

Reducing Qubit Requirements while Maintaining Numerical Precision for the Variational Quantum Eigensolver: A Basis-Set-Free Approach

Jakob S. Kottmann,* Philipp Schleich, Teresa Tamayo-Mendoza, and Alán Aspuru-Guzik*



Cite This: *J. Phys. Chem. Lett.* 2021, 12, 663–673



Read Online

ACCESS |

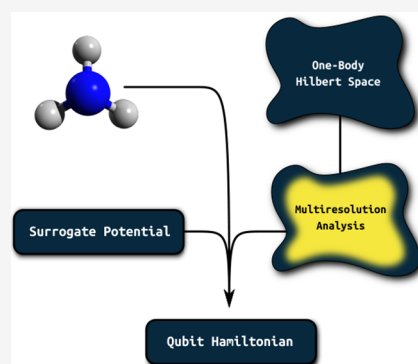


Metrics & More



Article Recommendations

ABSTRACT: We present a basis-set-free approach to the variational quantum eigensolver using an adaptive representation of the spatial part of molecular wave functions. Our approach directly determines system-specific representations of qubit Hamiltonians while fully omitting globally defined basis sets. In this work, we use directly determined pair-natural orbitals on the level of second-order perturbation theory. This results in compact qubit Hamiltonians with high numerical accuracy. We demonstrate initial applications with compact Hamiltonians on up to 22 qubits where conventional representation would for the same systems require 40–100 or more qubits. We further demonstrate reductions in the quantum circuits through the structure of the pair-natural orbitals.



Within the framework of the Born–Oppenheimer approximation, the electronic structure of molecules is described as a multidimensional wave function of electrons in an external potential, usually generated by the charges of the nuclear framework. Finding accurate approximations to describe those multidimensional wave functions is one of the key goals in quantum chemistry, and several types of models have emerged over the last decades.¹ Most models formally decompose the multidimensional electronic wave function into a linear combination of antisymmetrized tensor products (Slater determinants) of one-electron wave functions (spin-orbitals). The spin component can be described completely by the two spin-up and spin-down basis states, leaving only the spatial part of the orbitals to be represented. Traditional electronic structure packages usually use global sets of atom-centered basis functions to describe the spatial parts of the orbitals. Those basis functions mimic the solutions of the hydrogen atom, and the whole procedure is termed a linear combination of atomic orbitals (LCAO). The by far most prominent choice are atom-centered Gaussian functions multiplied by polynomial factors.^{2,3} The exponents of the Gaussian functions are globally defined for each atom, which leads to a large number of individual basis sets (the EMSL basis set exchange library⁴ lists currently 429 different basis sets for the carbon atom alone), making it a nontrivial task to pick the right basis set for the right computation. Alternatives to Gaussian basis sets, such as exponential functions⁵ (Slater-type basis sets) and Sturmians,⁶ exist and are a topic of ongoing research but still rely on globally defined basis sets.

Basis-set-free approaches represent the spatial part of the orbitals or other wave functions without the use of a globally defined basis set. Some examples are approaches based on Daubechies wavelets,^{7–9} Lagrange-sinc functions,¹⁰ or multi-resolution analysis (MRA).^{11–14} MRA offers an alternative to the traditional basis sets by representing the spatial parts of molecular wave functions on adaptive real-space grids, where wavelet-based numerical techniques allow adaptive refinement of the grid in a black-box fashion. In this representation, each function (orbital) is described individually by automatically constructed adaptive multiwavelets, making it a basis-set-free representation, because the numerical basis is individually constructed from a proper L^2 basis with a numerically well-defined truncation criterion. MRA allowed the development of highly accurate quantum chemistry algorithms for ground-state energies,^{11,15} excitation energies,^{16,17} polarizabilities,^{18,19} magnetic properties,^{20,21} as well as relativistic applications^{22,23} using mean-field and density functional theory (DFT) models. Initial treatment of correlated methods beyond density functional theory aimed at representing multielectron wave functions directly, resulting in basis-set-free and virtual-orbital-

Received: November 16, 2020

Accepted: December 21, 2020

Published: January 4, 2021



free approaches.^{24–27} Recently, an approach to directly determine MRA-represented pair-natural orbitals (MRA-PNOs) on the level of Møller–Plesset perturbation theory of second order (MP2) was demonstrated. This approach allows one to grow near-optimal system-adapted PNOs from scratch, omitting the use of global basis sets completely. In this work, we will apply those MRA-PNOs in a more general framework beyond the MP2 model, similar to PNO-based methods within LCAO approaches.^{28–31}

Classical quantum chemistry algorithms are highly optimized toward Gaussian basis sets (GBS), which makes them the dominant choice of representation. Quantum algorithms have not reached this highly optimized stage, and current research is exploring alternative representations, for example, plane-waves³² or Gausslets,³³ which offer advantages over Gaussian basis sets within the context of quantum computation. In this work, we introduce a basis-set-free approach to the variational quantum eigensolver (VQE),^{34,35} a class of algorithms that variationally minimize the expectation value of a qubit Hamiltonian using a parametrized quantum circuit. The basis-set-free qubit Hamiltonians are constructed from directly determined occupied Hartree–Fock orbitals and MRA-PNOs.³⁶ In contrast to global basis sets, the orbitals are optimized system-specific with a surrogate model (MP2), which already accounts for electron correlation. This allows for the construction of compact qubit Hamiltonians, which require a significantly lower number of qubits compared to their LCAO-based counterparts. In variational quantum algorithms, the reduction of the qubit resources has become one of the main objectives, particularly for applications in chemistry. Recent approaches include external corrections, such as the (virtual) quantum subspace expansion³⁷ or explicitly correlated approaches in the form of trans-correlated Hamiltonians,^{38,39} and often come with additional costs in quantum measurements and classical computation. These methods can as well be applied within the framework of this work and potentially lower the qubit resources even further. Here, we develop a way to directly construct system-specific compact qubit Hamiltonians with high numerical precision providing a path toward high-accuracy quantum chemistry with variational quantum algorithms.

When solving the electronic structure problem, one aims to find approximations for the eigenenergies of electronic Hamiltonians, which for N_e electrons with coordinates $\vec{r}_k \in \mathbb{R}^3$ are defined as

$$H(\vec{r}_1, \dots, \vec{r}_{N_e}) = \sum_{k=1}^{N_e} h(\vec{r}_k) + \frac{1}{2} \sum_{k \neq l}^{N_e} g(\vec{r}_k, \vec{r}_l) \quad (1)$$

where $h = T + V$ denotes the one-electron kinetic energy operator T with the external potential V , and $g(\vec{r}_k, \vec{r}_l)$ denotes the electron–electron Coulomb potential. For molecules, the external potential $V(\vec{r})$ is given by the sum of Coulomb potentials between the individual point charges of the nuclei and an electron at position \vec{r} . The eigenfunctions of the electronic Hamiltonian are antisymmetric, multidimensional functions in \mathbb{R}^{3N_e} , making brute-force grid-based computation an unfeasible task. The requirement of having an antisymmetrized wave function arises from the fermionic nature of the electrons and is usually handled by using Slater determinants or anticommuting second-quantized operators.

In order to tackle this challenge, a large family of approximations to the electronic wave function have been

introduced. The most prominent is the Hartree–Fock (HF) method, which variationally optimizes a single Slater determinant, therefore reducing the $3N_e$ -dimensional problem to N_e coupled, three-dimensional, nonlinear problems. Improvements upon Hartree–Fock, like configuration–interaction (CI) and coupled-cluster (CC) methods, add more antisymmetrized functions to the wave function ansatz, usually created by replacing n orbitals in the initial determinant by correlated n -electron functions. In conventional quantum chemistry, the spatial parts of the Hartree–Fock orbitals are represented with globally defined fixed LCAO basis sets for each individual atom. The Hartree–Fock algorithm is then simplified to optimize only the LCAO coefficients, resulting in the Roothaan equations, which often are still referred to as Hartree–Fock. An N -orbital basis set results in $2N$ -orthonormal spin–orbitals, of which the first N_e define the Hartree–Fock reference determinant. This leaves $2N - N_e$ virtual spin–orbitals free to represent the correlated electron functions used in CI and CC methods.²⁶

Within the language of second quantization, the description of these methods can be significantly simplified,^{3,40–42} by expressing the electronic Hamiltonian with abstract field operators⁴³ $\hat{\psi}^\dagger(x)$, $\hat{\psi}(x)$, which create or annihilate electron density at spin-coordinate $x = (\vec{r}, \sigma)$

$$H = \int dx \hat{\psi}^\dagger(x) h(x) \hat{\psi}(x) + \frac{1}{2} \int dx dy \hat{\psi}^\dagger(x) \hat{\psi}^\dagger(y) g(x, y) \hat{\psi}(y) \hat{\psi}(x) \quad (2)$$

Formal expansion of the field operators into an orthogonal set of spin–orbitals as $\hat{\psi}(x) = \sum_k \varphi_k(x) \hat{a}_k$ leads to the numerically more suitable form

$$H = \sum_{kl} h_{kl} \hat{a}_k^\dagger \hat{a}_l + \frac{1}{2} \sum_{klmn} g_{klmn} \hat{a}_k^\dagger \hat{a}_l^\dagger \hat{a}_n \hat{a}_m \quad (3)$$

where h_{kl} and g_{klmn} are integrals over spin–orbitals in Dirac notation. Within the scope of variational quantum algorithms, the second-quantized Hamiltonian can then be transformed to a qubit Hamiltonian using various encodings.^{44–46} Canonically, when global basis sets are used, the second-quantized Hamiltonian in eq 3 is constructed by the occupied and virtual Hartree–Fock orbitals. This can be interpreted as preoptimizing the orbitals within a fixed set of basis functions by a mean-field method. In this work, we construct the second-quantized Hamiltonian from the occupied Hartree–Fock orbitals, solved variationally within a multiresolution analysis representation,¹¹ combined with directly determined pair-natural orbitals,³⁶ optimized by MP2. In other words, we are preoptimizing the orbitals with a correlated method and within a basis-set-free adaptive representation. The orbitals represented by MRA are basis-set-free because they do not require a globally defined fixed set of basis functions but are instead individually represented by adaptively constructed sets of piecewise polynomials.^{11–13} The construction of the multiresolution representation on a grid is carried out automatically, and for most applications, the user does not require detailed knowledge about the underlying machinery. Hence, from a user perspective, this representation can be treated as *effectively basis-free* and is sometimes referred to as just *basis-free*. To build a Hamiltonian, it is necessary only to set a numerical accuracy threshold (in this work, this is $\epsilon = 10^{-4}$) and the required number of orbitals. In Figure 1, we illustrate the

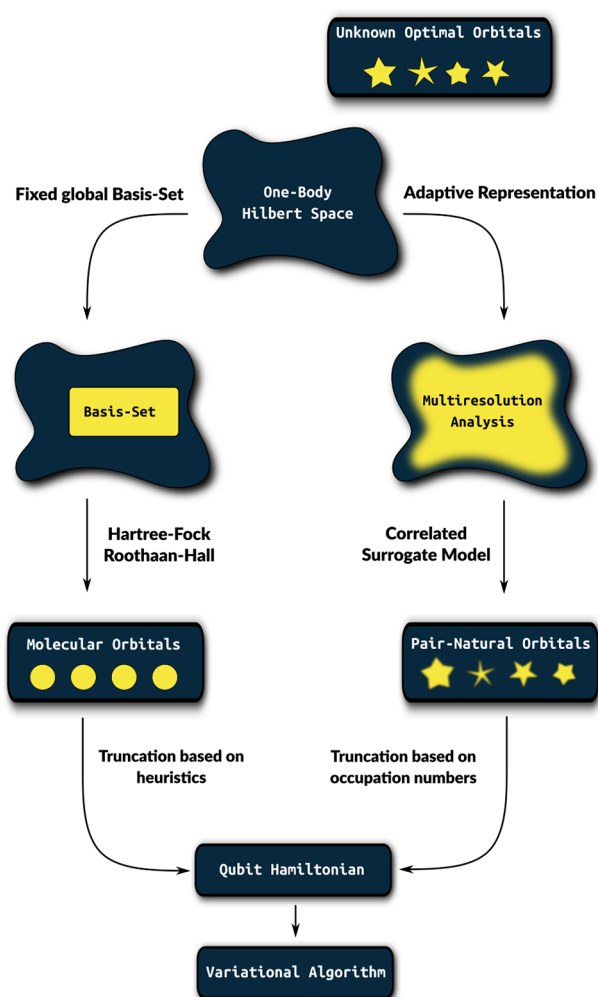


Figure 1. Constructing molecular qubit Hamiltonians: (Left) Representation of the spatial part of molecular orbitals by a fixed set of atom-centered Gaussian functions. This is a global basis set, where functions for each atom are globally defined throughout all possible molecules. (Right) The spatial part of the molecules is represented with multiresolution analysis (MRA) resulting in a locally adaptive representation. Pair-natural orbitals (PNOs) are directly determined and optimized within the MRA representation by a surrogate model (in this work MP2). Truncation of the representation to the available qubit number is naturally given by the occupation numbers of the pair-natural orbitals.

construction of the qubit Hamiltonians using the basis set and the MRA-PNO-based basis-set-free approach. Conceptually, the biggest difference to the canonical construction is that our approach does not rely on globally fixed sets, but rather optimizes the orbitals directly and system-specific. This allows the freedom to adapt to the molecule at hand, in order to find a close-to-optimal compact representation. In general, the underlying representation of the pair-natural orbitals can be chosen freely and does not necessarily need to be MRA. If a sufficiently large Gaussian basis set is chosen, the PNO representation using this basis set will lead to similar qubit Hamiltonians as obtained with MRA-PNOs. In other words, once computed, the large qubit Hamiltonians of Table 1 could be reduced to similar sizes as the qubit Hamiltonians constructed from MRA-PNOs, similar to ref 47, where such an approach is used with intrinsic atomic orbitals instead of pair-natural orbitals. There is, however, no guarantee that the

Table 1. Qubit Requirements of MRA and GBS Representations^a

system	metric	MRA	GBS	more
He	MAX	4	4–10	Figure 2
Be	MAX	10	10–18	Figure 2
H ₂	NPE	4	20–56	Figures 5 and 3
H ₂	NPE	8	20–56	Figures 5 and 3
H ₂	NPE	20	56–120	Figures 5 and 3
H ₂	MAX	4	8	Figures 5 and 3
H ₂	MAX	8	20–56	Figures 5 and 3
H ₂	MAX	20	56	Figures 5 and 3
LiH	NPE	12–22	38–88	Figures 5 and 3
LiH	MAX	12	38–88	Figures 5 and 3
LiH	MAX	22	170–288	Figures 5 and 3
BH	NPE	12–22	38–88	Figures 5 and 3
BH	MAX	12–22	38–88	Figures 5 and 3
BeH ₂	NPE	12	46–114	Figures 5 and 3
BeH ₂	MAX	12	24–46	Figures 5 and 3
NH ₃	ΔE	12–18	58–100	Figure 6

^aQubit requirements for the MRA Hamiltonians used in this work compared to qubit requirements using standard Dunning-type basis sets (GBS) that achieve comparable accuracy within different metrics. The employed metrics are non-parallelity (NPE) and maximum (MAX) errors on potential energy surfaces and a reaction barrier (ΔE).

chosen Gaussian basis set is sufficient, and picking the right basis set for the right task depends widely on heuristics and trial and error procedures. We chose MRA-PNOs, because they are usually close to optimally represented pair-natural orbitals, allowing us to focus on the surrogate model itself, without having to speculate about basis set effects. It was shown before that the direct construction of pair-natural orbitals in this way can reach accuracies beyond existing Gaussian basis sets within second-order methods.³⁶ Current implementations can run with molecular systems consisting of dozens of atoms, making it a realistic candidate for future large-scale applications.

Most MRA-based optimization protocols in quantum chemistry solve a quantum chemical model, defined by the potential V , by transforming the Schrödinger-like differential equation

$$\left(-\frac{\Delta}{2} + V\right)|\Psi\rangle = E|\Psi\rangle \quad (4)$$

into an integral equation using the bound-state Helmholtz Green's function G_E as kernel

$$\Psi(r) = -2 \int dr' G_E(r, r') V(r') \Psi(r') \quad (5)$$

leading to an iterative optimization of the wave function.^{48,49} The potential V depends on the underlying quantum chemical model and is usually an effective one-body potential from Hartree–Fock and Kohn–Sham theory¹¹ with corresponding excited-state variants,^{16,17} or an effective two-body potential from MP2^{24,25} and coupled-cluster approaches.^{26,27} While more general many-body potentials are possible in theory,⁵⁰ they are computationally challenging and have not been realized yet. In this work, the potential V is the PNO-MP2 Hylleraas potential, an effective one-body potential described in detail in ref 36. We refer to a recent review¹³ for further details on MRA-based methods.

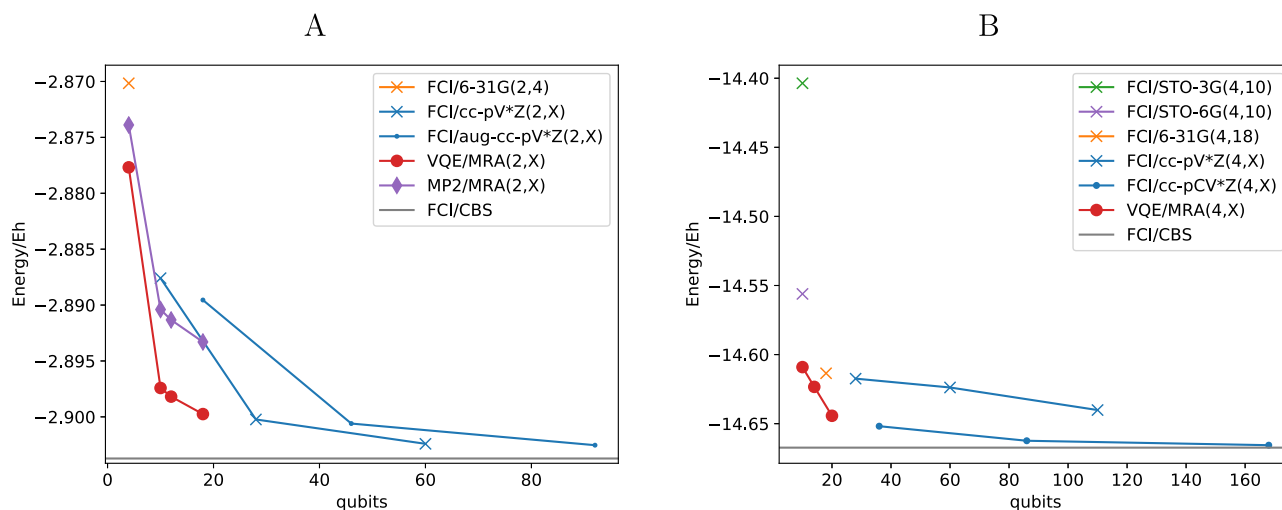


Figure 2. Helium (A) and beryllium (B) atoms: Classical full-CI energies compared with basis-set-free VQE energies for the helium (A) and beryllium (B) atoms. The VQE part uses the UpCCGSD ansatz. UpCCGSD results with small Gaussian basis sets are nearly indistinguishable from the FCI results and are omitted in the plots. The reference values FCI/CBS are -2.9037^{64-66} and -14.667^{65} in Hartree units.

In this approach, occupied and localized Hartree–Fock orbitals are optimized according to ref 11, and initial pair-specific guess functions for the PNOs are created by multiplying monomials onto the optimized Hartree–Fock orbitals. These initial PNOs are optimized according to eq 5, where V is determined by the PNO-MP2 Hylleraas functional.³⁶ The optimization results in multiple sets of pair-natural orbitals corresponding to pairs of occupied Hartree–Fock orbitals $|i\rangle$ and $|j\rangle$

$$S_{ij} = \bigcup_{a_{ij}=1}^{r_{ij}} \{|a_{ij}\rangle\} \quad (6)$$

where the ranks r_{ij} depend on the occupation numbers of the pair-natural orbitals, the truncation threshold, and the number of available initial guess functions. In this work, we used two cycles of guess function construction according to ref 36 and set the PNO ranks according to the desired number of qubits. In order to construct the qubit Hamiltonians, we first globally select the MRA-PNOs $|a_{ij}\rangle$ with the largest occupation numbers and orthonormalize them via Cholesky decomposition. Other orthogonalization methods, like the symmetric Loewdin approach, are also supported. However, we found that Cholesky behaves better within unitaries that exploit the PNO structure (see later sections), because it preserves the important pair-natural orbitals (ordered by their MP2 occupation number) better. Combining the occupied Hartree–Fock orbitals $|i\rangle$ with the globally orthonormalized pair-natural orbitals $|\tilde{a}_{ij}\rangle$, we obtain an orthonormal set of orbitals

$$S = S_{\text{HF}} \cup S_{\text{PNO}} \quad (7)$$

$$S_{\text{PNO}} = \bigcup_{i \leq j=1}^{N_e/2} \tilde{S}_{ij} \quad (8)$$

where S_{HF} denotes the occupied Hartree–Fock orbitals and \tilde{S}_{ij} the selected and orthonormalized PNOs from eq 6. Because of the orthogonalization, the orbitals in \tilde{S}_{ij} will differ from the original PNOs of the surrogate model. Finally, the second-

quantized Hamiltonian is constructed from the orbitals in S . This Hamiltonian can be transformed to a qubit Hamiltonian using standard qubit encodings.

The computational cost of the MRA-PNO-MP2 surrogate model formally scales as $O\left(\left(\frac{N_e}{2}\right)^3 R^2\right)$, where N_e is the number

of electrons and R is bounded from above by the maximal PNO rank $R_{\text{max}} = \max_{ij} r_{ij}$ and, in practice, behaves like the average rank of the PNO pairs.³⁶ Within this approach, the maximal ranks are fixed and determined by the number of desired qubits N_q in the qubit Hamiltonian but are in general not expected to grow asymptotically with system size.^{31,51,52} The formal scaling of the MRA-PNO-MP2 surrogate can then be anticipated to be cubic with system size, which is the same as the natural scaling of MP2 in its real-space formulation.²⁶ The orthonormalization of the PNOs using Cholesky or other techniques can be performed directly in the MRA representation, requiring only the computation of the PNO overlaps, $O\left(\left(\frac{N_q}{2}\right)^2\right)$, already required within the MRA-PNO-MP2 calculation, and a transformation of the orbitals, $O\left(\left(\frac{N_q}{2}\right)^3\right)$, afterwards. The Hartree–Fock calculation within MRA formally scales quadratically with system size and has the potential to be further reduced.^{13,53}

Because the qubit Hamiltonians of this work are constructed by occupied Hartree–Fock orbitals and pair-natural orbitals given in eq 6, this additional structure can be exploited in the construction of the quantum circuit. The PNO-MP2 surrogate model itself contains only double excitations from occupied Hartree–Fock orbitals $|i\rangle$, $|j\rangle$ to the pair-specific PNOs $|a_{ij}\rangle$, and this excitation structure can be employed to construct unitary quantum circuits. In the case of pair-excitation models, such as the k -UpCCGSD model, we can define the PNO-restricted doubles and generalized doubles models as

$$| \text{PNO-UpCCD} \rangle = U_{\tilde{D}} U_{\text{HF}} | 0 \rangle \quad (9)$$

$$| \text{PNO-UpCCGD} \rangle = U_{\tilde{G}\tilde{D}} U_{\tilde{D}} U_{\text{HF}} | 0 \rangle \quad (10)$$

using the PNO-restricted unitary operators

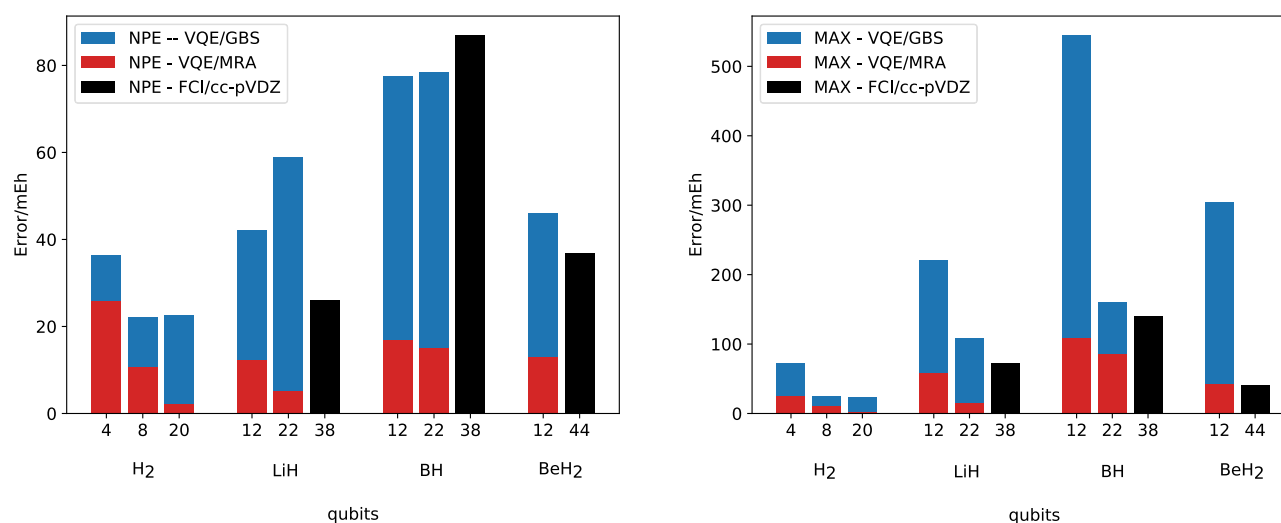


Figure 3. Nonparallelity (NPE) and maximum errors (MAX): Comparison of NPE and MAX errors of the potential energy surfaces of Figure 5 using standard Gaussian basis sets (GBS) or the basis-set-free VQE approach (MRA). The used reference values are FCI/cc-pVQZ(2,120) for H₂, FCI/cc-pCVQZ(4,228) for LiH, CISDTQ/cc-pCVTZ(6,114) for BH, and FCI/cc-pVQZ(4,224) for BeH₂. The GBS for the VQE are STO-3G, 6-31G, and cc-pVDZ.

$$U_{\tilde{D}} = \prod_{i=1}^{N_e/2} \prod_{a \in \tilde{S}_i} e^{-i\theta/2\tilde{G}_{iaia}} \quad (11)$$

$$U_{G\tilde{D}} = \prod_{i=1}^{N_e/2} \prod_{a,b \in \tilde{S}_i} e^{-i\theta/2\tilde{G}_{abab}} \quad (12)$$

and the pair excitation generator defined as

$$\tilde{G}_{iaia} = i(a_{a_i}^\dagger a_{i_i}^\dagger a_i a_{i_i} - h.c.) \quad (13)$$

The incorporation of (generalized) singles can be realized in the same way, with the single excitation generators corresponding to the used double-excitation generators. The PNO-UpCCD wave function has a product structure of electron-pair wave functions that themselves are supported by Hartree–Fock orbitals $|i\rangle$ and the corresponding PNOs $|a_{ii}\rangle$. The doubles-only variants (PNO-UpCCD and PNO-UpCCGD) of these low-depth approaches offer additional advantages. Because of the restriction to pair-excitations resulting from the same spatial orbitals, these wave functions allow direct encoding^{54,55} of spatial orbitals into qubits, reducing the qubit requirements by a factor of 2. For example, for the LiH molecule, this is possible without notable loss of accuracy (see Figure 5). In addition, the model depends only on diagonal pairs $|a_{ii}\rangle$, lowering the computational demands of the employed surrogate model. Detailed strategies in order to exploit the product structure of the wave function in quantum and classical simulations are currently investigated. Note that those additional reductions are not denoted in Table 1.

To evaluate the accuracy and efficiency of the proposed approach, we employ MRA-PNO-MP2 as a basis-set-free surrogate model to the variational quantum eigensolver. We employed the UpCCGSD model of ref 56 to construct the quantum circuits (Figure 2). All used model systems are chosen such that they are well-described by this ansatz, which allows us to focus on the numerical accuracy of the qubit Hamiltonians without worrying about the quality of the ansatz. An overview of the qubit requirements using MRA and GBS representations is given in Table 1, where we report significant

improvement for all systems and types of energy metrics investigated in this work. As energy metrics we used nonparallelity (NPE) and maximum (MAX) errors and a reaction barrier. Nonparallelity errors⁵⁶ are defined as the difference between the maximal and minimal error on a given potential energy surface (Figure 3). Note that, other than in ref 56, the reference values are here also chosen with respect to the underlying one-particle basis. The MRA-PNOs are optimized according to ref 36 using MADNESS.¹² Note that in this work the MRA-PNOs were optimized without regularizing the Coulomb singularity.

VQE calculations are performed with TEQUILA⁵⁷ using QULACS⁵⁸ as quantum backend, the BFGS optimizer of SCIPY,⁵⁹ and the qubit encodings of OPENFERMION.⁶⁰ Analytical gradients for the BFGS optimization were obtained automatically through the techniques described in ref 61. LCAO reference calculations are performed with psi4.⁶² In all VQE calculations the parameters are initialized as zero, that is, starting from the Hartree–Fock reference state. Representations of Hamiltonians are abbreviated with MRA(N_e , N_q) for MRA-PNOs and the acronym for standard LCAO basis sets. The values in parentheses represent the number of electrons N_e and qubits (spin–orbitals) N_q . Note that classical FCI calculations with basis sets corresponding to large qubit Hamiltonians with 50 or more qubits are possible, because these algorithms are not operating in the full Fock space of the second quantized Hamiltonian. For simplicity, we omitted known general compression schemes that allow reducing the number of qubits by two when combined with parity-based encodings,⁶³ because these would apply to all qubit Hamiltonians in this work in the same way. The numerical accuracy of the qubit Hamiltonian is independent of the encoding, and the results of this work were obtained with the Jordan–Wigner representation; our implementation within TEQUILA, however, does support other encodings. The qubit encoding can influence the results of possible future demonstrations on real quantum hardware, because it will result in different gate decomposition of the VQE unitary and therefore will have varying properties with regard to the specifics of the device noise characteristics.

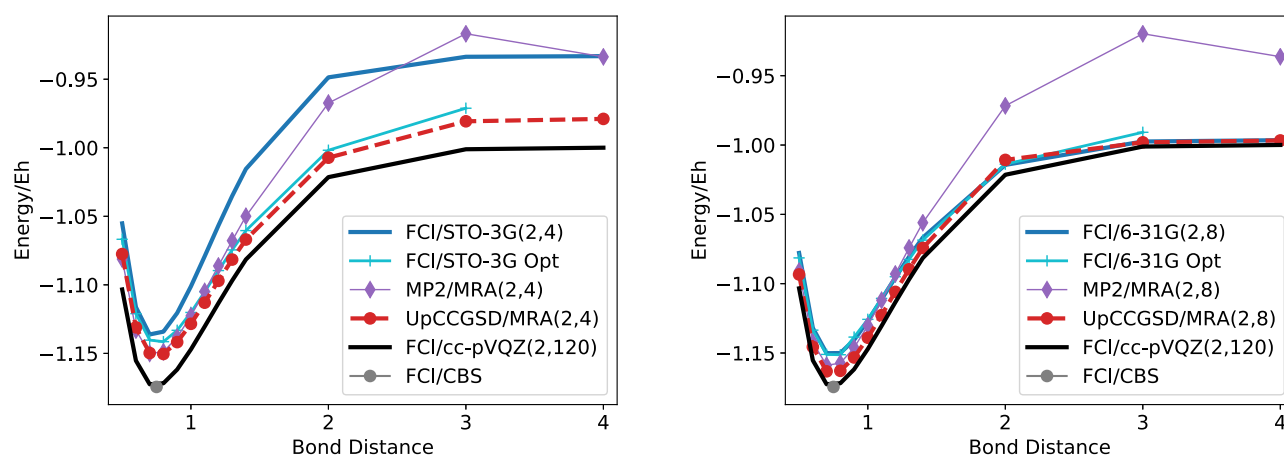


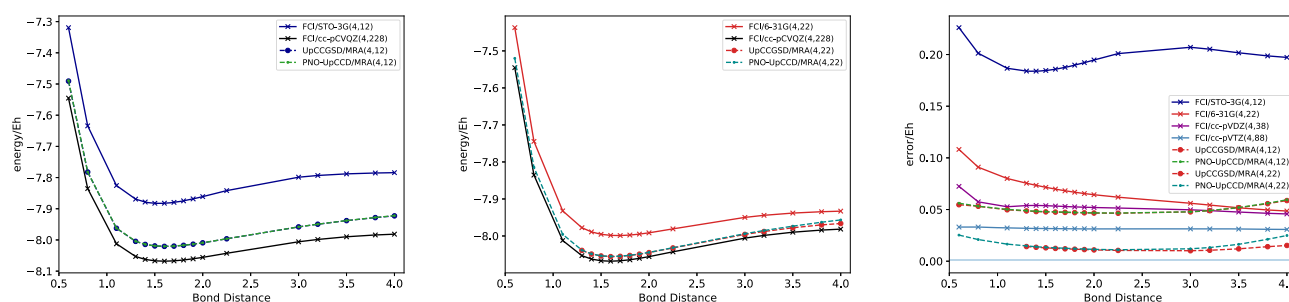
Figure 4. Dissociation curve of the hydrogen molecule. Energies are computed with a VQE (UpCCGSD) or classical methods (FCI). The Hamiltonians were constructed with basis-set-free methods (MRA) and Gaussian basis sets (GBS). All curves are denoted as method/basis(N_e , N_q) with number of electrons N_e and number of qubits N_q . Blue curves show GBS results with minimal (STO-3G) and slightly larger (6-31G) basis sets, while red curves show the UpCCGSD/MRA results with the same qubit requirements. Black curves show the best GBS results (the same in both panels). The two H_2 /GBS-Opt curves show results for the individually optimized Gaussian basis sets (exponents, contraction coefficients, centers) according to ref 70. The MP2/MRA(N_e , N_q) curves show the result of the classical surrogate model (PNO-MP2). The same orbitals are used for UpCCGSD/MRA(N_e , N_q), illustrating the energy differences between the surrogate model and the VQE. FCI/CBS points were taken from refs 65 and 71. Bond distances are given in Ångström.

As an initial test, we computed the energies of the helium and beryllium atoms, where the true energies close to the complete basis set (CBS) limit are known from different highly accurate numerical calculations in the literature.^{64–66} The results are shown in Figure 2, where we show the absolute energies with respect to the qubit requirements. MRA-MP2 results are shown for the helium atom to illustrate that the problem is not fully solved by the surrogate model. For atoms, LCAO basis sets are expected to perform well, and the basis-set-free Hartree–Fock orbitals and PNOs have the same shape as atomic orbitals. One of the main differences is the missing *nuclear cusp*,^{67,68} present in all molecular wave functions because of the singularities in the nuclear Coulomb potential, and nonrepresentable by primitive Gaussian functions. Gaussian basis sets mitigate this through contracted Gaussian basis functions. The STO-3G basis set for example uses three Gaussians to represent one contracted basis function while the STO-6G uses six, explaining the significantly better performance of STO-6G compared to STO-3G in Figure 2. The other basis sets are expected to represent the nuclear cusp well. Their slow convergence is a known phenomenon resulting from the *electronic cusp*^{67,69} in the wave functions, generated by the singularities in the electron–electron potential and, in general, hard to represent in a separated representation. Therefore, this is also the case for the MRA-PNOs in this work, because the many-electron wave function is still represented as a separated representation using one-electron wave functions. For the helium atom, a decrease in the slope of the convergence can already be observed, and we expect the same for the Beryllium atom with increasing number of qubits. The basis-set-free approach however achieves significantly more accurate results compared to the LCAO representations. In future work, the representation could be further improved by explicitly correlated approaches that correct the inaccurate description of the electronic cusp.⁶⁶ These techniques can already be applied in the MRA-PNO optimization,³⁶ and recently such methods have been introduced for VQEs in the form of transcorrelated Hamiltonians.^{38,39}

We simulated VQE energies, employing UpCCGSD, and the pair-specific PNO-UpCCD model, along the potential energy surfaces of the small molecules H_2 , LiH, BH, and BeH_2 and compared it with the best affordable variational methods using the largest Gaussian basis sets affordable. The potential energy surfaces are shown in Figures 4 and 5. Similar as for the helium atom in Figure 2, we show the PNO-MP2/MRA results in Figure 4 to illustrate that the problem is not fully solved by the surrogate model which performs poorly for the stretched geometries, and the improvements in the method are in this case within the same range as the improvements in numerical representation. For H_2 we furthermore show GBS calculations where the basis set was optimized for each point individually according to ref 70 using the DIFFICULT⁷² package. This can be seen as an intermediate approach using Hartree–Fock as a surrogate model and allowing the basis set to relax. In the (2,4) representation the optimal spatial orbitals can be well-approximated by two atom-centered s-type orbitals, leading to significant improvements in the optimized STO-3G representation if individual optimization is enabled. For the slightly larger 6-31G basis set, individual optimization does not lead to improvements. In this case, the optimal (4,8) spatial orbitals contain two π orbitals which intrinsically can not be represented by the 6-31G basis set but are well-approximated by MRA where the techniques of refs 16 and 36 ensure that the correct symmetries are present in the initial guess functions for the MRA-PNOs.

In order to accurately describe chemistry, obtaining consistent relative energies over different molecular structures is in most cases more important than accurate absolute energies. In Figure 3, we use nonparallelity (NPE) and maximum (MAX) errors as accuracy metrics with the best achievable method and Gaussian basis set as reference. We chose the three diatomic molecules here such that UpCCGSD is a good ansatz; that is, differences to FCI are below the millihartree threshold for STO-3G and 6-31G simulations. For the BeH_2 molecule, UpCCGSD did not always converge toward the best solution. We employed adaptive operator growth according to refs 73 and 61 in order to reach the lowest

LiH



BH

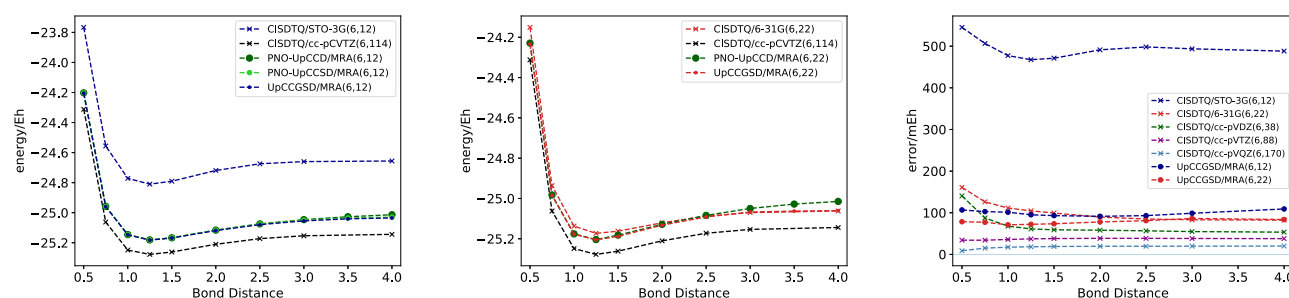
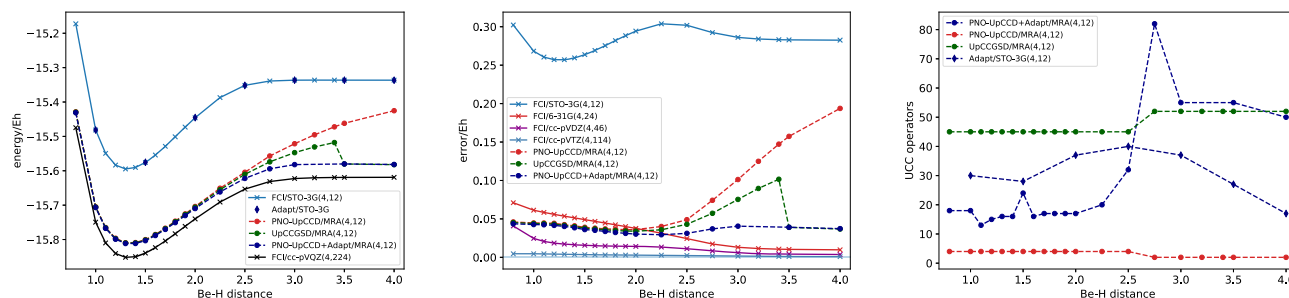

$$\text{BeH}_2$$


Figure 5. Bond dissociation curves. Total energies and errors with respect to the best affordable exact diagonalization (FCI) in the largest affordable Gaussian basis set (given in black). PNO-UpCC(S)D exploits the PNO structure of the surrogate model according to [eq 9](#).

energies of the given PNO Hamiltonians. As references, we used FCI/cc-pVQZ(2,120) for H_2 , FCI/cc-pCVQZ(2,288) for LiH , CISDTQ/cc-pCVTZ(6,114) for BH , and FCI/cc-pVQZ-(4,224) for BeH_2 . Comparison with accurate numerical results from refs 71 and 65 confirms that the reference values for H_2 and LiH are close to the basis set limit. In terms of NPE and MAX errors (Figure 3), the basis-set-free VQE clearly outperforms the traditional basis sets with the same number of qubits as well as the *cc-pVDZ* simulations, which use approximately twice the number of qubits. The error of the MRA-PNO representation tends to grow with larger bond distances. This result is not unexpected, because the MP2 surrogate model performs worst in this regime. The effect on the orbitals determined by the surrogate model are however not that severe. In contrast, the GBS representation performs best at large bond distances. This is also not surprising, because these basis sets are optimized for the atomic systems. The PNO-restricted unitaries perform well at not-too-far-stretched bond distances, where they offer significant savings in the number of operators used for the UCC ansatz (see Table 2). This ansatz can be viewed as an even further restricted form of

Table 2. Parameter and CNOT Counts^a

	A	B	C
LiH(4,12)	4 (192)	12 (352)	45 (1280)
BH(6,12)	3 (144)	9 (240)	45 (1280)
BeH ₂ (4,12)	4 (192)	12 (368)	45 (1280)
LiH(4,22)	9 (432)	27 (1216)	165 (6160)
BH(6,22)	7 (336)	21 (848)	165 (6160)

^aNumber of parameters (fermionic excitations) for PNO-UpCCD (A), PNO-UpCCSD (B), and UpCCGSD (C). See Figure 5 for the corresponding energies. Numbers in parentheses denote CNOT counts from a naive, non-optimized decomposition of the unitaries as a first estimate. More advanced circuit construction is anticipated to significantly reduce those counts.

the approach in refs 54 and 74, where all configurations in the UCC wave function are restricted to double occupancies. In contrast to ref 54, the PNO-UpCCD ansatz however performs well for all bond distances of LiH. The reason for this lies in the PNO structure, which is not present in canonical Hartree–

Fock orbitals. Note that this is independent of the underlying numerical representation of the orbitals.

As a last example, we simulate the umbrella reaction of ammonia, a small intramolecular reaction, where the umbrella-like molecular structure of ammonia is inverted passing a planar transition state. The activation barriers of this reaction were simulated using the basis-set-free VQE approach and different classical methods with large basis sets (cf. Figure 6).

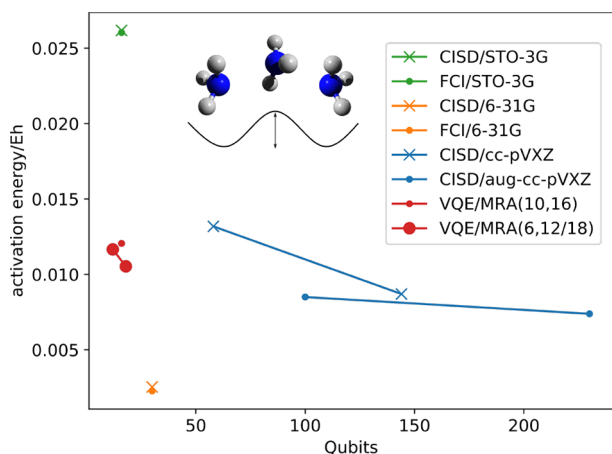


Figure 6. Umbrella reaction of ammonia. Energy barriers in relation to qubit requirements. Results are computed with different classical methods and the basis-set-free VQE (UpCCGSD ansatz).

For larger basis sets, exact diagonalization is already unfeasible here. CISD, however, provides an accurate model for this reaction type because possible size-inconsistency issues are negligible in this intramolecular reaction, witnessed by the accompanied FCI calculations performed for the smaller basis-sets. In addition to the full treatment of the 10 electrons of ammonia, we used a 6-electron active space, freezing the lowest occupied Hartree–Fock orbital and the orbital corresponding to the lone pair of ammonia. Simulations with the small basis sets STO-3G and 6-31G over- or underestimate the activation barrier by more than a factor of 2, while the basis-set-free approaches yield accurate energies. In this case, 12–18 qubits suffice for the basis-set-free approach to achieve a numerical accuracy, which would require 50–100 spin-orbitals with traditional basis sets.

In this work, we developed a basis-set-free approach by using MRA-PNO-MP2 as a surrogate model to construct compact, system-adapted qubit Hamiltonians with high numerical precision. Initial demonstrations for absolute energies of atoms, small potential energy surfaces, and a toy model for a chemical reaction show a clear advantage of this approach compared to standard basis sets throughout all used metrics, allowing the achievement of high numerical precision in the spatial representation of the wave function with significantly reduced number of qubits (see Table 1). Our approach is furthermore a promising candidate toward black-box quantum chemistry on quantum computers. In combination with adaptive circuit construction,^{73,75} this approach opens a path towards fully adaptive quantum chemistry independent of static basis sets and ansatz models. Combinations of adaptive circuit construction with generalized pair approximations^{56,74} in the spirit of ref 61 are promising candidates toward accurate, adaptive, and computationally feasible variational methods. Our approach is different in spirit as other approaches aiming

to reduce the qubit requirements by applying external corrections³⁷ or by using symmetries to obtain compressed representations.^{63,76} These approaches could be applied within the basis-set-free representation in the same way. Further improvements on qubit requirements and numerical accuracy can be expected in the context of explicitly correlated approaches, which were already applied in the original formulation of the MRA-PNO-MP2 optimization.³⁶ In the VQE context, approaches using trans-correlated Hamiltonians have been demonstrated recently using LCAO representations.^{38,39} These methods are promising candidates to further improve the performance of the basis-set-free VQE. We developed the first applications that exploit the structure of the pair-natural orbitals. Independent of the underlying numerical representation, this leads to low-depth quantum circuits, allowing savings in the qubit and gate requirements. These approaches have the potential to play an important role in future developments, either as benchmarks for small quantum hardware, or as initial states for more sophisticated quantum algorithms as illustrated for BeH₂ in Figure 5. In this work, we used MP2 as a correlated surrogate model, but other models are also possible within this framework and might bring additional advantages. A natural extension to MP2 is given by the coupled-cluster hierarchy, where models like CCD and CCSD first come to mind. More restricted models like pair-restricted coupled-cluster, which are also employed in quantum circuit construction,^{56,74} offer advantages through lower computational cost. In principle, more advanced surrogate models, that for example incorporate higher-order coupled-cluster terms only for the determination of specific orbitals, can be envisioned as well.

AUTHOR INFORMATION

Corresponding Authors

Jakob S. Kottmann – Chemical Physics Theory Group, Department of Chemistry, University of Toronto, Toronto, Canada; Department of Computer Science, University of Toronto, Toronto, Canada; orcid.org/0000-0002-4156-2048; Email: jakob.kottmann@utoronto.ca

Alán Aspuru-Guzik – Chemical Physics Theory Group, Department of Chemistry, University of Toronto, Toronto, Canada; Department of Computer Science, University of Toronto, Toronto, Canada; Vector Institute for Artificial Intelligence, Toronto, Canada; Canadian Institute for Advanced Research (CIFAR), Toronto, Canada; orcid.org/0000-0002-8277-4434; Email: aspuru@utoronto.ca

Authors

Philipp Schleich – Center for Computational Engineering Science, RWTH Aachen University, Aachen, Germany

Teresa Tamayo-Mendoza – Department of Computer Science, University of Toronto, Toronto, Canada; Department of Chemistry and Chemical Biology, Harvard University, Cambridge, Massachusetts, United States

Complete contact information is available at: <https://pubs.acs.org/10.1021/acs.jpclett.0c03410>

Notes

The authors declare no competing financial interest.

Our implementations are open-source and available online through the TEQUILA package under <https://github.com/aspuru-guzik-group/tequila>, where we provide initial tutorials

on the usage. A modified version of the MRA-PNO-MP2 implementation is available over a separate MADNESS branch under <https://github.com/kottmanj/madness/tree/tequila>. Feel free to contact J.S.K. for more information and updates.

■ ACKNOWLEDGMENTS

This work was supported by the U.S. Department of Energy under Award No. DE-SC0019374. A.A.-G. acknowledges the generous support from Google, Inc. in the form of a Google Focused Award. A.A.-G. also acknowledges support from the Canada Industrial Research Chairs Program and the Canada 150 Research Chairs Program. We thank the generous support of Anders G. Frøseth. P.S. acknowledges support from a fellowship within the IFI programme of the German Academic Exchange Service (DAAD). This research was enabled in part by support provided by Compute Canada. Computations were performed on the niagara supercomputer at the SciNet HPC Consortium.^{77,78} SciNet is funded by the Canada Foundation for Innovation; the Government of Ontario; Ontario Research Fund - Research Excellence; and the University of Toronto.

■ REFERENCES

- (1) Aspuru-Guzik, A.; Lindh, R.; Reiher, M. The Matter Simulation (R)evolution. *ACS Cent. Sci.* **2018**, *4*, 144–152. PMID: 29532014.
- (2) Fermann, J. T.; Valeev, E. F. *Fundamentals of Molecular Integrals Evaluation*; 2020.
- (3) Helgaker, T.; Jorgensen, P.; Olsen, J. *Molecular electronic-structure theory*; John Wiley & Sons, 2014.
- (4) Pritchard, B. P.; Altarawy, D.; Didier, B.; Gibson, T. D.; Windus, T. L. New Basis Set Exchange: An Open, Up-to-Date Resource for the Molecular Sciences Community. *J. Chem. Inf. Model.* **2019**, *59*, 4814–4820.
- (5) Van Lenthe, E.; Baerends, E. J. Optimized Slater-type basis sets for the elements 1–118. *J. Comput. Chem.* **2003**, *24*, 1142–1156.
- (6) Herbst, M. F.; Avery, J. E.; Dreuw, A. Quantum chemistry with Coulomb Sturmians: Construction and convergence of Coulomb Sturmian basis sets at the Hartree-Fock level. *Phys. Rev. A: At., Mol., Opt. Phys.* **2019**, *99*, 012512.
- (7) Genovese, L.; Neelov, A.; Goedecker, S.; Deutsch, T.; Ghasemi, S. A.; Willand, A.; Caliste, D.; Zilberberg, O.; Rayson, M.; Bergman, A.; et al. Daubechies wavelets as a basis set for density functional pseudopotential calculations. *J. Chem. Phys.* **2008**, *129*, 014109.
- (8) Mohr, S.; Ratcliff, L. E.; Genovese, L.; Caliste, D.; Boulanger, P.; Goedecker, S.; Deutsch, T. Accurate and efficient linear scaling DFT calculations with universal applicability. *Phys. Chem. Chem. Phys.* **2015**, *17*, 31360–31370.
- (9) Ratcliff, L. E.; Dawson, W.; Fiscaro, G.; Caliste, D.; Mohr, S.; Degomme, A.; Videau, B.; Cristiglio, V.; Stella, M.; D'Alessandro, M.; et al. Flexibilities of wavelets as a computational basis set for large-scale electronic structure calculations. *J. Chem. Phys.* **2020**, *152*, 194110.
- (10) Kang, S.; Woo, J.; Kim, J.; Kim, H.; Kim, Y.; Lim, J.; Choi, S.; Kim, W. Y. ACE-Molecule: An open-source real-space quantum chemistry package. *J. Chem. Phys.* **2020**, *152*, 124110.
- (11) Harrison, R. J.; Fann, G. I.; Yanai, T.; Gan, Z.; Beylkin, G. Multiresolution quantum chemistry: Basic theory and initial applications. *J. Chem. Phys.* **2004**, *121*, 11587–11598.
- (12) Harrison, R. J.; Beylkin, G.; Bischoff, F. A.; Calvin, J. A.; Fann, G. I.; Fosso-Tande, J.; Galindo, D.; Hammond, J. R.; Hartman-Baker, R.; Hill, J. C.; et al. MADNESS: A multiresolution, adaptive numerical environment for scientific simulation. *SIAM Journal on Scientific Computing* **2016**, *38*, S123–S142.
- (13) Bischoff, F. A. *Adv. Quantum Chem.*; Elsevier, 2019; Vol. 79, pp 3–52.
- (14) Frediani, L.; Fossgaard, E.; FlÅ, T.; Ruud, K. Fully adaptive algorithms for multivariate integral equations using the non-standard form and multiwavelets with applications to the Poisson and bound-state Helmholtz kernels in three dimensions. *Mol. Phys.* **2013**, *111*, 1143–1160.
- (15) Jensen, S. R.; Saha, S.; Flores-Livas, J. A.; Huhn, W.; Blum, V.; Goedecker, S.; Frediani, L. The Elephant in the Room of Density Functional Theory Calculations. *J. Phys. Chem. Lett.* **2017**, *8*, 1449–1457. PMID: 28291362.
- (16) Kottmann, J. S.; Höfener, S.; Bischoff, F. A. Numerically accurate linear response-properties in the configuration-interaction singles (CIS) approximation. *Phys. Chem. Chem. Phys.* **2015**, *17*, 31453–31462.
- (17) Yanai, T.; Fann, G. I.; Beylkin, G.; Harrison, R. J. Multiresolution quantum chemistry in multiwavelet bases: excited states from time-dependent Hartree-Fock and density functional theory via linear response. *Phys. Chem. Chem. Phys.* **2015**, *17*, 31405–31416.
- (18) Brakestad, A.; Jensen, S. R.; Wind, P.; D'Alessandro, M.; Genovese, L.; Hopmann, K. H.; Frediani, L. Static Polarizabilities at the Basis Set Limit: A Benchmark of 124 Species. *J. Chem. Theory Comput.* **2020**, *16*, 4874.
- (19) Sekino, H.; Yokoi, Y.; Harrison, R. J. A new implementation of dynamic polarizability evaluation using a multi-resolution multiwavelet basis set. *Journal of Physics. Conference Series (Online)* **2012**, *352*, 012014.
- (20) Bischoff, F. A. Structure of the H₃ molecule in a strong homogeneous magnetic field as computed by the Hartree-Fock method using multiresolution analysis. *Phys. Rev. A: At., Mol., Opt. Phys.* **2020**, *101*, 053413.
- (21) Jensen, S. R.; FlÅ, T.; Jonsson, D.; Monstad, R. S.; Ruud, K.; Frediani, L. Magnetic properties with multiwavelets and DFT: the complete basis set limit achieved. *Phys. Chem. Chem. Phys.* **2016**, *18*, 21145–21161.
- (22) Anderson, J.; Harrison, R. J.; Sundahl, B.; Thornton, W. S.; Beylkin, G. Real-space quasi-relativistic quantum chemistry. *Comput. Theor. Chem.* **2020**, *1175*, 112711.
- (23) Anderson, J.; Sundahl, B.; Harrison, R.; Beylkin, G. Dirac-Fock calculations on molecules in an adaptive multiwavelet basis. *J. Chem. Phys.* **2019**, *151*, 234112.
- (24) Bischoff, F. A.; Harrison, R. J.; Valeev, E. F. Computing many-body wave functions with guaranteed precision: The first-order Møller-Plesset wave function for the ground state of helium atom. *J. Chem. Phys.* **2012**, *137*, 104103.
- (25) Bischoff, F. A.; Valeev, E. F. Computing molecular correlation energies with guaranteed precision. *J. Chem. Phys.* **2013**, *139*, 114106.
- (26) Kottmann, J. S.; Bischoff, F. A. Coupled-Cluster in Real Space. 1. CC2 Ground State Energies Using Multiresolution Analysis. *J. Chem. Theory Comput.* **2017**, *13*, 5945–5955.
- (27) Kottmann, J. S.; Bischoff, F. A. Coupled-Cluster in Real Space. 2. CC2 Excited States Using Multiresolution Analysis. *J. Chem. Theory Comput.* **2017**, *13*, 5956–5965.
- (28) Sosa, C.; Geertsen, J.; Trucks, G. W.; Bartlett, R. J.; Franz, J. A. Selection of the reduced virtual space for correlated calculations. An application to the energy and dipole moment of H₂O. *Chem. Phys. Lett.* **1989**, *159*, 148–154.
- (29) DePrince, A. E., III; Sherrill, C. D. Accurate noncovalent interaction energies using truncated basis sets based on frozen natural orbitals. *J. Chem. Theory Comput.* **2013**, *9*, 293–299.
- (30) Riplinger, C.; Neese, F. An efficient and near linear scaling pair natural orbital based local coupled cluster method. *J. Chem. Phys.* **2013**, *138*, 034106.
- (31) Pinski, P.; Riplinger, C.; Valeev, E. F.; Neese, F. Sparse maps—A systematic infrastructure for reduced-scaling electronic structure methods. I. An efficient and simple linear scaling local MP2 method that uses an intermediate basis of pair natural orbitals. *J. Chem. Phys.* **2015**, *143*, 034108.
- (32) Babbush, R.; Berry, D. W.; McClean, J. R.; Neven, H. Quantum simulation of chemistry with sublinear scaling in basis size. *npj Quantum Information* **2019**, *5*, 92.
- (33) McClean, J.; Faulstich, F.; Zhu, Q.; O'Gorman, B.; Qiu, Y.; White, S. R.; Babbush, R.; Lin, L. Discontinuous Galerkin

discretization for quantum simulation of chemistry. *New J. Phys.* **2020**, *22*, 093015.

(34) Peruzzo, A.; McClean, J.; Shadbolt, P.; Yung, M.-H.; Zhou, X.-Q.; Love, P. J.; Aspuru-Guzik, A.; O'Brien, J. L. A variational eigenvalue solver on a photonic quantum processor. *Nat. Commun.* **2014**, *5*, 4213.

(35) McClean, J. R.; Romero, J.; Babbush, R.; Aspuru-Guzik, A. The theory of variational hybrid quantum-classical algorithms. *New J. Phys.* **2016**, *18*, 023023.

(36) Kottmann, J. S.; Bischoff, F. A.; Valeev, E. F. Direct determination of optimal pair-natural orbitals in a real-space representation: The second-order Møller–Plesset energy. *J. Chem. Phys.* **2020**, *152*, 074105.

(37) Takeshita, T.; Rubin, N. C.; Jiang, Z.; Lee, E.; Babbush, R.; McClean, J. R. Increasing the Representation Accuracy of Quantum Simulations of Chemistry without Extra Quantum Resources. *Phys. Rev. X* **2020**, *10*, 011004. Publisher: American Physical Society.

(38) Motta, M.; Gujarati, T. P.; Rice, J. E.; Kumar, A.; Masteran, C.; Latone, J. A.; Lee, E.; Valeev, E. F.; Takeshita, T. Y. Quantum simulation of electronic structure with transcorrelated Hamiltonian: increasing accuracy without extra quantum resources. *arXiv [physics, quant-ph]* **2020**, 2006.02488.

(39) McArdle, S.; Tew, D. P. Improving the accuracy of quantum computational chemistry using the transcorrelated method. *arXiv* **2020**, 2006.11181.

(40) Shavitt, I.; Bartlett, R. J. *Many-body methods in chemistry and physics: MBPT and coupled-cluster theory*; Cambridge University Press, 2009.

(41) Jørgensen, P. *Second quantization-based methods in quantum chemistry*; Elsevier, 2012.

(42) Surján, P. R. *Second quantized approach to quantum chemistry: an elementary introduction*; Springer Science & Business Media, 2012.

(43) Jordan, P.; Klein, O. Zum Mehrkörperproblem der Quantentheorie. *Eur. Phys. J. A* **1927**, *45*, 751–765.

(44) Bravyi, S. B.; Kitaev, A. Y. Fermionic Quantum Computation. *Ann. Phys.* **2002**, *298*, 210–226.

(45) Seeley, J. T.; Richard, M. J.; Love, P. J. The Bravyi-Kitaev transformation for quantum computation of electronic structure. *J. Chem. Phys.* **2012**, *137*, 224109.

(46) Setia, K.; Whitfield, J. D. Bravyi-Kitaev Superfast simulation of electronic structure on a quantum computer. *J. Chem. Phys.* **2018**, *148*, 164104.

(47) Barison, S.; Galli, D. E.; Motta, M. *Quantum simulations of molecular systems with intrinsic atomic orbitals*; 2020.

(48) Kalos, M. H. Monte Carlo Calculations of the Ground State of Three- and Four-Body Nuclei. *Phys. Rev.* **1962**, *128*, 1791–1795.

(49) Beylkin, G.; Mohlenkamp, M. J. Algorithms for Numerical Analysis in High Dimensions. *SIAM Journal on Scientific Computing* **2005**, *26*, 2133–2159.

(50) Kottmann, J. S. *Coupled-Cluster in Real Space*. Ph.D. Thesis. Humboldt-Universität zu Berlin, Mathematisch-Naturwissenschaftliche Fakultät, 2018.

(51) Neese, F.; Hansen, A.; Liakos, D. G. Efficient and accurate approximations to the local coupled cluster singles doubles method using a truncated pair natural orbital basis. *J. Chem. Phys.* **2009**, *131*, 064103.

(52) Werner, H.-J.; Knizia, G.; Krause, C.; Schwillk, M.; Dornbach, M. Scalable Electron Correlation Methods I: PNO-LMP2 with Linear Scaling in the Molecular Size and Near-Inverse-Linear Scaling in the Number of Processors. *J. Chem. Theory Comput.* **2015**, *11*, 484–507.

(53) Yanai, T.; Fann, G. I.; Gan, Z.; Harrison, R. J.; Beylkin, G. Multiresolution quantum chemistry in multiwavelet bases: Hartree–Fock exchange. *J. Chem. Phys.* **2004**, *121*, 6680–6688.

(54) Elfving, V. E.; Gámez, J. A.; Gogolin, C. Simulating quantum chemistry in the restricted Hartree-Fock space on a qubit-based quantum computing device. *arXiv* **2020**, 2002.00035.

(55) Khamoshi, A.; Evangelista, F. A.; Scuseria, G. E. Correlating AGP on a quantum computer. *arXiv* **2020**, 2008.06138.

(56) Lee, J.; Huggins, W. J.; Head-Gordon, M.; Whaley, K. B. Generalized unitary coupled cluster wave functions for quantum computation. *J. Chem. Theory Comput.* **2019**, *15*, 311–324.

(57) Kottmann, J. S.; Alperin-Lea, S.; Tamayo-Mendoza, T.; Cervera-Lierta, A.; Lavigne, C.; Yen, T.-C.; Verteletskyi, V.; Schleich, P.; Anand, A.; Degroote, M. et al. Tequila: A platform for rapid development of quantum algorithms. *arXiv* **2020**, 2011.03057.

(58) Suzuki, Y.; Kawase, Y.; Masumura, Y.; Hiraga, Y.; Nakadai, M.; Chen, J.; Nakanishi, K. M.; Mitarai, K.; Imai, R.; Tamiya, S. et al. Qulacs: a fast and versatile quantum circuit simulator for research purpose. *arXiv* **2020**, 2011.13524.

(59) Virtanen, P.; Gommers, R.; et al. SciPy 1.0: Fundamental Algorithms for Scientific Computing in Python. *Nat. Methods* **2020**, *17*, 261–272.

(60) McClean, J. R. et al., OpenFermion: The Electronic Structure Package for Quantum Computers. *arXiv* **2017**, 1710.07629.

(61) Kottmann, J. S.; Anand, A.; Aspuru-Guzik, A. A Feasible Approach for Automatically Differentiable Unitary Coupled-Cluster on Quantum Computers. *arXiv* **2020**, 2011.05938.

(62) Smith, D. G. A.; Burns, L. A.; Simmonett, A. C.; Parrish, R. M.; Schieber, M. C.; Galvelis, R.; Kraus, P.; Kruse, H.; Di Remigio, R.; Alenaizan, A.; et al. PSI4 1.4: Open-source software for high-throughput quantum chemistry. *J. Chem. Phys.* **2020**, *152*, 184108.

(63) Bravyi, S.; Gambetta, J. M.; Mezzacapo, A.; Temme, K. Tapering off qubits to simulate fermionic Hamiltonians. *arXiv* **2017**, 1701.08213.

(64) Dehesa, J. S.; Angulo, J. C.; Koga, T.; Matsui, K. Study of some interelectronic properties in helium-like atoms. *Z. Phys. D: At., Mol. Clusters* **1992**, *25*, 9–16.

(65) Nakatsuji, H. Discovery of a General Method of Solving the Schrödinger and Dirac Equations That Opens a Way to Accurately Predictive Quantum Chemistry. *Acc. Chem. Res.* **2012**, *45*, 1480–1490. PMID: 22686372.

(66) Bischoff, F. A. Regularizing the molecular potential in electronic structure calculations. II. Many-body methods. *J. Chem. Phys.* **2014**, *141*, 184106.

(67) Kato, T. On the eigenfunctions of many-particle systems in quantum mechanics. *Communications on Pure and Applied Mathematics* **1957**, *10*, 151–177.

(68) Bischoff, F. A. Regularizing the molecular potential in electronic structure calculations. I. SCF methods. *J. Chem. Phys.* **2014**, *141*, 184105.

(69) Kutzelnigg, W. r_{12} -Dependent terms in the wave function as closed sums of partial wave amplitudes for large l . *Theoretica chimica acta* **1985**, *68*, 445–469.

(70) Tamayo-Mendoza, T.; Kreisbeck, C.; Lindh, R.; Aspuru-Guzik, A. Automatic Differentiation in Quantum Chemistry with Applications to Fully Variational Hartree–Fock. *ACS Cent. Sci.* **2018**, *4*, 559–566.

(71) Bande, A.; Nakashima, H.; Nakatsuji, H. LiH potential energy curves for ground and excited states with the free complement local Schrödinger equation method. *Chem. Phys. Lett.* **2010**, *496*, 347–350.

(72) Tamayo-Mendoza, T. *DiffiQult*; 2018; <https://github.com/aspuru-guzik-group/DiffiQult.git>

(73) Grimsley, H. R.; Economou, S. E.; Barnes, E.; Mayhall, N. J. An adaptive variational algorithm for exact molecular simulations on a quantum computer. *Nat. Commun.* **2019**, *10*, 3007.

(74) Sokolov, I. O.; Barkoutsos, P. K.; Ollitrault, P. J.; Greenberg, D.; Rice, J.; Pistoia, M.; Tavernelli, I. Quantum orbital-optimized unitary coupled cluster methods in the strongly correlated regime: Can quantum algorithms outperform their classical equivalents? *J. Chem. Phys.* **2020**, *152*, 124107.

(75) Ryabinkin, I. G.; Yen, T.-C.; Genin, S. N.; Izmaylov, A. F. Qubit coupled cluster method: a systematic approach to quantum chemistry on a quantum computer. *J. Chem. Theory Comput.* **2018**, *14*, 6317–6326.

(76) Setia, K.; Chen, R.; Rice, J. E.; Mezzacapo, A.; Pistoia, M.; Whitfield, J. Reducing qubit requirements for quantum simulation using molecular point group symmetries. *arXiv* **2019**, 1910.14644.

(77) Ponce, M.; van Zon, R.; Northrup, S.; Gruner, D.; Chen, J.; Ertinaz, F.; Fedoseev, A.; Groer, L.; Mao, F.; Mundim, B. C. et al. *Proceedings of the Practice and Experience in Advanced Research Computing on Rise of the Machines (learning)*; 2019; pp 1–8.

(78) Loken, C.; Gruner, D.; Groer, L.; Peltier, R.; Bunn, N.; Craig, M.; Henriques, T.; Dempsey, J.; Yu, C.-H.; Chen, J.; et al. SciNet: lessons learned from building a power-efficient top-20 system and data centre. *Journal of Physics-Conference Series* **2010**, 256, 012026.