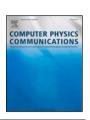
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$K\omega$ — Open-source library for the shifted Krylov subspace method of the form $(zI-H)x=b^{^{(+)},^{(+)}}$



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ABSTRACT

We develop $K\omega$, an open-source linear algebra library for the shifted Krylov subspace methods. The methods solve a set of shifted linear equations $(z_kI-H)\mathbf{x}^{(k)}=\mathbf{b}(k=0,1,2,\ldots)$ for a given matrix H and a vector \mathbf{b} , simultaneously. The leading order of the operational cost is the same as that for a single equation. The shift invariance of the Krylov subspace is the mathematical foundation of the shifted Krylov subspace methods. Applications in materials science are presented to demonstrate the advantages of the algorithm over the standard Krylov subspace methods such as the Lanczos method. We introduce benchmark calculations of (i) an excited (optical) spectrum and (ii) intermediate eigenvalues by the contour integral on the complex plane. In combination with the quantum lattice solver $\mathcal{H}\Phi$, $K\omega$ can realize parallel computation of excitation spectra and intermediate eigenvalues for various quantum lattice models.

Program summary

Program Title: $K\omega$ [kéi-óumig@]

CPC Library link to program files: https://dx.doi.org/10.17632/mt928nz5r3.1

Developer's repository link: https://github.com/issp-center-dev/Komega Licensing provisions: GNU Lesser General Public License Version 3.

Programming language: Fortran 90

External routines/libraries: BLAS library, LAPACK library (Used in the sample program), MPI library (Optional).

Nature of problem: Efficient algorithms, called shifted Krylov subspace algorithms, designed to solve the shifted linear equations.

Solution method: Shifted conjugate gradient method, Shifted conjugate orthogonal conjugate gradient method, shifted bi-conjugate gradient method.

Additional comments: The present paper is accompanied by a frozen copy of $K\omega$ release 2.0.0 that is made publicly available on GitHub (repository https://github.com/issp-center-dev/Komega, commit hash fd5455328b102ec4fa13432496e41c404a0f5a9d).

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1. Introduction

The response of physical observables to an external field or a perturbation is an essential probe in the experimental and theoretical studies of quantum many-body systems. Theoretically, such a quantity can be formulated as a selected element of the Green's function,

$$\mathbf{a}^{\dagger}G(z)\mathbf{b},$$
 (1)

with given states (or vectors) \mathbf{a} and \mathbf{b} , where z is the complex energy parameter and the Green's function G(z) is defined as the inverse of the shifted Hamiltonian H,

$$G(z) \equiv (zI - H)^{-1}. (2)$$

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The review of this paper was arranged by Prof. D.P. Landau.

常宜 This paper and its associated computer program are available via the Computer Physics Communication homepage on ScienceDirect (http://www.sciencedirect.com/science/journal/00104655).

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Table 1Classification of the Krylov subspace methods.

Type of $A + \sigma I$	Method	Solver name
Real symmetric	Shifted conjugate gradient (CG) method (using real vector)	komega_cg_r
Hermitian	Shifted conjugate gradient (CG) method (using complex vector)	komega_cg_c
Complex symmetric	Shifted conjugate orthogonal conjugate gradient (COCG) method	komega_cocg
Otherwise	Shifted bi-conjugate gradient (BiCG) method	komega_bicg

In computational quantum physics, the Hamiltonian H is usually represented by a real-symmetric or Hermitian $M \times M$ sparse matrix. A typical application of the Green's function formalism is calculating spectra in which the elements $\mathbf{a}^{\dagger}G(z_k)\mathbf{b}$ are calculated at sampling points z_k located near the real axis $(z_k \equiv \omega_k + \mathrm{i}\delta)$ with a tiny imaginary part δ or where the elements $\mathbf{a}^{\dagger}G(z_k)\mathbf{b}$ ($\mathbf{a} \neq \mathbf{b}$) are calculated at sampling points z_k for the numerical contour integral on the complex plane. The dimension M is large, e.g., $M \geq 10^9$, and efficient numerical methods are essential to treat such large matrices.

Formally, the Green's function (2) can be decomposed as

$$G(z) = \sum_{j} \frac{\mathbf{y}_{j} \mathbf{y}_{j}^{\dagger}}{z - \lambda_{j}} \tag{3}$$

where λ_j and \mathbf{y}_j are the *j*th eigenvalue and the corresponding eigenvector of the matrix H, respectively, i.e.,

$$H\mathbf{y}_{j} = \lambda_{j}\mathbf{y}_{j}. \tag{4}$$

In practice, however, the numerical evaluation of all the eigenpairs is too expensive [the operational cost is $O(M^3)$ and the memory cost is $O(M^2)$] to treat large-scale matrices ($M \geq 10^9$). In computational quantum physics, a Lanczos-based algorithm combined with the continued fraction technique has been proposed and widely used to calculate the Green's function [1,2]. The most expensive operation of the Lanczos-based algorithm is the matrix-vector product, with an operational cost of $O(M^2)$. Moreover, the sparseness of the Hamiltonian matrix reduces this cost to O(m) where m is the number of nonzero elements in the matrix. However, the calculation of the Green's function is restricted to diagonal components and it is not straightforward to evaluate the convergences of the obtained results.

In the present paper, we describe the numerical library $K\omega$ (https://github.com/issp-center-dev/Komega), which solves linear equations defined by

$$(z_k I - H) \mathbf{x}^{(k)} = \mathbf{b} \quad (k = 0, 1, 2, \dots, N_{eq} - 1),$$
 (5)

instead of Eq. (4). Here, \boldsymbol{b} is a given vector, $\{z_k\}$ are given scalars, I is the identity matrix, and $N_{\rm eq}$ is the number of linear equations. Eq. (5) is called the shifted linear equation since the matrix on the left hand side is different only by the 'shift' term, $z_k I$. In general, the scalar z is a complex number and the matrix of (zI-H) is non-Hermitian. The solution vector $\boldsymbol{x}^{(k)}$ is written as

$$\mathbf{x}^{(k)} = G(z_k)\mathbf{b},\tag{6}$$

and satisfies $\mathbf{a}^{\dagger}G(z_k)\mathbf{b} = \mathbf{a}^{\dagger}\mathbf{x}^{(k)}$. An iterative algorithm is adopted in $K\omega$, based on the property that the result of the nth iteration, regardless of z_k , can be spanned in the Krylov subspace. The operational cost is usually dominated by the matrix–vector multiplication procedure for a vector \mathbf{v} ($\mathbf{v} \Rightarrow H\mathbf{v}$).

The library $K\omega$ is based on efficient algorithms, called shifted Krylov subspace algorithms, designed to efficiently solve the shifted linear equations Eq. (5) with $N_{\rm eq} > 1$. The shifted Krylov subspace algorithms and related techniques are briefly explained in this paper and the details can be found in [3–17] and references therein. Currently, shifted Krylov subspace algorithms are used in many computational science fields such as quantum chromodynamics [3], electronic structure calculations [4,8,18,19], excited electron calculations [20], nuclear physics [21,22], transport

calculations with non-equilibrium Green's function theory [23], and nano-structured superconducting systems [24]. However, no open-source numerical library/solver of the shifted Krylov subspace methods has yet been developed to our knowledge.

In this paper, we first review the algorithm implemented in $K\omega$ briefly in Section 2. Next, the basic information such as installation and usage is introduced in Section 3. Then, some examples using $K\omega$ as a library or software are illustrated in Section 4. Finally, we summarize the paper in Section 5.

2. Algorithms

2.1. Overview

In this section, we briefly review the shifted Krylov subspace methods implemented in $K\omega$. The shifted linear equations of Eq. (5) can be rewritten as

$$(A + \sigma I)\mathbf{x}^{\sigma} = \mathbf{b} \tag{7}$$

with $A=z_0I-H$ and $\sigma\equiv z_k-z_0$ for $k=0,1,2,\ldots,N_{\rm eq}-1$. Here, the suffix k in Eq. (5) is dropped for simplicity. Hereafter, Eq. (7) with $\sigma=0$ is called the *seed equation*, while the other equations with $\sigma\neq 0$ are called the *shifted equations*. The accuracy of the approximate solution vector \mathbf{x}_n^σ can be checked by monitoring the residual vector

$$\mathbf{r}_n^{\sigma} \equiv \mathbf{b} - (A + \sigma I) \, \mathbf{x}_n^{\sigma}, \tag{8}$$

where n is the number of iteration steps.

Table 1 shows the four solver methods available in $K\omega$. The methods are classified according to the type of matrix $A + \sigma I$. Users need to select an appropriate method depending on whether the matrix H is real-symmetric or Hermitian and whether the shift constant values $\{z_k\}$ are complex or real [25].

The mathematical foundation of $K\omega$ is an iterative algorithm and the numerical solution of Eq. (7) at the nth iteration is obtained in the Krylov subspace defined as

$$K_n(A, \mathbf{b}) = \operatorname{span}[\mathbf{b}, A\mathbf{b}, A^2\mathbf{b}, \dots, A^{n-2}\mathbf{b}, A^{n-1}\mathbf{b}]. \tag{9}$$

The common mathematical foundation of the shifted Krylov subspace methods is the shift invariance property of the Krylov subspace

$$K_n(A + \sigma I, \mathbf{b}) = K_n(A, \mathbf{b}), \tag{10}$$

and the collinear residual theorem [3], which is explained later in this section.

This section explains the shifted conjugate gradient (CG) method in Table 1, as an example. The other methods, the shifted conjugate orthogonal CG method and the shifted BiCG method are explained in Appendices A and B, respectively.

2.2. Shifted conjugate gradient method

The shifted CG method is based on the CG method [26] that is used for solving $A\mathbf{x} = \mathbf{b}$, when the matrix A is real-symmetric (or Hermitian) and positive definite.

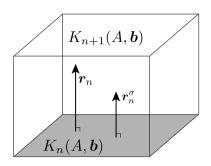


Fig. 1. Collinearity of residual vectors. The shaded bottom surface and the cube represent the *n*th Krylov subspace $K_n(A, \mathbf{b})$ and the (n + 1)th Krylov subspace $K_{n+1}(A, \mathbf{b})$, respectively. The two arrows depict the *n*th residuals, \mathbf{r}_n and \mathbf{r}_n^{σ} .

2.2.1. Seed equation

The seed equation [Eq. (7) with $\sigma = 0$] is denoted as $A\mathbf{x} = \mathbf{b}$ and the solution vector at the nth iterative step is denoted as \mathbf{x}_n , where \mathbf{x}_0 is set to be $\mathbf{x}_0 = \mathbf{0}$. The residual vector at the nth step is denoted as $\mathbf{r}_n = \mathbf{b} - A\mathbf{x}_n$. The three vectors, \mathbf{x}_n , \mathbf{r}_n , and \mathbf{p}_n , are computed by the following recurrence equations:

$$\rho_n = \mathbf{r}_n^{\dagger} \mathbf{r}_n \tag{11}$$

$$\alpha_n = \frac{\rho_n}{\mathbf{p}_n^{\dagger} A \mathbf{p}_n},\tag{12}$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + \alpha_n \mathbf{p}_n, \tag{13}$$

$$\boldsymbol{r}_{n+1} = \boldsymbol{r}_n - \alpha_n A \boldsymbol{p}_n, \tag{14}$$

$$\beta_n = \frac{\rho_{n+1}}{\rho_n},\tag{15}$$

$$\boldsymbol{p}_{n+1} = \boldsymbol{r}_{n+1} + \beta_n \boldsymbol{p}_n \tag{16}$$

with the initialization $\mathbf{p}_0 = \mathbf{r}_0 = \mathbf{b}$.

Kω uses the three-term recurrence formula [7]

$$\boldsymbol{r}_{n+1} = \left(1 + \frac{\alpha_n \beta_{n-1}}{\alpha_{n-1}} - \alpha_n A\right) \boldsymbol{r}_n - \frac{\alpha_n \beta_{n-1}}{\alpha_{n-1}} \boldsymbol{r}_{n-1},\tag{17}$$

which is obtained by eliminating \mathbf{p}_n from Eqs. (14) and (16). By taking the inner product between \mathbf{r}_n and Eq. (17), we also obtain

$$\alpha_n = \frac{\rho_n}{\mathbf{r}_n^{\dagger} A \mathbf{r}_n - \frac{\beta_{n-1}}{\alpha_{n-1}} \rho_n}.$$
 (18)

These two recurrence equations start with $\beta_{-1}/\alpha_{-1}=0$ because $\alpha_0=\rho_0/({\pmb p}_0^\dagger A {\pmb p}_0)=\rho_0/({\pmb r}_0^\dagger A {\pmb r}_0)$. The reason for this implementation style will be presented in the following subsubsections.

It is crucial that the *n*th residual \mathbf{r}_n belongs to the (n+1)th Krylov subspace $K_{n+1}(A, \mathbf{b})$ and is orthogonal to the *n*th Krylov subspace $K_n(A, \mathbf{b})$,

$$\boldsymbol{r}_n \in K_{n+1}(A, \boldsymbol{b}) \quad \text{and} \quad \boldsymbol{r}_n \perp K_n(A, \boldsymbol{b}).$$
 (19)

These mean that \mathbf{r}_n belongs to the one-dimensional orthogonal complementary space of $K_n(A, \mathbf{b})$ within $K_{n+1}(A, \mathbf{b})$, say $K_n^{\perp}(A, \mathbf{b})$.

2.2.2. Shifted equation and collinear residuals

When the CG method is applied to a shifted equation $((A + \sigma I) \mathbf{x} = \mathbf{b}, \sigma \neq 0)$ with $\mathbf{x}_0^{\sigma} = \mathbf{0}$ and $\mathbf{r}_0 = \mathbf{b}$, the residual vector \mathbf{r}_n^{σ} belongs to the same one-dimensional subspace $K_n^{\perp}(A, \mathbf{b})$ because of Eq. (10). Consequently, the residual \mathbf{r}_n^{σ} is parallel to that of the seed equation (\mathbf{r}_n)

$$\boldsymbol{r}_{n}^{\sigma} = \frac{1}{\pi_{n}^{\sigma}} \boldsymbol{r}_{n}, \tag{20}$$

where π_n^{σ} is a constant [3] called the collinearity factor. Fig. 1 is a schematic figure depicting the collinearity $(\mathbf{r}_n^{\sigma} \parallel \mathbf{r}_n)$.

Table 2

The operational costs per iteration are listed for (i) the conventional Krylov method, (ii) the shifted Krylov subspace method for the whole elements of the solution vectors, and (iii) the shifted Krylov subspace method for the projection of the solution vectors [4]. The operational costs of the sparse matrix–vector products (SpMV), the scalar–vector products (SV), and the inner products (Inner) are listed separately.

Method	SpMV	SV	Inner
Conventional	$MM_{\rm NZ}N_{\rm eq}$	3MN _{eq}	$3MN_{\rm eq}$
Shift (whole)	MM_{NZ}	$3MN_{\rm eq}$	3 <i>M</i>
Shift (projection)	MM_{NZ}	$3M + 3M_{\text{left}}(N_{\text{eq}} - 1)$	3 <i>M</i>

The collinearity factor π_n^{σ} satisfies the recurrence equation

$$\pi_{n+1}^{\sigma} = \left(1 + \frac{\alpha_n \beta_{n-1}}{\alpha_{n-1}} + \alpha_n \sigma\right) \pi_n^{\sigma} - \frac{\alpha_n \beta_{n-1}}{\alpha_{n-1}} \pi_{n-1}^{\sigma} \tag{21}$$

with the initialization $\pi_0^{\sigma}=\pi_{-1}^{\sigma}=1$, because of Eqs. (17) and (20). By substituting $\pi_n^{\sigma} {\bf r}_n^{\sigma}$ into ${\bf r}_n$ in Eq. (17), the recurrence equations for the shifted equations are derived as

$$\alpha_n^{\sigma} = \frac{\pi_n^{\sigma}}{\pi_{n+1}^{\sigma}} \alpha_n,\tag{22}$$

$$\beta_n^{\sigma} = \left(\frac{\pi_n^{\sigma}}{\pi_{n+1}^{\sigma}}\right)^2 \beta_n,\tag{23}$$

$$\mathbf{x}_{n+1}^{\sigma} = \mathbf{x}_{n}^{\sigma} + \alpha_{n}^{\sigma} \mathbf{p}_{n}^{\sigma}, \tag{24}$$

$$\boldsymbol{p}_{n+1}^{\sigma} = \frac{1}{\pi_{n+1}^{\sigma}} \boldsymbol{r}_{n+1} + \beta_n^{\sigma} \boldsymbol{p}_n^{\sigma}$$
 (25)

with $\mathbf{x}_0^{\sigma} = \mathbf{0}$ and $\mathbf{p}_0^{\sigma} = \mathbf{b}$. It should be noted that the recurrences of Eqs. (21), (22), (23), (24), and (25) require no expensive matrix–vector multiplication.

2.2.3. Cost and projection

Table 2 compares the operational costs of the conventional Krylov method and the shifted Krylov subspace methods [4]. The first row of Table 2 represents the conventional Krylov method, which solves the N_{eq} linear equations independently in each Krylov subspace. Here M and $M_{\rm NZ}$ are the dimension and the average number of non-zero elements per column of matrix A, respectively ($M_{NZ} \leq M$). The second row represents the cost of the shifted Krylov subspace method for the case where all the elements of the solution vectors \mathbf{x}^{σ} are calculated. We find that the operational cost for the sparse matrix-vector product (SpMV) is drastically reduced ($MM_{NZ}N_{eq} \rightarrow MM_{NZ}$), since the explicit SpMV appears only in the seed equation ($\sigma = 0$). The third row represents the cost of the shifted Krylov subspace method for the case where we do not need all the elements of the solution \mathbf{x}^{σ} , but only its projection, $\mathbf{y}^{\sigma} = P\mathbf{x}^{\sigma}$ with the $M_{\text{left}} \times M$ projection matrix P (1 $\leq M_{\text{left}} \leq M$). In the third case, we can replace the recurrence equations, Eqs. (24) and (25), by

$$\boldsymbol{y}_{n+1}^{\sigma} = \boldsymbol{y}_{n}^{\sigma} + \alpha_{n}^{\sigma} \boldsymbol{u}_{n} \tag{26}$$

$$\mathbf{u}_{n+1}^{\sigma} = \frac{1}{\pi_{n}^{\sigma}} P \mathbf{r}_{n} + \beta_{n}^{\sigma} \mathbf{u}_{n}^{\sigma}$$
 (27)

with $y_0^{\sigma} = \mathbf{0}$ and $\mathbf{u}_0^{\sigma} = P\mathbf{b}$. By this replacement, the number of scalar-vector products (SV) is reduced from $3MN_{\rm eq}$ to $3M + 3M_{\rm left}(N_{\rm eq} - 1)$. For example, the calculation of an element of the Green's function by Eq. (1)

$$G_{ab}(z) \equiv \boldsymbol{a}^{\dagger} (zI - H)^{-1} \boldsymbol{b} = \boldsymbol{a}^{\dagger} \boldsymbol{x}(z)$$
 (28)

is a case with $M_{\rm left}=1$. In typical applications, both $N_{\rm eq}$ and $M_{\rm left}$ are much smaller than M ($N_{\rm eq}$, $M_{\rm left}\ll M$), and thus the operational costs in the third row in Table 2 are reduced to those in the first row with $N_{\rm eq}=1$. In other words, the operational cost

in the third case is reduced, typically, to that for solving a single linear equation.

Kω implements the third case in Table 2, i.e., the shifted Krylov subspace method with projection. For the seed equation, the residual vector \mathbf{r}_n is updated by Eq. (17) and the coefficients ρ_n , α_n , and β_n are updated by Eqs. (11), (18), and (15), respectively. For the shifted equations, the projected solution vector \mathbf{y}_n^{σ} and the projected search direction vector \mathbf{u}_n^{σ} are updated by Eqs. (26) and (27), respectively, and the coefficients π_n^{σ} , α_n^{σ} , and β_n^{σ} are updated by Eqs. (21), (22), and (23), respectively. The solution vectors \mathbf{x}_n^{σ} for the shifted equations can be obtained by setting P = I or $M_{\text{left}} = M$, but users should accept the additional memory cost, as explained below.

The present implementation of $K\omega$ offers a great advantage not only in the operational cost but also in the memory cost, because an M-dimensional vector \boldsymbol{v} requires a large memory cost in typical applications. In the present implementation, only three M-dimensional vectors, the residual vectors for the seed equation $(\boldsymbol{r}_{n+1}, \boldsymbol{r}_n, \text{ and } \boldsymbol{r}_{n-1})$, are stored in the memory with a memory cost of $O(M^1M_{\text{eq}}^0)$. The residual vectors for the shifted equations, such as $\boldsymbol{r}_n^{\sigma}$, can be obtained by Eq. (20) with a negligible memory cost of $O(M^0M_{\text{eq}}^1)$. To store the other M-dimensional vectors for the shifted equations, such as $\boldsymbol{x}_n^{\sigma}$ and $\boldsymbol{p}_n^{\sigma}$, the required memory size will be $O(M^1M_{\text{eq}}^1)$, which can be huge.

2.2.4. Seed switching

A mathematical technique called seed switching [7] is adopted in $K\omega$ for an efficient convergence, because the convergence speed of the CG method can be different among the energy points $\{\sigma_k\}$. The residual vectors for several shifted equations (\mathbf{r}_n^σ , $\sigma \neq 0$) can sometimes remain large, while that for the seed equation is reduced to be negligible [4]. In this case, we can switch the seed equation as $A' = A + \sigma_{\text{seed}}I$, where σ_{seed} is the shift that gives the largest residual $|\mathbf{r}_n^{\sigma_{\text{seed}}}|$, or in other words, the smallest collinearity factor $|\pi_n^{\sigma_{\text{seed}}}|$. The residual vector for the new seed equation is obtained by Eq. (20) to be $\mathbf{r}_n^{\sigma_{\text{seed}}} = (1/\pi_n^{\sigma_{\text{seed}}})\mathbf{r}_n$. It is noted that the present implementation style does not require the solution and search direction vectors for the new seed equation ($\mathbf{x}_n^{\sigma_{\text{seed}}}$, $\mathbf{p}_n^{\sigma_{\text{seed}}}$). $K\omega$ always performs the seed switching after an update.

2.3. Comparison with other methods

Here, we briefly describe the merits of calculating $G_{ab}(z)$ by the shifted Krylov subspace method compared to the traditional way based on the Lanczos-based algorithm [1,2]. In the Lanczos-based algorithm, the Krylov subspace of $K_n(A, \mathbf{b})$ is generated by the Lanczos-type recurrence formula and the *diagonal* component G_{bb} is given by a continued fraction form. In contrast, using the shifted Krylov subspace method, both the *diagonal* components G_{bb} and the *off-diagonal* components G_{ab} can be calculated directly and simultaneously with the same order of operational cost as that of the Lanczos-based method. In addition, the accuracy of the obtained results can be evaluated by monitoring the residual vector of Eq. (8).

The characteristics of the shifted Krylov subspace method are more clarified when compared with (i) independent computation by the standard Krylov subspace method at each single shift point ('conventional method' in Table 2) on a massively parallel supercomputer and (ii) use of sparse-direct algorithms as in the PEXSI solver [27]. As an advantage of the present method, it meets the demand for extremely large problems, such as quantum lattice models with matrix size $M=2^{24}\sim 2^{32}$. Such problems are severe not only in the operational cost but also in the memory cost. In the case of $M=2^{32}\approx 4.3\times 10^9$, for example, the present method uses three double-precision complex vectors in the recurrence relations and thus requires

memory of 3×16 B \times $M \approx 200$ GB, while the others require much larger memory cost. As a limitation of the present $K\omega$ code, on the other hand, the code is not applicable to the generalized shifted equations (zB - H)x = b with a positive-definite matrix B, unlike the two methods. The generalized shifted equations appear in many physical problems like the electronic state calculation with the overlap matrix B. Some of the present authors proposed the shifted Krylov-subspace algorithms for the generalized shifted equations [8,11], and the implementation of the generalized algorithm in $K\omega$ remains as a future issue.

3. Usage of $K\omega$

In this section, the usage of $K\omega$ is introduced. We first introduce how to install $K\omega$. $K\omega$ provides libraries and a standalone program ShiftK.out. In Section 3.2, the procedures for using $K\omega$ are schematically shown.

3.1. Installation

The stable version of $K\omega$ is distributed on the release page with the Lesser General Public License (LGPL) version 3. To build $K\omega$ a Fortran compiler and a BLAS library are required. The following is an example of compiling $K\omega$:

```
$ ./configure --prefix=install_dir
$ make
$ make install
```

Here, install_dir indicates the full path of the directory where the library is installed.

3.2. Schematic flow of $K\omega$ usage

In this subsection, the usage of $K\omega$ as a library or a standalone program is explained. In Fig. 2, the schematic flow of the library usage (a) and the corresponding flowchart in the standalone program (b) are shown.

3.2.1. Library

(i) Preparation of a routine for matrix-vector multiplication

 $K\omega$ provides a reverse communication interface for the matrix-vector multiplication routine ($v \Rightarrow Hv$). The interface requires the preparation of a routine to perform the matrix-vector multiplication, to be called in $K\omega$. This interface allows extremely large matrices to be handled, since the matrix elements are internally generated in the 'matrix-vector multiplication' and not stored in the memory.

(ii) Selection of an appropriate solver

 $K\omega$ provides four kinds of numerical solvers. An appropriate solver should be selected depending on whether the type of Hamiltonian H and the frequency z are complex or real, as shown in Table 1. It is noted that H must be Hermitian or symmetric for a complex or real matrix, respectively. For an efficient calculation, the seed switching function [7] is introduced in all methods.

(iii) Calculation

The calculation is performed in the following steps:

- (a) Initialization using *_init functions to set and initialize internal variables in the library. For restart calculations, the initial values of the coefficients and the vector should be inputted at this step.
- (b) Updating of the results iteratively using *_update functions, which are called alternately with the matrix-vector product routine in the loop to update the solution. *_update also performs the seed switching. At each step, the values of the 2-norm

(a) Schematic flow of using library

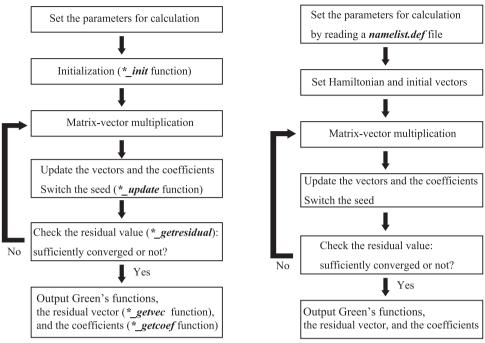


Fig. 2. (a) Schematic flow of use of library and (b) calculation flow in shiftK.out. Here, * indicates the name of the solver, such as komega_cg_r, komega_cg_c, komega_cocg, and komega_bicg.

of the residual vector at each shift point are obtained using *_getresidual functions.

(c) Finalization using *_finalize functions to release the memories of the arrays stored in the library.

Here, * indicates the name of the solver such as komega_cg_r, komega_cg_c, komega_cocg, and komega_bicg.

We summarize this subsection by showing a pseudo code in Algorithm 1, which solves $(zI-H)\mathbf{x}=\mathbf{b}$ by the shifted BiCG method (komega_bicg_*) implemented in K ω . Users can use the shifted CG method (komega_cg_*) or the shifted COCG method (komega_cocg_*) in a similar way (they do not require the shadow vectors, $\tilde{\mathbf{r}}$ and $\tilde{\mathbf{q}}$.) The application using the library is shown in Section 4.1.

Algorithm 1 Solve $(zI - H)\mathbf{x} = \mathbf{b}$ by Shifted BiCG method with \mathbf{K}_{ω}

```
procedure SHIFTEDBICG(\boldsymbol{x}, H, \{z_i\}, \boldsymbol{b})
    ndim \leftarrow dim(H)
    nz \leftarrow dim(\{z_i\})
                                                           ▶ # of frequencies
    x \leftarrow 0

    initialize output

    allocate r[ndim]
                                                             r \leftarrow b
    allocate \tilde{r}[ndim]

⊳ shadow residual vector

    \tilde{r} \leftarrow b^{\dagger}
    allocate q[ndim], \tilde{q}[ndim]
                                                        \triangleright storing Hr and H\tilde{r}
    call komega_bicg_init(ndim, nz, ndim, x, \{z_i\})
    repeat
        q = Hr
        \tilde{q} = H\tilde{r}
        call komega_bicg_update(q, r, \tilde{q}, \tilde{r}, x) \Rightarrow update and
seed switching
    until converged
    call komega_bicg_finilize()
    deallocate r, \tilde{r}, q, \tilde{q}
end procedure
```

3.2.2. Standalone program

(b) Schematic calculation flow in shiftK.out

As described in the Introduction, Green's functions are often calculated to investigate the response of physical observables in the quantum many-body systems. In $K\omega$ a simple standalone program ShiftK.out is provided, which computes the diagonal element of the Green's function:

$$G_{aa}(z) = \boldsymbol{a}^{\dagger} (zI - H)^{-1} \boldsymbol{a}. \tag{29}$$

Here, \boldsymbol{a} is expanded as $\sum_i a_i \boldsymbol{n}_i$ where \boldsymbol{n}_i is the ith basis vector of the Hilbert space. An off-diagonal element G_{ab} can be obtained by ShiftK.out, when one calculates G_{aa} , G_{bb} , G_{cc} , G_{dd} ($\boldsymbol{c} \equiv \boldsymbol{a} + \boldsymbol{b}$, $\boldsymbol{d} \equiv \boldsymbol{a} + i\boldsymbol{b}$), and uses the relation $G_{ab} = [(G_{cc} - G_{aa} - G_{bb}) + i(G_{dd} - G_{aa} - G_{bb})]/2$. [1] In the following, the usage of ShiftK.out is explained.

(i) Preparation of an input file

The input parameters for ShiftK.out are categorized into four sections: cg, dyn, filename, and ham.

The cg section sets the numerical condition for the CG (or COCG, BiCG) method. maxloops is the maximum number of iterations and convfactor is the threshold value for the convergence criterion of the residual norm, i.e., $\max_{\sigma} |\boldsymbol{r}_n^{\sigma}| < 10^{-\text{convfactor}}$.

The dyn section specifies the parameters for computing the spectrum. By setting the parameters omegamin, omegamax, and nomega, the target frequencies are given as $\omega_i = \text{omegamin} + i \times (\text{omegamax} - \text{omegamin})/\text{nomega}, (i = 0, \dots \text{nomega} - 1).$ In ShiftK.out, the calculation mode is chosen from normal, recalc, and restart by specifying the parameter calctype. normal is the mode for computing with the Krylov subspace from scratch. recalc is the mode for computing with the Krylov subspace generated in the previous calculation (See (iv) for details). restart is the mode for restarting the calculation from the previous run. In the computation of the Green's function, the shifted BiCG or shifted COCG method is automatically selected when H is real or complex, respectively.

(a) Input of the matrix of the Hamiltonian

The filename section specifies the name of the input files for the matrix of the Hamiltonian inham or the initial excited vector invec. The file format of both files is the Matrix Market format [28]. If invec is not specified, a random vector is used as the initial vector. An example input file for reading the Hamiltonian matrix and excited vector is as follows:

(b) Generation of the matrix of the Hamiltonian

For the trial use, the matrix of the Hamiltonian is generated in ShiftK.out mode. In this mode, the ham section is used instead of inham in the filename section. In the ham section, model parameters are specified to generate the Hamiltonian matrix for a one-dimensional spin chain model:

$$H = \sum_{i=1}^{\text{nsite}} \left(\begin{array}{ccc} S_x^{(i)} & S_y^{(i)} & S_z^{(i)} \end{array} \right) \left(\begin{array}{ccc} Jx & Dz & 0 \\ -Dz & Jy & 0 \\ 0 & 0 & Jz \end{array} \right) \left(\begin{array}{c} S_x^{(i+1)} \\ S_y^{(i+1)} \\ S_z^{(i+1)} \end{array} \right),$$
(30)

where $S_j^{(i)}$ is a spin-1/2 operator at a site i with component j=x,y,z. In the ham section, the following parameters are specified: the total site of the spin chain nsite and the parameters of the strength for spin-spin interactions Jx, Jy, Jz, and Dz. An example of using the mode to generate internally the Hamiltonian matrix is as follows:

```
&filename
/
&ham
Jx = 1d0
Jy = 1d0
Jz = 1d0
Dz = 1d0
/
&cg
   maxloops = 100
   convfactor = 6
/
&dyn
   calctype = "normal"
   nomega = 100
   outrestart = .TRUE.
/
```

(ii) Run

After preparing the input file, an executable ShiftK.out in terminal is run as follows:

\$ ShiftK.out namelist.def

Here, namelist.def is the name of the input file. The residual values at each step are output to the residual.dat file in the working directory.

(iii) Results

After running ShiftK.out, the output directory is automatically generated. In this directory, dynamical Green's functions, the residual vector, and the coefficients are output to dynamicalG.dat, ResVec.datO, and TriDiagComp.dat, respectively.

(iv) Recalculation for additional data (optional)

The standalone program ShiftK.out provides, as an optional function, a recalculation function for additional data, as explained below. After the successful completion of ShiftK.out, users might like to obtain the solution at additional points σ that are not obtained in the completed calculation. In such cases, users can calculate the solution at these points for the shifted equations with negligible operational costs, because the coefficients of $\{\alpha_n, \beta_n\}$ and the projected residual vectors $\{Pr_n\}$ are saved in the files TriDiagComp.dat and ResVec.dat, respectively, and the recurrence relations for the shifted equation Eqs. (21), (22), (23), (26), and (27) can be solved without any expensive matrix–vector multiplications.

4. Applications in material science

In this section, we present several numerical results of $K\omega$ applied to quantum lattice models, to demonstrate typical applications in computational quantum physics. The quantum lattice model plays a crucial role for quantum many-body systems and is described by a large sparse matrix H. The applied studies in this section use the quantum lattice solver package $\mathcal{H}\Phi$ [29]. $\mathcal{H}\Phi$ was developed as an exact diagonalization solver and can treat a wide range of quantum lattice models, such as the Hubbard model, the Kondo model, and the Heisenberg model. $K\omega$ can be called from $\mathcal{H}\Phi$ and the users of $\mathcal{H}\Phi$ can use the shifted Krylov subspace solvers. Here, three typical examples are explained.

4.1. Calculation of internal eigenpairs

The first example is the calculation of internal eigenvalues by a contour-integral method [30]. The method is an efficient method for obtaining eigenpairs (eigenvalues and eigenvectors) in a specified eigenvalue range. Although a number of software packages for this method have already been developed (for example, zPares [31] and FEAST [32,33]), this example illustrates how to use $K\omega$ in the contour-integral method.

Here, we briefly explain how to obtain eigenpairs by using the method following Ref. [34]. In the contour-integral method, the projection matrix on the *n*th eigenvector,

$$P_n = \mathbf{y}_n \mathbf{y}_n^{\dagger}, \tag{31}$$

plays a key role. By multiplying P_n by a vector $\phi = \sum_n a_n y_n$, we can extract the component of y_n as

$$P_n \boldsymbol{\phi} = a_n \boldsymbol{y}_n. \tag{32}$$

On the other hand, the projection matrix P_n can be expressed by an integration in the complex plane as

$$P_{\Gamma} = \sum_{\lambda \in \Gamma} P_n = \frac{1}{2\pi i} \oint_{\Gamma} \frac{1}{zI - H} dz, \tag{33}$$

where Γ represents a contour on the complex plain.

Using P_{Γ} , we can extract only the eigenvectors whose eigenvalues exist within Γ as

$$\mathbf{s}_{0,0} = P_{\Gamma} \boldsymbol{\phi_0} = \frac{1}{2\pi i} \oint_{\Gamma} \frac{1}{zI - H} \boldsymbol{\phi_0} dz$$

$$= \frac{1}{2\pi i} \oint_{\Gamma} \boldsymbol{\phi_0'} dz = \sum_{\lambda_n \in \Gamma} a_n \boldsymbol{y}_n.$$
(34)

In this process, we can use $K\omega$ to obtain $\phi'_0 = (zI - H)^{-1}\phi_0$. The kth moment can be simply expressed as follows:

$$\mathbf{s}_{k,0} = (H - z_0 I)^k P_{\Gamma} \phi_0 = \frac{1}{2\pi i} \int_{\Gamma} \frac{(z - z_0)^k}{zI - H} \phi_0 dz, \tag{35}$$

where z_0 is an arbitrary complex number. Thus, once we obtain $\phi_0' = (zl - H)^{-1}\phi_0$, we can calculate the Krylov subspace defined as

$$K_{N_{\nu} \times 1} = \text{span}[\mathbf{s}_{0,0}, \mathbf{s}_{1,0}, \dots, \mathbf{s}_{N_{\nu}-1,0}],$$
 (36)

where N_k represent the number of moments. Taking other vectors $\phi_l(l=1\cdots N_l-1)$, we can extend the Krylov subspace as

$$K_{N_k \times N_l} = \text{span}[\mathbf{s}_{0,0}, \dots, \mathbf{s}_{N_k-1,0}; \\ \mathbf{s}_{0,1}, \dots, \mathbf{s}_{N_k-1,1}; \dots \\ \mathbf{s}_{0,N_l-1}, \dots, \mathbf{s}_{N_k-1,N_l-1}].$$
 (37)

Then, the matrix S with vectors $\mathbf{s}_{k,l}$ is constructed as

$$S = (\mathbf{s}_{0,0}, \dots, \mathbf{s}_{N_k-1,0}, \\ \mathbf{s}_{0,1}, \dots, \mathbf{s}_{N_k-1,1}, \dots \\ \mathbf{s}_{0,N_l-1}, \dots, \mathbf{s}_{N_k-1,N_l-1}).$$
(38)

By performing singular value decomposition for S, we obtain

$$S = U \Sigma V^{\dagger}. \tag{39}$$

The number of non-zero singular values in Σ is just that of the independent vectors $(M_{\rm nz})$ in $N_{N_k \times N_l}$ spanned by $\mathbf{s}_{k,l}$. By using only the left-singular vectors with non-zero singular values, we can construct the matrix \tilde{U} ,

$$\tilde{U} = (\boldsymbol{u}_0, \dots, \boldsymbol{u}_{M_{nz}-1}). \tag{40}$$

Using \tilde{U} , we project the original Hamiltonian on the small matrix (dimension $M_{\rm nz} \times M_{\rm nz}$) whose eigenvalues exist within Γ as

$$\tilde{H} = \tilde{U}^{\dagger} H \tilde{U}. \tag{41}$$

By diagonalizing \bar{H} , we can obtain all the eigenvalues and eigenvectors which exist within Γ .

As an example, we apply the contour-integral method to the 12-site Heisenberg chain model defined in Eq. (45). For simplicity, we obtain the eigenvalues around the ground state (the lowest eigenvalue). To perform the integration on the complex plane, we use the following points:

$$z_{j} = \gamma + \rho \exp\left[\frac{2\pi i}{N_{z}} \left(j + \frac{1}{2}\right)\right], \tag{42}$$

where we take $\gamma = -5$, $\rho = 0.8$, and $N_z = 100$ as an example.

In Table 3, we show the result of the method for the L=12 Heisenberg chain, whose Hilbert dimension is 924. To see the convergence of the contour-integral method, we show the obtained eigenvalues for several different conditions. By taking about 50 different bases, we can obtain the correct eigenvalues including degeneracies. To perform the above calculations, the matrix data file in the Matrix Market format (Ham.dat) is generated using $\mathcal{H}\Phi$ as in Appendix C. Using Ham.dat, we can perform the contour-integral method by $K\omega$. Two sample codes are provided in the web site [35]: One is a simple code for obtaining ϕ_0' in Eq. (34) by BiCG method (see Algorithm 1) and the other is for performing the contour-integral method.

4.2. Calculation of spectra using ShiftK.out

The second example is a spectral calculation of Eq. (29) by the standalone program ShiftK.out explained in Section 3.2.2. A typical application for a quantum lattice model is the calculation of the dynamical correlation factor, since it is often used to

Table 3

Eigenvalues obtained by $\mathcal{H}\Phi$ (full diagonalization using LAPACK) and the contour-integral method. For the contour-integral method, we show results for several different N_k and N_l , denoted as $SS(N_k, N_l)$. When the rank of $SS(N_k, N_l)$ is smaller than n, we do not obtain the nth eigenvalues. This case is indicated as NA (not available) in the table. In the actual calculation, we discard singular values that are smaller than 10^{-3} .

E_n	$\mathcal{H}\Phi$	SS(10,1)	SS(10,2)	SS(10,5)
0	-5.387391	-5.387391	-5.387391	-5.387391
1	-5.031543	-5.031543	-5.031543	-5.031543
2	-4.777389	-4.777389	-4.777389	-4.777389
3	-4.569374	-4.569374	-4.569374	-4.569374
4	-4.569374	-4.297689	-4.569374	-4.569374
5	-4.297689	NA	-4.297689	-4.297689
6	-4.297689	NA	-4.297689	-4.297689

analyze low-energy structures. The dynamical correlation factor is defined as

$$D(\omega, \eta) = -\frac{1}{\pi} \text{Im} \left[\boldsymbol{\phi}_{\mathbf{0}}^{\dagger} A \left[(\omega - E_0 - i\eta)I - H \right]^{-1} B \boldsymbol{\phi}_{\mathbf{0}} \right], \tag{43}$$

where ϕ_0 is the ground state vector, A and B are the matrices to generate excited states, ω represents the frequency, and η represents the smearing factor. For example, by taking $A = B = S^z$, we can calculate the dynamical spin structure factors, which can be directly measured by the neutron scattering experiments. The key part in the calculation of the dynamical correlation factors is solving the linear equation defined as

$$\phi' = [(\omega - E_0 - i\eta)I - H]^{-1} B\phi_0. \tag{44}$$

As an example of a calculation of dynamical Green's functions by ShiftK.out, we present the calculation of the dynamical spin structure factors on a one-dimensional Heisenberg chain, whose Hamiltonian is defined as

$$H = J \sum_{i=0}^{L-1} \mathbf{S}_i \cdot \mathbf{S}_{i+1}, \tag{45}$$

where **S** represents the spin 1/2 operator and we take a magnetic interaction of J=1 and system size of L=12. The matrix data file of the Hamiltonian H is generated by $\mathcal{H}\Phi$ in the Matrix Market format. The dynamical spin structure factors are defined as

$$S(q, \omega, \eta) = -\frac{1}{\pi} \operatorname{Im} \left[\boldsymbol{\phi}_0^{\dagger} S^z(-q) \left[(\omega - E_0 - i\eta)I - H \right]^{-1} S^z(q) \boldsymbol{\phi}_0 \right], \tag{46}$$

where $S^z(q) = \sum_{j=0}^{L-1} e^{iqj} S_j^z$ and q is the wave vector. The input file for ShiftK.out is given as follows ($E_0=0,-5.5\leq\omega\leq0.0$, $\eta=0.02$):

Here, Ham. dat and excitedvec. dat are the input files for the Hamiltonian matrix H and the excited vector $S^z(q)\phi_0$. These files can be obtained using the quantum lattice solver $\mathcal{H}\Phi$. The details are shown in Appendix C.

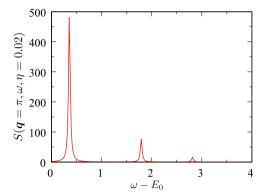


Fig. 3. Dynamical Green's function for the one-dimensional Heisenberg model. In this calculation, we take $\eta=0.02$, $\omega_{\min}=-5.5$, $\omega_{\max}=0.0$, and the energy shift $E_0=-5.387$, which is the energy of the ground state.

In Fig. 3, we show the numerical result of $S(q=\pi,\omega,\eta)$. Here, we shift ω by the ground state energy $E_0=-5.387$. The frequencies where $S(q=\pi,\omega,\eta)$ have sharp peaks correspond to the excitation energies induced by $S_z(q=\pi)$. The convergence of the residuals can be checked in the standard output.

4.3. Calculation of optical conductivity using $\mathcal{H}\Phi$

The third example is the calculation of a spectrum or dynamical correlation function using $\mathcal{H}\Phi$, demonstrating that using $K\omega$ in $\mathcal{H}\Phi$ can be a powerful tool for various problems. As an example, we show the optical conductivity in the extended Hubbard model, which is a typical model for strongly correlated electron systems. We note that the optical conductivity is often used for examining the metallic or insulating behavior of correlated electron systems. The optical conductivity can be calculated from the current–current correlation $I(\omega, \eta)$, which is defined as

$$j_{x} = i \sum_{i,\sigma} (c_{\mathbf{r}_{i}+\mathbf{e}_{x},\sigma}^{\dagger} c_{\mathbf{r}_{i},\sigma} - c_{\mathbf{r}_{i},\sigma}^{\dagger} c_{\mathbf{r}_{i}+\mathbf{e}_{x},\sigma}), \tag{47}$$

$$I(\omega, \eta) = \operatorname{Im} \left[\boldsymbol{\phi}_{\mathbf{0}}^{\dagger} j_{x} [H - (\omega - E_{0} - i\eta)I]^{-1} j_{x} \boldsymbol{\phi}_{\mathbf{0}} \right], \tag{48}$$

where \mathbf{e}_x is the unit translational vector in the x direction. The ground state vector $\boldsymbol{\phi}_0$ is calculated by an exact diagonalization solver built in $\mathcal{H}\Phi$. To obtain $\boldsymbol{\phi}_0' = [H - (\omega - E_0 - \mathrm{i}\eta)I]^{-1}j_x\boldsymbol{\phi}_0$, we use $K\omega$. From the current-current correlation, the regular part (i.e., without the Drude part at $\omega = 0$) of the optical conductivity is defined as

$$\sigma_{\text{reg}}(\omega) = \frac{I(\omega, \eta) + I(-\omega, -\eta)}{\omega N_{\text{s}}},\tag{49}$$

where N_s is the number of sites.

To directly compare the optical conductivity with previous studies [36], we calculate the optical conductivity in the extended Hubbard model defined as

$$H = -t \sum_{\langle i,j \rangle} \left[c_{i\sigma}^{\dagger} c_{j\sigma} + c_{j\sigma}^{\dagger} c_{i\sigma} \right] + U \sum_{i} n_{i\uparrow} n_{i\downarrow}$$

$$+ V \sum_{\langle i,j \rangle} N_{i} N_{j} + V' \sum_{\langle \langle i,j \rangle \rangle} N_{i} N_{j},$$
(50)

where $c_{i\sigma}$ and $c_{i\sigma}^{\dagger}$ denote the annihilation and creation operators of an electron at site i with spin σ , and $n_{i\sigma} = c_{i\sigma}^{\dagger} c_{i\sigma}$ represents the number operator of an electron at site i with spin σ . We note that this model at quarter filling is an effective model for organic

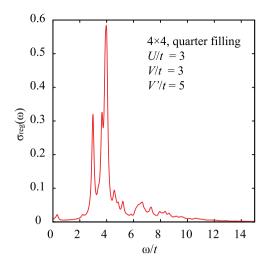


Fig. 4. Optical conductivity for the extended Hubbard model. In this calculation, we take $\eta=0.1$, $\omega_{\min}=0$, $\omega_{\max}=15.0$, and the number of ω as 2000.

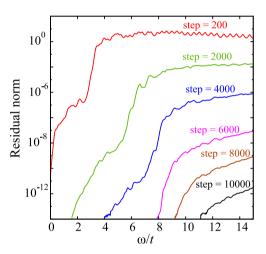


Fig. 5. Residual norm $\max_{\sigma} |\mathbf{r}_n^{\sigma}|$ of $I(\omega,\eta)$ for several iteration steps. For the extended Hubbard model, we find that the residual for the high-energy part remains large. This result indicates that it is necessary to perform more than 10^4 iterations to obtain a well-converged correlation function when the threshold value is set to 10^{-12} .

conductors. We perform a calculation for an $N_s = 4 \times 4$ system with t = 1, U = 4, V = 3, and V' = 5.

In Fig. 4, we show the result of the optical conductivity in the extended Hubbard model. This result is consistent with a previous study [36]. We also show the residuals of $I(\omega,\eta)$ in Fig. 5. By calculating the residuals, we can check whether the obtained dynamical correlation functions are well converged. As mentioned above, this is one of the main advantages of the shifted Krylov subspace method. Furthermore, by examining the ω dependence of the residuals we can also identify bottlenecks for the convergence.

It is noteworthy that the matrix size becomes huge for the calculation of realistic models. In such cases, the matrix-vector product dominates the numerical cost of the shifted-Krylov method. To address this problem, we designed the $K\omega$ library to not directly include the matrix-vector product in the function, as explained in Section 3.2.1. If users can prepare the matrix-vector product function with high parallelization efficiency, the calculation time is greatly suppressed. In fact, the calculation of

dynamical Green's functions by $\mathcal{H}\Phi$ shows high parallelization efficiency, since the matrix-vector product is well parallelized [29]. From this point of view, this library is considered to be suitable for large-scale calculations.

5. Summary

We developed the numerical library $K\omega$ for solving shifted linear equations with the shifted Krylov subspace methods. The present paper details applications for quantum many-body problems that appear in condensed matter physics. As a demonstration, numerical results including dynamical Green's functions, optical conductivity, and eigenvalues through the contourintegral method are shown. The method is also applicable to other computational physics areas, and hence $K\omega$ could be a useful numerical library in computational physics.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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Appendix A. Shifted conjugate orthogonal conjugate gradient method

When all the shifted matrices $A + \sigma I$ are complex symmetric, users should use the shifted conjugate orthogonal conjugate gradient (COCG) method [7]. In the COCG method [37], recurrence equations can be obtained by changing the inner products from $\mathbf{x}^{\dagger}\mathbf{y}$ to $\mathbf{x}^{T}\mathbf{y}$ in those for the CG method and in the definition of the orthogonality and A conjugate. The nth residual \mathbf{r}_n is orthogonal (defined by the ordinary inner product) not to $K_n(A, \mathbf{b})$ but to $K_n(A, \mathbf{b})^*$. The collinearity of the shifted residuals still holds [7], however, since \mathbf{r}_n belongs to the one-dimensional orthogonal complementary space of $K_n(A, \mathbf{b})^*$ within $K_{n+1}(A, \mathbf{b})$. In the result, the shifted equations have the same form as those for the shifted CG method.

Appendix B. Shifted bi-conjugate gradient method

When $A + \sigma I$ is not symmetric or Hermitian, users should use the shifted bi-conjugate gradient (BiCG) method [3].

B.1. Bi-conjugate gradient method

The BiCG method [38] generates sequences of approximate solutions x_n , residuals r_n , shadow residuals \tilde{r}_n , search directions p_n , and shadow search directions \tilde{p}_n so that the pair of residuals and shadow residuals forms a bi-orthogonal system, that is,

$$\tilde{\boldsymbol{r}}_{i}^{\dagger}\boldsymbol{r}_{i}=0 \quad (i\neq j); \quad \tilde{\boldsymbol{r}}_{i}^{\dagger}\boldsymbol{r}_{i}\neq 0.$$
 (B.1)

These vectors are iteratively generated by the following recurrence equations, as for the CG method,

$$\rho_n = \tilde{\boldsymbol{r}}_n^{\dagger} \boldsymbol{r}_n, \tag{B.2}$$

$$\alpha_n = \frac{\rho_n}{\tilde{\boldsymbol{p}}_n^{\dagger} A \boldsymbol{p}_n},\tag{B.3}$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + \alpha_n \mathbf{p}_n, \tag{B.4}$$

$$\mathbf{r}_{n+1} = \mathbf{r}_n - \alpha_n A \mathbf{p}_n, \tag{B.5}$$

$$\tilde{\boldsymbol{r}}_{n+1} = \tilde{\boldsymbol{r}}_n - \alpha_n^* A^{\dagger} \tilde{\boldsymbol{p}}_n, \tag{B.6}$$

$$\beta_n = \frac{\rho_{n+1}}{\rho_n},\tag{B.7}$$

$$\boldsymbol{p}_{n+1} = \boldsymbol{r}_{n+1} + \beta_n \boldsymbol{p}_n, \tag{B.8}$$

$$\tilde{\boldsymbol{p}}_{n+1} = \tilde{\boldsymbol{r}}_{n+1} + \beta_n^* \tilde{\boldsymbol{p}}_n, \tag{B.9}$$

with the initialization $\mathbf{r}_0 = \mathbf{p}_0 = \mathbf{b} - A\mathbf{x}_0$ and $\tilde{\mathbf{r}}_0 = \tilde{\mathbf{p}}_0 = \tilde{\mathbf{b}}$, where \mathbf{x}_0 is an input vector and $\tilde{\mathbf{b}}$ is an arbitrary vector that is not orthogonal to \mathbf{b} .

The residuals ${m r}$ and $\tilde{{m r}}$ and the coefficient ${m lpha}$ follow three term recurrences:

$$\mathbf{r}_{n+1} = \left(1 + \frac{\alpha_n \beta_{n-1}}{\alpha_{n-1}} - \alpha_n A\right) \mathbf{r}_n - \frac{\alpha_n \beta_{n-1}}{\alpha_{n-1}} \mathbf{r}_{n-1}, \tag{B.10}$$

$$\tilde{\boldsymbol{r}}_{n+1} = \left(1 + \left(\frac{\alpha_n \beta_{n-1}}{\alpha_{n-1}}\right)^* - \alpha_n^* A^{\dagger}\right) \tilde{\boldsymbol{r}}_n - \left(\frac{\alpha_n \beta_{n-1}}{\alpha_{n-1}}\right)^* \tilde{\boldsymbol{r}}_{n-1},$$
(B.11)

$$\alpha_n = \frac{\rho_n}{\tilde{\mathbf{r}}_n^{\dagger} A \mathbf{r}_n - \frac{\beta_{n-1}}{\alpha_{n-1}} \rho_n},\tag{B.12}$$

with $\beta_{-1}/\alpha_{-1}=0$. K ω adopts these three term recurrences.

B.2. Shifted equation

When starting the BiCG iteration for the seed equation $A\mathbf{x} = \mathbf{b}$ and some shifted equation $(A + \sigma I)\mathbf{x} = \mathbf{b}$ with the same initial guess $\mathbf{x}_0 = \mathbf{x}_0^{\sigma} = \mathbf{0}$ and shadow residuals $\tilde{\mathbf{r}}_0$, the *n*th residuals \mathbf{r}_n and \mathbf{r}_n^{σ} are parallel to each other [3],

$$\mathbf{r}_n^{\sigma} = \frac{1}{\pi_n^{\sigma}} \mathbf{r}_n. \tag{B.13}$$

The approximate solutions for the shifted equations are calculated via the same equations in the shifted CG method, Eqs. (21)–(25).

Appendix C. Generate input files by $\mathcal{H}\Phi$

This appendix shows the Hamiltonian matrix in the Matrix Market format and the excited vector generated using $\mathcal{H}\Phi$. These files can be read by shiftK.out. As an example, we focus on the 12-site one-dimensional Heisenberg chain, whose Hamiltonian is defined in Eq. (45) and the excited vector $S^z(q)\phi_0$ in Eq. (46).

C.1. Output Hamiltonian matrix

An input file for $\mathcal{H}\Phi$ is given as follows:

```
L = 12

model = "Spin"

lattice = "chain"

method = "FullDiag"

J = 1.0

2Sz = 0

HamIO = "out"
```

Using this input file, we can obtain the output file of the Hamiltonian matrix in the Matrix Market format (zvo_Ham.dat renamed as Ham.dat in Section 4).

C.2. Output the excited vector

To obtain the excited vector $S^z(q)\phi_0$ by $\mathcal{H}\mathcal{\Phi}$, we first calculate the ground state ϕ_0 and then calculate the excited vector by multiplying ϕ_0 by $S^z(q)$. The ground state ϕ_0 can be obtained by the following input file:

```
L = 12
model = "Spin"
lattice = "chain"
method = "CG"
J = 1.0
2Sz = 0
EigenVecIO = "Out"
```

The input file for calculating the excited state $S^z(q)\phi_0$ is given as follows:

```
L = 12
model = "Spin"
lattice = "chain"
method = "CG"
J = 1.0
2Sz = 0
CalcSpec="Normal"
SpectrumQL = 0.5
EigenVecI0 = "In"
OutputExcitedVec = "Out"
```

Here, the wave vector is specified using SpectrumQL. In the above case, $S^z(q=\pi)\phi_0$ is obtained. For details, see $\mathcal{H}\Phi$ user's manual. After the calculations are finished, the output file of the excited vector zvo_excitedvec_rank_0.dat (renamed as excitedvec.dat in Section 4) is obtained.

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