Performance-Portable Implicit Scale-Resolving Compressible Flow Using libCEED

SIAM CSE 2023

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Outline

- 1. What is libCEED?
- 2. libCEED Overview
- 3. Fluid Simulations with libCEED
- 4. Accuracy and Performance of High-Order Scale-Resolving Simulations

· C library for element-based discretizations



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 - · Computational backend selectable at runtime, using runtime compilation
- Geared toward high-order element discretizations

libCEED Overview

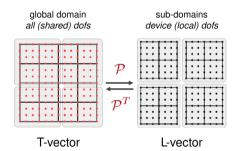
$$A = \mathcal{P}^T \mathcal{E}^T B^T D B \mathcal{E} \mathcal{P}$$

global domain all (shared) dofs

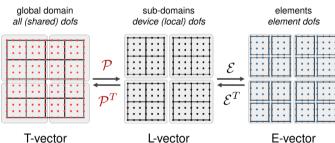


T-vector

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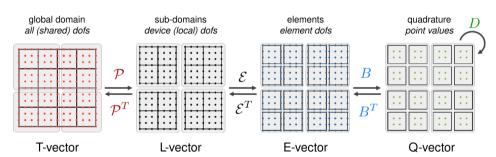


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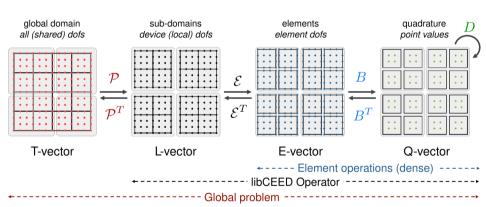


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$$A = \mathcal{P}^T \mathcal{E}^T B^T DB \mathcal{E} \mathcal{P}$$



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Fluid Simulations with libCEED



Compressible Navier-Stokes

$$\mathbf{A_0}\mathbf{Y}_{.t} + \mathbf{F}_{i.i}(\mathbf{Y}) - S(\mathbf{Y}) = 0$$

for

$$\mathbf{A_0} \underbrace{\begin{bmatrix} p \\ u_i \\ T \end{bmatrix}}_{\mathbf{Y}} = \begin{bmatrix} \rho \\ \rho u_i \\ \rho e \end{bmatrix}, \quad \mathbf{F}_i(\mathbf{Y}) = \underbrace{\begin{pmatrix} \rho u_i \\ \rho u_i u_j + p \delta_{ij} \\ (\rho e + p) u_i \end{pmatrix}}_{\mathbf{F}^{\text{adv}}} + \underbrace{\begin{pmatrix} 0 \\ -\sigma_{ij} \\ -\rho u_j \sigma_{ij} - kT_{,i} \end{pmatrix}}_{\mathbf{F}^{\text{diff}}}, \quad S(\mathbf{Y}) = -\begin{pmatrix} 0 \\ \rho \mathbf{g} \\ 0 \end{pmatrix}$$

Find
$$Y \in \mathcal{S}^h$$

$$\int_{\Omega} \boldsymbol{v} \cdot \left(\boldsymbol{A_0} \boldsymbol{Y}_{,t} - \boldsymbol{S}(\boldsymbol{Y}) \right) \, \mathrm{d}\Omega - \int_{\Omega} \boldsymbol{v}_{,i} \cdot \boldsymbol{F}_i(\boldsymbol{Y}) \, \mathrm{d}\Omega$$

$$+ \int_{\partial\Omega} \boldsymbol{v} \cdot \boldsymbol{F}_i(\boldsymbol{Y}) \cdot \widehat{\boldsymbol{n}}_i \, \mathrm{d}\partial\Omega$$

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$$+ \int_{\Omega} P(\mathbf{v})^T \left(\mathbf{A_0} \mathbf{Y}_{,t} + \mathbf{F}_{i,i}(\mathbf{Y}) - S(\mathbf{Y}) \right) d\Omega = 0, \ \forall \mathbf{v} \in \mathcal{V}^h$$
SUPG

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$$\mathcal{G}(\mathbf{Y}_{,t},\mathbf{Y})=0$$

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$$\Rightarrow \left[\mathbf{\mathcal{P}}^{T} \mathcal{E}^{T} B^{T} G B \mathcal{E} \mathbf{\mathcal{P}} \right] \begin{bmatrix} \mathbf{Y}_{,t} \\ \mathbf{Y} \end{bmatrix} = 0$$

Implicit Timestepping

Implicit timestepping requires solving:

$$\frac{\mathrm{d}\mathcal{G}(\boldsymbol{Y}_{,t},\boldsymbol{Y})}{\mathrm{d}\boldsymbol{Y}}\Delta\boldsymbol{Y} = \mathcal{G}(\boldsymbol{Y}_{,t},\boldsymbol{Y})$$

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- Finite difference matrix-free approximation:

$$\frac{\mathrm{d}\mathcal{G}(\boldsymbol{Y}_{\!,t},\boldsymbol{Y})}{\mathrm{d}\boldsymbol{Y}}\Delta\boldsymbol{Y}\approx\frac{\mathcal{G}(\boldsymbol{Y}_{\!,t},\boldsymbol{Y}+\epsilon\Delta\boldsymbol{Y})-\mathcal{G}(\boldsymbol{Y}_{\!,t},\boldsymbol{Y})}{\epsilon}$$

- · Pros: Just need a residual evaluation, cheap (in programming and computation)
- · Cons: Approximation, accuracy limited to $\sqrt{\epsilon_{
 m machine}}$

$$\frac{\mathrm{d}\mathcal{G}}{\mathrm{d}\boldsymbol{Y}}\Delta\boldsymbol{Y} = \frac{\mathrm{d}}{\mathrm{d}\boldsymbol{Y}}\left[\mathcal{G}(\boldsymbol{Y}_{t},\boldsymbol{Y})\right]\Delta\boldsymbol{Y}$$

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- Store intermediary data at quadrature points to improve efficiency ("taping")
 - · We store Y, viscous stress, and stabilization perturbation (P(v))
- · Pros: Exact Jacobian matrix-vector product (potentially faster convergence)
- · Cons: More expensive than residual evaluation (but not by too much)

Code Architecture

- PETSc used for handling everything libCEED doesn't
 - $\mathcal{P}, \mathcal{P}^T$ (Partition global-to-local operations)
 - · Time integration, linear, non-linear equation solving
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- libCEED Operator based on user-implemented CeedQFunctions (D)
 - Use different CeedQFunctions for volume vs boundary integrals
 - · Combined into a single **CeedOperator** to represent $\mathcal{G}(Y_t, Y)$



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- 3. PETSc gets $G^G = \mathcal{P}^T G^L$
- 4. PETSc uses $m{G}^G$ to compute new solution value ...or whatever else it wants

Accuracy and Performance of High-Order Scale-Resolving Simulations

Flat Plate Boundary Layer, Zero Pressure Gradient

Problem Description:

- $Re_{\theta} \approx 970$ boundary layer at inflow, $M \approx 0.1$
- Synthetic turbulence generation (STG) used for inflow structures
- · Internal damping layer (IDL) used in STG development region to prevent pressure wave growth
- asdf
- Domain size of $\{27 \times 24 \times 4\}\delta_0$

High-Order Approach

- Test 3 different order elements, Q_1, Q_2, Q_3 tensor-product hexes
- Maintain DOF resolution (DOFs per physical length)

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- DOF resolution for streamwise and spanwise was $\Delta x^+ = 30$ and $\Delta z^+ = 12$
 - For Q_1 , this is about half the resolution required for DNS resolution

Support and References

This work was supported by.... Add in sponsor support (DOE, ECP, etc)

