On Quantum Versions of Record-Breaking Algorithms for SAT

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Abstract

It is well known that a straightforward application of Grover's quantum search algorithm enables to solve SAT in $O(2^{n/2})$ steps. Ambainis (SIGACT News, 2004) observed that it is possible to use Grover's technique to similarly speed up a sophisticated algorithm for solving 3-SAT. In this note, we show that a similar speed up can be obtained for all major record-breaking algorithms for satisfiability. We also show that if we use Grover's technique only, then we cannot do better than quadratic speed up.

1 Quantum Computing and Satisfiability

Faster quantum algorithms for SAT. In the satisfiability problem (SAT), we are given a Boolean formula F in conjunctive normal form $C_1 \,\&\, \ldots \,\&\, C_m$, where each clause C_j is a disjunction $l_1 \vee \ldots \vee l_k$ of literals, i.e., variables or their negations. We need to find a truth assignment $x_1 = a_1, \ldots, x_n = a_n$ that makes F true. A simple exhaustive search can solve this problem in time $\sim 2^n$, where $\sim$ means equality modulo a term which is polynomial in the length of the input formula.

The main attraction of quantum computing is that it can speed up computations. In particular, Grover's quantum algorithm [9, 10, 11, 15] searches an unsorted list of N elements to find an element with a given property. In non-quantum computations, every such search algorithm requires, in the worst case, N steps; Grover's algorithm can find this element in time $O(\sqrt{N})$ with arbitrary high probability of success. Thus, a straightforward application of Grover's technique can solve SAT in time $\sim 2^{n/2}$.

Computer simulation of quantum computing suggests that it may be possible to solve SAT even faster [12]. Can we actually use quantum computing to solve SAT faster than in time $\sim 2^{n/2}$? In this note, we discuss some aspects of this question.

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Remark. We only consider quantum computing within the standard quantum physics. It is known that if we consider non-standard versions of quantum physics (e.g., a version in which it is possible to distinguish between a superposition of $|0\rangle$ and $|1\rangle$ and a pure state) then, in principle, we can solve NP-complete problems in polynomial time; see, e.g., [1] and references therein, and also [2, 14, 16].

Ambainis' observation. In [3], Ambainis considers algorithms for k-SAT, a restricted version of SAT where each clause has at most k literals. He shows that one of the fastest algorithms for k-SAT, namely, the algorithm proposed by Schöning [21], can be similarly sped up from time $T \sim (2-2/k)^n$ to $\sqrt{T} \sim (2-2/k)^{n/2}$.

Schöning's algorithm is a *multi-start random walk* algorithm that repeats the polynomial-time random walk procedure $\mathcal{S}$ exponentially many times. This procedure $\mathcal{S}$ takes an input formula F and does the following:

- Choose an initial assignment a uniformly at random.
- Repeat $3n$ times:
  - If F is satisfied by the assignment a, then return a and halt.
  - Otherwise, pick any clause C_j in F such that C_j is falsified by a; choose a literal l_s in C_j uniformly at random; modify a by flipping the value of the variable x_i from the literal l_s.

As shown in [21], if the formula F is satisfiable, then each random walk of length $3n$ finds a satisfying assignment with the probability $\geq (2-2/k)^{-n}$. Therefore, for any constant probability of success, after $O((2-2/k)^n)$ runs of the random walk procedure $\mathcal{S}$, we get a satisfying assignment with the required probability. Since $\mathcal{S}$ is a polynomial time procedure, the overall running time of this algorithm is also $T \sim (2-2/k)^n$. This upper bound is close to the best known upper bound for k-SAT (see below). Schöning's algorithm was derandomized in [6].

In Schöning's algorithm, there are $N \sim (2-2/k)^n$ results of different runs of $\mathcal{S}$, and we look for a result in which the input formula F is satisfied. Grover's algorithm enables us to find this result in time $\sim \sqrt{N}$. More exactly, this reduction comes from the modification of the original Grover's algorithm called *amplitude amplification*) [3, 5]. Thus, there exists a quantum algorithm that solves k-SAT in time $\sim \sqrt{T} \sim (2-2/k)^{n/2}$.

For 3-SAT, Schöning's algorithm was improved by Rolf [19] to $T \sim 1.330^n$. This improvement also consists of exponentially many runs of a polynomial-time algorithm. Therefore, Rolf's non-quantum running time $T \sim 1.330^n$ leads to the corresponding quantum time $\sqrt{T} \sim 1.154^n$.

SAT is a particular case of a more general discrete *constraint satisfaction problem* (CSP), where variables $x_1, \ldots, x_n$ can take $d \geq 2$ possible values, and constraints can be more general than clauses. In particular, we can consider k-CSP, in which every constraint contains $\leq k$ variables. Schöning's algorithm can be naturally extended to k-CSP [21]. The running time of the corresponding algorithm is $T \sim (d \cdot (1-1/k) + \varepsilon)^n$, where ε can be arbitrarily small. Similar to Schöning's algorithm for k-SAT, this extension to k-CSP can be quantized with

the running time $T_Q \sim \sqrt{T} \sim (d \cdot (1-1/k)+\varepsilon)^{n/2}$. A different quantum algorithm for 2-CSP is described in [4].

The fastest algorithm for k-SAT. The best known upper bound for k-SAT is given by the algorithm proposed by Paturi, Pudlák, Saks, and Zane [17, 18]; this algorithm is called PPSZ. This algorithm consists of exponentially many runs of a polynomial-time procedure. This procedure is based on the following approach:

- Pick a random permutation $\pi(1)$, $\pi(2)$, ..., $\pi(n)$ of the variables.
- Select a truth value of the variable $x_{\pi(1)}$ at random.
- Simplify the input formula as follows:
  - Substitute the selected truth value for $x_{\pi(1)}$.
  - If one of the clauses reduces to a single literal, simplify the formula again by using this literal.
  - Repeat such simplification while possible.
- Select a truth value of the first unassigned variable (in the order $\pi(1), \pi(2), \ldots$) at random.
- Simplify the formula as above.
- Continue this process until all n variables are assigned.

As shown in [18], the PPSZ algorithm runs in time $T \sim 2^{n \cdot (1-\mu_k/k)}$, where $\mu_k \to \pi^2/6$ as k increases. The PPSZ algorithm was derandomized in [20] for the case when there is at most one satisfying assignment.

Since the PPSZ algorithm also consists of exponentially many runs of a polynomial-time procedure, we can use Grover's technique to design its quantum version which requires time $T_Q \sim \sqrt{T}$.

A combination of the PPSZ and Shöning's approaches leads to the best known upper bound for 3-SAT: $T \sim 1.324^n$ (Iwama and Tamaki [13]). Similarly to the previous algorithms, this algorithm also consists of independent runs of a polynomial-time procedure. So, by applying Grover's algorithm, we can similarly get a quantum algorithm with time $\sqrt{T} \sim 1.151^n$.

The fastest algorithm for SAT with no restriction on clause length. The best known upper bound for SAT with no restriction on clause length is given in [8]. The corresponding algorithm is based on the *clause shortening* approach proposed by Schuler in [22]. This approach suggests exponentially many runs of the following polynomial-time procedure $\mathcal{S}$:

- Convert the input formula F to an auxiliary k-CNF formula F'. Namely, for each clause C_j longer than k, keep the first k literals and delete the other literals in C_j.

- Use a k-SAT algorithm, e.g., one random walk of Schöning's algorithm, to test satisfiability of F'. Assuming that F has a satisfying assignment a, there are two possible cases:
  - First, the k-SAT algorithm has found a; then we are done.
  - Second, some clause C'_j in F' is false under a. If we guess this clause, we can reduce the number of variables in F by substituting the corresponding truth values for the variables of C'_j. Therefore, we choose a clause in F' at random and simplify F by replacing the variables that occur in this clause with the corresponding truth values.
- Finally, we recursively apply $\mathcal{S}$ to the result of simplification.

The procedure $\mathcal{S}$ runs in polynomial time and finds a satisfying assignment (if any) with probability at least

$$2^{-n\cdot\left(1-\frac{1}{\ln\left(\frac{m}{n}\right)+O(\ln\ln(m))}\right)}.$$

This probability can be increased to a constant by repetition in the usual way, so the algorithm for SAT requires time

$$T\sim 2^{n\cdot\left(1-\frac{1}{\ln\left(\frac{m}{n}\right)+O(\ln\ln(m))}\right)}.$$

By using Grover's technique, we can produce a quantum version of this algorithm that requires time T_Q:

$$T_Q\sim\sqrt{T}\sim 2^{-(n/2)\cdot\left(1-\frac{1}{\ln\left(\frac{m}{n}\right)+O(\ln\ln(m))}\right)}.$$

2 How Much More Can Grover's Algorithm Help?

At most quadratic speed-up. So far, we have used Grover's technique to speed up the non-quantum computation time T to the quantum computation time $T_Q\sim\sqrt{T}$. Let us show that if Grover's technique is the only quantum technique that we use, then we cannot get a further time reduction. Informally speaking, let us call a quantum algorithm that uses only Grover's technique (and no other quantum ideas) *Grover-based*. We show that the following two statements hold:

- **Statement 1.** If we have a Grover-based quantum algorithm $\mathcal{A}_Q$ that solves a problem in time T_Q, then we can "dequantize" it into a non-quantum algorithm $\mathcal{A}$ that requires time $T=O(T_Q^2)$.
- **Statement 2.** If we have a non-quantum algorithm that solves a problem in time T, then any Grover-based quantum algorithm for solving this problem requires time at least $T_Q=\Omega(\sqrt{T})$.

First statement. Without loss of generality, we can assume that the time is measured in number of steps. Then $T_Q = t_0 + t_1 + \ldots + t_s$, where t_0 denotes the number of non-quantum steps in $\mathcal{A}_Q$, s denotes the number of Grover's searches, and t_i denotes the time required for i-th quantum search.

To show that the first statement holds, let us recall that the Grover's algorithm searches the list of N elements to find an element with the desired property. Exhaustive search can find this element by N calls to a procedure which checks whether a given element has this property. While the (worst-case) running time of exhaustive search is $r \cdot N$, where r is the running time of the checking procedure, Grover's algorithm enables us to find the desired element in $c \cdot \sqrt{N}$ calls to this procedure, where c is a constant determined by the required probability of success. So, the running time of Grover's algorithm is $r \cdot c \cdot \sqrt{N}$.

In the i-th Grover's search, $t_i = r_i \cdot c \cdot \sqrt{N_i}$, where N_i is the number of elements in the corresponding list and r_i is the running time of the corresponding checking procedure. So, we can conclude that

$$N_i = \frac{t_i^2}{r_i^2 \cdot c^2}.$$

Hence, by using (non-quantum) exhaustive search algorithm, we can perform the same search in time

$$t_i' = r_i \cdot N_i = \frac{t_i^2}{r_i \cdot c^2}.$$

Since $r_i \geq 1$, we conclude that $t_i' \leq c' \cdot t_i^2$, where $c' = \max(1, c^{-2})$.

Since t_0 is a non-negative integer, we have $t_0 \leq t_0^2$; since $c' \geq 1$, we have $t_0 \leq c' \cdot t_0^2$. Thus, by replacing each Grover's search by the non-quantum search, we get the time $T = t_0 + t_1' + \ldots + t_s'$. Here, $t_i' \leq c' \cdot t_i^2$ for all i, hence $T \leq c' \cdot (t_0^2 + t_1^2 + \ldots + t_s^2)$. Since

$$t_0^2 + \ldots + t_s^2 \leq (t_0 + \ldots + t_s)^2 = t_0^2 + \ldots + t_s^2 + 2 \cdot t_0 \cdot t_1 + \ldots,$$

we conclude that $T \leq c' \cdot T_Q^2$.

Second statement. Since $T \leq c' \cdot T_Q^2$, we have $T_Q \geq (1/\sqrt{c'}) \cdot \sqrt{T}$, i.e., $T_Q = \Omega(\sqrt{T})$.

Remark. Our observation is valid only if we restrict the use of quantum computation to Grover's algorithm. There are quantum techniques which lead to a faster speed-up. For example, the well-known Shor's algorithm for factoring large integers requires polynomial time [23, 24, 15], while all known non-quantum factorization algorithms require, in the worst case, exponential time. If we can use such techniques, we might get more than quadratic speed-up.

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