

Due: Thursday November 17, 2022

MATH 6800: Problem Set 7

1. A matrix  $D$  is block tridiagonal if it is of the form

$$D = \begin{bmatrix} B_1 & C_1 & & & \\ A_2 & B_2 & C_2 & & \\ & A_3 & B_3 & C_3 & \\ & & A_4 & B_4 & C_4 \\ & & & \ddots & \ddots & \ddots \\ & & & & A_n & B_n \end{bmatrix}$$

where each  $A_i, B_i$  and  $C_i$  is a small matrix of size  $p \times p$  ( $p$  is the block size). Derive a block LU decomposition (i.e. an LU decomposition that uses operations involving  $p \times p$  matrices instead of scalars), assuming no pivoting. What are the conditions you need for this LU decomposition to exist?

Assuming every block matrix  $(A_i, B_i, C_i)$  is nonsingular then let the operations be performed as:

$$\begin{bmatrix} I & & & \\ -A_2(B_1)^{-1} & I & & \\ & & I & \\ & & & \ddots \end{bmatrix} \begin{bmatrix} B_1 & C_1 & & \\ A_2 & B_2 & C_2 & \\ & A_3 & B_3 & C_3 \\ & & \ddots & \ddots \end{bmatrix} = \begin{bmatrix} B_1 & C_1 & & \\ (-A_2(B_1)^{-1}C_1 + B_2) & C_2 & & \\ A_3 & B_3 & C_3 & \\ & A_4 & B_4 & C_4 \\ & & \ddots & \ddots \end{bmatrix}$$

$$\begin{bmatrix} I & & & \\ & I & & \\ & (-A_3(-A_2(B_1)^{-1}C_1 + B_2)^{-1}) & I & \\ & & & \ddots \end{bmatrix} \begin{bmatrix} B_1 & C_1 & & \\ (-A_2(B_1)^{-1}C_1 + B_2) & C_2 & & \\ A_3 & B_3 & C_3 & \\ & \ddots & \ddots & \ddots \end{bmatrix} = \begin{bmatrix} B_1 & C_1 & & \\ (-A_2(B_1)^{-1}C_1 + B_2) & C_2 & & \\ & (-A_3(-A_2(B_1)^{-1}C_1 + B_2)^{-1}C_2 + B_3) & C_3 & \\ & & & \ddots \end{bmatrix}$$

Therefore the algorithm can be written as:

$$\begin{aligned} \alpha_1 &= \mathbf{0}_{p \times p}, & \beta_1 &= B_1 \\ \alpha_i &= A_i(\beta_{i-1})^{-1} & \beta_i &= -\alpha_i C_{i-1} + B_i, \quad i = 2, 3, 4, \dots, p \end{aligned}$$

In this case, we require every  $\beta_i$  matrix to be nonsingular.

$$\begin{bmatrix} I & & & \\ \alpha_2 & I & & \\ & \alpha_3 & I & \\ & & \ddots & \\ & & \alpha_n & I \end{bmatrix} \begin{bmatrix} \beta_1 & C_1 & & \\ & \beta_2 & C_2 & \\ & & \beta_3 & C_3 \\ & & & \ddots \\ & & & & \beta_n \end{bmatrix} = LU$$

2. **NLA 20.3** Suppose an  $m \times m$  matrix  $A \dots$

The matrix  $A \in \mathbb{C}^{m \times m}$  is written as a block matrix with  $A_{11}, A_{12}, A_{21}, A_{22}$  where  $A_{11} \in \mathbb{C}^{n \times n}$  and  $A_{22} \in \mathbb{C}^{m-n \times m-n}$ .

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$$

(a)

$$\begin{bmatrix} I & \mathbf{0} \\ -A_{21}(A_{11})^{-1} & I \end{bmatrix} \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ -\cancel{A_{21}}(A_{11})^{-1}A_{11} + \cancel{A_{21}} & A_{21}(A_{11})^{-1}A_{12} + A_{22} \end{bmatrix} \\ = \begin{bmatrix} A_{11} & A_{12} \\ \mathbf{0} & A_{22} - A_{21}(A_{11})^{-1}A_{12} \end{bmatrix}$$

(b)  $A_{21} \in \mathbb{C}^{m-n \times n}$ . Since there are  $n$  columns, there are  $n$  Gauss Elimination steps. Show that this still results in  $A_{22} - A_{21}(A_{11})^{-1}A_{12}$  in the bottom right.

3. The matrix  $A \in \mathbb{C}^{m \times m}$  is diagonally dominant if

$$|a_{ii}| > \sum_{j=1, j \neq i}^m |a_{ij}|, \quad i = 1, 2, \dots, m.$$

(a) Prove that if  $A$  is diagonally dominant, then any principle submatrix of  $A$  is diagonally dominant.

To generate any principle submatrix, if  $k$  rows are removed, then the same  $k$  columns are to be removed. So, if 2, 3, 5 rows are removed, then 2, 3, 5 columns are also removed. In which case, the resulting matrix is also a square matrix of the form  $\tilde{A} \in \mathbb{C}^{m-3 \times m-3}$ . When generalized, this resulting square matrix is  $\tilde{A} \in \mathbb{C}^{m-k \times m-k}$ . Now in this particular example, if we take any row  $r \neq 2, 3, 5$  from  $A$  matrix,

$$\sum_{j=1, j \neq 2, 3, 5}^m |a_{rj}| = |a_{r1}| + |a_{r2}| + \dots + |a_{rr}| + \dots + |a_{rm}| - |a_{r2}| - |a_{r3}| - |a_{r5}|$$

If  $|a_{rr}| > \sum_{j=1, j \neq r}^m |a_{rj}|$  then it is definitely greater than,  $\left(\sum_{j=1, j \neq r}^m |a_{rj}|\right) - |a_{r2}| - |a_{r3}| - |a_{r5}|$  since those are absolute values which are just positive values being subtracted. Therefore, if  $k$  rows and  $k$  columns are removed, then  $k$  positive values are subtracted from the summation. Hence, any principle submatrix is diagonally dominant.

- (b) Prove that if  $A$  is diagonally dominant, then  $A$  is nonsingular.

If a matrix is singular, then there exists a vector  $u \in \mathbb{C}^m \setminus \{0\}$  such that  $Au = \mathbf{0}$ . In which case, we have

$$\sum_{j=1}^m a_{ij}u_j = 0$$

In the worst case scenario, if for any row  $k$  if  $u_k = \|u\|_\infty$ ,

$$\begin{aligned} \sum_{j=1}^m a_{kj} \frac{u_j}{u_k} &= 0 \\ - \sum_{j=1, j \neq k}^m a_{kj} \frac{u_j}{u_k} &= a_{kk} \\ \left| \sum_{j=1, j \neq k}^m a_{kj} \frac{u_j}{u_k} \right| &= |a_{kk}| \end{aligned}$$

Then, we have

$$|a_{kk}| \leq \sum_{j=1, j \neq k}^m |a_{kj}| \left| \frac{u_j}{u_k} \right|$$

This is a contradiction because  $\left| \frac{u_j}{u_k} \right| \leq 1$  but  $|a_{kk}| > \sum_{j=1, j \neq k}^m |a_{kj}|$ . Therefore,  $A$  is nonsingular.

- (c) Prove that if  $A$  is diagonally dominant then it will have an LU decomposition (you may use the result of NLA 20.1).

In NLA 20.1 we have proved that when  $A$  is nonsingular, a LU decomposition exists if  $A_{1:k, 1:k}$  is nonsingular. From part (a), we proved that every principle submatrix is diagonally dominant. In part (b) we proved that any diagonally dominant matrix is nonsingular. Hence any principle submatrix  $A_{1:k, 1:k}$  will be nonsingular since they are all diagonally dominant. So, with NLA 20.1, we can prove that  $A$  has a LU decomposition.

4. Write a Matlab code  $[L, U, P] = \text{lufactor}(A)$  that takes an  $m \times m$  matrix  $A$  and computes the LU factorization,  $PA = LU$ , using partial pivoting. Write a second Matlab code  $x = \text{lusolve}(b, L, U, P)$  that solves the system  $Ax = b$ , for some  $x$  given  $b$ , using the output from  $\text{lufactor}$ . For this exercise you should only use elementary arithmetic, vector and matrix operations (e.g. no backlash operators to solve the triangular systems).

- (a) Test your  $\text{lufactor}(A)$  routine on this matrix,

$$A = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

Output  $A, L, U$  and  $P$ . Check that  $PA = LU$ .

The code to compute  $L, U$  and  $P$  from  $A$  is in Listing 1.

```

1 function [L,U,P] = lufactor(A)
2
3 [m,~] = size(A);
4 U = A;
5 L = eye(m);
6 P = eye(m);
7
8 for k=1:m-1
9     % finding maximum value in the column
10    max_val = abs(U(k,k));
11    max_id = k;
12    for i=k:m
13        if(abs(U(i,k)) > max_val)
14            max_val = abs(U(i,k));
15            max_id = i;
16        end
17    end
18
19    % interchanging two rows
20    t = U(k,k:m);
21    U(k,k:m) = U(max_id,k:m);
22    U(max_id,k:m) = t;
23
24    t = L(k,1:k-1);
25    L(k,1:k-1) = L(max_id,1:k-1);
26    L(max_id,1:k-1) = t;
27
28    t = P(k,:);
29    P(k,:) = P(max_id,:);
30    P(max_id,:) = t;
31
32    for j=k+1:m
33        L(j,k) = U(j,k)/U(k,k);
34        U(j,k:m) = U(j,k:m) - L(j,k)*U(k,k:m);
35    end
36
37 end
38
39 end

```

Listing 1: **lufactor**(A)

- (b) Test your function *lusolve* by solving  $Ax = b$  where  $A$  is from part (a) and  $b = [7, 23, 69, 79]^T$ . Output  $x$  and check that  $Ax = b$ .

This is solved in two parts, where

$$Ax = b$$

$$PAx = Pb$$

$$LUx = Pb$$

$$Ly = z$$

$$Ux = y$$

The code used to solve this system of equations is in Listing 2. Script in Listing 3 is used to verify for the given case in consideration and the solutions are attached below as well.

```

1 function x = lusolve(b,L,U,P)
2
3 [m,~] = size(L);
4
5 % part 1
6 z = P*b;
7 y = zeros(m,1);
8 for i=1:m
9     s = z(i);
10    for j=1:i
11        if j~=i
12            s = s - L(i,j)*y(j);
13        end
14    end
15    y(i) = s;
16 end
17
18 % part 2
19 x = zeros(m,1);
20 for i=m:-1:1
21     if i==m
22         x(i) = y(i)/U(i,i);
23     else
24         rv = U(i,i+1:m);
25         xv = x(i+1:m,1);
26         x(i) = (y(i)-rv*xv)/U(i,i);
27     end
28 end
29
30 end

```

Listing 2: lusolve(b,L,U,P)

```

1 clc
2 clear
3 %%
4 A = [2 1 1 0; 4 3 3 1; 8 7 9 5; 6 7 9 8];
5
6 [L,U,P] = lufactor(A);
7 fprintf('Norm(PA-LU): %f \n', norm(P*A-L*U));
8
9 b = [7 23 69 79]';
10 x = lusolve(b,L,U,P);
11 disp('x=');
12 disp(x);
13 disp('Ax = ');
14 disp(A*x);
15 fprintf('Norm(Ax-b): %f \n', norm(A*x-b));

```

Listing 3: script

Norm(PA-LU): 0.000000

x=

1.0000

2.0000

3.0000

4.0000

Ax =

7.0000

23.0000

69.0000

79.0000

Norm(Ax-b): 0.000000

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