

# Variational Monte Carlo

FK8029 - Computational Physics

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# 1 Introduction

In the last report, we solved the time independent Schrödinger equation using two methods, in order to obtain the eigenstates of the system. This method works for larger systems too, however, the computational cost increases drastically with the number of particles and dimension. Therefore, certain methods have been developed in order to solve such equations with less computational cost. One such way is the use of variational Monte Carlo.

In this report, we solve the time independent Schrödinger equation for the harmonic oscillator using the variational Monte Carlo method. Furthermore, to see the potential of the method, we also solve the system for a more complex system – quantum dots.

## Contents

<b>1</b>	<b>Introduction</b>	<b>1</b>
<b>2</b>	<b>Theory &amp; Method</b>	<b>2</b>
<b>3</b>	<b>Result &amp; Discussion</b>	<b>4</b>
3.1	Harmonic Oscillator . . . . .	4
3.2	Quantum Dots . . . . .	6
<b>4</b>	<b>Conclusion</b>	<b>8</b>

## 2 Theory & Method

The time independent Schrödinger equation is given by:

$$\hat{H}|\psi\rangle = E|\psi\rangle,$$

where  $|\psi\rangle$  is the state of the system, in this case, the wave-function. The Hamiltonian  $\hat{H}$  is system specific, and since we're solving for two systems, we have two different Hamiltonians'. In the first case, we use the Harmonic oscillator, where the Hamiltonian is given by, in reduced units:

$$\hat{H} = -\frac{1}{2} \frac{d}{dz} + \frac{1}{2} z^2, \quad (1)$$

and in the second case, we investigate quantum dots, with the following Hamiltonian, again in reduced units:

$$\hat{H} = \left( \sum_{i=1}^2 -\nabla_i^2 + |\mathbf{x}_i|^2 \right) + \frac{\lambda}{\sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}}. \quad (2)$$

In the above definition,  $\mathbf{x}_i = (x_i, y_i)$  is the position of particle  $i$ , and  $x_i$  and  $y_i$  are the coordinates of the particle. The above Hamiltonian is for two particles, and the sum over the contribution both particles feel from a potential well. The second term is a coupling factor, which couples the two particles to an electromagnetic field. From this, we define the local energy,  $E_{loc}$ , to be the energy obtained by the wave-function at a specific location instead of the entire domain. This is then written as:

$$E_{loc} = \frac{\hat{H}\psi_T}{\psi_T}.$$

Here  $\psi_T$  is the trail wave-function evaluated at a specific point. This however, implies that the trail wave-function is the true wave-function. Therefore, we introduce the variational principle, which states that we which minimize the local energy of the system, in order to find the ground state. We do this by a parameter  $\alpha$ , which we vary in order to find the minimum energy of the system.

$$E_{loc}^\alpha = \frac{\hat{H}\psi_T^\alpha}{\psi_T^\alpha}. \quad (3)$$

Thus, we solve for the ground state of the system by minimizing the local energy,  $E_{loc}$ , with respect to  $\alpha$ .

Monte Carlo methods build upon the idea of random sampling, with some probability weighting. The metropolis algorithm includes a random walk, where every state in the system is reachable from any other state, from a condition called detailed balance. The algorithm is as follows:

1. Pick a random atom
2. Propose a move
3. Calculate the probability of the move,  $p$
4. Accept or reject the move depending on  $\min(1, p)$
5. Update the system (if move was accepted, otherwise keep the system as it was).

These steps are then repeated for a number of iterations, until the system has reached equilibrium, e.g. when the system is ergodic. Furthermore, we also sample the systems observable, such as

the energy, in order to calculate the expectation value of the system, as well as its variance. The probability of the move,  $p$  is given by the ratio of the probabilities of the trail wave-functions,  $\psi_T$ :

$$p = \left| \frac{\psi_T^\alpha(\mathbf{X}')}{\psi_T^\alpha(\mathbf{X})} \right|^2, \quad (4)$$

where  $\mathbf{X}'$  is the new configuration, and  $\mathbf{X}$  is the old configuration.

In order to minimize the local energy, we could use various methods, such as golden search or steepest descent. In this report, we use golden search which is a method that finds the minimum of domain given four points. Suppose we have a potential given by the figure below, fig 1. If we define two points  $a$  and  $d$ , where there exists a single local minimum in the interval  $[a, d]$ , then we can define two points  $b$  and  $c$ , which constitute a sub-interval  $[b, c]$  in  $[a, d]$ . Then we compute the value as then we compute the value  $f(b)$  and  $f(c)$ , and if  $f(b) < f(c)$ , then we set  $d = c$ , otherwise we set  $a = b$ . We continue doing this until the difference between  $b$  and  $c$  is small enough such that we have found the point of the minima.

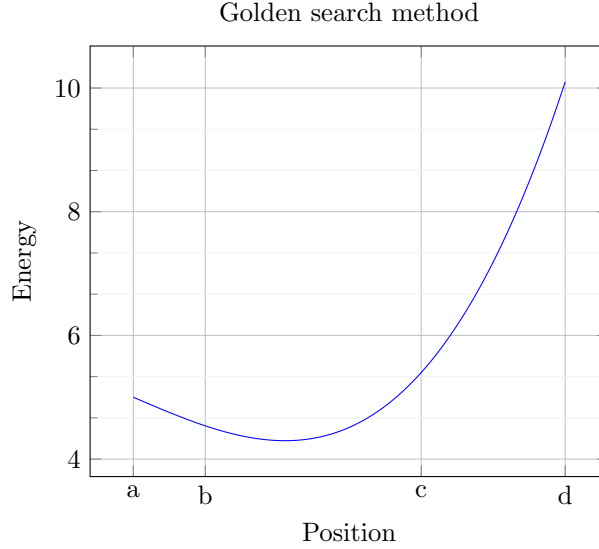


Figure 1: Visualization of the golden search method.

We use this method to find the local minimum of the observables in the system, where we minimize the free parameter  $\alpha$ , defined in eq (3). The observables,  $\langle E \rangle$  and  $\sigma(E)$ , are then calculated by the following equations:

$$\langle E \rangle = \frac{1}{N} \sum_{i=1}^N E_{loc}^\alpha, \quad (5)$$

$$\sigma(E) = \left( \frac{1}{N} \sum_{i=1}^N (E_{loc}^\alpha)^2 - \langle E \rangle^2 \right)^{\frac{1}{2}}, \quad (6)$$

where  $N$  is the number of samples taken.

### 3 Result & Discussion

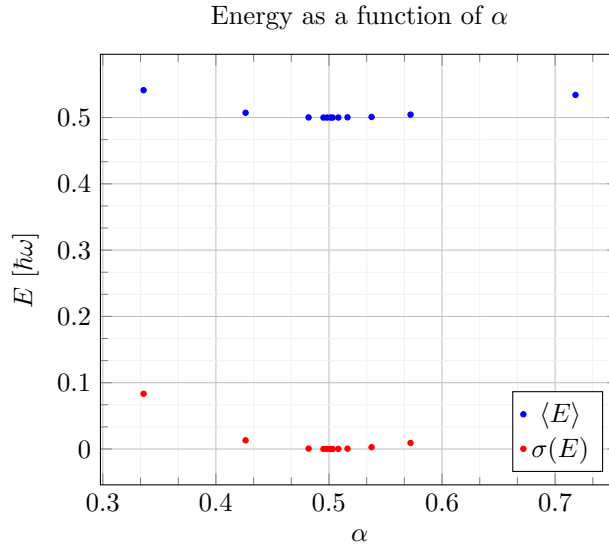
This section is dedicated to the presentation of the results obtained from the variational Monte Carlo simulation, and a discussion of the results. Moreover, the results were obtained simulation written in V.

#### 3.1 Harmonic Oscillator

In this section, we present the results of the variational Monte Carlo method for the harmonic oscillator case, eq 1, in a one-dimensional potential. The ground state of a harmonic oscillator is known to be a Gaussian, and thus we provide the following trial wave-function:

$$\psi_T(x) = \exp\left(-\alpha x^2\right),$$

where  $\alpha$  is the free parameter we wish to minimize. In principle, there is also a normalization constant in front of the expression, but this is neglected, since the minimization of the local energy is independent of the normalization constant. The average energy  $\langle E \rangle$ , and variance  $\sigma(E)$  was computed for a set of  $\alpha$  values, of which were obtained by the golden search algorithm. In doing so, we obtained the following results, fig 2a:

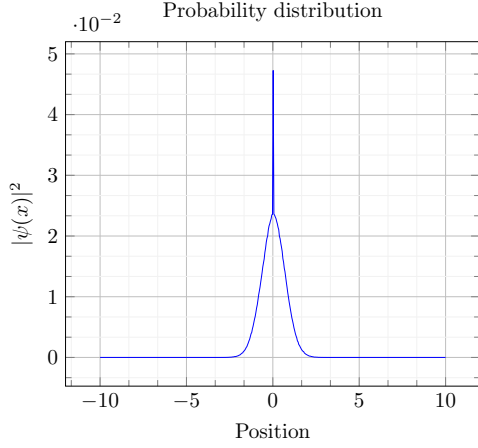


From the above figure, it's clear that the energy is minimized around  $\alpha = 0.5$ , which is the expected value for the harmonic oscillator, since the analytical solution of the ground-state is given by:

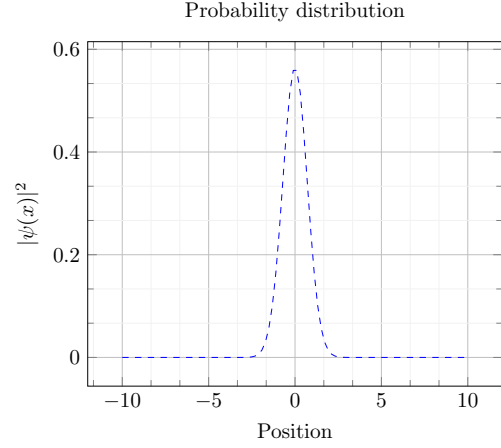
$$\psi(x) = c \exp\left(-\frac{x^2}{2}\right),$$

where  $c$  is a normalization constant. From the above figure, we also see that the average energy  $\langle E \rangle$  has its minima at  $E = \frac{1}{2}\hbar\omega$ , which is expected since the energy of the ground state is given by  $E_0 = (0 + \frac{1}{2})\hbar\omega$ .

If we discretize the domain into  $n$  bins, we can also plot the wave-function obtained from the ground-state, as the number of times a position been visited by the metropolis algorithm. In doing so, we obtain the figure below, fig 2a. The figure shows the ground state of the harmonic oscillator, a Gaussian, which is expected. However, there is a discrepancy between the analytical solution, and the one shown below. At the coordinate  $x = 0$ , the calculated wave-function exhibits



(a) Ground state of the harmonic oscillator, numerical solution



(b) Ground state of the harmonic oscillator, analytical solution

Figure 2: Analytical and numerical solution of the ground state of the harmonic oscillator.

a distinct peak. And in the close neighborhood of  $x = 0$ , the wave-function is not smooth, as the analytical solution. The reason for this is simply due to the discretization of the domain. Since the solution is localized at  $x = 0$  the particle has a very high probability of being located around  $x = 0$ , and thus the number of times a particle is located in the bin located around  $x = 0$  is greater much greater than the surrounding bins. Thus, increasing the number of bins would yield a smoother wave-function, and a more accurate representation of the ground state. The wave-function is normalized, where the normalization constant is given by:

$$\sum_{i=0}^N c_i \psi(x_i) = 1,$$

where  $N$  is the number of bins in the domain, and  $c_i$  is the bin count at position  $x_i$ .

### 3.2 Quantum Dots

In the definition of the Hamiltonian, eq (2), it's visible that we have two particles in a two-dimensional plane. The trial wave-function is defined to be:

$$\psi_T^\alpha(\mathbf{X}_1, \mathbf{X}_2) = \exp \left[ -\frac{1}{2} (|\mathbf{x}_1|^2 + |\mathbf{x}_2|^2) \right] \exp \left[ \frac{\lambda |\mathbf{x}_1 - \mathbf{x}_2|}{1 + \alpha |\mathbf{x}_1 - \mathbf{x}_2|} \right],$$

where  $\mathbf{x}_i$  is the position of particle  $i$ . Hence, we still have a parameter to minimize,  $\alpha$ , but we also have free parameter  $\lambda$  which then is a free parameter characterizing the coupling strength. Therefore,  $\lambda$  was set to equal two, unless otherwise specified. Since it's the ground-state the variance  $\sigma(E)$  is also close to 0 as visible in the figure below, fig 3b.

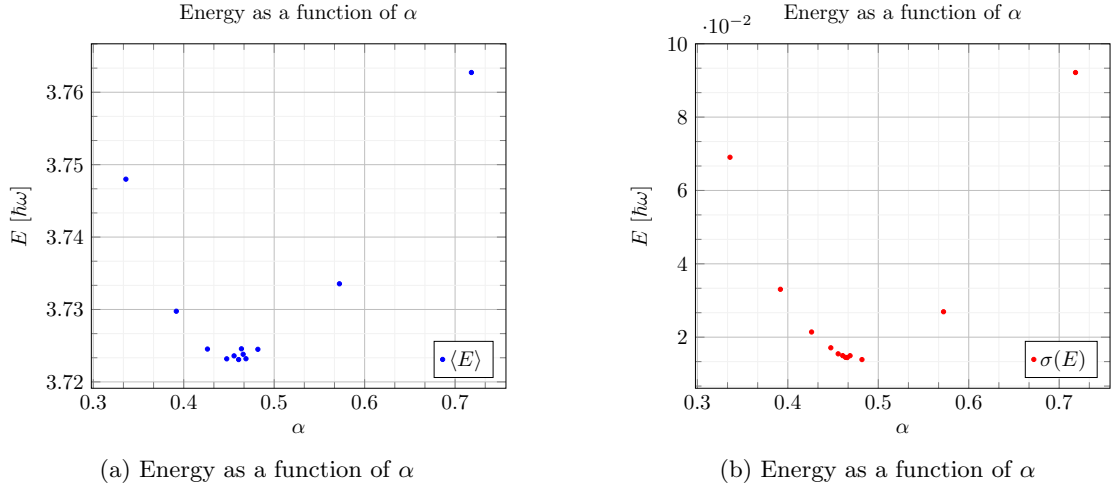
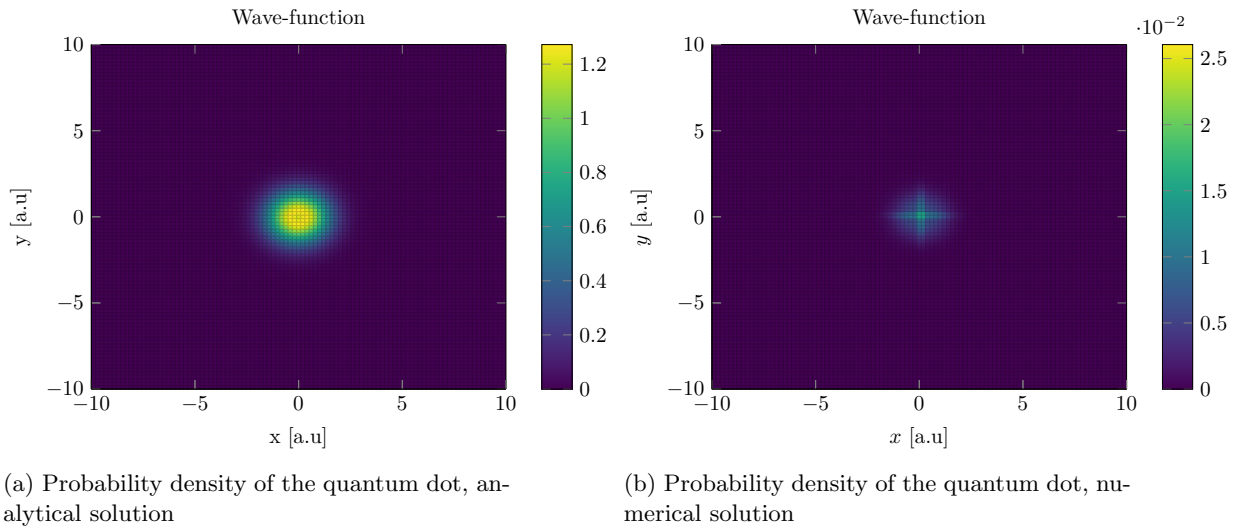


Figure 3: Minimization of  $\sigma(E)$  and  $\langle E \rangle$ ,  $\lambda = 2$

The average energy  $\langle E \rangle$  is computed to be approximately  $\langle E \rangle \approx 3.725$ . Given this, and the fact that  $\lambda$  is known, we can look at the wave-function for a static point. In the figure below, fig 4a, one point is static at the origin, and thus the figure shows the wave-function experienced by the other particle. The numerical solution exhibits the same behavior just dampened.



(a) Probability density of the quantum dot, analytical solution

(b) Probability density of the quantum dot, numerical solution

Varying the free parameter  $\lambda$ , which determines the coupling strength, we can investigate how the ground-state energy of the quantum dot changes. Performing the same procedure as before, we

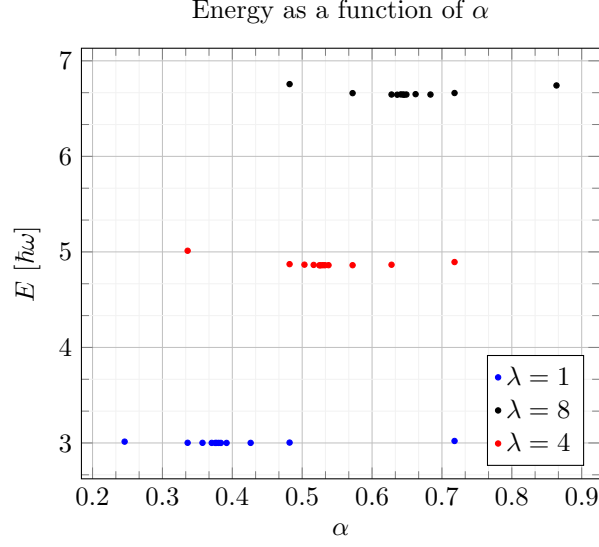


Figure 5: Energy as a function of  $\alpha$ , for different values of  $\lambda$

Table 1: Ground state energy of the quantum dot, for different values of  $\lambda$

$\lambda$	$\langle E \rangle [\hbar\omega]$	$\alpha$
1	3.000	0.379
4	4.860	0.528
8	6.646	0.642

obtain the following results, fig 5: The results above, and in table 1, show that the ground-state energy of the quantum dot increases with the coupling strength  $\lambda$ , which is expected. Moreover, when the coupling constant is zero, the ground-state energy is approximately  $2.0\hbar\omega$ , which is the expected value for the ground-state energy of the quantum dot with no coupling. The wave-function, with a particle located at the origin, for  $\lambda = 8$ , with the corresponding  $\alpha$  value located in table 1, is shown in the figure below, fig 6.

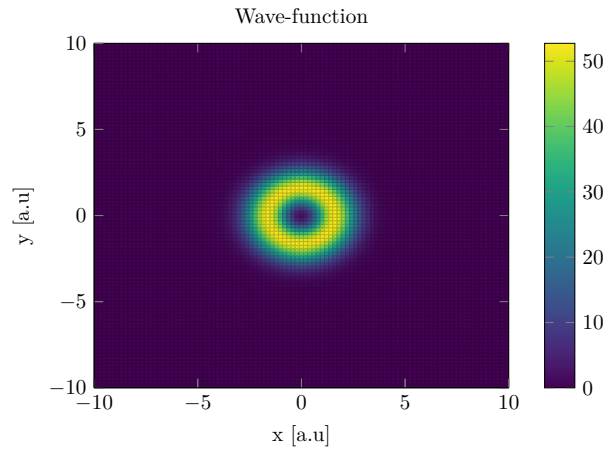


Figure 6: Wave-function of the quantum dot, when one particle is centered at the origin,  $\lambda = 8$

Comparing the above figure, fig 6, with the figure fig 4a, one sees that the annulus around the origin has a larger radius. This is due to the increased coupling strength, pushing the particles



apart, and thus increasing the probability of the particle being located further away from the origin.

## 4 Conclusion

In this report, the variational Monte Carlo method was used to solve the time independent Schrödinger equation for two different systems. Both methods were successful in finding the ground-state energy of the systems, and the wave-function of the systems were also obtained using the trial wave-function. For a simpler system, such as the harmonic oscillator in one dimension, the method was considerably slower than doing a direct method, where one also obtains all eigenstates. But for more complex systems, such as quantum dots, the method is more efficient, and the computational cost is significantly lower.

The method is also very versatile, if one can give a good estimate on the trial wave-function. In doing so, one could also have multiple arguments to minimize. However, this would then increase the computational cost, but the method would most likely still be more efficient than a direct method, or an iterative method.