

Numerical Analysis Fall 2015
MS/PhD Qualifying Examination

1. Let $A \in R^{n \times n}$ be an invertible, symmetric positive definite matrix, $b \in R^n$. This problem regards the method of steepest descent to find the solution \mathbf{x}^* of $A\mathbf{x} = \mathbf{b}$. Steepest descent is an iterative method that defines a sequence \mathbf{x}_n which converges to the minimizer of the function

$$\phi(x) = \frac{1}{2} \mathbf{x}^T A \mathbf{x} - \mathbf{x}^T \mathbf{b}.$$

- (a) Let $e(\mathbf{x}) = \mathbf{x} - \mathbf{x}^*$ and $\|x\|_A = \sqrt{x^T A x}$. where $\mathbf{x}^* = A^{-1} \mathbf{b}$ solves $A\mathbf{x} = \mathbf{b}$. Prove that \mathbf{x} minimizes $\phi(\mathbf{x})$ if and only if \mathbf{x} minimizes $\|e(\mathbf{x})\|_A$, and thus $\mathbf{x} = \mathbf{x}^*$, unique.
(b) Derive a formula for $-\nabla \phi$.
(c) The vector $-\nabla \phi(\mathbf{x})$ points in the direction of steepest descent of ϕ at \mathbf{x} . The method of steepest descent consists of iterating

$$\mathbf{x}_{n+1} = \mathbf{x}_n - \alpha_n \nabla \phi(\mathbf{x}_n)$$

starting from an initial guess \mathbf{x}_0 . That is, one steps from \mathbf{x}_n to \mathbf{x}_{n+1} by moving along the direction of steepest descent. Determine the optimal step length α_n that minimizes $\phi(\mathbf{x}_{n+1})$. Explain why the method always converges to the minimizer \mathbf{x}^* of ϕ .

- (d) Write down an algorithm for the full steepest descent iteration. There are three operations inside the main loop.
2. Consider the least squares problem $\min_{\mathbf{x}} \|A\mathbf{x} - \mathbf{b}\|_2$ with $A \in R^{m \times n}$, $m > n$, and $\mathbf{b} \in R^m$.
- (a) Assume A has full rank. Describe how to solve the least squares problem using (i) the normal equations, (ii) the QR factorization, and (iii) the SVD decomposition. Discuss the condition number and the number of operations for each method.
(b) Discuss what happens when the rank of A is less than n .

3. Determine the rate of convergence of the Rayleigh quotient $\mathbf{r}(v_k) = \mathbf{v}_k^T A \mathbf{v}_k$, to an eigenvalue of $A \in R^{n \times n}$, $A = A^T$, with vectors $\mathbf{v}_k \in R^n$ given by the normalized power method $\mathbf{v}_{k+1} = A\mathbf{v}_k / \|A\mathbf{v}_k\|$.
4. Let $f \in C_{[0,1]}$ be given, and u solve

$$-u_{xx} = f, \quad x \in [0, 1], \quad u(0) = u(1) = 0.$$

For given N , an approximation v_i of $u(x_i)$, $x_i = ih$, $h = 1/(N+1)$ is given by

$$-v_{i+1} + 2v_i - v_{i-1} = h^2 f_i, \quad i = 1, \dots, N, \quad h = 1/(N+1), \quad (1a)$$

$$v_0 = v_{N+1} = 0 \quad (1b)$$

where $f_i = f(x_i)$. Equations (1ab) form a linear system in v_i which can be written as

$$T_N[v_1, \dots, v_N]^T = h^2[f_1, \dots, f_N]^T.$$

- (a) Show that the eigenvalues and eigenvectors of T_N are $\lambda_j = 2(1 - \cos(\frac{\pi j}{N+1}))$ and \mathbf{z}_j , with components $z_j^k = \sqrt{\frac{2}{N+1}} \sin(\frac{\pi j k}{N+1})$.

- (b) Let $Z = [\mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_N]$, and $T_N = Z\Lambda Z^T$, and consider the equivalent problem in 2D,

$$u_{xx} + u_{yy} = f(x, y)$$

where $(x, y) \in D = [0, 1] \times [0, 1]$, and

$$u = 0 \quad \text{on} \quad \partial D,$$

with the approximation $v_{i,j}$ of $u(x_i, y_j)$ given by

$$-v_{i+1,j} - v_{i-1,j} - v_{i,j+1} - v_{i,j-1} + 4v_{i,j} = h^2 f_{i,j}, \quad (2a)$$

$$v_{i,0} = v_{i,N+1} = v_{0,j} = v_{N+1,j} = 0, \quad i, j = 1, \dots, N, \quad (2b)$$

where $f_{i,j} = f(x_i, y_j)$, $x_i = ih$, $y_j = jh$, $h = 1/(N+1)$. Show that equations (2ab) can be written as a matrix equation

$$T_N V + V T_N = h^2 F. \quad (3)$$

Here $V_{i,j} = v_{i,j}$ and $F_{i,j} = f_{i,j}$.

- (c) Let $V' = Z^T V Z$. Show that the solution V of equation (3) can be found by the following steps

$$F' = Z^T F Z, \quad (4a)$$

$$v'_{j,k} = \frac{h^2 f'_{j,k}}{\lambda_j + \lambda_k}, \quad (4b)$$

$$V = Z V' V^T. \quad (4c)$$

- (d) To leading order in N , how many operations does it require to solve (4)?
(e) Outline a method to solve (3) with complexity $O(N^2 \log N)$.

5. Let

$$A = \begin{pmatrix} 1 & 1 \\ 2 & 2 \\ 0 & 0 \end{pmatrix}$$

- (a) Find the reduced and a full singular value decomposition of A .
(b) Find the best rank-1 approximation B of A in the 2-norm.
(c) Find $\|A - B\|_2$.