchical WENO</u> <u>Reconstruction for Compressible Flows on Tetrahedral Grids.</u> *J. Comput. Phys.* 236:477–492, 2013.

Discontinuous Galerkin Method

Semi-discrete form

A system of ordinary differential equations (ODEs) in time

$$M\frac{d\mathbf{U}}{dt} = \mathbf{R}$$

where M is the mass matrix and R is the residual vector.

Three-stage TVD Runge-Kutta (TVDRK3) time stepping

$$\mathbf{U}^{(1)} = \mathbf{U}^{n} + \Delta t \mathbf{M}^{-1} \mathbf{R}(\mathbf{U}^{n})$$
$$\mathbf{U}^{(2)} = \frac{3}{4} \mathbf{U}^{n} + \frac{3}{4} \Big(\mathbf{U}^{(1)} + \Delta t \mathbf{M}^{-1} \mathbf{R}(\mathbf{U}^{(1)}) \Big)$$
$$\mathbf{U}^{n+1} = \frac{1}{3} \mathbf{U}^{n} + \frac{2}{3} \Big(\mathbf{U}^{(2)} + \Delta t \mathbf{M}^{-1} \mathbf{R}(\mathbf{U}^{(2)}) \Big)$$

Design of OpenACC Parallel Regions

Computing-intensive regions

1. ~65% Internal & boundary face integral: loop over mesh faces T_e

$$\int_{\Gamma_e} \mathbf{F}_k \mathbf{n}_k B_i \, d\Gamma \qquad \qquad \int_{\Gamma_e} \mathbf{G}_k \mathbf{n}_k B_i \, d\Gamma$$

2. ~25% Domain integral: loop over mesh elements Ω_e

$$\int_{\Omega_e} \mathbf{F}_k \frac{\partial B_i}{\partial \mathbf{x}_k} d\Omega \qquad \qquad \int_{\Omega_e} \mathbf{G}_k \frac{\partial B_i}{\partial \mathbf{x}_k} d\Omega$$

3. ~5% TVDRK3 time stepping: loop over mesh elements Ω_e

Design of OpenACC Parallel Regions

Example: a readily vectorizable region		
Domain integral: loop over mesh elements $\Omega_e \int_{\Omega_e} \mathbf{F}_k \partial B / \partial \mathbf{x}_k d\Omega$		
OpenMP version OpenACC version		
<pre>!\$omp parallel !\$omp do do ie = 1, Nelem do ig = 1, Ngp ! contribution to this element rhsel(:, :, ie) = rhsel(:, :, ie) + flux enddo enddo !\$omp end parallel</pre>	<pre>!\$acc parallel !\$acc loop do ie = 1, Nelem do ig = 1, Ngp ! contribution to this element rhsel(:, :, ie) = rhsel(:, :, ie) + flux enddo enddo !\$acc end parallel</pre>	

Design of OpenACC Parallel Regions

Example: a region that is not readily vectorizable		
Face integral: loop over mesh faces T_e	$\int_{\Gamma_e} \mathbf{F}_k \mathbf{n}_k B_i d\Gamma$	
OpenMP version	OpenACC version	
!\$omp parallel	!\$acc parallel	
!\$omp do	!\$acc do	
do ifa = Njfac+1, Nafac	do ifa = Njfac+1, Nafac	
do ig = 1, Ngp	do ig = 1, Ngp	
I contribution to the face-left element	I contribution to the face-left element	
rhsel(:, :, iel) = rhsel(:, :, iel) - flux	rhsel(:, :, iel) = rhsel(:, :, iel) - flux	
! contribution to the face-right element	I contribution to the face-right element	
rhsel(:, :, ier) = rhsel(:, :, ier) + flux	rhsel(:, :, ier) = rhsel(:, :, ier) + flux	
enddo enddo enddo		
enddo	enddo	
!\$omp end parallel	!\$acc end parallel	

"race condition" – multiple writes to the same elemental residual vector!

Design of OpenACC Parallel Regions

A common method for avoiding "race condition" in face integral

- Perform face integral at the element level
 - □ All the computing is implemented as loops over mesh elements
- Overheads
 - □ Lead to redundant computation of face integrals (doubled!)
 - □ Require an additional element-face connectivity array
- Reference articles on unstructured DG/FV methods (with CUDA)
 - A. Corrigan et al. <u>Running Unstructured Grid-based CFD Solvers on Modern</u> <u>Graphics Hardware</u>. Int. J. Numer. Methods Fluids, 66(2):221–229, 2011.
 - Tristan Cabel and Stephane Lanteri. <u>Discontinuous Galerkin Time-Domain Solver on</u> <u>GPU Based Systems</u>. Plafrim meeting, May 31, 2011

This design approach requires a major rebuild in code structures!

Design of OpenACC Parallel Regions

An alternative approach to avoiding "race condition" in face integral

Face renumbering (coloring method)

□ renumber the faces and split them into several groups.

Any two faces that share a common cell do not reside in the same group.

□ The original face loops are nested in a sequential loop over groups.

An example of "face renumbering & grouping" Before Physical boundary faces Partition boundary faces After 4 groups 3 groups 6 groups

Design of OpenACC Parallel Regions

Example: face loops after "face renumbering & grouping"

The original face loops are nested in a sequential loop over groups.

OpenMP version

```
Nfac1 = Njfac
do ipass = 1, Npass ift
  Nfac0 = Nfac1 + 1
  Nfac1 = fpass ift(ipass)
  !$omp parallel
  !$omp do
  do ifa = Nfac0, Nfac1
     do ig = 1, Ngp
       !... contribution to the face-left element
       rhsel(:, :, iel) = rhsel(:, :, iel) - flux
       !... contribution to the face-right element
       rhsel(:, :, ier) = rhsel(:, :, ier) + flux
     enddo
  enddo
  !$omp end parallel
enddo
```

OpenACC version

```
Nfac1 = Njfac
do ipass = 1, Npass ift
  Nfac0 = Nfac1 + 1
  Nfac1 = fpass_ift(ipass)
  !$acc parallel
  !$acc do
  do ifa = Nfac0, Nfac1
     do ig = 1, Ngp
       I... contribution to the face-left element
       rhsel(:, :, iel) = rhsel(:, :, iel) - flux
       I... contribution to the face-right element
       rhsel(:, :, ier) = rhsel(:, :, ier) + flux
     enddo
  enddo
  !$acc end parallel
enddo
```

Design of OpenACC Parallel Regions

"Face renumbering & grouping" vs. "element loops"

Advantages

- □ No redundant computation.
- □ Least intrusion to the original code structures.
- □ Recoverable to CPU parallel computing.

Disadvantages

Sequential groups lead to overheads in initializing more parallel kernels.

Remarks

- 1. No direct comparison of these two strategies is yet conducted for our solver. It is unknown which will render better performance on GPU.
- 2. Since **portability** is our design priority, the current approach is a preferred, although **optimal performance** might be compromised.

Hardware/Software Resource

CPU

- 2-way SMPs with AMD Opteron 6128 (Many Core) with 8 cores per socket (16 cores per node)
 - □ 32 GB DRAM
 - □ 2.0 GHz core-speed for 6128 Opteron (single core)

GPU

- NVIDIA Tesla K20c
 - □ Memory amount: **5.0 GB**
 - □ Stream processors: 2496

Software (64 bit)

- Operating system
 Operation S 5 7 Linux x
 - CentOS 5.7 Linux x86 64
- Compilation & runtime suite
 - PGI Accelerator Fortran Compiler (ver. 13.4)
 - OpenMPI (ver. 1.5.5)





SECOND ELECTRON



Performance Assessment

Timing measurements

Unit running time T_{unit}

$$T_{\text{unit}} = \frac{T_{\text{run}}}{\text{Ntime} \times \text{Nelem}} \times 10^6 \text{ (microsecond)}$$

where the running time $T_{\rm run}$ refers to the time recorded for completing the time marching loop with a given iteration number Ntime.

Control sets (simulations in double-precision arithmetic)		
GPU-1	Parallel computing with 1 GPU device	
CPU-1	Serial computing with 1 CPU core	
CPU-16	Parallel computing with 16 CPU cores (domain partitioning + MPI)	

Example 1. Subsonic Flow past a Sphere

Subsonic flow past a sphere at M_{∞} = 0.50 and α = 0°.



Figure. Pressure contours plotted on the surface triangular meshes of a sequence of four successively refined tetrahedral grids

	Unit time (microsecond)			Spee	edup
Nelem	GPU-1	CPU-1	CPU-16	vs. CPU-1	vs. CPU-16
2,426	20.2	176.8	14.8	8.8	0.73
16,467	10.7	182.8	12.6	17.0	1.18
124,706	9.3	182.8	13.0	19.6	1.40
966,497	8.8	198.9	13.1	22.6	1.49

Example 1. Subsonic Flow past a Sphere

Subsonic flow past a sphere at M_{∞} = 0.50 and α = 0°.



Example 1. Subsonic Flow past a Sphere



Anim. Comparison of runtime performance rendered by pressure contours on the surface meshes

Example 2. Transonic Flow over a Boeing 747

Transonic flow over a Boeing 747 aircraft at M_{∞} = 0.85 and α = 2°



Figure. Mach number contours plotted on the surface triangular meshes

	Unit time (microsecond)			Spe	edup
Nelem	GPU-1	CPU-1	CPU-16	vs. CPU-1	vs. CPU-16
253,577	11.5	243.8	16.4	21.2	1.43
1,025,170	10.6	249.4	16.6	24.5	1.57

Example 2. Transonic Flow over a Boeing 747

Transonic flow over a Boeing 747 aircraft at M_{∞} = 0.85 and α = 2°



Example 2. Transonic Flow over a Boeing 747

Transonic flow over a Boeing 747 aircraft at M_{∞} = 0.85 and α = 2 \circ



CPU-16

CPU-1

Anim. Comparison of runtime performance rendered by Mach number contours on the surface meshes

Example 3. Viscous Flow past a Sphere

Subsonic flow past a sphere at M_{∞} = 0.5 and Re = 118



Triangular surface meshes



Streamtraces on the symmetry plane

	Unit time (microsecond)			Spe	edup
Nelem	GPU-1	CPU-1	CPU-16	vs. CPU-1	vs. CPU-16
200,416	14.6	259.9	20.5	17.8	1.41
925,995	13.9	257.2	20.6	18.5	1.48

Example 3. Viscous Flow past a Sphere

Subsonic flow past a sphere at M_{\odot} = 0.5 and Re = 118 Time = 0.00800522579184411 Time = 0.00800522579184411 Time = 0.00800522579184411 Mach: 0.05 0.15 0.25 0.35 0.45 0.55 Mach: 0.05 0.15 0.25 0.35 0.45 0.55 Mach: 0.05 0.15 0.25 0.35 0.45 0.55 **GPU-1**

CPU-16

CPU-1

Anim. Comparison of runtime performance rendered by Mach number contours along with streamtraces on the symmetry plane

Example 4. Large Eddy Simulation (to be completed soon...)

Lid-driven cubical cavity at $M_{\rm b}$ = 0.2 and Re = 10,000



Concluding Remarks

RDGFLO with OpenACC

- Performance on single GPU
 - □ An average scaling factor of over **20.0x** vs. 1 CPU core
 - □ An average scaling factor of over **1.4x** vs. 16 CPU cores
- Extensibility
 - □ The current parallel strategy can be applied to more functionalities
- Verified compilers
 PGI Fortran
- Verified platform
 NVIDIA CUDA-enabled GPUs

Limitation

- Memory constraint on GPU device
- □ Balance between [optimal GPU parallelism] and [code portability]

Publications

Y. Xia, L. Luo, H. Luo, J. Edwards, J. Lou, and F. Mueller. <u>OpenACC-based GPU Acceleration of</u> <u>a 3-D Unstructured Discontinuous Galerkin Method.</u> 52nd AIAA Aerospace Sciences Meeting, AIAA-2014-1129, January 2014.

1. Multi-GPU parallelization via MPI

• Face grouping algorithm for balanced load over partition-domain regions



1. Multi-GPU parallelization via MPI (cont.)

- Atomic operation: !\$acc atomic
 - Declared in OpenACC 2.0
 - Not yet implemented in PGI Accelerator
 - □ Why we need it?
 - Significant simplification of programming
 - Potential huge improvement of scalability

Physical boundary faces	Partition boundary faces	Internal faces
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Future: atomic operation (no grouping needed), e.g., in 2 partition domains



2. Implicit time integration

- Preconditioned linear solver, e.g., GMRES-LU+SGS algorithm
 - The inverse of the approximate block diagonal Jacobian matrix is the least we need for preconditioning
 - □ Any direct algorithm of matrix inverse requires some auxiliary arrays, as below

OpenMP version	OpenACC version
real*8 A(20,20,Nel) real*8 B(20,20) real*8 C(20)	real*8 A(20,20,Nel) real*8 B(20,20) real*8 C(20)
<pre>!\$omp parallel !\$omp do do ie = 1, Nel Direct algorithm to invert A(20,20,ie) with the aid of B(20,20) and C(20) enddo !\$omp end parallel</pre>	<pre>!\$acc parallel !\$acc loop do ie = 1, Nel ! Direct algorithm to invert A(20,20,ie) ! with the aid of B(20,20) and C(20) enddo !\$acc end parallel</pre>

However, the direct algorithm in the OpenACC ver. is practically useless, due to very limited "shared memory" (48~64KB) / thread block on GPU!

2. Implicit time integration (cont.)

- Solution: iterative algorithm for matrix inverse
 - Pros
 - No need of auxiliary arrays, suitable for OpenACC parallel regions
 - Adequate solution efficiency for diagonal-dominant matrix

Cons

To be explored...

OpenMP version	OpenACC version
real*8 A(20,20,Nel) real*8 B(20,20) real*8 C(20)	real*8 A(20,20,NeI)
<pre>!\$omp parallel !\$omp do do ie = 1, Nel Direct algorithm to invert A(20,20,ie) with the aid of B(20,20) and C(20) enddo !\$omp end parallel</pre>	<pre>!\$acc parallel !\$acc loop do ie = 1, Nel ! Iterative algorithm to invert A(20,20,ie) ! with no need of auxiliary arrays enddo !\$acc end parallel</pre>

Development Plans

Recent work

Multi-GPU parallelization via MPI

□ A balanced face grouping algorithm is required over domain partitions.

Challenges for implementing implicit time integration

- □ Algorithms suitable for GPU parallel computing
- □ memory limitation on GPU devices.

Goals

A portable, GPU-accelerated high-order CFD toolkit

A complete framework for the reconstructed discontinuous Galerkin (RDG) method for compressible flows on hybrid grids

Impact

A potential, competitive parallel-computing model of CFD programming for the next-generation HPC hardware/software