

SF2520 — Applied numerical methods

Lecture 13

Numerical methods for hyperbolic equations

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Today's lecture

- Finite difference methods for hyperbolic equations
- Numerical analysis, smooth solutions
- Numerical analysis, discontinuous solutions

Hyperbolic PDEs

We want to solve the advection equation

$$u_t + au_x = 0, \quad x \in \mathbb{R}, \quad t > 0,$$

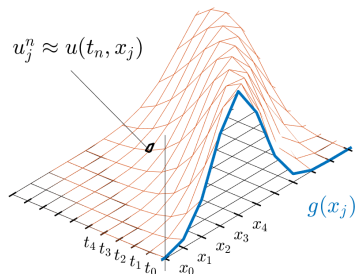
$$u(x, 0) = g(x),$$

numerically. (No boundary for now.)

- Semi-discretizations (MoL) path not so practical:
 - Most common methods not on this form.
 - Natural discretizations lead to unstable methods.
 - Absolute stability of MoL often not sufficient for stability of full discretization.
- Instead do full discretization directly,

$$x_j = j\Delta x, \quad t_n = n\Delta t,$$

$$u_j^n \approx u(x_j, t_n).$$



We want to solve the advection equation

$$\begin{aligned}u_t + au_x &= 0, & x \in \mathbb{R}, \quad t > 0, \\u(x, 0) &= g(x).\end{aligned}$$

- Introduce the CFL number λ as the relation between time and spatial step,

$$\lambda = \frac{\Delta t}{\Delta x} \quad \left(\sigma = \frac{a\Delta t}{\Delta x} \text{ in the book.}\right)$$

- In hyperbolic problems:
 - λ is held constant (typically) as Δt and Δx are refined.
(I.e. not $\Delta t/\Delta x^2$ as in parabolic case with explicit methods.)
 - Discretizations are **not stiff** so explicit methods are most often used.
(I.e. not implicit methods as in parabolic case.)

Examples of methods

1 A cautionary example

Consider the natural discretization

$$\underbrace{\frac{u_j^{n+1} - u_j^n}{\Delta t}}_{\approx u_t} + a \underbrace{\frac{u_{j+1}^n - u_{j-1}^n}{2\Delta x}}_{\approx u_x} = 0,$$

or

$$u_j^{n+1} = u_j^n - a \frac{\lambda}{2} (u_{j+1}^n - u_{j-1}^n).$$

- Based on forward differences in time and central differences in space ("FTCS")
- **This method is unstable for all fixed $\lambda = \Delta t / \Delta x$.** It cannot be used to solve the advection equation.

2 Lax–Friedrichs

Replace u_j^n in the previous method by the average of u_{j+1}^n and u_{j-1}^n ,

$$u_j^{n+1} = \frac{1}{2}(u_{j+1}^n + u_{j-1}^n) - a\frac{\lambda}{2}(u_{j+1}^n - u_{j-1}^n).$$

- This is stable for $|a|\lambda \leq 1$.

③ Upwind

Replace central differences in unstable method by backward/forward differences ("FTBS"/"FTFS")

$$\frac{u_j^{n+1} - u_j^n}{\Delta t} + \begin{cases} a \frac{u_j^n - u_{j-1}^n}{\Delta x}, & a > 0, \\ a \frac{u_{j+1}^n - u_j^n}{\Delta x}, & a < 0, \end{cases} = 0$$

or

$$u_j^{n+1} = \begin{cases} u_j^n - a\lambda(u_j^n - u_{j-1}^n), & a > 0, \\ u_j^n - a\lambda(u_{j+1}^n - u_j^n), & a < 0. \end{cases}$$

- This is stable for $|a|\lambda \leq 1$.
- Change makes method first order in space. Not big problem since it is only first order in time anyway.
- When $a > 0$ the approximation u_j^{n+1} only depends on the values in the points to the left: u_j^n and u_{j-1}^n . This mimics the exact solution where $u(x, t) = g(x - at)$ only depends on values to the left.
- This property is important when discontinuous solutions are computed. (It is e.g. not true for Lax–Friedrichs.)

❹ Lax–Wendroff

Subtract a small difference approximating u_{xx}

$$\frac{u_j^{n+1} - u_j^n}{\Delta t} + a \frac{u_{j+1}^n - u_{j-1}^n}{2\Delta x} - \underbrace{\Delta t \frac{a^2}{2} \frac{u_{j+1}^n - 2u_j^n + u_{j-1}^n}{\Delta x^2}}_{\approx \Delta t \frac{a^2}{2} u_{xx}} = 0,$$

or

$$u_j^{n+1} = u_j^n - a \frac{\lambda}{2} (u_{j+1}^n - u_{j-1}^n) + a^2 \frac{\lambda^2}{2} (u_{j+1}^n - 2u_j^n + u_{j-1}^n).$$

- This is stable for $|a|\lambda \leq 1$.
- Added term vanishes as $\Delta t \rightarrow 0$ but precisely cancels the leading part of error. (Note: $u_t = -au_x \Rightarrow u_{tt} = -au_{xt} = a^2 u_{xx}$.)
- Makes the method second order in time and space!

Linear systems of equations: $\mathbf{u}_t + A\mathbf{u}_x = 0$

- ② **Lax–Friedrichs** (as scalar case, just replace $a \rightarrow A$ and $u_j^n \rightarrow \mathbf{u}_j^n$)

$$\mathbf{u}_j^{n+1} = \frac{1}{2}(\mathbf{u}_{j+1}^n + \mathbf{u}_{j-1}^n) - \frac{\lambda}{2}A(\mathbf{u}_{j+1}^n - \mathbf{u}_{j-1}^n).$$

- ③ **Upwind**

Split A -matrix into positive and negative parts

$$A = S\Lambda S^{-1}, \quad A^+ = S\Lambda^+ S^{-1}, \quad A^- = S\Lambda^- S^{-1},$$

where Λ^\pm is diagonal with negative/positive eigenvalues replaced by zero. (Note: $\Lambda = \Lambda^+ + \Lambda^-$ and $A = A^+ + A^-$.) Then do backward difference with A^+ and forward difference with A^- :

$$\mathbf{u}_j^{n+1} = \mathbf{u}_j^n - \lambda A^+(\mathbf{u}_j^n - \mathbf{u}_{j-1}^n) - \lambda A^-(\mathbf{u}_{j+1}^n - \mathbf{u}_j^n).$$

- ④ **Lax–Wendroff** (as scalar case, just replace $a \rightarrow A$ and $u_j^n \rightarrow \mathbf{u}_j^n$)

$$\mathbf{u}_j^{n+1} = \mathbf{u}_j^n - \frac{\lambda}{2}A(\mathbf{u}_{j+1}^n - \mathbf{u}_{j-1}^n) + \frac{\lambda^2}{2}A^2(\mathbf{u}_{j+1}^n - 2\mathbf{u}_j^n + \mathbf{u}_{j-1}^n).$$

- These methods are stable if $\max |\mu_k| \lambda \leq 1$, where μ_k = eigenvalues of A .

Linear systems with source functions: $u_t + Au_x = f$

Let $f_j^n := f(x_j, t_n)$. Then

- ② **Lax–Friedrichs** (just add a term at the end)

$$u_j^{n+1} = \frac{1}{2}(u_{j+1}^n + u_{j-1}^n) - \frac{\lambda}{2}A(u_{j+1}^n - u_{j-1}^n) + \Delta t f_j^n$$

- ③ **Upwind** (just add a term at the end)

$$u_j^{n+1} = u_j^n - \lambda A^+(u_j^n - u_{j-1}^n) - \lambda A^-(u_{j+1}^n - u_j^n) + \Delta t f_j^n$$

- ④ **Lax–Wendroff** (more complicated)

$$\begin{aligned} u_j^{n+1} &= u_j^n - \frac{\lambda}{2}A(u_{j+1}^n - u_{j-1}^n) + \frac{\lambda^2}{2}A^2(u_{j+1}^n - 2u_j^n + u_{j-1}^n) + \Delta t \tilde{f}_j^n \\ \tilde{f}_j^n &= f_j^n - \frac{\lambda}{4}A(f_{j+1}^n - f_{j-1}^n) + \frac{1}{2}(f_j^{n+1} - f_j^n) \end{aligned}$$

Nonlinear systems of equations: $\mathbf{u}_t + \mathbf{F}(\mathbf{u})_x = 0$

- ② **Lax–Friedrichs** (as scalar case, just replace $au \rightarrow \mathbf{F}(\mathbf{u})$ and $u_j^n \rightarrow \mathbf{u}_j^n$)

$$\mathbf{u}_j^{n+1} = \frac{1}{2}(\mathbf{u}_{j+1}^n + \mathbf{u}_{j-1}^n) - \frac{\lambda}{2}(\mathbf{F}(\mathbf{u}_{j+1}^n) - \mathbf{F}(\mathbf{u}_{j-1}^n)).$$

- ③ **Upwind**

Much more complicated. In the nonlinear case the upwind scheme is referred to as the Godunov scheme.

- ④ **Lax–Wendroff**

More complicated. A two-step approach typically used.

Numerical analysis of hyperbolic PDEs

For hyperbolic equations we are concerned with convergence in two different settings:

- Convergence for smooth solutions.
This is governed by the usual order of accuracy and local truncation errors.
- Convergence for non-smooth, discontinuous, solutions.
Important since discontinuous solutions are common in applications. Governed by "modified equations".

Computer experiments with Lax-Friedrichs (LxF), Upwind and Lax-Wendroff (LW).

Conclusions:

- Smooth solutions: LW much more accurate than LxF and Upwind.
- Non-smooth solutions: Upwind best. Smears discontinuities less than LxF. LW introduces a lot of spurious oscillations.
- All three methods stable iff $|a|\lambda \leq 1$.

Theory for smooth solutions

Define the local truncation error (LTE), denoted ℓ_j^n , as the residual when the exact solution is entered into the scheme.

Example: Lax–Friedrichs

$$u(x_j, t_{n+1}) = \frac{u(x_{j+1}, t_n) + u(x_{j-1}, t_n)}{2} - a\Delta t \frac{u(x_{j+1}, t_n) - u(x_{j-1}, t_n)}{2\Delta x} + \underbrace{\ell_j^n}_{\text{LTE}}.$$

Convergence conditions are given by the Lax Equivalence Theorem, which in this case implies

Theorem (Lax)

Suppose $\lambda = \Delta t / \Delta x$ is fixed as $\Delta t \rightarrow 0$. Then

if the method is $|u_j^n - u(x_j, t_n)| = O(\Delta t^p),$

- Consistent with order p : $|\ell_j^n| = O(\Delta t^{p+1})$ with $p \geq 1$,
- Stable: $\sum_j (u_j^n)^2 \leq C \sum_j (u_j^0)^2$, with C independent of Δt and $n \leq \frac{T}{\Delta t}$.

Theory for smooth solutions

Need to check:

- Consistency/accuracy

- Done by Taylor expansions
- Derivations give:

Lax–Friedrichs $\ell_j^n = O\left(\Delta t(\Delta t + \Delta x^2 + \frac{\Delta x^2}{\Delta t})\right)$

Upwind $\ell_j^n = O(\Delta t(\Delta t + \Delta x))$

Lax–Wendroff $\ell_j^n = O(\Delta t(\Delta t^2 + \Delta x^2))$

- Hence, if $\Delta t/\Delta x = \text{fixed}$:

Lax–Friedrichs $|u_j^n - u(x_j, t_n)| = O(\Delta x),$ order of accuracy 1

Upwind $|u_j^n - u(x_j, t_n)| = O(\Delta x),$ order of accuracy 1

Lax–Wendroff $|u_j^n - u(x_j, t_n)| = O(\Delta x^2),$ order of accuracy 2

- Stability

Checked by

- CFL condition \Rightarrow necessary conditions
- von Neumann analysis \Rightarrow sufficient conditions (next course)

Stability – the CFL condition

CFL condition

For any consistent three-point scheme of the form

$$u_j^{n+1} = c_0 u_{j-1}^n + c_1 u_j^n + c_2 u_{j+1}^n, \quad u_j^0 = g(x_j),$$

the condition

$$|a|\lambda \leq 1$$

is **necessary** for stability.

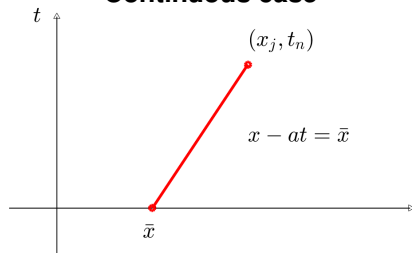
- This condition is also sufficient for Lax–Friedrichs, Upwind and Lax–Wendroff. (Although not for the unstable FTCS scheme.)
- For systems of d equations it is necessary that

$$|\mu_k|\lambda \leq 1, \quad k = 1, \dots, d,$$

where μ_k are the eigenvalues of the system matrix A . (Again sufficient for the methods mentioned above.)

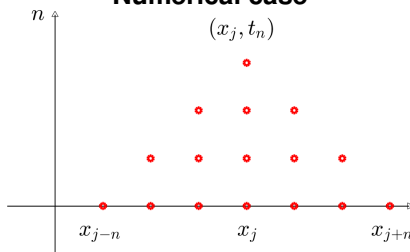
Stability – the CFL condition

Continuous case



Exact solution $u(x_j, t_n)$ depends on value of g in the point \bar{x} , only.
(= domain of dependence)

Numerical case



Numerical solution u_j^n depends on value of g in the points x_{j-n}, \dots, x_{j+n} .
(= numerical domain of dependence)

- The method is unstable if $\bar{x} \notin [x_{j-n}, x_{j+n}]$ because then it has no way of "knowing" the correct solution value $g(\bar{x}) \Rightarrow$ need

$$x_{j-n} \leq \bar{x} \leq x_{j+n} \Rightarrow x_j - n\Delta x \leq x_j - an\Delta t \leq x_j + n\Delta x$$

$$\Rightarrow -\Delta x \leq -a\Delta t \leq \Delta x \Rightarrow |a| \frac{\Delta t}{\Delta x} \leq 1.$$

- Require: domain of dependence \subset numerical domain of dependence.

Theory for non-smooth solutions

- For each method one can introduce a **modified equation** for which the method is more accurate, and the solution is smooth.
- The idea is to derive a precise expression for the local truncation error, and add this (scaled by Δt) to the original equation. Then the new local truncation error is smaller.
- **Example:** Lax–Friedrichs has local truncation error

$$\ell_j^n = \Delta t \frac{a}{2} \Delta x \left(\frac{1}{a\lambda} - a\lambda \right) u_{xx}(x_j, t_n) + O(\Delta t^3).$$

- Therefore, let v solve the modified PDE

$$v_t + av_x = \frac{a}{2} \Delta x \left(\frac{1}{a\lambda} - a\lambda \right) v_{xx}.$$

- Lax–Friedrichs is now a **second order** accurate approximation of v ,

$$|u_j^n - u(x_j, t_n)| = O(\Delta t), \quad |u_j^n - v(x_j, t_n)| = O(\Delta t^2)$$

and the difference $u - v$ describes the leading order error term in u_j^n .
(Also note that the v_{xx} term makes solution smooth.)

Modified equations for the three schemes above are:

$$v_t + av_x = \frac{a}{2} \Delta x \left(\frac{1}{a\lambda} - a\lambda \right) v_{xx}, \quad \text{Lax-Friedrichs,}$$

$$v_t + av_x = \frac{|a|}{2} \Delta x (1 - |a|\lambda) v_{xx}, \quad \text{Upwind,}$$

$$v_t + av_x = \frac{a}{2} \Delta x^2 \left((a\lambda)^2 - 1 \right) v_{xxx}, \quad \text{Lax-Wendroff.}$$

Modified equations

Modified equations for Lax–Friedrichs and Upwind are advection-diffusion equations of the form

$$v_t + av_x = \varepsilon v_{xx},$$

with $\varepsilon = \varepsilon_L$ and $\varepsilon = \varepsilon_U$ respectively

$$\varepsilon_L = \frac{a}{2} \Delta x \left(\frac{1}{a\lambda} - a\lambda \right), \quad \varepsilon_U = \frac{|a|}{2} \Delta x (1 - |a|\lambda).$$

- Schemes are unstable if $\varepsilon < 0$ = "backward heat equation". Stability condition $\varepsilon \geq 0$ is again $|a|\lambda \leq 1$.
- Solution to the advection-diffusion equation given by

$$v(x, t) = w(x - at, t), \quad w_t = \varepsilon w_{xx}, \quad w(x, 0) = g(x).$$

- Diffusion coefficient ε determines amount of smearing of discontinuities. (Exact u has $\varepsilon = 0$.) Upwind smears less than Lax–Friedrichs since

$$1 - |a|\lambda < \frac{1}{a\lambda} - a\lambda$$

- Larger $|a|\lambda$ gives less smearing.
- $a\lambda = 1$ special "magic time step" $\Delta t = \Delta x/a$. Then $\varepsilon = 0$. No smearing occurs, but only works for simple scalar constant coefficient problems.

Modified equations

Modified equation for Lax–Wendroff is a dispersive advection equation (linear KdV) of the form

$$v_t + av_x = \varepsilon v_{xxx},$$

with

$$\varepsilon = \frac{a}{2} \Delta x^2 \left((a\lambda)^2 - 1 \right).$$

- Here $\varepsilon = O(\Delta x^2)$, much smaller than for Lax–Friedrichs and Upwind.
- Stability condition not seen directly. (Need even more precise modified equation which also includes v_{xxxx} .)
- Solution to the dispersive equation given by

$$v(x, t) = w(x - at, t), \quad w_t = \varepsilon w_{xxx}, \quad w(x, 0) = g(x).$$

- Coefficient ε determines amount of dispersion seen at discontinuities.
- Larger $|a|\lambda$ gives less dispersion. Magic time step also here.