

October 10, 2013

Preconditioning:

$$r_k = P_k(A) r_0, \quad P_k(0) = 1$$

want $P_k(\lambda_i(A))$ small in magnitude

GMRES, basic form:

$$x_k = x_0 + Q_k z_k$$

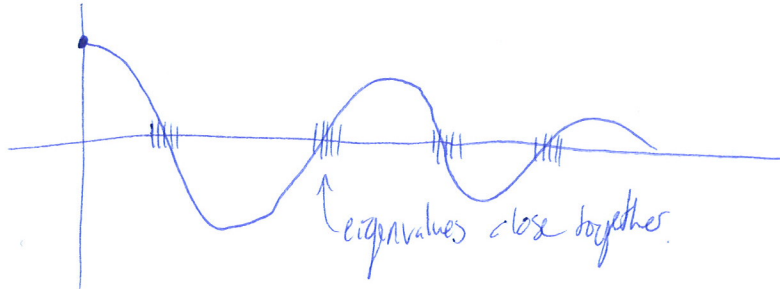
$$\text{where } z_k = \arg \min_z \| r_0 - H_{k+1,k} z \|_2 \quad \rho = \| r_0 \|_2$$

So we use a preconditioner M

$$M^{-1}Ax = M^{-1}b \quad \text{left preconditioning (could also } AM^{-1}u = b, \quad x = M^{-1}u \text{ right preconditioning)}$$

hope: $P_k(M^{-1}A)$ or $P_k(AM^{-1})$ is "good"

$P_k(\lambda_i(AM^{-1}))$ or $P_k(\lambda_i(M^{-1}A))$ small for all λ_i



left + right preconditioning

$$l: M^{-1}Ax = M^{-1}b$$

$$r: AM^{-1}u = b, \quad x = M^{-1}u$$

Notice that the corresponding preconditioned Krylov subspaces are a bit different

$$\text{right: } \text{span} \{ b, AM^{-1}b, \dots \}$$

$$\text{left: } \text{span} \{ M^{-1}b, (M^{-1}A)M^{-1}b \}$$

doesn't make a huge difference

M is non

M should be much easier to invert than A , but still $M^{-1} \approx A^{-1}$
in a certain way ... (clustering or less important, good condition number of $M^{-1}A$)

Note: instead of a matrix-vector product w/ A originally, now we have:

mat-vec prod

plus need to solve $Mz_k = S_k$

mat-vec prod w/ $M^{-1}A$

\hookrightarrow given y , need $\underbrace{M^{-1}Ay}_S = z$

* Compute $s = Ay$

* solve $Mz = s$

Also have split preconditioning: $\underbrace{M_1^{-1} A M_2^{-1}}_{\text{pre-conditioned matrix}} (\underbrace{M_2 x}_u) = \underbrace{M_1^{-1} b}_b$

$$M = M_1 M_2$$

eg: M is SPD

$$M = FF^T \text{ (Cholesky)}$$

\uparrow in factored form

Split preconditioning + CG:

$$Ax = b$$

A is SPD; need M SPD

recall CG: Lanczos

$$T_{k,k} z = p e_1$$

$$x_k = x_0 + Q_k z.$$

PCG: $M^{-1}A$ is not even symmetric

but if M is SPD: $M = Q \Lambda Q^T$

$$M^{1/2} = Q \Lambda^{1/2} Q^T$$

$$M^{1/2} M^{1/2} = Q \Lambda^{1/2} \Lambda^{1/2} Q^T = Q \Lambda Q^T = M$$

$$(M^{-1/2} A M^{-1/2}) (M^{1/2} x) = F^{-1/2} b$$

if $M = FF^T$

$$x^T F^T A F^{-1} x = (F^{-1} x)^T A (F^{-1} x)$$

$$= y^T A y$$

$$> 0.$$

if A is not symmetric

$$Ax = b$$

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

then $A^T A$ is SPD if A is nonsingular.

CGLS

but if the condition number of A is large \rightarrow the condition number of $A^T A$ squares it.

$A^T A \rightarrow$ apply CG only if A is well-conditioned

$\hookrightarrow A$: non-symmetric

Preconditioners:

we want $M^{-1} \approx A^{-1}$

but much easier to invert

Simple preconditioners:

$M = I$ or αI : Richardson

$M = \text{diag}(A)$: Jacobi

$M = \text{tril}(A)$: Gauss Seidel

Incomplete LU:

$$A \approx LU$$

static pattern ILU } And sparse L, U

dynamic pattern ILU } s.t. $A = LU$ in some places, $A \approx LU$ in some places, $A \neq LU$ in places we don't care about...



$$A = LU$$

$$A = \begin{pmatrix} \text{diagonal} & \text{upper triangular} \end{pmatrix} \begin{pmatrix} \text{lower triangular} & \text{diagonal} \end{pmatrix}$$

instead

$$L = \begin{pmatrix} \diagup & & \\ & \diagdown & \\ & & \diagup \end{pmatrix}$$

$$U = \begin{pmatrix} \diagdown & & \\ & \diagup & \\ & & \diagdown \end{pmatrix}$$

Impose: $A = LU$ only for $\{(i,j)\}$ for which $A_{ij} \neq 0$