

# Study Problems 2 Foundations of Computational Math 1 Fall 2024

All of these problems are based on the material in the notes on  $LU$  factorization and lectures. They are all of the basic type that you should be able to reproduce after studying the solutions. They include a derivation of Gauss-Jordan factorization based on elementary transformations related to Gauss Transformations that is close to the algorithm often taught as a way of computing Reduced Row Echelon Form. Also discussed is the use of pivoting for efficiency rather than numerical stability or existence of the factorization; an incremental or bordering form of producing a factorization based on partitioning; growth of elements in the factorization; and a key additional algebraic assumption, symmetric positive definiteness of  $A$ .

## Problem 2.1

Suppose you are computing a factorization of the  $A \in \mathbb{C}^{n \times n}$  with partial pivoting and at the beginning of step  $i$  of the algorithm you encounter the transformed matrix with the form

$$T^{-1}A = A^{(i-1)} = \begin{pmatrix} U_{11} & U_{12} \\ 0 & A_{i-1} \end{pmatrix}$$

where  $U_{11} \in \mathbb{R}^{i-1 \times i-1}$  and nonsingular, and  $U_{12} \in \mathbb{R}^{i-1 \times n-i+1}$  contain the first  $i-1$  rows of  $U$ . Show that if the first column of  $A_{i-1}$  is all 0 then  $A$  must be a singular matrix.

## Problem 2.2

Suppose you have the LU factorization of an  $i \times i$  matrix  $A_i = L_i U_i$  and suppose the matrix  $A_{i+1}$  is an  $(i+1) \times (i+1)$  matrix formed by adding a row and column to  $A_i$ , i.e.,

$$A_{i+1} = \begin{pmatrix} A_i & a_{i+1} \\ b_{i+1}^T & \alpha_{i+1,i+1} \end{pmatrix}$$

where  $a_{i+1}$  and  $b_{i+1}$  are vectors in  $\mathbb{R}^i$  and  $\alpha_{i+1,i+1}$  is a scalar.

**2.2.a.** Derive an algorithm that, given  $L_i$ ,  $U_i$  and the new row and column information, computes the LU factorization of  $A_{i+1}$  **and identify the conditions under which the step will fail.**

**2.2.b.** What computational primitives are involved?

**2.2.c.** Show how this basic step could be used to form an algorithm that computes the LU factorization of an  $n \times n$  matrix  $A$ .

## Problem 2.3

Suppose that  $A \in \mathbb{R}^{n \times n}$  is nonsingular and that  $A = LU$  is its  $LU$  factorization. Give an algorithm that can compute,  $e_i^T A^{-1} e_j$ , i.e., the  $(i, j)$  element of  $A^{-1}$  in approximately  $(n - j)^2 + (n - i)^2$  operations.

## Problem 2.4

(Restated Golub and Van Loan 3rd Ed. p. 103 Problem P3.2.5.)

Define the elementary matrix  $N_k^{-1} = I - y_k e_k^T \in \mathbb{R}^{n \times n}$ , where  $1 \leq k \leq n$  is an integer,  $y_k \in \mathbb{R}^n$  and  $e_k \in \mathbb{R}^n$  is the  $k$ -th standard basis vector.  $N_k^{-1}$  is a Gauss-Jordan transform if it is defined by requiring  $N_k^{-1} v = e_k \nu_k$  for a particular given vector  $v \in \mathbb{R}^n$  whose elements are denoted  $\nu_j = e_j^T v$ . For example, if  $n = 6$  and  $k = 3$  then

$$N_3^{-1} = \begin{pmatrix} 1 & 0 & * & 0 & 0 & 0 \\ 0 & 1 & * & 0 & 0 & 0 \\ 0 & 0 & * & 0 & 0 & 0 \\ 0 & 0 & * & 1 & 0 & 0 \\ 0 & 0 & * & 0 & 1 & 0 \\ 0 & 0 & * & 0 & 0 & 1 \end{pmatrix} \quad \text{and} \quad v = \begin{pmatrix} \nu_1 \\ \nu_2 \\ \nu_3 \\ \nu_4 \\ \nu_5 \\ \nu_6 \end{pmatrix}$$

where  $*$  indicates a value that must be determined.

**(2.4.a)** Determine how to choose  $y_k$  and define  $N_k^{-1}$  given a vector  $v \in \mathbb{R}^n$ , i.e., determine the values of the elements of  $y_k$  in terms of the values of the elements of  $v$  so that  $N_k^{-1} v = e_k \nu_k$ . For  $n = 6$  and  $k = 3$  then

$$N_3^{-1} v = \begin{pmatrix} 1 & 0 & * & 0 & 0 & 0 \\ 0 & 1 & * & 0 & 0 & 0 \\ 0 & 0 & * & 0 & 0 & 0 \\ 0 & 0 & * & 1 & 0 & 0 \\ 0 & 0 & * & 0 & 1 & 0 \\ 0 & 0 & * & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \nu_1 \\ \nu_2 \\ \nu_3 \\ \nu_4 \\ \nu_5 \\ \nu_6 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \nu_3 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

**(2.4.b)** Determine when  $N_k^{-1}$  exists and is nonsingular.

**(2.4.c)** Show how a series of  $N_k^{-1}$  can be used to transform a nonsingular matrix  $A \in \mathbb{R}^{n \times n}$  into a nonsingular diagonal matrix  $D \in \mathbb{R}^{n \times n}$ , i.e., all of the off-diagonal elements of  $D$  are 0 and all of the diagonal elements are nonzero. You may assume that  $A$  is such that all of the  $N_k^{-1}$  exist.

**(2.4.d)** Does the factorization that this transformation induces have any structure other than that in  $D$ ?

## Problem 2.5

Consider an  $n \times n$  real matrix where

- $\alpha_{ij} = e_i^T A e_j = -1$  when  $i > j$ , i.e., all elements strictly below the diagonal are  $-1$ ;
- $\alpha_{ii} = e_i^T A e_i = 1$ , i.e., all elements on the diagonal are  $1$ ;
- $\alpha_{in} = e_i^T A e_n = 1$ , i.e., all elements in the last column of the matrix are  $1$ ;
- all other elements are  $0$

For  $n = 4$  we have

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

**2.5.a.** Compute the factorization  $A = LU$  for  $n = 4$  where  $L$  is unit lower triangular and  $U$  is upper triangular.

**2.5.b.** What is the pattern of element values in  $L$  and  $U$  for any  $n$ ?

## Problem 2.6

Suppose  $A \in \mathbb{R}^{n \times n}$  is a nonsymmetric nonsingular matrix with the following nonzero pattern (shown for  $n = 6$ )

$$\begin{pmatrix} * & * & * & * & * & * \\ * & * & 0 & 0 & 0 & 0 \\ * & 0 & * & 0 & 0 & 0 \\ * & 0 & 0 & * & 0 & 0 \\ * & 0 & 0 & 0 & * & 0 \\ * & 0 & 0 & 0 & 0 & * \end{pmatrix}$$

Suppose  $A$  has an  $LU$  factorization that can be computed without partial or complete pivoting while being numerical reliable.

**2.6.a.** Suppose  $A = LU$  is computed without any pivoting for this pattern of nonzero elements. Given that the number of operations in the algorithm is of the form  $Cn^k + O(n^{k-1})$ , where  $C$  is a constant independent of  $n$  and  $k > 0$ , what are  $C$  and  $k$ ?

**2.6.b.** Now suppose you use pivoting to reduce complexity. Describe an algorithm that computes a factorization of a permuted  $A$  and then is used to solve  $Ax = b$  as efficiently as possible.

- 2.6.c. Given that the number of operations in the algorithm is of the form  $Cn^k + O(n^{k-1})$ , where  $C$  is a constant independent of  $n$  and  $k > 0$ , what are  $C$  and  $k$  for the efficient algorithm? Do you expect a significant reduction in computational time compared to the first algorithm that does not pivot?

## Problem 2.7

Consider a symmetric matrix  $A$ , i.e.,  $A = A^T$ .

- 2.7.a. Consider the use of Gauss transforms to factor  $A = LU$  where  $L$  is unit lower triangular and  $U$  is upper triangular. **You may assume that the factorization does not fail.** Show that  $A = LDL^T$  where  $L$  is unit lower triangular and  $D$  is a matrix with nonzeros on the main diagonal. i.e., elements in positions  $(i, i)$ , and zero everywhere else, by demonstrating that  $L$  and  $D$  can be computed by applying Gauss transforms appropriately to the matrix  $A$ .
- 2.7.b. For an arbitrary symmetric matrix the  $LDL^T$  factorization will not always exist due to the possibility of 0 in the  $(i, i)$  position of the transformed matrix that defines the  $i$ -th Gauss transform. Suppose, however, that  $A$  is a **positive definite** symmetric matrix, i.e.,  $x^T Ax > 0$  for any vector  $x \neq 0$ . Show that the diagonal element of the transformed matrix  $A$  that is used to define the vector  $l_i$  that determines the Gauss transform on step  $i$ ,  $M_i^{-1} = I - l_i e_i^T$ , is always positive and therefore the factorization will not fail. Combine this with the existence of the  $LDL^T$  factorization to show that, in this case, the nonzero elements of  $D$  are in fact positive.