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Code of this assignment are in the attachment, and they can also be found in GitHub: https://github.com/Orcuslc/Learning/tree/master/Numerical%20Algorithm%20and%20Case%20Studies/Homework2/code

Problem 1 & 2.

Abstract.

Details of the implementation of LU decomposition in MATLAB.

Introduction.

In most cases pivoting is necessary. If we randomly choose a matrix $M \in \mathbb{R}^{n \times n}$, the probability that $M(1,1) = \max_i M(i,1)$ is $\frac{1}{n}$. So the efficiency of pivoting is a great concern.

In the experiment below, we generate the matrix M by using MATLAB function $\mathrm{randn}(n, n)$; and we use separately

$$[L, U] = lu(A)$$
$$[L, U, P] = lu(A)$$

for the two problems.

Results & Discussion. After trying for several times with different n, the dimension of the random matrix, we found that in most cases, calling [L, U] = lu(A) leads to L not being a lower triangular matrix, but U is always an upper triangular matrix.

When we count the number of non-zero elements of each row in L, we can find out the list of the numbers is a permutation of [1, 2, ..., n].

Thus I tried to call [L1, U1, P] = lu(A), and the P is exactly the permutation matrix:

$$PL1 = L$$

. So it is obvious how MATLAB express the P: it simply multiplied it with the L; In order to represent this, I guess that MATLAB records the number of lines that need interchanging in each step; when the algorithm ends, it conbines the records to gain a permutation matrix.

In order to realize P*b, we should use the pernutation record: in each of the n steps, change the elements of two lines, and it will only need O(n) time, and no extra space.

Problem 3.

Abstract & Introduction. As is known to all, using Cholesky decomposition for a symmetric positive definite matrix will need half of the flops comparing LU decomposition; In the mean time, since there is no need of pivoting for a symmetric positive definite matrix, in each step n-1 times of comparing and 2n times of exchanging will be saved on average.

Result. We generate the symmetric matrix A by

$$A = randn(n, n); A = A + A';$$

and we did the experiment with n = 4, 6, 8, 10. The result of calling

$$L = chol(A)$$

Error using chol: Matrix must be positive definite.

Analysis. If $A \in \mathbb{R}^{n \times n}$ is not positive definite, then according to the properties of symmetric elementary transformations, PAP^{\top} is still not positive definite, where P is a permutation matrix.

Thus, we can assume A(1:2,1:2) is not positive definite. According to the algorithm, we have

$$l_{11} = \sqrt{a_{11}},$$

$$l_{21} = \frac{a_{21}}{l_{11}},$$

$$l_{22} = (a_{22} - l_{21}^2)^{1/2}.$$

That means $l_{22}=a_{22}-\frac{a_{21}^2}{a_{11}}=\frac{a_{22}a_{11}-a_{21}^2}{a_{11}}$. Since A is symmetric and not positive definite, we have $a_{22}a_{11}-a_{21}^2=a_{22}a_{11}-a_{21}a_{12}<0$. So l_{22} does not exist.

Problem 4.

Introduction. The LDL^{\top} decomposition is used for Hermitian indefinite matrices.

Discussion. I found an introduction to LDL^{\top} decomposition from Golub's *Matrix Computations*, Page 186. It said the method was developed by Bunch and Parlett. I searched their paper in Google Scholar, but it seemed that our school has no access to siam.org, so I cannot read the paper.

Reference

■ Gene H. Golub *et al*, *Matrix Computations*, Page 186-187.