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ABSTRACT

We develop a range-separated stochastic resolution of identity (RS-SRI) approach for the four-index electron repulsion integrals, where the larger terms (above a predefined threshold) are treated using a deterministic RI and the remaining terms are treated using a SRI. The approach is implemented within a second-order Green’s function formalism with an improved $O(N^3)$ scaling with the size of the basis set, $N$. Moreover, the RS approach greatly reduces the statistical error compared to the full stochastic version [T. Y. Takeshita et al., J. Chem. Phys. 151, 044114 (2019)], resulting in computational speedups of ground and excited state energies of nearly two orders of magnitude, as demonstrated for hydrogen dimer chains and water clusters.

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I. INTRODUCTION

Many-body perturbation theory (MBPT) based on Green’s function (GF) approaches [e.g., the Møller–Plesset (MP) perturbation theory,1 the second-order Green’s function (GF2) approach,2 and the GW3 approximation] have proven very useful in predicting the ground state properties beyond the limitations of density functional theory (DFT) and the Hartree–Fock (HF) method as well as in predicting quasi-particle and neutral excitation. In these methods, correlations are treated systemically by expanding the self-energy (which contains the information of correlations) in the Coulomb4−6 or screened Coulomb7−9 interactions. MBPT has been applied to a variety of molecular and bulk systems in predicting, e.g., correlation energies, ionization potentials and electron affinities,7 and excited states.7,15,23–26 Excluding several recent applications,7,15 MBPT has been limited to relatively small systems due to the steep computational scaling with the system size.
A particularly interesting implementation of MBPT, relevant to the applications reported below, is based on a second-order approximation to the electron self-energy,\textsuperscript{24,26} which has received increasing attention in recent years.\textsuperscript{26–30} In contrast to the GW approximation, dynamical exchange correlations are included explicitly in the GF2 self-energy to the second order in Coulomb interactions, providing accurate ground state energies\textsuperscript{30,37} and quasi-particle energies.\textsuperscript{26,28,30} Although the results of recent studies are extremely promising, the GF2 approach suffers from a high computational cost [$O(N^3)$], limiting its application to relatively small system sizes.

To overcome this limitation, two stochastic formulations were recently introduced to reduce the computational scaling. Neuhauser et al.\textsuperscript{31,32} developed a stochastic decomposition of the imaginary time GF to reduce the overall scaling of GF2 to $O(N^3)$. Takeshita et al.\textsuperscript{33} and Dou et al.\textsuperscript{34} proposed an approach that builds on the stochastic resolution of identity (SRI) for the electron repulsion integrals (ERIs)\textsuperscript{35} to describe both ground and quasi-particle excited states. Similar to the deterministic resolution of identity (RI),\textsuperscript{26,28,30} the SRI decouples the four-index ERIs. While the number of auxiliary bases increases with the system size for the RI, the number of stochastic realizations in the SRI is independent of the system size, resulting in an overall $O(N^3)$ scaling. However, the RI technique comes at the cost of introducing a statistical error in the energy and nuclear forces,\textsuperscript{36,37} which can be controlled by increasing the number of stochastic realizations, $N_s$. While the overall scaling of the stochastic formulations of GF2 is similar to DFT and HF, achieving chemical accuracy requires a large number of stochastic realizations, resulting in increasingly longer computational time, even for small systems.\textsuperscript{31,32}

In this work, we develop a range-separated stochastic resolution of identity (RS-SRI) approach to decouple the four-index ERIs, where the short-range ERIs (larger values) are treated deterministically using the resolution of identity (RI)\textsuperscript{33–38} and the remaining terms are treated using the stochastic resolution of identity (SRI).\textsuperscript{11} The RS-SRI approach allows for a significant reduction in the statistical error without the need to increase the number of stochastic realizations while maintaining the overall $O(N^3)$ scaling. We apply the RS-SRI technique to GF2 theory and demonstrate its ability to reduce the overall computational scaling from $O(N^3)$ to $O(N^2)$ as well as increase the sampling efficiency by nearly two orders of magnitude as compared to the SRI technique.

II. RANGE-SEPARATED STOCHASTIC RESOLUTION OF IDENTITY

Consider a generic many-body electronic Hamiltonian in the second-quantization representation,

$$
\hat{H} = \sum_{ij} h_{ij} \hat{a}_{i}^{\dagger} \hat{a}_{j} + \frac{1}{2} \sum_{ijkl} v_{ijkl} \hat{a}_{i}^{\dagger} \hat{a}_{j}^{\dagger} \hat{a}_{k} \hat{a}_{l},
$$

where $\hat{a}_{i}^{\dagger}$ and $\hat{a}_{i}$ are the Fermionic creation and annihilation operators, respectively, for an electron in orbital $\chi_{i}(\mathbf{r})$. In the applications below, $\chi_{i}(\mathbf{r})$ is chosen to be an atomic orbital, but we do not use the locality of the basis to reduce the scaling or do we introduce a cutoff to compute the ERIs [see Eq. (2)] or the overlap matrix [see Eq. (3)]. Therefore, the formalism and the resulting scaling reported below are general for any choice of basis. The creation and annihilation operators obey the following anti-commutation relationship:

$$
\{ \hat{a}_{i}, \hat{a}_{j}^{\dagger} \} = S_{ij},
$$

where $S_{ij} = \int \chi_{i}(\mathbf{r}) \chi_{j}(\mathbf{r}) d\mathbf{r}$ is the matrix element of the overlap matrix $S$. In Eq. (1), $h_{ij}$ is the matrix element of the one-body Hamiltonian and $v_{ijkl}$ is the four-index ERI ($v_{ijkl} \equiv (ij|kl)$),

$$
v_{ijkl} = \int d\mathbf{r}_{1} d\mathbf{r}_{2} \frac{\chi_{i}(\mathbf{r}_{1}) \chi_{j}(\mathbf{r}_{1}) \chi_{k}(\mathbf{r}_{2}) \chi_{l}(\mathbf{r}_{2})}{|\mathbf{r}_{1} - \mathbf{r}_{2}|}.
$$

Describing correlations within a many-body perturbation technique beyond the mean-field approximation relies on the contraction of $v_{ijkl}$ (or powers of $v_{ijkl}$), a task that becomes computationally intractable with increasing levels of accuracy. A common approach to reduce the computational complexity is based on the resolution of identity (RI), where the four-index ERIs in Eq. (3) are approximated by products of three-index ERIs and two-index ERIs\textsuperscript{34–38},

$$
v_{ijkl} \approx \sum_{AB} \left( (ij|A) \chi_{B}(\mathbf{r}) \right)^{\frac{1}{2}} B(kl).
$$

Here, $\chi_{i}(\mathbf{r})$ and $\chi_{j}(\mathbf{r})$ are the auxiliary orbitals, and $(ij|A)$ and $V_{AB}$ are the three-index and two-index ERIs respectively,

$$
(ij|A) = \int d\mathbf{r}_{1} d\mathbf{r}_{2} \frac{\chi_{i}(\mathbf{r}_{1}) \chi_{j}(\mathbf{r}_{1}) \chi_{A}(\mathbf{r}_{2})}{|\mathbf{r}_{1} - \mathbf{r}_{2}|},
$$

$$
V_{AB} = \int d\mathbf{r}_{1} d\mathbf{r}_{2} \frac{\chi_{A}(\mathbf{r}_{1}) \chi_{B}(\mathbf{r}_{2})}{|\mathbf{r}_{1} - \mathbf{r}_{2}|}.
$$

For convenience, we define a new set of three-index ERIs $K_{ij}^{Q}$,

$$
K_{ij}^{Q} = \sum_{A} \left( (ij|A) \right)^{\frac{1}{2}} \chi_{A}(\mathbf{r}_{i})^{\dagger},
$$

such that the four-index ERI can be expressed in terms of three-index ERIs only,

$$
v_{ijkl} = \sum_{Q} K_{ij}^{Q} K_{kl}^{Q}.
$$

The advantage of the above decomposition is that the resolution of identity reduces the number of two-body ERIs from $O(N^4)$ to $O(N^2 N_{aux})$, where $N$ is the size of the atomic basis and $N_{aux}$ is the size of the auxiliary basis. However, since $N_{aux}$ increases nearly linearly with the size of the atomic basis $N$ and since the calculation of $K_{ij}^{Q}$ scales as $O(N^4)$, the approach does not always reduce the computational scaling of the correlation energy for, e.g., MP2 and GF2.\textsuperscript{31–38}

Recently, we have introduced a stochastic version of the resolution of identity, which provides a framework to reduce the scaling for contraction within many-body perturbation techniques at the account of introducing a controlled statistical error in the calculated observables (e.g., the forces on the nuclei and the energy...
per electron). The balance between accuracy and efficiency is controlled by the number of stochastic realizations (N\textsubscript{i}) according to the central limit theorem. The stochastic RI approach utilizes the same set of three-index ERIs (ij|A) while circumventing the need to directly compute K\textsubscript{ij} by introducing a set of N\textsubscript{i} stochastic orbitals, \{|\theta\}, \xi = 1, 2, \ldots, N\textsubscript{i}. The stochastic orbitals are defined as arrays of length N\textsubscript{aux} with randomly selected elements 1 or -1, i.e., \theta\textsubscript{A} = \pm 1. Defining
\[
R\textsubscript{ij}^S = \sum_{Q} \langle ij|A|V\rangle \frac{1}{N} \delta_{Q} = \frac{1}{N} \sum_{A} \langle ij|A|V\rangle \delta_{Q},
\]
(9)
the expression for \nu_{ijkl} can be reduced to
\[
\nu_{ijkl} \approx \frac{1}{N} \sum_{\xi} R\textsubscript{ij}^S R\textsubscript{kl}^S \equiv \langle R_j R_k \rangle_{\beta},
\]
(10)
where \langle \cdots \rangle\textsubscript{\beta} implies a statistical average over the stochastic orbitals, \{\theta\}. The overall computational scaling of the R\textsubscript{ij} matrices is O(N\textsubscript{i}N\textsuperscript{3}), but N\textsubscript{i} is found to be independent of the system size for different applications.\textsuperscript{25,29,41–43,49,55} The SRI technique has been successfully used to reduce the scaling of the correlation energy within MP2 and GF2 theories, from O(N\textsuperscript{6}) to O(N\textsuperscript{5}).

The above approach has been implemented for simple molecules and for hydrogen chains of different lengths in order to assess its accuracy for large systems.\textsuperscript{41–43} To converge the results to chemical accuracy required a rather large number of stochastic orbitals (N\textsubscript{i} \approx 1000), which limits the application of the SRI technique to relatively small systems (due to the large "prefactor"), with N \to 1000, still exceedingly larger than the deterministic approach.\textsuperscript{41–43} In order to reduce the number of stochastic orbitals and to allow for a smaller statistical error, we first sort the ERIs (ij|A) according to their magnitude and keep only those that are larger than a threshold,
\[
\langle ij|A\rangle^S = \begin{cases} 
\langle ij|A\rangle & \text{if } \langle ij|A\rangle \geq \epsilon, \\
0 & \text{otherwise.}
\end{cases}
\]
(11)
Here, \langle \langle ij|A\rangle \rangle\textsubscript{ij}\textsuperscript{max} is the maximal value of \langle ij|A\rangle for each j and \epsilon is a predefined parameter. The superscript L (or S) denotes large (or small) values. By setting the cutoff threshold to depend on \frac{\epsilon}{N}, the number of nonzero elements in \langle ij|A\rangle\textsuperscript{S} for each j scales as O(N) [rather than O(N\textsuperscript{2}) if no threshold is used or O(1) if a fixed threshold \epsilon is used]. This implies that the total non-vanishing elements in (ij|A)\textsuperscript{S} scale as O(N\textsuperscript{2}). We then define [K\textsubscript{ij}\textsuperscript{S}]\textsuperscript{L} as
\[
[K\textsubscript{ij}\textsuperscript{S}]^L = \sum_{A} \langle ij|A|V\rangle \frac{1}{N} \delta_{Q},
\]
(12)
and keep only the terms that are larger than a predefined threshold, namely, we set [K\textsubscript{ij}\textsuperscript{S}]\textsuperscript{L} = 0 for values below the threshold according to
\[
[K\textsubscript{ij}\textsuperscript{S}]^L = \begin{cases} 
[K\textsubscript{ij}\textsuperscript{S}]^L & \text{if } [K\textsubscript{ij}\textsuperscript{S}]^L \geq \epsilon \langle[K\textsubscript{ij}\textsuperscript{S}]^L\rangle\textsubscript{ij}\textsuperscript{max} \\
0 & \text{otherwise.}
\end{cases}
\]
(13)

The calculation of [K\textsubscript{ij}\textsuperscript{S}]\textsuperscript{L} using the above procedure scales as O(N\textsuperscript{3}). We proceed by defining
\[
[R\textsubscript{ij}^L]^S = \sum_{\alpha} [K\textsubscript{ij}\textsuperscript{S}]^L \delta_{Q} = \frac{1}{N} \sum_{A} \langle ij|A|V\rangle \delta_{Q},
\]
(14)
\[
[R\textsubscript{ij}^S]^L = R\textsubscript{ij} - [R\textsubscript{ij}^L]^S,
\]
(15)
where R\textsubscript{ij} is defined above in Eq. (9) and the computational scaling for both terms, [R\textsubscript{ij}^S]^L and [R\textsubscript{ij}^S]^S, is O(N\textsuperscript{3}). Using these definitions, the four-index tensor \nu_{ijkl} can be rewritten as
\[
\nu_{ijkl} = \sum_{Q} [K\textsubscript{ij}\textsuperscript{S}]^L [K\textsubscript{kl}\textsuperscript{S}]^L + \langle R_j R_k \rangle_{\beta},
\]
(16)
Equation (16) is referred to as the range-separated stochastic resolution of identity (RS-SRI). The RS-SRI reduces to the SRI for \epsilon = 1 and to the deterministic RI for \epsilon = 0. This suggests that \epsilon can be used as a control parameter balancing the computational efficiency and statistical errors. For optimal choices of \epsilon, the contribution of \epsilon to the self-energy (in the closed shell case) is given by
\[
\Sigma = -\frac{1}{Z} \sum_{\alpha} \langle \hat{a}_\alpha | e^{\beta H} | \hat{a}_\alpha \rangle,
\]
(17)
where \hat{a}_\alpha and \hat{a}^\dagger_\alpha are defined above in Sec. II, T is a time ordering operator, and \tau is an imaginary time point along \tau \in (0, -\beta). In the above, we have used the Heisenberg picture for the operators: \hat{a}_\alpha(\tau) = e^{i(\beta-H)\tau} \hat{a}_\alpha e^{-i(\beta-H)\tau}, where N = \sum_{\alpha} \langle \hat{a}_\alpha | e^{i(\beta-H)\tau} | \hat{a}_\alpha \rangle is the number operator and H is the many-body Hamiltonian defined in Eq. (1). The average is taken with respect to the grand canonical partition function \langle \cdots \rangle = Z^{-1} \text{Tr}[\exp(-\beta H)\hat{S}], where Z = \text{Tr}[\exp(-\beta H)\hat{S}] is the normalization factor, \beta = 1/k_B T is the inverse temperature, and \mu is the chemical potential.

The Matsubara GF obeys the following Dyson equation:
\[
\begin{align*}
- \frac{1}{\beta} \partial_{\tau} G_\tau(\tau) & = \delta(\tau) + (F - \mu(S)) G(\tau) + \int_{0}^{\beta} d\tau_1 \Sigma(\tau - \tau_1) G(\tau_1),
\end{align*}
\]
(18)
where F is the Fock matrix given by
\[
F_{ij} = h_{ij} - 2 \sum_{\beta} G_{ij}(\beta^-) (v_{ijkl} - \frac{1}{2} v_{ij\beta}),
\]
(19)
and \Sigma is the self-energy. In the second-order Born approximation, the self-energy (in the closed shell case) is given by

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\[ \Sigma_\beta(\tau) = \sum_{klmnpq} v_{mkpq} (2u_{lpnj} - v_{npq}) G_{kl}(\tau) G_{mn}(\tau) G_{pq}(\beta - \tau). \] (20)

The above form scales as \(O(N^5)\) using the appropriate contraction.

The Matsubara Green’s function for the Fermionic systems obeys the following anti-symmetric relationship: \(G(\tau) = -G(\tau + \beta)\). The anti-symmetry feature allows for a Fourier representation of \(G(\tau)\) in imaginary frequency,

\[ \tilde{G}(i\omega_n) = \int_0^\beta e^{i\omega_n \tau} G(\tau). \] (21)

Here, \(i\omega_n = i(2n + 1)\pi/\beta\) are the Matsubara frequencies, and the inverse Fourier transform is defined by

\[ G(\tau) = \frac{1}{\beta} \sum_n e^{-i\omega_n \tau} \tilde{G}(i\omega_n). \] (22)

The Dyson equation [cf. Eq. (18)] can then be solved in the frequency domain,

\[ \tilde{G}(i\omega_n) = \left[ \tilde{G}_\beta(i\omega_n) \right]^{-1} - \tilde{\Sigma}(i\omega_n), \] (23)

where \(\tilde{\Sigma}(i\omega_n)\) is the Fourier transform of the self-energy [Eq. (20)] and \(\tilde{G}_\beta(i\omega_n)\) is the non-interacting GF,

\[ \tilde{G}_\beta(i\omega_n) = \left[ (\mu + i\omega_n)\sigma - \hat{S} - \hat{F} \right]^{-1}. \] (24)

Since the self-energy \(\tilde{\Sigma}(i\omega_n)\) depends on \(\tilde{G}(i\omega_n)\) itself, Eq. (23) as well as Eq. (20) must be solved self-consistently. This is done by first performing a Hartree–Fock calculation to obtain the overlap matrix \(S\), the Fock matrix \(F\), and the chemical potential \(\mu\). The Fock matrix can then be used for constructing the non-interacting GF [cf. Eq. (24)] that serves as our initial guess of \(\tilde{G}(i\omega_n) = \tilde{G}_\beta(i\omega_n)\). The next step involves the calculation of the self-energy, which is performed in the imaginary time domain [Eq. (20)]. The self-energy is then used to update the GF in Eq. (23), and the latter is used to update the Fock matrix in Eq. (19). It is often necessary to conserve the number of particles \(N_e = -2\Sigma_\beta(\tau = \beta) S_{hh}\). This can be achieved by tuning the chemical potential \(\mu\).

The computational bottleneck in GF2 is the calculation of the self-energy, which scales formally as \(O(N^5)\). Using the RS-SRI representation for \(v_{mkpq}\) given by Eq. (16), the self-energy can be written as

\[
\Sigma_\beta(\tau) = \sum_{klmnpq} G_{kl}(\tau) G_{mn}(\tau) G_{pq}(\beta - \tau) \\
\times \left( \sum_{Q}^{N_m} \left[ K_{Q}^L \right]^2 \left[ K_{Q}^L \right]^L + \left[ R_{mn}^L R_{pq}^L \right]_0 + \left[ R_{mn}^S R_{pq}^L \right]_0 + \left[ R_{mn}^S R_{pq}^S \right]_0 \right) \\
\times 2 \left( \sum_{Q}^{N_m} \left[ K_{Q}^L \right]^2 \left[ K_{Q}^L \right]^L + \left[ R_{pq}^L R_{pq}^L \right]_0 + \left[ R_{pq}^S R_{pq}^L \right]_0 + \left[ R_{pq}^S R_{pq}^S \right]_0 \right) \\
- \sum_{Q}^{N_m} \left[ K_{Q}^L \right]^2 \left[ K_{Q}^L \right]^L + \left[ R_{pq}^L R_{pq}^L \right]_0 + \left[ R_{pq}^S R_{pq}^L \right]_0 + \left[ R_{pq}^S R_{pq}^S \right]_0 \right].
\] (25)

In Sec. IV, we apply the RS-SRI to a series of hydrogen chain molecules and compare the results to deterministic RI as well as to SRI. We find, in practice, that the RS-SRI scales even better than the upper theoretical limit of \(O(N^3)\) and at the same time reduces the statistical error by about an order of magnitude as shown below.

IV. RESULTS AND DISCUSSION

In this section, we assess the performance of the RS-SRI-GF2 approach and compare the results to deterministic GF2 and SRI-GF2 for hydrogen dimer chains \(H_2\) of length \(N_H\) and on a series of water clusters. We begin with the hydrogen chain that has been the canonical model system to assess different electronic structure methods. The distance between strongly bonded hydrogen atoms was set to 0.74 Å, and the distance between weakly bonded hydrogen atoms was set to 1.26 Å. For each hydrogen, we used the STO-3G basis and the CC-pVDZ-R1 fitting basis for the resolution of identity in evaluating the self-energy as well as the CC-pVDZ-JKFFT fitting basis in evaluating the Fock matrix in Eq. (19). The inverse temperature used for the calculation of the GFs was set to \(\beta = 50\) inverse Hartree, sufficient to converge the results due to the large quasi-particle gap. We used the approach developed in Ref. 40 to perform the discrete Fourier transform with 20 000 Matsubara frequencies and 300 imaginary-time points.

We first test how \(\epsilon^\prime\) and \(\epsilon\) [in Eqs. (11) and (13)] affect the final results. In Fig. 1, we plot the correlation energy per electron, defined as

\[ E_{corr} = \frac{1}{N_e} \int_0^\beta d \tau \text{Tr} \left( \Sigma(\tau) G(\beta - \tau) \right) \] (26)

for \(H_2\) as a function of \(\epsilon^\prime\) for two values of \(\epsilon\). The statistical errors were estimated from the standard deviation \(\sigma\) over \(N_{run} = 10\) independent RS-SRI-GF2 runs. The RS-SRI-GF2 results agree with the deterministic GF2 results within the statistical error for all values of \(\epsilon\) and \(\epsilon^\prime\). In general, we find that the statistical errors decrease with the decreasing values of \(\epsilon\) and \(\epsilon^\prime\), consistent with the deterministic limit defined by \(\epsilon \to 0\) and \(\epsilon^\prime \to 0\), respectively. However, for the

![FIG. 1. Correlation energy per electron for \(H_2\) as a function of \(\epsilon\) for two values of \(\epsilon^\prime\) from RS-SRI-GF2 calculations (blue symbols and lines). Orange lines denote the deterministic results that are independent of \(\epsilon\) and \(\epsilon^\prime\). Note that the RS-SRI-GF2 results agree with the deterministic GF2 results within the statistical error bar. The statistical errors were estimated from the standard deviation \(\sigma\) over \(N_{run} = 10\) independent RS-SRI-GF2 runs.](image-url)
larger value of $\epsilon'$, we find an optimal value of $\epsilon$, suggesting that the approach to the deterministic limit may not be monotonic. Clearly, one needs to be careful when choosing the optimal thresholds for range separation. Below, for hydrogen molecule chains, we will set $\epsilon' = 0.02$ and $\epsilon = 0.1$ as our thresholds for RS-SRI calculations unless otherwise noted.

In Fig. 2, we plot the correlation energy per electron for a series of hydrogen dimer chains. We compare the results obtained using the RS-SRI-GF2 with SRI-GF2 and for small systems, with deterministic calculations. We find, as expected, that the correlation energy per electron is roughly independent of the length of the chain. Furthermore, both RS-SRI-GF2 with SRI-GF2 agree with the deterministic results within their statistical error. However, the statistical error for the same number of stochastic orbitals ($N_s$) is significantly smaller (by nearly an order of magnitude) for RS-SRI-GF2 compared to SRI-GF2 for the entire range of system sizes. The error bar was estimated as the standard deviation $\sigma$, where we have used $N_{\text{run}} = 10$ runs to estimate the statistical fluctuations.

In Fig. 3, we plot the correlation energy per electron as a function of the inverse of the number of stochastic orbitals ($1/N_s$) for $H_{80}$, $H_{200}$, and $H_{500}$. We find that the statistical fluctuations decrease as $1/N_s$, indicated by the decrease in the magnitude of the error bars. For $N_s = 800$, we compare the RS-SRI-GF2 with the SRI-GF2 (red symbol, Fig. 3) for $H_{500}$. Clearly, the statistical noise is much larger (by about a factor of 10) compared to the RS-SRI-GF2 result (green symbols). We also find that the statistical fluctuations in the correlation energy per electron are independent of the system size, consistent with $N_s$ being independent on $N$, similar to other implementation of stochastic orbital techniques.

A closer examination of the results shown in Fig. 3 suggests that for $H_{500}$, the correlation energy per electron depends linearly on $1/N_s$. This dependence has been discussed and analyzed in Ref. 40 and was attributed to the existence of a bias resulting from the non-linear dependence of the correlation energy on the stochastic process (see the Appendix for a complete derivation of the bias). In comparison to a previous work, we find that the bias is rather small, well within the statistical errors, and thus, its existence is questionable.

In Fig. 4, we plot the computational wall time of the different GF2 approaches (deterministic GF2, RS-SRI-GF2, and SRI-GF2) as a function of the length of the hydrogen atom chain, $N_H$. All calculations are performed on a single node with the 32-core Intel-Xeon processor E5-2698 v3 (“Haswell”) at 2.3 GHz. The deterministic GF2 scales as $O(N^{5.1})$, the SRI-GF2 scales as $O(N^{3.1})$, and the current
approach, for the same level of accuracy as in the SRI-GF2, scales as \(O(N^{3.5})\), slightly better than the theoretical limit of \(O(N^3)\). Note that the RS-SRI-GF2 approach has a much smaller total wall time compared to the other approaches, across the entire system range studied. As additional checks, the inset of Fig. 4 shows the scaling compared to the other approaches, across the entire system range of water clusters. We set \(\sigma = 0.005\) and \(\epsilon = 0.025\) in the RS-SRI-GF2 calculations.

Finally, we demonstrate that the range-separated stochastic resolution of identity method is also useful for three dimensional systems. In Fig. 5, we plot the correction function per electron for a series of water clusters. We set \(\epsilon' = 0.005\) and \(\epsilon = 0.025\) in the RS-SRI-GF2 calculations. Note that, again, the SRI-GF2 as well as the RS-SRI-GF2 results agree with the deterministic GF2 results within the statistical errors, which are significantly smaller for the RS-SRI-GF2 approach. The statistical error is estimated by the standard deviation over \(N_{\text{run}} = 10\) runs. The computational scaling for the water cluster is found numerically to be \(O(N^{2.5})\) (not shown), slightly better than the theoretical limit of \(O(N^3)\). This implies that the range-separated stochastic resolution of identity approach is also useful to reduce the computational scaling as well as statistical error for 3D structures.

V. CONCLUSIONS

We have developed a range-separated stochastic resolution of identity approach to decouple the four-index electron repulsion integrals and implemented the approach within the second-order Green’s function formalism, GF2. The RS-SRI technique can be viewed as a hybridization of the RI and SRI techniques, leveraging from both the accuracy of the RI and the reduced computational complexity of the SRI approaches. Results calculated for hydrogen dimer chains of varying lengths and for water clusters show an improved scaling of \(O(N^{3.5})\) with the size of the basis, \(N\) in comparison to our previous fully stochastic approach, the RS-SRI-GF2 approach significantly reduces the statistical error, resulting in computational wall times that are nearly two orders of magnitude shorter compared to the SRI-GF2. While we focused in this work on the specific implementation of the RS-SRI, the approach lends itself to higher-order approximations to the self-energy and for going beyond ground state properties. Future work should assess the performance of this RS-SRI technique for a wider range of geometries as well as its applicability to the calculation of excited state properties.

AUTHORS’ CONTRIBUTIONS

W.D. and M.C. contributed equally to this work.

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APPENDIX: BIAS AND STOCHASTIC ERROR

In this appendix, we provide a short analysis of the origin of the bias and its interplay with the statistical error. Consider the general case where \(g\) is the observable of interest given by

\[
g = G(\langle x \rangle),
\]

where \(G\) is some smooth function and \(\langle x \rangle\) is the expected value of a random variable \(x\), which also has the variance \(\sigma^2 = \langle (x - \langle x \rangle)^2 \rangle\).

Using \(N_i\) independent samples, \(x_n, n = 1, \ldots, N_i\), we define \(y_{N_i}\) as

\[
y_{N_i} = \frac{1}{N_i} \sum_{n=1}^{N_i} x_n,
\]

which has the following properties:

\[
\langle y_{N_i} - \langle x \rangle \rangle = 0,
\]

\[
\langle \langle y_{N_i} - \langle x \rangle \rangle^2 \rangle = \frac{\sigma_i^2}{N_i}.
\]

We now Taylor expand \(G(y_{N_i})\) near \(\langle x \rangle\),

\[
G(y_{N_i}) = G(\langle x \rangle) + G'(\langle x \rangle)(y_{N_i} - \langle x \rangle) + \frac{1}{2} G''(\langle x \rangle)(y_{N_i} - \langle x \rangle)^2 + O((y_{N_i} - \langle x \rangle)^3).
\]
According to the central limit theorem, the expectation value of \( g = G(x) \) is then given by

\[
\langle g_{\text{Num}} \rangle_{\text{Num}} = G(x) + \frac{1}{2} G''(x) \frac{\sigma^2}{N},
\]

and

\[
G'(x) - \frac{\sigma}{\sqrt{N N_{\text{Num}}}}.
\]

The second term on the right-hand side of the above equation is what we refer to as bias, and the last term is the statistical error. Note that bias depends linearly on \( 1/N \), whereas the statistical error depends on \( 1/\sqrt{N} \).

**DATA AVAILABILITY**

The data that support the findings of this study are available from the corresponding author upon reasonable request.

**REFERENCES**

1. C. Møller and M. S. Plesset, Phys. Rev. 46, 618 (1934).
2. L. S. Cederbaum, J. Phys. B: At. Mol. Phys. 8, 290 (1975).
3. L. Hedlin, Phys. Rev. 139, A3796 (1965).
4. L. J. Holleboom and J. G. Snijders, J. Chem. Phys. 93, 5826 (1990).
5. M. S. Hybertsen and S. G. Louie, Phys. Rev. Lett. 55, 1418 (1985).
6. M. M. Rieger, L. Steinbeck, I. D. White, H. N. Rojas, and R. W. Godby, Comput. Phys. Commun. 117, 211 (1999).

7. G. Onida, L. Reining, and A. Rubio, Rev. Mod. Phys. 74, 601 (2002).
8. N. E. Dahlen, R. van Leeuwen, and U. von Barth, Int. J. Quantum Chem. 101, 512 (2005).
9. Y.-y. Ohnishi and S. Ten-no, J. Comput. Chem. 37, 2447 (2016).
10. F. Pavlović, C. Peng, J. Ortiz, and E. F. Valeev, J. Chem. Phys. 147, 121101 (2017).
11. M. S. Hybertsen and S. G. Louie, Phys. Rev. B 34, 5390 (1986).
12. P. Rinke, A. Qteish, J. Neugebauer, C. Freysoldt, and M. Scheffler, New J. Phys. 7, 126 (2005).
13. P. Liao and E. A. Carter, Phys. Chem. Chem. Phys. 13, 15189 (2011).
14. J. B. Neaton, M. S. Hybertsen, and S. G. Louie, Phys. Rev. Lett. 97, 216405 (2006).
15. M. I. Tiago and J. R. Chelikowsky, Phys. Rev. B 73, 205334 (2006).
16. C. Friedrich, A. Schindlmayr, S. Blügel, and T. Kotani, Phys. Rev. B 74, 045104 (2006).
17. M. Grüninger, A. Marini, and A. Rubio, J. Chem. Phys. 124, 154108 (2006).
18. C. Rostgaard, K. W. Jacobsen, and K. S. Thygesen, Phys. Rev. B 81, 085103 (2010).
19. P. Koval, D. Foerster, and D. Sánchez-Portal, Phys. Rev. B 89, 155417 (2014).
20. J. L. Benedict, A. Puzder, A. J. Williamson, J. C. Grossman, G. Galli, and J. E. Bertini, J. Chem. Theory Comput. 11, 5665 (2015).
21. M. Rohlfing and S. G. Louie, Phys. Rev. B 62, 4927 (2000).
22. L. X. Benedict, A. Puzder, A. J. Williamson, J. C. Grossman, G. Galli, J. E. Kolepas, J.-Y. Roy, and O. Pankratov, Phys. Rev. B 68, 085130 (2003).
23. E. Rabani, R. Baer, and D. Neuhauser, Phys. Rev. B 91, 235302 (2015).
24. S. Rafaely-Abramson, R. Baer, and L. Kronik, Phys. Rev. B 84, 075144 (2011).
25. M. Shishkin and G. Kresse, Phys. Rev. B 75, 235102 (2007).
26. F. Caruso, P. Rinke, X. Ren, A. Rubio, and M. Scheffler, Phys. Rev. B 88, 075105 (2013).
27. D. Neuhauser, Y. Gao, C. Amsden, C. Karshenas, E. Rabani, and R. Baer, Phys. Rev. Lett. 113, 076402 (2014).
28. H.-V. Nguyen, T. A. Pham, D. Roca, and G. Galli, Phys. Rev. B 85, 081101 (2012).
29. J. Deslippe, G. Samsonidze, D. A. Strubbe, M. Jain, M. L. Cohen, and S. G. Louie, Comput. Phys. Commun. 183, 1269 (2012).
30. D. Foerster, P. Koval, and D. Sánchez-Portal, J. Chem. Phys. 135, 074105 (2011).
31. X. Gonze, B. Amadon, P.-M. Anglade, J.-M. Beuken, F. Bottin, P. Boulanger, F. Brunoval, D. Caliste, R. Caracas, M. Côté et al., Comput. Phys. Commun. 180, 2582 (2009).
32. G. Stefanucci and R. van Leeuwen, Nonequilibrium Many-Body Theory of Quantum Systems: A Modern Introduction (Cambridge University Press, 2013).
33. J. Phillips and D. Zgid, J. Chem. Phys. 140, 241101 (2014).
34. A. A. Kanatenaka, J. J. Phillips, and D. Zgid, J. Chem. Theory Comput. 12, 564 (2016).
35. A. A. Rusakov and D. Zgid, J. Chem. Phys. 144, 054106 (2016).
36. N. E. Dahlen and R. van Leeuwen, J. Chem. Phys. 122, 164102 (2005).
37. A. R. Welden, J. J. Phillips, and D. Zgid, arXiv:1505.05575 (2015).
38. D. Neuhauser, R. Baer, and D. Zgid, J. Chem. Theory Comput. 13, 5396 (2017).
39. T. Y. Takeshita, W. Dou, D. G. A. Smith, W. A. de Jong, R. Baer, D. Neuhauser, and E. Rabani, J. Chem. Phys. 151, 044114 (2019).
40. W. Dou, T. Y. Takeshita, M. Chen, R. Baer, D. Neuhauser, and E. Rabani, J. Chem. Theory Comput. 15, 6703 (2019).
41. T. Y. Takeshita, W. A. de Jong, D. Neuhauser, R. Baer, and E. Rabani, J. Chem. Theory Comput. 13, 4605 (2017).
42. J. L. Whitten, J. Chem. Phys. 58, 4496 (1973).
43. B. I. Dunlap, J. Chem. Phys. 78, 3140 (1983).
44. B. I. Dunlap, J. W. D. Connolly, and J. R. Sabin, J. Chem. Phys. 71, 3396 (1979).
45. O. Vahtras, J. Almlof, and M. W. Feyerisen, Chem. Phys. Lett. 213, 514 (1993).
46. M. Feyerisen, G. Fitzgerald, and A. Komornicki, Chem. Phys. Lett. 208, 359 (1993).
47. M. Chen, R. Baer, D. Neuhauser, and E. Rabani, J. Chem. Phys. 150, 034106 (2019).
48. O. Arnon, E. Rabani, D. Neuhauser, and R. Baer, J. Chem. Phys. 152, 161103 (2020).
49. Y. Cytter, E. Rabani, D. Neuhauser, and R. Baer, Phys. Rev. B 97, 115207 (2018).
50. Q. Ge, Y. Gao, R. Baer, E. Rabani, and D. Neuhauser, J. Phys. Chem. Lett. 5, 185 (2013).
51. Y. Gao, D. Neuhauser, R. Baer, and E. Rabani, J. Chem. Phys. 142, 034106 (2015).
52. Due to the steep scaling of \( O(N^3) \), we were reluctant to compute the correlation energies deterministically for \( N > 100 \).