Finite Volume Methods for Hyperbolic Problems

TVD Methods and Limiters

- Slope limiters vs. flux limiters
- Total variation for scalar problems
- Proving TVD in flux-limiter form
- Design of TVD limiters
- Sweby Region

- Methods that give good accuracy for smooth solutions Clawpack methods: at best second-order accuracy
- Do not have oscillations around discontinuities Not only ugly but can lead to nonlinear instabilities

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- Godunov-type methods based on Riemann solvers Wave-propagation algorithms with "limiters"

Limiters can eliminate oscillations

Step function data with minmod slope:



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Evolving solution and averaging maintains monotonicity:



Could make slope steeper and still be monotone

Step function data with MC slope (twice that of minmod):



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Monotonized centered (MC) limiter

Using the centered slope $(Q_{i+1}^n - Q_{i-1}^n)/(2\Delta x)$ gives second-order accuracy (Fromm's method) but not monotonicity.

Limit this slope based on twice the one-sided slopes.

$$\sigma_i^n = \mathsf{minmod}\left(\left(\frac{Q_{i+1}^n - Q_{i-1}^n}{2\Delta x}\right), \ 2\left(\frac{Q_i^n - Q_{i-1}^n}{\Delta x}\right), \ 2\left(\frac{Q_{i+1}^n - Q_i^n}{\Delta x}\right)\right).$$

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Rationale:

- Where solution is smooth, centered slope is smaller and chosen, hence maintains accuracy.
- Near jumps in solution, don't expect second-order but want to resolve discontinuities as sharply as possible.

TVD REA Algorithm

1 Reconstruct a piecewise linear function $\tilde{q}^n(x, t_n)$ defined for all x, from the cell averages Q_i^n .

$$\tilde{q}^n(x,t_n) = Q_i^n + \sigma_i^n(x-x_i)$$
 for all $x \in \mathcal{C}_i$

with the property that $TV(\tilde{q}^n) \leq TV(Q^n)$.

- 2 Evolve the hyperbolic equation exactly (or approximately) with this initial data to obtain $\tilde{q}^n(x, t_{n+1})$ a time k later.
- Average this function over each grid cell to obtain new cell averages

$$Q_i^{n+1} = \frac{1}{\Delta x} \int_{\mathcal{C}_i} \tilde{q}^n(x, t_{n+1}) \, dx.$$

Note: Steps 2 and 3 are always TVD.

MC slopes are not always a TVD reconstruction

Sample data with MC slope (twice that of minmod):



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Sample data with MC slope (twice that of minmod):



But evolving and averaging still maintains monotonicity (TVD):



Slope limiters and flux limiters

Slope limiter formulation for advection:

 $\tilde{q}^n(x, t_n) = Q_i^n + \sigma_i^n(x - x_i)$ for $x_{i-1/2} \le x < x_{i+1/2}$.

Applying REA algorithm gives (for u > 0):

$$Q_i^{n+1} = Q_i^n - \frac{u\Delta t}{\Delta x}(Q_i^n - Q_{i-1}^n) - \frac{1}{2}\frac{u\Delta t}{\Delta x}\left(\Delta x - u\Delta t\right)(\sigma_i^n - \sigma_{i-1}^n)$$

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Flux limiter formulation:

$$Q_i^{n+1} = Q_i^n - \frac{\Delta t}{\Delta x} (F_{i+1/2}^n - F_{i-1/2}^n)$$

with flux

$$F_{i-1/2}^{n} = uQ_{i-1}^{n} + \frac{1}{2}u(\Delta x - u\Delta t)\sigma_{i-1}^{n}$$

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$$F_{i-1/2}^n = uQ_{i-1}^n + \frac{1}{2}u(\Delta x - u\Delta t)\sigma_{i-1}^n = \frac{1}{\Delta t}\int_{t_n}^{t_{n+1}} u\tilde{q}(x_{i-1/2},t)\,dt.$$

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$$F_{i-1/2}^n = \begin{cases} uQ_{i-1}^n + \frac{1}{2}u(\Delta x - u\Delta t)\sigma_{i-1}^n & \text{ if } u > 0, \\ uQ_i^n - \frac{1}{2}u(\Delta x + u\Delta t)\sigma_i^n & \text{ if } u < 0. \end{cases}$$

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Lax-Wendroff:

$$F_{i-1/2}^{n} = u^{+}Q_{i-1}^{n} + u^{-}Q_{i}^{n} + \frac{1}{2}|u|(1 - |u|\Delta t/\Delta x)(Q_{i}^{n} - Q_{i-1}^{n})$$

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Flux limiter method: Replace $\Delta Q_{i-1/2}^n$ by limited version $\delta_{i-1/2}^n$

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Flux limiters and wave limiters

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For systems of equations:

• Solve Riemann problem to decompose $\Delta Q_{i-1/2}^n$ into waves

$$\Delta Q_{i-1/2} = \sum_{p} \mathcal{W}_{i-1/2}^{p} = \sum_{p} \alpha_{i-1/2}^{p} r^{p}$$

- Use wave propagation form of Godunov (first-order) update
- Apply limiters to waves to get $\widetilde{\mathcal{W}}_{i-1/2}^p = \tilde{\alpha}_{i-1/2}^p r^p$
- Use limited waves in "second-order" corrections

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Limiter based on the ratio

$$\theta_{i-1/2}^n = \frac{Q_I - Q_{I-1}}{Q_i - Q_{i-1}}$$

where I denotes the cell in the upwind direction:

$$I = \begin{cases} i-1 & \text{if } u > 0\\ i+1 & \text{if } u < 0. \end{cases}$$

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Note that:

- $\theta \approx 1 + \mathcal{O}(\Delta x)$ where the solution is smooth,
- $\theta < 0$ if slopes have different sign.

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$$\delta_{i-1/2}^{n} = \phi(\theta_{i-1/2}^{n}) \Delta Q_{i-1/2}^{n}$$

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Desirable properties:

- $\phi(\theta) = 0$ for $\theta \le 0$ (zero slope at extrema)
- + $\phi(1)=1$ so nearly using Lax-Wendroff where smooth

$$\begin{split} F_{i-1/2}^n &= u^+ Q_{i-1}^n + u^- Q_i^n + \frac{1}{2} |u| (1 - |u| \Delta t / \Delta x) \delta_{i-1/2}^n \\ \delta_{i-1/2}^n &= \phi(\theta_{i-1/2}^n) \Delta Q_{i-1/2}^n \quad \text{where} \quad \theta_{i-1/2}^n = \frac{Q_I - Q_{I-1}}{Q_i - Q_{i-1}} \end{split}$$
 Note that:

- $\phi(\theta) \equiv 0$ for all $\theta \implies$ upwind method
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- $\phi(\theta) = \theta \implies$ Beam-Warming: $\delta_{i-1/2}^n = Q_I Q_{I-1}$

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- $\phi(\theta) = \frac{1}{2}(1+\theta) \implies$ Fromm's method

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- $\bullet \ \phi(\theta) = \mathsf{minmod}(1,\theta) \implies \mathsf{Minmod} \ \mathsf{method}$

For $q_t + uq_x = 0$ with u > 0 and $\nu = u\Delta t/\Delta x$

$$Q_i^{n+1} = Q_i^n - \frac{\Delta t}{\Delta x} (F_{i+1/2}^n - F_{i-1/2}^n)$$

$$F_{i-1/2}^{n} = uQ_{i-1}^{n} + \frac{1}{2}u(1 - u\Delta t/\Delta x)\delta_{i-1/2}^{n}$$
$$= uQ_{i-1}^{n} + \frac{1}{2}u(1 - \nu)[\phi(\theta_{i-1/2})(Q_{i} - Q_{i-1})]$$

Can be written as:

$$Q_i^{n+1} = Q_i^n - \left[\nu + \frac{1}{2}\nu(1-\nu)\left(\frac{\phi(\theta_{i+1/2})}{\theta_{i+1/2}} - \phi(\theta_{i-1/2})\right)\right](Q_i^n - Q_{i-1}^n)$$

since $Q_{i+1} - Q_i = (1/\theta_{i+1/2})(Q_i - Q_{i-1})).$

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Use this part of Theorem 6.1 (Harten):

The method

$$Q_i^{n+1} = Q_i^n - C_{i-1}^n (Q_i^n - Q_{i-1}^n)$$

is TVD provided $0 \le C_i^n \le 1$ for all *i*, regardless of how these coefficients depend on Q^n , Δx , Δt .

$$Q_i^{n+1} = Q_i - C_{i-1}(Q_i - Q_{i-1}), \qquad TV(Q) = \sum |Q_{i+1} - Q_i|$$

$$Q_{i+1}^{n+1} - Q_i^{n+1} = (Q_{i+1} - Q_i) - C_i(Q_{i+1} - Q_i) + C_{i-1}(Q_i - Q_{i-1})$$
$$= (1 - C_i)(Q_{i+1} - Q_i) + C_{i-1}(Q_{i+1} - Q_i)$$

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$$|Q_{i+1}^{n+1} - Q_i^{n+1}| \le (1 - C_i)|Q_{i+1} - Q_i| + C_{i-1}|Q_i - Q_{i-1}|$$

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$$\sum |Q_{i+1}^{n+1} - Q_i^{n+1}| \le \sum (1 - C_i)|Q_{i+1} - Q_i| + \sum C_{i-1}|Q_i - Q_{i-1}|$$

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$$\sum |Q_{i+1}^{n+1} - Q_i^{n+1}| \le \sum (1 - C_i)|Q_{i+1} - Q_i| + \sum C_{i-1}|Q_i - Q_{i-1}|$$

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$$\begin{aligned} |Q_{i+1}^{n+1} - Q_i^{n+1}| &\leq (1 - C_i)|Q_{i+1} - Q_i| + C_{i-1}|Q_i - Q_{i-1}|\\ \sum |Q_{i+1}^{n+1} - Q_i^{n+1}| &\leq \sum (1 - C_i)|Q_{i+1} - Q_i| + \sum C_{i-1}|Q_i - Q_{i-1}|\\ &\leq \sum (1 - C_i)|Q_{i+1} - Q_i| + \sum C_i|Q_{i+1} - Q_i|\\ &\leq \sum (1 - C_i + C_i)|Q_{i+1} - Q_i| = TV(Q^n) \end{aligned}$$

The method

$$Q_i^{n+1} = Q_i^n - \left[\nu + \frac{1}{2}\nu(1-\nu)\left(\frac{\phi(\theta_{i+1/2})}{\theta_{i+1/2}} - \phi(\theta_{i-1/2})\right)\right](Q_i^n - Q_{i-1}^n)$$

is TVD provided

$$0 \le \left[\nu + \frac{1}{2}\nu(1-\nu)\left(\frac{\phi(\theta_1)}{\theta_1} - \phi(\theta_2)\right)\right] \le 1$$

for all values of θ_1 and θ_2 (provided $0 \le \nu \le 1$).

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for all values of θ_1 and θ_2 (provided $0 \le \nu \le 1$).

This is true if

$$-2 \le \left(\frac{\phi(\theta_1)}{\theta_1} - \phi(\theta_2)\right) \le 2$$

for all values of θ_1 and θ_2

So the method

$$Q_i^{n+1} = Q_i^n - \left[\nu + \frac{1}{2}\nu(1-\nu)\left(\frac{\phi(\theta_{i+1/2})}{\theta_{i+1/2}} - \phi(\theta_{i-1/2})\right)\right](Q_i^n - Q_{i-1}^n)$$

is TVD provided

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for all values of θ_1 and θ_2 .

Satisfied provided $\phi(\theta)$ satisfies:

$$0 \le \frac{\phi(\theta)}{\theta} \le 2, \qquad 0 \le \phi(\theta) \le 2,$$

or

$$0 \leq \phi(\theta) \leq \mathsf{minmod}(2, 2\theta).$$

Sweby diagram

If we plot $\phi(\theta)$, the curve must lie in the shaded region:



Sweby diagram

If we plot $\phi(\theta)$, the curve must lie in the shaded region:



Standard second order methods go outside this region.

Recall we want $\phi(1) = 1$ for good accuracy of smooth solutions.

Sweby diagram

Sweby's investigation suggested best methods lie between Lax-Wendroff and Beam-Warming (and inside the TVD region).

Sweby region:



 $\phi(\theta) = \min \operatorname{mond}(1, \theta)$ follows the lower limit of this region.

R. J. LeVeque, University of Washington FVMHP Fig. 6.6

Superbee method

The superbee limiter follows the upper limit:



 $\phi(\theta) = \max(0, \min(1, 2\theta), \min(2, \theta))$

MC method

The Monotonized Centered (MC) limiter follows Fromm's method near $\theta = 1$, and is smooth at $\theta = 1$:



 $\phi(\theta) = \max(0, \min((1+\theta)/2, 2, 2\theta))$

van Leer method

The van Leer limiter is a smoother version of MC



Some popular limiters

Linear methods:

 $\begin{array}{rl} \text{upwind}: & \phi(\theta)=0\\ \text{Lax-Wendroff}: & \phi(\theta)=1\\ \text{Beam-Warming}: & \phi(\theta)=\theta\\ \text{Fromm}: & \phi(\theta)=\frac{1}{2}(1+\theta) \end{array}$

High-resolution limiters:

$$\begin{array}{ll} \mbox{minmod}: & \phi(\theta) = \mbox{minmod}(1,\theta) \\ \mbox{superbee}: & \phi(\theta) = \mbox{max}(0, \mbox{min}(1,2\theta), \mbox{min}(2,\theta)) \\ \mbox{MC}: & \phi(\theta) = \mbox{max}(0, \mbox{min}((1+\theta)/2, \ 2, \ 2\theta)) \\ \mbox{van Leer}: & \phi(\theta) = \frac{\theta + |\theta|}{1 + |\theta|} \end{array}$$