Quantum Monte Carlo Simulation

We use random walks to obtain the ground state (i.e., the eigenstate with the lowest energy) of a quantum system [1]. To understand this method, we substitute $\tau = it$ into the time-dependent Schrödinger equation (in one dimension) to obtain

$$\frac{\partial}{\partial \tau}\psi(x,\tau) = -\frac{H}{\hbar}\psi(x,\tau) = \frac{\hbar}{2m}\frac{\partial^2}{\partial x^2}\psi(x,\tau) - \frac{V(x)}{\hbar}\psi(x,\tau).$$
(1)

If we retain only the first term in the right-hand side, this is the diffusion equation we have studied in the lecture on "Monte Carlo Simulation on Stochastic Processes". We can thus interpret the wave function $\psi(x,\tau)$ as a probability density with a diffusion constant $D = \hbar/2m$. We have learned that random walks can be used to find the solution to the diffusion equation. The second term, on the other hand, is a growth term, i.e., the population density of the random walkers at x grows or decays for a negative or positive potential V(x) value, respectively.

The general solution to Eq. (1) is written in terms of the eigenstates, $\phi_n(x)$, and eigenenergies, E_n , of the Hamiltonian *H*. (We sort the eigenenergies, E_n , in the ascending order, with E_0 being the ground state energy.)

$$\psi(x,\tau) = \exp(-H\tau/\hbar)\psi(x,0) = \sum_{n} c_n \phi_n(x) \exp(-E_n\tau/\hbar), \qquad (2)$$

where the weight of the initial wave function projected onto the *n*-th eigenstate is

$$c_n = \int dx \phi_n(x) \psi(x,0) \,. \tag{3}$$

Consider the long-time limit of Eq. (2). Noting that the ground state decays slowest among all the eigenstates, we obtain

$$\lim_{\tau \to \infty} \psi(x,\tau) = c_0 \phi_0(x) \exp(-E_0 \tau / \hbar).$$
(4)

Thus the histogram of random walkers will be proportional to the ground-state wave function, $\phi_0(x)$, in the long-time limit.

The only problem of the above procedure is that the population of walkers will decay to zero if $E_0 > 0$ or explode to infinity if $E_0 < 0$. To avoid this problem, we will measure the energy from a reference energy, V_{ref} , which is adjusted to achieve a stationary population of random walkers. As shown in Appendix, E_0 can be determined from the relation,

$$E_0 = \left\langle V \right\rangle = \frac{\int_{-\infty}^{\infty} V(x)\psi(x,\tau)dx}{\int_{-\infty}^{\infty} \psi(x,\tau)dx},$$
(5)

or it can be estimated by finite sampling,

$$E_0 = \left\langle V \right\rangle = \frac{1}{N} \sum_{i=1}^N V(x_i) \,, \tag{6}$$

where N is the number of random walkers and x_i is the position of the *i*-th walker.

Quantum Monte Carlo simulation discretizes imaginary time in unit of $\Delta \tau$. During one time step,

 $\Delta \tau$, each random walker moves either to the right or left by step length, Δs , which is related to $\Delta \tau$ by $(\Delta s)^2 = 2D\Delta \tau$. (Note that we have proven this relation in the lecture on "Monte Carlo Simulation of Stochastic Processes, when we derive the diffusion equation from random walks. The diffusion constant D = 1/2 in the atomic unit, $\hbar = m = 1$.) The reference potential can be estimated from the mean potential $\langle V \rangle$ and the change in the number of random walkers during $\Delta \tau$ as follows.

$$V_{\rm ref} = \left\langle V \right\rangle - \frac{N(\tau + \Delta \tau) - N(\tau)}{N(\tau)\Delta \tau},\tag{7}$$

as shown in Appendix.

This leads to a simple quantum Monte Carlo simulation algorithm shown below.

(Quantum Monte Carlo Algorithm)

- 1. Place N_0 walkers at the initial set of positions x_i .
- 2. Compute the reference energy, $V_{ref} = \sum_i V(x_i)/N_0$.
- 3. For each walker,
 - a. Randomly move the walker to the right or left by a fixed step length Δs .

b. Compute $\Delta V = V(x) - V_{ref}$ and a random number *r* in the range [0, 1]. If $\Delta V > 0$ and $r < \Delta V \Delta \tau$, then remove the walker. If $\Delta V < 0$ and $r < -\Delta V \Delta \tau$, then add another walker at *x*. Otherwise, just leave the walker at *x*.

- 4. Compute the mean potential energy (6) and the actual number of random walkers. The new reference potential is given by Eq. (7). The average $\langle V \rangle$ is an estimate of the ground state energy.
- 5. Repeat steps 3–4 until the estimates of the ground state energy $\langle V \rangle$ have reached a steady state value with only random fluctuations. Average $\langle V \rangle$ over many Monte Carlo steps to compute the ground state energy.

The program, qmwalk.c, implements this algorithm for a harmonic potential, $V(x) = x^2/2$, in the atomic unit. Initially, the walkers are randomly distributed within a distance w_0 of the origin. Input parameters are the desired number of walkers N_0 , the number of Monte Carlo steps per walker *mcs*, and the step length *ds*. The program computes the current number of walkers *N*, the current estimate of the ground state energy, and the value of V_{ref} . Initial part of the samples is discarded in the averages to approximate equilibration.

The following figure shows the time average of the mean potential $\langle V \rangle$ (calculated at the *imcs*-th Monte Carlo step as the accumulated sum of $\langle V \rangle$ over *imcs* steps, *Esum*, divided by *imcs*) and the estimate of V_{ref} for $N_0 = 50$, ds = 0.1, and mcs = 500.



Figure: Time average of the mean potential $\langle V \rangle$ (denoted as *Esum/imcs*) and the estimate of V_{ref} as a function of quantum Monte Carlo steps, *imsc*.

The following figure shows the histogram of the random walkers from the same simulation.



Figure: Histograms of random walkers in quantum Monte Carlo simulations for mcs = 500 and 50,000, compared with the exact solution, $\psi_0(x) = \pi^{-1/4} \exp(-x^2/2)$.

Appendix: Derivation of Equations (5) and (7)

Let us rewrite the imaginary-time Schrödinger equation (1), including the reference potential.

$$\frac{\partial}{\partial \tau}\psi(x,\tau) = \frac{\hbar}{2m}\frac{\partial^2}{\partial x^2}\psi(x,\tau) - \frac{V(x) - V_{\text{ref}}}{\hbar}\psi(x,\tau)$$
(A1)

By integrate this equation with respect to x, we obtain

$$\int_{-\infty}^{\infty} \frac{\partial}{\partial \tau} \psi(x,\tau) dx = \left[\frac{\hbar}{2m} \frac{\partial}{\partial x} \psi(x,\tau) \right]_{-\infty}^{\infty} - \int_{-\infty}^{\infty} \frac{V(x)}{\hbar} \psi(x,\tau) dx + \frac{V_{\text{ref}}}{\hbar} \int_{-\infty}^{\infty} \psi(x,\tau) dx \\ = -\int_{-\infty}^{\infty} \frac{V(x)}{\hbar} \psi(x,\tau) dx + \frac{V_{\text{ref}}}{\hbar} \int_{-\infty}^{\infty} \psi(x,\tau) dx$$
(A2)

The kinetic-energy term vanishes, because the wave function and hence its derivative are 0 at $x \rightarrow \pm \infty$.

Now note that, in the long-time limit, the wave function is

$$\psi(x,\tau) \approx c_0 \phi_0(x) \exp\left(-(E_0 - V_{\text{ref}})\tau / \hbar\right).$$
(A3)

By differentiating Eq. (A3) with respect to time, we obtain

$$\frac{\partial}{\partial \tau} \psi(x,\tau) \approx \frac{V_{\text{ref}} - E_0}{\hbar} \psi(x,\tau) \,. \tag{A4}$$

Substituting Eq. (A4) into (A2), we obtain

$$\int_{-\infty}^{\infty} \frac{V_{\text{ref}} - E_0}{\hbar} \psi(x, \tau) dx = -\int_{-\infty}^{\infty} \frac{V(x)}{\hbar} \psi(x, \tau) dx + \frac{V_{\text{ref}}}{\hbar} \int_{-\infty}^{\infty} \psi(x, \tau) dx \,. \tag{A5}$$

The terms proportional to V_{ref} cancel, and we finally obtain

$$\frac{E_0}{\hbar} \int_{-\infty}^{\infty} \psi(x,\tau) dx = \int_{-\infty}^{\infty} \frac{V(x)}{\hbar} \psi(x,\tau) dx, \qquad (A6)$$

which is equivalent to Eq. (5).

To derive Eq. (7), note that the number of random walkers $N(\tau)$ is written as

$$N(\tau) = \int_{-\infty}^{\infty} \psi(x,\tau) dx \,. \tag{A7}$$

Equation (A2) can then be written as

$$\frac{d}{d\tau}N = \frac{V_{\rm ref} - \langle V \rangle}{\hbar}N(\tau) \,. \tag{A8}$$

Suppose we have estimated the current rate of change in N. We will then set V_{ref} in order to counter-balance this change so that a steady population is, i.e.,

$$-\frac{d}{d\tau}N\bigg|_{\text{current}} = \frac{V_{\text{ref}}^{\text{new}} - \langle V \rangle}{\hbar}N(\tau), \qquad (A9)$$

or

$$\frac{V_{\rm ref}^{\rm new}}{\hbar} = \frac{\langle V \rangle}{\hbar} - \frac{dN / d\tau}{N(\tau)} \,. \tag{A10}$$

The time-discretized form of Eq. (A.10) gives Eq. (7).

References

1. H. Gould and J. Tobochnik, An Introduction to Computer Simulation Methods, Second Edition (Addison-Wesley, 1995).