

# On some strategies for Jacobian-free preconditioning of sequences of nonsymmetric linear systems

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# 1. Jacobian-free preconditioning

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Consider a **sequence** of large scale systems of nonlinear equations

$$F^{(j)}(x) = 0, \quad j = 0, 1, 2, \dots$$

with continuously differentiable functions

$$F^{(j)} : \mathbb{R}^n \rightarrow \mathbb{R}^n.$$

They arise in numerous scientific and industrial applications, e.g. **every time step  $t_j$  in a numerical simulation may require the solution of  $F^{(j)}(x) = 0$** . They arise in Navier-Stokes equations, heat conduction problems, differential-algebraic problems, many kinds of initial and boundary value problems and many others.



# 1. Jacobian-free preconditioning

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We restrict ourselves to **inexact Newton-Krylov methods** (see e.g. [Kelley - 1995]) where each Newton iteration has the form

$$J(x_k)(x_{k+1} - x_k) = -F^{(j)}(x_k), \quad k = 1, 2, \dots,$$

where  $J(x_k)$  represents the Jacobian of  $F^{(j)}$  evaluated at  $x_k$ .

If the linear systems with the Jacobians are solved by a transpose-free Krylov subspace method, one can use a **Jacobian-free implementation**:

The multiplication of a vector  $v$  with  $J(x_k)$  is replaced by a **finite-difference approximation**, for example the first-order approximation

$$J(x_k) \cdot v \approx \frac{F^{(j)}(x_k + h\|x_k\|v) - F^{(j)}(x_k)}{h\|x_k\|},$$

for some small  $h$ .



# 1. Jacobian-free preconditioning

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- Exploiting finite-difference approximation, the **system matrix needs not be stored and not even be computed**. Convergence is Newton is in general not significantly influenced.
- Such an implementation is sometimes called „**matrix-free**“, which is a little misleading: It is often necessary to store preconditioners and smaller matrices in some Krylov subspace methods.
- A possible definition of matrix-free is „free of matrices with storage comparable to that of the Jacobian“ [Knoll, Keyes - 2004].
- We will use the term „**Jacobian-free**“.



# 1. Jacobian-free preconditioning

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Preconditioning is often **crucial** for satisfactory performance of both the Krylov and the Newton iteration.

Some preconditioners that do **not need the entries of the Jacobians explicitly** in order to be applied are:

- Multigrid and additive Schwarz preconditioners based on domain decomposition and grid coarsening,
- Fast Poisson solvers and other simplifications of operators as preconditioners in e.g. convection-diffusion-reaction problems,
- Krylov subspace methods as preconditioners for the inner iteration of an outer Krylov subspace, e.g. flexible GMRES [Saad - 1993].



# 1. Jacobian-free preconditioning

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- For particular problems these preconditioners can be superior to incomplete factorizations, for some references see [Knoll, Keyes - 2004].
- In other cases one needs the **universality and robustness** of incomplete LU or Cholesky decomposition.
- These factorizations need the entries of the Jacobian **explicitly!**
- The only way to **estimate** these entries is through simulating matrix-vector products (matvecs) with selected test vectors  $v_t$ , e.g.

$$J(x_k) \cdot v_t \approx \frac{F^{(j)}(x_k + h\|x_k\|v_t) - F^{(j)}(x_k)}{h\|x_k\|}.$$

In [Curtis, Powell, Reid - 1978] it was first shown this may be done with a **small number of matvecs** if the sparsity pattern of the Jacobian is known.



# 1. Jacobian-free preconditioning

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For example a tridiagonal Jacobian of arbitrary size

$$\begin{pmatrix} * & * & & & & & \\ * & * & * & & & & \\ & * & * & * & & & \\ & & \ddots & \ddots & \ddots & & \\ & & & * & * & * & \\ & & & & * & * & \end{pmatrix}$$

can be estimated through **only three** matvecs, namely with

$$\begin{aligned} & (1, 0, 0, 1, 0, 0, \dots)^T, \\ & (0, 1, 0, 0, 1, 0, \dots)^T, \\ & (0, 0, 1, 0, 0, 1, \dots)^T. \end{aligned}$$



# 1. Jacobian-free preconditioning

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In general, one uses a **graph coloring algorithm** to estimate the entries of the Jacobian [Coleman, More - 1984].

- The algorithms work on the *intersection graph*, that is on  $G(J^T J)$
- The computational costs of graph coloring algorithms are linear in the number of nonzeros
- Every color defines a test vector whose matvec simulated with a finite difference approximation gives some entries of the Jacobian
- Heuristics are used to **minimize** the number of colors

Note that to construct good **preconditioners** it need not be necessary to have **all** nonzeros of the Jacobian. Partial estimation with modified graph coloring may be cheaper [Cullum, Tũma - 2006].



# 1. Jacobian-free preconditioning

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When solving the sequence of large scale systems of nonlinear equations

$$F^{(j)}(x) = 0, \quad j = 0, 1, 2, \dots$$

one will always try to **share part of the computational effort throughout the sequence**. An option is reusing the same preconditioner over several Newton iterations, i.e. **freeze the preconditioner**. Note that **recomputing** the preconditioner requires for every linear system:

- A number of **additional matvec simulations (i.e. function evaluations)** to estimate the current matrix,
- When the sparsity pattern changes during the sequence: **Rerunning** the graph coloring algorithm,
- **Rerunning** the incomplete factorization.

Unfortunately, a frozen preconditioner can deteriorate when the system matrix changes too much. A good compromise may be **approximate preconditioner updates**.



## 2. Approximate preconditioner updates

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Some proposed preconditioner updates include:

- In [Meurant - 2001] and [Bellavia, de Simone, di Serafino, Morini - 2011] we find approximate preconditioner updates of **incomplete Cholesky factorizations** for **shifted SPD** matrices.
- In Quasi-Newton methods the difference between system matrices is of small rank and preconditioners may be efficiently adapted with approximate **small-rank preconditioner updates**; this has been done in the symmetric positive definite case, see e.g. [Bergamaschi, Bru, Martínez, Putti - 2006, Nocedal, Morales - 2000].
- The preconditioner update can consist of adding **recycled (spectral) information from previously generated Krylov subspaces**. This can be beneficial in many applications, see e.g. [Parks, de Sturler, Mackey, Johnson, Maiti - 2006], [Giraud, Gratton, Martin - 2007], [Frank, Vuik - 2001].



## 2. Approximate preconditioner updates

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The previous factorization updates are more or less problem specific. We now describe a class of more **black-box approximate preconditioner updates** based on an idea in [Benzi, Bertaccini - 2003].

Notation: Consider two linear systems in the sequence,

$$Jx = b, \quad \text{and} \quad J^+x^+ = b^+$$

and let

$$\Delta \equiv J - J^+.$$

Further, let

$$J \approx LDU$$

be a **reference (seed) ILU factorization** for the reference Jacobian  $J$ .



## 2. Approximate preconditioner updates

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Then

$$J - LDU = J^+ - LDU + \Delta = J^+ - L(D + L^{-1}\Delta U^{-1})U,$$

i.e. the **preconditioner update**

$$L(D + L^{-1}\Delta U^{-1})U$$

is of the same accuracy as  $LDU$ .

This update is **not useful** as preconditioner because of the middle factor.

Assuming **fast decay** when moving away from the main diagonal in  $L^{-1}$  and  $U^{-1}$ , two ideas to modify the middle factor were proposed. Note that fast decay is often given in diagonally dominant problems, or it can be stimulated by reorderings that move large entries close to the main diagonal.



## 2. The considered preconditioner updates

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- Consider

$$L(D + \text{band}(L^{-1}\Delta U^{-1}))U$$

for a small bandwidth  $\Rightarrow$  additional inexpensive forming of  $\text{band}(L^{-1}\Delta U^{-1})$  to **compute** the preconditioner and additional inexpensive solving of a banded system to **apply** the preconditioner. This update was used in [Bertaccini - 2004].

- Consider

$$L(D + \Delta U^{-1})U = L(DU + \Delta) \approx L(DU + \text{triu}\Delta)$$

or

$$L(D + L^{-1}\Delta)U = (LD + \Delta)U \approx (LD + \text{tril}\Delta)U$$

$\Rightarrow$  **no additional costs** to compute or apply the preconditioner, but based on rather **greedy simplification**, see [DT, Tůma - 2007].



## 2. Approximate preconditioner updates

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Analogously, if

$$J^{-1} \approx W \tilde{D}^{-1} Z^T, \quad W, Z \text{ upper triangular,}$$

is a reference approximate **inverse** factorization for  $J$ , we can use the **banded** factorization update

$$W(\tilde{D} + \text{band}(Z^T \Delta W))^{-1} Z^T,$$

see [Benzi, Bertaccini - 2003] (symmetric case) and [Bellavia, Bertaccini, Morini - 2011], or we can use the **triangular** updates

$$W(\tilde{D} Z^T + \text{tril}(\Delta))^{-1}, \quad (\tilde{W} D + \text{triu}(\Delta))^{-1} Z^T.$$

Note that in general approximate *inverse* factorizations need more nonzeros than ILU or Cholesky to be efficient and thus are less suited for Jacobian-free environment.



## 2. Approximate preconditioner updates

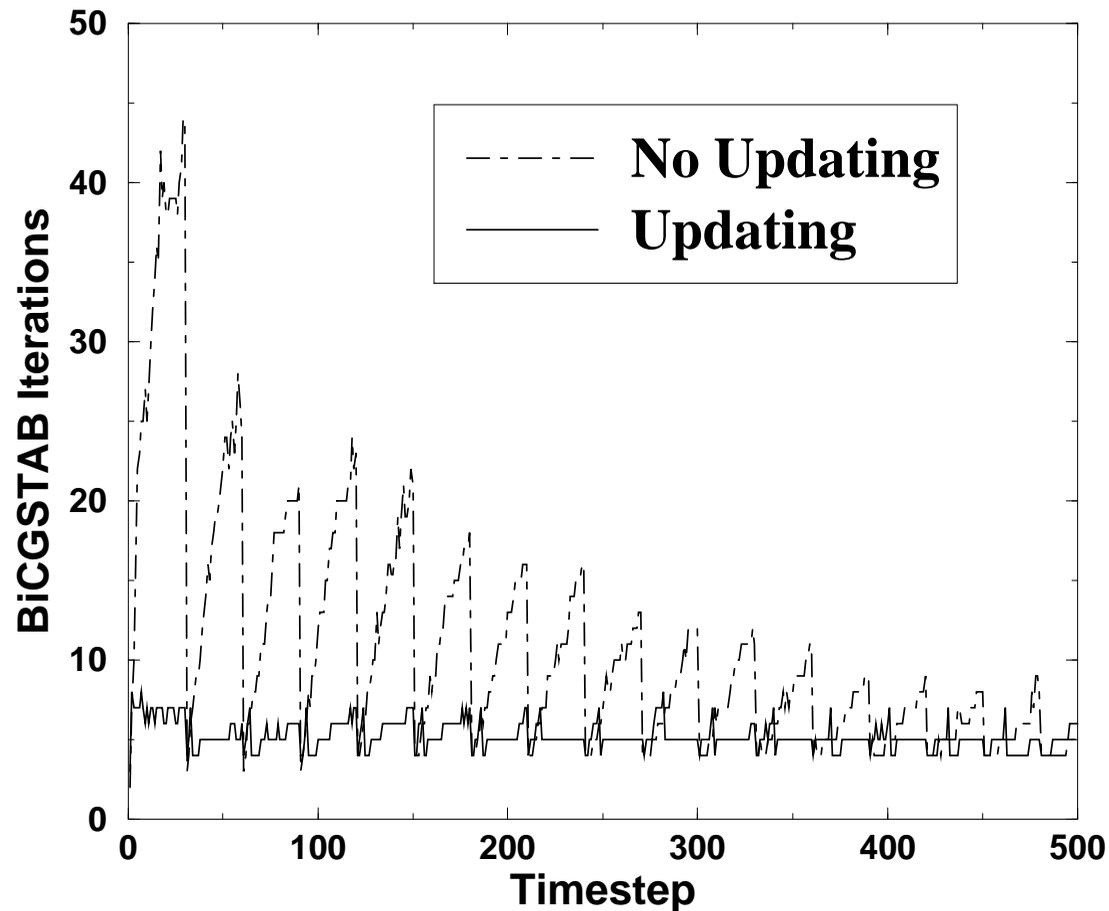
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Consider the following **CFD problem** (compressible supersonic flow):

- **Frontal flow with Mach-number 10 around a cylinder**, which leads to a steady state.
- 500 time steps of the implicit Euler method are performed.
- The grid consists of 20994 points, we use Finite Volume discretization and system matrices are of dimension 83976. The number of nonzeros is about  $1.33 \cdot 10^6$  for all matrices of the sequence.
- In the beginning, a strong shock detaches from the cylinder, which then slowly moves backward through the domain until reaching the steady state position.
- The iterative solver is BiCGSTAB with stopping criterion  $10^{-7}$ , the implementation is in C++.
- The **ILU preconditioner is recomputed for every 30th linear system.**



## 2. Approximate preconditioner updates



BiCGSTAB iterations for the first 500 systems in the [cylinder problem](#).



## 2. Approximate preconditioner updates

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Back to Jacobian-free implementation:

The described class of updates works with the **difference matrix**, that is with

$$\text{band}(Z^T \Delta W), \text{tril}(\Delta), \text{triu}(\Delta), \quad \text{where} \quad \Delta = J - J^+.$$

The **reference Jacobian  $J$  has usually been estimated** in order to obtain the reference factorization, but  $J^+$  has not.

Instead of direct estimation of the current Jacobian  $J^+$ , two directions for improvement were proposed:

- **Adapted graph coloring algorithms** to estimate only the necessary entries of  $J^+$ ,
- Cheap estimation of selected entries with **function component evaluations**.

First consider e.g. estimating  $\text{triu}(\Delta)$  with adapted graph coloring.



## 2. Approximate preconditioner updates

Academic example:

$$J^+ = \begin{pmatrix} * & & & & * \\ & * & & & * \\ & & \ddots & & \vdots \\ & & & \ddots & * \\ * & * & * & * & * \end{pmatrix}$$

- estimating the **whole matrix** asks for  $n$  matvecs with all unit vectors (i.e. there are  $n$  colors in the intersection graph);
- estimating the **upper triangular part** requires **only 2 matvecs** (i.e. there are 2 colors in the intersection graph),

$$(1, \dots, 1, 0)^T \quad \text{and} \quad (0, \dots, 0, 1)^T.$$



## 2. Approximate preconditioner updates

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Recall the graph coloring algorithm for a matrix  $C$  works on the *intersection graph*

$$G(C^T C).$$

We can prove [DT, Tůma - 2010]: The graph coloring algorithm for  $\text{triu}(C)$  works on

$$G(\text{triu}(C)^T \text{triu}(C)) \cup G_K, \quad \text{where}$$

$$G_K = \cup_{i=1}^n G_i, \quad G_i = (V_i, E_i) = (V, \{\{k, j\} \mid c_{ik} \neq 0 \wedge c_{ij} \neq 0 \wedge k \leq i < j\}).$$

- This graph may have less edges than  $G(C^T C)$  and **there may be needed significantly less matvecs to estimate  $\text{triu}(C)$  than to estimate  $C$**
- Similar modified coloring may be used to estimate the diagonals of  $J^+$  needed to form  $\text{band}(Z^T \Delta W)$
- These are instances of **partial graph coloring problems** [Pothen et al. - 2007].



## 2. Approximate preconditioner updates

As for the second strategy, assume the **components**  $F_i^+ : \mathbb{R}^n \rightarrow \mathbb{R}$  of the current function

$$F^+ = [F_1^+, \dots, F_n^+]^T : \mathbb{R}^n \rightarrow \mathbb{R}^n$$

can be easily **separated**. By this we mean that **a function component evaluation**  $F_i^+(x)$  **has the cost of about one  $n$ th of the full function evaluation**  $F^+(x)$ . (Note that in some Finite Element or Volume implementations different components  $F_i^+$  of  $F^+$  contain the same or simultaneously generated expressions and the evaluation of  $F^+(x)$  may be cheaper than  $n$  function component evaluations  $F_i^+(x)$ .)

With easily separable function component evaluations it may pay-off to **compute the needed entries of  $J^+$  individually** according to

$$e_i^T J^+(x_k) e_j \approx \frac{F_i^+(x_k + h\|x_k\|e_j) - F_i^+(x_k)}{h\|x_k\|}.$$



## 2. Approximate preconditioner updates

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By individual estimation with function component evaluations, we are able to **avoid running a graph coloring algorithm**.

Easily separable function components also give cheap estimations of *inner products with rows of the Jacobian*. For an arbitrary vector  $v$ , the inner product with the  $i$ th row of  $J^+$  can be computed as

$$e_i^T J^+(x_k)v \approx \frac{F_i^+(x_k + h\|x_k\|v) - F_i^+(x_k)}{h\|x_k\|}$$

$\Rightarrow$  the entries of  $J^+W$  needed to form

$$\text{band}(Z^T \Delta W) = \text{band}(Z^T (J - J^+)W)$$

can be obtained from inner products with the rows of  $J^+$  [Bellavia, Bertaccini, Morini - 2011]. This is in general faster than estimating the needed entries of  $J^+$  individually. For extracting bands of Hessians in truncated Newton with function component evaluations, see [Lukšan, Matonoha, Vlček - submitted].



## 2. Approximate preconditioner updates

For the preconditioner update  $L(DU - \text{triu}\Delta)$  (or  $(LD - \text{tril}\Delta)U$ ) we can avoid any estimation of entries of  $J^+$  (except for the main diagonal). We can apply  $L(DU - \text{triu}\Delta)$  as follows:

- The forward solves with  $L$  are done with the stored entries of  $L$ .
- For the backward solves, use a **mixed explicit-implicit solves**: Split

$$DU - \text{triu}\Delta = DU - \text{triu}(J) + \text{triu}(J^+) \equiv X + \text{triu}(J^+)$$

in the **explicitly given**  $X \equiv DU - \text{triu}(J)$  and the **implicit**  $\text{triu}(J^+)$ .

- We then have to solve the upper triangular systems

$$(DU - \text{triu}\Delta) z = (X + \text{triu}(J^+)) z = y,$$

yielding the standard backward substitution cycle

$$z_i = \frac{y_i - \sum_{j>i} X_{ij} z_j - \sum_{j>i} J_{ij}^+ z_j}{X_{ii} + J_{ii}^+}, \quad i = n, n-1, \dots, 1.$$



## 2. Approximate preconditioner updates

In

$$z_i = \frac{y_i - \sum_{j>i} X_{ij} z_j - \sum_{j>i} J_{ij}^+ z_j}{X_{ii} + J_{ii}^+}, \quad i = n, n-1, \dots, 1.$$

the inner product  $\sum_{j>i} J_{ij}^+ z_j$  with the  $i$ th row of  $J^+$  can be computed by the function component evaluation

$$\begin{aligned} \sum_{j>i} J_{ij}^+ z_j &= e_i^T J^+(0, \dots, 0, z_{i+1}, \dots, z_n)^T \\ &\approx \frac{F_i^+(x_k + h\|x_k\|(0, \dots, 0, z_{i+1}, \dots, z_n)^T) - F_i^+(x_k)}{h\|x_k\|}. \end{aligned}$$

The **diagonal**  $\{J_{11}^+, \dots, J_{nn}^+\}$  can be found by individual estimation

$$J_{ii}^+ = e_i^T J^+(x_k) e_i \approx \frac{F_i^+(x_k + h\|x_k\|e_i) - F_i^+(x_k)}{h\|x_k\|}, \quad 1 \leq i \leq n.$$



## 2. Approximate preconditioner updates

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- This last technique enables to do **backward or forward solves with  $J^+$  without storage of off-diagonal entries.**
- The same can be done for the reference Jacobian, hence we do not need to form or store off-diagonal entries of  $\Delta = J - J^+$
- This **gain in storage costs is paid by slightly higher computational costs:** Every forward or backward solve requires  $n$  function component evaluations, i.e. roughly one function evaluation, instead of  $n$  inner products.



## 2. Approximate preconditioner updates

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As an example consider a two-dimensional **nonlinear convection-diffusion model problem**: It has the form

$$-\Delta u + Ru \left( \frac{\partial u}{\partial x} + \frac{\partial u}{\partial y} \right) = 2000x(1-x)y(1-y), \quad (1)$$

on the unit square, discretized by 5-point finite differences on a uniform grid.

- The initial approximation is the discretization of  $u_0(x, y) = 0$ .
- We use here  $R = 500$ .
- We use the Newton method with a line search and solve the resulting linear systems with BiCGSTAB with right preconditioning.
- We use a flexible stopping criterion (see e.g. [Eisenstat, Walker - 1996]).
- Fortran implementation (embedded in the UFO - software for testing nonlinear solvers [Lukšan et al. - 2008]).



## 2. Approximate preconditioner updates

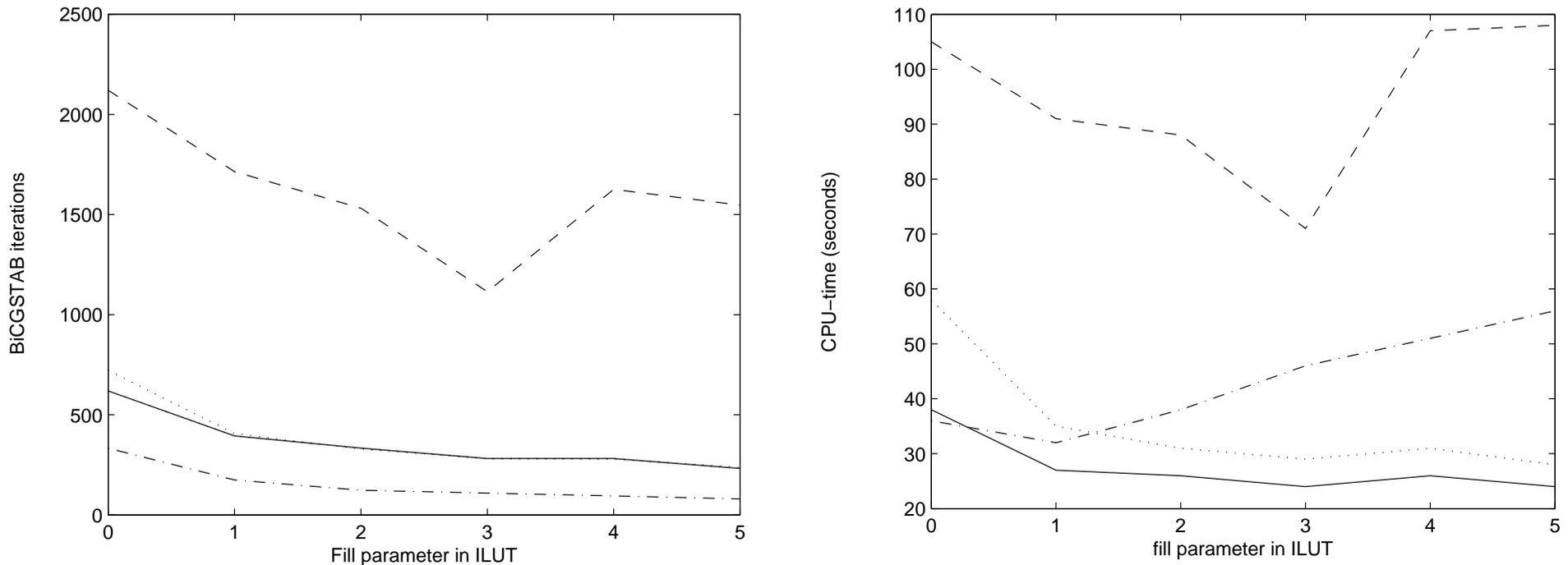


Figure 1: BiCGStab iterations and CPU-times for a  $250 \times 250$  grid (dimension 62 500) with varying sizes of ILUT-factorizations (depending on the fill parameter) for freezing (dashed lines), recomputing (dash-dotted lines), triangular updating with modified graph coloring (solid lines) and triangular updating with mixed explicit-implicit solves (dotted lines).



## 2. Approximate preconditioner updates

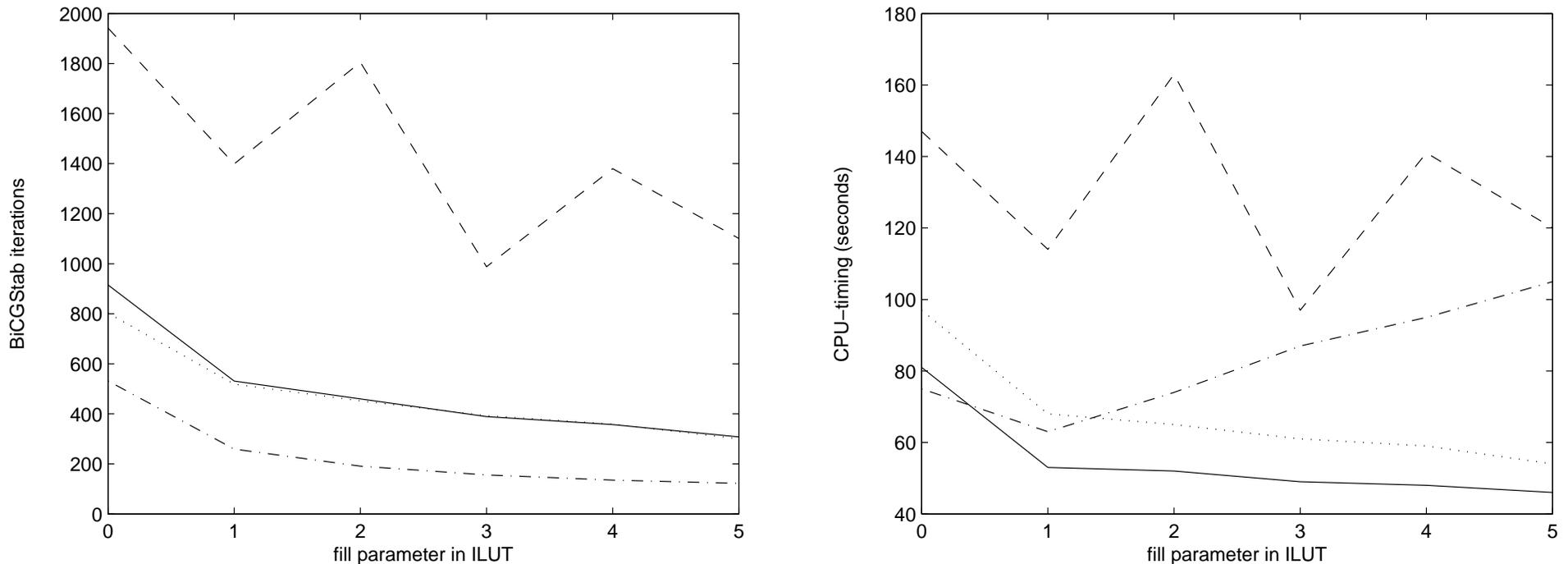


Figure 2: BiCGStab iterations and CPU-times for a  $310 \times 310$  grid (dimension 96 100) with varying sizes of ILUT-factorizations (depending on the fill parameter) for freezing (dashed lines), recomputing (dash-dotted lines), triangular updating with modified graph coloring (solid lines) and triangular updating with mixed explicit-implicit solves (dotted lines).



## 2. Approximate preconditioner updates

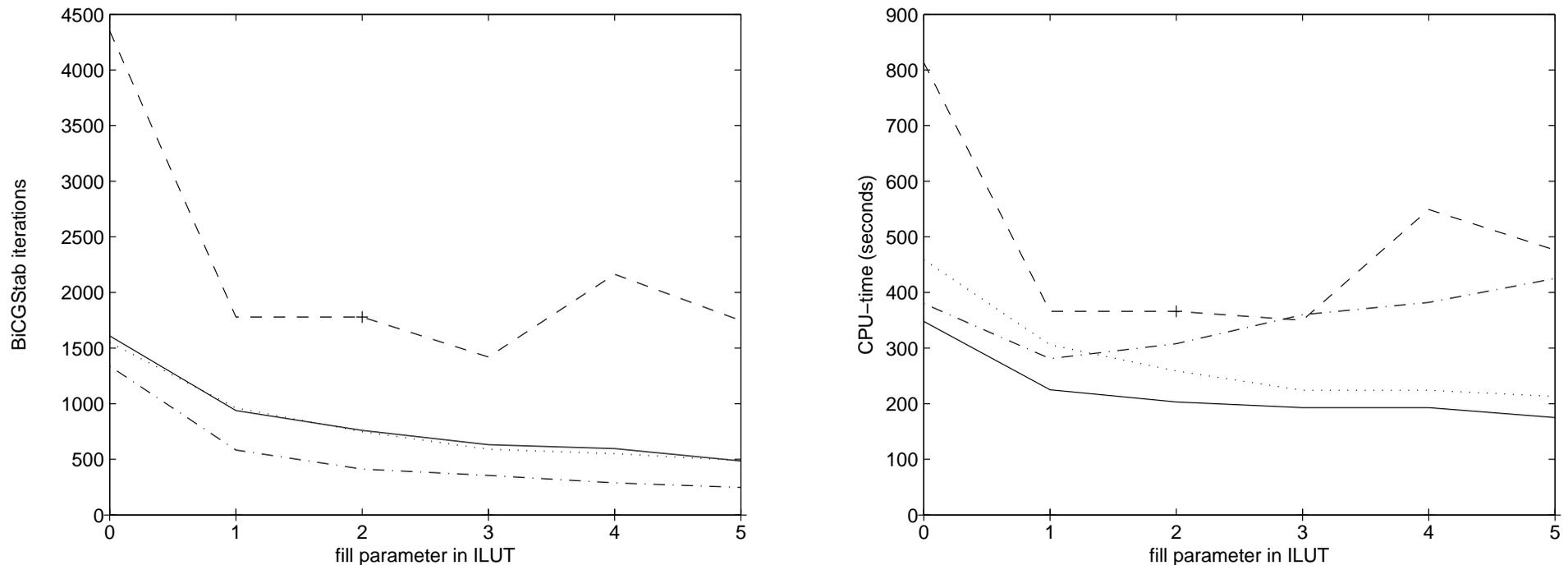


Figure 3: BiCGStab iterations and CPU-times for a  $490 \times 490$  grid (dimension 240 100) with varying sizes of ILUT-factorizations (depending on the fill parameter) for freezing (dashed lines), recomputing (dash-dotted lines), triangular updating with modified graph coloring (solid lines) and triangular updating with mixed explicit-implicit solves (dotted lines).



# Conclusion

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Conclusions and future work:

- The described preconditioner updates are **suitable for Jacobian-free sequences of linear systems**
- We showed that **triangular solves with the Jacobians** can be done while storing only their diagonal → all Gauss-Seidel type preconditioners can be implemented in this way
- For the far future: work towards **ILU preconditioning without storing the L and U factors** (or only their main diagonal) ??



For more details see:

- DUINTJER TEBBENS J, TŮMA M: *Preconditioner Updates for Solving Sequences of Linear Systems in Matrix-Free Environment*, Num. Lin. Alg. Appl. vol. 17, pp. 997–1019, 2010.
- BIRKEN PH, DUINTJER TEBBENS J, MEISTER A, TŮMA M: *Preconditioner Updates Applied to CFD Model Problems*, Applied Numerical Mathematics vol. 58, no. 11, pp. 1628–1641, 2008.
- DUINTJER TEBBENS J, TŮMA M: *Efficient Preconditioning of Sequences of Nonsymmetric Linear Systems*, SIAM J. Sci. Comput., vol. 29, no. 5, pp. 1918–1941, 2007.

# Thank you for your attention!

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