

New preconditioner updates in Newton-Krylov methods for nonlinear systems

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Joint work with
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11th EUROPT Workshop on Advances in Continuous Optimization
Firenze - June 26-28, 2013

Newton-Krylov methods for nonlinear systems

$$F(x) = 0$$

$F : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is continuously differentiable and J is the Jacobian matrix of F , n is large.

Given $x_0 \in \mathbb{R}^n$, a sequence $\{x_k\}$ is built:

- ▶ a **Krylov** method, e.g. GMRES, BiCGSTAB, is used to compute a step s_k that approximately solves

$$J(x_k)s = -F(x_k)$$

satisfying

$$\|J(x_k)s_k + F(x_k)\| \leq \eta_k \|F(x_k)\|, \quad \eta_k \in [0, 1)$$

Then $x_{k+1} = x_k + s_k$,



Sequence of linear systems

$$J(x_k)s = -F(x_k), \quad k = 0, 1, \dots$$

- ▶ By continuity, if $x_k \approx x_{k+1}$ then matrix $J(x_k)$ varies slowly from one step k to the next;
- ▶ Generally, $J(x_k)$ is not symmetric.
- ▶ **Matrix-free setting**: the action of the Jacobian times a vector can be provided by an operator or it can be approximated by finite differences:

$$J(x)v \approx \frac{F(x + \epsilon v) - F(x)}{\epsilon}$$

where ϵ is an appropriately chosen positive scalar.



Preconditioning & Matrix-free setting

- ▶ **Recomputing** an algebraic preconditioner at each nonlinear iterations requires forming the Jacobian matrices or approximating them and the resulting Newton-Krylov procedure is **no longer matrix-free**.
- ▶ Freezing the preconditioner slows the Krylov solver when the system matrix changes too much.
- ▶ A preconditioning strategy is considered to be matrix-free, or **nearly matrix-free** if
 - ▶ only a few full Jacobians are formed;
 - ▶ it needs matrices that are reduced in complexity with respect to the full Jacobians;
 - ▶ it takes advantage of matrix-vector product approximations by finite differences.

[Knoll, Keyes, 2004]



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Techniques for preconditioning sequences of matrices

- ▶ **Freezing the preconditioner** computed for the first matrix of the sequence.
- ▶ **Recomputing the preconditioner** from scratch for each matrix (costly and pointless accurate).



Updating an existing preconditioner computed for one matrix of the sequence (**seed** preconditioner) by inexpensive strategies.



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Existing updating approaches

Quasi-Newton Preconditioner:

- ▶ Broyden-type rank one updates for sequences arising in Newton methods [Bergamaschi, Bru, Martinez, Putti 2006]
- ▶ Limited memory Quasi-Newton updates (spd matrices) [Morales, Nocedal 2000].

Approximate preconditioner updated:

- ▶ Banded preconditioned updated for sequences of spd matrices [Benzi, Bertaccini 2003, Bertaccini 2004]
- ▶ AINV based preconditioners for sequences of unsymmetric linear systems [Bellavia, Bertaccini, Morini 2011]
- ▶ LDL^T based preconditioner updated for sequences of diagonally modified linear systems [Bellavia, De Simone, di Serafino, Morini 2010-2012]
- ▶ LDU based preconditioner for sequences of unsymmetric linear systems [Duintjer Tebbens, Tuma 2007, 2010]



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The Duintjer Tebbens & Tuma approach

- ▶ Let J_s be a specific Jacobian of the sequence and $P_s = LDU$ be an ILU factorization of J_s and let $J_k = J_s + (J_k - J_s)$.

The ideal updated preconditioner

$$P_k^I = L(D + L^{-1}(J_k - J_s)U^{-1})U$$

- ▶ Simple approximations to $L^{-1}(J_k - J_s)U^{-1}$: extracts triangular parts of the current J_k :

$$P_k = L(DU + \text{triu}(J_k - J_s))$$

$$P_k = (LD + \text{tril}(J_k - J_s))U$$

- ▶ **Key Assumption:** L or $U \approx I$ (L^{-1} or U^{-1} neglected in P_k^I)



The Diagonally Updated ILU (DU_ILU)

We take the main diagonal of the matrix $J_k - J_s$:

$$P_k^l \simeq LDU + \underbrace{\text{diag}(J_k - J_s)}_{\Delta_k}$$

- The factorization of $LDU + \Delta_k$ is impractical: approximate factorization $LDU + \Delta_k \approx P_k = L_k D_k U_k$

$$D_k = D + \Delta_k, \quad L_k = \text{eye}(n), \quad \text{off}(L_k) = \text{off}(L) S_k$$

$$U_k = \text{eye}(n), \quad \text{off}(U_k) = S_k \text{off}(U)$$

$$S_k = \text{diag}(s_{11}^k, \dots, s_{nn}^k), \quad s_{ii}^k = \frac{|d_{ii}|}{|d_{ii}| + |\delta_{ii}^k|}, \quad i = 1, \dots, n,$$

Generalization to the unsymmetric case of the update preconditioning presented in [Bellavia, De Simone, di Serafino, Morini, 2010-2012]



Properties of DU_ILU

- ▶ The **sparsity pattern** of the factors of P_S is preserved
- ▶ The **cost** to form P_k is **low**;
- ▶ By the form of the diagonal scaling matrix S_k , P_k **mimics the behaviour** of $J_S + \Delta_k$

$$s_{ii}^k = \frac{|d_{ii}|}{|d_{ii}| + |\delta_{ii}^k|}, \quad i = 1, \dots, n,$$

- ▶ The **conditioning** of the matrices L_k and U_k is at least as good as the conditioning of L and U respectively.



The Banded Updated ILU (BU_ILU)

We take banded parts of the matrices $J_k - J_s, L^{-1}, U^{-1}$:

$$P_k^l \approx P_k = L [D + [L^{-1}]_\gamma [J_k - J_s]_\delta [U^{-1}]_\gamma]_\beta U$$

with $\delta, \gamma, \beta > 0$, small.

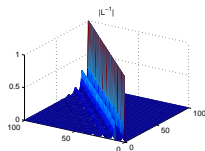
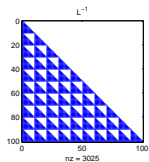
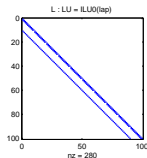
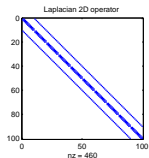
- ▶ We propose an algorithm for computing $[L^{-1}]_\gamma$ and $[U^{-1}]_\gamma$ exactly without the need of a complete inversion of L and U (suitable for parallel implementation).
- ▶ The application of P_k depends on β :
 - ▶ If $\beta = 0 \Rightarrow 1$ diagonal and 2 triangular systems must be solved.
 - ▶ Else, the **factorization of the banded middle factor** in P_k is necessary.

$[A]_\gamma$ denotes the banded matrix obtained extracting from A the main diagonal and γ upper and lower diagonals.



Motivation: matrices with decaying inverses

- ▶ banded spd and indefinite matrices [Demko et al. 1984][Meurant 1992];
- ▶ nonsymmetric block tridiagonal banded matrices [Nabben 1999];
- ▶ matrices $f(A)$ with A symm and banded and f analytic [Benzi, Golub 99].



Typically, the rate of decay of the entries of A^{-1} is fast if A is diag. dominant.



Approximation analysis of DU_ILU and BU_ILU

The quality of DU_ILU and BU_ILU measured as

$$\|J_k - P_k\|$$

depends on two terms:

- ▶ $\|J_s - P_s\|$, i.e. the quality of the seed preconditioner;
- ▶ $\|J_k - J_s\|$, i.e. distance of the current Jacobian from the seed Jacobian



$$P_s \approx J_s \text{ and } \{J_k\} \text{ is slowly varying} \Rightarrow P_k \approx J_k$$

[Bellavia, Morini, Porcelli 2013]



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Preconditioned Newton-Krylov method with linesearch: numerical comparison under Matlab

- ▶ 3 preconditioning strategies (DU_ILU , BU_ILU , TT) and the unpreconditioned (UP) case.
- ▶ Linear solver: BiCGSTAB, $LI_{\max} = 400$
- ▶ **Refresh**: when the backtracking strategy fails in producing an acceptable step a new reference matrix and preconditioner are initialized.
- ▶ **Finite difference approximation** for the computation of :
 - ▶ $J_s (= J_0)$
 - ▶ J_k times a vector
 - ▶ $[J_k]_{\delta}$ (and then stored)



Numerical comparison under Matlab c.ed

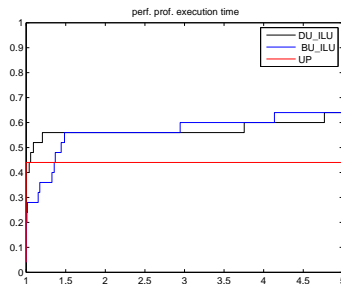
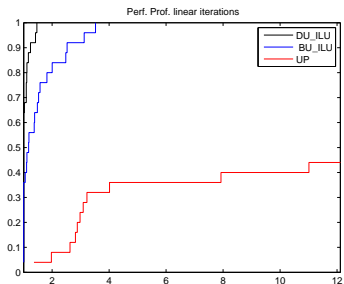
► Test Problems

- Nonlinear Convection-Diffusion (NCD), $Re = 750, 1000, 1250$
 - Flow in a Porous Medium (FPM),
 - CounterCurrent Reactor (CCR)
 - 2D Driven Cavity (2DC), $Re = 50, 100, 150$
- The dimension n of the nonlinear systems can be varied:
 $n = 4900, 8100, 10000, 15625, 22500$.
- The most time-consuming part of each updating strategy has been implemented as Fortran 90 mex-file with MATLAB interface.

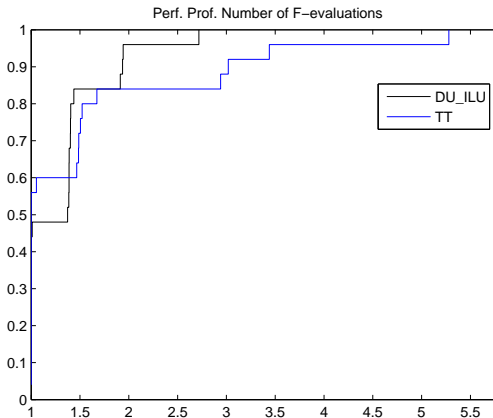


Comparison among DU_ILU , BU_ILU , UP

- ▶ Problems NCD, FPM, CCR: 25 runs
- ▶ BU_ILU strategy: $\delta = \beta = 0, \gamma = 20$



Comparison between DU_ILU and TT



Comparison between DU_ILU and BU_ILU

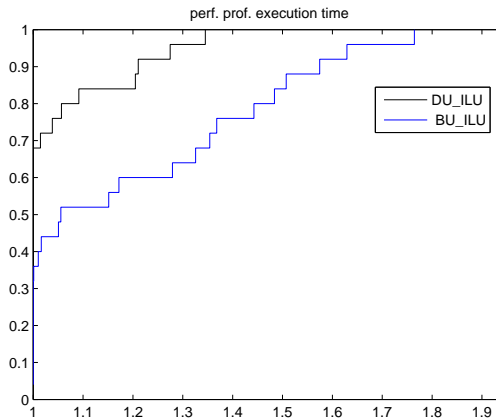


Figure: Performance profile in terms of execution time



2D Driven Cavity

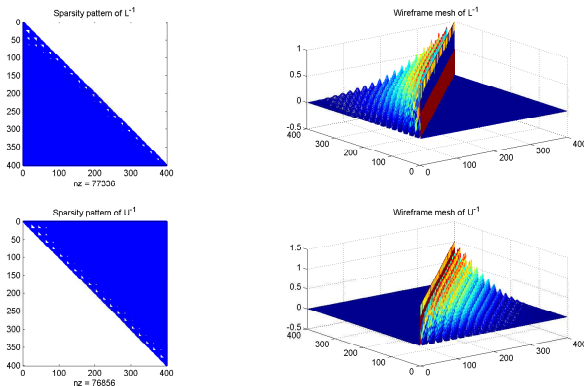


Figure: Sparsity pattern (on the left) and wireframe mesh (on the right) of the inverses of the L and U factors obtained from the ILU factorization of the Jacobian at the starting point ($n = 400$).



2D Driven Cavity c.ed

- ▶ The **Newton-Krylov method with UP** fails in all runs.
- ▶ The **DU_ILU** strategy is not applicable.
- ▶ The **Newton-Krylov method with TT** fails in 11 runs out of 15
 - ▶ Failures of preconditioned BiCGSTAB gave rise to poor descent directions and unsuccessful backtracking strategies
 - ▶ Preconditioner's refresh was not beneficial.
- ▶ The **Newton-Krylov method with BU_ILU** is successful in all runs and gives the best performance with $\gamma = 50$, $\delta = \beta = 1$: the tridiagonal part of J is used and a tridiagonal system has to be solved to apply the preconditioner.



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Final remarks

- ▶ The DU_ILU and BU_ILU preconditioners use matrices that are **reduced in complexity** with respect to the full J_k 's.
- ▶ The DU_ILU and BU_ILU preconditioners can be used for the solution of **general sequences of linear systems**

$$A_k x = b_k, \quad k = 1, 2, \dots$$

where $A_k \in \mathbb{R}^{n \times n}$ are large and sparse and $b_k \in \mathbb{R}^n$.

- ▶ Their applicability comprises the symmetric case since the updates preserve the symmetry of the preconditioners.
- ▶ They are **nearly matrix-free** whenever the function that performs the matrix-vector product is separable.



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Thanks for your attention

S. Bellavia, B. Morini, M. Porcelli, *New updates of incomplete LU factorizations and applications to large nonlinear systems*, Optimization Methods and Software (2013)

