

Class 7

- von Neumann analysis
 - implicit vs explicit schemes
 - iterative methods for linear systems
 - implicit operators in OpenFOAM
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Remainder of semester

- Incompressible N-S, projection methods, meshing (Class 8)
- Final projects, lift/drag, turbulence modeling
- Multiphase flow

Return HW this week.

- Feedback on reading period
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$$\begin{cases} \partial_t u + a \partial_x u = v \partial_{xx} u \\ u(0) = u(2\pi) \end{cases}$$

- Recall our FD scheme

$$\frac{u_i^{n+1} - u_i^n}{\Delta t} = -a \left(\frac{u_i^n - u_{i-1}^n}{\Delta x} \right) + v \left(\frac{u_{i+1}^n - 2u_i^n + u_{i-1}^n}{\Delta x^2} \right)$$

$$u_i^{n+1} = u_i^n - K_1 (u_i^n - u_{i-1}^n) + K_2 (u_{i+1}^n - 2u_i^n + u_{i-1}^n)$$

- From PDE, we know that linear PDE take the form for a single Fourier mode

$$u(x,t) = \hat{u}(t) \exp(ikx)$$

- Plugging this in

$$u_i^n = u(x_i, t^n)$$

$$u_i^n - u_{i-1}^n = \hat{u}(t^n) \left[\exp(ikx_n) - \exp(ik(x_n - \Delta x)) \right]$$

$$= \hat{u}(t^n) \exp(ikx_n) \left(1 - \exp(-ik\Delta x) \right)$$

$$u_{i+1}^n - 2u_i^n + u_{i-1}^n = \hat{u}(t^n) \left[\exp(ik(x_n + \Delta x)) - 2e^{ikx_n} + e^{ik(x_n - \Delta x)} \right]$$

$$= \hat{u}(t^n) e^{ikx_n} \left(e^{k\Delta x} - 2 + e^{-k\Delta x} \right)$$

$$2 \cos k\Delta x - 2$$

$$\hat{u}(t^{n+1}) e^{ikx} = \hat{u}(t^n) e^{ikx} \left[1 - K_1 (1 - \cos k\Delta x + i \sin k\Delta x) + 2K_2 (\cos k\Delta x - 1) \right]$$

What condition such that mode doesn't grow?

$$\left| \frac{\hat{u}(t^{n+1})}{\hat{u}(t^n)} \right| < 1 \quad ?$$

(Take $K_1 = 0$ for simplicity)

$$= \left| 1 - 2K_2 (\underbrace{\cos k\Delta x - 1}_{\text{between } 0 \text{ and } -2}) \right| < 1$$

$$\leq |1 - 4K_2|$$

$$K_2 \leq \frac{1}{2}$$

\Rightarrow

$$\frac{v \Delta t}{\Delta x^2} < \frac{1}{2}$$

Similarly (HW)

$$\frac{a \Delta t}{\Delta x} < 1$$

We can see the big issue... what if v is large?
Then we need a tiny timestep.

Implicit schemes

$$\frac{U_i^{n+1} - U_i^n}{\Delta t} = -a \left(\frac{U_i^{n+1} - U_{i-1}^{n+1}}{\Delta x} \right) + \nu \left(\frac{U_{i+1}^{n+1} - 2U_i^{n+1} + U_{i-1}^{n+1}}{\Delta x^2} \right)$$

Take $a=0$, Let $K_2 = \frac{\nu \Delta t}{\Delta x^2}$

$$U_i^{n+1} = U_i^n + K_2 \left(U_{i+1}^{n+1} - 2U_i^{n+1} + U_{i-1}^{n+1} \right)$$

Again, ansatz $u(x,t) = \hat{u}(t) e^{ikx}$

$$\hat{u}^{n+1} e^{ikx} = \hat{u}^n e^{ikx} + K_2 \hat{u}^{n+1} \left(e^{ik(x+\Delta x)} - 2e^{ikx} + e^{ik(x-\Delta x)} \right)$$

$$\hat{u}^{n+1} = \hat{u}^n + K_2 \hat{u}^{n+1} \underbrace{\left(e^{ik\Delta x} - 2 + e^{-ik\Delta x} \right)}_{2 \cos k\Delta x - 2}$$

$$\hat{u}^{n+1} \left(1 - K_2 (2 - 2 \cos k\Delta x) \right) = \hat{u}^n$$

$$\left| \frac{\hat{u}^{n+1}}{\hat{u}^n} \right| \leq \left| \frac{1}{1 + \underbrace{2K_2(1 - \cos k\Delta x)}_{\text{between } 0, 2}} \right| \leq 1$$

→ unconditionally ~~stable~~ stable

Almost there - how to solve?

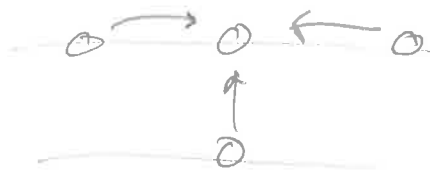
- $\vec{u}_a^{n+1} = A \setminus \vec{u}^n$ in Matlab will be extremely expensive (depending on how careful you are)
- Gauss elimination is $O(N^3)$ (dense solver)
- Need it to be fast enough that we're better off than using explicit scheme
 - explicit \rightarrow steps $\rightarrow \frac{t_{\text{final}}}{\Delta t}$
 $\Delta t \sim 1/N^2 \Rightarrow N^2$ steps, each w/ $O(N)$ updates $\Rightarrow O(N^3)$
 - implicit $\rightarrow \Delta t \sim 1/N \Rightarrow N$ steps, need $O(N^2)$ update or better to break even
- even storing the matrix is expensive ($O(N^2)$)

Iterative Sparse Solvers

While direct solvers provide an exact (to machine precision) solution in $O(N^3)$ time, iterative solvers take an initial guess x_0 and generate a sequence $\{x_i\}_{i=1,2,\dots}$ such that the residual $r_i = b - Ax_i$ satisfies $\|r_i\| \downarrow 0$ for large i , eg. 0.1%

Key idea Start with a good guess, terminate when $\frac{\|r_i\|}{\|b\|} < \epsilon$

So how do we implement this? Each point of the stencil relies on other stencil information



Collect terms

$$\underbrace{u_{i+1}^{n+1} (-K_2)}_{\alpha_{i,i+1}} + \underbrace{u_i^{n+1} (1 + K_1 + 2K_2)}_{\alpha_{i,i}} + \underbrace{u_{i-1}^{n+1} (-K_1 - K_2)}_{\alpha_{i,i-1}} = u_i^n$$

Build a big system of equations

$$A u^{n+1} = u^n$$



$$\begin{pmatrix}
 \alpha_{00} \alpha_{01} & & & & \\
 \alpha_{10} & \alpha_{11} & \alpha_{12} & & \\
 \alpha_{21} & \alpha_{22} & \alpha_{23} & & \\
 & \dots & \dots & \dots & \\
 \alpha_{N-1} & \alpha_{N} & & &
 \end{pmatrix}
 \begin{pmatrix}
 u_0^{n+1} \\
 u_1^{n+1} \\
 \vdots \\
 u_N^{n+1}
 \end{pmatrix} = \begin{pmatrix}
 u_0^n \\
 \vdots \\
 u_N^n
 \end{pmatrix}$$

Jacobi's method

idea Want an update formula

$$x_{n+1} = x_n + B r_n$$

what if we had A^{-1} ?

$$Ax = b$$

$$x_{n+1} = x_n + A^{-1}(b - Ax_n)$$

$$\begin{aligned} x_{n+1} &= x_n + x_{\text{exact}} - x_n \\ &= x_{\text{exact}} \end{aligned}$$

But if we knew A^{-1} we'd be done already....

Split $Ax = b$

$$Bx^{n+1} + (A-B)x^n = b$$

$$x^{n+1} = B^{-1}(b - (A-B)x^n)$$

Want B that's cheap to invert

Jacobi $\Rightarrow B = \text{diag}(A)$

for Take $x_0 = x^n$

$$x_i^{n+1} = \frac{1}{a_{ii}} \left(x_i^n - \alpha_{i,i-1} x_{i-1}^n - \alpha_{i,i+1} x_{i+1}^n \right)$$

w/

Algorithm (what goes in your update function)

update() {

last timestep
compute this without storing matrix

$$- X^n = X^{\text{old}}$$

$$- r = b - AX^{\text{old}}$$

$$- \epsilon = \|r\|_2$$

while $\epsilon < \text{TOL}$

for $i = 1 : N$

$$X_i^{n+1} = \frac{1}{\alpha_{ii}} \left(X_i^n - \alpha_{i,i+1} X_{i+1}^n - \alpha_{i,i-1} X_{i-1}^n \right)$$

end

$$r_i = b_i - A_{i,:} X^{n+1}$$

$$\epsilon = \|r\|_2$$

$$X_i^n = X_i^{n+1}$$

end

Some choices for stopping criteria

$$\epsilon = \|r\|_2$$

$$\epsilon = \frac{\|r\|}{\|b\|}$$

$$\epsilon = \frac{\|X_{n+1} - X_n\|}{\|X_n\|}$$