Diagrammatic Monte Carlo method for impurity models with general interactions and hybridizations

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We present a diagrammatic Monte Carlo method for quantum impurity problems with general interactions and general hybridization functions. Our method uses a recursive determinant scheme to sample diagrams for the scattering amplitude. Unlike in other methods for general impurity problems, an approximation of the continuous hybridization function by a finite number of bath states is not needed and accessing low temperatures does not incur an exponential cost. We test the method for the example of molecular systems, where we systematically vary temperature, interatomic distance, and basis set size. We further apply the method to an impurity problem generated by a self-energy embedding calculation of correlated antiferromagnetic NiO. We find that the method is ideal for quantum impurity problems with a large number of orbitals but only moderate correlations.

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1. INTRODUCTION

Quantum impurity models, originally introduced to describe magnetic impurities such as iron or copper atoms with partially filled d shells in a nonmagnetic host material [1], have since found applications in nanoscience as representations of quantum dots and molecular conductors [2] and in surface science to understand the adsorption of atoms on surfaces [3,4]. In addition, they form the central part of embedding theories such as the dynamical mean-field theory (DMFT) [5,6] and its variants [7–19], as well as the self-energy embedding theory (SEET) [20–22], where they describe the behavior of a few “strongly correlated” orbitals embedded into a weakly correlated or noninteracting background of other orbitals. These methods promise a systematic route for the simulation of strongly correlated quantum many-body problems [23].

While the original formulation of a quantum impurity model [1] only describes a single correlated orbital coupled to a noninteracting environment, in general the impurities occurring in the context of surface science and embedding theories contain many orbitals with general four-fermion interactions and few symmetries [24]. The time-dependent hybridization function describing the hopping between the impurity and its environment is typically such that it cannot be diagonalized for all frequencies at once.

Solving quantum impurity problems, i.e., obtaining the impurity Green’s function given an impurity Hamiltonian and a hybridization function, requires the use of numerical methods. A wide range of such methods exist. Hamiltonian-based methods, such as exact diagonalization (ED) [25–29] and its variants [30], configuration-interactions [31], or coupled cluster theory [32,33], solve the impurity problem by mapping the impurity problem onto a system with a local Hamiltonian and a finite number of auxiliary “bath” states chosen to fit the time-dependent hybridization function. The methods are limited to a relatively small set of strongly interacting sites or breakdown at moderate correlation strength. The bath fitting, which typically approximates a continuous bath dispersion by a nonlinear fit to a small number of delta-function peaks, introduces additional approximations [27,29]. Numerical renormalization techniques [34,35] overcome this issue by providing an almost continuous bath density of states but are in turn limited to a few orbitals in highly symmetrical situations.

A complementary approach is given by Monte Carlo techniques such as the continuous-time quantum Monte Carlo methods [36]. These methods are based on a stochastic sampling of the terms in a diagrammatic expansion of the partition function. For particle-hole symmetric systems with on-site density-density interactions, interaction expansion methods [37–39] can solve systems with hundreds of strongly correlated orbitals [40]. Away from particle-hole symmetry and at low temperature, they are typically limited to around eight orbitals, and their naive adaptation to general four-fermion operator terms suffers from a severe sign problem [41]. In contrast, a partition function expansion in the hybridization [42–44] is able to work with general local Hamiltonians of up to five orbitals, but is similarly restricted to diagonal hybridization functions. A reformulation [45] in terms of inchworm diagrams [46] overcomes the restriction of diagonal hybridizations, but so far remains limited to impurities with up to three orbitals.

There is therefore a need for impurity solver methods that can treat the problems of embedding theory and surface science, where several orbitals with general interactions and hybridizations occur. Diagrammatic Monte Carlo methods [47–51], which expand physical observables rather than partition functions, along with efficient ways of evaluating the resulting diagrammatic series via the connected determinant (CDet) approach [52–57], are promising. While these
methods suffer from other limitations, including divergences of the series in the strong correlation regime, they do not require to approximate the hybridization function by a fit, and are not based on a diagonalization of the local Hamiltonian.

In this paper, we show a formulation of the diagrammatic Monte Carlo method for impurity problems with general interactions and hybridizations based on the CDet framework. We test the method on the example of molecular systems, for which a broad range of very mature Hamiltonian methods exist. From the point of view of the algorithmic formulation, the molecular systems exhibit the full complexity of general impurity problems. The only difference between molecules and quantum impurities is that the latter are formulated with impurity problems. The only difference between molecules exist. From the point of view of the algorithmic formulation, we do not intend to present our method as a viable method for quantum chemistry systems without retardation effects. We carefully analyze the convergence behavior of the computational problem, the diagrammatic formulation, and the algorithmic description. In Sec. III, we present applications to molecular systems and benchmark results for quantum impurities. Finally, Sec. IV presents conclusions. Appendices A through E present technical details useful for implementing our algorithm and reproducing our results.

II. METHOD

A. Partition function expansion

We describe molecular electrons using the following Hamiltonian:

$$\hat{H} = \sum_{\alpha\beta} h_{\alpha\beta} \hat{c}_{\alpha}^\dagger \hat{c}_{\beta} + \frac{1}{4} \sum_{\alpha\beta\gamma\delta} U_{\alpha\beta\gamma\delta} \hat{c}_{\alpha}^\dagger \hat{c}_{\beta} \hat{c}_{\gamma}^\dagger \hat{c}_{\delta},$$

where $a, b, c, d$ denote spin orbitals, $1, \ldots, N$. We employ second quantization: $\hat{c}_{a}$ and $\hat{c}_{a}^\dagger$ annihilates and create, respectively, an electron in the spin-orbital $a$. The noninteracting term $\hat{H}_0$ is parametrized by the one-electron integrals $h_{\alpha\beta} = [a|\hat{h}|b]$, whereas the interacting term $\hat{H}_V$ is parametrized by the antisymmetrized two-electron integrals $U_{\alpha\beta\gamma\delta} = [ab|\hat{c}|cd] - [ad|\hat{c}|bc]$. We note that explicit antisymmetrization, $U_{\alpha\beta\gamma\delta} = -U_{\alpha\beta\gamma\delta}$, avoids ambiguities in the diagrammatic expansions below [59]. We orthonormalize the basis, $[\hat{c}_{a}^\dagger, \hat{c}_{b}] = \delta_{ab}$, as we empirically found this to improve the error bars in the subsequent Monte Carlo procedure. For completeness, we compiled the explicit expressions for $h$ and $U$ in Appendix A.

As we are going to perform series expansions later, it is convenient to introduce an expansion parameter $\xi$ into the Hamiltonian:

$$\hat{H}_\xi = \hat{H}_0 + \xi \hat{H}_V. \quad (2)$$

The noninteracting case is given by $\hat{H}_{\xi=0}$, whereas $\hat{H}_{\xi=1}$ recovers the full Hamiltonian Eq. (1).

We are primarily interested in calculating finite-temperature observables such as energies, densities, as well as the spectral function and other electronic correlation functions. We start with the grand-canonical partition function,

$$Z_\xi = \text{Tr} \exp[-\beta(\hat{H}_\xi - \mu \hat{N})], \quad (3)$$

where $\beta = 1/T$ is the inverse temperature, $\mu$ denotes the chemical potential, and $\hat{N} = \sum_a \hat{c}_a^\dagger \hat{c}_a$ is the density operator. Expanding Eq. (3) about $\xi = 0$ within the interaction picture [60] yields the Dyson series,

$$Z_\xi = Z_0 \sum_{k=0}^\infty \frac{(-\xi)^k}{k!} \int_0^\beta d\tau_1 \cdots \int_0^\beta d\tau_k \langle \hat{H}_V(\tau_1) \cdots \hat{H}_V(\tau_k) \rangle_0, \quad (4)$$

where $Z_0 := \text{Tr} \exp[-\beta(\hat{H}_0 - \mu \hat{N})]$ is the noninteracting partition function, $\tau$ denotes imaginary (Euclidean) time, $\langle \cdot \rangle_0$ denotes the noninteracting expectation value,

$$\langle \hat{X}_1 \cdots \hat{X}_k \rangle_0 := \frac{1}{Z_0} \text{Tr} [\text{e}^{-\beta(\hat{H}_0 - \mu \hat{N})} \mathcal{T} \langle \hat{X}_1 \cdots \hat{X}_k \rangle], \quad (5)$$

and $\mathcal{T}$ indicates path ordering in imaginary time. We note that for molecules, both $\hat{H}_0$ and $\hat{H}_V$ are bounded and thus away from zero temperature, the series expansion for the partition function Eq. (4) is absolutely convergent for all $\xi$.

Inserting Eq. (1) into Eq. (4) yields

$$\frac{Z_\xi}{Z_0} = \sum_{k=0}^\infty \frac{(-\xi)^k}{k!} \sum_{a_1 b_1 c_1 d_1} \cdots \sum_{a_k b_k c_k d_k} \int_0^\beta d\tau_1 \cdots \int_0^\beta d\tau_k \times \langle \hat{U}_{a_1 b_1 c_1 d_1} \cdots \hat{U}_{a_k b_k c_k d_k} \rangle_0. \quad (6)$$

To simplify our notation we combine four spin orbitals $a, b, c, d$ and an imaginary time $\tau$ into a single “vertex” $v = (a_v, b_v, c_v, d_v, \tau_v)$. We also introduce the following shorthands:

$$\int d^2 \mathcal{V} := \frac{1}{k!} \prod_{v \in \mathcal{V}} \sum_{a_v b_v c_v d_v} \int_0^\beta d\tau_v, \quad (7a)$$

$$D(\mathcal{V}) := \prod_{v \in \mathcal{V}} \left( -\frac{U_{a_v b_v c_v d_v}}{4} \right) \times \left( \prod_{v \in \mathcal{V}} \hat{c}_{a_v}^\dagger(\tau_v) \hat{c}_{b_v}^\dagger(\tau_v) \hat{c}_{d_v}(\tau_v) \hat{c}_{c_v}(\tau_v) \right) \rangle_0. \quad (7b)$$

Equation (7b) emphasizes the fact that the expectation value in Eq. (6) corresponds to the sum over all disconnected and connected Feynman diagrams with vertices $\mathcal{V} = (v_1, \ldots, v_k)$, while Eq. (7a) just corresponds to the sum over all internal
degrees of freedom of the diagrams. With these substitutions, Eq. (6) simplifies to
\[ Z_{\xi} = Z_0 + \sum_{k=0}^{\infty} \xi^k \int d^k \mathcal{V} D(\mathcal{V}). \]  
\[ (8) \]

To evaluate Eq. (8), we first introduce the noninteracting Green’s function:
\[ g_{ba}(\tau) = -\langle \hat{c}_b(\tau) \hat{c}_a(0) \rangle_0 = \langle \{ -\partial_\tau + \mu \} \mathbf{1} - h \rangle^{-1}_{ba}. \]  
\[ (9) \]
Given a diagram \( \mathcal{V} = (v_1, \ldots, v_k) \) with \( v_i = (a_i, b_i, c_i, d_i, \tau_i) \), we can use Wick’s theorem to write Eq. (7b) as
\[ D(\mathcal{V}) = \prod_{i=1}^{k} \left( - \frac{U_{a_i b_i c_i d_i}}{4} \right) \det \mathbf{G}(\mathcal{V}), \]  
\[ (10) \]
where \( \mathbf{G} \) is a \( 2k \times 2k \) matrix in which the rows (columns) correspond to the \( 2k \) annihilation (creation) operators. Introducing the column and row indices \( \alpha, \beta, \ldots \) such that
\[ \{ a_i \} := \{ a_1, c_1, a_2, c_2, \ldots, a_k, c_k \}, \]
\[ \{ b_j \} := \{ b_1, d_1, b_2, d_2, \ldots, b_k, d_k \}, \]
\[ \{ \tau_a \} := \{ \tau_1, \tau_1, \tau_2, \tau_2, \ldots, \tau_k, \tau_k \}, \]
we define the matrix elements
\[ [\mathbf{G}(\mathcal{V})]_{\beta a} := -\langle \hat{c}_b(\tau_\beta) \hat{c}_a(0) \rangle_0 = g_{b_\beta a_\alpha}(\tau_\beta - \tau_\alpha + 0^+). \]  
\[ (12) \]
The full matrix can be written in a block form as
\[ \mathbf{G}(\mathcal{V}) := \begin{bmatrix} g_{11} & g_{12} & \cdots & g_{1n} \\ g_{21} & g_{22} & \cdots & g_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ g_{n1} & g_{n2} & \cdots & g_{nn} \end{bmatrix}, \]  
\[ (13) \]
where each \( 2 \times 2 \) block is given by
\[ g_{ij} := \begin{bmatrix} g_{b_\beta a_\alpha}(\tau_\beta - \tau_\alpha + 0^+) & g_{d_\beta a_ \gamma}(\tau_\beta - \tau_\gamma + 0^+) \\ g_{b_\gamma c_\delta}(\tau_\gamma - \tau_\delta + 0^+) & g_{d_\gamma c_\varepsilon}(\tau_\gamma - \tau_\varepsilon + 0^+) \end{bmatrix}. \]  
\[ (14) \]

Equations (8) and (10) serve as the basis of interaction expansion continuous-time quantum Monte Carlo (CT-QMC): One generates random configurations \( (u_1, \ldots, u_k) \) and evaluates the corresponding weight by computing the determinant \( \det \mathbf{G}(\mathcal{V}) \). We can now consider the partition function \( Z_{\xi} \), defined as
\[ Z_{\xi} = \exp(-\beta \Omega_{\xi}). \]  
\[ (15) \]
\( \Omega_{\xi} \) serves as a cumulant-generating function for correlations functions \[ [36,37,41] \], and its power series in \( \xi \) is given by
\[ \Omega_{\xi} = \Omega_0 - \frac{1}{\beta} \sum_{k=1}^{\infty} \xi^k \int d^k \mathcal{V} D_{\xi}(\mathcal{V}), \]  
\[ (16) \]
where \( \Omega_0 \) is defined as \( Z_0 = \exp(-\beta \Omega_0) \).

The symbol \( D_{\xi} \) indicates that unlike in Eq. (8), the sum is to be performed over connected Feynman diagrams only. Using a recursion formula similar to the one introduced in Ref. [52], \( D_{\xi} \) can be defined recursively:
\[ D_{\xi}(\mathcal{V}) = D(\mathcal{V}) - \sum_{S \subseteq \mathcal{V}} \frac{|S|}{|V|} D_{\xi}(S) D(\mathcal{V} \setminus S). \]  
\[ (17) \]
A derivation is given in Appendix D. Equations (17) and (10) allow the computation of connected diagrams as a hierarchy of determinants at a cost of \( \mathcal{O}(3^k) \).

We note that even in simple cases, the convergence radius \( R \) of the series Eq. (16) is not infinite, with the value of \( R \) depending on \( h, U, \) and \( \beta. \) Whenever \( R < 1 \), an order-by-order summation of the series will fail. We will discuss strategies to extend the convergence radius in Sec. II E.

For convergent series \( (R > 1) \), one can employ the diagrammatic Monte Carlo algorithm to sample the series Eq. (16) by generating random vertices and computing the weight using the recursion Eq. (17). One observes that the relative statistical error diverges exponentially with diagrammatic order \( k \) \[ [62] \], which requires truncation of the series to a finite order \( k_{\text{max}} \).

### C. Scattering amplitude expansion

Other than free energy, we are primarily interested in the thermal correlation function of some operators \( \langle \hat{X}_1, \ldots, \hat{X}_m \rangle \),
\[ \langle \hat{X}_1 \cdots \hat{X}_m \rangle := \frac{1}{Z} \text{Tr} \left[ e^{-\beta (H - \mu \mathcal{N})} T(\hat{X}_1 \cdots \hat{X}_m) \right], \]  
\[ (18) \]
in particular, the single-particle Green’s function:
\[ G_{ab}(\tau) = -\langle \hat{c}_b(\tau) \hat{c}_a(0) \rangle. \]  
\[ (19) \]

One can write down a diagrammatic expansion for the Green’s function similar to Eq. (16) and a corresponding recursion relation \[ [52] \]. We instead choose to perform the expansion for a vertex-like object.

In the case of the expansion of the free energy, the corresponding one-particle vertex is the scattering amplitude \( M \) \[ [37,38] \], defined as
\[ G(\tau) = g(\tau) + \int_0^\beta dt_1 dt_2 g(\tau - t_1) M(\tau - t_1) g(\tau_2), \]  
\[ (20) \]
where multiplication is to be understood as matrix-matrix multiplication in spin orbitals. Sampling a one-particle vertex is advantageous because it is independent of the choice of “external legs” and thus allows measurements of both imaginary time-dependent quantities (\( G \) and \( \mathcal{X} \)) and fixed-time quantities (density, kinetic energy, etc.) in the same simulation.

\( M \) arises naturally as a functional derivative of the grand potential:
\[ M_{ab}(\tau) = \frac{\delta(\Omega - \Omega_0)}{\delta g_{ba}(-\tau)}. \]  
\[ (21) \]
We show this relation in Appendix B. Equation (21) expresses the fact that by removing one line from a (closed) free-energy diagram, we get an interaction correction to the Green’s function, which is exactly what the scattering amplitude encodes. Figure 1 provides a schematic example of diagrammatic representations of different quantities.

Combining Eq. (21) with Eq. (16) yields a series expansion for $M$:

$$M_{\xi,ab}(\tau) = -\frac{1}{\beta} \sum_{k=1}^{\infty} \xi^{k} \int d^{2}\nu \frac{\delta D_{c}(V)}{g_{ba}(-\tau)}.$$  \hspace{1cm} (22)

We thus need to evaluate the functional derivative of the recursion relation Eq. (17).

We start with the derivative of the sum of all diagrams $D(V)$, where we rely on the following identity:

$$\frac{\delta \det A}{\delta A_{a\beta}} = (\text{adj} A)_{\beta a} := (-1)^{a+\beta} \text{det} A_{a\beta},$$  \hspace{1cm} (23)

where $A$ is an $n \times n$ matrix, adj(A) denotes the $n \times n$ adjugate matrix of $A$, and $A_{a\beta}$ is the $(n - 1) \times (n - 1)$ submatrix of $A$ with the $a$th row and $\beta$th column removed. The adjugate matrix adj $A$ can be computed in $O(n^{3})$ time. The adjugate (or cofactor) matrix arises naturally in determinantal methods as a result of the Wick’s theorem [37,38,63], and is often absorbed into the inverse matrix if the matrix $A$ is not singular. In the context of CDet, however, care must be taken because $A$ may be singular while adj $A$ is still meaningful [64]. We elaborate on the numerical calculation in Appendix E.

Combining Eq. (10) with Eq. (23), we have

$$\frac{\delta D(V)}{g_{ba}(-\tau)} = \prod_{i=1}^{k} \left(-\frac{U_{a_{i}b_{i}a_{i}b_{i}}}{4}\right) \times \sum_{a,\beta} \left[\text{adj} G(V)\right]_{a\beta} \delta_{a_{i}a} \delta_{b_{i}b} \times \left[\delta(\tau_{a} - \tau_{\beta} - \tau) - \delta(\tau_{a} - \tau_{\beta} + \beta - \tau)\right]$$

for $0 < \tau \leq \beta$, where $a_{i}, b_{i}$, and $c_{b_{i}}(\tau_{\beta})$ takes the same meaning as in Eq. (12), and we have defined the $2k \times 2k$ matrix,

$$A(V) := -\prod_{i=1}^{k} \left(-\frac{U_{a_{i}b_{i}c_{i}d_{i}}}{4}\right) \text{adj} G(V),$$  \hspace{1cm} (25)

which includes all connected and disconnected amputated diagrams in which internal legs corresponding to $\hat{c}^{l}_{a_{i}}(\tau_{a})$ and $c_{b_{i}}(\tau_{\beta})$ are removed.

For the functional derivative of a connected free-energy diagram Eq. (26), the sum over all diagrams in Eq. (24) with amputated legs $A(V)$ needs to be replaced with the sum over connected diagrams with amputated legs $A_{c}(V)$,

$$\frac{\delta D_{c}(V)}{g_{ba}(-\tau)} = -\sum_{a,\beta} \left[\text{adj} G(V)\right]_{a\beta} \delta_{a_{i}a} \delta_{b_{i}b} \times \left[\delta(\tau_{a} - \tau_{\beta} - \tau) - \delta(\tau_{a} - \tau_{\beta} + \beta - \tau)\right],$$

for $0 < \tau \leq \beta$. The expansion of $M$ Eq. (22) can now be expressed in terms of $A_{c}$ as

$$[M_{\xi}(\tau)]_{a\beta} = \frac{1}{\beta} \sum_{k=1}^{\infty} \xi^{k} \int d^{2}\nu \sum_{a,\beta} \left[\text{adj} G(V)\right]_{a\beta} \delta_{a_{i}a} \delta_{b_{i}b} \times \left[\delta(\tau_{a} - \tau_{\beta} - \tau) - \delta(\tau_{a} - \tau_{\beta} + \beta - \tau)\right].$$

The sum over connected amputated diagrams $A_{c}(V)$ can be built up from a recursion technique similar to Eq. (17).
FIG. 2. Schematic illustration of the recursive removal of disconnected amputated diagrams. Empty boxes stand for the contribution of all diagrams (D or A), and filled ones for connected diagrams only \((A_0)\). Symbols inside boxes denote the set of vertices included in each component. The top relation shows all partitions of \([A(V)]_{\alpha \beta}\) into a subset fully connected to the amputated legs and a disconnected complement set. It is reorganized as the bottom relation which recursively defines \(A_\alpha(V)\) by removing all disconnected components.

Defining \(v_\alpha\) and \(v_\beta\) as vertices where the \(\alpha\)th and \(\beta\)th operators are located, respectively, diagrams in \([A(V)]_{\alpha \beta}\) can always be partitioned to a connected part which contains \(v_\alpha\) and \(v_\beta\), and the disconnected vacuum diagrams, i.e.,

\[
[A(V)]_{\alpha \beta} = \sum_{S \subseteq V} [A_\alpha(S)]_{\alpha' \beta'} D(V \setminus S),
\]

(28)

where \(\alpha', \beta'\) are row and column indices within \(S\) that correspond to the row and column indices \(\alpha, \beta\) in \(V\). Extracting the term with \(S = V\), we have the recursion relation for \(A_\alpha\) :

\[
[A_\alpha(V)]_{\alpha \beta} = [A(V)]_{\alpha \beta} - \sum_{S \subseteq V} [A_\alpha(S)]_{\alpha' \beta'} D(V \setminus S).
\]

(29)

This partitioning process is illustrated in Fig. 2. Since \(M\) captures the interaction correction to the Green’s function which starts at the first order in interaction, the zeroth order contribution \(A_\alpha(\emptyset) = 0\). For each fixed \(V\), we apply Eq. (29) to recursively compute \(A_\alpha(V)\), which in turn yields \(M\) following Eq. (26). Algorithmically, Eq. (29) can be evaluated by following Algorithm 1. Algorithm 1 runs in \(O(3^k k^2)\) time.

D. Observables from scattering amplitude

The electron self-energy \(\Sigma\) relates the Green’s function \(G\) to the noninteracting propagator \(g\) via the Dyson’s equation:

\[
G(\tau) = g(\tau) + \int_0^\beta d\tau_1 d\tau_2 g(\tau - \tau_1) \Sigma(\tau_1 - \tau_2) G(\tau_2).
\]

(30)

The expansion of the self-energy \(\Sigma\) can be interpreted as one-particle irreducible (1-PI) amputated diagrams, which stay connected even when any single propagator line is removed (cf. Fig. 1). The self-energy is thus not directly sampled, \(M\) and \(\Sigma\) are related to each other by [38]

\[
\int_0^\beta d\tau' \Sigma(\tau - \tau') G(\tau') = \int_0^\beta d\tau' M(\tau - \tau') g(\tau').
\]

(31)

Replacing \(G\) with Eq. (20), we have

\[
\Sigma^{-1}(iv) = M(iv)^{-1} + g(iv),
\]

(32)

where \(X(iv)\) denotes the Fourier transform of \(X(\tau)\) \((X = \Sigma, M, \ldots)\) and \(iv\) is a fermionic Matsubara frequency.

The one- and two-body contributions to the electronic energy follow from Eqs. (30) and (20):

\[
E = E_0 + E_V,
\]

(33a)

\[
E_0 = \langle \hat{H}_0 \rangle = \frac{1}{\beta} \text{Tr}[\hat{h}G] = \sum_{ab} h_{ab} \rho_{ab},
\]

(33b)

\[
E_V = \langle \hat{H}_V \rangle = \frac{1}{\beta} \text{Tr}[\Sigma G] = \frac{1}{2} \int_0^\beta d\tau \sum_{ab} M_{ab}(\tau) g_{ba}(-\tau).
\]

(33c)

Here \(\rho_{ij} = \langle \hat{c}_i^\dagger \hat{c}_j \rangle\) is the electron density matrix. Note that \(\hat{H}_0\) does not include the Hartree and Fock terms of the interaction. See also Appendix C.

E. Hartree-Fock-shifted Hamiltonian

In systems with significant electron-electron correlations where \(E_V\) has significant contribution to the full energy \(E\), the perturbation expansions in Eqs. (16) and (22) may not converge at \(\xi = 1\).

To achieve better convergence by starting from a “better” noninteracting solution such that \(\hat{H}_0\) is closer to \(\hat{H}\), we change the partition of the Hamiltonian \(\hat{H} = \hat{H}_0 + \hat{H}_V\) by adding physically motivated counterterms to \(\hat{H}_0\) and subtracting the same terms from \(\hat{H}_V\). Such an approach is referred to as the \(\alpha\)-shift [37] or as the shifted-action [65] in the action formalism.

Algorithm 1. Recursive evaluation of \(A_\alpha(V)\).

Require: Vertices \(V\), \(G(V)\) defined in Eq. (13).

1: function \text{RECURSION}(V, G)
2: if \(V = \emptyset\) then
3: \hspace{1em} return \(A_\alpha(\emptyset) = 0\).
4: \hspace{1em} else
5: \hspace{2em} Compute \(A(V)\) from \(G\) following Eq. (25).
6: \hspace{2em} Initialize \(A_\alpha(V) \leftarrow A(V)\).
7: \hspace{2em} for \(S \subseteq V\) do
8: \hspace{3em} Compute \(D(V \setminus S)\) following Eq. (10).
9: \hspace{3em} \(A_\alpha(S) \leftarrow \text{RECURSION}(S, G[S,S])\).
10: \hspace{2em} end for
11: \hspace{1em} return \(A_\alpha(V)\).
12: end function

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We start by adding the simplest counterterm in the quadratic form

$$\Delta H_\alpha = \sum_{ab} \alpha_{ab} \hat{c}_a \hat{c}_b$$  \hfill (34)

to $\hat{H}$ and subtract it from $\hat{H}_0$, such that

$$\hat{H}_{0,\alpha} = \sum_{ab} (\hat{h}_{ab} + \alpha_{ab}) \hat{c}_a \hat{c}_b,$$

$$\hat{H}_{\alpha,v} = \frac{1}{4} \sum_{abcd} U_{abcd} \hat{c}_a \hat{c}_d \hat{c}_b - \alpha_{ab} \hat{c}_a \hat{c}_b. $$  \hfill (35)

The total Hamiltonian $\hat{H} = \hat{H}_{0,\alpha} + \hat{H}_{\alpha,v}$ is unchanged, whereas the perturbation expansion of $\hat{H}_2 = \hat{H}_{0,\alpha} + \hat{\xi} \hat{H}_{\alpha,v}$ can be controlled by choosing different $\alpha$. The counterterm need not be quadratic in general. Though quadratic choices are convenient in the determinantal setup, recursion schemes have been developed for general counterterms [55].

The shifted noninteracting propagator

$$g^\mu_\alpha(\tau) = \left[(-\partial_\tau + \mu) I - \hat{h} - \alpha \right]^{-1} \hfill (37)$$

can be seen as a Green’s function with an $a$ priori self-energy $\alpha$.

In the molecular context, a significant contribution to electron correlations can be obtained by the Hartree-Fock approximation. We therefore choose $\alpha$ to be the Hartree-Fock self-energy, i.e., $\Sigma_{\text{HF}}$. $\Sigma_{\text{HF}}$ is given by the self-consistent equations at finite temperature:

$$[\Sigma_{\text{HF}}]_{ab} = \sum_{cd} U_{abcd} \rho_{cd}, \hfill (38a)$$

$$\rho = f(h + \Sigma_{\text{HF}} - \mu I). \hfill (38b)$$

Here $f(A) = [1 + \exp(\beta A)]^{-1}$ is the matrix-valued Fermi distribution function, and $\mu$ is the chemical potential which may be adjusted so the total number of electrons in the system is adjusted to charge neutrality.

Diagrammatically, the Hartree-Fock shift renormalizes the propagators to $g_\alpha$ and an additional effective two-point vertex $\alpha$ has to be included in diagrams. The effective vertex $\alpha$ cancels any diagram which has at least one vertex connecting to itself with exactly one propagator line. This removes all ”tadpole” diagrams in expansions of $G$ and $M$, as well as that of $\Omega$ except for the first-order diagram whose vertex connects to itself with two propagator lines. Figure 3 illustrates the cancellation of such diagrams.

Given a specific set of vertices $\mathcal{V}$, the removal of all tadpole diagrams is achieved by replacing the $G$ matrix Eq. (13) defined on internal vertices $\nu$ with

$$G(\mathcal{V}) := \begin{bmatrix} 0 & g^\nu_{12} & \cdots & g^\nu_{1n} \\ g^\nu_{21} & 0 & \cdots & g^\nu_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ g^\nu_{n1} & g^\nu_{n2} & \cdots & 0 \end{bmatrix},$$

$$g^\nu_{ij} := \begin{bmatrix} g^{\nu}_{b12}(\tau_i - \tau_j + 0^-) & g^{\nu}_{b13}(\tau_i - \tau_j + 0^-) \\ g^{\nu}_{b21}(\tau_i - \tau_j + 0^-) & g^{\nu}_{b23}(\tau_i - \tau_j + 0^-) \end{bmatrix}  \hfill (39)$$

i.e., by setting all $2 \times 2$ diagonal blocks (corresponding to self-connections of vertices) to zero, and replacing bare propagators with $g^{\nu}$. Using the modified definition of $G$ in Eqs. (10) and (25), one can carry out the same recursive calculations in Eq. (26) to obtain corresponding connected quantities.

Note that this introduces a bias in the free-energy evaluation by setting the first-order contribution (the “dumbbell” diagram) to zero, which needs to be corrected:

$$\Omega^{(1)}_\alpha = \frac{1}{2} \sum_{abcd} U_{abcd} g^{\nu}_{b1c}(0^-) g^{\nu}_{b2c}(0^+) - \sum_{ab} \alpha_{ab} g^{\nu}_{b1c}(0^-) = \frac{1}{2} \sum_{ab} [\Sigma_{\text{HF}}]_{ab} \rho_{ab}. \hfill (40)$$

In the remainder of this paper, we will always use a Hartree-Fock counterterm and omit the $\alpha$ subscripts.

### F. Monte Carlo integration of diagrammatic series

Evaluations of diagrammatic series, such as Eqs. (16) and (27), can be formally summarized as

$$X = \sum_{k=0}^{\infty} \int d^4 \mathcal{V} C(\mathcal{V}), \hfill (41)$$

where $X$ is the physical variable ($G, M, \ldots$), $\mathcal{V} = (v_1, \ldots, v_k)$ denotes spacetime indices of internal vertices, and $C$ the contribution of each fixed configuration of $\mathcal{V}$ to $X$. Here we take the “physical” value of the coupling constant $\xi = 1$. To perform a Monte Carlo integral, we introduce a cutoff $k_{\text{max}}$ of the expansion order, and an $a$ priori probability distribution of vertex spacetime indices $p(\mathcal{V})$ such that

$$p(\mathcal{V}) \geq 0, \quad \sum_{k=0}^{k_{\text{max}}} \int d^4 \mathcal{V} p(\mathcal{V}) \equiv 1. \hfill (42)$$

In addition, we require that $p(\mathcal{V}) > 0$ whenever $C(\mathcal{V}) \neq 0$. The order-$k_{\text{max}}$ approximation to $X$ can be estimated stochas-
and DIAGRAMMATIC MONTE CARLO METHOD FOR IMPURITY … PHYSICAL REVIEW RESEARCH 2, 033211 (2020)

With a large number following distribution fermionic Matsubara frequencies iν, implementation, we only measure the energy and from the scattering matrix for each frequency, where symbols with hats represent quantities in frequency representation as matrices in spin-orbital order contribution to the total energy:

\[
\langle \varepsilon(\tau) \rangle = \sum_{\beta}^{2|\nu|} \{ A_{\nu}(\nu) \} \left[ g_{\beta \alpha} \langle \tau_{\alpha} - \tau \rangle + \sum_{ab} [\Sigma_{HF}]_{\alpha \beta} g_{\beta \alpha} g_{\alpha \beta} \langle \tau_{\beta} - \tau \rangle \right],
\]

where \( \| \cdot \| \) denotes the Frobenius norm of a matrix, and \( \varepsilon(V) \) is the energy measurement defined in Table I.

\[
\varepsilon(V) = \frac{1}{2\beta} \sum_{\alpha=1}^{2|\nu|} \{ A_{\nu}(\nu) \}_{\alpha \beta} \left[ g_{\beta \alpha} \langle \tau_{\alpha} - \tau \rangle + \sum_{ab} [\Sigma_{HF}]_{\alpha \beta} g_{\beta \alpha} g_{\alpha \beta} \langle \tau_{\beta} - \tau \rangle \right].
\]

For efficient Monte Carlo simulations, it is important to choose the \textit{a priori} distribution \( p(V) \) to achieve importance sampling, such that the simulation samples more frequently when \( |C(V)| \) is large and less frequently otherwise. Since we measure multiple observables in one simulation, we need to define such a distribution that works for all measurements. We find in practice that the following choices provide efficient samplings for most measurements:

\[
X = \frac{C(V)}{C(V)}_{pE}, \quad W = \sum_{k=0}^{k_{\text{max}}} \int d^4V ||A_{\nu}(V)||, \quad p_{A}(V) = \frac{||A_{\nu}(V)||}{W_A}, \quad p_{E}(V) = \frac{|\varepsilon(V)|}{W_E},
\]

for each frequency, where symbols with hats represent quantities in frequency representation as matrices in spin-orbital indices. Resampling techniques such as the jackknife or the bootstrap are applied to avoid biased error estimations.

TABLE I. Measurements for observable quantities. The imaginary time convolution is defined as \( [f * g](\tau) = f(\tau') g(\tau') \). \( E_{0}^{HF} \) and \( E_{V}^{HF} \) are kinetic and potential energies from the Hartree-Fock solution.

| X | C(V) |
|---|---|
| \[e(V)\] | \[\varepsilon(V)\] |

with a large number \( N \) of Monte Carlo samples \( \{V_i\} \) generated following distribution \( p(V) \).

Since the Green’s function \( G \), self-energy \( \Sigma \), as well as the total electronic energy \( E = E_0 + E_V \) can all be derived from the scattering matrix \( M \) using Eqs. (20), (32), (33), it is sufficient to only keep track of the amputated diagrams \( A_{\nu}(V) \) and obtain all other observables as derived quantities. Table I summarizes some of these measurements. In our implementation, we only measure the energy and \( M(i\nu) \) with fermionic Matsubara frequencies \( i\nu \) on the fly, and construct \( G(i\nu) \) and \( \Sigma(i\nu) \) from \( M(i\nu) \), following:

\[
G(i\nu) = g(i\nu) + g(i\nu)M(i\nu)g(i\nu),
\]

\[
\Sigma(i\nu) = \Sigma_{HF}^{-1} = [M(i\nu)]^{-1} + g(i\nu)[g(i\nu)]^{-1},
\]

for each frequency, where symbols with hats represent quantities in frequency representation as matrices in spin-orbital indices. Resampling techniques such as the jackknife or the bootstrap are applied to avoid biased error estimations.

For efficient Monte Carlo simulations, it is important to choose the \textit{a priori} distribution \( p(V) \) to achieve importance sampling, such that the simulation samples more frequently when \( |C(V)| \) is large and less frequently otherwise. Since we measure multiple observables in one simulation, we need to define such a distribution that works for all measurements. We find in practice that the following choices provide efficient samplings for most measurements:

\[
p_{A}(V) = \frac{\|A_{\nu}(V)\|}{W_A}, \quad W_A = \sum_{k=0}^{k_{\text{max}}} \int d^4V \|A_{\nu}(V)\|, \quad p_{E}(V) = \frac{|\varepsilon(V)|}{W_E}, \quad W_E = \sum_{k=0}^{k_{\text{max}}} \int d^4V |\varepsilon(V)|,
\]

where \( \| \cdot \| \) denotes the Frobenius norm of a matrix, and \( \varepsilon(V) \) is the energy measurement defined in Table I.

\[
\varepsilon(V) = \frac{1}{2\beta} \sum_{\alpha=1}^{2|\nu|} \{ A_{\nu}(\nu) \}_{\alpha \beta} \left[ g_{\beta \alpha} \langle \tau_{\alpha} - \tau \rangle + \sum_{ab} [\Sigma_{HF}]_{\alpha \beta} g_{\beta \alpha} g_{\alpha \beta} \langle \tau_{\beta} - \tau \rangle \right].
\]

Here we have chosen \( p_{E} \) as the \textit{a priori} distribution. Any other measurements can now be estimated as:

\[
X = \frac{C(V)}{C(V)}_{pE}, \quad W = \sum_{k=0}^{k_{\text{max}}} \int d^4V |A_{\nu}(V)|, \quad p_{A}(V) = \frac{\|A_{\nu}(V)\|}{W_A},
\]

Similar relations apply when we use other choices of \textit{a priori} distributions or normalization measurements.

Once \( p(V) \) is defined, we generate Monte Carlo samples as a Markov chain via the Metropolis-Hastings algorithm. From each configuration \( V_i \), a new configuration \( V_j \) is proposed following some proposal probability distribution \( u^{\text{prop}}(V_j|V_i) \). To ensure detailed balance, an acceptance ratio \( R \) is calculated after each proposal as:

\[
R(V_j|V_i) = \frac{u^{\text{prop}}(V_j|V_i)p(V_j)}{u^{\text{prop}}(V_j|V_i)p(V_i)},
\]

The proposal \( V_i \rightarrow V_j \) is accepted with probability:

\[
u^{\text{acc}}(V_j|V_i) = \min(1, R(V_j|V_i)).
\]

This ensures the detailed balance of the Markov process, i.e.,

\[w(V_j|V_i)p(V_i) = w(V_i|V_j)p(V_j), \]

with \( p(V) \).

\[
E_{0}^{HF} - E_{0}^{HF}
\]

As mentioned in Table I, \( E_{0}^{HF} \) and \( E_{V}^{HF} \) are kinetic and potential energies from the Hartree-Fock solution.

\[
\sum_{\nu=0}^{\infty} \int d^4V \left[ g_{\beta \alpha} \langle \tau_{\alpha} - \tau \rangle + \sum_{ab} [\Sigma_{HF}]_{\alpha \beta} g_{\beta \alpha} g_{\alpha \beta} \langle \tau_{\beta} - \tau \rangle \right].
\]
where
\[
 w(V_j|V_i) = w^{\text{acc}}(V_j|V_i)w^{\text{prop}}(V_j|V_i)p(V_i). \tag{54}
\]
which guarantees samples obtain the equilibrium distribution \(p(V)\) after thermalization.

In molecular systems, due to the complexity in the multi-orbital Coulomb interaction tensor, as well as the energy differences in noninteracting energy levels, the configuration space of the Monte Carlo can be uneven and may lead to ergodicity problems in the random walk. We design the following set of updates which lead to an ergodic random walk in the configuration space for all systems we investigate in Sec. III:

1. **Vertex splitting**: Split a random vertex \(v = (a, b, c, d; \tau)\) to two new vertices \(v_1 = (a, b', c', d'; \tau')\) and \(v_2 = (a', b', c, d; \tau'')\). The new indices \(a', b', c', d', \tau'\) can be proposed by some *a priori* probability \(p^{\text{ms}}\). The proposal probability distribution for this update from order \(k\) to \(k + 1\) is
\[
 w^{\text{prop}}(v_1, v_2; k + 1|v; k) = \frac{p^{\text{ms}}(d', b', c', d', \tau')}{k} \tag{55}
\]
2. **Vertex merging**: Pick two random vertices \(v_1 = (a, b, c, d; \tau)\) and \(v_2 = (a', b', c, d; \tau')\) and merge them into \(v = (a, b, c, d; \tau)\). The proposal probability distribution from order \(k\) to \(k + 1\) is
\[
 w^{\text{prop}}(v; k|v_1, v_2; k + 1) = \frac{1}{k(k + 1)} \tag{56}
\]
3. **Vertex shift in time**: Update the time label \(\tau\) of a vertex \(v\) to a new value \(\tau'\).
4. **Vertex shift in orbitals**: Update one of the orbital labels \(a, b, c, d\) of a vertex \(v\) to a random new value.

Vertex shift in time or orbitals are self-balancing moves, hence the acceptance ratios shares the same form:
\[
 R(V_2|V_1) = \frac{p(V_2)}{p(V_1)}. \tag{57}
\]
Vertex splitting and merging are mutually inverse updates. The acceptance ratios are therefore
\[
 R(v_1, v_2; k + 1|v; k) = \frac{w^{\text{prop}}(v_1, v_2; k + 1|v; k) p(v_1, v_2; k + 1)}{w^{\text{prop}}(v_1, v_2; k|v; k) p(v_1, v_2; k + 1|v; k) [p(v_1, v_2; k + 1|v; k) p(v_1; k + 1|v; k)] \tag{58}
\]
There is considerable freedom in choosing \(p^{\text{ms}}\). For all systems we study in this paper, we choose \(p^{\text{ms}}\) such that
\[
 p^{\text{ms}}(d', b', c', d', \tau') = p^{\text{ub}}(d', b', c', d') p^{\text{time}}(\tau'), \tag{59}
\]
where \(p^{\text{ub}}(d', b', c', d')\) is uniformly distributed if the inserted indices can form nonzero propagator connections and zero otherwise, and
\[
 p^{\text{time}}(\tau') = \varphi(|\tau' - \bar{\tau}_V|), \tag{60}
\]
where \(\bar{\tau}_V = \frac{1}{N} \sum_{n=1}^{N} \tau_k\) is the average time coordinate of the existing vertices, and we choose \(\varphi(\tau)\) as a function in \([0, \beta]\) which has more weight near \(\tau = 0\) and \(\beta\) but still non-negligible weight in between. Since the Hartree-Fock propagators decay exponentially away from 0 and \(\beta\), this makes sure that the new vertex are more likely to stay close to existing vertices so the resulting configuration has a sizable contribution. In our implementation, we define
\[
 f(\tau) = \frac{\lambda}{\arctan(\beta \lambda)} \left[ \frac{1}{1 + ((\beta \tau) \lambda)^2} + \frac{1}{1 + [(\beta - \tau) \lambda]^2} \right], \tag{61}
\]
as a Lorentzian distribution where \(\lambda\) is an estimation of the overall energy scale of the system proportional to, e.g., the standard deviation of the Hartree-Fock energy levels.

### III. RESULTS

#### A. Series convergence

We first present a test of our method on a minimal molecular system: \(H_2\) in the STO-6g basis set [66]. Two hydrogen atoms are placed at distance \(r\) and finite temperature \(T = 1/\beta\). The basis set only contains the 1s orbital in each atom. This setup allows us to easily perform ED calculations of the full molecular Hamiltonian at any temperature, such that exact benchmark results for our CDet results are available.

In Fig. 4, we compare the total energy \(E_{\text{tot}}\) from CDet with order truncation \(k_{\text{max}}\) up to 6 to the ED energy at \(T = 50^{-1} E_h\), both as a function of \(r\). Around equilibrium distance \(r \approx 1.4 \alpha_0\), the CDet energy converges well to the ED solution. The system moves to the strongly correlated regime (i.e., a regime far from the Hartree-Fock solution), as we “stretch” the molecule by increasing \(r\). At \(r > 2.0 \alpha_0\), we start to observe significant systematic deviation at \(k_{\text{max}} = 6\). Since the kinetic energy of electrons moving between two atoms is significantly reduced as we increase \(r\) but the long-range Coulomb repulsion between electrons changes slowly, the electron-electron interaction becomes more important at larger \(r\), and hence it is expected that the perturbation expansion becomes more difficult to converge. This setup is standard in quantum chemistry [67] and is similar in spirit to lattice model setups in which a metal-to-insulator transition is induced by gradually increasing an on-site interaction.

Analytically, the convergence behavior is determined by the properties of the expanded quantity (e.g., \(E[\xi]\)) as a function of the coupling constant \(\xi\) on the complex plane, similar to the convergence analysis for many-body perturbation theory calculations at \(T = 0\) [68–72]. We evaluate the electron energy \(E[\xi]\) for complex values of \(\xi\) near \(\xi = 0\) using ED at \(r = 1.4, 2.0, 2.8,\) and \(3.6 \alpha_0\), following
\[
 Z[\xi] = \text{Tr}[e^{-\beta(\hat{H}_V + \hat{H}_a + \xi(\hat{H}_V - \hat{H}_a - \mu N)}], \tag{62}
\]
\[
 E[\xi] = \frac{1}{Z[\xi]} \text{Tr}[(\hat{H}_V + \hat{H}_a/2 + \xi(\hat{H}_V - \hat{H}_a/2)) \times e^{-\beta((\hat{H}_V + \hat{H}_a) + \xi(\hat{H}_V - \hat{H}_a - \mu N))}], \tag{63}
\]
where \(\hat{H}_a\) is the Hartree-Fock counterterm introduced in Sec. II E. One can show via a straightforward substitution that \(E[1]\) gives the physical electron energy and \(E[0]\) recovers the Hartree-Fock energy. Figure 5(a) shows the interaction correction \(E[\xi] - E[0]\) to the total energy, where the black dot represents the physical value at \(\xi = 1\). Since the convergence
radius of the power series around $\xi = 0$ is determined by the singularity (pole or branch cut) closest to the origin, the series is convergent at the physical point $\xi = 1$ if and only if there are no singularities in the unit circle (dashed circles in Fig. 5). At $r = 1.4 a_0$, all poles are far outside the unit circle, indicating a rapidly convergent series. As we increase $r$, poles move closer to the unit circle at $r = 2.0 a_0$, implying a slower convergence of the series, and finally enter the unit circle at $r = 2.8 a_0$ and $3.6 a_0$, resulting in divergent series at $\xi = 1$.

The analytic properties are reflected directly in the convergence behavior of the CDet results. For a direct comparison, we calculate the contribution of each order $k$ to the total energy $E_{\text{tot}}^{(k)}$ up to $k_{\text{max}} = 8$ for the same values of $r$, as shown in Fig. 6. At $r = 1.4 a_0$, $E_{\text{tot}}^{(k)}$ quickly converges to zero at $k > 4$. At $r = 2.0 a_0$, we observe a tendency to converge at $k = 8$ but nonzero systematic deviations remain. For $r = 2.8 a_0$ and $3.6 a_0$, no signs of convergence are observed up to $k = 8$.

The CDet approach can be applied to different temperatures without adding significant computational cost, as we will show in Sec. III B. This is fundamentally different from methods such as CT-QMC, where reaching lower $T$ is only possible at an exponential cost away from half filling [36]. In Fig. 7, we show the temperature dependence of the CDet total energy for $H_2$, STO-6g at $k_{\text{max}} = 6$, from $T = 10.0^{-1} E_h$ down to $T = 500.0^{-1} E_h$, in comparison to the ED solution at $T = 0$. All calculations use the same algorithmic setup and the same number of Monte Carlo steps. Convergence to the zero-temperature solution is observed as $T$ decreases, while the stochastic error estimation does not change significantly. Systematic deviations can be observed at similar locations ($r > 2.0 a_0$) for different temperatures, indicating similar convergence behavior for the same system at different temperature. This can be shown by the temperature dependence of the analytic structure of $E[\xi]$, as plotted in Fig. 5(b). As temperature is reduced, the spacing of the poles along the imaginary direction decreases proportionally, but the real-axis locations of the vertical “walls” of poles stay almost unchanged, which leads to similar convergence radii at different temperatures.

The Hartree-Fock shifted action plays an important role in achieving better series convergence in CDet. Figure 8 compares the ED analytic structure of the total energy $E[\xi]$ with and without the Hartree-Fock shift. Without the shift, even for the equilibrium distance $r = 1.4 a_0$ (usually considered weakly correlated), there are poles deep inside the unit circle, implying a highly divergent series at $\xi = 1$. In contrast, the Hartree-Fock shift pushes the poles away from the origin, which leads to a convergent series as seen in Figs. 4 and 6.

The CDet approach gives access to dynamic quantities, such as the Green’s function $G$ and the self-energy $\Sigma$, through the scattering amplitude $M$. The left column of Fig. 9 shows the CDet measurement of $\hat{M}(i\omega_n)$ in Matsubara frequency space up to $k_{\text{max}} = 6$ for $H_2$, STO-6g at $r = 1.4 a_0$ and $T = 50^{-1} E_h$. As we increase the expansion order, CDet results gradually converge to the ED solution (black lines), and at order 6 we observe only a small systematic error due to order truncation. The CDet self-energy $\Sigma$ is calculated from $M$ following Eq. (45). Both quantities exhibit similar behavior, as shown in the right column of Fig. 9. At order 3 and higher, the real part of $\Sigma(i\omega_n)$ takes nonzero value at the high-frequency limit, corresponding to the correction to the frequency-independent Hartree-Fock self-energy $\Sigma_{\text{HF}}$. The CDet Green’s function, derived from $M$ following Eq. (44), is shown in Fig. 10. Good agreement with ED is observed at $k_{\text{max}} = 6$ on the top panel, where both the Monte Carlo error estimation and the systematic error due to order truncation is much smaller than the symbol size. The bottom panel shows convergence of CDet Green’s function to ED by increasing $k_{\text{max}}$, with a small but visible systematic deviation at low frequency when $k_{\text{max}} = 6$.

The generality of our CDet implementation allows a straightforward extension to much larger basis sets. Going
FIG. 5. Analytic structure of electron total energy evaluated with ED as a function of the complex coupling constant $\xi$, $\text{H}_2$, STO-6g. Colors represent complex phases and brightness indicates the magnitude (see color bars). The black dot at $\xi = 1$ represents the “physical” result. The dashed black circle indicates minimum convergence radius necessary for the perturbation series to converge at $\xi = 1$. (a) Effect of changing $r$ at fixed temperature $T = 50^{-1} E_h$. For $r < 2.0 a_0$, CDet gives decent convergence to ED at $k_{\text{max}} = 4$, with both stochastic and systematic errors below 1 m$E_h$. The $2s$ and $2p$ orbitals added by cc-pVDZ basis and $3s$, $3p$, and $3d$ orbitals by cc-pVTZ basis are mostly unoccupied, and the most electron excitations occur near the lowest $1s$ orbitals. Consequently, the convergence behavior and computational cost of CDet do not change significantly from the minimal basis STO-6g.

Finally, we extend our method to bigger molecules by adding more hydrogen atoms to the system. We consider a chain of ten hydrogen atoms on a straight line with equal spacing $r$, the same benchmark system used in Ref. [59]. At minimal basis STO-6g, all ten $1s$ orbitals contribute equally to the active space of ten electrons. Compared to $\text{H}_2$ with cc-pVDZ, which has the same number of orbitals, $\text{H}_{10}$ with STO-6g has more orbitals relevant to electron correlations, and the cost of CDet is higher (for a detailed analysis see Sec. III B). The left column of Fig. 12 plots the CDet total energy up to $k_{\text{max}} = 4$ in comparison to ED solution at $T = 50^{-1} E_h$. Convergence within 5 m$E_h$ is achieved at $k_{\text{max}} = 4$ for $r < 2.4 a_0$, and systematic deviations are evident for $r > 2.4 a_0$. Similar behavior can be found in the the zero-temperature coupled cluster (CCSD) result (dotted lines), as both methods rely on the perturbative expansions of electron-electron interactions in different forms. The computational cost becomes much higher as we go to a bigger basis for $\text{H}_{10}$. With cc-pVDZ, there are 50 atomic orbitals in total, with potential excitations to the empty orbitals from all ten electrons. As shown in the right column Fig. 12, CDet still agrees with the reference method (MRCI+Q data from Ref. [59] at $T = 0$) for small values of $r$, but the Monte Carlo errors are significantly larger. Although our generic implementation has achieved decent extensibility without fine-tuning for each specific system, more efficient Monte Carlo estimators and sampling schemes as well as analytical resummation techniques should advance the limit of CDet to more complex systems.
Fig. 7. Temperature dependence of CDet total energy $E_{\text{tot}}$ for $\text{H}_2$, STO-6g, $k_{\text{max}} = 6$. Top panel: Comparison of ED at $T = 0$ and CDet at different $T$. Middle panel: Total energy with Hartree-Fock contribution removed. Bottom panel: Difference between ED and CDet.

B. Analysis of computational cost

The computational cost of a Markov chain Monte Carlo simulation, measured as the computational time needed for reaching a result for observable $X$ within a desired accuracy $\Delta X$, is determined by three factors. First, the cost of each individual update, which is $O(k^3 2^k)$ for a configuration at expansion order $k$ according to Algorithm 1. Second, the number of configuration updates needed to reach an independent sample by transversing a Markov chain of potentially correlated configurations, described by the integrated autocorrelation time $\tau_{\text{int}}$. Finally, the variance $\text{Var}(X)$ of the estimator of the quantity of interest (Table I), such that

$$\Delta X = \sqrt{\frac{\text{Var}(X)}{N}} \left(2 \tau_{\text{int}} + 1\right).$$

To assess the computational cost of our CDet implementation for reaching a certain uncertainty level, as well as how the effort changes with respect to temperature, choice of basis set, and system size, we perform a series of simulations of convergent series for the hydrogen chain $\text{H}_n$ with the same Monte Carlo updates and measurements for a fixed number of Markov chain iterations. In Fig. 13, we show estimates of autocorrelation effects, actual computational costs, and stochastic uncertainties in total energy, as functions of temperature $T$, the number of orbitals $N_{\text{orb}}$, or the number of

Fig. 8. Analytic structure of electron total energy as a function of a complex coupling constant $\xi$ with and without Hartree-Fock shift. Left (right) panel presents values of $E[\xi] - E[0]$ evaluated on the complex plane for $\text{H}_2$, STO-6g, $T = 50^{-1} E_h$ and $r = 1.4 a_0$ with (without) Hartree-Fock shift.

Fig. 9. CDet dynamic quantities for $\text{H}_2$, STO-6g, $T = 50^{-1} E_h$, $r = 1.4 a_0$. Shadings indicate Monte Carlo error estimates. Left column: Measured CDet Re$M(i\omega_n)$ (top panel) and Im$M(i\omega_n)$ (bottom panel) compared to ED. Right column: CDet self-energy in comparison to ED (excluding Hartree-Fock contribution $\Sigma_{\text{HF}}$), top (bottom) panel showing real (imaginary) part of $\Sigma(i\omega_n)$. Here we show the diagonal matrix element at orbital 1 for both $M$ and $\Sigma$. 
FIG. 10. CDet Green’s function in comparison to ED. H₂, STO-6g, T \(= 50^{-1} E_h\), \(r = 1.4 \text{a}_0\). Top panel: Values of \(\hat{G}(i\omega_n)\) at orbital 1. CDet results at \(k_{\text{max}} = 6\) are plotted as symbols and ED values as lines. Error bars are indicated but much smaller than symbol size. Bottom panel: Deviations of CDet results from ED at different \(k_{\text{max}}\). Solid (dashed) lines represent real (imaginary) part of \(\hat{G}_{11}(i\omega_n)\). Shadings indicate stochastic uncertainties of CDet.

hydrogen atoms \(n\) in log-log plots. We rescale the \(y\) values by an arbitrary factor to emphasize the respective scaling of these quantities in the same plot.

Figure 13(a) shows the temperature dependence of CDet simulations of a fixed system (H₂, STO-6g, \(r = 1.4 \text{a}_0\)) at \(k_{\text{max}} = 6\). We observe that the simulation time does not change significantly as we decrease temperature, indicating similar distributions of the expansion order (usually tilted to the highest order). The error estimate in total energy follows almost the same tendency as that of the autocorrelation effect \(\sqrt{2\tau_{\text{int}} + 1}\), indicating the underlying energy estimator does not have strong temperature dependence. The autocorrelation effect shows a slow power-law increase as temperature is lowered, implying that our Monte Carlo updates remain efficient at low temperature.

A similar analysis is shown in Fig. 13(b) for the basisset dependence of the same system (H₂, \(r = 1.4 \text{a}_0\)) at fixed temperature. We perform CDet simulations with \(k_{\text{max}} = 4\) for basis sets STO-6g, cc-pVDZ, and cc-pVTZ, with 2, 10, and 28 atomic orbitals, respectively. As we add more “virtual” orbitals to the system, the computational time increases slowly and the autocorrelation time even decreases as the additional orbitals improve the connectivity of Monte Carlo configurations. However, the stochastic error shows a different trend from the autocorrelation effect and increases (a fit with a power law results in \(\sim N_{\text{orb}}^{1.22}\)), meaning that the additional orbitals introduce more diagrammatic configurations with alternating signs that lead to stronger Monte Carlo fluctuations.

As we increase the systems size in Fig. 13(c) by adding more hydrogen atoms, the stochastic error (normalized by the system size \(n\)) at fixed computational time increases with a much larger power law than Fig. 13(b) (fitted \(\sim n^{3.48}\)), while the autocorrelation time barely changes. This implies that adding electrons that contribute to excitations near the Fermi level rapidly increases the complexity of the diagrammatics. The result is very different from the situation where additional basis states for the same number of electrons are added [Fig. 13(b)].

The behavior illustrated in Fig. 13(c) also differs from diagrammatic Monte Carlo applications with short-range or on-site interactions, which are formulated directly in the thermodynamic limit [47–49] and usually do not show strong scaling dependencies on system size. We suspect the difference is caused by the long-range nature of the bare Coulomb interaction, which introduces significant nonlocal electronic correlations as the system size increases. In this case, the use of bold (or screened) interactions instead of the bare Coulomb interactions, as performed in Ref. [59], may alleviate the problem. However, bold methods must deal with intrinsic issues of misconvergence to unphysical solutions [73]. Moreover, adapting such a method to the CDet framework requires further algorithmic development. This topic is under active development [55,65].

Thus, through the empirical analysis above, we have shown that for convergent series, the computational cost of our CDet implementation is not very sensitive to changes in temperature

FIG. 11. Total energy \(E_{\text{tot}}\) with Monte Carlo errors for H₂ with cc-pVDZ (left column) and cc-pVTZ (right column) basis sets, \(T = 50^{-1} E_h\). Top panels: Comparison of ED and CDet at different \(k_{\text{max}}\). Middle panel: Total energy with Hartree-Fock contribution removed. Bottom panels: Difference between ED and CDet.
or basis sets, but depends strongly on the size of the system or, more specifically, on the number of valence electrons directly participating in electron excitations.

C. Realistic impurity: SEET for NiO

Finally, we test our CDet implementation in a general quantum impurity problem setup that includes the coupling to a noninteracting bath. We employ the SEET framework [23] for the antiferromagnetic compound NiO, which was studied by Mott [74] as one of the original correlated insulators. Following the computational setup in Ref. [58], we choose fcc NiO with lattice constant \( a = 4.1705\text{Å} \) at temperature \( T \sim 451 \text{ K} \) (\( \beta = 700 E_h^{-1} \)). The unit cell is doubled along the [111] direction to capture the antiferromagnetic ordering, which contains two nickel atoms and two oxygen atoms. We use a \( 4 \times 4 \times 4 \) momentum discretization and GTH-DZVP-MOLOPT-SR basis set [75] with GTH-PBE pseudopotential [76]. The Coulomb integral is decomposed using density fitting with the DEF2-SVP-RI auxiliary basis [77]. For benchmark purposes, we select the \( e_g \) orbitals of both Ni atoms in the unit cell as the strongly correlated “impurities,” which is the minimal choice of impurities to capture correlation effects. This yields two independent impurities each with two orbitals. Noninteracting impurity propagators are generated from a converged GW simulation of the complete unit cell following the SEET framework (for details of the computational setup, see Ref. [58]).

As a benchmark, we compare our CDet impurity solver to the ED [78] results used in Ref. [58]. ED requires the discretization of the continuous bath spectrum and its approximation by a few states. To separate ED bath fitting errors from the performance of the CDet method, we run our method for the noninteracting impurity Green’s function \( g \) corresponding to the discretized noninteracting problem solved by ED. A

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FIG. 12. Total energy \( E_{\text{tot}} \) with Monte Carlo errors for H\(_{10}\) with STO-6g (left column) and cc-pVDZ (right column) basis. ED results are used as reference for STO-6g and MRCI+Q (\( T = 0 \)) from Ref. [59] for cc-pVDZ. Top panels: Comparison of reference data and CDet at different \( k_{\text{max}} \) at finite temperature \( T = 50^{-1} E_h \), along with ED and CCSD results at \( T = 0 \) for STO-6g basis. Middle panel: Total energy with Hartree-Fock contribution removed. Bottom panels: Difference between CDet and reference data at finite temperature (for STO-6g), in comparison to difference between CCSD and ED at zero temperature.

FIG. 13. Empirical cost analysis of CDet simulations of hydrogen chain H\(_n\) at \( r = 1.4 a_0 \). In each panel, all simulations are carried out using the same setup of Monte Carlo updates and number of iterations. We estimate the contribution of integrated autocorrelation time \( \tau_{\text{aut}} \) to the stochastic error (blue), the computational cost (orange), and the total stochastic uncertainty of energy \( \Delta E_{\text{tot}} \) (green) for each simulation, and scale them to the same range on double-logarithmic plots. (a) Temperature dependence, H\(_2\), STO-6g, \( k_{\text{max}} = 6 \). (b) Basis set dependence, H\(_2\), \( T = 50^{-1} E_h \), \( k_{\text{max}} = 4 \). (c) System size dependence, H\(_n\), cc-pVDZ, \( T = 50^{-1} E_h \), \( k_{\text{max}} = 4 \).
FIG. 14. Matsubara Green’s function for a NiO $e_g$ impurity. Top (bottom) row shows the real (imaginary) part of the Green’s functions, and left (right) column shows values for spin up (down). Red solid lines: Impurity Green’s function from CDet with $k_{\text{max}} = 8$. Monte Carlo error estimations are plotted as color shadings but smaller than line width. Blue dashed lines: Impurity Green’s function from ED, mostly overlapping with the CDet lines within line width. Green dash-dotted lines: Noninteracting impurity propagator with discretized hybridization from ED. Purple dotted lines: Impurity Green’s function with Hartree-Fock counterterm as the starting point of CDet.

FIG. 15. Difference of CDet impurity Green’s functions to ED at different truncation order $k_{\text{max}}$. Top (bottom) row shows the real (imaginary) part, and left (right) column shows values for spin up (down). Shadings indicate stochastic uncertainties of CDet.

The precise listing of all parameters and input Green’s functions is given in the Supplemental Material [79]. Figure 14 shows the impurity Green’s functions for one of the two $e_g$ impurities. At $k_{\text{max}} = 8$, the impurity Green’s function from CDet agrees with the ED solution within line width, and the stochastic uncertainty is almost invisible. The spin polarization due to the antiferromagnetic ordering is greatly enhanced in both the ED and the CDet solutions, indicating that dynamical correlations plays an important role and are well captured by the selected impurity.

In Fig. 15, we take a closer look at the convergence of CDet series in comparison to ED by plotting the differences of CDet impurity Green’s functions to ED at different order truncations up to $k_{\text{max}} = 8$. We observe that for both spins, the CDet result consistently converges to the ED result, giving agreement to within a percent for $k_{\text{max}} = 8$. The convergence of the spin-down component is slower than spin up, which is consistent to what can be observed in Fig. 14, i.e., the Hartree-Fock contribution already accounts for a greater part of the overall interaction contribution for the spin-up component than for spin down.

The comparison to ED illustrates that our CDet solver can be reliably applied to general impurity problems as part of a quantum embedding theory using the exact same framework as developed for molecules. We emphasize that, at the same complexity, systems with continuous bath states can be solved. Our method is therefore a controlled method for quantum impurities with general multiorbital interactions and hybridizations, not limited by the systematic error introduced by the bath discretization procedure. The application of the solver to more complex impurities, where ED calculations are impractical, is a topic of subsequent publications.

IV. CONCLUSION

In conclusion, we have presented a diagrammatic Monte Carlo method for quantum impurity models with general interactions and hybridizations using the connected determinant formalism [52]. We have tested the method at the example of molecular systems, which presents a systematic way of changing correlation strength, system size, basis size, and temperature. We have also tested our method for impurity problems occurring in realistic quantum impurity calculations.

Our method is formulated in the language of Green’s functions and self-energies. As a grand-canonical finite-temperature method, it is able to describe systems with particle-number fluctuations and excited states. However, similar to other perturbative methods, the diagrammatic series breaks down in the strong correlation regime. This breakdown is clearly evident in the order-by-order convergence of the series and, as we have shown in detail, can be traced back to the pole structure of the diagram series.

Our method fills a crucial need of impurity solvers able to treat general four-fermion interaction and general off-diagonal hybridizations in large multiorbital problems. It should therefore find applications in moderately correlated real-material simulations such as those occurring in DMFT [5,6] and SEET [20–23].

Further methodological progress, such as the use of higher order counterterms [55], better integration methods [80], complex conformal mapping techniques [81–83], and other types of Monte Carlo updates will expand the accessible parameter
regime of the method and may make simulations in the strongly correlated regime possible.

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APPENDIX A: HAMILTONIAN

We describe the molecular electrons using the second quantized Hamiltonian

\[ \hat{H} = \hat{H}_0 + \hat{H}_V, \]

\[ \hat{H}_0 = \sum_{pq} \sum_{\sigma} h_{pq}^\dagger c_{pq}^\sigma c_{pq}^\sigma, \]

\[ \hat{H}_V = \frac{1}{2} \sum_{pqrs} \sum_{\sigma\sigma'} V_{pqrs} c_{pq}^\sigma c_{rs}^\sigma c_{sr}^\sigma c_{qp}^\sigma, \]

where \( \phi_p^{(1)} \) is the electron annihilation (creation) operator associated with orbital \( \phi_p \) and spin \( \sigma \). The one- and two-body ‘matrix elements’ are defined as

\[ h_{pq} = \int d\mathbf{r} \phi_p^\dagger(\mathbf{r}) \left( -\frac{\nabla^2}{2m_e} - \sum_{i=1}^{N_n} \frac{Z_i}{|\mathbf{r} - \mathbf{R}_i|} \right) \phi_q(\mathbf{r}), \]

\[ V_{pqrs} = \int d\mathbf{r} d\mathbf{r}' \phi_p^\dagger(\mathbf{r}) \phi_q(\mathbf{r}) \phi_s^\dagger(\mathbf{r}') \phi_r(\mathbf{r}'), \]

where \( \{\mathbf{R}_i\} \) are the coordinates of the \( N_n \) nuclei, each with charge \( Z_i \), and \( m_e \) is the electron mass. Equations (A1) define a quantum many-body problem which can be studied with a wide range of approximate or exact theoretical and numerical methods.

To combine spin and orbital indices in Eqs. (A1), we introduce the compound notation \{a, b, \ldots\} such that

\[ h_{ab} = h_{pq \langle \sigma \rho \sigma' \rho' \rangle} \equiv h_{pq} \delta_{\sigma \sigma'}, \]

\[ V_{abcd} = V_{pqrs \langle \sigma \rho \sigma' \rho' \rangle} \equiv V_{pqrs} \delta_{\sigma \sigma'} \delta_{\rho \rho'}. \]

The fermionic antisymmetry can be explicitly encoded in the antisymmetrized interaction \( \hat{U}_{abcd} = \hat{V}_{abcd} - \hat{V}_{dcab} \), such that Eq. (A1) becomes

\[ \hat{H}_0 + \hat{H}_V = \sum_{ab} h_{ab} c_a^\dagger c_b + \frac{1}{4} \sum_{abcd} U_{abcd} c_a^\dagger c_b^\dagger c_c^\dagger c_d. \]

APPENDIX B: SCATTERING AMPLITUDE

The intuition in defining the \( M \) object is similar to the relation between the self-energy \( \Sigma \) and the Luttinger-Ward functional \( \Phi[G] \),

\[ \Sigma(x', x) = \frac{\delta \Phi[G]}{\delta G(x, x')}, \]

which gives the 1PI amputated diagrams with the bold propagator \( G \) [85]. Here we have employed the compound spacetime indices \( x = (a, \tau) \). To get the connected amputated diagrams with the bare propagator instead, we define a similar relation:

\[ M(x', x) = \frac{\delta (\Omega - \Omega_0)}{\delta g(x, x')} = -\frac{\delta \log Z_0}{\delta g(x, x')} \]

We show that by carrying out this functional derivative, we will recover the definition of \( M \) as in Eq. (20).

Switching to the action formalism using coherent state path-integrals of Grassmann variables [61], we rewrite the partition function as

\[ Z = \int \mathcal{D}[\bar{c}, c] e^{-S[\bar{c}, c]}, \]

where the action is given as

\[ S = S_0 + S_V, \]

\[ S_0 = -\int dy dy' \bar{c}(y)g^{-1}(y', y)c(y), \]

\[ S_V = \frac{1}{4} \int d\tau \sum_{abcd} U_{abcd} \bar{c}_a(\tau) \bar{c}_c(\tau) c_d(\tau) c_b(\tau). \]

Using the fact that for an invertible matrix \( A \),

\[ (A + \delta A)^{-1} - A^{-1} = -A^{-1} \delta A A^{-1}, \]

we have

\[ \frac{\delta \log Z}{\delta g(x, x')} = -\int dy dy' g^{-1}(y', y)G(y, y')g^{-1}(y', x). \]

Similarly, in the noninteracting case,

\[ \frac{\delta \log Z_0}{\delta g(x, x')} = -\int dy dy' g^{-1}(y', y)g(y, y')g^{-1}(y', x) = -g^{-1}(x', x). \]

Putting it all together, we have

\[ M(x', x) = -\frac{\delta \log Z_0}{\delta g(x, x')} = \int dy dy' g^{-1}(x', y') [G(y, y') - g(y, y')] g^{-1}(y', x). \]
Therefore,
\[ G(y, y') = g(y, y') + \int dx dx' g(y, x') \kappa^{x'}(x, y) g(x, y') \quad (B11) \]
which is exactly the same as Eq. (20).

Expanding compound indices \( x, y, \ldots \), Eq. (B2) can be rewritten as
\[ M_{ab}(t_1, t_2) = \beta \frac{\delta(\Omega - \Omega_0)}{g_{ba}(-\tau)} \delta(t_2 - t_1). \quad (B12) \]

In practice, we usually work with the time-translational invariant functions \( M(\tau) \) and \( g(\tau) \) instead of their two-variable form. To that effect, we consider for \( 0 < \tau \leq \beta \),
\[ \beta \frac{\delta(\Omega - \Omega_0)}{g_{ba}(-\tau)} = \sum_{a'b'} \int_0^\beta dt_1 dt_2 \left[ \beta \frac{\delta(\Omega - \Omega_0)}{g_{ba'}(t_2, t_1)} \frac{\delta(t_2 - t_1)}{g_{ba}(-\tau)} \right] \]
\[ = \sum_{a'b'} \int_0^\beta dt_1 dt_2 M_{a'b'}(t_1, t_2) \frac{\delta(t_2 - t_1)}{g_{ba}(-\tau)} \quad (B13) \]
in which
\[ \frac{\delta(t_2 - t_1)}{g_{ba}(-\tau)} = \frac{\delta(t_2 - t_1)}{g_{ba}(-\tau)} - \frac{\delta(t_2 - t_1 + \tau)}{g_{ba}(-\tau)} \Theta(t_1 - t_2) - \delta(t_2 - t_1 + \tau - \beta) \Theta(t_2 - t_1). \quad (B14) \]

Therefore,
\[ \beta \frac{\delta(\Omega - \Omega_0)}{g_{ba}(-\tau)} = \sum_{a'b'} \int_0^\beta dt_1 dt_2 M_{a'b'}(t_1, t_2) \]
\[ \times \frac{\delta(t_2 - t_1 + \tau)}{g_{ba}(t_1 - t_2)} \Theta(t_1 - t_2) - \delta(t_2 - t_1 + \tau - \beta) \Theta(t_2 - t_1)] \]
\[ = \int_0^\beta dt_1 dt_2 \left[ M_{ab}(t) \left\{ \delta(t_2 - t_1 + \tau) \Theta(t_1 - t_2) - \delta(t_2 - t_1 + \tau - \beta) \Theta(t_2 - t_1) \right\} \right] \]
\[ - M_{ab}(\tau) \int_0^\beta dt_1 dt_2 \left\{ \delta(t_2 - t_1 + \tau) \Theta(t_1 - t_2) + \delta(t_2 - t_1 + \tau - \beta) \Theta(t_2 - t_1) \right\} \]
\[ = M_{ab}(\tau) \int_0^\beta dt_1 \int_{t_1}^{t_1 + \beta} dt_2 \delta(t_2 - t_1 + \tau - \beta) \Theta(t_2 - t_1) \]
\[ = \beta M_{ab}(\tau), \quad (B15) \]
which gives Eq. (21).

**APPENDIX C: THERMAL EXPECTATION VALUE OF THE ELECTRON ENERGY**

The one-body energy is straightforward:
\[ E_0 = \langle \hat{H}_0 \rangle = \sum_{ab} \epsilon_a \rho_{ab} = \sum_{ab} \rho_{ab}. \quad (C1) \]

Expression for the two-body energy term can be derived in multiple ways such as using the equation of motion or the Schwinger-Dyson equation. Here we provide a simple derivation following Ref. [86]. We introduce a coupling constant \( \xi \) to the action defined in Eq. (B4) such that \( S_\xi = S_0 + \xi S_V \) and \( \xi \to 1 \) recovers the physical results. Now we have
\[ \frac{dZ_\xi}{d\xi} \bigg|_{\xi=1} = - \int D[\vec{c}], c e^{-S_0 - S_V} \bigg|\bigg|_{\xi=1} \]
\[ = -Z \int \tau \frac{\delta(t_2 - t_1)}{g_{ba}(-\tau)} \Theta(t_1 - t_2) - \delta(t_2 - t_1 + \tau - \beta) \Theta(t_2 - t_1)] \]
\[ = -Z \beta \langle \hat{H}_V \rangle. \quad (C2) \]

Introducing a change of variables such that \( c \to c/\xi^{1/2} \) and \( \tilde{c} \to \tilde{c}/\xi^{1/2} \), then
\[ D[\tilde{c}, c] = \lim_{N \to \infty} \prod_{a=1}^N d\tilde{c}_a d\epsilon_a \]
\[ \to \lim_{N \to \infty} \xi^{1/2} \prod_{a=1}^N d\tilde{c}_a d\epsilon_a \]
\[ = \xi^{-1/2} D[\tilde{c}, c], \quad (C3) \]
where the indices \( a \) denote states at each discretized time point on the integration path, \( \text{Tr}[I] = \int dx \delta(x, x) \) in which \( x \) is the compound spacetime index, and the plus sign on the exponent is due to the nature of Grassmann integrals. The partition function is unaffected by the change of variables, which now takes the form
\[ Z_\xi = \xi^{1/2} \int D[\tilde{c}, c] e^{-\xi^{1/2} S_0 - S_V}. \quad (C4) \]

Therefore,
\[ \frac{dZ_\xi}{d\xi} \bigg|_{\xi=1} = \frac{\text{Tr}[I]}{2} Z + \int D[\tilde{c}, c] e^{-\xi^{-3/2} S_0 e^{-\xi^{1/2} S_V - S_V}} \bigg|\bigg|_{\xi=1} \]
\[ = \frac{\text{Tr}[I]}{2} Z - \frac{1}{2} \int D[\tilde{c}, c] e^{g^{-1}(x', x) \tilde{c}(x') c(x) e^{-S}} \]
\[ = \frac{\text{Tr}[I]}{2} Z - Z g^{-1}(x', x) G(x, x'). \quad (C5) \]
Comparing Eqs. (C2) and (C5), we have
\[ \langle \hat{H}_V \rangle = \frac{1}{2\beta} \text{Tr}[g^{-1} G - I] = \frac{1}{2\beta} \text{Tr}[g^{-1} - G^{-1}] G \]
\[ = \frac{1}{2\beta} \text{Tr}[\Sigma G]. \quad (C6) \]

**APPENDIX D: RECURSION RELATIONS FOR THE GRAND POTENTIAL**

The expansion of the grand potential, Eq. (16), consists of connected vacuum diagrams as shown in Fig. 1. As a result of Wick’s theorem Eq. (10), all vacuum diagrams \( D(V) \) for a fixed vertex configuration \( V \) can be partitioned into a connected subdiagram and the remainder of the vacuum components. Since no external legs exist to serve as reference
The initial condition is the zeroth order contribution

\[ D(\mathcal{V}) = \sum_{S \subseteq \mathcal{V}} D_c(S) D(\mathcal{V} \setminus S). \]  

(D1)

As the choice of \( v \) is arbitrary, it can be any of the \( k = |\mathcal{V}| \) vertices in \( \mathcal{V} \), therefore

\[ D(\mathcal{V}) = \frac{1}{|\mathcal{V}|} \sum_{v \in \mathcal{V}} \sum_{S \subseteq \mathcal{V}} D_c(S) D(\mathcal{V} \setminus S). \]  

(D2)

This is equivalent to iterating all possible subsets \( S \) of \( \mathcal{V} \), where the reference \( v \) can be any vertex in \( S \):

\[ D(\mathcal{V}) = \frac{1}{|\mathcal{V}|} \sum_{S \subseteq \mathcal{V}} \sum_{v \in S} D_c(S) D(\mathcal{V} \setminus S) \]

\[ = \sum_{S \subseteq \mathcal{V}} \frac{|S|}{|\mathcal{V}|} D_c(S) D(\mathcal{V} \setminus S). \]  

(D3)

We now extract the term where \( \mathcal{V} = \mathcal{V} \) from the right-hand side and obtain the recursive formula for \( D_c(\mathcal{V}) \):

\[ D_c(\mathcal{V}) = D(\mathcal{V}) - \sum_{S \subseteq \mathcal{V}} \frac{|S|}{|\mathcal{V}|} D_c(S) D(\mathcal{V} \setminus S). \]  

(D4)

The initial condition is the zeroth order contribution \( D_c(\emptyset) = 0 \).

A more general framework of deriving the recursion relations using idempotent polynomials is described in Ref. [53]. This framework does not resort to topological arguments.

**APPENDIX E: NUMERICAL COMPUTATION OF THE ADJUGATE MATRIX**

We calculate the adjugate \( \text{adj}(A) \) of a matrix \( A \in \mathbb{R}^{n \times n} \) numerically by first performing a rank-revealing factorization on the matrix \([87]\), such as the pivoted QR via the Householder algorithm

\[ A = QDRP \]  

(E1)

where \( Q \) is an orthogonal matrix of Householder reflections, \( D \) is a diagonal matrix, \( R \) is an upper triangular matrix in which all diagonal elements equal 1, and \( P \) is a permutation matrix of the columns. The rank of the matrix \( r = \text{rank}(A) \) is determined by the number of nonzero diagonal elements of \( D \).

If \( A \) is not singular, i.e., \( r = n \), then the adjugate is given by

\[ \text{adj}(A) = \det(A) A^{-1} \]

\[ = \det(P) \det(D) \det(Q) P^T R^{-1} D^{-1} Q^T \]

(E2)

where \( n_P \) is the number of transpositions in the permutation \( P \), \( n_Q \) is the number of Householder reflections in \( Q \), and \( d_i \) are diagonal elements of \( D \). The scaling as a function of \( n \) for the complexity of calculating the adjugate is the same as the one for calculating \( A^{-1} \), and we obtain \( \det(A) \) at the same time.

If \( A \) is singular, i.e., \( r < n \), \( \det(A) \) becomes zero, and Eq. (E2) is replaced by

\[ \text{adj}(A) = \left[ (-1)^{r n_Q} \prod_{i=1}^n d_i \right] P^T R^{-1} \text{adj}(D) Q^T. \]  

(E3)

If \( r = n - 1 \), there is one zero in the diagonal of \( D \). Assuming \( d_n = 0 \) and \( d_i \neq 0 \) for \( i = 1, \ldots, n - 1 \), the adjugate of \( D \) follows directly from the definition Eq. (23):

\[ \text{adj}(D) = \text{diag} \left( 0, \ldots, 0, \prod_{i=1}^{n-1} d_i \right). \]  

(E4)

If \( r < n - 1 \), \( \text{adj}(D) = 0 \), therefore \( \text{adj}(A) = 0 \).

In the presence of off-diagonal propagators, it is possible that the amputated diagram \( A(\mathcal{V}, \emptyset) \) is nonzero while the vacuum diagrams \( D(\mathcal{V}, \emptyset) \) vanish. Therefore, it is crucial to implement the adjugate of singular matrices as discussed above.

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