Numerical Computations - Chapter #3: Solving Systems of Linear Equations

Numerical Computations

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Chapter's Topics

- Overview of Systems of Linear Equations
- Solution by inverse matrices (Cramer's rule)
- The Gaussian Method
- The Gaussian Compact Method
- The Modification of Crout-Doolittle
- The Method of Principal Elements
- The Scheme of Khaletsky
- Method of successive approximations
 - · Jacobi's Method
 - · Gauss-Siedel Method

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Overview

Solving systems of linear equations may be required for solving problems with different variables in different fields.

Methods of solving systems of linear equations are divided mainly into two groups:

Exact methods: Computations are carried out exactly; thus they yield exact values of the unknowns variables.

Iterative methods: Begin with an initial guess and improve the answer in each iteration.

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Systems of Linear Equations

Suppose we have a system of n linear equations in n unknown variables:

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = a_{1(n+1)} \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = a_{2(n+1)} \\ & : \\ a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = a_{n(n+1)} \end{cases}$$

Denote the matrix of the coefficients, variables, constant term of the above system, respectively, by A, b and x where

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}, b = \begin{bmatrix} a_{1(n+1)} \\ a_{2(n+1)} \\ \dots \\ a_{n(n+1)} \end{bmatrix}, x = \begin{bmatrix} x_1 \\ x_2 \\ \dots \\ x_n \end{bmatrix}.$$

So, this system can be written in the matrix form of Ax = b

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Solution by inverse matrices (Cramer's rule)

If matrix A is <u>non-singular</u>, then

$$\det A = \begin{vmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{vmatrix} \neq 0$$

then, the system has a unique solution, i.e.,

$$x = A^{-1} b$$
.

Cramer's rule: The *j*th component of $x = A^{-1} b$ is the ratio

$$x_j = \frac{\det B_j}{\det A}$$

where

$$B_{j} = \begin{bmatrix} a_{11} & a_{12} & \cdots & \mathbf{b_{1}} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & \mathbf{b_{2}} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots & \cdots & \vdots \\ a_{n1} & a_{n2} & \cdots & \mathbf{b_{n}} & \cdots & a_{nn} \end{bmatrix}$$

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Systems of Linear Equations (cont'd)

But Cramer's rule as a method of solving a linear system with nunknowns leads to computing n + 1 determinants, which is quite a laborious operation especially when the number n is rather large.

It is shown that the complexity of computing determinant is $\mathcal{O}(n^{\omega})$ for some $\omega \geq 2$.

The fastest matrix-multiplication algorithms (Coppersmith-Winograd) can be used with $\mathcal{O}(n^{\sim 2.376})$ arithmetic operations, but use heavy mathematical tools and are often impractical.

The Gaussian Method



Carl Fredrich Gauss (1777-1855)

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The Gaussian Method

Example.
$$\begin{cases} x_1 + 2x_2 + 3x_3 = 6 \\ 2x_1 - 1x_2 + 4x_3 = 8 \\ -x_1 + 8x_2 + 2x_3 = 12 \end{cases}$$

Solution.

Solution.
$$\begin{bmatrix} 1 & 2 & 3 & 6 \\ 2 & -1 & 4 & 8 \\ -1 & 8 & 2 & 12 \end{bmatrix} \xrightarrow{R_2 - 2R_1} \begin{bmatrix} 1 & 2 & 3 & 6 \\ 0 & -5 & -2 & -4 \\ 0 & 10 & 5 & 18 \end{bmatrix} \xrightarrow{R_3 + 2R_2}$$

$$\begin{bmatrix} 1 & 2 & 3 & 6 \\ 0 & -5 & -2 & -4 \\ 0 & 0 & 1 & 10 \end{bmatrix} \implies x_3 = 10, \quad x_2 = -\frac{16}{5}, \quad x_1 = -\frac{88}{5}$$

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The Gaussian Method (cont'd)

The most common technique for the solution of systems of linear equations is via an algorithm for the successive elimination of the unknowns. This method is called the Gaussian method.

For the below system of 4 linear equations:

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + a_{14}x_4 = a_{15} \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + a_{24}x_4 = a_{25} \\ a_{31}x_1 + a_{32}x_2 + a_{33}x_3 + a_{34}x_4 = a_{35} \\ a_{41}x_1 + a_{42}x_2 + a_{43}x_3 + a_{44}x_4 = a_{45} \end{cases}$$

Let $a_{11} \neq 0$ (the leading element).

Dividing the first equation of the system by a_{11} , we get

$$x_1 + b_{12}x_2 + b_{13}x_3 + b_{14}x_4 = b_{15}$$

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The Gaussian Method (cont'd)

$$\begin{cases} x_1 + b_{12}x_2 + b_{13}x_3 + b_{14}x_4 = b_{15} \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + a_{24}x_4 = a_{25} \\ a_{31}x_1 + a_{32}x_2 + a_{33}x_3 + a_{34}x_4 = a_{35} \\ a_{41}x_1 + a_{42}x_2 + a_{43}x_3 + a_{44}x_4 = a_{45} \end{cases}$$

Using the previous equation, eliminate the unknown x_1 from the second, third, and fourth equations of the system.

We get a system consisting of three equations as:

$$a_{22}^{(1)}x_2 + a_{23}^{(1)}x_3 + a_{24}^{(1)}x_4 = a_{25}^{(1)}$$

$$a_{32}^{(1)}x_2 + a_{33}^{(1)}x_3 + a_{34}^{(1)}x_4 = a_{35}^{(1)}$$

$$a_{42}^{(1)}x_2 + a_{43}^{(1)}x_3 + a_{44}^{(1)}x_4 = a_{45}^{(1)}$$

$$a_{ij}^{(1)} = a_{ij} - a_{i1}b_{1j} \; (i=2,3,4; \; j=2,3,4,5)$$

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The Gaussian Method (cont'd)

$$\begin{cases} x_1 + b_{12}x_2 + b_{13}x_3 + b_{14}x_4 &= b_{15} \\ 0x_1 + a_{22}^{(1)}x_2 + a_{23}^{(1)}x_3 + a_{24}^{(1)}x_4 &= a_{25}^{(1)} \\ 0x_1 + a_{32}^{(1)}x_2 + a_{33}^{(1)}x_3 + a_{34}^{(1)}x_4 &= a_{35}^{(1)} \\ 0x_1 + a_{42}^{(1)}x_2 + a_{43}^{(1)}x_3 + a_{44}^{(1)}x_4 &= a_{45}^{(1)} \end{cases}$$

Dividing the first equation of the new system by the leading element $a_{22}^{\left(1\right)}$, we get

$$x_2 + b_{23}^{(1)} x_3 + b_{24}^{(1)} x_4 = b_{25}^{(1)}$$

$$b_{2j}^{(1)} = \frac{a_{2j}^{(1)}}{a_{22}^{(1)}}$$
 for $j = 3, 4, 5$.

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The Gaussian Method (cont'd)

Thus

$$\begin{cases} x_1 + b_{12}x_2 + b_{13}x_3 + b_{14}x_4 &= b_{15} \\ 0x_1 + x_2 + b_{23}^{(1)}x_3 + b_{24}^{(1)}x_4 &= b_{25}^{(1)} \\ 0 x_1 + a_{32}x_2 + a_{33}x_3 + a_{34}x_4 &= a_{35} \\ 0 x_1 + a_{42}x_2 + a_{43}x_3 + a_{44}x_4 &= a_{45} \end{cases}$$

Eliminating x_2 in the same way that we eliminated x_1 , we arrive at the following new system of equations:

$$a_{33}^{(2)}x_3 + a_{34}^{(2)}x_4 = a_{35,}^{(2)}$$

$$a_{43}^{(2)}x_3 + a_{44}^{(2)}x_4 = a_{45,}^{(2)}$$

$$a_{ij}^{(2)} = a_{ij}^{(1)} - a_{i2}^{(1)}b_{2j}^{(1)}(i = 3,4; j = 3,4,5)$$

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The Gaussian Method (cont'd)

Thus

$$\begin{cases} x_1 + b_{12}x_2 + b_{13}x_3 + b_{14}x_4 &= b_{15} \\ 0x_1 + x_2 + b_{23}^{(1)}x_3 + b_{24}^{(1)}x_4 &= b_{25}^{(1)} \\ 0 x_1 + 0x_2 + a_{33}^{(2)}x_3 + a_{34}^{(2)}x_4 &= a_{35}^{(2)}, \\ 0 x_1 + 0x_2 + a_{43}^{(2)}x_3 + a_{44}^{(2)}x_4 &= a_{45}^{(2)}, \end{cases}$$

Dividing the first equation by the leading element $a_{33}^{(2)}$ we get

$$x3 + b_{34}^{(2)}x4 = b_{35}^{(2)}$$
$$b_{3j}^{(2)} = \frac{a_{3j}^{(2)}}{a_{33}^{(2)}} (j = 4,5)$$

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The Gaussian Method (cont'd)

Thus

$$\begin{cases} x_1 + b_{12}x_2 + b_{13}x_3 + b_{14}x_4 &= b_{15} \\ 0x_1 + x_2 + b_{23}^{(1)}x_3 + b_{24}^{(1)}x_4 &= b_{25}^{(1)} \\ 0 x_1 + 0x_2 + x_3 + b_{34}^{(2)}x_4 &= b_{35}^{(2)} \\ 0 x_1 + 0x_2 + a_{43}^{(2)}x_3 + a_{44}^{(2)}x_4 &= a_{45}^{(2)}, \end{cases}$$

We eliminate x_3 from the second equation. We get the equation

$$a_{44}^{(3)}x_4 = a_{45}^{(3)}$$
 $a_{4j}^{(3)} = a_{4j}^{(2)} - a_{43}^{(2)}b_{3j}^{(2)}$ for $(j = 4, 5)$

$$\begin{cases} x_1 + b_{12}x_2 + b_{13}x_3 + b_{14}x_4 &= b_{15} \\ 0x_1 + x_2 + b_{23}^{(1)}x_3 + b_{24}^{(1)}x_4 &= b_{25}^{(1)} \\ 0 x_1 + 0x_2 + x_3 + b_{34}^{(2)}x_4 &= b_{35}^{(2)} \\ 0 x_1 + 0x_2 + 0x_3 + a_{44}^{(3)}x_4 &= a_{45}^{(3)} \end{cases}$$

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The Gaussian Method (cont'd)

We have reduced system to an equivalent system with a triangular matrix (left) that can easily be solved (right):

$$x_1 + b_{12}x_2 + b_{13}x_3 + b_{14}x_4 = b_{15}$$

$$x_2 + b_{23}^{(1)}x_3 + b_{24}^{(1)}x_4 = b_{25}^{(1)}$$

$$x_3 + b_{34}^{(2)}x_4 = b_{35}^{(2)}$$

$$a_{44}^{(3)}x_4 = a_{45}^{(3)}$$



$$x_{4} = a_{45}^{(4)} / a_{44}^{(4)}$$

$$x_{3} = b_{35}^{(2)} - b_{34}^{(2)} x_{4}$$

$$x_{2} = b_{25}^{(1)} - b_{24}^{(1)} x_{4} - b_{23}^{(1)} x_{3}$$

$$x_1 = b_{15} - b_{14}x_4 - b_{13}x_3 - b_{12}x_2$$

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The Gaussian Method (cont'd)

How many separate arithmetical operations does elimination require, for n equations in Type equation here.

$$\begin{cases} a_{11}x_1 + \cdots + a_{1n}x_n = b_1 & a_{11} - \cdots - a_{1n} & b_1 \\ \vdots & \vdots & \vdots & = \vdots & \vdots & \vdots \\ a_{n1}x_1 + \cdots + a_{nn}x_n = b_n & a_{n1} - \cdots - a_{nn} & b_n \end{cases}$$

$$\frac{\text{Left Side}}{\text{Right Side}} \frac{(n^2 - n) + ((n - 1)^2 - (n - 1)) + \dots + 1 = \frac{n^3 - n}{3}}{(n - 1) + (n - 2) + \dots + 1 = \frac{n(n - 1)}{2}}$$

$$\frac{\text{Solution}}{\text{Total}} \frac{1 + \dots + (n - 1) + n = \frac{n(n + 1)}{2}}{\frac{n^3 - n}{3} + n^2 + n} \approx \frac{1}{3}n^3$$

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The Gaussian Method (cont'd)

Example.

$$2.0x_1 + 1.0x_2 - 0.1x_3 + 1.0x_4 = 2.7,$$

$$0.4x_1 + 0.5x_2 + 4.0x_3 - 8.5x_4 = 21.9,$$

$$0.3x_1 - 1.0x_2 + 1.0x_3 + 5.2x_4 = -3.9,$$

$$1.0x_1 + 0.2x_2 + 2.5x_3 - 1.0x_4 = 9.9,$$

Dividing the first equation of the system by $a_{11} = 2$, we get $x_1 + 0.5 x_2 - 0.05 x_3 = 1.35$

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The Gaussian Method (cont'd)

Compute the coefficients $a_{ij}^{(1)}$ and form the new system. For i=2, we have

$$a_{22}^{(1)} = a_{22} - a_{21}b_{12} = 0.5 - 0.4 \times 0.5 = 0.3,$$

$$a_{23}^{(1)} = a_{23} - a_{21}b_{13} = 4 + 0.4 \times 0.05 = 4.02,$$

$$a_{24}^{(1)} = a_{24} - a_{21}b_{14} = -8.5 - 0.4 \times 0.5 = -8.7,$$

$$a_{25}^{(1)} = a_{25} - a_{21}b_{15} = 21.9 - 0.4 \times 1.35 = 21.36.$$

For i = 3 and 4 computations are performed in a similar way.

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The Gaussian Method (cont'd)

Thus, we get a system in three unknowns:

$$0.3x_2 + 4.02x_3 - 8.7x_4 = 21.36,$$

-1.15x₂ + 1.015x₃ + 5.05x₄ = -4.305,
-0.3x₂ + 2.55x₃ - 1.5x₄ = 8.55.

Dividing the first equation of the obtained system by $a_{22}^{(1)}=0.3$ we get

$$x_2 + 13.4 x_3 - 29.00 x_4 = 71.20$$

$$b_{23}^{(1)} = 13.40;$$
 $b_{24}^{(1)} = -29.00;$ $b_{25}^{(1)} = 71.20$

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The Gaussian Method (cont'd)

Compute the coefficients $a_{ij}^{(2)}$ and form the new system. For i=3, we have

$$a_{33}^{(2)} = a_{33}^{(1)} - a_{32}^{(1)}b_{23}^{(1)} = 1.015 + 1.15 \times 13.40 = 16.425,$$

$$a_{34}^{(2)} = a_{34}^{(1)} - a_{32}^{(1)} b_{24}^{(1)} = 5.05 - 1.15 \times 29.00 = -28.300,$$

$$a_{35}^{(2)} = a_{35}^{(1)} - a_{32}^{(1)} b_{25}^{(1)} = -4.305 + 1.15 \times 71.20 = 77.575.$$

For i=4 computations are performed in a similar way. We get a system in two unknowns:

$$16.425x_3 - 28300x_4 = 77.575,$$

$$6.570x_3 - 10.200x_4 = 29.910.$$

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The Gaussian Method (cont'd)

Dividing the first equation of the obtained system by $a_{33}^{(2)}=16.425$ we get:

$$x_3 - 1.7229 x_4 = 4.72298$$

 $b_{34}^{(2)} = -1.72298; b_{35}^{(2)} = 4.72298$

Find the coefficients $a_{4j}^{(3)}$:

$$a_{44}^{(3)} = a_{44}^{(2)} - a_{43}^{(2)}b_{34}^{(2)} = -10.200 + 6.570 \times 1.72298 = 1.11998,$$

$$a_{45}^{(3)} = a_{45}^{(2)} - a_{43}^{(1)}b_{35}^{(2)} = 29.910 - 6.570 \times 4.72298 = -1.11998$$

Write one equation in one unknown:

$$1.11998 x_4 = -1.11998$$

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The Gaussian Method (cont'd)

The equivalent system is

$$x1 + 0.5x_2 - 0.05x_3 + 0.5x_4 = 1.35$$

 $x_2 + 13.40x_3 - 29.00x_4 = 71.20$
 $x_3 - 1.72298x_4 = 4.72298$
 $1.11998x_4 = -1.11998$

Reverse procedure:

$$x_4 = -1.00000,$$

 $x_3 = 4.72298 - 1.72298 = 3.00000,$
 $x_2 = 71.20 - 13.40 \times 3 + 29.0 = 2.00000,$
 $x_1 = 1.35 - 0.5 \times 2 + 0.05 \times 3 + 0.5 = 1.00000.$

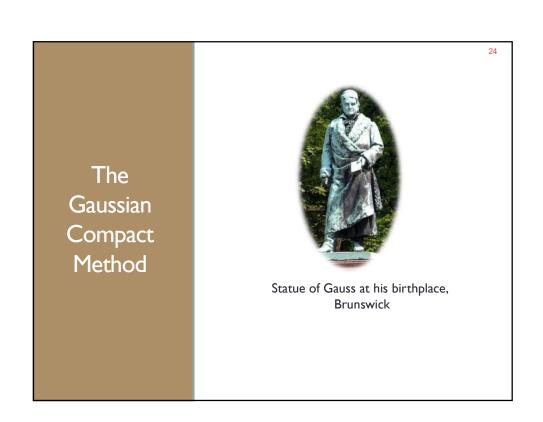
Numerical Computations - Chapter #3: Solving Systems of Linear Equations The Gaussian Method (cont'd) The computations are checked by so-called "check sums". The system equation: For each $1 \le j \le n$, let Ax = b $a_{1(n+1)}$ $a_{i(n+2)} = \sum_{i=1}^{n} a_{ij} + a_{i(n+1)}$ b=

$$\sum_{j=1}^{a_{n}} a_{tj} + a_{t(n+1)}$$

After finding the value of unknown variables $x_1, ..., x_n$, we obtain

$$\sum_{j=1}^{n} a_{ij}(x_j+1) = \sum_{j=1}^{n} a_{ij}x_j + \sum_{j=1}^{n} a_{ij} = \sum_{j=1}^{n} a_{ij} + a_{i(n+1)} = a_{i(n+2)}.$$

for each $1 \le i \le n$. Then in the absence of errors in the computations, the sums of elements of the rows of the matrix of original system including the constant terms serves as a check on this direct procedure!!!

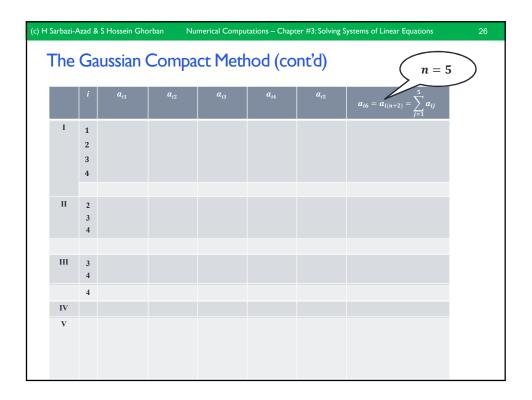




If computations are performed on a key computer by means of the scheme of unique division, then much time is spent on recording intermediate results. The Gaussian compact scheme provides an economical method of recording.

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Let us consider the sequence of forming this scheme for previous system. All the results of computations we shall enter into one table.



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The Gaussian Compact Method (cont'd)

Consider a system of 4 equations:

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + a_{14}x_4 = a_{15} \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + a_{24}x_4 = a_{25} \\ a_{31}x_1 + a_{32}x_2 + a_{33}x_3 + a_{34}x_4 = a_{35} \\ a_{41}x_1 + a_{42}x_2 + a_{43}x_3 + a_{44}x_4 = a_{45} \end{cases}$$

Direct procedure: Write down the coefficients of the given system in four rows and five columns in the following table:

	i	a _{i1}	A_{i2}	A_{i3}	a _{i4}	a_{i5}	$a_{i6} = \sum_{j=1}^5 a_{ij}$
I	1	a ₁₁	a_{12}	a ₁₃	a ₁₄	a ₁₅	$\sum_{j=1}^{5} a_{ij}$
	2	a ₂₁	a_{22}	a ₂₃	a ₂₄	a ₂₅	$\sum_{j=1}^5 a_{2j}$
	3	a ₃₁	a ₃₂	a ₃₃	a ₃₄	a ₃₅	$\sum_{j=1}^{5} a_{3j}$
	4	a ₄₁	a_{42}	a ₄₃	a ₄₄	a_{45}	$\sum_{j=1}^{5} a_{4j}$

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The Gaussian Compact Method (cont'd)

Divide all the numbers of the first row by a_{11} and enter the results into the fifth row of the given system of equations:

	i	a_{i1}	a_{i2}	a_{i3}	a _{i4}	a _{i5}	a _{i6}
I	1	a_{11}	a_{12}	a_{13}	a_{14}	a ₁₅	$\sum a_{1j} = a_{16}$
	2	a_{21}	a_{22}	a_{23}	a_{24}	a_{25}	$\sum \mathbf{a_{2j}} = \mathbf{a_{26}}$
	3	a_{31}	a ₃₂	a_{33}	a_{34}	a_{35}	$\sum a_{3j} = a_{36}$
	4	a_{41}	a_{42}	a_{43}	a_{44}	a ₄₅	$\sum a_{4j} = a_{46}$
		1	b_{12}	b_{13}	b_{14}	b_{15}	$b_{16} = a_{16}/a_{11}$
b_{1}	$j = \frac{a}{a}$	<u>u_{1j}</u> for	j = 2, 3, 4	7 4, 5		$\frac{a_{16}}{a_{11}} = \frac{a_{11} + \sum_{i=1}^{n} a_{i}}{a_{i}}$	$\sum_{j=2}^{5} a_{1j} = 1 + \sum_{j=2}^{5} b_{1j}$

Compute $\sum b_{1j}$ and carry out a check. If the computations are performed with a constant number of decimal digits, then should not differ by more than one unit of the last retained digit; otherwise, check the operations of the last step.

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The Gaussian Compact Method (cont'd)

Using this formula, compute the coefficients

$$a_{ij}^{(1)} = a_{ij} - a_{i1}b1j$$
 $(i = 2, 3, 4; j = 2, 3, 4, 5)$

	i	a_{i1}	a_{i2}	a_{i3}	a_{i4}	a _{i5}	a_{i6}
I	1	a ₁₁	a ₁₂	a_{13}	a_{14}	a ₁₅	$\sum a_{1j} = a_{16}$
	2	a_{21}	a ₂₂	a_{23}	a_{24}	a ₂₅	$\sum a_{2j} = a_{26}$
	3	a_{31}	a_{32}	a_{33}	a_{34}	a_{35}	$\sum a_{3j} = a_{36}$
	4	a_{41}	a_{42}	a_{43}	a_{44}	a_{45}	$\sum \mathbf{a_{4j}} = \mathbf{a_{46}}$
		1	b_{12}	b_{13}	b_{14}	b_{15}	$a_{16}/a_{11} = b_{16}$
II	2		$a_{22}^{(1)}$	$a_{23}^{(1)}$	$a_{24}^{(1)}$	$a_{25}^{(1)}$	$a_{26}^{(1)}$
	3		$a_{32}^{(1)} \ a_{42}^{(1)}$	$a^{(1)}_{23} \ a^{(1)}_{33} \ a^{(1)}_{43}$	$a_{34}^{(1)}$	$a^{(1)}_{25} \ a^{(1)}_{35} \ a^{(1)}_{45}$	$a_{36}^{(1)}$
	4		a ₄₂	$a_{43}^{(1)}$	$a_{44}^{(1)}$	$a_{45}^{(1)}$	$a_{46}^{(1)}$

Make a check. The sum of the elements of each row

 $\sum a_{ij}^{(1)}(i=2,3,4)$ must not differ from $a_{i6}^{(1)}$ by more than one unit of the last retained digit.

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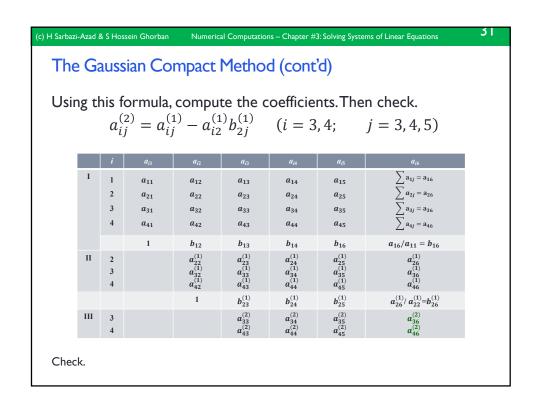
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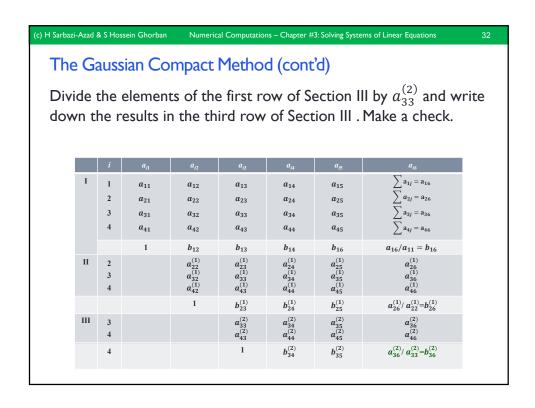
The Gaussian Compact Method (cont'd)

Divide all elements of the first row of Section II by $a_{22}^{(1)}$ and write down the results in the fourth row of last part:

	i	a_{i1}	a_{i2}	a_{i3}	a_{i4}	a ₁₅	a _{i6}
I	1	a ₁₁	a_{12}	a ₁₃	a_{14}	a_{15}	$\sum a_{1j} = a_{16}$
	2	a_{21}	a_{22}	a_{23}	a_{24}	a_{25}	$\sum a_{2j} = \mathbf{a_{26}}$
	3	a_{31}	a_{32}	a_{33}	a_{34}	a_{35}	$\sum a_{3j} = a_{36}$
	4	a_{41}	a_{42}	a ₄₃	a_{44}	a_{45}	$\sum a_{4j} = a_{46}$
		1	b_{12}	b_{13}	b_{14}	b_{16}	$a_{16}/a_{11} = b_{16}$
II	2 3 4		$a_{22}^{(1)} \ a_{32}^{(1)} \ a_{42}^{(1)}$	$a^{(1)}_{23} \ a^{(1)}_{33} \ a^{(1)}_{43}$	$a^{(1)}_{24} \ a^{(1)}_{34} \ a^{(1)}_{44}$	$a_{25}^{(1)} \ a_{35}^{(1)} \ a_{45}^{(1)}$	$a_{26}^{(1)} \ a_{36}^{(1)} \ a_{46}^{(1)}$
			1	$b_{23}^{(1)}$	$b_{24}^{(1)}$	$b_{25}^{(1)}$	$a_{26}^{(1)}/a_{22}^{(1)}=b_{26}^{(1)}$

Check.





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The	Gau	ssian C	ompaci	t Metho	od (con	ıt'd)		
			•			,		
Com	pute	$a_{4j}^{(3)}$ and	d enter	the res	ults in S	Section	IV.	
	i	a_{i1}	a_{i2}	a_{i3}	a_{i4}	a_{i5}	a_{i6}	
I	1	a ₁₁	a_{12}	a_{13}	a_{14}	a ₁₅	$\sum a_{1j} = a_{16}$	
	2	a_{21}	a_{22}	a_{23}	a_{24}	a_{25}	$\sum a_{2j} = a_{26}$ $\sum a_{3j} = a_{36}$	
	3	a ₃₁	a_{32}	a_{33}	a_{34}	a_{35}		
	4	a_{41}	a_{42}	a_{43}	a_{44}	a_{45}	$\sum a_{4j} = a_{46}$	
		1	b_{12}	b_{13}	b_{14}	b_{16}	$a_{16}/a_{11} = b_{16}$	
II	2		$a_{22}^{(1)}$	$a_{23}^{(1)}$	$a_{24}^{(1)}$	$a_{25}^{(1)}$	$a_{26}^{(1)}$	
	3 4		$a^{(1)}_{22} \ a^{(1)}_{32} \ a^{(1)}_{42}$	$a^{(1)}_{23} \ a^{(1)}_{33} \ a^{(1)}_{43}$	$a^{(1)}_{24} \ a^{(1)}_{34} \ a^{(1)}_{44}$	$a_{25}^{(1)} \ a_{35}^{(1)} \ a_{45}^{(1)}$	$a^{(1)}_{26} \ a^{(1)}_{36} \ a^{(1)}_{46}$	
	4		1	$b_{23}^{(1)}$	$b_{24}^{(1)}$	$b_{25}^{(1)}$	$a_{46}^{(1)}/a_{22}^{(1)}=b_{26}^{(1)}$	
ш			•					
111	3			$a_{33}^{(2)} \ a_{43}^{(2)}$	$a_{34}^{(2)} \ a_{44}^{(2)}$	$a_{35}^{(2)} \ a_{45}^{(2)}$	$a^{(2)}_{36} \ a^{(2)}_{46}$	
	4			1	b ₃₄ ⁽²⁾	b ₃₅ ⁽²⁾	$a_{36}^{(2)}/a_{33}^{(2)}=b_{36}^{(2)}$	
IV					$a_{44}^{(3)}$	$a_{45}^{(3)}$	$a_{46}^{(3)}$	
					44	45	4 46	

The Gaussian Compact Method (cont'd)

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Reverse procedure

1. Write down the unities in Section V as is indicated in Table.

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

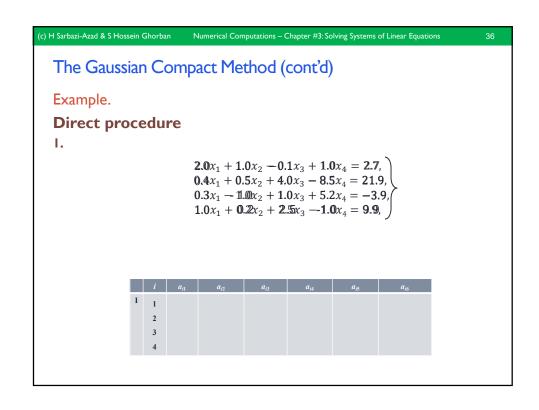
- 2. Compute x_4 .
- 3. For computing the values x_3 , x_2 , x_1 use only the rows of Sections I, II, III containing units, beginning with the last.

Thus, to compute x_3 multiply x_4 by $b_{34}^{(2)}$ and subtract the result from $b_{35}^{(2)}$. The units put in Section V help us to find the corresponding coefficients for x_i for (i=3,2,1) in the marked rows.

$$x_3 = b_{35}^{(2)} - b_{34}^{(2)} x_4$$

Numerical Computations – Chapter #3: Solving Systems of Linear Equations The Gaussian Compact Method (cont'd) Reverse procedure (cont'd) 4. Compute x_2 for which purpose use the elements of the marked row of Section II: $x_2 = b_{25}^{(1)} - b_{24}^{(1)} x_4 - b_{23}^{(1)} x_3$ 5. Compute x_1 for which purpose use the elements of the marked row of Section I: $x_1 = b_{15} - b_{14}x_4 - b_{13}x_3 - b_{12}x_2$ In the check scheme the reverse procedure is carried out in a similar way. The solutions of this scheme must differ from those of the given scheme by I. $\sum_{i=1}^{n} a_{ij}(x_j+1) = \sum_{i=1}^{n} a_{ij}x_j + \sum_{i=1}^{n} a_{ij}$ $\overline{x}_i = x_i + 1. (i = 1, 2, 3, 4)$

 $= \sum a_{ij} + a_{i(n+1)} = a_{i(n+2)}.$



Numerical Computations – Chapter #3: Solving Systems of Linear Equations

The Gaussian Compact Method (cont'd)

2. Compute the sums of the coefficients along the row.

$$\sum_{i=1}^{5} a_{1j} = 2.0 + 1.0 - 0.1 + 1.0 + 2.7 = 6.6$$

	i	a_{iI}	a_{i2}	a _{i3}	a _{i4}	a_{i5}	a_{i6}
I	1	2.0	1.0	-0.1	1.0	2.7	6.6
	2	0.4	0.5	4.0	-8.5	21.9	18.3
	3	0.3	-1.0	1.0	5.2	-3.9	1.6
	4	1.0	0.2	2.5	-1.0	9.9	12.6
		1	0.50	-0.05	0.50	1.35	3.30

- 3. Divide all the numbers of the first row by $a_{11}=2.0$ and enter the results into the fifth row of Section 1.
- 4. Checking: computing the sum of the first five numbers obtained in (3), we get 3.30, which completely coincides with the number obtained in the last column.

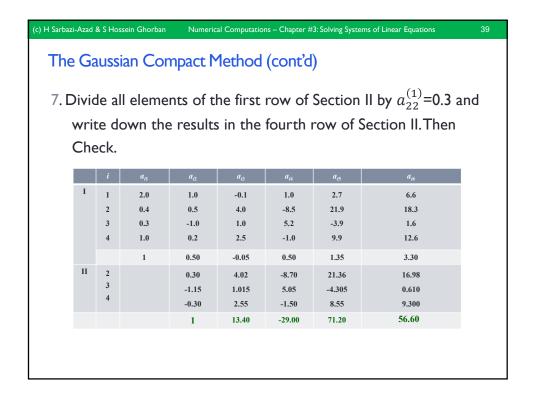
Numerical Computations – Chapter #3: Solving Systems of Linear Equations

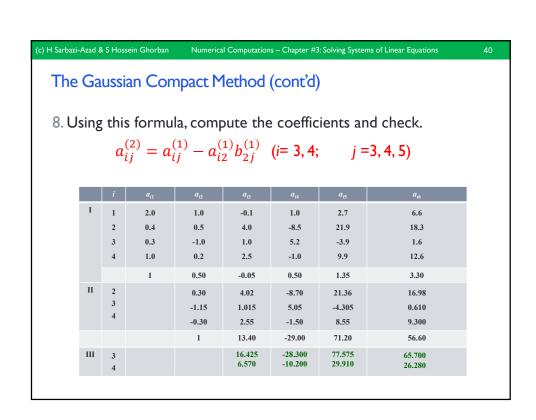
The Gaussian Compact Method (cont'd)

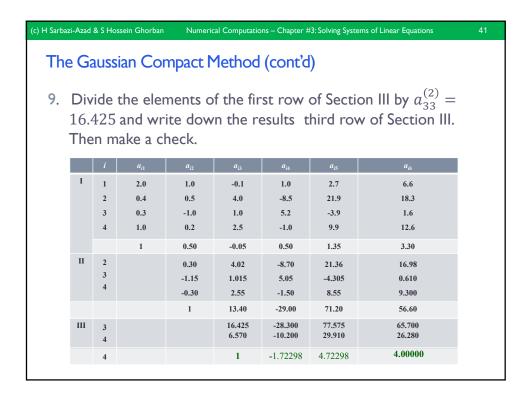
5. Using this formula, compute the coefficients
$$a_{ij}^{(1)}=aij-ai_1b_{1j} \ (i=2,3,4; j=2,3,4,5)$$

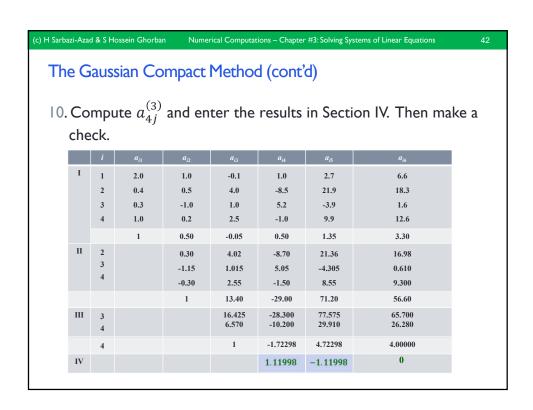
	a_{i1}	a_{i2}	a_{i3}	a_{i4}	a _{i5}	a_{i1}	a_{i6}
I	1	2.0	1.0	-0.1	1.0	2.7	6.6
	2	0.4	0.5	4.0	-8.5	21.9	18.3
	3	0.3	-1.0	1.0	5.2	-3.9	1.6
	4	1.0	0.2	2.5	-1.0	9.9	12.6
		1	0.50	-0.05	0.50	1.35	3.30
II			0.30	4.02	-8.70	21.36	16.98
	3		-1.15	1.015	5.05	-4.305	0.610
			-0.30	2.55	-1.50	8.55	9.300

6. Make a check.









The Gaussian Compact Method (cont'd)

Reverse procedure

Following the sequence of operations in the reverse procedure, we get the values of the unknowns as $x_4 = -1.0000; \\ x_2 = 2.0000; \\ x_1 = 1.0000;$ The solution of the check system: $\bar{x}_4 = 0.00000; \\ \bar{x}_3 = 3.00000; \\ x_1 = 3.00000; \\ x_2 = 3.00000; \\ x_3 = 3.00000; \\ x_4 = 0.00000; \\ \bar{x}_3 = 3.00000; \\ \bar{x}_3 = 3.00000$

 $\bar{x_1} = 2.00000;$

The Modification of Crout-Doolittle



Alan Mathison Turing (1912-1954)

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

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The Modification of Crout-Doolittle

By the Gaussian method, we obtain

$$x_1 + b_{12}x_2 + b_{13}x_3 + b_{14}x_4 = b_{15}$$

$$x_2 + b_{23}^{(1)}x_3 + b_{24}^{(1)}x_4 = b_{25}^{(1)}$$

$$x_3 + b_{34}^{(2)}x_4 = b_{35}^{(2)}$$

$$a_{44}^{(3)}x_4 = a_{45}^{(3)}$$

which gives us the value of x_1, \dots, x_4 :

$$x_4 = a_{45}^{(3)} / a_{44}^{(3)}$$

$$x_3 = b_{35}^{(2)} - b_{34}^{(2)} x_4$$

$$x_2 = b_{25}^{(1)} - b_{24}^{(1)} x_4 - b_{23}^{(1)} x_3$$

$$x_1 = b_{15} - b_{14} x_4 - b_{13} x_3 - b_{12} x_2$$

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Numerical Computations – Chapter #3: Solving Systems of Linear Equations

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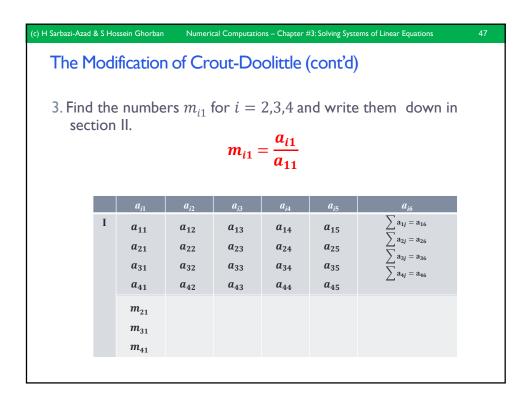
The Modification of Crout-Doolittle

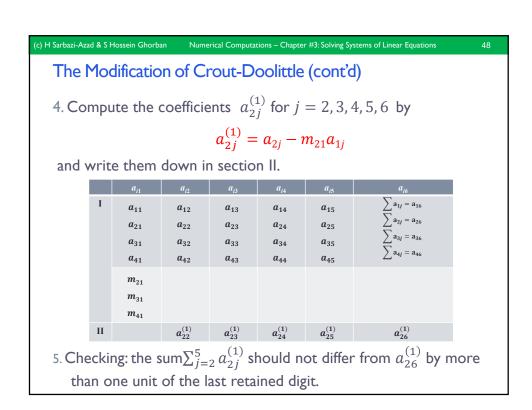
Direct procedure

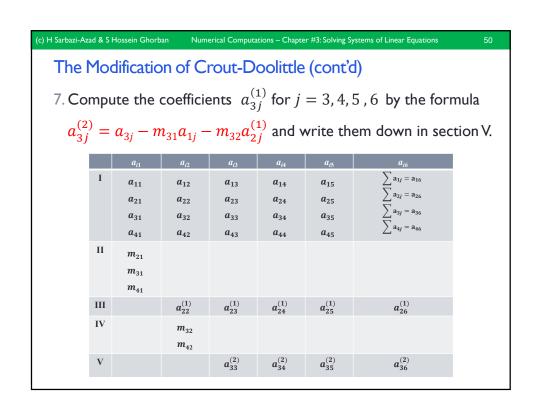
I. Write the coefficients of the system a_{ij} for $i=1,2,3,4;\ j=1,2,3,4,5$ in Section I of Table.

	a_{i1}	a_{i2}	a_{i3}	a_{i4}	a _{i5}	a_{i6}
I	a_{11}	a ₁₂	a ₁₃	a ₁₄	a ₁₅	$\sum a_{1j} = a_{16}$
	a_{21}	a_{22}	a_{23}	a_{24}	a_{25}	$\sum \mathbf{a}_{2j} = \mathbf{a}_{26}$
	a_{31}	a_{32}	a_{33}	a_{34}	a_{35}	$\sum_{i} a_{3j} = a_{36}$
	a_{41}	a_{42}	a_{43}	a_{44}	a_{45}	$\sum_{i} a_{4j} = a_{46}$
						_

2. Sum the coefficients of each row and enter the results in the sum column as a_{i6} (i=1,2,3,4)







Numerical Computations – Chapter #3: Solving Systems of Linear Equations The Modification of Crout-Doolittle (cont'd) 8. Checking: compare $\sum_{j=3}^{5} a_{3j}^{(2)}$ with the number $a_{36}^{(2)}$ a_{11} a_{12} a_{13} a_{14} a_{15} a_{21} a_{22} a_{23} a_{24} a_{25} a_{32} a_{33} a_{35} a_{31} a_{41} a_{42} a_{43} a_{44} a_{45} m_{21} m_{31} m_{41} $a_{26}^{(1)}$ Ш $a_{22}^{(1)}$ $a_{23}^{(1)}$ $a_{24}^{(1)}$ $a_{25}^{(1)}$ IV m_{32} m_{42} V $a_{36}^{(2)}$ $a_{33}^{(2)}$ $a_{34}^{(2)}$ $a_{35}^{(2)}$

The Modification of Crout-Doolittle (cont'd)

9. Find the numbers

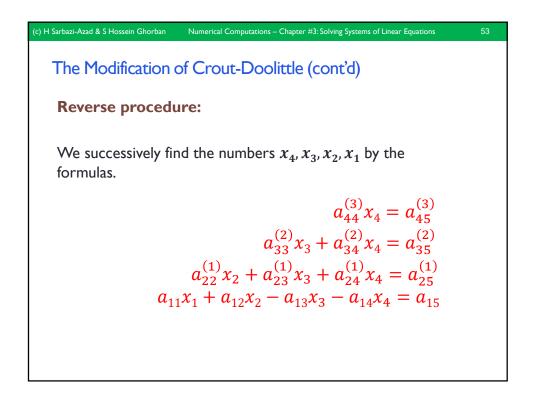
$$m_{43} = \frac{\left(a_{43} - m_{41}a_{13} - m_{42}a_{23}^{(1)}\right)}{a_{33}^{(2)}}$$

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

10. Find the coefficients $a_{4j}^{(3)}$ for j=4,5, 6 by the following formula and write them down in section VII. $a_{4j}^{(3)}=a_{4j}-m_{41}a_{1j}-m_{42}a_{2j}^{(1)}-m_{43}a_{3j}^{(2)}$

$$a_{4j}^{(3)} = a_{4j} - m_{41}a_{1j} - m_{42}a_{2j}^{(1)} - m_{43}a_{3j}^{(2)}$$

II. Checking: the sum $a_{44}^{(3)}$ + $a_{45}^{(3)}$ with the number $a_{46}^{(3)}$.



(c) H Sar	bazi-Aza	d & S Hosseir	n Ghorban	Numerical C	Computations –	Chapter #3: So	olving Systems of Linear Equation	ons 54
Т	he I	Modific	cation c	of Crou	ıt-Dool	ittle (co	ont'd)	
		a_{i1}	a_{i2}	a_{i3}	a_{i4}	a_{i5}	a_{i6}	
	I	a_{11}	a_{12}	a_{13}	a_{14}	a_{15}	$\sum a_{1j} = a_{16}$	
		a_{21}	a_{22}	a_{23}	a_{24}	a_{25}	$\sum a_{2j} = a_{26}$	
		a_{31}	a_{32}	a_{33}	a_{34}	a_{35}	$\sum a_{3j} = a_{36}$	
		a_{41}	a_{42}	a_{43}	a_{44}	a_{45}	$\sum a_{4j} = a_{46}$	
	II	$m_{21} \ m_{31} \ m_{41}$						
	III		$a_{22}^{(1)}$	$a_{23}^{(1)}$	$a_{24}^{(1)}$	$a_{25}^{(1)}$	$a_{26}^{(1)}$	$\sum_{n=1}^{\infty} a_{n}(x+1)$
	IV		$m_{32} \ m_{42}$					$\begin{vmatrix} \sum_{j=1}^{n} a_{ij}(x_j+1) \\ = \sum_{j=1}^{n} a_{ij}x_j \\ + \sum_{j=1}^{n} a_{ij} \\ = \sum_{j=1}^{n} a_{ij} + a_{i(n+1)} \\ = a_{i(n+2)}. \end{vmatrix}$
	V			$a_{33}^{(2)}$	$a_{34}^{(2)}$	$a_{35}^{(2)}$	$a_{36}^{(2)}$	$=\sum_{j=1}^{\infty}a_{ij}x_{j}$
	VI			m_{43}				$+\sum_{i=1}^{n}a_{ii}$
	VII				$a_{44}^{(3)}$	$a_{45}^{(3)}$	$a_{46}^{(3)}$	j=1 n
	VIII					x_4	$\overline{x_4}$	$=\sum_{j=1}a_{ij}+a_{i(n+1)}$
						x_3	$\overline{x_3}$	$=a_{i(n+2)}.$
						x_2	$\overline{x_2}$	
						x_1	$\overline{x_1}$	

The Method of Principal Elements



Richard Courant (1888-1972)

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The Method of Principal Elements

Example. Solve a linear system of equations where

$$\begin{cases} 0.0003x_1 + 1.566x_2 = 1.569 \\ 0.3454x_1 - 2.436x_2 = 1.018 \end{cases}$$

Solution.

$$\begin{bmatrix} 0.0003 & 1.566 & | & 1.569 \\ 0.3454 & -2.436 & | & 1.018 \end{bmatrix} \rightarrow \begin{bmatrix} 0.0003 & 1.566 & | & 1.569 \\ 0 & -1804 & | & -1018 \end{bmatrix} \rightarrow$$

$$x_2 = \frac{-1805}{-1804} = 1.001, \quad x_1 = \frac{0.001}{0.0003} = 3.333.$$

$$\begin{bmatrix} 0.3454 & -2.436 & | & 1.018 \\ 0.0003 & 1.566 & | & 1.569 \end{bmatrix} \rightarrow \begin{bmatrix} 0.3454 & -2.436 & | & 1.018 \\ 0.0003 & 1.566 & | & 1.569 \end{bmatrix} \rightarrow$$

$$x_2 = \frac{1.568}{1.568} = 1,$$
 $x_1 = \frac{3.454}{0.3454} = 10$

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

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The Method of Principal Elements

Example. Solve a linear system of equations where

$$\begin{cases} 10^{-5}x_1 + x_2 = 1 \\ x_1 + x_2 = 2 \end{cases}$$

Solution.

$$\begin{bmatrix} 10^{-5} & 1 & | & 1 \\ 1 & 1 & | & 1 \end{bmatrix} \stackrel{R_2 - \frac{10^5}{R_1}}{\overset{\frown}{\hookrightarrow}} \begin{bmatrix} 10^{-5} & 1 & | & 1 \\ 0 & (1 - 10^5) & | & 2 - 10^5 \end{bmatrix} \xrightarrow{}$$

$$x_2 = \frac{2 - 10^5}{1 - 10^5} = 1 + \frac{1}{1 - 10^5} = 0.9999899999 \approx 1$$

$$x_1 = 1 - \left(1 + \frac{1}{1 - 10^5}\right) = \frac{1}{1 - 10^5} = -0.0000100001 \approx 0$$

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The Method of Principal Elements

Example (cont'd).

$$\begin{cases} 10^{-5}x_1 + x_2 = 1 \\ x_1 + x_2 = 2 \end{cases}$$

Solution (cont'd).

$$\begin{bmatrix} 1 & 1 & | & 2 \\ \mathbf{10}^{-5} & 1 & | & 2 \end{bmatrix} \overset{R_2 - \mathbf{10}^{-5} R_1}{\overset{\frown}{\hookrightarrow}} \begin{bmatrix} & 1 & 1 & | & 2 \\ 0 & 1 - \mathbf{10}^{-5} & | & 1 - 2 \times \mathbf{10}^{-5} \end{bmatrix} \xrightarrow{}$$

$$x_2 = \frac{1 - 2 \times 10^{-5}}{1 - 10^{-5}} = 1 - \frac{10^{-5}}{1 - 10^{-5}}, = 1 - 0.0000100001 \approx 1$$

$$x_1 = 2 - 1 - 0.0000100001 \approx 1$$

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The Method of Principal Elements

Example (cont'd).

$$\begin{cases} 10^{-5}x_1 + x_2 = 1 \\ x_1 + x_2 = 2 \end{cases}$$

As a result:

$$\begin{bmatrix} 10^{-5} & 1 & | & 1 \\ 1 & 1 & | & 1 \end{bmatrix} \stackrel{R_2 - \frac{10^5}{5} \times R_1}{\stackrel{\frown}{\hookrightarrow}} \begin{bmatrix} 10^{-5} & 1 & | & 1 \\ 0 & (1 - 10^5) & | & 2 - 10^5 \end{bmatrix} \rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \approx \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & | & 2 \\ 10^{-5} & 1 & | & 2 \end{bmatrix} \stackrel{R_2 - \frac{10^{-5}}{5} \times R_1}{\stackrel{\frown}{\hookrightarrow}} \begin{bmatrix} & 1 & 1 & | & 2 \\ 0 & 1 - 10^{-5} & | & 1 - 2 \times 10^{-5} \end{bmatrix} \rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \approx \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

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The Method of Principal Elements (cont'd)

Consider a linear system of equations

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = a_{1,n+1} \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = a_{2,n+1} \\ & \vdots \\ a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = a_{n,n+1} \end{cases}$$

Write the augmented rectangular matrix consisting of the coefficients of the system

$$M = \begin{pmatrix} a_{11} \dots a_{1q} \dots a_{1n} a_{1,n+1} \\ \dots & \dots & \dots \\ a_{p1} \dots a_{pq} \dots a_{pn} a_{p,n+1} \\ \dots & \dots & \dots \\ a_{n1} \dots a_{nq} \dots a_{nn} a_{n,n+1} \end{pmatrix}$$

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

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The Method of Principal Elements (cont'd)

Direct procedure

Choose a nonzero (as a rule, the numerically largest) element a_{pq} (of matrix M not belonging to the column of constant terms $(q \neq n+1)$) this element being called the principal element. Compute the multipliers

$$mi1 = \frac{a_{iq}}{a_{pq}} \quad (i \neq p)$$

The row of M with index p which contains the principal element is called the <u>principal row</u>.

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The Method of Principal Elements (cont'd)

- I. From each *ith* non-principal row subtract term wise the principal row multiplied by m_i .
- 2. Obtain a new matrix in which all the elements of the qth column (with the exception of a_{pq}) are equal to zero.
- 3. Discarding this column and the principal row, we obtain a new matrix M_1 with the number of rows and columns diminished by unity.
- 4. Repeat these operations with matrix M_1 to get matrix M_2 , and so on. Thus, we obtain a sequence of matrices M, M_1, \dots, M_{n-1}
- 5. Combine into a system all the principal rows beginning with the last.
- 6. After an appropriate replacement they form a triangular matrix which is equivalent to the initial one.

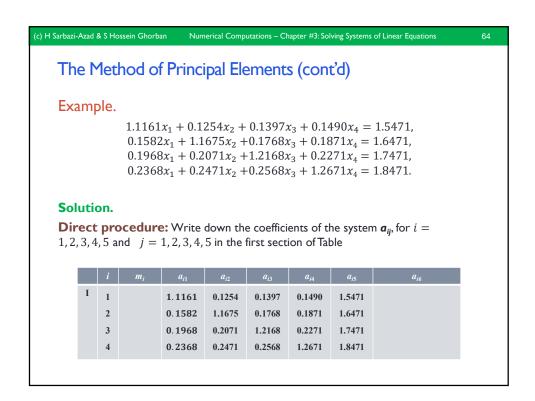


Reverse procedure

Solving the system with the obtained matrix of coefficients, we find, step by step, the values of the unknowns x_i (i = 1, 2, ..., n)

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

All the above described computations can be arranged in one table, which will be similar to the Gaussian compact scheme, with a check provided for each stage of computations.



The Method of Principal Elements (cont'd)

2. Compute the sums of the coefficients along the row. $\Sigma = ai_6$

	i	m_i	a_{il}	a_{i2}	a_{i3}	a _{i4}	a _{i5}	a_{i6}
I	1	0.11759	1.1161	0.1254	0.1397	0.1490	1.5471	3.07730
	2	0.14766	0.1582	1.1675	0.1768	0.1871	1.6471	3.33760
	3	0.17923	0.1968	0.2071	1.2168	0.2271	1.7471	3.59490
	4		0.2368	0.2471	0.2568	1.2671	1.8471	3.58490

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

3. Find the principal element. In the given system it will be the coefficient

$$a_{44} = 1.26710 (p = 4, q = 4).$$

4. Find the numbers m_i (i=l,2,3). To this end divide the elements of the column a_{i4} by a_{44} and write down the results in the column m_i Section I.

$$m_1 = \frac{a_{14}}{a_{44}} = \frac{0.14900}{1.26710} = 0.11759; \qquad m_2 = \frac{a_{24}}{a_{44}} = \frac{0.18710}{1.26710} = 0.14766; \\ m_3 = \frac{a_{34}}{a_{44}} = \frac{0.22710}{1.26710} = 0.17923$$

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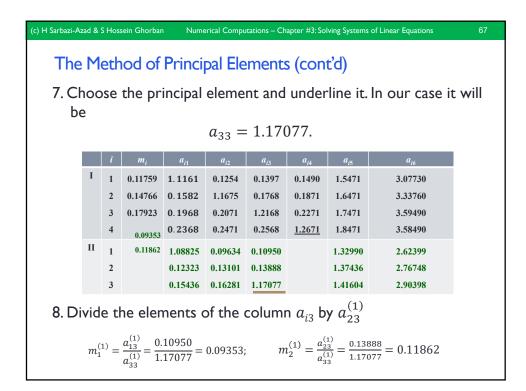
The Method of Principal Elements (cont'd)

5. Compute the coefficients of the new matrix. From each row $i\ (i=1,2,3)$ subtract the principal row multiplied by the corresponding element m_i .

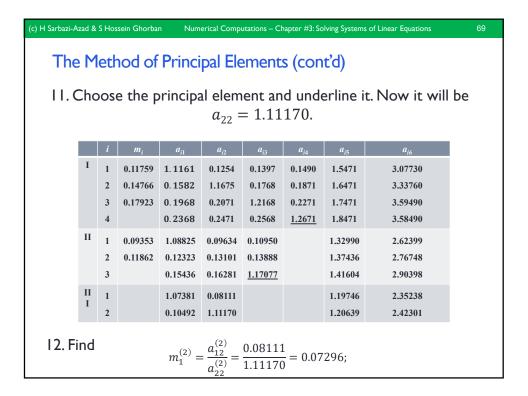
$$a_{ij}^{(1)} = aij - m_i a_{4j} (i = 2,3,4; j = 1,2,3,5,6)$$

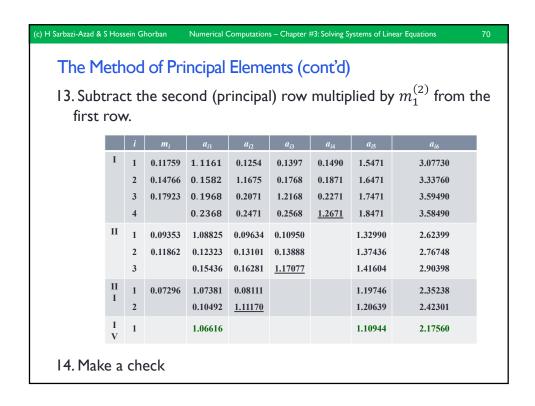
	i	m_i	a_{i1}	<i>a</i> _{i2}	<i>a</i> _{i3}	<i>a</i> _{i4}	<i>a</i> _{i5}	a_{i6}
I	1	0.11759	1.1161	0.1254	0.1397	0.1490	1.5471	3.07730
	2	0.14766	0.1582	1.1675	0.1768	0.1871	1.6471	3.33760
	3	0.17923	0.1968	0.2071	1.2168	0.2271	1.7471	3.59490
	4		0.2368	0.2471	0.2568	<u>1.2671</u>	1.8471	3.58490
II	1		1.08825	0.09634	0.10950		1.32990	2.62399
	2		0.12323	0.13101	0.13888		1.37436	2.76748
	3		0.15436	0.16281	1.17077		1.41604	2.90398

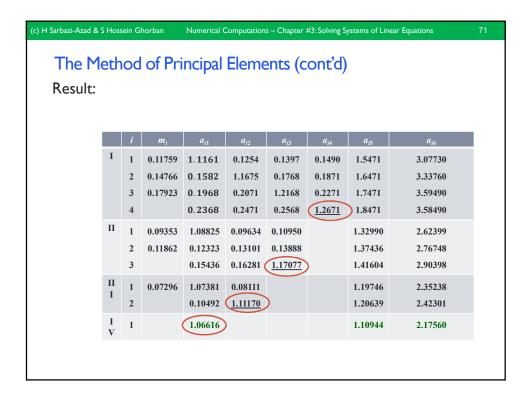
6. Make a check.

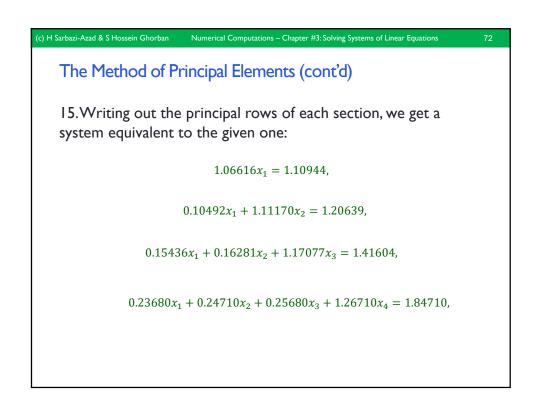


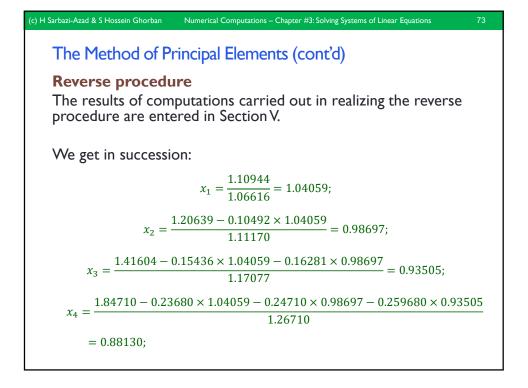
Sarbazi-Azao	1 & S F	Hosseii	n Ghorban	Numerio	cal Computa	tions – Chapi	ter #3: Solvir	ng Systems of	Linear Equations	6
The	Me	thc	od of P	rincipa	al Elen	nents	(cont	d)		
9. Co	mp	ute	the co	oefficie	nts $a_i^{(i)}$	²⁾ . For	this p	urpose	from each lir	ne
i (i =	1,	2) sub	otract 1	the pri	ncipal	row m	nultiplie	ed by the	
СО	rre	spc	nding	m_i .						
		i	m_i	a_{i1}	a_{i2}	a_{i3}	a_{i4}	a _{i5}	a _{i6}	
	I	1	0.11759	1.1161	0.1254	0.1397	0.1490	1.5471	3.07730	
		2	0.14766	0.1582	1.1675	0.1768	0.1871	1.6471	3.33760	
		3	0.17923	0.1968	0.2071	1.2168	0.2271	1.7471	3.59490	
		4		0.2368	0.2471	0.2568	<u>1.2671</u>	1.8471	3.58490	
	II	1	0.09353	1.08825	0.09634	0.10950		1.32990	2.62399	
		2	0.11862	0.12323	0.13101	0.13888		1.37436	2.76748	
		3		0.15436	0.16281	<u>1.17077</u>		1.41604	2.90398	
	II I	1		1.07381	0.08111			1.19746	2.35238	
	1	2		0.10492	1.11170			1.20639	2.42301	











) H Sarbaz	i-Aza	d & S F	Hossein Ghor	ban Nui	merical Comp	outations – Ch	apter #3: Solv	ving Systems of	Linear Equations	74
Th	The Method of Principal Floments (cont'd)									
- 11	The Method of Principal Elements (cont'd)									
	v	ı	m_i	a_{i1}	a_{i2}	<i>a</i> _{i3}	a_{i4}	a_{i5}	a_{i6}	
	I	1	0.11759	1.1161	0.1254	0.1397	0.1490	1.5471	3.07730	
		2	0.14766	0.1582	1.1675	0.1768	0.1871	1.6471	3.33760	
		3	0.17923	0.1968	0.2071	1.2168	0.2271	1.7471	3.59490	
		4		0.2368	0.2471	0.2568	<u>1.2671</u>	1.8471	3.58490	
	II	1	0.09353	1.08825	0.09634	0.10950		1.32990	2.62399	
		2	0.11862	0.12323	0.13101	0.13888		1.37436	2.76748	
		3		0.15436	0.16281	1.17077		1.41604	2.90398	
1	Ш	1	0.07296	1.07381	0.08111			1.19746	2.35238	
		2		0.10492	1.11170			1.20639	2.42301	
]	IV	1		1.06616				1.10944	2.17560	
	V	1		1				1.04059	2.04059	
		2			1			0.98697	1.98697	
		3				1		0.93505	1.93505	
		4					1	0.88130	1.88130	

The Scheme of Khaletsky

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

The Scheme of Khaletsky

Consider a system of linear equations written in matrix notation as Ax = b where $A = (a_{ij})$ is a square matrix i, j = 1, 2, ..., n and

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \dots \\ x_n \end{bmatrix} \quad b = \begin{bmatrix} a_{1,n+1} \\ a_{2,n+1} \\ \dots \\ a_{n,n+1} \end{bmatrix}$$

are column vectors.

Represent matrix A in the form of a product A = BC, where

$$B = \begin{bmatrix} b_{11} & 0 & 0 \dots 0 \\ b_{21} & b_{22} & 0 \dots 0 \\ \vdots & \vdots & \vdots & \vdots \\ b_{n1} & b_{n2} & b_{n3} \dots b_{nn} \end{bmatrix} \qquad C = \begin{bmatrix} 1 & c_{12} \dots & c_{1n} \\ 0 & 1 & \dots & c_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 1 \end{bmatrix}$$

$$C = \begin{bmatrix} 1 & c_{12} \dots & c_{1n} \\ 0 & 1 & \dots & c_{2n} \\ \dots & \dots & \dots & \dots \\ 0 & 0 & 1 \end{bmatrix}$$

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

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The Scheme of Khaletsky (cont'd)

Then the elements b_{ij} and c_{ij} are determined from the formulas

$$b_{i1}=a_{i1},$$

$$b_{ij} = a_{ij} - \sum_{k=1}^{j-1} b_{ik} c_{kj} \ (i \ge j > 1)$$

and

$$c_{1j} = a_{1j}/b_{11},$$

$$c_{ij} = \frac{1}{b_{ii}} \left(a_{ij} - \sum_{k=1}^{i-1} b_{ik} c_{kj} \ (1 < i < j) \right)$$

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Numerical Computations – Chapter #3: Solving Systems of Linear Equations

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The Scheme of Khaletsky (cont'd)

Whence the desired vector x may be computed from the chain of equations By = b, Cx = y.

Since the matrices B and C are triangular, systems are solved, namely:

$$y_1 = a_{1,n+1}/b_{11},$$

$$y_i = (a_{i,n+1} - \sum_{k=1}^{i-1} b_{ik} y_k)$$
 for $(i > 1)$

and

$$x_n = y_n,$$

$$x_i = y_i - \sum_{k=i+1}^{n} c_{ik} x_k \ (i < n)$$

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

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The Scheme of Khaletsky (cont'd)

Example.

$$3x_1 + x_2 - x_3 + 2x_4 = 6,$$

$$-5x_1 + x_2 + 3x_3 - 4x_4 = -12,$$

$$2x_1 + x_3 - x_4 = 1,$$

$$x_1 - 5x_2 + 3x_3 - 3x_4 = 3.$$

Solution.

1. Write down the matrix of the coefficients of the system, its constant terms and the check sums in Section I of Table .

	x_1	x_2	x_3	x_4		$\sum_{a_{ij}}$
I	3	1	-1	2	6	11
	-5	1	3	-4	-12	-17
	2	0	1	1	1	3
	1	-5	3	-3	3	-1

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The Scheme of Khaletsky (cont'd)

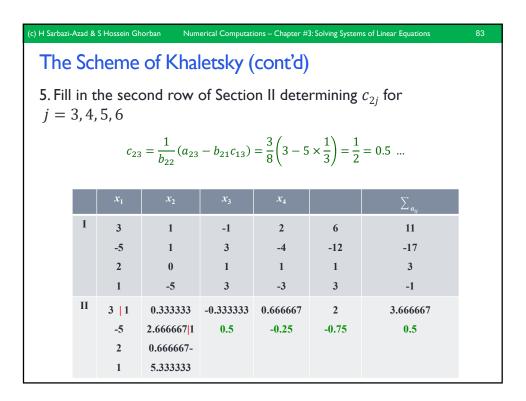
2. Transfer the elements of the column \mathbf{x}_i from Section I to Section II since

$$b_{i1} = a_{i1}; \quad i = 1, 2, 3, 4$$

	x_1	x_2	x_3	x_4		$\sum_{a_{ij}}$
I	3	1	-1	2	6	11
	-5	1	3	-4	-12	-17
	2	0	1	1	1	3
	1	-5	3	-3	3	-1
II	3					
	-5					
	2					
	1					

The Scheme of Khaletsky (cont'd) 3. Divide all elements of the first row of Section I by the element $a_{11}=b_{11}$. We have: $c_{12}=\frac{1}{3}=0.333333,\ c_{13}=-\frac{1}{3}=-0.333333,\ c_{14}=\frac{2}{3}=0.666667,\ c_{15}=\frac{6}{3}=2,\ c_{16}=\frac{11}{3}=3.666667$

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The	The Scheme of Khaletsky (cont'd)									
	4. Fill in the column x_2 of Section II beginning with the second row. We determine b_{j2} : $b_{22}=a_{22}-b_{21}c_{12}=1-\left(-5\times\frac{1}{3}\right)=\frac{8}{3}=2.6666667\dots$									
		x_1	x_2	<i>X</i> ₃	x_4		$\sum_{a_{ij}}$			
	I	3	1	-1	2	6	11			
		-5	1	3	-4	-12	-17			
		2	0	1	1	1	3			
		1	-5	3	-3	3	-1			
	II	3 1	0.333333	-	0.666667	2	3.666667			
		-5	2.666667	0.333333						
		2	0.666667-							
		1	5.333333							



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The	The Scheme of Khaletsky (cont'd)									
6. Fi	6. Fill in the column x_3 , computing its elements b_{33} and b_{43}									
		x_1	x_2	<i>x</i> ₃	x_4		$\sum_{a_{ij}}$			
	I	3	1	-1	2	6	11			
		-5	1	3	-4	-12	-17			
		2	0	1	1	1	3			
		1	-5	3	-3	3	-1			
	II	3 1	0.333333	-0.333333	0.666667	2	3.666667			
		-5	2.666667 1	0.5	-0.25	-0.75	0.5			
		2	0.666667-	2						
		1	5.333333	6						

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The Scheme of Khaletsky (cont'd)

7. Proceed analogously until Section II is filled in completely.

We thus get a staircase arrangement in Section II:

	x_1	x_2	x_3	x_4		$\sum_{a_{ij}}$
I	3	1	-1	2	6	11
	-5	1	3	-4	-12	-17
	2	0	1	1	1	3
	1	-5	3	-3	3	-1
II	3 1	0.333333	-0.333333	0.666667	2	3.666667
	-5	2.666667 1	0.5	-0.25	-0.75	0.5
	2	0.666667-	2 1	-1.25	-1.75	-2
	1	5.333333	6	2.5 1	3	4

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The Scheme of Khaletsky (cont'd)

8. Determine y_i and x_i (i=1,2,3,4), and enter them in Section III:

$$y_1 = \frac{a_{15}}{b_{11}} = \frac{6}{3} = 2;$$

$$y_2 = \frac{(a_{25} b_{21}y_1)}{b_{22}} = \frac{-12 + 5 \times 5}{2.666667} = -0.75;$$

$$y_3 = \frac{(a_{35} b_{31}y_1 - b_{32}y_2)}{b_{33}} = \frac{1 - 2 \times 2 - 0.666667 \times 0.75}{2} = -1.75;$$

$$y_4 = \frac{(a_{45} b_{41}y_1 - b_{42}y_2 b_{43}y_3)}{b_{44}} = (3 - 2 - 5.333333 \times 0.75 + 6 \times 1.75) = 3;$$

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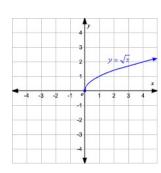
The Scheme of Khaletsky (cont'd)

8'. Determine y_i and x_i (i = 1, 2, 3, 4), and enter them in Section III:

$$\begin{aligned} x_4 &= y_4 = 3; \\ x_3 &= y_3 - C_{34} x_4 = -1.75 + 1.25 \times 3 = 2; \\ x_2 &= y_2 - C_{23} x_3 - C_{24} x_4 = -0.75 - 0.5 \times 2 + 0.25 \times 3 = -1; \\ x_1 &= y_1 - C_{12} x_2 - C_{13} x_3 - C_{14} x_4 = (2 + 0.3333333 + 0.3333333 \times 2 - 0.666667 \times 3) = 1. \end{aligned}$$

9. Intermediate checking is done by means of the I column, which is involved in the same operations as is the Σ column of constant terms.

The Square-Root Method



Numerical Computations - Chapter #3: Solving Systems of Linear Equations

The Square-Root Method

The square-root method is used for solving a linear system Ax = B

where $A = [a_{ij}]$ is a symmetric matrix, i.e.,

$$a_{ij} = a_{ji}$$
 for $i, j = 1, 2, ... n$.

If A is non-singular, then the LU decomposition is unique. Consequently, if A is symmetric, then $\mathbf{L}=U^T$, so $A=U^TU$

where
$$U = \begin{bmatrix} u_{11} & u_{12} & \cdots & u_{1n} \\ 0 & u_{22} & \cdots & u_{2n} \\ \vdots & 0 & \ddots & \vdots \\ 0 & \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & u_{nn} \end{bmatrix}$$

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The Square-Root Method(cont'd)

Multiplying together the matrices U^T and U and equating the product to the matrix A, we get:

$$a_{ij} = \left(U^T U\right)_{ij} = \sum_{k=1}^n u_{ki} u_{kj}$$

Consequently,

$$u_{11} = \sqrt{a_{11}}, \qquad u_{1j} = \frac{a_{1j}}{u_{11}} \qquad (2 < j \le n)$$

$$\sum_{k=1}^{i-1} u_{ik}^2 + u_{ii}^2 = a_{ii} \implies u_{ii} = \sqrt{a_{ii} - \sum_{k=1}^{i-1} u_{ik}^2} \quad (1 < i \le n)$$

$$\sum_{k=1}^{i} u_{ki} u_{kj} = a_{ij} \implies u_{ii} u_{ij} = a_{ii} - \sum_{k=1}^{i-1} u_{ki} u_{kj} \implies$$

$$u_{ij} = \frac{a_{ii} - \sum_{k=1}^{i-1} u_{ki} u_{kj}}{u_{ii}} \quad (i < j)$$

$$u_{ij} = 0 \qquad (i > j)$$

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

The Square-Root Method(cont'd)

Note that we consider that A is non-singular matrix, then $\det A \neq 0$.

Since

$$\det A = \det U^T \det U = (u_{11} \times \cdots \times u_{nn})^2 \neq 0$$

we have $u_{ii} \neq 0$ for each $1 \leq i \leq n$,

On finding the matrix U, we have $Ax = U^TU x = b$.

Thus, we can replace the system by two equivalent systems with triangular matrices:

$$U^T y = b$$
 and $Ux = y$

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Numerical Computations – Chapter #3: Solving Systems of Linear Equations

The Square-Root Method(cont'd)

Write new systems in the expanded form:

$$U^{T}y = b \implies \begin{cases} u_{11}y_{1} & = b_{1} \\ u_{12}y_{1} + u_{22}y_{2} & = b_{2} \\ u_{1n}y_{1} + u_{2n}y_{2} + \cdots + u_{nn}y_{n} = b_{n} \end{cases}$$

$$Ux = y \Longrightarrow \left\{ \begin{array}{c} u_{11}x_1 + u_{12}y_2 + \cdots + u_{1n}x_n = y_1 \\ u_{22}x_2 + \cdots + u_{2n}x_n = y_2 \\ \vdots \\ u_{nn}x_n = y_n \end{array} \right.$$

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

The Square-Root Method(cont'd)

We successively find

$$y_1 = \frac{b_1}{t_{11}}$$

$$y_{1} = \frac{1}{t_{11}}$$

$$y_{i} = \frac{b_{i} - \sum_{k=1}^{i-1} t_{ki} y_{k}}{t_{ii}} \qquad (1 < i \le n)$$

$$x_n = \frac{y_n}{t_{nn}}$$

$$x_n = \frac{y_n}{t_{nn}}$$

$$x_i = \frac{y_i - \sum_{k=i+1}^n t_{ik} x_k}{t_{ii}} (1 < i \le n)$$

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

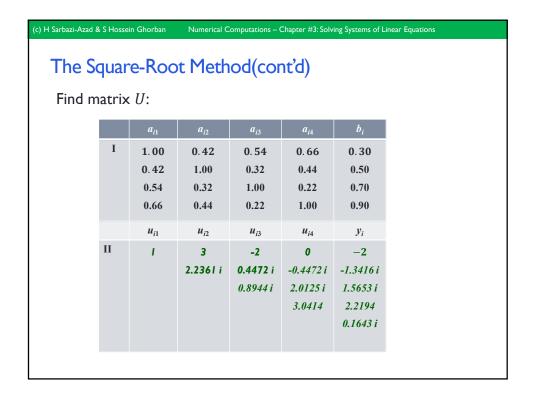
The Square-Root Method(cont'd)

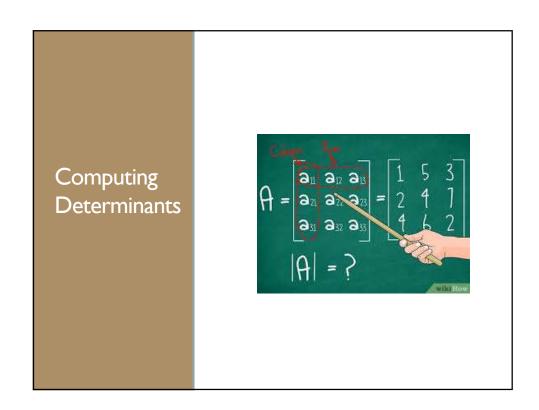
Example.

$$\begin{cases} 1x_1 + 3x_2 - 2x_3 + 0x_4 - 2x_5 = 0.5\\ 3x_1 + 4x_2 - 5x_3 + 1x_4 - 3x_5 = 5.4\\ -2x_1 - 5x_2 + 3x_3 - 2x_4 + 2x_5 = 0.5\\ 0x_1 + 1x_2 - 2x_3 + 5x_4 + 3x_5 = 7.5\\ -2x_1 - 3x_2 + 2x_3 + 3x_4 + 4x_5 = 3.3 \end{cases}$$

Solution. Write down the coefficients of the system a_{ij} , for i,j=1, 2, 3, 4, 5 in the first section of the table

	a_{i1}	a_{i2}	<i>a</i> _{i3}	<i>a</i> _{i4}	a_{i5}	a_{i6}	S_i
Ι	1	3	-2	0	-2	0.5	0.5
	3	4	-5	1	-3	5.4	5.4
	-2	-5	3	-2	2	0.5	1
	0	1	-2	5	3	7.5	14.5
	-2	-3	2	3	4	3.3	7.3





Numerical Computations - Chapter #3: Solving Systems of Linear Equations

Computing Determinants

If we can use of the Gaussian method, successfully, then the determinant of matrix A is equal to the product of the leading elements of the corresponding Gaussian scheme.

$$\Delta = \det A = a_{11}a_{22}^{(1)} \dots a_{nn}^{(n-1)}$$

Note that by using the method we obtain A=LU where L is a square lower unit triangular matrix, and U a rectangular matrix.

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Numerical Computations – Chapter #3: Solving Systems of Linear Equations

Computing Determinants (cont'd)

$$\Delta = \begin{vmatrix} 1.1161 & 0.1254 & 0.1397 & 0.1490 \\ 0.1582 & 1.1675 & 0.1768 & 0.1871 \\ 0.1968 & 0.2071 & 1.2168 & 0.2271 \\ 0.2368 & 0.2471 & 0.2568 & 1.2671 \end{vmatrix}$$

The given determinant is equal to the determinant of the system solved in a example by the Gaussian method with the principal element chosen. Forming the product of the leading elements, we get the required value of the determinant





If A is a symmetric matrix, then it is advisable to use the square-root method for evaluating the determinant of this matrix.

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

Thus

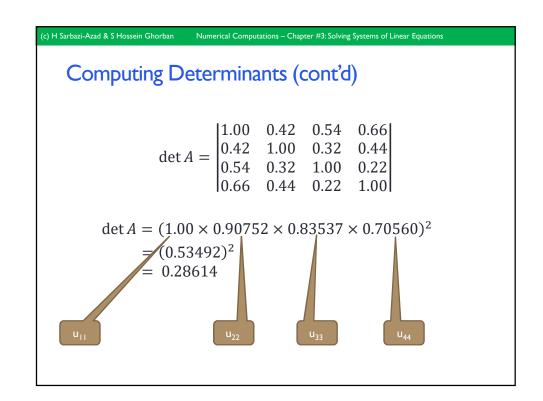
$$A = U^T U$$

where

$$T = \begin{pmatrix} u_{11} & u_{12} \dots & u_{1n} \\ 0 & u_{22} \dots & u_{2n} \\ \dots & \dots & \dots \\ 0 & 0 & \dots & u_{nn} \end{pmatrix}.$$

Then

$$\det A = \det U^T \det U = (\det U)^2 = (u_{11} \ u_{22} \dots \ u_{nn})^2$$



Computing the Elements of an Inverse Matrix by the Gaussian Method



Wilhelm Jordan (1842-1899)

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Numerical Computations - Chapter #3: Solving Systems of Linear Equations

Computing the Elements of an Inverse Matrix

Definition. A square matrix A is called non-singular if its determinant $\det A$ is non-zero.

Theorem. Every nonsingular matrix has an inverse.

Sketch of proof. It is shown that $(adj \ A)A = A(adj \ A) = (\det A) \ I$.

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

Computing the Elements of an Inverse Matrix (cont'd)

Let A is non-singular and $AA^{-1} = I_n$ where

$$A^{-1} = \begin{bmatrix} x_{11} & x_{12} & \dots & x_{1n} \\ x_{21} & x_{22} & \dots & x_{2n} \\ \dots & \dots & \dots & \dots \\ x_{n1} & x_{n2} & \dots & x_{nn} \end{bmatrix}.$$

Consider

$$x_i = \begin{bmatrix} x_{1i} \\ x_{2i} \\ \vdots \\ x_{ni} \end{bmatrix} (1 \le i \le n) \qquad I = \begin{bmatrix} e_1 & \dots & e_n \end{bmatrix}.$$

Thus, $AA^{-1} = A [x_1 \quad \cdots \quad x_n] = [Ax_1 \quad \ldots \quad Ax_n] = [e_1 \quad \cdots \quad e_n].$

Consequently, we get n systems with n equations totally in n^2 unknowns x_{ij} .

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C	om	puting	the Ele	ments	of an I	nverse	Matrix	k (con	ťd)	
		a_{i1}	a_{i2}	a_{i3}	a_{i4}	<i>a</i> _{i5,I}	<i>a</i> _{i5,II}	$a_{i5,\mathrm{III}}$	<i>a</i> _{i5,IV}	
	I	a_{11}	a_{12}	a_{13}	a_{14}	1	0	0	0	
		a_{21}	a_{22}	a_{23}	a_{24}	0	1	0	0	
		a_{31}	a_{32}	a_{33}	a_{34}	0	0	1	0	
		a_{41}	a_{42}	a_{43}	a_{44}	0	0	0	1	
		1	b_{12}	b_{13}	b_{14}	b_{15}	0	0	0	
	II		$a_{22}^{(1)}$	$a_{23}^{(1)}$	$a_{24}^{(1)}$	$a_{25}^{(1)}$, I	1	0	0	
			$a_{32}^{(1)} \ a_{42}^{(1)}$	$a^{(1)}_{23} \ a^{(1)}_{33} \ a^{(1)}_{43}$	$a^{(1)}_{24} \ a^{(1)}_{34} \ a^{(1)}_{44}$	$a_{35}^{(1)}$, I	0	1	0	
			$a_{42}^{(1)}$	$a_{43}^{(1)}$	$a_{44}^{(1)}$	$a_{45}^{(1)}$, I	0	0	1	
			1	b ₂₃	b ₂₄	b ₂₅ , I	b ₂₅ , II	0	0	
	III			$a_{33}^{(2)}$	$a^{(2)}_{34} \ a^{(2)}_{44}$	$a^{(2)}_{35} \ a^{(2)}_{45}$	$a_{35}^{(2)}$, II	1	0	
				$a^{(2)}_{33} \ a^{(2)}_{43} \ 1$	$a_{44}^{(2)}$ b_{34}	$a_{45}^{(2)}$ b_{35} , I	$a_{45}^{(2)}$, II b_{35} , II	0	1 0	
	IV			1				(3)		
	IV				$a_{44}^{(3)}$	$a_{45}^{(3)} \ b_{45}$, I	$a_{45}^{(3)}$, II b_{45} , II	$a_{45}^{(3)}$, III b_{45} , III	1 b ₂₅ , IV	
	V								20	
	v					x ₄₁	x ₄₂	x ₄₃	x ₄₄	
						x_{31} x_{21}	$x_{32} \\ x_{22}$	$x_{33} = x_{23}$	$x_{34} \\ x_{24}$	
						x ₁₁	x ₁₂	x ₁₃	x ₁₄	

Numerical Computations - Chapter #3: Solving Systems of Linear Equation

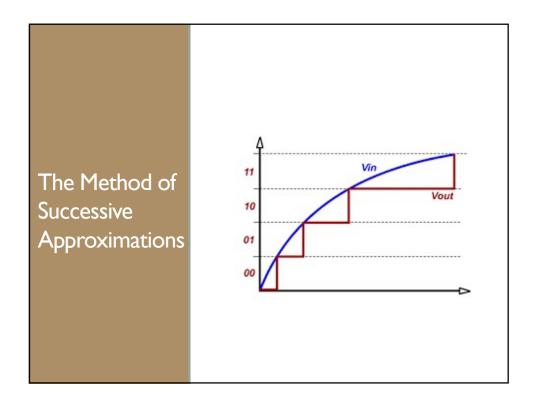
Computing the Elements of an Inverse Matrix (cont'd)

Find the inverse matrix for the matrix

$$A = \begin{bmatrix} 1.8 & -3.8 & 0.7 & -3.7 \\ 0.7 & 2.1 & -2.6 & -2.8 \\ 7.3 & 8.1 & 1.7 & -4.9 \\ 1.9 & -4.3 & -4.9 & -4.7 \end{bmatrix}$$

The computations are given in the table. The last column of the table consists of the sums of elements for each row.

Parting the Elements of an Inverse Matrix (cont'd) a_{II}		iting th	ne Eler	ments	of an	Inverse	e Mati	rix (co	ont'd)
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$		0						•	,
1.8		a_{i1}	a_{i2}	a_{i3}	a_{i4}	a _{i5,I}	$a_{i5,\mathrm{II}}$	$a_{i5,\mathrm{III}}$	$a_{i5,\mathrm{IV}}$
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	I	1.8	-3.8	0.7	-3.7	1	0	0	0
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		0.7	2.1	-2.6	-2.8	0	1	0	0
1		7.3	8.1	1.7	-4.9	0	0	1	0
III 3.57778 -2.87222 -1.36111 -0.38885 1 0 0 23.51110 -1.13890 10.10559 -4.05551 0 1 0 -0.28889 -5.63889 -0.79444 -1.05554 0 0 1 1 -0.80279 -0.38043 -0.10868 0.27950 0 0 III 17.73577 19.04992 -1.50032 -6.57135 1 0 -5.87081 -0.90434 -1.08694 0.08074 0 1 1.07411 -0.08459 -0.37108 0.05638 0 IV 5.40155 -1.58355 -2.09780 0.33100 1 V 1 -0.29316 -0.38837 0.06128 0.18513 0.23030 0.04607 -0.00944 -0.19885		1.9	-4.3	-4.9	-4.7	0	0	0	1
111		1	-2.11111	0.38889	-2.05556	0.55556	0	0	0
10	II		3.57778	-2.87222	-1.36111	-0.38885	1	0	0
III				-1.13890			0	1	0
III -0.80279 -0.38043 -0.10868 0.27950 0 0 17.73577 19.04992 -1.50032 -6.57135 1 0 -5.87081 -0.90434 -1.08694 0.08074 0 1 1 1.07411 -0.08459 -0.37108 0.05638 0 IV 5.40155 -1.58355 -2.09780 0.33100 1 V 1 -0.29316 -0.38837 0.06128 0.18513 0.23030 0.04607 -0.00944 -0.19885			-0.28889					0	
17.75377 15.04392 -1.30032 -6.37133 1 0 -5.87081 -0.90434 -1.08694 0.08074 0 1 1 1.07411 -0.08459 -0.37108 0.05638 0 IV 5.40155 -1.58355 -2.09780 0.33100 1 V 1 -0.29316 -0.38837 0.06128 0.18513 0.23030 0.04607 -0.00944 -0.19885			1	-0.80279	-0.38043	-0.10868	0.27950	0	0
1 1.07411 -0.08459 -0.37108 0.05638 0 IV 5.40155 -1.58355 -2.09780 0.33100 1 V 1 -0.29316 -0.38837 0.06128 0.18513 0.23030 0.04607 -0.00944 -0.19885	III			17.73577	19.04992	-1.50032	-6.57135	1	0
IV 5.40155 -1.58355 -2.09780 0.33100 1 V 1 -0.29316 -0.38837 0.06128 0.18513 0.23030 0.04607 -0.00944 -0.19885				-5.87081	-0.90434	-1.08694	0.08074	0	1
V 1 -0.29316 -0.38837 0.06128 0.18513 0.23030 0.04607 -0.00944 -0.19885				1	1.07411	-0.08459	-0.37108	0.05638	0
0.23030 0.04607 -0.00944 -0.19885	IV				5.40155	-1.58355	-2.09780	0.33100	1
	V				1	-0.29316	-0.38837	0.06128	0.18513
-0.03533						0.23030	0.04607	-0.00944	-0.19885
						-0.03533	0.16873	0.01573	-0.08920



Numerical Computations - Chapter #3: Solving Systems of Linear Equations

Method of successive approximations

An Iterative technique to solve the $n \times n$ linear system Ax = b starts with an initial approximation $x^{(0)}$ to the solution x and generates a sequence of vectors $\{x^{(k)}\}_{k=0}^{\infty}$.

The process stops if the approximations "stabilize", i.e., if the difference between successive approximations becomes negligible.

One of the possible stopped criterion is to iterate until

$$\frac{\left\| \pmb{x}^{(p+1)} - \pmb{x}^{(p)} \right\|}{\left\| \pmb{x}^{(p+1)} \right\|}$$

Is smaller than some prescribed tolerance.

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

Method of successive approximations

The Jacobia and the Gauss-Seidel iterative methods are classical methods.

Since the time required for sufficient accuracy exceeds that required time for direct techniques such as Gaussian elimination, iterative techniques are seldom used for solving linear system of small dimension.

For system equation with large number of equations and a high percentage of 0 entries, the iterative techniques are efficient in terms of both computer storage and computation.

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Numerical Computations – Chapter #3: Solving Systems of Linear Equations

Jacobi's Method

Given a linear system:

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{14}x_4 = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{24}x_4 = b_2 \\ \vdots + \vdots + \dots + \vdots = \vdots \\ a_{n1}x_1 + a_{n2}x_2 + a_{n3}x_3 + a_{n4}x_4 = b_n \end{cases}$$

Assuming that the diagonal coefficients are zero, $a_{ii} \neq 0$ for all $1 \leq i \leq n$.

$$\begin{array}{l} x_1 = \beta_1 + b_{12}x_2 + b_{13}x_3 + \cdots + \alpha_{1n}x_n \\ x_2 = \beta_2 + \ b_{21}x_1 + b_{23}x_3 + \cdots + b_{24}x_4 \\ & \vdots \\ x_n = \beta_n + \ b_{n1}x_1 + b_{n2}x_2 + \cdots + b_{n(n-1)}x_{n-1} \end{array}$$

where $\, eta_i = rac{b_i}{a_{ii}} \, {
m and} \, \, b_{{
m ij}} = -rac{a_{ij}}{a_{ii}} \, \, {
m for \, all} \, \, i
eq j \, .$

Numerical Computations - Chapter #3: Solving Systems of Linear Equation

lacobi's Method (cont'd)

So we can write the given system equation as

$$x = \beta + Bx$$

where

$$\mathbf{B} = \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ b_{21} & b_{22} & \cdots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \cdots & b_{nn} \end{bmatrix} \text{ and } \boldsymbol{\beta} = \begin{bmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_n \end{bmatrix}.$$

We take $x^{(0)} = \beta$, then we consecutively construct the column matrices

$$\mathbf{x}^{(1)} = \beta + B\mathbf{x}^{(0)}.$$

as a first approximation, then

$$\mathbf{x}^{(2)} = \mathbf{\beta} + B\mathbf{x}^{(1)}$$

And second approximation and generally speaking any k+1 the approximation is computed from $x^{(k+1)} = \beta + Bx^k$.

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

lacobi's Method (cont'd)

Consequently, the Jacobi method is written in the form

$$\pmb{x}^{(k+1)} = \beta + B \pmb{x}^k.$$

For solving the equation system Ax = b.

The coefficient matrix A can be split into its diagonal and off-diagonal parts (as summation of one strictly lower-triangular part of A, and the strictly upper-triangular part of A. That means:

$$\underbrace{\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \cdots & \cdots & \cdots & \cdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}}_{\hat{A}} = \underbrace{\begin{bmatrix} a_{11} & 0 & \cdots & 0 \\ 0 & a_{22} & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & 0 \end{bmatrix}}_{\hat{D}} - \underbrace{\begin{bmatrix} 0 & 0 & \cdots & 0 \\ -a_{21} & 0 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ -a_{n1} & -a_{n2} & \cdots & 0 \end{bmatrix}}_{\hat{L}} - \underbrace{\begin{bmatrix} 0 & -a_{12} & \cdots & -a_{1n} \\ 0 & 0 & \cdots & -a_{2n} \\ 0 & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & 0 \end{bmatrix}}_{\hat{U}} - \underbrace{\begin{bmatrix} 0 & -a_{12} & \cdots & -a_{1n} \\ 0 & 0 & \cdots & -a_{2n} \\ 0 & 0 & \cdots & 0 \end{bmatrix}}_{\hat{U}} - \underbrace{\begin{bmatrix} 0 & -a_{12} & \cdots & -a_{1n} \\ 0 & 0 & \cdots & -a_{2n} \\ 0 & 0 & \cdots & 0 \end{bmatrix}}_{\hat{U}} - \underbrace{\begin{bmatrix} 0 & -a_{12} & \cdots & -a_{1n} \\ 0 & 0 & \cdots & -a_{2n} \\ 0 & 0 & \cdots & 0 \end{bmatrix}}_{\hat{U}} - \underbrace{\begin{bmatrix} 0 & -a_{12} & \cdots & -a_{1n} \\ 0 & 0 & \cdots & -a_{2n} \\ 0 & 0 & \cdots & 0 \end{bmatrix}}_{\hat{U}} - \underbrace{\begin{bmatrix} 0 & -a_{12} & \cdots & -a_{1n} \\ 0 & 0 & \cdots & -a_{2n} \\ 0 & 0 & \cdots & 0 \end{bmatrix}}_{\hat{U}} - \underbrace{\begin{bmatrix} 0 & -a_{12} & \cdots & -a_{1n} \\ 0 & 0 & \cdots & -a_{2n} \\ 0 & 0 & \cdots & 0 \end{bmatrix}}_{\hat{U}}$$

The equation Ax = b, or (D - L - U)x = b, is then transformed into Dx = b + (L + U)x. If D^{-1} exists, i.e., $a_{ii} \neq 0$ for each i, then $x = \underbrace{D^{-1}b}_{\beta} + \underbrace{D^{-1}(L + U)}_{B}x$

$$x = \underbrace{D^{-1}b}_{\beta} + \underbrace{D^{-1}(L+U)}_{B} x$$

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

Jacobi's Method (cont'd)

We have

$$\lim_{k \to \infty} x^{(k+1)} = \beta + B \lim_{k \to \infty} x^{(k)}$$

If the sequence of approximations $x^0, ..., x^k, ...$ has a limit, then

$$\lim_{k\to\infty} x^{(k)} = x \text{ and } x = \beta + Bx.$$

In the other words, the solution of the system is a fixed point for the function $G: \mathbb{R}^n \to \mathbb{R}^n$ such that $G(x) = \beta + Bx$

Thus, the formulation of the approximations is as follows

$$x_i^{(0)} = \beta_i$$

$$x_i^{(k+1)} = \beta_i + \sum_{i=1}^n b_{ij} x_j^{(k)}$$

for $1 \le i \le n$. When we have n equations in n unknowns, then $\alpha_{ii} = 0$.

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

Jacobi's Method (cont'd)

Example. Solve the system

$$\begin{cases} 20.9x_1 + 1.2x_2 + 2.1x_3 + 0.9x_4 = 21.70 \\ 1.2x_1 + 21.2x_2 + 1.5x_3 + 2.5x_4 = 27.46 \\ 2.1x_1 + 1.5x_2 + 19.8x_3 + 1.3x_4 = 28.76 \\ 0.9x_1 + 2.5x_2 + 1.3x_3 + 32.1x_4 = 49.72 \end{cases}$$

Solution. Reduce the above system to the following form:

$$x_{1} = \frac{1}{a_{11}}(b_{1} - a_{12}x_{2} - \dots - a_{1n}x_{n})$$

$$x_{2} = \frac{1}{a_{22}}(b_{2} - a_{21}x_{1} - a_{23}x_{3} - \dots - a_{2n}x_{n})$$

$$\dots$$

$$x_{n} = \frac{1}{a_{nn}}(b_{n} - a_{n1}x_{1} - a_{n2}x_{2} - \dots - a_{nn-1}x_{n-1})$$

$$x_{1} = \frac{1}{a_{11}}(b_{1} - a_{12}x_{2} - \dots - a_{1n}x_{n})$$

$$x_{2} = \frac{1}{a_{22}}(b_{2} - a_{21}x_{1} - a_{23}x_{3} - \dots - a_{2n}x_{n})$$

$$\dots$$

$$x_{n} = \frac{1}{a_{nn}}(b_{n} - a_{n1}x_{1} - a_{n2}x_{2} - \dots - a_{nn-1}x_{n-1})$$

$$x_{1} = \frac{1}{20.9}(21.70 - 1.2x_{2} - 2.1x_{3} - 0.9x_{4}),$$

$$x_{2} = \frac{1}{21.2}(27.46 - 1.2x_{1} - 1.5x_{3} - 2.5x_{4}),$$

$$x_{3} = \frac{1}{19.8}(28.76 - 2.1x_{1} - 1.5x_{2} - 1.3x_{4}),$$

$$x_{4} = \frac{1}{32.1}(49.72 - 0.9x_{1} - 2.5x_{2} - 1.3x_{3}).$$

Jacobi's Method (cont'd)

Ist iteration:

$$x_1^{(1)} = \frac{1}{20.9}(21.70 - 1.560 - 3.045 - 1.395) = 0.75,$$

$$x_2^{(1)} = \frac{1}{21.2}(27.46 - 1.248 - 2.175 - 3.875) = 0.95,$$

$$x_3^{(1)} = \frac{1}{19.8}(28.76 - 2.184 - 1.950 - 2.015) = 1.14,$$

$$x_4^{(1)} = \frac{1}{32.1}(49.72 - 0.936 - 3.250 - 1.885) = 1.36$$

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

Jacobi's Method (cont'd)

2nd iteration:

$$x_1^{(2)} = \frac{16.942}{20.9} = 0.8106,$$

$$x_1^{(2)} = \frac{16.942}{20.9} = 0.8106,$$
 $x_2^{(2)} = \frac{21.450}{21.2} = 1.0118,$

$$x_3^{(2)} = \frac{23.992}{19.8} = 1.2117$$

$$x_3^{(2)} = \frac{23.992}{19.8} = 1.2117, \qquad x_4^{(2)} = \frac{45.188}{32.1} = 1.4077,$$

3rd iteration:

$$x_1^{(3)} = \frac{16.67434}{20.9} = 0.7978, \qquad x_2^{(3)} = \frac{21.15048}{21.2} = 0.9977,$$

$$x_3^{(3)} = \frac{23.71003}{19.8} = 1.1975, \qquad x_4^{(3)} = \frac{44.88575}{32.1} = 1.3983,$$

Jacobi's Method (cont'd)

4th iteration:

$$x_1^{(4)} = \frac{16.7295}{20.9} = 0.8004, \qquad x_2^{(4)} = \frac{21.2106}{21.2} = 1.0005,$$

$$x_3^{(4)} = \frac{23.7703}{19.8} = 1.2005, \qquad x_4^{(4)} = \frac{44.9510}{32.1} = 1.4003$$

Calculate the moduli of the differences of the values of $x_i^{(k)}$ for k = 3 and k = 4:

$$\left|x_1^{(3)} - x_1^{(4)}\right| = 0.0026, \qquad \left|x_2^{(3)} - x_2^{(4)}\right| = 0.0028,$$

$$\left|x_2^{(3)} - x_2^{(4)}\right| = 0.0028$$

$$\left|x_3^{(3)} - x_3^{(4)}\right| = 0.0030,$$
 $\left|x_4^{(3)} - x_4^{(4)}\right| = 0.0020,$

$$\left| x_4^{(3)} - x_4^{(4)} \right| = 0.0020$$

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

lacobi's Method (cont'd)

Since all of them exceed the pre-assigned number $\varepsilon = 10^{-3}$, the process of iteration is continued. We get for k = 5:

$$x_1^{(5)} = \frac{16.71808}{20.9} = 0.7999, \quad x_2^{(5)} = \frac{21.19802}{21.2} = 0.9999,$$

$$x_3^{(5)} = \frac{23.75802}{19.8} = 1.1999, \qquad x_4^{(5)} = \frac{44.93774}{32.1} = 1.3999,$$

Find the moduli of the differences of the values of $x_i^{(k)}$ for k=4and $k = 5 ||x^{(5)}|| = 2.24479$:

$$\left|x_1^{(4)} - x_1^{(5)}\right| = 0.0005,$$
 $\left|x_2^{(4)} - x_2^{(5)}\right| = 0.0006,$

$$\left|x_3^{(4)} - x_3^{(5)}\right| = 0.0006, \qquad \left|x_4^{(4)} - x_4^{(5)}\right| = 0.0004.$$

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

Jacobi's Method (cont'd)

They are less than the given number ϵ therefore we take the following as the solution:

$$x_1 \approx 0.7999$$
, $x_2 \approx 0.9999$, $x_3 \approx 1.1999$, $x_4 \approx 1.3999$

In accordance with previous estimate the errors of these values should not exceed $1/3 \times 0.0006 = 0.0002$.

For comparison we give the exact values of the unknowns:

$$x_1 = 0.8$$

$$x_2 = 1.0$$

$$x_3 = 1.2$$

$$x_1 = 0.8$$
, $x_2 = 1.0$, $x_3 = 1.2$, $x_4 = 1.4$

The Seidel Method



Philipp Ludwig von Seidel (1821-1896)

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

The Gauss-Seidel Method

The **Gauss-Seidel method** is a certain modification of the method of simple iteration. The principal idea behind it is that in computing the (k+1)th approximation of the unknown x_i for i>1, the earlier computed (k+1)th approximations of the unknowns $x_1, x_2, \ldots, x_{i-1}$ are taken into account.

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The Seidel Method and the Jacobi Method

Jacobi method

In its (p+1)-th iteration, the value of $x_1^{(p)}, \dots, x_n^{(p)}$ are substituted into every equation of the re-arranged system simultaneously to obtain $x^{(p+1)}$

Seidel method

This method differs from the Jacobi Method in which immediately after a new $x_i^{(p+1)}$ value is obtained from the ith equation, it is used I n place of the old value in successive substitutions.

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The Seidel Method (cont'd)

Example. Solve the following system of equations by the Seidel method:

$$\begin{cases} 10x_1 + x_2 + x_3 = 12 \\ 2x_1 + 10x_2 + x_3 = 13 \\ 2x_1 + 2x_2 + 10x_3 = 14 \end{cases}$$

Solution. Reduce the system, to a form convenient for iteration:

$$\begin{cases} x_1 = 1.2 - 0.1x_2 - 0.1 x_3 \\ x_2 = 1.3 - 0.2x_1 - 0.1x_3 \\ x_3 = 1.4 - 0.2 x_1 - 0.2 x_2 \end{cases}$$

Let
$$x^{(k)} = \begin{bmatrix} x_1^{(k)} \\ x_3^{(k)} \\ x_3^{(k)} \end{bmatrix}$$
. For the zeroth approximations of the roots take $x^{(0)} = \begin{bmatrix} 1.2 \\ 0 \\ 0 \end{bmatrix}$.

Then
$$x^{(1)} = \begin{bmatrix} x_1^{(1)} \\ x_3^{(1)} \\ x_3^{(1)} \end{bmatrix} = \begin{bmatrix} 1.2 - 0.1 \times 0 - 0.1 \times 0 \\ 1.3 - 0.2 \times 1.2 - 0.1 \times 0 \\ 1.4 - 0.2 \times 1.2 - 0.2 \times 1.6 \end{bmatrix} = \begin{bmatrix} 1.2 \\ 1.6 \\ 0.948 \end{bmatrix}.$$

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The Seidel Method (cont'd)

We obtain
$$x^{(1)} = \begin{bmatrix} x_1^{(1)} \\ x_3^{(1)} \\ x_3^{(1)} \end{bmatrix} = \begin{bmatrix} 1.2 \\ 1.6 \\ 0.948 \end{bmatrix}$$
. Now,

$$x^{(2)} = \begin{bmatrix} x_1^{(2)} \\ x_3^{(2)} \\ x_2^{(2)} \end{bmatrix} = \begin{bmatrix} 1.2 - 0.1 \times 1.6 - 0.1 \times 0.948 \\ 1.3 - 0.2 \times 0.9992 - 0.1 \times 0.948 \\ 1.4 - 0.2 \times 0.9992 - 0.2 \times 1.00536 \end{bmatrix} = \begin{bmatrix} 0.9992 \\ 1.00536 \\ 0.999098 \end{bmatrix}$$

By continuing this process, we obtain:

k	$x_1^{(k)}$	$x_{2}^{(k)}$	$x_3^{(k)}$
0	1.2000	0.0000	0.0000
1	1.200	1.0600	0.9480
2	0.9992	1.0054	0.991
3	0.9996	1.0001	1.0001
4	1.0000	1.0001	1.0001
5	1.0000	1.0000	1.0001

The exact values of the roots are

$$x_1 = 1, x_2 = 1, x_3 = 1$$

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The Seidel Method (cont'd)

Simple Iteration vs. Seidel:

- Seidel method usually improves the rate of convergence (not always!)
- As a programmer, Seidel method consumes less memory:
 - Seidel: The old value of a variable can be overwritten as soon as a new value is obtained.
 - Simple: All values from the last iteration must be kept.
 - Result: In Simple Iteration, twice as much storage is needed.
- However, Simple Iteration is perfectly suited to parallel programming, whereas the Gauss-Seidel method is not.

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The Seidel Method (cont'd)

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General Iteration Methods

To study the convergence of general iteration techniques, we need to analyze the formula

$$\mathbf{x}^{(p+1)} = \mathbf{c} + T\mathbf{x}^{(p)}$$

for p = 1,2,... where $x^{(0)}$ is arbitrary.

Definition. The spectral radius $\rho(A)$ of a matrix A is defined by $\rho(A) = \max |\lambda|$

where λ is an eigenvalue of A.

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General Iteration Methods (cont'd)

Theorem. For any $x^{(0)} \in \mathbb{R}^n$, the sequence $\{x^{(p)}\}_{p=0}^\infty$ defined by

$$\mathbf{x}^{(p)} = \mathbf{c} + T\mathbf{x}^{(p-1)}$$

for each $p \ge 1$, converges to the unique solution of

$$x = c + Tx$$

if and only if $\rho(T) < 1$.

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General Iteration Methods (Just for fun!)

We present a proof for (\Leftarrow) . For this proof, we need some Lemmas.

Lemma. For an $n \times n$ matrix A, the following statements are equivalent:

(i)
$$\rho(A) < 1$$
.

(ii)
$$\lim_{n\to\infty} A^n x = 0$$
, for every x .

Lemma. If
$$\rho(A) < 1$$
, then $(I - T)^{-1}$ exists, and $(I - T)^{-1} = I + T + T^2 + \dots = \sum_{j=0}^{\infty} T^j$.

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General Iteration Methods (Just for fun!)

Lemma. If $\rho(A) < 1$, then $(I - T)^{-1}$ exists, and

$$(I-T)^{-1} = I + T + T^2 + \dots = \sum_{j=0}^{\infty} T^j$$
.

Proof. If λ is an eigenvalue for T, then $1 - \lambda$ is an eigenvalue for $I - \lambda$ T. $(Tx = \lambda x \rightarrow (I - T)x = (1 - \lambda)x.)$

Let

$$S_m = I + T + T^2 + \dots + T^m.$$

Then

$$(I-T)S_{m} = (I+T+T^{2}+\cdots+T^{m})-(T+T^{2}+\cdots+T^{m+1}) = I-T^{m+1}.$$

Since
$$\rho(A) < 1$$
, we have $\lim_{n \to \infty} T^m = 0$. Thus
$$\lim_{m \to \infty} (I - T) S_m = \lim_{m \to \infty} I - T^{m+1} \Longrightarrow (I - T) \sum_{j=0}^{\infty} T^j = I.$$

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General Iteration Methods (Just for fun!) (cont'd)

Proof.(\Leftarrow) Assume $\rho(T)$ < 1. Then,

$$x^{(p)} = c + Tx^{(p-1)}$$

$$= c + T(c + Tx^{(p-2)})$$

$$= (T+I)c + T^2x^{(p-2)})$$

$$\vdots$$

$$= (T^{p-1} + \dots + T+I)c + T^px^{(0)}.$$

Since
$$ho(T) < 1$$
, we have $\lim_{p \to \infty} T^p x^{(0)} = 0$. Thus
$$\lim_{p \to \infty} x^{(p)} = \left(\sum_{j=0}^{\infty} T^j\right) c + \lim_{p \to \infty} T^p x^{(0)}$$

$$\lim_{p \to \infty} x^{(p)} = (I - T)^{-1} c$$

Hence, the sequence $\{x^{(p)}\}_{p=0}^{\infty}$ converge to the vector $x = (I - T)^{-1}c$.

Numerical Computations – Chapter #3: Solving Systems of Linear Equations

General Iteration Methods (cont'd)

Definition. The $n \times n$ matrix A is said to be diagonally dominant when

$$|a_{ii}| \ge \sum_{i \ne i} a_{ij}$$

hold for each i = 1, 2, ..., n. A diagonally dominant matrix is said to be strictly diagonally dominant when the above inequality is strict for each n, i.e.,

$$|a_{ii}| > \sum_{j \neq i} a_{ij}$$

hold for each i = 1, 2, ..., n.

Numerical Computations - Chapter #3: Solving Systems of Linear Equations

General Iteration Methods (cont'd)

Theorem. If A is strictly diagonally dominant, i.e.,

$$|a_{ii}| > \sum_{j \neq i} a_{ij}$$

hold for each $i=1,2,\ldots,n$, then for <u>any</u> choice of $x^{(0)}$, both the Jaconi and Gauss-Seidel methods give sequences $\{x^{(k)}\}_{k=0}^{\infty}$ that converge to the unique solution of Ax=b.

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General Iteration Methods (cont'd)

Definition. Let A be a $n \times n$ matrix. The norm of it is defined by vector norm as follows:

$$||A|| = \max_{||x||=1} ||Ax||.$$

Note that for any $\mathbf{z} \neq \mathbf{0}$, the vector $\mathbf{x} = \frac{\mathbf{z}}{\|\mathbf{z}\|}$ is a unit vector. Hence

$$\max_{\|\boldsymbol{x}\|=1}\|A\boldsymbol{x}\|=\max_{\boldsymbol{z}\neq\boldsymbol{0}}\left\|A\left(\frac{\boldsymbol{z}}{\|\boldsymbol{z}\|}\right)\right\|=\max_{\boldsymbol{z}\neq\boldsymbol{0}}\frac{\|A\boldsymbol{z}\|}{\|\boldsymbol{z}\|}.$$

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General Iteration Methods (cont'd)

Theorem. If ||T|| < 1, then the sequences $\{x^{(k)}\}_{k=0}^{\infty}$ that converges the for <u>any</u> choice of $x^{(0)}$, to a vector $x \in \mathbb{R}^n$, and the following error bounds hold:

$$||x - x^{(k)}|| \le \frac{||T||^k}{1 - ||T||} ||x^{(1)} - x^{(0)}||.$$

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General Iteration Methods (cont'd)

Remark 1. No general results exist to tell which of the two techniques, Jacobi or Gauss-Seidel, will be most successful for an arbitrary linear system.

Remark 2. Under the conditions

 $(a_{ij} \le 0, \text{ for each } i = j \text{ and } a_{ii} > 0, \text{ for each } i = 1, 2, ..., n)$:

- one method gives convergence, then both give convergence, and the Gauss-Seidel method converges faster than the Jacobi method.
- owhen one method diverges then both diverge, and the divergence is more pronounced for the Gauss-Seidel method.

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General Iteration Methods (cont'd)

Remark. For a $n \times n$ matrix $A, \rho(A) \leq ||A||$.

Remark. The rate of convergence of an iterative technique depends on the spectral radius of the matrix associated with the method. It is shown that

$$||x^{(k)} - x|| \le \rho(A)^k ||x^{(0)} - x||.$$

Remark. One way to select a procedure to accelerate convergence is to choose a method whose associated matrix has minimal spectral radius.

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Successive Over-Relaxation (SOR)

Let

$$T_{\omega} = (D - \omega L)^{-1}[(1 - \omega)D + \omega U]$$

$$c_{\omega} = \omega (D - \omega L)^{-1} b$$

$$\boldsymbol{x}^{(k)} = c_{\omega} + T_{\omega} \boldsymbol{x}^{(k-1)}$$

For $0 < \omega < 1$.

This method accelerate the convergence for linear system that are convergent by the Gauss-Seidel thechnique.

