## 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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- Easy to combine with adaptive mesh refinement (AMR)
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- Godunov-type methods based on Riemann solvers Wave-propagation algorithms with "limiters"

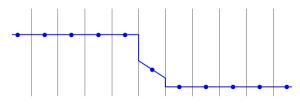
### Limiters can eliminate oscillations

Step function data with minmod slope:

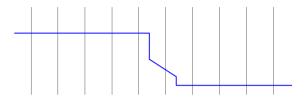


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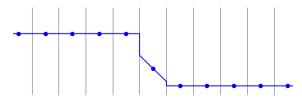


Evolving solution and averaging maintains monotonicity:



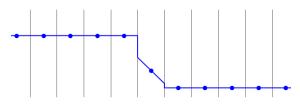
# Could make slope steeper and still be monotone

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

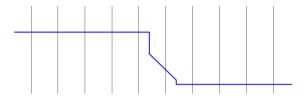


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## Monotonized centered (MC) limiter

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.

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

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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  $TV(\tilde{q}^n) \leq TV(Q^n)$ .

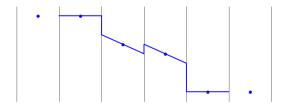
- **2** Evolve the hyperbolic equation exactly (or approximately) with this initial data to obtain  $\tilde{q}^n(x,t_{n+1})$  a time k later.
- 3 Average this function over each grid cell to obtain new cell averages

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

Note: Steps 2 and 3 are always TVD.

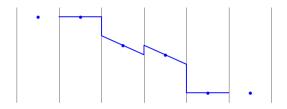
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Sample data with MC slope (twice that of minmod):



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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):

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

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

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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  $\Delta Q^n_{i-1/2}$  by limited version  $\delta^n_{i-1/2}$ 

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### Flux limiters and wave limiters

Flux limiter method: Replace  $\Delta Q^n_{i-1/2}$  by limited version  $\delta^n_{i-1/2}$ 

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### For systems of equations:

 $\bullet\,$  Solve Riemann problem to decompose  $\Delta Q^n_{i-1/2}$  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:

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

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

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Limiter function: Define  $\phi(\theta)$  and then

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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

#### Flux limiter method:

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$$\delta_{i-1/2}^n = \phi(\theta_{i-1/2}^n) \Delta Q_{i-1/2}^n \qquad \text{where} \quad \theta_{i-1/2}^n = \frac{Q_I - Q_{I-1}}{Q_i - Q_{i-1}}$$

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

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

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  $Q^n, \ \Delta x, \ \Delta t.$ 

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

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

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

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

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

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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  $\theta_1$  and  $\theta_2$  (provided  $0 \le \nu \le 1$ ).

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

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  $\theta_1$  and  $\theta_2$  (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  $\theta_1$  and  $\theta_2$ 

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

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

for all values of  $\theta_1$  and  $\theta_2$ .

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

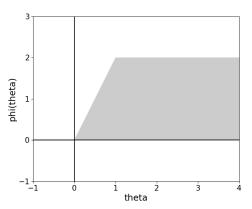
$$0 \le \frac{\phi(\theta)}{\theta} \le 2, \qquad 0 \le \phi(\theta) \le 2,$$

or

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

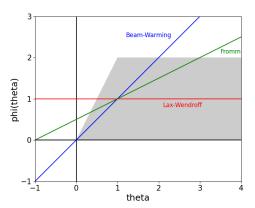
## Sweby diagram

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:



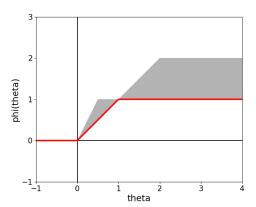
Standard second order methods go outside this region.

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).

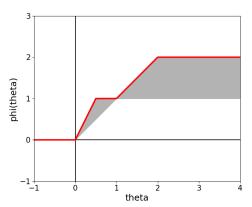
### Sweby region:



 $\phi(\theta) = \mathsf{minmod}(1, \theta)$  follows the lower limit of this region.

# Superbee method

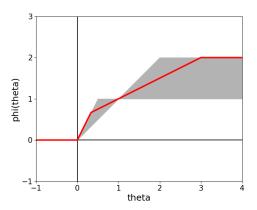
### The superbee limiter follows the upper limit:



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

### MC method

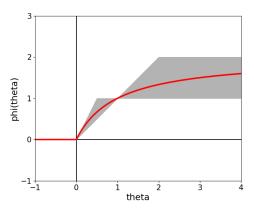
The Monotonized Centered (MC) limiter follows Fromm's method near  $\theta = 1$ , and is smooth at  $\theta = 1$ :



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

### van Leer method

### The van Leer limiter is a smoother version of MC



$$\phi(\theta) = \frac{\theta + |\theta|}{1 + |\theta|}$$
.

# Some popular limiters

#### Linear methods:

$$upwind: \quad \phi(\theta) = 0$$

$${\bf Lax\text{-}Wendroff}: \quad \phi(\theta)=1$$

$$\text{Beam-Warming}: \quad \phi(\theta) = \theta$$

Fromm : 
$$\phi(\theta) = \frac{1}{2}(1+\theta)$$

### **High-resolution limiters:**

$$\mathsf{minmod}: \quad \phi(\theta) = \mathsf{minmod}(1,\theta)$$

$$\mathsf{superbee}: \ \phi(\theta) = \max(0, \, \min(1, 2\theta), \, \min(2, \theta))$$

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

van Leer : 
$$\phi(\theta) = \frac{\theta + |\theta|}{1 + |\theta|}$$