

SYSTEMS OF LINEAR EQUATIONS

EXERCISE 7

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# Systems of Linear Equations Report

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# 1 Introduction

In this exercise I am going to use solve a linear system  $A\vec{x} = \vec{y}$  using LU decomposition. The goal is to verify that first element of the unknowns vector,  $x_1$ , will contain an approximation of  $e - 2$ .

As we've seen in class, there are multiple ways of solving a linear system  $AX = B$ . Assume  $A$  is a  $n \times n$  square matrix,  $B$  is a "constant" term matrix  $n \times h$ , and  $X$  is a  $n \times h$  unknown matrix. To solve for  $X$ , we could compute the inverse of  $A$  and find  $x = A^{-1}y$ . We've seen that this approach, however, requires more computations than necessary and returns a less accurate result.

On the other hand, the LU decomposition technique is a way to represent the matrix  $A$  in the form of simpler matrices,  $L$  and  $U$  (lower triangular and upper triangular matrices, respectively):

$$PA = LU$$

This method uses forward substitution (solving for  $Y$  from  $LY = B$ ) and backward substitution (solving for  $X$  from  $UX = Y$ ). I'll be specifically be solving the system by using Gaussian Elimination *with partial pivoting*, which reduces round-offs errors compared to its naive implementation, and then provide observations on the accuracy of the results obtained.

## 2 Tools

The following programming language and libraries have been used in this exercise:

- C
- GSL (GNU Scientific Library)

The following double-precision GSL data types have been used in the exercise:

- `gsl_vector`
- `gsl_matrix`
- `gsl_permutation`

The following GSL methods have been used in the exercise:

- `gsl_matrix_alloc(size1, size2)`
- `gsl_matrix_set_zero(matrix)`
- `gsl_matrix_set(matrix, row, column, value)`
- `gsl_matrix_get(matrix, row, column)`
- `gsl_vector_alloc(size)`
- `gsl_vector_set_zero(vector)`
- `gsl_vector_set(vector, index, value)`
- `gsl_vector_get(vector, index)`
- `gsl_matrix_memcpy(matrixToCopyFrom, matrix)`
- `gsl_linalg_SV_decomp(A, V, S, workspaceVector)`

In order to factorize a matrix into the LU decomposition, and then solve the square system  $Ax = y$  using the decomposition of A, I've used the following methods:

- `gsl_linalg_LU_decomp(A, permutation, signum)`
- `gsl_linalg_LU_solve(LU, permutation, b, x)`
- `gsl_permutation_alloc(size)`

### 3 Solving the linear system

In order to solve the system  $Ax = y$ , I first need to build the matrix  $A$  by understanding how it's build. The requirements are to build a tridiagonal matrix with the values  $-1$  on the adjacent upper diagonal, the entries  $+1$  on the adjacent lower diagonal, and on the main diagonal the values  $b_i$ , with  $i = 1, \dots, n$  given by

$$b_i = \frac{2(i+1)}{3}, \quad i+1 = 3, 6, 9, \dots$$

$$b_i = 1, \quad i+1 = 2, 4, 5, 7, 8, \dots$$

By looking closely at the first rule, we see that the  $i+1$  are all multiples of 3 ( $i+1 = 3 * k$ , for some  $k$ ). Hence the  $i$  are of the form  $i = 3 * k - 1$ , for some  $k$ . For  $n = 5$ , for example, this is what the matrix looks like:

$$\begin{bmatrix} 1.0000000000 & -1.0000000000 & 0.0000000000 & 0.0000000000 & 0.0000000000 \\ 1.0000000000 & 2.0000000000 & -1.0000000000 & 0.0000000000 & 0.0000000000 \\ 0.0000000000 & 1.0000000000 & 1.0000000000 & -1.0000000000 & 0.0000000000 \\ 0.0000000000 & 0.0000000000 & 1.0000000000 & 1.0000000000 & -1.0000000000 \\ 0.0000000000 & 0.0000000000 & 0.0000000000 & 1.0000000000 & 4.0000000000 \end{bmatrix}$$

The coefficients matrix  $A$  is first allocated by using the `gsl_matrix_alloc` method, then I set all the elements to zero with `gsl_matrix_set_zero` and finally nested `for` loops fill the diagonal values by checking the indexes. The coefficients reported above on the diagonal have 5 significant digits for improve the readability of this report.

I used the `gsl_vector_alloc` method to create an instance of the vector. All of its elements were set to zero by using `gsl_vector_set_zero(vector)`. The exercise asks us to set the first element of the  $y$  vector to one, so I used `gsl_vector_set(vector, 0, 1)` to assign the value 1 to index 0. For  $n = 5$ , we have:

$$\vec{y} = \begin{bmatrix} 1.0000000000 \\ 0.0000000000 \\ 0.0000000000 \\ 0.0000000000 \\ 0.0000000000 \end{bmatrix}$$

Given the  $Ax = y$  system, my goal is now to find the vector of the unknowns  $x$ . To do so, I first factorize  $A$  into its LU decomposition by allocating a new matrix (so that the matrix which represents  $A$  doesn't get overridden) using `gsl_matrix_memcpy` and then by calling `gsl_linalg_LU_decomp`. This method utilizes Gaussian Elimination with partial pivoting to compute the decomposition. The following is the  $LU$  matrix for  $n = 5$ :

$$LU = \begin{bmatrix} 1.0000000000 & -1.0000000000 & 0.0000000000 & 0.0000000000 & 0.0000000000 \\ 1.0000000000 & 3.0000000000 & -1.0000000000 & 0.0000000000 & 0.0000000000 \\ 0.0000000000 & 0.3333333333 & 1.3333333333 & -1.0000000000 & 0.0000000000 \\ 0.0000000000 & 0.0000000000 & 0.7500000000 & 1.7500000000 & -1.0000000000 \\ 0.0000000000 & 0.0000000000 & 0.0000000000 & 0.5714285714 & 4.5714285714 \end{bmatrix}$$

I can now use the  $LU$  matrix to solve the system by passing  $LU$ ,  $x$ , a permutation structure `gsl_permutation` and  $y$  to `gsl_linalg_LU_solve`. This method uses forward and back-substitution to modify the contents of the  $x$  vector given in input, which now looks like this (for  $n = 5$ ):

$$\vec{x} = \begin{bmatrix} 0.7187500000 \\ -0.2812500000 \\ 0.1562500000 \\ -0.1250000000 \\ 0.0312500000 \end{bmatrix}$$

The first element looks contains an approximation of  $e - 2$ . By increasing the size of the matrix  $n$ ,  $x_i$  becomes increasingly more precise. For  $n = 10$ , for example:

$$\vec{x} = \begin{bmatrix} 0.7182817183 \\ -0.2817182817 \\ 0.1548451548 \\ -0.1268731269 \\ 0.0279720280 \\ -0.0149850150 \\ 0.0129870130 \\ -0.0019980020 \\ 0.0009990010 \\ -0.0009990010 \end{bmatrix}$$

For  $n = 20$ :

$$\vec{y} = \begin{bmatrix} 0.7182818285 \\ -0.2817181715 \\ 0.1548454854 \\ -0.1268726862 \\ 0.0279727992 \\ -0.0149814893 \\ 0.0129913099 \\ -0.0019901794 \\ 0.0010502335 \\ -0.0009399460 \\ 0.0001102875 \\ -0.0000576459 \\ 0.0000526416 \\ -0.0000050043 \\ 0.0000025983 \\ -0.0000024061 \\ 0.0000001922 \\ -0.0000000994 \\ 0.0000000928 \\ -0.0000000066 \end{bmatrix}$$

Then, I calculate the condition number to say if this is a well-conditioned or an ill-conditioned linear system. In GSL there is no direct function that calculates the condition number, but it's possible to use the ratio of the largest singular value of matrix  $A$ ,  $\sigma_n(A)$ , to the smallest  $\sigma_1(A)$ :

$$\kappa(A) := \frac{\sigma_n(A)}{\sigma_1(A)} = \frac{\|A\|}{\|A^{-1}\|^{-1}}$$

I proceed to factorize  $A$  into its singular value decomposition  $SVD$  using the `gsl_linalg_SV_decomp` method. For  $n = 5$ , the condition number is

NOT SURE. Iterative refinement, a technique used to improve the approximate solution  $\vec{x}$  of our system, could be helpful to improve these results. GSL

## 4 Observations