

## Gauss Seidel

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Error Analysis:

$x^*$  is a solution to

$$Ax = b.$$

$$\text{Then, } \|x^* - x^{(k+1)}\| \leq \|N^{-1}P\|^{k+1} \|x^* - x^{(0)}\|$$

$$\text{where } Nx^{(k+1)} = b + Px^{(k)} \quad k=0,1,\dots$$

for given choice of  $x^{(0)}$ .

$$P = N - A$$

For Jacobi,

$$N = \begin{bmatrix} a_{11} & 0 & 0 \\ 0 & a_{22} & 0 \\ 0 & 0 & a_{33} \end{bmatrix}$$

$$P = N - A = \begin{bmatrix} 0 & -a_{12} & -a_{13} \\ -a_{21} & 0 & -a_{23} \\ -a_{13} & -a_{32} & 0 \end{bmatrix}$$

Any iterative method: given  $x^{(0)}$

$$Nx^{(k+1)} = b + Px^{(k)} \quad k=0,1,\dots$$

converges if  $\|N^{-1}P\| < 1$ .

example: Given  $2x_1 + x_2 = 1$   
 $x_1 + 2x_2 = 5$

without calculating any iterates  $x^{(1)}, x^{(2)}, \dots$   
 based on  $x^{(0)}$ , check if Jacobi method  
 converges.

$$Ax = b$$

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \quad b = \begin{bmatrix} 1 \\ 5 \end{bmatrix}$$

Jacobi Method: Guess  $x^{(0)}$

$$N^J x^{(k+1)} = b + P^J x^{(k)} \quad k=0,1,\dots$$

$$N^J = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

$$P^J = N^J - A = \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix} \quad \text{All remaining entries of } A \text{ with a neg. sign.}$$

$$= \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$$

$$\|(N^J)^{-1} P^J\| < 1 ?$$

$$N^J = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \rightarrow (N^J)^{-1} = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix}$$

$$(N^J)^{-1} P^J = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & -0.5 \\ -0.5 & 0 \end{bmatrix} \begin{matrix} 0+0.5 = 0.5 \\ 0.5+0 = 0.5 \end{matrix}$$

$$\|(N^J)^{-1} P^J\| = \max \{0.5, 0.5\} = 0.5 < 1$$

$\Rightarrow$  The Jacobi Method converges since  $\|(N^J)^{-1} P^J\| < 1$ .

$$\|x^* - x^{(k+1)}\| \leq \underbrace{\|(N^J)^{-1} P^J\|}_{R+1} \underbrace{\|x^* - x^{(0)}\|}_{R+1}$$

$$\|x^* - x^{(n+1)}\| \leq \|(\text{N}^{-1})^{-1} P J\| \underbrace{\|x^* - x^{(n)}\|}_{\substack{\text{error in} \\ \text{initial guess}}}$$

$0 < \|(\text{N}^{-1})^{-1} P J\|^{k+1}$  if  $\|(\text{N}^{-1})^{-1} P J\| < 1$   
 $\hookrightarrow 0 \text{ as } k \rightarrow \infty.$

Example: #12 from 6.6

Consider the iterative method:

$$x^{(n+1)} = b + \alpha \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} x^{(n)}$$

$n=0, 1, \dots$

where  $\alpha$  is an unknown number.

Find the values of  $\alpha$  for which the above method converges. Irrespective of  $A, b$  &  $x^{(0)}$ .

$$N x^{(n+1)} = b + P x^{(n)}$$

$$\text{where } N = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, P = \alpha \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}.$$

Convergence of given iterative method:

$$x^{(n+1)} = b + \alpha \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} x^{(n)}$$

depends on  $\|N^{-1} P\| < 1$  or not!

$$N^{-1} P = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2\alpha & \alpha \\ \alpha & 2\alpha \end{bmatrix}$$

Row sum  
1 1 1 1

$$\hookrightarrow \text{...} = \begin{bmatrix} 0 & 1 & \alpha & -\alpha \\ 2\alpha & \alpha & \xrightarrow{\substack{\text{Row sum} \\ |2\alpha| + |\alpha|}} & \\ \alpha & 2\alpha & \xrightarrow{\substack{\text{Row sum} \\ |\alpha| + |2\alpha|}} & \end{bmatrix}$$

For convergence, we need  $\|N^{-1}P\| < 1$

$$\begin{aligned}\|N^{-1}P\| &= \max \{ |2\alpha| + |\alpha|, |\alpha| + |2\alpha| \} \\ &= 3|\alpha|\end{aligned}$$

choose  $\alpha$  such that

$$3|\alpha| < 1$$

$$|\alpha| < \frac{1}{3}$$

$$-\frac{1}{3} < \alpha < \frac{1}{3}.$$

### Gauss Seidel Method (Method of Successive Replacements)

$$\begin{array}{lll} x_1^{(1)} & 2x_1 - x_2 & = 0 \rightarrow (1) \\ x_1^{(1)}, x_2^{(1)} & -x_1 + 2x_2 - x_3 & = 1 \rightarrow (2) \\ & -x_1 + 2x_3 & = 2 \rightarrow (3) \end{array}$$

formula for GS (Gauss Seidel)?

Diagonally Dominant matrix  $A \Rightarrow$  GS method converge.

$$N^{GS} x^{(k+1)} = b + P^{GS} x^{(k)}$$

$$\| (N^{GS})^{-1} P^{GS} \| < 1 \text{ is}$$

equivalent to  $A$  being strictly diagonally dominant

equivalent to A being strictly diagonally dominant.

Deriving NGS & PGS (remember PGS = NGS - A !)

Given  $X^{(0)} = \begin{bmatrix} b_1/a_{11} \\ b_2/a_{22} \\ b_3/a_{33} \end{bmatrix}$

$$b = \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} \quad a_{11}=2 \\ a_{22}=2 \\ a_{33}=2.$$

$$X^{(0)} = \begin{bmatrix} 0 \\ 1/2 \\ 1 \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ 1 \\ 1.5 \end{bmatrix} \dots \rightarrow \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix}$$

Derive GS  $X^{(1)}$  ?

use  $2x_1 - x_2 = 0$  to update  $x_1^{(1)}$

$$2x_1^{(1)} - x_2^{(0)} = 0 \rightarrow x_1^{(1)} = 0$$

use  $x_1^{(1)} = 0$  and eqn ② to update  $x_2^{(1)}$ .

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

$$0 + 2x_2^{(1)} - 1 = 1$$

$$x_2^{(1)} = 1$$

use  $x_1^{(1)} = 0$  &  $x_2^{(1)} = 1$  in eqn ③ and update  $x_3^{(1)}$

$$-x_2 + 2x_3 = 2$$

$$-x_2^{(1)} + 2x_3^{(1)} = 2$$

$$-1 + 2x_3^{(1)} = 2$$

$$-1 + 2x_3^{(1)} = 2$$

$$x_3^{(1)} = \frac{3}{2}$$

What are NGS and PGS for any:

①

$$a_{11}x_1^{(0)} + a_{12}x_2^{(0)} + a_{13}x_3^{(0)} = b_1$$

②

$$a_{21}x_1^{(1)} + a_{22}x_2^{(1)} + a_{23}x_3^{(0)} = b_2$$

③

$$a_{31}x_1^{(1)} + a_{32}x_2^{(1)} + a_{33}x_3^{(1)} = b_3$$

$$Ax = b$$

$$\text{NGS } x^{(1)} = b + \text{PGS } x^{(0)}$$

express ① - ③

NGS

$$\begin{bmatrix} a_{11} \\ a_{21} \\ a_{31} \end{bmatrix} \quad \begin{bmatrix} 0 \\ a_{22} \\ a_{32} \end{bmatrix}$$

$$\begin{bmatrix} 0 \\ 0 \\ a_{32} \end{bmatrix} \quad \begin{bmatrix} x_1^{(1)} \\ x_2^{(1)} \\ x_3^{(1)} \end{bmatrix}$$

$$\begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$$\begin{bmatrix} a_{12}x_2^{(0)} - a_{13}x_3^{(0)} \\ -a_{23}x_3^{(0)} \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} + \begin{bmatrix} 0 & -a_{12} & -a_{13} \\ 0 & 0 & -a_{23} \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1^{(0)} \\ x_2^{(0)} \\ x_3^{(0)} \end{bmatrix}$$

$$\text{NGS } x^{(k+1)} = b + \text{PGS } x^{(k)}$$

where

$$\text{NGS} = \begin{bmatrix} a_{11} & 0 & 0 \\ a_{21} & a_{22} & 0 \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

where  $N = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$

$$P_{GS} = \begin{bmatrix} 0 & -a_{12} & -a_{13} \\ 0 & 0 & -a_{23} \\ 0 & 0 & 0 \end{bmatrix} = N^{GS} - A$$

$$\|(N^{GS})^{-1} P_{GS}\| < 1$$

For GS iterative method,

if  $A$  is strictly d.d i.e

$$|a_{12}| + |a_{13}| < |a_{11}|$$

$$|a_{21}| + |a_{23}| < |a_{22}|$$

$$|a_{31}| + |a_{32}| < |a_{33}|$$

$$\Rightarrow \|(N^{GS})^{-1} P_{GS}\| < 1$$

example:

$$\begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

here  $A$  is strictly D.D  $\Rightarrow$

GS method converges!

LOOK LECTURE SLIDES

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$$AX = b$$

if  $\det A \neq 0$   
 $\Rightarrow$  solution  $x$  exists  
 $A^{-1}b$ .

Sometimes, the quality of  $A$   
 small changes in  $b$  cause large deviations  
 in  $x$ .

example:  $\begin{bmatrix} 5 & 7 \\ 7 & 10 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.7 \\ 1 \end{bmatrix}$

$A \quad x \quad b$

Replace  $b$  with  $\begin{bmatrix} 0.69 \\ 1.01 \end{bmatrix} = \hat{b}$  Perturbed load vector/RHS vector

Solution to  $A\hat{x} = \hat{b}$

is  $\hat{x} = \begin{bmatrix} -0.17 \\ 0.22 \end{bmatrix}$

while solution to original problem :

$$Ax = b$$

is  $x = \begin{bmatrix} 0 \\ 0.1 \end{bmatrix}$

magnitude of perturbation  $\|b - \hat{b}\| = 0.01$

magnitude of perturbation  $\|b - \hat{b}\| = 0.01$

Ill Conditioned matrix : A is said to be ill conditioned if small changes in b

$$b \rightarrow \hat{b} \text{ where } \|b - \hat{b}\| = \epsilon \ll 1$$

(much smaller than 1)

result in large deviations in the

Solutions to  $Ax = b$   
 $A\hat{x} = \hat{b}$ .

Condition Number of a matrix :

$$\text{cond}(A, \inf) = \|A\| \|A^{-1}\|$$

$$\|A * A^{-1}\| = \|I\| \quad I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \approx 1$$

$$\|A\| \|A^{-1}\| \approx 1$$

$\|A\| \|A^{-1}\| \gg 1$  then A is

said to be ill conditioned.

$$A = \begin{bmatrix} 5 & 7 \\ 7 & 10 \end{bmatrix}_{12}^{17}$$

$$A^{-1} = \begin{bmatrix} 10 & -7 \\ -7 & 5 \end{bmatrix} \xrightarrow{\text{Row sum}} \begin{bmatrix} 17 \\ 12 \end{bmatrix}$$

$$\|A\| = \max\{12, 17\} = 17 \quad \|A^{-1}\| = 17$$

$$\text{Cond}(A) = 17^2$$

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$$\text{cond}(A) = 17^2$$

$\Rightarrow A$  is ill conditioned matrix & solutions will be sensitive to changes in  $b$ .

Homework 05 :

$$A_n = \begin{bmatrix} 1 & -1 & \dots & -1 \\ 1 & 1 & -1 & \dots \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix}$$

Input  $n \rightarrow$  returns  $A_n^\top$

$\text{cond}(A_n) \rightarrow \text{cond}(A_n, \inf)$

(III)  $|\alpha| + |\beta| < 1$

$$\|N^{-1}P\| = \|M\|.$$