

Lecture 7: Simple PDEs

Supporting textbook chapters for week 8: Chapters 9.1 - 9.3

Topics:

- Classifying PDEs
- Elliptic equation solvers: Jacobi, Gauss-Seidel, overrelaxation
- Parabolic equation solver: Explicit FTCS (Forward Time, Centered Space)
- Parabolic and hyperbolic equation solver: Implicit FTCS, Crank-Nicolson

Classification and General Approach

- Solving partial differential equations is one of the pinnacles of computational physics, bringing together many methods.
- Each type comes with design decisions on how to discretize and implement numerical methods,
- Stability is crucial.
- Accuracy is crucial too.

Recall conical equations in geometry:

$$\alpha x^2 + \beta xy + \gamma y^2 + \delta x + \varepsilon y = f,$$

classified using $\Delta = \beta^2 - 4\alpha\gamma$.

1. $\Delta = 0$: equation for a parabola,
2. $\Delta < 0$: equation for an ellipse,
3. $\Delta > 0$: equation for a hyperbola.

Similar for PDEs:

$$\alpha \frac{\partial^2 \phi}{\partial x^2} + \beta \frac{\partial^2 \phi}{\partial x \partial y} + \gamma \frac{\partial^2 \phi}{\partial y^2} + \delta \frac{\partial \phi}{\partial x} + \varepsilon \frac{\partial \phi}{\partial y} = f.$$

With $\Delta = \beta^2 - 4\alpha\gamma$,

1. $\Delta = 0$: parabolic PDE,
2. $\Delta < 0$: elliptic PDE,
3. $\Delta > 0$: hyperbolic PDE.

Physics Examples

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1. Canonical parabolic PDE: the diffusion equation, $\kappa \frac{\partial^2 T}{\partial x^2} - \frac{\partial T}{\partial t} = 0$,

$$x \rightarrow x, \quad y \rightarrow t, \quad \alpha \rightarrow \kappa, \quad \varepsilon \rightarrow -1, \quad \beta, \gamma, \delta, f \rightarrow 0 \quad \Rightarrow \quad \beta^2 - 4\alpha\gamma = 0.$$

2. Canonical elliptic PDE: the Poisson equation, $\nabla^2 \phi = \rho$,

$$x \rightarrow x, \quad y \rightarrow y, \quad \alpha, \gamma \rightarrow 1, f \rightarrow \rho, \beta, \delta, \varepsilon \rightarrow 0 \quad \Rightarrow \quad \beta^2 - 4\alpha\gamma = -4 < 0.$$

- e.g. 2D electrostatics, with electric potential ϕ s.t. $\vec{E} = \nabla \phi$, in the absence of charges ($\rho \equiv 0$), have Gauss' law: $\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} = 0$

3. Canonical hyperbolic PDE: the wave equation, $c^2 \frac{\partial^2 \phi}{\partial x^2} - \frac{\partial^2 \phi}{\partial t^2} = 0$.

$$x \rightarrow x, \quad y \rightarrow t, \quad \alpha \rightarrow c^2, \quad \gamma \rightarrow -1, \quad \beta, \delta, \varepsilon, f \rightarrow 0 \quad \Rightarrow \quad \beta^2 - 4\alpha\gamma = 4c^2 > 0.$$

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Note: we use these categorizations even when the spatial operator is $\nabla^2 = \partial_x^2 + \partial_y^2 + \partial_z^2$, i.e., for 4D PDEs.

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General approach

- Discretize system spatially and temporally: can use finite difference, spectral coefficients, etc.
- \Rightarrow set of coupled ODEs that you need to solve in an efficient way.
- Spatial derivatives bring information in from neighbouring points \Rightarrow coupling,
- \Rightarrow errors depend on space and time and can get wave-like characteristics.
- For 2nd derivatives, recall central difference calculation (§5.10.5, p.197):

$$f''(x) = \frac{f(x+h) - 2f(x) + f(x-h)}{h^2} - \frac{1}{12} h^2 f^{(4)}(x) + \dots$$

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Elliptic Equations

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Start with simplest case of Gauss's Law with 2D Laplacian:

$$0 = \nabla^2 \phi = \frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2},$$

On regular square grid of cell side length a , finite difference form is

$$\begin{aligned} \frac{\partial^2 \phi}{\partial x^2} &\approx \frac{\phi(x+a, y) - 2\phi(x, y) + \phi(x-a, y)}{a^2}, \\ \frac{\partial^2 \phi}{\partial y^2} &\approx \frac{\phi(x, y+a) - 2\phi(x, y) + \phi(x, y-a)}{a^2}. \end{aligned}$$

Gauss's law then becomes:

$$0 \approx \phi(x+a, y) + \phi(x-a, y) + \phi(x, y+a) + \phi(x, y-a) - 4\phi(x, y)$$

at each location (x, y) .

- Put together a series of equations of the form

$$\phi(x+a, y) + \phi(x-a, y) + \phi(x, y+a) + \phi(x, y-a) - 4\phi(x, y) = 0$$

for each x and y , subject to boundary conditions.

- ϕ or derivative $\partial\phi/\partial\xi$ ($\xi = x, y$, or both) given on boundary.
- If ϕ given, use this value for adjacent points.
- If $\partial\phi/\partial\xi$ given, find algebraic relationship between points near to boundary using finite difference.
- Could solve using matrix methods $\mathbf{L}\phi = \mathbf{R}\phi$, but a simpler method is possible.

Jacobi relaxation method

$$\phi(x+a, y) + \phi(x-a, y) + \phi(x, y+a) + \phi(x, y-a) - 4\phi(x) = 0$$

- Iterate the rule $\phi_{new}(x, y) = \frac{1}{4}[\phi(x+a, y) + \phi(x-a, y) + \phi(x, y+a) + \phi(x, y-a)]$.
- Much like the relaxation method for finding solutions of $f(x) = x$,
- For this problem it turns out that Jacobi Relaxation is always stable and so always gives a solution!

Overrelaxation method

$$\phi_{new}(x, y) =$$

$$(1 + \omega) \left[\frac{\phi(x+a, y) + \phi(x-a, y) + \phi(x, y+a) + \phi(x, y-a)}{4} \right] - \omega\phi(x, y)$$

- When it works, it usually speeds up the calculation.
- Not always stable! How to choose ω is not always reproducible.
- see Newman's exercise 6.11 for a similar problem for finding $f(x) = x$.

Gauss-Seidel method

- Replace function on the fly as in

$$\phi(x, y) \leftarrow \frac{\phi(x+a, y) + \phi(x-a, y) + \phi(x, y+a) + \phi(x, y-a)}{4}.$$

- Crucial difference: the LHS is ϕ , not ϕ_{new} : we use newer values as they are being computed (Jacobi used only old values to compute new one).
- This can be shown to run faster.
- Can be combined with overrelaxation.

Parabolic PDEs

- Consider the 1D heat equation:

$$\frac{\partial T}{\partial t} = \kappa \frac{\partial^2 T}{\partial x^2},$$

- B.Cs.:

$$T(x=0, t) = T_0, \quad T(x=L, t) = T_L.$$

- I.C.:

$$T(x, t=0) = T_0 + (T_L - T_0) \left(\frac{f(x) - f(0)}{f(L) - f(0)} \right)$$

Explicit Forward Time Centred Space method

Step 1: Discretize in space

$$x_m = \frac{m}{M}L = am, \quad m = 0 \dots M, \quad a = \frac{L}{M},$$

$$T_m(t) = [T_0(t), \dots, T_M(t)]$$

$$\left. \frac{\partial^2 T}{\partial x^2} \right|_{x=x_m, t} \approx \frac{T_{m+1} - 2T_m + T_{m-1}}{a^2} \quad (\text{"centred space", CS})$$

Step 2: Discretize in time

$$\frac{dT_m}{dt} \approx \kappa \frac{T_{m+1} - 2T_m + T_{m-1}}{a^2}, \quad m = 1 \dots, M-1$$

Let $t_n = nh$, h the time step. Let $T_m(t_n) \equiv T_m^n$.

$$\Rightarrow \left. \frac{\partial T}{\partial t} \right|_{x=ma, t=nh} \approx \frac{T_m^{n+1} - T_m^n}{h} \equiv \kappa \frac{T_{m+1}^n - 2T_m^n + T_{m-1}^n}{a^2} \quad (\text{"Forward Time", FT}).$$

\Rightarrow **Explicit FTCS method:**

$$T_m^{n+1} = T_m^n + \frac{\kappa h}{a^2} (T_{m+1}^n - 2T_m^n + T_{m-1}^n).$$

It may be easier to understand by writing the problem as a set of ODEs

$$\frac{\partial \phi_m}{\partial t} = \psi_m, \quad \text{and} \quad \frac{\partial \psi_m}{\partial t} = \frac{c^2}{a^2} (\phi_{m+1} - 2\phi_m + \phi_{m-1})$$

and the discretization in time as:

$$\begin{bmatrix} \phi_m^{n+1} \\ \psi_m^{n+1} \end{bmatrix} = \begin{bmatrix} 1 & +h \\ -\frac{2hc^2}{a^2} & 1 \end{bmatrix} \begin{bmatrix} \phi_m^n \\ \psi_m^n \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{c^2 h}{a^2} (\phi_{m+1}^n + \phi_{m-1}^n) \end{bmatrix}$$

Implicit FTCS Method

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Evaluate the RHS of the above at time $t + h$ instead of t

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- first do $h \rightarrow -h$ (from the current time step, compute the *previous* one):

$$\phi_m^{n-1} = \phi_m^n - h\psi_m^n,$$

$$\psi_m^{n-1} = \psi_m^n - h \frac{c^2}{a^2} (\phi_{m-1}^n + \phi_{m+1}^n - 2\phi_m^n),$$

- Then, $n \rightarrow n + 1$ (one shift forward in time):

$$\phi_m^n = \phi_m^{n+1} - h\psi_m^{n+1},$$

$$\psi_m^n = \psi_m^{n+1} - h \frac{c^2}{a^2} (\phi_{m-1}^{n+1} + \phi_{m+1}^{n+1} - 2\phi_m^{n+1}),$$

or

$$\begin{bmatrix} \phi_m^n \\ \psi_m^n \end{bmatrix} = \begin{bmatrix} 1 & -h \\ +\frac{2hc^2}{a^2} & 1 \end{bmatrix} \begin{bmatrix} \phi_m^{n+1} \\ \psi_m^{n+1} \end{bmatrix} - \begin{bmatrix} 0 \\ \frac{c^2 h}{a^2} (\phi_{m+1}^{n+1} + \phi_{m-1}^{n+1}) \end{bmatrix}$$

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"Implicit": we now have a set of simultaneous equations relating the values of ϕ , ψ at t to their values at $t + h$.

Why bother solving these simultaneous equations, rather than using an "explicit" expression for the values of ϕ , ψ at $t + h$ given their values at t ?

Because in certain cases, this is numerically stable while the explicit FTCS is not! (More about this later)

Note, it does often suffer from accuracy issues, where solutions decay to 0 over time.

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Crank-Nicolson

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Average of explicit and implicit methods.

Explicit ('forward'):

$$\phi_m^{n+1} = \phi_m^n + h\psi_m^n, \quad \psi_m^{n+1} = \psi_m^n + h \frac{c^2}{a^2} (\phi_{m-1}^n + \phi_{m+1}^n - 2\phi_m^n).$$

Implicit ('backward'):

$$\phi_m^{n+1} - h\psi_m^{n+1} = \phi_m^n, \quad \psi_m^n = \psi_m^{n+1} - h \frac{c^2}{a^2} (\phi_{m-1}^{n+1} + \phi_{m+1}^{n+1} - 2\phi_m^{n+1}).$$

Crank-Nicolson (C-N):

$$\begin{aligned} \phi_m^{n+1} - \frac{h}{2} \psi_m^{n+1} &= \phi_m^n + \frac{h}{2} \psi_m^n \\ \psi_m^{n+1} - \frac{h}{2} \frac{c^2}{a^2} (\phi_{m-1}^{n+1} + \phi_{m+1}^{n+1} - 2\phi_m^{n+1}) &= \psi_m^n + \frac{h}{2} \frac{c^2}{a^2} (\phi_{m-1}^n + \phi_{m+1}^n - 2\phi_m^n). \end{aligned}$$

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C-N is 2nd-order accurate in time, while both explicit and implicit methods are 1st-order accurate. So, C-N often solves the 'decaying to 0' issues encountered with the implicit method.

Hyperbolic PDEs

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Explicit FTCS is always unstable. Use C-N, or spectral methods (next time)

