Parabolic PDEs: Finite Difference Methods

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

Michaelmas Term 2025, 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 $\theta = 1/2$)

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

Finite Difference Schemes

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

We must also discretise the initial and boundary conditions as

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

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

$$
(I - \mu\theta A_2)\mathbf{U}^{m+1} = (I + \mu(1-\theta)A_1)\mathbf{U}^m + \mu\theta\mathbf{g_2}^{m+1} + \mu(1-\theta)\mathbf{g_2}^m .
$$

Here, $\mu = \Delta t/\Delta x^2$, $\mathbf{U}^m = (U_1^m, U_2^m, \ldots, U_{N-1}^m)^T$, I is the $(N-1) \times (N-1)$ identity matrix, $A_2 =$ triadiag$(1, -2, 1)$, and $\mathbf{g_2}^m = (u_a(t_m), 0, \ldots, 0, u_b(t_m))^T$.

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 $m = 0, 1, 2, \ldots$ 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 $i = 0, 1, \ldots, N_x$, $j = 0, 1, \ldots, N_y$ 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 $i = 0, 1, \ldots, N_x$, $j = 0, 1, \ldots, N_y$ and $m = 0, 1, 2, \ldots$.

Finite Difference Schemes

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

$$
\begin{aligned}
D_x^+ D_x^- U_{i,j} &= \frac{U_{i+1,j} - 2U_{i,j} + U_{i-1,j}}{\Delta x^2} , \\
D_y^+ D_y^- U_{i,j} &= \frac{U_{i,j+1} - 2U_{i,j} + U_{i,j-1}}{\Delta y^2} .
\end{aligned}
$$

Then we may write

▶ Forward Euler (or Explicit Euler)

$$
\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = D_x^+ D_x^- U_{i,j}^m + D_y^+ D_y^- U_{i,j}^m
$$

▶ Backward Euler (or Implicit Euler)

$$
\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = D_x^+ D_x^- U_{i,j}^{m+1} + D_y^+ D_y^- U_{i,j}^{m+1}
$$

Finite Difference Schemes

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

$$
\begin{aligned}
\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} &= D_x^+ D_x^- \left(\theta U_{i,j}^{m+1} + (1 - \theta) U_{i,j}^m \right) \\
&\quad + D_y^+ D_y^- \left(\theta U_{i,j}^{m+1} + (1 - \theta) U_{i,j}^m \right) \quad (1)
\end{aligned}
$$

Finite Difference Schemes

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

We must also discretise the initial and boundary conditions as

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

Forward Euler Scheme

The forward Euler scheme is

$$\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} \;=\; D_x^+ D_x^- U_{i,j}^m + D_y^+ D_y^- U_{i,j}^m$$

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

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 $i = 1, \ldots, N_x - 1$, $j = 1, \ldots, N_y - 1$ and $m = 0, 1, \ldots$.

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

θ-Method

The θ-method is

$$\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = D_x^+ D_x^- \left(\theta U_{i,j}^{m+1} + (1-\theta)U_{i,j}^m\right)$$
$$+ D_y^+ D_y^- \left(\theta U_{i,j}^{m+1} + (1-\theta)U_{i,j}^m\right) .$$

We may re-arrange the scheme to get

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

for $i = 1, \ldots, N_x - 1$, $j = 1, \ldots, Ny - 1$ and $m = 0, 1, \ldots$.

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

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

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

$$A = \left.\begin{pmatrix} B & C & & & \\ C & B & C & & \\ & \ddots & \ddots & \ddots & \\ & & C & B & C \\ & & & C & B \end{pmatrix}\right\} N_x - 1 \text{ blocks}$$

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

$$
\begin{aligned}
T_{i,j}^m &= \frac{u_{i,j}^{m+1} - u_{i,j}^m}{\Delta t} - D_x^+ D_x^- \left(\theta u_{i,j}^{m+1} + (1-\theta) u_{i,j}^m \right) \\
&\quad - D_y^+ D_y^- \left(\theta u_{i,j}^{m+1} + (1-\theta) u_{i,j}^m \right) .
\end{aligned}
$$

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, Δx, and Δy:

- 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:

$$\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} = \frac{1}{2} D_x^+ D_x^- \left(U_{i,j}^{m+1} + U_{i,j}^m \right) + \frac{1}{2} D_y^+ D_y^- \left(U_{i,j}^{m+1} + U_{i,j}^m \right)$$

or equivalently

$$\left(1 - \frac{\Delta t}{2} D_x^+ D_x^- - \frac{\Delta t}{2} D_y^+ D_y^- \right) U_{i,j}^{m+1}$$
$$= \left(1 + \frac{\Delta t}{2} D_x^+ D_x^- + \frac{\Delta t}{2} D_y^+ D_y^- \right) U_{i,j}^m .$$

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

ADI Method

We write this approximation as

$$
\left(1 - \frac{\Delta t}{2} D_x^+ D_x^-\right)\left(1 - \frac{\Delta t}{2} D_y^+ D_y^-\right) U_{i,j}^{m+1}
$$
$$
= \left(1 + \frac{\Delta t}{2} D_x^+ D_x^-\right)\left(1 + \frac{\Delta t}{2} D_y^+ D_y^-\right) U_{i,j}^m .
$$

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

$$
\left(1 - \frac{\Delta t}{2} D_x^+ D_x^-\right) U_{i,j}^{m+1/2} = \left(1 + \frac{\Delta t}{2} D_y^+ D_y^-\right) U_{i,j}^m ,
$$
$$
\left(1 - \frac{\Delta t}{2} D_y^+ D_y^-\right) U_{i,j}^{m+1} = \left(1 + \frac{\Delta t}{2} D_x^+ D_x^-\right) U_{i,j}^{m+1/2} .
$$

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

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

(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

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

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

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

Results with $\Delta x^2 = \Delta y^2$ and $\Delta t = \Delta x^2/4$
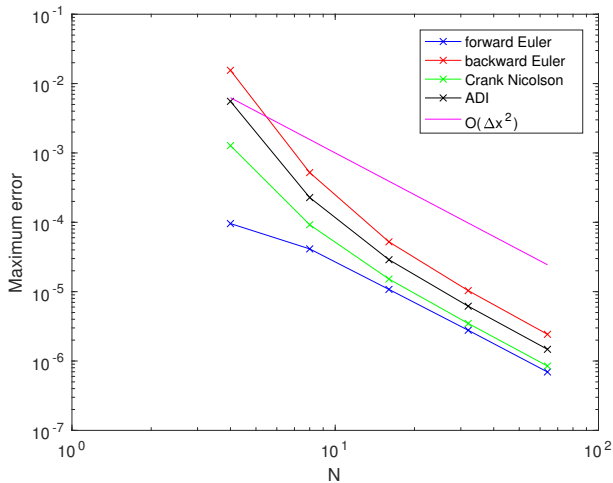
Solution $\Delta x \neq \Delta y$
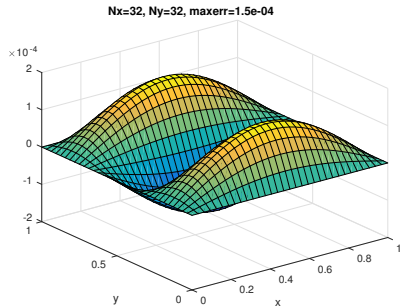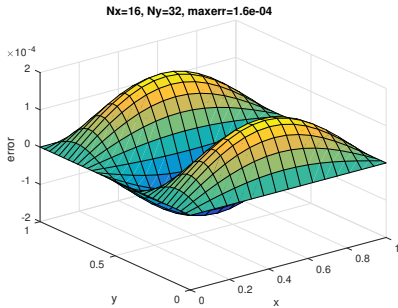
Coupled Problems

Can use the types of methods already discussed to solve coupled systems of PDEs.

Recall that for the heat equation with homogeneous Dirichlet boundary conditions, we can write the θ-method in matrix form as

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

Now suppose we want to solve a coupled system of the form

$$\frac{\partial u}{\partial t} = \nabla^2 u + \alpha v$$
$$\frac{\partial v}{\partial t} = \nabla^2 v + \beta u$$

for $t > 0$, and $x \in \Omega \subset \mathbb{R}^2$, with homogeneous Dirichlet boundary conditions on both u and v, and initial conditions

$$u(x, y, 0) = u_0(x, y), \qquad v(x, y, 0) = v_0(x, y)$$

for $x \in \Omega$.

Coupled Problems

Using the same mesh and timestep as before, we can write a θ-method for the u equation as

$$
\begin{aligned}
\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} \;=\; & D_x^+ D_x^- \left(\theta U_{i,j}^{m+1} + (1-\theta)U_{i,j}^m \right) \\
& + D_y^+ D_y^- \left(\theta U_{i,j}^{m+1} + (1-\theta)U_{i,j}^m \right) \\
& + \theta \alpha V_{i,j}^{m+1} + (1-\theta)\alpha V_{i,j}^m,
\end{aligned}
$$

which can be written in matrix form as

$$
\begin{aligned}
(I - \theta A)\mathbf{U}^{m+1} - \theta \alpha \Delta t \mathbf{V}^{m+1} \;=\; & (I + (1-\theta)A)\mathbf{U}^m \\
& + (1-\theta)\alpha \Delta t \mathbf{V}^m.
\end{aligned}
$$

Coupled Problems

Writing a similar finite difference equation for v also leads to a matrix form

$$
\begin{aligned}
(I - \theta A)\mathbf{V}^{m+1} - \theta\beta\Delta t\mathbf{U}^{m+1} &= (I + (1-\theta)A)\mathbf{V}^m \\
&\quad + (1-\theta)\beta\Delta t\mathbf{U}^m.
\end{aligned}
$$

This can be written as a big matrix system

$$
\begin{pmatrix} I - \theta A & -\theta\alpha\Delta tI \\ -\theta\beta\Delta tI & I - \theta A \end{pmatrix} \begin{pmatrix} \mathbf{U}^{m+1} \\ \mathbf{V}^{m+1} \end{pmatrix}
$$

$$
= \begin{pmatrix} I + (1-\theta)A & (1-\theta)\alpha\Delta tI \\ (1-\theta)\beta\Delta tI & I + (1-\theta)A \end{pmatrix} \begin{pmatrix} \mathbf{U}^m \\ \mathbf{V}^m \end{pmatrix}.
$$

Nonlinear Problems

We can also extend these ideas to nonlinear problems. Consider a problem of the form

$$\frac{\partial u}{\partial t} = \nabla^2 u + f(u),$$

for $t > 0$, and $x \in \Omega \subset \mathbb{R}^2$, with homogeneous Dirichlet boundary conditions, and initial condition

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

for $x \in \Omega$.

Nonlinear Problems

We can write a finite difference scheme of the form

$$
\begin{aligned}
\frac{U_{i,j}^{m+1} - U_{i,j}^m}{\Delta t} &= D_x^+ D_x^- \left(\theta U_{i,j}^{m+1} + (1-\theta) U_{i,j}^m \right) \\
&\quad + D_y^+ D_y^- \left(\theta U_{i,j}^{m+1} + (1-\theta) U_{i,j}^m \right) \\
&\quad + \theta f(U_{i,j}^{m+1}) + (1-\theta) f(U_{i,j}^m),
\end{aligned}
$$

along with the usual initial and boundary conditions.

The drawback to this is that, unless the function f is linear, we now have to solve a very large nonlinear system at each timestep. This nonlinear system takes the form

$$
\begin{aligned}
(I - \theta A)\mathbf{U}^{m+1} - \theta \Delta t f(\mathbf{U}^{m+1}) &= (I + (1-\theta)A)\mathbf{U}^m \\
&\quad + (1-\theta)\Delta t f(\mathbf{U}^m).
\end{aligned}
$$

Nonlinear Problems

An alternative is to treat the linear terms implicitly and the nonlinear terms explicitly so that the finite difference scheme becomes, in matrix form,

$$(I - A)\mathbf{U}^{m+1} = \mathbf{U}^m + \Delta t f(\mathbf{U}^m).$$

This has the advantage of only requiring a linear solve at each timestep. The approach often works well in practice and it is possible to use a larger timestep size than the simple explicit Euler scheme would have required.

Coupled Nonlinear Example

We consider the Cahn-Hilliard equation which was originally proposed to model phase separation in binary alloys. This is a 4th order problem but can be written as a system of two 2nd order equations

$$
\frac{\partial c}{\partial t} - \nabla^2 w = 0
$$

$$
w - \frac{1}{\epsilon}\Phi'(c) + \epsilon\nabla^2 c = 0
$$

with homogeneous Neumann boundary conditions for both c and w. Usually Φ is a double well potential, e.g. $\Phi(c) = (1 - c^2)^2/4$.

Here c has steady state ± 1 corresponding to pure phase A and pure phase B. In addition, ϵ represents the thickness of the interface between areas where $c = 1$ and areas where $c = -1$.

Coupled Nonlinear Example

If we let A be the matrix representing the Laplacian operator with Neumann boundary conditions (so a slightly different matrix to earlier) then we can use the method of lines to write

$$\frac{d\mathbf{C}}{dt} - A\mathbf{W} = 0$$

$$\mathbf{W} - \frac{1}{\epsilon}\Phi'(\mathbf{C}) + \epsilon A\mathbf{C} = 0.$$

Using an implicit scheme for the linear terms and an explicit scheme for the nonlinear terms, we must solve

$$\frac{\mathbf{C}^{m+1} - \mathbf{C}^m}{\Delta t} - A\mathbf{W}^{m+1} = 0$$

$$\mathbf{W}^{m+1} - \frac{1}{\epsilon}\Phi'(\mathbf{C}^m) + \epsilon A\mathbf{C}^{m+1} = 0,$$

or, as a system we can write this as

$$\begin{pmatrix} I & -\Delta t A \\ \epsilon A & I \end{pmatrix} \begin{pmatrix} \mathbf{C}^{m+1} \\ \mathbf{W}^{m+1} \end{pmatrix} = \begin{pmatrix} \mathbf{C}^m \\ \Phi'(\mathbf{C}^m)/\epsilon \end{pmatrix}.$$

Coupled Nonlinear Example

We can take an initial condition where $c = 1$ in a cross in the centre of the domain and $c = -1$ outside this region.
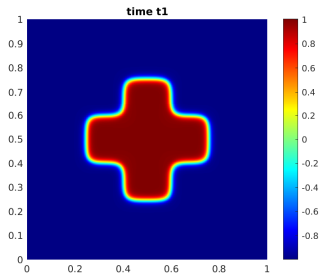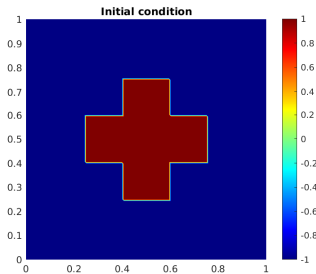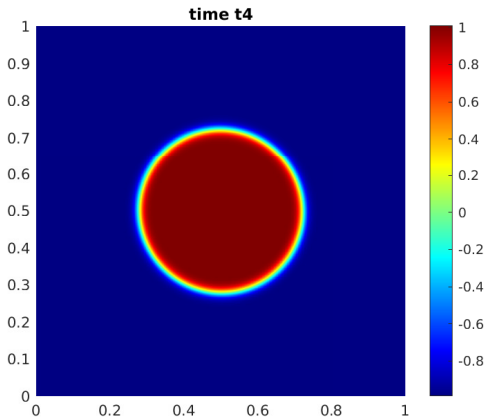
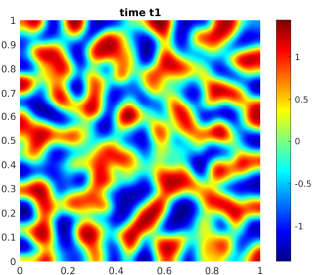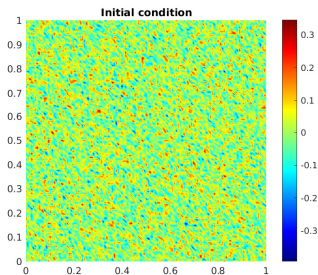# Coupled Nonlinear Example

The edges of the cross smooth out.
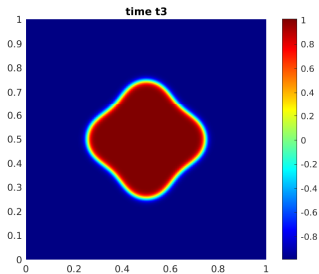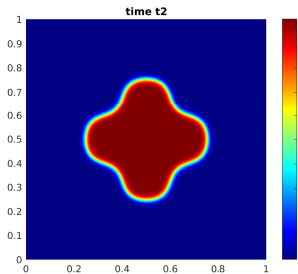
Coupled Nonlinear Example

The steady state has an interface in the shape of a circle.

Coupled Nonlinear Example

Alternatively we can take a random initial condition. At each grid point we set c to be a number drawn from a normal distribution with mean zero and variance one, then scaled by 0.1.
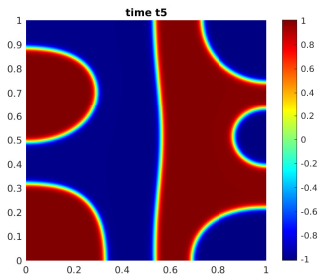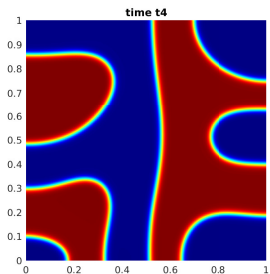
Coupled Nonlinear Example

The solution has patches where it is 1 and patches where it is -1
and the boundaries of these regions are preferentially straight edges
or circles.
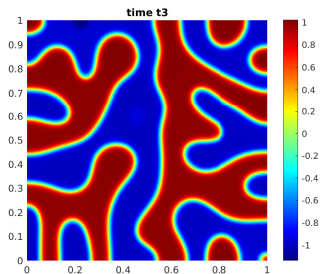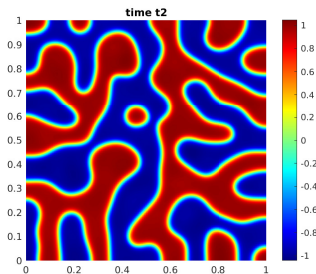
Coupled Nonlinear Example

The solution has patches where it is 1 and patches where it is -1 and the boundaries of these regions are preferentially straight edges or circles.

Coupled Nonlinear Example

The steady state solution (for this initial data) is -1 in the left half of the domain and 1 in the right half of the domain with an interface of width $\mathcal{O}(\epsilon)$.