

Numerical methods – lecture 5

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Iteration methods for solving system of linear equations

$$A\mathbf{x} = \mathbf{b} \quad \longrightarrow \quad \mathbf{x} = T\mathbf{x} + \mathbf{g}$$

Iteration process:

$$\mathbf{x}^{k+1} = T\mathbf{x}^k + \mathbf{g}, \quad k = 0, 1, \dots$$

Solution:

$$\hat{\mathbf{x}} = (E - T)^{-1}\mathbf{g}$$

Theorem

The sequence $\{\mathbf{x}^k\}_{k=0}^{\infty}$ determined by the iteration process $\mathbf{x} = T\mathbf{x} + \mathbf{g}$ converges for every initial iteration $\mathbf{x}^0 \in \mathbb{R}^n \iff \rho(T) < 1$. In this case

$$\lim_{k \rightarrow \infty} \mathbf{x}^k = \hat{\mathbf{x}}, \quad \hat{\mathbf{x}} = T\hat{\mathbf{x}} + \mathbf{g}$$

Jacobi iteration method

System of linear equations:

$$A\mathbf{x} = \mathbf{b}$$

i -th equation:

$$a_{i1}x_1 + \cdots + a_{ii}x_i + \cdots + a_{in}x_n = b_i$$

The component x_i is expressed

$$x_i = - \sum_{\substack{j=1 \\ j \neq i}}^n \frac{a_{ij}}{a_{ii}} x_j + \frac{b_i}{a_{ii}},$$

and it is used as the new $(k + 1)$ -th iteration

$$x_i^{k+1} = - \sum_{\substack{j=1 \\ j \neq i}}^n \frac{a_{ij}}{a_{ii}} x_j^k + \frac{b_i}{a_{ii}},$$

Matrix notation

$$\begin{pmatrix} x_1^{k+1} \\ x_2^{k+1} \\ \vdots \\ x_n^{k+1} \end{pmatrix} = \begin{pmatrix} 0 & -\frac{a_{12}}{a_{11}} & \dots & -\frac{a_{1n}}{a_{11}} \\ -\frac{a_{21}}{a_{22}} & 0 & & -\frac{a_{2n}}{a_{22}} \\ \vdots & & \ddots & \vdots \\ -\frac{a_{n1}}{a_{nn}} & -\frac{a_{n2}}{a_{nn}} & \dots & 0 \end{pmatrix} \begin{pmatrix} x_1^k \\ x_2^k \\ \vdots \\ x_n^k \end{pmatrix} + \begin{pmatrix} \frac{b_1}{a_{11}} \\ \frac{b_2}{a_{22}} \\ \vdots \\ \frac{b_n}{a_{nn}} \end{pmatrix}.$$

$$A\mathbf{x} = \mathbf{b}, \quad A = D + L + U,$$

$$A\mathbf{x} = (D + L + U)\mathbf{x} = \mathbf{b}$$

$$D = \begin{pmatrix} a_{11} & & 0 \\ & \ddots & \\ 0 & & a_{nn} \end{pmatrix},$$

$$L = \begin{pmatrix} 0 & & & 0 \\ a_{21} & \ddots & & \\ \vdots & \ddots & \ddots & \\ a_{n1} & \cdots & a_{n,n-1} & 0 \end{pmatrix},$$

$$U = \begin{pmatrix} 0 & a_{12} & \cdots & a_{1n} \\ & \ddots & \ddots & \vdots \\ & & \ddots & a_{n-1,n} \\ 0 & & & 0 \end{pmatrix}.$$

$$\mathbf{x} = -D^{-1}(L + U)\mathbf{x} + D^{-1}\mathbf{b}.$$

$$\mathbf{x}^{k+1} = -D^{-1}(L + U)\mathbf{x}^k + D^{-1}\mathbf{b}.$$

$$\mathbf{x}^{k+1} = T_J \mathbf{x}^k + D^{-1} \mathbf{b},$$

$$T_J = -D^{-1}(L + U), \quad t_{ij} = -\frac{a_{ij}}{a_{ii}} \text{ for } i \neq j, \quad t_{ii} = 0.$$

$$T_J = \begin{pmatrix} 0 & -\frac{a_{12}}{a_{11}} & \dots & -\frac{a_{1n}}{a_{11}} \\ -\frac{a_{21}}{a_{22}} & 0 & & -\frac{a_{2n}}{a_{22}} \\ \vdots & & \ddots & \vdots \\ -\frac{a_{n1}}{a_{nn}} & -\frac{a_{n2}}{a_{nn}} & \dots & 0 \end{pmatrix}, \quad D^{-1} \mathbf{b} = \begin{pmatrix} \frac{b_1}{a_{11}} \\ \frac{b_2}{a_{22}} \\ \vdots \\ \frac{b_n}{a_{nn}} \end{pmatrix}.$$

Gauss–Seidel iteration method

The component of the new iteration is used in the following step:

$$x_1^{k+1} = \frac{1}{a_{11}} (b_1 - a_{12}x_2^k - a_{13}x_3^k - a_{14}x_4^k - \dots,)$$

$$x_2^{k+1} = \frac{1}{a_{22}} (b_2 - a_{21}x_1^{k+1} - a_{23}x_3^k - a_{24}x_4^k - \dots,)$$

$$x_3^{k+1} = \frac{1}{a_{33}} (b_3 - a_{31}x_1^{k+1} - a_{32}x_2^{k+1} - a_{34}x_4^k - \dots,)$$

⋮

$$x_i^{k+1} = \frac{1}{a_{ii}} \left(b_i - \sum_{j=1}^{i-1} a_{ij}x_j^{k+1} - \sum_{j=i+1}^n a_{ij}x_j^k \right)$$

Matrix notation:

$$A\mathbf{x} = \mathbf{b}$$

$$(D + L + U)\mathbf{x} = \mathbf{b}$$

$$(D + L)\mathbf{x} = -U\mathbf{x} + \mathbf{b}$$

$$\mathbf{x} = -(D + L)^{-1}U\mathbf{x} + (D + L)^{-1}\mathbf{b}$$

$$T_G = -(D + L)^{-1}U, \quad \mathbf{x}^{k+1} = T_G\mathbf{x}^k + (D + L)^{-1}\mathbf{b}.$$

Theorem: If A is diagonally dominant matrix, i.e.

$$|a_{ii}| > \sum_{j \neq i} |a_{ij}| \quad \text{or} \quad |a_{ii}| > \sum_{j \neq i} |a_{ji}|$$

then Jacobi and Gauss–Seidel methods converge.

Relaxation (Successive over-relaxation (SOR)) method

x^k – k -th iteration

x_{GS}^{k+1} – the following iteration acquired by the Gauss–Seidel method

$\omega \in (0, 2)$ – relaxation parameter

$$x^{k+1} = (1 - \omega)x^k + \omega x_{GS}^{k+1}$$