

[4] Discretization

→ In PDEs, we need to discretize both

- (i) time t , and
- (ii) space, x, y, z

→ Notations are given as:

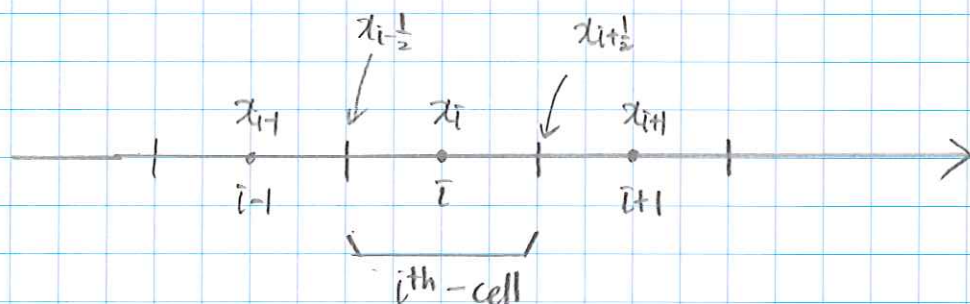
(i) $t^n = n\Delta t$, $n=0, \dots, M$,

(ii) $x_i = (i-\frac{1}{2})\Delta x$, $i=1, 2, \dots, N$.

→ In addition, we also adopt the "half-integer":

(iii) $x_{i+\frac{1}{2}} = x_i + \frac{\Delta x}{2}$

x_i : cell-center
 $x_{i+\frac{1}{2}}$: cell-interface



Def. Let $u_i^n = u(x_i, t^n)$ be the pointwise values of the exact soln u of a given PDE at discrete pts (x_i, t^n) .

→ No error is associated with u_i^n .

Def. Let U_i^n be the numerical approx to the exact soln u_i^n at discrete points (x_i, t^n) :

$$\left\{ \begin{array}{l} U_i^n \approx u_i^n \quad \text{for finite difference methods,} \\ U_i^n \approx \frac{1}{\Delta x} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} u(x) dx \quad \text{for finite volume methods.} \end{array} \right.$$

Def. D_i^n : the exact soln of the associated difference eqn of the PDE, e.g., for FTBS (forward in time backward in space) for the advection PDE:

$$\frac{\partial u}{\partial t} + a \frac{\partial u}{\partial x} = 0 \quad \dots \textcircled{1}$$

the difference eqn, using FTBS is

$$\frac{U_i^{n+1} - U_i^n}{\Delta t} + a \frac{U_i^n - U_{i-1}^n}{\Delta x} = 0, \quad \dots \textcircled{2}$$

then D_i^n satisfies $\textcircled{2}$ exactly:

$$\frac{D_i^{n+1} - D_i^n}{\Delta t} + a \frac{D_i^n - D_{i-1}^n}{\Delta x} = 0,$$

$\rightarrow D_i^n$ is the exact soln of the difference eqn with no round-off errors.

Def Discretization error at (x_i, t^n) :

$$\varepsilon_{d,i}^n = u_i^n - D_i^n \quad (= \text{truncation errors} + \text{BC errors})$$

Def Round-off error at (x_i, t^n) :

$$\varepsilon_{r,i}^n = D_i^n - U_i^n \quad \left(= \begin{array}{l} \text{repeated computer arithmetic} \\ \text{operations with rounding off} \\ \text{numbers to some significant} \\ \text{digits} \end{array} \right)$$

Def Global error at (x_i, t^n) :

$$\varepsilon_{g,i}^n = u_i^n - U_i^n$$

Rule $\varepsilon_{g,i}^n = \varepsilon_{d,i}^n + \varepsilon_{r,i}^n$.

Def The numerical method is convergent at t^n in $\|\cdot\|$ if

$$\lim_{\Delta x, \Delta t \rightarrow 0} \|\varepsilon_g^n\| = 0$$

$$\begin{aligned} \text{(ex)} \quad \|\varepsilon_g^n\|_1 &= \Delta x \sum_{i=1}^N |\varepsilon_{g,i}^n| \\ &= \Delta x \sum_{i=1}^N |u_i^n - U_i^n| \end{aligned}$$

$$\text{(ex)} \quad \|\varepsilon_g^n\|_p = \left(\Delta x \sum_{i=1}^N |\varepsilon_{g,i}^n|^p \right)^{\frac{1}{p}}$$

5] The Fundamental Theorem of Numerical Methods

The Lax Equivalence Theorem for linear PDE.

Thm. (only for linear PDEs)

Consistency + ^{absolute} stability \Leftrightarrow convergence.

┌──────────────────┐

└─ easier to check
without knowing
the exact, analytical
soln.

┌──────────────────┐

└─ requires to
know the
exact soln.

Def. Let N be the linear numerical operator mapping the approximate soln at t^n to the approximate soln at t^{n+1} .

(e.g) $U^{n+1} = N(U^n) \rightarrow$ one step explicit.

Def. The one-step error $\mathbb{E}_{1\text{step},i}^{n+1}$ at (x_i, t^{n+1}) ,

$$\mathbb{E}_{1\text{step},i}^{n+1} = u_i^{n+1} - N(u_i^n), \quad i=1, \dots, N.$$

Def. The local truncation error $\mathbb{E}_{L,i}^{n+1}$ at (x_i, t^{n+1})

$$\mathbb{E}_{L,i}^{n+1} = \frac{1}{\Delta t} \mathbb{E}_{1\text{step},i}^{n+1}, \quad i=1, \dots, N.$$

Def. The numerical method is of order p in both t & x for all sufficiently smooth data with compact support

$$\Leftrightarrow \sum_{L, i}^{n+1} = \begin{cases} \mathcal{O}(\Delta t^p + \Delta x^p) & \text{in 1D} \\ \mathcal{O}(\Delta t^p + \Delta x^p + \Delta y^p) & \text{in 2D} \\ \mathcal{O}(\Delta t^p + \Delta x^p + \Delta y^p + \Delta z^p) & \text{in 3D.} \end{cases}$$

Remark In practice one may have a method that is

$\left\{ \begin{array}{l} p^{\text{th}} \text{ order accurate in time, \& \\ r^{\text{th}} \text{ order accurate in space} \end{array} \right.$

$$\Rightarrow \sum_{L, i}^{n+1} = \mathcal{O}(\Delta t^p + \Delta x^r)$$

\rightarrow In this case, the numerical soln in a fully resolved state, both temporally and spatially, will exhibit the convergence rate dominated by the larger error between the two:

$$\Rightarrow \sum_{L, i}^{n+1} = \max[\mathcal{O}(\Delta t^p), \mathcal{O}(\Delta x^r)] \dots \textcircled{1}$$

4/29/16

Def. The numerical method is consistent in $\|\cdot\|$ with the target PDE if

$$\lim_{\Delta t, \Delta x \rightarrow 0} \|\Sigma_{\Delta t}^n\| = 0$$

for all smooth fns $u(x,t)$ that satisfies the given PDE,

Stability Theory for "Linear" PDEs

→ The stability Theory for nonlinear PDEs need yet different approaches and we are not going to study them in AMS 213B, (But in AMS 260)

→ For linear PDEs, we need to bound the global error $\Sigma_{g,i}^{n+1} = u_i^{n+1} - U_i^{n+1}$ using a recurrence relation.

$$\begin{aligned} \rightarrow \Sigma_{g,i}^{n+1} &= u_i^{n+1} - U_i^{n+1} \\ &= u_i^{n+1} - \mathcal{N}(U_i^n) \\ &= u_i^{n+1} - \mathcal{N}(u_i^n - \Sigma_{g,i}^n) \\ &= \underbrace{u_i^{n+1} - \mathcal{N}(u_i^n)}_{\Sigma_{1st step, i}^{n+1}} + \mathcal{N}(\Sigma_{g,i}^n) \\ &= \Delta t E_{\Delta t, i}^{n+1} + \mathcal{N}(\Sigma_{g,i}^n) \quad \text{--- (1)} \end{aligned}$$

Recall from the ODE theory, that we have:

Def. The linear numerical method defined by the linear operator N is stable in $\|\cdot\|$ if

$$\exists C \text{ s.t. } \|N^n\| \leq C, \quad \forall n \text{ s.t. } \leq T,$$

for each T .

Prop. N^n is the n th power of the linear operator N obtained by the repeated applications of the linear operations of N .

→ this is only true for linear operators, but not for nonlinear operators.

Prop. We also can say the numerical method is stable if $\|N\| \leq 1$, since

$$\|N^n\| \leq \|N\|^n \leq 1.$$

Thm. The Lax Equivalence Thm for linear difference methods

For a well-posed consistent linear method, (absolute) stability is necessary and sufficient condition for convergence:

$$\boxed{\text{Consistency} + \text{absolute stability} \Leftrightarrow \text{convergence}}$$

proof (\Leftarrow) Richtmyer and Morton,

Difference Methods for IVPs, Wiley-Interscience, 1967.

(\Rightarrow) Claim: $\lim_{\Delta t, \Delta x \rightarrow 0} \|E_g^{n+1}\| = 0.$

$$\left[\begin{array}{l} \text{Note here that, if we use } L\text{-norm,} \\ \|E_g^{n+1}\| = \|E_g^{n+1}\|_1 \\ = \Delta x \sum_{i=1}^N |E_{g,i}^{n+1}| \end{array} \right]$$

From Eq. ①:

$$\begin{aligned} \|E_g^{n+1}\| &\leq \Delta t \|E_{\tau}^{n+1}\| + \|N(E_g^n)\| && \rightarrow \|AB\| \leq \|A\| \|B\| \\ &\leq \Delta t \|E_{\tau}^{n+1}\| + \|N\| \|E_g^n\| && \rightarrow \|N\| \leq \tilde{C} \\ &\leq \Delta t \|E_{\tau}^{n+1}\| + C \|E_g^n\| && (\text{① stable}) \\ &\leq \Delta t \|E_{\tau}^{n+1}\| + C (\Delta t \|E_{\tau}^n\| + C \|N\| \|E_g^{n-1}\|) \end{aligned}$$

$$\begin{aligned} \|E_{\tau}\| &\equiv \max_j \|E_{\tau}^j\|, \\ \text{for some constants } \tilde{C}, \tilde{D}, & \left(\begin{aligned} &\leq \Delta t \sum_{j=1}^{n+1} C^{n+1-j} \|E_{\tau}^j\| + C^{n+1} \|E_g^0\| \\ &\leq \tilde{D} (n+1) \Delta t \|E_{\tau}\| + \tilde{C} \|E_g^0\| \\ &= \tilde{D} \tau^{n+1} \|E_{\tau}\| + \tilde{C} \|E_g^0\| \end{aligned} \right) \quad \dots \text{②} \end{aligned}$$

In Eqn (2),

(i) $\|E_g^0\| \rightarrow 0$ as $\Delta t, \Delta x \rightarrow 0$; otherwise it means that the problem is ill-posed due to the nonzero global error in resolving the initial data.

(ii) $\|E_T\| \rightarrow 0$ as $\Delta t \rightarrow 0$, since the method is consistent by assumption.

Therefore, we prove

$$\lim_{\Delta t, \Delta x \rightarrow 0} \|E_g^{nth}\| = 0 \quad \square$$

Remark. We note that the above proof is only available when N is linear, hence the theorem is only valid for linear PDEs.

→ For nonlinear PDEs, we don't have any general form of stability theory, and we often just show nonlinear stability by showing that a suitable form of stability holds, using different techniques depending on the character of the PDEs.