Programming linear solver

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Linear problem

We want to find x such that

$$Ax = b \tag{1}$$

A is symmetric positive definite. The function

$$J(y) = \frac{1}{2}y^{\top}Ay - b^{\top}y$$

is strictly convex and has unique minimum x which is characterized by

$$J'(x) = Ax - b = 0$$

The solution of (1) is the unique minimizer of J.

Steepest descent method

Descent direction

$$d_n = -J'(x_n) = b - Ax_n =: r_n$$

Determine step size

$$J(x_n + \omega_n d_n) = \min_{\omega} J(x_n + \omega d_n) \implies \omega_n = \frac{d_n^{\top} r_n}{d_n^{\top} A d_n}$$

Update

$$x_{n+1} = x_n + \omega_n d_n$$

Conjugate direction method

In first step, descent direction

$$d_0 = -J'(x_0) = b - Ax_0$$

Then for n = 1, 2, ..., choose next descent direction d_n to be A-orthogonal to all previous descent directions

$$d_j^{\top} A d_n = 0, \qquad 0 \le j \le n - 1$$

Determine step size

$$J(x_n + \omega_n d_n) = \min_{\omega} J(x_n + \omega d_n) \implies \omega_n = \frac{d_n^{\top} r_n}{d_n^{\top} A d_n}$$

Update

$$x_{n+1} = x_n + \omega_n d_n$$

CG algorithm

Assume initial guess x_0 is given. Otherwise set $x_0 = 0$ Set n = 0

$$r_0 = b - Ax_0, \qquad d_0 = r_0$$

While $||r_n|| > \varepsilon ||r_0||$ and $n < N_{max}$

• If n > 0

$$d_n = r_n + \frac{\|r_n\|^2}{\|r_{n-1}\|^2} d_{n-1}$$

- $\omega_n = \frac{\|r_n\|^2}{d_n^\top A d_n}$
- $\bullet \ x_{n+1} = x_n + \omega_n d_n$
- $r_{n+1} = r_n \omega_n A d_n$
- n = n + 1

Need matrix-vector product and dot product of two vectors.

CG algorithm: Computer version

Assume initial guess x is given. Otherwise set x = 0Set n = 0

$$r = b - Ax$$
, $d = r$, $\rho_0 = ||r_0||^2$

While $\sqrt{\rho_n} > \varepsilon \sqrt{\rho_0}$ and $n < N_{max}$

• If n > 0

$$d \leftarrow r + \frac{\rho_n}{\rho_{n-1}} d$$

- v = Ad
- $\omega = \frac{\rho_n}{d^\top v}$
- $x \leftarrow x + \omega d$
- $r \leftarrow r \omega v$
- $\rho_{n+1} = ||r||^2$
- $n \leftarrow n+1$

Remark: Need storage for three vectors: r, d, v

Jacobi method

Solve i'th equation of

$$Ax = b$$

for x_i

$$x_i = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1, j \neq i}^{N} a_{ij} x_j \right]$$

Jacobi method: Make an initial guess x^0 and then iterate

$$x_i^{n+1} = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1, j \neq i}^{N} a_{ij} x_j^n \right]$$

Rewrite

$$x_i^{n+1} = x_i^n + \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^N a_{ij} x_j^n \right]$$

Jacobi method

Algorithm:

Make initial guess x, set n = 0

- $1 r = b Ax, \ \rho = ||r||$
- 2 If n=0, then $\rho_0=\rho$
- **3** If $\rho < \varepsilon \rho_0$, then stop.
- **4** For $i = 1, 2, \dots, N$

$$x_i \leftarrow x_i + \frac{r_i}{a_{ii}}$$

6 $n \leftarrow n+1$, go to step (1)

Remark: Need additional storage for vector r.

Remark: Very good for parallelization

Gauss-Seidel (GS) method

Solve i'th equation of

$$Ax = b$$

for x_i

$$x_i = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1, j \neq i}^{N} a_{ij} x_j \right]$$

GS method: Make an initial guess x^0 and then iterate

$$x_i^{n+1} = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{n+1} - \sum_{j=i+1}^{N} a_{ij} x_j^n \right]$$

Rewrite

$$x_i^{n+1} = x_i^n + \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{n+1} - \sum_{j=i}^{N} a_{ij} x_j^n \right]$$

Gauss-Seidel (GS) method

Algorithm:

Make initial guess x, set $\rho = \rho_0 = 1$

- **1** If $\rho < \varepsilon \rho_0$, then stop.
- **2** $\rho = 0$
- **3** For $i = 1, 2, \dots, N$

$$r = b_i - \sum_{j=1}^{N} a_{ij} x_j$$
$$x_i \leftarrow x_i + \frac{r}{a_{ii}}$$
$$\rho \leftarrow \rho + r * r$$

- $\bullet \rho \leftarrow \sqrt{\rho}$
- **6** If n = 0, $\rho_0 = \rho$
- **6** $n \leftarrow n+1$, go to step (1)

Remark: No additional storage is required.

Remark: Cannot be parallelized.

Successive Over Relaxation (SOR) method

A variation on GS method. Choose $\omega > 1$

For i = 1, 2, ..., N

$$x_{i}^{*} = \frac{1}{a_{ii}} \left[b_{i} - \sum_{j=1}^{i-1} a_{ij} x_{j}^{n+1} - \sum_{j=i}^{N} a_{ij} x_{j}^{n} \right] + x_{i}^{n}$$

$$x_{i}^{n+1} = x_{i}^{n} + \omega(x_{i}^{*} - x_{i}^{n})$$

or equivalently

$$x_i^{n+1} = x_i^n + \frac{\omega}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{n+1} - \sum_{j=i}^{N} a_{ij} x_j^n \right]$$

Remark: $\omega = 1$ yields the GS method.

Successive Over Relaxation (SOR) method

Algorithm:

Make initial guess x, set $\rho = \rho_0 = 1$

- **1** If $\rho < \varepsilon \rho_0$, then stop.
- **2** $\rho = 0$
- **3** For $i = 1, 2, \dots, N$

$$r = b_i - \sum_{j=1}^{N} a_{ij} x_j$$
$$x_i \leftarrow \omega \frac{r}{a_{ii}} + x_i$$
$$\rho \leftarrow \rho + r * r$$

- $\bullet \rho \leftarrow \sqrt{\rho}$
- **6** If n = 0, $\rho_0 = \rho$
- **6** $n \leftarrow n+1$, go to step (1)

Symmetric Gauss-Seidel (SGS) method

For i = 1, 2, ..., N

$$x_i^* = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^* - \sum_{j=i}^{N} a_{ij} x_j^n \right] + x_i^n$$

For i = N, N - 1, ..., 1

$$x_i^{n+1} = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^* - \sum_{j=i}^{N} a_{ij} x_j^{n+1} \right] + x_i^*$$

Remark: Symmetric SOR (SSOR) method is obtained by doing over-relaxation in SGS method.