



Technische
Universität
Braunschweig



Introduction to PDEs and Numerical Methods

Lecture 7.

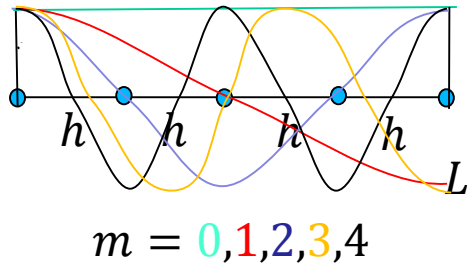
Finite difference methods – stability, consistency, convergence

Dr. Noemi Friedman, 30.11.2016.

Von Neumann Stability Analysis

Stability checking with Von Neumann Stability Analysis (Fourier stability analysis)

Let's suppose our solution has the form of:



$$u(t, x) = \sum_{m=0}^{\infty} A_m(t) e^{ik_m x} \quad (\text{Fourier-expansion})$$

$$e^{ik_m x} = \cos(k_m x) + i \sin(k_m x)$$

With the wave number: $k_m = \frac{m\pi}{L} \quad m = 0..M \quad M = \frac{L}{h}$
 (Shannon's theorem)

Let's suppose that the solution in time changes exponentially

$$A_m(t) = e^{\alpha_m t} \quad \text{where} \quad \alpha_m: \text{constant}$$

The solution takes the form after discretisation: $t = n\Delta t, \quad x = jh$

$$u(n, j) = \sum_{m=0}^M G(k_m)^n e^{ik_m jh}$$

$$G(k_m)^n = A_m(t) = e^{\alpha_m n \Delta t} = (e^{\alpha_m \Delta t})^n \quad \text{gain factor/ amplifier}$$

Von Neumann Stability Analysis – stability of Euler forward

in simpler form

$$u(n, j) = \sum_{k=0}^M G(k)^n e^{ikjh}$$

$u_{n,j}(k) = G(k)^n e^{ikjh}$ for one frequency

Example: let's check the stability of the following scheme for the instationary heat equation:

$$u_{n+1,j} - u_{n,j} = \Delta t \frac{\beta^2}{h^2} (u_{n,j-1} - 2u_{n,j} + u_{n,j+1}) \quad (\text{Euler forward, three point spatial discr.})$$

$$G(k)^{n+1} e^{ikjh} - G(k)^n e^{ikjh} = \Delta t \frac{\beta^2}{h^2} (G(k)^n e^{ik(j-1)h} - 2G(k)^n e^{ikjh} + G(k)^n e^{ik(j+1)h})$$

/ $G(k)^n$

$$G(k) e^{ikjh} - e^{ikjh} = \Delta t \frac{\beta^2}{h^2} (e^{ik(j-1)h} - 2e^{ikjh} + e^{ik(j+1)h})$$

/ e^{ikjh}

$$G(k) - 1 = \Delta t \frac{\beta^2}{h^2} (e^{-ikh} - 2 + e^{ikh})$$

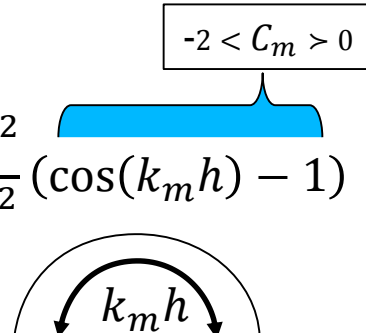
$$e^{ikh} + e^{-ikh} = 2\cos(kh)$$

$$G(k) = 1 + 2\Delta t \frac{\beta^2}{h^2} (\cos(kh) - 1)$$

Von Neumann Stability Analysis – stability of Euler forward

The gain factor:
$$G(k) = 1 + 2\Delta t \frac{\beta^2}{h^2} (\cos(kh) - 1)$$

In a more precise form:

$$G(k_m) = 1 + 2\Delta t \frac{\beta^2}{h^2} (\cos(k_m h) - 1)$$


$$k_m = \frac{m\pi}{L}$$

$$m = 0..M$$

$$M = \frac{L}{h}$$

Stability requirement: $|G(k_m)| \leq 1$

$\max(G(k_m))$: lowest frequency $m = 0 \longrightarrow G(k_0) = 1 + 2\Delta t \frac{\beta^2}{h^2} (1 - 1) = 1$

$\min(G(k_m))$: highest frequency $m = M \longrightarrow G(k_M) = 1 + 2\Delta t \frac{\beta^2}{h^2} (-1 - 1) = 1 - 4\Delta t \frac{\beta^2}{h^2}$

Von Neumann Stability Analysis – stability of Euler forward

Stability requirement: $G(k_M) = 1 - 4\Delta t \frac{\beta^2}{h^2} \geq -1$

$$4\Delta t \frac{\beta^2}{h^2} \leq 2$$

$$\Delta t \leq \frac{h^2}{2\beta^2}$$

Scheme for the heat equation is only stable if this condition is satisfied. (conditionally stable)

If the gain factor is positive, the solution will not oscillate in time:

$$G(k_m) \geq 0$$

$$G(k_M) = 1 - 4\Delta t \frac{\beta^2}{h^2} \geq 0$$

$$\frac{h^2}{4\beta^2} \geq \Delta t$$

Solution will give oscillatory solution if this condition is not satisfied.

Von Neumann Stability Analysis – stability of Euler backward

Example 2: let's check the stability of the following scheme for the instationary heat equation:

$$\frac{u_{j,n+1} - u_{j,n}}{\Delta t} = \theta \frac{\beta^2}{h^2} (u_{j-1,n+1} - 2u_{j,n+1} + u_{j+1,n+1}) + \quad (\text{Theta method})$$

$$+ (1 - \theta) \frac{\beta^2}{h^2} (u_{j-1,n} - 2u_{j,n} + u_{j+1,n}) + O(h^2) + O(\Delta t^p)$$

$$G(k)^{n+1} e^{ikjh} - G(k)^n e^{ikjh} = \Delta t \theta \frac{\beta^2}{h^2} (G(k)^{n+1} e^{ik(j-1)h} - 2G(k)^{n+1} e^{ikjh} + G(k)^{n+1} e^{ik(j+1)h})$$

$$+ \Delta t (1 - \theta) \frac{\beta^2}{h^2} (G(k)^n e^{ik(j-1)h} - 2G(k)^n e^{ikjh} + G(k)^n e^{ik(j+1)h}) \quad /G(k)^n$$

$$G(k) e^{ikjh} - e^{ikjh} =$$

$$= \Delta t \theta \frac{\beta^2}{h^2} G(k) (e^{ik(j-1)h} - 2e^{ikjh} + e^{ik(j+1)h}) + \Delta t (1 - \theta) \frac{\beta^2}{h^2} (e^{ik(j-1)h} - 2e^{ikjh} + e^{ik(j+1)h}) \quad /e^{ikjh}$$

$$G(k) - 1 = \Delta t \theta \frac{\beta^2}{h^2} G(k) (e^{-ikh} - 2 + e^{ikh}) + \Delta t (1 - \theta) \frac{\beta^2}{h^2} (e^{-ikh} - 2 + e^{ikh})$$

Von Neumann Stability Analysis – stability of Euler backward

Example 2: let's check the stability of the following scheme for the instationary heat equation:

$$G(k) - 1 = \Delta t \theta \frac{\beta^2}{h^2} G(k) (e^{-ikh} - 2 + e^{ikh}) + (1 - \theta) \frac{\Delta t \beta^2}{h^2} (e^{-ikh} - 2 + e^{ikh})$$

$$G(k) - \underbrace{\theta \Delta t \frac{\beta^2}{h^2}}_{:= r} G(k) (e^{-ikh} - 2 + e^{ikh}) = (1 - \theta) \underbrace{\frac{\Delta t \beta^2}{h^2}}_{:= r} (e^{-ikh} - 2 + e^{ikh}) + 1$$

$$e^{ikh} + e^{-ikh} = 2\cos(kh)$$

$$G(k) - \theta r G(k) (2\cos(kh) - 2) = (1 - \theta) r (2\cos(kh) - 2) + 1$$

$$G(k) (1 + 2\theta r - 2\theta r \cos(kh)) = 2(1 - \theta) r (\cos(kh) - 1) + 1$$

$$G(k) = \frac{2(1 - \theta) r (\cos(kh) - 1) + 1}{(1 + 2\theta r - 2\theta r \cos(kh))}$$

Von Neumann Stability Analysis – stability of Euler backward

The gain factor: $G(k_m) = \frac{2(1 - \theta)r(\cos(k_m h) - 1) + 1}{(1 + 2\theta r - 2\theta r \cos(k_m h))}$

Lowest frequency: $k_0 = 0$

$$G(0) = \frac{2(1 - \theta)r(1 - 1) + 1}{(1 + 2\theta r - 2\theta r)} = 1$$

Highest frequency: $k_M = M \frac{\pi}{L} = \frac{L\pi}{Lh}$

$$G(k_M) = \frac{2(1 - \theta)r(-1 - 1) + 1}{(1 + 2\theta r + 2\theta r)} = \frac{1 - 4(1 - \theta)r}{1 + 4\theta r}$$

Stability requirement: $|G(k_M)| \leq 1$

$$\frac{1 - 4(1 - \theta)r}{1 + 4\theta r} \geq -1 \quad \Rightarrow \quad (1 - 2\theta)r \leq 1/2$$

$\theta \geq 1/2$
Unconditionally stable

$\theta < 1/2$

Restriction on the timestep:

$$r < \frac{1}{2(1 - 2\theta)}$$

Von Neumann Stability Analysis – stability of Euler backward check positivity

Requirement to get **non-oscillatory solution**: $G(k_M) \geq 0$ $\frac{1 - 4(1 - \theta)r}{1 + 4\theta r} \geq 0 \implies (1 - 2\theta)r \leq 1/2$

Restriction on the timestep:

$$r < \frac{1}{4(1 - 2\theta)}$$

Requirement to get **positive solution**:

$$\frac{u_{j,n+1} - u_{j,n}}{\Delta t} = \theta \frac{\beta^2}{h^2} (u_{j-1,n+1} - 2u_{j,n+1} + u_{j+1,n+1}) + (1 - \theta) \frac{\beta^2}{h^2} (u_{j-1,n} - 2u_{j,n} + u_{j+1,n}) + O(h^2) + O(\Delta t^p)$$

$$u_{j,n+1} = \theta \Delta t \frac{\beta^2}{h^2} (u_{j-1,n+1} - 2u_{j,n+1} + u_{j+1,n+1}) + (1 - \theta) \Delta t \frac{\beta^2}{h^2} (u_{j-1,n} - 2u_{j,n} + u_{j+1,n}) + u_{j,n} + O(h^2) + O(\Delta t^p)$$

$$r < \frac{1}{2(1 - 2\theta)}$$

Stability, consistency, convergence – clear up definitions

Important definitions:

Well-posedness (in the sense of Hadamard)

- solution exists
- the solution is unique
- continuous dependence on the initial data

- e.g.: heat equation, Laplace-equation

Ill-posed problems

That are not well-posed in the sense of Hadamard

e.g.: inverse problems, like the inverse of the heat equation

Stability, consistency, convergence - introduction

Well-posedness differently: $\mathcal{L}x = y \quad \mathcal{L}: X \rightarrow Y$

Surjective

$\mathcal{L}: X \rightarrow Y$ is surjective, if every element y in Y has a corresponding element x in X such that $f(x) = y$. The function f may map more than one element of X to the same element of Y . (For all $y \in Y$ I can find a solution in X)

$$\forall y \in Y, \exists x \in X \quad \mathcal{L}x = y$$

+

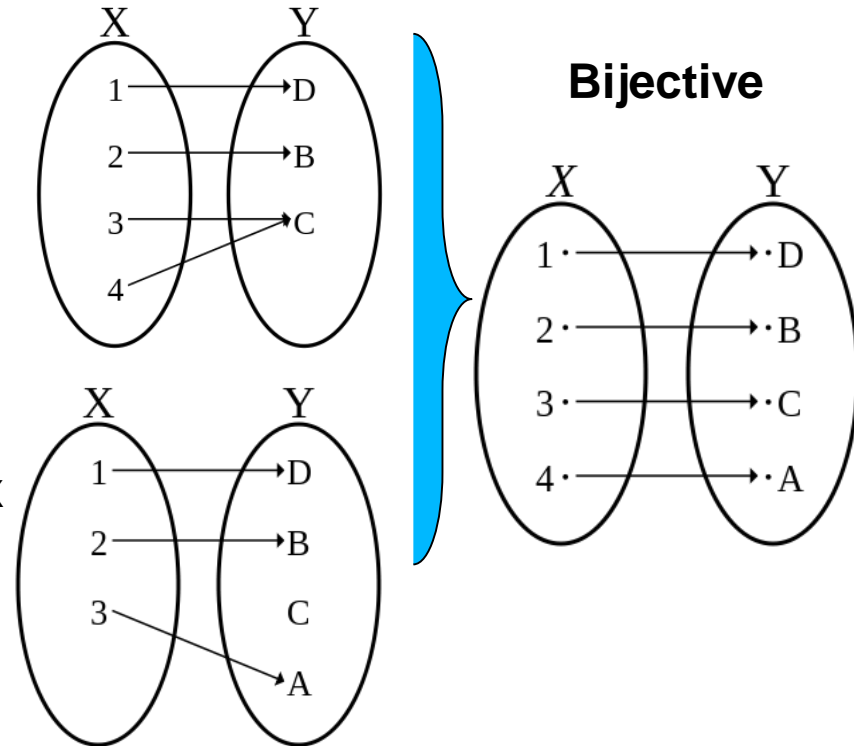
The function ~~$g: \mathbf{R} \rightarrow \mathbf{R}$~~ defined by $g(x) = x^2$
 The function $g: \mathbf{R} \rightarrow \mathbf{R}^+$ defined by $g(x) = x^2$

Injective (one-to-one mapping)

every element of Y is the image of at most one element of X

+

$$\mathcal{L}x_1 = \mathcal{L}x_2 \implies x_1 = x_2$$



Bijective

Continuous dependence on the inital data

The inverse/solution operator is uniformly bounded

$$\|\mathcal{L}^{-1}\| < C \implies \|A^{-1}y\| < C \|y\|$$

Norm of an operator(example): $Ax = b \quad \|A\| = \sup_{\substack{x \in X \\ x \neq 0}} \frac{\|Ax\|}{\|x\|}$

Stability, consistency, convergence - introduction

$$\mathcal{L}u = \dot{u} + u_{xx} = f \quad \mathcal{L}_h u_h = f_h \quad (\text{discretized in space})$$

Numerical stability

Even if an operator is well-posed in the sense of Hadamard, it may suffer from **numerical instability** when solved with finite precision, or with errors in the data.

$$\|\mathcal{L}^{-1}\| \leq C \quad \text{but} \quad \|\mathcal{L}_h^{-1}\| \leq C$$


A method is **numerically instable** if the round-off or truncation **errors** can be **amplified**, causing the error to grow exponentially

Ill-conditioned

A well-posed operator may be **ill-conditioned**, that is a small error in the initial data can result in much larger errors in the answers.

(indicated by a large condition number)

Stability, consistency, convergence - introduction

Consistency

A certain finite difference method is consistent if:

$$\lim_{\Delta t, h \rightarrow 0} \|\mathcal{L}(u) - \mathcal{L}_{\Delta t, h}(u)\| = 0 \quad (\text{method approximates the differential equation})$$

where $\mathcal{L}(u)$: original operator

$\mathcal{L}_{\Delta t, h}(u)$: approximated operator (discretized)

For example:

$$\mathcal{L}(u) = u'$$

from the Taylor expansion $u' = \frac{u(x+h) - u(x)}{h} + O(h)$

$$\mathcal{L}_h(u) = \frac{u(x+h) - u(x)}{h} \quad (\text{first order method})$$

$$\left\| u' - \frac{u(x+h) - u(x)}{h} \right\| \leq Ch \quad \Rightarrow \quad \lim_{h \rightarrow 0} \|\mathcal{L}(u) - \mathcal{L}_h(u)\| = 0$$

Stability, consistency, convergence - introduction

Convergence

A finite difference method is convergent if:

$$\lim_{\Delta t, h \rightarrow 0} \|u - u_h\| = 0$$

where u : analytical solution

u_h : approximated solution

Solution of the FD method (numerical approximation) gets closer to the exact solution of the PDE as the discretisation is made finer.

Difficult to show, **but**

Lax Richtmyer theorem

A **consistent** finite difference method for a **well-posed, linear** initial value problem is convergent if and only if it is **stable**.



Instead of analysing convergency
check consistency and stability

Stability, consistency, convergence - introduction

Lax Richtmyer theorem

A **consistent** finite difference method for a **well-posed, linear** initial value problem is convergent if and only if it is **stable**.

$$\|u - u_h\| = \|\mathcal{L}^{-1}f - \mathcal{L}_h^{-1}f_h\|$$

convergency:

$$\lim_{h \rightarrow 0} \|u - u_h\| = 0$$

$$\|u - u_h\| = \|\underbrace{\mathcal{L}^{-1}f - \mathcal{L}_h^{-1}f}_{=0} + \mathcal{L}_h^{-1}f - \mathcal{L}_h^{-1}f_h\|$$

$$\leq \|\mathcal{L}^{-1}f - \mathcal{L}_h^{-1}f\| + \|\mathcal{L}_h^{-1}f - \mathcal{L}_h^{-1}f_h\| \quad (\text{triangular inequality})$$

$$= \|\mathcal{L}_h^{-1}(\mathcal{L}_h u - \mathcal{L}u)\| + \|\mathcal{L}_h^{-1}(f - f_h)\| \quad (\text{linearity of } \mathcal{L}_h^{-1})$$

$$\leq C \underbrace{\|\mathcal{L}_h u - \mathcal{L}u\|}_{\rightarrow 0 \text{ as } h \rightarrow 0} + C \underbrace{\|f - f_h\|}_{\rightarrow 0 \text{ as } h \rightarrow 0} \quad (\text{stability})$$

(consistency)

$$\|\mathcal{L}^{-1}f - \mathcal{L}_h^{-1}f\| = \|\mathcal{L}^{-1}\mathcal{L}u - \mathcal{L}_h^{-1}\mathcal{L}u\| = \|u - \mathcal{L}_h^{-1}\mathcal{L}u\| = \|\underbrace{\mathcal{L}_h^{-1}\mathcal{L}_h u}_{=1} - \mathcal{L}_h^{-1}\mathcal{L}u\| = \|\mathcal{L}_h^{-1}(\mathcal{L}_h u - \mathcal{L}u)\| \quad (\text{linearity of } \mathcal{L}_h^{-1})$$

Consistency

Check consistency

Derivatives are approximated with the help of the Taylor series

1. Derivation of a consistent finite difference operator

Example: derivation of $u'(x)$ used in the Richardson scheme

$$u(x+h) = u(x) + u'(x)h + \frac{1}{2}u''(x)h^2 + O(h^3) \quad (1)$$

$$u(x-h) = u(x) - u'(x)h + \frac{1}{2}u''(x)h^2 + O(h^3) \quad (2)$$

Subtracting from eq. (1) eq. (2) results in:

$$u(x+h) - u(x-h) = 2u'(x)h + O(h^3)$$

$$u'(x) = \frac{u(x+h) - u(x-h)}{2h} + O(h^2)$$

$$u'_k = \frac{u_{k+1} - u_{k-1}}{2h} + O(h^2)$$

Consistency

2. Check consistency of an already defined scheme

Example: prove consistency of the DuFort-Frankel scheme

$$\frac{u_{n+1,j} - u_{n-1,j}}{2\Delta t} - \frac{\beta^2}{h^2} (u_{n,j-1} - (u_{n-1,j} + u_{n+1,j}) + u_{n,j+1}) = 0$$

$$u_{n+1,j} = u(t + \Delta t, x) = u_{n,j} + \frac{\partial u}{\partial t} \Delta t + \frac{1}{2} \frac{\partial^2 u}{\partial t^2} \Delta t^2 + O(\Delta t^3)$$

$$u_{n-1,j} = u(t - \Delta t, x) = u_{n,j} - \frac{\partial u}{\partial t} \Delta t + \frac{1}{2} \frac{\partial^2 u}{\partial t^2} \Delta t^2 + O(\Delta t^3)$$

$$u_{n,j+1} = u(t, x + h) = u_{n,j} + \frac{\partial u}{\partial x} h + \frac{1}{2} \frac{\partial^2 u}{\partial x^2} h^2 + O(h^3)$$

$$u_{n,j-1} = u(t, x - h) = u_{n,j} - \frac{\partial u}{\partial x} h + \frac{1}{2} \frac{\partial^2 u}{\partial x^2} h^2 + O(h^3)$$

$$\frac{\frac{\partial u}{\partial t} \Delta t + O(\Delta t^3)}{\Delta t} - \frac{\beta^2 (2u_{n,j} + \frac{\partial^2 u}{\partial x^2} h^2 + O(h^3))}{h^2} + \frac{\beta^2 (2u_{n,j} + \frac{\partial^2 u}{\partial t^2} \Delta t^2 + O(\Delta t^3))}{h^2} = 0$$

Consistency

$$\frac{\frac{\partial u}{\partial t} \Delta t + O(\Delta t^3)}{\Delta t} - \frac{\beta^2 (2u_{n,j} + \frac{\partial^2 u}{\partial x^2} h^2 + O(h^3))}{h^2} + \frac{\beta^2 (2u_{n,j} + \frac{\partial^2 u}{\partial t^2} \Delta t^2 + O(\Delta t^3))}{h^2} = 0$$

$$\frac{\partial u}{\partial t} - \beta^2 \frac{\partial^2 u}{\partial x^2} + E = 0$$

$$E = \frac{\beta^2 \Delta t^2}{h^2} \frac{\partial^2 u}{\partial t^2} + O(\Delta t^2) + O(h)$$

The method is consistent if: $\lim_{\Delta t, h \rightarrow 0} |E| = 0$

The second and the last term will tend to zero as discretisation is refined, but the last term will only be zero if

$$\lim_{\Delta t, h \rightarrow 0} \left| \frac{\Delta t}{h} \right| = 0$$

For example if the stability condition of Euler forward is satisfied:

$$\Delta t < \frac{h^2}{2\beta^2} \quad \longrightarrow \quad \Delta t = O(h^2) \quad \longrightarrow \quad \text{scheme is consistent}$$

Stability

Stability checking from eigenvalue analysis:

- Method of lines

$$\mathbf{u}(t) = \sum_{j=1}^{N-1} \beta_j(t) \mathbf{v}_j \quad \beta_j(t) = \beta_j^0 e^{\lambda_j t}$$

$$\text{eig}(\mathbf{A}) \leq 0$$



unconditionally satisfied

- Euler forward method

$$\mathbf{u}_{n+1} = \underbrace{(\mathbf{I} + \Delta t \mathbf{A})}_{\mathbf{B}} \mathbf{u}_n$$

$$|\text{eig}(\mathbf{B})| \leq 1$$



$$\Delta t < \frac{h^2}{2\beta^2}$$

- Euler backward method

$$\mathbf{u}_n = \underbrace{(\mathbf{I} - \Delta t \mathbf{A})}_{\mathbf{B}_1} \mathbf{u}_{n+1}$$

$$|\text{eig}(\mathbf{B}_1^{-1})| \leq 1$$



unconditionally satisfied

- Theta method

$$\underbrace{(\mathbf{I} - \theta \Delta t \mathbf{A})}_{\mathbf{B}_1} \mathbf{u}_{n+1} = \underbrace{(\mathbf{I} + (1 - \theta) \Delta t \mathbf{A})}_{\mathbf{B}_2} \mathbf{u}_n$$

$$|\text{eig}(\mathbf{B}_1^{-1} \mathbf{B}_2)| \leq 1$$



for $\theta \geq 1/2$: unconditionally stable

for $\theta < 1/2$: $\frac{\beta^2 \Delta t}{h^2} < \frac{1}{2(1 - 2\theta)}$

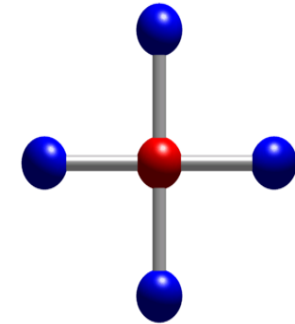
Instationary heat equation in 3D

Let's consider the heat equation in 2 dimensions: $\frac{\partial u}{\partial t} - \beta^2 \left(\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right) = f$

$$\begin{aligned} \frac{\partial^2 u_{j,l}}{\partial x^2} &= \frac{1}{\Delta x^2} (u_{j-1,l} - 2u_{j,l} + u_{j+1,l}) & x &= j \cdot \Delta x \\ \frac{\partial^2 u_{j,l}}{\partial y^2} &= \frac{1}{\Delta y^2} (u_{j,l-1} - 2u_{j,l} + u_{j,l+1}) & y &= l \cdot \Delta y. \end{aligned}$$

$$\frac{\partial u_{j,l}}{\partial t} - \frac{\beta^2}{\Delta x^2} (u_{j-1,l} - 2u_{j,l} + u_{j+1,l}) - \frac{\beta^2}{\Delta y^2} (u_{j,l-1} - 2u_{j,l} + u_{j,l+1}) = f$$

If $dx=dy$:
$$\frac{\partial u_{j,l}}{\partial t} - \frac{\beta^2}{h^2} (-4u_{j,l} + u_{j-1,l} + u_{j+1,l} + u_{j,l-1} + u_{j,l+1}) = f.$$



After time discr. with theta method:

$$\begin{aligned} \frac{u_{j,l}^{n+1} - u_{j,l}^n}{\Delta t} &= \frac{\beta^2}{h^2} \left((1 - \theta) (-4u_{j,l}^n + u_{j-1,l}^n + u_{j+1,l}^n + u_{j,l-1}^n + u_{j,l+1}^n) + \right. \\ &\quad \left. \theta (-4u_{j,l}^{n+1} + u_{j-1,l}^{n+1} + u_{j+1,l}^{n+1} + u_{j,l-1}^{n+1} + u_{j,l+1}^{n+1}) \right) + f \end{aligned}$$

Instationary heat equation in 3D

A: triangular matrix \longrightarrow $\left. \begin{array}{l} \mathbf{B}_1 = (\mathbf{I} - \Delta t \mathbf{A}) \\ \mathbf{B}_{1\theta} = (\mathbf{I} - \theta \Delta t \mathbf{A}) \end{array} \right\}$ tridiagonal matrices

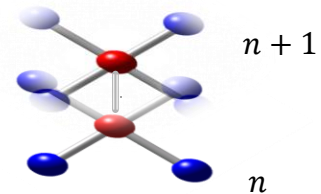
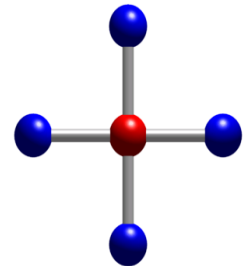
Recall 2D instationary heat equation: $\frac{\partial u}{\partial t} - \beta^2 \left(\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right) = f$ (See Tutorial 3.)

$$\frac{\partial u_{j,l}}{\partial t} - \frac{\beta^2}{\Delta x^2} (u_{j-1,l} - 2u_{j,l} + u_{j+1,l}) - \frac{\beta^2}{\Delta y^2} (u_{j,l-1} - 2u_{j,l} + u_{j,l+1}) = f$$

If $dx=dy$: $\frac{\partial u_{j,l}}{\partial t} - \frac{\beta^2}{h^2} (-4u_{j,l} + u_{j-1,l} + u_{j+1,l} + u_{j,l-1} + u_{j,l+1}) = f.$

After time discr. with theta method:

$$\frac{u_{j,l}^{n+1} - u_{j,l}^n}{\Delta t} = \frac{\beta^2}{h^2} \left((1 - \theta) (-4u_{j,l}^n + u_{j-1,l}^n + u_{j+1,l}^n + u_{j,l-1}^n + u_{j,l+1}^n) + \theta (-4u_{j,l}^{n+1} + u_{j-1,l}^{n+1} + u_{j+1,l}^{n+1} + u_{j,l-1}^{n+1} + u_{j,l+1}^{n+1}) \right) + f$$



Instationary heat equation in 3D

$$f := 0 \quad \Delta t \frac{\beta^2}{h^2} := r$$

$$\begin{aligned} u^{n+1}_{j,l} - r\theta(-4u^{n+1}_{j,l} + u^{n+1}_{j-1,l} + u^{n+1}_{j+1,l} + u^{n+1}_{j,l-1} + u^{n+1}_{j,l+1}) &= \\ = u^n_{j,l} + r(1 - \theta)(-4u^n_{j,l} + u^n_{j-1,l} + u^n_{j+1,l} + u^n_{j,l-1} + u^n_{j,l+1}) \end{aligned}$$

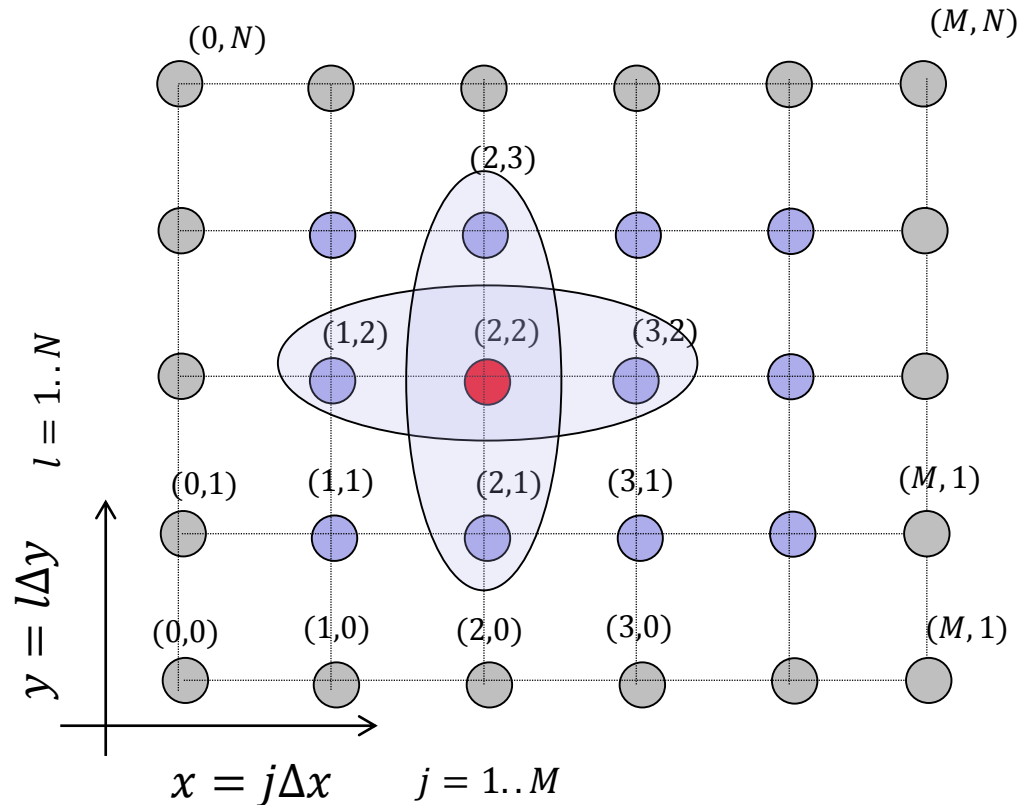
$\theta = 1$ Euler backward method

$$u^{n+1}_{j,l} - r(-4u^{n+1}_{j,l} + u^{n+1}_{j-1,l} + u^{n+1}_{j+1,l} + u^{n+1}_{j,l-1} + u^{n+1}_{j,l+1}) = u^n_{j,l}$$

Instationary heat equation in 3D

$$\mathbf{u}^n = \begin{bmatrix} u_{1,1}^n \\ u_{2,1}^n \\ u_{3,1}^n \\ \vdots \\ u_{M-1,1}^n \\ u_{1,2}^n \\ u_{2,2}^n \\ \vdots \\ u_{M-1,2}^n \\ \vdots \\ u_{1,N-1}^n \\ \vdots \\ u_{M-1,N-1}^n \end{bmatrix}$$

$$\mathbf{u}^n = \begin{bmatrix} u_{1,1}^n \\ u_{2,1}^n \\ u_{3,1}^n \\ u_{4,1}^n \\ u_{1,2}^n \\ \mathbf{u_{2,2}^n} \\ u_{3,2}^n \\ u_{4,2}^n \\ u_{1,3}^n \\ u_{2,3}^n \\ u_{3,3}^n \\ u_{4,3}^n \end{bmatrix}$$



with homogenous Dirichlet BC.

$$u_{j,l}^{n+1} - r(-4u_{j,l}^{n+1} + u_{j-1,l}^{n+1} + u_{j+1,l}^{n+1} + u_{j,l-1}^{n+1} + u_{j,l+1}^{n+1}) = u_{j,l}^n$$

Instationary heat equation in 3D

$$u^{n+1}_{j,l} - r(-4u^{n+1}_{j,l} + u^{n+1}_{j-1,l} + u^{n+1}_{j+1,l} + u^{n+1}_{j,l-1} + u^{n+1}_{j,l+1}) = u^n_{j,l}$$

with homogenous Dirichlet BC.

4	-1				-1	-1				$\begin{bmatrix} u^{n+1}_{1,1} \\ u^{n+1}_{2,1} \\ u^{n+1}_{3,1} \\ u^{n+1}_{4,1} \\ u^{n+1}_{1,2} \\ u^{n+1}_{2,2} \\ u^{n+1}_{3,2} \\ u^{n+1}_{4,2} \\ u^{n+1}_{1,3} \\ u^{n+1}_{2,3} \\ u^{n+1}_{3,3} \\ u^{n+1}_{4,3} \end{bmatrix}$					
-1	4	-1					-1								
	-1	4	-1					-1							
		-1	4	-1							-1				
-1			4	1				-1			-1				
+r 0	-1	0	0	-1	4	-1	0	0	-1		0	0			
		-1					-1	4	-1						
				-1	4				-1						
				-1				4	-1						
				-1	-1	4	-1					-1			
				-1				-1	4		-1				
				-1				-1	-1		4				
				-1				-1	-1		4				

Sparse matrix with bandwidth: $2M-1$ (here 9) BUT the band itself is sparse, only five diagonals are nonzero

Instationary heat equation in 3D

$$u^{n+1}_{j,l} - r(-4u^{n+1}_{j,l} + u^{n+1}_{j-1,l} + u^{n+1}_{j+1,l} + u^{n+1}_{j,l-1} + u^{n+1}_{j,l+1}) = u^n_{j,l}$$

$$\mathbf{B}_1 \mathbf{u}^{n+1} = \mathbf{u}^n$$

where

$$\mathbf{B}_1 = \begin{bmatrix} B & C & & \\ C & B & C & \\ & \ddots & \ddots & \ddots \\ & & C & B \end{bmatrix}$$

$$\mathbf{B} = \begin{bmatrix} 1 + 4r & -r & & & \\ -r & 1 + 4r & -r & & \\ & \ddots & \ddots & \ddots & \\ & & -r & 1 + 4r & \end{bmatrix}$$

$$\mathbf{C} = \begin{bmatrix} -r & & & & \\ & -r & & & \\ & & \ddots & & \\ & & & -r & \end{bmatrix}$$

Reminder:

Stationary heat equation – what to solve?

Instationary heat equation with constant BC, and source term approaches a stationary state:

$$\frac{\partial}{\partial t}u(x,y,z,t) - \beta^2\Delta u(x,y,z,t) = f(x,y,z) \quad (\text{parabolic})$$
$$u(x,y,z,t) \rightarrow \tilde{u}(x,y,z) \quad \text{as } t \rightarrow \infty$$

Equilibrium equation (stationary heat equation):

$$\frac{\partial}{\partial t}\tilde{u}(x,y,z) = 0 \quad \Rightarrow \quad -\beta^2\Delta\tilde{u}(x,y,z) = f(x,y,z) \quad (\text{elliptic})$$

Discretised form:

$$\frac{\partial}{\partial t}\mathbf{u} + \mathbf{A}\mathbf{u} = \mathbf{f} \quad \longrightarrow \quad \mathbf{A}\mathbf{u} = \mathbf{f}$$

Reminder:

Stationary heat equation – what to solve?

Conclusion

- instationary heat equation with implicit FD methods (Euler backward, Theta method)
- stationary heat equation



System of linear equations:

$$\mathbf{G}\mathbf{x} = \mathbf{b}$$

solve for \mathbf{x}

Where the G matrix is in general

- **sparse, banded**
- can get very **large** with refined spatial and temporal discretisation
- for 1D heat equation with three-point-stencils: **tridiagonal**
- for 1D heat equation with five-point-stencils: **pentadiagonal**
- for 2D heat equation: **banded with sparse band**