

Numerical Analysis Screening Exam, Spring 2018

This exam has 4 problems. Please write down your names on all pages of your answer and mark clearly the order of pages. Start a new problem on a new page.

I. Direct Methods

Consider the matrix

$$A = \left[\begin{array}{c|c} R & v \\ \hline u^T & 0 \end{array} \right]$$

where R is an $n \times n$ invertible upper triangular matrix and u and v are vectors in \mathbb{R}^n .

- Consider the LU decomposition of A , $A = LU$, where L is unit lower triangular and U is upper triangular. Find the matrices L and U .
- Find necessary and sufficient conditions on R , u and v for A to be non-singular.
- Suppose A is non-singular. Use the LU-decomposition from (a) to formulate an economical algorithm for solving the system of equations $Ax = b$ and determine its computational cost.

II. Iterative Methods

Consider the system of equations $Ax = b$ where

$$A = D - E - F$$

is an $n \times n$ Hermitian positive definite matrix, D is diagonal, $-E$ is strictly lower triangular and $-F$ is strictly upper triangular. (In other words, both E and F have zeros on the diagonal.) To solve the system using the Gauss-Seidel method with successive over-relaxation (SOR) we write $A = M - N$ where

$$M = \frac{1}{\omega}D - E$$

and

$$N = \left(\frac{1}{\omega} - 1 \right) D + F$$

where ω is a real number. The iteration matrix is

$$T_\omega = M^{-1}N = (D - \omega E)^{-1} [(1 - \omega)D + \omega F].$$

- For any matrix B let B^* be its conjugate transpose. Show that if $|\omega - 1| < 1$ then $M^* + N$ is positive definite.
- Consider the vector norm $\|\cdot\|_A$ be defined by

$$\|x\|_A = \sqrt{x^*Ax}.$$

and the corresponding matrix norm $\|\cdot\|_A$ it induces on $\mathbb{C}^{n \times n}$. Show that if $M^* + N$ is positive definite then $\|M^{-1}N\|_A < 1$.

- What can you deduce about the convergence of the SOR method in the case when $|\omega - 1| < 1$?
- Suppose the SOR method converges for all initial vectors x_0 . Show that $|\omega - 1| < 1$. (*Hint: compute $\det(T_\omega)$.*)

IV. Least Square Problem

Let $A \in \mathbb{R}^{m \times n}$ be a matrix and let $A = U\Sigma V^T$ be its singular value decomposition. We want to find k vectors $x_i \in \mathbb{R}^m$ and $y_i \in \mathbb{R}^n$ such that

$$\left\| A - \sum_{i=1}^k x_i y_i^T \right\|_F^2$$

is minimal, where $\|A\|_F^2 := \sum_{i=1}^m \sum_{j=1}^n a_{ij}^2$ is the Frobenius norm.

- (a) Let u_i and v_j the i -th and j -th column of U and V respectively and σ_r the r -th singular value of A . Show that

$$x_i = \sqrt{\sigma_i} u_i, \quad y_i = \sqrt{\sigma_i} v_i$$

is a solution of the optimization problem .

Hint: For any orthogonal matrix O and S we have $\|OAS\|_F = \|A\|_F$.

- (b) Assume that $m = n$ is large and k small. How many values do you have to store for A and how many for its approximation $\sum_{i=1}^k x_i y_i^T$?
- (c) Suppose that the matrix A represents a database of rankings of n movies by m movie viewers such that $a_{ij} \in \{0, \dots, 5\}$ is the star ranking of movie j by viewer i . Is it reasonable that A can be well approximated by some $\sum_{i=1}^k x_i y_i^T$ for some comparatively small k ?

Hint: Discuss the case that there are k groups of users with similar interests and therefore similar movie rankings.