Mathematical Tripos Part II: Michaelmas Term 2023

Numerical Analysis – Examples' Sheet 3

21. As discussed in the Lectures periodicity is necessary for spectral convergence. Suppose that an analytic function f on [-1,1] is not periodic, yet f(-1) = f(+1) and f'(-1) = f'(1). Integrating by parts the Fourier coefficients \hat{f}_n show that $\hat{f}_n = \mathcal{O}(n^{-3})$. Show that the rate of convergence of the *N*-terms truncated Fourier expansion of f is hence $\mathcal{O}(N^{-2})$.

Now, suppose $f(-1) \neq f(1)$. We can force the values at the endpoints to be equal. Set $f(x) = \frac{1}{2}(1-x)f(-1) + \frac{1}{2}(1+x)f(+1) + g(x)$, where $g(x) = f(x) - \frac{1}{2}(1-x)f(-1) - \frac{1}{2}(1+x)f(+1)$. Verify that $g(\pm 1) = 0$ and that if f is analytic then so is g. The idea is now to represent f as a linear function plus the Fourier expansion of g, i.e.,

$$f(x) = \frac{1}{2}(1-x)f(-1) + \frac{1}{2}(1+x)f(+1) + \sum_{n=-\infty}^{\infty} \hat{g}_n e^{i\pi nx}.$$

We can iterate this idea: To do so, construct a function h for which $h(\pm 1) = h'(\pm 1) = 0$ and verify that $\hat{h}_n = \mathcal{O}(n^{-3})$. [*Hint: In the second construction the function f will be represented as a cubic function plus the Fourier expansion of h.*]

22. Consider the following boundary value problem for the heat equation

$$\begin{cases} u_t = u_{xx}, & -1 \le x \le 1, \ t > 0\\ u(-1,t) = u(1,t), \ u_x(-1,t) = u_x(1,t), & t > 0\\ u(x,0) = e^{i\pi M x}, & -1 < x < 1. \end{cases}$$

where $M \in \mathbb{Z}$. By separation of variables one can compute the exact solution and get

$$u(x,t) = e^{-\pi^2 M^2 t} e^{i\pi M x}.$$

Now, approximate the solution u by its N-term truncated Fourier series and solve the spectral approximation for the heat equation, i.e.,

$$\sum_{n=-N/2+1}^{N/2} \frac{d\hat{u}_n}{dt}(t)e^{i\pi nx} = \sum_{n=-N/2+1}^{N/2} \hat{u}_n(t)\frac{d^2}{dx^2}e^{i\pi nx}.$$

What do you receive? What is the error of this method with the correct choice of *N*?

- 23. By Theorem 4.12, the Gauss-Seidel method for the solution of Ax = b converges whenever the matrix A is symmetric and positive definite. Show, however, by a 3×3 counterexample, that the Jacobi method for such an A need not converge. [Warning: For Jacobi, it is not enough to construct a positive definite A such that 2D A is not positive definite, because we did not prove that the Householder-John theorem gives a criterion. So, you need also to prove that $\rho(D^{-1}(A D)) > 1$.]
- 24. Let the Gauss-Seidel method be applied to the equations Ax = b when A is the nonsymmetric 2×2 matrix

$$A = \left[\begin{array}{rrr} 10 & -3 \\ 3 & 1 \end{array} \right].$$

Find the spectral radius of the iteration matrix. Then show that the relaxation method, described in Lecture 17, can reduce the spectral radius by a factor of 2.9. Further, show that iterating twice with Gauss-Seidel with this relaxation decreases the error $\|\boldsymbol{x}^{(k)} - \boldsymbol{x}^{(\infty)}\|$ by more than a factor of ten. Estimate the number of iterations of the original Gauss-Seidel method that would be required to achieve this decrease in the error.

25. The function u(x) = x (x - 1), $0 \le x \le 1$, is defined by the equations u''(x) = 2, $0 \le x \le 1$, and u(0) = u(1) = 0. A difference approximation to the differential equation provides the estimates $u_m \approx u(mh)$, m = 1, 2, ..., M - 1, through the system of equations

$$\begin{cases} u_{m-1} - 2u_m + u_{m+1} = 2h^2, & m = 1, 2, \dots, M-1 \\ u_0 = u_M = 0, \end{cases}$$

where h = 1/M, and M is a large positive integer. Show that the exact solution of the system is just $u_m = u(mh)$, m = 1, 2, ..., M - 1.

We employ the notation $u_m^{(\infty)} = u(mh)$, because we let the system be solved by the Jacobi iteration, using the starting values $u_m^{(0)} = 0$, m = 1, 2, ..., M - 1. Prove that the iteration matrix H has the spectral radius $\rho(H) = \cos(\pi/M)$. Further, by regarding the initial error vector $\mathbf{u}^{(0)} - \mathbf{u}^{(\infty)}$ as a linear combination of the eigenvectors of H, show that the largest component of $\mathbf{u}^{(k)} - \mathbf{u}^{(\infty)}$ for large k is approximately $(8/\pi^3) \cos^k(\pi/M)$. Hence deduce that the Jacobi method requires about $2.5M^2$ iterations to achieve $\|\mathbf{u}^{(k+1)} - \mathbf{u}^{(\infty)}\|_{\infty} \leq 10^{-6}$.

- 26. Implement using your favourite language (Matlab, Python, Julia, etc.) the multigrid method as seen in lecture to solve the 1D Poisson equation u'' = f on [0, 1] with zero Dirichlet boundary conditions u(0) = u(1) = 0. Try your method on a grid of size $m = 2^{10} 1$, with a forcing term containing high and low frequencies. Try changing the parameters of the algorithm (Jacobi vs. Gauss-Seidel, etc.) and comment.
- 27. Apply the standard form of the conjugate gradient method to the linear system

[1]	0	0		[1]	
0	2	0	x =	1	,
0	0	3		1	

starting as usual with $x^{(0)} = 0$. Verify that the residuals $r^{(0)}$, $r^{(1)}$ and $r^{(2)}$ are mutually orthogonal, that the search directions $d^{(0)}$, $d^{(1)}$ and $d^{(2)}$ are mutually conjugate, and that $x^{(3)}$ satisfies the equations.

- 28. Let the standard form of the conjugate gradient method be applied when *A* is positive definite. Express $\mathbf{d}^{(k)}$ in terms of $\mathbf{r}^{(i)}$ and $\beta^{(i)} > 0$, i = 0, 1, ..., k. Then deduce in a few lines from the formula $\mathbf{x}^{(k+1)} = \sum_{i=0}^{k} \omega^{(i)} \mathbf{d}^{(i)}$, from $\omega^{(i)} > 0$, and from the fact that $\mathbf{r}^{(i)}$ are orthogonal, that the sequence $\{ \| \mathbf{x}^{(k)} \| : k = 0, 1, ... \}$ increases monotonically.
- 29. The polynomial $p(x) = x^m + \sum_{i=0}^{m-1} c_i x^i$ is the *minimal polynomial* of the $n \times n$ matrix A if it is the polynomial of lowest degree that satisfies p(A) = 0. Note that $m \leq n$ holds because of the Cayley-Hamilton theorem.

Give an example of a 3×3 symmetric positive definite matrix with a quadratic minimal polynomial.

Prove that (in exact arithmetic) the conjugate gradient method requires at most m iterations to calculate the exact solution of Av = b, where m is the degree of the minimal polynomial of A.