

# A Positivity-preserving Strategy for Entropy Stable Discretizations of the Compressible Euler and Navier-Stokes equations

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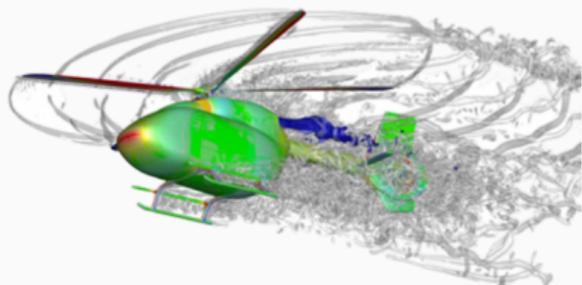
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# High order discontinuous Galerkin methods for PDEs

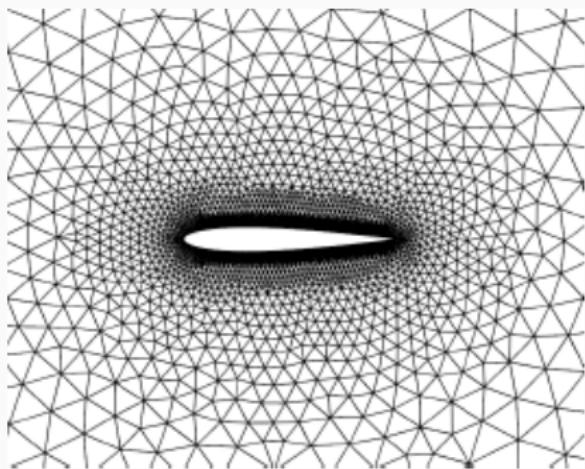
- Physical phenomena governed by PDE: aerospace engineering, nuclear engineering



**Figure 1:** Vortex structures from a helicopter simulation

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- High accuracy computational fluid dynamics on complex geometries



**Figure 1:** Unstructured mesh for NACA 0012 foil

# High order discontinuous Galerkin methods for PDEs

- Physical phenomena governed by PDE: aerospace engineering, nuclear engineering
- High accuracy computational fluid dynamics on complex geometries
- More accurate per degrees of freedom than low order methods (for smooth solutions)

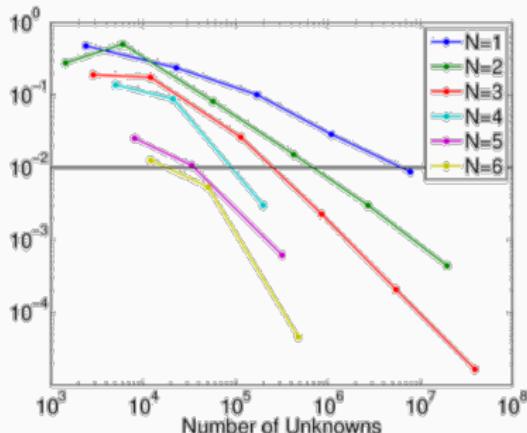


Figure 1: high order methods achieve better accuracy more efficiently

# Compressible Euler and Navier-Stokes equations

- Compressible Euler and Navier-Stokes equations

$$\frac{\partial \mathbf{U}}{\partial t} + \underbrace{\sum_{i=1}^3 \frac{\partial \mathbf{f}_i(\mathbf{U})}{\partial \mathbf{x}_i}}_{\text{inviscid flux}} = \underbrace{\sum_{i=1}^3 \frac{\partial \mathbf{g}_i(\mathbf{U})}{\partial \mathbf{x}_i}}_{\text{viscous flux}}$$

- Entropy variables symmetrizes the viscous fluxes:

$$\sum_{i=1}^d \frac{\partial \mathbf{g}_i}{\partial \mathbf{x}_i} = \sum_{i,j=1}^d \frac{\partial}{\partial \mathbf{x}_i} \left( \mathbf{K}_{ij} \frac{\partial \mathbf{v}}{\partial \mathbf{x}_j} \right),$$

$$\mathbf{K} = \begin{bmatrix} \mathbf{K}_{11} & \dots & \mathbf{K}_{1d} \\ \vdots & \ddots & \vdots \\ \mathbf{K}_{d1} & \dots & \mathbf{K}_{dd} \end{bmatrix} = \mathbf{K}^T, \quad \mathbf{K} \succeq 0$$

## Continuous Entropy Balance

- With convex entropy  $\eta$ , entropy variable  $v = \frac{\partial \eta(u)}{\partial u}$  and entropy potential  $\psi_i$ . We can derive an [entropy balance](#)

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$$\int_{\Omega} \mathbf{v}^T \frac{\partial \mathbf{u}}{\partial t} + \sum_{i=1}^d \int_{\Omega} \mathbf{v}^T \frac{\partial f_i(\mathbf{u})}{\partial x_i} = \sum_{i=1}^d \int_{\Omega} \mathbf{v}^T \frac{\partial g_i(\mathbf{u})}{\partial x_i} \quad \text{Test by } \mathbf{v}$$

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$$\int_{\Omega} \frac{\partial \eta(\mathbf{u})}{\partial t} + \sum_{i=1}^d \int_{\partial \Omega} n_i \left( F_i(\mathbf{u}) - \frac{1}{c_v T} \kappa \frac{\partial T}{\partial x_i} \right) = - \int_{\Omega} \sum_{i,j=1}^d \left( \frac{\partial \mathbf{v}}{\partial x_i} \right)^T \left( \kappa_{ij} \frac{\partial \mathbf{v}}{\partial x_j} \right)$$

Integration by parts and chain rule

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- Loss of chain rule at discrete level (discrete effects, inexact quadrature)  
⇒ Loss of entropy stability

# High order entropy stable DG discretization

- Discretize the variational form of invscid term

$$\int_{\widehat{D}} \frac{\partial f}{\partial x} \vec{l} \xrightarrow{\text{Flux Differencing}} 2(\mathbf{Q} \circ \mathbf{F}_S) \mathbf{1}, \quad (\mathbf{F}_S)_{ij} = f_S(u_i, u_j)$$

Summation-by-parts  $\mathbf{Q} = \mathbf{M}\mathbf{D}$ ,  $\mathbf{Q} + \mathbf{Q}^T = \mathbf{B}$ ,  $\mathbf{Q}\mathbf{1} = 0$

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- Discretize the symmetrized viscous term

$$\frac{\partial g}{\partial x} = \frac{\partial}{\partial x} \left( \kappa \frac{\partial v}{\partial x} \right) \xrightarrow{\text{Discretize}} \begin{cases} (\Theta, \varphi)_\Omega = \left( \frac{\partial v}{\partial x}, \varphi \right)_\Omega + \langle [v] n_i, \varphi \rangle_{\partial\Omega} \\ (\sigma, \eta)_\Omega = (\kappa \Theta, \eta)_\Omega \\ (G_{\text{visc}}, \psi)_\Omega = - \left( \sigma, \frac{\partial \psi}{\partial x} \right)_\Omega + \langle \{\sigma\} n_i, \psi \rangle_{\partial\Omega} \end{cases}$$

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- The scheme is entropy stable

$$\sum_k \left( \frac{\partial \eta(\mathbf{u})}{\partial t}, \mathbf{1} \right)_{D^k} = \sum_k \sum_{i,j=1}^d - (\kappa_{ij} \boldsymbol{\Theta}_j, \boldsymbol{\Theta}_i)_{D^k} \leq 0$$

## Current work: Positivity Limiting for nodal ESDG

- The entropy is well-defined only if densities and pressures are positive.

$$v_1 = (\gamma + 1 - s) - \frac{(\gamma - 1) E}{p}, \quad s = \log \left( \frac{p}{\rho^\gamma} \right)$$

# Current work: Positivity Limiting for nodal ESDG

- Strong shock forms - Negative densities

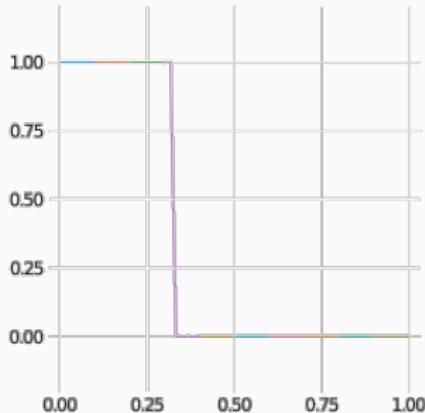


Figure 2: Exact solution

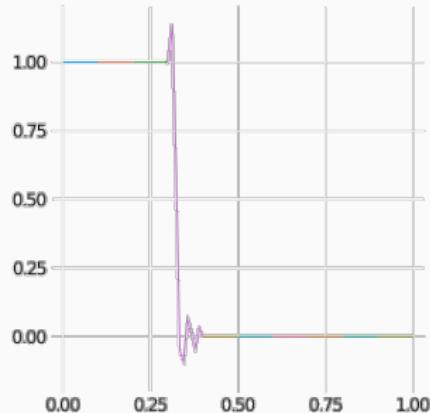


Figure 3: Solution in polynomial basis

- Oscillation by Gibbs phenomenon leads to negative density

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  - Low order positivity-preserving and ESDG:

$$\frac{m_i}{\tau} (u_i^{L,n+1} - u_i^n) + r^{L,i} = 0$$
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- Choose suitable parameter  $l_i \in [0, 1]$  to satisfy positivity

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- $l_i = 1 \implies$  recovers ESDG.
- $l_i = 0 \implies$  recovers low order positivity-preserving scheme.

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- Elementwise limiting parameter

$$l = \min_{i \in D^k} l_i$$

# Positivity preserving discretization

- Low order positivity preserving method could be written as

$$\underbrace{m_i \frac{\partial u}{\partial t} + \sum Q_{ij} (f(u_j) - \sigma_j)}_{\text{low order nodal DG on LGL nodes}} - \underbrace{\sum d_{ij}(u_j - u_i)}_{\text{graph viscosity}} = 0$$

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- Weighted differentiation matrix  $\mathbf{Q}$  is a sparse low order (SBP) operator:

$$\begin{bmatrix} -\frac{1}{2} & \frac{1}{2} & 0 & 0 & \dots & 0 \\ -\frac{1}{2} & 0 & \frac{1}{2} & 0 & \dots & 0 \\ 0 & -\frac{1}{2} & 0 & \frac{1}{2} & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix}$$

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- Define the graph viscosity coefficients:

$$d_{ij} = \max \{ \beta(u_i, u_j, n_{ij}) \|Q_{ij}\|, \beta(u_j, u_i, n_{ji}) \|Q_{ji}\| \}, n_{ij} = Q_{ij}/\|Q_{ij}\|$$

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- Zhang's positivity preserving flux

$$\beta(u_i, u_j, \sigma_i, \sigma_j, n_{ij}) = |n \cdot u| + \frac{1}{2\rho^2 e} \left( \sqrt{\rho^2 (q \cdot n)^2 + 2\rho^2 e \|n \cdot \tau - p n\|^2} + \rho |q \cdot n| \right)$$

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- Positivity preserving under the CFL condition:

$$\tau \leq \frac{\mathbf{m}_i}{2 \sum d_{ij}}$$

## Positivity preserving discretization - Tensor product elements

- Interpretation: subcell Lax-Friedriches type dissipation

The diagram illustrates a 1D finite difference stencil for a central difference operator. A horizontal black line represents a grid. Blue circular markers at the ends represent ghost points. Orange square markers are placed on the grid at positions  $u_{i-1}$ ,  $u_i$ , and  $u_{i+1}$ . Arrows point from each orange square to its corresponding expression above it:

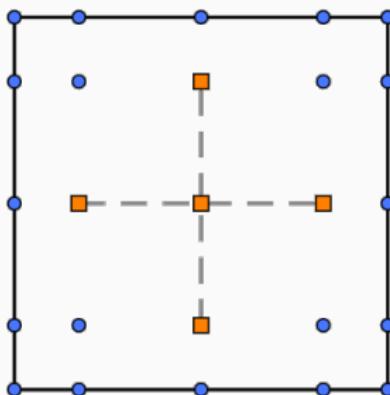
$$\frac{1}{2} (f(u_{i-1}) + f(u_i) - \beta(u_{i-1} - u_i))$$
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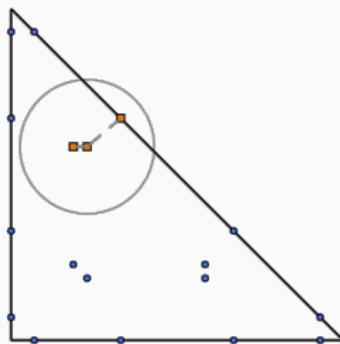
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- Extension to tensor product elements



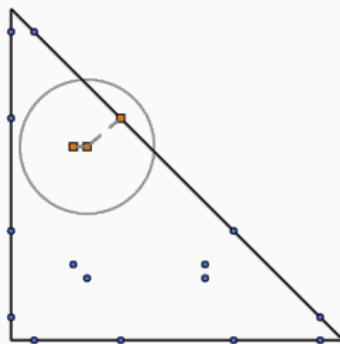
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- Build connectivity graph



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- Generalized sparse low order SBP operator

$$\mathbf{Q}_r^L \mathbf{1} = 0$$

$$\text{s.t. } \left( \frac{\mathbf{Q}_r^L - (\mathbf{Q}_r^L)^T}{2} \right)_{ij} = \begin{cases} 0 & \text{if } A_{ij} = 0 \\ \psi_j - \psi_i & \text{otherwise} \end{cases}.$$

$$\mathbf{Q}_r^L = \frac{\mathbf{Q}_r^L - (\mathbf{Q}_r^L)^T}{2} + \frac{1}{2} \mathbf{E}^T \mathbf{B} \mathbf{E}, \quad \psi^T \mathbf{1} = 0$$

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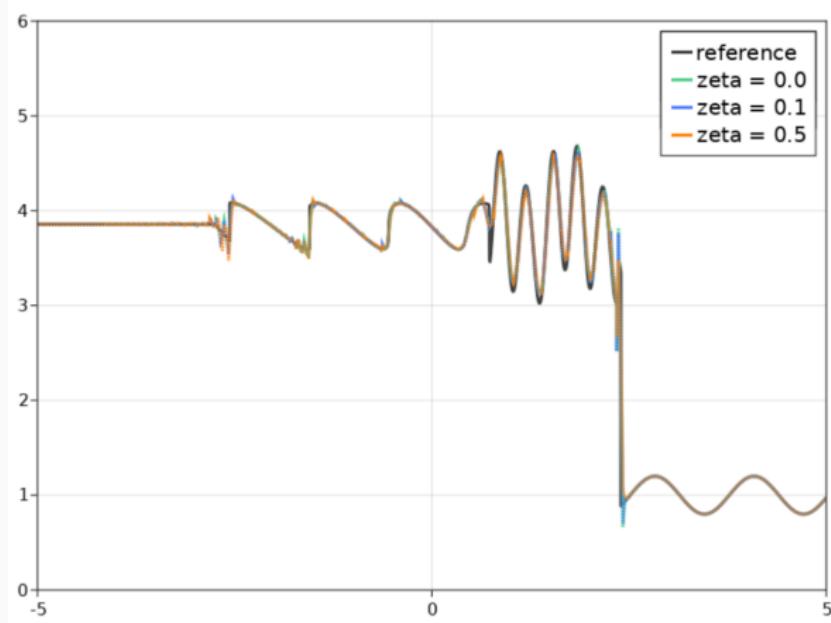
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- Entropy stable with the new dissipation coefficient  $\beta \rightarrow \max \{\beta, \lambda_{\max}\}$
- Preserves high order accuracy (for smooth problems)
- Shock capturing: replace  $l$  with a shock indicator.

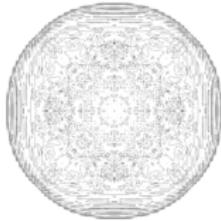
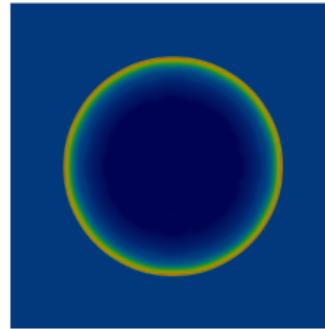
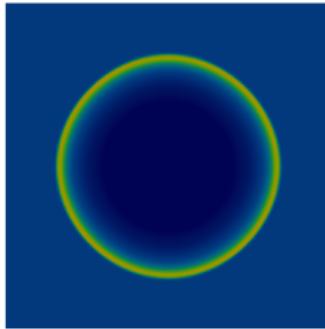
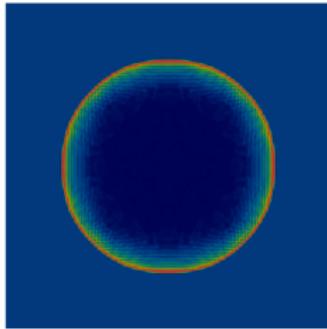
# Shu-Osher shocktube

- $N = 3$ , 128 elements,  $T = 0.2$ , limiting with the nodal bound  
 $\rho > \zeta \rho^L, \rho e > \zeta (\rho e)^L$



# Sedov blast wave

- $N = 3, 100 \times 100$  elements, limiting with the nodal bound  
 $\rho > \zeta \rho^L, \rho e > \zeta (\rho e)^L$



$\zeta = 0.1$ , without shock capturing

$\zeta = 0.1$ , with shock capturing

$\zeta = 0.5$ , without shock capturing

# Double Mach Reflection - Compressible Euler

- $N = 3$ ,  $1000 \times 250$  elements,  $T = 0.2$ , limiting with the nodal bound  
 $\rho > \zeta \rho^L, \rho e > \zeta (\rho e)^L$

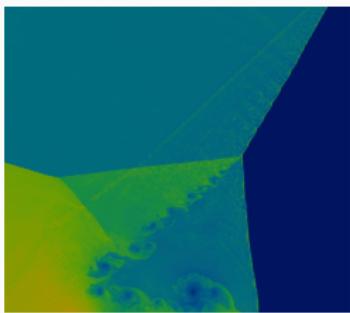


Figure 4:  $\zeta = 0.1$  without shock capturing

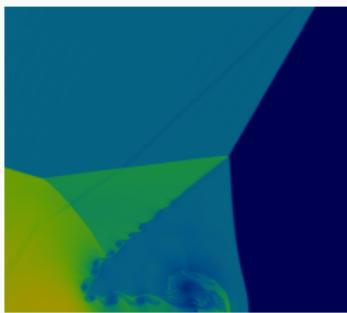


Figure 5:  $\zeta = 0.1$  with shock capturing

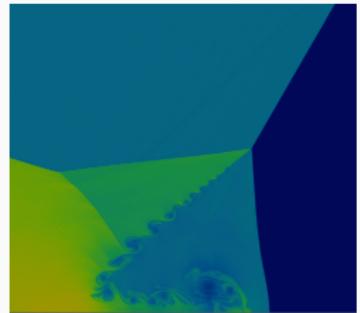


Figure 6:  $\zeta = 0.5$  without shock capturing

# Double Mach Reflection - Compressible Euler

- $N = 3$ ,  $1000 \times 250$  elements,  $T = 0.2$ , limiting with the nodal bound  
 $\rho > \zeta \rho^L, \rho e > \zeta (\rho e)^L$

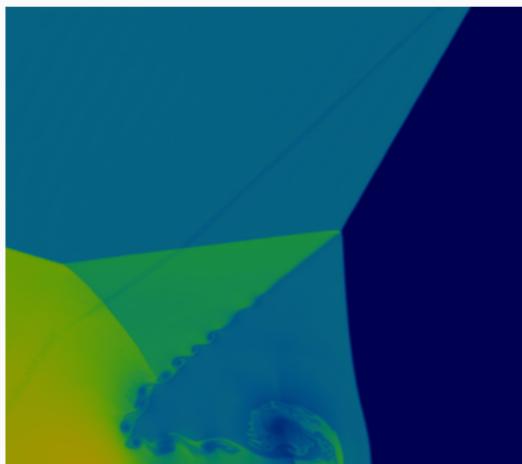


Figure 4: (ESDG)  $\zeta = 0.1$  with shock capturing

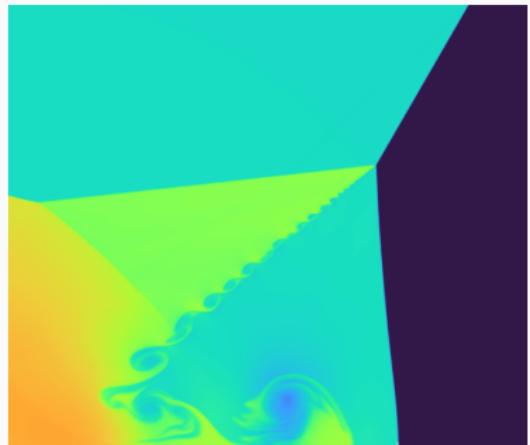


Figure 5: (Pazner, Sprase IDP, enforced minimum entropy principle)  $N = 3, 2400 \times 600$  elements,  
 $T \approx 0.275$

# Daru-Tenaud shocktube - Compressible Navier-Stokes

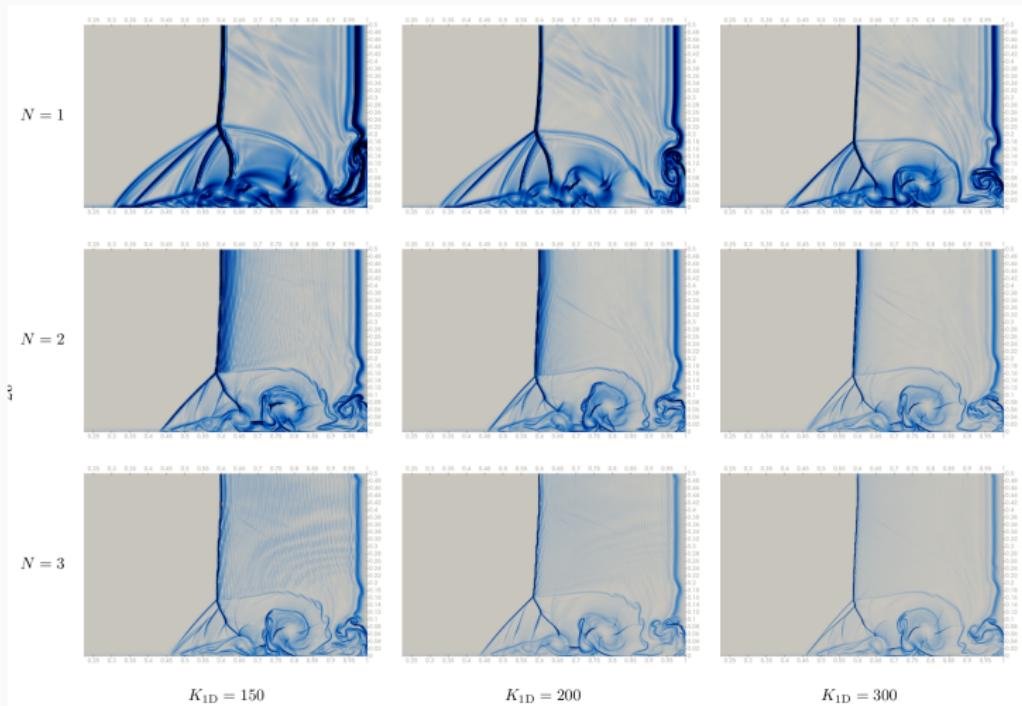


Figure 6:  $Re = 1000$ , uniform quad mesh,  $\eta = 0.1$  without shock capturing

# Daru-Tenaud shocktube - Compressible Navier-Stokes

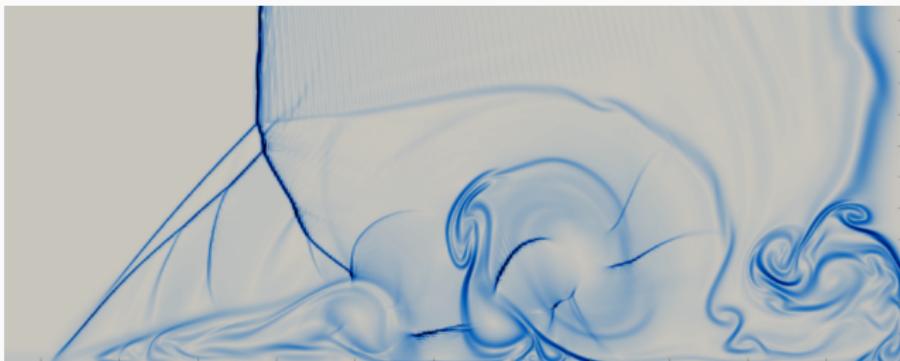


Figure 6:  $Re = 1000, T = 1.0$  Result:  $N = 3,300 \times 150$  uniform quad elements without shock capturing

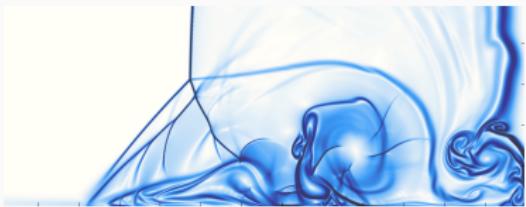


Figure 7: Invariant domain discretization,  
2048  $\times$  1024 elements

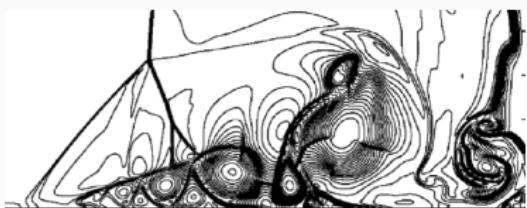


Figure 8: WENO5-RK4, 1000  $\times$  500 grid points

# Daru-Tenaud shocktube - Compressible Navier-Stokes



Figure 6:  $Re = 1000, T = 1.0$  Result:  $N = 3,300 \times 150$  uniform quad elements without shock capturing

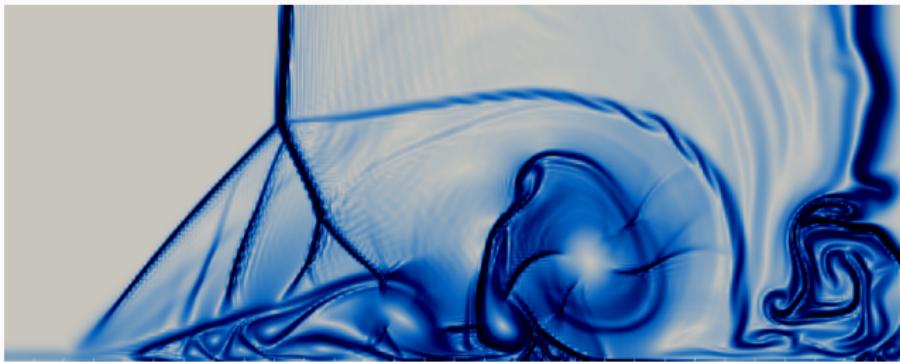


Figure 7:  $Re = 1000, T = 1.0$  Result:  $N = 3,300 \times 150$  uniform quad elements with shock capturing

## Summary and future works

- We present a positivity limiting strategy for nodal ESDG based on graph viscosity.
- Future work: Positivity limiting for modal ESDG. Implicit timestepping.

Thank you!