2020年3月25日 上午 01:17 Let of A. positive definite (SPD). L is bi-diagonal.

Due date: 29 March, Sunday $\mathbf{A} = \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix}$ Find the LU decompsotion with partial pivoting of A.

MATH3322 Matrix Computation

Homework 2

Instead of the LU decomposition, we can also use a UL decomposition to solve the system of linear equations. In particular, given $A \in \mathbb{R}^{n \times n}$, we decompose A = UL, where $U \in \mathbb{R}^{n \times n}$ is unit upper triangular and $U \in \mathbb{R}^{n \times n}$ is lower triangular. Propose an algorithm for computing the UL decomposition 3. Let $A \in \mathbb{R}^{n \times n}$ be a tridiagonal matrix, i.e., $a_{ij} = 0$ if |i - j| > 1. We also assume that A is symmetric

(a) Prove that the Cholesky decomposition $A = LL^T$ satisfies $l_{ij} = 0$ for all i - j > 1. In other words,

(b) Propose an O(n) algorithm for computing the Cholesky decomposition of A. What is the number of operations needed of your algorithm? Your answer should be in the form of Cn + O(1) with explicit constant C. (c) Based on the Cholesky decomposition, construct an O(n) algorithm to solve Ax = b. Express the number of operations needed in the form of Cn + O(1) with explicit C. 4. We consider a discrete 1-D Laplacian equation Ax = b, where $\boldsymbol{A} = \begin{bmatrix} 2 & -1 & & & & \\ -1 & 2 & -1 & & & \\ & \ddots & \ddots & \ddots & \\ & & \ddots & \ddots & -1 \\ & & & -1 & 2 \end{bmatrix} \in \mathbb{R}^{n \times n} \qquad \boldsymbol{b} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} \in \mathbb{R}^n$

1

spdtridiagsolver.m is provided. Plot the solution you obtained with n = 500.

(b) Since A is also tridiagonal, the algorithms in Question 3 can be applied. Write a Matlab code to implement your algorithm in Question 3(b)(c) for solving Ax = b where A template file

(a) Prove that A is SPD.

 $A = \begin{bmatrix} 3 & 8 & 14 \\ \frac{1}{3} & \frac{-2}{3} & -\frac{2}{3} \\ \frac{2}{3} & -1 & 3 \end{bmatrix}$

Therefore,

Explanation:

and $J^2 = I$

2. Algorithm for computing A = UL decomposition: a. Let $J = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$, compute JAJ, denote the result by M. b. Using LU decomposition algorithm learnt in class to solve M=LU, which is: for k = 1, 2, ..., nM(k, k: n) = M(k, k: n) - M(k, 1: k - 1)M(1: k - 1, k: n)M(k + 1: n, k) =(M(k + 1: n, k) - M(k + 1: n, 1: k - 1)M(1: k - 1, k))/M(k, k)end c. which L is the lower triangular matrix and U is the upper triangular matrix

d. Then, Compute JLJ = U' and JUJ = L'

 $JAJ = JLUJ = JLJ^2UJ = (JLJ)(JUJ) = U'L'$

So, we can obtain the above algorithm.

e. Finally, we obtain A = (JLJ)(JUJ) = U'L'

 $\begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{3} & 1 & 0 \\ \frac{2}{3} & -1 & 1 \end{bmatrix} \begin{bmatrix} 3 & 8 & 14 \\ 0 & \frac{-2}{3} & -\frac{2}{3} \\ 0 & 0 & 3 \end{bmatrix}$

3. Let $\mathbf{A} \in \mathbb{R}^{n \times n}$ be a tridiagonal matrix, i.e., $a_{ij} = 0$ if |i - j| > 1. We also assume that \mathbf{A} is symmetric positive definite (SPD). (a) Prove that the Cholesky decomposition $\mathbf{A} = \mathbf{L}\mathbf{L}^T$ satisfies $l_{ij} = 0$ for all i - j > 1. In other words, Using the Cholesky algorithm learnt in class to decompose A. 3.(a)

Since JLJ is upper triangular matrix and JUJ is lower triangular matrix

At step 1, $L(1,1) = \int A(1,1) = \sqrt{d}$ L(d:n,1) = A(2:n,1)

Since A(3:n/1) = 0, A(2:n/1) = A(2,1). $L(211) = \frac{A(211)}{L(111)} = \frac{C1}{\sqrt{d_1}}$

So, L(1:n,1), there is only 2 entires with value 70. The bidiagonal property holds at step i. The bi-diagonal property holds at step 1, i.e. L(i,1)= I for all

1-171. Assume the bidiagonal property holds for Step 1 to Step k-1. i.e. for i=k-1, uij = 0 for j-i>1 1r j=k-1. lij = 0 for i-j>1

 $L(k+2:n,k) = \left[A(k+2:n,k) - L(k+2:n,l:k-1)L(k+2:n,l:k-1)\right]$

The chloesky decomposition A=LL satisfies

bidiagonal property, bij =0 for all i-j>1.

L(K/K)

which is $DL(k,k) = A(k,k)^{\frac{1}{2}}$ Same as DL(k+1,k) = A(k+1,k) L(k,k)

For step k, (when i=k)

=> i=k-1, lji=0 for j-i>1.

= (0-0(L(K,1:K-1)))/L(K,K) i. The bidiagonal property holds for step k.

... Using mathematical inductions

36. Algorithm:

For k = 1: n

is needed.

needed.

3(c).

 $y_1 = \frac{b_1}{A(1,1)} / for j = 1$

 $y_j = \frac{b_j - A(j, j - 1)y_{j-1}}{A(j, j)}$

 $x_i = \frac{y_i - A(i, i + 1)y_{i+1}}{A(i, i)}$

After overwriting, algorithm becomes:

for j = 2: n

 $b_1 = \frac{b_1}{A(1,1)}$

end

end

end

 $L(k, k) = \sqrt{A(k, k) - L(k, k - 1)^2}$ L(k + 1, k) = A(k + 1, k)/L(k, k)End After overwriting, algorithm becomes: For k = 1: n $A(k, k) = \sqrt{A(k, k) - A(k, k - 1)^2}$ A(k + 1, k) = A(k + 1, k)/A(k, k)End

Operation needed: if taking square root is counted as one operation,

4 operations are needed for first line, except when k=1, 2 operation

(c) Based on the Cholesky decomposition, construct an O(n) algorithm to solve Ax = b. Express

1 operation for second line, except when k=n, no operation is

the number of operations needed in the form of Cn + O(1) with explicit C.

First, decompose A in LL^T , where L and L^T are both bi –

Note that L matrix can be replaced by A.

So it is (4 + 1)n - 2 - 1 = 5n - 3 cost.

diagonal matrix using the algorith in (b).

For k = 1: n $A(k, k) = \sqrt{A(k, k) - A(k, k - 1)^2}$ A(k + 1, k) = A(k + 1, k)/A(k, k)End Then, we use forward substitution to compute Ly = b where L is the lower — bidiaginal matrix. Pseudo code for forward substitution:

for j = 2: n $b_j = \frac{b_j - A(j, j - 1)b_{j-1}}{A(i, j)}$ end Finally, after we obtain y, we use backward substitution to compute $L^{T}x = y$ Pseudo Code for backward substitution: $x_n = \frac{y_n}{A(n,n)} / for i = n$ for i = n - 1:1

After overwriting, algorithm becomes: $b_{n} = \frac{b_{n}}{A(n, n)}$ for i = n - 1:1 $b_i = \frac{b_i - A(i, i + 1)b_{i+1}}{A(i, i)}$ **Running Cost Analysis:**

Using the result of (b), Cholesky decomposition costs 5n - 3

Forward substitution costs 3(n-1) + 1 = 3n - 2Backward substition costs 3(n-1) + 1 = 3n - 2Total cost = 11n - 7Q4(a). 4. We consider a discrete 1-D Laplacian equation Ax = b, where

(a) Prove that A is SPD. (b) Since A is also tridiagonal, the algorithms in Question 3 can be applied. Write a Matlab code to implement your algorithm in Question 3(b)(c) for solving Ax = b where A template file spdtridiagsolver.m is provided. Plot the solution you obtained with n = 500. 4(a). Prove $\forall x, x^T Ax > 0$

Let $x = \begin{pmatrix} x_1 \\ x_2 \\ \dots \end{pmatrix} \in \mathbb{R}^n$ $\mathbf{x}^{\mathrm{T}}\mathbf{A}\mathbf{x} = \begin{pmatrix} \mathbf{x}_1 & \mathbf{x}_2 & \dots & \mathbf{x}_n \end{pmatrix}$ $x^{T}Ax$ $= (2x_1 - x_2, -x_1 + 2x_2 - x_3, -x_2 + 2x_3 - x_4, ..., -x_{n-1})$

 $x^{T}Ax$ $=2x_1^2-x_1x_2-x_1x_2+2x_2^2-x_3x_2-x_2x_3+2x_3^2-x_4x_3+\cdots$ $-x_{n-1}x_n + 2x_n^2$ $x^{T}Ax = x_{1}^{2} + (x_{1} - x_{2})^{2} + (x_{2} - x_{3})^{2} + \dots + (x_{n-1} - x_{n})^{2} + x_{n}^{2}$ And $x^T A x = 0$ only if $x_1 = x_2 = \cdots = x_n = 0$ $x \neq 0$ $\therefore \mathbf{x}^{\mathrm{T}} \mathbf{A} \mathbf{x} > 0$ And also we can observe that $A = A^{T}$ ∴ A is a SPD. 4(b).

3 2.5 1.5 0.5 250 300 350

3.5