

3 Examples for Lectures 9 – 10

(Numerical methods for solving linear systems)

3.1 Direct methods

Example 3.1 Solve the system

$$\begin{cases} 2x_1 + 4x_3 + x_4 = 7 \\ 2x_2 + 4x_3 + x_4 = 7 \\ 2x_1 + 4x_2 + 3x_3 = 9 \\ x_1 + 2x_2 + 2x_4 = 5 \end{cases}$$

using the *Gauss method with partial pivoting*.

We start by writing the matrix A that contains the coefficients of the unknowns (x_1, x_2, x_3, x_4) . We also write \bar{A} which contains also the column vector b (the result of each equation), since the modifications should be performed on this column too.

$$A = \begin{pmatrix} 2 & 0 & 4 & 1 \\ 0 & 2 & 4 & 1 \\ 2 & 4 & 3 & 0 \\ 1 & 2 & 0 & 2 \end{pmatrix} \quad \text{and} \quad \bar{A} = \left(\begin{array}{cccc|c} \color{blue}{2} & 0 & 4 & 1 & 7 \\ 0 & 2 & 4 & 1 & 7 \\ 2 & 4 & 3 & 0 & 9 \\ 1 & 2 & 0 & 2 & 5 \end{array} \right).$$

On the first column of \bar{A} , the pivot (maximum element in absolute value) is $a_{11} = 2$, so we do not interchange any rows. $a_{21} = 0$ so we let it the same, and to obtain $a_{31} = 0$ and $a_{41} = 0$, we have to perform $L_3 - L_1$ and $L_4 - \frac{1}{2}L_1$. (Don't forget to change also the column of free term b !)

$$\bar{A} \sim \left(\begin{array}{cccc|c} 2 & 0 & 4 & 1 & 7 \\ 0 & 2 & 4 & 1 & 7 \\ 0 & 4 & -1 & -1 & 2 \\ 0 & 2 & -2 & \frac{3}{2} & \frac{3}{2} \end{array} \right)$$

On the second column (below the main diagonal - and including it), the maximum element in absolute value is $a_{32} = 4$, so we interchange L_2 and L_3 .

$$\bar{A} \sim \left(\begin{array}{cccc|c} 2 & 0 & 4 & 1 & 7 \\ 0 & \color{blue}{4} & -1 & -1 & 2 \\ 0 & 2 & 4 & 1 & 7 \\ 0 & 2 & -2 & \frac{3}{2} & \frac{3}{2} \end{array} \right)$$

The pivot is now $a_{22} = 4$. Now, to obtain 0 below the main diagonal on the second column, we need $a_{32} = 0$ and $a_{42} = 0$, so we perform $L_3 - \frac{1}{2}L_2$ and $L_4 - \frac{1}{2}L_2$, obtaining

$$\bar{A} \sim \left(\begin{array}{cccc|c} 2 & 0 & 4 & 1 & 7 \\ 0 & 4 & -1 & -1 & 2 \\ 0 & 0 & \color{blue}{\frac{9}{2}} & \frac{3}{2} & 6 \\ 0 & 0 & -\frac{3}{2} & 2 & \frac{1}{2} \end{array} \right)$$

On the third column, the maximum element in absolute value below the main diagonal (including it) is $a_{33} = \frac{9}{2}$, so we don't interchange anything. To obtain 0 below the main diagonal, we need $a_{43} = 0$, so we have to compute $L_4 + \frac{1}{3}L_3$, obtaining

$$\bar{A} \sim \left(\begin{array}{cccc|c} 2 & 0 & 4 & 1 & 7 \\ 0 & 4 & -1 & -1 & 2 \\ 0 & 0 & \frac{9}{2} & \frac{3}{2} & 6 \\ 0 & 0 & 0 & \frac{5}{2} & \frac{5}{2} \end{array} \right)$$

Now, using **backward substitution**, we obtain

$$\begin{aligned}\frac{5}{2}x_4 &= \frac{5}{2} \implies x_4 = 1 \\ \frac{9}{2}x_3 + \frac{3}{2} \cdot 1 &= 6 \implies x_3 = 1 \\ 4x_2 - 1 \cdot 1 - 1 \cdot 1 &= 2 \implies x_2 = 1 \\ 2x_1 + 0 \cdot 1 + 4 \cdot 1 + 1 \cdot 1 &= 7 \implies x_1 = 1\end{aligned}$$

Remark 3.2 The theory can be found in [Course 9, slides 17–20](#). Other examples using the Gauss elimination with partial pivoting are in [Course 9, slides 21–25](#).

Example 3.3 Solve the system

$$\begin{cases} x_1 + x_2 - 3x_3 = -9 \\ 4x_1 + x_2 + 2x_3 = 9 \\ 2x_1 + 4x_2 - x_3 = -5 \end{cases}$$

using the *Gauss-Jordan method*.

The idea here is to make zeros below and above the main diagonal, reducing the system to something of the form

$$\begin{pmatrix} \hat{a}_{11} & 0 & \dots & 0 \\ 0 & \hat{a}_{22} & \dots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & \hat{a}_{nn} \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} \hat{b}_1 \\ \hat{b}_2 \\ \vdots \\ \hat{b}_n \end{pmatrix} \text{ such that } x_i = \frac{\hat{b}_i}{\hat{a}_{ii}}, i = \overline{1, n}.$$

$$\begin{aligned}\overline{A} &= \left(\begin{array}{ccc|c} 1 & 1 & -3 & -9 \\ 4 & 1 & 2 & 9 \\ 2 & 4 & -1 & -5 \end{array} \right) \xrightarrow[L_3 \sim L_3 - 2L_1]{L_2 \sim L_2 - 4L_1} \left(\begin{array}{ccc|c} 1 & 1 & -3 & -9 \\ 0 & -3 & 14 & 45 \\ 0 & 2 & 5 & 13 \end{array} \right) \xrightarrow[L_3 \sim L_3 + \frac{2}{3}L_2]{L_3 \sim L_3 + \frac{2}{3}L_2} \left(\begin{array}{ccc|c} 1 & 1 & -3 & -9 \\ 0 & -3 & 14 & 45 \\ 0 & 0 & \frac{43}{3} & 43 \end{array} \right) \xrightarrow[L_3 \sim \frac{1}{43}L_3]{L_1 \sim L_1 + \frac{1}{3}L_2} \left(\begin{array}{ccc|c} 1 & 0 & \frac{5}{3} & 6 \\ 0 & -3 & 14 & 45 \\ 0 & 0 & \frac{1}{3} & 1 \end{array} \right) \xrightarrow[L_2 \sim L_2 - 3 \cdot \frac{1}{3}L_3]{L_1 \sim L_1 - 5L_3} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 1 \\ 0 & -3 & 0 & 3 \\ 0 & 0 & \frac{1}{3} & 1 \end{array} \right) \implies \begin{cases} x_1 = 1 \\ x_2 = -1 \\ x_3 = 3 \end{cases}.\end{aligned}$$

Remark 3.4 In this case we did not use the partial pivoting, we have only solved this system with the usual Gauss-Jordan method.

Remark 3.5 For the LU factorization (modified forms of Gauss elimination), see [Course 10, slides 1–3](#) for theory and [slides 4–6](#) for an example.

3.2 Iterative methods

Example 3.6 Determine the approximate solution for the system

$$\begin{cases} 5x_1 + x_2 - x_3 = 7 \\ x_1 + 5x_2 + x_3 = 7 \\ x_1 + x_2 + 5x_3 = 7 \end{cases}$$

with the initial approximation $x^{(0)} = (0, 0, 0)^T$ using

a) **Jacobi method** in 3 steps ;

We can see that the matrix

$$A = \begin{pmatrix} 5 & 1 & -1 \\ 1 & 5 & 1 \\ 1 & 1 & 5 \end{pmatrix}$$

is strictly diagonally dominant since

$$\begin{aligned}|a_{11}| &= |5| > |a_{12}| + |a_{13}| = |1| + |-1| = 2 \\|a_{22}| &= |5| > |a_{21}| + |a_{23}| = |1| + |1| = 2 \\|a_{33}| &= |5| > |a_{31}| + |a_{32}| = |1| + |1| = 2\end{aligned}$$

hence all the three methods will converge (no matter what the initial approximation $x^{(0)}$ is).

To apply the methods, we have to express the unknown x_k from the equation k with respect to the other unknowns. So, we have

$$\begin{cases} x_1 = \frac{7 - x_2 + x_3}{5} \\ x_2 = \frac{7 - x_1 - x_3}{5} \\ x_3 = \frac{7 - x_1 - x_2}{5} \end{cases} \quad (3.1)$$

Now, the Jacobi method consists in expressing $x^{(k)}$ (the unknown x at step k) using the previous approximations $x^{(k-1)}$. We have:

$$\begin{cases} x_1^{(1)} = \frac{7 - x_2^{(0)} + x_3^{(0)}}{5} = \frac{7 - 0 + 0}{5} = \frac{7}{5} = 1.4 \\ x_2^{(1)} = \frac{7 - x_1^{(0)} - x_3^{(0)}}{5} = \frac{7 - 0 - 0}{5} = \frac{7}{5} = 1.4 \\ x_3^{(1)} = \frac{7 - x_1^{(0)} - x_2^{(0)}}{5} = \frac{7 - 0 - 0}{5} = \frac{7}{5} = 1.4 \end{cases}$$

Next, on the second iteration we have:

$$\begin{cases} x_1^{(2)} = \frac{7 - x_2^{(1)} + x_3^{(1)}}{5} = \frac{7 - \frac{7}{5} + \frac{7}{5}}{5} = \frac{7}{5} = 1.4 \\ x_2^{(2)} = \frac{7 - x_1^{(1)} - x_3^{(1)}}{5} = \frac{7 - \frac{7}{5} - \frac{7}{5}}{5} = \frac{21}{25} = 0.84 \\ x_3^{(2)} = \frac{7 - x_1^{(1)} - x_2^{(1)}}{5} = \frac{7 - \frac{7}{5} - \frac{7}{5}}{5} = \frac{21}{25} = 0.84 \end{cases}$$

and the last one:

$$\begin{cases} x_1^{(3)} = \frac{7 - x_2^{(2)} + x_3^{(2)}}{5} = \frac{7 - \frac{21}{25} + \frac{21}{25}}{5} = \frac{7}{5} = 1.4 \\ x_2^{(3)} = \frac{7 - x_1^{(2)} - x_3^{(2)}}{5} = \frac{7 - \frac{7}{5} - \frac{21}{25}}{5} = \frac{119}{125} = 0.952 \\ x_3^{(3)} = \frac{7 - x_1^{(2)} - x_2^{(2)}}{5} = \frac{7 - \frac{7}{5} - \frac{21}{25}}{5} = \frac{119}{125} = 0.952 \end{cases}$$

b) **Gauss-Seidel method** in 2 steps ;

The difference between Jacobi and Gauss-Seidel is that in this case, we have to replace the unknowns with their most recent approximations. So, if we are at the step k , when we compute $x_3^{(k)}$, we won't use x_1 and x_2 from the previous step ($x_1^{(k-1)}$, $x_2^{(k-1)}$), but instead we will use their values from the current step, since we have already determined them. Using again (3.1), we obtain

$$\begin{cases} x_1^{(1)} = \frac{7 - x_2^{(0)} + x_3^{(0)}}{5} = \frac{7 - 0 + 0}{5} = \frac{7}{5} = 1.4 \\ x_2^{(1)} = \frac{7 - x_1^{(1)} - x_3^{(0)}}{5} = \frac{7 - \frac{7}{5} - 0}{5} = \frac{28}{25} = 1.12 \\ x_3^{(1)} = \frac{7 - x_1^{(1)} - x_2^{(1)}}{5} = \frac{7 - \frac{7}{5} - \frac{28}{25}}{5} = \frac{112}{125} = 0.896 \end{cases}$$

Next, we have:

$$\begin{cases} x_1^{(2)} = \frac{7 - x_2^{(1)} + x_3^{(1)}}{5} = \frac{7 - \frac{28}{25} + \frac{112}{125}}{5} = \frac{847}{625} = 1.3552 \\ x_2^{(2)} = \frac{7 - x_1^{(2)} - x_3^{(1)}}{5} = \frac{7 - \frac{847}{625} - \frac{112}{125}}{5} = \frac{2968}{3125} = 0.94976 \\ x_3^{(2)} = \frac{7 - x_1^{(2)} - x_2^{(2)}}{5} = \frac{7 - \frac{847}{625} - \frac{2968}{3125}}{5} = \frac{14672}{15625} = 0.939008 \end{cases}$$

c) **SOR method for $\omega = \frac{1}{2}$** in 2 steps .

It is similar to Gauss-Seidel method. First, we compute an intermediary point $\tilde{x}^{(k)}$ as in Gauss-Seidel and then $x^{(k)} = \omega \tilde{x}^{(k)} + (1 - \omega)x^{(k-1)}$. So, for (3.1), we have:

$$\begin{cases} \tilde{x}_1^{(1)} = \frac{7 - x_2^{(0)} + x_3^{(0)}}{5} = \frac{7 - 0 + 0}{5} = \frac{7}{5} = 1.4 \\ x_1^{(1)} = \omega \tilde{x}_1^{(1)} + (1 - \omega)x_1^{(0)} = \frac{1}{2} \cdot 1.4 + \frac{1}{2} \cdot 0 = 0.7 \\ \tilde{x}_2^{(1)} = \frac{7 - x_1^{(1)} - x_3^{(0)}}{5} = \frac{7 - 0.7 - 0}{5} = 1.26 \\ x_2^{(1)} = \omega \tilde{x}_2^{(1)} + (1 - \omega)x_2^{(0)} = \frac{1}{2} \cdot 1.26 + \frac{1}{2} \cdot 0 = 0.63 \\ \tilde{x}_3^{(1)} = \frac{7 - x_1^{(1)} - x_2^{(1)}}{5} = \frac{7 - 0.7 - 0.63}{5} = 1.134 \\ x_3^{(1)} = \omega \tilde{x}_3^{(1)} + (1 - \omega)x_3^{(0)} = \frac{1}{2} \cdot 1.134 + \frac{1}{2} \cdot 0 = 0.567 \end{cases}$$

And the second iteration is

$$\begin{cases} \tilde{x}_1^{(2)} = \frac{7 - x_2^{(1)} + x_3^{(1)}}{5} = \frac{7 - 0.63 + 0.567}{5} = 1.3874 \\ x_1^{(2)} = \omega \tilde{x}_1^{(2)} + (1 - \omega)x_1^{(1)} = \frac{1}{2} \cdot 1.3874 + \frac{1}{2} \cdot 0.7 = 1.0437 \\ \tilde{x}_2^{(2)} = \frac{7 - x_1^{(2)} - x_3^{(1)}}{5} = \frac{7 - 1.0437 - 0.567}{5} = 1.07786 \\ x_2^{(2)} = \omega \tilde{x}_2^{(2)} + (1 - \omega)x_2^{(1)} = \frac{1}{2} \cdot 1.07786 + \frac{1}{2} \cdot 0.63 = 0.85393 \\ \tilde{x}_3^{(2)} = \frac{7 - x_1^{(2)} - x_2^{(2)}}{5} = \frac{7 - 1.0437 - 0.85393}{5} = 1.020474 \\ x_3^{(2)} = \omega \tilde{x}_3^{(2)} + (1 - \omega)x_3^{(1)} = \frac{1}{2} \cdot 1.020474 + \frac{1}{2} \cdot 0.567 = 0.793737 \end{cases}$$

Remark 3.7 The exact solution is $(1.4; 0.9(3); 0.9(3))$.

Remark 3.8 See the theory and other examples for the three Iterative methods in [Course 10, slides 7–22](#).