

Lectures Références

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Numerical Methods for PDE: Finite Differences and Finite Volumes

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- 1 Finite Difference(FD) and Finite volume(FV) : Overview
- 2 Modelization and Simplified models of PDE.
- 3 Scalar Advection-Diffusion Eqation.
- 4 Approximation of a Scalar 1D ODE.
- 5 FD for 1D scalar poisson equation (elliptic).
- 6 FD for 1D scalar difusion equation (parabolic).
- 7 FD for 1D scalar advection-diffusion equation.
- 8 Scalar Nonlinear Conservation law : 1D (hyperbolic).
- 9 FV for scalar nonlinear Conservation law : 1D
- 10 Multi-Dimensional extensions

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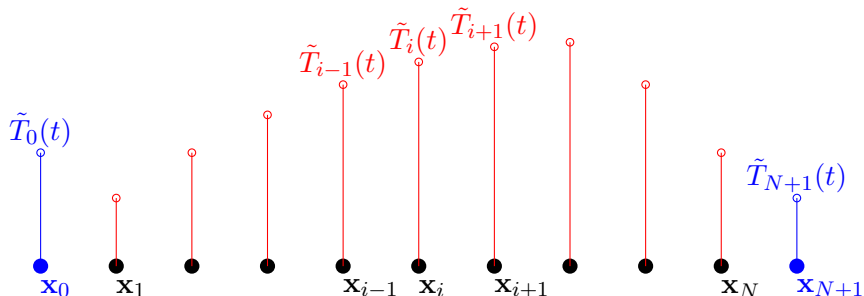
1D Scalar advection-diffusion Equation (periodic)

$$\left\{ \begin{array}{ll} \frac{\partial T}{\partial t} + c \frac{\partial T}{\partial x} = \lambda \frac{\partial^2 T}{\partial \mathbf{x}^2}, & \forall \mathbf{x} \in (0, 1), \quad t > 0, \\ T(t, \mathbf{x}) = T_0(\mathbf{x}) & \forall \mathbf{x} \in (0, 1), \quad t = 0, \\ T(t, \mathbf{x} = 0) \equiv T(t, \mathbf{x} = 1) & \forall t \geq 0 \end{array} \right.$$

Properties

- well posedness : *bb*.
- Existence : The solution $T(\mathbf{x})$ always exists and is unique.
- Regularity : The solution $T(\mathbf{x})$ is regular and uniformly bounded
- Positivity :
- Maximum principle : .

1D mesh for the discretization (approximated solution)



$$\mathbf{x}_1 = 0, \quad \mathbf{x}_{N+1} = 1, \quad \mathbf{x}_i = (i-1)\delta x, \quad \delta x = \frac{1}{N}$$

$$\tilde{T}_i = \tilde{T}(\mathbf{x}_i) \text{ for } 1 \leq i \leq N, \quad \tilde{T}_0 \equiv \tilde{T}_N, \quad \tilde{T}_{N+1} \equiv \tilde{T}_1$$

Finite difference strategy : advection operator

Taylor's expansions :

$$T(t, \mathbf{x} + \beta \delta \mathbf{x}) = T(t, \mathbf{x}) + \beta \delta \mathbf{x} \frac{\partial T}{\partial \mathbf{x}}(t, \mathbf{x}) + \beta^2 \frac{\delta \mathbf{x}^2}{2} \frac{\partial^2 T}{\partial \mathbf{x}^2}(t, \mathbf{x}) + \dots + R_m(t, \mathbf{x})$$

Discrete approximations of $\frac{\partial T}{\partial x}$:

- **Central Finite Difference scheme** : $\vartheta = \{-1, 1\}$.

$$\frac{\partial T}{\partial x} \simeq \frac{T(t, \mathbf{x} + \delta \mathbf{x}) - T(t, \mathbf{x} - \delta \mathbf{x})}{2\delta \mathbf{x}}$$

- **Upwind Finite Difference scheme** : $\vartheta \subset \{-1, 0, 1\}$.

$$\frac{\partial T}{\partial x} \simeq \frac{T(t, \mathbf{x} + \delta \mathbf{x}) - T(t, \mathbf{x} - \delta \mathbf{x})}{2\delta \mathbf{x}} - \frac{|c|}{c} \frac{T(t, \mathbf{x} + \delta \mathbf{x}) - 2T(t, \mathbf{x}) + T(t, \mathbf{x} - \delta \mathbf{x})}{2\delta \mathbf{x}}$$

Finite difference strategy : some Difference operators

First order derivatives :

$$\delta_{\xi}^{\beta} \tilde{T}(\xi) = \frac{\tilde{T}(\xi + \beta \delta \xi) - \tilde{T}(\xi)}{\beta \delta \xi}, \quad \bar{\delta}_{\xi}^{\beta} \tilde{T}(\xi) = \frac{\delta_{\xi}^{\beta} + \delta_{\xi}^{-\beta}}{2} \tilde{T}(\xi)$$

- Forward finite difference :

$$\delta_{\mathbf{x}}^{+} \tilde{T}_j = \frac{\tilde{T}_{j+1} - \tilde{T}_j}{\delta \mathbf{x}} = \left. \frac{\partial T}{\partial \mathbf{x}} \right|_j + R_1^{+}(\delta \mathbf{x})$$

- Backward finite difference :

$$\delta_{\mathbf{x}}^{-} \tilde{T}_j = \frac{\tilde{T}_j - \tilde{T}_{j-1}}{\delta \mathbf{x}} = \left. \frac{\partial T}{\partial \mathbf{x}} \right|_j + R_1^{-}(\delta \mathbf{x})$$

- Centered finite difference :

$$\delta_{\mathbf{x}} \tilde{T}_j = \frac{\tilde{T}_{j+1} - \tilde{T}_{j-1}}{2\delta \mathbf{x}} = \left(\frac{\delta_{\mathbf{x}}^{+} + \delta_{\mathbf{x}}^{-}}{2} \right) \tilde{T}_j = \left. \frac{\partial T}{\partial \mathbf{x}} \right|_j + R_1^0(\delta \mathbf{x}^2)$$

Second order derivatives :

- $\delta_{\mathbf{xx}} \tilde{T}_j = \frac{\tilde{T}_{j+1} - 2\tilde{T}_j + \tilde{T}_{j-1}}{\delta \mathbf{x}^2} = \left. \frac{\partial^2 T}{\partial \mathbf{x}^2} \right|_j + R_2^0(\delta \mathbf{x}^2)$

One step Finite difference scheme for advection/diffusion

$$\begin{aligned} \frac{\tilde{T}_j^{n+1} - \tilde{T}_j^n}{\delta t} + c \left(\theta_c \delta_{\mathbf{x}} \tilde{T}_j^{n+1} + (1 - \theta_c) \delta_{\mathbf{x}} \tilde{T}_j^n \right) \\ - \zeta \delta_{\mathbf{x}} \left(\theta_c \delta_{\mathbf{xx}} \tilde{T}_j^{n+1} + (1 - \theta_c) \delta_{\mathbf{xx}} \tilde{T}_j^n \right) \\ = \lambda \left(\theta_d \delta_{\mathbf{xx}} \tilde{T}_j^{n+1} + (1 - \theta_d) \delta_{\mathbf{xx}} \tilde{T}_j^n \right) \end{aligned}$$

ζ is the numerical diffusion coefficient.

- $\zeta = 0$ for the centered scheme
- $\zeta = \frac{|c|}{2}$ for the upwind scheme.

The parameters $\theta \equiv \theta_c$ or θ_d can takes the values :

- $\theta = 0$ for explicit approximation of space derivatives.
- $\theta = 1$ for implicit approximation of space derivatives.
- $\theta = \frac{1}{2}$ for Crank-Nicholson approximation of space derivatives.

Truncation Error

$$\begin{aligned}
 \mathcal{E}_j^{n+1} &= \frac{T_j^{n+1} - T_j^n}{\delta t} + c \left(\theta_c \delta_{\mathbf{x}} T_j^{n+1} + (1 - \theta_c) \delta_{\mathbf{x}} T_j^n \right) \\
 &\quad - \zeta \delta_{\mathbf{x}} \left(\theta_c \delta_{\mathbf{xx}} T_j^{n+1} + (1 - \theta_c) \delta_{\mathbf{xx}} T_j^n \right) \\
 &\quad - \lambda \left(\theta_d \delta_{\mathbf{xx}} T_j^{n+1} + (1 - \theta_d) \delta_{\mathbf{xx}} T_j^n \right) \\
 &= \left(\partial_t T \right)_j^n + \frac{\delta t}{2} \left(\partial_t^2 T \right)_j^n + O(\delta t^2 \partial_t^3) \\
 &\quad + c \theta_c \left(\partial_{\mathbf{x}} T \right)_j^{n+1} + c(1 - \theta_c) \left(\partial_{\mathbf{x}} T \right)_j^n + O(\delta \mathbf{x}^2 \partial_{\mathbf{x}}^3) \\
 &\quad - \zeta \delta_{\mathbf{x}} \left[\theta_c \left(\partial_{\mathbf{x}}^2 T \right)_j^{n+1} + (1 - \theta_c) \left(\partial_{\mathbf{x}}^2 T \right)_j^n \right] + \zeta \delta_{\mathbf{x}} O(\delta \mathbf{x}^2 \partial_{\mathbf{x}}^4) \\
 &\quad - \lambda \left[\theta_d \left(\partial_{\mathbf{x}}^2 T \right)_j^{n+1} + (1 - \theta_d) \left(\partial_{\mathbf{x}}^2 T \right)_j^n \right] + O(\delta \mathbf{x}^2 \partial_{\mathbf{x}}^4)
 \end{aligned}$$

Truncation Error

$$\begin{aligned}
 \mathcal{E}_j^{n+1} &= (\partial_t T)_j^n + \frac{\delta t}{2} (\partial_t^2 T)_j^n + O(\delta t^2 \partial_t^3) \\
 &\quad + c (\partial_x T)_j^n + c \theta_c \delta t (\partial_t \partial_x T)_j^n + O(\delta \mathbf{x}^2 \partial_x^3, \delta t^2 \partial_t^2 \partial_x) \\
 &\quad + \zeta \delta \mathbf{x} \left[-\partial_x^2 T + \theta_c \delta t \partial_t \partial_x^2 T \right]_j^n + \zeta \delta \mathbf{x} O(\delta \mathbf{x}^2 \partial_x^4 + \delta t^2 \partial_t^2 \partial_x^2) \\
 &\quad - \lambda (\partial_x^2 T)_j^n + \theta_d \lambda \delta t (\partial_t \partial_x^2 T)_j^n + O(\delta \mathbf{x}^2 \partial_x^4 + \delta t^2 \partial_t^2 \partial_x^2) \\
 &= (\partial_t T + c \partial_x T - \lambda \partial_x^2 T)_j^n \\
 &\quad + \left[\delta t \partial_t \left(\frac{1}{2} \partial_t T + \theta_c c \partial_x T - \theta_d \lambda \partial_x^2 T \right) - \zeta \delta \mathbf{x} (\partial_x^2 T) \right]_j^n \\
 &\quad + \left[\partial_t^3 + \partial_t^2 \partial_x + \partial_t^2 \partial_x^2 + \zeta \delta \mathbf{x} \partial_t^2 \partial_x^2 \right] O(\delta t^2) \\
 &\quad + \left[\partial_x^3 + \partial_x^4 + \zeta \delta \mathbf{x} \partial_x^4 \right] O(\delta \mathbf{x}^2)
 \end{aligned}$$

$$\mathcal{E}_j^{n+1} = \delta t \left[\partial_t \left(\frac{1}{2} \partial_t T + \theta_c c \partial_x T - \theta_d \lambda \partial_x^2 T \right) \right]_j^n - \zeta \delta \mathbf{x} (\partial_x^2 T)_j^n + O(\delta h^2 \partial^3 +$$

blue terms are principal component of the truncation error.

Truncation Error : Conclusions for regular solutions.

$$\mathcal{E}_j^{n+1} = \delta t \left[\partial_t \left(\frac{1}{2} \partial_t T + \theta_c c \partial_x T - \theta_d \lambda \partial_x^2 T \right) \right]_j^n - \zeta \delta \mathbf{x} (\partial_x^2 T)_j^n + O(\delta h^2 \partial^3 +$$

- These schemes are always consistent : $\lim_{\substack{\delta t \rightarrow 0 \\ \delta \mathbf{x} \rightarrow 0}} \|\mathcal{E}^{n+1}\| = 0$.
- Upwind schemes (for advection : $\zeta \neq 0$) are always first order accurate in space.
- Crank-Nicholson schemes for both advection and diffusion ($\theta_c = \theta_d = \frac{1}{2}$) are second order accurate in time.
- Centered(for advection)/Crank-Nicholson scheme is second order accurate both in time and space.

Modified equation

We assume now that we can define a smooth approximated solution $V(x, t)$ such as $V(x_j, t^n) = \tilde{T}_j^n$ for all j and n .

Therefore,

$$\begin{aligned} & \frac{V(x_j, t^{n+1}) - V(x_j, t^n)}{\delta t} \\ & + c\theta_c \frac{V(x_{j+1}, t^{n+1}) - V(x_{j-1}, t^{n+1})}{2\delta x} + c(1 - \theta_c) \frac{V(x_{j+1}, t^n) - V(x_{j-1}, t^n)}{2\delta x} \\ & - \left[\zeta \delta \mathbf{x} \theta_c + \lambda \theta_d \right] \frac{V(x_{j+1}, t^{n+1}) - 2V(x_j, t^{n+1}) + V(x_{j-1}, t^{n+1})}{\delta x^2} \\ & - \left[\zeta \delta \mathbf{x} (1 - \theta_c) + \lambda (1 - \theta_d) \right] \frac{V(x_{j+1}, t^n) - 2V(x_j, t^n) + V(x_{j-1}, t^n)}{\delta x^2} = 0 \end{aligned}$$

We can apply the Taylor expansions for $V(x, t)$ to obtain the modified equation satisfied by the approximate solution.

Modified equation

$$\begin{aligned}
 & (\partial_t V)_j^n + \frac{\delta t}{2} (\partial_t^2 V)_j^n + \frac{\delta t^2}{6} (\partial_t^3 V)_j^n + O(\delta t^3 \partial_t^4) \\
 & + c \theta_c \left[(\partial_{\mathbf{x}} V)_j^{n+1} + \frac{\delta x^2}{6} (\partial_x^3 V)_j^{n+1} \right] + O(\delta \mathbf{x}^4 \partial_{\mathbf{x}}^5) \\
 & + c(1 - \theta_c) \left[(\partial_{\mathbf{x}} V)_j^n + \frac{\delta x^2}{6} (\partial_x^3 V)_j^n \right] + O(\delta \mathbf{x}^4 \partial_{\mathbf{x}}^5) \\
 & - \left[\zeta \delta \mathbf{x} \theta_c + \lambda \theta_d \right] \left[(\partial_{\mathbf{x}}^2 V)_j^{n+1} + O(\delta \mathbf{x}^2 \partial_{\mathbf{x}}^4) \right] \\
 & - \left[\zeta \delta \mathbf{x} (1 - \theta_c) + \lambda (1 - \theta_d) \right] \left[(\partial_{\mathbf{x}}^2 V)_j^n + O(\delta \mathbf{x}^2 \partial_{\mathbf{x}}^4) \right] = 0
 \end{aligned}$$

Modified equation

$$\begin{aligned}
 & \left[\partial_t V + \frac{\delta t}{2} \partial_t^2 V + \frac{\delta t^2}{6} \partial_t^3 V + O(\delta t^3 \partial_t^4) \right]_j^n \\
 & + c \theta_c \left[\delta t \partial_{\mathbf{x}} \partial_t V + \frac{\delta t^2}{2} \partial_{\mathbf{x}} \partial_t^2 V + O(\delta h^3 \partial^4) \right]_j^n \\
 & + c \left[\partial_{\mathbf{x}} V + \frac{\delta x^2}{6} \partial_{\mathbf{x}}^3 V + O(\delta h^4 \partial^5) \right]_j^n \\
 & - \left[\zeta \delta \mathbf{x} \theta_c + \lambda \theta_d \right] \left[\delta t \partial_{\mathbf{x}}^2 \partial_t V + O(\delta h^2 \partial^4) \right]_j^n \\
 & - \left[\zeta \delta \mathbf{x} + \lambda \right] \left[\partial_{\mathbf{x}}^2 V + O(\delta \mathbf{x}^2 \partial_{\mathbf{x}}^4) \right]_j^n = 0
 \end{aligned}$$

$$\begin{aligned}
 & \partial_t V + c \partial_{\mathbf{x}} V - \lambda \partial_{\mathbf{x}}^2 V = \\
 & - \frac{\delta t}{2} \partial_t^2 V - c \theta_c \delta t \partial_{\mathbf{x}} \partial_t V + \zeta \delta \mathbf{x} \partial_{\mathbf{x}}^2 V + \lambda \theta_d \delta t \partial_{\mathbf{x}}^2 \partial_t V + O(\delta h^2) \\
 & \partial_t^2 V = c^2 \partial_{\mathbf{x}}^2 V - 2c\lambda \partial_{\mathbf{x}}^3 V + \lambda^2 \partial_{\mathbf{x}}^4 V + O(\delta h) \\
 & \partial_{\mathbf{x}} \partial_t V = -c \partial_{\mathbf{x}}^2 V + \lambda \partial_{\mathbf{x}}^3 V + O(\delta h)
 \end{aligned}$$

modified equation

$$\begin{aligned} \partial_t V + c \partial_x V - \lambda \partial_x^2 V &= \\ -\frac{\delta t}{2} \partial_t^2 V - c \theta_c \delta t \partial_x \partial_t V + \zeta \delta x \partial_x^2 V + \lambda \theta_d \delta t \partial_x^2 \partial_t V + O(\delta h^2) \\ \partial_t^2 V &= c^2 \partial_x^2 V - 2c\lambda \partial_x^3 V + \lambda^2 \partial_x^4 V + O(\delta h) \\ \partial_x \partial_t V &= -c \partial_x^2 V + \lambda \partial_x^3 V + O(\delta h) \end{aligned}$$

The modified equation is then written as

$$\partial_t V + c \partial_x V - \lambda \partial_x^2 V = \mathcal{T}(\partial_x) V$$

$$\mathcal{T}(\partial_x) \equiv \beta_2 \partial_x^2 + \beta_3 \partial_x^3 + \beta_4 \partial_x^4 - \dots \quad \text{where}$$

- $\beta_{2m} \partial_x^{2m} V$ are associated to **amplitude error** \equiv dissipation.
- $\beta_{2m+1} \partial_x^{2m+1} V$ are associated to **phase error** \equiv dispersion..

Modified Equation

$$\partial_t V + c \partial_x V - \lambda \partial_x^2 V = \sum_{k \geq 2} \beta_k \partial_x^k V$$

Solving this equation, by using Fourier transform in space :

$$\widehat{V}(\theta, t) = \exp(\mu(\theta)t) \widehat{V}(\theta, 0) = e^{\mu_a t} e^{i\mu_\phi t} \widehat{V}(\theta, 0)$$

$$\text{where } \mu(\theta) = -ic\theta - \lambda\theta^2 + \sum_{k \geq 2} \beta_k (i\theta)^k$$

$$\mu_a = \text{Real} [\mu(\theta)] = -\lambda\theta^2 + \sum_{m \geq 1} \beta_{2m} (-1)^m \theta^{2m}$$

$$\mu_\phi = \text{Imag} [\mu(\theta)] = -c\theta + \sum_{m \geq 1} \beta_{2m+1} (-1)^m \theta^{2m+1}$$

- $-\lambda\theta^2 - \mu_a$ is the **amplitude error** \equiv dissipation.
- $-c\theta - \mu_\phi$ is the **phase error** \equiv dispersion..

Equivalent equation (regular solution) : Explicit scheme

$$\mathcal{T}(\partial_x) \equiv -\frac{c^2 \delta t}{2} \partial_x^2 + \zeta \delta x \partial_x^2 + O(\delta h^2 \partial^3 \dots)$$

Appropriate choice of ζ leads to the elimination of the second order dissipation vanish.

The Lax-Wendroff scheme :: $\zeta = \frac{c^2 \delta t}{2 \delta x}$

This scheme is second order in space and time.

VN Stability (L_2) : Discrete Fourier mode & transformations.

$$\hat{\mathbf{T}}(t, \theta) = \beta(t, \theta) \hat{\mathbf{T}}(\theta) \quad \text{with} \quad \hat{\mathbf{T}}_j(\theta) = \exp(ij\theta) \quad \text{for } 1 \leq j \leq N$$

$$\frac{\hat{T}_j^{n+1} - \hat{T}_j^n}{\delta t} = \frac{\beta^{n+1} - \beta^n}{\delta t} \hat{T}_j(\theta),$$

$$\delta_{\mathbf{x}} \hat{T}_j(t) = \frac{\hat{T}_{j+1}(t) - \hat{T}_{j-1}(t)}{2\delta \mathbf{x}} = \beta(t) \frac{e^{i\theta} - e^{-i\theta}}{2\delta \mathbf{x}} \hat{T}_j = i\beta(t) \frac{\sin \theta}{\delta \mathbf{x}} \hat{T}_j(\theta)$$

$$\delta_{\mathbf{xx}} \hat{T}_j(t) = -\beta(t) \frac{4 \sin^2 \frac{\theta}{2}}{\delta \mathbf{x}^2} \hat{T}_j(\theta)$$

Von Neumann Stability Analysis : Discrete Fourier mode

$$\begin{aligned}
 & \frac{\tilde{T}_j^{n+1} - \tilde{T}_j^n}{\delta t} + c \left(\theta_c \delta_{\mathbf{x}} \tilde{T}_j^{n+1} + (1 - \theta_c) \delta_{\mathbf{x}} \tilde{T}_j^n \right) \\
 & \quad - \zeta \delta_{\mathbf{x}} \left(\theta_c \delta_{\mathbf{xx}} \tilde{T}_j^{n+1} + (1 - \theta_c) \delta_{\mathbf{xx}} \tilde{T}_j^n \right) \\
 & \quad = \lambda \left(\theta_d \delta_{\mathbf{xx}} \tilde{T}_j^{n+1} + (1 - \theta_d) \delta_{\mathbf{xx}} \tilde{T}_j^n \right) \\
 \implies & \frac{\beta^{n+1} - \beta^n}{\delta t} + \left[\theta_c \beta^{n+1} + (1 - \theta_c) \beta^n \right] \frac{ic \sin \theta}{\delta_{\mathbf{x}}} \\
 & \quad + \left[\theta_c \beta^{n+1} + (1 - \theta_c) \beta^n \right] \frac{4\zeta \sin^2 \frac{\theta}{2}}{\delta_{\mathbf{x}}} \\
 & \quad = - \left[\theta_d \beta^{n+1} + (1 - \theta_d) \beta^n \right] \frac{4\lambda \sin^2 \frac{\theta}{2}}{\delta_{\mathbf{x}}^2} \\
 g(\theta) = & \frac{1 - (1 - \theta_c) \left(\frac{ic\delta t \sin \theta}{\delta_{\mathbf{x}}} + \frac{4\zeta\delta t \sin^2 \frac{\theta}{2}}{\delta_{\mathbf{x}}} \right) - (1 - \theta_d) \frac{4\lambda\delta t \sin^2 \frac{\theta}{2}}{\delta_{\mathbf{x}}^2}}{1 + \theta_c \left(\frac{ic\delta t \sin \theta}{\delta_{\mathbf{x}}} + \frac{4\zeta\delta t \sin^2 \frac{\theta}{2}}{\delta_{\mathbf{x}}} \right) + \theta_d \frac{4\delta t \lambda \sin^2 \frac{\theta}{2}}{\delta_{\mathbf{x}}^2}}
 \end{aligned}$$

Von Neumann Stability Analysis

$$g(\theta) \equiv g(X) = \frac{1 - (1 - \theta_c) \frac{4\sigma\zeta}{c} X - 4\nu(1 - \theta_d)X - i\sigma(1 - \theta_c) \sin \theta}{1 + \theta_c \frac{4\sigma\zeta}{c} X + 4\nu\theta_d X + i\sigma\theta_c \sin \theta}$$

$$\text{where } \sigma = \frac{c\delta t}{\delta \mathbf{x}}, \quad 0 \leq X = \sin^2 \frac{\theta}{2} \leq 1, \quad \nu = \frac{\delta t \lambda}{\delta \mathbf{x}^2}$$

Using the fact that $\sin^2 \theta = 4(X - X^2)$ and setting

$\alpha_i = 2\sigma(1 - \theta_c)$, $\bar{\alpha}_i = 2\sigma\theta_c$, $\alpha_r = (1 - \theta_c) \frac{4\sigma\zeta}{c} + 4\nu(1 - \theta_d)$ and $\bar{\alpha}_r = \theta_c \frac{4\sigma\zeta}{c} + 4\nu\theta_d$, we obtain

$$|g(X)|^2 = \frac{(1 - \alpha_r X)^2 + \alpha_i^2 (X - X^2)}{(1 + \bar{\alpha}_r X)^2 + \bar{\alpha}_i^2 (X - X^2)} = \frac{1 + X \left[(\alpha_i^2 - 2\alpha_r) + (\alpha_r^2 - \alpha_i^2) X \right]}{1 + X \left[(\bar{\alpha}_i^2 + 2\bar{\alpha}_r) + (\bar{\alpha}_r^2 - \bar{\alpha}_i^2) X \right]}$$

Von Neumann Stability Analysis

$$|g(\theta)|^2 \equiv |g(X)|^2 = \frac{1 + X \left[(\alpha_i^2 - 2\alpha_r) + (\alpha_r^2 - \alpha_i^2)X \right]}{1 + X \left[(\bar{\alpha}_i^2 + 2\bar{\alpha}_r) + (\bar{\alpha}_r^2 - \bar{\alpha}_i^2)X \right]}$$

Von Neumann Stability condition : $|g(\theta)| \leq 1, \forall \theta$

Here it is equivalent, for $0 \leq X \leq 1$, to

$$(\alpha_i^2 - 2\alpha_r) + (\alpha_r^2 - \alpha_i^2)X \leq (\bar{\alpha}_i^2 + 2\bar{\alpha}_r) + (\bar{\alpha}_r^2 - \bar{\alpha}_i^2)X$$

$$\Leftrightarrow \begin{cases} \alpha_i^2 - 2\alpha_r \leq \bar{\alpha}_i^2 + 2\bar{\alpha}_r & \because X = 0 \\ \alpha_r^2 - 2\alpha_r \leq \bar{\alpha}_r^2 + 2\bar{\alpha}_r & \because X = 1 \end{cases}$$

VN Stability : Explicit centered (advection) scheme

$$\text{VN condition : } \alpha_i^2 - 2\alpha_r \leq \bar{\alpha}_i^2 + 2\bar{\alpha}_r \quad \text{and} \quad \alpha_r^2 - 2\alpha_i \leq \bar{\alpha}_r^2 + 2\bar{\alpha}_i$$

$$\text{Explicit centered scheme : } \alpha_i = 2\frac{\delta t c}{\delta \mathbf{x}}, \quad \alpha_r = 4\frac{\delta t \lambda}{\delta \mathbf{x}^2}, \quad \bar{\alpha}_i = \bar{\alpha}_r = 0.$$

Then the VN Stability condition is satisfied when δt and $\delta \mathbf{x}$ satisfies the following conditions :

$$\nu = \frac{\delta t \lambda}{\delta \mathbf{x}^2} \leq \frac{1}{2} \quad \text{and} \quad Pe = \frac{c^2 \delta t}{\lambda} \leq 2$$

- **advection dominated asymptotic when $\lambda \rightarrow 0$** : the VN condition require a time step $\delta t \rightarrow 0$: unstable!!!
- **diffusion dominated asymptotic when $c \rightarrow 0$** : the VN condition require a time step $\frac{\delta t \lambda}{\delta \mathbf{x}^2} \leq \frac{1}{2}$. see previous lectures.

L_∞ Stability

L_∞ Stability condition

Let us consider a numerical scheme given by

$$\tilde{T}_j^{n+1} = \sum_k \alpha_k \tilde{T}_{j(k)}^n$$

This numerical scheme is L_∞ when (sufficient condition) :

$$\sum_k \alpha_k = 1 \quad \text{and} \quad \alpha_k \geq 0, \quad \forall k$$

Indeed, These conditions insure that

$$\|\tilde{\mathbf{T}}^{n+1}\|_\infty \leq \|\tilde{\mathbf{T}}^n\|_\infty$$

L_∞ Stability of Explicit schemes ($\|\tilde{T}\|_\infty = \max_j |\tilde{T}_j|$)

$$\tilde{T}_j^{n+1} - \tilde{T}_j^n + \sigma \left(\tilde{T}_{j+1}^n - \tilde{T}_{j-1}^n \right) = (\nu + \xi) \left(\tilde{T}_{j+1}^n - 2\tilde{T}_j^n + \tilde{T}_{j-1}^n \right)$$

where $\sigma = \frac{c\delta t}{\delta x}$, $\xi = \frac{\zeta\delta t}{\delta x}$ and $\nu = \frac{\lambda\delta t}{\delta x^2}$

$$\tilde{T}_j^{n+1} = \left[\nu + \xi - \sigma \right] \tilde{T}_{j+1}^n + \left[1 - 2(\nu + \xi) \right] \tilde{T}_j^n + \left[\nu + \xi + \sigma \right] \tilde{T}_{j-1}^n$$

The scheme is L_∞ stable if $\nu + \xi \pm \sigma \geq 0$ and $\nu + \xi \leq \frac{1}{2}$.

$$\nu + \xi \leq \frac{1}{2} \iff \frac{c\delta t}{\delta x} + \frac{\lambda\delta t}{\delta x^2} \leq \frac{1}{2}$$

Indeed (in this case)

$$\begin{aligned} |\tilde{T}_j^{n+1}| &\leq \left[\nu + \xi - \sigma \right] |\tilde{T}_{j+1}^n| + \left[1 - 2(\nu + \xi) \right] |\tilde{T}_j^n| + \left[\nu + \xi + \sigma \right] |\tilde{T}_{j-1}^n| \\ \|\tilde{T}^{n+1}\|_\infty &= \|\mathcal{G}(\tilde{T}^n)\|_\infty \leq \|\tilde{T}^n\|_\infty \end{aligned}$$

L_1 Stability of Explicit schemes ($\|\tilde{T}\|_1 = \delta x \sum_j |\tilde{T}_j|$)

$$\tilde{T}_j^{n+1} = [\nu + \xi - \sigma] \tilde{T}_{j+1}^n + [1 - 2(\nu + \xi)] \tilde{T}_j^n + [\nu + \xi + \sigma] \tilde{T}_{j-1}^n$$

The scheme is L_1 stable if $\nu + \xi \pm \sigma \geq 0$ and $\nu + \xi \leq \frac{1}{2}$. Indeed (in this case) ($\|\tilde{T}\|_\infty$)

$$|\tilde{T}_j^{n+1}| \leq [\nu + \xi - \sigma] |\tilde{T}_{j+1}^n| + [1 - 2(\nu + \xi)] |\tilde{T}_j^n| + [\nu + \xi + \sigma] |\tilde{T}_{j-1}^n|$$

$$\|\tilde{T}^{n+1}\|_1 = \|\mathcal{G}(\tilde{T}^n)\|_1 \leq \|\tilde{T}^n\|_1$$

Then

$$\|\mathcal{G}\|_1 = \sup_{\tilde{T}^n \neq 0} \left(\frac{\|\mathcal{G}(\tilde{T}^n)\|_1}{\|\tilde{T}^n\|_1} \right) \leq 1$$

L_2 Stability of Explicit schemes ($\|\tilde{T}\|_2^2 = \delta x^2 \sum_j |\tilde{T}_j|^2$)

$$\tilde{T}_j^{n+1} = \alpha_{+1} \tilde{T}_{j+1}^n + \alpha_0 \tilde{T}_j^n + \alpha_{-1} \tilde{T}_{j-1}^n$$

with $\alpha_{+1} = \nu + \xi - \sigma$, $\alpha_0 = 1 - 2(\nu + \xi)$, $\alpha_{-1} = \nu + \xi + \sigma$,
 Extension of the numerical scheme to functions

$$\tilde{T}^{n+1}(\mathbf{x}) = \alpha_{+1} \tilde{T}^n(\mathbf{x} + \delta \mathbf{x}) + \alpha_0 \tilde{T}^n(\mathbf{x}) + \alpha_{-1} \tilde{T}^n(\mathbf{x} - \delta \mathbf{x})$$

Fourier transform of this extension (assumed periodic) :

$$\begin{aligned} \hat{T}^{n+1}(k) &= \left[\alpha_{+1} \exp(i k \delta \mathbf{x}) + \alpha_0 + \alpha_{-1} \exp(-i \xi \delta \mathbf{x}) \right] \hat{T}^n(k) \\ &= g(\theta_k, \delta \mathbf{x}, \delta t) \hat{T}^n(k) \end{aligned}$$

$$\theta_k = k \delta \mathbf{x}$$

$$g(\theta, \delta \mathbf{x}, \delta t) = 1 + (\nu + \xi) \left[-2 + 2 \cos(\theta) \right] - 2i\sigma \sin(\theta)$$

L_2 Stability of Explicit schemes ($\|\tilde{T}\|_2^2 = \delta x^2 \sum_j |\tilde{T}_j|^2$)

$$\hat{T}^{n+1}(k) = g(\theta_k, \delta \mathbf{x}, \delta t) \hat{T}^n(k)$$

where the amplification factor is

$$\begin{aligned} g(\theta, \delta \mathbf{x}, \delta t) &= 1 + (\nu + \xi) \left[-2 + 2 \cos(\theta) \right] - 2i\sigma \sin(\theta) \\ &= 1 - 4(\nu + \xi) \sin^2\left(\frac{\theta}{2}\right) - 2i\sigma \sin(\theta) \end{aligned}$$

The scheme is L_2 stable if $\forall \theta$ we have $|g(\theta, \delta \mathbf{x}, \delta t)| \leq 1$ (VN Condition). Indeed, in this case (using Plancherel relation)

$$\begin{aligned} \|\tilde{T}^{n+1}\|_2^2 &= \|\hat{T}^{n+1}\|_2^2 = \delta x^2 \sum_k |\hat{T}^{n+1}(k)|^2 = \delta x^2 \sum_k |\hat{T}^n(k)|^2 \\ &\leq \max_k \left[|g(\theta_k, \delta \mathbf{x}, \delta t)| \right] \delta x^2 \sum_k |\hat{T}^n(k)|^2 \\ &\leq \max_k \left[|g(\theta_k, \delta \mathbf{x}, \delta t)| \right] \|\tilde{T}^n\|_2^2 \end{aligned}$$

L_2 Stability of Explicit schemes ($\|\tilde{T}\|_2^2 = \delta x^2 \sum_j |\tilde{T}_j|^2$)

$$\|\tilde{T}^{n+1}\|_2^2 \leq \max_k \left[|g(\theta_k, \delta \mathbf{x}, \delta t)| \right] \|\tilde{T}\|_2^2$$

$$\|\mathcal{G}\|_2^2 = \sup_{\tilde{T}^n \neq 0} \left(\frac{\|\mathcal{G}(\tilde{T}^n)\|_2^2}{\|\tilde{T}^n\|_2^2} \right) \leq \max_k \left[|g(\theta_k, \delta \mathbf{x}, \delta t)| \right]$$

$$g(\theta_k, \delta \mathbf{x}, \delta t) = \left[\alpha_{+1} \exp(i\theta_k) + \alpha_0 + \alpha_{-1} \exp(-i\theta_k) \right]$$

If $\alpha_{+1} \geq 0$, $\alpha_0 \geq 0$ and $\alpha_{-1} \geq 0$ (Courant-Friedrichs-Lewy condition) then the scheme is L_2 stable. Indeed, in this case

$$|g(\theta_k, \delta \mathbf{x}, \delta t)| \leq \alpha_{+1} |\exp(i\theta_k)| + \alpha_0 + \alpha_{-1} |\exp(-i\theta_k)| = 1$$

L_∞ Stability of Implicit schemes ($\|\tilde{T}\|_\infty = \max_j |\tilde{T}_j|$)

$$\tilde{T}_j^{n+1} - \tilde{T}_j^n + \sigma \left(\tilde{T}_{j+1}^{n+1} - \tilde{T}_{j-1}^{n+1} \right) = (\nu + \xi) \left(\tilde{T}_{j+1}^{n+1} - 2\tilde{T}_j^{n+1} + \tilde{T}_{j-1}^{n+1} \right)$$

If we assume $\sigma = \frac{c\delta t}{\delta x} \geq 0$, $\xi = \frac{\zeta\delta t}{\delta x}$, $\nu = \frac{\lambda\delta t}{\delta x^2}$ and $\nu + \xi \geq 0$

$$\tilde{T}_j^{n+1} + \sigma\tilde{T}_{j+1}^{n+1} + (\nu + \xi) \left(\tilde{T}_{j+1}^{n+1} + \tilde{T}_{j-1}^{n+1} \right) = \tilde{T}_j^n + \sigma\tilde{T}_{j+1}^{n+1} + 2(\nu + \xi)\tilde{T}_j^{n+1}$$

Therefore

$$\|\tilde{T}^{n+1}\|_\infty (1 + \sigma + 2(\nu + \xi)) \leq \|\tilde{T}^n\|_\infty + (\sigma + 2(\nu + \xi)) \|\tilde{T}^{n+1}\|_\infty$$

Do it for

- a) $\sigma \leq 0$ and $\nu + \xi \leq 0$.
- b) $\sigma \leq 0$ and $\nu + \xi \geq 0$.
- c) $\sigma \geq 0$ and $\nu + \xi \leq 0$.

$$\implies \|\tilde{T}^{n+1}\|_\infty \leq \|\tilde{T}^n\|_\infty$$

unconditionnal stability for implicit schemes.

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- 3 Scalar Advection-Diffusion Equation.
- 4 Approximation of a Scalar 1D ODE.
- 5 FD for 1D scalar poisson equation (elliptic).
- 6 FD for 1D scalar diffusion equation (parabolic).
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