ML(n)BiCGStab: Reformulation, Analysis and Implementation^{*}

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> Abstract. With the aid of index functions, we re-derive the ML(n)BiCGStab algorithm in [Yeung and Chan, SIAM J. Sci. Comput., 21 (1999), pp. 1263-1290] systematically. There are *n* ways to define the ML(n)BiCGStab residual vector. Each definition leads to a different ML(n)BiCGStab algorithm. We demonstrate this by presenting a second algorithm which requires less storage. In theory, this second algorithm serves as a bridge that connects the Lanczos-based BiCGStab and the Arnoldi-based FOM while ML(n)BiCG is a bridge connecting BiCG and FOM. We also analyze the breakdown situation from the probabilistic point of view and summarize some useful properties of ML(n)BiCGStab. Implementation issues are also addressed.

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1. Introduction

Consider the solution of the linear system

$$\mathbf{A}\mathbf{x} = \mathbf{b},\tag{1.1}$$

where $\mathbf{A} \in \mathbb{C}^{N \times N}$ and $\mathbf{b} \in \mathbb{C}^{N}$. If we express the BiCG [4, 15] residual as $\mathbf{r}_{k}^{BiCG} = p_{k}(\mathbf{A})\mathbf{r}_{0}$ in terms of a polynomial $p_{k}(\lambda)$ of degree k and the initial residual \mathbf{r}_{0} , the residual vector \mathbf{r}_{k} of a Lanczos-type product method[†] based on BiCG is defined to be $\mathbf{r}_{k} = \phi_{k}(\mathbf{A})p_{k}(\mathbf{A})\mathbf{r}_{0}$, where $\phi_{k}(\lambda)$ is some polynomial of degree k with $\phi_{k}(0) = 1$. In CGS [28], $\phi_{k} = p_{k}$. Since, in every iteration, CGS searches for an approximate solution in a larger Krylov subspace, it

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[†]For this type of Krylov subspace methods, one can consult [9]. They are called hybrid BiCG methods in [27].

often converges much faster than BiCG. However, CGS usually behaves irregularly due to a lack of a smoothing mechanism. In BiCGStab [31], the ϕ_k is

$$\phi_k(\lambda) = \begin{cases} 1 & \text{if } k = 0, \\ (1 - \omega_k \lambda) \phi_{k-1}(\lambda) & \text{if } k > 0. \end{cases}$$
(1.2)

Here ω_k is a free parameter selected to minimize the 2-norm of $\mathbf{r}_k^{BiCGStab}$ in the *k*th iteration. As a result, BiCGStab is generally more stable and robust than CGS. BiCGStab has been extended to BiCGStab2 [7] and BiCGStab(*l*) [23,27] through the use of minimizing polynomials of higher degree. In BiCGStab2, the ϕ_k is defined by the recursion

$$\phi_k(\lambda) = \begin{cases} 1 & \text{if } k = 0, \\ (1 - \omega_k \lambda) \phi_{k-1}(\lambda) & \text{if } k \text{ is odd,} \\ ((\alpha_k \lambda + \beta_k)(1 - \omega_{k-1}\lambda) + 1 - \beta_k) \phi_{k-2}(\lambda) & \text{if } k \text{ is even.} \end{cases}$$

The parameters are again chosen to minimize BiCGStab2 residuals. Likewise, BiCGStab(*l*) defines its ϕ_k as

$$\phi_k(\lambda) = \begin{cases} 1 & \text{if } k = 0, \\ (1 + \sum_{j=1}^l \alpha_j \lambda^j) \phi_{k-l}(\lambda) & \text{if } k \text{ is a multiple of } l, \end{cases}$$

where the parameters in the factor $1 + \sum_{j=1}^{l} \alpha_j \lambda^j$ yields an *l*-dimensional minimization in every *l*th step. BiCGStab2 and BiCGStab(*l*) usually converge faster than BiCGStab because of smaller residuals in magnitude while avoiding near-breakdowns caused by a possibly too small ω_k . CGS, BiCGStab and BiCGStab2 have been summarized and generalized by GPBi-CG [40] where ϕ_k is

$$\phi_k(\lambda) = \begin{cases} 1 & \text{if } k = 0, \\ 1 - \omega_1 \lambda & \text{if } k = 1, \\ (1 + \beta_k - \omega_k \lambda) \phi_{k-1}(\lambda) - \beta_k \phi_{k-2}(\lambda) & \text{if } k > 1. \end{cases}$$

GPBi-CG will become CGS, BiCGStab or BiCGStab2 when the α, β, ω are appropriately chosen. For detailed descriptions of these and other product-type methods, one is referred to [6,8,20,22,32] and the references therein. Moreover, a history of product-type methods can be found in [10]. The history starts three decades ago with IDR [36] method which can be considered as the predecessor of CGS and BiCGStab [24]. Recently, IDR has been generalized to IDR(*s*) with a shadow space of higher dimension, see [24, 30, 34]. IDR(*s*) has close relations with ML(*s*)BiCGStab.

Generalizations of BiCGStab to methods based on the generalizations of BiCG have been made. For example, BL-BiCGStab [3] is a BiCGStab variant built on the BL-BiCG [16] for the solution of systems with multiple right-hand sides. ML(n)BiCGStab [39] is another BiCGStab variant built on ML(n)BiCG, a BiCG-like method derived from a variant of the band Lanczos process described in [1] with *n* left-starting vectors and a single right-starting vector.