# A Pairing Method Which Improves Convergence in Monte-Carlo Estimation of Quantum Mechanical Expectation Values 

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#### Abstract

The "pairing" method reduces the variance in Monte-Carlo estimates of expectation values by combining the locations of the largest and smallest contributors to these estimates. The number of paired locations and hence the variance reduction increase with the number of wave function evaluations $N_{\mathrm{e}}$. This allows Monte-Carlo-like calculations to converge at rates faster than $V_{e}^{1.2}$. In particular, an evaluation of the expectation values of the potential, $\langle V\rangle$. for $\mathrm{H}_{2}^{+}$has a standard deviation which decreases as $N_{c}^{-0.72}$ and an evaluation of $\langle V\rangle$ for spinaligned $\mathrm{H}_{2}$ at the minimum in its potential well has a standard deviation which decreases as $\therefore$ N62. $\because 1985$ Academic Press. Inc


## Introndection

An $n$-dimensional integral ready for Monte-Carlo integration can have the form

$$
\begin{equation*}
I=\int_{0}^{1} \cdots \int_{0}^{1} f(\mathbf{x}) d x_{1} \cdots d x_{n}=\langle f\rangle \tag{1}
\end{equation*}
$$

where $\mathbf{x}$ is an $n$-dimensional vector. A Monte-Carlo estimate of $I$ using $N$ random $x_{i}$ 's is

$$
\begin{equation*}
I_{\mathrm{MC}}=\frac{1}{N} \sum_{i=1}^{N} f\left(\mathbf{x}_{i}\right) \tag{2}
\end{equation*}
$$

where $\mathbf{x}_{i}$ is the $i$ th choice of the $n$ components of $\mathbf{x}$ with each component being chosen uniformly and randomly between 0 and 1 . The standard deviation of the mean value of $I_{\mathrm{MC}}$ is $\sigma_{\mathrm{MC}}$, where

$$
\begin{equation*}
\sigma_{\mathrm{MC}}^{2}=\frac{\left\langle(\langle f\rangle-f)^{2}\right\rangle}{N} . \tag{3}
\end{equation*}
$$

[^0]For our purposes it is necessary to allow the argument of $f$ to be outside the region 0 to 1 . Defining $f(\mathbf{x}+\mathrm{I})=f(\mathbf{x})$ where I is an $n$-dimensional vector of integers makes $f$ explicitly periodic so that

$$
\begin{equation*}
I_{\mathrm{MC}}=\frac{1}{N} \sum_{i=1}^{N} f\left(\mathbf{x}_{i}+\mathbf{a}\right) \tag{4}
\end{equation*}
$$

where a is any $n$-dimensional vector. This allows the introduction of a set of $M$ vectors $\mathbf{a}_{q}$ similar to the antithetic variates of Hammersly and Hanscomb [1], which can be held constant while $\mathbf{x}_{p}$ is chosen randomly to yield

$$
\begin{equation*}
f_{A}\left(\mathbf{x}_{p}\right)=\frac{1}{M} \sum_{q=1}^{M} f\left(\mathbf{x}_{p}+\mathbf{a}_{q}\right) \tag{5}
\end{equation*}
$$

This allows us to define

$$
\begin{equation*}
I_{\mathrm{A}}=\frac{1}{N} \sum_{p-1}^{N} f_{\mathrm{A}}\left(\mathbf{x}_{p}\right) \tag{6}
\end{equation*}
$$

with standard deviation

$$
\begin{equation*}
\sigma_{A}^{2}=\frac{\left\langle\left(\langle f\rangle-f_{A}\right)^{2}\right\rangle}{N} \tag{7}
\end{equation*}
$$

In Eq. (7) the fact that $\left\langle f_{A}\right\rangle=\langle f\rangle$ has been utilized and it should be noted that $\left\langle f_{\mathrm{A}}^{2}\right\rangle$ is not equal to $\left\langle f^{2}\right\rangle$. Only $N$ independent random vectors enter Eqs. (6) and (7) but the number of function evaluations $N_{c}$ is $N$ times $M$. Pairing is preferable to unbiased Monte-Carlo if $M \sigma_{A}^{2}<\sigma_{M C}^{2}$. In this work, a method for systematically finding new $\mathbf{a}_{q}$ 's for which $\sigma_{A}^{2}$ decreases faster than $1 / M$ will be described.

A scheme for finding these $\mathbf{a}_{q}$ 's that essentially finds different pseudo-random number generators in each dimension and minimizes $\sigma^{2}$ for a class of "worst" functions has been developed by Haselgrove [2]. The scheme is supposed to yield $\sigma^{2}=x / N^{2}$ and indeed for some functions this appears to be achieved. The method was extended by Burdick [3] to include the use of biased selection methods, quantum mechanical expectation values, and standard deviation cstimates similar to Eq. (7). The method has $\sigma_{\mathrm{A}}^{2}=\alpha / N^{2}$ for small values of $N$, but with $\sigma_{\mathrm{A}}^{2}>\sigma_{\mathrm{MC}}^{2}$. For $N$ on the order of the optimization set $\sigma_{\mathrm{A}}^{2}$ becomes smaller than $\sigma_{\mathrm{MC}}^{2}$, but after about a factor of 2 gain changes slope and becomes proportional to $1 / N$.

The new ingredient in this work is abandonment of the notion of a universal set of $\mathbf{a}_{q}$ 's. Indeed $\sigma_{A}^{2}$ for other functions often rises while that for the function of interest drops. The method is simple, intuitive, and easy to use. It probably works because it utilizes in a small way the locations of the Monte-Carlo function evaluations, information which is usually discarded.

## The Method

The largest and smallest values of $f\left(\mathbf{x}_{i}\right)$ contribute the largest amounts to $N \sigma_{\mathrm{MC}}^{2}$ in Eq. (3). If they could be paired so that they can only be found for a single $\mathbf{x}_{i}$ value in Eq. (5), the resulting $f_{\mathrm{A}}(\mathbf{x})$ which contains these two values contributes much less to $\sigma_{\mathrm{A}}^{2}$ in Eq. (7) while contributing exactly the same amount to $I_{\mathrm{A}}$ in Eq. (6).

The pairing method utilizes a number of $N$ point estimators of $I_{\mathrm{A}}$. The first level 0 is made by selecting $N$ vectors $\mathbf{x}_{i}$ uniformly and randomly and averaging as in Eq. (2). The vector $\mathbf{b}_{1}^{1}$ is made equal to the $\mathbf{x}_{i}$ for which $f\left(\mathbf{x}_{i}\right)$ is largest and the vector $\mathbf{b}_{2}$ to the $\mathbf{x}_{i}$ for which $f\left(\mathbf{x}_{i}\right)$ is smallest. Then at level 1

$$
\begin{align*}
& \mathbf{a}_{1}=\mathbf{b}_{1}^{1}  \tag{8}\\
& \mathbf{a}_{2}=\mathbf{b}_{2}^{1} \tag{9}
\end{align*}
$$

The second $N$ point estimator consists of an independent set of $\mathbf{x}_{p}$ 's used to average the $f_{A}\left(\mathbf{x}_{p}\right)$ in Eq. (5) as in Eq. (6) with the $\mathbf{a}_{q}$ 's given in Eqs. (8) and (9). Note that an $\mathbf{x}_{p}$ of zero in this set would yield

$$
\begin{equation*}
f_{\mathrm{A}}(0)=\frac{1}{2}\left(f\left(\mathbf{a}_{1}\right)+f\left(\mathbf{a}_{2}\right)\right) \tag{10}
\end{equation*}
$$

which as a pair contributes less to the sum in Eq. (7) than these same terms occurring independently contribute to the sum in Eq. (3).

The $\mathbf{x}_{p}$ 's for which $f_{\mathrm{A}}\left(\mathbf{x}_{p}\right)$ is extreme are saved as $\mathbf{b}_{1}^{2}$ and $\mathbf{b}_{2}^{2}$. This enables us to generate the set of a's for level 2 as

$$
\begin{align*}
\mathbf{a}_{1} & =\mathbf{b}_{1}^{1}+\mathbf{b}_{1}^{2} \\
\mathbf{a}_{2} & =\mathbf{b}_{1}^{1}+\mathbf{b}_{2}^{2} \\
\mathbf{a}_{3} & =\mathbf{b}_{2}^{1}+\mathbf{b}_{1}^{2}  \tag{11}\\
\mathbf{a}_{4} & =\mathbf{b}_{2}^{1}+\mathbf{b}_{2}^{2}
\end{align*}
$$

to be used in the next $N$ point estimator of $i$. Finally, the $L$ th set of $a$ 's to be used in the $L+1$ level estimator are

$$
\begin{equation*}
\mathbf{a}_{j}=\sum_{i=1}^{L} \mathbf{b}_{p i}^{\prime} \tag{12}
\end{equation*}
$$

where $p j$ is the $j$ th set of the $2^{L}$ possible sets of $\{1,2,1, \ldots, 2\}$. Note that only $2 L$ terms need be stored to generate the $M=2^{L}$ vectors $\mathbf{a}_{q}$.

The result of these efforts is that $f_{\mathrm{A}}\left(x_{i}\right)$ may be written

$$
\begin{align*}
f_{\mathrm{A}}\left(x_{i}\right) & =\frac{1}{M} \sum_{j=1}^{M} f\left(\mathbf{x}_{i}+\mathbf{a}_{j}\right)  \tag{13}\\
& =\frac{1}{M} \sum_{j=1}^{2} \sum_{k=1}^{2} \cdots \sum_{l=1}^{2} f\left(\mathbf{x}_{i}+\mathbf{b}_{j}^{1}+\mathbf{b}_{k}^{2}+\cdots+\mathbf{b}_{l}^{L}\right) .
\end{align*}
$$

Note that each additional level utilizes a new set of $N$ random $\mathbf{x}_{p}$ 's and doubles the total number of function evaluations.

Before introducing the expectation values used in testing the pairing method, it can be illustrated with a simple $2 d$ lattice function. Let space consist only of the 16 lattice points shown in Fig. 1 along with some of its periodic repeats. The function $f(x, y)$ is 4 at points $(0.25,0.25),(0.25,0.50),(0.50,0.25),(0.50,0.50)$ and zero for the other 12 . The average value $\langle f\rangle=1$ and $\left\langle f^{2}\right\rangle=4$ which implies $N \sigma_{\mathrm{MC}}^{2}=3$. A pairing method estimation of $(f)$ at the end of level 0 would thus have the usual Monte-Carlo standard deviation $\sigma^{2}=3 / N$. In addition the code would have found $\mathbf{b}_{1}^{1}=(0.25,0.25)$ or its equivalent and $\mathbf{b}_{2}^{1}$ would be $(0.0,0.50)$ or one of the 12 other $x, y$ value sets for which $f(x, y)=0$. This one will be followed through a number of levels.

The vectors $\mathbf{a}_{q}$ at level 1 are $(0.25,0.25)$ and $(0,0.50)$, there is one $x$ for which $f_{\mathrm{A}}=4$, which makes $b_{1}^{2}=(0.250)$, six $x$ 's for which $f_{\mathrm{A}}=2$, and nine for which $f_{\mathrm{A}}=0$, so that $b_{2}^{2}$ can be picked as $(0,0.75)$. For this choice at level $1, N \sigma_{\mathrm{A}}^{2}=\frac{3}{2}$, exactly the same as for this many random function evaluations. The vector $\mathbf{a}_{q}$ at level 2 is $(0.50,0.25),(0.25,0.50)(0.25,0),(0,0.25)$. There are four $x$ 's with $f_{\mathrm{A}}=2$, eight with $f_{\mathrm{A}}=1$, and four with $f_{\mathrm{A}}=0$. The $b$ 's may be chosen to be $b_{1}^{3}=(0,0.25)$ and $b_{2}^{3}=(0.5,0.75)$. At level $2, N \sigma_{\mathrm{A}}^{2}=\frac{1}{2}$, which is less than the value of $\frac{3}{4}$ for this many random evaluations. The vectors $\mathbf{a}_{q}$ at level 3 are $(0.50,0.50),(0.25,0.75),(0.50,0)$, $(0.25,0.25),(0,0.50),(0.75,0.75),(0,0)$, and $(0.75,0.25)$ which is the checkerboard pattern covering every other square shown in Fig. 1. For all $x, f_{\mathrm{A}}=1$ and thus at level $3 N \sigma_{\mathrm{A}}^{2}=0$.


Fig. 1. A simple function $f(x, y)$ defined on the lattice points. In the shaded regions $f(x, y)=4$, elsewhere $f(x, y)=0$. The $A^{\prime}$ 's are the $\mathbf{a}_{q}$ vectors at level 3 in the example.

The example is so simple that the correct four trapezoidal rule points also give an exact answer. Furthermore only one of the multitude of ways of reaching level 3 has been followed in detail. A bit of promise, however, is present in the example.

## Expectation Values

A Monte-Carlo estimate of the expectation value of $f$ with respect to a trial wave function $\psi_{1}$ may be written as

$$
\begin{equation*}
\langle f\rangle=\frac{\int_{0}^{1} F(\mathbf{x}) d \mathbf{x}}{\int_{0}^{1} G(\mathbf{x}) d \mathbf{x}}=\frac{1_{i}^{\prime} N \sum_{i=1}^{N} F_{A}\left(\mathbf{x}_{i}\right)}{1 / N \sum_{i=1}^{N} G_{A}\left(\mathbf{x}_{i}\right)} \tag{14}
\end{equation*}
$$

wherc

$$
\begin{equation*}
F=f \psi_{t}^{2}, \quad G=\psi_{t}^{2} \tag{15}
\end{equation*}
$$

and $F_{\mathrm{A}}$ and $G_{\mathrm{A}}$ are defined by Eq. (5). The same $\mathbf{x}_{i}$ vectors are used in both the numerator and denominator which makes the estimate more accurate than performing two independent estimates of the integrals and dividing by the result [9], though the convergence rate is still the $N^{-1: 2}$ typical of Monte-Carlo methods.

The standard deviation $\sigma$ in the estimate of an expectation value using (14) may be written as

$$
\begin{equation*}
\sigma_{\Lambda}^{2}=\frac{1}{N\left\langle G_{\Lambda}\right\rangle^{2}}\left\langle\left(F_{\Lambda}-\langle f\rangle G_{\Lambda}\right)^{2}\right\rangle \tag{16}
\end{equation*}
$$

which is equivalent to Eq. (20) derived in Appendix A.
The biased selection method described in Appendix B changes a simple MonteCarlo estimate to

$$
\begin{equation*}
I_{\mathrm{MC}}=\frac{1}{N} \sum_{i} f\left(\mathbf{x}\left(\mathbf{q}_{i}\right)\right)_{i} w\left(\mathbf{q}_{i}\right) \tag{17}
\end{equation*}
$$

where $\mathbf{q}_{i}$ is chosen randomly and $\mathbf{x}$ is functionally related to $\mathbf{q}$ instead of being chosen randomly as in Eq. (2). The expectation values in Eq. (14) remain the same except Eq. (15) becomes

$$
\begin{equation*}
F=f(\mathbf{x}(\mathbf{q})) \psi_{1}(\mathbf{x}(\mathbf{q})) / \bar{w}(\mathbf{q}), \quad G=\psi_{1}^{2}(\mathbf{x}(\mathbf{q})) / \bar{w}(\mathbf{q}) \tag{18}
\end{equation*}
$$

where $\bar{w}(\mathbf{q})$ is given explicitly for $\mathrm{H}_{2}^{+}$in Eq. (28).
The fact that for expectation values the standard deviation is proportional to $\left\langle\left(F_{\mathrm{A}}-\langle f\rangle G_{\mathrm{A}}\right)^{2}\right\rangle$ introduces a slight complication in the method, an estimate of $\langle f\rangle$ enters into the function being extremized. A crude cstimate of $\langle f\rangle$ needs to be found before the first level, then subsequently the estimate of $\langle f\rangle$ from the last completed level can be used.

## Results

Expectation values of the energy $\left\langle H \psi_{\mathrm{t}} \mid \psi_{\mathrm{t}}\right\rangle$, the kinetic energy $\left.\langle T\rangle=\langle | \nabla \psi /\left.\psi\right|^{2}\right\rangle$, and the potential energy $\langle V\rangle$ were estimated for $H_{2}^{+}$and spinaligned $\mathrm{H}_{2}$ with nuclei $7.8 a_{\mathrm{B}}$ apart using the trial wave functions described in Appendix C and the biased selection method of Appendix B combined with pairing of the randomly chosen $\mathbf{q}$ vectors. The $\mathbf{q}$ vectors for $\mathbf{H}_{2}^{+}$are four dimensional, three electron coordinates plus one extra from the biased selection method. The $\mathbf{q}$ vectors for $\mathrm{H}_{2}$ are eight dimensional, six electron coordinates plus two extra from the biased selection method. Though in principle the $\mathrm{H}_{2}^{+}$integrals can be reduced to two dimensional and the $\mathrm{H}_{2}$ integrals to five dimensional, owing to the fact that Monte-Carlo methods are very insensitive to such reductions no attempts were made in this direction.
Table I and Fig. 2 give the results for $\mathrm{H}_{2}^{+}$from two independent runs each involving slightly more than $4 \times 10^{6}$ wave function evaluations. The expectation value of $\langle V\rangle$ for both runs was calculated using Eqs. (13) and (14) and the standard deviation was calculated using Eq. (16). The $M$ in Eq. (13) was 1 for levels - 1 and 0 and $2^{l}$ for higher levels. The first run used $N=1024$ in Eq. (13) and the final $M$ was $2^{11}$, and the second run used $N=64$ in Eq. (13) and the final $M$ was $2^{15}$. Note that $N_{\mathrm{e}}$ is $N$ times the sum of all $M_{l}$ up to and including the last level. At each level the standard deviation was calculated using Eq. (16) with $N=1024$ for the first pass and $N=64$ for the second pass. Since for large $M$ the functions $F_{\mathrm{A}}$ and $\mathrm{G}_{\mathrm{A}}$ are themselves almost integral estimates, this is roughly equivalent to evaluating these integrals 1024 times for $N=1024$ and 64 times for $N=64$ and determining the standard deviations from the resulting variations. For small $M$, using $N=64$ may result in too few function evaluations to adequately sample the exceptionally large and small regions resulting in standard deviation estimates with rather large standard deviations. As $M$ increases, however, $F_{\mathrm{A}}$ and $G_{\mathrm{A}}$ rapidly become very smooth owing to the fact that such regions are by design becoming very unlikely to be missed. The final estimates incorporated the results of all levels by utilizing Eq. (22) for the expectation values and Eq. (23) for their standard deviations.
Since the expensive part of the calculation is finding $\mathbf{x}(\mathbf{q})$ and evaluating $\psi(\mathbf{x})$, in

TABLE I
Results for $\mathrm{H}_{2}^{+}$( $4.19 \times 10^{6}$ Wave Function Evaluations)

|  | $\langle V\rangle$ | $\left.\left(\langle T\rangle=\langle \| \nabla \psi /\left.\psi\right\|^{2}\right\rangle\right)$ | $\langle H \psi / \psi\rangle$ <br> $(\mathrm{Ry})$ | $(\langle H \psi / \psi\rangle-\langle V\rangle-\langle T\rangle)$ <br> $N$ |
| :---: | :---: | :---: | :---: | :---: |
| $(\mathrm{Ry})$ | 0.969459 | -1.00600545 | $-6.045 \times 10^{-5}$ |  |
| 1024 | -1.975404 | $\pm 0.000043$ | $\pm 0.00000008$ | $\pm 8.045 \times 10^{-5}$ |
|  | $\pm 0.000068$ | 0.969446 | -1.00600543 | $+3.66 \times 10^{-5}$ |
| 64 | -1.975488 | $\pm 0.000022$ | $\pm 0.00000004$ | $\pm 5.28 \times 10^{-5}$ |
|  |  |  |  |  |



Fig. 2. Standard deviation in $\langle V\rangle$ for $\mathrm{H}_{2}{ }^{-}$versus the number of wave function evaluations (pairing with respect to $V$ ). (-) $\sigma_{\mathrm{Mc}}$ without pairing. ( -- ) Line for which $\sigma_{\mathrm{A}} \times N_{r}^{-0.72}$. (+) Results using $N=1024$ in Eq. (16). (o) Results using $N=64$ in Eq. (16).
addition to $\langle V\rangle$, expectation values of $\langle T\rangle=\left\langle\mid \nabla \psi i \psi^{2}\right\rangle$ and of $\langle H \psi / \psi\rangle$ were estimated in the same runs. However, in the term $F_{\mathrm{A}}-\langle f\rangle G_{\mathrm{A}}$, whose extremal values determine the $\mathbf{q}$ 's which are to be paired, $f$ was kept equal to $V$. The final results of the two runs are given in Table I along with the very satisfying fact that to within the error $\langle H \psi / \psi\rangle=\langle T\rangle+\langle V\rangle$. This last is an important test of the integration since an integration by parts has removed the singularity in $T$ which otherwise nearly cancels the singularity in $V$ to give an $H \psi / \psi$ with almost no singularity. Thus $\langle T\rangle,\langle V\rangle$, and $\langle H \psi / \psi\rangle$ emphasize somewhat different regions of space and different properties of $\psi$ and will normally not be equal if there are errors either in calculating the derivatives of $\psi$ or in evaluating the integrals. Indeed the ability to make this test is one reason for wanting an accurate $\langle V\rangle$.

Figure 2 shows the standard deviation in $\langle V\rangle$ for $\mathrm{H}_{2}{ }^{+}$as a function of the number of wave function evaluations. The second to the last point plotted required 700 CPU seconds on an IBM 3081, and the last point 1400 CPU seconds. The standard deviation is decreasing as $N_{e}^{-0.72}$. If the desired quantity were $\langle V\rangle$ with a standard deviation less than $10{ }^{4} \mathrm{Ry}$, simple Monte-Carlo would require nearly 100 times as many CPU seconds to reach the desired accuracy.
A more detailed examination of the two runs is shown in Fig. 3, where the ratio of the standard deviation per wave function evaluation at the last level calculated to that of the unpaired Monte-Carlo result is plotted both for $\langle V\rangle$ and also for $\langle H \psi \psi \psi$ calculated at the same time using a vectors selected to minimize the standard deviation in $\langle V\rangle$. Note that the abscissa in this plot is the level number $l$, not the number of function evaluations. Even though the standard deviation in the $N=64$ run at, for example, level 7 is much higher than that for the $N=1024$ run, the number of function evaluations to find the $N=64$ result is so much smaller that, as seen in Fig. 2, the $N=64$ result has the smaller total standard deviation. The standard deviation for $\langle H \psi / \psi\rangle$ and also for $\langle T\rangle$ (not shown because its ratio


Fig. 3. Ratios $\sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle V\rangle$ and $\langle H \psi / \psi\rangle$ in $\mathrm{H}_{2}^{+}$versus level when pairing is with respect to $V$. $(--) \sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle H \psi / \psi\rangle .(--) \sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle V\rangle .(+)$ Values for $N=1024$ in Eq. (16). (O) Values for $N=64$ in Eq. (16).
to $\sigma_{\mathrm{MC}}$ is very nearly the same as that for $\langle H \psi / \psi\rangle$ ) at low levels is constant at the unpaired value, then at about level 5 begins to decrease at nearly the same rate as that in $\langle V\rangle$, indicating that the set of a's being constructed in this case lowers the standard deviations for all three quantities.

If one only wanted to know $\langle H \psi / \psi\rangle$, one would use $H \psi / \psi$ as the $f$ in $F_{\mathrm{A}}-\langle f\rangle G_{\mathrm{A}}$ not $V$. The results of an $N=400$ run doing this are shown in Fig. 4. The standard deviation in $\langle H \psi / \psi\rangle$, especially for levels less than 5 , is less than when the pairing was carried out with respect to $V$. The standard deviation in $\langle V\rangle$ using these a's, however, is larger and moreover appears to be decreasing for large levels as $N_{\mathrm{e}}^{-0.57}$ rather than $N_{\mathrm{e}}^{-0.72}$. The reason for this is probably the constancy of $H \psi / \psi$ for $\mathbf{x}(\mathbf{q})$ 's which cause the potential to vary over wide ranges.


Fig. 4. Ratios $\sigma_{A} / \sigma_{\mathrm{MC}}$ for $\langle V\rangle$ and $\langle H \psi / \psi\rangle$ in $\mathrm{H}_{2}^{+}$versus level when pairing is with respect to $H \psi / \psi$. (--) $\sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle H \psi / \psi\rangle$. ( - ) $\sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle V\rangle$. (O) Values for $\langle H \psi / \psi\rangle, N=400$ in Eq. (16). ( + ) Values for $\langle V\rangle, N=400$ (same run).


Fig. 5. Standard deviation in $\langle V\rangle$ for spin-aligned $\mathrm{H}_{2}$ versus the number of wave function evaluations (pairing with respect to $V$ ). (- -) $\sigma_{\text {mc }}$ without pairing. ( $-\cdots$ ) Line for which $\sigma_{A} \times N_{\mathrm{v}}{ }^{0.62}$. ( $\times$ ) Results using $N=1024$ in Eq. (16). (O) Results using $N=64$ in Eq. (16).

The $\mathrm{H}_{2}^{+}$calculations were made as four-dimensional calculations, but there is an underlying two-dimensional structure. The pairing method in part is simply uncovering this structure and utilizing it to achieve a relatively slow, by two-dimensional standards, convergence rate which is still faster than unpaired Monte-Carlo. The spin-aligned $\mathrm{H}_{2}$ system which is treated here as an eight-dimensional integral cannot be reduced below five dimensions and is therefore a much more stringent test of the method.

The standard deviation in $\langle V\rangle$ for spin-aligned $\mathrm{H}_{2}$ as a function of the number of wave function evaluations is shown in Fig. 5. Approximately $2 \frac{1}{2}$ hours of IBM 3081 CPU time were required to reach the last $x$ in this figure. The standard deviation $\sigma_{\mathrm{A}}$ is decreasing as $N_{\mathrm{e}}^{-0.62}$ and is approximately half the size of $\sigma_{\mathrm{Mc}}$ at the highest numbers of function evaluations considered. The results of these calculations are shown in Table II. If the goal were to find $\langle V\rangle$ with $\sigma_{\mathrm{A}}$ less than $10^{4} \mathrm{Ry}$ approximately $9 \times 10^{7}$ function evaluations would be required assuming

TABLE II
Results for $\mathrm{H}_{2}$

| N | Ne | $\begin{aligned} & \langle V\rangle \\ & (\mathrm{Ry}) \end{aligned}$ | $\begin{gathered} (\langle T\rangle:= \\ \left.\left.\left.\langle \| \mathbf{V} \psi i \psi\right\|^{2}\right\rangle\right) \\ (\mathrm{Ry}) \end{gathered}$ | $\begin{gathered} \langle H \dot{\psi} ; \dot{\psi}\rangle \\ (\mathrm{Ry}) \end{gathered}$ | $\begin{gathered} (\langle H \psi i \psi\rangle- \\ \langle V\rangle-\langle T\rangle) \\ (R y) \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 1024 | $4.2 \times 10^{6}$ | -4.00002 | 2.0000873 | -2.0000364 | $+1.0 \times 10^{4}$ |
|  |  | $\because 0.00070$ | $\pm 0.0000078$ | $\pm 0.0000012$ | $\pm 7.0 \times 10^{-4}$ |
| 1024 | $2.1 \times 10^{6}$ | -3.99874 | 2.000109 | - 2.0000366 | $+14.1 \times 10^{-4}$ |
|  |  | $\pm 0.00095$ | $\pm 0.000012$ | $\pm 0.0000017$ | $\pm 9.5 \times 10^{4}$ |
| 64 | $2.1 \times 10^{6}$ | -4.00139 | 2.0000945 | -2.0000378 | $-12.6 \times 10^{-}=$ |
|  |  | $\pm 0.00113$ | $\pm 0.0000091$ | $\pm 0.0000014$ | $\pm 11.3 \times 10^{-4}$ |



Fig. 6. Ratio $\sigma_{A} / \sigma_{M C}$ for $\langle V\rangle$ for spin-aligned $\mathrm{H}_{2}$ when pairing is with respect to $V .(+, \times)$ Two independent runs with $N=1024$ in Eq. (16). (O) $N=64$ in Eq. (16).
the continuance of this convergence rate. This is less than the number required to reduce $\sigma_{\mathrm{MC}}$ to this size by a factor of 8.5 .

In Fig. 6, the ratio of $\sigma_{\mathrm{A}}$ as given by Fq: (16) to $\sigma_{\mathrm{MC}}$ the unpaired result is shown for two $N=1024$ runs along with one $N=64$ run. The $N=64$ run is not only higher than the $N=1024$ runs, it is so much higher that the factor of 32 or 5 level advantage in the number of function evaluations is almost exactly canceled, leaving the $N=1024$ and $N=64 \sigma_{\mathrm{A}}$ 's almost equal as already seen in Fig. 5. An interpretation of this is that the $N$ evaluations must contain rather rare events to properly pair them. Note, however, that the ratio with $N=64$ drops for the first four levels then fluctuates but remains approximately constant for the next five levels. This implies the existence of gross features which pairing can account for relatively early coupled with finer details which require many more function evaluations.
If one is solely interested in the expectation value of $\langle H \psi / \psi\rangle, H \psi / \psi$ should be used as the $f$ in the $F_{\mathrm{A}}-\langle f\rangle G_{\mathrm{A}}$, which determines the locations paired rather than $V$. An $N=1024$ run of spin-aligned $\mathrm{H}_{2}$ carried out to level 9 in which this was done is shown in Fig. 7. The ratio $\sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ decreases steadily from 1.0 at level 0 to 0.9 at level 9 , indicating a slightly improved convergence rate of $N_{\mathrm{e}}^{-0.51}$. In the same run the ratio of $\sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle V\rangle$ using a vectors selected to improve $\sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle H \psi / \psi\rangle$ increases rather dramatically to a maximum of 2.76 at level 8 . At level 8 , $2.6 \times 10^{6}$ wave function evaluations yield the same accuracy in $\langle H \psi / \psi\rangle$ as $3.2 \times 10^{6}$


Fig. 7. Ratios $\sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle V\rangle$ and $\langle H \psi / \psi\rangle$ for spin-aligned $\mathbf{H}_{2}$ when pairing is with respect to $\langle H \psi / \psi\rangle, N=1024$ in Eq. (16). (O) $\sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle V\rangle$. (×) $\sigma_{\mathrm{A}} / \sigma_{\mathrm{MC}}$ for $\langle H \psi / \psi\rangle$.
unpaired wave function evaluations and also the same accuracy in $\langle V\rangle$ as $3.4 \times 10^{5}$ unpaired wave function evaluations. This comes about because $(H \psi / \psi-E) \psi^{2}$ has both its largest positive and negative values in the same regions where $(V-\langle V\rangle) \psi^{2}$ is all of one sign. Thus the values that cancel to shrink $\sigma_{\mathrm{A}}$ for $H \psi i \psi$ add to enlarge $\sigma_{\Lambda}$ for $V$.

## Conclusion

The pairing method constructs vectors $\mathbf{a}_{j}$ containing the coordinate values giving rise to the most extreme values of the $F$ defined in Eq. (18). These vectors contain information about the coordinate values giving rise to $F$ which is usually ignored in Monte-Carlo schemes for estimating integrals. Only a slight modification of MonteCarlo, i.e., the replacement of $F\left(x_{i}\right)$ by $F_{\mathrm{A}}\left(x_{i}\right)$ as defined in Eq . (13), is required. The overhead is so small as to be almost undetectable while the accuracy gains can be large. In summary, pairing should be considered when truly accurate expectation values are needed and the number of function evaluations is large compared with the expected oscillations in the function being evaluated. Accuracy increases of a factor of 10 or computer time savings of a factor of 100 have been shown in the present work.

## Appendix A: Error Estimation

The statistical error which results from estimating the expectation value, $E$, using $N$ random values can be determined, as explained in [4], by summing the squares of the differences between the $N$ evaluations of $E$ over $N-1$ values which result from omitting each $x_{i}$ in turn. The difference in the estimate due to the omission of the $i$ th term is

$$
\delta_{i}=E-\frac{\left[\langle F\rangle-(1 / N) F_{\mathrm{A}}\left(x_{i}\right)\right]}{\left[\langle G\rangle-(1 / N) G_{\mathrm{A}}\left(x_{i}\right)\right]}
$$

which noting that $E=\langle F\rangle /\langle G\rangle$ and combining terms becomes

$$
\delta_{i}=\frac{1}{N} \frac{\left[F_{\mathrm{A}}\left(x_{i}\right)-E G_{\mathrm{A}}\left(x_{i}\right)\right]}{\left[(G)-(1 / G) G_{\mathrm{A}}\left(x_{i}\right)\right]}
$$

where $E, F, G, F_{A}\left(x_{i}\right)$, and $G_{A}\left(x_{i}\right)$ are defined in the second section. Since $G_{\mathrm{A}}\left(x_{i}\right) \ll N\{G\}$, we approximate $1-G_{\mathrm{A}}\left(x_{i}\right) / N\{G\}$ by 1 and simplify to obtain

$$
\begin{equation*}
\delta_{i}=\frac{1}{N\langle G\rangle}\left[F_{\mathrm{A}}\left(x_{i}\right)-E G_{\mathrm{A}}\left(x_{i}\right)\right] \tag{19}
\end{equation*}
$$

so that the variance is given by

$$
\begin{equation*}
\sigma^{2}=\sum_{i=1}^{N} \sigma_{i}^{2}=\frac{1}{N\langle G\rangle^{2}}\left[\left\langle F^{2}\right\rangle-2 E\langle F G\rangle+E^{2}\left\langle G^{2}\right\rangle\right] . \tag{20}
\end{equation*}
$$

Note that, as in Eq. (5), $M$ wave function evaluations are required for each of the $N$ values of $F_{\mathrm{A}}\left(x_{i}\right)$ and $G_{A}\left(x_{i}\right)$ for a total of $N_{\mathrm{e}}=N * M$ wave function evaluations to determine $E$. The variations of the $N$ values of $F_{\mathrm{A}}$ and $G_{\mathrm{A}}$ which are statistically independent are then used by Eq. (20) to determine $\sigma$.

The average over estimates with varying accuracies is done in the standard fashion [5] which minimizes $\chi^{2}$ with respect to the estimate of $E$ :

$$
\begin{equation*}
\frac{\hat{C}\left(\chi^{2}\right)}{\hat{\partial} E}=0=2 \sum_{i=1}^{L}\left(\frac{E-E_{i}}{\sigma_{i}^{2}}\right) \tag{21}
\end{equation*}
$$

Solving the above expression for $E$ yields

$$
\begin{equation*}
E=\sum_{i=1}^{L}\left(\frac{E_{i}}{\sigma_{i}^{2}}\right) / \sum_{i=1}^{N}\left(\frac{1}{\sigma_{i}^{2}}\right) \tag{22}
\end{equation*}
$$

so the weight factor for each estimate is equal to $1 / \sigma_{i}^{2}$. The $\sigma^{2}$ at each stage is given by

$$
\begin{equation*}
\frac{1}{\sigma^{2}}=\sum_{i=1}^{1} \frac{1}{\sigma_{i}^{2}} \tag{23}
\end{equation*}
$$

A word of caution is required here in case an estimate results with an artificially low variance. The incorrect low estimate could dominate both Eqs. (22) and (23) normally giving rise to both an incorrect $E$ and $\sigma$. This case is detectable owing to the incorrect $E$ which causes $\chi^{2}$ to exceed its normal value. The case where the $E_{i}$ associated with the incorrect $\sigma_{i}$ is not itself in error is much harder to detect. For $N$ sufficiently large, one need not worry about this situation because the error in the variance converges as $1 / N$, but one does need to be aware of this possibility in the early levels. We checked our results over the initial stages by performing independent evaluations and comparing the results for consistency at each of the levels.

## APPENDIX B: Thf Biased Selection Mrthod

The biased selection method, as used here, is a specialized form of information sampling involving two centers. Multidimensional points, $q_{i}$, are selected with relative probabilities $w_{i}$ to ensure that all regions of importance are sampled. This method is adapted from the work done by R. L. Coldwell and R. E. Lowther [6].

The $\mathrm{H}_{2}^{+}$ion is positioned in a box of maximum side length of 12.8 Bohr radii. The axis which contains the two nuclei is along the longest diagonal possible
between two corners of the box. The arrangement does not make use of the radial symmetry of the $\mathrm{H}_{2}^{+}$ion. We chose to do this to test whether the code could utilize this information without it being explicitly written in the code. We felt that this situation is more in line with what we would expect the code to do for more complex systems, where underlying symmetries are not known.

The electronic coordinates are selected with respect to one of the nuclei. Since the two nuclei are identical, the nucleus used to specify the placement of the electron is a function of a random value, $q_{1}$, where nucleus A is used for $q_{1} \leqslant 0.5$, and nucleus B for $q_{1} \geqslant 0.5$.

The electron distance from the nucleus, $r$, is selected such that

$$
\begin{equation*}
q_{2}=\int_{0}^{r} h\left(r^{\prime}\right) d r^{\prime} \int_{0}^{1} h\left(r^{\prime}\right) d r^{\prime} \tag{24}
\end{equation*}
$$

i.e., with a density proportional to $h(r)$. The function $h(r)$ is obtained by doing a linear interpolation between 64 points spaced $0.2 a_{\mathrm{B}}$ apart of the function

$$
\begin{equation*}
g(r)=r^{2} \sum_{i=1}^{3} x_{i}^{3} e^{-x_{i} r} \tag{25}
\end{equation*}
$$

with $x==\{4,2,0.5\}$. The first point $h(0)$ is set equal to $h(0.2)$ so that the resulting weight function near the nuclei contains an $r^{2}$ that more than concels the $1 / r$ in the potential. Values for $\phi$ and $\mu=\cos \theta$ are specified by $\phi=2 \pi q_{3}$ and $\mu=2 q_{4}-1$, so that the cartesian coordinates of the electron are

$$
\begin{align*}
x\left(q_{1}, q_{2}, q_{3}, q_{4}\right) & =r\left(q_{2}\right) \sin \left[\phi\left(q_{3}\right)\right] \sqrt{1-\left[\mu\left(q_{4}\right)\right]^{2}}+X\left(q_{1}\right) \\
y\left(q_{1}, q_{2}, q_{3}, q_{4}\right) & =r\left(q_{2}\right) \cos \left[\phi\left(q_{3}\right)\right] \sqrt{1-\left[\mu\left(q_{4}\right)\right]^{2}}+Y\left(q_{1}\right)  \tag{26}\\
z\left(q_{1}, q_{2}, q_{3}, q_{4}\right) & =r\left(q_{2}\right) \mu\left(q_{4}\right)+Z\left(q_{1}\right) . \\
X\left(q_{1}\right) & =Y\left(q_{1}\right)=Z\left(q_{1}\right)= \pm r_{\mathrm{AB}} / 2 \sqrt{3}
\end{align*}
$$

where the plus sign is used with one nucleus and the minus sign is used with the other and $r_{A B}$ is the distance between the two nuclei expressed in Bohr radii.

The relative probability of choosing this position in cartesian coordinates from all possible positions is

$$
\begin{equation*}
\omega(\mathbf{q})=\frac{h(r)}{4 \pi r^{2}} \tag{27}
\end{equation*}
$$

Since the position of the electron may be chosen with respect to either of the two nuclei, the average probability for selecting positions is

$$
\begin{equation*}
\bar{\omega}(\mathbf{q})=\frac{1}{2}\left[\frac{h\left(r_{\mathrm{wA}}\right)}{4 \pi r_{\mathrm{A}}^{2}}+\frac{h\left(r_{\mathrm{B}}\right)}{4 \pi r_{\mathrm{B}}^{2}}\right] \tag{28}
\end{equation*}
$$

where the subscripts $A$ and $B$ indicate which of the two nuclei is being used for the placement of the electron.

Using the biased selection method in this manner allows an integral to be written

$$
\begin{equation*}
I=\int_{0}^{1} f(\mathbf{x}) d \mathbf{x}=\int_{0}^{1}\left\{\frac{f[x(\mathbf{q})]}{\bar{\omega}[x(\mathbf{q})]}\right\} d \mathbf{q} \tag{29}
\end{equation*}
$$

The MC estimate of this integral is then $\langle f / \omega\rangle$ and its variance is that of $(f / \bar{\omega})$. Note that $f[x(q)] / w[x(q)]$ will tend to be constant, so that when a Monte-Carlo estimate of Eq. (29) is obtained, the variance between individual evaluations of the integrand will be small which will reduce the inherent Monte-Carlo error.

The biased selection procedure for spin-aligned $\mathrm{H}_{2}$ uses an $h(r)$ in which only $\alpha=2$ and allows for a $7 \%$ probability of placing each electron completely at random within the large box, a $3 \%$ probability of placing both electrons with respect to the same nucleus, and a $12 \%$ probability of placing the second electron at a random distance between 0 and $1.3 a_{\mathrm{B}}$ from the first. This last probability removes the singularity in the electron electron part of the potential. The resulting $w(q)$ is of course averaged over every possible way of reaching the final configuration exactly as in Eq. (28) above.

## APPENDIX C: The Trial. Wave Functions

The pairing method was tested on two systems consisting of two protons, $A$ and $\mathrm{B}, 7.8 \mathrm{Bohr}$ radii apart. The first system, $\mathrm{H}_{2}^{+}$, has only one electron, while the second, spin-aligned $\mathrm{H}_{2}$, has two spin-up electrons. This nuclear separation distance is approximately the potential minimum of the ${ }^{3} \Sigma_{g}^{+}$state of $\mathrm{H}_{2}$, spin-aligned hydrogen. The onc-clectron wave function is

$$
\begin{equation*}
\psi=\chi\left(r_{1 \mathrm{~A}}+r_{1 \mathrm{~B}}\right) \phi\left(r_{1 \mathrm{~A}}-r_{1 \mathrm{~B}}\right) \tag{30}
\end{equation*}
$$

with

$$
\begin{equation*}
\chi(x)=e^{-(1 / 2) x} \sum_{i=0}^{4} a_{i} x^{i} \tag{31}
\end{equation*}
$$

and

$$
\begin{equation*}
\phi(x)=\left(e^{(1 / 2) x}+e^{-(1 / 2) x}\right) \sum_{i=0}^{2} a_{(i+5)} x^{2 i}+\left(e^{(1 / 2) x}-e^{-(1 / 2) x}\right) x \sum_{i=0}^{1} a_{i+8} x^{2 i+1} \tag{32}
\end{equation*}
$$

where the parameters given in Table III were found by minimizing

$$
\begin{equation*}
\sigma^{2}=N\left\{\sum_{i=1}^{N} \psi_{i}^{4} / \omega_{i}^{2}\left(H \psi_{i /} \psi_{i}-\langle E\rangle\right)^{2}\right\} /\left\{\sum_{i=1}^{N} \psi_{i}^{2} / \omega_{i}\right\}^{2} \tag{33}
\end{equation*}
$$

TABLE III
The Coefficients for $\gamma(x)$ and $\phi(x)$ in Eq. (31) and Eq. (32)

| $i$ | 0 | 1 | 2 | 3 | 4 |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $a_{i}$ | 1.0 | $-9.54565 \times 10^{3}$ | $-1.75124 \times 10^{3}$ | $1.06063 \times 10^{4}$ | $-3.38318 \times 10^{6}$ |  |
| $i$ | 5 | 6 | 7 | 8 | 9 |  |
| $a_{i}$ | 1.0 |  |  |  | $1.07657 \times 10^{5}$ |  |

with respect to these parameters for $N=1000$. The $\omega_{i}$ refers to the bias used in selecting the $N$ configurations as described in Appendix B. The $\sqrt{N \sigma}$ found is $2.8 \times 10^{-4} \mathrm{Ry}$ with $\langle E\rangle=-1.00600545 \pm 8 \times 10^{8} \mathrm{Ry}$ differing from that found by an exact solution, which is possible for $\mathrm{H}_{2}^{+}$, by less than the final standard deviation. (The code which found the exact result of -1.0060549 Ry is due to Hyun-Joon Shin [7] using the methods of Barber and Hasse [8, 9].)

The spin-aligned $\mathrm{H}_{2}$ wave function is

$$
\begin{align*}
\psi= & A_{\mathrm{A}, \mathrm{~B}: 1,9}\left\{e ^ { - r _ { 1 \mathrm { A } } - r _ { 2 \mathrm { B } } } \left[\frac{1}{2}+v_{p}\left(r_{1 \mathrm{~B}}, r_{12}\right)\left(a_{1}+a_{2}\left(r_{\mathrm{i} 2}-r_{\mathrm{AB}}\right)\right.\right.\right. \\
& \left.\left.\left.+a_{3} r_{1 \mathrm{~A}}+a_{4} r_{2 \mathrm{~B}}+a_{5} r_{1 \mathrm{~A}} r_{2 \mathrm{~B}}+a_{6} r_{2 \mathrm{~B}}^{2}+a_{7}\left(r_{12}-r_{\mathrm{AB}}\right)^{2}\right)\right]\right\} \tag{34}
\end{align*}
$$

wherc $A_{\mathrm{A}, \mathrm{B} .1,2}$ is an operator which antisymmetrizes with respect to nuclei A and B and electrons 1 and 2 . The term $v_{p}$ is given by

$$
\begin{equation*}
v_{p}\left(r_{1 \mathrm{~B}}, r_{12}\right)=s\left(r_{12}\right) / r_{12}-2 s\left(r_{1 \mathrm{~B}}\right) / r_{1 \mathrm{~B}}+1 / r_{\mathrm{AB}} \tag{35}
\end{equation*}
$$

with

$$
\begin{equation*}
s(x)=1-4(1-x)_{+}^{3} / 3+8(0.5-x)_{+}^{3} / 3 \tag{36}
\end{equation*}
$$

where

$$
\begin{array}{rlrl}
(x)_{+} & =0, & x<0  \tag{37}\\
& =x, & & x>0 .
\end{array}
$$

This form is suggested by the fact that the Hamiltonian for two separated atoms can be written as

$$
\begin{align*}
H= & -\nabla_{1}^{2}-2 / r_{1 \mathrm{~A}}+\left(1 / r_{12}-2 / r_{1 \mathrm{~B}}+1 / r_{\mathrm{AB}}\right)  \tag{38}\\
& -\nabla_{2}^{2}-2 / r_{2 \mathrm{~B}}+\left(1 / r_{12}-2 / r_{2 \mathrm{~A}}+1 / r_{\mathrm{AB}}\right)
\end{align*}
$$

where the terms in parentheses are small for electron 1 close to nucleus A and electron 2 close to nucleus $\mathbf{B}$. The splines in $v_{p}$ remove the singularities while leaving $v_{p}$ identical to the term in parentheses for $r_{12}$ and $r_{1 \mathrm{~B}}$ on the order of $7.8 a_{\mathrm{B}}$.

The parameters $a_{1}$ through $a_{7}$ were found in two stages. First $\sigma^{2}$ was minimized with respect to the $a$ 's using $N=400$. The resulting $\psi$ had

TABLE IV
The Cueflicients in $\psi\left(\mathbf{r}_{1}, \mathbf{r}_{2}\right)$ (Eq. (34))

| $i$ | 1 | 2 | 3 | 4 |
| :--- | :--- | :--- | :--- | :--- |
| $a_{i}$ | -3.024050 | $-3.872121 \times 10^{-2}$ | $-4.98549 \times 10^{-2}$ | $-3.965802 \times 10^{-2}$ |
| $i$ | 5 | 6 | 7 |  |
| $a_{i}$ | $-1.501606 \times 10^{-2}$ | $5.380148 \times 10^{-3}$ | $3.331075 \times 10^{-3}$ |  |

$\langle E\rangle=-2+(-3.36 \pm 0.17) \times 10^{-5} \mathrm{Ry}$ and $\sigma \approx 1.4 \times 10^{-3} \mathrm{Ry}$. This $\psi$ was used to find a set of 256 configurations using the pairing method, plus another 144 "worst" set of configuration picked from 28,000 evaluations of $H \psi / \psi$. These configurations provide a capsulized picture of those areas where $\psi$ is not as good as it could be and allow more minimization of $E$ over a small set of points than would otherwise be possible. Then $\langle E\rangle+25 \sigma$ was minimized with respect to these 400 configurations. The resulting parameters are given in Table IV. For this $\psi$, $\sqrt{N \sigma}=2.54 \times 10^{-3} \mathrm{Ry}$ and when finally evaluated in the process of this work $\langle E\rangle=-2+(-3.65 \pm 0.10) \times 10^{-5} \mathrm{Ry}$, which should be compared with the value of $-4.09 \times 10^{-5}$ Ry calculated by Kolos and Wolniewicz [10] using a 60parameter wave function.

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