

Ch22

Linear Complexity “Sweeping” Scheme

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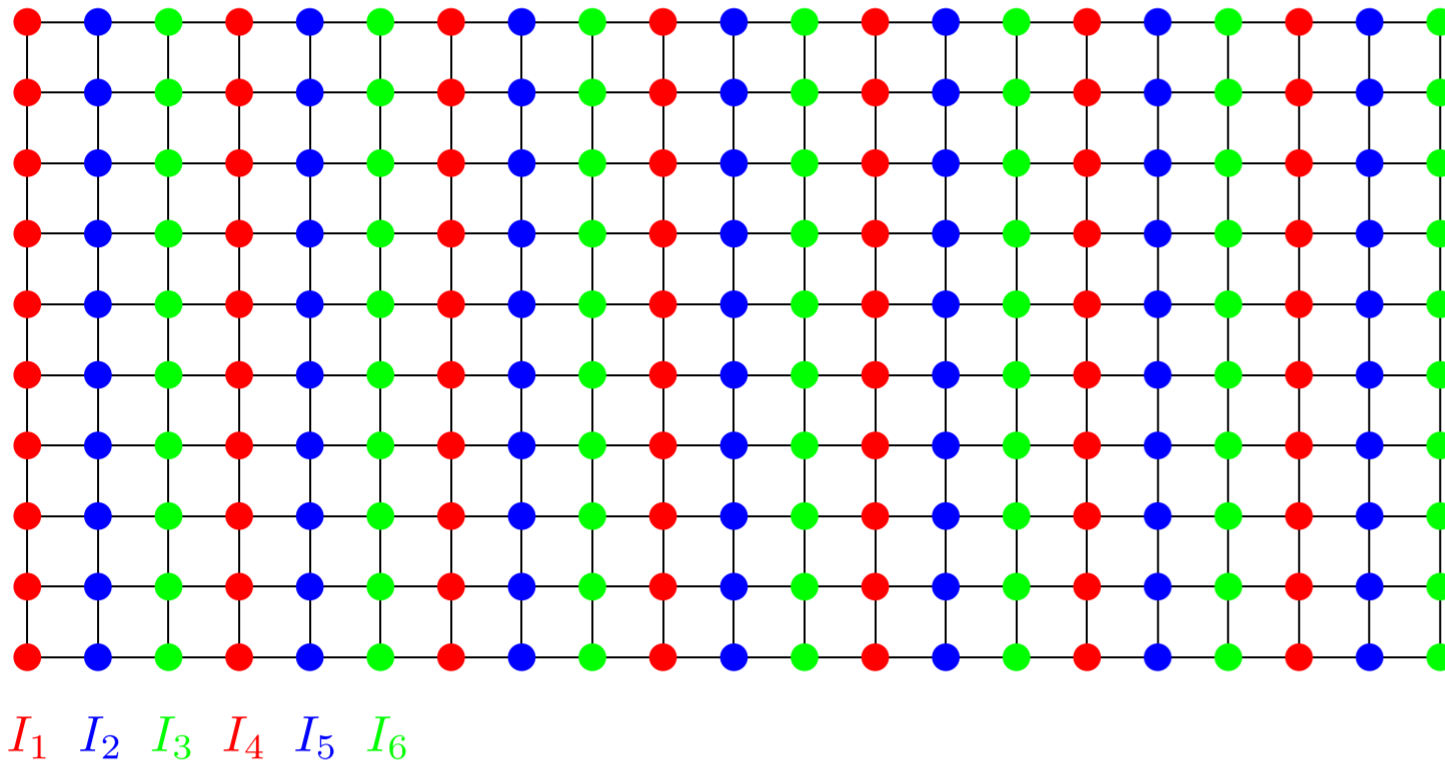
WARNING!

The algorithm is....

- Neither the most efficient one
- Nor the most general one that exists!
- But simple both to describe and to implement when using uniform grid!

Ch22.1 A basic sweeping scheme (uniform rectangular grids)

Partition of nodes



$$I = I_1 \cup I_2 \cup \cdots \cup I_{n_1},$$

The grid:
 n_1 columns, n_2 rows

Each I_k holds the
points in the k th
column of the grid

Linear system

$$\begin{bmatrix} \mathbf{A}_{1,1} & \mathbf{A}_{1,2} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{A}_{2,1} & \mathbf{A}_{2,2} & \mathbf{A}_{2,3} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{A}_{3,2} & \mathbf{A}_{3,3} & \mathbf{A}_{3,4} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{A}_{4,3} & \mathbf{A}_{4,4} & \mathbf{A}_{4,5} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \\ \mathbf{u}_3 \\ \mathbf{u}_4 \\ \vdots \end{bmatrix} = \begin{bmatrix} \mathbf{f}_1 \\ \mathbf{f}_2 \\ \mathbf{f}_3 \\ \mathbf{f}_4 \\ \vdots \end{bmatrix} .$$

Notice that

- Points in I_k only interact with I_{k-1} and I_{k+1} !
- Each $A_{k,j}$ is also banded!

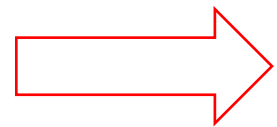
Linear system

$$\begin{bmatrix} \mathbf{A}_{1,1} & \mathbf{A}_{1,2} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{A}_{2,1} & \mathbf{A}_{2,2} & \mathbf{A}_{2,3} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{A}_{3,2} & \mathbf{A}_{3,3} & \mathbf{A}_{3,4} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{A}_{4,3} & \mathbf{A}_{4,4} & \mathbf{A}_{4,5} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \\ \mathbf{u}_3 \\ \mathbf{u}_4 \\ \vdots \end{bmatrix} = \begin{bmatrix} \mathbf{f}_1 \\ \mathbf{f}_2 \\ \mathbf{f}_3 \\ \mathbf{f}_4 \\ \vdots \end{bmatrix}.$$

Schur complement (i)

$$\mathbf{A}_{1,1}\mathbf{u}_1 + \mathbf{A}_{1,2}\mathbf{u}_2 = \mathbf{f}_1,$$

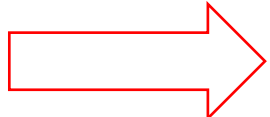
$$\mathbf{A}_{2,1}\mathbf{u}_1 + \mathbf{A}_{2,2}\mathbf{u}_2 + \mathbf{A}_{2,3}\mathbf{u}_3 = \mathbf{f}_2.$$

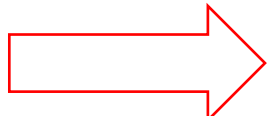

$$\left(\mathbf{A}_{2,2} - \mathbf{A}_{2,1}\mathbf{A}_{1,1}^{-1}\mathbf{A}_{1,2}\right)\mathbf{u}_2 + \mathbf{A}_{2,3}\mathbf{u}_3 = \mathbf{f}_2 - \mathbf{A}_{2,1}\mathbf{A}_{1,1}^{-1}\mathbf{f}_1.$$

Schur complement (ii)

$$\mathbf{A}_{1,1}\mathbf{u}_1 + \mathbf{A}_{1,2}\mathbf{u}_2 = \mathbf{f}_1,$$

$$\mathbf{A}_{2,1}\mathbf{u}_1 + \mathbf{A}_{2,2}\mathbf{u}_2 + \mathbf{A}_{2,3}\mathbf{u}_3 = \mathbf{f}_2.$$


$$\left(\mathbf{A}_{2,2} - \mathbf{A}_{2,1}\mathbf{A}_{1,1}^{-1}\mathbf{A}_{1,2}\right)\mathbf{u}_2 + \mathbf{A}_{2,3}\mathbf{u}_3 = \mathbf{f}_2 - \mathbf{A}_{2,1}\mathbf{A}_{1,1}^{-1}\mathbf{f}_1.$$


$$\mathbf{S}_2\mathbf{u}_2 + \mathbf{A}_{2,3}\mathbf{u}_3 = \tilde{\mathbf{f}}_2.$$

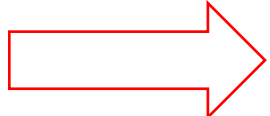
With $\mathbf{S}_2 = \mathbf{A}_{2,2} - \mathbf{A}_{2,1}\mathbf{A}_{1,1}^{-1}\mathbf{A}_{1,2}$, $\tilde{\mathbf{f}}_2 = \mathbf{f}_2 - \mathbf{A}_{2,1}\mathbf{A}_{1,1}^{-1}\mathbf{f}_1$,

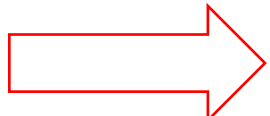
Schur complement (iii)

$$\mathbf{S}_2 \mathbf{u}_2 + \mathbf{A}_{2,3} \mathbf{u}_3 = \tilde{\mathbf{f}}_2.$$

Keep going...

$$\mathbf{A}_{3,2} \mathbf{u}_2 + \mathbf{A}_{3,3} \mathbf{u}_3 + \mathbf{A}_{3,4} \mathbf{u}_4 = \mathbf{f}_3.$$


$$(\mathbf{A}_{3,3} - \mathbf{A}_{3,2} \mathbf{S}_2^{-1} \mathbf{A}_{2,3}) \mathbf{u}_3 + \mathbf{A}_{3,4} \mathbf{u}_4 = \mathbf{f}_3 - \mathbf{A}_{3,2} \mathbf{S}_2^{-1} \tilde{\mathbf{f}}_2.$$

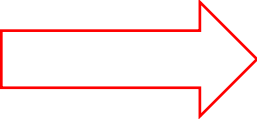

$$\mathbf{S}_3 \mathbf{u}_3 + \mathbf{A}_{3,4} \mathbf{u}_4 = \tilde{\mathbf{f}}_3.$$

With $\mathbf{S}_3 = \mathbf{A}_{3,3} - \mathbf{A}_{3,2} \mathbf{S}_2^{-1} \mathbf{A}_{2,3}$, $\tilde{\mathbf{f}}_3 = \mathbf{f}_3 - \mathbf{A}_{3,2} \mathbf{S}_2^{-1} \tilde{\mathbf{f}}_2$,

Schur complement (iv)

$$\mathbf{S}_{n_1} \mathbf{u}_{n_1} = \tilde{\mathbf{f}}_{n_1},$$

End of 'rightward' sweeping


$$\mathbf{u}_{n_1} = \mathbf{S}_{n_1}^{-1} \tilde{\mathbf{f}}_{n_1}.$$

Start of 'leftward' sweeping

Schur complement (v)

Sequentially, we have

$$\mathbf{S}_{n_1} \mathbf{u}_{n_1} = \tilde{\mathbf{f}}_{n_1}, \quad \Rightarrow \quad \mathbf{u}_{n_1} = \mathbf{S}_{n_1}^{-1} \tilde{\mathbf{f}}_{n_1}.$$

$$\mathbf{S}_{n_1-1} \mathbf{u}_{n_1-1} + \mathbf{A}_{n_1-1, n_1} \mathbf{u}_{n_1} = \tilde{\mathbf{f}}_{n_1-1}, \quad \Rightarrow \quad \mathbf{u}_{n_1-1} = \mathbf{S}_{n_1-1}^{-1} (\tilde{\mathbf{f}}_{n_1-1} - \mathbf{A}_{n_1-1, n_1} \mathbf{u}_{n_1}).$$

⋮

$$\mathbf{A}_{1,1} \mathbf{u}_1 + \mathbf{A}_{1,2} \mathbf{u}_2 = \mathbf{f}_1, \quad \Rightarrow \quad \mathbf{u}_1 = \mathbf{A}_{1,1}^{-1} (\mathbf{f}_1 - \mathbf{A}_{1,2} \mathbf{u}_2).$$

Schur complement (vi)

The iteration of Schur Complements
(Actually we care about its inverse!)

Define $\mathbf{X}_k = \mathbf{S}_k^{-1}$

$$\mathbf{X}_k = \begin{cases} \mathbf{A}_{1,1}^{-1}, & \text{when } k = 1, \\ \left(\mathbf{A}_{k,k} - \mathbf{A}_{k,k-1} \mathbf{X}_{k-1} \mathbf{A}_{k-1,k} \right)^{-1}, & \text{when } k > 1. \end{cases}$$

INVERTING AND SOLVING A BLOCK-TRIDIAGONAL LINEAR SYSTEM

Build all solution operators in a rightward sweep:

$$\mathbf{X}_1 = \mathbf{A}(I_1, I_1)^{-1}$$

for $i = 2 : m$

$$\mathbf{X}_i = (\mathbf{A}(I_i, I_i) - \mathbf{A}_{i,i-1} \mathbf{X}_{i-1} \mathbf{A}_{i-1,i})^{-1}$$

end for

Given a load \mathbf{f} , compute equivalent loads in a rightward sweep:

$$\tilde{\mathbf{f}}(I_1) = \mathbf{f}(I_1)$$

for $i = 2 : m$

$$\tilde{\mathbf{f}}_i = \mathbf{f}(I_i) - \mathbf{A}(I_i, I_{i-1}) \mathbf{X}_{i-1} \tilde{\mathbf{f}}(I_{i-1})$$

end for

Given the equivalent loads, compute solutions in a leftward sweep:

$$\mathbf{u}(I_m) = \mathbf{X}_m \tilde{\mathbf{f}}(I_m)$$

for $i = (m - 1) : (-1) : 1$

$$\mathbf{u}(I_i) = \mathbf{X}_i (\tilde{\mathbf{f}}(I_i) - \mathbf{A}_{i,i+1} \mathbf{u}(I_{i+1}))$$

end for

Algorithm

Complexity:

$$O(n_1 n_2) = O(N)$$

Linear!

Important Message

The algorithm right now...

- Nothing more than traditional Gaussian elimination with column-by-column ordering
- **Can be accelerated since these Schur complements are highly compressible!**

Remarks

- An alternative solver is based on **LU factorization**

$$\mathbf{A} = \mathbf{LU} = \begin{bmatrix} \mathbf{L}_{1,1} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{L}_{2,1} & \mathbf{L}_{2,2} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{L}_{3,2} & \mathbf{L}_{3,3} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{L}_{4,3} & \mathbf{L}_{4,4} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix} \begin{bmatrix} \mathbf{U}_{1,1} & \mathbf{U}_{1,2} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{U}_{2,2} & \mathbf{U}_{2,3} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{U}_{3,3} & \mathbf{U}_{3,4} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{U}_{4,4} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix} .$$

Remarks

It leads to the recursive formula:

$$[\mathbf{L}_{1,1}, \mathbf{U}_{1,1}] = \text{lu}(\mathbf{A}_{1,1}).$$

$$\mathbf{L}_{i,i-1} = \mathbf{A}_{i,i-1} \mathbf{U}_{i-1,i-1}^{-1},$$

$$\mathbf{U}_{i-1,i} = \mathbf{L}_{i-1,i-1}^{-1} \mathbf{A}_{i-1,i},$$

$$[\mathbf{L}_{i,i}, \mathbf{U}_{i,i}] = \text{lu}(\mathbf{A}_{i,i} - \mathbf{L}_{i,i-1} \mathbf{U}_{i-1,i}).$$

PS: We can always use a Cholesky factorization when A is SPD!

LU FACTORIZATION AND SOLVE OF A BLOCK-TRIDIAGONAL SYSTEM

Build the LU factorization through a rightward sweep:

$$[\mathbf{L}_{1,1}, \mathbf{U}_{1,1}] = \text{lu}(\mathbf{A}_{1,1})$$

for $i = 2 : m$

$$\mathbf{S}_i = \mathbf{A}_{i,i} - \mathbf{A}_{i,i-1} \mathbf{U}_{i-1,i-1}^{-1} \mathbf{L}_{i-1,i-1}^{-1} \mathbf{A}_{i-1,i}$$

$$[\mathbf{L}_{i,i}, \mathbf{U}_{i,i}] = \text{lu}(\mathbf{S}_i)$$

end for

Given a load \mathbf{f} , solve $\mathbf{L}\mathbf{y} = \mathbf{f}$ in a rightward sweep:

$$\mathbf{y}(I_1) = \mathbf{L}_{1,1}^{-1} \mathbf{f}(I_1)$$

for $i = 2 : m$

$$\mathbf{y}(I_i) = \mathbf{L}_{i,i}^{-1} (\mathbf{f}(I_i) - \mathbf{A}_{i,i-1} \mathbf{U}_{i-1,i-1}^{-1} \mathbf{u}(I_{i-1}))$$

end for

Given \mathbf{y} , solve $\mathbf{U}\mathbf{u} = \mathbf{y}$ in a leftward sweep:

$$\mathbf{u}(I_m) = \mathbf{U}_{m,m}^{-1} \mathbf{y}(I_m)$$

for $i = (m - 1) : (-1) : 1$

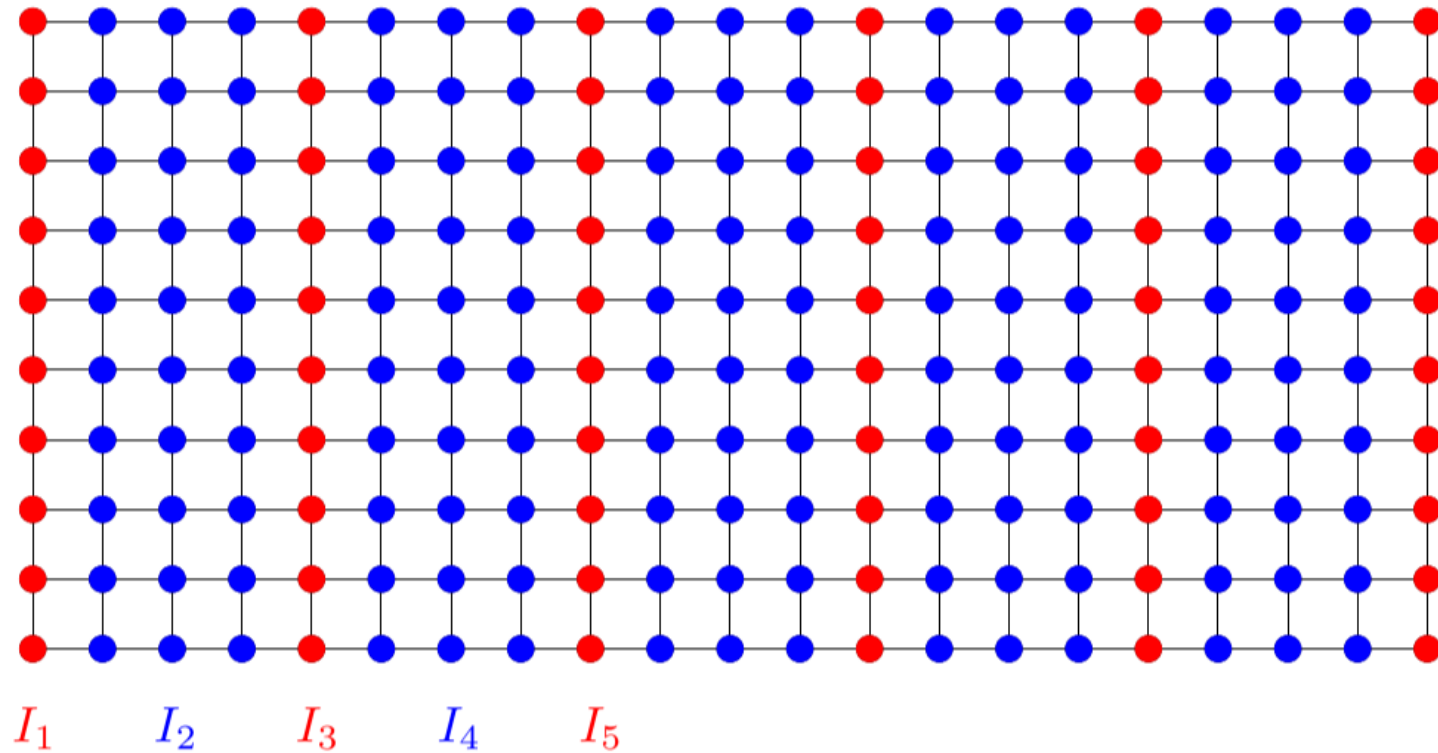
$$\mathbf{u}(I_i) = \mathbf{U}_{i,i}^{-1} (\mathbf{y}(I_i) - \mathbf{L}_{i,i}^{-1} \mathbf{A}_{i,i+1} \mathbf{u}(I_{i+1}))$$

end for

Algorithm
(LU)

Ch22.2 Buffered sweeping schemes

Partition of nodes (Buffered)



$$I = I_1 \cup I_2 \cup I_3 \cup I_4 \cup I_5 \cup \dots \cup I_{2m+1}.$$

The grid:
 n_1 columns, n_2 rows

Buffers: I_2, I_4, I_6, \dots
Each holds b
columns of the grid

$$n_1 = 1 + m(b + 1)$$

Linear system

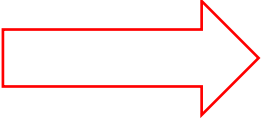
$$\begin{bmatrix} \mathbf{A}_{1,1} & \mathbf{A}_{1,2} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{A}_{2,1} & \mathbf{A}_{2,2} & \mathbf{A}_{2,3} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{A}_{3,2} & \mathbf{A}_{3,3} & \mathbf{A}_{3,4} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{A}_{4,3} & \mathbf{A}_{4,4} & \mathbf{A}_{4,5} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{A}_{5,4} & \mathbf{A}_{5,5} & \mathbf{A}_{5,6} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{A}_{6,5} & \mathbf{A}_{6,6} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \\ \mathbf{u}_3 \\ \mathbf{u}_4 \\ \mathbf{u}_5 \\ \mathbf{u}_6 \\ \vdots \end{bmatrix} = \begin{bmatrix} \mathbf{f}_1 \\ \mathbf{f}_2 \\ \mathbf{f}_3 \\ \mathbf{f}_4 \\ \mathbf{f}_5 \\ \mathbf{f}_6 \\ \vdots \end{bmatrix} .$$

Linear system


$$\begin{bmatrix} \mathbf{A}_{1,1} & \mathbf{A}_{1,2} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{A}_{2,1} & \mathbf{A}_{2,2} & \mathbf{A}_{2,3} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{A}_{3,2} & \mathbf{A}_{3,3} & \mathbf{A}_{3,4} & \mathbf{0} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{A}_{4,3} & \mathbf{A}_{4,4} & \mathbf{A}_{4,5} & \mathbf{0} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{A}_{5,4} & \mathbf{A}_{5,5} & \mathbf{A}_{5,6} & \cdots \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{A}_{6,5} & \mathbf{A}_{6,6} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \\ \mathbf{u}_3 \\ \mathbf{u}_4 \\ \mathbf{u}_5 \\ \mathbf{u}_6 \\ \vdots \end{bmatrix} = \begin{bmatrix} \mathbf{f}_1 \\ \mathbf{f}_2 \\ \mathbf{f}_3 \\ \mathbf{f}_4 \\ \mathbf{f}_5 \\ \mathbf{f}_6 \\ \vdots \end{bmatrix} .$$

Schur complement (i)

$$\mathbf{A}_{2,1}\mathbf{u}_1 + \mathbf{A}_{2,2}\mathbf{u}_2 + \mathbf{A}_{2,3}\mathbf{u}_3 = \mathbf{f}_2.$$

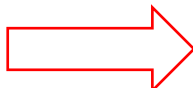

$$\mathbf{u}_2 = \mathbf{A}_{2,2}^{-1}(\mathbf{f}_2 - \mathbf{A}_{2,1}\mathbf{u}_1 - \mathbf{A}_{2,3}\mathbf{u}_3).$$

Schur complement (ii)



$$\begin{bmatrix}
 (\mathbf{A}_{1,1} - \mathbf{A}_{1,2}\mathbf{A}_{2,2}^{-1}\mathbf{A}_{2,1}) & -\mathbf{A}_{1,2}\mathbf{A}_{2,2}^{-1}\mathbf{A}_{2,3} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \cdots \\
 -\mathbf{A}_{3,2}\mathbf{A}_{2,2}^{-1}\mathbf{A}_{2,1} & (\mathbf{A}_{3,3} - \mathbf{A}_{3,2}\mathbf{A}_{2,2}^{-1}\mathbf{A}_{2,3}) & \mathbf{A}_{3,4} & \mathbf{0} & \mathbf{0} & \cdots \\
 \mathbf{0} & \mathbf{A}_{4,3} & \mathbf{A}_{4,4} & \mathbf{A}_{4,5} & \mathbf{0} & \cdots \\
 \mathbf{0} & \mathbf{0} & \mathbf{A}_{5,4} & \mathbf{A}_{5,5} & \mathbf{A}_{5,6} & \cdots \\
 \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{A}_{6,5} & \mathbf{A}_{6,6} & \cdots \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots
 \end{bmatrix}
 \begin{bmatrix}
 \mathbf{u}_1 \\
 \mathbf{u}_3 \\
 \mathbf{u}_4 \\
 \mathbf{u}_5 \\
 \mathbf{u}_6 \\
 \vdots
 \end{bmatrix}
 =
 \begin{bmatrix}
 \mathbf{f}_1 - \mathbf{A}_{1,2}\mathbf{A}_{2,2}^{-1}\mathbf{f}_2 \\
 \mathbf{f}_3 - \mathbf{A}_{3,2}\mathbf{A}_{2,2}^{-1}\mathbf{f}_2 \\
 \mathbf{f}_4 \\
 \mathbf{f}_5 \\
 \mathbf{f}_6 \\
 \vdots
 \end{bmatrix}
 .$$

Schur complement (ii)



$$\begin{bmatrix}
 (\mathbf{A}_{1,1} - \mathbf{A}_{1,2}\mathbf{A}_{2,2}^{-1}\mathbf{A}_{2,1}) & -\mathbf{A}_{1,2}\mathbf{A}_{2,2}^{-1}\mathbf{A}_{2,3} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \cdots \\
 -\mathbf{A}_{3,2}\mathbf{A}_{2,2}^{-1}\mathbf{A}_{2,1} & (\mathbf{A}_{3,3} - \mathbf{A}_{3,2}\mathbf{A}_{2,2}^{-1}\mathbf{A}_{2,3}) & \mathbf{A}_{3,4} & \mathbf{0} & \mathbf{0} & \cdots \\
 \mathbf{0} & \mathbf{A}_{4,3} & \mathbf{A}_{4,4} & \mathbf{A}_{4,5} & \mathbf{0} & \cdots \\
 \mathbf{0} & \mathbf{0} & \mathbf{A}_{5,4} & \mathbf{A}_{5,5} & \mathbf{A}_{5,6} & \cdots \\
 \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{A}_{6,5} & \mathbf{A}_{6,6} & \cdots \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots
 \end{bmatrix}
 \begin{bmatrix}
 \mathbf{u}_1 \\
 \mathbf{u}_3 \\
 \mathbf{u}_4 \\
 \mathbf{u}_5 \\
 \mathbf{u}_6 \\
 \vdots
 \end{bmatrix}
 =
 \begin{bmatrix}
 \mathbf{f}_1 - \mathbf{A}_{1,2}\mathbf{A}_{2,2}^{-1}\mathbf{f}_2 \\
 \mathbf{f}_3 - \mathbf{A}_{3,2}\mathbf{A}_{2,2}^{-1}\mathbf{f}_2 \\
 \mathbf{f}_4 \\
 \mathbf{f}_5 \\
 \mathbf{f}_6 \\
 \vdots
 \end{bmatrix}
 .$$

Keep eliminating u_4, u_6, \dots
in an analogous manner!

New Linear system (i)

$$\begin{bmatrix} \tilde{\mathbf{A}}_{1,1} & \tilde{\mathbf{A}}_{1,3} & \mathbf{0} & \mathbf{0} & \cdots \\ \tilde{\mathbf{A}}_{3,3} & \tilde{\mathbf{A}}_{3,3} & \tilde{\mathbf{A}}_{3,5} & \mathbf{0} & \cdots \\ \mathbf{0} & \tilde{\mathbf{A}}_{5,3} & \tilde{\mathbf{A}}_{5,5} & \tilde{\mathbf{A}}_{5,7} & \cdots \\ \mathbf{0} & \mathbf{0} & \tilde{\mathbf{A}}_{7,5} & \tilde{\mathbf{A}}_{7,7} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} \mathbf{u}_1 \\ \mathbf{u}_3 \\ \mathbf{u}_5 \\ \mathbf{u}_7 \\ \vdots \end{bmatrix} = \begin{bmatrix} \tilde{\mathbf{f}}_1 \\ \tilde{\mathbf{f}}_3 \\ \tilde{\mathbf{f}}_5 \\ \tilde{\mathbf{f}}_7 \\ \vdots \end{bmatrix},$$

New Linear system (ii)

Diagonal :

$$\begin{aligned}\tilde{\mathbf{A}}_{2k+1,2k+1} = & \mathbf{A}_{2k+1,2k+1} - \mathbf{A}_{2k+1,2k} \mathbf{A}_{2k,2k}^{-1} \mathbf{A}_{2k,2k+1} \\ & - \mathbf{A}_{2k+1,2k+2} \mathbf{A}_{2k+2,2k+2}^{-1} \mathbf{A}_{2k+2,2k+1},\end{aligned}$$

Off-Diagonal :

$$\tilde{\mathbf{A}}_{2k-1,2k+1} = - \mathbf{A}_{2k-1,2k} \mathbf{A}_{2k,2k}^{-1} \mathbf{A}_{2k,2k+1},$$

$$\tilde{\mathbf{A}}_{2k+1,2k-1} = - \mathbf{A}_{2k+1,2k} \mathbf{A}_{2k,2k}^{-1} \mathbf{A}_{2k,2k-1}.$$

And

$$\tilde{\mathbf{A}}_{1,1} = \mathbf{A}_{1,1} - \mathbf{A}_{1,2} \mathbf{A}_{2,2}^{-1} \mathbf{A}_{2,1},$$

$$\tilde{\mathbf{A}}_{2m+1,2m+1} = \mathbf{A}_{2m+1,2m+1} - \mathbf{A}_{2m+1,2m} \mathbf{A}_{2m,2m}^{-1} \mathbf{A}_{2m,2m+1}.$$

New Linear system (iii)

Given the new reduced system

$$\tilde{\mathbf{A}}\tilde{\mathbf{u}} = \tilde{\mathbf{f}}.$$

We use the algorithm in Chapter 22.1 to solve it!

Now it's much smaller and more practicable!

A BUFFERED SWEEPING SCHEME

Initialize the block-tridiagonal matrix $\tilde{\mathbf{A}}$ by copying over the diagonal blocks from \mathbf{A} :

for $k = 1 : (m + 1)$
 $\tilde{\mathbf{A}}_{2k-1,2k-1} = \mathbf{A}_{2k-1,2k-1}$
end for

Eliminate all buffer nodes (the loop can be executed in any order):

for $k = 1 : m$
 $\tilde{\mathbf{A}}_{2k-1,2k-1} = \tilde{\mathbf{A}}_{2k-1,2k-1} - \mathbf{A}_{2k-1,2k} \mathbf{A}_{2k,2k}^{-1} \mathbf{A}_{2k,2k-1}$
 $\tilde{\mathbf{A}}_{2k+1,2k+1} = \tilde{\mathbf{A}}_{2k+1,2k+1} - \mathbf{A}_{2k+1,2k} \mathbf{A}_{2k,2k}^{-1} \mathbf{A}_{2k,2k+1}$
 $\tilde{\mathbf{A}}_{2k-1,2k+1} = \mathbf{A}_{2k-1,2k} \mathbf{A}_{2k,2k}^{-1} \mathbf{A}_{2k,2k+1}$
 $\tilde{\mathbf{A}}_{2k+1,2k-1} = \mathbf{A}_{2k+1,2k} \mathbf{A}_{2k,2k}^{-1} \mathbf{A}_{2k,2k-1}$
end for

**Fully
parallelizable!**

Factorize the block-tridiagonal system in a rightward sweep:

$\mathbf{X}_1 = \tilde{\mathbf{A}}_{1,1}^{-1}$
for $k = 1 : m$
 $\mathbf{X}_{2k+1} = \left(\tilde{\mathbf{A}}_{2k+1,2k+1} - \tilde{\mathbf{A}}_{2k+1,2k-1} \mathbf{X}_{2k-1} \tilde{\mathbf{A}}_{2k-1,2k+1}^{-1} \right)^{-1}$
end for

Algorithm
(i)

Algorithm
(ii)

Given a load \mathbf{f} , compute equivalent loads for the block-tridiagonal system:

for $k = 1 : (m + 1)$

$$\tilde{\mathbf{f}}_{2k-1} = \mathbf{f}(I_{2k-1})$$

end for

for $k = 1 : m$

$$\tilde{\mathbf{f}}_{2k-1} = \tilde{\mathbf{f}}_{2k-1} - \mathbf{A}_{2k-1,2k} \mathbf{A}_{2k,2k}^{-1} \mathbf{f}(I_{2k})$$

$$\tilde{\mathbf{f}}_{2k+1} = \tilde{\mathbf{f}}_{2k+1} - \mathbf{A}_{2k+1,2k} \mathbf{A}_{2k,2k}^{-1} \mathbf{f}(I_{2k})$$

end for

Given the equivalent loads, compute solutions on the interfaces in a rightward sweep:

$$\mathbf{u}(I_{2m+1}) = \mathbf{X}_{2m+1} \tilde{\mathbf{f}}_{2m+1}$$

for $k = m : (-1) : 1$

$$\mathbf{u}(I_{2k-1}) = \mathbf{X}_{2k-1} (\tilde{\mathbf{f}}_{2k-1} - \tilde{\mathbf{A}}_{2k-1,2k+1} \mathbf{u}(I_{2k+1}))$$

end for

Solve for the solutions on all buffer nodes (the loop can be executed in any order):

for $k = 1 : m$

$$\mathbf{u}(I_{2k}) = \mathbf{A}_{2k,2k}^{-1} (\mathbf{f}(I_{2k}) - \mathbf{A}_{2k,2k-1} \mathbf{u}(I_{2k-1}) - \mathbf{A}_{2k,2k+1} \mathbf{u}(I_{2k+1}))$$

end for

Implementation considerations

- Each matrix $\mathbf{A}_{2k,2k}$ for two-dimensional problems will be **banded**. $\mathbf{A}_{2k,2k}^{-1}$ can be applied very rapidly.
- **Parallel implementation**: Computing the block-tridiagonal matrix $\tilde{\mathbf{A}}$ is trivially parallelizable since **each buffer region can be eliminated independently**.

Thank you!

Q&A time