

A Bound-Preserving Fourth Order Compact Finite Difference Scheme for Scalar Convection Diffusion Equations

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1 Introduction

- Bound Preserving
- Conservative Scheme
- Time Discretization
- Monotone Scheme

2 Compact Finite Difference

- Weak Monotonicity

3 Further Extension

- Different Boundary Conditions
- Higher Order Scheme
- Non-uniform Grids

4 Numerical Test

Stability: Compressible Euler Equations in Gas Dynamics

$$\begin{pmatrix} \rho \\ m \\ E \end{pmatrix}_t + \begin{pmatrix} m \\ \rho u^2 + p \\ (E + p)u \end{pmatrix}_x = 0,$$

with

$$m = \rho u, \quad E = \frac{1}{2}\rho u^2 + \rho e, \quad p = (\gamma - 1)\rho e.$$

The speed of sound is given by $c = \sqrt{\gamma p / \rho}$ and the three eigenvalues of the Jacobian are $u, u \pm c$.

If either $\rho < 0$ or $p < 0$, then the sound speed is imaginary and the system is no longer hyperbolic. Thus the initial value problem is ill-posed. This is why it is computationally unstable.

Bound Preserving for Scalar Conservation Laws

Consider the initial value problem

$$u_t + \nabla \cdot \mathbf{F}(u) = 0, u(\mathbf{x}, 0) = u^0(\mathbf{x}), \mathbf{x} \in \mathbb{R}^n$$

for which the unique entropy solution $u(\mathbf{x}, t)$ satisfies

$$\min_{\mathbf{x}} u(\mathbf{x}, t_0) \leq u(\mathbf{x}, t) \leq \max_{\mathbf{x}} u(\mathbf{x}, t_0), \quad \forall t \geq t_0. \quad \text{Maximum Principle}$$

In particular,

$$\min_{\mathbf{x}} u_0(\mathbf{x}) = m \leq u(\mathbf{x}, t) \leq M = \max_{\mathbf{x}} u_0(\mathbf{x}). \quad \text{Bound Preserving}$$

It is also a desired property for numerical solutions due to

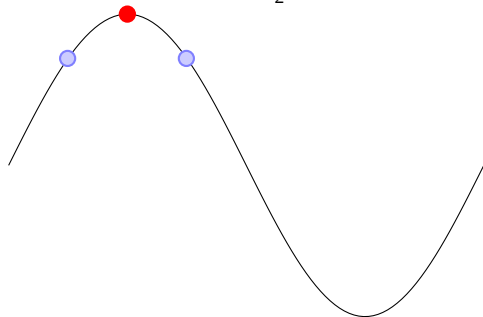
- 1 **Physical meaning:** vehicle density (traffic flow), mass percentage (pollutant transport), probability distribution (Boltzmann equation) and etc.
- 2 **Stability for systems:** positivity of density and pressure (gas dynamics), water height (shallow water equations), particle density for describing electrical discharges (a convection-dominated system) and etc.

For numerical schemes, this is a completely DIFFERENT problem from discrete maximum principle in solving elliptic equations.

Scalar Equations

- IVP: $u_t + f(u)_x = 0$, $u(x, 0) = u^0(x)$.
- Maximum Principle (Bound Preserving): $u(x, t) \in [m, M]$ where $m = \min u_0(x)$, $M = \max u_0(x)$.
- For finite difference, any scheme satisfying $\min_j u_j^n \leq u_j^{n+1} \leq \max_j u_j^n$ can be at most first order accurate.

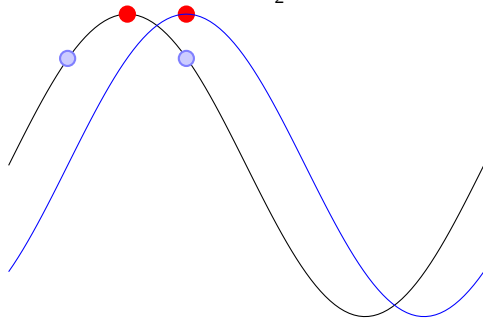
Harten's Counter Example: consider $u_t + u_x = 0$, $u(x, 0) = \sin x$. Put the grids in a way such that $x = \frac{\pi}{2}$ is in the middle of two grid points.



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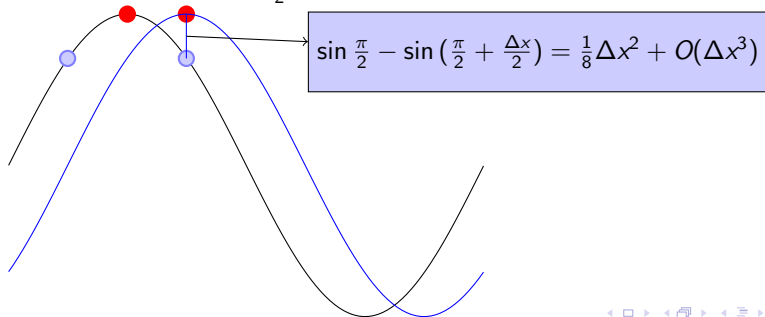
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Bound-Preserving Schemes

- 1 First order monotone schemes.
- 2 FD/FV schemes satisfying $\min_j u_j^n \leq u_j^{n+1} \leq \max_j u_j^n$: can have any formal order of accuracy in the monotone region but are only first order accurate around the extrema. E.g.,
 - Conventional total-variation-diminishing (TVD) schemes.
 - High Resolution schemes such as the MUSCL scheme.
- 3 FV schemes satisfying $\min_x u^n(x) \leq u^{n+1}(x) \leq \max_x u^n(x)$:
 - R. Sanders, 1988: a third order finite volume scheme for 1D.
 - X.Z. and Shu, 2009: higher order (up to 6th) extension of Sanders' scheme.
 - Liu and Osher, 1996: a third order FV scheme for 1D (can be proven bound-preserving only for linear equations).
 - Noelle, 1998; Kurganov and Petrova, 2001: 2D generalization of *Liu and Osher*.
 - All schemes in this category use the exact time evolution.

Bound-Preserving Schemes

Practical/popular high order schemes are NOT bound-preserving.

It was unknown previously how to construct a high order bound-preserving scheme for 2D nonlinear equations.

Conservative Eulerian Schemes

Conservative Schemes: The scheme must have the following form:

$$U_j^{n+1} = U_j^n - \frac{\Delta t}{\Delta x_j} \left[\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}} \right].$$

Global Conservation: $\sum_j \bar{u}_j^{n+1} \Delta x_j = \sum_j \bar{u}_j^n \Delta x_j.$

We insist on using conservative schemes:

- 1 **Lax-Wendroff Theorem:** If converging (as mesh sizes go to zero), the converged solution of a conservative scheme is a weak solution.
- 2 **The shock location will be wrong** if the conservation is violated.
- 3 If a scheme is conservative and positivity preserving, then we have **L1-stability:** $\sum_j |\bar{u}_j^{n+1}| \Delta x_j = \sum_j \bar{u}_j^{n+1} \Delta x_j = \sum_j \bar{u}_j^n \Delta x_j = \sum_j |\bar{u}_j^n| \Delta x_j.$
 - In Euler equations, if density and pressure are positive, then we have L1-stability for density and total energy.
 - **Crude replacement of negative values by positive ones is simply unacceptable and unstable** because it destroys the local conservation.

Objective

We want to construct a scheme which is

- ① genuinely high order accurate (at least third order)
- ② conservative
- ③ positivity/bound preserving
- ④ Practical concern: cost effective, multi-dimensions, unstructured meshes, parallelizability and etc.

Why high order schemes?

- For smooth solutions, the computational cost of higher order schemes is smaller to reach the same accuracy.
- Near shocks, the error in L^∞ norm of any scheme will at most half order. For a lot of problems: accuracy is still high order away from discontinuities.

Motivation: nonlinear stability is one of the main reasons why high order schemes have not been widely used for more real world problems.

Time Discretization: SSP Runge-Kutta or Multi-Step Method

High order strong stability preserving (SSP) Runge-Kutta or multi-step method is a convex combination of several forward Euler schemes. E.g., the 5 stages 4th order SSP Runge-Kutta method for solving $u_t = F(u)$ is given by

$$u^{(1)} = u^n + 0.391752226571890\Delta t F(u^n)$$

$$u^{(2)} = 0.444370493651235u^n + 0.555629506348765u^{(1)} \\ + 0.368410593050371\Delta t F(u^{(1)})$$

$$u^{(3)} = 0.620101851488403u^n + 0.379898148511597u^{(2)} \\ + 0.251891774271694\Delta t F(u^{(2)})$$

$$u^{(4)} = 0.178079954393132u^n + 0.821920045606868u^{(3)} \\ + 0.544974750228521\Delta t F(u^{(3)})$$

$$u^{n+1} = 0.517231671970585u^{(2)} + 0.096059710526147u^{(3)} \\ + 0.063692468666290\Delta t F(u^{(3)}) + 0.386708617503269u^{(4)} \\ + 0.226007483236906\Delta t F(u^{(4)})$$

This method was numerically found to be optimal.

- If the forward Euler is bound-preserving, then so is the high order Runge-Kutta/Multi-Step.
- SSP time discretization has been often used to construct positivity preserving schemes.

First Order Monotone Schemes

Let $\lambda = \frac{\Delta t}{\Delta x}$, a monotone scheme for $u_t + f(u)_x = 0$ is given by

$$\begin{aligned}u_j^{n+1} &= u_j^n - \lambda \left[\widehat{f}(u_j^n, u_{j+1}^n) - \widehat{f}(u_{j-1}^n, u_j^n) \right] \\ &= H(u_{j-1}^n, u_j^n, u_{j+1}^n).\end{aligned}$$

where the numerical flux $\widehat{f}(\uparrow, \downarrow)$ is monotonically increasing w.r.t. the first variable and decreasing w.r.t. the second variable. E.g., the Lax-Friedrichs flux

$$\widehat{f}(u, v) = \frac{1}{2}(f(u) + f(v) - \alpha(v - u)), \alpha = \max_u |f'(u)|.$$

If $m \leq u_j^n \leq M$ for all j , then

$$u_j^{n+1} = H(u_{j-1}^n, u_j^n, u_{j+1}^n) = H(\uparrow, \uparrow, \uparrow) \tag{1}$$

implies

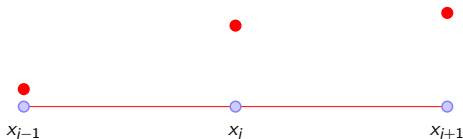
$$m = H(m, m, m) \leq u_j^{n+1} \leq H(M, M, M) = M.$$

The Weighting Operator for Convection

If we regard W as an operator mapping a vector to another vector, then

$$(Wu)_j = \frac{1}{6}u_{j+1} + \frac{4}{6}u_j + \frac{1}{6}u_{j-1},$$

which happens to be the Simpson's rule (or 3-point Gauss-Lobatto Rule) in quadrature.



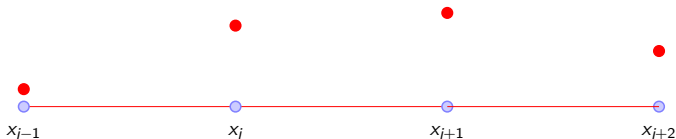
Locally, for each interval $[x_{j-1}, x_{j+1}]$, there exists a cubic polynomial $p_j(x)$, obtained through interpolation at $x_{j-1}, x_j, x_{j+1}, x_{j+2}$ (or $x_{j-2}, x_{j-1}, x_j, x_{j+1}$)

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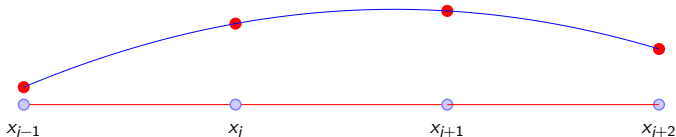
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The Fourth Order Compact Finite Difference Scheme for Convection

Let $\bar{u}_i = (Wu)_i = \frac{1}{6}u_{i-1} + \frac{4}{6}u_i + \frac{1}{6}u_{i+1}$. The fourth order compact finite difference for $u_t + f(u)_x = 0$ can be written as

$$\bar{u}_i^{n+1} = \bar{u}_i^n - \frac{\Delta t}{\Delta x} \frac{1}{2} [f(u_{i+1}^n) - f(u_{i-1}^n)] = \bar{u}_i^n - \frac{\Delta t}{\Delta x} (\hat{f}_{i+\frac{1}{2}}^n - \hat{f}_{i-\frac{1}{2}}^n),$$
$$\hat{f}_{i+\frac{1}{2}}^n = \frac{1}{2} (f(u_{i+1}) + f(u_i)).$$

The **weak monotonicity** holds under the CFL constraint $\lambda \max_u |f'(u)| \leq \frac{1}{3}$:

$$\begin{aligned}\bar{u}_i^{n+1} &= \frac{1}{6}u_{i-1}^n + \frac{4}{6}u_i^n + \frac{1}{6}u_{i+1}^n + \frac{1}{2}\lambda [f(u_{i+1}^n) - f(u_{i-1}^n)] \\ &= \frac{1}{6}[u_{i-1} - 3\lambda f(u_{i-1}^n)] + \frac{1}{6}[u_{i+1}^n + 3\lambda f(u_{i+1}^n)] + \frac{4}{6}u_i^n \\ &= H(u_{i-1}^n, u_i^n, u_{i+1}^n) = H(\uparrow, \uparrow, \uparrow).\end{aligned}$$

Thus $m \leq u_i^n \leq M$ implies $m = H(m, m, m) \leq \bar{u}_i^{n+1} \leq H(M, M, M) = M$.

How to enforce bounds?

Consider enforcing the positivity (general lower/upper bounds can be similarly treated). Given point values u_i , if $\frac{1}{6}u_{i-1} + \frac{4}{6}u_i + \frac{1}{6}u_{i+1} \geq 0$ for any i , then the following hold:

- For any three consecutive points, at least one point value is non-negative.
- If $u_i < 0$, local average is non-negative: $\frac{1}{3}u_{i-1} + \frac{1}{3}u_i + \frac{1}{3}u_{i+1} \geq 0$. This means that it is possible to eliminate the undershoot without changing $\frac{1}{3}u_{i-1} + \frac{1}{3}u_i + \frac{1}{3}u_{i+1}$.
- If $u_i < 0$, then either $u_i + \frac{1}{2}u_{i+1} \geq 0$ or $u_i + \frac{1}{2}u_{i-1} \geq 0$.

So a simple local limiter can enforce the positivity (or general lower/upper bounds):

$$\begin{array}{ll} \text{if } u_{i+1} \geq u_{i-1}, & \text{set } u_{i+1} \leftarrow u_{i+1} + u_i, u_i \leftarrow 0, \\ \text{if } u_{i+1} < u_{i-1}, & \text{set } u_{i-1} \leftarrow u_{i-1} + u_i, u_i \leftarrow 0. \end{array}$$



2D Convection

Let $u_{i,j}$ denote point values on a 2D uniform grid. Define two weighting operators:

$$(W_x u)_{i,j} = \frac{1}{6} u_{i-1,j} + \frac{4}{6} u_{i,j} + \frac{1}{6} u_{i+1,j},$$

$$(W_y u)_{i,j} = \frac{1}{6} u_{i,j-1} + \frac{4}{6} u_{i,j} + \frac{1}{6} u_{i,j+1}.$$

Let \bar{u} denote $W_x W_y u$. The fourth order compact finite difference for $u_t + f(u)_x + g(u)_y = 0$ can be written as

$$\bar{u}_{i,j}^{n+1} = \bar{u}_{i,j}^n + \frac{\Delta t}{\Delta x} \frac{1}{2} W_y [f(u_{i+1,j}^n) - f(u_{i-1,j}^n)] + \frac{\Delta t}{\Delta y} \frac{1}{2} W_x [g(u_{i,j+1}^n) - g(u_{i,j-1}^n)]$$

$\bar{u}_{i,j}^{n+1}$ has weak monotonicity thus $u_{i,j}^n \geq 0$ implies $\bar{u}_{i,j}^{n+1} \geq 0$.

A dimension by dimension fashion 2D limiter:

- Introduce a new variable $v = W_x u$. So $\bar{u} = W_y v$.
- Given $\bar{u} \geq 0$, apply 1D limiter in y-direction on v to enforce $v \geq 0$.
- Given $v \geq 0$, apply 1D limiter in x-direction on u to enforce $u \geq 0$.

$$\frac{1}{12}(u''_{i+1} + 10u''_i + u''_{i-1}) = \frac{u_{i+1} - 2u_i + u_{i-1}}{\Delta x^2} + \mathcal{O}(\Delta x^4)$$

Let $\tilde{u}_i = (Wu)_i = \frac{1}{12}(u_{i-1} + 10u_i + u_{i+1})$. The fourth order compact finite difference for $u_t = g(u)_{xx}$ can be written as

$$\tilde{u}_i^{n+1} = \tilde{u}_i^n + \frac{\Delta t}{\Delta x^2} [g(u_{i+1}^n) - 2g(u_i^n) + g(u_{i-1}^n)],$$

Assuming $g'(u) \geq 0$. The weak monotonicity holds under the CFL constraint $\frac{\Delta t}{\Delta x^2} \max_u |f'(u)| \leq \frac{1}{6}$. The same 1D limiter can be used for enforcing bounds.

Remarks

For approximating the 2D Laplacian, the fourth order compact finite difference scheme is also known as the 9-point discrete Laplacian (the second order centered difference is 5-point Laplacian). For solving heat equation, Crank-Nicolson time discretization and the 9-point Laplacian give an M-matrix, thus it can preserve bounds/positivity. However, such a result cannot be easily extended to nonlinear cases.

For the equation $u_t + f(u)_x = g(u)_{xx}$, define two weighting operators:

$$(W_1 u)_j = \frac{1}{6}(u_{j+1} + 4u_j + u_{j-1}),$$

$$(W_2 u)_j = \frac{1}{12}(u_{j+1} + 10u_j + u_{j-1}).$$

Let $\tilde{u}_i = (W_1 W_2 u)_i = (W_2 W_1 u)_i$, then the scheme is

$$\tilde{u}_j^{n+1} = \tilde{u}_j^n - \lambda \frac{1}{2} W_2 [f(u_{j+1}^n) - f(u_{j-1}^n)] + \mu W_1 [g(u_{j+1}^n) - 2g(u_j^n) + g(u_{j-1}^n)]$$

\tilde{u}_j^{n+1} has weak monotonicity, thus $u_j^n \geq 0$ implies $\tilde{u}_j^{n+1} \geq 0$.

The same 1D limiter can be applied twice to for enforcing the bound.

- The result is still true if adding the TVB limiter in Bernardo Cockburn and Chi-Wang Shu. Nonlinearly stable compact schemes for shock calculations. SIAM Journal on Numerical Analysis, 31(3):607-627, 1994.
- Using both TVB limiter and the simple local bound-preserving limiter can eliminate overshoot/undershoot and oscillations without losing accuracy/conservation.

2D Incompressible Flow

The stream function formulation of 2D incompressible Navier-Stokes equations:

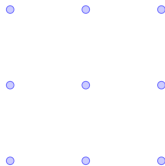
$$\omega_t + (u\omega)_x + (v\omega)_y = \Delta\omega$$

$$u = -\psi_y, v = \psi_x$$

$$\psi = \Delta\omega$$

The weak monotonicity holds. But to have $H(M, M, M) = M$, we need a discrete incompressible condition:

$$\frac{1}{\Delta x} \left[\frac{1}{6} u_{i+1,j-1} + \frac{4}{6} u_{i+1,j} + \frac{1}{6} u_{i+1,j+1} - \frac{1}{6} u_{i-1,j-1} - \frac{4}{6} u_{i-1,j} - \frac{1}{6} u_{i-1,j+1} \right]$$
$$+ \frac{1}{\Delta y} \left[\frac{1}{6} v_{i+1,j+1} + \frac{4}{6} v_{i,j+1} + \frac{1}{6} v_{i-1,j+1} - \frac{1}{6} v_{i+1,j-1} - \frac{4}{6} v_{i,j-1} - \frac{1}{6} v_{i-1,j-1} \right] = 0$$



Inflow-outflow boundary conditions

For the equation $u_t + f(u)_x = 0$ on $[0, 1]$ with initial value $u(x, 0) = u^0(x)$ and inflow boundary condition $u(0, t) = g(t)$. We have the 4th-order compact finite difference

$$\frac{d}{dt} \frac{1}{6} \begin{pmatrix} 1 & 4 & 1 & & & & & & \\ & 1 & 4 & 1 & & & & & \\ & & 1 & 4 & 1 & & & & \\ & & & \ddots & \ddots & \ddots & & & \\ & & & & 1 & 4 & 1 & & \\ & & & & & 1 & 4 & 1 & \end{pmatrix} \begin{pmatrix} u_0 \\ u_1 \\ u_3 \\ \vdots \\ u_N \\ u_{N+1} \end{pmatrix} = -\frac{1}{2\Delta x} \begin{pmatrix} -1 & 0 & 1 & & & & & & \\ & -1 & 0 & 1 & & & & & \\ & & -1 & 0 & 1 & & & & \\ & & & \ddots & \ddots & \ddots & & & \\ & & & & -1 & 0 & 1 & & \\ & & & & & -1 & 0 & 1 & \end{pmatrix} \begin{pmatrix} f_0 \\ f_1 \\ f_2 \\ \vdots \\ f_N \\ f_{N+1} \end{pmatrix}$$

Inflow-outflow boundary conditions

With forward euler time discretization the scheme is

$$\bar{u}_i^{n+1} = \bar{u}_i^n - \frac{\Delta t}{2\Delta x}(f_{i+1} - f_{i-1}), 1 \leq i \leq N.$$

Since u_0 is known, to solve the previous linear system, we need to find a good approximation of u_{N+1} . A cubic polynomial $p_j(x)$, obtained through enforcing its integral equals \bar{u}_i over interval $[x_{i-1}, x_{i+1}]$ for $i = N-3, N-2, N-1, N$. We get

$$u_{N+1} = -\frac{2}{3}\bar{u}_{N-3} + \frac{17}{6}\bar{u}_{N-2} + \frac{14}{3}\bar{u}_{N-1} + \frac{7}{2}\bar{u}_N.$$

A filter

$$u_{N+1} = \max(\min(u_{N+1}, ubound), lbound)$$

should be used to make $u_{N+1} \in [m, M]$. Next we can compute $u_i, 1 \leq i \leq N$. Apply the previous bound-preserving limiter to u_i to make sure $u_i \in [m, M]$. Then we can update the value of $\bar{u}_i, 1 \leq i \leq N$ and compute the solution at next level.

Dirichlet boundary conditions

Consider $u_t + f(u)_x = g(u)_{xx}$ on $[0, L]$ with initial value $u(x, 0) = u^0(x)$ and Dirichlet boundary condition $u(0, t) = L(t)$, $u(L, t) = R(t)$. Define $W := W_1 W_2 = W_2 W_1$, we will have

$$\frac{d}{dt}(WU) + W_2 D_x f + W_2 F = W_1 D_{xx} g + W_1 G. \quad (2)$$

and

$$-W_2 F + W_1 G = \begin{pmatrix} -\frac{1}{18} u_{t,0} + \frac{1}{12} f_{x,0} + \frac{5}{12\Delta x} f_0 + \frac{2}{3\Delta x^2} g_0 \\ -\frac{1}{72} u_{t,0} + \frac{1}{24\Delta x} f_0 + \frac{1}{6\Delta x^2} g_0 \\ 0 \\ \vdots \\ 0 \\ -\frac{1}{72} u_{t,N+1} - \frac{1}{24\Delta x} f_{N+1} + \frac{1}{6\Delta x^2} g_{N+1} \\ -\frac{1}{18} u_{t,N+1} + \frac{1}{12} f_{x,N+1} - \frac{5}{12\Delta x} f_{N+1} + \frac{2}{3\Delta x^2} g_{N+1} \end{pmatrix} \quad (3)$$

Dirichlet boundary conditions

We need to approximate $f(u)_{x,0}$ and $f(u)_{x,N+1}$. But to have weak monotonicity, only 3rd-order approximation can be made.

$$f(u)_{x,0} = \frac{1}{\Delta x} \left(-\frac{11}{6} f_0 + 3f_1 - \frac{3}{2} f_2 + \frac{1}{3} f_3 \right)$$

and

$$f(u)_{x,N+1} = \frac{1}{\Delta x} \left(-\frac{1}{3} f_{N-2} + \frac{3}{2} f_{N-1} - 3f_N + \frac{11}{6} f_{N+1} \right).$$

Then we plug these approximations in and rearrange (2). Finally we have the space discretization

$$\frac{d}{dt} \tilde{W}\tilde{u} = -\tilde{D}_x \tilde{f} + \tilde{D}_{xx} \tilde{g}.$$

$$\tilde{W} = \frac{1}{72} \begin{pmatrix} \frac{24}{5} & \frac{246}{5} & \frac{84}{5} & \frac{6}{5} & & & \\ 1 & 14 & 42 & 14 & 1 & & \\ & \ddots & \ddots & \ddots & \ddots & \ddots & \\ & & 1 & 14 & 42 & 14 & 1 \\ & & & \frac{6}{5} & \frac{84}{5} & \frac{246}{5} & \frac{24}{5} \end{pmatrix}_{N \times (N+2)}, \tilde{u} = \begin{pmatrix} u_0 \\ u_1 \\ \vdots \\ u_N \\ u_{N+1} \end{pmatrix}_{(N+2) \times 1}, \quad (4)$$

$$\tilde{D}_x = \frac{1}{24\Delta x} \begin{pmatrix} -\frac{38}{5} & -\frac{42}{5} & \frac{78}{5} & \frac{2}{5} & & & \\ -1 & -10 & 0 & 10 & 1 & & \\ & \ddots & \ddots & \ddots & \ddots & \ddots & \\ & & -1 & -10 & 0 & 10 & 1 \\ & & & -\frac{2}{5} & -\frac{78}{5} & \frac{42}{5} & \frac{38}{5} \end{pmatrix}_{N \times (N+2)}, \tilde{f} = \begin{pmatrix} f_0 \\ f_1 \\ \vdots \\ f_N \\ f_{N+1} \end{pmatrix}_{(N+2) \times 1}, \quad (5)$$

$$\tilde{D}_{xx} = \frac{1}{6\Delta x^2} \begin{pmatrix} \frac{24}{5} & -\frac{42}{5} & \frac{12}{5} & \frac{6}{5} & & & \\ 1 & 2 & -6 & 2 & 1 & & \\ & \ddots & \ddots & \ddots & \ddots & \ddots & \\ & & 1 & 2 & -6 & 2 & 1 \\ & & & \frac{6}{5} & \frac{12}{5} & -\frac{42}{5} & \frac{24}{5} \end{pmatrix}_{N \times (N+2)}, \tilde{g} = \begin{pmatrix} g_0 \\ g_1 \\ \vdots \\ g_N \\ g_{N+1} \end{pmatrix}_{(N+2) \times 1}, \quad (6)$$

Weak monotonicity and consistency hold. The previous limiter can be applied to preserve the bound. [2D case can be treated similarly.](#)

6th and 8th Order Scheme

6th order first derivative approximation with the weak monotonicity:

$$\beta f'_{i-2} + \alpha f'_{i-1} + f'_i + \alpha f'_{i+1} + \beta f'_{i+2} = b \frac{f_{i+2} - f_{i-2}}{4\Delta x} + a \frac{f_{i+1} - f_{i-1}}{2\Delta},$$

$$\beta = \frac{1}{12}(-1 + 3\alpha), \quad a = \frac{2}{9}(8 - 3\alpha), \quad b = \frac{1}{18}(-17 + 57\alpha), \quad \alpha > \frac{1}{3}.$$

If we let $\alpha = \frac{4}{9}$, $\beta = \frac{1}{36}$, $a = \frac{40}{27}$, $b = \frac{25}{54}$, we actually have the 8th order with the weak monotonicity.

6th-order second derivative approximation with the weak monotonicity:

$$\beta f''_{i-2} + \alpha f''_{i-1} + f''_i + \alpha f''_{i+1} + \beta f''_{i+2} = b \frac{f_{i+2} - 2f_i + f_{i-2}}{4\Delta x^2} + a \frac{f_{i+1} - 2f_i + f_{i-1}}{\Delta^2},$$

$$a = \frac{-78\alpha + 48}{31}, \quad b = \frac{291\alpha - 36}{62}, \quad \alpha > 0.$$

If we let $\alpha = \frac{344}{1179}$, we actually have the 8th order scheme with the weak monotonicity.

Non-uniform Grids

Consider a mapping for the physical non-uniform grid points x_j on the interval $[a, b]$ to uniform reference grid points ξ_j on the same interval $[a, b]$. Let $\xi(x)$ denote such a map and $\frac{d\xi(x)}{dx} \neq 0$. Then a convection diffusion equation in the form of

$$u_t + f(u)_x = g(u)_{xx}$$

can be written as:

$$\frac{\partial}{\partial t} \hat{u} + \left(\frac{d\xi}{dx} - \frac{d^2\xi}{dx^2} \right) \frac{\partial}{\partial \xi} f(\hat{u}) = \left(\frac{d\xi}{dx} \right)^2 \frac{\partial^2}{\partial \xi^2} g(\hat{u}).$$

Therefore, it is equivalent to solve the following equation

$$u_t + c(x)f(u)_x = d(x)g(u)_{xx}, \quad d(x) > 0.$$

Accuracy Test

SSP 4th-order multistep for 2D linear convection:

Mesh	L^∞ error	order	L^1 error	order
20×20	5.81E-5	-	3.68E-5	-
40×40	3.52E-6	4.05	2.25E-6	4.03
80×80	2.17E-7	4.02	1.38E-7	4.03
160×160	1.35E-8	4.01	8.57E-9	4.01

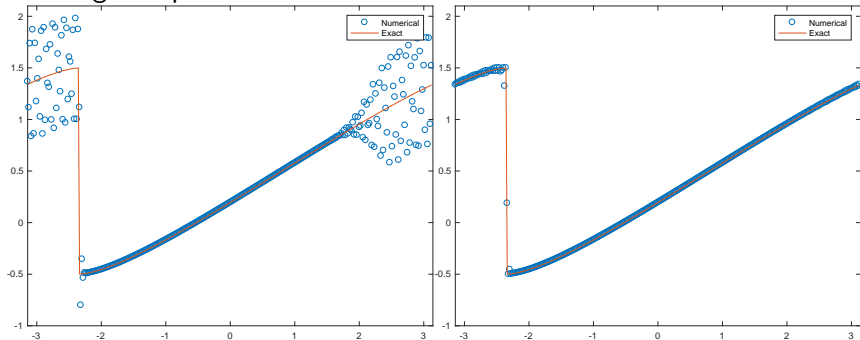
SSP RK4 for 1D linear convection:

Mesh	L^∞ error	order	L^1 error	order
40	2.72E-5	-	1.67E-5	-
80	1.77E-6	3.95	1.08E-6	3.94
160	2.52E-7	2.81	8.37E-8	3.69
320	1.17E-7	1.11	1.20E-8	2.81

SSP 4th-order multistep for 1D convection-diffusion equation:

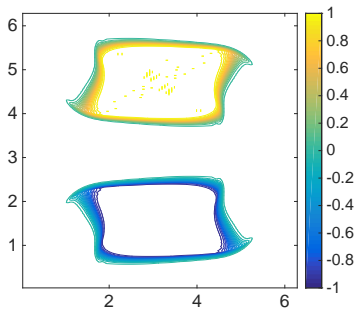
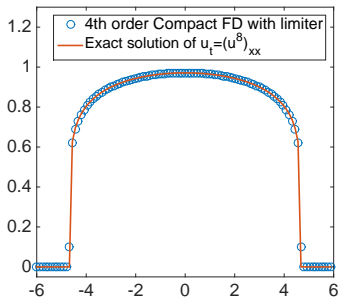
Mesh	L^∞ error	order	L^1 error	order
40	1.55E-6	-	9.92E-7	-
80	9.72E-8	4.00	6.19E-8	4.00
160	6.07E-9	4.00	3.87E-9	4.00
320	3.76E-10	4.01	2.40E-10	4.01

1D burgers equation when the exact solution has a shock



Left: numerical solutions without bound-preserving limiter.

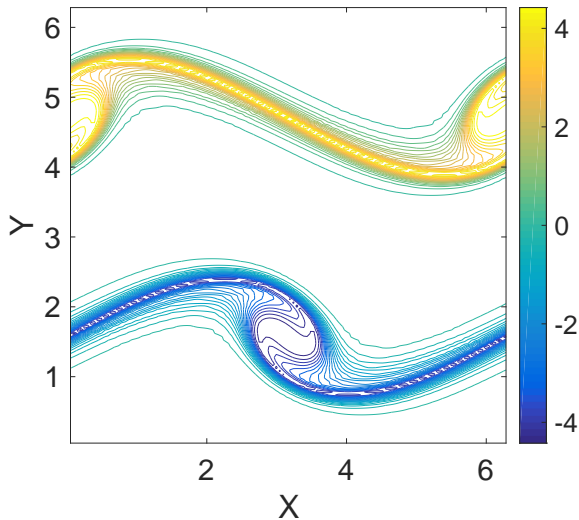
Right: numerical solutions with bound-preserving limiter.



Left: degenerate nonlinear parabolic equation $u_t = (u^8)_{xx}$, the numerical solution is strictly nonnegative.

Right: the vortex patch test for 2D incompressible NS equation with Reynolds number 3000, the solution is strictly bounded by 1 and -1 .

2D Incompressible flow with Reynolds number 3000: double shear layer



The solution is strictly in the bound.

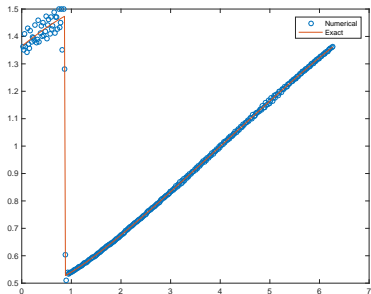
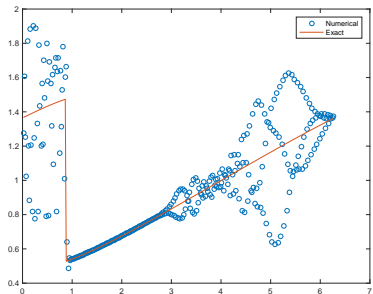
8th order scheme for linear convection equation:

Mesh	L^∞ error	order	L^1 error	order
20	3.14E-6	-	1.99E-6	-
40	7.13E-8	5.46	4.55E-8	5.45
80	2.71E-10	8.04	1.73E-10	8.04
160	1.43E-12	7.57	7.24E-12	7.90

8th order scheme for burgers equation:

Mesh	L^∞ error	order	L^1 error	order
20	1.18E-3	-	3.37E-4	-
40	2.57E-5	5.52	2.23E-6	7.24
80	5.23E-8	8.94	2.97E-9	9.56
160	1.77E-10	8.21	9.90E-12	8.23

Inflow-outflow Boundary condition:



Burgers equation $u_t + \left(\frac{1}{2}u^2\right)_{xx} = 0$, when the solution is positive .

Left: numerical solution without bound-preserving limiter

Right: numerical solution with bound-preserving limiter. The solution is strictly bounded by 0.5 and 1.5.

SSP 4th-order multistep for 1D convection diffusion equation with Dirichlet boundary condition:

Mesh	L^∞ error	order	L^1 error	order
40	3.79E-5	-	2.03E-5	-
80	3.16E-6	3.58	1.16E-6	4.13
160	3.79E-7	3.06	1.26E-7	3.20
320	4.81E-8	2.98	1.86E-8	2.75
640	1.26E-9	5.25	4.72E-10	5.31

Generalization to Systems?

Try to use the idea of FV?

The End