

Parallel Scientific Computing

Course AMS301 — Fall 2022 — Lecture 3

Iterative methods for linear systems (1)

Stationary methods & Application to finite differences

Find $\mathbf{x} \in \mathbb{R}^n$ such that $\boxed{\mathbf{Ax} = \mathbf{b}}$ with $\mathbf{A} \in \mathbb{R}^{n \times n}$ and $\mathbf{b} \in \mathbb{R}^n$.

Motivation

Many computational procedures require the solution of linear systems:

- for linear physical problems,
- for non-linear physical problems,
- for optimization procedures,
- ...

for many applications:

- electromagnetic compatibility, aeroacoustic studies, CFD,
- medical imaging, geophysical imaging,
- ...

From a mathematical point of view:

- discretized elliptic problem
⇒ linear system to solve!
- discretized hyperbolic problem with an implicit time stepping scheme
⇒ linear system to solve at each time step!

Find $\mathbf{x} \in \mathbb{R}^n$ such that $\boxed{\mathbf{Ax} = \mathbf{b}}$ with $\mathbf{A} \in \mathbb{R}^{n \times n}$ and $\mathbf{b} \in \mathbb{R}^n$.

Solution procedures

- ▶ **Direct methods:** Factorization of \mathbf{A} into triangular and diagonal matrices (ex. $\mathbf{A} = \mathbf{LU}$) and solution of simpler problems.

$$\mathbf{Ax} = \mathbf{b} \quad \Leftrightarrow \quad \mathbf{LUx} = \mathbf{b} \quad \Leftrightarrow \quad \begin{cases} \mathbf{Ly} = \mathbf{b} \\ \mathbf{Ux} = \mathbf{y} \end{cases}$$

Advantages: exact solution known after a given number of operations

Difficulties: heavy computational cost (*operations/memory*), hard to parallelize

- ▶ **Iterative methods:** Iterative procedure to minimizing an error $\|\mathbf{x}^{(\ell)} - \mathbf{x}_{\text{ref}}\|$ and/or a residual $\|\mathbf{Ax}^{(\ell)} - \mathbf{b}\|$.

$$\begin{cases} \mathbf{x}^{(0)} = \text{Iter}_{(0)}(\mathbf{A}, \mathbf{b}) \\ \mathbf{x}^{(\ell+1)} = \text{Iter}^{(\ell+1)}(\mathbf{x}^{(\ell)}, \mathbf{x}^{(\ell-1)}, \dots, \mathbf{A}, \mathbf{b}), \quad \text{pour } \ell \geq 0 \end{cases}$$

Advantages: limited cost per iteration (*operations/memory*), easy to parallelize

Difficulties: approximate solution, control of the convergence of the process

Find $\mathbf{x} \in \mathbb{R}^n$ such that $\boxed{\mathbf{Ax} = \mathbf{b}}$ with $\mathbf{A} \in \mathbb{R}^{n \times n}$ and $\mathbf{b} \in \mathbb{R}^n$.

► **Iterative methods:**

$$\left| \begin{array}{l} \mathbf{x}^{(0)} = \text{Iter}_{(0)}(\mathbf{A}, \mathbf{b}) \\ \mathbf{x}^{(\ell+1)} = \text{Iter}^{(\ell+1)}(\mathbf{x}^{(\ell)}, \mathbf{x}^{(\ell-1)}, \dots, \mathbf{A}, \mathbf{b}), \quad \text{pour } \ell \geq 0 \end{array} \right.$$

The **order** of the method is the numb. of steps which the current iter. depends on.
Stationary method if the functions $\text{Iter}^{(\ell)}$ are indep. of ℓ , otherwise **nonstationary**
Linear method if the functions $\text{Iter}^{(\ell)}$ are linear, otherwise **nonlinear**

Today, we consider stationary linear iterative schemes of first order:

$$\left| \begin{array}{l} \mathbf{x}^{(0)} \text{ given} \\ \mathbf{x}^{(\ell+1)} = \mathbf{B}\mathbf{x}^{(\ell)} + \mathbf{f}, \quad \ell \geq 0 \end{array} \right.$$

where $\mathbf{B} \in \mathbb{R}^{n \times n}$ is the **iteration matrix** and $\mathbf{f} \in \mathbb{R}^n$ depends on \mathbf{b} .

Iterative methods for linear systems

Stationary methods

System arising from a finite difference discretization

Find $\mathbf{x} \in \mathbb{R}^n$ such that $\boxed{\mathbf{Ax} = \mathbf{b}}$ with $\mathbf{A} \in \mathbb{R}^{n \times n}$ and $\mathbf{b} \in \mathbb{R}^n$.

We consider a general procedure:

$$\left| \begin{array}{l} \mathbf{x}^{(0)} \text{ given} \\ \mathbf{x}^{(\ell+1)} = \mathbf{B}\mathbf{x}^{(\ell)} + \mathbf{f}, \quad \ell \geq 0 \end{array} \right.$$

where $\mathbf{B} \in \mathbb{R}^{n \times n}$ is the **iteration matrix** and $\mathbf{f} \in \mathbb{R}^n$ depends on \mathbf{b} .

Definitions and properties

- **Consistent** method if the solution is a fixed point of the scheme (*i.e.* $\mathbf{x} = \mathbf{B}\mathbf{x} + \mathbf{f}$).

$$\boxed{\text{OK}} \text{ if and only if } \mathbf{f} = (\mathbf{I} - \mathbf{B})\mathbf{A}^{-1}\mathbf{b}$$

- **Convergent** method if $\lim_{\ell \rightarrow \infty} \mathbf{x}^{(\ell)} = \mathbf{x}$ for all $\mathbf{x}^{(0)}$.

$$\text{For a consistent method, } \boxed{\text{OK}} \text{ if and only if } \boxed{\rho(\mathbf{B}) < 1}$$

The spectral radius $\rho(\mathbf{B})$ is the max. of the absolute values of the eigenval. of \mathbf{B} .

Find $\mathbf{x} \in \mathbb{R}^n$ such that $\boxed{\mathbf{Ax} = \mathbf{b}}$ with $\mathbf{A} \in \mathbb{R}^{n \times n}$ and $\mathbf{b} \in \mathbb{R}^n$.

We consider a general procedure:

$$\left\{ \begin{array}{l} \mathbf{x}^{(0)} \text{ given} \\ \mathbf{x}^{(\ell+1)} = \mathbf{B}\mathbf{x}^{(\ell)} + \mathbf{f}, \quad \ell \geq 0 \end{array} \right.$$

where $\mathbf{B} \in \mathbb{R}^{n \times n}$ is the **iteration matrix** and $\mathbf{f} \in \mathbb{R}^n$ depends on \mathbf{b} .

Definitions and properties

- We use a **stopping criteria** on the number of iterations and the norm of the residual:

$$\ell \leq \ell_{\text{tol}} \quad \text{and} \quad \|\mathbf{r}^{(\ell)}\| / \|\mathbf{r}^{(0)}\| \leq \varepsilon_{\text{tol}}$$

with the **residual vector** $\boxed{\mathbf{r}^{(\ell)} := \mathbf{b} - \mathbf{Ax}^{(\ell)}}$.

For stationary methods, one has:

$$\begin{aligned} \mathbf{A}\mathbf{e}^{(\ell)} &= \mathbf{Ax} - \mathbf{Ax}^{(\ell)} = \mathbf{r}^{(\ell)} \\ \Rightarrow \|\mathbf{r}^{(\ell)}\| &\leq \|\mathbf{A}\| \|\mathbf{e}^{(\ell)}\| \quad \text{et} \quad \|\mathbf{e}^{(\ell)}\| \leq \|\mathbf{A}^{-1}\| \|\mathbf{r}^{(\ell)}\| \end{aligned}$$

→ *Choice of \mathbf{B} for fast convergence and efficient computation?*

Stationary methods — Standard methods [1/3]

We consider a **regular decomposition**: $\mathbf{A} = \mathbf{M} - \mathbf{N}$ where $\mathbf{M} \in \mathbb{R}^{n \times n}$ is invertible.

Stationary method

$\mathbf{x}^{(0)} \in \mathbb{C}^n$

for $\ell = 0, 1, \dots$ **do**

 | $\mathbf{M}\mathbf{x}^{(\ell+1)} = \mathbf{N}\mathbf{x}^{(\ell)} + \mathbf{b}$

end

Choices

	By points	By blocks
Jacobi	$\mathbf{M} = \mathbf{D}$	$\mathbf{M} = \mathbf{D}^{\text{blk}}$
Gauss-Seidel	$\mathbf{M} = \mathbf{D} + \mathbf{L}$	$\mathbf{M} = \mathbf{D}^{\text{blk}} + \mathbf{L}^{\text{blk}}$

$$\underbrace{\begin{bmatrix} \times & \times & \times \\ \times & \times & \times \\ \times & \times & \times \end{bmatrix}}_{\mathbf{A}} = \underbrace{\begin{bmatrix} \times & & \\ & \times & \\ & & \times \end{bmatrix}}_{\mathbf{D}} + \underbrace{\begin{bmatrix} & \times & \\ \times & \times & \\ & & \end{bmatrix}}_{\mathbf{L}} + \underbrace{\begin{bmatrix} & \times & \times \\ & & \times \\ & & \end{bmatrix}}_{\mathbf{U}}$$

Stationary methods — Standard methods [2/3]

We consider a **regular decomposition**: $\mathbf{A} = \mathbf{M} - \mathbf{N}$ where $\mathbf{M} \in \mathbb{R}^{n \times n}$ is invertible.

Stationary method *with relaxation*

$\mathbf{x}^{(0)} \in \mathbb{C}^n$

for $\ell = 0, 1, \dots$ **do**

$$\mathbf{M}\tilde{\mathbf{x}} = \mathbf{N}\mathbf{x}^{(\ell)} + \mathbf{b}$$

$$\mathbf{x}^{(\ell+1)} = \omega\tilde{\mathbf{x}} + (1 - \omega)\mathbf{x}^{(\ell)} \quad (\omega \text{ is a real parameter})$$

end

Choices

	By points	By blocks
Jacobi over relaxation (JOR)	$\mathbf{M} = \mathbf{D}$	$\mathbf{M} = \mathbf{D}^{\text{blk}}$
Successive over relaxation (SOR)	$\mathbf{M} = \mathbf{D} + \mathbf{L}$	$\mathbf{M} = \mathbf{D}^{\text{blk}} + \mathbf{L}^{\text{blk}}$

$$\underbrace{\begin{bmatrix} [\times] & [\times] & [\times] \\ [\times] & [\times] & [\times] \\ [\times] & [\times] & [\times] \end{bmatrix}}_{\mathbf{A}^{\text{blk}}} = \underbrace{\begin{bmatrix} [\times] & & \\ & [\times] & \\ & & [\times] \end{bmatrix}}_{\mathbf{D}^{\text{blk}}} + \underbrace{\begin{bmatrix} [\times] & & \\ [\times] & [\times] & \\ & & \end{bmatrix}}_{\mathbf{L}^{\text{blk}}} + \underbrace{\begin{bmatrix} & [\times] & [\times] \\ & & [\times] \\ & & \end{bmatrix}}_{\mathbf{U}^{\text{blk}}}$$

Convergence of stationary methods

- ▶ Convergence if and only if $\rho(\mathbf{B}) < 1$ with $\mathbf{B} = \mathbf{M}^{-1}\mathbf{N}$.
- ▶ If \mathbf{A} is a strict diagonal dominant matrix (i.e. $|a_{ii}| > \sum_{j \neq i} |a_{ij}|, \forall i$)
 - Jacobi converges
 - GS converges
 - SOR converges if $0 < \omega \leq 1$
- ▶ Si \mathbf{A} is a symmetric positive definite matrix (i.e. $\mathbf{A} = \mathbf{A}^*$ et $(\mathbf{A}\mathbf{x}, \mathbf{x}) > 0, \forall \mathbf{x} \neq 0$)
 - Jacobi converges if $(2\mathbf{D} - \mathbf{A})$ is a symmetric positive definite matrix
 - GS converges
 - SOR converges if and only if $0 < \omega < 2$

What we generally expect for the convergence rates:

Jacobi < Gauss-Seidel < SOR By points < By blocks
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Jacobi method ($\mathbf{M} = \mathbf{D}$ and $\mathbf{N} = -(\mathbf{L} + \mathbf{U})$)

```

 $\mathbf{x}^{(0)} \in \mathbb{R}^n$ 
for  $\ell = 0, 1, \dots$  do
  |  $\mathbf{D}\mathbf{x}^{(\ell+1)} = \mathbf{b} - (\mathbf{L} + \mathbf{U})\mathbf{x}^{(\ell)}$ 
end

```

Jacobi method (*rewriting*)

```

 $x_i^{(0)} \in \mathbb{R}$  for  $i = 1 \dots n$ 
for  $\ell = 0, 1, \dots$  do
  | for  $i = 1 \dots n$  do
    | |  $x_i^{(\ell+1)} = a_{ii}^{-1} \left( b_i - \sum_{i \neq j} a_{ij} x_j^{(\ell)} \right)$ 
    | end
  | end
end

```

Discussion

- Matrix-vector product with dense matrix ($\mathbf{L} + \mathbf{U}$)
- Linear combinations because \mathbf{D} is diagonal
- The iterations of the interior loop are independent.

(BLAS in //)
(Lin. combi. in //)

Gauss-Seidel method ($M = L + D$ and $N = -U$) $\mathbf{x}^{(0)} \in \mathbb{R}^n$ **for** $\ell = 0, 1, \dots$ **do**

$$\left| \begin{array}{l} (\mathbf{D} + \mathbf{L})\mathbf{x}^{(\ell+1)} = \mathbf{b} - \mathbf{U}\mathbf{x}^{(\ell)} \\ \Leftrightarrow \mathbf{D}\mathbf{x}^{(\ell+1)} = \mathbf{b} - \mathbf{L}\mathbf{x}^{(\ell+1)} - \mathbf{U}\mathbf{x}^{(\ell)} \end{array} \right.$$

endGauss-Seidel method (*rewriting*) $x_i^{(0)} \in \mathbb{R}$ for $i = 1 \dots n$ **for** $\ell = 0, 1, \dots$ **do****for** $i = 1 \dots n$ **do**

$$\left| \begin{array}{l} x_i^{(\ell+1)} = a_{ii}^{-1} \left(b_i - \sum_{j < i} a_{ij} x_j^{(\ell+1)} - \sum_{i < j} a_{ij} x_j^{(\ell)} \right) \end{array} \right.$$

end**end****Discussion**

- For each ℓ , solution of a inferior triangular system (Descent in //)
- The iterations of the interior loop are **dependent**:
For each i , the solution is updated by using the last available values.

Block Jacobi/Gauss-Seidel methods (*rewriting*)
 $\mathbf{x}_I^{(0)} \in \mathbb{R}^{n_I}$ for $I = 1 \dots n^{\text{blk}}$
for $\ell = 0, 1, \dots$ **do**

 | **for** $I = 1 \dots n^{\text{blk}}$ **do**

| | $\mathbf{A}_{II}\mathbf{x}_I^{(\ell+1)} = \mathbf{b}_I - \sum_{I \neq J} \mathbf{A}_{IJ}\mathbf{x}_J^{(\ell)}$ *if block Jacobi*

| | $\mathbf{A}_{II}\mathbf{x}_I^{(\ell+1)} = \mathbf{b}_I - \sum_{J < I} \mathbf{A}_{IJ}\mathbf{x}_J^{(\ell+1)} - \sum_{I < J} \mathbf{A}_{IJ}\mathbf{x}_J^{(\ell)}$ *if block G.-S.*

 | **end**
end
Discussion

- Interior loop over n^{blk} blocks of $\mathbf{x}^{(\ell)}$, with $n^{\text{blk}} \leq n$.
- Jacobi: Matrix-vector product with dense matrix ($\mathbf{L}^{\text{blk}} + \mathbf{U}^{\text{blk}}$) (BLAS in //)
- Jacobi: Solution of a block diagonal system (Blocks solved in //)
- G.-S.: Solution of a block inferior triangular system (Descent in //)

Iterative methods for linear systems

Stationary methods

System arising from a finite difference discretization

Definition of the problem

The field $u(x, y)$ is governed by the Poisson equation on a square domain:

$$\begin{cases} \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = f(x, y), & \text{for } (x, y) \in \Omega =]a, b[\times]a, b[, \\ u = 0, & \text{for } (x, y) \in \partial\Omega. \end{cases}$$

Discretization and numerical scheme

The problem is discretized on a regular grid:

- Discretization points: $(x_i, y_j) = (a + ih, a + jh)$ ($i, j = 0, \dots, n + 1$)
- Spatial step: $h = (b - a)/(n + 1)$
- Approximate field: $u_{i,j} \approx u(x_i, y_j)$

We consider a standard finite difference scheme with a **5-point stencil**:

$$\boxed{\frac{1}{h^2} (u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{i,j}) = f_{i,j}} \quad (i, j = 1, \dots, n)$$

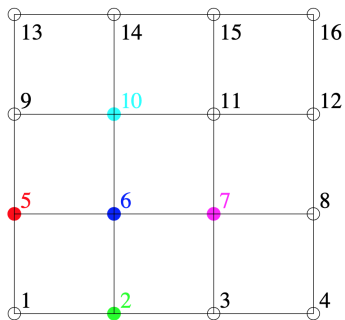
with $u_{i,j} = 0$ (i and/or $j \in \{0, n + 1\}$) and $f_{i,j} = f(x_i, y_j)$.

Asymptotic accuracy: $\mathcal{O}(h^2)$

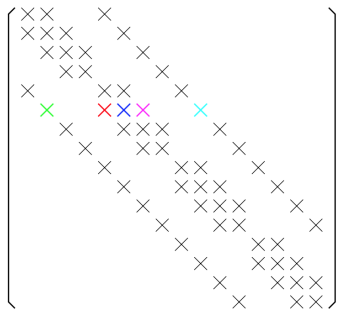
Finite difference scheme — Description [2/3]

$$\begin{cases} \frac{1}{h^2} (u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{i,j}) = f_{i,j} & (i, j = 1, \dots, n) \\ u_{i,j} = 0 & (i \text{ and/or } j \in \{0, n+1\}) \end{cases}$$

Matrix representation of the problem: $\mathbf{Au} = \mathbf{f}$ $\mathbf{A} \in \mathbb{R}^{n^2 \times n^2}$ $\mathbf{u}, \mathbf{f} \in \mathbb{R}^{n^2}$



Finite difference grid



Matrix of the system

$$\mathbf{M}\mathbf{x}^{(\ell+1)} = \mathbf{N}\mathbf{x}^{(\ell)} + \mathbf{b}$$

$$\text{with } \mathbf{A} = (\mathbf{M} - \mathbf{N}) = -\frac{4}{h^2}\mathbf{I} - \frac{1}{h^2}$$

$$\begin{bmatrix} 0 & -1 & & -1 & & & & & & \\ & -1 & 0 & -1 & & & & & & \\ & & \ddots & \ddots & \ddots & & & & & \\ & & & \ddots & \ddots & \ddots & & & & -1 \\ -1 & & & & \ddots & \ddots & \ddots & & & \\ & \ddots & & & & & & & & \\ & & & & & & -1 & 0 & -1 & \\ & & & & & & & & & \\ & & & & & & & & & -1 \\ & & & & & & & & & -1 & 0 \end{bmatrix}$$

Sequential algorithm with Jacobi

$u_{i,j}^{(0)} \in \mathbb{R}$ for $i, j = 1 \dots n$

for $\ell = 0, 1, \dots$ **do**

for $i = 1, \dots, n$ **do**

for $j = 1, \dots, n$ **do**

$$-\frac{4}{h^2}u_{i,j}^{(\ell+1)} = -\frac{1}{h^2} \left(u_{i+1,j}^{(\ell)} + u_{i-1,j}^{(\ell)} + u_{i,j+1}^{(\ell)} + u_{i,j-1}^{(\ell)} \right) + f_{i,j}$$

end

end

end

$$\mathbf{M}\mathbf{x}^{(\ell+1)} = \mathbf{N}\mathbf{x}^{(\ell)} + \mathbf{b}$$

$$\text{avec } \mathbf{A} = (\mathbf{M} - \mathbf{N}) = -\frac{4}{h^2}\mathbf{I} - \frac{1}{h^2}$$

$$\begin{bmatrix} 0 & -1 & & -1 & & & & & & \\ & -1 & 0 & -1 & & & & & & \\ & & \ddots & \ddots & \ddots & & & & & \\ & & & \ddots & \ddots & \ddots & & & & -1 \\ -1 & & & & \ddots & \ddots & \ddots & & & \\ & \ddots & & & & & & & & \\ & & & & & & -1 & 0 & -1 & \\ & & & & & & & & & \\ & & & & -1 & & & & & 0 \end{bmatrix}$$

Sequential algorithm with Jacobi (*rewriting*)

$u_{i,j}^{(0)} \in \mathbb{R}$ for $i, j = 1 \dots n$

for $\ell = 0, 1, \dots$ **do**

for $i = 1, \dots, n$ **do**

for $j = 1, \dots, n$ **do**

$$u_{i,j}^{(\ell+1)} = \frac{1}{4} \left(u_{i+1,j}^{(\ell)} + u_{i-1,j}^{(\ell)} + u_{i,j+1}^{(\ell)} + u_{i,j-1}^{(\ell)} \right) - \frac{h^2}{4} f_{i,j}$$

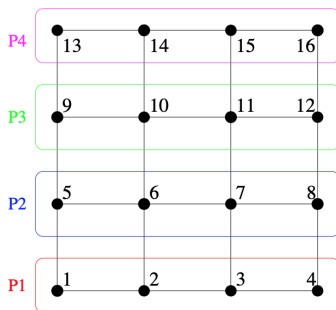
end

end

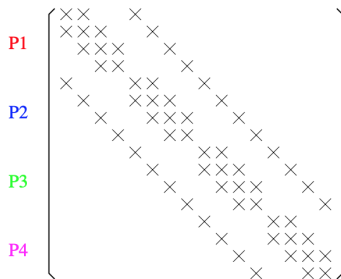
end

Parallelization

- ▶ Domain partitioning and matrix partitioning:



Finite difference grid



Matrix of the system

- ▶ Analysis of communications:
 - Each process has to communicate with both neighbors.
 - Sending/reception of n updated unknowns with each neighbor.
 - Only (local) point-to-point communications.

Parallel algorithm with Jacobi (1D partition)

On each process p : $u_{i,j}^{(0)} \in \mathbb{R}$ for $i = i_{\text{start},p}, \dots, i_{\text{end},p}$ and $j = 1, \dots, n$ **for $\ell = 0, 1, \dots$ do****Communication phase:**

- If $p > 0$: send $u_{i_{\text{start},p},\star}$ to process $p - 1$
- If $p > 0$: receive $u_{i_{\text{start},p-1},\star}$ from process $p - 1$
- If $p < (P - 1)$: send $u_{i_{\text{end},p},\star}$ to process $p + 1$
- If $p < (P - 1)$: receive $u_{i_{\text{end},p+1},\star}$ from process $p + 1$

\\ Update of unknowns

for $i = i_{\text{start},p}, \dots, i_{\text{end},p}$ do**for $j = 1, \dots, n$ do**

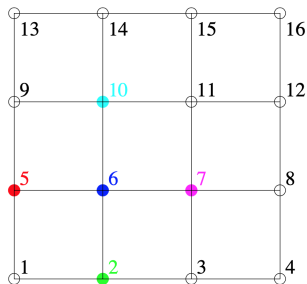
$$u_{i,j}^{(\ell+1)} = \frac{1}{4} \left(u_{i+1,j}^{(\ell)} + u_{i-1,j}^{(\ell)} + u_{i,j+1}^{(\ell)} + u_{i,j-1}^{(\ell)} \right) - \frac{h^2}{4} f_{i,j}$$

end**end****end***(In the communications, \star indicates that the whole line is sent.)*

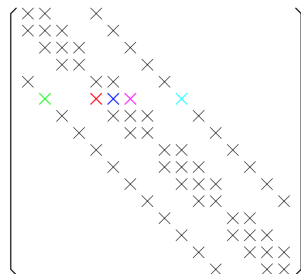
Parallelization of the scheme

The Gauss-Seidel method uses the last available values for the update.

⇒ This procedure is (*a priori*) sequential



Finite difference grid

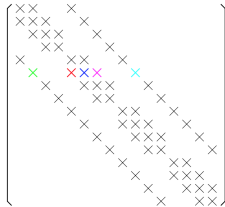
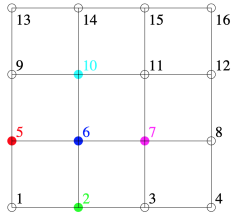


Matrix of the system

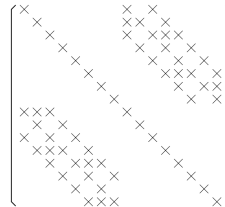
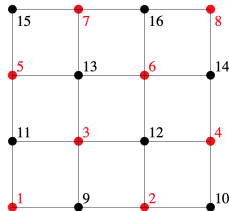
Idea: change the order of evaluation of the unknowns (*i.e. permutation of lines*) to make this procedure parallelizable

Parallelization of the scheme (*with coloring*)

Natural numbering



Numbering with
“red-black” coloring



The unknowns associated to a given color can be updated in parallel.

Sequential algorithm with Gauss-Seidel (with red-black coloring)

$u_{i,j}^{(0)} \in \mathbb{R}$ for $i, j = 1 \dots n$

for $\ell = 0, 1, \dots$ **do**

 \\ Update of red unknowns

for $i = 1, \dots, n$ **do**

for $j = 1, \dots, n$ **do**

If (i, j) **red:** $u_{i,j}^{(\ell+1)} = \frac{1}{4} \left(u_{i+1,j}^{(\ell)} + u_{i-1,j}^{(\ell)} + u_{i,j+1}^{(\ell)} + u_{i,j-1}^{(\ell)} \right) - \frac{h^2}{4} f_{i,j}$

end

end

 \\ Update of black unknowns

for $i = 1, \dots, n$ **do**

for $j = 1, \dots, n$ **do**

If (i, j) **black:**

$u_{i,j}^{(\ell+1)} = \frac{1}{4} \left(u_{i+1,j}^{(\ell+1)} + u_{i-1,j}^{(\ell+1)} + u_{i,j+1}^{(\ell+1)} + u_{i,j-1}^{(\ell+1)} \right) - \frac{h^2}{4} f_{i,j}$

end

end

end

Parallel algorithm with Gauss-Seidel (with red-black coloring)

For each process p :

$u_{i,j}^{(0)} \in \mathbb{R}$ for $i = i_{\text{start},p}, \dots, i_{\text{end},p}$ and $j = 1, \dots, n$

for $\ell = 0, 1, \dots$ do

Communication phase (as for Jacobi)

\\ Update of red unknowns

for $i = i_{\text{start},p}, \dots, i_{\text{end},p}$ do

for $j = 1, \dots, n$ do

If (i, j) red: $u_{i,j}^{(\ell+1)} = \frac{1}{4} \left(u_{i+1,j}^{(\ell)} + u_{i-1,j}^{(\ell)} + u_{i,j+1}^{(\ell)} + u_{i,j-1}^{(\ell)} \right) - \frac{h^2}{4} f_{i,j}$

end

end

Communication phase (as for Jacobi)

\\ Update of black unknowns

for $i = i_{\text{start},p}, \dots, i_{\text{end},p}$ do

for $j = 1, \dots, n$ do

If (i, j) black: $u_{i,j}^{(\ell+1)} = \frac{1}{4} \left(u_{i+1,j}^{(\ell+1)} + u_{i-1,j}^{(\ell+1)} + u_{i,j+1}^{(\ell+1)} + u_{i,j-1}^{(\ell+1)} \right) - \frac{h^2}{4} f_{i,j}$

end

end

end

Comments on parallelization strategies with coloring

- ▶ Basic idea:
 - Each color = Unknowns updated in parallel
 - Communication phase between each color
- ▶ Different numbering, so . . .
 - Different algorithm, but still Gauss-Seidel
 - Different numerical solution, but scheme with the same properties
- ▶ Some extensions:
 - If larger stencil → Coloring with more colors
 - If unstructured mesh → Algorithms for automatic coloring

Ressources

- ▶ *Méthodes Numériques : Algorithmes, analyse et applications*
A. Quarteroni, R. Sacco, F. Saleri (2007), Springer
- ▶ *Calcul scientifique parallèle*
F. Magoulès et F.-X. Roux (2017), Dunod
- ▶ *Calcul scientifique parallèle*
P. Ciarlet and E. Jamelot, polycopié de cours