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PyFR: An Open Source Python Framework for High-Order CFD on Many-Core Platforms

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Introduction

- Various types of high-order accurate numerical methods for unstructured grids
- Continuous Galerkin (CG), Discontinuous Galerkin (DG), Spectral Difference (SD), Spectral Volume (SV), Flux Reconstruction (FR)

Introduction

Various types of high-order accurate numerical methods for unstructured grids

 Continuous Galerkin (CG), Discontinuous Galerkin (DG), Spectral Difference (SD), Spectral Volume (SV), Flux Reconstruction (FR)

Introduction

- FR approach was first proposed by Huynh at NASA Glenn in 2007 [1]
- Based on differential form of governing system
- Unifying, can cast various known + new schemes within a single framework

[1] H.T. Huynh. A flux reconstruction approach to high-order schemes including discontinuous Galerkin methods. AIAA Paper 2007-4079. 2007

Theory

- Consider ID scalar conservation law $\frac{\partial u}{\partial t} + \frac{\partial f}{\partial x} = 0$
- Divide ID domain into elements

$$\Omega = \bigcup_{n=0}^{N-1} \Omega_n \qquad \bigcap_{n=0}^{N-1} \Omega_n = \emptyset$$



$\Omega_n = \{ x | x_n < x < x_{n+1} \}$



Theory

- Consider ID scalar conservation law $\frac{\partial u}{\partial t} + \frac{\partial f}{\partial x} = 0$
- Represent solution by order k piecewise discontinuous polynomials in each element



Theory

- Consider ID scalar conservation law $\frac{\partial u}{\partial t} + \frac{\partial f}{\partial x} = 0$
- Represent flux by order k+1 piecewise continuous polynomials within each element



Theory

• Map each element to a standard element

$$\Omega_S = \{r| - 1 \le r \le 1\} \qquad r = 2\left(\frac{x - x_n}{x_{n+1} - x_n}\right) - 1$$

$$\frac{\partial \hat{u}^{\delta}}{\partial t} + \frac{\partial \hat{f}^{\delta}}{\partial r} = 0 \qquad \hat{u}^{\delta}$$

$$h_n = (x_{n+1})$$



$$= u_n^\delta \qquad \hat{f}^\delta = \frac{f_n^\delta}{h_n}$$

$$(x_n)/2$$

Theory

 Represent order k solution within standard element using a nodal basis



Theory



Theory

• Evaluate solution at element boundaries

$$\hat{u}^{\delta}(-1) = \sum_{i=0}^{k} \hat{u}_{i}^{\delta} l_{i}(-1)$$



$$\hat{u}^{\delta}(1) = \sum_{i=0}^{k} \hat{u}_i^{\delta} l_i(1)$$

Theory

• Calculate common interface fluxes



Overview | Introduction | Theory | Implementation | Results | Summary

Define order k+1 'correction function'

 $g_L(-1) = 1, \quad g_L(1) = 0$



Theory

• Scale this correction function...





Theory

• ... and add it to the discontinuous flux





• Repeat from the right...







• Repeat from the right...





• Repeat from the right...



Theory

• Evaluate divergence of the continuous flux at the solution points







Theory

- Nature of FR scheme depends on location of solution points, interface flux, correction function
- Can recover nodal Discontinuous Galerkin (DG) method, and any Spectral Difference (SD) method via judicious choice of correction function [1]

[1] H.T. Huynh. A flux reconstruction approach to high-order schemes including discontinuous Galerkin methods. AIAA Paper 2007-4079. 2007



Theory

In 2011 Vincent, Castonguay and Jameson identified a range of stable correction functions for all orders of accuracy using an energy method [2][3]

[2] P. E. Vincent, P. Castonguay, A. Jameson. A New Class of High-Order Energy Stable Flux Reconstruction Schemes. Journal of Scientific Computing. 2011 [3] A. Jameson, P. E. Vincent, P. Castonguay. On the Non-Linear Stability of High-Order Flux Reconstruction Schemes. Journal of Scientific Computing. 2011

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Theory

- Most operations point/element local
- No explicit quadrature (c.f. DG)
- No global assembly (c.f. FEM)
- No matrix inversions



Recent rapid evolution of hardware available for scientific computing – exciting times



PyFR

- Multi-core CPUs now ubiquitous (with 10's of cores per device)
- Rely extensively on vectorization for throughput







PyFR

Many-core accelerators a hot topic (100s - 1000s cores per device)





PyFR

GPUs may or may not be the future... but many-core + low memory-per-core quite likely is





Nvidia K20

Intel Phi







AMD S10000

PyFR

- What approach have we adopted?
- Write platform-independent code in Python and interface with accelerator via bespoke kernels
- Leverage the near universal availability of optimized **BLAS** libraries for CPUs/accelerators





CUDA



Implementation

- Setup (SymPy + NumPy)
- Kernel generation (Mako)
- Kernel invocation (PyCUDA &c.)
- Communication (mpi4py)

C/OpenMP



- Current status
- 4000 lines of Python
- 800 lines of Mako/CUDA
- 800 lines of Mako/C
- Slated for release under a BSD license in late summer

Theory of VCJH correction functions $\eta_k \in \left\{ 0, \frac{k}{k+1}, \frac{k+1}{k}, \dots \right\} \qquad g'_R =$

Implementation using SymPy

def diff vcjh correctionfn(k, eta, sym): # Expand shorthand forms of eta k for common schemes etacommon = dict(dg='0', sd='k/(k+1)', hu='(k+1)/k')

eta k = sy.S(etacommon.get(eta, eta), locals=dict(k=k))

lkm1, lk, lkp1 = [sy.legendre poly(m, sym) for m in [k-1, k, k+1]]

Correction function derivatives, Eq. 3.46 and 3.47 diffgr = (sy.S(1)/2 * (lk + (eta k*lkm1 + lkp1)/(1 + eta k))).diff()diffgl = -diffgr.subs(sym, -sym)

return diffgl, diffgr



$$= \frac{1}{2} \frac{d}{dr} \left[L_k + \frac{\eta_k L_{k-1} + L_{k+1}}{1 + \eta_k} \right]$$

Kernel generation using Make templates

```
<%include file='idx of.cu.mak' />
 global void
negdivconf(int nupts, int neles,
          ${dtype}* restrict tdivtconf,
          const ${dtype}* restrict rcpdjac,
          int ldt, int ldr)
   int eidx = blockIdx.x * blockDim.x + threadIdx.x;
   if (eidx < neles)
       for (int uidx = 0; uidx \leq nupts; ++uidx)
           ${dtype} s = -rcpdjac[IDX OF(uidx, eidx, ldr)];
       % for i in range(nvars):
           tdivtconf[U IDX OF(uidx, eidx, ${i}, neles, ldt)] *= s;
       % endfor
```



Kernel generation using Mako templates

```
#ifndef PYFR IDX OF
#define PYFR IDX OF
#define IDX OF(i, j, ldim) ((i)*(ldim) + (j))
#define U IDX OF(upt, ele, var, nele, ldim) \
   IDX OF(upt, nele*var + ele, ldim)
#endif // PYFR IDX OF
 global void
negdivconf(int nupts, int neles,
           double* restrict tdivtconf,
           const double* restrict rcpdjac,
           int ldt, int ldr)
   int eidx = blockIdx.x * blockDim.x + threadIdx.x;
   if (eidx < neles)</pre>
        for (int uidx = 0; uidx < nupts; ++uidx)
            double s = -rcpdjac[IDX OF(uidx, eidx, ldr)];
            tdivtconf[U IDX OF(uidx, eidx, 0, neles, ldt)] *= s;
            tdivtconf[U IDX OF(uidx, eidx, 1, neles, ldt)] *= s;
            tdivtconf[U IDX OF(uidx, eidx, 2, neles, ldt)] *= s;
            tdivtconf[U IDX OF(uidx, eidx, 3, neles, ldt)] *= s;
            tdivtconf[U IDX OF(uidx, eidx, 4, neles, ldt)] *= s;
    }
```



Queued compute and MPI kernel execution

```
def get negdivf(self):
    runall = self. backend.runall
    q1, q2 = self. queues
    q1 << self. disu fpts kerns()</pre>
    q1 << self. mpi inters scal fpts0 pack kerns()</pre>
    runall([q1])
    q1 << self. tdisf upts kerns()</pre>
    q1 << self. tdivtpcorf upts kerns()</pre>
    q1 << self. int inters rsolve kerns()</pre>
    q1 << self. bc inters rsolve kerns()
    q2 << self. mpi inters scal fpts0 send kerns()</pre>
    q2 << self. mpi inters scal fpts0 recv kerns()</pre>
    q2 << self. mpi inters scal fpts0 unpack kerns()</pre>
    runall([q1, q2])
    q1 << self. mpi inters rsolve kerns()</pre>
    q1 << self. tdivtconf upts kerns()</pre>
    q1 << self. negdivconf upts kerns()</pre>
    runall([q1])
```



Results

Some preliminary results on our 4 × K20 workstation



Results

Super 2k + I accuracy for a 2D Euler vortex with k = 3 [4]



[4] P. E. Vincent, P. Castonguay, A. Jameson. Insights from von Neumann analysis of high-order flux reconstruction schemes. Journal of Computational Physics. 2011

Results

• Flow over cylinder at Ma = 0.2 Re = 3900



Results

• 5th order accurate with 2.9 x 10⁷ DOFs



Results

Double precision



Results

- Scaling on Emerald, UK's largest multi-GPU cluster
- 372 × Nvidia M2090s each with 512 CUDA cores



est multi-GPU cluster ith 512 CUDA cores

Results

3D Navier-Stokes weak scaling on Emerald





Results

• 3D Navier-Stokes strong scaling on Emerald



Results

• T106c Low Pressure Turbine Blade at Re = 80,000





Summary

- Interested in FR methods
- Interested in simple and efficient Python based implementations that target multiple hardware platforms
- Interested in application to solve real world industrial flow problems



Summary

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- Any questions?
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