

### Definition 5.7: Two-Level Linear Finite-Difference Scheme

A finite-difference scheme that can be written as,

$$P_h \mathbf{u}_{\ell+1} = Q_h \mathbf{u}_\ell + h_t \mathbf{b}_\ell, \quad (5.5)$$

is called a two-level linear finite-difference scheme. Each iteration depends only on two instances of time. Examples are given in Example 5.8.

ETBS: 
$$u_{k,\ell+1} - u_{k,\ell} + a \frac{u_{k,\ell} - u_{k-1,\ell}}{h_x} = 0$$

$$u_{k,\ell+1} - u_{k,\ell} + \frac{ah_t}{h_x} (u_{k,\ell} - u_{k-1,\ell}) = 0$$

$$u_{k,\ell+1} = \left(1 - \frac{ah_t}{h_x}\right) u_{k,\ell} + \frac{ah_t}{h_x} u_{k-1,\ell}$$

$$\Rightarrow P_n = I$$

$$Q_n = \text{tridiag} \left( \frac{ah_t}{h_x}, 1 - \frac{ah_t}{h_x}, 0 \right)$$

$u_{k+1,\ell}$

**Definition 5.10: Truncation Error**

The local truncation error,  $\tau_{k,\ell}$ , is the error that remains when a finite-difference method is applied to the exact solution,  $u(x_k, t_\ell)$ .

**Example 5.12: ETFS Truncation Error**

$$\begin{aligned}\tau_{k,\ell} &= \frac{u(x_k, t_{\ell+1}) - u(x_k, t_\ell)}{h_t} + a \frac{u(x_{k+1}, t_\ell) - u(x_k, t_\ell)}{h_x} \\ &= \frac{1}{h_t} \left( u(x_k, t_\ell) + u_t(x_k, t_\ell)h_t + u_{tt}(x_k, \varsigma) \frac{h_t^2}{2} - u(x_k, t_\ell) \right) \\ &\quad + \frac{a}{h_x} \left( u(x_k, t_\ell) + u_x(x_k, t_\ell)h_x + u_{xx}(\xi^+, t_\ell) \frac{h_x^2}{2} - u(x_k, t_\ell) \right) \\ &= u_{tt}(x_k, \varsigma) \frac{h_t}{2} + a u_{xx}(\xi^+, t_\ell) \frac{h_x}{2} \\ &= \mathcal{O}(h_t, h_x)\end{aligned}$$

### Example 5.14: ETCS Truncation Error (Matrix Form)

For ETCS,  $P_h = I$  and

$$(P_h \mathbf{U}_{\ell+1})_k = u(x_k, t_{\ell+1}) = u(kh_x, (\ell+1)h_t),$$

$$(Q_h \mathbf{U}_\ell)_k = u(kh_x, \ell h_t) - \frac{ah_t}{2h_x} \left( u((k+1)h_x, \ell h_t) - u((k-1)h_x, \ell h_t) \right).$$

Thus,

$$\begin{aligned} (P_h \mathbf{U}_{\ell+1} - Q_h \mathbf{U}_\ell)_k &= u_t(kh_x, \ell h_t)h_t + u_{tt}(kh_x, \varsigma) \frac{h_t^2}{2} + ah_t u_x(kh_x, \ell h_t) \\ &\quad + \frac{ah_t}{12} \left( u_{xxx}(\xi^+, \ell h_t) + u_{xxx}(\xi^-, \ell h_t) \right) h_x^2 \\ &= \left( u_{tt}(kh_x, \varsigma) \frac{h_t}{2} + \frac{a}{12} (u_{xxx}(\xi^+, \ell h_t) + u_{xxx}(\xi^-, \ell h_t)) h_x^2 \right) h_t \\ &= \tau_{k,\ell} h_t. \end{aligned}$$

In general,  $P_n \mathbf{U}_{\ell+1} = Q_n \mathbf{U}_\ell + \mathbf{b}_{\ell+1} + \boldsymbol{\tau}_{\ell+1} h_t$

$$\boldsymbol{\tau}_\ell = \begin{bmatrix} \tau_{1\ell} \\ \tau_{0\ell} \\ \tau_{1\ell} \end{bmatrix}$$

$$\underline{b=0}$$

$$P_n U_{x+1} = Q_n U_x + \tau_x h_x$$

$$P_n u_{x+1} = Q_n u_x$$

$$\Rightarrow P_n e_{x+1} = Q e_x + \tau_x h_x$$

$$\Rightarrow e_{x+1} = P_n^{-1} Q_n u_x + P_n^{-1} \tau_n h_x$$

- the max-norm ( $l_\infty$ ):  $\|e\|_\infty = \max_{k,l} |e_{k,l}|$ ; or

- the scaled Euclidean norm ( $l_2$ ):  $\|e\|_2 = \left( \sum_k \sum_l h_x h_t (e_{k,l})^2 \right)^{\frac{1}{2}}$

Demo

$$n_x = 64$$



$$n_x = 128$$

## Definition 5.15: Consistency, Stability, and Convergence

Let  $\frac{\partial^\mu u}{\partial x^\mu}$  denote the  $\mu$ -th partial derivative in  $x$ , and  $\frac{\partial^\nu u}{\partial t^\nu}$  denote the  $\nu$ -th partial derivative in  $t$ . Assume that  $\frac{\partial^\mu u(x, \hat{t})}{\partial x^\mu}, \frac{\partial^\nu u(x, \hat{t})}{\partial t^\nu} \in L^2(\mathbb{R})$ , for all  $\hat{t} < t^*$ .

A two-level linear finite-difference scheme,  $P_h \mathbf{u}_{\ell+1} = Q_h \mathbf{u}_\ell + \mathbf{b}_\ell h_t$ , is

- consistent in the  $L^2$ -norm with order  $\nu$  in time and  $\mu$  in space if

$$\max_{\ell, \ell h_t \leq t^*} \|\tau_\ell\| = \mathcal{O}(h_x^\mu, h_t^\nu);$$

- convergent in the  $L^2$ -norm with order  $\nu$  in time and  $\mu$  in space if

$$\max_{\ell, \ell h_t \leq t^*} \|\mathbf{e}_\ell\| = \mathcal{O}(h_x^\mu, h_t^\nu);$$

- stable in the  $L^2$ -norm if  $\exists c > 0$ , independent of  $h_t$  and  $h_x$ , such that  $\|(P_h^{-1} Q_h)^\ell P_h^{-1}\| \leq c$  for all  $\ell$  and  $h_t$  such that  $\ell h_t \leq t^*$ .

- consistent if the scheme is "right" for the PDE.
- convergence is a statement on the error.

Linear algebra recall.

Solve  $Ax=b$

→ a stable algorithm gives  
a small residual:  $r = b - Ax$

But

$$\frac{\| \Delta x \|}{\| x \|} \leq \text{cond}(A) \frac{\| r \|}{\| A \| \| x \|}$$

How do we use this?

Two tools:

### **Theorem 5.19: Lax Convergence Theorem**

*If a two-level linear finite-difference scheme,  $P_h \mathbf{u}_{\ell+1} = Q_h \mathbf{u}_\ell + \mathbf{b}_\ell h_t$ , is consistent in the  $L^2$ -norm with order  $\nu$  in time and  $\mu$  in space **and** stable in the  $L^2$ -norm, then it is convergent in the  $L^2$ -norm with order  $\nu$  in time and  $\mu$  in space.*



Stronger :

**Theorem 5.20: Lax Equivalence Theorem**

*Let the two-level linear finite-difference scheme,  $P_h \mathbf{u}_{\ell+1} = Q_h \mathbf{u}_\ell + h_t \mathbf{b}_\ell$ , be consistent in the  $L^2$ -norm with order  $\nu$  in time and  $\mu$  in space, then it is stable in the  $L^2$ -norm if and only if it is convergent in the  $L^2$ -norm with order  $\nu$  in time and  $\mu$  in space.*

main point

consistency + stability  $\Rightarrow$  convergence

↑  
easy

↑  
hard

## Norms

Consider matrix  $T$ .

What is  $\|T\|$ ?

$$\rightarrow \|T\| = \max_i |\sigma_i(T)|$$

↑  
singular values

$$\text{also } \|T\| = \sup \frac{\|Tx\|}{\|x\|}$$

$$\|Tx\| \leq \|T\| \|x\|$$



## How to establish stability

$$\text{Need } \|(P_n^{-1} Q_n)^L P_n^{-1}\| \leq c \quad \forall L$$
$$\forall Lh < t^*$$

↓

$$\|(P_n^{-1} Q)^L\| \|P_n^{-1}\|$$

$$\text{So we need } \|P^{-1} Q\| < 1$$

$$\text{we } \|P^{-1}\| < c$$

"Lax-Richtmyer" stability

ETBS

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

$$\gamma = a \frac{h_t}{h_x}$$

$$\rightarrow u_{k, \ell+1} = \gamma u_{k-1, \ell} + (1-\gamma) u_{k, \ell}$$

$$\rightarrow \vec{P}_h = I$$

$$Q_h = \text{tridiag}(\gamma, 1-\gamma, 0)$$

(consider  $u(0, t) = f(t)$  as a boundary condition)

First  $P_n$ :  $P = I \Rightarrow \|P^{-1}\| = 1$

Next  $P_n^{-1} Q_n = Q_n$ :

consider

$$\|Q_n\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^n |(Q_n)_{ij}|$$

or

$$\|Q_n\|_\infty = \max_i \sum_{j=1}^n |(Q_n)_{ij}|$$

$$\|Q\|_1 = \|Q\|_\infty = |\gamma| + |1 - \gamma|$$

if  $0 \leq \gamma \leq 1$   
then  $\|Q\|_1 = 1$

if  $\gamma \notin [0, 1]$   
then  $\|Q\|_1 > 1$

(similar for  $l_2$  using Gerschgorin's theorem)

$\Rightarrow$  conditional stability

if  $0 \leq \sigma \leq 1$ , then stable.

$a > 0$

$$\Rightarrow \frac{a h_x}{h_x} \leq 1 \Rightarrow h_t \leq \frac{h_x}{a}$$

CFL condition



Ok this is complicated.

### Definition 5.28: Symbol of a Two-Level Finite-Difference Scheme

Given a two-level linear finite-difference scheme,  $P_h \mathbf{u}_{\ell+1} = Q_h \mathbf{u}_\ell + \mathbf{b}_\ell h_t$ , such that  $p_k$  and  $q_k$  are the coefficients of the scheme corresponding to  $P_h$  and  $Q_h$ , respectively, then,

$$\hat{p}(\varphi) = \sum_k p_k e^{-i\varphi k} \text{ and } \hat{q}(\varphi) = \sum_k q_k e^{-i\varphi k}.$$

The symbol of the two-level finite-difference method is  $s(\varphi) = \frac{\hat{q}(\varphi)}{\hat{p}(\varphi)}$ .

### Definition 5.29: von Neumann Stability

Consider a two-level linear finite-difference scheme,  $P_h \mathbf{u}_{\ell+1} = Q_h \mathbf{u}_\ell + \mathbf{b}_\ell h_t$ , with symbol,  $s(\varphi)$ . If

$$\max_{\varphi} |s(\varphi)| \leq 1 \quad \text{and} \quad \max_{\varphi} \left| \frac{1}{\hat{p}(\varphi)} \right| \leq c,$$

for some constant  $c > 0$ , then we say the scheme is von Neumann stable.

What is  $p_k, q_k$ ?

Why?

next we.