

Iterative solvers for linear equations

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August 21, 2014

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Given a square system of n linear equations:

$$\mathbf{Ax} = \mathbf{b}. \quad (1)$$

Some (very slow) direct techniques exist that involve algebra. We will solve this equation by an iterative (also called relaxation) method. We will use a sequence of $\mathbf{x}^{(k)}$ that converges toward the fixed point \mathbf{x} , solution of our problem.

In conclusion, we want to write, for a given $\mathbf{x}^{(0)}$, a sequence $\mathbf{x}^{(k+1)} = F(\mathbf{x}^{(k)})$ with $k \in \mathbb{N}$.

1 Jacobi's method

We decompose $A = D - L - U$ where D is an invertible matrix.

$$\mathbf{Ax} = \mathbf{b} \Leftrightarrow \mathbf{Dx} = (\mathbf{L} + \mathbf{U})\mathbf{x} + \mathbf{b} \quad (2)$$

$$\Leftrightarrow \mathbf{x} = \mathbf{D}^{-1}(\mathbf{L} + \mathbf{U})\mathbf{x} + \mathbf{D}^{-1}\mathbf{b} \quad (3)$$

$$= F(\mathbf{x}) \quad (4)$$

where F is a linear function of \mathbf{x} .

It is shown that we can solve this equation by iteratively solving, where k indicates the iteration number

$$\mathbf{x}^{(k+1)} = \mathbf{D}^{-1}(\mathbf{L} + \mathbf{U})\mathbf{x}^{(k)} + \mathbf{D}^{-1}\mathbf{b}. \quad (5)$$

Since $\text{diag}((\mathbf{L} + \mathbf{U})\mathbf{x}^{(k)}) = \mathbf{0}$, each $x_i^{(k+1)}$ is independent of $x_i^{(k)}$. It is instead a weighted average of other $x_{j \neq i}$.

Since we don't know the solution *a priori*, we define the residual at each step k as $\mathbf{r}^{(k)} = \mathbf{b} - \mathbf{Ax}^{(k)}$.

References

- [1] http://fr.wikipedia.org/wiki/M%C3%A9thode_de_Jacobi
- [2] <http://mathworld.wolfram.com/JacobiMethod.html>
- [3] http://en.wikipedia.org/wiki/Jacobi_method
- [4] http://sfb649.wiwi.hu-berlin.de/fedc_homepage/xplore/ebooks/html/csa/node38.html

Algorithm 1 Jacobi's algorithm

Output

- \mathbf{x} , an approximate solution to $\mathbf{Ax} = \mathbf{b}$ where \mathbf{A} and \mathbf{b} are known.

Inputs

- \mathbf{A}
- \mathbf{b}
- $\mathbf{x}^{(0)}$ including boundary conditions. No restriction on $\mathbf{x}^{(0)}$.
- ϵ , the tolerance or convergence criteria. A suitable tolerance might be $\|\mathbf{r}^{(k)}\| < \epsilon$ or $\|\mathbf{r}^{(k)} - \mathbf{r}^{(k-1)}\| < \epsilon$.

Restrictions

- \mathbf{A} must be strictly diagonally dominant, *i.e.*, $|A_{i,i}| > \sum_{j=1, n; j \neq i} |A_{i,j}| \Rightarrow A_{i,i} \neq 0$.
- $\epsilon > 0$

Algo

1. $\mathbf{D} \leftarrow \text{diag}(\mathbf{A})$
 2. $\mathbf{R} \leftarrow \mathbf{D} - \mathbf{A}$
 3. $\mathbf{D}^{-1} \leftarrow 1/\mathbf{D}$
 4. do while $\xi \geq \epsilon$
 - (a) $\mathbf{x}_{\text{old}} \leftarrow \mathbf{x}$
 - (b) $\mathbf{x} \leftarrow \mathbf{D}^{-1} \mathbf{R} \mathbf{x}_{\text{old}} + \mathbf{D}^{-1} \mathbf{b}$
 - (c) $\mathbf{r} \leftarrow \mathbf{Ax} - \mathbf{b}$
 - (d) $\xi \leftarrow \|\mathbf{r}\|$
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