

Math 551

4/15

- No class on 4/20 (w schedule)
- HW5 returned
- HW7 due 4/21

Recall:

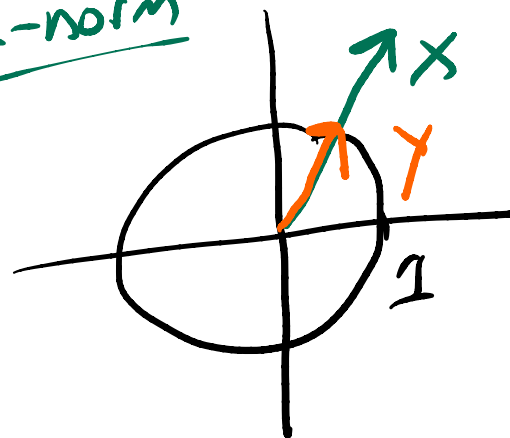
$$\|A\| \stackrel{\text{defn}}{=} \max_{x \neq 0} \frac{\|Ax\|}{\|x\|}$$

$$= \max_{x \neq 0} \frac{\frac{1}{\|x\|}}{\frac{1}{\|x\|}} \frac{\|Ax\|}{\|x\|}$$

$$= \max_{x \neq 0} \frac{\|A(\frac{x}{\|x\|})\|}{\|\frac{x}{\|x\|}\|} = 1$$

$$= \max_{\|y\|=1} \frac{\|Ay\|}{1} = \max_{\|x\|=1} \|Ax\|$$

2-norm



$$y = \frac{1}{\|x\|} x$$

$$\Rightarrow \|y\| = 1$$

Iterative Methods for $Ax=b$

$A = N - P$ splitting where N^{-1} exists

$$Ax=b \Rightarrow (N-P)x=b$$

$$\Rightarrow Nx = Px + b$$

$$\Rightarrow x = N^{-1}(Px + b)$$

$$\Rightarrow x = N^{-1}Px + N^{-1}b$$

Choose $x^{(0)}$: $x^{(k+1)} = N^{-1}Px^{(k)} + N^{-1}b$

In
Practice

$$Nx^{(k+1)} = Px^{(k)} + b \quad k=1, 2, 3, \dots$$

Ex: Jacobi's Method

$$A = L + D + U = \underbrace{D}_N + \underbrace{(L+U)}_{-P}$$

Not $LU=A$

$$Ax = b \Rightarrow (L + D + U)x = b$$

$$\Rightarrow Dx = -(L+U)x + b$$

$$\text{iteration: } Dx^{(k+1)} = -(L+U)x^{(k)} + b$$

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

$$x^{(k+1)} = M_J x^{(k)} + D^{-1}b, \quad \underbrace{M_J = -D^{-1}(L+U)}_{\text{Jacobi Iteration Matrix}}$$

Ex: $A = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$, $b = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$, $x = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

$L = \begin{bmatrix} 0 & 0 \\ -1 & 0 \end{bmatrix}$, $D = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$, $U = \begin{bmatrix} 0 & -1 \\ 0 & 0 \end{bmatrix}$

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

or $x^{(k+1)} = -D^{-1}(L+U) x^{(k)} + D^{-1}b$
 $= M_J x^{(k)} + \underbrace{D^{-1}b}_{\tilde{b}}$

where $M_J = -D^{-1}(L+U) = -\begin{bmatrix} 1/2 & 0 \\ 0 & 1/2 \end{bmatrix} \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$
 $= \begin{bmatrix} 0 & 1/2 \\ 1/2 & 0 \end{bmatrix}$

Iteration:
$$\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} x_1^{(k+1)} \\ x_2^{(k+1)} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_1^{(k)} \\ x_2^{(k)} \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

or $2x_1^{(k+1)} = x_2^{(k)} + 1$

$2x_2^{(k+1)} = x_1^{(k)} + 1$

$2x_1^{(2)} = x_2^{(1)} + 1$
 $= \frac{1}{2} + 1 = \frac{3}{2}$

k	$x_1^{(k)}$	$x_2^{(k)}$
0	0	0

initial guess

1 $\frac{1}{2}$ $\frac{1}{2}$

2 $\frac{3}{4}$ $\frac{3}{4}$

3 $\frac{7}{8}$ $\frac{7}{8}$

4 $\frac{15}{16}$ $\frac{15}{16}$

$\downarrow$ $\downarrow$ $\downarrow$
 ∞ 1 1

$\Rightarrow$ Convergence!

$\|e^{(3)}\|_{\infty} = \left\| \begin{bmatrix} 1 \\ 1 \end{bmatrix} - \begin{bmatrix} \frac{7}{8} \\ \frac{7}{8} \end{bmatrix} \right\|_{\infty}$
 $= \left\| \begin{bmatrix} \frac{1}{8} \\ \frac{1}{8} \end{bmatrix} \right\|_{\infty} = \frac{1}{8}$

Error: $Ax = b$ $X^{(k+1)} = M_J X^{(k)} + \tilde{b}$
 $X = M_J X + \tilde{b}$

$$\begin{aligned} e^{(k)} &\stackrel{\text{defn}}{=} X - X^{(k)} \\ &= (M_J X + \tilde{b}) - (M_J X^{(k)} + \tilde{b}) \\ &= M_J (X - X^{(k)}) \end{aligned}$$

$$e^{(k)} = M_J e^{(k-1)}$$

$$\text{or } e^{(k)} = M_J e^{(k-1)} = M_J (M_J e^{(k-2)}) = M_J^2 e^{(k-2)}$$

$$e^{(k)} = \dots = M_J^k e^{(0)}$$

So

$$\|e^{(k)}\| = \|M_J^k e^{(0)}\|$$

$$\|AB\| \leq \|A\| \|B\|$$

(6)

$$\|Ax\| \leq \|A\| \|x\|$$

$$\leq \|M_J^k\| \|e^{(0)}\|$$

Fixed
vector

$$\|e^{(k)}\| \leq \|M_J\|^k \|e^{(0)}\| \quad e^{(0)} = x - x^{(0)}$$

So, it is SUFFICIENT to show
convergence if we can find an INDUCED
matrix norm such that

$$\|M_J\|_\infty = \left\| \begin{bmatrix} 0 & 1/2 \\ 1/2 & 0 \end{bmatrix} \right\|_\infty = \frac{1}{2} < 1$$

Gauss-Seidel $A = L + D + U$, $\boxed{d_{i,i}} = a_{i,i} \neq 0$
 $1 \leq i \leq n$

$$Ax = b \Rightarrow (L + D + U)x = b$$

$$\Rightarrow (L + D)x = -Ux + b$$

Given $x^{(0)}$:

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

$$\begin{aligned} \Rightarrow x^{(k+1)} &= -(L + D)^{-1} U x^{(k)} + (L + D)^{-1} b \\ &= M_{GS} x^{(k)} + \tilde{b} \end{aligned}$$

where $M_{GS} = -(L + D)^{-1} U$ is the

Gauss-Seidel Iteration matrix.

Back to our example: $A = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$

$$M_{GS} = -(L+D)^{-1}U = -\begin{bmatrix} 2 & 0 \\ -1 & 2 \end{bmatrix}^{-1} \begin{bmatrix} 0 & -1 \\ 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1/2 \\ 0 & 1/4 \end{bmatrix}$$

$$\|M_{GS}\|_{\infty} = \frac{1}{2} < \|M_{GS}\|_1 = \frac{3}{4}$$

and $(L+D)X^{(k+1)} = -UX^{(k)} + b$

$$\begin{bmatrix} 2 & 0 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} x_1^{(k+1)} \\ x_2^{(k+1)} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1^{(k)} \\ x_2^{(k)} \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\Rightarrow \left. \begin{aligned} 2x_1^{(k+1)} &= x_2^{(k)} + 1 \\ 2x_2^{(k+1)} &= x_1^{(k+1)} + 1 \end{aligned} \right\} \begin{array}{l} \text{Since you have } x_1^{(k+1)} \\ \text{use it for any} \\ \text{occurrence of} \\ x_1 \end{array}$$

Iteration

$$2x_1^{(k+1)} = x_2^{(k)} + 1$$

$$2x_2^{(k+1)} = x_1^{(k+1)} + 1$$

$$\begin{aligned} 2x_1^{(1)} &= x_2^{(0)} + 1 \\ &= 0 + 1 \end{aligned}$$

$$\begin{aligned} 2x_2^{(1)} &= x_1^{(1)} + 1 \\ &= \frac{1}{2} + 1 = \frac{3}{2} \end{aligned}$$

k	$x_1^{(k)}$	$x_2^{(k)}$
0	0	0
1	$\frac{1}{2}$	$\frac{3}{4}$
2	$\frac{7}{8}$	$\frac{15}{16}$
3	$\frac{31}{32}$	$\frac{63}{64}$
4	$\frac{127}{128}$	$\frac{255}{256}$
$\downarrow$	$\downarrow$	$\downarrow$
∞	1	1

Convergence

Note: For Jacobi $\|e^{(3)}\|_{\infty} = \frac{1}{8}$

$$\begin{aligned}\text{Gauss-Seidel } \|e^{(3)}\|_{\infty} &= \left\| \begin{bmatrix} 1 \\ 1 \end{bmatrix} - \begin{bmatrix} 3/32 \\ 63/64 \end{bmatrix} \right\|_{\infty} \\ &= \left\| \begin{bmatrix} 1/32 \\ 1/64 \end{bmatrix} \right\|_{\infty} = \frac{1}{32}\end{aligned}$$

So we see for our $Ax=b$

GS is much faster than Jacobi!!

Why? $\|M_J\|_{\infty} = \frac{1}{2}$ and $\|M_{GS}\|_{\infty} = \frac{1}{2}$

They are equal, so WHY is GS faster!