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An Algebraic Sparsified Nested Dissection Algorithm using Low-Rank Approximations

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Sparse Linear Systems

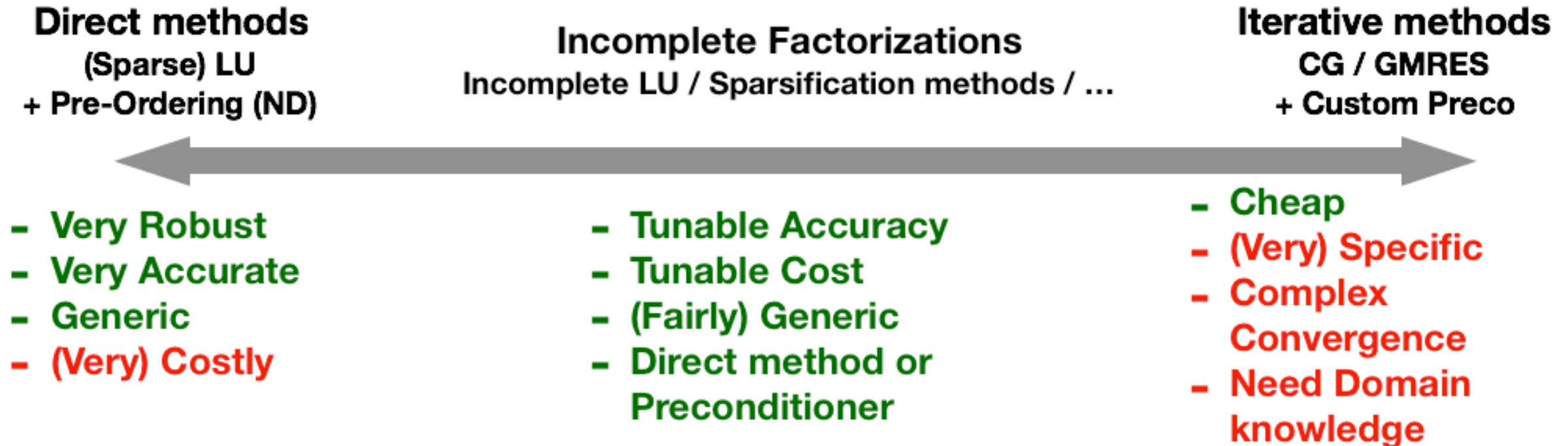
We want to solve

$$Ax = b$$

When A of size $N \times N$ is large, SPD and sparse

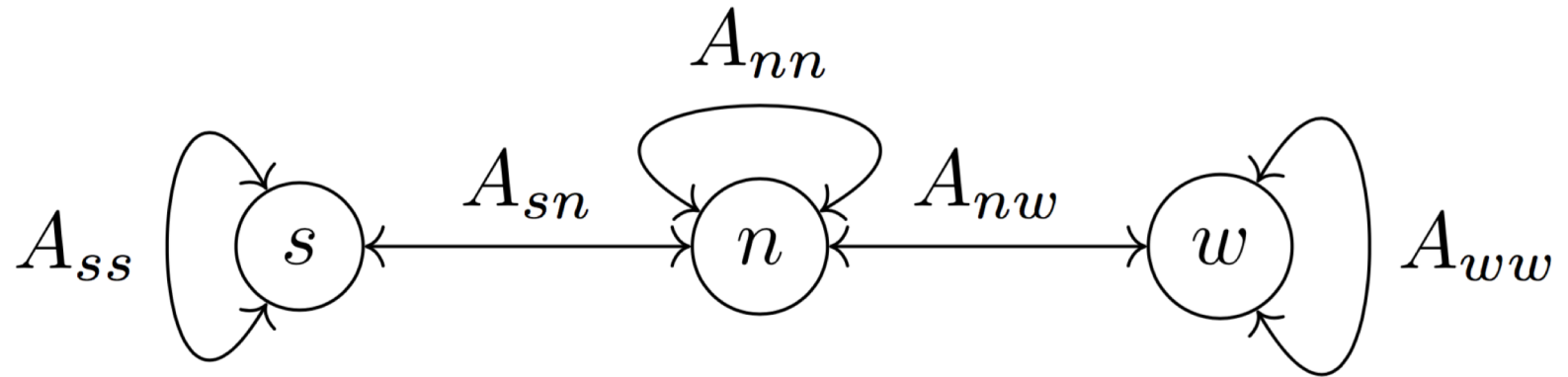
Bottleneck in many physical applications

Sparse Linear Systems

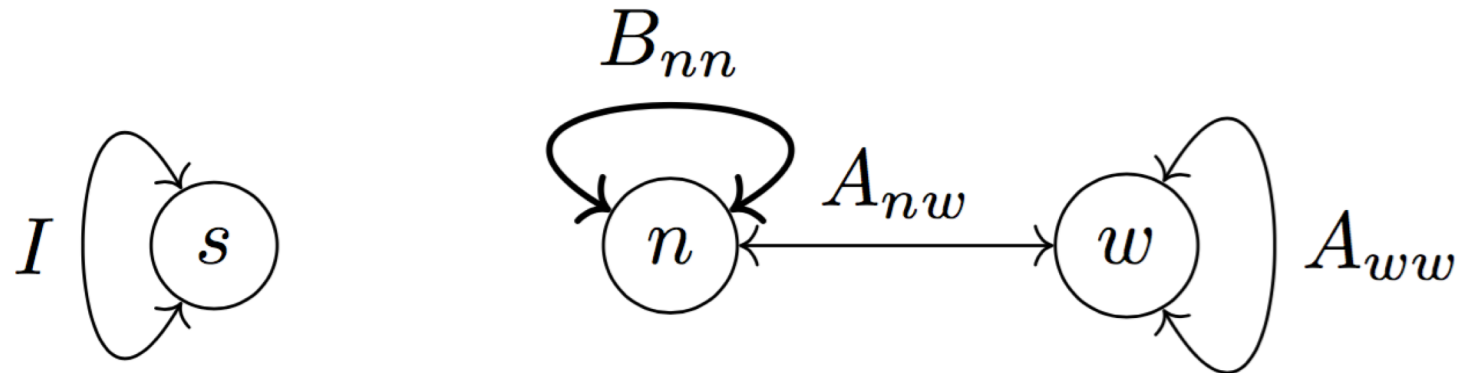


Sparse Linear Systems

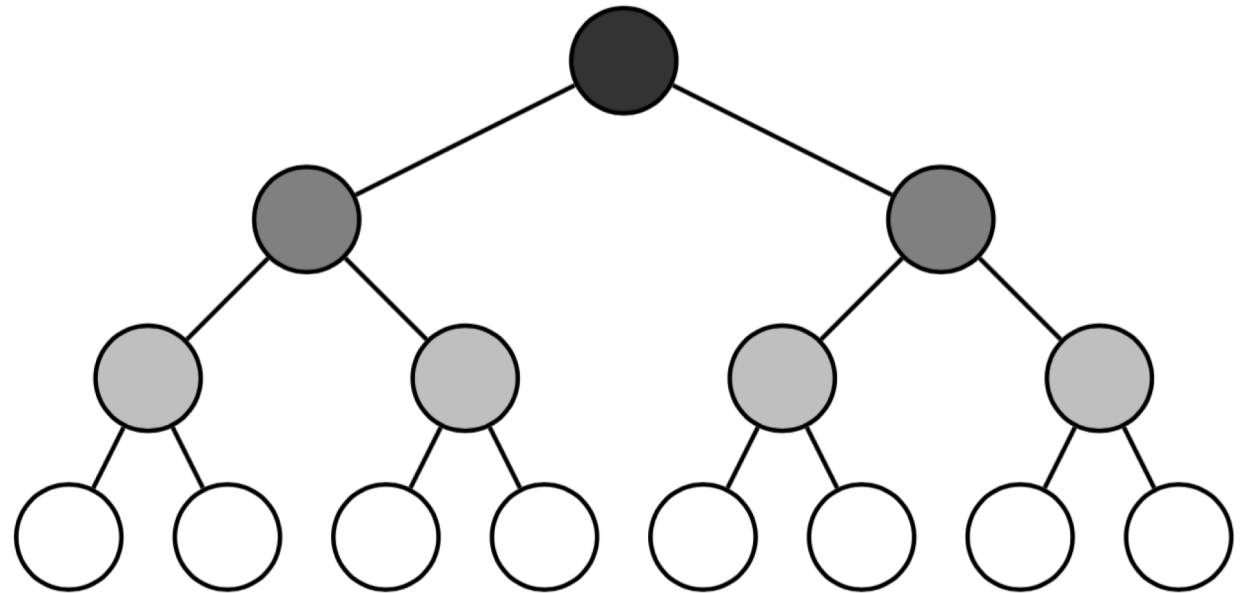
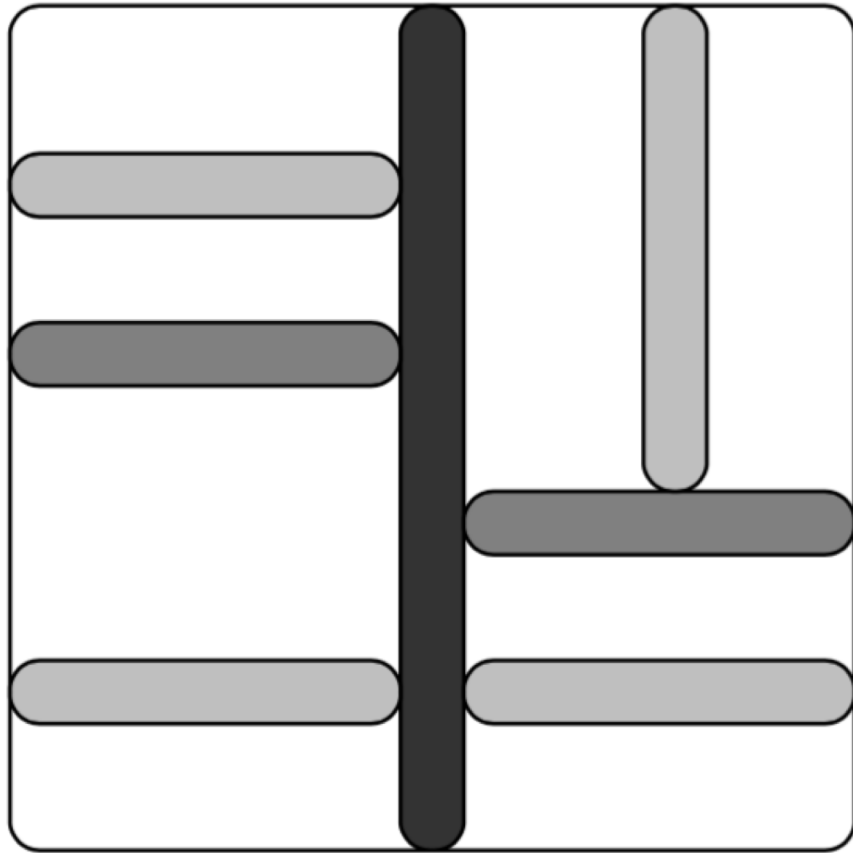
$$\begin{bmatrix} A_{ss} & & A_{sn} \\ & A_{ww} & A_{wn} \\ A_{ns} & A_{nw} & A_{nn} \end{bmatrix}$$



$$\begin{bmatrix} I & & \\ & A_{ww} & A_{wn} \\ & A_{nw} & B_{nn} \end{bmatrix}$$



Nested Dissection

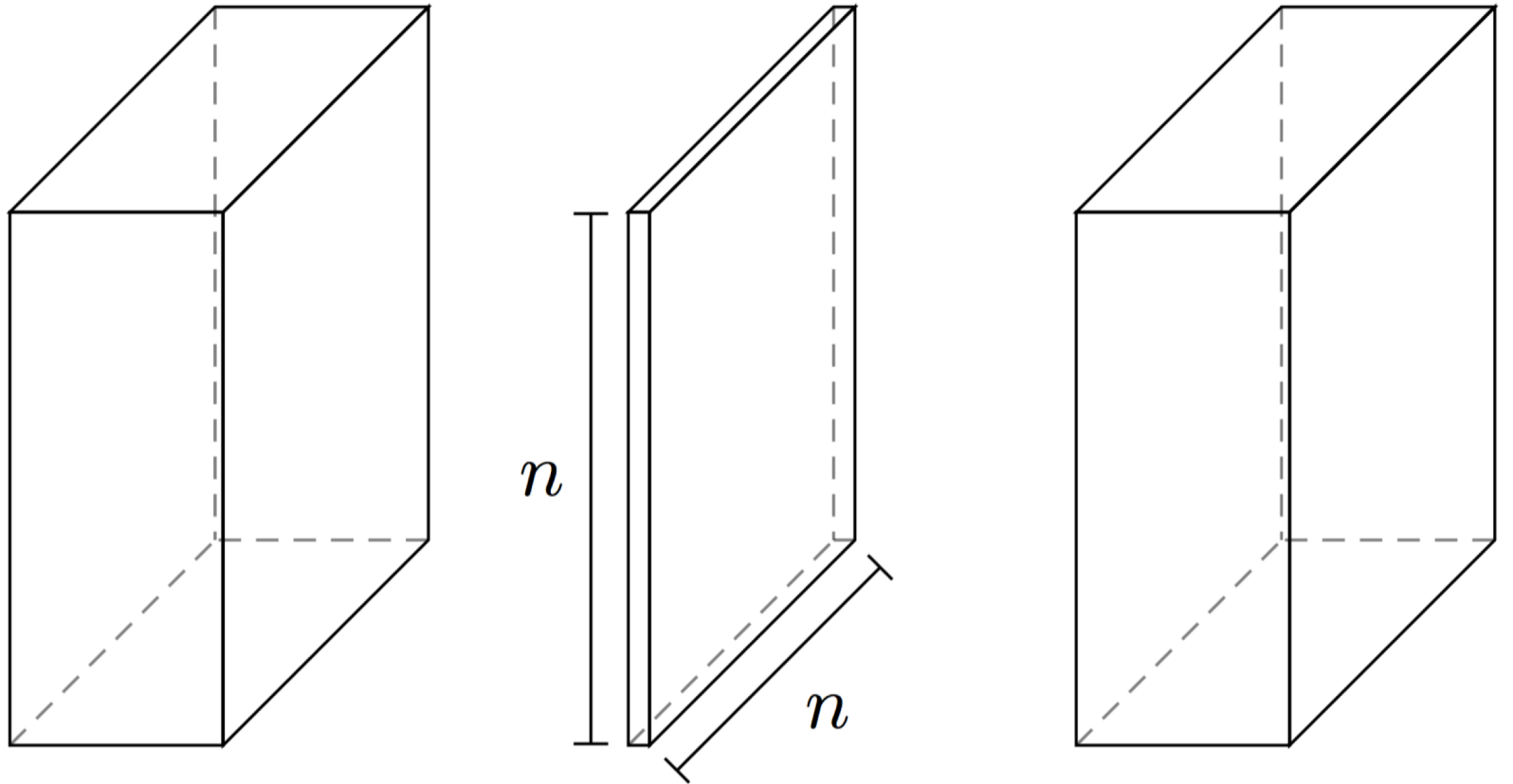


Nested Dissection

Issue: separators are small, but still too big on typical 3D problems

Separator: n^2

Cost: $n^{2 \cdot 3} = N^2$

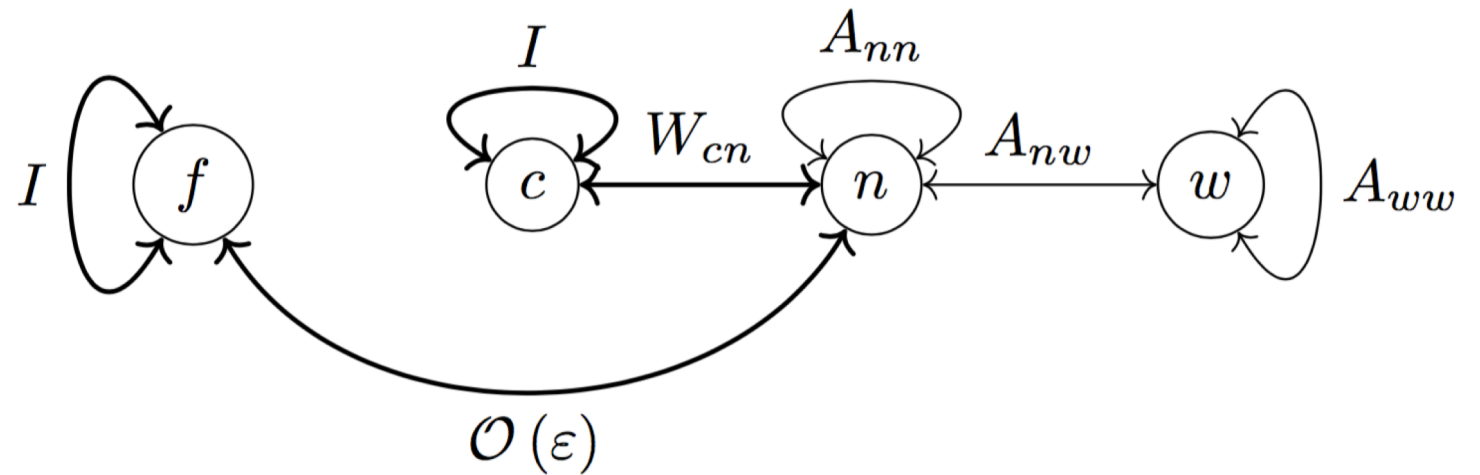


Sparsification

$$\begin{bmatrix} I & & A_{sn} \\ & A_{ww} & A_{wn} \\ A_{ns} & A_{nw} & A_{nn} \end{bmatrix}$$

$$A_{sn} = Q_{sc}W_{cn} + \varepsilon$$

$$\begin{bmatrix} I & & & W_{cn} \\ & I & & \varepsilon \\ & & A_{ww} & A_{wn} \\ W_{cn}^T & \varepsilon & A_{nw} & A_{nn} \end{bmatrix}$$

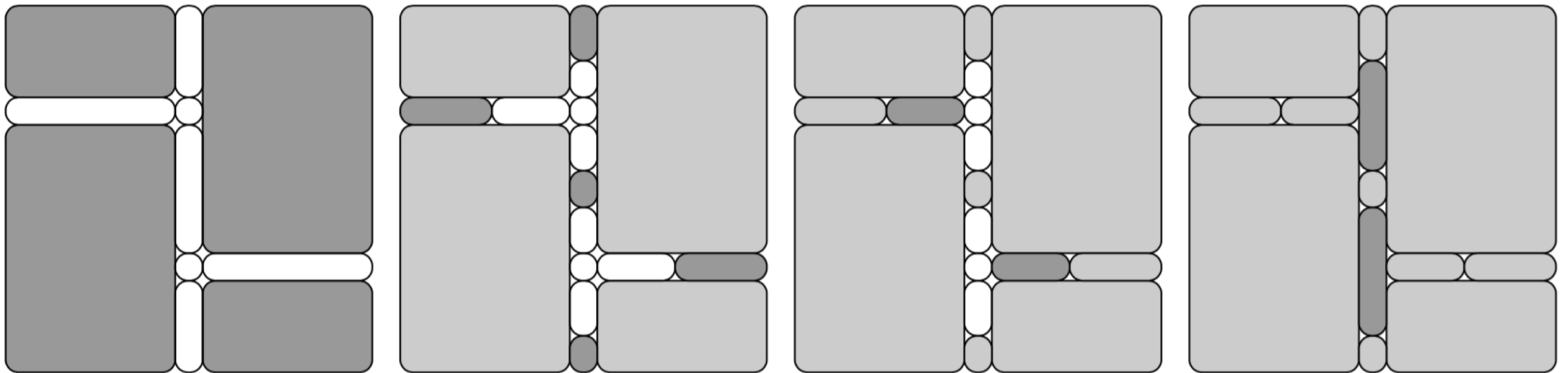


Sparsified Nested Dissection

For level ℓ , from leaves to top

Eliminate interiors at level ℓ

Sparsify (Scale & Compress) interfaces at level ℓ



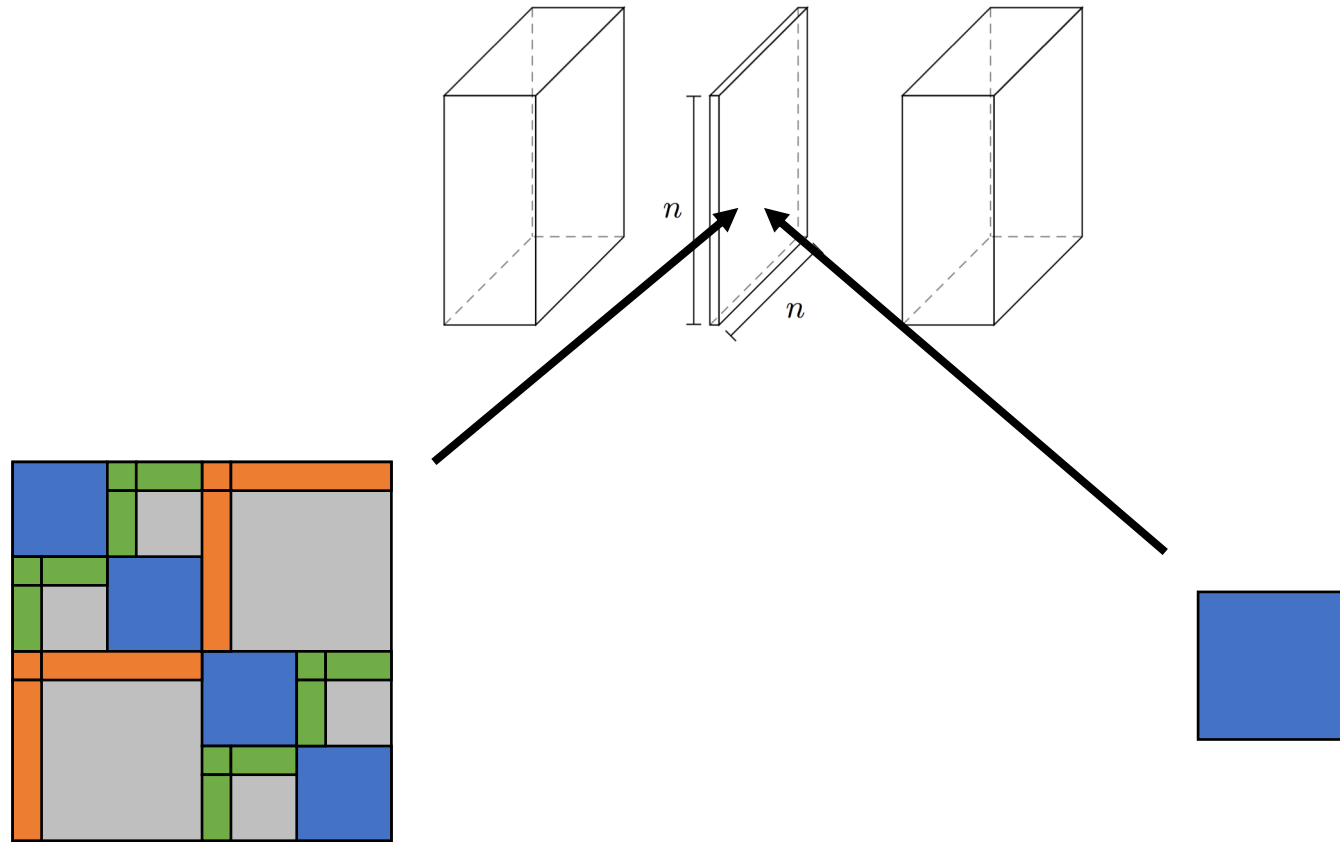
Sparsified Nested Dissection

We effectively build a preconditioner M such that

$$M^T A M \approx I \quad M = \prod_{\ell} \left(\prod_s Q_s^T \prod_s E_s \right)$$

We then use M as a preconditioner for CG or GMRES.

Different from fast-algebra techniques



Low-Rank Compression: three variants

$$\begin{bmatrix} A_{ss} & & A_{sn} \\ & A_{ww} & A_{wn} \\ A_{ns} & A_{nw} & A_{nn} \end{bmatrix}$$

$$A_{ns} = [A_{nc} \quad A_{nc}T_{cf}] + \varepsilon$$

OR

$$\begin{bmatrix} I & & A_{sn} \\ & A_{ww} & A_{wn} \\ A_{ns} & A_{nw} & A_{nn} \end{bmatrix}$$

$$A_{ns} = [A_{nc} \quad A_{nc}T_{cf}] + \varepsilon$$

$$A_{sn} = Q_{sc}W_{cn} + \varepsilon$$

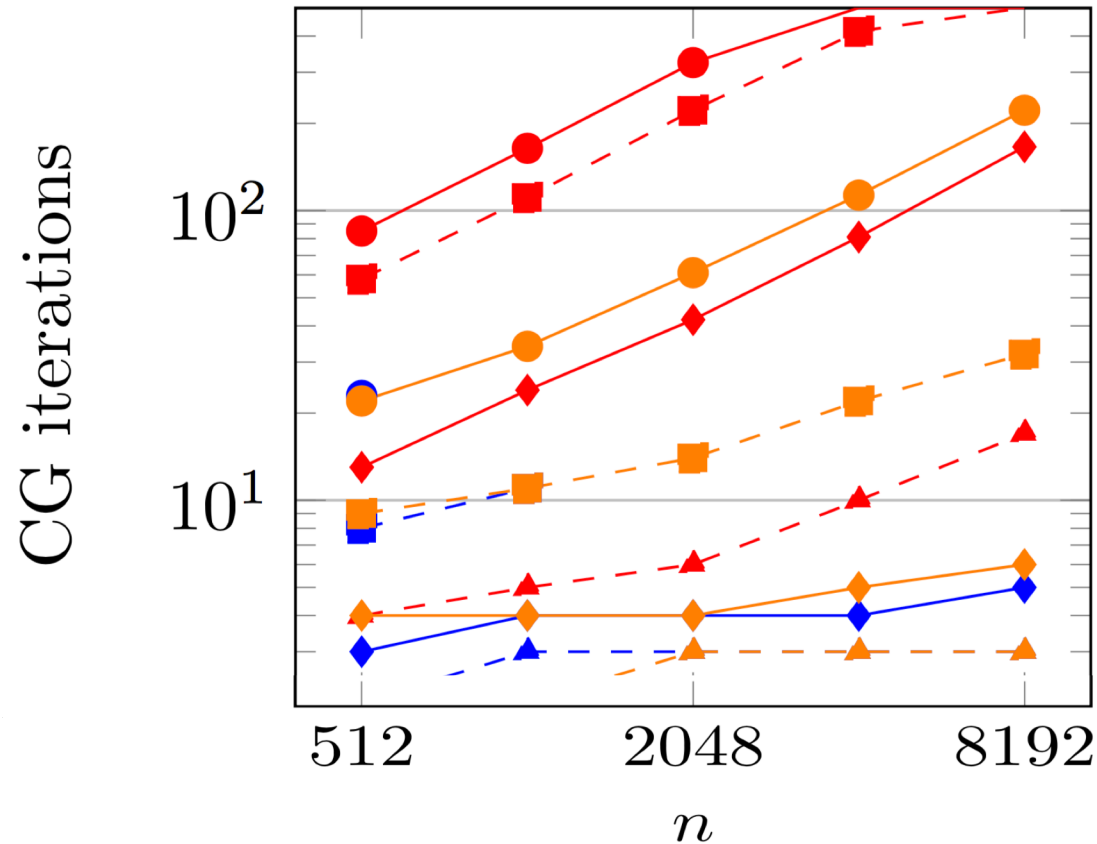
Low-Rank Compression: three variants

$$\begin{bmatrix} \mathbf{I} & & A_{sn} \\ & A_{ww} & A_{wn} \\ A_{ns} & A_{nw} & A_{nn} \end{bmatrix} \quad A_{sn} = Q_{sc}W_{cn} + \overset{O(\varepsilon)}{\boxed{Q_{sf}W_{fn}}}$$

$$\boxed{S_{nn} = A_{nn} - W_{cn}^\top W_{cn}} - W_{cf}^\top W_{cf}$$

Schur Complement over (n,n)

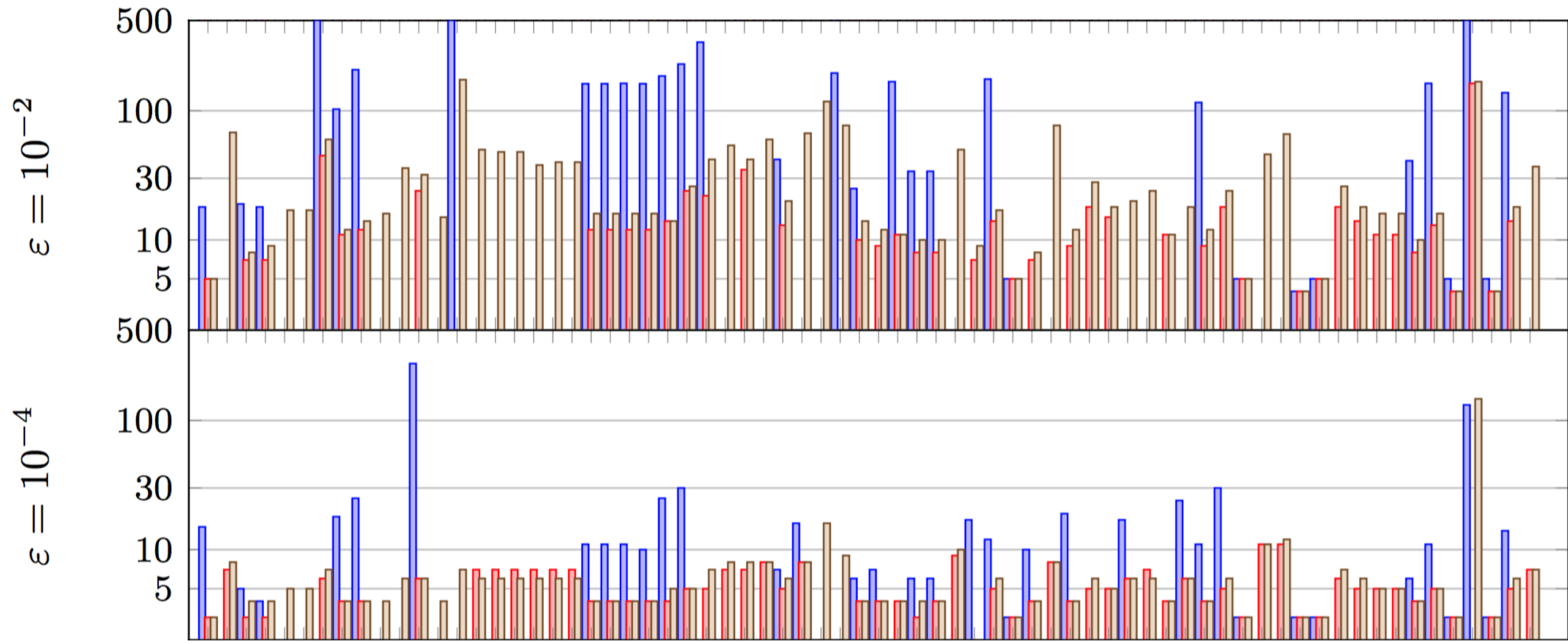
Low-Rank Compression: three variants



Interpolative, no scaling
Interpolative, with scaling
Orthogonal, with scaling

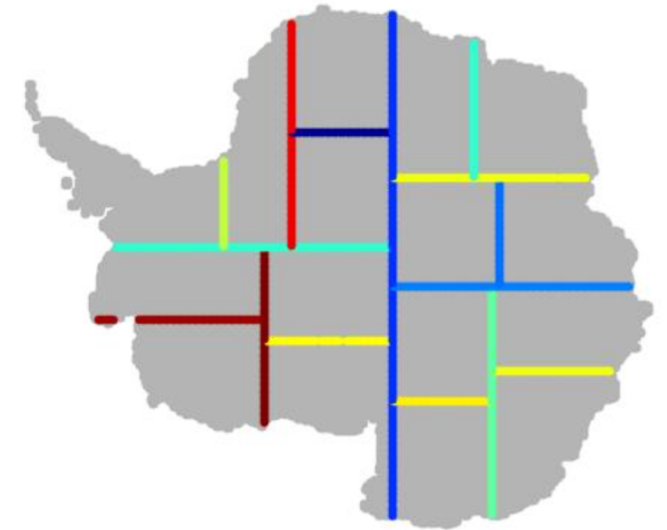
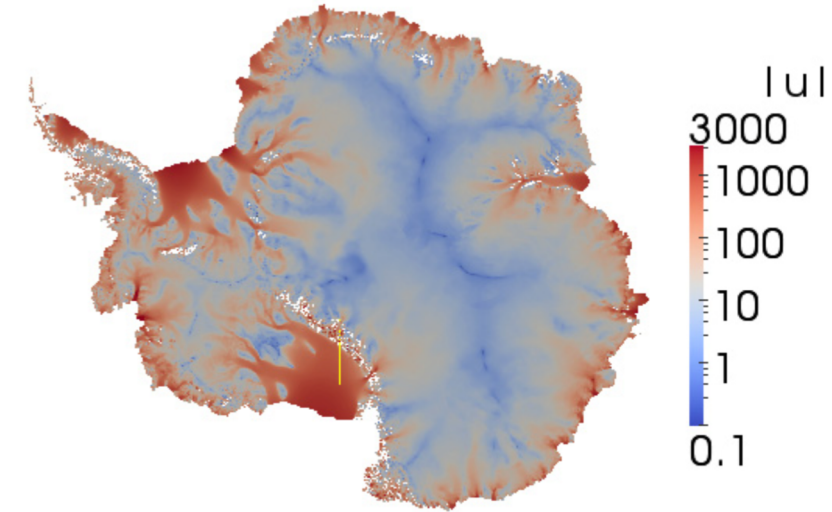
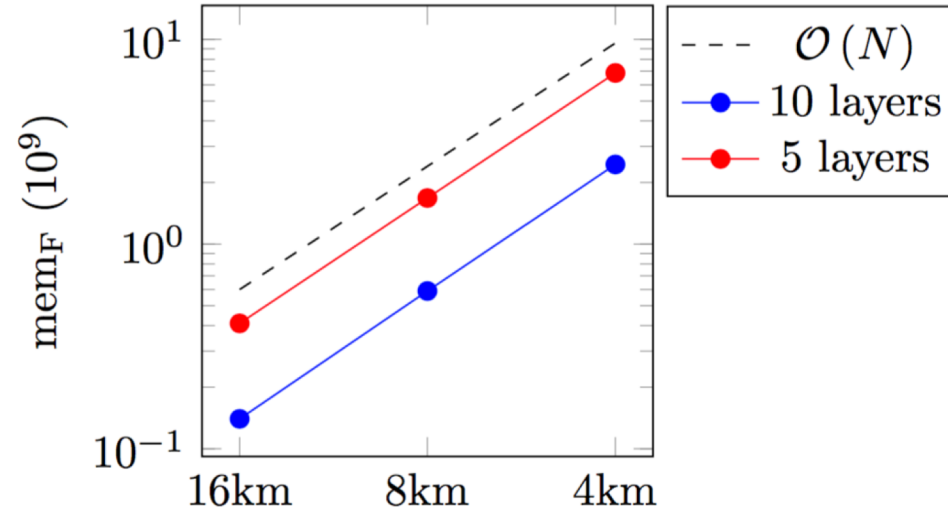
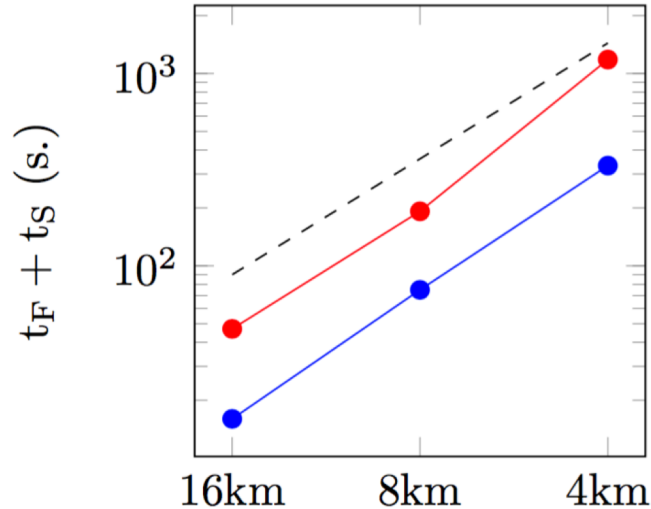
$$\varepsilon = 10^{-6} \rightarrow 10^{-1}$$

All SPD problems from SuiteSparse



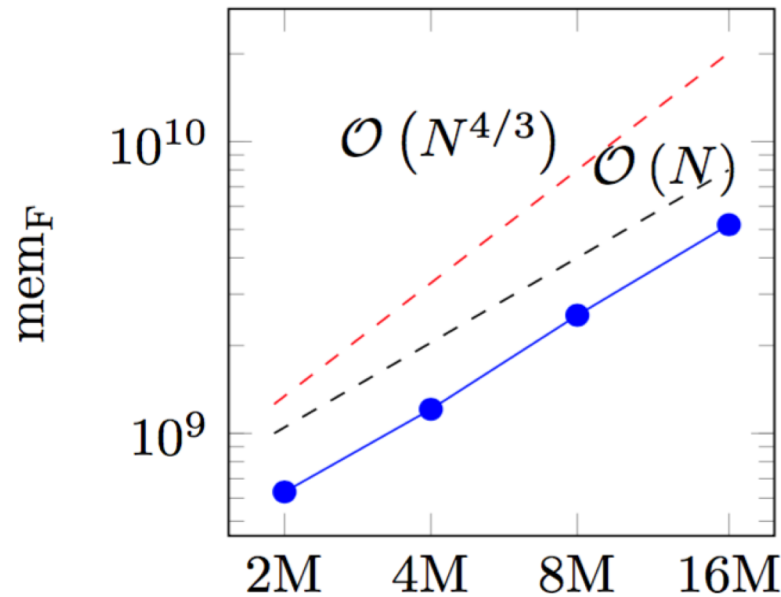
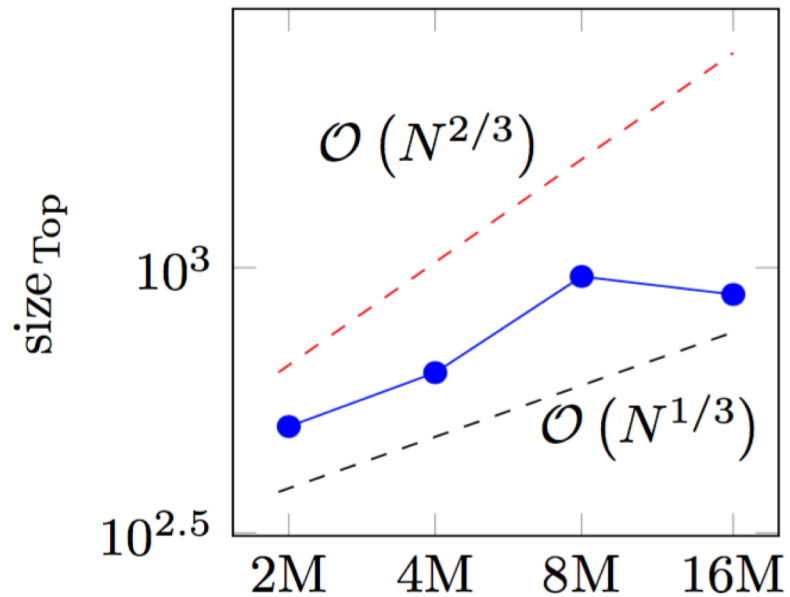
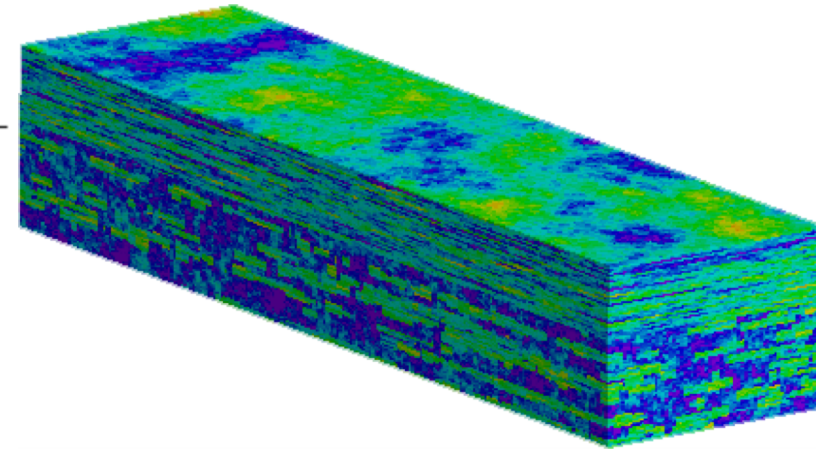
A very ill-conditioned problem

N	spaND					Direct	ILU
	t_F (s.)	t_S (s.)	n_{CG}	$size_{Top}$	mem_F (10^9)	$t_F + t_S$ (s.)	$t_F + t_S$ (s.) (GMRES)
5 layers							
629 544 (16 km)	13	3	7	76	0.14	22	150 (45)
2 521 872 (8 km)	55	20	8	89	0.59	206	801 (76)
10 096 080 (4 km)	217	115	10	100	2.45	1578	
10 layers							
1 154 164 (16 km)	39	8	7	136	0.41	90	524 (90)
4 623 432 (8 km)	148	44	8	148	1.68	710	
18 509 480 (4 km)	798	384	10	159	6.86	—	



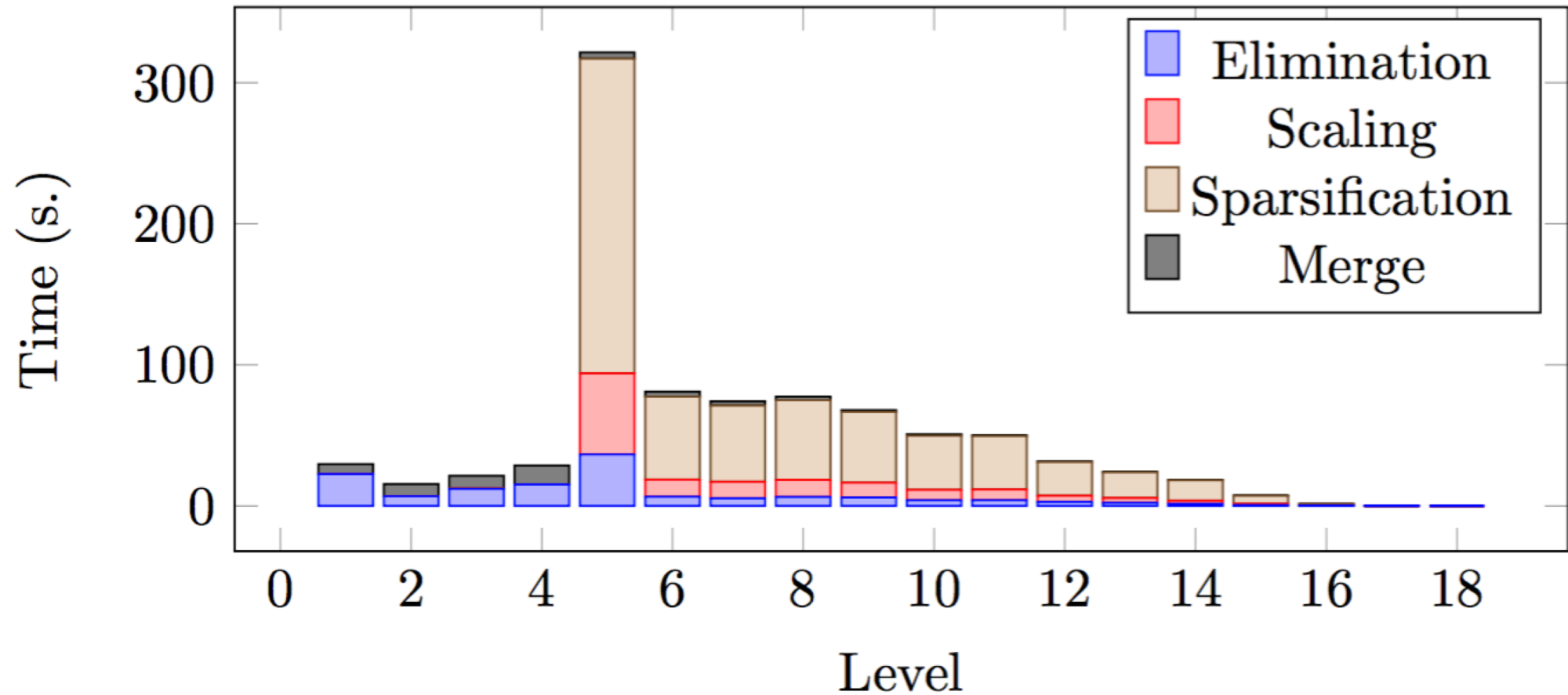
A 'cubic-like' problem

n	$N = n^3$	spaND					Direct.
		t_F (s.)	t_S (s.)	n_{CG}	$size_{Top}$	mem_F (10^9)	$t_F + t_S$ (s.)
128	2 097 152	61	23	12	502	0.63	686
160	4 096 000	175	46	13	634	1.21	—
200	8 000 000	287	158	16	962	2.54	—
252	16 003 008	963	369	16	890	5.19	—



Top separator bloc
would be 32 GB
otherwise!

Profiling: RRQR takes (lots) of time



Acknowledgements & Funding

- References:
 - K. L. Ho and L. Ying, Hierarchical interpolative factorization for elliptic operators: differential equations, Communications on Pure and Applied Mathematics, 69 (2016), pp. 1415–1451
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