

MATH 231 : Numerical ODEs

Random Examples

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Question 1: Eigenvalues of special tridiagonal matrices

This question is about finding eigenvalues of tridiagonal linear systems arising from applications, specifically finding the eigenvalues of an $n \times n$ matrix of the form,

$$A = \begin{pmatrix} a & b & & & \\ c & a & b & & \\ & \ddots & \ddots & \ddots & \\ & & c & a & b \\ & & & c & a \end{pmatrix}$$

where a, b, c are real numbers with $bc > 0$ (i.e. b and c have the same signs).

- (a) Show that the eigenvalue problem of A is equivalent to the equations

$$cv_{j-1} + (a - \lambda)v_j + bv_{j+1} = 0, \quad j = 1, \dots, n$$

$$v_0 = 0 = v_{n+1}$$

where $\mathbf{v} = (v_1, \dots, v_n)^T$ is an eigenvector of A associated with the eigenvalue λ .

- (b) The recurrence relation (1) is a second order linear difference equation and can be solved similar to second order linear differential equations. By guessing $v_j = r^j$ for some constant r , show that r satisfies

$$r_{\pm} = \frac{\lambda - a \pm \sqrt{(\lambda - a)^2 - 4bc}}{2b}, \quad \text{with} \quad r_+ r_- = \frac{c}{b}$$

- (c) Show by contradiction that r_{\pm} must be distinct.

Hint: if $r_{\pm} = r$ are repeated, then $v_j = Ar^j + Br^j$ for some constants A, B .

- (d) Since r_{\pm} are distinct, the general solution for (1) is $v_j = Ar_+^j + Br_-^j$ for constants A, B . Use this to conclude from (2) and (3) that,

$$\left(\frac{br_+^2}{c}\right)^{n+1} = 1$$

- (e) From part (c), (3) and (4), show that r_{\pm} must be complex valued and conclude that (4) has the solutions for $k = 1, \dots, n$,

$$r_{\pm, k} = \sqrt{\frac{c}{b}} \exp\left(\frac{\pm i k \pi}{n+1}\right), \quad \text{where } i = \sqrt{-1}$$

- (f) Using part (e), conclude that the eigenvalues of A is given by

$$\lambda_k = a + 2 \operatorname{sgn} \sqrt{bc} \cos\left(\frac{\pi k}{n+1}\right), \quad k = 1, \dots, n$$

- (g) Find the eigenvalues of the $n \times n$ finite difference matrix $A_h = \frac{1}{h^2} \begin{pmatrix} 2 & -1 & & & \\ -1 & 2 & \ddots & & \\ & \ddots & \ddots & \ddots & \\ & & -1 & 2 & -1 \\ & & & -1 & 2 \end{pmatrix}$, where

$$h = \frac{1}{n+1}.$$

Conclude that A_h is symmetric positive definite and find its condition number $\kappa(A_h)$ with respect to $\|\cdot\|_2$. Show that $\kappa(A_h) = \mathcal{O}(h^{-2})$ as number of grid points n increases. What does this mean for solving $A_h \mathbf{x} = \mathbf{b}$ when n is large?

Solution

- (a) Let (λ, \vec{v}) be an eigenpair of A

$$\begin{aligned} & \Rightarrow A\vec{v} = \lambda\vec{v} \\ & \Rightarrow (A - \lambda I)\vec{v} = \vec{0} \\ & \Rightarrow \begin{pmatrix} (a - \lambda)v_1 + bv_2 \\ cv_1 + (a - \lambda)v_2 + bv_3 \\ \vdots \\ cv_{n-2} + (a - \lambda)v_{n-1} + b_n \\ c_{n-1} + (a - \lambda)v_n \end{pmatrix} = \vec{0}. \end{aligned}$$

We can write the above relation as the following,

$$cv_{j-1} + (a - \lambda)v_j + bv_{j+1} = 0. \quad (1)$$

Where $0 \leq j \leq n + 1$ and $v_0 = 0 = v_{n+1}$ \ominus

- (b) Using the hint we guess the following form of the solution $v_j = r^j$. Substituting in 1,

$$\begin{aligned} cr^{j-1} + (a - \lambda)r^j + br^{j+1} &= 0 \\ c + (a - \lambda)r + br^2 &= 0 \\ &\cdot \end{aligned}$$

Using the quadratic formula, we get

$$r_{\pm} = \frac{\lambda - a \pm \sqrt{(a - \lambda)^2 - 4bc}}{2b}..$$

- (c) If 1 has a repeated root, say $r_{\pm} = r$, then solution to the recursion would look like,

$$v_j = Ar^j + Bjr^j.$$

Question 2: Classical iterative methods for strictly diagonally dominant matrices

- (a) Show that the diagonal part of any strictly diagonally dominant (S.D.D.) matrix is invertible.
 (b) Recall the Gershgorin's theorem below, which can give useful information about the eigenvalues of a matrix. The eigenvalues of a complex valued matrix A lies in the union of n discs $\bigcup_{i=1}^n D_i$ on the complex plane, where

$$D_i = \left\{ z \in \mathbb{C} : |z - a_{ii}| \leq \sum_{j \neq i} |a_{ij}| \right\}$$

Using Gershgorin's theorem, conclude S.D.D. matrices are invertible. Hint: Show that $0 \notin D_i$ for all $i = 1, \dots, n$. The next two parts are about showing convergence of Jacobi and Gauss-Seidel iterations for S.D.D. matrices.

- (c) Recall the matrix $-M^{-1}N$ associated with the Jacobi iteration takes the form $-D^{-1}(L + U)$, where $A = L + D + U$.
 (d) Let A be S.D.D. and λ be any eigenvalue of $-D^{-1}(L + U)$. Show that $\det(L + U + \lambda D) = 0$ using part (a). (ii) Now suppose $|\lambda| \geq 1$. Deduce from A being S.D.D. that $L + U + \lambda D$ must also be S.D.D. (iii) Deduce a contradiction by applying the result from part (b) to $L + U + \lambda D$, and conclude that $|\lambda| < 1$. (iv) Combine parts (i)-(iii) to conclude that Jacobi iteration converges for S.D.D. matrices.

- (e) Follow a similar argument as part (c) to show that Gauss-Seidel iterations converges for S.D.D. matrices.

Question 3: Classical iterative methods for symmetric positive definite matrices

This question is about coding and comparing classical iterative methods for the S.P.D. matrix A_h from Q1(g).

- (a) Write a pseudocode for the classical iterative methods: Richardson, optimal Richardson, Jacobi, Gauss-Seidel, S.O.R., and optimal S.O.R.
- (b) Implement a program to solve $A_h \mathbf{x} = \mathbf{b}$ with $\mathbf{b} = (1, \dots, 1)^T \in \mathbb{R}^{20}$ and $\mathbf{x}_0 = (1, 0, \dots, 0)^T \in \mathbb{R}^{20}$ using Richardson (with $\omega = \lambda_{\max}^{-1}$), optimal Richardson, Jacobi, Gauss-Seidel, S.O.R. (with $\theta = 1.2$) and optimal S.O.R. Generate a plot comparing the log of their residual in ℓ_2 norm versus iterations up to 5000. Rank the performance of each method by comparing the iterations needed to reach the residual tolerance of 10^{-14} . Use sparse representation when appropriate. Hint: Use Q1(g) to find parameters for Richardson and vary θ to find an approximate optimal parameter for S.O.R.
- (c) Comment on the decreases in performance when $n = 1000$. Explain briefly how this relates to $\kappa(A_h) = \mathcal{O}(h^{-2})$.

Question 4: Steepest Descent and Conjugate Gradient

- (a) Let A be a S.P.D. matrix. Show that $(\mathbf{x}, \mathbf{y})_A := \mathbf{x}^T A \mathbf{y}$ for $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ forms an inner product on \mathbb{R}^n .
- (b) Using part (a), conclude that $\|\mathbf{x}\| = (\mathbf{x}, \mathbf{x})_A^{1/2}$ for $\mathbf{x} \in \mathbb{R}^n$ is a norm on \mathbb{R}^n .
Hint: You can assume the Cauchy-Schwarz inequality $|\mathbf{x}, \mathbf{y})_A| \leq \|\mathbf{x}\|_A \|\mathbf{y}\|_A$ holds.
- (c) For the method of Steepest Descent, show that $\nabla f(\mathbf{x}_k)$ and $\nabla f(\mathbf{x}_{k+1})$ are orthogonal (i.e. zig-zaging behavior), where $f(\mathbf{y}) = \frac{1}{2} \mathbf{y}^T A \mathbf{y} - \mathbf{y}^T \mathbf{b}$. Hint: Recall how the step size for Steepest Descent is determined.
- (d) Repeat the experiment from Q3(b) with $\mathbf{b} = (1, \dots, 1)^T \in \mathbb{R}^{1000}$ and $\mathbf{x}_0 = (1, 0, \dots, 0)^T \in \mathbb{R}^{1000}$ using the method of Steepest Descent and Conjugate Gradient. Generate a plot comparing the log of their residual in ℓ_2 norm versus iterations up to 5000. Rank their performance by comparing the iterations needed to reach the residual tolerance of 10^{-14} , as well as versus the classical iterative methods. Verify your CG method terminates after the desired number of iterations. Use sparse representation when appropriate.