Linear response strength functions with iterative Arnoldi diagonalization

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We report on an implementation of a new method to calculate random phase approximation (RPA) strength functions with iterative non-Hermitian Arnoldi diagonalization method, which does not explicitly calculate and store the RPA matrix. We discuss the treatment of spurious modes, numerical stability, and how the method scales as the used model space is enlarged. We perform the particle-hole RPA benchmark calculations for double magic nucleus ¹³²Sn and compare the resulting electromagnetic strength functions against those obtained within the standard RPA.

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I. INTRODUCTION

The linear response theory (LRT) obtained from the linearized time-dependent mean-field method is an important tool for calculating properties of excited states of many-fermion systems, such as nuclear giant resonances. In its chargechanging version, it can also give access to the beta-decay strengths. This method is especially important in heavy nuclei, where the shell-model or configuration-interaction approaches are intractable. An advantage is also that LRT does not require the knowledge of an interaction and can therefore be used both within density functional theory (DFT) and phenomenological energy density functional (EDF) approaches, giving rise to a set of equations of RPA type. Below, for simplicity we refer to this method and associated equations simply as RPA method or equations. Strength functions obtained in this way probe new aspects of the EDFs and thus have a potential of constraining parameters in phenomenological nuclear EDFs.

The purpose of the present study is to present an implementation of an efficient RPA algorithm that is based on the local nuclear EDF. For electronic systems, similar methods have been used since many years (see, e.g., the recent Ref. [1] for a review), and they also constitute parts of standardized computer packages such as GAMESS [2,3]. There are two essential elements of these methods, which are at the heart of their efficiency and scalability, namely, (i) the RPA equations are solved iteratively and (ii) the RPA matrix does not have to be explicitly calculated. The second of these elements is particularly important; it is based on the observation that the action of the RPA matrix on the vector of RPA amplitudes can proceed through the calculation of the mean fields corresponding to these amplitudes.

In nuclear physics context, probably the first study that used the concept of mean fields in the RPA method was that by P.-G. Reinhard [4]. Iterative solutions of the RPA equations were introduced by Johnson *et al.* [5], and applied to the case of separable interactions, but in fact these methods can also be applied in more complicated situations, as we show here. Strangely enough, these very efficient methods have not yet been used in practical applications. Only very recently, Nakatsukasa *et al.* [6,7] have implemented the analogous approach within the so-called finite amplitude method (FAM). Our present implementation pertains to the spherical symmetry with neglected pairing correlations—thus it constitutes only a proof-of-principle study. The real challenge is in solving the quasiparticle RPA (QRPA) problem in deformed nuclei. Although at present, a few implementations that are based on solving the standard QRPA equations already exist [8,9] or begin to emerge [10,11], such a route is bound to be blocked by the shear dimensionality of the problem. On the other hand, as we show here, methods based on the iterative solutions using mean fields have much better scalability properties and are potentially very promising.

The article is organized as follows. In Secs. II and III we lay down the essential features of the method by presenting the use of mean fields and iterative solution of the RPA method, respectively. Then, in Sec. IV we present the method to remove the spurious RPA states, which is tailored to be used within the iterative approach. Secs. V and VI present the convergence and scalability properties of our method, respectively, and summary and conclusions are given in Sec. VII.

II. RPA FROM LINEARIZED TDHF

To be concise, in the following we present a less general derivation than the standard method [12,13] to derive the RPA equations from linearized time-dependent Hartree-Fock (TDHF) equations. For density-independent forces or functionals with terms quadratic in density, the density matrix and mean field of a time-dependent nuclear state are expressed as

$$\rho(t) = \rho_0 + \tilde{\rho}_{\omega} e^{\iota \omega t} + \tilde{\rho}_{\omega}^{\dagger} e^{-\iota \omega t}, \qquad (1)$$

$$h(t) = h_0 + \tilde{h}_{\omega} e^{i\omega t} + \tilde{h}_{\omega}^{\dagger} e^{-i\omega t}, \qquad (2)$$

where ρ_0 and h_0 are the Hartree-Fock (HF) ground-state density matrix and mean field, respectively. Inserting $\rho(t)$ and h(t) of Eqs. (1) and (2) into the TDHF equation, and keeping only terms linear in the fluctuating quantities $\tilde{\rho}$ and \tilde{h} , we get a linearized TDHF equation, or the RPA equations:

$$\hbar\omega\tilde{\rho}_{\omega,mi} = (\epsilon_m - \epsilon_i)\,\tilde{\rho}_{\omega,mi} + \tilde{h}_{\omega,mi},\tag{3}$$

$$\hbar\omega\tilde{\rho}_{\omega,im} = (\epsilon_i - \epsilon_m)\,\tilde{\rho}_{\omega,im} - \tilde{h}_{\omega,im},\tag{4}$$

where we use the letter *m* for particle states and *i* for hole states, and where $\epsilon_{m,i}$ are the HF single-particle energies. The fields

 \tilde{h} are the first functional derivatives of the used EDF, evaluated using the density amplitudes $\tilde{\rho}$ of Eq. (1). Density dependence of the used EDF beyond quadratic gives rise to rearrangement fields in \tilde{h} . These rearrangement parts of \tilde{h} must be linearized around ρ_0 to make our RPA equations explicitly first order in $\tilde{\rho}$.

One way to achieve this is by calculating functional derivatives of the rearrangement parts of \tilde{h} with respect to density, which technically makes our mean-field routine differ from the standard HF routines. Since in our implementation we use the standard Skyrme forces that have simple density dependencies of the coupling constants, the explicit functional differentiation does not cause any mathematical or performance problems. Had an EDF with more complex density dependence been used it would have been an advantage to instead use the FAM method [6] for linearization.

If the matrix elements of \tilde{h} in Eqs. (3) and (4) are expanded in terms of the particle-hole (p-h) and hole-particle (h-p) matrix elements of $\tilde{\rho}$, we obtain the traditional RPA equations. In this work, we do not construct the RPA matrix, but directly solve Eqs. (3) and (4) by calculating the matrix elements of fields \tilde{h} using an HF mean-field routine that uses the timereversal-invariance breaking density matrix $\tilde{\rho}$. Since the same routine is used to evaluate the HF and RPA mean fields, the method is always fully self-consistent [14,15]. In the following equations, we use the standard abbreviations $X_{mi} = \tilde{\rho}_{\omega,mi}$ and $Y_{mi} = \tilde{\rho}_{\omega,im}$. The density vector that contains the p-h matrix elements of $\tilde{\rho}^{\omega}$ is defined as $\mathcal{X}^{\omega} = (X_{m_1,i_1}, X_{m_2,i_2}, \dots, X_{m_D,i_D})$, and similarly for the vector \mathcal{Y} of h-p elements, where D is the number of allowed p-h configurations. Overlaps of RPA vectors are defined as

$$\langle X, Y | X', Y' \rangle = (\mathcal{X}^*, \mathcal{Y}^*) \begin{pmatrix} \mathcal{X}'^T \\ -\mathcal{Y}'^T \end{pmatrix},$$
(5)

and the minus sign results from the RPA norm matrix.

III. ITERATIVE SOLUTION OF THE RPA EQUATIONS

The RPA equations [Eqs. (3) and (4)] constitute a non-Hermitian eigenproblem with non-positive-definite norm. We solve this problem by using an iterative method that during each iteration only needs to know the product of the RPA matrix and a density vector, that is, the right-hand sides of Eqs. (3) and (4):

$$W_{mi}^{k} = (\epsilon_{m} - \epsilon_{i}) X_{mi}^{k} + \tilde{h}_{mi}(\mathcal{X}^{k}, \mathcal{Y}^{k}), \qquad (6)$$

$$W_{mi}^{\prime k} = (\epsilon_i - \epsilon_m) Y_{mi}^k - \tilde{h}_{im}(\mathcal{X}^k, \mathcal{Y}^k),$$
(7)

where index *k* labels iterations and the mean fields $\tilde{h}(\mathcal{X}^k, \mathcal{Y}^k)$ depend linearly on the density vectors \mathcal{X}^k and \mathcal{Y}^k . Expressed through the standard RPA matrices *A* and *B* [12], Eqs. (6) and (7) for a positive norm basis vector and for its opposite norm partner vector read:

$$\begin{pmatrix} \mathcal{W}_{+}^{k} \\ \mathcal{W}_{+}^{\prime} \end{pmatrix} = \begin{pmatrix} A & B \\ -B^{\prime *} - A^{\prime *} \end{pmatrix} \begin{pmatrix} \mathcal{X}^{k} \\ \mathcal{Y}^{k} \end{pmatrix}, \tag{8}$$

$$\begin{pmatrix} \mathcal{W}_{-}^{k} \\ \mathcal{W}_{-}^{\prime \, k} \end{pmatrix} = \begin{pmatrix} A & B \\ -B^{\prime *} & -A^{\prime *} \end{pmatrix} \begin{pmatrix} \mathcal{Y}^{k *} \\ \mathcal{X}^{k *} \end{pmatrix}.$$
(9)

In exact arithmetic A = A' and B = B' and therefore either Eqs. (8) or (9) could be used in the iteration procedure with equivalent results. Nevertheless, below we use them both to stabilize the iteration process.

Various iterative methods, which only need to know the products of the diagonalized matrix and vectors, exist for non-Hermitian matrix eigenvalue equations, and good examples with pseudocode are shown in Ref. [1]. For our RPA solver we modified the non-Hermitian Lanczos method of Ref. [5]. It has the advantage of conserving all odd-power energy weighed sum rules (EWSR) if the starting vector (pivot) of iteration is chosen correctly. To satisfy odd-power EWSRs, our RPA solver in this work always starts from a pivot vector that has its elements set to the matrix elements of electromagnetic multipole operator,

$$X_{mi}^{1} = \frac{e}{\sqrt{N^{1}}} \langle \phi_{m} | r^{p} Y_{JM} | \phi_{i} \rangle, \quad Y_{mi}^{1} = 0,$$
(10)

where p = 2 and J = 0 for the 0^+ mode, p = 1 and J = 1 for the 1^- mode, and p = 2 and J = 2 for the 2^+ mode. The constant N^1 is used to normalize the pivot vector to unity. For the IS 1^- mode we only present results obtained with the operator,

$$\left(r^3 - \frac{5}{3}\langle r^2 \rangle r\right)Y_{1M},\tag{11}$$

to stay consistent with Refs. [16] and [17]. For the choice of pivot in Eq. (10) one can prove [5] that odd-power EWSRs are satisfied throughout a Lanczos iteration procedure. The proof is also valid with our modified Arnoldi iteration method, because it does not assume anything about the details of used iterative diagonalization method.

Since our RPA treatment is fully self-consistent in the sense that the HF ground state is a true energy minimum and not a saddle point, and our RPA calculation uses the same EDF as in the HF calculation, we expect to obtain only real nonzero physical RPA frequencies. Therefore, all complex RPA eigenvalues in our method result from inaccurate numerics (note that spurious solutions have zero real eigenvalues). Occurrence of complex eigenvalues we call numerical instability, and in the following we explain why the numerical instabilities happen and what methods we use to remove the instabilities from our RPA solver.

We stabilize our iterative RPA solution method by modifying the method of Ref. [5] in two ways. First, we use the non-Hermitian Arnoldi method instead of the non-Hermitian Lanczos method. The advantage of Arnoldi method is that it orthogonalizes each new basis vector against all previous basis vectors and their opposite norm partners, that is,

$$\begin{pmatrix} \tilde{\mathcal{X}}^{k+1} \\ \tilde{\mathcal{Y}}^{k+1} \end{pmatrix} = \begin{pmatrix} \mathcal{W}^{k}_{+} \\ \mathcal{W}^{\prime k}_{+} \end{pmatrix} - \sum_{i=1}^{k} \begin{pmatrix} \mathcal{X}^{i} \\ \mathcal{Y}^{i} \end{pmatrix} a_{ik} + \sum_{i=1}^{k} \begin{pmatrix} \mathcal{Y}^{i*} \\ \mathcal{X}^{i*} \end{pmatrix} b_{ik},$$

$$\begin{pmatrix} \tilde{\mathcal{Y}}^{k+1*} \\ \tilde{\mathcal{X}}^{k+1*} \end{pmatrix} = - \begin{pmatrix} \mathcal{W}^{k}_{-} \\ \mathcal{W}^{\prime k}_{-} \end{pmatrix} + \sum_{i=1}^{k} \begin{pmatrix} \mathcal{X}^{i} \\ \mathcal{Y}^{i} \end{pmatrix} b_{ik}^{\prime*} - \sum_{i=1}^{k} \begin{pmatrix} \mathcal{Y}^{i*} \\ \mathcal{X}^{i*} \end{pmatrix} a_{ik}^{\prime*},$$

$$(13)$$

where the overlap matrices a_{ik} , b_{ik} , a'^*_{ik} , and b'^*_{ik} are calculated as in Eq. (5). Again, in exact arithmetic, Eqs. (12) and (13) are equivalent and the lower matrix elements a_{ik}^{**} and b_{ik}^{**} in Eq. (13) are exact complex conjugates of the elements of the upper Krylov-space [18] RPA matrices. We assume this to keep the Krylov-space RPA matrix in the standard form.

In the Lanczos method, only the tridiagonal parts of Krylov-space RPA matrices are calculated, and matrix elements outside the tridiagonal part are set exactly to zero. In exact arithmetic the Lanczos iteration procedure produces an orthonormal basis. However, due to finite numerical precision and the fact that each new vector is only explicitly orthogonalized against the two previous vectors, the Lanczos basis vectors lose their orthogonality after a few tens of iterations and complex RPA eigenvalues can appear. Therefore the orthogonality of the generated basis must be continuously monitored and when orthogonality is lost the vectors must be re-orthogonalized. The standard way to do this is to use Gram-Schmidt orthogonalization, which is done after each Lanczos orthogonalization step. This correction therefore affects the basis vectors but does not affect the elements of the tridiagonal Krylov-space RPA matrices, which creates an inconsistency between basis vectors and matrix elements.

In contrast, the Arnoldi method explicitly orthogonalizes each new obtained basis vector against all previous vectors. The Krylov-space RPA matrix elements outside the tridiagonal parts have very small magnitudes, but they ensure accurate orthogonality and therefore improve the stability as compared to the Lanczos method.

Standard RPA method that constructs and diagonalizes the full RPA matrix can ensure that the lower matrices in the RPA supermatrix are exact complex conjugates of the upper matrices. Because we calculate the RPA matrix-vector products by using the mean-field method, and not with a precalculated RPA matrix, we introduce small, but significant numerical noise to the resulting vectors. Small differences in the implicitly used upper and lower RPA matrices appear due to finite numerical precision. The consequence of this is that we will in general have $a'_{ij} \neq a_{ij}$ and $b'_{ij} \neq b_{ij}$ for the Krylov-space RPA matrices. This spoils the consistency of Eqs. (12) and (13). If corrective measures are not used to remove or reduce this noise, the iteration method produces complex RPA eigenvalues early on in the iteration.

The numerical errors in the matrix-vector products can be reduced by symmetrization. In our second stabilization method, we calculate the RPA fields twice, first using the densities of a positive norm basis vector ($\mathcal{X}^k, \mathcal{Y}^k$), and second using the densities of negative norm vector ($\mathcal{Y}^{k*}, \mathcal{X}^{k*}$). The two resulting vectors are subtracted from each other to get the final stabilized RPA matrix-vector product,

$$\begin{pmatrix} \mathcal{W}^k \\ \mathcal{W}'^k \end{pmatrix} = \frac{1}{2} \begin{pmatrix} \mathcal{W}^k_+ - \mathcal{W}'^{\ k*}_- \\ \mathcal{W}'^{\ k}_+ - \mathcal{W}^{k*}_- \end{pmatrix},$$
(14)

to be used in Eq. (12). To conclude, we note that using Arnoldi method alone without symmetrized matrix products was not enough for us to produce numerically stable RPA modes. Furthermore, the re-orthogonalizing Lanczos method used with symmetrized matrix products was not a stable method. We found that when used together the Arnoldi iteration method and symmetrization of matrix-vector products stabilized our mean-

field-based iterative RPA diagonalization in the sense that hundreds of iteration steps can be made without the appearance of complex RPA frequencies or loss of orthogonality.

The norm of the obtained new residual vector in Eq. (12) can be either positive or negative. We do not in practice use Eq. (13), which in exact arithmetic would duplicate the results of Eq. (12). Instead, we store only the positive-norm basis states and use a similar method as in Ref. [5] to change sign of the norm in case the norm of the residual vector in Eq. (12) is negative. Thus, explicitly, for the positive norm of the residual vector $\tilde{N}^{k+1} = \langle \tilde{X}^{k+1}, \tilde{Y}^{k+1} | \tilde{X}^{k+1}, \tilde{Y}^{k+1} \rangle$, we define the new normalized positive-norm basis vector as

$$X_{mi}^{k+1} = \frac{1}{\sqrt{\tilde{N}^{k+1}}} \tilde{X}_{mi}^{k+1}, \quad Y_{mi}^{k+1} = \frac{1}{\sqrt{\tilde{N}^{k+1}}} \tilde{Y}_{mi}^{k+1}.$$
 (15)

If $\tilde{N}^{k+1} < 0$, the new normalized positive norm basis vector is defined as

$$X_{mi}^{k+1} = \frac{1}{\sqrt{-\tilde{N}^{k+1}}} \tilde{Y}_{mi}^{k+1*}, \quad Y_{mi}^{k+1} = \frac{1}{\sqrt{-\tilde{N}^{k+1}}} \tilde{X}_{mi}^{k+1*}.$$
 (16)

When the maximum number of iterations has been made or the iteration has been stopped, the generated Krylov-space RPA matrix, with dimension $d \ll D$, is diagonalized with standard methods, that is, we solve

$$\begin{pmatrix} a & b \\ -b^* & -a^* \end{pmatrix} \begin{pmatrix} x^k \\ y^k \end{pmatrix} = \hbar \omega_k \begin{pmatrix} x^k \\ y^k \end{pmatrix}, \quad (17)$$

where the definition of Krylov-space RPA matrix elements follows from Eq. (12) and the fact that all basis vectors are mutually orthogonal:

$$a_{ik} = (\mathcal{X}^{i*}, \mathcal{Y}^{i*}) \begin{pmatrix} \mathcal{W}_{+}^{kT} \\ -\mathcal{W}_{+}^{\prime kT} \end{pmatrix}, \qquad (18)$$

$$b_{ik} = (\mathcal{Y}^i, \mathcal{X}^i) \begin{pmatrix} \mathcal{W}_+^{kT} \\ -\mathcal{W}'_+^{kT} \end{pmatrix}, \qquad (19)$$

and in similar fashion for a'_{ik} and b'_{ik} using Eq. (13). The approximate RPA solutions are then obtained by transforming the Krylov-space basis vectors,

$$X_{mi}^{k} = \sum_{l=1}^{d} \left(X_{mi}^{l} x_{l}^{k} + Y_{mi}^{l*} y_{l}^{k*} \right),$$
(20)

$$Y_{mi}^{k} = \sum_{l=1}^{d} \left(Y_{mi}^{l} x_{l}^{k*} + X_{mi}^{l*} y_{l}^{k} \right), \qquad (21)$$

for all k = 1, ..., d, and these vectors are used to evaluate the strength functions.

IV. TREATMENT OF SPURIOUS RPA MODES

For the discussion of various spurious modes in the RPA method we refer the reader to, for example, Ref. [13]. In the present study, we only consider spherical ground states neglecting pairing correlations, so the only spurious excitation is generated by the total linear momentum. Therefore, the only affected RPA mode is the isoscalar 1⁻ mode. In traditional RPA

calculations that construct and diagonalize the full RPA matrix, the spurious 1⁻ mode is typically removed *after* the RPA diagonalization. Often a modified transition operator [Eq. (11)] is used, which has the property of $\langle HF | [\hat{F}, \hat{P}_{cm}] | HF \rangle = 0$, as long as the commutator is evaluated within a complete set of basis states. In a finite model space built of localized orbitals this relation is no more exactly valid, and the corrected operator does not remove spurious components exactly.

To remove the spurious isoscalar 1^- mode from our physical RPA excitations we use the same method as in Ref. [6], where the basis vectors are orthogonalized against the spurious translational mode **P** and its conjugate "boost" operator **R**, which have the form,

$$\hat{P}_{\mu} = \frac{1}{\sqrt{3}} \sum_{mi} (i(\phi_m ||\nabla_1||\phi_i) [c_m^{\dagger} \tilde{c}_i]_{1\mu} + \text{h.c.}), \quad (22)$$

$$\hat{R}_{\mu} = \frac{1}{\sqrt{3}} \sum_{mi} ((\phi_m ||r_1||\phi_i) [c_m^{\dagger} \tilde{c}_i]_{1\mu} + \text{h.c.}).$$
(23)

The spurious RPA vectors $(\mathcal{P}, \mathcal{P}^*)$ and $(\mathcal{R}, \mathcal{R}^*)$ contain the p-h and h-p matrix elements of Eqs. (22) and (23), respectively. Our method differs from that of Ref. [6] in the fact that we orthogonalize our basis *during* the Arnoldi iteration, which fits naturally with the iterative solution method and guarantees that the obtained approximate RPA excitations have exact zero overlaps with spurious modes. This is equivalent to diagonalizing the full RPA matrix in the subspace orthogonal to the spurious states. In our implementation, each generated new Arnoldi basis vector is orthogonalized as

$$\begin{pmatrix} \mathcal{X}_k \\ \mathcal{Y}_k \end{pmatrix}_{\text{phys.}} = \begin{pmatrix} \mathcal{X}_k \\ \mathcal{Y}_k \end{pmatrix} - \lambda \begin{pmatrix} \mathcal{P} \\ \mathcal{P}^* \end{pmatrix} - \mu \begin{pmatrix} \mathcal{R} \\ \mathcal{R}^* \end{pmatrix}, \quad (24)$$

where the overlaps λ and μ are defined as

$$\lambda = \frac{\langle R, R^* | X^k, Y^k \rangle}{\langle R, R^* | P, P^* \rangle},\tag{25}$$

$$\mu = -\frac{\langle P, P^* | X^k, Y^k \rangle}{\langle R, R^* | P, P^* \rangle}.$$
(26)

When more symmetries are broken, formulas equivalent to Eqs. (24)–(26) can be used to remove spurious components coming from each broken symmetry of the mean field.

V. CONVERGENCE OF STRENGTH FUNCTIONS

The iterative Arnoldi method is meaningful for the calculation of strength functions only if the number of iterations needed for accurate results is significantly less than the full RPA dimension. To study how many Arnoldi iterations we need for good accuracy, we calculated electromagnetic isoscalar (IS) and isovector (IV) strength functions [16] for doubly magic nuclei. All calculations were performed by implementing the RPA iterative solutions within the computer program HOSPHE [19], which solves the self-consistent equations in the spherical harmonic-oscillator (HO) basis. We studied both the convergence of smoothed strength functions as a function of number of Arnoldi iterations and as a function of the number of HO shells.

We used the same definitions of the 0^+ , 1^- , and 2^+ transition operators as in Ref. [17] and the Skyrme functional SkM*. The weight function $\gamma(E, R_{\text{box}}, \lambda_n)$ we used to smooth the strength functions was also the same as in Ref. [16], with $R_{\text{box}} = 20$ fm and λ_n the neutron Fermi energy, which in our calculations was -8.142 MeV. The smoothed strength function has the form,

$$S_J(E) = \frac{1}{\pi} \sum_{k=1}^d \frac{\gamma(E_k) |(J,k||\hat{F}_J||0)|^2}{(E_k - E)^2 - \gamma^2(E_k)}.$$
 (27)

Because the HF ground state of 132 Sn is spherically symmetric, our approximate RPA phonons have good angular momentum. We tested the use of large basis sets up to 40 HO shells. The HF ground-state energies were well converged for all double magic nuclei when 25 HO shells were used. Below, we present the results only for 132 Sn.

A. Convergence as a function of the number of Arnoldi iterations

1. The 0^+ strength functions

Figure 1 shows the 0⁺ IS and IV smoothed strength functions for ¹³²Sn calculated with 100 Arnoldi iterations compared with the standard RPA results from Ref. [17]. Agreement between the strength functions is excellent. The four panels of Fig. 2 show the convergence of the smoothed strength functions of Fig. 1 as the number of Arnoldi iterations increases. The panels show differences of the strength functions calculated at 20 iteration intervals. The convergence is quite satisfactory after 100–120 iterations.

In Fig. 3 we show the 0^+ IV strength function for three different numbers of Arnoldi iterations. Initially at n = 10 the IV strength is concentrated on two states. After 50 iterations the strength has fragmented to five states, and the fragmentation continues until after n = 100 so many RPA modes have







FIG. 2. Convergence of the 132 Sn 0⁺ strength functions of Fig. 1. Solid lines are for the IS and dashed lines are for the IV strength functions. Each panel shows the difference of two strength functions, one with *n* iterations and the other calculated with n - 20 iterations.

appreciable transition strength that the shape of the smoothed strength function has almost fully converged.

2. The 2⁺ strength functions

Figures 4 and 5 show similar results as Figs. 1 and 2, but for the 2⁺ strength functions in ¹³²Sn. As for the 0⁺ case, the IS and IV strength functions from the Arnoldi iteration agree very well with the strength functions of Ref. [17]. The convergence of strength functions is as fast as for 0⁺; after 120 iterations the smoothed strength functions change only by about 5%. We thus have to make only 120 iterations to calculate reasonably accurate 2⁺ strength functions for the RPA problem whose dimension is D = 1020. The large double spikes observed in Fig. 5 below 10 MeV are due to the lowest RPA phonons, which by the smoothing procedure acquire $\simeq 100$ -keV widths and move slightly down in excitation energy.



FIG. 3. The ¹³²Sn 0⁺ isovector strength function calculated with n = 10, 50, 100 Arnoldi iterations. Dotted line shows the strength after 10 iterations, dashed line after 50 iterations, and solid line after 100 iterations.



FIG. 4. Similar to Fig. 1 but for the 2^+ strength functions. All results were calculated for the SkM* functional.

3. The 1^- strength functions

Figure 6 compares the 1^- strength functions of our iterative method (solid line) with the strength functions from Ref. [17] (dashed line). The solid line shows the IS strength calculated when the generated Arnoldi basis is orthogonalized against spurious mode during iteration and dotted line corresponds to similar iterative calculation without orthogonalization. The low-lying state at 0.72 MeV, which has a large overlap with the spurious IS 1^- mode disappears when the orthogonalization method of Eqs. (24)–(26) is used. Furthermore, for the 1^- strength function, 100–120 Arnoldi iterations were needed to produce reasonably accurate results, see Fig. 7.

When no orthogonalization is made against the spurious 1^{-1} IS state, the obtained excitations contain small components of the spurious state. This mixing affects the physical part of the IS strength distribution very little when the transition operator of Eq. (11) is used. The standard RPA strength function of Ref. [17] was not corrected for the spuriosity but only the strength of the lowest-lying state that has a large overlap with the spurious IS mode was omitted. The resulting strength still agrees well with our corrected strength and justifies the claim made in Ref. [16] that spurious mixing is small. Without orthogonalization against the spurious mode we need 140 iterations instead of 100 to get acceptably converged strength function. The orthogonalization method improves the



FIG. 5. Similar to Fig. 2 but for the 2^+ strength functions.



FIG. 6. (Main panel) The 1^- strength functions in 132 Sn, calculated using 100 Arnoldi iterations and with spurious IS mode removed (solid line), and results of the standard RPA from Ref. [17] (dashed line). Dotted line shows results of 140 Arnoldi iterations without orthogonalization against the spurious IS mode. (Inset) Same as in the main panel, but for the IV strength functions. All results were calculated for the SkM* functional.

convergence of the strength function, because spurious mode is not present and the higher lying 8.3-MeV 1^- excitation converges first.

The ground-state expectation value of the commutator of transition operator of Eq. (11) and the total linear momentum operator is exactly zero when the operators are expressed in position representation. Therefore, in position representation the spurious translational RPA mode is exactly orthogonal against the operator of Eq. (11). In finite harmonic oscillator basis this orthogonality is not exact, but approximate. Because of that the dotted line in Fig. 6 shows a small peak of IS strength at the spurious mode. We thus verify that despite being incomplete, the HO basis is in practice sufficient to approximately represent the orthogonality of Eq. (11) and the spurious 1^- IS mode. However, had the normal dipole operator \vec{r} been used instead of Eq. (11), we would have obtained very large IS strength for the spurious state, and



FIG. 7. Similar to Fig. 5 but for the 1^- strength functions. The IV strength-function differences were multiplied by the factor of 200 fm⁴.



FIG. 8. Similar to Fig. 2 but the convergence of the 0^+ strength functions as a function of the number of HO shells N_0 .

without the orthogonalization method the spurious admixtures in the physical RPA excitations would have produced larger errors in the obtained strength function.

B. Convergence in function of the number of HO shells

In Sec. V A, we showed our strength functions calculated with 25 HO shells. This was found to be satisfactory, and using more shells did not appreciably change the obtained strength functions. The effect of using more oscillator shells was that we need to use slightly more Arnoldi iterations to produce well-converged results. In the case of 40 shells, about 20 more iterations were needed, compared to calculations made with 25 shells. In Figs. 8, 9, and 10, we show the convergence of strength functions for the 0⁺, 2⁺, and 1⁻ modes, respectively. Each panel shows the difference of two strength functions obtained in the intervals of $\Delta N_0 = 4$ HO shells, between N_0 of 22 and 38.

These plots overstress the variations of strength functions in the sense that slight shifts of peaks create the oscillating patterns in the difference plots. To illustrate this point, in Fig. 11 we show the 1⁻ strength functions calculated for $N_0 = 22$, 26, 30, 34, and 38 HO shells. Slow convergence of the IS surface mode as a function of N_0 creates some uncertainty in the position and width of the high-energy



FIG. 9. Similar to Fig. 5 but the convergence of the 2^+ strength functions as a function of the number of HO shells N_0 .



FIG. 10. Similar to Fig. 7 but for the convergence of the 1^- strength functions as a function of the number of HO shells N_0 .

bump. This lack of convergence is a result of the fact that in Fig. 11 we for simplicity use a fixed number of Arnoldi iterations. Because calculations in larger HO basis converge more slowly, the IS strengths in Fig. 11 are at slightly different stages of convergence, which enhances the oscillation of strength functions. The oscillations could be minimized by using an algorithm that uses more iterations with larger bases, or monitors the convergence of lowest RPA modes, for example.

As noted in Ref. [5], well before the maximum number of iterations (equal to the RPA dimension D) is reached, the iteratively generated RPA matrix in the Krylov space can become singular. In that case, the stabilized iteration method of Eqs. (12)–(14) protects us from obtaining complex RPA eigenvalues, but the condition number (the ratio of largest and smallest eigenvalue) of the Krylov-space RPA matrix approaches infinity, because one or more of the RPA matrix eigenvalues collapse nearly to zero.

In the standard RPA method the RPA matrix is calculated by using the bare p-h basis states. In our method, we instead start from the pivot vector of Eq. (10) and the Arnoldi iteration then produces the rest of our basis vectors composing the



FIG. 11. Similar to Fig. 6 but for the 1⁻ strength functions calculated for the numbers of HO shells $N_0 = 22$, 26, 30, 34, and 38.

TABLE I. Spherical RPA and QRPA dimensions D as functions of the number of HO shells N_0 .

N_0	RPA			QRPA		
	0+	1-	2^{+}	0+	1-	2^{+}
10	70	195	261	390	1040	1510
20	205	555	766	2880	8180	12720
25	273	734	1020	5538	15912	25088
30	340	915	1271	9470	27420	43630

Krylov subspace. This subspace is spanned by the eigenstates of the RPA matrix which have nonzero overlap with the pivot vector. Thus, in general, the Arnoldi iterations can only be continued until this subspace is exhausted in which case the condition number goes to infinity. However, with finite numerical precision this maximum limit of Arnoldi iterations is further reduced.

In a typical iteration, during the first few iterations the condition number of the Krylov-space RPA matrix fluctuates, then approaches a stable plateau, and finally suddenly goes toward infinity. When that happens, the iteration must be stopped and one must backtrack to the iteration where the condition number was still acceptable. Therefore, the number of Arnoldi iterations can depend on the size of the HO basis.

VI. SCALING OF ITERATIVE SOLUTION METHOD

We illustrate the benefits of the iterative solution of the RPA or QRPA equations over the traditional method by comparing how the numerical work increases in the iterative method as the HO basis is increased.

As can be seen in Table I, the RPA dimensions D for doubly magic spherical nuclei increase almost linearly with the number of oscillator shells N_0 . This is easy to understand, because in this case, only the number of particle states increases and the number of hole states always stays constant. Therefore the time to solve the full RPA eigenproblem in this case scales approximately as N_0^3 . In the spherical QRPA, the dimensions scale roughly between N_0^2 and N_0^3 , and the full QRPA scales approximately as N_0^6 or N_0^9 . The physically interesting and computationally challenging calculations are for deformed nuclei with pairing, and we should therefore compare the N_0 scaling of iterative and standard QRPA diagonalizations.

In the case of all symmetries of the mean field being broken, the QRPA dimension D is

$$D = \frac{1}{9} [(N_0 + 1)(N_0 + 2)(N_0 + 3)]^2.$$
(28)

This dimension increases very steeply (N_0^6) as the number of HO shells is increased. For $N_0 = 14$, the QRPA dimension is $D = 1.8496 \times 10^6$, for example. The corresponding standard QRPA solution scales as N_0^{18} and is thus untractable. Therefore, for the QRPA calculations in deformed nuclei, we must truncate the single-particle space. The best method is to use the two-basis method [21], by which one solves the HFB equations in the basis generated by the HF part of the



FIG. 12. Times to calculate 100 Arnoldi iterations for the spherical QRPA method applied to 132 Sn as functions of N_0 . Squares and circles show results for the 1^- and 2^+ modes, respectively, and lines show cubic fits.

HFB matrix, and truncate the basis using a cutoff on the obtained pseudo-HF single-particle energies. But even then, the QRPA calculations scale as the sixth power of the number of useful single-particle states, and are thus prohibitively difficult.

To illustrate the scaling properties of the iterative QRPA method, we calculated the corresponding matrix-vector products with our developmental QRPA code, where the pairing has been set to zero. The RPA fields \tilde{h} only depend on the normal RPA density matrix and the calculation of pairing part of the QRPA matrix-vector product is very fast due to a small number of the pairing coupling constants and the simple density dependence of typical pairing EDF. Therefore the time to calculate the pairing part is negligible. The only significant increase of running time in spherical ORPA compared to RPA comes from the need to handle higher-dimensional basis vectors in the calculation of various overlaps and vector additions during the Arnoldi iteration. Therefore, as we keep all particle-particle and hole-hole RPA amplitudes in the calculation, but set them to zero, our timing results accurately reflect the timing of the spherical QRPA calculation.

In Fig. 12, we show the scaling properties of the spherical QRPA calculation with the iterative Arnoldi method. It is clear that the scaling of our iterative method is as N_0^3 , that is, it is linear with respect to the QRPA dimension *D*. Of course, the prefactor itself is linearly proportional to the number of Arnoldi iterations. However, as discussed in the previous section, the Arnoldi iteration method cannot in practice go full dimension before the generated Krylov-space matrices become singular. As long as we are satisfied with a few hundred iterations at most, the iterative method gives us a vast speed improvement. In the full RPA or QRPA diagonalization, the calculation and storage of a very large dense RPA or QRPA matrices also take considerable additional time – steps that the iterative method avoid completely.

In addition to the moment-method-based iteration, which is ideal for strength functions, the iterative method can also be modified to be suitable for different kinds of other calculations. If we are interested in a number of very well-converged lowest RPA eigenmodes, restarted Arnoldi methods [22] can be used. These methods use more iterations than basis states, that is, after a maximum number d of basis vectors is generated, new approximations for the wanted d' < d eigenmodes are calculated, and iteration is then continued to generate new improved basis states from d' + 1 to d again. The restarting can be made as many times as needed to produce the wanted number of well-converged lowest excitations.

Methods such as Arnoldi or Lanczos produce convergence at the extreme ends of the excitation energy spectrum. If eigenmodes away from the extremes are looked for, shift and invert methods [23,24] can be used. These methods allow iterative methods to be used to find RPA eigenmodes anywhere inside the RPA excitation spectrum.

VII. SUMMARY AND CONCLUSIONS

We have presented a method to calculate accurate RPA response functions by using the iterative Arnoldi diagonalization related to the sum-rule conserving Lanczos method of Ref. [5]. We used strictly the same EDF for the ground-state calculation and RPA excitations. We have showed how the Arnoldi method must be stabilized in order to apply it reliably to the RPA eigenvalue problem. The resulting electromagnetic strength functions are in good agreement with the standard RPA results and are obtained with numerical effort smaller by orders of magnitude.

Our method closely resembles the FAM of Nakatsukasa *et al.* [6,7], except that our iterative method is different and that we use the HO basis instead of the mesh in coordinate space. The FAM and our method both allow the existing EDF mean-field codes to be used for the calculation of the RPA or QRPA matrix-vector products. With minor modifications, mostly pertaining to the full implementation of the time-odd mean fields, these codes can easily be extended to RPA/QRPA. In particular, our future implementation of the deformed QRPA solution will be based on the code HFODD [25].

We also implemented the method to remove components of the spurious RPA modes from the calculated strength functions that keeps the physical excitations exactly orthogonal against the spurious excitations in any finite model space.

The smaller numerical effort of the iterative Arnoldi method, and the fact that in this method one does not have to calculate and store the RPA or QRPA matrices, allows our method to be applied to the calculation of electromagnetic and beta decay strengths and strength functions for deformed heavy nuclei. Work to extend our formalism and codes to deformed superfluid nuclei is in progress.

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