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Convergence

If $c(x) > 0$, then $\max_i |u_i - \phi_i| \leq \frac{1}{96} |\max \phi^{(9)}(x)| \cdot \epsilon$

Error Analysis

$$AX = b, \quad A: \text{invertible}$$

x : exact solution, $x = A^{-1}b$

\tilde{x} : approximate solution

$e = x - \tilde{x}$: error

$r = b - A\tilde{x}$: residual

Claim: $Ae = r$

Pf

$$Ae = A(x - \tilde{x}) = Ax - A\tilde{x} = b - A\tilde{x} = r$$

Note

$$e = 0 \Leftrightarrow r = 0$$

However, small residual \checkmark doesn't always imply that error \hat{e} is also small

Def A vector norm $\|x\|$ has the following properties

1. $\|x\| \geq 0$ and $\|x\| = 0$ iff $x = 0$

2. $\|\alpha x\| = |\alpha| \cdot \|x\|$ where α is a scalar

3. $\|x + y\| \leq \|x\| + \|y\|$: triangle inequality

Ex let $x = (x_1, \dots, x_n)^T$.

$$\|x\|_{\infty} = \max \{ |x_i|, i=1, \dots, n \} \quad \text{l}_{\infty} \text{ norm}$$

$$\|x\|_2 = \left(\sum_{i=1}^n |x_i|^2 \right)^{1/2} \quad \text{l}_2 \text{ norm}$$

$$\|x\|_1 = \sum_{i=1}^n |x_i| \quad \text{l}_1 \text{ norm}$$

Ex $x = (1, 1, 1)$ $\|x\|_{\infty} = 1$, $\|x\|_2 = \sqrt{3}$, $\|x\|_1 = 3$

Check that $\|x\|_{\infty}$ is indeed a norm

Pr

1. Clearly $\|x\|_{\infty} \geq 0$

$$\|x\|_{\infty} = 0 \Leftrightarrow \max_i |x_i| = 0 \Leftrightarrow |x_i| = 0 \Leftrightarrow x_i = 0$$

$$\Leftrightarrow x = 0$$

2. $\|\alpha x\|_{\infty} = \max_i |\alpha x_i|, i=1, \dots, n =$

$$= |\alpha| \cdot \max \{ |x_i|, i=1, \dots, n \} = |\alpha| \cdot \|x\|_{\infty}$$

3. $\|x+y\|_{\infty} = \max \{ |x_i+y_i|, i=1, \dots, n \} \leq$

$$\leq \max \{ |x_i| + |y_i|, i=1, \dots, n \} \leq$$

$$\leq \max \{ |x_i|, i=1, \dots, n \} + \max \{ |y_i|, i=1, \dots, n \} =$$

$$= \|x\|_{\infty} + \|y\|_{\infty}$$

$$\underline{\text{Ex}} \quad \overbrace{\begin{pmatrix} 1.01 & 0.99 & 2 \\ 0.99 & 1.01 & 2 \end{pmatrix}}^A \quad x = \begin{pmatrix} 1 \\ 1 \end{pmatrix} : \text{exact solutions}$$

$$b = \begin{pmatrix} 2 \\ 2 \end{pmatrix}$$

$$\tilde{x}_1 = \begin{pmatrix} 1.01 \\ 1.01 \end{pmatrix} \quad e_1 = x - \tilde{x}_1 = \begin{pmatrix} -0.01 \\ -0.01 \end{pmatrix} \quad r_1 = b - A\tilde{x}_1 = \begin{pmatrix} -0.02 \\ -0.02 \end{pmatrix}$$

$$\tilde{x}_2 = \begin{pmatrix} 2 \\ 0 \end{pmatrix} \quad e_2 = x - \tilde{x}_2 = \begin{pmatrix} -1 \\ 1 \end{pmatrix} \quad r_2 = b - A\tilde{x}_2 = \begin{pmatrix} -0.02 \\ 0.02 \end{pmatrix}$$

thus $\|e_1\|_\infty = 0.01, \quad \|e_2\|_\infty = 1$

$$\|r_1\|_\infty = \|r_2\|_\infty = 0.02$$

$$\|x\|_\infty = 1$$

$$\|b\|_\infty = 2$$

$$\frac{\|e_1\|_\infty}{\|x\|_\infty} = \frac{\|r_1\|_\infty}{\|b\|_\infty} \quad \text{but} \quad \frac{\|e_2\|_\infty}{\|x\|_\infty} = 100 \cdot \frac{\|r_2\|_\infty}{\|b\|_\infty}$$

$$\frac{0.01}{1} = \frac{0.02}{2} \quad \frac{1}{1} = 100 \cdot \frac{0.02}{2} = 0.01$$

Q What is the relation between relative error

$$\frac{\|e\|}{\|x\|} \quad \text{and} \quad \text{relative residual} \quad \frac{\|r\|}{\|b\|} \quad ?$$

Def A matrix norm $\|A\|$ has the following properties:

1. $\|A\| \geq 0$ and $\|A\| = 0 \Leftrightarrow A = 0$
2. $\|\alpha A\| = |\alpha| \cdot \|A\|$, where α is a scalar
3. $\|A+B\| \leq \|A\| + \|B\|$
4. $\|A \cdot B\| \leq \|A\| \cdot \|B\|$

Given a vector norm $\|x\|$, the subordinate (or induced) matrix norm is defined by

$$\|A\| = \max_{x \neq 0} \frac{\|Ax\|}{\|x\|} \quad \left(\|A\| = \max_{\|x\|=1} \|Ax\| \right)$$

The subordinate matrix norm has an additional property:

5. $\|Ax\| \leq \|A\| \cdot \|x\|$ for all vectors x .

Ex

$$A = \begin{pmatrix} 4 & 1 & 0 \\ 1 & 4 & 1 \\ 0 & 1 & 4 \end{pmatrix}$$

$$\|A\|_{\infty} = \max_{x \neq 0} \frac{\|Ax\|_{\infty}}{\|x\|_{\infty}} = 6$$

(proof later)

$$x_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \quad Ax_1 = \begin{pmatrix} 4 \\ 1 \\ 0 \end{pmatrix} \quad \Rightarrow \quad \frac{\|Ax_1\|_{\infty}}{\|x_1\|_{\infty}} = \frac{4}{1} = 4$$

$$x_2 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \quad Ax_2 = \begin{pmatrix} 5 \\ 5 \\ 1 \end{pmatrix} \Rightarrow \frac{\|Ax_2\|_\infty}{\|x_2\|_\infty} = \frac{5}{1} = 5$$

$$x_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \quad Ax_3 = \begin{pmatrix} 5 \\ 6 \\ 5 \end{pmatrix} \Rightarrow \frac{\|Ax_3\|_\infty}{\|x_3\|_\infty} = \frac{6}{1} = 6$$

Claim $\|A\|_\infty = \max_i \sum_{j=1}^n |a_{ij}|$ ("max row sum")