

Mathematical Tripos Part II: Michaelmas Term 2021

Numerical Analysis – Lecture 14

Method 3.23 (The spectral method for evolutionary PDEs) We consider the problem

$$\begin{cases} \frac{\partial u(x,t)}{\partial t} = \mathcal{L}u(x,t), & x \in [-1, 1], \quad t \geq 0, \\ u(x,0) = g(x), & x \in [-1, 1], \end{cases} \quad (3.20)$$

with appropriate boundary conditions on $\{-1, 1\} \times \mathbb{R}_+$ and where \mathcal{L} is a linear operator (acting on x), e.g., a differential operator. We want to solve this problem by the method of lines (semi-discretization), using a spectral method for the approximation of u and its derivatives in the spatial variable x . Then, in a general spectral method, we seek solutions $u_N(x, t)$ with

$$u_N(x, t) = \sum_{\#\{n\}=N} c_n(t) \varphi_n(x), \quad (3.21)$$

where $c_n(t)$ are expansion coefficients and φ_n are basis functions chosen according to the specific structure of (3.20). For example, we may take

- 1) the *Fourier expansion* with $c_n(t) = \hat{u}_n(t)$, $\varphi_n(x) = e^{i\pi n x}$ for periodic boundary conditions,
- 2) a polynomial expansion such as the *Chebyshev expansion* with $c_n(t) = \check{u}_n(t)$, $\varphi_n(x) = T_n(x)$ for other boundary conditions.

The spectral approximation in space (3.21) results into a $N \times N$ system of ODEs for the expansion coefficients $\{c_n(t)\}$:

$$\mathbf{c}' = B\mathbf{c}, \quad (3.22)$$

where $B \in \mathbb{R}^{N \times N}$, and $\mathbf{c} = \{c_n(t)\} \in \mathbb{R}^N$. We can solve it with standard ODE solvers (Euler, Crank-Nikolson, etc.) which as we have seen are approximations to the matrix exponent in the exact solution $\mathbf{c}(t) = e^{tB}\mathbf{c}(0)$.

Example 3.24 (The diffusion equation) Consider the diffusion equation for a function $u = u(x, t)$,

$$\begin{cases} u_t = u_{xx}, & (x, t) \in [-1, 1] \times \mathbb{R}_+, \\ u(x, 0) = g(x), & x \in [-1, 1]. \end{cases} \quad (3.23)$$

with the periodic boundary conditions $u(-1, t) = u(1, t)$, $u_x(-1, t) = u_x(1, t)$, and standard normalisation $\int_{-1}^1 u(x, t) dx = 0$, both imposed for all values $t \geq 0$.

For each t , we approximate $u(x, t)$ by its N -th order partial Fourier sum in x ,

$$u(x, t) \approx u_N(x, t) = \sum_{n \in \Gamma_N} \hat{u}_n(t) e^{i\pi n x}, \quad \Gamma_N := \{-N/2+1, \dots, N/2\}.$$

Then, from (3.23), we see that each coefficient \hat{u}_n fulfills the ODE

$$\hat{u}'_n(t) = -\pi^2 n^2 \hat{u}_n(t). \quad n \in \Gamma_N \quad (3.24)$$

Its exact solution is $\hat{u}_n(t) = e^{-\pi^2 n^2 t} \hat{g}_n$ for $n \neq 0$ and we set $\hat{u}_0(t) = 0$ due to the normalisation condition, so that

$$u_N(x, t) = \sum_{n \in \Gamma_N} \hat{g}_n e^{-\pi^2 n^2 t} e^{i\pi n x},$$

which is the exact solution truncated to N terms.

Here, we were able to find the exact solution without solving ODE numerically due to the special structure of the Laplacian. However, for more general PDE we will need a numerical method, and thus the issue of stability arises, so we consider this issue on that simplified example.

Analysis 3.25 (Stability analysis) The system (3.24) has the form

$$\hat{\mathbf{u}}' = B\hat{\mathbf{u}}, \quad B = \text{diag} \{-\pi^2 n^2\}, \quad n \in \Gamma_N,$$

and we note that (a) all the eigenvalues of B are negative, and that (b) they consist of the eigenvalues $\lambda_n^{(2)}$ of the second order differentiation operator, with $\max |\lambda_n^{(2)}| = (\frac{N}{2})^2$.

If we approximate this system with the Euler method:

$$\hat{\mathbf{u}}^{k+1} = (I + \tau B)\hat{\mathbf{u}}^k, \quad \tau := \Delta t,$$

then we see that, for stability condition $\|I + \tau B\| \leq 1$, we need to scale the time step $\tau = \Delta t \sim N^{-2}$.

Note that, for the Crank-Nikolson scheme, since the spectrum of B is negative, we get stability for any time step $\tau > 0$.

For general linear operator \mathcal{L} in (3.20) with constant coefficients, the matrix B is again diagonal (hence normal), and provided that its spectrum is negative, for stability we must scale the time step $\tau \sim N^{-m}$, where m is the maximal order of differentiation.

The scaling $\tau \sim N^{-2}$ may seem similar to the scaling $k \sim h^2$ in difference methods which we viewed as a disadvantage, however in spectral methods we can take N , the order of partial Fourier or Chebyshev sums to achieve a good approximation, rather small. (We may still need to choose τ small enough to get a desired accuracy.)

Example 3.26 (The diffusion equation with non-constant coefficient) We want to solve the diffusion equation with a non-constant coefficient $a(x) > 0$ for a function $u = u(x, t)$

$$\begin{cases} u_t = (a(x)u_x)_x, & (x, t) \in [-1, 1] \times \mathbb{R}_+, \\ u(x, 0) = g(x), & x \in [-1, 1], \end{cases} \quad (3.25)$$

with boundary and normalization conditions as before. Approximating u by its partial Fourier sum results in the following system of ODEs for the coefficients \hat{u}_n

$$\hat{u}'_n(t) = -\pi^2 \sum_{m \in \Gamma_N} mn \hat{a}_{n-m} \hat{u}_m(t), \quad n \in \Gamma_N.$$

For the discretization in time we may apply the Euler method, this gives

$$\hat{u}_n^{k+1} = \hat{u}_n^k - \tau \pi^2 \sum_{m \in \Gamma_N} mn \hat{a}_{n-m} \hat{u}_m^k, \quad \tau = \Delta t,$$

or in the vector form

$$\hat{\mathbf{u}}^{k+1} = (I + \tau B)\hat{\mathbf{u}}^k,$$

where $B = (b_{m,n}) = (-\pi^2 mn \hat{a}_{n-m})$. For stability of Euler method, we again need $\|I + \tau B\| \leq 1$, but analysis here is less straightforward.

Matlab demo: See the online documentation *Using Chebyshev Spectral Methods* at <http://www.damtp.cam.ac.uk/user/hf323/M21-II-NA/demos/chebyshev/chebyshev.html> for a simple example of how boundary conditions can be installed.