

Liam Keegan  
liam@keegan.ch

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## 1 Introduction

This document contains a description of the implemented algorithms. They are all variants of CG inverters, where the inversion may include  $N_{\text{shifts}}$  shifts and act simultaneously on  $n_{\text{RHS}}$  vectors. Notation conventions used:

- $\mathbf{A}$  is the hermitian positive definite  $L \times L$  matrix to be inverted
- $x, r, p$ , etc. are  $L$ -component complex vectors
- $X, R, P$ , etc. are  $L \times n_{\text{RHS}}$  complex block-vectors
- $\alpha, \beta, \zeta$ , etc. are  $n_{\text{RHS}} \times n_{\text{RHS}}$  complex matrices
- $\sigma$  is a positive real scalar shift

The norm used here is the L2 norm, defined for a vector  $x$  as

$$|x| \equiv (x^\dagger x)^{1/2} \quad (1.1)$$

## 2 Solvers

### 2.1 CG

The standard CG solver for  $\mathbf{A}x = b$ , with initial guess  $x_0 = 0$  and criterion for convergence that the relative norm of the residual is smaller than some number  $\epsilon$ , i.e.  $|\mathbf{A}x_k - b|/|b| < \epsilon$  is given in Alg. 1.

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**Algorithm 1** CG: Solve  $\mathbf{A}x = b$

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1:  $x, p, r, t \in \mathcal{C}^L; \alpha, \beta \in \mathcal{R}$ 
2:  $x_0 = t_0 = 0, r_0 = p_0 = b$ 
3: for  $k = 1, 2, \dots$  until  $|r_k|/|r_0| < \epsilon$  do
4:    $t_k \leftarrow \mathbf{A}p_k$ 
5:    $\beta_k \leftarrow (p_k^\dagger t_k)^{-1}(r_{k-1}^\dagger r_{k-1})$ 
6:    $r_k \leftarrow r_{k-1} - t_k \beta_k$ 
7:    $\alpha_k \leftarrow (r_{k-1}^\dagger r_{k-1})^{-1}(r_k^\dagger r_k)$ 
8:    $x_k \leftarrow x_{k-1} + p_k \alpha_k$ 
9:    $p_k \leftarrow r_k + p_k \alpha_k$ 
10: end for
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## 2.2 SCG

Since the Krylov basis is shift-invariant, the residual  $r_k^\sigma$  of the solution of the shifted equation  $(\mathbf{A} + \sigma)x_k^\sigma = b$ , where  $\sigma$  is a real and positive scalar, can be related to the residual of the unshifted equation,  $r_k^\sigma = r_k \zeta_k$ . This allows the construction of a shifted CG (SCG) solver [1], which can solve multiple shifts without additional applications of the matrix  $\mathbf{A}$ , as described in Alg. 2. For the case  $N_{\text{shifts}} = 1$  this reduces to the CG solver of Alg. 1.

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**Algorithm 2** SCG: Solve  $(\mathbf{A} + \sigma_j)x^{\sigma_j} = b$ , where  $j = 0, \dots, N_{\text{shifts}} - 1$

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1:  $x^{\sigma_j}, p^{\sigma_j}, r, t \in \mathcal{C}^L$ ;  $\alpha, \beta, \zeta^{\sigma_j} \in \mathcal{R}$ 
2:  $x_0^{\sigma_j} = t_0 = 0, r_0 = p_0^{\sigma_j} = b; \alpha_0 = 0, \beta_0 = \zeta_{-1}^{\sigma_j} = \zeta_0^{\sigma_j} = 1$ 
3: for  $k = 1, 2, \dots$  until  $|r_k|/|r_0| < \epsilon$  do
4:    $t_k \leftarrow (\mathbf{A} + \sigma_0)p_k^{\sigma_0}$ 
5:    $\beta_k \leftarrow ((p_k^{\sigma_0})^\dagger t_k)^{-1}(r_{k-1}^\dagger r_{k-1})$ 
6:    $r_k \leftarrow r_{k-1} - t_k \beta_k$ 
7:    $\alpha_k \leftarrow (r_{k-1}^\dagger r_{k-1})^{-1}(r_k^\dagger r_k)$ 
8:    $x_k^{\sigma_0} \leftarrow x_{k-1}^{\sigma_0} + p_k^{\sigma_0} \beta_k$ 
9:    $p_k^{\sigma_0} \leftarrow r_k + p_k^{\sigma_0} \alpha_k$ 
10:  for  $j = 1, \dots, N_{\text{shifts}} - 1$  do
11:     $\zeta_k^{\sigma_j} \leftarrow \zeta_{k-1}^{\sigma_j} [1 + (\sigma_j - \sigma_0)\beta_k + \alpha_{k-1}(\beta_k/\beta_{k-1})(1 - \zeta_{k-1}^{\sigma_j}/\zeta_{k-2}^{\sigma_j})]^{-1}$ 
12:     $x_k^{\sigma_j} \leftarrow x_{k-1}^{\sigma_j} + p_k^{\sigma_j} \beta_k (\zeta_k^{\sigma_j}/\zeta_{k-1}^{\sigma_j})$ 
13:     $p_k^{\sigma_j} \leftarrow r_k \zeta_k^{\sigma_j} + p_k^{\sigma_j} \alpha_k (\zeta_k^{\sigma_j}/\zeta_{k-1}^{\sigma_j})^2$ 
14:  end for
15: end for
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## 3 Block Solvers

### 3.1 BCG

The BlockCG (BCG) [2] algorithm is a straightforward extension of the CG algorithm to multiple RHS vectors, described in Alg. 3.

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**Algorithm 3** [IN PROGRESS] BCG: Solve  $AX = B$

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1:  $X, P, R, T \in \mathcal{C}^{L \times n_{\text{RHS}}}$ ;  $\alpha, \beta \in \mathcal{C}^{n_{\text{RHS}} \times n_{\text{RHS}}}$ 
2:  $X_0 = T_0 = P_0 = 0, R_0 = B, \alpha_0 = B^\dagger B$ 
3: for  $k = 1, 2, \dots$  until  $|AX_k - B|/|B| < \epsilon$  do
4:    $P_k \leftarrow R_{k-1} + P_{k-1} \alpha_{k-1}$ 
5:    $T_k \leftarrow AP_k$ 
6:    $\beta_k \leftarrow (P_k^\dagger T_k)^{-1}(R_{k-1}^\dagger R_{k-1})$ 
7:    $R_k \leftarrow R_{k-1} - T_k \beta_k$ 
8:    $X_k \leftarrow X_{k-1} + P_k \beta_k$ 
9:    $\alpha_k \leftarrow (R_{k-1}^\dagger R_{k-1})^{-1}(R_k^\dagger R_k)$ 
10: end for
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### 3.2 BCGrQ

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**Algorithm 4** BCGrQ: Solve  $AX = B$ 

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1:  $X, P, R, T \in \mathcal{C}^{L \times n_{\text{RHS}}}$ ;  $\alpha, \beta, \delta \in \mathcal{C}^{n_{\text{RHS}} \times n_{\text{RHS}}}$ 
2:  $X_0 = T_0 = P_0 = 0$ ,  $\{R_0, \delta_0\} = \text{qr}(B)$ ,  $\alpha_0 = 1$ 
3: for  $k = 1, 2, \dots$  until  $\max_i \sqrt{\delta_k(i, i) / \delta_0(i, i)} < \epsilon$  do
4:    $P_k \leftarrow R_{k-1} + P_{k-1} \alpha_{k-1}^\dagger$ 
5:    $T_k \leftarrow AP_k$ 
6:    $\beta_k \leftarrow (P_k^\dagger T_k)^{-1}$ 
7:    $\{R_k, \alpha_k\} \leftarrow \text{qr}(R_{k-1} - T_k \beta_k)$ 
8:    $X_k \leftarrow X_{k-1} + P_k \beta_k \delta_{k-1}$ 
9:    $\delta_k \leftarrow \alpha_k \delta_{k-1}$ 
10: end for
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## References

- [1] B. Jegerlehner, *Krylov space solvers for shifted linear systems*, [hep-lat/9612014](#).
- [2] D. P. O’Leary, *The block conjugate gradient algorithm and related methods*, *Linear Algebra and its Applications* **29** (1980) 293 – 322. Special Volume Dedicated to Alson S. Householder.