

I Méthode de Richardson

$\alpha > 0$

$$\begin{cases} x_0 \in \mathbb{R}^m \\ x_{h+1} = x_h + \alpha (b - Ax_h) \end{cases}$$

1) Posons $r_h = b - Ax_h$

on a $x_{h+1} = x_h + \alpha r_h$

On pose $M = \frac{1}{\alpha} I_m$ inverse

des

$$x_{h+1} = x_h + M^{-1} r_h$$

D'autre $A = M - N \Rightarrow N = M - A$

$$N = \frac{1}{\alpha} I_m - A$$

2) $\lambda \in \mathbb{C}$

~~On suppose que~~

$$\begin{aligned} M^{-1} N &= \alpha \left(\frac{1}{\alpha} I_m - A \right) \\ &= I_m - \alpha A \end{aligned}$$

D'autre

$$\det(M^{-1} N - \mu I_m) = 0 \Leftrightarrow \det\left(\frac{1}{\alpha} I_m - A - \mu I_m\right) = 0$$

$$\Leftrightarrow \det\left(\left(\frac{1-\mu}{\alpha}\right) I_m - A\right) = 0$$

$$\Leftrightarrow \alpha^m \det\left(\left(\frac{1-\mu}{\alpha}\right) I_m - A\right) = 0$$

$$\Leftrightarrow (-\alpha)^m \det\left(A - \frac{(1-\mu)}{\alpha} I_m\right) = 0$$

$$\Leftrightarrow \frac{1-\mu}{\alpha} \text{ vp de } A$$

$$\Leftrightarrow \exists \lambda \text{ vp de } A \text{ tq}$$

$$\lambda = \frac{1-\mu}{\alpha}$$

$$\Leftrightarrow \exists \lambda \text{ vp de } A \text{ tq } \mu = 1 - \alpha \lambda$$

D'au $\boxed{\lambda \text{ vp de } A \Rightarrow 1-\alpha\lambda \text{ vp de } \tilde{N}}$

3) $\forall \lambda \text{ vp de } A \Rightarrow \lambda \in \mathbb{R}$

$\forall \alpha \in \mathbb{R}^m \text{ convergence} \Rightarrow \rho(\tilde{N}) < 1$

$\Rightarrow |1-\alpha\lambda| < 1$

$\forall \lambda \text{ vp de } A$

$\Rightarrow -1 \leq 1-\alpha\lambda \leq 1$

$\forall \lambda \text{ vp de } A$

$\boxed{\forall \alpha \in \mathbb{R}^m \text{ convergence} \Rightarrow 0 \leq \alpha \leq 2}$
 $\forall \lambda \text{ vp de } A$

4) AGS method dit partii

• Near $f(x) = \frac{1}{2} x^T A x - b^T x$

ou a $Df(x) = Ax - b$

D'au le schéma s'écrit :

$\begin{cases} x_0 \in \mathbb{R}^n \\ x_{h+1} = x_h - \alpha Df(x_h) \end{cases}$

• Steepest descent? $\alpha_h = \frac{\|Df(x_h)\|^2}{\|Df(x_h)\|_A^2}$

my

II Préconditionnement

$P^T A x = P^T b$

5) $x_{h+1} = x_h + \alpha (P^T b - P^T A x_h)$

$\boxed{x_{h+1} = x_h + \tilde{P}^T (b - A x_h)}$

$$x_{h+1} = x_h + P^{-1} r_h$$

D'au $\boxed{N = P}$ inverse

on a alors $N = M - A$

$$\boxed{N = P - A}$$

7) a) $P = D = \text{diag}(A) \rightarrow$ Jacobi
inverse n' est pas

on a $x_{h+1} = x_h + D^{-1}(b - Ax_h)$

D'au

$$D x_{h+1} = D x_h + b - A x_h$$

on $A = D - E - F$ on

D'au $D x_{h+1} = (D - D + E + F) x_h + b$

$$\boxed{D x_{h+1} = (E + F) x_h + b}$$

Jacobi

b) $P = D - E$ inverse n' est pas

D'au

$$\begin{aligned} (D - E) x_{h+1} &= (D - E) x_h + b - A x_h \\ &= (D - E - D + E + F) x_h + b \\ &= F x_h + b \end{aligned}$$

$$\boxed{(D - E) x_{h+1} = F x_h + b}$$

Gauss-Seidel