

# NUMERICAL ANALYSIS QUALIFYING EXAM

Each Problem Counts 25 Points

①

January 1994

Name: \_\_\_\_\_

1. (i) State the *fundamental theorem of linear algebra*.
- (ii) Find the *LU* decomposition of the following matrix  $A$ .
- (iii) Find bases for each of the four fundamental subspaces associated with  $A$  (that is,  $\mathcal{R}(A)$ ,  $\mathcal{R}(A^T)$ ,  $\mathcal{N}(A)$ , and  $\mathcal{N}(A^T)$ ), and state the dimension of these subspaces.

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

2. (i) Describe the *singular value decomposition* of an  $m \times n$  matrix  $A$ . Define all matrices that you introduce.
- (ii) For the following matrix  $A$  and vector  $b$ , find the singular value decomposition, the pseudoinverse  $A^+$ , and the minimum length least squares solution  $x^+$  of  $Ax = b$ .

$$A = \begin{bmatrix} \sqrt{3} & 0 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}, \quad b = \begin{bmatrix} 4 \\ 3 \\ 8 \end{bmatrix}.$$

3. A real square matrix  $A$  is called positive definite symmetric (PDS) if it is symmetric,  $A = A^T$ , and, for any  $x \neq 0$ :

$$x^T A x > 0.$$

- (i) Show that all eigenvalues of a PDS matrix are real and positive.
- (ii) Let  $A^{(m)}$  be the  $m \times m$  matrix obtained by intersecting the first  $m$  rows and columns of a PDS matrix,  $A$ . Show that  $A^{(m)}$  is also PDS.
- (iii) Use the results above to show that if Gaussian elimination without pivoting is applied to a PDS matrix, only positive pivots are encountered. (Hint: consider the relationship between the pivots, the determinant, and the eigenvalues.)
- (iv) Use (iii) to prove the existence of the Cholesky decomposition of a PDS matrix:  $A = LL^T$  where  $L$  is lower triangular.

4. The power method for computing an eigenvalue of a matrix,  $A$ , is defined by:

$$y_{n+1} = Ax_n, \quad \lambda_{n+1} = y_{n+1}^T x_n, \quad x_{n+1} = \frac{y_{n+1}}{\sqrt{y_{n+1}^T y_{n+1}}},$$

where  $x_0$  satisfying  $x_0^T x_0 = 1$  is otherwise arbitrary.

- (i) Show that if there is a single extreme eigenvalue, that is a simple eigenvalue  $\lambda$  such that  $|\alpha| < |\lambda|$  for all other eigenvalues  $\alpha$ , then the power method converges, that is  $\lambda_n \rightarrow \lambda$ , for most  $x_0$ . (To simplify your arguments, assume that  $A$  is diagonalizable.)
- (ii) Describe the typical behavior of the method if the extreme eigenvalues correspond to a conjugate imaginary pair. In particular show that the sequence  $\lambda_n$  may converge to a number which is not an eigenvalue. (An example will do.)
- (iii) Let  $A$  be a real skew-symmetric matrix, i.e.  $A = -A^T$ . Show that all eigenvalues of  $A$  are imaginary. What can you say about the eigenvalues of  $A^2$ ?
- (iv) Suggest a modification of the power method to compute extreme eigenvalues of a skew-symmetric matrix.