Recall: Iterative solvers

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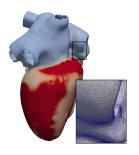




What we want to solve

We want to solve partial differential equations (PDEs)

- on parallel computers (distributed memory + shared memory)
- with $10^3 10^4$ CPU cores
- for $10^6 10^8$ discretization points.





Considered Problem Classes

Find
$$u:$$
 $Lu(x) = f(x)$ $\forall x \in \Omega$ $lu(x) = g(x)$ $\forall x \in \partial \Omega$

variational \downarrow formulation

$$\mathsf{Find} \quad u \in \mathbb{V} \colon \qquad \mathsf{a}(u,v) = \ \langle \mathsf{F}, v \rangle \qquad \forall v \in \mathbb{V}$$

FEM, FDM ↓ FVM, FIT

Solve
$$K_h \cdot \underline{u}_h = \underline{f}_h$$
 $\underline{u}_h \in \mathbb{R}^{N_h}$

With K_h

- symmetric, i.e., $K_h \equiv K_h^T$
- positive definite, i.e., $\langle K_h \underline{u}_h, \underline{u}_h \rangle > 0$ $\forall \underline{u}_h \not\in \ker(K_h)$

- (linear) 2nd order problem.
 - Poisson equation (temperature)
 - Lamé equation (deformation)
 - Maxwell's equations (magnetic field)
- Matrix K_h is sparse, positive definite (symmetric, large dimension)
- non-linear and time-dependent problems.

Richardson iteration to solve Ku = f

Richardson iteration:

$$\underline{u}^{k+1} := \underline{u}^k + \tau \left(\underline{f} - K \cdot \underline{u}^k\right)$$

with
$$au_{opt} = rac{2}{\lambda_{max}(K) + \lambda_{min}(K)}$$
.

Preconditioned Richardson iteration:

$$\underline{u}^{k+1} := \underline{u}^k + C^{-1} \left(\underline{f} - K \cdot \underline{u}^k\right)$$

with $C \approx K$ and $C\underline{u} = \underline{f}$ can be faster solved than $K\underline{u} = \underline{f}$.

Choice $C:=\frac{1}{\tau}I$ (scaled Identity) results in classical Richardson.

Richardson iteration: algorithm

Algorithm 1 Preconditioned Richardson for Ku = f

1: Choose
$$u^0 = 0$$
; $k = 0$

initial guess
 ini

2:
$$s0 = \langle C^{-1} \cdot f, f \rangle$$

3: repeat

4:
$$r := f - K \cdot \mu^k$$

▷ defect

5: Solve
$$Cw = r$$

6:
$$u^{k+1} := u^k + w$$

⊳ update

7: **until**
$$\langle w, r \rangle > s0 \cdot \varepsilon^2$$

▷ relative accuracy

Basic operations:

- sparse matrix-vector product
- vector operations, inner product
- preconditioning system Cw = r

Jacobi iteration

Special Richardson iteration with preconditioner

$$C = \frac{1}{\omega}D$$

denoting D := diag(K).

• Jacobi iteration (solver: $\omega := 1$):

$$\underline{u}^{k+1} := \underline{u}^k + \omega D^{-1} \left(\underline{f} - K \cdot \underline{u}^k \right)$$

with the preconditioning step

$$w := \omega D^{-1} \cdot r$$

ILU iteration

Special Richardson iteration with preconditioner

$$C = L \cdot U \iff L \cdot U \cdot w = r$$

resulting from an incomplete factorization $K = L \cdot U + R$, with a remainder R.

ILU iteration:

$$\underline{u}^{k+1} := \underline{u}^k + U^{-1}L^{-1}\left(\underline{f} - K \cdot \underline{u}^k\right)$$

• with the preconditioning steps

solve
$$L \cdot v = r$$

solve $U \cdot w = v$

Preconditioned cg iteration to solve Ku = f

Choose
$$\underline{u}^0$$
 $\underline{r} := \underline{f} - K \cdot \underline{u}^0$
 $\underline{w} := \underline{C^{-1} \cdot \underline{r}}$
 $\underline{s} := \underline{w}$
 $\sigma := \sigma_{\text{old}} := \sigma_0 := (\underline{w}, \underline{r})$
repeat
$$\underline{v} := K \cdot \underline{s}$$
 $\alpha := \sigma/(\underline{s}, \underline{v})$
 $\underline{u} := \underline{u} + \alpha \cdot \underline{s}$
 $\underline{r} := \underline{r} - \alpha \cdot \underline{v}$
 $\underline{w} := \underline{C^{-1} \cdot \underline{r}}$
 $\sigma := (\underline{w}, \underline{r})$
 $\beta := \sigma/\sigma_{\text{old}}$
 $\sigma_{\text{old}} := \sigma$
 $\underline{s} := \underline{w} + \beta \cdot \underline{s}$
until $\sigma < \varepsilon^2 \cdot \sigma_0$

Preconditioning opportunities

Many opportunities for choosing C in $Cw = r \iff w = C^{-1}r$

- C = I: no preconditioning, or scaled as $C = \frac{1}{\omega}I$
- $C = \frac{1}{\omega}D$: Jacobi preconditioning / diagonal scaling (one Jacobi iteration with initial guess $u^0 = 0$)
- $C = L \cdot U$: ILU-preconditioning
- Approximativ inverse as SPAI
- Additive/multiplicative Schwarz methods (ASM/MSM) incl.
 - Multilevel (ASM)
 - Multigrid (MSM)
 - Domain decomposition methods (DDM)
- Multigrid: $C^{-1} = (I_\ell M_\ell)K^{-1}$ with

$$M_k = S_k^{post} (I_k - P_k(I_{k-1} - M_{k-1})K_{k-1}^{-1}R_kK_k) S_k^{pre}$$

or as two-grid: $\mathit{M}_1 = \mathit{S}_1^{\mathit{post}} \left(\mathit{I}_1 - \mathit{P}_1 \mathit{K}_0^{-1} \mathit{R}_1 \mathit{K}_1 \right) \mathit{S}_1^{\mathit{pre}}$