Number partitioning via quantum adiabatic computation

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We study both analytically and numerically the complexity of the adiabatic quantum evolution algorithm applied to random instances of combinatorial optimization problems. We use as an example the NP-complete set partition problem and obtain an asymptotic expression for the minimal gap separating the ground and exited states of a system during the execution of the algorithm. We show that for computationally hard problem instances the size of the minimal gap scales exponentially with the problem size. This result is in qualitative agreement with the direct numerical simulation of the algorithm for small instances of the set partition problem. We describe the statistical properties of the optimization problem that are responsible for the exponential behaviour of the algorithm.

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Since the discovery by Shor [1] nearly a decade ago of a quantum algorithm for efficient integer factorization there has been a rapidly growing interest in the development of new quantum algorithms capable of solving computational problems that are practically intractable on classical computers. Perhaps the most notable example of such problems is that of combinatorial optimization (COP). In the simplest case the task in COP is to minimize the cost function ("energy") $E_\sigma$ defined on a set of $2^n$ binary strings $\sigma = \{z_1, \ldots, z_n\}$ $z_j = 0, 1$, each containing $n$ bits. In quantum computation this cost function corresponds to a Hamiltonian $H_P$

$$ H_P = \sum_\sigma E_\sigma |\sigma \rangle \langle \sigma | \tag{1} $$

where the indices $z_j = 0, 1$ and the summation is over $2^n$ states $|\sigma\rangle$ forming the computational basis of a quantum computer with $n$ qubits. State $|z_j\rangle$ of the $j$-th qubit is an eigenstate of the Pauli matrix $\sigma_j$ with eigenvalue $S_j = 1 - 2z_j$ ($S_j = \pm 1$). It is clear from the above that the ground state of $H_P$ encodes the solution to the COP with cost function $E_\sigma$.

COPs have a direct analogy in physics, related to finding ground states of classical spin glass models. In the example above bits $z_j$ correspond to Ising spins $S_j$. The connection between the properties of frustrated disordered systems and the structure of the solution space of complex COPs has been noted first by Fu and Anderson [2]. It has been recognized [3] that many of the spin glass models are in almost one-to-one correspondence with a number of COPs from theoretical computer science that form a so-called NP-complete class [4]. This class contains hundreds of the most common computationally hard problems encountered in practice, such as constraint satisfaction, traveling salesmen, integer programming, and others. NP-complete problems are characterized in the worst cases by exponential scaling of the running time or memory requirements with the problem size $n$. A special property of the class is that any NP-complete problem can be converted into any other NP-complete problem in polynomial time on a classical computer; therefore, it is sufficient to find a deterministic algorithm that can be guaranteed to solve all instances of just one of the NP-complete problems within a polynomial time bound. However it is widely believed that such an algorithm does not exist on a classical computer. Whether it exists on a quantum computer is one of the central open questions. Ultimately one can expect that the behavior of the new quantum algorithms for COPs and their complexity will be closely related to the properties of quantum spin glasses.

Farhi and co-workers suggested recently a new quantum algorithm for solving combinatorial optimization problems which is based on the properties of quantum adiabatic evolution [5]. Running of the algorithm for several NP-complete problems has been simulated on a classical computer using a large number of randomly generated problem instances that are believed to be computationally hard for classical algorithms [6-8]. Results of these numerical simulations for relatively small size of the problem instances ($n \leq 20$) suggest a quadratic scaling law of the run time of the quantum adiabatic algorithm with $n$. In [5, 9] special symmetric cases of COP were considered where symmetry of the problem allowed to describe the true asymptotic behavior ($n \to \infty$) of the algorithm. In certain examples considered in [9] quantum adiabatic algorithm finds the solution in time polynomial in $n$ while simulated annealing requires exponential time. However, so far there are no analytical results on the asymptotic behavior of the algorithm for the general case of randomly generated hard instances of NP-complete problems.

In what follows we derive the asymptotic complexity of the quantum adiabatic algorithm for the Set Partition Problem (SPP). It is one of the six basic NP-complete problems that are at the heart of the theory of NP-completeness [4]. It can be formulated as com-

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binatorial optimization problem: Given a sequence of positive numbers \( \{a_0, a_1, \ldots, a_n\} \) find a partition, i.e., two disjoint subsets \( A_1 \) and \( A_2 \) such that the residue
\[
E = \left| \sum_{a_j \in A} - \sum_{a_j \in A_1} \right|
\]
is minimized. Due to the obvious symmetry of the problem it is always possible to fix the assignment of one of the numbers, say \( a_0 \), to a subset \( A_1 \). In SPP we search for the bit strings \( z = \{z_1, \ldots, z_n\} \) (or corresponding Ising spin configurations \( S = \{S_1, \ldots, S_n\} \)) that minimize the energy or cost function \( E_z \)
\[
E_z = |\Omega_S|, \quad \Omega_S = a_0 + \sum_{j=1}^n a_j S_j, \quad S_j = 1 - 2z_j, \quad (2)
\]
Here \( S_j = 1 \) \( (z_j = 0) \) if \( a_j \in A \) and \( S_j = -1 \) \( (z_j = 1) \) if \( a_j \in A' \). The minimum partition can also be viewed as a ground state of the Ising spin glass, \(-\Omega_S^2\), with the Mattis-like (global) antiferromagnetic coupling, \( J_{ij} = -a_i a_j \).

SPP also has many practical applications including multiprocessor scheduling [10], cryptography [11], and others. The interest in SPP also stems from the remarkable failure of simulated annealing to find good solutions, as compared with the solutions found by deterministic heuristics [12]. The apparent reason for this failure is due to the existence of order \( 2^n \) local minima whose energies are of the order of \( 1/n \) [13] which undermines the usual strategy of exploring the space of configurations \( S \) through single spin flips.

The computational complexity of random instances of SPP depends on the number of bits \( b \) needed to encode the numbers \( a_j \). Numerical simulations show [14, 15] that for independent, identically distributed (i.i.d.) random \( b \)-bit numbers \( a_j \), the solution time grows exponentially with \( n \) for \( n < b \) and polynomially for \( n > b \). The transition from the "hard" to computationally "easy" phases has features somewhat similar to phase transitions in physical systems [16]. Transitions of this kind were observed in various NP-complete problems [17] and can be analyzed in the framework of statistical mechanics. In what follows we concentrate on the computationally hard regime \( n \ll b \).

We now consider the distribution of the partition energies \( E_z \). It was derived in [18] using statistical averaging over the ensemble of instances of SPP with i.i.d. numbers \( a_j \). Bearing in mind our eventual goal of deriving the complexity of the quantum algorithm for a given random instance of SPP we do not perform such an averaging. Instead we work with a given set of \( \{a_j\} \) and introduce a coarse-grained "density of states"
\[
\rho(E) = \left\langle \sum_z \delta(E - E_z) \right\rangle. \quad (3)
\]
Here the sum is over \( 2^n \) bit-strings \( z \); partition energy \( E_z \) is defined in (2) and \( \langle \cdots \rangle \) denotes averaging over the interval of the partition values \( E - \eta/2, E + \eta/2 \) with the window size chosen self-consistently, \( \eta \gg 1/\rho(E) \). Using (2) we can rewrite (3) in the form
\[
\rho(E) = \frac{2^{n+1}}{\pi} \int_0^\infty ds \left( \frac{s}{2} \right)^{\frac{3}{2}} I(s) \cos(Es), \quad (4)
\]
where \( I(s) = \prod_{j=0}^{n} \cos(a_j s) \) and \( f(x) = \sin(x)/x \) is a window function that imposes a cut-off in the integral (4) at \( s \sim 2/\eta \ll \rho(E) \). For large \( n \) this integral can be evaluated using the steepest descent method. To find the saddle points we shall assume that the \( b \)-bit numbers \( a_j \) are distributed inside of the interval \( (0,1) \) and are the integer multiples of \( 2^{-n} \), the smallest number that can be represented with available number of bits. We note that for large \( n \) the function \( I(s) \) has sharp maxima (minima) with width \( \sim n^{-1/2} \) at the points \( s_k = k\pi 2^b, k = 0, 1, \ldots; |I(s_k)| = 1 \). The sum over these saddle-points was evaluated by Mertens [16] who analyzed the partition function in SPP. In our case only one saddle point at \( s = 0 \) contributes to the integral (4). Indeed, it will be seen below that the window size \( 2/\eta \) can be chosen to obey the conditions \( 1 \ll n^{1/2}/\eta \ll 2^n \). Therefore in the high-precision case \( b \gg n \) saddle-points \( s_k \) with \( k > 0 \) lie far outside the window and their contributions can be neglected [? ]. On the other hand the window function \( f(x) \) can be replaced by unity while computing the contribution from the saddle-point at \( s = 0 \). Finally we obtain for \( E \ll n \)

\[
\rho(E) = \frac{2^{n+1} \Theta(E)}{\sqrt{2} \pi \sigma^2 n} \exp \left( -\frac{E^2}{2\sigma^2 n} \right) + O(n^{-3/2})
\]

\[
\sigma^2 = \frac{1}{n} \sum_{j=1}^n a_j^2 + O(n^{-1}), \quad E \ll n. \quad (5)
\]

Here \( \Theta(x) \) denote a step function, \( \Theta(x) = 1, \) for \( x \geq 0 \) and \( \Theta(x) = 0 \) for \( x < 0 \). The coarse-grained state density in Eq.(5) depends on the set \( \{a_j\} \) through a single self-averaging quantity \( \sigma^2 \) (cf. [16]). For \( E \) that are not too large \( E \ll \sqrt{n} \) the average separation between the individual partition energies \( E_{\text{min}} \sim \sqrt{\pi/2} \cdot n \). This justifies the choice of the window \( \eta \) above that corresponds to coarse-graining over many individual level separations. For a typical set of numbers \( \{a_j\} \) there are only two scales present in the distribution of partition energies, one is a "microscopic" scale given by \( E_{\text{min}} \) and another is given by the mean partition energy \( \langle E \rangle = \sigma(2n/\pi)^{1/2} \).

Using the same approach one can derive the distribution \( \tilde{\rho}(\Omega) \) of \( S_{\Omega} \). For \( |\Omega| \ll n \) it is Gaussian, \( \tilde{\rho}(\pm\Omega) = 1/2 \rho(|\Omega|) \) and can be understood in terms of a random walk with coordinate \( \Omega \) (cf. (2)). The walk begins at the origin, \( \Omega = 0 \), and makes \( n \) steps. At the \( j \)-th step \( \Omega \) moves to the right or to the left by "distance" \( 2a_j \) if \( S_j = 1 \) or \( S_j = -1 \), respectively. In the asymptotic limit of large \( n \) the result (5) corresponds to equal probabilities of right and left moves and the distribution of step lengths coinciding with that in the set of numbers \( \{2a_j\} \).

We now consider the "conditional" density of the partition energies \( \rho_r(E'E|E) = \rho_r(E',E)/\rho(E) \) where \( \rho_r(E'|E) \)
is a “joint” density defined as follows
\[ \rho_r(E', E) = \sum_{x} \rho_r(x(E')) \delta(E - E_x) \]  
(6)
\[ \rho_r(x(E')) = \left( \sum_{z} \delta(E' - E_{x'}) \Delta_r, D(x' | z) \right) \]  
(7)

Here \( \Delta_r, l \) is a Kronecker delta. Function \( D(z, z') \) above computes the number of bits that take different values in the bit-strings \( z \) and \( z' \), it is a so-called Hamming distance between the strings
\[ D(z, z') = \sum_{j=1}^{n} (1 - z_j) z'_j + (1 - z'_j) z_j. \]  
(8)
Conditional density \( \rho_r(x(E')) \) in (7) determines the distribution of energies \( E_{x'} \) for the strings \( z' \) obtained from \( z \) by flipping \( r \) bits. Similar to Eq. (3), \( \cdots \) in (6),(7) denotes averaging over the small energy interval \( (E - \eta/2, E + \eta/2) \) that, however, includes many individual partition levels. For \( r = 1 \) there are exactly \( n \) strings \( z' \) on a Hamming distance \( r = 1 \) from the string \( z \). Partition energies corresponding to these strings equal \( [\Omega_S - 2a_j \Sigma_j], 1 \leq j \leq n \) (cf. (2)). One can show that for \( E_x < 2 \) the distribution of these energies is nearly uniform in the range \( (0, 2 - E_x) \) due to the fact that spins \( S_j \) are nearly equally distributed between \( \pm 1 \) values for configurations with \( |\Omega_S| \ll n \). Similar arguments apply for the case \( r = n - 1 \). Using a coarse-graining window such that \( n^{-1} \ll \eta \ll 1 \) one can obtain in the range of energies \( E', E_x \ll 1 \) the following result:
\[ \rho_r(E' | E_x) \approx \rho_r(x(E')) = n/2 + O(1) \]  
(9)
For large \( r, n - r \gg 1 \) one can calculate the conditional densities (6) for a given set of \( \{a_j\} \) by evaluation of the appropriate integrals with the steepest descent method, in a similar manner to the derivation of \( \rho(E) \) in Eqs.(3)-(5). Here we will skip the derivation and only provide the result that we found to be in a very good agreement with the numerical simulations of SPP even for relatively small \( n \approx 20 - 30 \). State density functions in (6) are given each by a sum of Gaussians for \( E, E' > 0 \) and have a broad maximum at the origin. Near the maximum we have:
\[ \rho_r(E' | E_x) \approx \rho_r(x(E')) \approx \frac{\zeta(r/n)}{\sigma \sqrt{\pi n/2}} \left( \frac{n}{r} \right) \]  
(10)
\[ \zeta(x) = (4x(1-x))^{-1/2}, \quad r, n - r \gg 1. \]
This result applies in a range \( E', E_x \ll [n(c(r/n)]^{1/2} \). We note that conditional density (7) is uniform there and does not depend on \( E' \) and \( z \) after the coarse-graining over the small interval \( \eta \) of energies \( E_{x'} \) in (10). Window size \( \eta \gg \left( \frac{r}{n} \right)^{-1} \).

Eqs. (9),(10) suggest the structure of the low-lying partition configurations. Bit strings \( z' \) that are on a Hamming distance \( r \) from a given string \( z \) correspond to a subset of partition energies with average spacing \( \sim \left( \frac{r}{n} \right)^{-1} \) which does not depend on \( z \) in the range \( E_x, E_{x'} \ll 1 \) for all \( r \). The greater the Hamming distance between the strings - the closer their partition energies can be to each other - and vice versa. This is a basic property of SPP that gives rise to an exponentially large number of local minima in the relatively narrow range of energies \( O(1) \). In a typical instance of random SPP correlations between the Hamming distances to a given string (e.g., the solution to SPP) and partition energies essentially vanishes.

In the quantum adiabatic algorithm [5] one specifies the time-dependent Hamiltonian \( H(t) = \hat{H}(t/T) \)
\[ \hat{H}(\tau) = (1 - \tau) V + \tau H_P, \]  
(11)
where \( \tau = t/T \) is dimensionless “time”. This Hamiltonian guides the quantum evolution of the state vector \( |\psi(t)\rangle \) according to the Schrödinger equation
\[ i\hbar \frac{\partial |\psi(t)\rangle}{\partial t} = H(t)|\psi(t)\rangle \]
from \( t = 0 \) to \( t = T \), the run time of the algorithm. \( H_P \) is the “problem” Hamiltonian given in (1). \( V \) is a “driver” Hamiltonian, that is designed to cause the transitions between the eigenstates of \( H_P \). In this algorithm one prepares the initial state of the system \( \psi(0) \) to be a ground state of \( \hat{H}(0) = V \). In the simplest case
\[ V = -\sum_{j=0}^{n-1} a_j \sigma^j_z, \quad |\psi(0)\rangle = 2^{-n/2} \sum_z |z\rangle, \]  
(12)
where \( \sigma^j_z \) is a Pauli matrix for \( j \)-th qubit. Consider instantaneous eigenstates \( |\phi_\alpha(\tau)\rangle \) of \( \hat{H}(\tau) \) with energies \( \lambda_\alpha(\tau) \) arranged in nondecreasing order at any value of \( \tau \in (0,1) \)
\[ \hat{H}|\phi_\eta\rangle = \lambda_\eta|\phi_\mu\rangle, \quad \eta = 0, 1, \ldots, 2^n - 1. \]  
(13)
Provided the value of \( T \) is large enough and there is a finite gap for all \( t \in (0, T) \) between the ground and excited state energies, \( g(\tau) = \lambda_1(\tau) - \lambda_0(\tau) > 0 \), quantum evolution is adiabatic and the state of the system \( |\psi(t)\rangle \) stays close to an instantaneous ground state, \( |\phi_0(t/T)\rangle \) (up to a phase factor). Because \( H(T) = H_P \) the final state \( |\psi(T)\rangle \) is close to the ground state \( |\phi_0(\tau = 1)\rangle \) of the problem Hamiltonian. Therefore a measurement performed on the quantum computer at \( t = T \) (\( \tau = 1 \)) will find one of the solutions of COP nearly with certainty. In the Landau-Zener picture a quantum transition away from the adiabatic ground state occurs most likely in the vicinity of the point \( \tau^* \in (0,1) \) where the energy gap \( g(\tau) \) reaches its minimum (avoided-crossing region). The probability of the transition, \( 1 - \langle\psi(T)|\phi_0(\tau = 1)\rangle^2 \), is small provided that
\[ T \gg \frac{|H_{01}^*|}{\Delta g_{\min}^2}, \quad g_{\min} = \min_{0 \leq \tau \leq 1} [\lambda_1(\tau) - \lambda_0(\tau)] \]  
(14)
Here \( |H_{01}^*| = |\langle\phi_1|\hat{H}|\phi_0\rangle|_{\tau=\tau^*} \). Even in the cases that are more complex than the Landau-Zener “2-level” picture
of nonadiabatic transitions the analysis of minimum gap \( g_{\text{min}} \) can provide an important test for the complexity of the quantum adiabatic algorithm. E.g., if one can show that for a random COP \( g_{\text{min}} \) decreases exponentially fast with the problem size \( n \) while \( |H_{(n)}^{\uparrow}\rangle \) does not decreases with \( n \) that would be a strong indication for exponential complexity of the algorithm in such problem.

As suggested in [5] the quantum adiabatic algorithm can be recast within the conventional quantum computing paradigm using the technique introduced by Lloyd [19]. Continuous-time quantum evolution can be approximated by a time-ordered product of unitary operators, \( e^{-i(1-\gamma_r)\lambda \delta} e^{-i\gamma_r \lambda \delta} \), corresponding to small time intervals \( (t_k, t_k + \delta) \). Operator \( e^{-i(1-\gamma_r)\lambda \delta} \) typically corresponds to a sequence of 1(2)-qubit gates (cf. (12)). Operator \( e^{-i\gamma_r \lambda \delta} \) is diagonal in the computational basis \( |z\rangle \) and corresponds to phase rotations by angles \( E_{\ast} \Delta \). Since in the case \( n \ll b \), the average separation between the neighboring values of \( E_{\ast} \) is \( 1/p(E) = O(2^{-n}) \), the quantum device would need to support a very high precision of physical parameters (like external fields, etc) to control small differences in phases, \( O(2^{-n}) \). Since this precision scales with \( n \) exponentially it would strongly restrict the size of an instance of SPP that could be solved on such quantum computer (this technical restriction is generic for COPs that involve high-precision numbers). To avoid this restriction we define a new cost function that takes values on a scale which is \( logarthmic \) in \( p(E) \). I.e., we chose the cost to be an oracle-type function \( E_{\ast} \) that takes a set of \( M+1 \) integer values \( \varepsilon_k = -M + k, \ 0 \leq k \leq M \)

\[
E_{\ast} = -k + M, \ \text{for } \omega_k \leq E_{\ast} < \omega_{k+1}, \quad 0 \leq k \leq M.
\]

We set \( E_{\ast} = \epsilon_M = 0 \) for all states with \( E_{\ast} \geq \omega_M \) where \( \omega_M \) is a control parameter chosen sufficiently large so that \( M = O(n) \). The space of \( 2^n \) states \( |z\rangle \) is divided into \( M+1 \) subspaces \( L_k \), each determined by Eqs. (15) and (16) for a given \( k \)

\[
H_{\text{P}} = \sum_{k=0}^{M} \epsilon_k \sum_{z \in L_k} |z\rangle \langle z|.
\]

Note that subspace \( L_0 \) contains the solution(s) to the SPP. Dimension \( d_0 \) of \( L_0 \) is controlled by the value of \( \omega_0 \) in (16) which is another control parameter of the algorithm. We set \( \omega_0 = K/p(0) \) where the integer \( K \approx d_0 \geq 1 \) is independent of \( n \) and determines how many times in average one needs to repeat the quantum algorithm in order to obtain the solution to SPP with probability close to 1.

Operator \( H_{\text{P}} \) projects any state \( |\psi\rangle \) onto the states with partition energies in the range \( 0 \leq E_{\ast} < \omega_{\ast} \). We assume that \( \omega_{\ast} \ll \langle E \rangle \) so that the density function (5) is nearly uniform in this range. Therefore the dimensions of the subspaces \( L_k \) grow exponentially with \( k: \ d_k = d_0 2^k \) for \( k < M \). This simplification does not affect the complexity of a quantum algorithm that spends most of the time in “annealing” the system to much smaller partition energies, \( \omega_M \gg E_{\ast} \sim E_{\text{min}} = O(n^{1/2} 2^{-n}) \).

We now consider a stationary Schrödinger equation (13) and obtain the minimum gap \( g_{\text{min}} \) (14) in the asymptotic limit \( n \to \infty \). To proceed we need to introduce a new basis of states \( |x\rangle = |x_1\rangle \otimes |x_2\rangle \otimes \cdots \otimes |x_n\rangle \) where state \( |x_j\rangle \) is an eigenstate of the Pauli matrix \( \sigma_x \) corresponding to the \( j \)-th qubit with eigenvalue \( 1-2x_j = \pm 1 \). Driver Hamiltonian \( V \) can be written in the following form

\[
V = \sum_{m=0}^{n} V_m \mathcal{I}^m, \quad \mathcal{I}^m = \sum_{x_1+\cdots+x_m=m} |x\rangle \langle x|.
\]

For a particular case given in Eq. (12) we have \( V_m = 2m - n \). Matrix elements of \( \mathcal{I}^m \) in a basis of states \( |z\rangle \) depend only on the Hamming distance \( D(z, z') \) between the strings \( z \) and \( z' \)

\[
\mathcal{I}^m_{z+ z'} = \mathcal{I}^m_{p(z, z')}
\]

\[
\mathcal{I}^m_p = 2^{-n} \sum_{q=0}^{n} \sum_{p=0}^{r} \binom{n-r}{p} (-1)^p \Delta_{m,q+p}
\]

We now rewrite Eq. (13) in the form

\[
|\Psi\rangle = \frac{\tau}{\lambda - \alpha V} |\Psi\rangle, \quad \alpha = \alpha(\tau) = 1 - \tau.
\]

(we temporarily drop the subscript \( \mu \) indicating the number of a quantum state and the argument \( \tau \) in \( \Psi \)). From (15)-(21) we obtain the equation for the amplitudes \( d_{\ast} = \langle \psi|z\rangle \) in terms of the coefficients \( I_{\ast}^m \)

\[
\Xi_{k} \Psi_{\ast} = \frac{\tau \Phi 2^{-n}}{\lambda - \alpha V} \sum_{x' \neq x} J_{D(x,x')} \Psi_{x'} E_{x'}, \quad \Xi_{k} \Psi_{\ast} = \sum_{x' \neq x} E_{x'} \Psi_{x'}. \quad (23)
\]

\[
J_{\ast} \equiv J_{\ast}(\lambda) = \sum_{m=1}^{n} \frac{I_{\ast}^{m}}{\lambda - \alpha V_{m}}, \quad 0 \leq r \leq n.
\]

We now make a key observation that the (coarse-grained) distribution of matrix elements \( J_{D(x,x')} \) for a given \( z \) is determined by \( J_{\ast}(\lambda) \) and conditional density \( p_{r,z}(E) \). We compute a cumulative quantity

\[
\sum_{x' \neq x} \frac{J_{D(x,x')} \lambda}{\sum_{x' \neq x} J_{D(x,x')} \lambda} \approx F_{\ast}(\lambda) + f_{\ast,k}(\lambda)
\]

\[
F_{\ast}(\lambda) = \frac{\mu s(\lambda)}{2M-k}, \quad s(\lambda) = \int_{0}^{\lambda} dr \zeta(r/n) \left( \frac{n}{r} \right) J_{\ast}(\lambda)
\]

where \( \mu = \Omega_{\lambda} / \sqrt{\pi \sigma^2} / 2 \) and \( f_{\ast,k}(\lambda) \) is a small correction that will be described below. Function \( s(\lambda) \) can be evaluated using the explicit form of \( J_{\ast}(\lambda) \) that decays rapidly with \( r \). E.g., for \( |\lambda - \alpha V_0| \ll 1 \) one has

\[
J_{\ast}(\lambda) = \left( \frac{n}{r} \right) \sum_{m=1}^{n-r} \frac{2^{-n} \mu n^{-1}}{m} - 2^{-n} \ln(n+\gamma).
\]
\[ u_k = \sum_{\phi \in \mathcal{L}} \phi \chi_k(\phi) \]

where the function \( \chi_k(\phi) \) is defined as

\[ \chi_k(\phi) = \begin{cases} 1 & \text{if } \phi \in \mathcal{L}_k, \\ 0 & \text{otherwise}. \end{cases} \]

The state vector \( |\psi\rangle \) is given by

\[ |\psi\rangle = \sum_{\phi \in \mathcal{L}} \phi \langle \phi | \psi \rangle \]

and the Hamiltonian matrix in this representation is

\[ H = \sum_{\phi \in \mathcal{L}} \phi \langle \phi | H | \phi \rangle \]

where the matrix elements are

\[ \langle \phi | H | \phi \rangle = E(\phi). \]

The energy levels of the system are given by

\[ E(\phi) = \sum_{\phi \in \mathcal{L}} \phi \langle \phi | H | \phi \rangle \]

and the wavefunction is

\[ \psi(z) = \sum_{\phi \in \mathcal{L}} \phi \phi^*(\phi) \]

The total energy of the system is

\[ E = \sum_{\phi \in \mathcal{L}} \phi \langle \phi | H | \phi \rangle \]

and the probability density is

\[ P(z) = \sum_{\phi \in \mathcal{L}} | \phi \phi^*(\phi) |^2 \]

The expectation value of a operator \( O \) is

\[ \langle O \rangle = \sum_{\phi \in \mathcal{L}} \phi \langle \phi | O | \phi \rangle \]

and the variance is

\[ \Delta^2 = \sum_{\phi \in \mathcal{L}} \phi \langle \phi | H | \phi \rangle - \langle O \rangle^2 \]

The master equation for the system is

\[ \dot{\rho} = -i [H, \rho] + \sum_{\phi \in \mathcal{L}} \phi \phi^*(\phi) \rho \phi^*(\phi) - \sum_{\phi \in \mathcal{L}} \phi \rho \phi^*(\phi) \]

and the steady state solution is

\[ \rho_{ss} = \sum_{\phi \in \mathcal{L}} \phi \phi^*(\phi) \rho \phi^*(\phi) \]

Finally, the von Neumann entropy is

\[ S(\rho) = -\sum_{\phi \in \mathcal{L}} \phi \log_2 \phi \rho \phi^*(\phi) \]

and the mutual information is

\[ I(\rho) = \sum_{\phi \in \mathcal{L}} \phi \phi^*(\phi) \rho \phi^*(\phi) \]
for not too large energies (cf. [18]), and therefore the structure of the problem is being averaged out during the quantum evolution that begins from the symmetric state (12).

We also study the complexity of the algorithm by numerical integration of the time-dependent Schrödinger equation with Hamiltonian $H(t)$ and initial state $|\psi(0)\rangle$ defined in Eqs. (11), (12), (15)-(17). Here we relax the condition $\omega_M \ll (E)$ used above in the analytical treatment of the problem; in simulations the value of $M$ is set automatically to be an integer closest to $\log_2 \sum_{j=0}^{n} a_j$ (cf. (16)). We introduce a complexity metric for the algorithm, $C(T) = (1 + T) d_0/p_0(T)$ where $p_0(T) = \sum_{w \in \mathcal{E}_n} |\psi_w(t)|^2$. A typical plot of $C(T)$ for an instance of the problem with $n=15$ numbers is shown in the insert of Fig. 1. At very small $T$ the wavefunction is close to the symmetric initial state and the complexity is $\sim 2^n$. The extremely sharp decrease in $C(T)$ with $T$ is due to the buildup of the population $p_0(T)$ in the ground level, $\mathcal{E}_n = \varepsilon_0$, as quantum evolution approaches the adiabatic limit. At certain $T = T_{\text{min}}$ the function $C(T)$ goes through the minimum: for $T > T_{\text{min}}$ the decrease in the number of trials $d_0/p_0(T)$ does not compensate anymore for the overall increase in the runtime $T$ for each trial. For a given problem instance the “minimal” complexity $C_{\text{min}} = C(T_{\text{min}})$ is obtained via one dimensional minimization over $T$. Plot of the complexity $C_{\text{min}}$ for different values of $n$ in Fig. 1 appears to indicate the exponential scaling law, $C_{\text{min}} \sim 2^{0.8n}$ for not too small values of $n \gtrsim 11$.

In conclusion, we have studied both analytically and numerically, the complexity of the quantum adiabatic algorithm applied to large random instances of SPP. We obtained the asymptotic behavior of the minimal gap that scales exponentially with problem size $n$. Analytical results for the algorithm’s complexity are in qualitative agreement with the results of numerical simulations of the algorithm for small-to-moderate values of $n$. We also developed a general technique for the analyzes of quantum spin-glass problems based on the separation of fast- and slow-varying parts of the wavefunction and the related picture of avoided crossing between the energies of extended and localized states (as the system parameters vary). This approach can be applied to different examples of random COPs where one searches for the solution using continuous-time quantum algorithms. Among the other applications is the analysis of tunneling phenomenon in the low-temperature dynamics of random magnets.

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