

# Chapter 9: Stability of Difference Schemes

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Mathematical Models, Analysis and Simulation, Part I

# Hadamard's Concept of Well-Posedness

A given problem is well-posed if its solution depends continuously on the data.

- Note: This notion depends essentially on
  - the allowed data,
  - the type of solutions searched for,
  - the measure of continuity (“the norm”).
- Consider a linear homogeneous initial value problem,

$$\frac{d}{dt}u = Lu, \quad u(0) = u_0.$$

$L$  may be a matrix, a linear differential operator w r t space variables etc.

- Consider only initial values as data.

**Definition.** *The IVP is well-posed w r t the initial data, if there exist constants  $K, C$  independent of  $u_0$  such that the IVP is uniquely solvable, and*

$$\|u(t)\| \leq Ke^{Ct} \|u_0\|.$$

## Example: Advection Equation

- Consider
  - $u_t = cu_x, \quad x \in \mathbb{R}, \quad t > 0.$
  - Special case of the transport equation in 1D.
  - Did we already meet in connection with d'Alembert's solution of the wave equation.
  - It is a *hyperbolic equation*.
- Multiply by  $u$  and integrate:

$$\begin{aligned} 0 &= \int_{-\infty}^{+\infty} (uu_t - cu_x u) dx \\ &= \frac{d}{dt} \int_{-\infty}^{+\infty} \frac{1}{2} u^2 dx - \underbrace{c \frac{1}{2} u^2}_{=0} \Big|_{-\infty}^{+\infty} \\ &= \frac{1}{2} \frac{d}{dt} \|u(\cdot, t)\|^2 \end{aligned}$$

Hence:

$$\|u(\cdot, t)\| = \|u_0\|$$

## Example (cont.)

- Growth estimate:  $\|u(\cdot, t)\| \leq 1 \cdot e^{0 \cdot t} \|u_0\|$
- Well-posedness with
  - $K = 1$  and  $C = 0$
  - data from  $L^2(\mathbb{R})$ , solutions from  $C^1([0, \infty), L^2(\mathbb{R}))$ .
- *Hyperbolic equations are well-posed.*

## Well-Posedness for Nonlinear Systems of ODEs

- Consider

$$\mathbf{y}' = \mathbf{f}(\mathbf{y}), \quad \mathbf{y}(0) = \mathbf{y}^0, \quad 0 \leq t \leq T.$$

- Lipschitz condition*: There exists a  $L$  such that

$$\|\mathbf{f}(\mathbf{y}) - \mathbf{f}(\mathbf{x})\| \leq L\|\mathbf{y} - \mathbf{x}\|$$

for all  $\mathbf{y}, \mathbf{x}$  in a neighborhood of  $\mathbf{y}_0$ .

- Theorem of Picard-Lindelöf*: For a sufficiently small  $T$  there is a unique solution which depends continuously on the data  $\mathbf{y}_0$ .
- Limit on the growth rate:

$$\|\mathbf{y}(t) - \mathbf{y}^0\| \leq t e^{Lt} \|\mathbf{f}(\mathbf{y}^0)\|.$$

- Consider now two solutions  $\mathbf{y}, \mathbf{x}$  subject to initial conditions  $\mathbf{y}^0, \mathbf{x}^0$ :

$$\|\mathbf{y} - \mathbf{x}\| \leq e^{Lt} \|\mathbf{y}^0 - \mathbf{x}^0\|.$$

These inequalities are a consequence of *Gronwall's lemma*.

## Euler Discretizations

Read: Strang, p 461–466

- Consider

$$\mathbf{y}' = \mathbf{A}\mathbf{y}, \quad \mathbf{y}(0) = \mathbf{y}^0, \quad t_n = n\Delta t.$$

- Let first  $\mathbf{y} = y \in \mathbb{R}^1, \mathbf{A} = a$ ,

$$y(t_n) = e^{t_n a} y^0.$$

- Explicit Euler reads

$$\frac{y^n - y^{n-1}}{\Delta t} = a y^{n-1}, \quad \text{hence } y^n = (1 + \Delta t a)^n y^0.$$

- Since

$$\lim_{n \rightarrow \infty} (1 + \Delta t a)^n = \lim_{n \rightarrow \infty} \left(1 + \frac{t_n}{n} a\right)^n = e^{a t_n},$$

it holds

$$e_n = y(t_n) - y^n \rightarrow 0,$$

that is *convergence*.

## Euler Discretizations: Cont

- Observe

$$|e^{\Delta t a}| < 1 \text{ iff } a < 0$$

- This shall be modeled by the discrete system:

$$a < 0 \implies |1 + \Delta t a| < 1.$$

This holds true only if

$$-a\Delta t < 2$$

Notation: *asymptotic stability*.

- If  $y^n$  shall *non-oscillatory*,  $1 + \Delta t a > 0$  must hold such that even

$$-a\Delta t < 1$$

is required!

Note the difference between: Convergence, asymptotic stability, non-oscillation!

## Explicit Euler For Systems

- Define the matrix exponential by

$$e^{\mathbf{B}} = \sum_{j=0}^{\infty} \frac{\mathbf{B}^j}{j!}$$

- Solutions:

$$\mathbf{y}(t_n) = e^{t_n \mathbf{A}} \mathbf{y}^0, \quad \mathbf{y}^n = (1 + \Delta t \mathbf{A})^n \mathbf{y}^0$$

- Convergence can be proved as before!
- “Growth factor”:

$$\mathbf{G} = \mathbf{I} + \Delta t \mathbf{A}.$$

## Explicit Euler for Systems: Cont

Q: Under which conditions holds:  $\|\mathbf{G}\| < 1$ ?

- Let  $\mathbf{A}$  be symmetric with eigenvalues  $\lambda_i$ . Then:
  - Eigenvalues of  $\mathbf{G}$ :  $\mu_i = 1 + \Delta t \lambda_i$ .
  - $\|\mathbf{G}\| = \max_i |\mu_i|$ .
  - $\mathbf{y}(t) \rightarrow 0$  iff  $-\mathbf{A}$  is positive definite.
  - For spd  $-\mathbf{A}$ :  $\|\mathbf{G}\| < 1$  iff

$$\Delta t |\lambda_{\max}| < 2$$

- For general  $\mathbf{A}$ :

$$\mathbf{y}(t) \rightarrow 0 \text{ iff } \Re(\lambda_i) < 0$$

$$\mathbf{y}^n \rightarrow 0 \text{ iff } \max_i |1 - \Delta t \lambda_i| < 1.$$

## Explicit Euler For PDE Discretizations

- Consider a parabolic problem

$$u_t - \operatorname{div}(k \nabla u) = 0, \quad k > 0$$

$$u_\Gamma = 0, \quad u(\cdot, 0) = u^0.$$

- Galerkin approximation:

$$\mathbf{M} \frac{d}{dt} \boldsymbol{\tau}_h + \mathbf{A} \boldsymbol{\tau}_h = 0, \quad \boldsymbol{\tau}_h^0 = \Pi u^0.$$

- Need the eigenvalues of  $\mathbf{M}^{-1} \mathbf{A}$ . It holds
  - This matrix is pd.
  - $0 < \lambda_{\min} = O(1)$  and  $\lambda_{\max} = O(h^{-2})$ .
- Asymptotic stability requirement  $\Delta t \lambda_{\max} < 2$  becomes

$$\Delta t < 2ch^2$$

This requirement makes explicit Euler rather expensive!

## Implicit Euler Discretization

- Discretization

$$\frac{\mathbf{y}^n - \mathbf{y}^{n-1}}{\Delta t} = \mathbf{A} \mathbf{y}^n, \text{ hence } \mathbf{y}^n = (\mathbf{I} - \Delta t \mathbf{A})^{-n} \mathbf{y}^0.$$

- Growth factor:

$$\mathbf{G} = (\mathbf{I} - \Delta t \mathbf{A})^{-1}$$

- For a symmetric  $\mathbf{A}$ :  $\mu_i = (1 - \Delta t \lambda_i)^{-1}$ .
- Asymptotic stability iff

$$\max_i \frac{1}{|1 - \Delta t \lambda_i|} < 1$$

- If  $-\mathbf{A}$  is spd, there are no restrictions on  $\Delta t$ !
- The price to pay: Solve a linear system in each step:

$$(\mathbf{I} - \Delta t \mathbf{A}) \mathbf{y}^n = -\mathbf{y}^{n-1}.$$

## Detailed Stability Analysis

Read: Strang, p 481–482

**Theorem. [Lax Equivalence Theorem]** Consistency and stability are necessary and sufficient for convergence.

	equilibrium problems	evolution problems
continuous	$Lu = f$	$u_t = Lu$
discrete	$L_h u_h = f_h$	$u_h^n = G_h u_h^{n-1}$
example	FEM	MOL/Rothe

**Convergence** Does the discrete solution converge towards the continuous one?

**Consistency** Does the discrete equation approximate the continuous counterpart?  $\Rightarrow$  Easy!

Example:

$$L_h u_h \rightarrow Lu, \quad f_h \rightarrow f?$$

**Stability** Need to distinguish:

- equilibrium: Is  $L_h^{-1}$  uniformly bounded?
- evolution: Is the discrete evolution uniformly bounded, i.e.,  $|G_h^n| \leq e^{Kn\Delta t}$ ?

This is the hard part!

## Detailed Stability Analysis: Cont

For PDEs, this coarse description is not sufficient. We need a more detailed study.

**Tool** Fourier analysis

**Result** von Neumann stability analysis

## Why Does it Work? An Example

Read: Strang, Ch. 6.3

- Consider the hyperbolic Cauchy problem

$$u_t = cu_x, \quad x \in \mathbb{R},$$

$$u(x, 0) = u^0(x), \quad u^0 \in L^2(\mathbb{R}).$$

- Fourier transform

$$\hat{f}(k) = \int_{-\infty}^{+\infty} f(x)e^{-ikx} dx, \quad k \in \mathbb{R}$$

- Plancherel's identity

$$\|\hat{f}\|^2 = 2\pi\|f\|^2.$$

- Taking the Fourier transform in  $x$ :

$$\int_{-\infty}^{+\infty} \frac{\partial}{\partial t} u(x, t) e^{-ikx} dx = c \int_{-\infty}^{+\infty} \frac{\partial}{\partial x} u(x, t) e^{-ikx} dx$$

$$\frac{\partial}{\partial t} \int_{-\infty}^{+\infty} u(x, t) e^{-ikx} dx = -c \int_{-\infty}^{+\infty} u(x, t) \frac{\partial}{\partial x} e^{-ikx} dx + \mathbf{b.t.}$$

$$\hat{u}_t(k) = cik\hat{u}(k)$$

## An Example: Cont

- The transformed equation is a set of ode's w r t  $t$  parametrized by  $k$ :

$$\hat{u}_t = cik\hat{u} \implies \hat{u}(k) = e^{cikt} \hat{u}^0(k).$$

- Taking norms:

$$\|\hat{u}\|^2 = \int_{-\infty}^{+\infty} |e^{cikt} \hat{u}^0(k)|^2 dk = \|\hat{u}^0\|^2.$$

- Using Plancherel's identity, we obtain the stability estimate (slide 3) once again:

$$\|u(t)\| = \|u^0\|.$$

## Example: The Discrete Version

- We use the direct *complete* discretization:

$$x_j = j \cdot \Delta x, t_n = n \cdot \Delta t, u_j^n \approx u(x_j, t_n).$$

- First-order accurate difference approximations ("explicit Euler"):

$$\frac{u_j^{n+1} - u_j^n}{\Delta t} = c \frac{u_{j+1}^n - u_j^n}{\Delta x}$$

More explicit:

$$u_j^{n+1} = u_j^n + r(u_{j+1}^n - u_j^n), \quad r = \frac{c\Delta t}{\Delta x}.$$

$r$  is called the *Courant number*.

- Interpolate  $u_j^n$  by a smooth function  $v(x, t)$ . Then

$$v(x, t + \Delta t) - v(x, t) - r(v(x + \Delta x, t) - v(x, t)) = 0$$

## The Discrete Version: Cont

- It holds

$$\widehat{f(\cdot + \Delta x)} = \int_{-\infty}^{+\infty} f(x + \Delta x) e^{-ikx} dx = \int_{-\infty}^{+\infty} f(x + \Delta x) e^{-ik(x + \Delta x)} e^{ik\Delta x} dx = \widehat{f} e^{ik\Delta x}$$

- Applied to our recursion:

$$\hat{v}(k, t + \Delta t) - \hat{v}(k, t) - r(e^{ik\Delta x} \hat{v}(k, t) - \hat{v}(k, t)) = 0$$

$$\hat{v}(k, t + \Delta t) = G(\theta, r) \hat{v}(k, t)$$

with the growth factor

$$G(\theta, r) = 1 + r(e^{i\theta} - 1)$$

depending on

- $\theta = k\Delta x$ , the phase shift per cell,
- $r = c\Delta t/\Delta x$ , the Courant number.

## Conclusions

- The norm estimate becomes

$$\|\hat{v}(\cdot, t + \Delta t)\| \leq \|G\|_{\infty} \|\hat{v}(\cdot, t)\|.$$

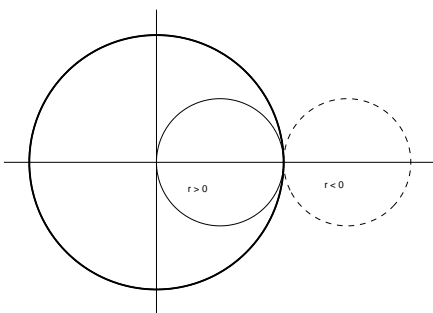
- Taking into account the stability properties of the continuous problem, we want to model it discretely. This leads to the *stability requirement*

$$\|G\|_{\infty} \leq 1$$

- The actual necessary and sufficient condition is  $|G| \leq 1 + O(\Delta t)$ .
- Fixing  $r$ ,  $G(\cdot, r)$  describes a circle with center  $1 - r$  and radius  $r$ .

## Conclusions: Cont

Stability Function



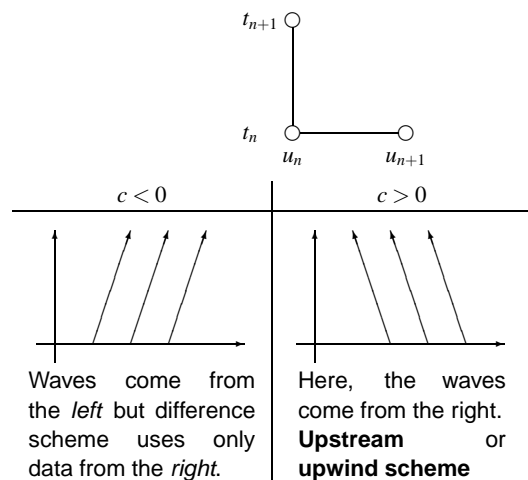
- Hence:
  - $c > 0$  is necessary for stability.
  - $r \leq 1$  is necessary, which is equivalent to

$$\frac{c\Delta t}{\Delta x} \leq 1$$

- This condition is the celebrated *Courant-Friedrichs-Lewy condition* (short: CFL condition):  
*If it is violated, then there cannot be convergence.*
- Here, it is a necessary (and sufficient) condition for *stability*.

## A Geometrical Interpretation

Difference stencil:



Waves come from the *left* but difference scheme uses only data from the *right*.

Here, the waves come from the *right*. **Upstream** or **upwind scheme**

## Higher Order Schemes

- Try a scheme *forward in time, centered in space* (FTCS):

$$u_j^{n+1} - u_j^n - \frac{r}{2}(u_{j+1}^n - u_{j-1}^n) = 0$$

It holds  $G(\theta, r) = 1 + ir \sin \theta$  such that the scheme is *unstable*,

$$|G| \geq 1$$

- *Lax-Friedrichs scheme* (first order)

$$u_j^{n+1} - \frac{1}{2}(u_{j+1}^n + u_{j-1}^n) = \frac{r}{2}(u_{j+1}^n - u_{j-1}^n)$$

Amplification factor  $G(\theta, r) = \cos \theta + ir \sin \theta$ . This is an ellipse with half axes 1 and  $r$ . So

$$|G| \leq 1 \text{ iff } |r| \leq 1$$

This scheme works independent of the sign of  $c$ !  
Necessary for the wave equation with two characteristics of opposite directions.

## Higher Order Schemes: Cont

- *Lax-Wendroff scheme* (second order)

Start by using the Taylor expansion:

$$u(t + \Delta t) = u(t) + \Delta t u_t(t) + \frac{(\Delta t)^2}{2} u_{tt}(t) + O((\Delta t)^3).$$

$u_t$  can be replaced by  $cu_x$  using the differential equation.  
What to do with  $u_{tt}$ ? Use the equation once again:

$$u_{tx} = cu_{xx} \text{ and } u_{tt} = cu_{xt} \Rightarrow u_{tt} = c^2 u_{xx}.$$

Hence,

$$u(t + \Delta t) \approx u(t) + \Delta t cu_x + \frac{(\Delta t)^2}{2} c^2 u_{xx}.$$

The last term is additional diffusion which has a stabilizing effect!

Discretization:

$$u_j^{n+1} - u_j^n = \frac{r}{2}(u_{j+1}^n - u_{j-1}^n) + \frac{r^2}{2}(u_{j+1}^n - 2u_j^n + u_{j-1}^n)$$

Amplification factor  $G(\theta, r) = 1 + ir \sin \theta - 2r^2 \sin^2 \theta$ .

One can prove that

$$|G| \leq 1 \text{ iff } |r| \leq 1$$