

# Performance-Portable Implicit Scale-Resolving Compressible Flow Using libCEED

SIAM CSE 2023

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Smead Aerospace  
UNIVERSITY OF COLORADO **BOULDER**

1. libCEED Overview
2. Compressible Fluid Equations in libCEED
3. Viscous Outflow Boundary Conditions
4. Efficient Implicit Timestepping



# libCEED Overview

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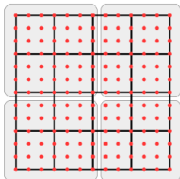
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- Geared toward high-order finite element discretizations



# Finite Element Operator Decomposition

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

global domain  
all (shared) dofs

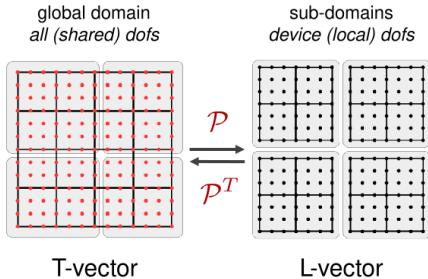


T-vector



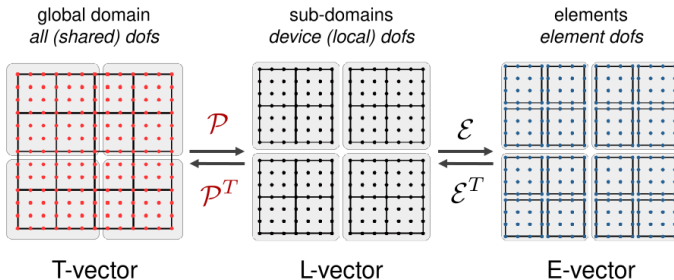
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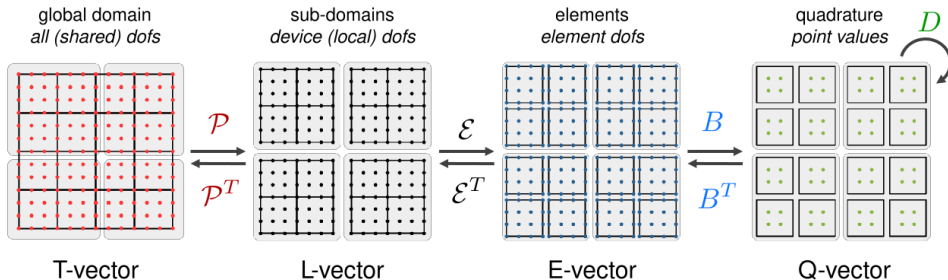
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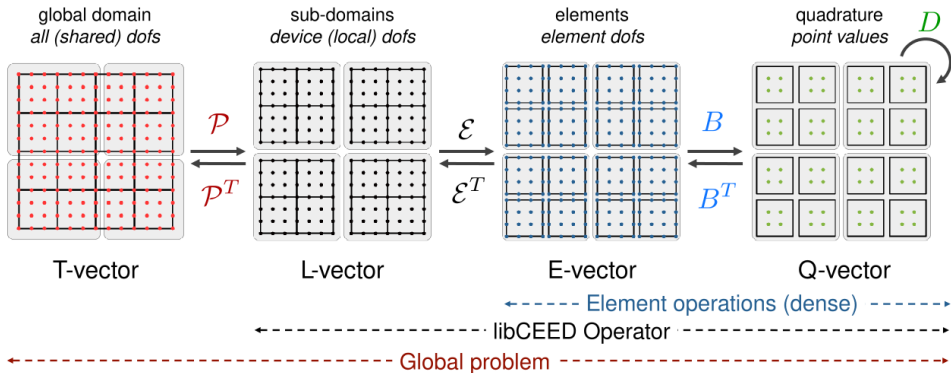
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# Compressible Fluid Equations in libCEED

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$$\mathbf{A}_0 \mathbf{Y}_{,t} + \mathbf{F}_{i,i}(\mathbf{Y}) - S(\mathbf{Y}) = 0$$

for

$$\underbrace{\mathbf{A}_0 \begin{bmatrix} p \\ u_i \\ T \end{bmatrix}}_{\mathbf{Y}} = \begin{bmatrix} \rho \\ \rho u_i \\ \rho e \end{bmatrix}, \quad \mathbf{F}_i(\mathbf{Y}) = \underbrace{\begin{pmatrix} \rho u_i \\ \rho u_i u_j + p \delta_{ij} \\ (\rho e + p) u_i \end{pmatrix}}_{\mathbf{F}_i^{\text{adv}}} + \underbrace{\begin{pmatrix} 0 \\ -\sigma_{ij} \\ -\rho u_j \sigma_{ij} - k T_{,i} \end{pmatrix}}_{\mathbf{F}_i^{\text{diff}}}, \quad S(\mathbf{Y}) = - \begin{pmatrix} 0 \\ \rho \mathbf{g} \\ 0 \end{pmatrix}$$



# Compressible Navier-Stokes for Continuous-Galerkin FEM

Find  $\mathbf{Y} \in \mathcal{S}^h$ ,  $\forall \mathbf{v} \in \mathcal{V}^h$

$$\int_{\Omega} \mathbf{v} \cdot (\mathbf{A}_0 \mathbf{Y}_{,t} - \mathbf{S}(\mathbf{Y})) \, d\Omega + \int_{\Omega} \mathbf{v} \cdot \mathbf{F}_{i,i}(\mathbf{Y}) \, d\Omega$$



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Simplify into residual form:

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



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# Viscous Outflow Boundary Conditions

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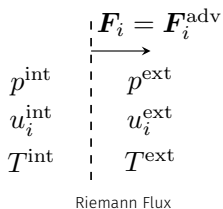


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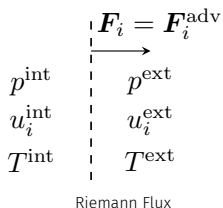
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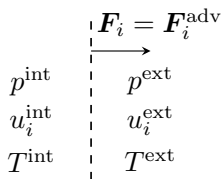


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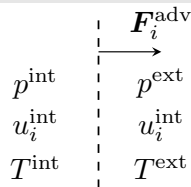
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Riemann Flux



Riemann Outflow Boundary Condition Flux



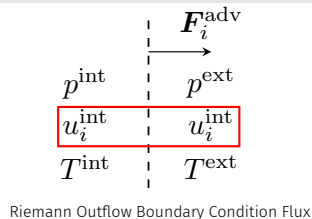
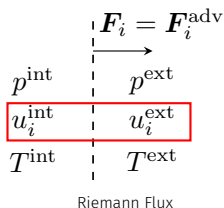


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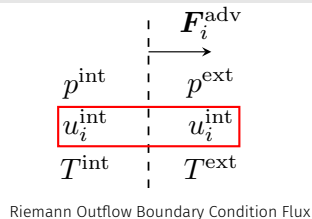
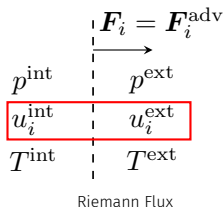


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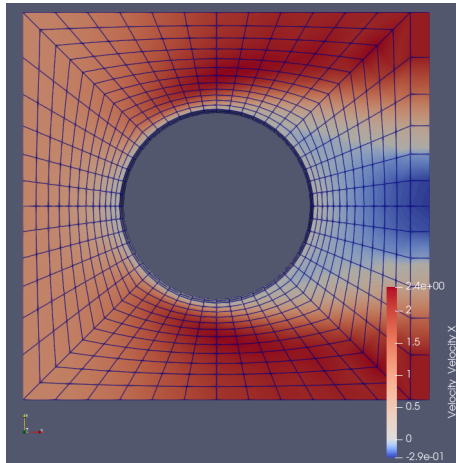
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## Solution

- Calculate  $\mathbf{F}_i^{\text{adv}}$  via Riemann solver, setting  $u_i^{\text{ext}} = u_i^{\text{int}}$
- Calculate  $\mathbf{F}_i^{\text{diff}}$  from solution, then  $\mathbf{F}_i = \mathbf{F}_i^{\text{adv}} + \mathbf{F}_i^{\text{diff}}$



# Viscous Outflow Boundary Conditions Demonstration



- Cylinder in cross flow torture test
- Stable despite recirculation
- Only  $p$  and  $T$  set at boundary



# Efficient Implicit Timestepping

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Implicit timestepping requires solving:

$$\frac{d\mathcal{G}(\mathbf{Y}_t, \mathbf{Y})}{d\mathbf{Y}} \Delta \mathbf{Y} = -\mathcal{G}(\mathbf{Y}_t, \mathbf{Y})$$



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## Bottom Line

Cost of  $\frac{d\mathcal{G}(\mathbf{Y}_t, \mathbf{Y})}{d\mathbf{Y}} \Delta \mathbf{Y}$  dominates implicit timestepping cost



# Jacobian Matrix-Vector Multiply Options

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  - **Cons:** Is large, expensive to store
- Finite difference matrix-free approximation:

$$\frac{d\mathcal{G}(\mathbf{Y}_t, \mathbf{Y})}{d\mathbf{Y}} \Delta \mathbf{Y} \approx \frac{\mathcal{G}(\mathbf{Y}_t, \mathbf{Y} + \epsilon \Delta \mathbf{Y}) - \mathcal{G}(\mathbf{Y}_t, \mathbf{Y})}{\epsilon}$$

- **Pros:** Just need a residual evaluation, cheap (in programming and computation)
- **Cons:** Accuracy limited to  $\sqrt{\epsilon_{\text{machine}}}$ , preconditioning require partial assembly



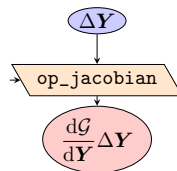
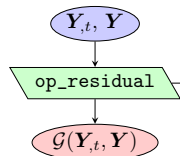
# Exact Matrix-Free Jacobian via CeedOperator

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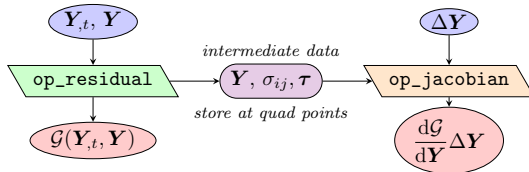
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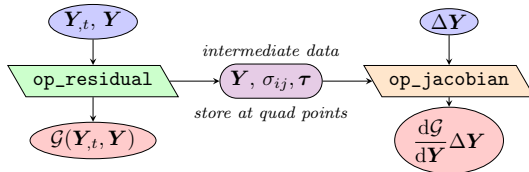
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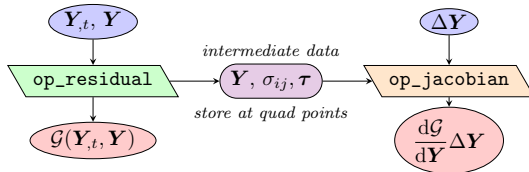


- **Pros:** Exact Jacobian matrix-vector product<sup>1</sup> (potentially faster convergence)
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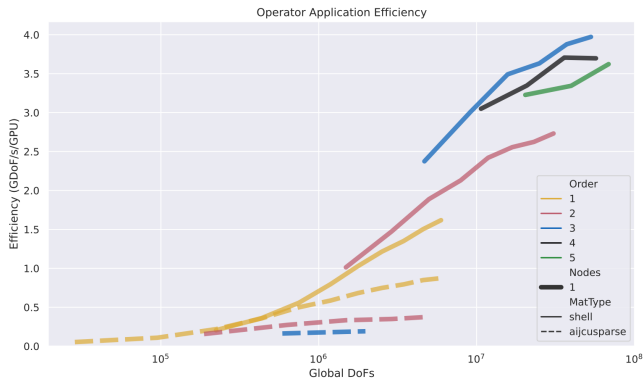
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<sup>1</sup>There's also the option of leaving out terms from the Jacobian. We do this with the fluids code

# Performance Exact Matrix-Free Jacobian via CeedOperator



Matrix-Free Application Comparison (Reproduced from Brown *et al.* 2022). Not from fluids code, but representative of libCEED matrix-free



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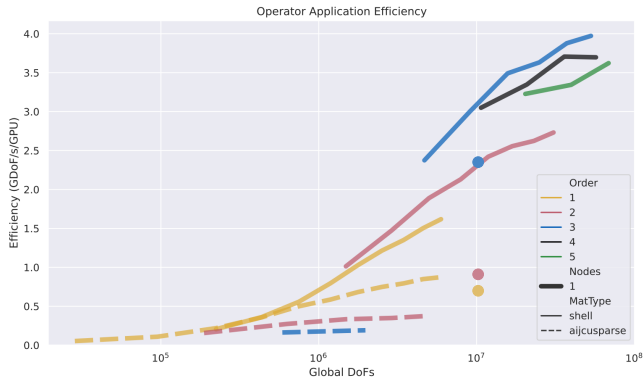


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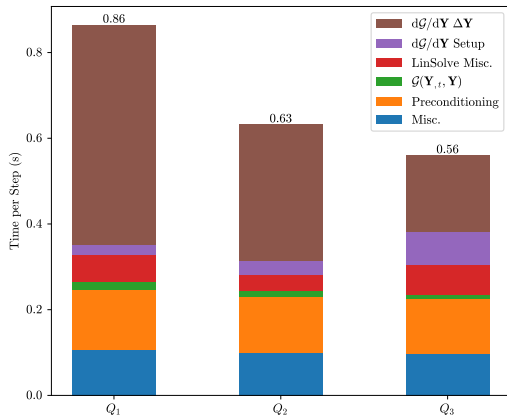


- Fluids run over 2 nodes  $\rightarrow$  network latency present
- Significant increase in efficiency for  $Q_1, Q_2 \rightarrow Q_3$

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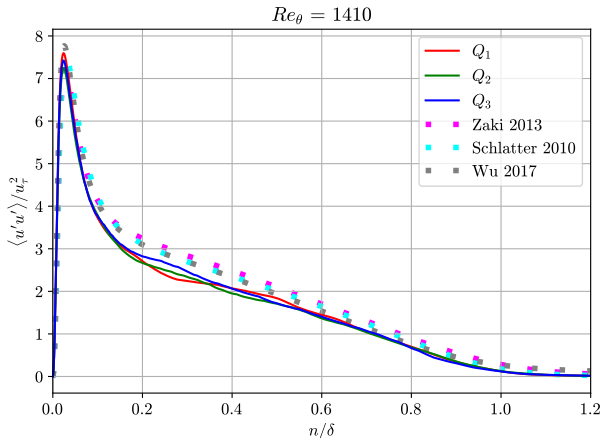
# Fluids Performance Analysis



other stuff



# Brief Results



- Spanwise statistics implemented to verify scale-resolving results
- Results *not* converged, but show realistic stress profiles



# Support and References

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

