

Algebraic Linearity Preserving Flux Limiting for Systems of PDEs

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Motivation

$$\frac{\partial u}{\partial t} + \nabla \cdot F(u) = 0$$

- Many physically relevant problems
 - Fluids, Drift-diffusion, Species transport
- Stability of solution is challenging
- Galerkin does poorly for shocks and steep gradients



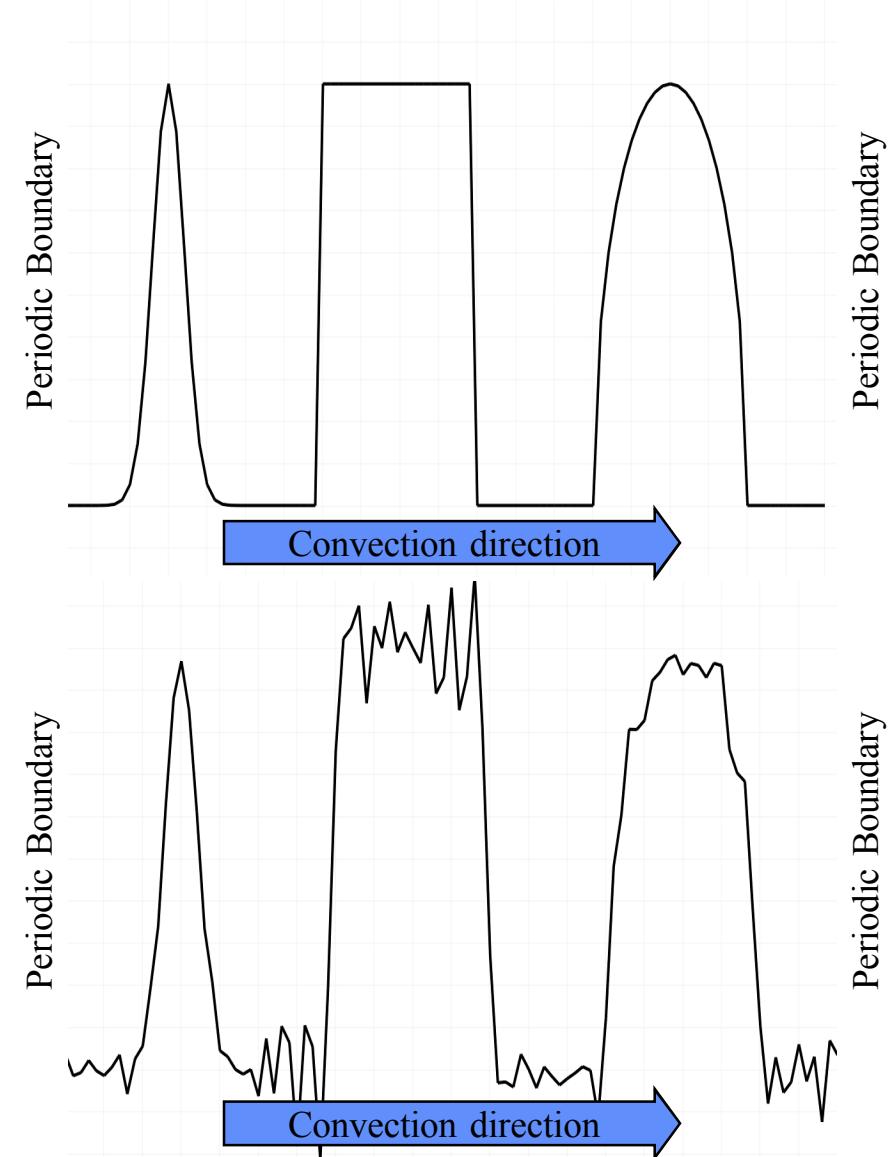
Shu-Osher Problem

1D Transient Problem:

- smooth regions
- shock-like gradients
- sharp peak
- periodic solution

Galerkin solution (1 period):

- Crank-Nicolson
- clearly unstable
- square wave error disrupts hump





A High-Order and Low-Order Discretization

Galerkin semi-discretized problem:

$$M \frac{du}{dt} + Ku = 0$$

Define a “low-order” diffusive problem algebraically:

$$M_L \frac{du}{dt} + \tilde{K}u = 0 \quad \text{where} \quad \tilde{K} = K - D$$

Diffusion operator ‘D’ defined algebraically:

$$d_{ij} = \max(k_{ij}, 0, k_{ji}) \quad \text{and} \quad d_{ii} = - \sum_{j \neq i} d_{ij}$$



Anti-Diffusive Flux

High-order problem is sum of the low-order and
“anti-diffusive flux”

$$M\dot{u} + Ku =$$

$$M_L\dot{u} + \tilde{K}u + (M - M_L)\dot{u} + (K - \tilde{K})u$$

Can rewrite as a sum of numerical fluxes

$$((M - M_L)\dot{u} + (K - \tilde{K})u)_i =$$

$$- \sum_{j \neq i} m_{ij}(\dot{u}_i - \dot{u}_j) - \sum_{j \neq i} d_{ij}(u_i - u_j)$$



Flux Limiting

The flux limited residual is

$$R(u) = M_L \dot{u} + Ku - \bar{f}$$

where

$$\bar{f}_i = - \sum_{j \neq i} \alpha_{ij}^M m_{ij} (\dot{u}_i - \dot{u}_j) - \sum_{j \neq i} \alpha_{ij}^D d_{ij} (u_i - u_j)$$



Flux Limiting

The limiters satisfy

$$0 \leq \alpha_{ij}^M, \alpha_{ij}^D \leq 1$$

and are chosen so the LED criteria is met

$$q_i^M(u_i^{min} - u_i) \leq \sum_{j \neq i} \alpha_{ij}^M m_{ij}(\dot{u}_i - \dot{u}_j) \leq q_i^M(u_i^{max} - u_i)$$

$$q_i^D(u_i^{min} - u_i) \leq \sum_{j \neq i} \alpha_{ij}^D d_{ij}(u_i - u_j) \leq q_i^D(u_i^{max} - u_i)$$

where N_i is the set of nodes neighboring i

$$u_i^{min} = \min_{j \in N_i}(u_j), \quad u_i^{max} = \max_{j \in N_i}(u_j)$$



Flux Limiting: Algorithm

Flux limiting algorithm assures that LED bounds are met:

1. Separate positive and negative anti-diffusive fluxes:

$$P_i^- = \sum_{j \neq i} \min(0, f_{ij}) \quad \text{and} \quad P_i^+ = \sum_{j \neq i} \max(0, f_{ij})$$

2. Build bounding quantities:

$$Q_i^- = q_i(u_i^{min} - u_i) \quad \text{and} \quad Q_i^+ = q_i(u_i^{max} - u_i)$$

3. Build limiting ratios (with symmetry):

$$R_i^- = \min(1, \frac{Q_i^-}{P_i^-}) \quad \text{and} \quad R_i^+ = \min(1, \frac{Q_i^+}{P_i^+})$$

4. Construct limiters:

$$\alpha_{ij} = \begin{cases} \min(R_i^-, R_j^+) & f_{ij} \leq 0 \\ \min(R_i^+, R_j^-) & f_{ij} \geq 0 \end{cases}$$



Linearity Preserving Flux Limiting

The bounds in the LED criteria are designed so that linear functions are not limited (thus the high-order method is used)

$$u_i - u_j = \nabla u_i \cdot (x_i - x_j)$$

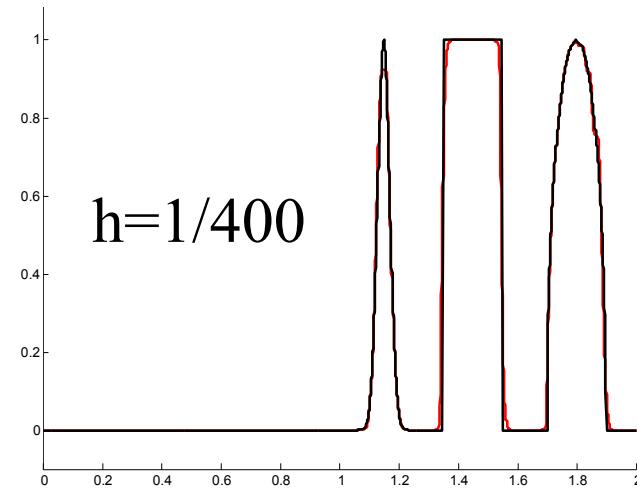
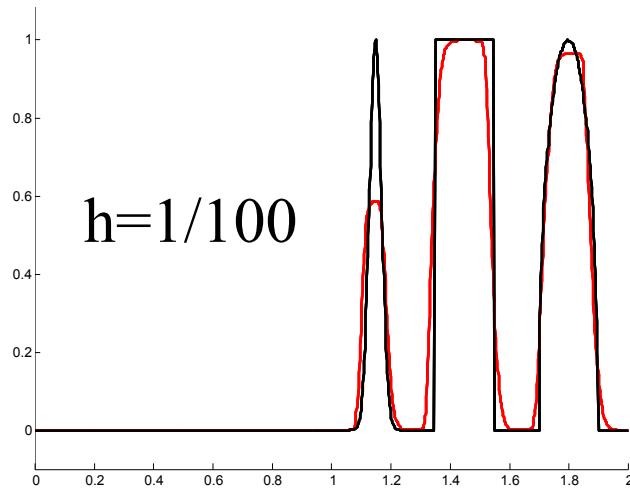
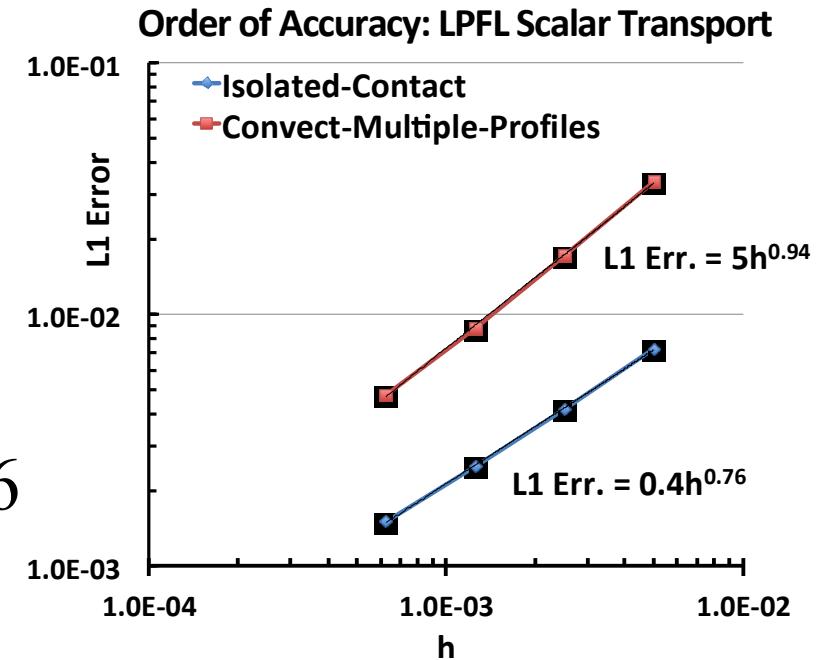
$$= \frac{1}{m_i} \sum_{k \neq i} \int \phi_i \nabla \phi_k (u_k - u_i) \cdot (x_i - x_j)$$

$$\leq \left(\frac{1}{m_i} \sum_{k \neq i} \left| (x_i - x_j) \cdot \int \phi_i \nabla \phi_k \right| \right) (u_i^{\max} - u_i)$$

$$q_i = \sum_{j \neq i} d_{ij} \left(\frac{1}{m_i} \sum_{k \neq i} \left| (x_i - x_j) \cdot \int \phi_i \nabla \phi_k \right| \right)$$

Results for a Scalar PDE

- Two different problems
 1. Isolated Contact
 2. Multiple Profiles
- Spatial convergence rate: 0.76





Coupled Linear PDEs

Use scalar dissipation to construct ‘D’

1. Compute maximum e.v. of \mathbf{K}_{ij} (power method)
2. Construct diffusive operator by

$$\mathbf{D}_{ij} = \lambda_{max}(\mathbf{K}_{ij}) \mathbf{I} \quad \text{and} \quad \mathbf{D}_{ii} = - \sum_{j \neq i} \mathbf{D}_{ij}$$

1. Limiters applied to each coupled entry

$$\sum_{j \neq i} \alpha_{ij} \mathbf{D}_{ij} (U_i - U_j)$$

2. Use synchronized LPFL (D and M-M_L are diagonal)

$$\alpha_{ij} = \min(\alpha_{ij}^{u1}, \alpha_{ij}^{u2}, \dots)$$



A Convected Wave

In primitive variables (two coupled PDEs):

$$\frac{\partial}{\partial t} \begin{bmatrix} u \\ v \end{bmatrix} + \begin{bmatrix} a & c \\ c & a \end{bmatrix} \frac{\partial}{\partial x} \begin{bmatrix} u \\ v \end{bmatrix} = 0$$

In characteristic variables (two decoupled PDEs):

$$\frac{\partial}{\partial t} \begin{bmatrix} q \\ r \end{bmatrix} + \begin{bmatrix} a + c & \\ & a - c \end{bmatrix} \frac{\partial}{\partial x} \begin{bmatrix} q \\ r \end{bmatrix} = 0$$

where

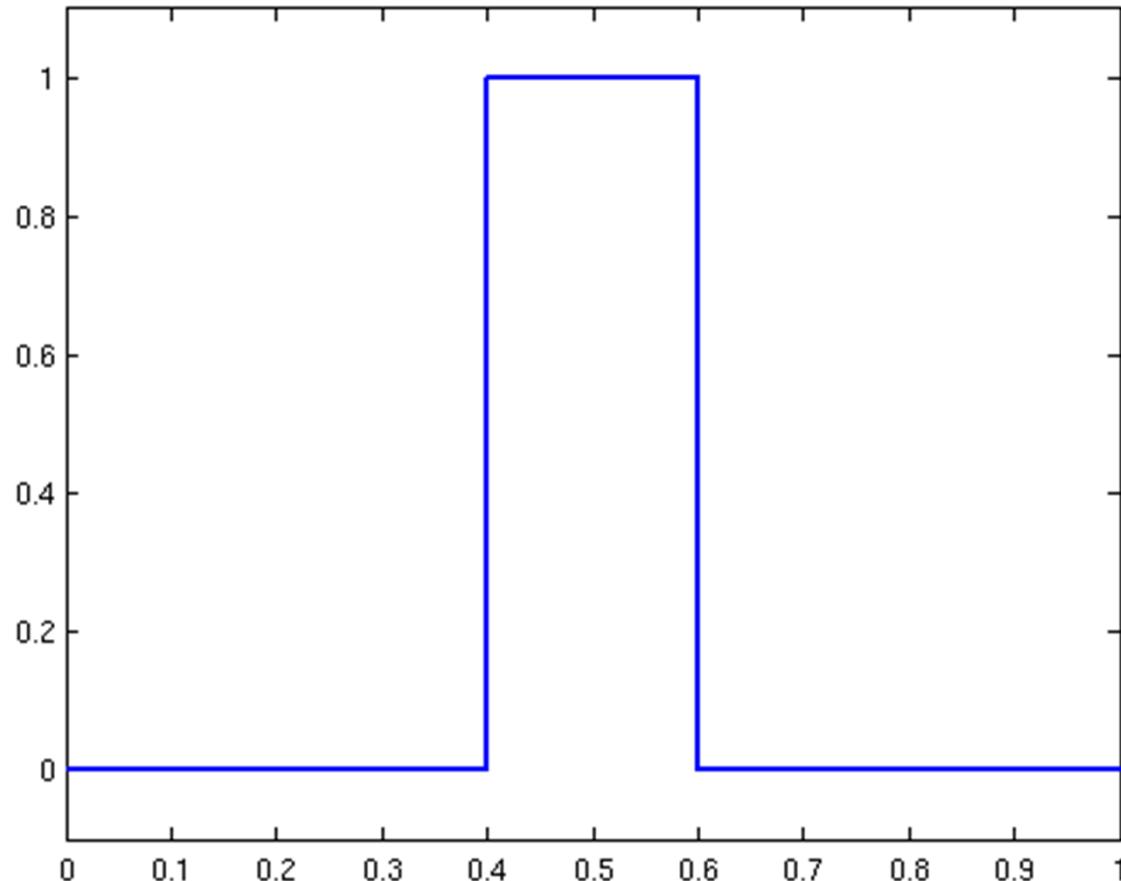
$$\begin{bmatrix} u \\ v \end{bmatrix} = 0.5 \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} q \\ r \end{bmatrix}$$



A Convected Wave

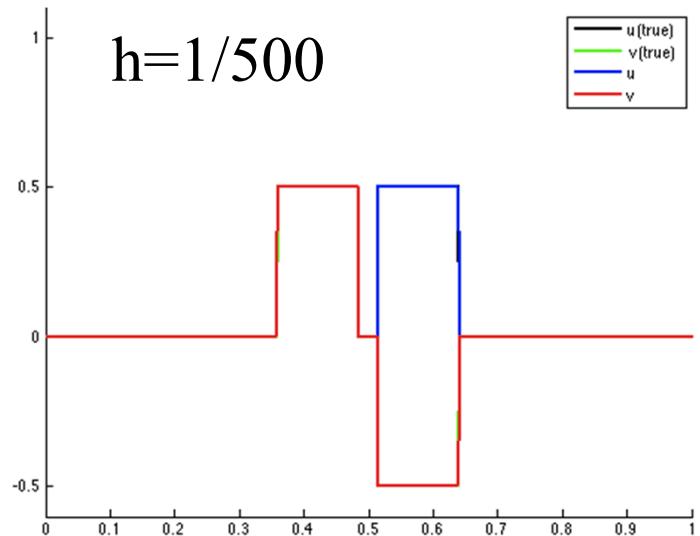
Two speeds seen in plot of 'u':

1. $a+c$ (moving to the right)
2. $a-c$ (moving to the left)

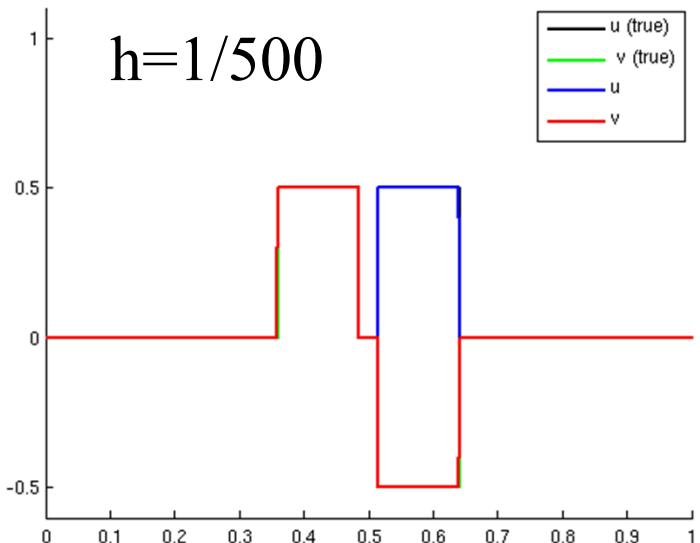


A Convected Wave: Primitive Variables

Limiting primitive variables
with coupled limiting and
scalar dissipation:

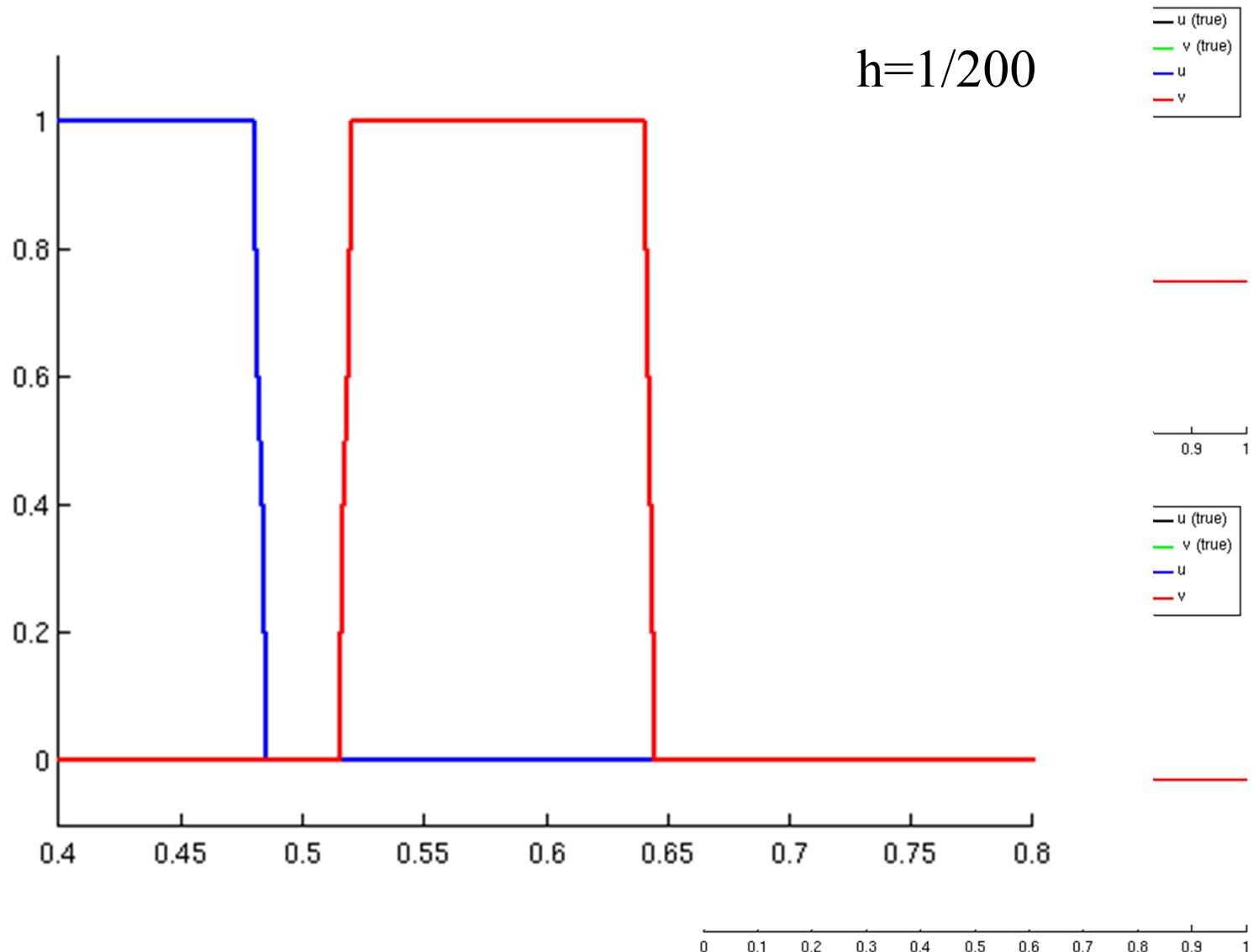


Limiting characteristic
variables with algebraic
linearity preserving FCT:



A Convected Wave: Characteristic Variables

Lim
with
scal

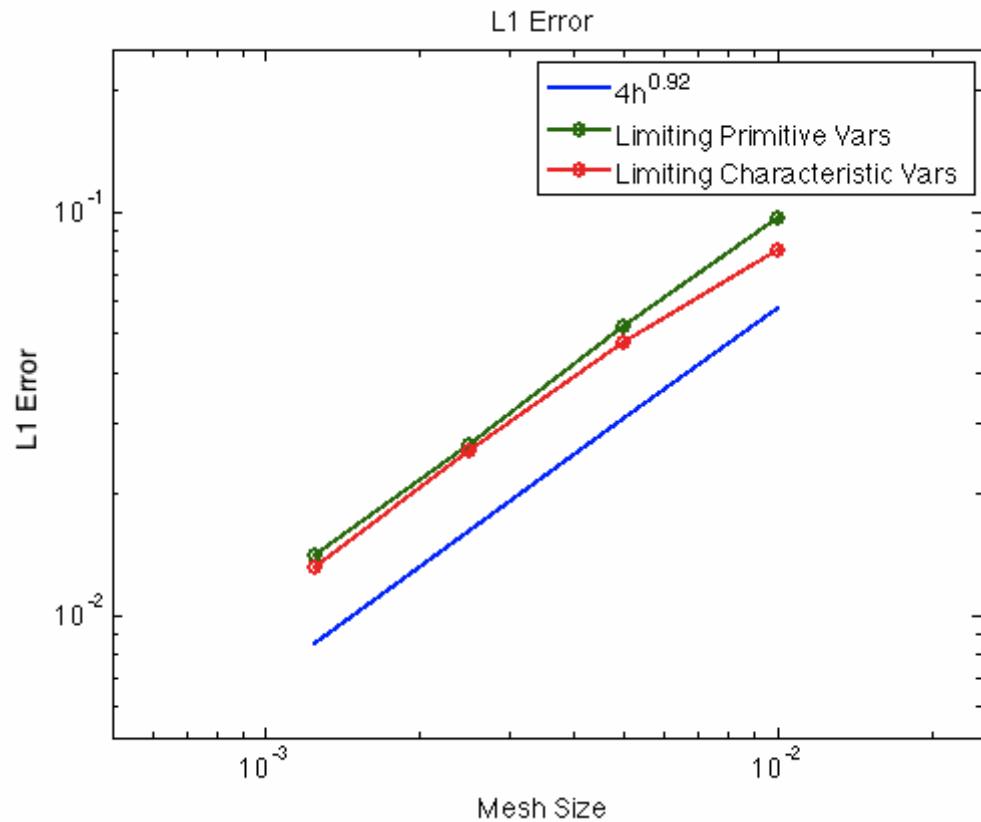


Lim
vari
line

A Convected Wave: Convergence

Convergence rate of 0.9

- Limiting characteristics slightly more accurate
- Over/undershoots characteristics with coupled limiting
- Caveat: Only a linear problem





Summary

- Presented an algebraic framework for flux corrected transport
- Discussed a linearity preserving limiter
- Showed results for a Shu-Osher problem
- Talked about a coupled wave problem:
 - Presented an algebraic diffusion operator based on scalar dissipation
 - Results comparing limiting of characteristic equations and coupled equations