

The importance of Structure in Algebraic Preconditioners

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August 14, 2009,

Hong Kong Baptist University

Outline

- 1 Introduction: Preconditioned iterative methods
- 2 Goal of this talk
- 3 Algebraic preconditioners - direct incomplete decompositions
- 4 The importance of having structure
- 5 Conclusions

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Preconditioned iterative methods

Solving large, sparse SPD systems by iterative methods

$$Ax = b$$

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In particular: **Incomplete decompositions**

- **As usual**, should be cheap, fast to compute, implying fast converging preconditioned iterative method
- **sparse** enough
- providing **just sufficient** approximation of the algebraic problem if this makes computations faster
- **Our target is robustness, not a fragile power**

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- Some notes on the history of the structure-based preconditioners

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- Some notes on the history of the structure-based preconditioners
- Basic ways of improvements
- Experimental results showing some structure-based effects

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Incomplete decompositions

First encounter

Stencil-based advent

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- **stencils \leftrightarrow local interpolation \leftrightarrow elimination**
- starting with first order factorizations $N = M - A = O(h)$

Incomplete decompositions

First encounter (continued)

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- Nice results related to conditioning of $M^{-1}A$, subsequent diagonal modifications and their “algebraizations” (Dupont, Kendall, Rachford (1968), Gustafsson (1978), Axelsson, Lindsgog (1986), van der Vorst (1989) and a lot later work!).

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Incomplete decompositions

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More general patterns

- Crucial moment: paper by Meijerink and van der Vorst (1977) recognizing the potential of incomplete decompositions for preconditioning.
- Incomplete decompositions classified by adding (ℓ) after the name. Starting to denote them by number of additional diagonals in simple problems $\rightarrow IC(\ell)$.

Incomplete decompositions

General patterns

Matrix-based approach

Incomplete decompositions

General patterns

Matrix-based approach

- Matrix \rightarrow graph

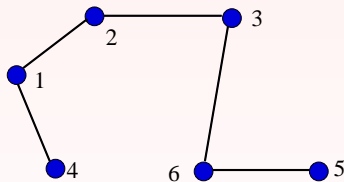
Incomplete decompositions

General patterns

Matrix-based approach

- Matrix \rightarrow graph

$$A = \begin{bmatrix} 1 & 1 & & & & & \\ & 1 & 1 & & & & \\ & & 1 & 1 & & & \\ 1 & & & & 1 & & \\ & & & & & 1 & 1 \\ & & 1 & & 1 & 1 & \\ & & & & & & 1 \end{bmatrix}$$



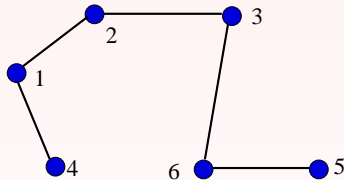
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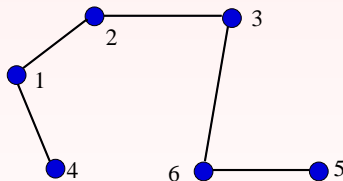
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- Entries of the Cholesky factor $l_{ij}, i > j$ are nonzero if and only if there is a **fill path** joining i and j in G .

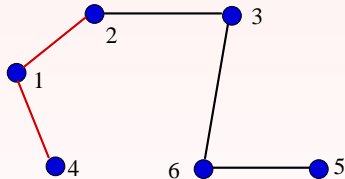
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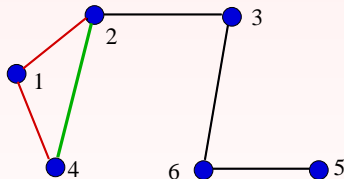
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$$A = \begin{bmatrix} 1 & 1 & & 1 & & \\ & 1 & 1 & 1 & \textcolor{red}{1} & \\ & & 1 & 1 & & 1 \\ 1 & \textcolor{red}{1} & & 1 & & \\ & & & & 1 & 1 \\ & & 1 & & 1 & 1 \end{bmatrix}$$



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- Allowing fill up to a maximum length ℓ of any fill path (Watts III, (1981)).

Incomplete decompositions

General patterns

Matrix-based approach

- Allowing fill up to a **maximum length** ℓ of any fill path (Watts III, (1981)).
- Practically: A fill entry is permitted provided $level(i, j) \leq \ell$.

$$level(i, j) = \min_{1 \leq l \leq \min\{i, j\}} \{level(i, l) + level(l, j) + 1\}$$

(one of more slightly different definitions)

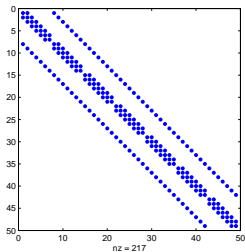
Incomplete decompositions

General patterns: an example

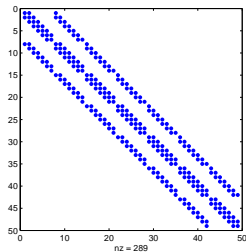
Example: Structure-based preconditioners

Incomplete decompositions

General patterns: an example



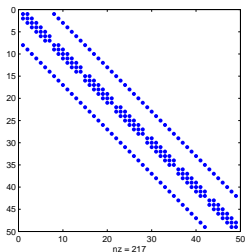
IC(0)



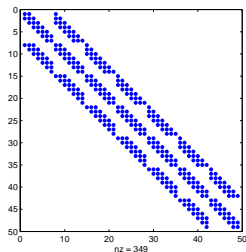
IC(1)

Incomplete decompositions

General patterns: an example



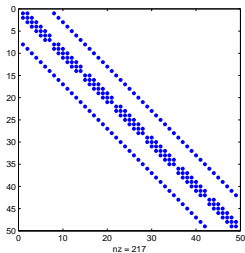
IC(0)



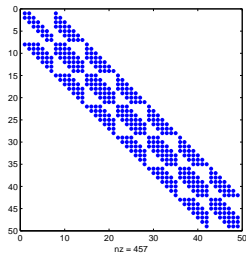
IC(2)

Incomplete decompositions

General patterns: an example



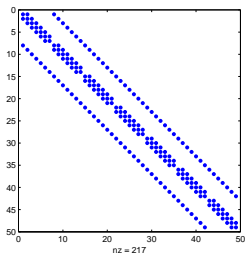
IC(0)



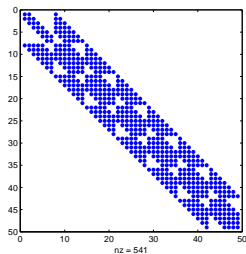
IC(3)

Incomplete decompositions

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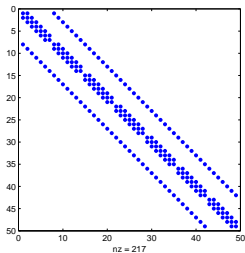
IC(0)



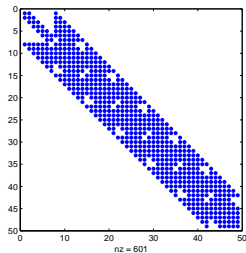
IC(4)

Incomplete decompositions

General patterns: an example



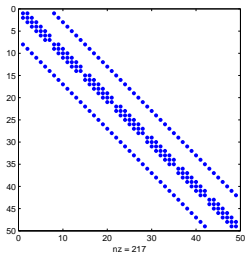
IC(0)



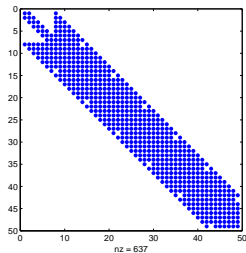
IC(5)

Incomplete decompositions

General patterns: an example



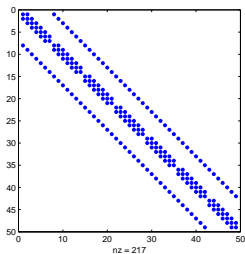
IC(0)



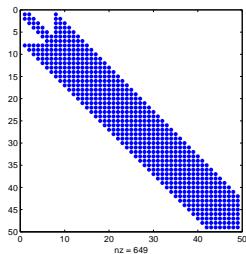
IC(6)

Incomplete decompositions

General patterns: an example



IC(0)



IC(7)

Incomplete decompositions

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- While the error $R = A - LL^T$ inside the pattern is zero, outside can be large.

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Matrix-based approach

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- While the error $R = A - LL^T$ inside the pattern is zero, outside can be large.
- But: Decay of entries away from diagonal may help a lot.

Incomplete decompositions

Natural competitor of level-based methods: considering values

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- Dropping entries with “smaller magnitudes” (absolutely/relatively) (Zlatev et al. (1978), Munksgaard (1980), Axelsson (1972, 1983 et al. etc.)

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- Plassman, Jones (1995): no structure, just the memory predictability, see also Freund, Nachtigal, (1990). Similarly Lin, Moré with extended memory. ILUT by Saad, (1994).
- The importance of error matrix $E = A - LL^T$ understood (Duff, Meurant, (1989)) and exploited (D'Azevedo, Forsyth, Tang, 1992)

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- First simple **combination** of level-based approaches with dropping: D'Azevedo, Forsyth, Tang, (1992a).

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- Generally, convergence behavior can be far from predictable
- The **real breakthrough** in level-based approaches: cheap predictions by Hysom, Pothen, (2002)
- Our MI22 preconditioner is a new way to use **level-based information, memory prediction and dropping** at the same time.

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Preassign levels to the entries individually

- Computing the absolute values of the smallest and largest entries of A : $msmall$ and $mbig$.
- Distribute nonzero entries uniformly by $\log |a_{ij}|$ into $ngroup0 = \lceil \log(mbig) - \log(msmall) \rceil + 1$ groups. Shrink zero groups to get $ngroup$ of them.

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- Set $level(i, j)$ for individual entries: For $\ell < ngroup$:
 $level(i, j) = (\ell - 1) * (l/ngroup) + 1$ where l ($1 \leq l \leq ngroup0$) is the index of the group a_{ij} belongs to, and slightly differently otherwise.

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- During the $IC(\ell)$ decomposition, entries of the factor L that correspond to nonzero entries of A are assigned the level $level(i, j)$.
- Each potential fill entry l_{ij} is assigned a level

$$level(i, j) = \min_{1 \leq l \leq \min\{i, j\}} \{level(i, l) + level(l, j) + 1\}.$$

A fill entry is permitted provided $level(i, j) \leq k$.

First component of our approach: new setting of levels

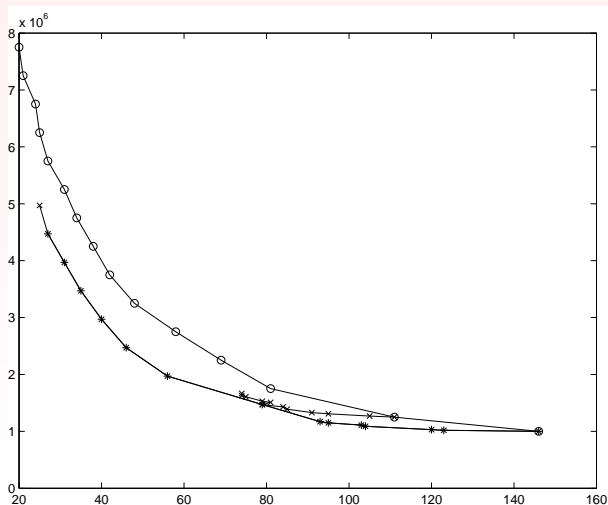
Experiments: Kohn-Sham equation, $n=250500$

Effects individual level preassignments

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Notes on the experiments

- (+) Settings do not increase timings significantly.
- (-) The improvements are **often small**. We intend to construct a robust strategy which should be used as a default value.
- **Open problem:** determine more sophisticated rules to preassign levels.

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- The final size parametrized by **memory multiplier** $0 \leq \theta$.

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- Only very small entries from the structure are not kept. The space is then freed and can be further used.
- The final size parametrized by **memory multiplier** $0 \leq \theta$.
- Additional space distributed **(1) uniformly** or **(2) nonuniformly**.

Second component of our approach: keeping structure

Integrate the predefined factor structure with dropping

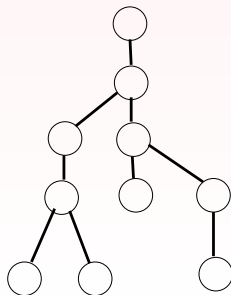
- Symbolic part of our modified (MI22) $IC(\ell)$ predefines the structure.
 - Only very small entries from the structure are not kept. The space is then freed and can be further used.
 - The final size parametrized by **memory multiplier** $0 \leq \theta$.
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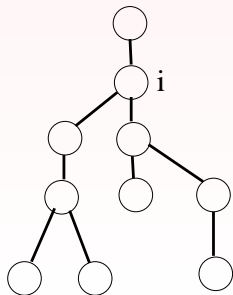


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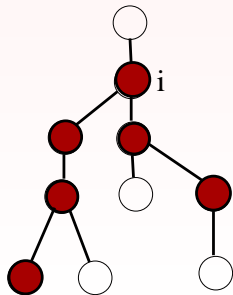


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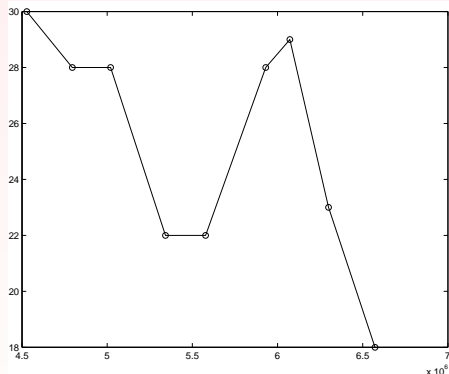
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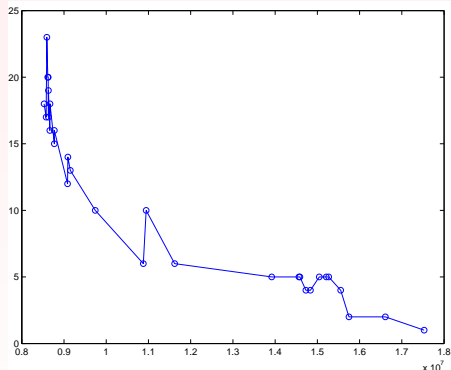
Second component of our approach: keeping structure

Experiments with memory multiplier $\theta = 1$: more difficult problem



MI22

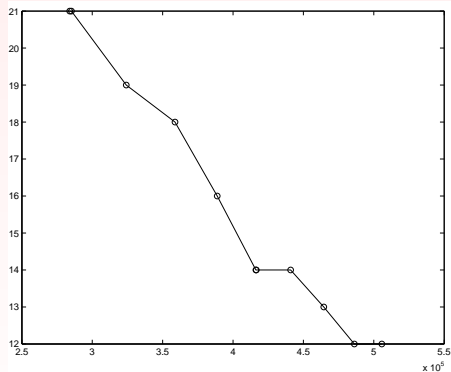
NASASRB, structural mechanics, $n=54870$



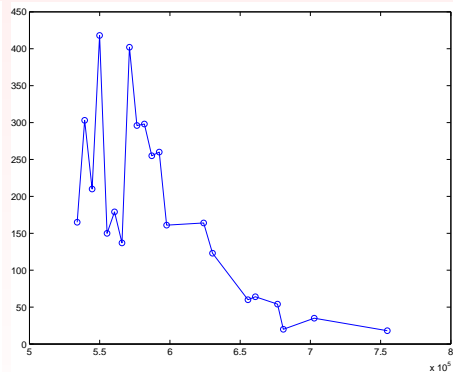
ILUT (ICT)

Second component of our approach: keeping structure

Experiments with memory multiplier $\theta = 1$: simpler problem



MI22



ILUT (ICT)

S1RMT3M1, cylindrical shell problem, $n=5489$

MI22 with levels versus $IC(\tau)$ (also via MI22)

TUBE1, cylindrical shell, $n=21498$

struc	drop=0.0		drop= 10^{-7}	
level	size	its	size	its
5	1250952	†	1227570	†
6	1660827	429	1618808	423
7	1807337	405	1756733	408
8	2178312	272	2104496	281
9	2368289	260	2280081	267
10	3026431	184	2873613	185
11	3968731	426	3656826	335
12	4874629	†	4398086	†
13	5849563	†	5178688	†
14	6840871	664	5938543	647
15	7838623	262	6680235	215

$IC(\tau)$	size	its
55	280626	†
50	1458024	†
45	2076970	†
40	2252687	†
1e-3	16139618	†
1e-4	9001342	†
5e-5	9649083	471
2e-5	9610841	87
1e-5	10050227	18
5e-6	10741254	6
1e-6	12451396	2
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But: Reorderings may minimize the effect.

MI22 and memory multiplier $\theta < 1$

simple problem, 2D POISSON on a square, $n=10000$

memory	drop=0.0		drop= 10^{-4}	
0.2	10000	160	10000	160
0.3	11880	226	11880	226
0.4	15840	205	15840	205
0.5	19800	155	19800	155
0.6	27729	141	27721	142
0.7	35597	111	35524	111
0.8	39583	63	39583	63
0.9	39584	58	39584	63
1	39601	41	39601	41
1.5	59401	42	59401	43
2	79202	42	79202	42
3	118803	42	118803	40
4	158404	42	158404	42
5	198005	44	198005	39
8	316808	42	316808	32
10	396010	42	396010	22
15	594015	37	471092	6

Outline

- 1 Introduction: Preconditioned iterative methods
- 2 Goal of this talk
- 3 Algebraic preconditioners - direct incomplete decompositions
- 4 The importance of having structure
- 5 **Conclusions**

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