

Parabolic PDEs: Finite Difference Methods

M.Sc. in Mathematical Modelling & Scientific Computing,
Practical Numerical Analysis

Michaelmas Term 2018, Lecture 9

1D Parabolic PDEs

1D Heat Equation

Last week we considered the simplest parabolic PDE in the form of the heat equation:

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

for $t > 0$ and $x \in [a, b]$ with an initial condition

$$u(x, 0) = u_0(x) ,$$

for $x \in [a, b]$. We began by considering Dirichlet boundary conditions

$$u(a, t) = u_a(t) ,$$

$$u(b, t) = u_b(t) ,$$

for $t > 0$.

Finite Difference Schemes

Common finite difference schemes are

- ▶ Forward Euler (or Explicit Euler)

$$\frac{U_j^{m+1} - U_j^m}{\Delta t} = \frac{U_{j+1}^m - 2U_j^m + U_{j-1}^m}{\Delta x^2}$$

- ▶ Backward Euler (or Implicit Euler)

$$\frac{U_j^{m+1} - U_j^m}{\Delta t} = \frac{U_{j+1}^{m+1} - 2U_j^{m+1} + U_{j-1}^{m+1}}{\Delta x^2}$$

- ▶ θ -Method (Crank Nicolson when $\theta = 1/2$)

$$\begin{aligned} \frac{U_j^{m+1} - U_j^m}{\Delta t} = & \theta \frac{U_{j+1}^{m+1} - 2U_j^{m+1} + U_{j-1}^{m+1}}{\Delta x^2} \\ & + (1 - \theta) \frac{U_{j+1}^m - 2U_j^m + U_{j-1}^m}{\Delta x^2} \end{aligned}$$

Finite Difference Schemes

All these finite difference schemes hold for $j = 1, \dots, N - 1$ and $m = 0, 1, \dots$

We must also discretise the initial and boundary conditions as

$$\begin{aligned}U_j^0 &= u_0(x_j), \quad j = 0, 1, \dots, N \\U_0^m &= u_a(t_m), \quad m = 1, 2, \dots \\U_N^m &= u_b(t_m), \quad m = 1, 2, \dots\end{aligned}$$

For the θ -method for $\theta > 0$ we have to solve a linear system at each timestep of the form

$$(I - \mu\theta A)\mathbf{U}^{m+1} = (I' + \mu(1 - \theta)A)\mathbf{U}^m + \mathbf{g}^{m+1}.$$

Here, $\mu = \Delta t / \Delta x^2$, $\mathbf{U}^m = (U_0^m, U_1^m, \dots, U_N^m)^T$, I is the $(N + 1) \times (N + 1)$ identity matrix, I' is the $(N + 1) \times (N + 1)$ identity matrix but with the $(1, 1)$ and $(N + 1, N + 1)$ entries being zero, and $\mathbf{g}^{m+1} = (u_a(t_{m+1}), 0, \dots, 0, u_b(t_{m+1}))^T$.

2D Parabolic PDEs

2D Heat Equation

The heat equation in 2D is given by

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

for $t > 0$ and $x \in \Omega \subset \mathbb{R}^2$ with an initial condition

$$u(x, y, 0) = u_0(x, y) ,$$

for $x \in \Omega$. We consider Dirichlet boundary conditions

$$u(x, y, t) = u_D(x, y, t) \quad \text{for } (x, y) \in \partial\Omega , \quad t > 0.$$

The Mesh

We define a sequence of uniform timesteps by

$$t_m = m\Delta t$$

for $m = 0, 1, 2, \dots$ where $\Delta t > 0$ is the constant timestep size.

For the spatial mesh, we assume that the domain Ω is a rectangle, namely $\Omega = (a, b) \times (c, d)$ so that $x \in [a, b]$ and $y \in [c, d]$. We then define a set of uniform mesh points by

$$x_i = a + i\Delta x,$$

$$y_j = c + j\Delta y,$$

for $i = 0, 1, \dots, N_x$, $j = 0, 1, \dots, N_y$ and with the meshsizes $\Delta x = (b - a)/N_x$ and $\Delta y = (d - c)/N_y$.

We write $u(x_i, y_j, t_m) = u_{i,j}^m$ and seek to approximate $u_{i,j}^m$ by $U_{i,j}^m$ for $i = 0, 1, \dots, N_x$, $j = 0, 1, \dots, N_y$ and $m = 0, 1, 2, \dots$

Finite Difference Schemes

We can write down finite difference schemes in an analogous way to the 1D case. First define

$$\begin{aligned}\delta_x^2 U_{i,j} &= U_{i+1,j} - 2U_{i,j} + U_{i-1,j} , \\ \delta_y^2 U_{i,j} &= U_{i,j+1} - 2U_{i,j} + U_{i,j-1} .\end{aligned}$$

Then we may write

- ▶ Forward Euler (or Explicit Euler)

$$\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = \frac{\delta_x^2 U_{i,j}^m}{\Delta x^2} + \frac{\delta_y^2 U_{i,j}^m}{\Delta y^2}$$

- ▶ Backward Euler (or Implicit Euler)

$$\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = \frac{\delta_x^2 U_{i,j}^{m+1}}{\Delta x^2} + \frac{\delta_y^2 U_{i,j}^{m+1}}{\Delta y^2}$$

Finite Difference Schemes

- θ -Method (Crank Nicolson when $\theta = 1/2$)

$$\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = \frac{\theta \delta_x^2 U_{i,j}^{m+1} + (1 - \theta) \delta_x^2 U_{i,j}^m}{\Delta x^2} + \frac{\theta \delta_y^2 U_{i,j}^{m+1} + (1 - \theta) \delta_y^2 U_{i,j}^m}{\Delta y^2} \quad (1)$$

Finite Difference Schemes

All these finite difference schemes hold for $i = 1, \dots, N_x - 1$, $j = 1, \dots, N_y - 1$ and $m = 0, 1, \dots$

We must also discretise the initial and boundary conditions as

$$\begin{aligned}U_{i,j}^0 &= u_0(x_i, y_j), \quad i = 0, 1, \dots, N_x, \quad j = 0, 1, \dots, N_y \\U_{0,j}^m &= u_D(a, y, t_m), \quad j = 0, 1, \dots, N_y, \quad m = 1, 2, \dots \\U_{N_x,j}^m &= u_D(b, y, t_m), \quad j = 0, 1, \dots, N_y, \quad m = 1, 2, \dots \\U_{i,0}^m &= u_D(x, c, t_m), \quad i = 1, \dots, N_x - 1, \quad m = 1, 2, \dots \\U_{i,N_y}^m &= u_D(x, d, t_m), \quad i = 1, \dots, N_x - 1, \quad m = 1, 2, \dots\end{aligned}$$

Forward Euler Scheme

The forward Euler scheme is

$$\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = \frac{\delta_x^2 U_{i,j}^m}{\Delta x^2} + \frac{\delta_y^2 U_{i,j}^m}{\Delta y^2}$$

for $i = 1, \dots, N_x - 1$, $j = 1, \dots, N_y - 1$ and $m = 0, 1, \dots$. Writing $\mu_x = \Delta t / \Delta x^2$ and $\mu_y = \Delta t / \Delta y^2$, we may re-arrange the scheme to get

$$\begin{aligned} U_{i,j}^{m+1} = & U_{i,j}^m + \mu_x (U_{i+1,j}^m - 2U_{i,j}^m + U_{i-1,j}^m) \\ & + \mu_y (U_{i,j+1}^m - 2U_{i,j}^m + U_{i,j-1}^m) \end{aligned}$$

for $i = 1, \dots, N_x - 1$, $j = 1, \dots, N_y - 1$ and $m = 0, 1, \dots$

As in 1D, this is very simple to implement.

θ -Method

The θ -method is

$$\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = \frac{\theta \delta_x^2 U_{i,j}^{m+1} + (1 - \theta) \delta_x^2 U_{i,j}^m}{\Delta x^2} + \frac{\theta \delta_y^2 U_{i,j}^{m+1} + (1 - \theta) \delta_y^2 U_{i,j}^m}{\Delta y^2}.$$

(Recall this includes the backward Euler scheme if we take $\theta = 1$.)

We may re-arrange the scheme to get

$$\begin{aligned} -\mu_x \theta (U_{i+1,j}^{m+1} + U_{i-1,j}^{m+1}) - \mu_y \theta (U_{i,j+1}^{m+1} + U_{i,j-1}^{m+1}) + (1 + 2\theta(\mu_x + \mu_y)) U_{i,j}^{m+1} \\ = \mu_x (1 - \theta) (U_{i+1,j}^m + U_{i-1,j}^m) + \mu_y (1 - \theta) (U_{i,j+1}^m + U_{i,j-1}^m) \\ + (1 - 2(1 - \theta)(\mu_x + \mu_y)) U_{i,j}^m \end{aligned}$$

for $i = 1, \dots, N_x - 1$, $j = 1, \dots, N_y - 1$ and $m = 0, 1, \dots$

θ -Method — Linear System

In the case of homogeneous Dirichlet boundary conditions we have $U_{0,j}^{m+1} = U_{N_x,j}^{m+1} = U_{i,0}^{m+1} = U_{i,N_y}^{m+1} = 0$ and we may write the vector of unknowns as

$$\mathbf{U}^{m+1} = (U_{1,1}^{m+1}, U_{1,2}^{m+1}, \dots, U_{1,N_y-1}^{m+1}, U_{2,1}^{m+1} \dots U_{N_x-1,N_y-1}^{m+1})^T.$$

We may then write a linear system

$$(I - \theta A)\mathbf{U}^{m+1} = (I + (1 - \theta)A)\mathbf{U}^m,$$

where A is a matrix with $(N_x - 1)(N_y - 1)$ rows and columns and I is the identity matrix of the same size.

θ -Method — Linear System

The structure of A is

$$A = \left(\begin{array}{cccc} B & C & & \\ C & B & C & \\ & \ddots & \ddots & \ddots \\ & & C & B & C \\ & & & C & B \end{array} \right) \Bigg\} N_x - 1 \text{ blocks}$$

where $B, C \in \mathbb{R}^{(N_y-1) \times (N_y-1)}$ are given by

$$B = \begin{pmatrix} -2(\mu_x + \mu_y) & \mu_y & & \\ \mu_y & -2(\mu_x + \mu_y) & \mu_y & \\ \ddots & \ddots & \ddots & \\ & \mu_y & -2(\mu_x + \mu_y) & \end{pmatrix},$$

and $C = \mu_x I_{N_y-1}$ with I_{N_y-1} being the identity matrix of size $N_y - 1$.

Truncation Error

The truncation error for the θ -method is given by

$$T_{i,j}^m = \frac{u_{i,j}^{m+1} - u_{i,j}^m}{\Delta t} - \frac{\theta \delta_x^2 u_{i,j}^{m+1} + (1-\theta) \delta_x^2 u_{i,j}^m}{\Delta x^2} - \frac{\theta \delta_y^2 u_{i,j}^{m+1} + (1-\theta) \delta_y^2 u_{i,j}^m}{\Delta y^2}.$$

It is standard to perform Taylor series approximations about the point $(x_i, y_j, t_{m+1/2})$. This gives

$$T_{i,j}^m = \left(\frac{1}{2} - \theta\right) \Delta t u_{tt} - \frac{1}{12} (\Delta t^2 u_{ttt} + \Delta x^2 u_{xxxx} + \Delta y^2 u_{yyyy}).$$

Thus for θ independent of Δt and Δx :

- ▶ in general, the θ -method is first order in Δt and second order in Δx and Δy ;
- ▶ for the particular case $\theta = 1/2$, the Crank Nicolson method is second order in Δt , Δx and Δy .

Stability

Stability can be assessed by inserting the Fourier mode $U_{i,j}^m = [\lambda(k_x, k_y)]^m e^{i(k_x x_i + k_y y_j)}$ into the numerical scheme. The scheme is then practically stable if $|\lambda(k_x, k_y)| \leq 1$. Substituting such a Fourier mode into the θ -method (1) and simplifying gives

$$\lambda(k_x, k_y) = \frac{1 - 4(1 - \theta)(\mu_x \sin^2(k_x \Delta x/2) + \mu_y \sin^2(k_y \Delta y/2))}{1 + 4\theta(\mu_x \sin^2(k_x \Delta x/2) + \mu_y \sin^2(k_y \Delta y/2))}$$

for $k_x \in [-\pi/\Delta x, \pi/\Delta x]$ and $k_y \in [-\pi/\Delta y, \pi/\Delta y]$ and where $\mu_x = \Delta t/\Delta x^2$ and $\mu_y = \Delta t/\Delta y^2$.

Clearly this satisfies $\lambda(k_x, k_y) \leq 1$ for all k_x and k_y . For $\lambda(k_x, k_y) \geq -1$ we require

$$2(\mu_x \sin^2(k_x \Delta x/2) + \mu_y \sin^2(k_y \Delta y/2))(1 - 2\theta) \leq 1.$$

This is clearly true for all $\theta \geq 1/2$, but for $\theta < 1/2$ this gives a restriction on Δt .

Stability

Thus for the θ -method we have

- ▶ If $\theta \geq 1/2$ the method is unconditionally stable. In particular this means that the backward Euler and Crank-Nicolson schemes are unconditionally stable.
- ▶ If $\theta < 1/2$ the method is only conditionally stable. The values of Δt , Δx and Δy must be chosen so that

$$\Delta t \leq \frac{\Delta x^2 \Delta y^2}{\Delta x^2 + \Delta y^2} \frac{1}{2(1 - 2\theta)} .$$

In particular this means that the forward Euler method is only conditionally stable and, in the case where $\Delta x = \Delta y$, the condition for stability is that $\Delta t \leq \Delta x^2/4$.

ADI Method

Consider the Crank Nicolson scheme for the 2D heat equation:

$$\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = \frac{1}{2} \frac{\delta_x^2 U_{i,j}^{m+1} + \delta_x^2 U_{i,j}^m}{\Delta x^2} + \frac{1}{2} \frac{\delta_y^2 U_{i,j}^{m+1} + \delta_y^2 U_{i,j}^m}{\Delta y^2},$$

or equivalently

$$\left(1 - \frac{1}{2}\mu_x\delta_x^2 - \frac{1}{2}\mu_y\delta_y^2\right) U_{i,j}^{m+1} = \left(1 + \frac{1}{2}\mu_x\delta_x^2 + \frac{1}{2}\mu_y\delta_y^2\right) U_{i,j}^m.$$

ADI schemes are based on approximately factorising the operators on the left and right of this equation.

ADI Method

We write this approximation as

$$\left(1 - \frac{1}{2}\mu_x\delta_x^2\right)\left(1 - \frac{1}{2}\mu_y\delta_y^2\right)U_{i,j}^{m+1} = \left(1 + \frac{1}{2}\mu_x\delta_x^2\right)\left(1 + \frac{1}{2}\mu_y\delta_y^2\right)U_{i,j}^m.$$

By introducing an intermediate time level $U^{m+1/2}$ we may write this in an equivalent form

$$\begin{aligned}\left(1 - \frac{1}{2}\mu_x\delta_x^2\right)U_{i,j}^{m+1/2} &= \left(1 + \frac{1}{2}\mu_y\delta_y^2\right)U_{i,j}^m, \\ \left(1 - \frac{1}{2}\mu_y\delta_y^2\right)U_{i,j}^{m+1} &= \left(1 + \frac{1}{2}\mu_x\delta_x^2\right)U_{i,j}^{m+1/2}.\end{aligned}$$

The advantage of doing this is that, instead of one large system of equations, we have many smaller tridiagonal systems.

ADI Method: Truncation Error

It can be shown that the truncation error for the ADI method is

$$T_{i,j}^m = -\frac{1}{12} (\Delta t^2 u_{ttt} + \Delta x^2 u_{xxxx} + \Delta y^2 u_{yyyy}) + \frac{1}{4} \Delta t^2 u_{xxyyt}$$

(i.e. the terms of the truncation error for Crank Nicolson with one extra term added coming from the fact that the approximation of Crank Nicolson is inexact).

ADI Method: Stability

Inserting the Fourier mode $U_{i,j}^m = [\lambda(k_x, k_y)]^m e^{i(k_x x_i + k_y y_j)}$ into the numerical scheme gives

$$\lambda(k_x, k_y) = \frac{(1 - 2\mu_x \sigma_x^2)(1 - 2\mu_y \sigma_y^2)}{(1 + 2\mu_x \sigma_x^2)(1 + 2\mu_y \sigma_y^2)},$$

where

$$\begin{aligned}\sigma_x^2 &= \sin^2\left(\frac{k_x \Delta x}{2}\right), \\ \sigma_y^2 &= \sin^2\left(\frac{k_y \Delta y}{2}\right).\end{aligned}$$

It is easy to see that $|\lambda(k_x, k_y)| \leq 1$ for all values of μ_x and μ_y so that the scheme is unconditionally stable.

Example

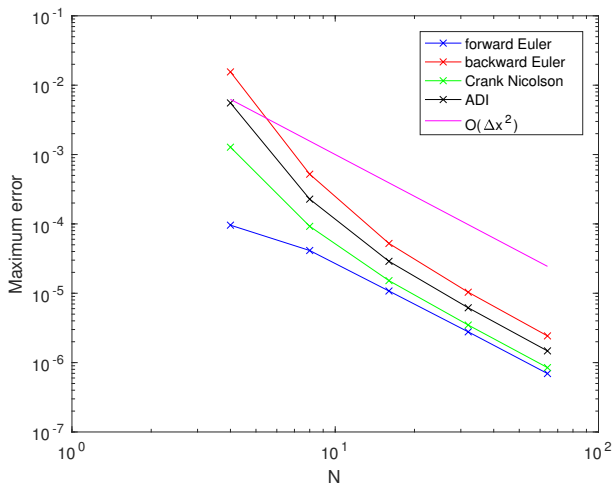
Solve the heat equation $u_t = u_{xx} + u_{yy}$ in the unit square $[0, 1]^2$ with homogeneous Dirichlet boundary conditions and initial condition

$$u(x, y, 0) = \sin(\pi x) \sin(3\pi y) .$$

The exact solution is

$$u(x, y, t) = e^{-10\pi^2 t} \sin(\pi x) \sin(3\pi y) .$$

Results with $\Delta x^2 = \Delta y^2$ and $\Delta t = \Delta x^2/4$



Solution $\Delta x \neq \Delta y$

