

553.385 Numerical Linear Algebra, Spring 2022

Section 9

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April 7, 2022

Theorem 1 (Banach fixed point). *Let (X, d) be a complete metric space and $F : X \mapsto X$ a contractive mapping ($\exists \lambda < 1$ such that $d(F(x), F(y)) \leq \lambda d(x, y)$ for all $x, y \in X$). Then F admits a unique fixed point. Furthermore, it can be obtained as the limit of any sequence defined by $x_{n+1} = F(x_n)$ with $x_0 \in X$.*

Proof. Let $x_0 \in X$ and consider the sequence defined iteratively by $x_{n+1} = F(x_n)$ for all $n \in \mathbb{N}$, we can show that $\forall n \in \mathbb{N}$ and $\forall p \in \mathbb{N}$ we have

$$d(x_n, x_{n+p}) \leq \frac{\lambda^n}{1 - \lambda} d(x_0, x_1)$$

$\lambda < 1$ results that $\forall \varepsilon > 0, \exists N \in \mathbb{N}$ such that $\forall n \geq N$ and $p \in \mathbb{N}$, $d(x_n, x_{n+p}) \leq \varepsilon$. This shows that (x_n) is a Cauchy sequence and as X is complete, it means that (x_n) is converging in X , i.e. $\exists x_* \in X$ such that $x_n \rightarrow x_*$. Now, taking the limit in $x_{n+1} = F(x_n)$ and using the fact that F is continuous (since it is contractive), we have $x_* = F(x_*)$ which shows that x_* is a fixed point of F .

As for the uniqueness, if \tilde{x} is another fixed point of F , we have

$$d(x_*, \tilde{x}) = d(F(x_*), F(\tilde{x})) \leq \lambda d(x_*, \tilde{x}) \implies d(x_*, \tilde{x}) = 0, \text{ i.e. } \tilde{x} = x_*$$

□

Theorem 2 (Convergence theorems in lecture notes). $\forall \mathbf{x}^0 \in \mathbb{R}^n$, the iteration $\mathbf{x}^{k+1} = \mathbf{T}\mathbf{x}^k + \mathbf{c}$ converges to the unique solution \mathbf{x} of $\mathbf{x} = \mathbf{T}\mathbf{x} + \mathbf{c}$ iff $\rho(\mathbf{T}) < 1$.

Proof. See lecture notes.

□

Iterative techniques for solving linear systems

Given $A \in \mathbb{R}^{n \times n}$, $b \in \mathbb{R}^n$, consider solving $Ax = b$. We write A as $A = D - L - U$, that is

$$\underbrace{\begin{pmatrix} a_{11} & \cdots & \cdots & a_{1n} \\ a_{21} & \cdots & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & \cdots & \cdots & a_{n,n} \end{pmatrix}}_A = \underbrace{\begin{pmatrix} a_{11} & 0 & 0 & 0 \\ 0 & \ddots & \ddots & 0 \\ \vdots & \ddots & \ddots & 0 \\ 0 & \cdots & 0 & a_{n,n} \end{pmatrix}}_D - \underbrace{\begin{pmatrix} 0 & 0 & 0 & 0 \\ -a_{21} & \ddots & 0 & 0 \\ \vdots & \ddots & \ddots & 0 \\ -a_{n1} & \cdots & -a_{n,n-1} & 0 \end{pmatrix}}_L - \underbrace{\begin{pmatrix} 0 & -a_{12} & \cdots & -a_{1n} \\ 0 & \ddots & \ddots & \vdots \\ 0 & 0 & \ddots & -a_{n-1,n} \\ 0 & 0 & 0 & 0 \end{pmatrix}}_U$$

Jacobi method

$$\begin{aligned}
 A\mathbf{x} = \mathbf{b} &\implies (D - L - U)\mathbf{x} = \mathbf{b} \\
 &\implies D\mathbf{x} = (L + U)\mathbf{x} + \mathbf{b} \\
 &\implies \mathbf{x} = \underbrace{D^{-1}(L + U)\mathbf{x}}_{T_J} + \underbrace{D^{-1}\mathbf{b}}_{\mathbf{c}}
 \end{aligned}$$

By theorem 2, our iteration is

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

that is,

$$x_i^{(k+1)} = \frac{-\sum_{j \neq i} a_{ij}x_j^{(k)} + b_i}{a_{ii}}, \quad i = 1, 2, \dots, n$$

or

$$x_i^{(k+1)} - x_i^{(k)} = \frac{-\sum_{j=1}^n a_{ij}x_j^{(k)} + b_i}{a_{ii}}, \quad i = 1, 2, \dots, n$$

The second iteration structure is what we used in demo_alg071.m, and the stop criterion is given by

$$\|\mathbf{x}^{(k+1)} - \mathbf{x}^{(k)}\| \leq \text{TOL}$$

Gauss-Seidel method

$$\begin{aligned}
 A\mathbf{x} = \mathbf{b} &\implies (D - L - U)\mathbf{x} = \mathbf{b} \\
 &\implies (D - L)\mathbf{x} = U\mathbf{x} + \mathbf{b} \\
 &\implies \mathbf{x} = \underbrace{(D - L)^{-1}U\mathbf{x}}_{T_G} + \underbrace{(D - L)^{-1}\mathbf{b}}_{\mathbf{c}}
 \end{aligned}$$

By theorem 2, our iteration is

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

that is,

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

Comments: For stop criterion, given the tolerance $\text{TOL} = \varepsilon$, when k large enough,

$$\|\mathbf{x}^{(k+1)} - \mathbf{x}^{(k)}\| \approx \|\mathbf{x} - \mathbf{x}^{(k)}\| \approx [\rho(T)]^k \|\mathbf{x} - \mathbf{x}^{(0)}\| < \varepsilon$$

which is equivalent to

$$\begin{aligned}
 k \underbrace{\log(\rho(T))}_{<0} + \log(\|\mathbf{x} - \mathbf{x}^{(0)}\|) &< \log(\varepsilon) \\
 \iff k > \frac{\log(\varepsilon)}{\log(\rho(T))} - \underbrace{\frac{\log(\|\mathbf{x} - \mathbf{x}^{(0)}\|)}{\log(\rho(T))}}_{\text{constant}}
 \end{aligned}$$

Theorem 3. If A strictly diagonally dominant, then for any choice of $\mathbf{x}^{(0)}$ both Jacobi and Gauss-Seidel methods converge and $\|T_G\|_\infty \leq \|T_J\|_\infty < 1$

Definition 1 (Irreducible matrix). A matrix A is said to be irreducible if \nexists permutation matrix P such that

$$P^T A P = \begin{bmatrix} \tilde{A}_{11} & \tilde{A}_{12} \\ O & \tilde{A}_{22} \end{bmatrix}, \quad \tilde{A}_{11} \in \mathbb{E}(p, p), \quad p > 0$$

Definition 2 (Irreducibly diagonally dominant). A matrix is said to be Irreducibly diagonally dominant if it is irreducible and

$$|a_{ii}| \geq \sum_{j \neq i} |a_{ij}| \quad \text{and} \quad |a_{ss}| > \sum_{j \neq s} |a_{sj}| \quad \text{for at least one } s$$

Theorem 4. A matrix A is irreducible iff $G(A)$ is connected.

Theorem 5. If A is irreducibly diagonally dominant, then the Jacobi method converges.

Relaxation Methods

SOR (Successive Over-Relaxation)

$$\begin{aligned} A\mathbf{b} = \mathbf{b} &\implies (D - L - U)\mathbf{x} = \mathbf{b} \\ &\implies (\omega D - \omega L - \omega U)\mathbf{x} = \omega \mathbf{b} \\ &\implies (D - (1 - \omega)D - \omega L - \omega U)\mathbf{x} = \omega \mathbf{b} \\ &\implies (D - \omega L)\mathbf{x} = [(1 - \omega)D + \omega U]\mathbf{x} + \omega \mathbf{b} \\ &\implies \mathbf{x} = \underbrace{(D - \omega L)^{-1}[(1 - \omega)D + \omega U]\mathbf{x}}_{T(\omega)} + \underbrace{(D - \omega L)^{-1}\omega \mathbf{b}}_{\mathbf{c}} \end{aligned}$$

In general, the iteration is

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

then,

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

Recall our Gauss-Seidel iteration, then

$$\begin{aligned} \mathbf{x}^{(k+1)} &= (1 - \omega) \underbrace{\mathbf{x}^{(k)}}_{\text{last iteration}} + \omega \underbrace{D^{-1} [L\mathbf{x}^{(k+1)} + U\mathbf{x}^{(k)} + \mathbf{b}]}_{\text{Gauss-Seidel iteration}} \\ &= (1 - \omega)\mathbf{x}^{(k)} + \omega \mathbf{x}_{GS}^{(k+1)} \end{aligned}$$

which is a “weighted average” (or more precisely, an affine combination) of the new Gauss-Seidel value with the one obtained during the previous iteration. Here ω is the so-called relaxation parameter. If $\omega = 1$ we simply have the Gauss-Seidel method. For $\omega > 1$ of overrelaxation (accelerated convergence of GS), and for $\omega < 1$ of underrelaxation (can converge if GS non-convergent).