

# Entropy stable reduced order modeling of nonlinear conservation laws using high order DG methods

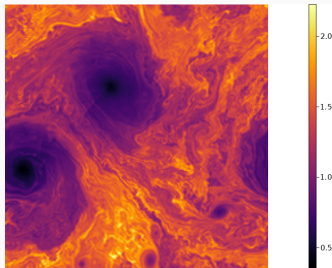
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Ray Qu, Jesse Chan

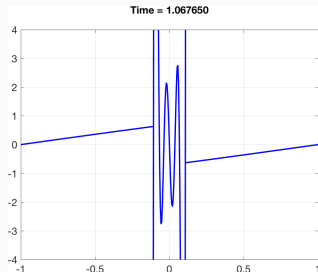
Dept. of Computational Applied Mathematics and Operations Research  
Rice University

2024 Finite Element Rodeo

# Why entropy stable reduced order modeling (ROM)?



(a) Kelvin-Helmholtz instability



(b) 1D Burgers'

- Extreme-scale nonlinear evaluations in high-fidelity simulations.
- ROM enables efficient many-query contexts.
- High order methods blow up around shocks and turbulence for both full order models (FOMs) and ROMs.

# Entropy stability for nonlinear problems

- Energy balance for **nonlinear** conservation laws (Burgers', shallow water, compressible Euler + Navier-Stokes).

$$\frac{\partial \mathbf{u}}{\partial t} + \sum_i^d \frac{\partial \mathbf{f}_i(\mathbf{u})}{\partial x_i} = 0. \quad (1)$$

- Continuous entropy inequality: convex **entropy** function  $S(\mathbf{u})$ , "entropy potential"  $\psi(\mathbf{u})$ , entropy variables  $\mathbf{v}(\mathbf{u})$

$$\int_{\Omega} \mathbf{v}^T \left( \frac{\partial \mathbf{u}}{\partial t} + \sum_i^d \frac{\partial \mathbf{f}_i(\mathbf{u})}{\partial x_i} \right) = 0, \quad \boxed{\mathbf{v}(\mathbf{u}) = \frac{\partial S}{\partial \mathbf{u}}}$$
$$\implies \int_{\Omega} \frac{\partial S(\mathbf{u})}{\partial t} + \sum_i^d (\mathbf{v}^T \mathbf{f}_i(\mathbf{u}) - \psi_i(\mathbf{u})) \Big|_{-1}^1 \leq 0.$$

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# Talk outline

1. Full order model (FOM) construction
2. Reduced order model (ROM) construction
3. Numerical Experiments

# Full order model (FOM) construction

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# Entropy stable high order DG formulation

- A global DG formulation of (1) is

$$\mathbf{M} \frac{d\mathbf{u}}{dt} + \sum_{i=1}^d (2(\mathbf{Q}^i \circ \mathbf{F}^i) \mathbf{1} + \mathbf{E}^T \mathbf{B}^i (\mathbf{f}^{i,*} - \mathbf{f}^i(\mathbf{u}))) = \text{dissipation},$$

where  $(\mathbf{F}^i)_{j,k} = \mathbf{f}^i(\mathbf{u}_j, \mathbf{u}_k)$  is a matrix of nonlinear flux evaluations,  $\mathbf{E}$  is a boundary extraction matrix, and  $\mathbf{Q}^i$  is a **global summation-by-parts** (SBP) operator  $(\mathbf{Q}^i + (\mathbf{Q}^i)^T = \mathbf{E}^T \mathbf{B}^i \mathbf{E})$  with zero row sum  $(\mathbf{Q}^i \mathbf{1} = \mathbf{0})$ .

- We can prove a **semi-discrete entropy stability** condition

$$\mathbf{1}^T \mathbf{M} \frac{dS(\mathbf{u})}{dt} \leq 0.$$

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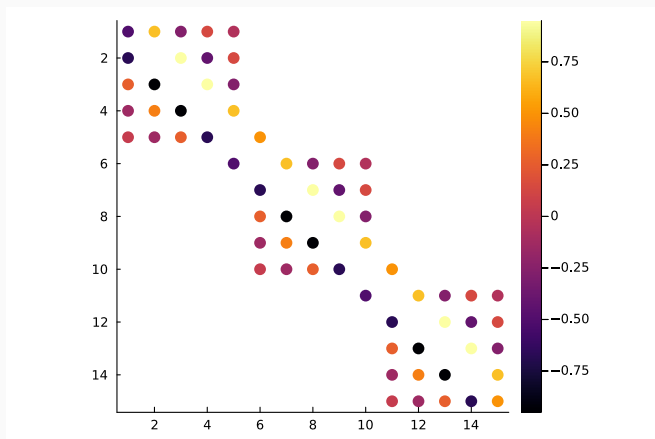
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# An example of a global SBP operator



Spy plot of  $Q^1$  (1D domain, 3 elements with 5 nodes in each)

# Reduced order model (ROM) construction

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# Galerkin projection ROM

- Galerkin projection ROM ( $\mathbf{V}_N$  is the **POD** basis):

$$\mathbf{M}_N \frac{d\mathbf{u}_N}{dt} + \sum_{i=1}^d (2\mathbf{V}_N^T (\mathbf{Q}^i \circ \mathbf{F}^i) \mathbf{1} + \mathbf{V}_b^T \mathbf{B}^i (\mathbf{f}^{i,*} + \mathbf{f}^i(\mathbf{u}_N))) = \mathbf{0},$$

$$\mathbf{u} \approx \mathbf{V}_N \mathbf{u}_N \text{ and } \mathbf{V}_b = \mathbf{E} \mathbf{V}_N .$$

- To achieve entropy stability, use entropy projection

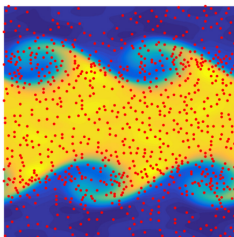
$$\tilde{\mathbf{u}} = \mathbf{u}(\mathbf{V}_N \mathbf{V}_N^\dagger \mathbf{v}(\mathbf{V}_N \mathbf{u}_N)) = \mathbf{u}(\tilde{\mathbf{v}}), \quad (\mathbf{F}^i)_{j,k} = \mathbf{f}^i(\tilde{\mathbf{u}}_j, \tilde{\mathbf{u}}_k).$$

- Still has **high** computational cost! Needs **hyper-reduction**.  
Hyper-reduction on volume and boundary terms are **independent**.

# Hyper-reduction on volume terms

Main idea: find a set of volume nodes and positive weights  $(I_v, \mathbf{w}_v)$

$$\mathbf{V}_N^T g(\mathbf{V}_N \mathbf{u}_N) \approx \mathbf{V}_N(I_v, :)^T \text{diag}(\mathbf{w}_v) g(\mathbf{V}_N(I_v, :)\mathbf{u}_N).$$



We use the greedy algorithm for empirical cubature.

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Chan (2020). *Entropy stable reduced order modeling of nonlinear conservation laws.*

Hernández et al. (2017) *Dimensional hyper-reduction of nonlinear finite element models via empirical cubature.*

## Hyper-reduction on boundary terms

Assume hyper-reduction on **boundary**  $(I_b, \mathbf{w}_b)$  and denote

$$\bar{\mathbf{B}}^i = \text{diag}(\mathbf{n}^i) \text{diag}(\mathbf{w}_b).$$

Suppose  $\mathbf{V}^i$  is some ROM basis matrix for the  $i$ th coordinate,

$$\mathbf{1}^T \mathbf{Q}^i \mathbf{V}^i = \mathbf{1}^T \bar{\mathbf{B}}^i \mathbf{V}^i (I_b, :),$$

is a matrix form of the fundamental theorem of calculus, and we need to **enforce this equality** to preserve entropy stability.

A natural way to do this: Carathéodory's pruning.

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# Carathéodory's pruning

**Carathéodory's theorem** states that, for any  $M$ -point positive quadrature rule exact on space  $\mathbf{V}$  with  $\dim(\mathbf{V}) = N$ , we can always generate a new  $N$ -point interpolatory positive rule to **preserve all moments**.

In our case,

$$\mathbf{1}^T \mathbf{B}^i \mathbf{V}^i = \int \phi_j^i(\mathbf{x}_k) \mathbf{n}^i = \sum_{k=1}^M \mathbf{w}_{b,j} \mathbf{n}_j^i \phi_j(\mathbf{x}_k).$$

In practice, we **concatenate all dimensions**

$$[\text{diag}(\mathbf{n}^1) \mathbf{V}^1 \quad \dots \quad \text{diag}(\mathbf{n}^d) \mathbf{V}^d],$$

which yields  $\mathcal{O}(dN)$  hyper-reduced positive boundary weights  $\mathbf{w}_b$  and node indices  $I_b$ .

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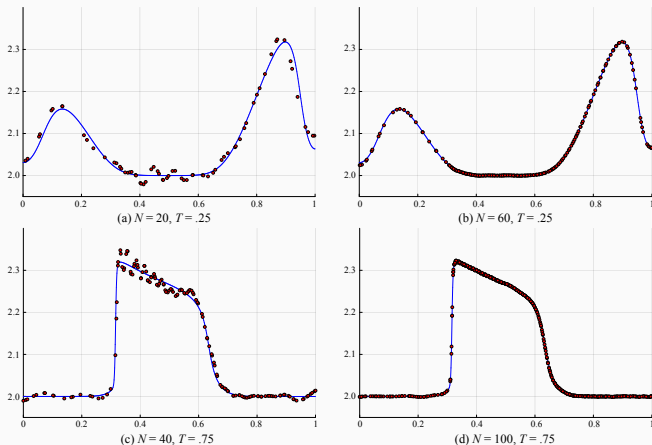
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# Numerical Experiments

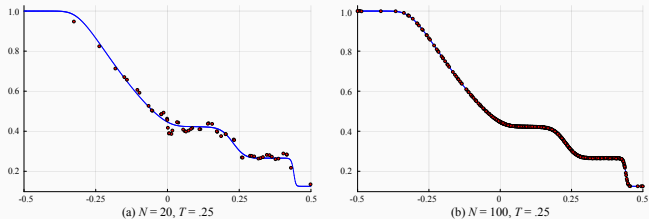
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# Example 1 - 1D Euler



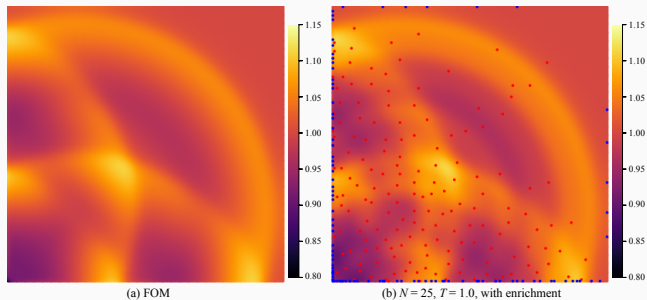
**Figure 1:** 1D Compressible Euler (reflective wall). FOM dim: 2048. Viscosity:  $2 \times 10^{-4}$ . Runtime  $T = .75$ .

## Example 2 - sod shock tube



**Figure 2:** FOM dim: 2048. Viscosity:  $2 \times 10^{-3}$ . Runtime  $T = .25$ .

## Example 3 - Gaussian



**Figure 3:** 2D compressible Euler (reflective wall). FOM dim: 6400. Viscosity:  $1 \times 10^{-3}$ . Run time  $T = 1.0$ . Boundary hyper-reduced by Carathéodory pruning.

# Summary and acknowledgement

In this work, we

- present an entropy stable reduced order modeling of nonlinear conservation laws based on high order DG methods.
- develop structure-preserving hyper-reduction techniques (Carathéodory pruning) which preserve entropy stability.

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