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Let

Due date: 29 March, Sunday

MATH3322 Matrix Computation

Homework 2

 $\mathbf{A} = \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix}$

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 of A. 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 positive definite (SPD).

Find the LU decompsotion with partial pivoting of A.

L is bi-diagonal. (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

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

 $\boldsymbol{A} = \begin{bmatrix} 2 & -1 & & & & \\ -1 & 2 & -1 & & & \\ & \ddots & \ddots & \ddots & \\ & & \ddots & \ddots & -1 \\ & & & -1 & 2 \end{bmatrix} \in \mathbb{R}^{n \times n} \qquad \boldsymbol{b} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} \in \mathbb{R}^n$

(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.

$$\begin{bmatrix} 2 & 4 \\ 2 & 4 \\ 6 & 13 \end{bmatrix}, P = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 14 \\ 4 \\ 13 \end{bmatrix}$$

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

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

$$\begin{bmatrix} 3 & 8 & 14 \\ 1 & -2 & 4 \end{bmatrix}$$

Therefore,

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

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

$$A = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} \\ \frac{2}{3} & -1 & 13 \end{bmatrix}$$

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

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

M(k, k: n) = M(k, k: n) - M(k, 1: k - 1)M(1: k - 1, k: n)

-M(k+1:n,1:k-1)M(1:k-1,k))/M(k,k)

c. which L is the lower triangular matrix and U is the upper

class to solve M=LU, which is:

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

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

M(k + 1: n, k) =

(M(k + 1: n, k))

for k = 1, 2, ..., n

triangular matrix

end

At step 1,
$$L(1,1) = \sqrt{A(1,1)} = \sqrt{d}i$$

$$L(\alpha;n,1) = A(2;n,1)$$

$$L(111)$$

Since $A(3:n,1) = 0$, $A(2:n,1) = A(2,1)$.

$$L(211) = \frac{A(211)}{L(111)} = \frac{C1}{\sqrt{01}}$$
So, $L(1:n,1)$, there is only 2 enthies with value $\neq 0$.

Assume the bidiagonal property bolds for Step 1 to Step k-1.

i.e. for
$$i=k-1$$
, $U_{ij}=0$ for $j-i>1$

for $j=k-1$. $U_{ij}=0$ for $i-j>1$

$$D = L(k,k) = (A(k,k) - L(k,1:k-1)(L(k,1:k-1))^{T})$$

$$O = L(k+1:n,k) = (A(k+1:n,k) - L(k+1:n,1:k-1)(L(k,1:k-1))^{T})/L(k,k)$$
Which is
$$D = L(k,k) = A(k,k)^{\frac{1}{2}}$$
Same as
$$O = L(k+1,k) = A(k+1,k)$$

$$L(k,k) = A(k+1,k)$$

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

For step k, (when i = k)

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

$$L(k,k)$$

$$= \left(0 - O(L(k,l:k-1)^{T}) / L(k,k) \right)$$

The chloesky decomposition
$$A = LL^T$$
 satisfies bidiagonal property, $l_{ij} = 0$ for all $i-j>1$.

i. The bidiagonal property holds for step k.

... Using mathematical inductions

 $L(k, k) = \sqrt{A(k, k) - L(k, k - 1)^2}$

L(k + 1, k) = A(k + 1, k)/L(k, k)

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

Using the result of (b), Cholesky decomposition costs 4n-3. Forward

(b) Since A is also tridiagonal, the algorithms in Question 3 can be applied. Write a Matlab code

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

to implement your algorithm in Question 3(b)(c) for solving Ax = b where A template file

Then, we use forward substitution to compute Ly =

diagonal matrix using the algorith in (b).

b where L is the lower — bidiaginal matrix.

Pseudo code for forward substitution:

3(b). Algorithm: For k = 1: n

End

3(c).

end

end

Running Cost Analysis:

(a) Prove that **A** is SPD.

 $x^{T}Ax$

 $x^{T}Ax$

 ≥ 0

 $x \neq 0$

4(b).

substitution costs 3(n-1) + 1 = 3n - 1

2. Backward substition costs 3(n-1) + 1 = 3n - 2

Running cost: 3 operations for first line, except when k=1, 1 operation is needed. 1 operation for second line, except when k=n, no operation is needed. So it is $(3 + 1)n - 2 - 1 = 4n - 3 \cos t$. (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.

for j = 1: n $y_j = \frac{b_j - L(j, j - 1)y_{j-1}}{L(j, j)}$ Finally, after we obtain y, we use backward substitution to compute $L^{T}x = y$ Pseudo Code for backward substitution: for i = n: 1 $x_i = \frac{y_i - L(i, i + 1)y_{i+1}}{L(i, i)}$

Total cost = 10n - 7Q4(a). 4. We consider a discrete 1-D Laplacian equation Ax = b, where

4(a). Prove $\forall x, x^T Ax > 0$ Let $x = \begin{pmatrix} x_1 \\ x_2 \\ \dots \end{pmatrix} \in \mathbb{R}^n$ $\mathbf{x}^{T}\mathbf{A}\mathbf{x} = (\mathbf{x}_{1} \quad \mathbf{x}_{2} \quad \dots \quad \mathbf{x}_{n}) \begin{pmatrix} 2 & -1 & & & \\ -1 & 2 & -1 & & & \\ & \dots & \dots & \dots & \\ & & \dots & \dots & -1 \\ & & & -1 & 2 \end{pmatrix} \begin{pmatrix} \mathbf{x}_{1} \\ \mathbf{x}_{2} \\ \dots \\ \mathbf{x}_{n} \end{pmatrix}$

 $= (2x_1 - x_2, -x_1 + 2x_2 - x_3, -x_2 + 2x_3 - x_4, ..., -x_{n-1})$ $+2x_n$ $\begin{pmatrix} x_2 \\ \dots \end{pmatrix}$ $=2x_1^2-x_1x_2-x_1x_2+2x_2^2-x_3x_2-x_2x_3+2x_3^2-x_4x_3+\cdots$

 $-x_{n-1}x_n + 2x_n^2$ $x^{T}Ax = x_{1}^{2} + (x_{1} - x_{2})^{2} + (x_{2} - x_{3})^{2} + \dots + (x_{n-1} - x_{n})^{2} + x_{n}^{2}$ And $\mathbf{x}^{\mathrm{T}}\mathbf{A}\mathbf{x} = 0$ only if $\mathbf{x}_1 = \mathbf{x}_2 = \cdots = \mathbf{x}_n = 0$ $\therefore x^{T}Ax > 0$ And also we can observe that $A = A^{T}$ ∴ A is a SPD.