## Multi-determinant Compression Algorithm for Quantum Monte Carlo

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## Multi-determinant expansions

Multi-determinant expansions are a tool to construct very accurate wave functions, and are frequently used in quantum Chemistry methods.

Slater determinants constitute a basis for anti-symmetric functions in $\mathscr{R}^{3 N}$, thus an expansion in Slater determinants converges to the exact wave function of the system as the number of determinants tends to $\infty$.

## The multi-determinant expansion

## Multi-determinant expansion

$$
\Psi_{\mathrm{MD}}(\mathbf{R})=\sum_{k=1}^{N_{s}} c_{k} \Phi_{k}^{\uparrow}\left(\mathbf{R}_{\uparrow}\right) \Phi_{k}^{\downarrow}\left(\mathbf{R}_{\downarrow}\right)
$$

- $\Phi_{k}^{\uparrow}\left(\mathbf{R}_{\uparrow}\right)=\operatorname{det}\left[\phi_{a_{i, k}^{\uparrow}}\left(\mathbf{r}_{j}^{\uparrow}\right)\right]$
- $\Phi_{k}^{\downarrow}\left(\mathbf{R}_{\downarrow}\right)=\operatorname{det}\left[\phi_{a_{i, k}^{\downarrow}}\left(\mathbf{r}_{j}^{\downarrow}\right)\right]$
- $a_{i, k}^{\sigma}$ is an orbital-selecting index


## Cost and limitations

Multi-determinant expansions have the drawback that the number of determinants $N_{s}$ required to obtain a wave function of a given accuracy grows very rapidly with system size $N$.

- This is why methods such as Cl typically scale as $N^{7}$
- Restricts multi-determinant wave fuctions to small systems

The cost of running a given number of steps in a QMC calculation is proportional to $N_{s}$.

Methods to speed up the evaluation of multi-determinant wave functions extend the range of system sizes where this tool can be used.

## Notation

For convenience, we use a compact notation to represent determinants:

## Notation for determinants

$$
\left[\phi_{a_{1}}, \phi_{a_{2}}, \ldots, \phi_{a_{n}}\right] \equiv\left|\begin{array}{cccc}
\phi_{a_{1}}\left(\mathbf{r}_{1}\right) & \phi_{a_{2}}\left(\mathbf{r}_{1}\right) & \cdots & \phi_{a_{n}}\left(\mathbf{r}_{1}\right) \\
\phi_{a_{1}}\left(\mathbf{r}_{2}\right) & \phi_{a_{2}}\left(\mathbf{r}_{2}\right) & \cdots & \phi_{a_{n}}\left(\mathbf{r}_{2}\right) \\
\vdots & \vdots & \ddots & \vdots \\
\phi_{a_{1}}\left(\mathbf{r}_{n}\right) & \phi_{a_{2}}\left(\mathbf{r}_{n}\right) & \cdots & \phi_{a_{n}}\left(\mathbf{r}_{n}\right)
\end{array}\right|
$$

## De-duplicating determinants

Multi-determinant wave functions are commonly written as a sum of Configuration State Functions (CSFs), which are themselves a sum of determinant products obeying a certain symmetry.
Each determinant product may appear in more than one CSF, and hence a multi-determinant expansion may contain multiple instances of a given determinant product. We combine these into a single term:

$$
\begin{aligned}
& \text { De-duplication example } \\
& c_{1} \Phi^{\uparrow} \Phi^{\downarrow}+c_{2} \Phi^{\uparrow} \Phi^{\downarrow}=\left(c_{1}+c_{2}\right) \Phi^{\uparrow} \Phi^{\downarrow}=c_{1}^{\prime} \Phi^{\uparrow} \Phi^{\downarrow}
\end{aligned}
$$

This operation may involve more than two terms. The de-duplication process is quick and simple, and independent from what follows.

## Combining determinants

The basic operation at the core of our compression algorithm is:

## Basic combination example, simplified

$$
\begin{array}{cc}
{\left[\phi_{a_{1}}, \phi_{a_{3}}, \ldots, \phi_{a_{4}}\right]+\left[\phi_{a_{2}}, \phi_{a_{3}}, \ldots, \phi_{a_{4}}\right]} & = \\
{\left[\phi_{a_{1}}+\phi_{a_{2}}, \phi_{a_{3}}, \ldots, \phi_{a_{4}}\right]} & = \\
{\left[\tilde{\phi}_{\tilde{a}_{1}}, \phi_{a_{3}}, \ldots, \phi_{a_{4}}\right]} &
\end{array}
$$

That is, we can "compress" two determinants which differ by a single orbital into one determinant by defining a new orbital which is a linear combination of two of the orbitals in the original pool of orbitals.

## Combining determinants

In practice, we need to consider the other-spin determinant, which must be equal in both terms, and the (de-duplicated) expansion coefficients:

## Basic combination example

$$
\begin{array}{cc}
c_{1}^{\prime}\left[\phi_{a_{1}}, \phi_{a_{3}}, \ldots, \phi_{a_{4}}\right] \Phi^{\downarrow}+c_{2}^{\prime}\left[\phi_{a_{2}}, \phi_{a_{3}}, \ldots, \phi_{a_{4}}\right] \Phi^{\downarrow} & = \\
{\left[c_{1}^{\prime} \phi_{a_{1}}+c_{2}^{\prime} \phi_{a_{2}}, \phi_{a_{3}}, \ldots, \phi_{a_{4}}\right] \Phi^{\downarrow}} & = \\
\tilde{c}_{1}\left[\tilde{\phi}_{\tilde{a}_{1}}, \phi_{a_{3}}, \ldots, \phi_{a_{4}}\right] \Phi^{\downarrow}
\end{array}
$$

This operation may involve more than two terms.

## Recursive combination

In some cases determinants which have already been compressed can be compressed again:

## Recursive combination example

$$
\begin{array}{cc}
{\left[\phi_{a_{1}}, \phi_{a_{3}}\right]+\left[\phi_{a_{2}}, \phi_{a_{3}}\right]+\left[\phi_{a_{1}}, \phi_{a_{4}}\right]+\left[\phi_{a_{2}}, \phi_{a_{4}}\right]} & = \\
{\left[\phi_{a_{1}}+\phi_{a_{2}}, \phi_{a_{3}}\right]+\left[\phi_{a_{1}}+\phi_{a_{2}}, \phi_{a_{4}}\right]} & = \\
{\left[\phi_{a_{1}}+\phi_{a_{2}}, \phi_{a_{3}}+\phi_{a_{4}}\right]} &
\end{array}
$$

This operation may involve several terms.

## Generalization

A compressed multi-determinant expansion may be expressed as

$$
\begin{aligned}
& \text { General form of compressed expansion } \\
& \Psi_{\mathrm{MD}}(\mathbf{R})=\sum_{k=1}^{N_{c}} \tilde{c}_{k} \operatorname{det}\left[\tilde{\phi}_{\tilde{a}_{i k}^{\uparrow}}\left(\mathbf{r}_{j}^{\uparrow}\right)\right] \operatorname{det}\left[\tilde{\phi}_{\tilde{a}_{i k}^{\downarrow}}\left(\mathbf{r}_{j}^{\downarrow}\right)\right] \\
& \tilde{c}_{k}= \pm c_{v_{k 1}}^{\prime} \prod_{p=2}^{P_{k}} \frac{c_{v_{k p}}^{\prime}}{c_{\delta_{k p}}^{\prime}} \\
& \tilde{\phi}_{a}(\mathbf{r})=\sum_{x=1}^{X_{a}} \pm \prod_{q=1}^{Q_{a x}} \frac{c_{n_{a x q}}^{\prime}}{c_{d_{a x q}}^{\prime}} \phi_{\mu_{a x}}(\mathbf{r})
\end{aligned}
$$

## Properties of the expansion

From the form of the expansion it can be seen that:

- The compressed expansion is a regular multi-determinant expansion and can be evaluated with the usual algorithms.
- The evaluation of the expansion is reduced by a factor of $N_{c} / N_{s}$, although there is the additional step of calculating the set of compressed orbitals from the original orbitals.
- The compressed expansion coefficients can be easily recalculated if the original coefficients change, i.e., during optimization.
NB, we optimize the original expansion coefficients to be able to keep the constraints on the parameters, e.g., proportionality between coefficients in the same CSF.


## Selecting operations

It may be possible to combine a given term in an expansion with more than one other term:

## Multiple choices

$$
\begin{aligned}
{\left[\phi_{a_{1}}, \phi_{a_{3}}\right]+\left[\phi_{a_{2}}, \phi_{a_{3}}\right]+\left[\phi_{a_{1}}, \phi_{a_{4}}\right] } & = \\
{\left[\phi_{a_{1}}+\phi_{a_{2}}, \phi_{a_{3}}\right]+\left[\phi_{a_{1}}, \phi_{a_{4}}\right] } & = \\
{\left[\phi_{a_{1}}, \phi_{a_{3}}+\phi_{a_{4}}\right]+\left[\phi_{a_{2}}, \phi_{a_{3}}\right] } &
\end{aligned}
$$

Choosing one grouping over another gives a different compressed expansion. One needs to choose the groupings so as to minimize the number of determinants in the compressed expansion $N_{c}$.

## Set covering problem

This can be reformulated as a set-covering problem: find the least number of subsets of a set that cover the entire set.


## The greedy algorithm

The set-covering problem can be approximately solved in polynomial time using the "greedy algorithm".

## The greedy algorithm

## Pick the subsets from largest to smallest until the set is covered.

The greedy algorithm gives very acceptable results in practice, but it does not give the optimal solution to the set-covering problem in general.

## Linear programs

The set-covering problem can be solved exactly using a binary linear program.

A binary linear program is an optimization problem where an objective function of binary unknowns is to be optimized subject to a set of linear constraints.

In the case of compression:

- the unknowns determine whether a subset is picked or not
- the constraints make sure we pick each term exactly once
- the objective function is the number of terms in the compressed expansion.


## Potential issues with the cost of compression

Linear programs are in principle NP-hard. However in practice it is possible to solve many of them quickly.

The construction of the complete set of operations that can be performed on the multi-determinant expansion is potentially very costly. This can be easily avoided, except in the presence of recursively-compressible terms where the full set of recursive operations need to be explicitly listed.

It is possible to deal approximately with recursively-compressible terms by applying the compression algorithm to the compressed expansion iteratively until not further gains are obtained.

## Limitations

We do not take into account the ability to do something like:

## Using a term multiple times

$$
\begin{gathered}
{\left[\phi_{a_{1}}, \phi_{a_{2}}\right]+2\left[\phi_{a_{1}}, \phi_{a_{3}}\right]+\left[\phi_{a_{2}}, \phi_{a_{3}}\right]+\left[\phi_{a_{1}}, \phi_{a_{4}}\right]+2\left[\phi_{a_{2}}, \phi_{a_{4}}\right]+\left[\phi_{a_{3}}, \phi_{a_{4}}\right]=} \\
{\left[\phi_{a_{1}}, \phi_{a_{2}}+\phi_{a_{3}}\right]+\left[\phi_{a_{1}}+\phi_{a_{2}}, \phi_{a_{3}}\right]+\left[\phi_{a_{1}}+\phi_{a_{2}}, \phi_{a_{4}}\right]+\left[\phi_{a_{2}}+\phi_{a_{3}}, \phi_{a_{4}}\right]=} \\
{\left[\phi_{a_{1}}-\phi_{a_{4}}, \phi_{a_{2}}+\phi_{a_{3}}\right]+\left[\phi_{a_{1}}+\phi_{a_{2}}, \phi_{a_{3}}+\phi_{a_{4}}\right]}
\end{gathered}
$$

where the second and fifth terms have been used twice. This type of operation would complicate matters quite a bit.

## Implementation

We provide a GPLv3 implementation of the compression algorithm,

## Our implementation

http://www.tcm.phy.cam.ac.uk/~pl275/DetCompress

This implementation offers three levels of compression:

- "Quick": greedy algorithm, iterative recursion
- "Good": linear program, iterative recursion
- "Best": linear program with full recursion

Less-costly levels are available in case the exponential cost of the potentially problematic parts of the algorithm manifests itself.

## Results: number of determinants

|  | $N_{\text {CSF }}$ | Original | De-dupe | "Quick" | "Good" | "Best" |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: |
| $\mathrm{Be}_{2}$ | 61 | 200 | 200 | 100 | 97 | 97 |
| N | 50 | 1271 | 764 | 324 | 324 | 324 |
| O | 100 | 3386 | 1271 | 535 | 534 | 534 |
| Li | 500 | 8140 | 5824 | 1226 | 1210 | 1210 |
| B | 500 | 14057 | 5703 | 530 | 529 | 529 |
| Be | 500 | 14212 | 10600 | 2218 | 2177 | 2163 |
| Ne | 400 | 22827 | 16260 | 1844 | 1805 | 1805 |
| F | 600 | 57456 | 17174 | 2801 | 2749 | 2747 |

The reduction in the number of determinants is very significant. All three levels offer very similar performance.

## Results: number of orbitals

|  | $N_{\text {CSF }}$ | Original | De-dupe | "Quick" | "Good" | "Best" |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: |
| $\mathrm{Be}_{2}$ | 61 | 12 | 12 | 60 | 72 | 73 |
| N | 50 | 51 | 51 | 191 | 188 | 188 |
| O | 100 | 53 | 53 | 365 | 361 | 361 |
| Li | 500 | 105 | 105 | 1023 | 1036 | 1046 |
| B | 500 | 105 | 105 | 629 | 629 | 631 |
| Be | 500 | 105 | 105 | 1174 | 1188 | 1198 |
| Ne | 400 | 105 | 105 | 1182 | 1191 | 1191 |
| F | 600 | 105 | 105 | 2553 | 2622 | 2627 |

The increase in the number of orbitals is likewise very significant.

## Results: compression CPU time

|  | $N_{\text {CSF }}$ | De-dupe | "Quick" | "Good" | "Best" |
| :--- | ---: | :---: | :---: | :---: | :---: |
| $\mathrm{Be}_{2}$ | 61 | $0.001(1)$ | $0.011(1)$ | $0.022(1)$ | $0.023(1)$ |
| N | 50 | $0.012(1)$ | $0.038(2)$ | $0.044(1)$ | $0.054(1)$ |
| O | 100 | $0.058(1)$ | $0.125(1)$ | $0.144(1)$ | $0.143(1)$ |
| Li | 500 | $0.332(6)$ | $1.270(2)$ | $1.430(3)$ | $3.144(3)$ |
| B | 500 | $0.79(1)$ | $1.802(2)$ | $1.884(7)$ | $3.73(1)$ |
| Be | 500 | $1.18(1)$ | $4.63(1)$ | $5.17(2)$ | $7.82(3)$ |
| Ne | 400 | $4.56(3)$ | $14.79(8)$ | $15.04(6)$ | $18.77(4)$ |
| F | 600 | $11.43(1)$ | $22.3(2)$ | $22.5(2)$ | $24.18(4)$ |

The three levels take very similar CPU times.

## Results: QMC CPU time

|  | $N_{s} / N_{b}$ | $T_{s} / T_{b}$ |
| :--- | ---: | ---: |
| $\mathrm{Be}_{2}$ | 2.06 | $1.885(3)$ |
| N | 3.92 | $3.718(9)$ |
| O | 6.34 | $6.09(1)$ |
| Li | 6.73 | $6.50(2)$ |
| B | 26.57 | $25.23(4)$ |
| Be | 6.57 | $6.48(1)$ |
| Ne | 12.65 | $13.17(2)$ |
| F | 20.92 | $21.77(5)$ |

CPU times closely mirror compression ratios. Computation of compressed orbitals has an insignificant impact in our tests.

## Results: scaling of QMC CPU time

The dependence of the cost on $N_{s}$ appears to become sub-linear when compression is used:


## Summary

Compression method for reducing the cost of evaluating multi-determinant wave functions.

- Leaves evaluation algorithms unchanged.
- Leaves original coefficients optimizable.
- Achieves compression ratios of $2-25$ in tests.
- Compression ratio directly translates to QMC speed-up.
- Sub-linear scaling of cost of QMC with expansion size.
- Future work: combine with accelerated evaluation methods.
- Implementation:
http://www.tcm.phy.cam.ac.uk/~pl275/DetCompress
- Paper: PRE 89, 023304 (2014)

Introduction
Compression Results and conclusion

Implementation
Results
Summary

## End

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