

Invariant domain preserving discretization-independent schemes and convex limiting for hyperbolic systems

J.-L. Guermond, B. Popov

Department of Mathematics
Texas A&M University

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Outline



Hyperbolic systems



The PDEs

- Hyperbolic system

$$\begin{aligned}\partial_t \mathbf{u} + \nabla \cdot \mathbf{f}(\mathbf{u}) &= \mathbf{S}(\mathbf{u}), & (\mathbf{x}, t) \in D \times \mathbb{R}_+, \\ \mathbf{u}(\mathbf{x}, 0) &= \mathbf{u}_0(\mathbf{x}), & \mathbf{x} \in D.\end{aligned}$$

- D open polyhedral domain in \mathbb{R}^d .
- $\mathbf{f} \in C^1(\mathbb{R}^m; \mathbb{R}^{m \times d})$, the flux. $\mathbf{S} : \mathbb{R}^m \rightarrow \mathbb{R}^m$ source term
- \mathbf{u}_0 , admissible initial data.
- Periodic BCs or \mathbf{u}_0 has compact support (to simplify BCs)



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Formulation of the problem

Assumptions for the hyperbolic part

- \exists admissible set $\mathcal{A} \subset \mathbb{R}^m$ s.t. for all $(\mathbf{u}_l, \mathbf{u}_r) \in \mathcal{A}$ the 1D Riemann problem

$$\partial_t \mathbf{v} + \partial_x (\mathbf{f}(\mathbf{v}) \mathbf{n}) = 0, \quad \mathbf{v}(x, 0) = \begin{cases} \mathbf{u}_l & \text{if } x < 0 \\ \mathbf{u}_r & \text{if } x > 0. \end{cases}$$

has a unique “entropy” solution $\mathbf{u}(\mathbf{u}_l, \mathbf{u}_r)(x, t)$ for all $\mathbf{n} \in \mathbb{R}^d$, $\|\mathbf{n}\|_{\ell^2} = 1$.

- There exists an invariant set $\mathcal{B} \subset \mathcal{A} \subset \mathbb{R}^m$, i.e.,

$$((\mathbf{u}_l, \mathbf{u}_r) \in \mathcal{B}) \Rightarrow (\mathbf{v}(\mathbf{u}_l, \mathbf{u}_r)(x, t) \in \mathcal{B}, \forall t \geq 0, \forall x \in \mathbb{R})$$

- \mathcal{B} is convex.

Assumptions for source term

- \mathcal{B} is invariant w.r.t. \mathbf{S}

$$\exists \Delta t_0, \quad (\mathbf{u} \in \mathcal{B}) \Rightarrow (\mathbf{u} + \Delta t \mathbf{S}(\mathbf{u}) \in \mathcal{B}, \quad \forall \Delta t \leq \Delta t_0)$$



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Global strategy



Global strategy

Strategy for high-order method

- ① Construct low-order invariant domain preserving method.
- ② Construct a high-order scheme that may not be invariant domain preserving.
- ③ Apply convex limiting with correct bounds inferred from low-order solution to get a high-order method that is invariant domain preserving.



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Outline



Invariant domain
preserving
discretization



Discretization-independent invariant domain approximation

- Current time t^n ; time step Δt
- Discrete scalar space $X_h = \text{span}\{\varphi_i\}_{i \in \mathcal{V}}$
- Vector-valued space $\mathbf{X}_h = (X_h)^m$
- $\mathbf{u}_h(\cdot, t^n)$ approximated by $\sum_{i \in \mathcal{V}} \mathbf{U}_i^n \varphi_i$, $\mathbf{U}_i^n \in \mathbb{R}^m$
- Assume high-order approximation of $\mathbf{u}(\cdot, t^{n+1})$ as follows:

$$\frac{m_i}{\Delta t} (\mathbf{U}_i^{G,n+1} - \mathbf{U}_i^n) + \sum_{j \in \mathcal{I}(i)} \mathbf{f}(\mathbf{U}_j^n) \mathbf{c}_{ij} = m_i \mathbf{S}_i^n,$$

- Here $\mathcal{I}(i)$ is the local stencil. List of dofs interacting with i .
- Think of above approximation as Galerkin approximation, or centered approximation, or inviscid approximation. (No stabilization, no artificial viscosity.)



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Discretization-independent invariant domain approximation

Assumptions

- Lumped mass is positive

$$m_i > 0, \quad \forall i \in \mathcal{V}$$

- Conservation

$$\mathbf{c}_{ij} = -\mathbf{c}_{ji}, \quad \text{and} \quad \sum_{j \in \mathcal{I}(i)} \mathbf{c}_{ij} = 0.$$

Lemma

- Conservation

$$\sum_{i \in \mathcal{V}} m_i \mathbf{U}_i^{G,n+1} = \sum_{i \in \mathcal{V}} m_i \mathbf{U}_i^n.$$



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Discretization-independent invariant domain approximation

Key idea

- Consider undirected graph $(\mathcal{V}, \mathcal{E})$, $((i, j) \in \mathcal{E}) \Leftrightarrow (i \in \mathcal{I}(j), \text{ and } j \in \mathcal{I}(i))$.
- Introduce artificial viscosity using graph Laplacian on $(\mathcal{V}, \mathcal{E})$.

Guaranteed Maximum Speed Graph Viscosity (GMS-GV)

$$\frac{m_i}{\Delta t}(\mathbf{U}_i^{\mathbb{L},n+1} - \mathbf{U}_i^n) + \sum_{j \in \mathcal{I}(i)} \mathbf{f}(\mathbf{U}_j^n) \mathbf{c}_{ij} - \sum_{j \in \mathcal{I}(i) \setminus \{i\}} d_{ij}^{\mathbb{L},n} (\mathbf{U}_j^n - \mathbf{U}_i^n) = m_i \mathbf{S}(\mathbf{U}_i^n),$$



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GMS-GV scheme

Assumptions

- Graph viscosity is positive

$$d_{ij}^{\text{L},n} \geq 0, \quad i \neq j$$

- Conservation. Assume

$$d_{ij}^{\text{L},n} = d_{ji}^{\text{L},n}.$$

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- Symmetry $c_{ij} = -c_{ji}$ and $d_{ij} = d_{ji}$ implies conservation

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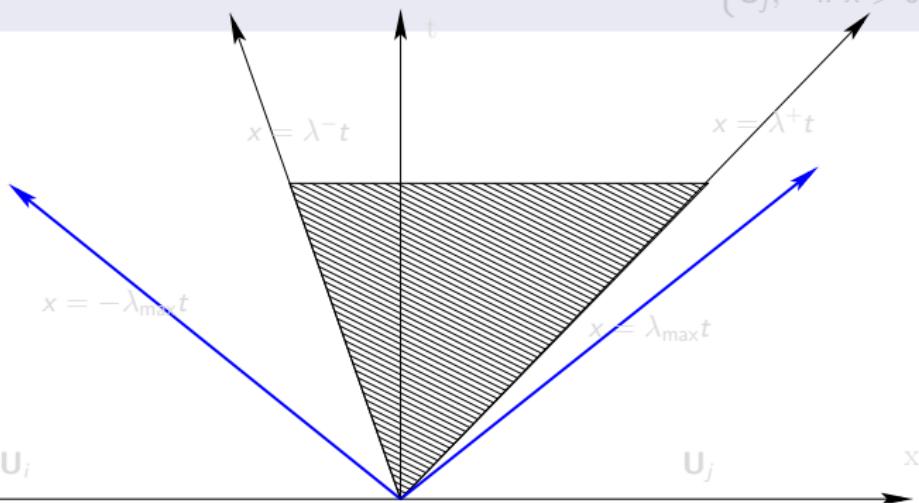


GMS-GV scheme

Key ideas (Lax, Perthame, Tadmor, Shu, etc.)

- Let $i \in \mathcal{V}, j \in \mathcal{I}(i)$.
- Define unit vector $\mathbf{n}_{ij} := \frac{\mathbf{c}_{ij}}{\|\mathbf{c}_{ij}\|_{\ell^2}}$
- Introduce $\lambda_{\max}(\mathbf{n}_{ij}, \mathbf{U}_i^n, \mathbf{U}_j^n)$, upper bound on maximum wave speed in 1D Riemann problem:

$$\partial_t \mathbf{u} + \partial_x (\mathbf{f}(\mathbf{u}) \mathbf{n}) = 0, \quad (x, t) \in \mathbb{R} \times \mathbb{R}_+, \quad \mathbf{u}(x, 0) = \begin{cases} \mathbf{U}_i^n, & \text{if } x < 0 \\ \mathbf{U}_j^n, & \text{if } x > 0, \end{cases}$$

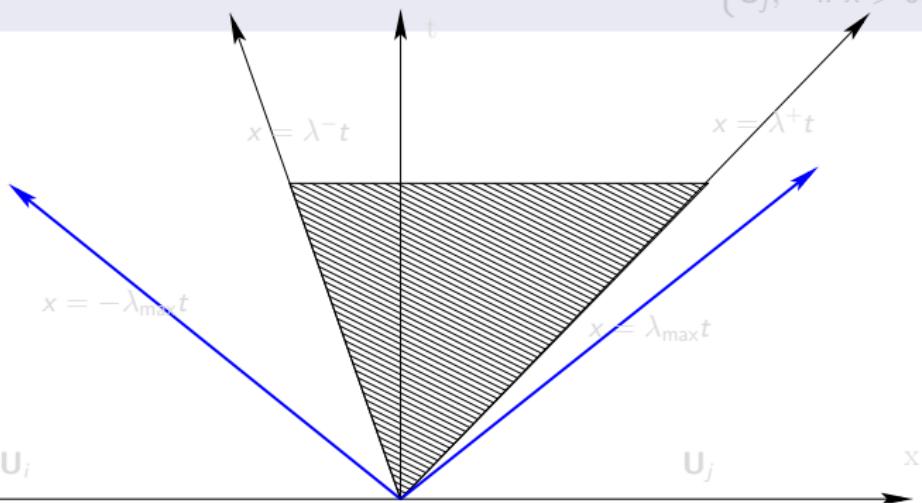


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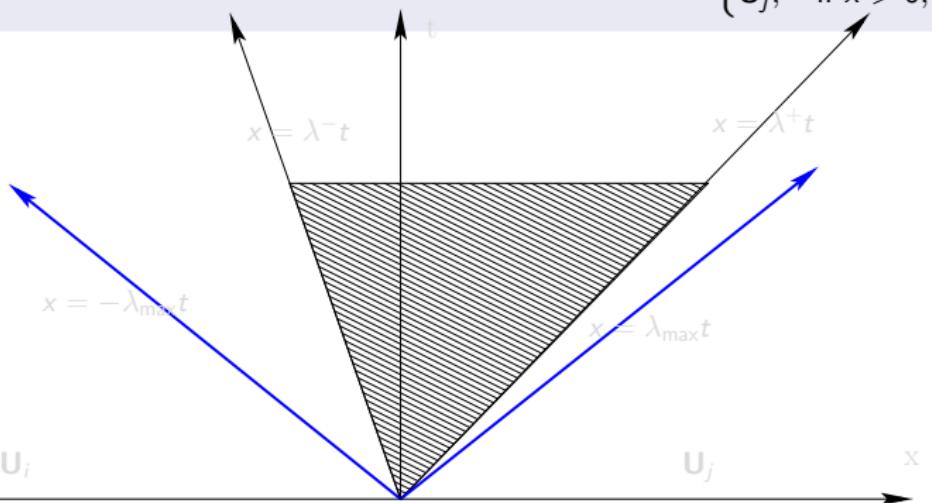


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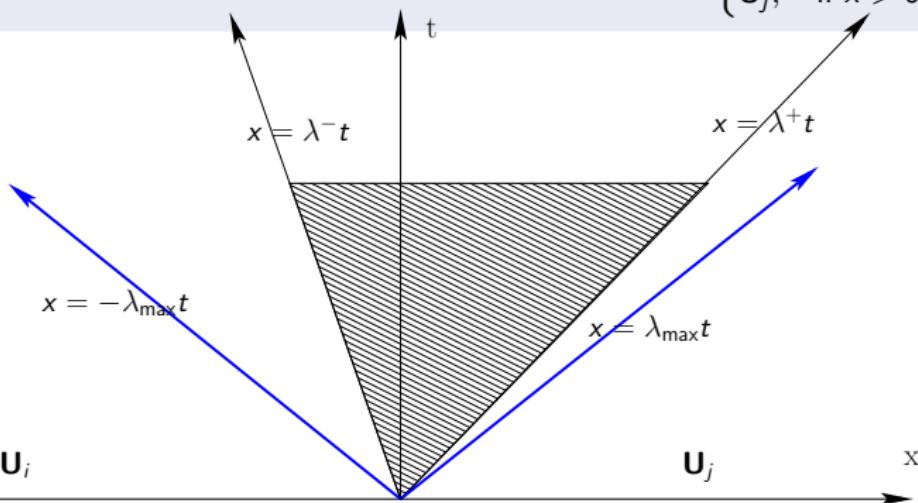


GMS-GV scheme

Key ideas (Lax, Perthame, Tadmor, Shu, etc.)

- Let $i \in \mathcal{V}, j \in \mathcal{I}(i)$.
- Define unit vector $\mathbf{n}_{ij} := \frac{\mathbf{c}_{ij}}{\|\mathbf{c}_{ij}\|_{\ell^2}}$
- Introduce $\lambda_{\max}(\mathbf{n}_{ij}, \mathbf{U}_i^n, \mathbf{U}_j^n)$, **upper bound** on maximum wave speed in 1D Riemann problem:

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GMS-GV scheme

Key ideas

- Introduce

$$d_{ij}^{\mathbb{L},n} := \max(\lambda_{\max}(\mathbf{n}_{ij}, \mathbf{U}_i^n, \mathbf{U}_j^n) \|\mathbf{c}_{ij}\|_{\ell^2}, \lambda_{\max}(\mathbf{n}_{ji}, \mathbf{U}_j^n, \mathbf{U}_i^n) \|\mathbf{c}_{ji}\|_{\ell^2}).$$

- Introduce auxiliary states (“bar states”)

$$\overline{\mathbf{U}}_{ij}^n := \frac{1}{2}(\mathbf{U}_i^n + \mathbf{U}_j^n) - (\mathbf{f}(\mathbf{U}_j^n) - \mathbf{f}(\mathbf{U}_i^n)) \frac{\mathbf{c}_{ij}}{2d_{ij}^{\mathbb{L},n}}.$$

Lemma (Invariance of the auxiliary states, JLG+BP (2016-2018))

Let $\mathcal{B} \subset \mathcal{A}$ be *any* convex invariant set such that $\mathbf{U}_i^n, \mathbf{U}_j^n \in \mathcal{B}$. Then, the state $\overline{\mathbf{U}}_{ij}^n$ belongs to \mathcal{B} .



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GMS-GV scheme

Key observation

- Convex combination

$$\mathbf{U}_i^{\text{L},n+1} = \frac{1}{2} \left(\left(1 + 4\Delta t \frac{d_{ii}^{\text{L},n}}{m_i} \right) \mathbf{U}_i^n + \sum_{j \in \mathcal{I}(i) \setminus \{i\}} \frac{4\Delta t d_{ij}^{\text{L},n}}{m_i} \bar{\mathbf{U}}_{ij}^n \right) + \frac{1}{2} (\mathbf{U}_i^n + 2\Delta t \mathbf{S}(\mathbf{U}_i^n)).$$

Theorem (GMS-GV, Local invariance, JLG+BP (2016-2018))

Let $n \geq 0$ and let $i \in \mathcal{V}$. Assume that Δt is small enough so that

$1 - 4\Delta t \frac{\sum_{j \in \mathcal{I}(i) \setminus \{i\}} d_{ij}^{\text{L},n}}{m_i} \geq 0$ and $2\Delta t \leq \tau_0$. Let $\mathcal{B} \subset \mathcal{A}$ be a convex invariant set such that $\mathbf{U}_j^n \in \mathcal{B}$ for all $j \in \mathcal{I}(i)$, then $\mathbf{U}_i^{\text{L},n+1} \in \mathcal{B}$.

- This is the generalization of the maximum principle for any discretization (any mesh), in any space dimension, for any hyperbolic system.
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Approximation (time and space)

Theorem (GMS-GV, Entropy inequality, JLG+BP (2016-2018))

Let $n \geq 0$ and $i \in \mathcal{V}$. Assume also that the local CFL condition holds

$1 - 2\Delta t \frac{\sum_{j \in \mathcal{I}(i) \setminus \{i\}} d_{ij}^{\text{L},n}}{m_i} \geq 0$ and $2\Delta t \leq \tau_0$, then the following local entropy inequality holds true for any entropy pair (η, \mathbf{q}) of the system:

$$\begin{aligned} \frac{m_i}{\Delta t} (\eta(\mathbf{U}_i^{\text{L},n+1}) - \eta(\mathbf{U}_i^n)) + \sum_{j \in \mathcal{I}(i)} \mathbf{q}(\mathbf{U}_j^n) \mathbf{c}_{ij} - d_{ij}^{\text{L},n} (\eta(\mathbf{U}_j^n) - \eta(\mathbf{U}_i^n)) \\ \leq m_i \mathbf{S}(\mathbf{U}_i^n) \cdot \nabla \eta(\mathbf{U}_i^{\text{L},n+1}). \end{aligned}$$

Literature: Lax (1954), Chueh, Conley, Smoller (1973), Hoff (1979, 1985), Perthame-Shu (1996), Frid (2001), Zhang-Shu (2010), (2011), ...



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Example 1: Continuous finite elements

Example (Continuous FE)

- $\{\mathcal{T}_h\}_{h>0}$ shape regular conforming mesh sequence
- Shape functions $\{\varphi_1, \dots, \varphi_I\}$ + partition of unity ($\sum_{j \in \{1:I\}} \varphi_j = 1$)
- Ex: \mathbb{P}_1 , \mathbb{Q}_1 , Bernstein polynomials (any degree)
- Lumped mass matrix ($m_i = \sum_{j \in \mathcal{I}(S_i)} \int_D \varphi_i \varphi_j \, dx$)

$$m_i := \int_D \varphi_i \, dx .$$

- “Stiffness” matrix for transport

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Example 3: Finite Volumes

Example

Finite volumes

$$m_i := |K_i|, \quad \mathbf{c}_{ij} := \frac{|\Gamma_{ij}|}{2} \mathbf{n}_{ij}, \quad \forall j \in \mathcal{I}(i) \setminus \{i\}, \quad \mathbf{c}_{ii} := \mathbf{0}.$$

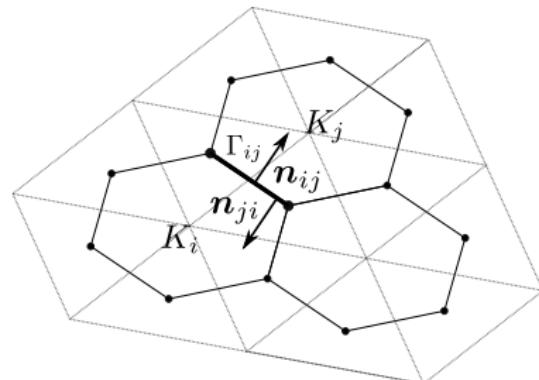
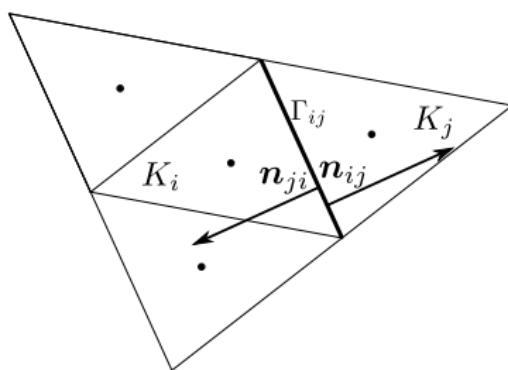


Figure: Finite volume patch arising from a cell-centered discretization (left) and a vertex-centered discretization (right).



Outline



High-order viscosity



High-order viscosity (II): Commutator-based entropy viscosities

High-order viscosity (II)

- Key idea: measure smoothness of an entropy using the [chain rule](#).

$$\left(\nabla \mathbf{F}(\mathbf{u}) = (\nabla \eta(\mathbf{u}))^\top \nabla \mathbf{f}(\mathbf{u}) \right) \Rightarrow \boxed{\nabla \cdot (\mathbf{F}(\mathbf{u})) = (\nabla \eta(\mathbf{u}))^\top \nabla \cdot (\mathbf{f}(\mathbf{u}))}$$

- Given an entropy pair $(\eta(\mathbf{v}), \mathbf{F}(\mathbf{v}))$, set
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- Commutator-based [entropy viscosity](#) is defined by setting

$$N_i^n := \sum_{j \in \mathcal{I}(i)} (\mathbf{F}(\mathbf{U}_j^n) - (\eta'(\mathbf{U}_i^n))^\top \mathbf{f}(\mathbf{U}_j^n)) \mathbf{c}_{ij},$$

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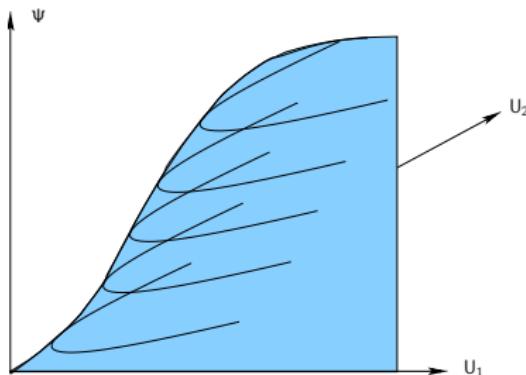
Outline



Convex limiting



Quasiconcavity



Definition (Quasiconcavity)

Given a convex set $\mathcal{B} \subset \mathbb{R}^m$, we say that a function $\Psi : \mathcal{B} \rightarrow \mathbb{R}$ is **quasiconcave** if the set $\{\mathbf{U} \in \mathcal{B}; \Psi(\mathbf{U}) \geq \chi\}$ is convex for any $\chi \in \mathbb{R}$.



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Strategy

- Let $\Psi : \mathcal{B} \rightarrow \mathbb{R}$ be a quasiconcave functional.
- If low-order update satisfies $\Psi(\mathbf{U}_i^{\text{L},n+1}) \geq 0$, then we want to “limit” the high-order update $\mathbf{U}_i^{\text{H},n+1} \rightarrow \mathbf{U}_i^{n+1}$ so that $\boxed{\Psi(\mathbf{U}_i^{n+1}) \geq 0}$.

Fundamental questions

- What should be limited?
- How can that be done so that the result is guaranteed.



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Example

- The negative of every entropy is quasiconcave (concave, actually).
- The negative of every specific entropy is quasiconcave (not concave).
- The negative of kinetic energy is quasiconcave (not concave).
- Internal energy is quasiconcave (concave, actually).
- Specific internal energy is quasiconcave (not concave).
- Every scalar component of \mathbf{U} is quasiconcave (concave and convex, actually).
(Think density, water height, etc.)

Example (with matrices, $\mathbb{R}^m = \mathbb{R}^{k^2}$, $m = k^2$.)

- Log-determinant of SPD matrices is concave.
- Negative of largest eigenvalue of SPD matrices is concave.



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Lemma (Quasiconcavity)

Let $\mathcal{B} \subset \mathbb{R}^m$ be convex set.

$\Psi : \mathcal{B} \rightarrow \mathbb{R}$ is quasiconcave



for every finite set $S \subset \mathbb{N}$, every set of convex coefficients $\{\lambda_j\}_{j \in S}$ (i.e., $\sum_{j \in S} \lambda_j = 1$ and $\lambda_j \geq 0$ for all $j \in S$), and every collection of vectors $\{\mathbf{U}_j\}_{j \in S}$ in \mathcal{B} , the following holds true:

$$\Psi\left(\sum_{j \in S} \lambda_j \mathbf{U}_j\right) \geq \min_{j \in S} \Psi(\mathbf{U}_j).$$



Fundamental bounds

Lemma (Fundamental bounds on the GMS-GV scheme, JLG+BP+IT (2018))

- Let $\mathcal{B} \subset \mathcal{A} \subset \mathbb{R}^m$ be any convex set.
- Let $\Psi : \mathcal{B} \rightarrow \mathbb{R}$ be any quasiconcave functional.
- Let $n \geq 0$, $i \in \mathcal{V}$. Assume that $1 - 4\Delta t \frac{\sum_{j \in \mathcal{I}(i)} d_{ij}^{L,n}}{m_i} \geq 0$ and $2\Delta t \leq \tau_0$.
- Let $\{\bar{\mathbf{U}}_{ij}^n\}_{j \in \mathcal{I}(i)}$ be the auxiliary states. Define the following quantity:

$$\Psi_i^{\min} := \min \left(\min_{j \in \mathcal{I}(i)} \Psi(\bar{\mathbf{U}}_{ij}^n), \Psi(\mathbf{U}_i^n + 2\Delta t \mathbf{S}(\mathbf{U}_i^n)) \right)$$

- Assume that $\mathbf{U}_j^n \in \mathcal{B}$ for all $j \in \mathcal{I}(i)$.

Then, the first-order update $\mathbf{U}_i^{L,n+1}$ computed with the GMS-GV scheme satisfies the following inequality:

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- Let \mathbf{U}^n be current solution, $n \geq 0$.
- Compute invariant domain preserving low-order solution $\mathbf{U}^{\text{L},n+1}$.
- Consider family of quasi-concave functionals $\Psi^l : \mathcal{B}^l \rightarrow \mathbb{R}$, $l \in \mathcal{L}$.
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- Invariant domain preserving low-order solution:

$$\frac{m_i}{\Delta t}(\mathbf{U}_i^{\text{L},n+1} - \mathbf{U}_i^n) + \sum_{j \in \mathcal{I}(i)} \mathbf{F}_{ij}^{\text{L},n} = m_i \mathbf{S}(\mathbf{U}_i^n), \quad \mathbf{F}_{ij}^{\text{L},n} = -\mathbf{F}_{ji}^{\text{L},n}.$$

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$$\mathbf{U}_i^{\text{H},n+1} = \mathbf{U}_i^{\text{L},n+1} + \sum_{j \in \mathcal{I}(i) \setminus \{i\}} \mathbf{A}_{ij}^n, \quad \mathbf{A}_{ij}^n = \frac{\Delta t}{m_i} (\mathbf{F}_{ij}^{\text{H},n} - \mathbf{F}_{ji}^{\text{L},n}).$$

Lemma (Limiter set not empty)

Limiter set is not empty because $\ell_{ij} = 0$ is an admissible limiter.

The above program is meaningful!



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$$\frac{m_i}{\Delta t}(\mathbf{U}_i^{\text{L},n+1} - \mathbf{U}_i^n) + \sum_{j \in \mathcal{I}(i)} \mathbf{F}_{ij}^{\text{L},n} = m_i \mathbf{S}(\mathbf{U}_i^n), \quad \mathbf{F}_{ij}^{\text{L},n} = -\mathbf{F}_{ji}^{\text{L},n}.$$

- High-order solution:

$$\frac{m_i}{\Delta t}(\mathbf{U}_i^{\text{H},n+1} - \mathbf{U}_i^n) + \sum_{j \in \mathcal{I}(i)} \mathbf{F}_{ij}^{\text{H},n} = m_i \mathbf{S}(\mathbf{U}_i^n), \quad \mathbf{F}_{ij}^{\text{H},n} = -\mathbf{F}_{ji}^{\text{H},n}.$$

- Difference, compute limiter $\ell_{ij} = \ell_{ji} \in [0, 1]$ as large as possible s.t.:

$$\boxed{\Psi'_i(\mathbf{U}_i^{\text{L},n+1} + \sum_{j \in \mathcal{I}(i) \setminus \{i\}} \ell_{ij} \mathbf{A}_{ij}^n) \geq 0}, \quad \mathbf{A}_{ij}^n = \frac{\Delta t}{m_i} (\mathbf{F}_{ij}^{\text{H},n} - \mathbf{F}_{ji}^{\text{L},n}).$$

Lemma (Limiter set not empty)

Limiter set is not empty because $\ell_{ij} = 0$ is an admissible limiter.

The above program is meaningful!



Abstract framework

Strategy

Divide and conquer: Take care of each pair (i, j) separately.

Lemma

- Let $\{\lambda_j\}_{j \in \mathcal{I}(i) \setminus \{i\}}$ be any set of strictly positive convex coefficients.
- Let $\mathbf{P}_{ij}^n := \frac{1}{\lambda_j} \mathbf{A}_{ij}^n$.
- Let $\Psi_i^l : \mathcal{B} \rightarrow \mathbb{R}$ be a quasiconcave function.
- Assume that the limiting parameters $\ell_{ij} \in [0, 1]$ are such that $\Psi_i^l(\mathbf{U}_i^{L,n+1} + \ell_{ij} \mathbf{P}_{ij}^n) \geq 0$, for all $j \in \mathcal{I}(i) \setminus \{i\}$.

Then the following inequality holds true:

$$\boxed{\Psi_i^l(\mathbf{U}_i^{n+1}) := \Psi_i^l\left(\sum_{j \in \mathcal{I}(i) \setminus \{i\}} \lambda_j (\mathbf{U}_i^{L,n+1} + \ell_{ij} \mathbf{P}_{ij}^n)\right) \geq 0}$$



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Example

Convex coefficients $\{\lambda_j\}_{j \in \mathcal{I}(i) \setminus \{i\}}$

- Example 1: $\lambda_j = \frac{1}{\text{card}(\mathcal{I}(i) \setminus \{i\})}$.
- Example 2: $\lambda_j = \frac{m_{ij}}{\sum_{k \in \mathcal{I}(i) \setminus \{i\}} m_{ik}}$.
- Example 3: $\lambda_j = \frac{\|\mathbf{c}_{ij}\|_{\ell^2}}{\sum_{k \in \mathcal{I}(i) \setminus \{i\}} \|\mathbf{c}_{ik}\|_{\ell^2}}$.
- Example 4: symmetry-preserving, linearity-preserving, etc. (a lot of choices to explore)



Abstract framework

Theorem (JLG+BP+IT (2018))

For every $i \in \mathcal{V}$ and $j \in \mathcal{I}(i)$, let ℓ_j^i be defined by

$$\ell_j^i = \begin{cases} 1 & \text{if } \Psi_i(\mathbf{U}_i^{\text{L},n+1} + \mathbf{P}_{ij}^n) \geq 0, \\ \max\{\ell \in [0, 1]; \Psi_i(\mathbf{U}_i^{\text{L},n+1} + \ell \mathbf{P}_{ij}^n) \geq 0\} & \text{otherwise.} \end{cases}$$

The following statements hold true:

- ① ℓ_j^i is uniquely defined.
- ② $\Psi_i(\mathbf{U}_i^{\text{L},n+1} + \ell \mathbf{P}_{ij}^n) \geq 0$ for every $\ell \in [0, \ell_j^i]$;
- ③ Setting $\ell_{ij} = \min(\ell_j^i, \ell_i^j)$, we have $\Psi_i(\mathbf{U}_i^{\text{L},n+1} + \ell_{ij} \mathbf{P}_{ij}^n) \geq 0$ and $\ell_{ij} = \ell_{ji}$.

Take home message

- For every i , limiter computed for every pair (i,j) in the stencil of i .
- Computation done by line search.



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Implementation

Claim

- If $\ell \rightarrow \Psi_i(\mathbf{U}^L + \ell\mathbf{P})$ is linear or quadratic, \Rightarrow
line search = **solving linear or quadratic equation**.
- If $\ell \rightarrow \Psi_i(\mathbf{U}^L + \ell\mathbf{P})$ is concave for all $\ell \in [0, \ell_{\max}]$, \Rightarrow
line search can be done **very efficiently** with **Newton-Secant** algorithm.
- Alg. always return limiter ℓ^{out} that satisfies $\Psi_i(\mathbf{U}^L + \ell^{\text{out}}\mathbf{P}) > 0$ for **any** tolerance.



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Numerical illustrations



Numerical
illustrations



1D smooth wave (Euler+ γ -law)

- 1D smooth wave (Euler+ γ -law) in $D = (0, 1)$.
- Solution: $v(x, t) = 1$, $p(x, t) = 1$ and

$$\rho(x, t) = \begin{cases} 1 + 2^6(x_1 - x_0)^{-6}(x - t - x_0)^3(x_1 - x + t)^3 & \text{if } x_0 \leq x - t < x_1 \\ 1 & \text{otherwise} \end{cases}$$

Table: \mathbb{P}_1 meshes, Convergence tests with Code 1([Entropy commutator GV](#)) and Code 3([smoothness-based GV](#)), CFL = 0.25.

# nodes	Code 1	
	δ_∞	rate
100	9.02E-03	
200	1.34E-04	6.07
400	1.01E-05	3.72
800	1.12E-06	3.18
1600	1.23E-07	3.19
3200	1.33E-08	3.21
6400	1.42E-09	3.22



2D isentropic vortex (Euler+ γ -law)

- $\rho_\infty = P_\infty = T_\infty = 1$, $\mathbf{u}_\infty = (u_\infty, v_\infty)^\top$, $u_\infty = 1$, $v_\infty = 1$.

$$\rho(\mathbf{x}, t) = (T_\infty + \delta T)^{1/(\gamma-1)}, \quad \mathbf{u}(\mathbf{x}, t) = \mathbf{u}_\infty + \delta \mathbf{u}, \quad p(\mathbf{x}, t) = \rho^\gamma$$

$$\delta \mathbf{u}(\mathbf{x}, t) = \frac{\beta}{2\pi} e^{\frac{1-r^2}{2}} (-\bar{x}_2, \bar{x}_1), \quad \delta T(\mathbf{x}, t) = -\frac{(\gamma-1)\beta^2}{8\gamma\pi^2} e^{1-r^2},$$

- With $\bar{\mathbf{x}} = (x_1 - x_1^0 - u_\infty t, x_2 - x_2^0 - v_\infty t)$, $r^2 = \|\bar{\mathbf{x}}\|_{\ell^2}^2$, $\gamma = \frac{7}{5} = 1.4$, $\beta = 5$, $(x_1^0, x_2^0) = (4, 4)$.

Table: Isentropic vortex, \mathbb{P}_1 meshes, convergence tests with limiting and without limiting, $t = 2$.
Code 2, CFL = 0.1.

	# nodes	$\delta_1(t)$	rate	$\delta_2(t)$	rate	$\delta_\infty(t)$	rate
Limiting	441	2.46E-02	—	9.61E-02	—	1.31E+00	—
	1681	8.50E-03	1.59	3.36E-02	1.57	4.45E-01	1.62
	6561	1.99E-03	2.13	7.45E-03	2.21	9.40E-02	2.28
	25921	4.16E-04	2.28	1.50E-03	2.33	2.05E-02	2.21
	103041	8.11E-05	2.37	2.92E-04	2.37	4.68E-03	2.14
No limiting	441	2.47E-02	—	9.61E-02	—	1.31E+00	—
	1681	8.48E-03	1.60	3.34E-02	1.58	4.52E-01	1.59
	6561	1.99E-03	2.13	7.45E-03	2.20	9.40E-02	2.31
	25921	4.16E-04	2.28	1.50E-03	2.33	2.05E-02	2.21
	103041	8.11E-05	2.37	2.92E-04	2.37	4.68E-03	2.14



Usual suspects (Continuous \mathbb{P}_1 finite elements)

# nodes	Code 3		Low-order	
	$\delta_1(t)$	rate	$\delta_1(t)$	rate
100	1.49E-01		2.61E-01	
200	9.01E-02	0.72	1.94E-01	0.43
400	4.92E-02	0.87	1.41E-01	0.46
800	2.61E-02	0.91	9.95E-02	0.50
1600	1.34E-02	0.96	6.74E-02	0.56
3200	6.83E-03	0.97	4.40E-02	0.62
6400	3.42E-03	0.99	2.78E-02	0.66
12800	1.70E-03	1.00	1.73E-02	0.68

Table: Leblanc shocktube, \mathbb{P}_1 meshes, Convergence tests with Code 1 and Code 3, CFL = 0.25.

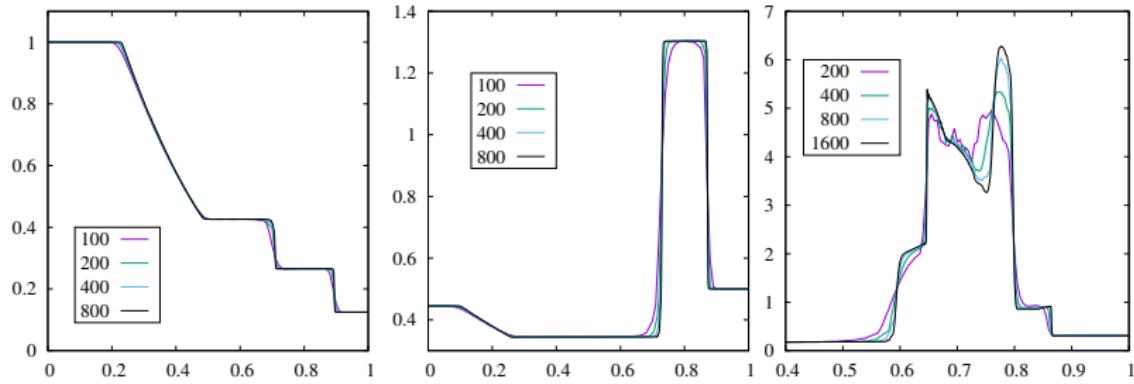


Figure: Left: Sod shocktube; Center: Lax shocktube; Right: Woodward-Colella blast wave.



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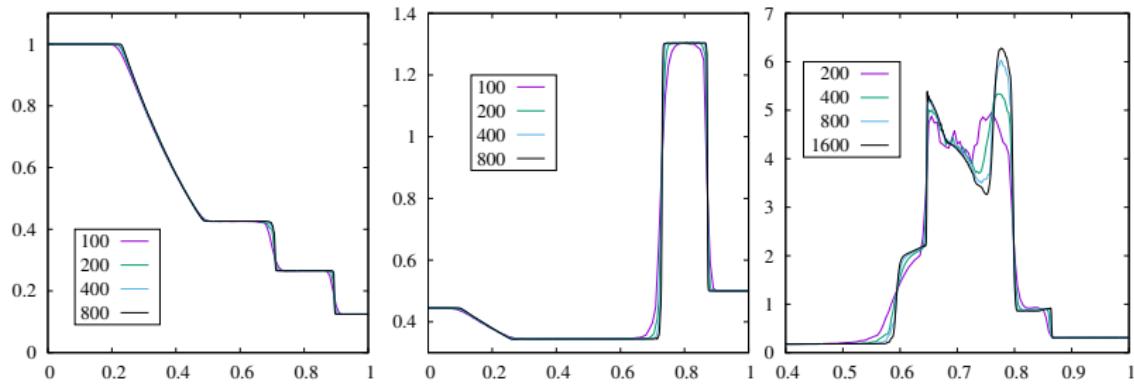


Figure: Left: Sod shocktube; Center: Lax shocktube; Right: Woodward-Collela blast wave.



Forward facing step, Mach 3 (Continuous \mathbb{P}_1 and \mathbb{Q}_1 finite elements)

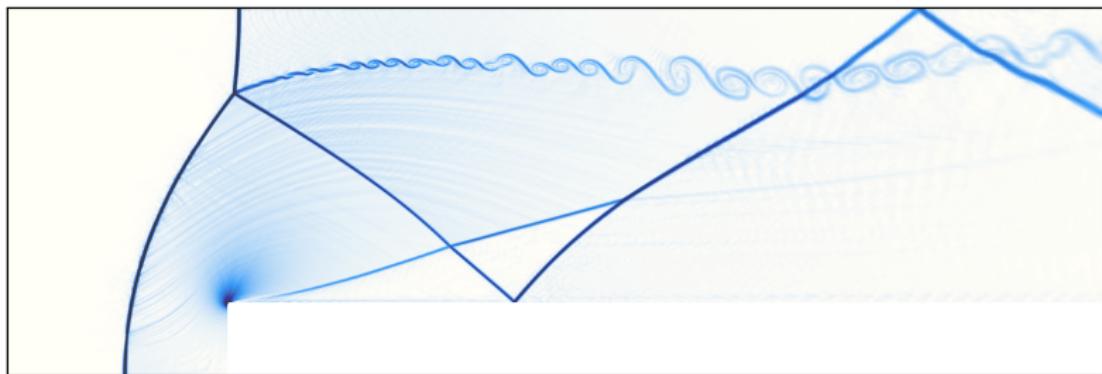


Figure: Mach 3 step, $t = 4$, density. Top: unstructured grid, 207340 \mathbb{P}_1 nodes. Limiting done on density (local min and max) and specific entropy (local min).



Cylinder in a channel (Continuous \mathbb{P}_1 finite elements)

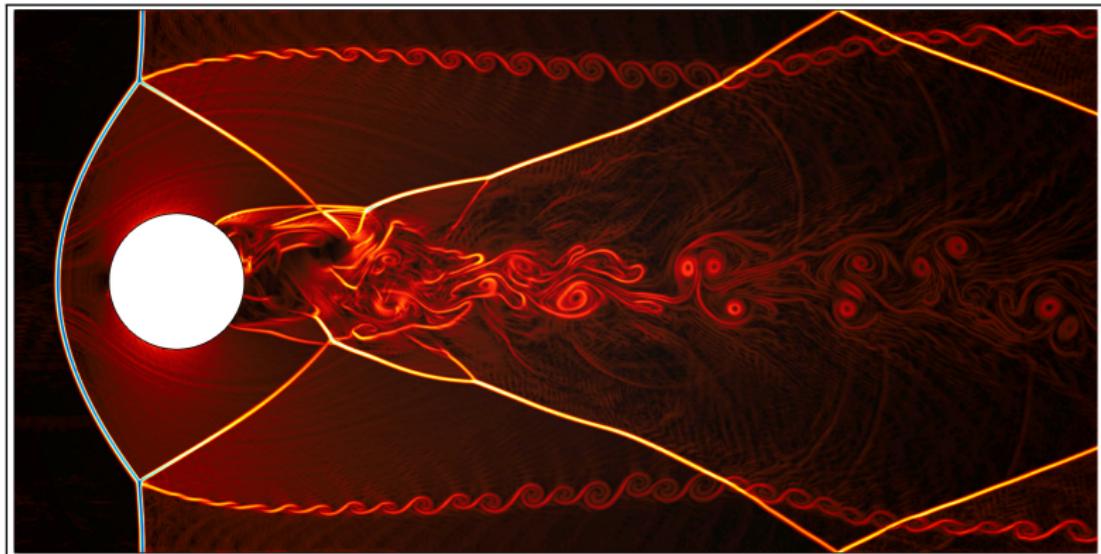
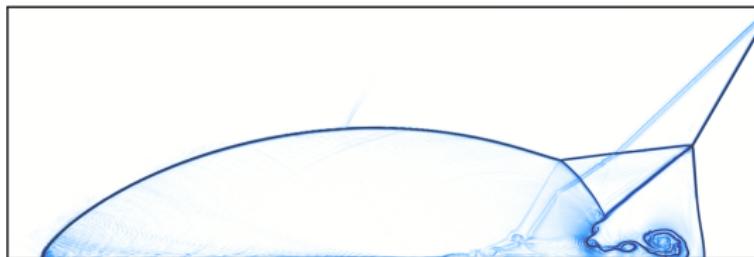


Figure: Cylinder in a channel 366764 \mathbb{P}_1 nodes. Limiting done on density (local min and max) and specific entropy (local min).

▶ Cylinder at Mach 3



Double Mach reflection, Mach 10 (Continuous \mathbb{P}_1 and \mathbb{Q}_1 finite elements)



Limiting done on density (local min and max) and specific entropy (local min).

▶ Mach 10



Outline



Beyond second-order

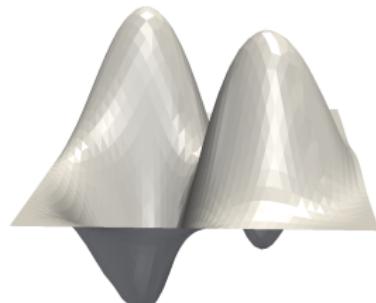


First-order solution: Dependence w.r.t polynomial degree k

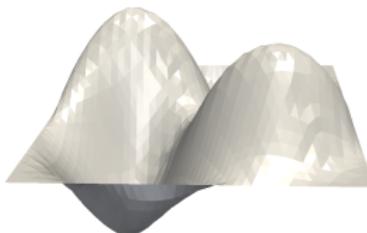
- Swirling flow: $\partial_t u + \beta \cdot \nabla u = 0$

$$\beta(x, t) = \cos(\pi t)(-\sin(2\pi x_2) \sin^2(\pi x_1) e_1 + \sin(2\pi x_1) \sin^2(\pi x_2) e_2)$$

- Solution is periodic in time with period 1.
- Initial data: $u(x, t) = 2^8(x_1(1-x_1)x_2(1-x_2))^2 \sin(2\pi x_1) \sin(2\pi x_2)$.



\mathbb{P}_1 , 1927 nodes,



\mathbb{P}_2 1945 nodes,



\mathbb{P}_3 1888 nodes

- Origin of the problem: Size of stencil increases with k .

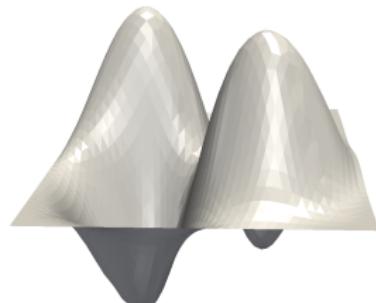


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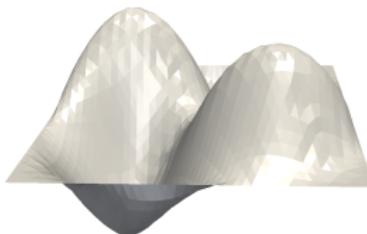
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\mathbb{P}_1 , 1927 nodes,



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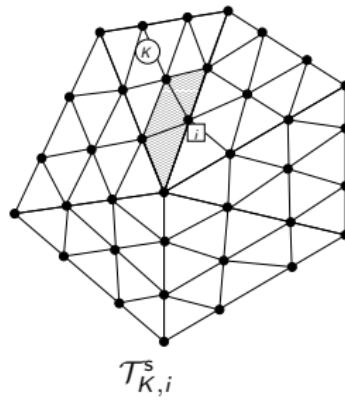
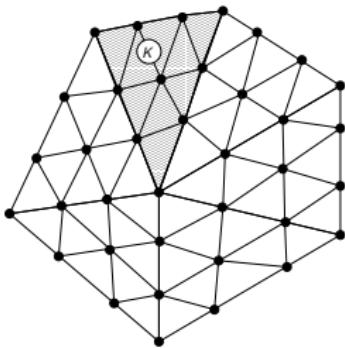
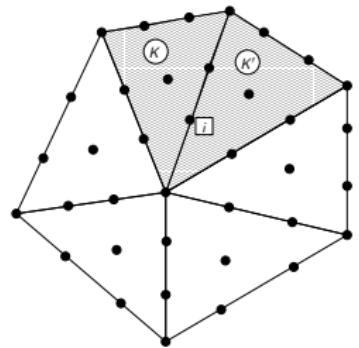


\mathbb{P}_3 1888 nodes

- Origin of the problem: **Size of stencil increases with k .**



Hierarchical structure



Hierarchical structure

- Introduce mesh \mathcal{T}_h
- Introduce high-order finite element $(\hat{K}, \hat{P}^H, \hat{\Sigma}^H)$ and high-order space

$$\boxed{P^H(\mathcal{T}_h) := \{v \in C^0(D; \mathbb{R}); v|_K \circ T_K \in \hat{P}^H, \forall K \in \mathcal{T}_h\}}$$

- Not necessarily Lagrange elements (could be Bernstein).

- Introduce subdivided mesh \mathcal{T}_h^s
- Introduce low-order Lagrange finite element $(\hat{K}, \hat{P}^L, \hat{\Sigma}^L)$ and low-order space

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Hierarchical structure

Assumptions

- (If $P^H(\mathcal{T}_h)$ is a Lagrange space) For all $K \in \mathcal{T}_h$, there exists $\{\beta^T\}_{T \in \mathcal{T}_K^s}$ s.t. the following holds true for every $\mathbf{p} \in \mathbf{P}_K^H := \{\hat{\mathbf{p}} \circ \mathbf{T}_K^{-1}; \hat{\mathbf{p}} \in \hat{\mathbf{P}}^H\}$:

$$\int_K \nabla \cdot (\mathbf{p}) \, dx = \sum_{T \in \mathcal{T}_K^s} \beta^T \int_T \nabla \cdot (\Pi_T^{\text{Lag}}(\mathbf{p})) \, dx.$$

- For all $K \in \mathcal{T}_h$ and all $T \in \mathcal{T}_K^s$, there exists $\{\alpha_j^T\}_{j \in \mathcal{I}^L(T)}$ s.t.

$$\forall j \in \mathcal{I}(K), \quad \sum_{T \in \mathcal{T}_{K,j}^s} \alpha_j^T = 1,$$

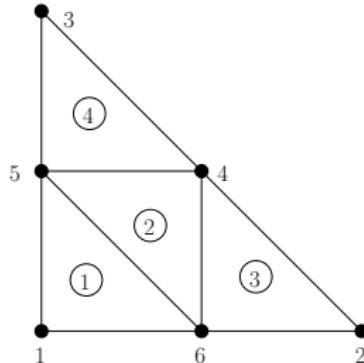
$$\forall T \in \mathcal{T}_K^s, \quad \sum_{j \in \mathcal{I}^L(T)} \alpha_j^T m_j^K = \beta^T |T|.$$

Literature

Hierarchical structure similar to high-order Residual Distribution method (**Abgrall et al. (2017)**)



Hierarchical structure \mathbb{P}_2 in 2D

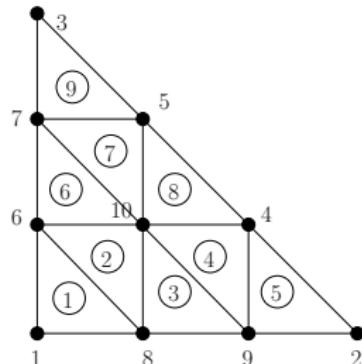


I	1	2	3	4
$\widehat{\beta}(I)$	$\frac{2}{3}$	$\frac{2}{3}$	$\frac{2}{3}$	2
$\widehat{\alpha}(1, I)$	1	$\frac{1}{2}$	1	1
$\widehat{\alpha}(2, I)$	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$	$\frac{1}{4}$
$\widehat{\alpha}(3, I)$	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$	$\frac{1}{4}$
$\widehat{j}(1, I)$	1	4	2	3
$\widehat{j}(2, I)$	5	5	4	4
$\widehat{j}(3, I)$	6	6	6	5

Table: $\widehat{\alpha}$ and $\widehat{\beta}$ for \mathbb{P}_2 polynomials in two dimensions.



Hierarchical structure \mathbb{P}_3 in 2D



$I \in \widehat{\mathcal{L}}$	1	2	3	4	5	6	7	8	9
$\widehat{\beta}(I)$	$\frac{3}{4}$	$\frac{3}{2}$	$\frac{3}{4}$	$\frac{3}{2}$	$\frac{3}{4}$	$\frac{3}{4}$	$\frac{3}{2}$	$\frac{3}{4}$	$\frac{3}{4}$
$\widehat{\alpha}(1, I)$	1	y	x	y	1	x	y	x	1
$\widehat{\alpha}(2, I)$	$\frac{1}{3}$	y	x	y	$\frac{1}{3}$	x	y	x	$\frac{1}{3}$
$\widehat{\alpha}(3, I)$	$\frac{1}{3}$	z	$\frac{1}{3} - z$	z	$\frac{1}{3}$	$\frac{1}{3} - z$	z	$\frac{1}{3} - z$	$\frac{1}{3}$
$\widehat{j}(1, I)$	1	6	8	4	2	6	5	4	3
$\widehat{j}(2, I)$	6	8	9	9	4	7	7	5	5
$\widehat{j}(3, I)$	8	10	10	10	9	10	10	10	7

Table: $\widehat{\alpha}$ and $\widehat{\beta}$ for \mathbb{P}_3 polynomials in two dimensions (uniform lattice). One parameter family $y \in [0, \frac{2}{3}]$, $x = \frac{2}{3} - y$ and $z = \frac{1}{3}(\frac{10}{9} - y)$.



Low-order scheme

- Define

$$\mathbf{c}_{ij}^L := \sum_{K \in \mathcal{T}(i)} m_i^K \sum_{T \in \mathcal{T}_K^s} \alpha_i^T \frac{1}{|T|} \int_T \nabla \varphi_j^{\text{Lag}, L} \, dx.$$

- Low-order update

$$m_i \frac{\mathbf{U}_i^{L,n+1} - \mathbf{U}_i^n}{\Delta t} = m_i \mathbf{S}(\mathbf{U}_i^n) - \sum_{j \in \mathcal{I}^L(i)} \mathbf{f}(\mathbf{U}_j^n) \mathbf{c}_{ij}^L + \sum_{j \in \mathcal{I}^L(i)} d_{ij}^{L,n} (\mathbf{U}_j^n - \mathbf{U}_i^n).$$

Theorem (Conservation)

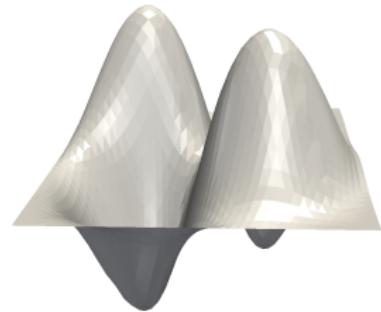
If $\mathbf{S} \equiv \mathbf{0}$, the scheme has the following conservation property:

$$\int_D \mathbf{u}_h^{L,n+1} \, dx = \int_D \mathbf{u}_h^n \, dx + \Delta t \int_D \nabla \cdot (\Upsilon_h^H(\mathbf{f}_h(\mathbf{u}_h^n))) \, dx.$$

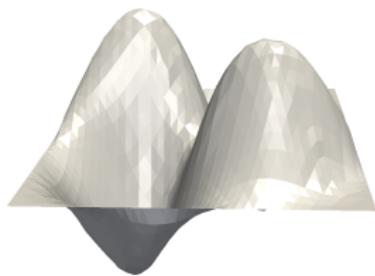
Rk: Here $\Upsilon_h \equiv \Pi_h^{\text{Lag}}$ if high-order FE are Lagrange.



Low-order scheme



\mathbb{P}_1 , 1927 nodes,



\mathbb{P}_2 1945 nodes,



\mathbb{P}_3 1888 nodes



High-order scheme

- If high-order FE is not Lagrange, make change of basis

$$\mathbf{U}_{j(m,K)}^{\text{Lag},n} := \mathbf{u}_h^n(\mathbf{a}_{j(m,K)}) = \sum_{I \in \widehat{\mathcal{N}}} b_{Im} \mathbf{U}_{j(I,K)}^n.$$

- Define Lagrange interpolant of flux

$$\Pi_h^{\text{Lag}}(\mathbf{f}(\mathbf{u}_h^n)) := \sum_{i \in \mathcal{V}} \mathbf{f}(\mathbf{U}_{j(m,K)}^{\text{Lag},n}) \varphi_i^{\text{Lag}}.$$

- Define

$$c_{ij} := \int_D (\nabla \varphi_j^{\text{Lag}}) \psi_i \, dx,$$

- High-order update

$$m_i \frac{\mathbf{U}_i^{\text{H},n+1} - \mathbf{U}_i^n}{\Delta t} = - \sum_{j \in \mathcal{I}(i)} \mathbf{f}(\mathbf{U}_j^{\text{Lag},n}) c_{ij} + \sum_{j \in \mathcal{I}^{\text{L}}(i)} d_{ij}^{\text{H},n} (\mathbf{U}_j^n - \mathbf{U}_i^n).$$



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Limiting

- Compute $m_i \mathbf{U}_i^{\text{H},n+1} - m_i \mathbf{U}_i^{\text{L},n+1}$

$$m_i \mathbf{U}_i^{\text{H},n+1} = m_i \mathbf{U}_i^{\text{L},n+1} - \sum_{K \in \mathcal{K}(i)} \mathbf{B}_{K,i} + \sum_{j \in \mathcal{I}^{\text{L}}(i)} \mathbf{A}_{ij}^n$$

- with **dof to dof** viscosity contribution

$$\mathbf{A}_{ij}^n := \Delta t (d_{ij}^{\text{H},n} - d_{ij}^{\text{L},n}) (\mathbf{U}_j^n - \mathbf{U}_i^n)$$

- and **dof to cell** flux contribution

$$\mathbf{B}_{i,K}^n := \Delta t \int_K \nabla \cdot (\Pi_h^{\text{Lag}}(\mathbf{f}(\mathbf{u}_h^n))) \psi_i \, dx - \Delta t m_i^K \sum_{T \in \mathcal{T}_{K,i}^s} \frac{\alpha_i^T}{|T|} \int_T \nabla \cdot (\Upsilon_h^{\text{L}}(\mathbf{f}_h(\mathbf{u}_h^n))) \, dx.$$



Limiting

Lemma

Let $\ell_K \in [0, 1]$ for all $K \in \mathcal{T}_h$. Let $\ell_{ij} \in [0, 1]$ for all $i \in \mathcal{V}$, $j \in \mathcal{I}^L(j)$ with the assumption that $\ell_{ij} = \ell_{ji}$. Let \mathbf{U}_i^{n+1} be defined by

$$m_i \mathbf{U}_i^{n+1} = m_i \mathbf{U}_i^{L,n+1} + \sum_{j \in \mathcal{I}^L(i)} \ell_{ij} \mathbf{A}_{ij}^n + \sum_{K \in \mathcal{T}(i)} \ell_K \mathbf{B}_{i,K}^n.$$

Assume that either

- the high-order finite element $(\hat{K}, \hat{P}^H, \hat{\Sigma}^H)$ is a [Lagrange](#) finite element
- or the flux \mathbb{f} is [linear](#).

Then \mathbf{u}_h^{n+1} and $\mathbf{u}_h^{L,n+1}$ carry the same mass, i.e., $\int_D \mathbf{u}_h^{n+1} dx = \int_D \mathbf{u}_h^{L,n+1} dx$.



Limiting

Claim

Convex limiting can be used to compute limiting coefficients ℓ_{ij} and ℓ_K .
⇒ High-order solution is **conservative** and **invariant-domain preserving**.

Question

Question is open with Bernstein FE **if flux is nonlinear.**



The end

The end

