

Lecture 18: Time stepping

Logistics: - Australia 3/30 to 4/26

⇒ next week is last in person

after either zoom or pre-recorded!

reschedule office hours or Piazza

- HW 6 due (9/17)?

- HW 7 will be posted

Last time: - Diffusion

$$\frac{\partial c}{\partial t} - \nabla \cdot [D_m \nabla c] = 0$$

$$\frac{\partial c}{\partial t} - D \frac{\partial^2 c}{\partial x^2} = 0$$

- Boltzmann variable:

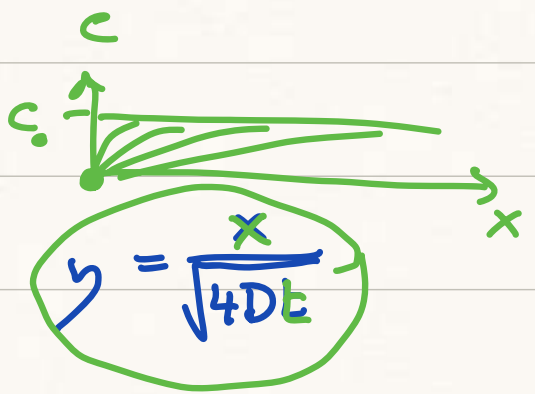
⇒ similarity solution

PDE → ODE

- analytic solution

$$c(x,t) = c_0 \operatorname{erf}\left(\frac{x}{\sqrt{4Dt}}\right)$$

$Pe \rightarrow 0$



Today: - Numerical solution of transient problems ⇒ time stepping

Transient Diffusion → Numerical solution

Transient PDE

$$\phi \frac{\partial c}{\partial t} - \nabla \cdot [\phi D_m \nabla c] = f_s \quad \underline{Kd} = \phi D_m \underline{\Gamma}$$

$$\underline{Lc} = - \underline{D} * \underline{Kd} * \underline{G} * \underline{c}$$

⇒ just need to discretize time derivative

Simple finite difference: $\frac{\partial c}{\partial t} \approx \frac{c^{n+1} - c^n}{\Delta t}$

$$\Delta t = t^{n+1} - t^n$$

substitute into PDE:

$$\underline{\Phi} \frac{c^{n+1} - c^n}{\Delta t} + \underline{L} c = f_s$$

$$\underline{\Phi} = \text{spdiags}(\underline{\phi}, 0, Nx, Nx)$$

$$\underline{\Phi} (c^{n+1} - c^n) + \Delta t \underline{L} c = \Delta t f_s$$

↑
what time level?

Theta method

Need to decide time level of $\underline{\underline{L}} \underline{\underline{c}}$

choose: $\underline{\underline{c}}^\theta = \theta \underline{\underline{c}}^n + (1-\theta) \underline{\underline{c}}^{n+1}$ $0 \leq \theta \leq 1$

substitute

$$\underline{\underline{\Phi}} (\underline{\underline{c}}^{n+1} - \underline{\underline{c}}^n) + \Delta t \underline{\underline{L}} (\theta \underline{\underline{c}}^n + (1-\theta) \underline{\underline{c}}^{n+1}) = \Delta t \underline{\underline{f}}_s$$

$\underline{\underline{c}}^n$ is known but $\underline{\underline{c}}^{n+1}$ needs to be calculated

collect unknown $\underline{\underline{c}}^{n+1}$ on l.h.s.

$$\underline{\underline{\Phi}} \underline{\underline{c}}^{n+1} + \Delta t (1-\theta) \underline{\underline{L}} \underline{\underline{c}}^{n+1} = \underline{\underline{\Phi}} \underline{\underline{c}}^n - \Delta t \theta \underline{\underline{L}} \underline{\underline{c}}^n + \Delta t \underline{\underline{f}}_s$$

pull out $\underline{\underline{c}}^{n+1}$ and $\underline{\underline{c}}^n$

$$\underbrace{[\underline{\underline{\Phi}} + \Delta t (1-\theta) \underline{\underline{L}}]}_{\underline{\underline{IM}}} \underline{\underline{c}}^{n+1} = \underbrace{[\underline{\underline{\Phi}} - \Delta t \theta \underline{\underline{L}}]}_{\underline{\underline{EX}}} \underline{\underline{c}}^n + \Delta t \underline{\underline{f}}_s$$

Linear system for single time step:

$$\underline{\underline{IM}} \underline{\underline{c}}^{n+1} = \underline{\underline{EX}} \underline{\underline{c}}^n + \Delta t \underline{\underline{f}}_s$$

implicit matrix: $\underline{\underline{IM}} = \underline{\underline{\Phi}} + \Delta t (1-\theta) \underline{\underline{L}}$ $\underline{\underline{L}} = -\underline{\underline{D}} \times |\underline{\underline{K}}| \times \underline{\underline{G}}$

$$\underline{\underline{EX}} = \underline{\underline{\Phi}} - \Delta t \theta \underline{\underline{L}}$$

Note: In general, we need to solve linear systems at every time step

Properties of Theta Method

$\theta = 1$: Forward Euler Method

$\underline{\underline{M}} = \underline{\underline{\Phi}}$ is diagonal

$$\Rightarrow \underline{\underline{c}}^{n+1} = \underline{\underline{\Phi}}^{-1} (\underline{\underline{E}}_x \underline{\underline{c}}^n + \Delta t \underline{\underline{f}}_s)$$

- explicit method
- only requires matrix vector multiply (cheap)
- conditionally stable $\Delta t \leq \frac{\Delta x^2}{2D_m}$
- first-order accurate

$\theta = 0$: Backward Euler Method

$\underline{\underline{E}}_x = \underline{\underline{\Phi}}$ diagonal

$\underline{\underline{M}}$ not diagonal

$$\underline{\underline{M}} \underline{\underline{c}}^{n+1} = \underline{\underline{E}}_x \underline{\underline{c}}^n + \Delta t \underline{\underline{f}}_s \quad \text{implicit}$$

- solve linear system \rightarrow solve_lbvpr.m

\Rightarrow expensive

- unconditionally stable
- first-order accurate

$\theta = \frac{1}{2}$: Crank-Nicholson Method

$$\cdot \underline{IM} \underline{c}^{n+1} = \underline{EX} \underline{c}^n + \Delta t \underline{f}_s$$

→ implicit method → solve linear system

$$\cdot c^\theta = \frac{1}{2} \underline{c}^n + \frac{1}{2} \underline{c}^{n+1} = \frac{\underline{c}^n + \underline{c}^{n+1}}{2} \text{ average}$$

⇒ central difference in time

⇒ second order accurate

• unconditionally stable

but has an oscillation limit

Amplification matrix

$$\underline{IM} \underline{c}^{n+1} = \underline{EX} \underline{c}^n + \Delta t \underline{f}_s + \underset{\downarrow}{\Delta t} \underline{f}_u + \underline{f}_D$$

In absence of r.h.s terms, i.e. no sink/source ($\underline{f}_s = 0$)

and homogeneous BC's ($\underline{f}_u = \underline{f}_D = 0$)

$$\underline{IM} \underline{c}^{n+1} = \underline{EX} \underline{c}^n$$

$$\underbrace{\underline{IM}^{-1} \underline{IM}}_{\underline{I}} \underline{c}^{n+1} = \underline{IM}^{-1} \underline{EX} \underline{c}^n$$

$$\underline{c}^{n+1} = \underbrace{\underline{IM}^{-1} \underline{EX}}_{\underline{A}} \underline{c}^n$$

\underline{A} = amplification matrix

$$\underline{c}^{n+1} = \underline{A} \underline{c}^n = \underline{A} (\underline{A} \underline{c}^{n-1}) = \underline{A}^2 \underline{c}^{n-1} = \underline{A}^3 \underline{c}^{n-2} = \dots = \underline{A}^n \underline{c}^0$$

where $\underline{c}^0 =$ initial condition ↑

To evolve the solution we just keep multiplying by \underline{A}

Evaluate matrix exponential \underline{A}^n using

the spectral decomposition: $\underline{A} = \underline{Q} \underline{\Lambda} \underline{Q}^{-1}$

where \underline{Q} is square matrix of eigenvectors (columns)

$\underline{\Lambda}$ is diagonal matrix of eigenvalues

$$\underline{A}^2 = \underline{A} \underline{A} = \underline{Q} \underline{\Lambda} \underbrace{\underline{Q}^{-1} \underline{Q}}_{\underline{I}} \underline{\Lambda} \underline{Q}^{-1} = \underline{Q} \underline{\Lambda}^2 \underline{Q}^{-1}$$

in general: $\underline{A}^n = \underline{Q} \underline{\Lambda}^n \underline{Q}^{-1}$

The solution decays, i.e., method is stable, if

$$\max |\lambda_n| \leq 1$$