
Aditi Ghai
**An Empirical Comparison of Krylov Subspace Methods
for Nonsymmetric Linear Systems from PDEs**

Department of Applied Math
Stony Brook University
Stony Brook
NY 11794
aditi.ghai@stonybrook.edu
Lu Cao
Jiao Xiangmin

Krylov subspace methods are widely used in solving partial differential equations and large sparse linear systems. For symmetric systems, CG [5] and MINRES [6] are almost always the optimal choices as efficient solvers. However, for nonsymmetric systems, such as those arising from convection-diffusion equations, the best choice of solvers is much less clearer. Various methods have been proposed for nonsymmetric systems over the years. Notable methods include GMRES [7], QMR [4], TFQMR [3], Bi-CGSTAB [8], QMRCGSTAB [2], etc. Unlike for symmetric case, there is no overall best method, and there is a need of guidelines in choosing these different types methods. The purpose of this study is to perform a systematic comparison of the various methods for linear systems arising from PDE discretizations. We focus on two classes of methods that are based on either nonsymmetric Lanczos iterations or Arnoldi iterations. We analyze the algorithm complexity and present numerical performances of these methods with and without preconditioners, for linear systems from finite difference or finite element discretizations in 2-D and 3-D.

In this study, we first report some theoretical comparisons of the methods with three-term recurrences, including QMR, TFQMR, Bi-CGSTAB and QMR-CGSTAB, as well as the methods based on n -term recurrence, such as GMRES. Table 1 compares the operation counts per iteration and the storage requirements for these methods, which augment the comparison in [1] with TFQMR and QMRCGSTAB. There is a tradeoff depending on the average number of nonzeros in the matrix versus the number of iterations. We present simple performance models for different classes of linear systems from PDE discretizations based on finite difference or finite element discretizations in 2-D and 3-D to provide practical guidelines in choosing these different Krylov subspace solvers.

To compliment the theoretical analysis, we also report some empirical comparisons of the different methods in terms of their convergence rates with and without preconditioners. We evaluate three preconditioners, including incomplete LU, Gauss Seidel, and Chebyshev polynomials. From the results, we

Table 1: Comparison of operations per iteration and memory requirements of various methods. m denotes the number of rows, n denotes the average number of nonzeros per row, and k denotes the iteration count. The cost for computing the residual norm is not included.

	Matrix-vector prods.	DAXPY	Inner prods.	FLOPs	Storage
QMR	2	12	2	$4m(n+7)$	$m(n+16)$
TFQMR	2	10	4	$4m(n+7)$	$m(n+8)$
Bi-CGSTAB	2	6	4	$4m(n+5)$	$m(n+10)$
QMRCGSTAB	2	8	6	$4m(n+7)$	$m(n+13)$
GMRES	1	$k+1$	$k+1$	$2m(n+2k+2)$	$m(n+k+5)$

observe that methods GMRES is more robust, with monotonically decreasing residuals, QMRCGSTAB, QMR, and TFQMR are less robust with nearly monotonically decreasing residual, and Bi-CGSTAB is the most efficient but also the least robust, with oscillatory residuals. For these different methods, we observe different accelerations with different preconditioners for different classes of problems. These results help provide some practical guidelines in choosing different preconditioners for different types of linear systems, and also motivate the development of hybrid solvers for sparse linear systems arising from PDE discretizations.

Bibliography

- [1] R. Barrett, M. Berry, T. F. Chan, J. Demmel, J. Donato, J. Dongarra, V. Eijkhout, R. Pozo, C. Romine, and H. Van der Vorst. *Templates for the Solution of Linear Systems: Building Blocks for Iterative Methods*. SIAM, Philadelphia, PA, 2nd edition, 1994.
- [2] T. F. Chan, E. Gallopoulos, V. Simoncini, T. Szeto, and C. H. Tong. A quasi-minimal residual variant of the Bi-CGSTAB algorithm for non symmetric systems. *SIAM J. Sci. Comput.*, 15(2):338–347, 1994.
- [3] R. W. Freund. A transpose-free quasi-minimal residual algorithm for non-hermitian linear systems. *SIAM J. Sci. Comput.*, 14(2):470–482, 1993.
- [4] R. W. Freund and N. M. Nachtigal. An implementation of the QMR method based on coupled two-term recurrences. *SIAM J. Sci. Comput.*, 15(2):313–337, 1994.
- [5] M. R. Hestenes and E. Stiefel. Methods of conjugate gradients for solving linear systems. *J. Res. Nat. Bur. Stand.*, 49(6), 1952.
- [6] C. C. Paige and M. A. Saunders. Solution of sparse indefinite systems of linear equations. *SIAM J. Numer. Anal.*, 12(4):617–629, 1975.
- [7] Y. Saad and M. H. Schultz. GMRES: A generalized minimal residual algorithm for solving non symmetric linear systems. *SIAM J. Sci. Comput.*, 7(3):856–869, 1986.
- [8] H. A. Van der Vorst. Bi-CGSTAB: A fast and smoothly convergent variant of Bi-CG for solution of non symmetric linear systems. *SIAM J. Sci. Comput.*, 13(2):631–644, 1992.