## CLAWPACK

Randy Leveque has developed a general-purpose software package based on precedize-linear FV methods for 10,20, and even 30 hme-dependent conservation laws. We use it more in Amoth 571, but it may be a useful resource for your regearch or other classes, too. See http://www.amath.washington.edu/~claw.

It was used to generate the plots below and on next page.

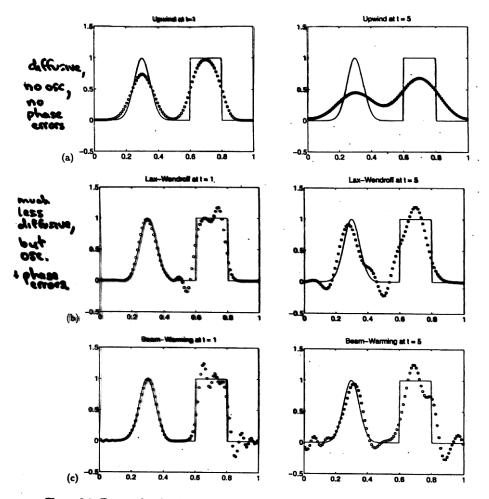


Figure 6.1. Tests on the advection equation with different linear methods. Results at time t=1 and t=5 are shown, corresponding to 1 and 5 revolutions through the domain in which the equation  $q_t+q_x=0$  is solved with periodic boundary conditions. Top: Upwind. Middle: Lax-Wendroff. Bottom: Beam-Warming. [class/booh/chap6/comparants]

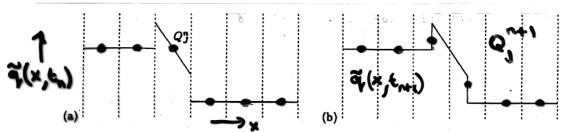


Figure 6.4. (a) Grid values  $Q^n$  and reconstructed  $\tilde{q}^n(\cdot,t_n)$  using Lax-Wendroff slopes. (b) After advection with  $\tilde{u}\Delta t = \Delta x/2$ . The dots show the new cell averages  $Q^{n+1}$ . Note the overshoot.

## Slope-limiters to grantee monotonic pos definiteadvection

A huge advantage of the FV approach is a natural way to suppress

(good for chemical concentrations that physically should remain within given bounds)

Spurious oscillations. Mathematically, we may quantity how much a continuous function  $\psi(x,t)$  oscillates over a region in terms of its

 $TV[\psi] = \int_{\mathbf{p}} \int_{-\infty}^{\infty} |dx| dx$ 



Geometrically, TV is the sum of all the upward and downward excursions of  $\psi(x,t)$  in  $0 \leq x \leq b$ , (I)+(II)+(III) in the example above.

Thus if y is a solution of the advection eqn. Yt + cy = 0, ψ(x, E) = ψ(x-ct, O) [waveforms will just then -

be translated a distance of I so on a periodic domain,

TV [
$$\psi(x,e)$$
] =  $\int_{\alpha}^{b} \psi(x,e)dx = \int_{\alpha}^{b} \psi(x-ce,o)dx = \int_{\alpha}^{b} \psi(x,o)dx$   
[by periodic rearrangement]

is conserved. Similarly on an infinite domain. For a discrete method, the analogous quantity is

$$TV\left[\vec{\phi}^{n}\right] = \sum_{i=1}^{N} |\phi_{i}^{n} - \phi_{i-i}^{n}|$$



A discrete method can be assured of not introducing spurious oscillations that increase TV it it is total-variation diminishing (TVD),

In particular, TVD methods are monotonicity-preserving or "monotonic", 7.e. For monotone unitial data  $\phi_{j}^{n} \geq \phi_{j+1}^{n} \; \forall j, TV[\tilde{\phi}^{n}] = \phi_{N} - \phi_{0}$ 

so it we consider a ramp with \$= \$\phi\_0, 1 \le 0\$ and  $\phi_1 = \phi_{N,1} \ge N$ , and a TVD method

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 $\phi_{j}^{n+1} \rightarrow \phi_{N} \text{ as } j \rightarrow +\infty$   $TV \left[ \phi_{N}^{n+1} \right] = \phi_{N} - \phi_{0} + \text{new internal oscillations}$ so TVD =TV(new internal oscillations) = O.



So, can be construct a FV-REA method that is TVD?

Yes! If the reconstructed function \( \vert(x,t\_n) \) is monotonic between each pair of gridpoints (x,-1,x,), this is a sufficient condition for TVD.

Proof (1) For such a reconstruction

$$\int_{0}^{\infty} \int_{0}^{\infty} \int_{0}^{\infty} \left[ \left( x_{i} + \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n} \left[ \left( \frac{1}{n} \right) - \left( \frac{1}{n} \right) \right] = \sum_{n$$

- (2) Evolvingto \$\psi(x, t\_{n+1})\$ does not change TV since it is just advection of reconstructed profile to right bey cat
- (3) Averaging \( \psi(x, \mathbb{k}\_{n+1}) \) across gridgells \( \mathbb{E}\_{j} \) can only reduce \( \mathbb{T} \mathbb{V} \)

A variety of reconstruction schemes with linear slopes satisfy this constraint, so are TVD, but are also 2nd-order accurate for monotone data. Example:

minmod slope limiter:  $\sigma_{j}^{n} = minmod \left(\frac{\phi_{j}^{n} - \phi_{j-1}^{n}}{\Delta x}, \frac{\phi_{j+1}^{n} - \phi_{j}^{n}}{\Delta x}\right)$ minmod  $(a,b) = \left\{a, \frac{|a| < |b|}{b}, \frac{|a| < |a|}{a,b} \right\}$ i.e. take smaller of slopes

we one-sided slopes

More accurate: MC - limiter

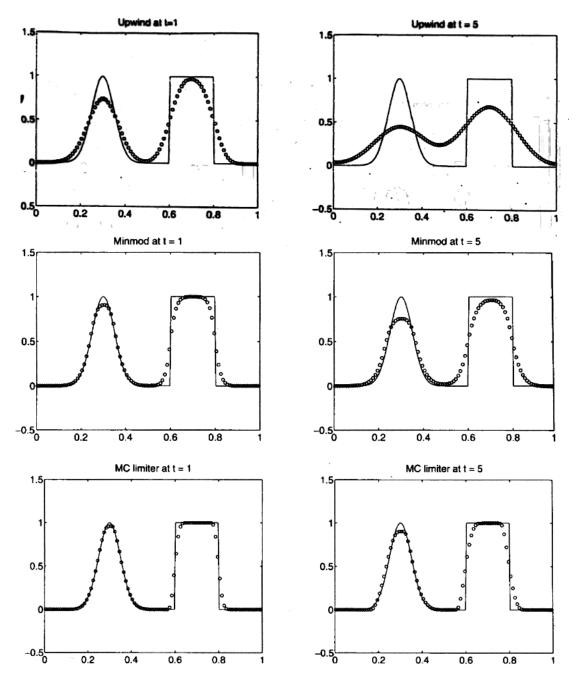
$$\sigma_{j}^{N} = minmod \left( \frac{\varphi_{j+1}^{N} - \varphi_{j-1}^{N}}{2\Delta x} \right) 2 \frac{\varphi_{j}^{N} - \varphi_{j-1}^{N}}{\Delta x} \right) 2 \frac{\varphi_{j+1}^{N} - \varphi_{j}^{N}}{\Delta x}$$

(Monotonized)

one-sided slopes is less than half as large, in which case the centered slope would possibly create a new extremom. At an extremol

Ψ(×, ε, n)

x φ, n × χ



Solutions to advection equation after 1 (left) and 5 (right) time units using different TVD REA methods. The MC slope limiter provides best overall performance.

## Accuracy of slope-limiter methods

On smooth fields such as a sine wave, Lax-Wendroff, Beam-Warming and Fromm are 2<sup>nd</sup>-order accurate. Slope-limiter methods such as MC are more accurate for coarsely-resolved oscillations but have slightly lower-order convergence (~1.7-order for a sinusoid). The top panels of the plot below show this behavior; the plot was generated by script LW\_MC\_errconv.m on the class web page. The left panel shows the reconstructed functional form of the approximate solution using both methods for the coarsest resolution tested. It shows how LW gives a nonmonotonic reconstruction between gridpoints, while MC gives a monotonic reconstruction between gridpoints that guarantees it is TVD.

On fields with sharp discontinuities, MC gives lower rms errors compared to Lax-Wendroff in addition to preventing overshoots. Both methods converge much more slowly, because regardless of resolution the rms errors are dominated by poorly-resolved grid-scale wavenumbers.

