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Projected density matrix embedding theory with applications to the two-dimensional Hubbard model

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ABSTRACT

Density matrix embedding theory (DMET) is a quantum embedding theory for strongly correlated systems. From a computational perspective, one bottleneck in DMET is the optimization of the correlation potential to achieve self-consistency, especially for heterogeneous systems of large size. We propose a new method, called projected DMET (p-DMET), which achieves self-consistency without needing to optimize the correlation potential. We demonstrate the performance of p-DMET on the two-dimensional Hubbard model.

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I. INTRODUCTION

Strong correlation effects play an important role in many quantum systems and require treatment beyond the level of mean-field theories. However, the absence of a mean-field starting point means that a direct treatment of strong correlations, for example, by full configuration interaction (FCI)1–3 or exact diagonalization (ED),4,5 scales exponentially with respect to system size. This motivates the development of quantum embedding theories,6–8 which partition the global system into a series of fictitious and strongly correlated fixed-size “impurities,” which are then treated accurately via a high-level theory (such as FCI or ED). The solutions from all impurities are then coupled together via a lower-level theory. This procedure often needs to be performed self-consistently. Examples of such quantum embedding theories include dynamical mean field theory (DMFT)9–12 and density matrix embedding theory (DMET).13–17 DMET has been successfully applied to compute phase diagrams of a number of strongly correlated models, such as the one-band Hubbard model on different lattices,18–21 quantum spin models,22,23 and some prototypical correlated molecular problems.13,17,24 This work focuses on improving numerical algorithms to achieve self-consistency in the context of DMET. In DMET, each impurity problem consists of two parts: a fragment and a bath. The low-level theory is used to construct the bath, which is defined by the Schmidt decomposition of the low-level wave function between the fragment and the remaining part of the global system. The standard choice in DMET is to choose the low-level theory to be a mean-field theory (such as Hartree-Fock theory) so that the one-particle reduced density matrix (1-RDM, also simply referred to as the density matrix here) is idempotent, and the bath space is fully spanned by a set of one-particle bath orbitals, which can be efficiently obtained by factorizing the 1-RDM between the one-particle degrees of freedom associated with the fragment and those in the global system. The high-level theory is then used to evaluate the 1-RDM and the two-particle reduced density matrix (2-RDM) associated with each impurity, which can be assembled
according to the DMET democratic partitioning protocol 17 to evaluate the total energy and other physical observables, such as correlation functions.

In many systems, using the bath orbitals generated from the Hartree-Fock Slater determinant yields energies and physical observables from DMET that already significantly improve on those from Hartree-Fock theory alone. Such calculations will be referred to as “single-shot” DMET calculations. 17 On the other hand, when the physical system undergoes a phase transition not predicted by mean-field theory, we expect that a mean-field theory will produce the wrong order parameter and the resulting bath orbitals will be very poor. In such a scenario, it is necessary to perform DMET self-consistently to improve the bath orbitals. The self-consistency condition is usually defined such that the 1-RDMs obtained from the low-level and high-level theories match each other according to some criterion, such as matching the density matrix in the impurity problem, 17 on the fragment only, 17 or simply matching the diagonal elements of the density matrix (i.e., the electron density). 17 Self-consistency can be achieved by optimizing a single-body potential, termed the correlation potential, in the low-level theory. Each optimization step requires diagonalizing a matrix, similar to in a self-consistent field (SCF) iteration step in the solution of the Hartree-Fock equations.

Nonetheless, there are two outstanding numerical issues associated with the optimization of the correlation potential. First, the optimization procedure may require a large number of iterations to converge. It is not uncommon for the number of iterations to be 100 ~ 1000 especially for systems that are not translationally invariant. Hence, when the system size becomes moderately large (a few hundred sites), the cost of the correlation potential optimization may exceed the cost of the impurity solver for small impurities. Second, the bath construction procedure of DMET requires the 1-RDM to be an idempotent matrix, and the corresponding low-level Hamiltonian should have a finite HOMO-LUMO gap. However, even if the strongly correlated global system is gapped, it is often the case that the low-level Hamiltonian associated with a given correlation potential in the optimization procedure becomes gapless. The derivative of the bath orbitals with respect to the correlation potential will then become infinite and the optimization cannot properly proceed. This work aims at addressing the first problem, namely, the cost associated with the correlation potential optimization. The second problem should be addressed by properly considering the zero temperature limit of a finite temperature generalization of DMET, which will be studied in the future.

We will introduce an alternate procedure to self-consistently determine the bath orbitals, which completely avoids the need to optimize the correlation potential. In the standard DMET, the bath orbitals are uniquely determined by the corresponding idempotent 1-RDM obtained from a low-level theory, denoted $D^0$. The goal of the self-consistent DMET can then be formulated, in an abstract way, as finding the solution of the following fixed point problem:

$$D^0 = \mathcal{F} \left[ \mathcal{D}(D^0) \right].$$

(1)

Here, the mapping $\mathcal{D}[\cdot]$ takes the idempotent 1-RDM as the input, generates the corresponding bath orbitals, and solves all impurity problems to obtain the 1-RDM evaluated from the high-level theory. The mapping $\mathcal{F}$ takes the high-level correlated 1-RDM, denoted by $D^h := \mathcal{D}(D^0)$ as the input, and generates another idempotent 1-RDM. The correlation potential optimization can be viewed as one way of achieving self-consistency as required by (1). To see this, we only need to define the mapping $\mathcal{F}$ to be the minimization procedure in the standard DMET, which uses a correlation potential to minimize the discrepancy between $D^h$ and $D^0$ evaluated on the impurity problems.

The perspective from the fixed point equation (1) suggests that other forms of $\mathcal{F}$ are possible which map $D^h$ to $D^0$ more efficiently. We propose that $D^0$ can be obtained by directly projecting $D^h$ onto the set of idempotent matrices with a given rank. The modified method is therefore called the projected density matrix embedding theory (p-DMET). The solution of p-DMET will not be identical to that of DMET, since they are defined using different mappings $\mathcal{F}$. In particular, unlike DMET, which can be defined to only use information from $D^h$ on the fragments during self-consistency, p-DMET requires the construction of $D^h$ on the global domain in order to define the projection operation.

Using the two-dimensional one-band Hubbard model and restricting to magnetic and nonmagnetic self-consistent solutions, we demonstrate that the results of p-DMET and DMET at self-consistency are very similar within the antiferromagnetic (AFM) and the paramagnetic (PM) phases. The discrepancy between the two methods is the largest near the phase boundary and for larger on-site interactions. We show that p-DMET significantly lowers the computational cost to achieve self-consistency for large lattices without translational invariance. For example, even for a moderately sized lattice with 128 sites and using 4 site impurities (without translational invariance), the correlation potential fitting procedure in standard DMET requires about 20 000 s of CPU time, which is reduced to about 1 s in the p-DMET approach.

The rest of this paper is organized as follows: In Sec. II, we briefly review DMET and the correlation potential optimization problem. We then discuss p-DMET and the associated numerical issues in Sec. III. We demonstrate the performance of p-DMET for the two-dimensional Hubbard model in Sec. IV, before deriving conclusions in Sec. V. The procedure for constructing the bath orbitals from the low-level RDM is summarized in the Appendix.

II. REVIEW OF DENSITY MATRIX EMBEDDING THEORY

Let us consider a system of $L_A$ sites and $N_e$ electrons, which is partitioned into $N_f$ nonoverlapping fragments. Without loss of generality, we assume that the system is spinless; spinful systems can be represented as spinless systems by doubling the system size. We assume that all of the fragments contain $L_A$ lattice sites. Figure 1 shows a $10 \times 10$ lattice partitioned into fragments of size $2 \times 2$ ($L_A = 4$). The (particle number-conserving) Hamiltonian of the global system takes the form

$$\hat{H} = \sum_{pq} t_{pq} \hat{a}_p^\dagger \hat{a}_q + \frac{1}{2} \sum_{pqrs} (pr|qp) \hat{a}_p^\dagger \hat{a}_r^\dagger \hat{a}_q \hat{a}_s,$$

(2)

where $\hat{a}_q$ ($\hat{a}_s$) are electron creation (annihilation) operators and $t_{pq}$ and $(pr|qp)$ are one- and two-electron integrals, respectively.
For any given fragment $A$, the ground state wave function of the global system $|Ψ⟩$ can always be partitioned as

$$|Ψ⟩ = ∑_{i=1}^{N_A} ∑_{j=1}^{N_B} Ψ_{ij}(A_i|B_j).$$  (3)

Here, $|A_i⟩$ represents a state in the fragment and $|B_j⟩$ represents a state in the environment $B$. $N_A$ and $N_B$ are the numbers of states in the fragment and environment, respectively. In general, $N_B ≫ N_A$.

Conceptually, we may perform a singular value decomposition (also called the Schmidt decomposition in the current context) of the coefficient matrix $Ψ_{ij}$, which yields

$$|Ψ⟩ = ∑_{i=1}^{N_A} ∑_{j=1}^{N_B} U_{ij} V_{ji} |A_i⟩⟨B_j|$$  (4)

Here, $|A_i⟩$ are the singular values. The unitary matrices $U$ and $V$ are absorbed into the transformed states $|A_i⟩$ and $|B_j⟩$. $|B_α⟩$ defines a bath state for fragment $A$. Equation (4) suggests that the maximum number of bath states is bounded by $N_A^2$.

For each fragment, we define the impurity Hilbert space to be the product space of the fragment states $|A_α⟩$ and the associated bath states. Hence, a basis for the impurity part (excluding the bath) of the global system according to a chosen partitioning, such as the democratic partitioning for the global system according to a chosen partitioning, such as the democratic partitioning. In general, $N_B ≫ N_A$.

Given the impurity Hamiltonians (5), we can solve for each of the impurity ground-states. In the special case of translational invariant systems, we can solve for a single impurity ground-state and use the fact that expectation values of all other impurities can be obtained by translation. From 1-RDMs computed from these ground-states, we can assemble a high-level correlated 1-RDM $D^{hl}$ for the global system according to a chosen partitioning, such as the democratic partitioning (also see Sec. III). Many numerical experiments show that obtaining the correct number of electrons from $D^{hl}$ is extremely important for the accuracy of observables. However, the number of electrons computed from the high-level correlated 1-RDM $D^{hl}$, that is, $\text{Tr}(D^{hl})$, generally does not match the target number of electrons in the system $N_e$. A global chemical potential $μ$ is often introduced on the fragment part (excluding the bath) of each impurity Hamiltonian. More specifically,

$$H^{emb}_f = H^{emb}_f - \begin{bmatrix} I_{N_A} & 0 \\ 0 & 0 \end{bmatrix}.$$  (6)

Here, $I_{N_A}$ is an $N_A \times N_A$ identity matrix. The condition $\text{Tr}(D^{hl}) = N_e$ is achieved by adjusting the global chemical potential. The global chemical potential can be viewed as a Lagrangian multiplier to enforce the correct number of electrons in $D^{hl}$ in the optimization problem.

To formulate DMET in a self-consistent fashion, we then refer to Eq. (1). For a given correlated 1-RDM $D^{hl}$, the mapping $F$ is defined as

$$D^{hl} := F[D^{hl}] = \arg \min_{D^{hl}} \| (D^{hl} - D) \circ W \|_F.$$  (7)

Here, $⊙$ is the elementwise product and $W$ is a weight matrix. If we choose $W$ to be the identity matrix, then Eq. (7) only measures the discrepancy of the 1-RDMs on the diagonal entries. This gives rise to the density embedding theory. We may also let $W_{pq} = 0$ if the indices $p$, $q$ do not belong to the same fragment and otherwise $W_{pq} = 1$. In this case, $W$ is a block diagonal matrix and each block is a matrix of ones. Equation (7) measures the sum of discrepancies on each fragment.

In standard DMET, the admissible set $Ω$ is the subset of idempotent density matrices generated by a correlation potential added to a one-particle Hamiltonian. More precisely, let $u_{pq}$ be a sparse Hermitian matrix, satisfying the condition $u_{pq} = 0$ if $p$, $q$ do not belong to the same fragment. The effective one-particle Hamiltonian of the low-level theory is

$$\hat{H}^{hl} = \hat{a}^\dagger (t + u) \hat{a},$$  (8)

where

$$\hat{a}^\dagger = [\hat{a}_1^\dagger \ldots \hat{a}_n^\dagger], \quad \hat{a} = [\hat{a}_1 \ldots \hat{a}_n]^T$$  (9)

define the collection of all creation operators and annihilation operators, and $u$ is called the correlation potential. $D^{hl}$ is the ground-state of the global Slater determinant, and the procedure is summarized in the Appendix. The projector $\tilde{P}^{(s)}$ can be replaced by a projector onto the Hilbert space of the fragment and bath orbitals. The contribution of the core orbitals amounts to an embedding potential in $H^{emb}_f$. We refer the reader to Ref. 17 for a more detailed description of these steps.
1-RDM corresponding to \( \hat{D}^B \) satisfying \( \text{Tr}[D^B] = N_e \), where \( N_e \) is the number of electrons in the system. For this procedure to be well defined, there should be a positive gap between the \( N_e \)th and \( (N_e + 1) \)th eigenvalues of \( \hat{D}^B \). Under this condition, the optimization problem (7) can be equivalently formulated as finding the optimal correlation potential. When \( u \) is a block diagonal Hermitian matrix, the number of independent variables is \( (\frac{(L_e+1)L_e}{2}) = \frac{(L_e+1)L_e}{2} \). The minimization can be carried out using derivatives with respect to \( u \), which can be explicitly calculated using perturbation theory. The optimization problem can then be numerically solved by quasi-Newton or conjugate gradient type methods. As discussed in Sec. I, the optimization of the correlation potential can be a bottleneck in larger systems, particularly when translational invariance is not present.

III. PROJECTED DENSITY MATRIX EMBEDDING THEORY

Motivated by the fixed point formulation of self-consistency (1), we propose the following procedure to obtain \( D^B \):

\[
D^B = \mathcal{F}[D^B] := \arg \min_{D=D^B, D^B = D, \text{Tr}(D) = N_e} \left\| \left( D - D^B \right) \otimes W \right\|_F. \tag{10}
\]

Compared to (7), the main simplification of Eq. (10) comes from the fact that the admissible set is now the set of idempotent density matrices with \( N_e \) electrons without further constraints. To further simplify the method, we let each entry of \( W \) be 1, i.e., \( W_{pq} \equiv 1 \). In other words, we measure the discrepancy of all entries of the density matrix on the same footing and solve

\[
D^B = \arg \min_{D=D^B, D^B = D, \text{Tr}(D) = N_e} \left\| D - D^B \right\|_F. \tag{11}
\]

Equation (11) has a simple analytic solution. Let \( \Psi^B \) be the eigenvectors corresponding to the largest \( N_e \) eigenvalues of \( \hat{D}^B \), i.e., \( \Psi^B \) consists of the leading \( N_e \) natural orbitals. Then, the solution to Eq. (11) is

\[
D^B = \Psi^B (\Psi^B)^\dagger. \tag{12}
\]

If \( D^B \) is fixed, this is the closest projection operator to \( D^B \) measured by the Frobenius norm. Informally, \( \Psi^B \) is the single determinant that best captures the information contained in all the density matrices of the fragments. Once \( D^B \) is obtained, we may compute the bath orbitals and the core orbitals according to the procedure in the Appendix and proceed to solve for the ground-state of the impurity problems as in the standard DMET procedure. Hence, we refer to this method as projected density matrix embedding theory (p-DMET). Again to make this procedure well defined, we require that there is a positive gap between the \( N_e \)th and \( (N_e + 1) \)th eigenvalue of the correlated density matrix \( \hat{D}^B \). p-DMET assumes that the high-level 1-RDM \( D^{(1)} \) of the global system has been computed. As mentioned above, such a global 1-RDM can be constructed from the high-level 1-RDMs in each impurity using the democratic partitioning. This procedure was briefly described in Ref. 17, and we give a more explicit description for completeness now.

Let \( C_x \) be the collection of fragment and bath orbitals of the impurity \( x \) as in Eq. (A8). Then, we define

\[
\begin{bmatrix}
D^{(s)} \\
D^{(e)} \\
\vdots \\
D^{(N)}
\end{bmatrix} = \bar{D}^{x(s)} C_x^\dagger, \tag{13}
\]

where \( \bar{D}^{x(s)} \) is the 1-RDM of the impurity problem with size \( (2L_A) \times (2L_A) \). \( \bar{D}^{x(s)} C_x^\dagger \) is a matrix of size \( (2L_A) \times L \). \( D^{(1)} \) is obtained by extracting the first \( L_A \) rows of \( \bar{D}^{x(s)} C_x^\dagger \), which is a block row of the density matrix corresponding to the fragment part of the impurity \( x \). Since the fragments collectively form a nonoverlapping partitioning of the global system, the global 1-RDM can be formed as

\[
D^B = \begin{bmatrix}
D^{(1)} \\
D^{(2)} \\
\vdots \\
D^{(N)}
\end{bmatrix}. \tag{14}
\]

Figure 2 illustrates the procedure of constructing \( D^B \) for one-dimensional model with 12 sites partitioned into 6 fragments.

Since each block row of \( D^B \) is obtained from the corresponding impurity problem independently, in general, \( D^B \) is not a Hermitian matrix. Therefore, after Eq. (14), we symmetrize the 1-RDM as

\[
D^B \leftarrow \frac{D^B + (D^B)^\dagger}{2}. \tag{15}
\]

This symmetrization procedure corresponds to the choice of "democratic partitioning" for constructing the 1-RDM (and its contribution to the total energy) in DMET. In general, \( D^B \) will not be an idempotent matrix. Hence, there is a nonzero discrepancy between \( D^B \) and \( D^B \). In order to solve the fixed point problem (1) in p-DMET, an extrapolation or mixing scheme is usually beneficial to accelerate the convergence. In p-DMET, we choose \( D^B \) as the mixing variable. Let \( D^{B,(k)} \) be the correlated 1-RDM at the beginning of the

![FIG. 2](image-url) Construction of a global density matrix. Each fragment contributes a rectangular block row of the global density matrix. Each block row has the same length as the global density matrix.
In order to obtain the mixing coefficient \( \alpha \), we usually converge when \( < \| \alpha_k - \alpha_{k+1} \| \), and solve the impurity problem. From this, we obtain an output correlated density matrix corresponding bath orbitals, and solve the impurity problem. From this, we obtain an output correlated density matrix \( \tilde{D}^\text{hl} \) as the mixing variable. However, there arises a practical problem related to this choice, namely that the linear combination of a few (or even two) idempotent matrices is generally not an idempotent matrix. Note that the same problem already existed in the context of Hartree-Fock calculations. Some of us have recently developed the proposed commutator DIIS (PC-DIIS) method, which accelerates Hartree-Fock calculations within a large basis set such as plane waves. The idea of PC-DIIS is to apply the idempotent density matrix \( D^\text{p} \) to a gauge-fixing matrix \( D^\text{ref} = F \Phi^\text{ref} \). It is clear that the information in \( D^\text{p} \) and \( \Phi \) is equivalent. In particular, \( D^\text{p} \) can be reconstructed from \( \Phi \) (Löwdin orthogonalization) as

\[
D^\text{p} = \Phi (\Phi^\dagger \Phi)^{-1} \Phi^\dagger.
\]

Thus, PC-DIIS uses \( \Phi \) as the mixing variable and reconstructs the idempotent density matrix using Eq. (20).

\[ \{ \alpha_j \} = \arg \min_{\{ \alpha \}} \| \sum_{j=k-\ell}^k a_j \alpha_k \|_F, \text{ s.t. } \sum_{j=k-\ell}^k \alpha_j = 1. \]  

A pseudocode implementation for the p-DMET and DMET algorithms is provided in Algorithm 1.

In principle, one could also choose the mean field density matrix \( D^\text{mf} \) as the mixing variable. However, there arises a practical question related to this choice, namely that the linear combination of a few (or even two) idempotent matrices is generally not an idempotent matrix. Note that the same problem already existed in the context of Hartree-Fock calculations. Some of us have recently developed the proposed commutator DIIS (PC-DIIS) method, which accelerates Hartree-Fock calculations within a large basis set such as plane waves. The idea of PC-DIIS is to apply the idempotent density matrix \( D^\text{p} \) to a gauge-fixing matrix \( D^\text{ref} = F \Phi^\text{ref} \). It is clear that the information in \( D^\text{p} \) and \( \Phi \) is equivalent. In particular, \( D^\text{p} \) can be reconstructed from \( \Phi \) (Löwdin orthogonalization) as

\[
D^\text{p} = \Phi (\Phi^\dagger \Phi)^{-1} \Phi^\dagger.
\]

Thus, PC-DIIS uses \( \Phi \) as the mixing variable and reconstructs the idempotent density matrix using Eq. (20).

**Algorithm 1.** A unified pseudocode for the projected density matrix embedding theory (p-DMET) and density matrix embedding theory (DMET).

**Input:** Initial guess of the correlated 1-RDM \( D^\text{hl}(0) \) and chemical potential \( \mu^{(0)} \).

**Output:** Converged correlated 1-RDM \( D^\text{hl} \) and ground state energy \( E \).

1. **for** \( k = 0, \ldots, \) **do**
2. \( \mu^{(k+1)} \leftarrow \mu^{(k)} \)
3. **for** \( m = 0, \ldots, \) **do**
4. **for each impurity** \( \times \) **do**
5. Compute fragment orbitals \( (C_i) \) and bath orbitals \( (C_j) \) \( \triangleright \) Section: Appendix
6. Formulate the impurity Hamiltonian, \( \hat{H}_\text{emb}^{(c)} \) \( \triangleright \) Eq. (16) and (17) in Ref. 17
7. Solve the impurity problem with \( \mu^{(k,m)} \) \( \triangleright \) via solvers such as FCI or DMRG
8. **end for**
9. Compute the total number of electrons, \( \text{Tr}(D^\text{hl}) \)
10. if convergence is not reached, update \( \mu^{(k,m+1)} \) by Newton’s iterations
11. **end for**
12. \( \mu^{(k+1)} \leftarrow \mu^{(k,m)} \)
13. Construct the correlated 1-RDM \( D^\text{hl} \) \( \triangleright \) Eq. (14) followed by symmetrization
14. Compute energy \( E^{(k)} \) from \( D^\text{hl} \) as well as the related 2-RDM \( \triangleright \) Eq. (28) in Ref. 17
15. if convergence is reached, exit the loop
16. if embedding method is p-DMET **then**
17. Perform mixing scheme to obtain \( D^{\text{hl}(k+1)} \)
18. Compute low-level density matrix \( D^\text{p} \) \( \triangleright \) Eq. (12)
19. **else if** embedding method is DMET **then**
20. Solve the minimization problem \( \triangleright \) Eq. (7)
21. Perform mixing scheme to obtain a new correlation potential \( \alpha^{(k+1)} \)
22. **end if**
23. **end for**
24. Set \( D^\text{hl} \leftarrow D^{\text{hl}(k)} \), \( E \leftarrow E^{(k)} \)
Following the PC-DIIS method, we can then choose \( \Phi \) as the mixing variable in the self-consistent p-DMET. The gauge-fixing matrix \( \Phi^{\text{ref}} \) can be chosen to be, for instance, the Hartree-Fock occupied orbital coefficient matrix. We refer to this method as the projected density matrix embedding theory with a fixed gauge (p-DMET-f). Note that from the perspective of Eq. (1), p-DMET-f solves the same fixed point problem as p-DMET. The only difference is the choice of the mixing variable.

For translational invariant systems, note that \( D_{\text{hl}} \) constructed from democratic partitioning does not break the translational symmetry among fragments. Therefore, the \( D_{\text{hl}} \) can be represented in \( k \)-space, denoted by \( D_{\text{hl}}(k) \). The corresponding \( \Psi_{\text{hl}}(k) \) is generated per \( k \) sector. The extrapolation over \( D_{\text{hl}}(k) \) also conserves the crystal momentum \( k \). Finally, we note that p-DMET-f can be formulated in a similar way by introducing a gauge-fixing matrix per \( k \) sector.

IV. NUMERICAL EXPERIMENTS

In this section, we investigate the performance of p-DMET and p-DMET-f for a 2D Hubbard model with periodic boundary conditions. The mean-field theory is chosen to be the unrestricted Hartree-Fock (UHF) theory, and the impurity Hamiltonian in DMET and p-DMET is defined within the interacting bath formulation as in (5). The impurity ground-states were computed using the FCI method implemented in the PYSCF package and the density matrix renormalization group (DMRG) method,32 as implemented in the BLOCK program,33 using a bond dimension of \( M = 1000 \), the split-localized orbital strategy described in Ref. 19, and the genetic algorithm for orbital ordering.21 The fragments were chosen to be \( 2 \times 2 \) clusters, treated without translational invariance, to allow a comparison between p-DMET and DMET in a general setting, except in the case of the cluster size convergence tests, where fragments of up to \( 4 \times 4 \) (16 sites) were used, and translational invariance was assumed. We used a convergence criterion on the energy difference between two consecutive iterations of less than \( 10^{-8} \) for the \( 2 \times 2 \) clusters and \( 10^{-9} \) for the larger clusters. All energies are reported in units of hopping (\( t \)).

A. Accuracy

To investigate the accuracy of p-DMET, we plot the phase diagram of a 2D Hubbard model with \( 40 \times 40 \) sites with periodic boundary conditions. This system has been studied in Ref. 19 using a translationally invariant implementation of DMET. The initial 1-RDM is produced by a converged UHF calculation. Figure 3 compares the phase diagrams generated by UHF, DMET, and p-DMET, respectively, evaluated on a \( 21 \times 21 \) grid with respect to the on-site interaction strength \( U \), as well as the filling factor \( n \).

The phase diagram is divided into two regions distinguished by their spin polarization, i.e., the antiferromagnetic (AFM) phase and the paramagnetic (PM) phase. The phase diagrams obtained from p-DMET and DMET are qualitatively similar. The two diagrams agree well with each other when \( U \leq 4.0 \) and larger discrepancies between p-DMET and DMET are observed in the region \( U > 4.0 \) and \( 0.6 \leq n \leq 0.8 \). We also observe that the phase boundary obtained from p-DMET is slightly softer, i.e., the decay of the spin polarization from the AFM phase to the PM phase is slower than that in DMET.

![Phase diagrams of the 2D Hubbard model from UHF, DMET, and p-DMET.](image)

**TABLE I.** Energy per site of the 2D Hubbard model by UHF, p-DMET, and DMET as a function of \( U \) and doping (\( n \)). The self-consistent p-DMET and DMET calculations use the converged UHF solution as the initial guess.

| \( U \) | \( n = 1.000 \) | \( n = 0.875 \) | \( n = 0.750 \) |
|---|---|---|---|
| | UHF | p-DMET | DMET | UHF | p-DMET | DMET | UHF | p-DMET | DMET |
| 2.0 | -1.13886 | -1.17999 | -1.17985 | -1.22470 | -1.27817 | -1.27799 | -1.27655 | -1.32270 | -1.32275 |
| 4.0 | -0.79703 | -0.86792 | -0.86856 | -0.88440 | -1.03002 | -1.03450 | -0.99530 | -1.16862 | -1.16707 |
| 6.0 | -0.59270 | -0.66099 | -0.66188 | -0.66592 | -0.87265 | -0.87395 | -0.75936 | -1.04709 | -1.06860 |
| 8.0 | -0.46588 | -0.52262 | -0.52393 | -0.52665 | -0.77299 | -0.77149 | -0.60439 | -0.97734 | -0.98954 |
A quantitative comparison of the total energy per site can be found in Table I. Overall, the discrepancy between p-DMET and DMET is much smaller than that between UHF and DMET. The energies of p-DMET and DMET agree very well (the difference is less than $10^{-3}$) inside the AFM/PM phases. The largest discrepancy occurs at $U = 6.0$, $n = 0.750$, again near the phase boundary, where the difference of the energy is 0.022. We remark that neither p-DMET nor DMET is variational, so we cannot determine from this single calculation which is better.

### B. Convergence

We observe in Fig. 3 that the softer phase boundary in p-DMET coincides with the region where UHF and DMET predict different phases. Since the UHF solution only enters p-DMET as an initial guess, we may wonder whether the fixed-point of p-DMET depends on the initial guess. Below, we demonstrate that the converged p-DMET/p-DMET-f solution can indeed depend on the initial guess, at least in certain parts of the phase diagram. We consider a 2D Hubbard system with $6 \times 6$ sites with periodic boundary conditions. The onsite interaction $U$ is set to 4.0, and we consider two fillings: $n = 1.0$ (half filling, $N_e = 36$) and $n = 0.722$ ($N_e = 26$). In both cases, the energy gap at the mean-field level is positive, and the self-consistent procedure for all methods is well defined without any finite temperature smearing.

In the first example ($n = 1.0$), DMET suggests that at convergence the system is in the AFM phase. Therefore, we start from the PM phase and break the spin symmetry of the initial density by alternately adding/subtracting a small number ($10^{-3}$) of odd/even sites to create slightly polarized spin-up and spin-down densities. Starting from this initial density, UHF converges within 20 steps using DIIS. We input the initial 1-RDM for p-DMET/p-DMET-f after performing 1, 5, 10, and 20 UHF iterations, respectively. For p-DMET-f, the gauge-fixing matrix is also obtained from the same 1-RDM. The convergence of the energies is reported in Fig. 4. We find that the convergence curves of p-DMET and p-DMET-f are very similar and almost coincide with each other in all cases. Both the converged energy and the spin polarization from p-DMET/p-DMET-f depend on the initial guess of the 1-RDM. Table II suggests that at convergence, UHF predicts an overpolarized spin configuration. However, starting from a significantly underpolarized 1-RDM obtained from one iteration of UHF, the converged solution of p-DMET underestimates the spin polarization (by 0.022 relative to the converged DMET result). If we improve the initial guess by using more UHF iterations, both the converged energy and spin polarization of p-DMET are closer to the results of DMET.

After 5-steps of UHF for the initial guess, p-DMET provides converged results in terms of energy and spin polarization. Remarkably, the solution of DMET is very robust with respect to the choice of the initial guess, even though neither DMET nor p-DMET/p-DMET-f guarantees a unique solution to the nonlinear fixed point problem \textit{a priori} (Table III).

In the second example, we set $n = 0.722$ ($N_e = 26$). We start from an initial guess that exhibits AFM order, where the spin up component of the density is set to be 0.1444 and 0.5778 on alternate sites and the spin down component is arranged alternately in the opposite direction.

![Figure 4](https://example.com/figure4.png)

**FIG. 4.** The convergence of the energy per site for p-DMET-f, p-DMET, and DMET with different initial guesses. The system is a $6 \times 6$ Hubbard model at $U = 4.0$, $n = 1.0$.

| Initial UHF | $m$ | UHF | $m$ | UHF | $m$ | UHF | $m$ | UHF | $m$ | UHF | $m$ | UHF |
|-------------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| Initial UHF | 0.001 | 0.34876 | 0.30812 | 0.30812 | 0.30812 | 0.30812 | 0.28578 | 0.30851 | 0.31000 |

**TABLE III.** Spin polarization for the $6 \times 6$ Hubbard model at $U = 4.0$, $n = 0.278$ obtained from converged UHF, DMET, and p-DMET calculations. $\#UHF$ stands for the number of UHF steps used to obtain the initial 1-RDM for the DMET and p-DMET calculations.

| Initial UHF | $m$ | UHF | $m$ | UHF | $m$ | UHF | $m$ | UHF | $m$ | UHF | $m$ | UHF |
|-------------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| $n = 1.0$ | 0.2167 | $2.166 \times 10^{-1}$ | $4.961 \times 10^{-8}$ | $1.365 \times 10^{-7}$ | $5.440 \times 10^{-8}$ | $4.175 \times 10^{-2}$ | $2.719 \times 10^{-2}$ | $4.865 \times 10^{-9}$ |
AFM phase with a small spin-polarization of 0.04. Both the energy and spin polarization improve quickly as the number of UHF iterations used to define the initial guess increases. Eventually, p-DMET also predicts a PM phase. Again, no initial guess dependence is observed in DMET.

In both examples above, UHF and DMET predict the same phase of matter. We find that self-consistent p-DMET can significantly reduce the error of physical observables starting from converged or unconverged UHF solutions, but there is some initial guess dependence. Therefore, when UHF and DMET predict different phases, p-DMET can reduce but not eliminate the (presumed) error from UHF. Hence, the phase diagram in Fig. 3 is similar to that of DMET, but softer near the boundary region.

C. Effect of fragment size

To understand the fragment size dependence of the physical observables in DMET and p-DMET, we carried out a number of 2D Hubbard model calculations (of 40 $\times$ 40 sites) at half-filling for different interaction strengths ($U = 2$, 4, 6, and 8) and with different fragment sizes ($2 \times 2$, $2 \times 4$, and $4 \times 4$) using translational invariance. The same cluster sizes were previously considered in Ref. 19 where translationally invariant DMET is in the noninteracting bath (NIB) formulation only. We use the data from Ref. 19 as the reference. DMET in the interacting bath formulation, as used everywhere else in this work, is here denoted DMET (IB). Given a set of 2D
clusters with $L_A$ sites, the DMET energy can be extrapolated to the thermodynamic limit (TDL) as a power series in $L_A^{-2}$:

$$E(L) = E(\infty) + a_0 L_A^{-2} + b_0 \left( L_A^{-1} \right)^2 + \ldots$$

(21)

We use the average of linear regression and a quadratic fit as the extrapolated result, and the error bar is defined to be the difference of the two fits.

Figure 6 presents the calculated energy of three methods [p-DMET, DMET (IB), and DMET (NIB)] and the corresponding extrapolations. All three methods give reasonable extrapolated energy values, compared to the benchmark data (gray shaded region). For p-DMET, the largest error is about $5 \times 10^{-3}$ at $U = 8$, while for DMET (IB), the largest error is about $3 \times 10^{-3}$ at $U = 4$. The behavior of DMET and p-DMET as a function of $L_A^{-2}$ is relatively similar, while that of DMET (NIB) is quite different. This reflects the energy influence of the different choice of the impurity Hamiltonian. In fact, extrapolations using DMET (NIB) have smaller error bars in general, and they are slightly more accurate than those of DMET (IB) and p-DMET. The results also indicate that after extrapolation, the performance of DMET and p-DMET is comparably reliable.

**D. Efficiency**

To demonstrate the efficiency of p-DMET and p-DMET-f, we analyze two factors that affect the overall computational cost, i.e., the total number of iterations for self-consistent convergence and the (average) time cost per iteration.

We first consider the number of iterations required for convergence. We extracted the number of iterations from the calculations for the preceding phase diagram (Fig. 3) and plot the distribution of the number of iterations in Fig. 7. As shown in Fig. 7, all three methods (p-DMET, p-DMET-f, and DMET) have a similar average convergence rate, with an average iteration number of 12 required to achieve an energy accuracy of $10^{-6}$. In most cases, the iteration number is less than 20. We also remark that at least for systems studied in this work, the distribution in the case of p-DMET is slightly narrower with fewer outliers, and hence, the self-consistent iteration is more stable.

Acquiring a similar average number of self-consistent iterations in p-DMET and DMET, we now discuss the total computational time. We performed a series of tests on lattices of size $2 \times N$ ($N$ is the number of sites in the $y$ direction ranging from 8 to 64). We set $U = 2.0$ and $n = 1.0$ (half filling). We chose this quasi-one-dimensional structure to ensure that the mean-field problem always has a positive energy gap, which is not the case for arbitrary 2D lattices. All the tests are performed on 36 Intel Broadwell vCPUs of the Google Cloud Platform (GCP).

We measured the CPU time of DMET, p-DMET, and p-DMET-f, spent on single-particle type computations to solve the impurity problems (many-body computations). When the fragment size is fixed, the cost to construct and solve the impurity problems (high-level computations) always scales linearly with respect to the global system size. In DMET, the single-particle type computations (low-level computations) include the diagonalization of the mean-field Hamiltonian and optimization of the correlation potential; in p-DMET, the cost of the single-particle computations is mainly due to the diagonalization to obtain the projected 1-RDM.

**FIG. 7.** Distribution of the number of self-consistent iterations of p-DMET, p-DMET-f, and DMET required to reach $10^{-6}$ in the energy across the phase diagram in Fig. 3. All calculations used DIII.
In DMET, the single-particle computational cost significantly increases with $N$ (Fig. 8). In each step of DMET, the correlation potential optimization typically requires more than 100 iterations to converge, and the number of iterations also increases with respect to the system size. When the system becomes moderately large (number of sites larger than 64), the cost at the mean-field level is much more expensive than solving the $2 \times 2$ impurity problems. When the number of sites is 128, the single particle computations in DMET take in total $\sim 20,000$ s. On the other hand, the main single-particle cost in p-DMET is only the eigenvalue decomposition to obtain $D^L$. Similarly, p-DMET-f only needs to perform single-particle type matrix multiplications and inversions. For the system with 128 sites, the computational cost at the mean-field level of p-DMET is reduced to only $\sim 1$ s. This demonstrates that p-DMET/p-DMET-f provides a significant reduction in computational cost relative to DMET for large, heterogeneous systems.

V. CONCLUSION

An important computational bottleneck in DMET calculations for large systems is the correlation potential optimization required to achieve self-consistency. In this optimization, each evaluation of the cost function requires the diagonalization of a mean-field Hamiltonian and each derivative evaluation amounts to a response calculation. Thus, for a moderately sized system, the correlation potential optimization procedure can be expensive. In this work, we viewed the self-consistent DMET as the only one formulation of a more general fixed point problem to obtain the high-level global density matrix $D^{hl}$. From this general perspective, we proposed the projected density matrix embedding method (p-DMET) as a simpler and more efficient way to achieve self-consistency. We found that for the 2D Hubbard model, compared to the unrestricted Hartree-Fock (UHF) solution, p-DMET significantly improved the accuracy of the total energy and behavior of the spin polarization across the entire parameter space. The phase diagrams predicted by p-DMET and DMET qualitatively agreed with each other, but the phase boundary obtained from p-DMET was softer. Further investigation showed that this was because the self-consistent solution of p-DMET retained a weak dependence on the initial guess. On the other hand, the cost associated with achieving self-consistency in p-DMET was negligible compared to that needed to optimize the correlation potential in DMET.

There are a number of directions that should now be pursued. First, we would like to identify the root of the initial guess dependence of p-DMET. Second, we plan to generalize p-DMET to superconducting systems, where the low-level theory requires the solution of the Bogoliubov-de Gennes (BdG) equations. Third, a remaining numerical issue in DMET associated with self-consistency is the appearance of vanishing gaps in the DMET mean-field Hamiltonian during the optimization. This should then be treated using a zero temperature limit of a finite temperature formulation of DMET. We will report these works in future publications.

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APPENDIX: OBTAINING BATH ORBITALS

Let $D^R$ be a mean-field density matrix. Without loss of generality, let us introduce a rotation matrix $R$, for each fragment $x$ such that the rotated density matrix $D = R_x D^R R_x^\dagger$ can be written as

$$D = CC^\dagger = \begin{bmatrix} D_{11} & D_{12} \\ D_{21} & D_{22} \end{bmatrix}.$$  \hfill (A1)

where $D_{11}$ corresponds to the fragment $x$ only. We divide the matrix $C$, which contains the occupied orbitals, into two matrices,

$$C = \begin{bmatrix} C_A \\ C_B \end{bmatrix}.$$  \hfill (A2)

$C_A$ is the rectangular matrix $(L_A \times N_f)$ with rows in the fragment (not fragment orbitals), and $C_B$ is the rectangular matrix $[(L - L_A) \times N_f]$ with rows in the environment (not bath orbitals). By diagonalizing the matrix $D_{11}$, we have

$$D_{11} = U_A \Sigma_A U_A^\dagger,$$  \hfill (A3)

where $\Sigma_A$ is a diagonal matrix containing all $L_A$ singular values of $C_A$. $U_A$ is a $L_A \times L_A$ unitary matrix. We can simply choose $V_A = I_{L_A}$ such that the above diagonalization corresponds to the singular value decomposition (SVD) on $C_A$,

$$C_A = U_A \Sigma_A V_A^\dagger.$$  \hfill (A4)

By the orthogonality of the orbitals, $C_B$ can be uniquely expressed as

$$C_B = U_B \Sigma_B V_A^\dagger,$$  \hfill (A5)

where $U_B$ is to be determined. Thus, the bath-fragment density matrix can be written as

$$D_{21} = C_B C_A^\dagger = U_B \Sigma_B \Sigma_A U_A^\dagger.$$  \hfill (A6)

The unitary matrix $U_B$ can be calculated by normalizing all the columns of the matrix $D_{21} U_A$, since

$$U_B \Sigma_B \Sigma_A = D_{21} U_A.$$  

The diagonal elements of $\Sigma_A \Sigma_B$ are the corresponding norms of the columns. As a result, $\Sigma_B$ is also obtained with the known $\Sigma_A$ in (A3). Thus, we have the fragment orbitals and bath orbitals, respectively,

$$C_f = \begin{bmatrix} U_A \\ 0 \end{bmatrix} \quad C_B = \begin{bmatrix} 0 \\ U_B \end{bmatrix}.$$  \hfill (A7)
The fragment orbitals $C_f$ and bath orbitals $C_b$ are then combined into the DMET impurity problem orbitals,

$$C_A = [C_f C_b] .$$

(A8)

Since the core orbitals do not explicitly appear in the DMET formulation, they do not have to be computed explicitly. The contribution of the core orbitals is reflected by the core density matrix. The core density matrix is expressed as

$$D_{\text{core}} = D_{22} - U_B \Sigma_B^* U_B^\dagger.$$  

(A9)

The above implementation is equivalent to the diagonalization of the density matrix $D_{22}$. In fact,

$$C_b C_b^\dagger = \begin{bmatrix} I_{L_b} - \Sigma_B^\dagger & 0 \\ 0 & I_{N_b - L_b} \end{bmatrix} \begin{bmatrix} U_B^\dagger \\ U_b \end{bmatrix} .$$

(A10)

During the above construction of the bath orbitals, we only need to perform an SVD on the matrix $C_b C_b^\dagger$ instead of diagonalizing the matrix $C_b C_b^\dagger$ or performing SVD on $C_b^\dagger C_b$. This process reduces the computational cost of bath orbital construction from $O(N_b^3)$ or $O(N_b L_b^2)$ to $O(L_b^3)$ when $N_b \gg L_b$.

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