

2021 Spring MAS 365
Chapter 6: Direct Methods for Solving Linear Systems

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Introduction

- This chapter studies **direct methods** for solving a linear system of n equations in n variables.

$$E_1 : a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1,$$

$$E_2 : a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2,$$

$$\vdots$$

$$E_n : a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n = b_n.$$

- Direct techniques are methods that theoretically give the exact solution to the system in a finite number of steps. In practice, it is affected by the round-off error, which we study in this chapter.
- Chapter 7 considers methods of approximating the solution to linear systems using iterative methods.

1 6.1 Linear Systems of Equations

2 6.2 Pivoting Strategies

Elementary Operations

- Multiply E_i by $\lambda \neq 0$: $(\lambda E_i) \rightarrow (E_i)$
 - Multiply E_j by λ and add to E_i : $(E_i + \lambda E_j) \rightarrow (E_i)$
 - Interchange E_i and E_j : $(E_i) \leftrightarrow (E_j)$
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- By a sequence of these operations, a linear system will be systematically transformed into a new linear system that is more easily solved and has the same solutions.

Gaussian Elimination with Backward Substitution

- Form the augmented matrix \tilde{A}

$$\tilde{A} = [A, \mathbf{b}] = \left[\begin{array}{cccc|c} a_{11} & a_{12} & \cdots & a_{1n} & a_{1,n+1} \\ a_{21} & a_{22} & \cdots & a_{2n} & a_{2,n+1} \\ \vdots & \vdots & & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} & a_{n,n+1} \end{array} \right]$$

- Provided $a_{11} \neq 0$, perform

$$(E_j - (a_{j1}/a_{11})E_1) \rightarrow (E_j) \quad \text{for each } j = 2, 3, \dots, n$$

to eliminate the coefficient of x_1 in each of these rows.

- For simplicity, we will continue to use the notation a_{ij} , even though it is expected to change after the operations.

Gaussian Elimination with Backward Substitution (cont'd)

- Provided $a_{ii} \neq 0$ we perform the following operation sequentially

$$(E_j - (a_{ji}/a_{ii})E_i) \rightarrow (E_j) \quad \text{for each } j = i + 1, i + 2, \dots, n.$$

- We then have

$$\tilde{A} = \left[\begin{array}{cccc|c} a_{11} & a_{12} & \cdots & a_{1n} & a_{1,n+1} \\ 0 & a_{22} & \cdots & a_{2n} & a_{2,n+1} \\ \vdots & \ddots & \ddots & \vdots & \vdots \\ 0 & \cdots & 0 & a_{nn} & a_{n,n+1} \end{array} \right]$$

Gaussian Elimination with Backward Substitution (cont'd)

- The new linear system is triangular

$$a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = a_{1,n+1},$$

$$a_{22}x_2 + \cdots + a_{2n}x_n = a_{2,n+1},$$

$$\vdots$$

$$a_{nn}x_n = a_{n,n+1}.$$

- Backward substitution

$$x_n = \frac{a_{n,n+1}}{a_{nn}}$$

$$x_i = \frac{a_{i,n+1} - \sum_{j=i+1}^n a_{ij}x_j}{a_{ii}}, \quad \text{for each } i = n-1, n-2, \dots, 1.$$

Gaussian Elimination with Backward Substitution (cont'd)

- Describe Gaussian elimination by forming a sequence of augmented matrices $\tilde{A}^{(1)} = \tilde{A}, \tilde{A}^{(2)}, \dots, \tilde{A}^{(n)}$, where $\tilde{A}^{(k)}$ has entries $a_{ij}^{(k)}$:

$$a_{ij}^{(k)} = \begin{cases} a_{ij}^{(k-1)}, & i = 1, 2, \dots, k-1, \text{ and } j = 1, 2, \dots, n+1, \\ 0, & i = k, k+1, \dots, n, \text{ and } j = 1, 2, \dots, k-1, \\ a_{ij}^{(k-1)} - \frac{a_{i,k-1}^{(k-1)}}{a_{k-1,k-1}^{(k-1)}} a_{k-1,j}^{(k-1)}, & i = k, k+1, \dots, n, \text{ and } j = k, k+1, \dots, n+1. \end{cases}$$

- Thus

$$\tilde{A}^{(k)} = \left[\begin{array}{ccccccccc|c} a_{11}^{(1)} & a_{12}^{(1)} & a_{13}^{(1)} & \cdots & a_{1,k-1}^{(1)} & a_{1k}^{(1)} & \cdots & a_{1n}^{(1)} & \vdots & a_{1,n+1}^{(1)} \\ 0 & a_{22}^{(2)} & a_{23}^{(2)} & \cdots & a_{2,k-1}^{(2)} & a_{2k}^{(2)} & \cdots & a_{2n}^{(2)} & \vdots & a_{2,n+1}^{(2)} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \ddots & a_{k-1,k-1}^{(k-1)} & a_{k-1,k}^{(k-1)} & \cdots & a_{k-1,n}^{(k-1)} & \vdots & a_{k-1,n+1}^{(k-1)} \\ \vdots & \vdots & \vdots & \ddots & 0 & a_{kk}^{(k)} & \cdots & a_{kn}^{(k)} & \vdots & a_{k,n+1}^{(k)} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & \cdots & \cdots & \cdots & 0 & a_{nk}^{(k)} & \cdots & a_{nn}^{(k)} & \vdots & a_{n,n+1}^{(k)} \end{array} \right]$$

Gaussian Elimination with Backward Substitution (cont'd)

- This will fail if one of the elements $a_{11}^{(1)}, a_{22}^{(2)}, \dots, a_{nn}^{(n)}$ is zero because the step

$$\left(E_i - \frac{a_{i,k}^{(k)}}{a_{kk}^{(k)}} (E_k) \right) \rightarrow E_i$$

either cannot be performed (if one of $a_{11}^{(1)}, a_{22}^{(2)}, \dots, a_{n-1,n-1}^{(n-1)}$ is zero)
or the backward substitution cannot be accomplished (if $a_{nn}^{(n)} = 0$).

- We need some change in the method.

Gaussian Elimination with Backward Substitution (cont'd)

Ex. Represent the linear system

$$E_1 : x_1 - x_2 + 2x_3 - x_4 = -8,$$

$$E_2 : 2x_1 - 2x_2 + 3x_3 - 3x_4 = -20,$$

$$E_3 : x_1 + x_2 + x_3 = -2,$$

$$E_4 : x_1 - x_2 + 4x_3 + 3x_4 = 4,$$

as an augmented matrix and use Gaussian Elimination to find its solution.

Sol. The first two augmented matrices are

$$\tilde{A}^{(1)} = \left[\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 2 & -2 & 3 & -3 & -20 \\ 1 & 1 & 1 & 0 & -2 \\ 1 & -1 & 4 & 3 & 4 \end{array} \right] \quad \text{and} \quad \tilde{A}^{(2)} = \left[\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & -1 & -4 \\ 0 & 2 & -1 & 1 & 6 \\ 0 & 0 & 2 & 4 & 12 \end{array} \right]$$

- $a_{22}^{(2)}$, called the **pivot element**, is 0.

The pivot element for a specific column is the entry that is used to place zeros in the other entries in that column.

Gaussian Elimination with Backward Substitution (cont'd)

Sol. Since $a_{32}^{(2)} \neq 0$, we perform $(E_2) \leftrightarrow (E_3)$, which yields

$$\tilde{A}^{(2)'} = \left[\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 2 & -1 & 1 & 6 \\ 0 & 0 & -1 & -1 & -4 \\ 0 & 0 & 2 & 4 & 12 \end{array} \right]$$

- We now already have $\tilde{A}^{(3)} = \tilde{A}^{(2)'}$, the next step gives

$$\tilde{A}^{(4)} = \left[\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 2 & -1 & 1 & 6 \\ 0 & 0 & -1 & -1 & -4 \\ 0 & 0 & 0 & 2 & 4 \end{array} \right]$$

and the backward substitution provides the solution.

Gaussian Elimination with Backward Substitution (cont'd)

- When $a_{kk}^{(k)} = 0$, search the first nonzero entry of the k th column of $\tilde{A}^{(k)}$ from the k th row to the n th row.
- 1. If $a_{pk}^{(k)} \neq 0$ for some p , with $k + 1 \leq p \leq n$, then $(E_k) \leftrightarrow (E_p)$ is performed to obtain $\tilde{A}^{(k)'$.
- 2. If $a_{pk}^{(k)} = 0$ for each p , with $k \leq p \leq n$, it can be shown that the linear system does not have a unique solution and the procedure stops.

1 6.1 Linear Systems of Equations

2 6.2 Pivoting Strategies

Pivoting

- To reduce round-off error, it is often necessary to perform row interchanges even when the pivot elements are not zero. When and why?

Pivoting (cont'd)

- If $a_{kk}^{(k)}$ is small in magnitude compared to $a_{jk}^{(k)}$, then the magnitude of the multiplier

$$m_{jk} = \frac{a_{jk}^{(k)}}{a_{kk}^{(k)}}$$

will be much larger than 1. Round-off error introduced in the computation of one of the terms $a_{kl}^{(k)}$ is multiplied by m_{jk} when computing $a_{jl}^{(k+1)}$.

Pivoting (cont'd)

- Also, when performing the backward substitution for

$$x_k = \frac{a_{k,n+1}^{(k)} - \sum_{j=k+1}^n a_{kj}^{(k)}}{a_{kk}^{(k)}}$$

with a small value of $a_{kk}^{(k)}$, any error in the numerator can be dramatically increased.

Pivoting (cont'd)

Ex. Apply Gaussian elimination to the system

$$E_1 : 0.003000x_1 + 59.14x_2 = 59.17,$$

$$E_2 : 5.291x_1 - 6.130x_2 = 46.78,$$

using four-digit arithmetic with rounding, and compare the results to the exact solution $x_1 = 10.00$ and $x_2 = 1.000$.

Pivoting (cont'd)

Sol. The first pivot element, $a_{11}^{(1)} = 0.003000$, is small, and its associated multiplier,

$$m_{21} = \frac{5.291}{0.003000} = 1763.\bar{6} \approx 1764.$$

Performing $(E_2 - m_{21}E_1) \rightarrow (E_2)$ with appropriate round gives

$$\begin{aligned} 0.003000x_1 + 59.14x_2 &\approx 59.17, \\ -104300x_2 &= -104400, \end{aligned}$$

instead of the exact system

$$\begin{aligned} 0.003000x_1 + 59.14x_2 &= 59.17, \\ -104309.37\bar{6}x_2 &= -104309.37\bar{6}. \end{aligned}$$

Pivoting (cont'd)

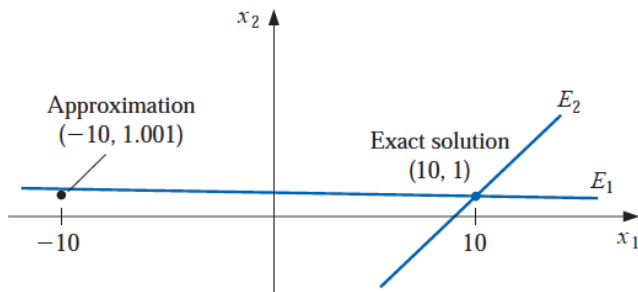
Sol. Backward substitution yields

$$x_2 \approx 1.001,$$

which is close to the actual value $x_2 = 1.000$. However, we have

$$x_1 \approx \frac{59.17 - (59.14)(1.001)}{0.003000} = -10.00,$$

where the small error 0.001 is multiplied by $\frac{59.14}{0.003000} \approx 20000$. The actual value is $x_1 = 10.00$.



Pivoting (cont'd)

- Difficulties can arise when the pivot element $a_{kk}^{(k)}$ is small relative to the entries $a_{ij}^{(k)}$ for $k \leq i \leq n$ and $k \leq j \leq n$.
- To avoid this, pivoting is performed by selecting an element $a_{pq}^{(k)}$ with a larger magnitude as the pivot. Then we interchange the k th and p th rows, followed by the interchange of the k th and q th columns, if necessary.
- Partial pivoting (maximal column pivoting)
- Scaled partial pivoting (scaled column pivoting)
- Complete pivoting (maximal pivoting)

Partial Pivoting

- Partial pivoting (maximal column pivoting):

Determine the smallest $p \geq k$ such that

$$|a_{pk}^{(k)}| = \max_{k \leq i \leq n} |a_{ik}^{(k)}|$$

and perform $(E_k) \leftrightarrow (E_p)$.

Partial Pivoting (cont'd)

Ex. Apply Gaussian elimination to the system

$$E_1 : 0.003000x_1 + 59.14x_2 = 59.17,$$

$$E_2 : 5.291x_1 - 6.130x_2 = 46.78,$$

using partial pivoting and four-digit arithmetic with rounding, and compare the results to the exact solution $x_1 = 10.00$ and $x_2 = 1.000$.

Partial Pivoting (cont'd)

Sol. The multiplier for this system is

$$m_{21} = \frac{a_{21}^{(1)}}{a_{11}^{(1)}} = 0.0005670,$$

and the operation $(E_2 - m_{21}E_1) \rightarrow (E_2)$ yields

$$\begin{aligned} 5.291x_1 - 6.130x_2 &\approx 46.78, \\ 59.14x_2 &\approx 59.14, \end{aligned}$$

which gives the correct values.

Partial Pivoting (cont'd)

Ex. Apply Gaussian elimination to the system

$$E_1 : 30.00x_1 + 591400x_2 = 591700,$$

$$E_2 : 5.291x_1 - 6.130x_2 = 46.78,$$

using partial pivoting and four-digit arithmetic with rounding.

Sol. The maximal value in the first column is 30.00 and the multiplier

$$m_{21} = \frac{5.291}{30.00} = 0.1764$$

yields

$$30.00x_1 + 591400x_2 \approx 591700,$$

$$-104300x_2 \approx -104400,$$

which has the inaccurate solution $x_1 = -10.00$ and $x_2 = 1.001$.

Scaled Partial Pivoting

- Scaled partial pivoting (scaled column pivoting): place the element in the pivot position that is largest relative to the entries in its row.

1. Define a scale factor s_i for each row as

$$s_i = \max_{1 \leq j \leq n} |a_{ij}|.$$

Let $s_i \neq 0$ for all i . Otherwise, the system has no unique solution.

2. Choose the least integer p with

$$\frac{|a_{p1}|}{s_p} = \max_{1 \leq k \leq n} \frac{|a_{k1}|}{s_k}$$

and perform $(E_k) \leftrightarrow (E_p)$. This ensures that the largest element in each row has a relative magnitude of 1 before the comparison for row interchange is performed.

Scaled Partial Pivoting (cont'd)

3. Similarly, before eliminating the variable x_i using $E_k - m_{ki}E_i$, we select the smallest integer $p \geq i$ with

$$\frac{|a_{pi}|}{s_p} = \max_{1 \leq k \leq n} \frac{|a_{ki}|}{s_k}$$

and perform $(E_i) \leftrightarrow (E_p)$ if $i \neq p$.

Scaled Partial Pivoting (cont'd)

Ex. Apply Gaussian elimination to the system

$$E_1 : 30.00x_1 + 591400x_2 = 591700,$$

$$E_2 : 5.291x_1 - 6.130x_2 = 46.78,$$

using scaled partial pivoting and four-digit arithmetic with rounding.

Complete Pivoting

- Complete (or maximal) pivoting: at the k th step searches all the entries a_{ij} , for $i = k, k + 1, \dots, n$ and $j = k, k + 1, \dots, n$, to find the entry with the largest magnitude. Both row and column interchanges are performed to bring this entry to the pivot position.
- The total time required to incorporate complete pivoting is

$$\sum_{k=2}^n (k^2 - 1) = \frac{n(n-1)(2n+5)}{6} \text{ comparisons.}$$

- Scaled partial pivoting requires a total of

$$n(n-1) + \sum_{k=1}^{n-1} k = \frac{3}{2}n(n-1) \text{ comparisons, and}$$

$$\sum_{k=2}^n k = \frac{1}{2}(n-1)(n+2) \text{ divisions.}$$