

A Matrix View of Two-Level Stencil Schemes

Numerical solution vectors:

\vec{v}_ℓ ↘

$$\vec{v}_\ell = \begin{bmatrix} u_{1,\ell} \\ \vdots \\ u_{N_x,\ell} \end{bmatrix}, \quad \mathbf{v} = \begin{bmatrix} \mathbf{v}_1 \\ \vdots \\ \mathbf{v}_{N_t} \end{bmatrix}.$$

True solution vectors:

$$\mathbf{u}_\ell = \begin{bmatrix} u(x_1, t_\ell) \\ \vdots \\ u(x_{N_x}, t_\ell) \end{bmatrix}, \quad \mathbf{u} = \begin{bmatrix} \mathbf{u}_1 \\ \vdots \\ \mathbf{u}_{N_t} \end{bmatrix}.$$

- hw1 now due Feb 11
- hw2 out before
Wed
- quiz?

Definition (Two-Level Finite Difference Scheme)

A finite difference scheme that can be written as

$$P_h \vec{v}_{\ell+1} = Q_h \vec{v}_\ell + h_t \mathbf{b}_\ell$$

is called a **two-level linear finite difference scheme**.

- Mostly $\mathbf{b}_\ell = \vec{0}$
- $P_h(h_x, h_t) \quad Q_h(h_x, h_t)$

Rewriting Schemes in Matrix Form (1/2)

$$P_h \mathbf{v}_{l+1} = Q_h \mathbf{v}_l + h_t \mathbf{b}_l'$$

Find P_h and Q_h for ETCS:

$$\text{ETCS: } \frac{u_{k,l+1} - u_{k,l}}{h_t} + a \frac{u_{k+1,l} - u_{k-1,l}}{2h_x} = 0$$

$$u_{k,l+1} = u_{k,l} + \frac{ah_t}{2h_x} (u_{k+1,l} - u_{k-1,l})$$

$$P_h = I \quad / \quad Q_h = \text{tridiag} \left(\frac{ah_t}{2h_x}, 1, -\frac{ah_t}{2h_x} \right)$$

Rewriting Schemes in Matrix Form (2/2)

Find P_h and Q_h for Crank-Nicolson:

$$P_h = \text{tridiag} \left(\frac{-ah_t}{4h_x}, 1, \frac{ah_t}{4h_x} \right)$$

$$Q_h = \text{tridiag} \left(\frac{ah_t}{4h_x}, 1, -\frac{ah_t}{4h_x} \right)$$

Truncation Error

Definition (Truncation Error)

The local truncation error $\tau_{k,e}$ is the error that remains when the FD method is applied to an exact sol.

Demo: Truncation Error Analysis via sympy [cleared]

Error and Error Propagation

$$P_n \vec{v}_{l+1} = Q_n \vec{v}_l$$

Express definition of truncation error in our two-level framework:

$$P_h \vec{u}_{l+1} = Q_h \vec{u}_l + \underbrace{\vec{\tau}_e}_\text{trunc. error} h_t$$

Define $\mathbf{e}_l = \mathbf{u}_l - \mathbf{v}_l$. Understand the error as accumulation of truncation error:

$$\begin{aligned} \mathbf{e}_0 &= \mathbf{0} \\ P_h \vec{e}_{l+1} &= Q_h \vec{e}_l + \vec{\tau}_e h_t \quad \left| \begin{array}{l} P_h \vec{u}_{l+1} = Q_h \vec{u}_l + \vec{\tau}_e h_t \\ \hline P_h \vec{v}_{l+1} = Q_h \vec{v}_l \end{array} \right. \\ \vec{e}_{l+1} &= P_h^{-1} Q_h \vec{e}_l + P_h^{-1} \vec{\tau}_e h_t \end{aligned}$$

Discrete and Continuous Norms

To measure properties of numerical solutions we need **norms**. Define a discrete L^∞ norm.

$$\|\vec{e}\|_\infty = \max_{k,l} |e_{k,l}|$$

Define a discrete L^2 norm.

$$\|\vec{e}\|_2 = \sqrt{\sum_{k,l} h_x h_t e_{k,l}^2}$$

\vec{e}

$$\|v\|_2 = \sqrt{\sum_{i=1}^n v_i^2}$$

$v \in \mathbb{R}^n$

l^2

Important features:

$$\|f\| = \sqrt{\int f^2}$$

- $\|\vec{e}\|_{2/\infty}$ should not change wildly as we consider limits $h_x, h_t \rightarrow 0$

Consistency and Convergence

Assume $u, (\partial_x^{q_x})u, (\partial_t^{q_t})u \in L^2(\mathbb{R} \times [0, t^*])$.

Definition (Consistency)

A two-level scheme is **consistent** in the L^2 -norm with order q_t in time and q_x in space if

$$\max_{\ell, \ell h_\ell \leq t^*} \|\tilde{\tau}_\ell\| = O(h_x^{q_x} + h_\ell^{q_\ell})$$

(as $h_x, h_\ell \rightarrow 0$)

Definition (Convergence)

A two-level scheme is **convergent** in the L^2 -norm with order q_t in time and q_x in space if

$$\max_{\ell, \ell h_\ell \leq t^*} \|\tilde{e}_\ell\| = O(h_x^{q_x} + h_\ell^{q_\ell})$$

(as $h_x, h_\ell \rightarrow 0$)

Analyzing ETFS (1/2)

$$\frac{u_{k,l+1} - u_{k,l}}{h_t} + a \frac{u_{k+1,l} - u_{k,l}}{h_x} = 0$$

Let's understand more precisely what happens for this scheme. Assume

$a > 0$

Rewrite

$$u_{k,l+1} = u_{k,l} - \frac{a h_t}{h_x} (u_{k+1,l} - u_{k,l}) = (1+\lambda) u_{k,l} - \lambda u_{k+1,l}$$

Analyzing ETFS (2/2)



$$u_{k,l+1} = (1 + \lambda)u_{k,l} - \lambda u_{k+1,l}$$

Consider $u(x, 0) = 1_{[-1,0]}(x)$. Predict solution behavior.

$$u_{0,0} = 1 \quad u_{1,\dots,0} = 0$$

$$u_{0,1} = 1 + \lambda \quad u_{1,\dots,1} = 0$$

$$u_{0,2} = (1 + \lambda)^2 \quad u_{1,\dots,2} = 0$$

⋮

⋮

$$\left(1 + \frac{\lambda}{n}\right)^n \rightarrow e^\lambda$$

$$u(0,t) \approx u_{0,t/h_x} = \left(1 + \frac{\lambda h_x}{h_x}\right)^{t/h_x} = \left(1 + \frac{\lambda t/h_x}{t/h_x}\right)^{t/h_x} \rightarrow \exp\left(\frac{\lambda t}{h_x}\right)$$

Demo: Methods for 1D Advection [cleared] (Revisit ETFS)

Stability

$$P_h \mathbf{v}_{\ell+1} = Q_h \mathbf{v}_\ell$$

Write down a matrix product to bring \mathbf{v}_0 to \mathbf{v}_ℓ :

$$\vec{v}_\ell = (P_h^{-1} Q_h)^\ell \vec{v}_0$$

Definition (Stability)

A two-level scheme is **stable** in the L^2 -norm if there exists a constant $c > 0$ independent of h_t and h_x so that

$$\left\| (P_h^{-1} Q_h)^\ell P_h^{-1} \right\| \leq c$$

for all ℓ and h_t such that $\ell h_t \leq t^*$.

Lax Convergence Theorem

Theorem (Lax Convergence)

If a two-level FD scheme is

- ▶ **consistent** in the L^2 -norm with order q_t in time and q_x in space, and
- ▶ **stable** in the L^2 -norm, then

it is **convergent** in the L^2 -norm with order q_t in time and q_x in space.

- Stronger result holds: "if and only if"
Lax equiv. thm. / Lax-Richtmyer thm.
- "fundamental thm of numerical analysis"
consistency + stability \Rightarrow convergence
- A related result holds for ODEs
(Dahlquist)

Lax Convergence: Proof (1/2)

Lax Convergence: Proof (2/2)

$$\|e_\ell\| \leq h_t \sum_{m=1}^{\ell} \|(P_h^{-1} Q_h)^{\ell-m} P_h^{-1}\| \tau^{m-1}.$$

Conditions for Stability

$$\left\| (P_h^{-1} Q_h)^\ell P_h^{-1} \right\| \leq c$$

Give a simpler, sufficient condition:

How can we show bounds on these matrix norms?