

Lecture 1: Introduction to finite difference methods

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Parabolic heat equation

We start with the simple 1D parabolic PDE which describes the change in non-dimensional temperature of a 1D rod

$$\frac{\partial V}{\partial t} = \frac{\partial^2 V}{\partial x^2}$$

to be solved on $0 < x < 1$, subject to some initial conditions $V_0(x)$ at time $t=0$, and $V(0, t)=V(1, t)=0$ on the two ends.

Parabolic heat equation

The exact solution can be expressed as a combination of Fourier modes:

$$V(x, t) = \sum_{m>0} A_m \sin(m\pi x) \exp(-m^2\pi^2 t)$$

in which the amplitudes are given by

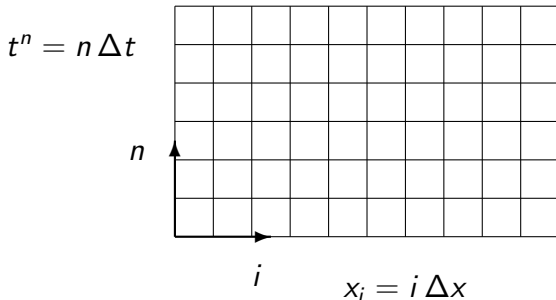
$$A_{m'} = 2 \int_0^1 V_0(x) \sin(m'\pi x) dx$$

since

$$\int_0^1 \sin(m\pi x) \sin(m'\pi x) dx = \begin{cases} \frac{1}{2}, & \text{if } m=m' \\ 0, & \text{otherwise} \end{cases}$$

Finite Difference Approximation

Suppose we use a computational grid with spacing $\Delta x = 1/K$ and timestep Δt :



Finite Difference Approximation

We want to construct an approximation $V_i^n \approx V(x_i, t^n)$.

To do this, we note that

$$V(x_i, t^n + \Delta t) \approx V(x_i, t^n) + \Delta t \left. \frac{\partial V}{\partial t} \right|_{(x_i, t^n)}$$

$$\implies \frac{\partial V}{\partial t} \approx \frac{1}{\Delta t} (V_i^{n+1} - V_i^n)$$

and also

$$V(x_i \pm \Delta x, t^n) \approx V(x_i, t^n) \pm \Delta x \left. \frac{\partial V}{\partial x} \right|_{(x_i, t^n)} + \frac{1}{2} \Delta x^2 \left. \frac{\partial^2 V}{\partial x^2} \right|_{(x_i, t^n)}$$

$$\implies \frac{\partial^2 V}{\partial x^2} \approx \frac{1}{\Delta x^2} (V_{i+1}^n - 2 V_i^n + V_{i-1}^n)$$

Finite Difference Approximation

Putting these together gives the explicit forward-time central-space approximation of the heat equation:

$$\frac{1}{\Delta t} (V_i^{n+1} - V_i^n) = \frac{1}{(\Delta x)^2} (V_{i+1}^n - 2V_i^n + V_{i-1}^n)$$

which, setting $\lambda = \Delta t / (\Delta x)^2$, can be re-arranged to give

$$\begin{aligned} V_i^{n+1} &= V_i^n + \lambda (V_{i+1}^n - 2V_i^n + V_{i-1}^n) \\ &= (1 - 2\lambda) V_i^n + \lambda (V_{i+1}^n + V_{i-1}^n) \end{aligned}$$

together with zero values for V_i^n on the two ends.

Finite Difference Approximation

The finite difference solution also has a Fourier mode decomposition of the form

$$V_i^n = \sum_{0 < m < 1/\Delta x} A_m^n \sin(m\pi x_i)$$

where the amplitudes A_m^n satisfy the equation

$$A_m^{n+1} = \left(1 - 4\lambda \sin^2\left(\frac{1}{2}m\Delta x\right)\right) A_m^n$$

We know the amplitudes should decay exponentially – the condition for this to happen is

$$-1 < \left(1 - 4\lambda \sin^2\left(\frac{1}{2}m\Delta x\right)\right) < 1$$

which requires

$$4\lambda \leq 2 \quad \implies \quad \Delta t \leq \frac{1}{2}(\Delta x)^2$$

Finite Difference Approximation

This timestep stability limit also follows directly from

$$V_i^{n+1} = (1 - 2\lambda)V_i^n + \lambda(V_{i+1}^n + V_{i-1}^n)$$

since for $\lambda \leq 1/2$ we get

$$\begin{aligned} |V_i^{n+1}| &\leq (1 - 2\lambda)|V_i^n| + \lambda|V_{i+1}^n| + \lambda|V_{i-1}^n| \\ &\leq (1 - 2\lambda) \max_{i'} |V_{i'}^n| + \lambda \max_{i'} |V_{i'}^n| + \lambda \max_{i'} |V_{i'}^n| \\ &\leq \max_{i'} |V_{i'}^n| \end{aligned}$$

so

$$\max_i |V_i^{n+1}| \leq \max_i |V_i^n|$$

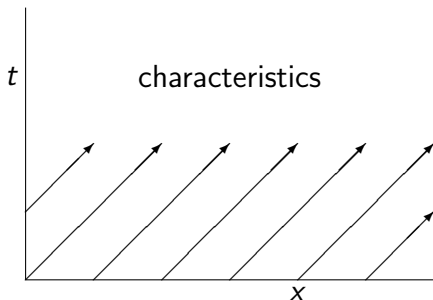
Hyperbolic equation

Another simple example is the 1D hyperbolic PDE which models convection:

$$\frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} = 0$$

to be solved again on $0 < x < 1$, subject to some initial conditions $V_0(x)$ at time $t=0$, and $V(0, t)=0$ on the left-hand end.

The exact solution has the form
 $V(x, t) = V_0(x-t)$
with $V_0(-t) = 0$.



Finite Difference Approximation

Using

$$\frac{\partial V}{\partial t} \approx \frac{1}{\Delta t} (V_i^{n+1} - V_i^n)$$

and

$$\frac{\partial V}{\partial x} \approx \frac{1}{\Delta x} (V_i^n - V_{i-1}^n)$$

gives the explicit upwind discretisation

$$\frac{1}{\Delta t} (V_i^{n+1} - V_i^n) + \frac{1}{\Delta x} (V_i^n - V_{i-1}^n) = 0$$

Finite Difference Approximation

Setting $\lambda = \Delta t / \Delta x$, this can be re-arranged to give

$$\begin{aligned} V_i^{n+1} &= V_i - \lambda (V_i^n - V_{i-1}^n) \\ &= (1 - \lambda) V_i^n + \lambda V_{i-1}^n \end{aligned}$$

together with $V_0^n = 0$.

If $\lambda \leq 1$ this is stable since we again get

$$\max_i |V_i^{n+1}| \leq \max_i |V_i^n|$$

Parabolic heat equation

As a 2D model problem, we consider the simple parabolic PDE which describes the change in non-dimensional temperature of a 2D plate

$$\frac{\partial V}{\partial t} = \frac{\partial^2 V}{\partial x^2} + \frac{\partial^2 V}{\partial y^2}$$

to be solved on the unit square $0 < x < 1$, $0 < y < 1$ subject to some initial conditions $V_0(x, y)$ at time $t=0$, and $V=0$ on the boundaries

Parabolic heat equation

The exact solution can be expressed as a combination of Fourier modes:

$$V(x, y, t) = \sum_{m,n>0} A_{m,n} \sin(m\pi x) \sin(n\pi y) \exp(-(m^2+n^2)\pi^2 t)$$

in which the amplitudes are given by

$$A_{m',n'} = 4 \int_0^1 \int_0^1 V_0(x, y) \sin(m'\pi x) \sin(n'\pi y) dx dy$$

since

$$\begin{aligned} \int_0^1 \int_0^1 \sin(m\pi x) \sin(n\pi y) \sin(m'\pi x) \sin(n'\pi y) dx dy \\ = \begin{cases} \frac{1}{4}, & \text{if } m=m', n=n' \\ 0, & \text{otherwise} \end{cases} \end{aligned}$$

Laplace equation

Suppose $U(x, y)$ is the solution to the 2D Laplace equation

$$\frac{\partial^2 U}{\partial x^2} + \frac{\partial^2 U}{\partial y^2} = 0$$

subject to specified values on the boundary.

If $W(x, y, t)$ is the solution of the parabolic PDE subject to those same boundary values, then $V \equiv W - U$ satisfies the parabolic PDE with zero boundary conditions.

Since $V(x, y, t) \rightarrow 0$ as $t \rightarrow \infty$, $W(x, y, t) \rightarrow U(x, y)$ as $t \rightarrow \infty$, which gives us one approach to approximating solutions of the Laplace equation.

Finite Difference Approximation

Using a similar explicit forward-time central-space approximation of the heat equation, with $\Delta x = \Delta y$, gives:

$$\begin{aligned}\frac{1}{\Delta t} (V_{i,j}^{n+1} - V_{i,j}^n) &= \frac{1}{(\Delta x)^2} (V_{i+1,j}^n - 2V_{i,j}^n + V_{i-1,j}^n) \\ &\quad + \frac{1}{(\Delta x)^2} (V_{i,j+1}^n - 2V_{i,j}^n + V_{i,j-1}^n)\end{aligned}$$

which, setting $\lambda = \Delta t/(\Delta x)^2$, can be re-arranged to give

$$\begin{aligned}V_{i,j}^{n+1} &= V_{i,j}^n + \lambda (V_{i+1,j}^n - 2V_{i,j}^n + V_{i-1,j}^n) \\ &\quad + \lambda (V_{i,j+1}^n - 2V_{i,j}^n + V_{i,j-1}^n) \\ &= (1 - 4\lambda)V_{i,j}^n + \lambda (V_{i+1,j}^n + V_{i-1,j}^n + V_{i,j+1}^n + V_{i,j-1}^n)\end{aligned}$$

together with zero values for $V_{i,j}^n$ on the boundary.

Finite Difference Approximation

It can be shown that the finite difference solution also has a Fourier mode decomposition of the form

$$V_{i,j}^n = \sum_{0 < k, m < 1/\Delta x} A_{k,m}^n \sin(k\pi x_i) \sin(m\pi y_j)$$

where the amplitudes $A_{k,m}^n$ satisfy the equation

$$A_{k,m}^{n+1} = \left(1 - 4\lambda \sin^2\left(\frac{1}{2}k\Delta x\right) - 4\lambda \sin^2\left(\frac{1}{2}m\Delta x\right)\right) A_{k,m}^n$$

We know the amplitudes should decay exponentially – the condition for this to happen is

$$-1 < \left(1 - 4\lambda \sin^2\left(\frac{1}{2}k\Delta x\right) - 4\lambda \sin^2\left(\frac{1}{2}m\Delta x\right)\right) < 1$$

which requires

$$8\lambda \leq 2 \implies \Delta t \leq \frac{1}{4}(\Delta x)^2$$

Finite Difference Approximation

When trying to solve the Laplace equation it is best to make λ (and hence Δt) as big as possible, while remaining stable.

Putting $\lambda = 1/4$ gives the Jacobi iteration

$$V_{i,j}^{n+1} = \frac{1}{4} (V_{i+1,j}^n + V_{i-1,j}^n + V_{i,j+1}^n + V_{i,j-1}^n)$$

This is applied to interior points; the values of the boundary points are fixed so for those we use simply

$$V_{i,j}^{n+1} = V_{i,j}^n$$

(There are other more efficient iterative methods, such as Conjugate Gradient and Multigrid, but we won't cover those here.)

Black-Scholes PDE in finance

The Black-Scholes PDE in mathematical finance for the value of a European option based on two financial assets each modelled by Geometric Brownian Motion is

$$\begin{aligned} \frac{\partial V}{\partial t} = & rV - rS_1 \frac{\partial V}{\partial S_1} - rS_2 \frac{\partial V}{\partial S_2} \\ & - \sigma^2 \left(\frac{1}{2} S_1^2 \frac{\partial^2 V}{\partial S_1^2} + \rho S_1 S_2 \frac{\partial^2 V}{\partial S_1 \partial S_2} + \frac{1}{2} S_2^2 \frac{\partial^2 V}{\partial S_2^2} \right) \end{aligned}$$

where r is the risk-free interest rate, σ is the volatility, and ρ is the correlation between the motion of the two assets

This is solved backwards in time on $(0, S_{max}) \times (0, S_{max}) \times (0, T)$ with the final value $V(S_1, S_2, T)$ equal to the payoff function, and subject to some boundary conditions at $S_1 = S_{max}$, $S_2 = S_{max}$, to get the value $V(S_1, S_2, 0)$ at the initial time $t = 0$.

Black-Scholes PDE

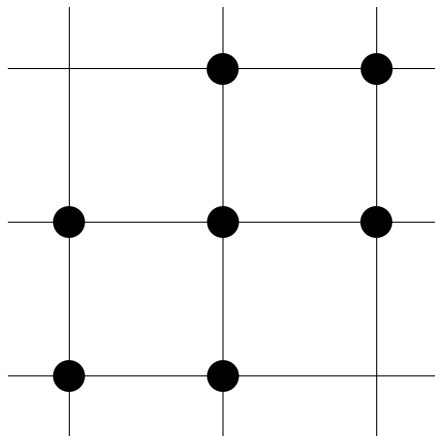
If $\rho > 0$ then a simple explicit Euler central space discretisation on a uniform grid is

$$\begin{aligned} V_{i,j}^{n+1} = & (1 - r\Delta t)V_{i,j}^n \\ & + \frac{r\Delta t}{2\Delta S} (S_{1,i} (V_{i+1,j}^n - V_{i-1,j}^n) + S_{2,j} (V_{i,j+1}^n - V_{i,j-1}^n)) \\ & + \frac{\sigma^2 S_{1,i}^2 \Delta t}{2\Delta S^2} (V_{i+1,j}^n - 2V_{i,j}^n + V_{i-1,j}^n) \\ & + \frac{\sigma^2 S_{2,j}^2 \Delta t}{2\Delta S^2} (V_{i,j+1}^n - 2V_{i,j}^n + V_{i,j-1}^n) \\ & + \frac{\rho\sigma^2 S_{1,j} S_{2,i} \Delta t}{2\Delta S^2} ((V_{i+1,j+1}^n - V_{i+1,j}^n - V_{i,j+1}^n + V_{i,j}^n) \\ & \quad + (V_{i,j}^n - V_{i-1,j}^n - V_{i,j-1}^n + V_{i-1,j-1}^n)) \end{aligned}$$

with time going backwards so that $t^n \equiv T - n\Delta t$.

Black-Scholes PDE

This gives a 7-point stencil:



Black-Scholes PDE

No boundary conditions are need when $i=0$ or $j=0$, because $V_{-1,j}$ and $V_{i,-1}$ are not used.

The boundary conditions on $i=I$ and $j=J$ are trickier, and a little unusual.

On $i=I$ usually omit term $(V_{I+1,j}^n - 2V_{I,j}^n + V_{I-1,j}^n)$ on the basis that $\frac{\partial^2 V}{\partial S_1^2} \approx 0$, for most European options, and replace

$(V_{I+1,j+1}^n - V_{I+1,j}^n - V_{I,j+1}^n + V_{I,j}^n)$ by $(V_{I,j}^n - V_{I-1,j}^n - V_{I,j-1}^n + V_{I-1,j-1}^n)$,
and $V_{I+1,j}^n - V_{I,j}^n$ with $V_{I,j}^n - V_{I-1,j}^n$ with similar changes when $j=J$.