COURSE 11

4.2. Iterative methods for solving linear systems

Because of round-off errors, direct methods become less efficient than iterative methods for large systems (>100~000~variables).

An iterative scheme for linear systems consists of converting the system

$$Ax = b \tag{1}$$

to the form

$$x = \tilde{b} - Bx.$$

After an initial guess for $x^{(0)}$, the sequence of approximations of the solution $x^{(0)}, x^{(1)}, ..., x^{(k)}, ...$ is generated by computing

$$x^{(k)} = \tilde{b} - Bx^{(k-1)}$$
, for $k = 1, 2, 3, ...$

4.2.1. Jacobi iterative method

Consider the $n \times n$ linear system,

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \dots \\ a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n, \end{cases}$$

where we assume that the diagonal terms $a_{11}, a_{22}, \ldots, a_{nn}$ are all nonzero.

We begin our iterative scheme by solving each equation for one of the variables:

$$\begin{cases} x_1 = u_{12}x_2 + \dots + u_{1n}x_n + c_1 \\ x_2 = u_{21}x_1 + \dots + u_{2n}x_n + c_2 \\ \dots \\ x_n = u_{n1}x_1 + \dots + u_{nn-1}x_{n-1} + c_n, \end{cases}$$

where
$$u_{ij} = -\frac{a_{ij}}{a_{ii}}, \ c_i = \frac{b_i}{a_{ii}}, \ i = 1,...,n.$$

Let $x^{(0)}=(x_1^{(0)},x_2^{(0)},...,x_n^{(0)})$ be an initial approximation of the solution. The k+1-th approximation is:

$$\begin{cases} x_1^{(k+1)} = u_{12}x_2^{(k)} + \dots + u_{1n}x_n^{(k)} + c_1 \\ x_2^{(k+1)} = u_{21}x_1^{(k)} + u_{23}x_3^{(k)} + \dots + u_{2n}x_n^{(k)} + c_2 \\ \dots \\ x_n^{(k+1)} = u_{n1}x_1^{(k)} + \dots + u_{nn-1}x_{n-1}^{(k)} + c_n, \end{cases}$$

for k = 0, 1, 2, ...

An algorithmic form:

$$x_i^{(k)} = \frac{b_i - \sum\limits_{j=1, j \neq i}^{n} a_{ij} x_j^{(k-1)}}{a_{ii}}, \ i = 1, 2, ..., n, \ \text{for } k \ge 1.$$

The iterative process is terminated when a convergence criterion is satisfied.

Stopping criterions: $\left|x^{(k)}-x^{(k-1)}\right|<\varepsilon$ or $\frac{\left|x^{(k)}-x^{(k-1)}\right|}{\left|x^{(k)}\right|}<\varepsilon$, with $\varepsilon>0$ - a prescribed tolerance.

Example 1 Solve the following system using the Jacobi iterative method. Use $\varepsilon = 10^{-3}$ and $x^{(0)} = (0\ 0\ 0\ 0)$ as the starting vector.

$$\begin{cases} 7x_1 - 2x_2 + x_3 & = 17 \\ x_1 - 9x_2 + 3x_3 - x_4 & = 13 \\ 2x_1 + 10x_3 + x_4 & = 15 \\ x_1 - x_2 + x_3 + 6x_4 & = 10. \end{cases}$$

These equations can be rearranged to give

$$x_1 = (17 + 2x_2 - x_3)/7$$

$$x_2 = (-13 + x_1 + 3x_3 - x_4)/9$$

$$x_3 = (15 - 2x_1 - x_4)/10$$

$$x_4 = (10 - x_1 + x_2 - x_3)/6$$

and, for example,

$$x_1^{(1)} = (17 + 2x_2^{(0)} - x_3^{(0)})/7$$

$$x_2^{(1)} = (-13 + x_1^{(0)} + 3x_3^{(0)} - x_4^{(0)})/9$$

$$x_3^{(1)} = (15 - 2x_1^{(0)} - x_4^{(0)})/10$$

$$x_4^{(1)} = (10 - x_1^{(0)} + x_2^{(0)} - x_3^{(0)})/6.$$

Substitute $x^{(0)} = (0,0,0,0)$ into the right-hand side of each of these equations to get

$$x_1^{(1)} = (17 + 2 \cdot 0 - 0)/7 = 2.428 571 429$$
 $x_2^{(1)} = (-13 + 0 + 3 \cdot 0 - 0)/9 = -1.444 444 444$
 $x_3^{(1)} = (15 - 2 \cdot 0 - 0)/10 = 1.5$
 $x_4^{(1)} = (10 - 0 + 0 - 0)/6 = 1.666 666 667$

and so $x^{(1)} = (2.428\ 571\ 429, -1.444\ 444\ 444, 1.5, 1.666\ 666\ 667)$. The Jacobi iterative process:

$$x_{1}^{(k+1)} = \left(17 + 2x_{2}^{(k)} - x_{3}^{(k)}\right) / 7$$

$$x_{2}^{(k+1)} = \left(-13 + x_{1}^{(k)} + 3x_{3}^{(k)} - x_{4}^{(k)}\right) / 9$$

$$x_{3}^{(k+1)} = \left(15 - 2x_{1}^{(k)} - x_{4}^{(k)}\right) / 10$$

$$x_{4}^{(k+1)} = \left(10 - x_{1}^{(k)} + x_{2}^{(k)} - x_{3}^{(k)}\right) / 6, \qquad k \ge 1.$$

We obtain a sequence that converges to

 $\mathbf{x}^{(9)} = (2.000127203, -1.000100162, 1.000118096, 1.000162172).$

4.2.2. Gauss-Seidel iterative method

Almost the same as Jacobi method, except that each x-value is improved using the most recent approx. of the other variables.

For a $n \times n$ system, the k + 1-th approximation is:

$$\begin{cases} x_1^{(k+1)} = u_{12}x_2^{(k)} + \ldots + u_{1n}x_n^{(k)} + c_1 \\ x_2^{(k+1)} = u_{21}x_1^{(k+1)} + u_{23}x_3^{(k)} + \ldots + u_{2n}x_n^{(k)} + c_2 \\ \ldots \\ x_n^{(k+1)} = u_{n1}x_1^{(k+1)} + \ldots + u_{nn-1}x_{n-1}^{(k+1)} + c_n, \end{cases}$$
 with $k = 0, 1, 2, \ldots; \ u_{ij} = -\frac{a_{ij}}{a_{ii}}, \ c_i = \frac{b_i}{a_{ii}}, \ i = 1, \ldots, n \ (as in Jacobi method)$

method).

Algorithmic form:

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

for each i = 1, 2, ...n, and for $k \ge 1$.

Stopping criterions: $\left|x^{(k)}-x^{(k-1)}\right|<\varepsilon$, or $\frac{\left|\mathbf{x}^{(k)}-\mathbf{x}^{(k-1)}\right|}{\left|\mathbf{x}^{(k)}\right|}<\varepsilon$, with ε - a prescribed tolerance, $\varepsilon>0$.

Remark 2 Because the new values can be immediately stored in the location that held the old values, the storage requirements for \mathbf{x} with the Gauss-Seidel method is half than that for Jacobi method and the rate of convergence is faster.

Example 3 Solve the following system using the Gauss-Seidel iterative method. Use $\varepsilon = 10^{-3}$ and $\mathbf{x}^{(0)} = (0\ 0\ 0\ 0)$ as the starting vector.

$$\begin{cases} 7x_1 - 2x_2 + x_3 & = 17 \\ x_1 - 9x_2 + 3x_3 - x_4 & = 13 \\ 2x_1 + 10x_3 + x_4 & = 15 \\ x_1 - x_2 + x_3 + 6x_4 & = 10 \end{cases}$$

We have

$$x_1 = (17 + 2x_2 - x_3)/7$$

$$x_2 = (-13 + x_1 + 3x_3 - x_4)/9$$

$$x_3 = (15 - 2x_1 - x_4)/10$$

$$x_4 = (10 - x_1 + x_2 - x_3)/6,$$

and, for example,

$$x_1^{(1)} = (17 + 2x_2^{(0)} - x_3^{(0)})/7$$

$$x_2^{(1)} = (-13 + x_1^{(1)} + 3x_3^{(0)} - x_4^{(0)})/9$$

$$x_3^{(1)} = (15 - 2x_1^{(1)} - x_4^{(0)})/10$$

$$x_4^{(1)} = (10 - x_1^{(1)} + x_2^{(1)} - x_3^{(1)})/6,$$

which provide the following Gauss-Seidel iterative process:

$$x_{1}^{(k+1)} = \left(17 + 2x_{2}^{(k)} - x_{3}^{(k)}\right) / 7$$

$$x_{2}^{(k+1)} = \left(-13 + x_{1}^{(k+1)} + 3x_{3}^{(k)} - x_{4}^{(k)}\right) / 9$$

$$x_{3}^{(k+1)} = \left(15 - 2x_{1}^{(k+1)} - x_{4}^{(k)}\right) / 10$$

$$x_{4}^{(k+1)} = \left(10 - x_{1}^{(k+1)} + x_{2}^{(k+1)} - x_{3}^{(k+1)}\right) / 6, \quad \text{for } k \ge 1.$$

Substitute $\mathbf{x}^{(0)} = (0,0,0,0)$ into the right-hand side of each of these equations to get

$$x_1^{(1)} = (17 + 2 \cdot 0 - 0)/7 = 2.428 571 429$$

 $x_2^{(1)} = (-13 + 2.428 571 429 + 3 \cdot 0 - 0)/9 = -1.1746031746$
 $x_3^{(1)} = (15 - 2 \cdot 2.428 571 429 - 0)/10 = 1.0142857143$
 $x_4^{(1)} = (10 - 2.428 571 429 - 1.1746031746 - 1.0142857143)/6$
 $= 0.8970899472$

and so

 $\mathbf{x}^{(1)} = (2.428571429 - 1.1746031746, 1.0142857143, 0.8970899472).$

Similar procedure generates a sequence that converges to

 $\mathbf{x}^{(5)} = (2.000025, -1.000130, 1.000020.0.999971).$

4.2.3. Relaxation method

In case of convergence, the Gauss-Seidel method is faster than Jacobi method. The convergence can be more improved using **relaxation method (SOR method)** (SOR=Succesive Over Relaxation)

Algorithmic form of the method:

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

for each i = 1, 2, ...n, and for $k \ge 1$.

For $0 < \omega < 1$ the procedure is called **under relaxation method**, that can be used to obtain convergence for systems which are not convergent by Gauss-Siedel method.

For $\omega > 1$ the procedure is called **over relaxation method**, that can be used to accelerate the convergence for systems which are convergent by Gauss-Siedel method.

By Kahan's Theorem follows that the method converges for $0 < \omega < 2$.

Remark 4 For $\omega = 1$, relaxation method is Gauss-Seidel method.

Example 5 Solve the following system, using relaxation iterative method. Use $\varepsilon = 10^{-3}$, $\mathbf{x}^{(0)} = (1\ 1\ 1)$ and $\omega = 1.25$,

$$4x_1 + 3x_2 = 24$$

 $3x_1 + 4x_2 - x_3 = 30$
 $-x_2 + 4x_3 = -24$

We have

$$x_1^{(k)} = 7.5 - 0.937x_2^{(k-1)} - 0.25x_1^{(k-1)}$$

$$x_2^{(k)} = 9.375 - 9.375x_1^{(k)} + 0.3125x_3^{(k-1)} - 0.25x_2^{(k-1)}$$

$$x_3^{(k)} = -7.5 + 0.3125x_2^{(k)} - 0.25x_3^{(k-1)}, \text{ for } k \ge 1.$$

The solution is (3, 4, -5).

4.2.4 The matriceal formulations of the iterative methods

Split the matrix A into the sum

$$A = D + L + U,$$

where D is the diagonal of A, L the lower triangular part of A, and U the upper triangular part of A. That is,

$$D = \begin{bmatrix} a_{11} & 0 & \cdots & 0 \\ 0 & \cdots & \ddots & \vdots \\ \vdots & \cdots & \cdots & 0 \\ 0 & \cdots & 0 & a_{nn} \end{bmatrix}, \quad L = \begin{bmatrix} 0 & \cdots & 0 \\ a_{21} \vdots & & \ddots \\ \vdots & \cdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{n,n-1} & 0 \end{bmatrix},$$

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

The system Ax = b can be written as

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

The **Jacobi method** in matriceal form is given by:

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

the Gauss-Seidel method in matriceal form is given by:

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

and the relaxation method in matriceal form is given by:

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

Convergence of the iterative methods

Remark 6 The convergence (or divergence) of the iterative process in the Jacobi and Gauss-Seidel methods does not depend on the initial guess, but depends only on the character of the matrices themselves. However, a good first guess in case of convergence will make for a relatively small number of iterations.

A sufficient condition for convergence:

Theorem 7 (Convergence Theorem) If A is strictly diagonally dominant, then the Jacobi, Gauss-Seidel and relaxation methods converge for any choice of the starting vector $\mathbf{x}^{(0)}$.

Example 8 Consider the system of equations

$$\begin{bmatrix} 3 & 1 & 1 \\ -2 & 4 & 0 \\ -1 & 2 & -6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 1 \\ 2 \end{bmatrix}.$$

The coefficient matrix of the system is strictly diagonally dominant since

$$|a_{11}| = |3| = 3 > |1| + |1| = 2$$

 $|a_{22}| = |4| = 4 > |-2| + |0| = 2$
 $|a_{33}| = |-6| = 6 > |-1| + |2| = 3$.

Hence, if the Jacobi or Gauss-Seidel method are used to solve the system of equations, they will converge for any choice of the starting vector $\mathbf{x}^{(0)}$.

Example 9 Consider the linear system

$$4x_1 + x_2 = 3$$
$$2x_1 + 5x_2 = 1.$$

Perform two iterations of Jacobi, Gauss-Seidel and relaxation methods to this system, beginning with the vector x = [3, 11] and for $\omega = 1.25$.

(Solutions of the system are 7/9 and -1/9).