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# Tensor representation techniques for full configuration interaction: A Fock space approach using the canonical product format

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In this proof-of-principle study, we apply tensor decomposition techniques to the Full Configuration Interaction (FCI) wavefunction in order to approximate the wavefunction parameters efficiently and to reduce the overall computational effort. For this purpose, the wavefunction ansatz is formulated in an occupation number vector representation that ensures antisymmetry. If the canonical product format tensor decomposition is then applied, the Hamiltonian and the wavefunction can be cast into a multilinear product form. As a consequence, the number of wavefunction parameters does not scale to the power of the number of particles (or orbitals) but depends on the rank of the approximation and linearly on the number of particles. The degree of approximation can be controlled by a single threshold for the rank reduction procedure required in the algorithm. We demonstrate that using this approximation, the FCI Hamiltonian matrix can be stored with N<sup>5</sup> scaling. The error of the approximation that is introduced is below Millihartree for a threshold of  $\epsilon = 10^{-4}$  and no convergence problems are observed solving the FCI equations iteratively in the new format. While promising conceptually, all effort of the algorithm is shifted to the required rank reduction procedure after the contraction of the Hamiltonian with the coefficient tensor. At the current state, this crucial step is the bottleneck of our approach and even for an optimistic estimate, the algorithm scales beyond N<sup>10</sup> and future work has to be directed towards reduction-free algorithms. *Published by AIP* Publishing. [http://dx.doi.org/10.1063/1.4953665]

#### I. INTRODUCTION

Since the early days of quantum chemistry it has been known that the full configuration interaction (FCI) method allows one to calculate the exact solution of the Schrödinger equation in principle. FCI is conceptually simple and obtaining the energies of the desired system only requires solving an eigenvalue problem. Hence one could follow Dirac's famous statement that "the fundamental laws necessary for the mathematical treatment of a large part of physics and the whole of chemistry are thus completely known, and the difficulty lies only in the fact that application of these laws leads to equations that are too complex to be solved".

Indeed, as in FCI the wavefunction is parametrized using the weights of all possible configuration state functions (or all possible slater determinants) the number of parameters grows exponentially with the number of particles. Therefore, FCI is usually only used for benchmark calculations and in spite of the remarkable increase of computational power in recent decades solving the FCI problem for few states using Davidson-type algorithms is still only feasible for systems with much less than 20 electrons. However, FCI should not be regarded as an approach with a tiny niche in quantum chemistry. As the central part of a complete active space (CAS) treatment<sup>1</sup> it is essential for solving the most challenging multi-reference problems in chemistry when other methods fail.

And in spite the paradigm that solving Schrödinger's equation demands exponential effort, several attempts have been made to accelerate the method and make it feasible for larger system sizes and larger active spaces. One important improvement was achieved through the insight that the Hamiltonian matrix is data sparse, and only a few elements are required for performing FCI, which means that the Hamiltonian matrix does not need to be constructed explicitly. This, in turn, led to the development of direct CI<sup>2</sup> which can be applied to larger systems, <sup>3–5</sup> but the factorial scaling behavior itself remains. In recent years, novel approaches have been developed to solve the Schrödinger equation for larger systems. One method of reducing the computational effort in FCI is to neglect certain configurations. Several schemes for selecting such configurations have been developed.<sup>6-11</sup> Other approaches are not based on solving an eigenvalue problem but instead apply Monte Carlo simulations, which has resulted in the development of approaches like Diffusion Quantum Monte Carlo (DMC)<sup>12,13</sup> or Full Configuration Interaction Quantum Monte Carlo (FCIOMC)<sup>14–16</sup> which can treat much larger systems than conventional FCI algorithms. Another alternative approach to solve the Schrödinger equation is the density matrix renormalization group (DMRG) method, in which the electron density is approximated using a tensor representation.<sup>17,18</sup> This Ansatz can also be applied to rather large systems, but convergence and accuracy in the energy depend on the ordering of the sites. 19–24 So, recent years have brought about approaches that are extremely efficient and have extended the applicability of methods that aim at FCI accuracy significantly.

The approach that we explore in this publication is the application of tensor decomposition techniques in the canonical product (CP)<sup>35,36</sup> format to FCI. In principle, tensor decomposition techniques are widely known in the field of quantum chemistry from approaches such as density fitting<sup>25–31</sup> or Cholesky decomposition<sup>32–34</sup> and have also been applied in various ways to post-Hartree-Fock methods. One example is the tensor hypercontraction format which has been applied to Møller-Plesset perturbation theory<sup>37</sup> and coupled cluster methods.<sup>38–40</sup> Tensor decompositions may also be used in multiresolution analysis representations for Hartree-Fock and MP2,<sup>41,42</sup> where both operators and coefficient tensors are decomposed. Furthermore, tensor decomposition techniques have also been applied to reduce the scaling of the FCI method.<sup>43</sup>

Driven by the potential of tensor decomposition techniques to break the curse of dimensionality in quantum chemistry, our recent work has focused on the canonical product tensor format, which has been applied to post-Hartree-Fock methods like MP2 and CCD, 44–46 leading to a reduction of the scaling in comparison to index based algorithms. Furthermore, we have applied the matrix product state format to represent two electron integrals and MP2 amplitudes and studied the similarities as well as the advantages and disadvantages in comparison with the CP format. 46,47

In the present proof-of-principle study, we present a FCI algorithm based on the canonical product tensor format, which is used to represent the Fock space. In Section II A we introduce the basic ideas of FCI. In Sections II B and II C it is shown how the representation of the coefficient and Hamiltonian tensors is constructed. Sections II D and II E focus on how the crucial steps of the tensor contraction

of the coefficient tensor with the Hamiltonian tensor and the denominator update are performed. Section II F gives an overview over the applied rank reduction algorithm and in Section II G an outline of the current algorithm is given including timings of the crucial steps. To estimate the reliability of our algorithm, results of benchmark calculations are presented in Section III A and the scaling behavior with increasing system size as well as the bottlenecks of the algorithm is discussed in Section III B.

#### **II. THEORY**

# A. Introduction of the FCI approach

The full CI wave function  $|\Phi^{(FCI)}\rangle$  is expanded as a linear combination of slater determinants,

$$|\Phi^{(FCI)}\rangle = \sum_{\mathbf{K}} C_{\mathbf{K}} |\mathbf{K}\rangle.$$
 (1)

Each slater determinant  $|\mathbf{K}\rangle$  can be represented by an occupation number vector (ONV)  $|\mathbf{K}\rangle$  containing the occupation numbers  $k_p$  of spin orbitals  $\chi_p$ . For a given basis of n molecular orbitals the ONV has the following form: <sup>48</sup>

$$|\mathbf{K}\rangle = |k_1, k_2, \dots, k_n\rangle$$
 with  $k_p = \begin{cases} 1 \text{ if } \chi_p \text{ is occupied} \\ 0 \text{ if } \chi_p \text{ is unoccupied} \end{cases}$ .

If all  $|\mathbf{K}\rangle$  constitute an orthonormal basis, the application of the variational principle in Eq. (1) leads to the following eigenvalue problem:

$$HC = EC. (3)$$

The elements of the Hamiltonian matrix **H** are given in Eq. (4), where  $\hat{H}$  is expressed in second quantisation,

$$\langle \mathbf{K} | \hat{H} | \mathbf{K}' \rangle = \langle \mathbf{K} | \left( \sum_{pp'} h(p, p') a_p^{\dagger} a_{p'} + \frac{1}{2} \sum_{pqp'q'} \langle pq | | p'q' \rangle a_p^{\dagger} a_q^{\dagger} a_{p'} a_{q'} \right) | \mathbf{K}' \rangle. \tag{4}$$

Here h(p, p') is the one-electron integral

$$h(p,p') = \int \chi_p^*(r_1) \left( -\frac{1}{2} \nabla_1^2 - \sum_{i} \frac{Z_A}{r_{A1}} \right) \chi_{p'}(r_1) dr_1 \quad (5)$$

and  $\langle pq||p'q'\rangle = \langle pq|p'q'\rangle - \langle pq|q'p'\rangle$  corresponds to the antisymmetrised two electron integral with

$$\langle pq|p'q'\rangle = \int \chi_p^*(r_1)\chi_q^*(r_2)\frac{1}{r_{12}}\chi_{p'}(r_1)\chi_{q'}(r_2)dr_1dr_2. \tag{6}$$

The creation and annihilation operators  $a_p^{\dagger}$  and  $a_p$  in second quantisation are defined as follows:<sup>48</sup>

$$a_{p}^{\dagger}|k_{1},k_{2},\ldots,0_{p},\ldots,k_{n}\rangle = \Gamma_{p}^{\mathbf{g}}|k_{1},k_{2},\ldots,1_{p},\ldots,k_{n}\rangle, 
a_{p}^{\dagger}|k_{1},k_{2},\ldots,1_{p},\ldots,k_{n}\rangle = 0, 
a_{p}|k_{1},k_{2},\ldots,1_{p},\ldots,k_{n}\rangle = \Gamma_{p}^{\mathbf{g}}|k_{1},k_{2},\ldots,0_{p},\ldots,k_{n}\rangle, 
a_{p}|k_{1},k_{2},\ldots,0_{p},\ldots,k_{n}\rangle = 0.$$
(7)
with  $\Gamma_{p}^{\mathbf{g}} = \prod_{q=1}^{p-1} (-1)^{k_{q}}.$ 

To obtain the proper coefficients in the wave function expansion (Eq. (1)) and the corresponding energies, the eigenvalue problem of Eq. (3) needs to be solved. In most cases only the lowest eigenvalue E, which corresponds to the

ground state energy, is of interest, so that only one coefficient vector *C* needs to be obtained.

# B. Canonical tensor representation of the Fock space

While many approaches can be formulated to apply tensor decomposition techniques to FCI, the aim of breaking the curse of dimensionality for a many-particle approach for fermions leads to difficulties in many cases. Indeed, we have tested two further approaches. In the first one, the wave function is expanded hierarchically. Starting from a reference determinant, additional determinants are incorporated into the wave function expansion, where orbitals that are occupied in a reference determinant are replaced by virtual orbitals. In this way, the coefficients belonging to determinants where one, two, ..., orbitals have been replaced are stored in tensors according to the number of replacements. However, this results in the problem that for higher substitutions the tensor contractions get more and more complicated. Therefore, we have tested a further approach that generalises the previous one by storing all coefficients in one representation. In this case there are still problems due to unphysical elements in the Hamiltonian tensor representation that were supposed to be zero when antisymmetry is correctly treated. In Ref. 49 we further discuss these two approaches.

In this section, we follow the ansatz by Legeza *et al.* and Szalay *et al.*<sup>50,51</sup> that will be discussed in this work as it does not lead to the aforementioned difficulties.

As shown in Eq. (2), the ONV possesses n entries  $k_p$ , which can each take one of two values. In order to address the proper coefficients, the corresponding coefficient tensor can be arranged in a way that follows a similar logic as is used for the ONVs. As a consequence, the coefficients are elements of an n dimensional tensor  $\mathfrak C$ . Each dimension  $\mu$  of this tensor can be assigned to one orbital  $|\chi_{\mu}\rangle$  and there exist two entries in each dimension corresponding to an occupied  $(k_{\mu}=1)$  or an unoccupied orbital  $(k_{\mu}=0)$ .

The coefficient tensor can be represented in the canonical product tensor format as shown in Eqs. (8) and (9) where for each of the J ranks there exist n representing vectors  $c_j^{(\mu)}$  and each representing vector holds two elements  $c_j^{(\mu)}(0)$  and  $c_j^{(\mu)}(1)$ ,

$$\mathfrak{C} = \sum_{i=1}^{J} c_j^{(1)} \otimes c_j^{(2)} \otimes \cdots \otimes c_j^{(n)}$$
 (8)

$$= \sum_{i=1}^{J} \bigotimes_{\mu=1}^{n} c_{j}^{(\mu)}. \tag{9}$$

Below, we illustrate the ansatz using a four-orbital (n = 4), two-electron (N = 2) example like  $H_2$ , which is also used later in this article. In order to use a transparent nomenclature, the coefficients  $c_{k_i,k_a,k_{\bar{i}},k_{\bar{a}}}$  are named in the same way as the ONVs, where  $k_i$ ,  $k_a$ ,  $k_{\bar{i}}$ , and  $k_{\bar{a}}$  are the corresponding occupation numbers of orbitals  $\chi_i$ ,  $\chi_a$ ,  $\chi_{\bar{i}}$ , and  $\chi_{\bar{a}}$ . As shown in Eq. (1), to each determinant can be assigned one coefficient leading in the present example to a total number of

 $\binom{n}{N} = \frac{4 \cdot 3}{2 \cdot 1} = 6$  coefficients, which are stored in a vector  $\vec{c}$  in conventional FCI as shown below

$$\vec{c} = \begin{pmatrix} c_{0,1,0,1} \\ c_{1,0,0,1} \\ c_{0,1,1,0} \\ c_{1,1,0,0} \\ c_{0,0,1,1} \\ c_{1,0,1,0} \end{pmatrix}.$$
 (10)

The same configuration space can be spanned in a CP format representation of an n dimensional tensor as follows:

$$\mathfrak{C} = \sum_{i=1}^{J} c_j^{(i)} \otimes c_j^{(a)} \otimes c_j^{(\bar{i})} \otimes c_j^{(\bar{a})}, \tag{11}$$

where each coefficient  $c(k_i, k_{\bar{i}}, k_a, k_{\bar{a}})$  can be obtained from

$$c(k_i, k_a, k_{\bar{i}}, k_{\bar{a}}) = \sum_{i=1}^{J} c_j^{(i)}(k_i) \cdot c_j^{(a)}(k_a) \cdot c_j^{(\bar{i})}(k_{\bar{i}}) \cdot c_j^{(\bar{a})}(k_{\bar{a}}).$$
 (12)

Suppose that only one coefficient c(1,0,1,0) that belongs to the Hartree-Fock determinant is equal to one while all the other coefficients are zero as, for example, in the initial guess of an FCI procedure. In this case the vector  $\vec{c}$  contains six elements

$$\vec{c} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \tag{13}$$

while in the representation of the coefficient tensor it takes the form of a rank one expansion,

$$\mathfrak{C} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix}. \tag{14}$$

Note that such a tensor representation covers the whole Fock space instead of an N particle subspace of the Fock space. For example, elements belonging to the vacuum state  $|0,0,0,0\rangle$  are accessible in the representation as

$$c(0,0,0,0) = \sum_{i=1}^{J} c_j^{(i)}(0) \cdot c_j^{(a)}(0) \cdot c_j^{(\bar{i})}(0) \cdot c_j^{(\bar{a})}(0). \quad (15)$$

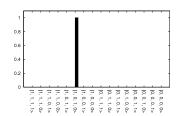
There are many instances of coefficient tensors that can occur for a four-orbital, two-electron system. Some examples of representations of the coefficient tensors are given in Fig. 1.

The wave function is obtained by inserting the coefficient tensor representation (8) into Eq. (1),

$$|\Phi^{(FCI)}\rangle = \sum_{k} C(k_1, k_2, \dots, k_n) |k_1, k_2, \dots, k_n\rangle$$
(16)  
$$= \sum_{k} \sum_{j=1}^{J} c_j^{(1)}(k_1) \cdot c_j^{(2)}(k_2) \cdot \dots$$
$$\cdot c_j^{(n)}(k_n) |k_1, k_2, \dots, k_n\rangle.$$
(17)

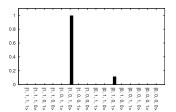
# **HF-Solution of H**<sub>2</sub>

$$\begin{split} |\Psi\rangle &= 1\cdot |1,0,1,0\rangle \\ \mathfrak{C} &= \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} \end{split}$$



#### FCI-Solution of H<sub>2</sub>

$$\begin{split} |\Psi\rangle &= 0.99 \cdot |1,0,1,0\rangle - 0.11 \cdot |0,1,0,1\rangle \\ \mathfrak{C} &= \left( \begin{array}{c} 0 \\ 0.99 \end{array} \right) \otimes \left( \begin{array}{c} 1 \\ 0 \end{array} \right) \otimes \left( \begin{array}{c} 0 \\ 1 \end{array} \right) \otimes \left( \begin{array}{c} 1 \\ 0 \end{array} \right) \\ &+ \left( \begin{array}{c} -0.11 \\ 0 \end{array} \right) \otimes \left( \begin{array}{c} 0 \\ 1 \end{array} \right) \otimes \left( \begin{array}{c} 0 \\ 1 \end{array} \right) \otimes \left( \begin{array}{c} 0 \\ 1 \end{array} \right) \end{split}$$



### **Equal Coefficients for all Determinants in the Fock-Space**

$$\begin{split} |\Psi\rangle &= \sum_{k_i} \sum_{k_a} \sum_{k_{\bar{i}}} \sum_{k_{\bar{a}}} 1 \cdot |k_i, k_a, k_{\bar{i}}, k_{\bar{a}}\rangle \\ \mathfrak{C} &= \begin{pmatrix} 1\\1 \end{pmatrix} \otimes \begin{pmatrix} 1\\1 \end{pmatrix} \otimes \begin{pmatrix} 1\\1 \end{pmatrix} \otimes \begin{pmatrix} 1\\1 \end{pmatrix} \end{split}$$



#### **Equal Coefficients for all two-electron Determinants**

$$\begin{split} |\Psi\rangle &= 1\cdot |1,0,1,0\rangle + 1\cdot |0,1,0,1\rangle + 1\cdot |0,1,1,0\rangle \\ &= 1\cdot |1,0,0,1\rangle + 1\cdot |1,1,0,0\rangle + 1\cdot |0,0,1,1\rangle \\ \mathfrak{C} &= \begin{pmatrix} 0.58 \\ 1.22 \end{pmatrix} \otimes \begin{pmatrix} 0.58 \\ 1.22 \end{pmatrix} \\ &+ \begin{pmatrix} 0.58 \\ -1.22 \end{pmatrix} \otimes \begin{pmatrix} -0.58 \\ 1.22 \end{pmatrix} \otimes \begin{pmatrix} 0.58 \\ -1.22 \end{pmatrix} \otimes \begin{pmatrix} -0.58 \\ 1.22 \end{pmatrix} \\ &+ \begin{pmatrix} 0 \\ 1.45 \end{pmatrix} \otimes \begin{pmatrix} 0 \\ -1.45 \end{pmatrix} \otimes \begin{pmatrix} 0 \\ -1.45 \end{pmatrix} \otimes \begin{pmatrix} 0 \\ -1.45 \end{pmatrix} \\ &+ \begin{pmatrix} 0.69 \\ 0 \end{pmatrix} \otimes \begin{pmatrix} -0.69 \\ 0 \end{pmatrix} \otimes \begin{pmatrix} -0.69 \\ 0 \end{pmatrix} \otimes \begin{pmatrix} -0.69 \\ -0 \end{pmatrix} \end{split}$$

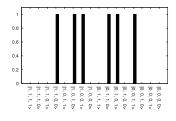


FIG. 1. Examples of coefficient tensor representations of the four-orbital, two-electron example. Shown are an explicit form of the wave function using ONVs, the representation of the coefficient tensor, and a bar plot of the absolute values of the coefficients.

Using the tensor representation of the coefficient tensor for the aforementioned four-orbital, two-electron example would yield the explicit form of the wave function

$$|\Phi^{(FCI)}\rangle = \sum_{k} \sum_{j=1}^{J} c_{j}^{(1)}(k_{1}) \cdot c_{j}^{(2)}(k_{2}) \cdot c_{j}^{(3)}(k_{3})$$
$$\cdot c_{j}^{(4)}(k_{4})|k_{1}, k_{2}, k_{3}, k_{4}\rangle. \tag{18}$$

# C. Decomposed representation of the Hamiltonian matrix using canonical products

As introduced in Section II A, to obtain the appropriate coefficients, an eigenvalue problem of a matrix  ${\bf H}$  needs to be solved. To find a tensor representation of the Hamiltonian matrix, the representation needs to be constructed in an analogous way as the matrix elements in Eq. (4). In a first step tensor representations are set up that are analogues of the annihilation and creation operators  $a_p^{\dagger}$  and  $a_p$ . Therefore,

matrices  $A^T$ , A, 1, and G are introduced,  $^{50,51}$ 

$$\mathbf{A}^T = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix},\tag{19}$$

$$\mathbf{A} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix},\tag{20}$$

$$\mathbf{G} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix},\tag{21}$$

$$\mathbf{1} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}. \tag{22}$$

Using these matrices, tensors  $\mathfrak{A}_p^\dagger$  and  $\mathfrak{A}_p$  representing the creation and annihilation operators can be constructed as follows:

$$\mathfrak{A}_{p}^{\dagger} = \mathbf{G}^{(1)} \otimes \mathbf{G}^{(2)} \otimes \cdots \otimes \mathbf{G}^{(p-1)}$$

$$\otimes \mathbf{A}^{(p)T} \otimes \mathbf{1}^{(p+1)} \otimes \cdots \otimes \mathbf{1}^{(n)},$$

$$\mathfrak{A}_{p} = \mathbf{G}^{(1)} \otimes \mathbf{G}^{(2)} \otimes \cdots \otimes \mathbf{G}^{(p-1)}$$
(23)

$$\otimes \mathbf{A}^{(p)} \otimes \mathbf{1}^{(p+1)} \otimes \cdots \otimes \mathbf{1}^{(n)}.$$
 (24)

In analogy to Eqs. (7), these tensors can be contracted with tensor representations of the occupation number vectors. Again, consider the four-orbital, two-electron example. Application of a creation operator  $a_a^{\dagger}$  on the ONV  $|1,0,1,0\rangle$  gives

$$a_a^{\dagger}|1,0,1,0\rangle = -1|1,1,1,0\rangle.$$
 (25)

If a tensor representation & of this occupation number vector

$$\mathfrak{E} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} \tag{26}$$

is used, the elements of the well known form of the ONV are obtained by setting the occupation numbers  $k_p$  equal to the elements  $e^{(p)}(1)$  of the representing vectors. In analogy to the operation shown in Eq. (25) a corresponding tensor contraction  $\mathfrak{A}_{\sigma}^{\dagger}\mathfrak{E}$  is done as follows:

$$\mathfrak{A}_{a}^{\dagger}\mathfrak{E} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\otimes \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \tag{27}$$

$$= \begin{pmatrix} 0 \\ -1 \end{pmatrix} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix}. \tag{28}$$

Note that the resulting representation of an ONV is exactly the same as the well known form of the ONV, where the prefactor is multiplied with the representing vector of the first dimension. In order to set up the Hamiltonian tensor, the contributions of all one- and two-electron integrals from all ONVs  $\langle \mathbf{K}|$  and  $|\mathbf{K}'\rangle$  need to be included. For simplicity, just consider the one-electron part  $\sum_{pp'}h(p,p')a_p^{\dagger}a_{p'}$  of Eq. (4). In analogy to the term  $h(p,p')a_p^{\dagger}a_{p'}$ , a rank one tensor representation  $\mathfrak{S}^{pp'}$  can be set up that contains only elements  $\langle \mathbf{K}|h(p,p')a_p^{\dagger}a_{p'}|\mathbf{K}'\rangle$  as in Eq. (29). The contraction

is done as shown in Eq. (30) where the representations of  $\mathfrak{A}_p^{\dagger}$  and  $\mathfrak{A}_{p'}$  from Eqs. (23) and (24) are applied, and the representing matrices are multiplied with each other in each dimension,

$$\mathfrak{H}^{pp'} = h(p,p')\mathfrak{A}_{p}^{\dagger}\mathfrak{A}_{p'}$$

$$= h(p,p') \cdot \mathbf{G}^{(1)}\mathbf{G}^{(1)} \otimes \cdots \otimes \mathbf{G}^{(p'-1)}\mathbf{G}^{(p'-1)}$$

$$\otimes \mathbf{G}^{(p')}\mathbf{A}^{(p')} \otimes \mathbf{G}^{(p'+1)}\mathbf{1}^{(p'+1)} \otimes \cdots \otimes \mathbf{G}^{(p-1)}\mathbf{1}^{(p-1)}$$

$$\otimes \mathbf{A}^{(p)T}\mathbf{1}^{(p)} \otimes \mathbf{1}^{(p+1)}\mathbf{1}^{(p+1)} \otimes \cdots \otimes \mathbf{1}^{(n)}\mathbf{1}^{(n)}.$$
(30)

The contributions of the two electron integrals to all elements of the Hamiltonian tensor can be obtained in the same way as for the one-electron integrals. Therefore, for each one- and two-electron integral, a rank one tensor can be constructed in the same manner.

The four-orbital, two-electron example introduced above yields tensor representations with four representing matrices for each one- and two-electron integral. Two examples  $\mathfrak{H}^{ia}$  and  $\mathfrak{H}^{i\bar{i}i\bar{i}}$  of such tensor representations are given below. These representations  $\mathfrak{H}^{ia}$  and  $\mathfrak{H}^{i\bar{i}i\bar{i}}$  constitute tensors that possess only elements equal to  $\langle \mathbf{K} | h(i,a) a_i^{\dagger} a_a | \mathbf{K}' \rangle$  or  $\langle \mathbf{K} | \langle i\bar{i} | | i\bar{i} \rangle a_i^{\dagger} a_i^{\dagger} a_i a_i | \mathbf{K}' \rangle$ , respectively, where  $\langle \mathbf{K} |$  and  $| \mathbf{K}' \rangle$  denote arbitrary ONVs,

$$\mathfrak{H}^{ia} = \underbrace{h(i,a) \cdot \mathbf{A}^{(i)T} \mathbf{G}^{(i)}}_{\mathbf{h}_{ia}^{(i)}} \otimes \underbrace{\mathbf{1}^{(a)} \mathbf{A}^{(a)}}_{\mathbf{h}_{ia}^{(a)}} \otimes \underbrace{\mathbf{1}^{(\bar{i})} \mathbf{1}^{(\bar{i})}}_{\mathbf{h}_{ia}^{(\bar{i})}} \otimes \underbrace{\mathbf{1}^{(\bar{a})} \mathbf{1}^{(\bar{a})}}_{\mathbf{h}_{ia}^{(\bar{a})}},$$
(31)

$$= \begin{pmatrix} 0 & 0 \\ h(i,a) & 0 \end{pmatrix} \otimes \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$
(32)  

$$\mathfrak{J}^{i\bar{i}t\bar{i}} = \underbrace{\langle i\bar{i} | | i\bar{i} \rangle \cdot \mathbf{A}^{(i)T} \mathbf{G}^{(i)} \mathbf{G}^{(i)} \mathbf{A}^{(i)}}_{h_{i\bar{i}t\bar{i}}} \otimes \underbrace{\mathbf{1}^{(a)} \mathbf{G}^{(a)} \mathbf{G}^{(a)} \mathbf{1}}_{h_{i\bar{i}t\bar{i}}}$$

$$\otimes \underbrace{\mathbf{1}^{(\bar{i})} \mathbf{A}^{(\bar{i})T} \mathbf{A}^{(\bar{i})} \mathbf{1}^{(\bar{i})}}_{\mathbf{h}_{i\bar{i}i\bar{i}}^{(\bar{i})}} \otimes \underbrace{\mathbf{1}^{(\bar{a})} \mathbf{1}^{(\bar{a})} \mathbf{1}^{(\bar{a})} \mathbf{1}^{(\bar{a})}}_{\mathbf{h}_{i\bar{i}i\bar{i}}^{(\bar{a})}}$$
(33)

$$= \begin{pmatrix} 0 & 0 \\ 0 & \langle i\bar{i}| |i\bar{i}\rangle \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}. \quad (34)$$

As mentioned above, the contraction of the representing matrices in each dimension leads to new matrices that are named  $\mathbf{h}_{ia}^{(i)}$ ,  $\mathbf{h}_{ia}^{(a)}$ ,  $\mathbf{h}_{ia}^{(\bar{i})}$ , and  $\mathbf{h}_{ia}^{(\bar{a})}$  for the representation of the one-electron integral h(i,a). The same is done for the two-electron integral contribution.

Now it becomes clear that for all integrals h(p,p') and  $\langle pq | | p'q' \rangle$  that occur in an N-electron, n-orbital system, a rank one tensor is set up with n representing matrices. By summing up all rank one representing tensors of the one- and two- electron contributions, a tensor representation  $\mathfrak{H}$  is built. This representation becomes the *Hamiltonian tensor* given in Eqs. (35)–(38) and corresponds to a 2n dimensional tensor representation of the Hamiltonian matrix. This representation of matrices by decomposed tensors is similar to the decomposition shown by Beylkin. So Note again the similarity between Eqs. (4) and (35),

$$\mathfrak{H} = \sum_{p=1}^{n} \sum_{p'=1}^{n} \mathfrak{H}_{pp'}^{pp'} + \sum_{p=1}^{n} \sum_{p'=1}^{n} \sum_{q=1}^{p} \sum_{q'=1}^{p'} \mathfrak{H}_{pqp'q'}^{pp'}$$

$$= \sum_{p=1}^{n} \sum_{p'=1}^{n} \mathbf{h}_{pp'}^{(1)} \otimes \mathbf{h}_{pp'}^{(2)} \otimes \cdots \otimes \mathbf{h}_{pp'}^{(n)}$$

$$+ \sum_{p=1}^{n} \sum_{p'=1}^{n} \sum_{q=1}^{p} \sum_{q'=1}^{p'} \mathbf{h}_{pqp'q'}^{(1)} \otimes \mathbf{h}_{pqp'q'}^{(2)} \otimes \cdots \otimes \mathbf{h}_{pqp'q'}^{(n)}$$
(36)

$$= \sum_{k=1}^{K} \mathbf{h}_{k}^{(1)} \otimes \mathbf{h}_{k}^{(2)} \otimes \cdots \otimes \mathbf{h}_{k}^{(n)}$$
(37)

$$=\sum_{k=1}^{K}\bigotimes_{\mu=1}^{n}\mathbf{h}_{k}^{(\mu)}.$$
(38)

Because all one- and two-electron integral contributions are taken into account, this representation is exact. The rank K of this representation is dominated by the two electron contributions, as their number rises with  $n^4$  (if no sparsity in the integrals is assumed). Because of the  $n^4$  dependence of the rank and the  $n \ 2 \times 2$  representing matrices, the overall scaling of this representation with the system size is  $\sim O(N^5)$  under the assumption that the number of molecular orbitals n is proportional to the number of electrons N. This is a significant reduction from a factorial scaling of a  $\binom{n}{N} \times \binom{n}{N}$  matrix to a  $\sim O(N^5)$  scaling of an n dimensional tensor.

For the four-orbital, two-electron example each element  $\langle k_i, k_a, k_{\bar{i}}, k_{\bar{a}}|\hat{H}|k'_i, k'_a, k'_{\bar{i}}, k'_{\bar{a}}\rangle$  of the Hamiltonian matrix can be obtained via the corresponding ONVs,

$$\langle k_{i}, k_{a}, k_{\bar{i}}, k_{\bar{a}} | \hat{H} | k'_{i}, k'_{a}, k'_{\bar{i}}, k'_{\bar{a}} \rangle = \sum_{k=1}^{K} h_{k}^{(i)}(k_{i}, k'_{i}) \cdot h_{k}^{(a)}(k_{a}, k'_{a}) \cdot h_{k}^{(\bar{i})}(k_{\bar{i}}, k'_{\bar{i}}) \cdot k_{k}^{(\bar{a})}(k_{\bar{a}}, k'_{\bar{a}}).$$
(39)

Again, it is important to mention that the present representation covers the whole Fock space of a given basis and not just an N electron subspace. Thus, for the aforementioned example also elements that even belong to the vacuum state  $|0,0,0,0\rangle$  (N=0) are accessible and are not necessarily equal to zero as shown below,

 $\langle 0, 0, 0, 0 | \hat{H} | 0, 0, 0, 0 \rangle$ 

$$= \sum_{k=1}^{K} h_k^{(i)}(0,0) \cdot h_k^{(a)}(0,0) \cdot h_k^{(\bar{i})}(0,0) \cdot h_k^{(\bar{a})}(0,0). \tag{40}$$

# D. Tensor contraction

As known from conventional algorithms, one of the crucial steps in FCI algorithms is the contraction  $\sigma = HC.^{48}$  This means that every element  $\sigma_K$  is obtained by contracting the Hamiltonian matrix with the coefficient vector,

$$\sigma_{\mathbf{K}} = \sum_{\mathbf{K}'} \langle \mathbf{K} | \hat{H} | \mathbf{K}' \rangle C_{\mathbf{K}'}. \tag{41}$$

However, in the present approach this step is considerably simplified because the contraction needs to be done only for the representing matrices and vectors rather than of the whole tensors. Furthermore, the coefficient tensor retains the same structure as an ONV in the tensor representation so that each coefficient can be addressed in the same way. In order to illustrate the contraction, the four-orbital, two-electron example is used again. Here the ONVs are given explicitly instead of  $\mathbf{K}'$  and  $\mathbf{K}$ . By inserting the explicit elements of the Hamiltonian and the coefficient tensors (Eqs. (39) and (12)) into Eq. (41) and rearranging the summations, the contraction in the chosen tensor format is carried out by performing simple matrix-vector multiplications in each dimension,

$$\sigma_{k_{i},k_{a},k_{\bar{i}},k_{\bar{a}}} = \sum_{k'_{i}} \sum_{k'_{a}} \sum_{k'_{\bar{i}}} \sum_{k'_{\bar{a}}} \langle k_{i},k_{a},k_{\bar{i}},k_{\bar{a}} | \hat{H} | k'_{i},k'_{a},k'_{\bar{i}},k'_{\bar{a}} \rangle$$

$$\cdot C_{k'_{i},k'_{a},k'_{i},k'_{\bar{a}}} \qquad (42)$$

$$= \sum_{k'_{i}} \sum_{k'_{a}} \sum_{k'_{\bar{i}}} \sum_{k'_{\bar{a}}} \sum_{k=1}^{K} h_{k}^{(i)}(k_{i},k'_{i}) \cdot h_{k}^{(a)}(k_{a},k'_{a})$$

$$\cdot h_{k}^{(\bar{i})}(k_{\bar{i}},k'_{\bar{i}}) \cdot h_{k}^{(\bar{a})}(k_{\bar{a}},k'_{\bar{a}})$$

$$\cdot \sum_{j=1}^{J} c_{j}^{(i)}(k'_{i}) \cdot c_{j}^{(a)}(k'_{a}) \cdot c_{j}^{(\bar{i})}(k'_{\bar{i}}) \cdot c_{j}^{(\bar{a})}(k'_{\bar{a}}) \qquad (43)$$

$$= \sum_{k=1}^{K} \sum_{j=1}^{J} \sum_{k'_{i}} h_{k}^{(i)}(k_{i},k'_{i})c_{j}^{(i)}(k'_{i}) \cdot \sum_{k'_{a}} h_{k}^{(a)}(k_{a},k'_{a})$$

$$\cdot c_{j}^{(a)}(k'_{a}) \cdot \sum_{k'_{\bar{i}}} h_{k}^{(\bar{i})}(k_{\bar{i}},k'_{\bar{i}})c_{j}^{(\bar{i})}(k'_{\bar{i}})$$

$$\cdot \sum_{k'_{a}} h_{k}^{(\bar{a})}(k_{\bar{a}},k'_{\bar{a}})c_{j}^{(\bar{a})}(k'_{\bar{a}}) \qquad (44)$$

$$= \sum_{m=1}^{M=J-K} s_{m}^{(i)}(k_{i}) \cdot s_{m}^{(a)}(k_{a}) \cdot s_{m}^{(\bar{i})}(k'_{\bar{i}}) \cdot s_{m}^{(\bar{a})}(k'_{\bar{a}}). \qquad (45)$$

A more general form of the contraction is shown in Eq. (46). For each dimension  $\mu$ , the contraction  $\mathbf{h}_k^{(\mu)}c_j^{(\mu)}$  must be carried out for all ranks of the Hamiltonian tensor K and for all ranks of the coefficient tensor J. Hence, the rank M of the resulting tensor  $\mathfrak S$  is increased and needs to be reduced afterwards,

$$\mathfrak{S} = \sum_{k=1}^{K} \sum_{i=1}^{J} \bigotimes_{j=1}^{n} \left( \mathbf{h}_{k}^{(\mu)} c_{j}^{(\mu)} \right) \tag{46}$$

$$=\sum_{m=1}^{M=J\cdot K} \bigotimes_{\mu=1}^{n} s_{m}^{(\mu)}.$$
 (47)

The scaling of this contraction will be  $K \cdot J \cdot n$ . Note that there is no further logic required for the ordering of coefficients or Hamiltonian tensor elements.

Beside this contraction in a decomposed format, further mathematical operations in a decomposed format are required for an eigenvalue solver. For further operations such as analogous to the scalar product or summations, interested readers are referred to the corresponding literature. <sup>53,54</sup>

#### E. Denominator update

The Davidson algorithm<sup>55</sup> requires the construction of a matrix  $(\mathbf{H} - \lambda Id)^{-1}$ . To set up a tensor representation of this matrix, some assumptions are made because there exists no

trivial solution. Because a representation of such a tensor in n dimensions should possess a low rank, the representation is approximated by exponential sums. Therefore, the diagonal dominance of Hamiltonian matrix is exploited by considering only the elements  $\langle \Psi_K | \hat{H} | \Psi_K \rangle$  in the representation. All other elements are assumed to be zero. For example, one element  $(H_{i,\bar{i}} - \lambda Id)^{-1}$  for the four-orbital, two-electron case is given below

$$(\langle 1, 0, 1, 0 | \hat{H} | 1, 0, 1, 0 \rangle - \lambda Id)^{-1}$$

$$= \frac{1}{\epsilon_i + 0 + \epsilon_{\bar{i}} + 0 - \lambda}$$
(48)

$$= \int_{0}^{\infty} e^{-(\epsilon_{i}+0+\epsilon_{\bar{i}}+0-\lambda)t} dt$$

$$\approx \sum_{s} \omega_{s} \cdot e^{(\alpha_{s}\lambda)} \cdot e^{(-\alpha_{s}\cdot\epsilon_{i})}$$
(49)

$$\cdot e^{(-\alpha_s \cdot 0)} \cdot e^{(-\alpha_s \cdot \epsilon_{\bar{i}})} \cdot e^{(-\alpha_s \cdot 0)} \tag{50}$$

$$\approx \sum_{s} \omega_{s} \cdot e^{(\alpha_{s}\lambda)} \cdot e^{(-\alpha_{s}\cdot\epsilon_{i})} \cdot 1 \cdot e^{(-\alpha_{s}\cdot\epsilon_{i})} \cdot 1. \quad (51)$$

More generally, the tensor approximation for the inverse is

$$(\mathfrak{H} - \lambda \mathfrak{Ib})^{-1} \approx \sum_{s=1}^{S} \bigotimes_{\mu=1}^{n} i_{s}^{(\mu)}.$$
 (52)

The representing vectors consist of the following elements:

$$i_s^{(x\neq 1)} = \begin{pmatrix} e^{(-\alpha_s \cdot \epsilon_x)} \\ e^0 \end{pmatrix}$$

and

$$i_s^{(1)} = \begin{pmatrix} \omega_s \cdot e^{(\alpha_s \cdot \lambda)} \cdot e^{(-\alpha_s \cdot \epsilon_1)} \\ \omega_s e^{(\alpha_s \cdot \lambda)} \end{pmatrix}. \tag{53}$$

The use of exponential sums allows the representation of a sum in the denominator by a sum of products where each summand carries one index. However, the diagonal elements of the Hamiltonian matrix contain a contribution of the two electron integrals that still depend on two indices  $\langle pq | |pq \rangle$ , meaning a connection of two dimensions. Because of this, the exponential sums currently do not allow the inclusion of the two electron integrals. Therefore, the orbital energies  $\epsilon_i$  are used as a compromise to include the contribution of electron-electron interactions.

# F. Rank reduction—Pivotised alternating steepest descent procedure

The efficient approximation of quantities by low-rank tensor representations is a crucial step in our approach. Usually, the numerical solution of this approximation problem needs most of the computational time of the whole simulation (see Scheme 2). While in a previous work we have applied an alternating least square scheme, in the following we outline a more efficient alternative, which is named Pivotised Alternating Steepest Descend (PASD) and which is applied in all rank reduction steps.

Let  $\mathfrak{S} = \sum_{i=1}^R \bigotimes_{\mu=1}^n s_i^{(\mu)}$  and r < R be given. We are looking for  $\hat{\xi} = \sum_{j=1}^r \bigotimes_{\mu=1}^d \hat{\xi}_j^{(\mu)}$  such that

$$\|\mathfrak{S} - \hat{\xi}\| = \min_{\xi \in \mathcal{T}_r} \|\mathfrak{S} - \xi\|,\tag{54}$$

under the constraints

$$\|\hat{\xi}_{j}^{(\mu)}\| = \|\hat{\xi}_{j}^{(\nu)}\|, \quad \text{for all } \mu, \nu \in \{1, \dots, d\}, \ j \in \{1, \dots, r\},$$

$$(55)$$

where  $\mathcal{T}_r$  denotes the set of tensors which can be represented with r bounded terms. The our applications it is a priori not obvious how to choose the size or the representation rank r. Rather, a desired approximation accuracy  $\varepsilon$  is of importance. Hence, the following extended approximation problem is stated. For a given accuracy  $\varepsilon > 0$  we have to find minimal  $r_\varepsilon \leq R$  and  $\hat{\xi}_\varepsilon \in \mathcal{T}_{r_\varepsilon}$  such that

$$\|\mathfrak{S} - \hat{\xi}_{\varepsilon}\| \le \varepsilon,\tag{56}$$

$$\|\mathfrak{S} - \hat{\xi}_{\varepsilon}\| = \min_{\xi \in \mathcal{T}_{r_{\varepsilon}}} \|\mathfrak{S} - \xi\|, \tag{57}$$

under the constraints from Eq. (55). The solution of the extended approximation problem is closely related to the choice of the initial guess for an efficient optimisation method. Here, we are following a scheme which was introduced in Refs. 58 and 59, where the extended optimisation problem was solved by successive use of a low-rank approximation method. With the specific definition of the initial guess described in Refs. 58 and 59, we ensure that the approximation error will not increase during the low-rank approximation of  $\mathfrak{S} = \sum_{i=1}^R \bigotimes_{\mu=1}^n s_i^{(\mu)}$ . For a fixed target rank r, the approximation problem from Eq. (54) is solved by the pivotised alternating steepest descent (PASD) method. Since the low-rank approximation in the canonical tensor format is not well defined, we have to consider the following stabilised target function:

$$f_{\lambda}(\xi_1^{(1)}, \dots, \xi_r^{(d)}) = \frac{1}{2} \|\xi\|^2 - \langle \xi, b \rangle + \frac{\lambda}{2} \sum_{i=1}^r \prod_{\mu=1}^d \|\xi_j^{(\mu)}\|^2, \quad (58)$$

where  $\lambda$  is a positive real number. In Scheme 1 we briefly outline the PASD method, for details and convergence analysis we refer the reader to Ref. 60.

Choose the initial guess as described in [58, 59] and define k := 1. while Stop Condition

$$\mu := \operatorname{argmax}_{1 \le \nu \le d} \left\| \frac{\partial f_{\lambda}}{\partial \xi_{\nu}}(\xi_{k,\nu}) \right\|_{\infty}$$

$$\begin{split} \xi_{k,\mu} &:= (\xi_1^{k+1}, \dots, \xi_{\mu-1}^{k+1}, \xi_{\mu+1}^k, \dots, \xi_d^k) \\ d_{\mu}^k &:= \frac{\partial f_{\lambda}}{\partial \xi_{\nu}} (\xi_{k,\nu}) \\ \lambda_{\mu}^k &:= \frac{\left\langle d_{\mu}^k, d_{\mu}^k \right\rangle}{\left\langle G_{k,\mu} d_{\mu}^k, d_{\mu}^k \right\rangle} \\ \xi_{\mu}^{k+1} &:= \xi_{\mu}^k - \lambda_{\mu}^k d_{\mu}^k \\ \xi_{k,\mu+1} &:= (\xi_1^{k+1}, \dots, \xi_{\mu}^{k+1}, \xi_{\mu+1}^k, \dots, \xi_d^k) \end{split}$$

balance the representation system  $\xi_{k,\mu+1}$  such that Eq. (55) is satisfied  $k \mapsto k + 1$ 

SCHEME 1. Pivotised Alternating Steepest Descent Algorithm (PASD).

```
Create C_1 with ||C_1|| = 1, K = 1
                                                                                                         rank
                                                                                                                                          time in s[61]
\tilde{\sigma}_1 = \mathbf{H}C_1, \, \sigma_1 = \mathrm{Approx}(\tilde{\sigma}_1)
                                                                                                         rank(\mathbf{H}) = 918
\lambda_1 = \langle \sigma_1, C_1 \rangle, r_1 = (\sigma_1 - \lambda_1 C_i)
while (\|res_K\| > \epsilon_{res})
          \delta = -(H - \lambda^K Id)^{-1} res_K
                                                                                                                                           0.30
          \tilde{C}_{K+1} = \prod_{i=1}^{K} (\mathbf{1} - C_i C_i^T) \delta
                                                                                                         rank(\tilde{C}_{K+1}) = 594
          C_{K+1} = \operatorname{Approx}(\tilde{C}_{K+1})
                                                                                                         \operatorname{rank}(C_{K+1}) = 44
                                                                                                                                          2.62
          C_{K+1}/=\|C_{K+1}\|
           \tilde{\sigma}_{K+1} = \mathbf{H}C_{K+1}
                                                                                                         rank(\tilde{\sigma}_{K+1}) = 5120 \ 0.99
                                                                                                        rank(\sigma_{K+1}) = 109 219.00
          \sigma_{K+1} = \operatorname{Approx}(\tilde{\sigma}_{K+1})
          \lambda_{K+1} = \langle \sigma_{K+1}, C_{K+1} \rangle \ r_{K+1} = (\sigma_{K+1} - \lambda_{K+1} C_{K+1})
          for i=1 \dots K
            for j=1...K
                                                                                                                                          0.01
             A_{i,j} = \langle \sigma_j, C_i \rangle,
           Diagonalise A, store the lowest eigenvalue \lambda_K
          and the corresponding eigenvector \alpha
          res_K = \sum_{i=1}^{K} (\alpha_i \sigma_i - \alpha_i \lambda_K C_i)
```

SCHEME 2. Workflow of the Davidson algorithm currently used as FCI eigenvalue solver (left) with the ranks (middle) and the timings in seconds (right) of the corresponding steps for the example of (H<sub>2</sub>)<sub>3</sub> given for iteration 2.

### G. Application of the Davidson algorithm

The Davidson algorithm is used to solve the eigenvalue problem posed in FCI. This algorithm constitutes a very efficient algorithm, since it was developed for sparse and diagonal dominant matrices. The eigenvector is expanded as a linear combination in an orthonormal basis. In each iteration the basis is extended by adding an additional vector, which is derived from the residual vector of the eigenvalue problem. An overview over the algorithm is given in Scheme 2. For further details, see Ref. 55.

In the present implementation the algorithm proceeds as follows: in a first step a guess tensor  $C_1$  is set up which contains only one coefficient equal to one. This coefficient is assigned to the Hartree-Fock-Slater determinant. All other coefficients are equal to zero (an example for such a tensor is given in Eq. (14) and in Fig. 1). Next, the coefficient tensor is contracted with the Hamiltonian tensor as discussed in Section II D. Subsequently, the rank of the resulting tensor  $\tilde{\sigma}_1$  is reduced to  $\sigma_1$ . Note that this rank reduction is the most time consuming step in all iterations. The scalar product  $\langle \sigma_1, C_1 \rangle$  provides the first approximation  $\lambda_1$  to the eigenvalue and using this value the residual res<sub>1</sub> will be calculated. If the norm of the residual is larger than the threshold  $\epsilon_{res}$ , a tensor  $\delta$ is constructed. This tensor is orthogonalised to all coefficient tensors of the previous iterations. After normalisation the tensor is added to the basis in which the coefficient tensor is expanded. Now, the contraction with the Hamiltonian tensor and the rank reduction is done in the same way as shown for the first iteration. A matrix A is constructed whose elements are obtained by the scalar products of all coefficient tensors with all  $\sigma$ -tensors. The eigenvalue  $\lambda_{K+1}$  constitutes the next approximation of the desired eigenvalue while the elements of the eigenvectors are used as the expansion coefficients for the desired coefficient tensors in a tensor representation. Using these coefficients, a new residual is constructed. In case that the residual  $res_K$  does not fulfill the convergence criterion  $||res_K|| > \epsilon_{res}$ , a new iteration is initiated.

#### **III. RESULTS AND DISCUSSION**

#### A. Benchmarks

Below we present benchmark calculations of a pilot implementation and investigate the accuracy, the convergence behavior, and the scaling of the tensor decomposed FCI algorithm.

The goal of the present algorithm is to achieve high accuracy in the calculation of the energy of the systems—the so called "chemical accuracy." This means that the deviations from the exact energies should lie in the range of  $10^{-3}$  E<sub>h</sub>. In order to set the convergence criterion  $\epsilon_{res}$  (see Scheme 2) appropriately, the correlation between the accuracy of the energy and  $||res_K||$  needs to be investigated.

Using a test set of molecules it is confirmed that  $||res_K||$  is about hundred times larger than the deviation of the energy from the exact value (in  $E_h$ ) (see Table I). Therefore, to obtain results in the range of the "chemical accuracy," the iterations should continue until  $||res_K||$  is smaller than  $10^{-1}$ .

TABLE I. Deviation of the energy from the exact eigenvalues  $\Delta E$  for various convergence thresholds  $\epsilon_{res}$  for various examples <sup>62</sup> that were calculated without any tensor approximation using the STO-3G basis set. <sup>63</sup>

Example	$\Delta E$ in $E_h$ At $  res_K   \le 10^{-1}$	$\Delta E \text{ in } E_h$ At $  res_K   \le 10^{-2}$		
H <sub>2</sub>	0.000 001	0.000 001		
He	0.000 584	0.000 019		
$(H_2)_2$	0.000 532	0.000 004		
He <sub>2</sub>	0.000 165	0.000 007		
$(H_2)_3$	0.001 535	0.000 010		
He <sub>3</sub>	0.000 440	0.000 018		
LiH	0.001 574	0.000 010		
Be	0.000 667	0.000 014		
$H_2O$	0.000 304	0.000 012		
$H_4$	0.000 288	0.000 004		
BeH <sub>2</sub>	0.001 331	0.000 044		
HF	0.000 532	0.000 420		

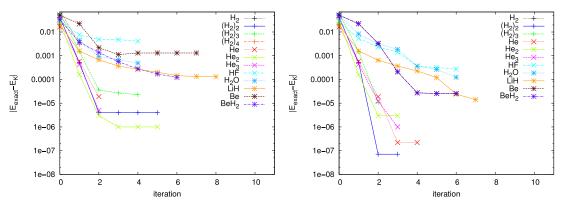


FIG. 2. Deviation in the energy from the exact values for various molecules. The accuracy in the rank reduction algorithm is  $\epsilon = 10^{-3}$  (left) and  $\epsilon = 10^{-4}$  (right).

As the parameter  $\epsilon$  is the only parameter that controls the accuracy of the decomposition scheme, the influence of this parameter is investigated in the following. For different accuracies in the rank reduction algorithm of  $\epsilon = 10^{-3}$  (left) and  $\epsilon = 10^{-4}$  (right), the convergence of the energy using the Davidson algorithm is shown in Fig. 2. Note that for this estimation the algorithm was not stopped when the convergence criterion  $||res_K|| \le \epsilon_{res}$  was fulfilled because the maximal achievable accuracy for a given  $\epsilon$  was to be reached. In both plots of Fig. 2, a rapid convergence towards the lowest eigenvalue in a decomposed tensor format is obtained for all examples that are shown. This is in agreement with the conventional Davidson algorithm. However, in the present algorithm the eigenvalues are obtained only to a certain accuracy. Additional iterations do not improve the accuracy any further. For the examples shown here, the maximal accuracy in the energy is better than the desired "chemical accuracy" of  $10^{-3}$  E<sub>h</sub>. This holds for both criteria  $\epsilon = 10^{-3}$ and  $\epsilon = 10^{-4}$ . A closer look shows that for these examples the largest deviation (in  $E_h$ ) lies in the range of  $\epsilon$ , which is applied for the rank reduction. This means that the expected

error in the energy for  $\epsilon = 10^{-3}$  can lie in the range of  $10^{-3}$  E<sub>h</sub> and for  $\epsilon = 10^{-4}$  in the range of  $10^{-4}$  E<sub>h</sub>. As shown above, the deviation in the energy is about two orders of magnitude larger than the criterion  $||res_K||$ . According to this and the data shown above, a convergence criterion of  $||res_K|| \le 10^{-1}$  can be expected to be fulfilled in case of  $\epsilon = 10^{-3}$  and  $||res_K|| \le 10^{-2}$  will be fulfilled for  $\epsilon = 10^{-4}$ .

Because the deviation in the energy is mostly influenced by the threshold  $\epsilon$  in the rank reductions, the errors that can be expected for various values of  $\epsilon$  are estimated by calculating the energies for a set of molecules. Both the energies and the deviations to the desired eigenvalues for the test set are given in Table II. All molecules were calculated using the STO-3G basis set, and the iterations were stopped when  $\|res_K\|$  was smaller than the corresponding threshold estimated above. For  $\epsilon = 10^{-3}$ , the mean absolute deviation is  $1.95 \cdot 10^{-3}$  E<sub>h</sub>, and the mean square deviation is  $2.95 \cdot 10^{-3}$  E<sub>h</sub>. For  $\epsilon = 10^{-4}$ , the mean absolute error amounts to  $1.73 \cdot 10^{-4}$  E<sub>h</sub>, and the mean square error amounts to  $3.98 \cdot 10^{-4}$  E<sub>h</sub>. In order to obtain reasonable results, the accuracy in the rank reductions should be set to  $\epsilon = 10^{-3}$ . However, for smaller deviations, especially

TABLE II. Energies  $E_{calc}$  and deviations  $|\Delta E|$  to the desired energies E, the number of electrons N, the number of orbitals n, the size of the Hilbert space  $\binom{n}{N}$  and the rank J of the coefficient tensor for various molecules <sup>64</sup> calculated using the STO-3G basis set. <sup>63</sup>

Molecule	E	$\epsilon = 10^{-3},   res_K   \le 10^{-1}$			$\epsilon = 10^{-4},   res_K   \le 10^{-2}$					
		$E_{calc}$	$ \Delta E $	$\overline{J}$	$E_{calc}$	$ \Delta E $	J	n	N	$\binom{n}{N}$
H <sub>2</sub>	-1.849 93	-1.849 93	$9.87 \cdot 10^{-7}$	2	-1.849 93	$9.87 \cdot 10^{-7}$	2	2	4	6
FH	-103.72787	-103.72024	$7.63 \cdot 10^{-3}$	10	-103.72647	$1.39 \cdot 10^{-3}$	17	12	10	66
$(H_2)_2$	-4.27697	-4.27644	$5.32 \cdot 10^{-4}$	11	-4.27697	$4.04 \cdot 10^{-6}$	18	4	8	70
$H_4$	-4.64653	-4.64624	$2.88 \cdot 10^{-4}$	11	-4.64652	$4.11 \cdot 10^{-6}$	12	4	8	70
$He_2$	-6.44588	-6.44571	$1.65 \cdot 10^{-4}$	12	-6.44587	$6.64 \cdot 10^{-6}$	14	4	8	70
Be	-14.40366	-14.40299	$6.67 \cdot 10^{-4}$	11	-14.40367	$1.42 \cdot 10^{-5}$	16	4	10	210
OH	-78.69955	-78.694 14	$5.41 \cdot 10^{-3}$	9	-78.699 15	$3.96 \cdot 10^{-4}$	21	12	9	220
LiH	-8.875 61	-8.87465	$9.58 \cdot 10^{-4}$	27	-8.875 33	$2.80 \cdot 10^{-4}$	58	12	4	495
NH	-57.831 39	-57.82793	$3.47 \cdot 10^{-3}$	14	-57.83097	$4.29 \cdot 10^{-4}$	26	12	8	495
ВеН	-16.51306	-16.51093	$2.13 \cdot 10^{-3}$	64	-16.51297	$8.74 \cdot 10^{-5}$	65	12	5	792
CH	-40.62149	-40.61768	$3.81 \cdot 10^{-3}$	32	-40.62147	$2.18 \cdot 10^{-5}$	66	12	7	792
BH	-26.94869	-26.94624	$2.45 \cdot 10^{-3}$	59	-26.94836	$6.62 \cdot 10^{-5}$	94	12	6	924
$(H_2)_3$	-6.98821	-6.98667	$1.53 \cdot 10^{-3}$	45	-6.98820	$1.02 \cdot 10^{-5}$	224	6	12	924
He <sub>3</sub>	-10.37441	-10.37397	$4.40 \cdot 10^{-4}$	42	-10.37440	$1.84 \cdot 10^{-5}$	89	6	12	924
H <sub>2</sub> O	-84.207 45	-84.207 14	$3.04 \cdot 10^{-4}$	20	-84.20746	$1.24 \cdot 10^{-5}$	116	10	14	1001
BeH <sub>2</sub>	-19.08012	-19.07879	$1.33 \cdot 10^{-3}$	72	-19.08008	$4.38 \cdot 10^{-5}$	133	6	14	3003

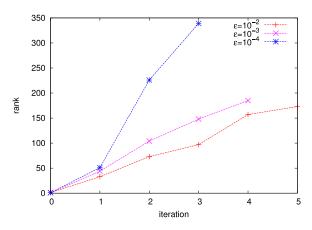


FIG. 3. Ranks of the coefficient tensors for the example of a  $H_2$  trimer that was calculated using the STO-3G basis set.<sup>63</sup>

smaller than the chemical accuracy, a threshold of  $\epsilon = 10^{-4}$  is recommended.

# B. Scaling behavior

A decisive variable for the scaling of the algorithm is the rank of the coefficient tensors. Because for each iteration a new coefficient tensor is constructed, each of them possesses its own rank. These ranks are shown in Fig. 3 for the example of a  $H_2$  trimer. Starting from an initial guess of rank one, the ranks of the coefficient tensors seem to increase linearly over the iterations. The ranks become larger as the accuracy of the rank reductions  $\epsilon$  increases.

To estimate the scaling with the system size of the whole algorithm, the most time consuming step has to be considered. This step is the rank reduction of  $\tilde{\sigma}$  after the contraction  $\tilde{\sigma} = \mathbf{H}c$  (Eqs. (46) and (47), Scheme 2 Ref. 61). As mentioned in Section II F, the rank reduction algorithm scales with  $\sim O(r_{in} \cdot r_{out}^2 \cdot n)$ . The initial rank  $r_{in}$  results from the contraction  $\tilde{\sigma} = \mathbf{H}C_K$  and scales as  $r_{in} = K \cdot J$ . K is the rank of the Hamiltonian tensor which scales with  $\sim O(N^4)$ , as can be seen in Eq. (35), and N is the number of electrons, which is assumed to be proportional to the number of basis functions n.

As the rank J of the coefficient tensor is not known in advance, calculations of a chain of H<sub>2</sub> molecules with a bond distance of 0.7425 Å that are separated by a distance of 3 Å are performed, for an optimistic estimate. The threshold for the rank reduction was set to  $\epsilon = 10^{-3}$  and the convergence criterion was set to  $\epsilon_{res} = 10^{-1}$  to reach the chemical accuracy. The length of the chain increases up to 4 molecules. Because there are several tensors used to represent the coefficient tensors that each possess a different rank, the overall rank needs to be determined for each system. This is done by setting up the linear expansion of the coefficient tensor  $\tilde{C}$  that is obtained from the Davidson algorithm and performing a successive rank reduction to obtain C, which corresponds to the decomposed expression of the FCI solution (Eqs. (3) and (9)). The rank of C is used to investigate the scaling behavior with the system size.

The ranks of the tensors C with increasing chain length are shown in Fig. 4. The data points are fitted by a curve of the

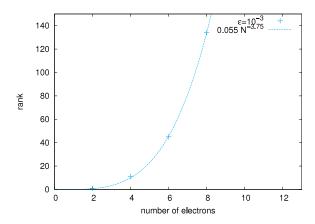


FIG. 4. Ranks of the coefficient tensors of chains of n = 1, 2, 3, 4 H<sub>2</sub> molecules calculated using the STO-3G basis set and  $\epsilon = 10^{-3}$ .

function  $f(x) = ax^b$  leading to the values  $a = 0.055 \pm 0.007$  and  $b = 3.75 \pm 0.06$ . Here, f(x) corresponds to rank(C) and x corresponds to N. This leads to a scaling of the rank of the C tensor of  $\sim O(N^4)$ . While this limited benchmark arguably is too small to make general statements, it allows to make an optimistic estimate of a scaling of the rank with  $N^4$ .

Hence, the initial rank of the most time consuming rank reduction step scales  $\sim O(N^4 \cdot N^4) = O(N^8)$ , which corresponds to the scaling of the rank of the  $\tilde{\sigma}$  tensor. Furthermore, the rank of the  $\sigma$  tensor, which corresponds to  $r_{out}$ , has the same order of magnitude as the ranks of the C tensor. This leads to an overall scaling of the FCI algorithm of  $\sim O(N^8 \cdot (N^4)^2 \cdot N) \approx O(N^{17})$ . Although the required storage shows a scaling of only  $\sim O(N^5)$  for both the coefficient and Hamiltonian tensors for this admittedly small test set of molecules, the overall scaling of the algorithm is much steeper. As an example, the largest system calculated with the present algorithm is  $(H_2)_4$ , where rank $(C_1) = 134$  and  $rank(\mathbf{H}) = 2912$ . While in conventional FCI-algorithms the coefficient vector consists of 12 870 elements, there are only 4288 elements to be stored in the CP format. Nevertheless, much larger ranks occur during the iterations, which need to be reduced further, e.g., rank( $\tilde{\sigma}_1$ ) = 42 552 (note that without a prescreening the rank would be 390 208). At the current state of the development of the reduction algorithm, the rank reduction of tensors with such large ranks is very time consuming. Therefore, this approach is only feasible for small systems and further improvements of the rank reduction algorithm are required to reduce the total scaling. In comparison to conventional algorithms, the effort of the contraction  $\sigma = \mathbf{H}C$  is shifted to the rank reduction of the  $\sigma$  tensor in the present algorithm. Hence, another promising alternative that will be explored in future work is algorithms in which rank reduction steps are avoided altogether.<sup>49</sup>

#### IV. CONCLUSIONS

In the present proof-of-principle study, a tensor representation of the Fock space in the canonical product format is used to approximate the FCI problem. The coefficient tensor and the Hamiltonian tensor are represented in a canonical product format based on an occupation number

vector formulation. A modified Davidson algorithm is used to solve the FCI problem in the decomposed format. Both the scaling behavior and the robustness of the algorithm are tested in a series of benchmark calculations. The number of elements in the representation of the Hamiltonian tensor scales subexponentially ( $\sim O(N^5)$ ). Note that in the present work, the representation of the Hamiltonian tensor is exact. Benchmark calculations show that the error in the energy in  $E_h$  lies in the range of  $\epsilon$  and hence, the desired "chemical accuracy" can be achieved by choosing a value of  $\epsilon \simeq 10^{-4}$ . The algorithm converges rapidly and in most of our test cases in less than five iterations. Because of these properties, the algorithm is a robust approximation, and its accuracy is controllable via the single parameter  $\epsilon$ .

Although the number of elements in the representation of the Hamiltonian and coefficient tensors scales with  $\sim O(N^5)$ for some small examples, the overall scaling of the algorithm is much steeper ( $>O(N^{10})$ ). This is because the rank increases after certain operations, especially after the contraction **H**C and the increased rank needs to be reduced afterwards. As the rank reduction algorithm scales with the rank to be reduced and the rank after reduction, the actual scaling of the whole algorithm is even for optimistic estimates much steeper than the scaling of the Hamiltonian/coefficient tensor representations. While the results presented here show that using the current algorithm still only very small systems can be treated, they still prove that the basic requirements for a viable approach are fulfilled—robust convergence of the iterations and systematic control of the approximation. However, the investigation of the computational bottlenecks also reveals that the next step in arriving at a more practical approach has to be the improvement of the rank reduction procedure—or its avoidance altogether.

Hence, future work will be directed at either improving the efficiency of the rank reduction algorithm or devising novel reduction free algorithms. 49 The second approach could include using an initial guess to define an initial tensor space so that fewer iterations are required, or one could optimise the parameters within the tensor representation rather than the wave function parameters. This way, the **H**C contraction and the subsequent rank reduction can be avoided, but an accurate guess for C would be required. For such an algorithm, the results of computationally less demanding methods like CCSD, CISD, CIPSI, 65,66 or ICE-CI11 could provide an initial guess for the coefficient tensors. This way, a hybrid method could also be constructed in analogy to SQMC that is a hybrid for FCIQMC.67

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- <sup>4</sup>J. Olsen, P. Jørgensen, and J. Simons, Chem. Phys. Lett. **169**, 463 (1990).
- <sup>5</sup>C. W. Bauschlicher and P. R. Taylor, J. Chem. Phys. **85**, 2779 (1986).
- <sup>6</sup>J. C. Greer, J. Chem. Phys. **103**, 1821 (1995).
- <sup>7</sup>A. O. Mitrushenkov, Chem. Phys. Lett. **217**, 559 (1994).
- <sup>8</sup>D. R. Alcoba, A. Torre, L. Lain, O. B. Oña, P. Capuzzi, M. Van Raemdonck, P. Bultinck, and D. Van Neck, J. Chem. Phys. 141, 244118 (2014).
- <sup>9</sup>M. C. Troparevsky and A. Franceschetti, J. Phys.: Condens. Matter 20, 055211 (2008).
- <sup>10</sup>M. Sambataro, D. Gambacurta, and L. Lo Monaco, Phys. Rev. B 83, 045102 (2011).
- <sup>11</sup>F. Neese, "Reliably, efficiently and accurately approaching the full configuration interaction limit with an iterative coniguration interaction expansion"
- <sup>12</sup>M. D. Towler, Phys. Status Solidi B **243**, 2573 (2006).
- <sup>13</sup>B. M. Austin, D. Y. Zubarev, J. William, and A. Lester, Chem. Rev. 112, 263 (2012).
- <sup>14</sup>G. H. Booth, A. J. W. Thom, and A. Alavi, J. Chem. Phys. **131**, 054106 (2009).
- <sup>15</sup>G. H. Booth, D. Cleland, A. J. W. Thom, and A. Alavi, J. Chem. Phys. 135, 084104 (2011).
- <sup>16</sup>J. J. Shepherd, G. E. Scuseria, and J. S. Spencer, Phys. Rev. B 90, 155130
- <sup>17</sup>S. R. White, Phys. Rev. B **48**, 10345 (1993).
- <sup>18</sup>G. K. L. Chan, J. J. Dorando, D. Ghosh, J. Hachmann, E. Neuscamman, H. Wang, and T. Yanai, Front. Quantum Syst. Chem. Phys. 18, 49 (2008).
- <sup>19</sup>O. Legeza, J. Röder, and B. A. Hess, Phys. Rev. B **67**, 125114 (2003).
- <sup>20</sup>G. Moritz, B. A. Hess, and M. Reiher, J. Chem. Phys. **122**, 024107 (2005).
- <sup>21</sup>G. K.-L. Chan and M. Head-Gordon, J. Chem. Phys. **116**, 4462 (2002).
- <sup>22</sup>V. Murg, F. Verstraete, O. Legeza, and R. M. Noack, Phys. Rev. B **82**, 205105
- <sup>23</sup>V. Murg, F. Verstraete, R. Schneider, P. R. Nagy, and O. Legeza, J. Chem. Theory Comput. 11, 1027 (2015).
- <sup>24</sup>W. Mizukami, Y. Kurashige, and T. Yanai, J. Chem. Theory Comput. 9, 401
- <sup>25</sup>F. R. Manby, J. Chem. Phys. **119**, 4607 (2003).
- <sup>26</sup>M. Schütz and F. R. Manby, Phys. Chem. Chem. Phys. **5**, 3349 (2003).
- <sup>27</sup>H. J. Werner, F. R. Manby, and P. J. Knowles, J. Chem. Phys. **118**, 8149 (2003).
- <sup>28</sup>R. Ahlrichs, Phys. Chem. Chem. Phys. **6**, 5119 (2004).
- <sup>29</sup>M. Feyereisen, G. Fitzgerald, and A. Komornicki, Chem. Phys. Lett. 208, 359 (1993).
- <sup>30</sup>R. Kendall and H. Früchtl, Theor. Chem. Acc. **97**, 158 (1997).
- <sup>31</sup>O. Vahtras, J. Almlöf, and M. Feyereisen, Chem. Phys. Lett. 213, 514 (1993).
- <sup>32</sup>N. H. F. Beebe and J. Linderberg, Int. J. Quantum Chem. **12**, 683 (1977).
- <sup>33</sup>H. Koch, A. S. de Merás, and T. B. Pedersen, J. Chem. Phys. 118, 9481
- <sup>34</sup>I. Røeggen and E. Wisløff-Nilssen, Chem. Phys. Lett. **132**, 154 (1986).
- <sup>35</sup>J. Carroll and J.-J. Chang, Psychometrika **35**, 283 (1970).
- <sup>36</sup>R. A. Harshman and M. E. Lundy, Comput. Stat. Data Anal. **18**, 39 (1994).
- <sup>37</sup>E. G. Hohenstein, R. M. Parrish, and T. J. Martínez, J. Chem. Phys. 137, 044103 (2012).
- <sup>38</sup>E. G. Hohenstein, R. M. Parrish, C. D. Sherrill, and T. J. Martínez, J. Chem. Phys. 137, 221101 (2012).
- <sup>39</sup>E. G. Hohenstein, S. I. L. Kokkila, R. M. Parrish, and T. J. Martínez, J. Chem. Phys. 138, 124111 (2013).
- <sup>40</sup>R. M. Parrish, C. D. Sherrill, E. G. Hohenstein, S. I. L. Kokkila, and T. J. Martínez, J. Chem. Phys. 140, 181102 (2014).
- <sup>41</sup>F. A. Bischoff and E. F. Valeev, J. Chem. Phys. **134**, 104104 (2011).
- <sup>42</sup>F. A. Bischoff and E. F. Valeev, J. Chem. Phys. **139**, 114106 (2013).
- <sup>43</sup>W. Uemura and O. Sugino, Phys. Rev. Lett. **109**, 253001 (2012).
- <sup>44</sup>U. Benedikt, A. A. Auer, M. Espig, and W. Hackbusch, J. Chem. Phys. 134, 054118 (2011).
- <sup>45</sup>U. Benedikt, K.-H. Böhm, and A. A. Auer, J. Chem. Phys. **139**, 224101
- $^{46}$ U. Benedikt, "Low-rank tensor approximation in post Hartree-Fock methods," Ph.D. thesis, TU Chemnitz, 2013.
- <sup>47</sup>U. Benedikt, H. Auer, W. Hackbusch, and A. Auer, Mol. Phys. 111, 2398
- <sup>48</sup>T. Helgaker, P. Jørgensen, and J. Olsen, *Molecular Electronic Structure* Theory, 1st ed. (Wiley and Sons, 2000).
- <sup>49</sup>K.-H. Böhm, Ph.D. thesis, TU Chemnitz, 2016.
- <sup>50</sup>O. Legeza, T. Rohwedder, R. Schneider, and S. Szalay, "Tensor product approximation (DMRG) and coupled cluster method in quantum chemistry," in Many-Electron Approaches in Physics, Chemistry and Mathematics,

<sup>&</sup>lt;sup>1</sup>B. O. Roos, P. R. Taylor, and P. E. Siegbahn, Chem. Phys. 48, 157 (1980).

<sup>&</sup>lt;sup>2</sup>B. Roos, Chem. Phys. Lett. **15**, 153 (1972).

<sup>&</sup>lt;sup>3</sup>E. Rossi, G. L. Bendazzoli, S. Evangelisti, and D. Maynau, Chem. Phys. Lett. 310, 530 (1999).

- Mathematical Physics Studies, edited by V. Bach and L. Delle Site (Springer International Publishing, 2014), pp. 53–76.
- <sup>51</sup>S. Szalay, M. Pfeffer, V. Murg, G. Barcza, F. Verstraete, R. Schneider, and O. Legeza, Int. J. Quantum Chem. 115, 1342 (2015).
- <sup>52</sup>G. Beylkin and M. J. Mohlenkamp, Proc. Natl. Acad. Sci. U. S. A. **99**, 10246 (2002).
- <sup>53</sup>G. Beylkin and M. J. Mohlenkamp, SIAM J. Sci. Comput. **26**, 2133 (2005).
- <sup>54</sup>T. G. Kolda and B. W. Bader, SIAM Rev. **51**, 455 (2009).
- <sup>55</sup>E. R. Davidson, J. Comput. Phys. **17**, 87 (1975).
- <sup>56</sup>D. Braess and W. Hackbusch, IMA J. Numer. Anal. **25**, 685 (2005).
- <sup>57</sup>Especially in high dimensions it is a good advise to balance the representation system of every tensor which is stored in real computer implementations. Otherwise, one cannot avoid a number overflow in some entries of the representation system. For example consider the rank-one tensor  $v := \bigotimes_{\mu=1}^n v_\mu \in \bigotimes_{\mu=1}^n \mathbb{R}^t$  with  $(v_\mu)_l = 1$  for all  $l \in \{1, \ldots, t\}$ . If one introduces the following constrains for the representation system:  $\|v_\mu\| = 1$  for all  $\mu \in \{2, \ldots, d\}$ , we have  $\|v_1\| = t^n$ . If n and t are large enough, we will produce a number overflow on every computer system. With (55) we have a balanced and solid representation system. Furthermore, with (55) we avoid unnecessary scaling influences in our objective function and consequently their second derivative with respect to the representation system.

- <sup>58</sup>M. Espig, "Effiziente Bestapproximation mittels Summen von Elementartensoren in hohen Dimensionen," Ph.D. thesis, Universität Leipzig, 2008.
- <sup>59</sup>M. Espig and W. Hackbusch, Numer. Math. **122**, 489 (2012).
- <sup>60</sup>M. Espig, "The alternating steepest descent method for solving linear systems in tensor format representations" (unpublished).
- <sup>61</sup>Calculation done on one node with an Intel® Xenon® CPU X5650 @ 2,67 GHz.
- <sup>62</sup>The geometries are H<sub>2</sub>  $r_{H-H}$  = 0.7425 Å, LiH  $r_{Li-H}$  = 1.5950 Å, H<sub>2</sub>O  $r_{O-H}$  = 0.9572 Å,  $\theta_{HOH}$  = 104.52°, H<sub>4</sub> square geometry  $r_{H-H}$  = 1.0584 Å, BeH<sub>2</sub>  $r_{Be-H}$  = 1.2905 Å  $\theta_{HBeH}$  = 180.00°, HF  $r_{F-H}$  = 0.9327 Å. Intermolecular distance is always 3.0 Å.
- <sup>63</sup>W. J. Hehre, R. F. Stewart, and J. A. Pople, J. Chem. Phys. **51**, 2657 (1969).
- <sup>64</sup>The geometries are LiH  $r_{Li-H}=1.5983$  Å, BeH  $r_{Be-H}=1.3601$  Å, BH  $r_{B-H}=1.2372$  Å, CH  $r_{C-H}=1.1301$  Å, NH  $r_{N-H}=1.0447$  Å, OH  $r_{O-H}=0.9821$  Å, FH  $r_{F-H}=0.9285$  Å, H<sub>2</sub>  $r_{H-H}=0.7425$  Å, H<sub>2</sub>O  $r_{O-H}=0.9572$  Å,  $\theta_{HOH}=104.52^{\circ}$ , H<sub>4</sub> square geometry  $r_{H-H}=1.0584$  Å, BeH<sub>2</sub>  $r_{Be-H}=1.2905$  Å  $\theta_{HBeH}=180.00^{\circ}$ . Intermolecular distance is always 3.0 Å.
- <sup>65</sup>B. Huron, J. Malrieu, and P. Rancurel, J. Chem. Phys. **58**, 5745 (1973).
- <sup>66</sup>S. Evangelisti, J. Daudey, and J. Malrieu, Chem. Phys. **75**, 91 (1983).
- <sup>67</sup>F. R. Petruzielo, A. A. Holmes, H. J. Changlani, M. P. Nightingale, and C. J. Umrigar, Phys. Rev. Lett. **109**, 230201 (2012).