AMCS 394E: Contemp. Topics in Computational Science. Computing with the finite element method

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Solving hyperbolic conservation laws

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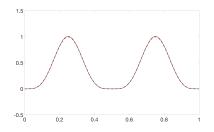
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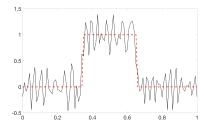
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- * We need to consider weak solutions.
- * The weak formulation admits multiple solutions.

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The solution $u_h(\mathbf{x}, t)$ must be mass conservative.

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Then mass conservation means:

$$\frac{\partial}{\partial t} \int_{\Omega} u_h d\mathbf{x} = 0 \iff \int_{\Omega} u_h(\mathbf{x}, t) d\mathbf{x} = \int_{\Omega} u_h(\mathbf{x}, 0) d\mathbf{x}.$$

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$$\partial_t \int_{\Omega} u \phi d\mathbf{x} = \int_{\Omega} \mathbf{f}(u) \cdot \nabla \phi d\mathbf{x} + \int_{\partial \Omega} \mathbf{n} \cdot \mathbf{f}(u) \phi d\mathbf{x}$$

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Since $\phi = 1$ is in the space V_h , we can use it in the formulation to get

$$\partial_t \int_{\Omega} u d\mathbf{x} = \underbrace{\int_{\Omega} \mathbf{f}(u) \cdot \nabla \phi d\mathbf{x}}_{=0}$$

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In general, we must guarantee

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Main problems when solving hyperbolic conservation laws

Presence of non-physical oscillations near large gradients

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* Combinations like enriched Galerkin FEs DG^0 - $CG^p(\Omega)$.

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- * Methods based on a (non)linear weak formulation of a dissipative term.
- * Algebraic methods based on flux corrections.

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The idea is to consider

$$\partial_t u + \nabla \cdot \mathbf{f}(u) = \nabla \cdot \mathbf{c}(u) \nabla u,$$

where
$$c(u) = \mathbb{O}(h^p)$$
. That is, $c(u) \to 0$ as $h \to 0$.

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- * The linearity of c(u).

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A linear and first-order dissipative discretization is

$$\partial_t \int_{\Omega} u \phi d\mathbf{x} + \int_{\Omega}
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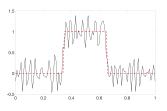
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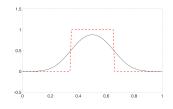
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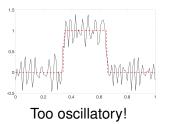


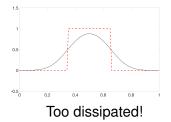
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- * Shock capturing methods.
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We can consider combinations. For instance,

SUPG + shock capturing

High-order artificial dissipation via SUPG

The SUPG method is given as follows:

$$\int_{\Omega} \left(\frac{\partial \textit{\textbf{u}}_{\textit{h}}}{\partial \textit{\textbf{t}}} + \nabla \cdot \textbf{\textbf{f}}(\textit{\textbf{u}}_{\textit{h}}) \right) \varphi \textit{\textbf{d}} \textbf{\textbf{x}} + \sum_{\textit{\textbf{k}}} \textit{\textbf{s}}_{\textit{\textbf{K}}}(\textit{\textbf{u}}_{\textit{\textbf{h}}}, \varphi) = 0,$$

where

$$s_K(u_h,\varphi) = \nu_K \int_{\mathcal{K}} \left(\frac{\partial u_h}{\partial t} + \nabla \cdot \mathbf{f}(u_h) \right) \left(\mathbf{f}'(u) \cdot \nabla \varphi \right) d\mathbf{x}.$$

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Here ν_K is a coefficient given as follows

$$\nu_{K} = \frac{ch_{K}}{||\mathbf{f}'(u_{h})||_{L^{\infty}(K)}}$$

where c is a user defined constant.

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If the solution is non-smooth, we need to consider instead

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Integrate over the domain and integrate by parts the RHS:

$$\int_{\Omega} \left[\frac{\partial \eta(u)}{\partial t} + \nabla \cdot \mathbf{q}(u) \right] d\mathbf{x} = -\epsilon \int_{\Omega} \eta''(u) |\nabla u|^2 d\mathbf{x} \le 0$$

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If the solution is non-smooth

$$\int_{\Omega} \left[\underbrace{\frac{\partial \eta(u)}{\partial t} + \nabla \cdot \mathbf{q}(u)}_{=R(u)} \right] d\mathbf{x} \le 0$$

High-order artificial dissipation via Entropy Viscosity (EV)

The idea of the EV method is to add dissipation if some local measurement of R(u) is non-zero.

High-order artificial dissipation via Entropy Viscosity (EV)

One version of the EV method is given as follows:

$$\int_{\Omega} \left(\frac{\partial \textit{u}_{\textit{h}}}{\partial \textit{t}} + \nabla \cdot \textbf{f}(\textit{u}_{\textit{h}}) \right) \varphi \textit{d}\textbf{x} + \sum_{\textit{K}} \textbf{s}_{\textit{K}}^{\text{EV}}(\textit{u}_{\textit{h}}, \varphi) = 0,$$

where

$$s_K^{\mathrm{EV}}(u_h, \varphi) = \nu_K^{\mathrm{EV}} \int_K \nabla u_h \cdot \nabla \varphi d\mathbf{x}$$

Here

$$\nu_K^{\rm EV} = \min \left(\frac{c_1 h_K || \mathbf{f}'(u_h) ||_{L^\infty(K)}}{2}, \; \frac{c_2 h_K^2 || R(u_h) ||_{L^\infty(K)}}{|| \eta(u_h) - \bar{\eta}(u_h) ||_{L^\infty(\Omega)}} \right)$$

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A first order method (for the advection equation) Consider

$$m_i\left(\frac{U_i^{n+1}-U_i^n}{\Delta t}\right)+a\sum_j c_{ij}U_j^n-\sum_j d_{ij}U_j^n=0$$

which guarantees

$$U_i^{\min} = \min_{j \in N_i} U_j^n \le U_i^{n+1} \le \max_{j \in N_i} U_j^n = U_i^{\max}$$

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which guarantees

$$\min_{\mathbf{x}} u_h(\mathbf{x}, t=0) \leq U_i^{n+1} \leq \max_{\mathbf{x}} u_h(\mathbf{x}, t=0)$$

Second order Flux Corrected Transport (FCT)

Consider a low-order solution; e.g.,

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$$m_i\left(\frac{U_i^H-U_i^n}{\Delta t}\right)+a\sum_i c_{ij}U_j^n=0$$

Second order Flux Corrected Transport (FCT)

Consider a low-order solution; e.g.,

$$U_i^L = U_i^n - \frac{a\Delta t}{m_i} \sum_i c_{ij} U_j^n + \frac{\Delta t}{m_i} \sum_i d_{ij} U_j^n$$

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and a high-order solution; e.g.,

$$U_i^H = U_i^L + \frac{1}{m_i} \sum_i f_{ij},$$

Here $f_{ij} := -\Delta t d_{ij} (U_j^n - U_i^n)$ is an anti-diffusive flux.

The idea with the FCT method is to add flux limiters $0 \le \alpha_{ij} \le 1$ that guarantee

$$U_i^{\min} = \min_{j \in N_i} U_j^n \le U_i^{n+1} \le \max_{j \in N_i} U_j^n = U_i^{\max},$$

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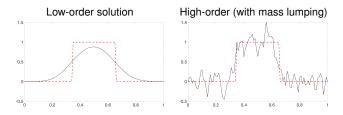
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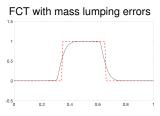
Consider the 1D advection equation $u_t + u_x = 0$.



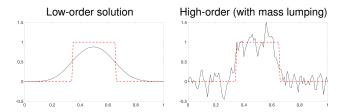


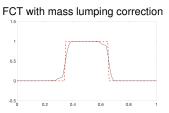
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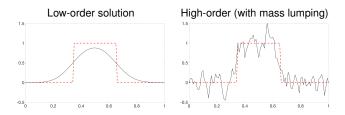


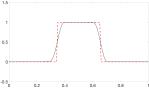
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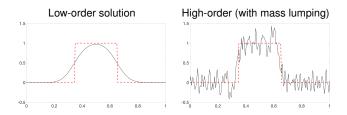


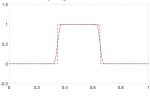
Consider the 1D advection equation $u_t + u_x = 0$ with ($\Delta x = 0.01$).



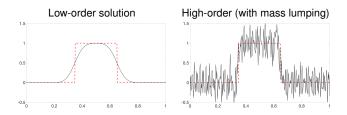


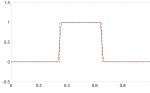
Consider the 1D advection equation $u_t + u_x = 0$ with $(\Delta x = 0.01/2)$.





Consider the 1D advection equation $u_t + u_x = 0$ with $(\Delta x = 0.01/4)$.





Consider the 1D advection equation $u_t + u_x = 0$ with $(\Delta x = 0.01/8)$.

