

A Fixed-Phase Quantum Search Algorithm with More Flexible Behavior

Xin Li*, Panchi Li

School of Computer & Information Technology, Northeast Petroleum University, Daqing, China

Email: *lixin_dq@163.com

Received February 6, 2012; revised March 18, 2012; accepted April 20, 2012

ABSTRACT

When the Grover's algorithm is applied to search an unordered database, the probability of success usually decreases with the increase of marked items. To address this phenomenon, a fixed-phase quantum search algorithm with more flexible behavior is proposed. In proposed algorithm, the phase shifts can be fixed at the different values to meet the needs of different practical problems. If research requires a relatively rapid speed, the value of the phase shifts should be appropriately increased, if search requires a higher success probability, the value of the phase shifts should be appropriately decreased. When the phase shifts are fixed at 0.1π , the success probability of at least 99.38% can be obtained in $O(\sqrt{N/M})$ iterations.

Keywords: Quantum Computing; Quantum Searching; Grover Algorithm; Fixed Phase Shifting

1. Introduction

Grover's quantum search algorithm [1] is one of the most important developments in quantum computation. For searching a marked state in an unordered database, it achieves quadratic speed up over classical search algorithms. At present, Grover's quantum search algorithm has been greatly noticed and has become a challenging research field. However, the Grover's algorithm also has some limitations. When the fraction of marked items is greater than a quarter of the total items in the database, the success probability will rapidly decrease, and when the fraction of marked items is greater than half of the total items in the database, the algorithm will be disabled.

Up to now, many efforts in improving Grover's original algorithm have been done. The Grover's original algorithm consists of inversion of the amplitude in the desired state and inversion-about-average operation [1]. In [2], Grover presented a general algorithm: $Q = -I\gamma U^+ I\tau U$, where U is any unitary operation, U^+ is the adjoint of U , $I\gamma = I - 2|\gamma\rangle\langle\gamma|$, $I\tau = I - 2|\tau\rangle\langle\tau|$, $|\gamma\rangle$ is an initial state and $|\tau\rangle$ is a desired state. When $U^{-1} = U = W$, where W is the Walsh-Hadamard transformation, and $|\gamma\rangle = 0$, the general algorithm becomes the original algorithm. Long extended Grover's algorithm [3]. In Long's algorithm, $I\gamma$ and $I\tau$ are expressed as $I\gamma = I - (1 - e^{i\theta})|\gamma\rangle\langle\gamma|$ and $I\tau = I - (1 - e^{i\phi})|\tau\rangle\langle\tau|$, respectively. Long further studied two-phase matching

conditions [4]. When $\theta = \phi = \pi$, Long's algorithm becomes Grover's original algorithm. In [5], Grover presented the new algorithm by replacing the selective inversions by selective phase shifts of $\pi/3$. Just like classical search algorithms the algorithm has a fixed point in state-space toward which it preferentially converges. Li *et al.* studied the changes of the approximation error in the fixed-point search algorithm obtained by replacing equal phase shifts of $\pi/3$ by different phase shifts [6]. In [7], by applying Partial Diffusion Operator, Younes *et al.* proposed a quantum search algorithm and showed that the performance of the algorithm is more reliable than other quantum search algorithms especially for multiple matches within the search space. The minimum success probability of this algorithm is 84.72% when

$M/N = 0.3084$. In [8], by regarding the searching engine as a three-dimensional rotation in the $SO(3)$ picture, Long presented a modified version of Grover's algorithm that searches a marked state with full successful probability. However, two phase shifts in Long's algorithm are determined by the number of iterations and the number of marked items, not a fixed value. Building quantum devices using fixed operators is a must to simplify the hardware construction. Quantum search engine is not an exception [9]. Therefore, this algorithm is not conducive to the quantum computer hardware implementation. Younes presented a fixed phase quantum search algorithm in [9]. By selecting phase shifts of 1.825π in the standard amplitude amplification, the success probability

*Corresponding author.

of at least 98% is obtained in $O(\sqrt{N/M})$, which is better than other fixed operator quantum search algorithms. In [10], a novel information geometric characterization of Grover's quantum search algorithm is presented, and the possible deviations from Grover's algorithm within this quantum information geometric setting are discussed. In [11], a method for implementing the Grover search algorithm based on multi-level systems is proposed.

The methods mentioned above cannot solve the problem that the algorithm efficiencies decrease as the marked items increase. In this paper, we study the phase matching in Grover's algorithm, and propose an adaptive matching, namely, $\theta = -\phi = f(\lambda)$, where λ is the fraction of marked items, and $f(\lambda)$ is the polynomial for λ . With application of the new phase matching, when λ is a rational number in range $(3-\sqrt{5})/8$ to $1/4$, the probability of getting correct results is equal to 1 with two Grover iterations, and when λ is greater than $1/4$, the probability of getting correct results is equal to 1 with only one Grover iteration.

In this letter, with application of two operators with arbitrary phase rotations in [3] and the phase matching conditions in [4], we will present a general Grover algorithm with fixed-phased shifts. The different algorithms can be derived from this algorithm according to the different phase shifts. When the phase shifts are fixed at 0.1π , the success probability of at least 99.38% can be obtained in $O(\sqrt{N/M})$ iterations.

2. Grover's Algorithm and Its Problem

2.1. Grover's Algorithm Summary

Suppose we wish to search through a search space of N elements. Rather than search the elements directly, we concentrate on the index to those elements, which is just a number in range 0 to $N-1$. For convenience we assume $N = 2^n$, so the index can be stored in n qubits, and that the search problem has exactly M solutions, with $1 \leq M \leq N$. The algorithm begins with the state $|0\rangle^{\otimes n}$. The Walsh-hadamard transform is used to put the state $|0\rangle^{\otimes n}$ in the equal superposition state,

$$|\varphi\rangle = \frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x\rangle. \quad (1)$$

The Grover quantum search algorithm then consists of repeated application of a quantum subroutine, known as the Grover iteration or Grover operator, which we denote G . The Grover iteration may be broken up into four steps.

1) Apply the oracle O . The oracle is a unitary operator defined by its action on the computational basis

$$O(|x\rangle|q\rangle) \rightarrow |x\rangle|q \oplus f(x)\rangle, \quad (2)$$

where $|x\rangle$ is the index register, \oplus denotes addition modulo 2, and the oracle qubit $|q\rangle$ is a single qubit

which is flipped if $f(x)=1$, and is unchanged otherwise.

2) Applying the Walsh-Hadamard transform $H^{\otimes n}$.

3) Perform a conditional phase shift, with every computational basis state except $|0\rangle$ receiving a phase shift of -1 , $|x\rangle \rightarrow -(-1)^{\delta_{x0}}|x\rangle$.

4) Applying the Walsh-Hadamard transform $H^{\otimes n}$.

It is useful to note that the combined effect of steps 2), 3), and 4) is

$$H^{\otimes n} (2|0\rangle\langle 0| - I) H^{\otimes n} = 2|\varphi\rangle\langle\varphi| - I, \quad (3)$$

where $|\varphi\rangle$ is the equally weighted superposition of states, (1). Thus the Grover iteration, G , may be written as $G = (2|\varphi\rangle\langle\varphi| - I)O$.

Let $\lambda = M/N$, and $\text{CI}(x)$ denote the integer closest to the real number x , where by convention we round halves down. Then repeating the Grover iteration

$$R = \text{CI} \left(\frac{\arccos(\sqrt{\lambda})}{2 \arcsin(\sqrt{\lambda})} \right) \quad (4)$$

times rotates $|\varphi\rangle$ to within an angle $\arcsin\sqrt{\lambda} \leq \pi/4$ of a superposition of marked states [12]. Observation of the state in the computational basis then yields a solution to the search problem with probability at least one-half.

2.2. Grover's Algorithm Success Probability

In fact, the Grover iteration can be regarded as a rotation in the two-dimensional space spanned by the starting vector $|\varphi\rangle$ and the state consisting of a uniform superposition of solutions to the search problem. Let $|\alpha\rangle$ represent a normalized states of a sum over all which are not solutions to the search problem, and $|\beta\rangle$ represent a normalized states of a sum over all which are solutions to the search problem. Simple algebra shows that the initial state $|\varphi\rangle$ may be re-expresses as

$$|\varphi\rangle = \cos(t)|\alpha\rangle + \sin(t)|\beta\rangle, \quad (5)$$

where $t = \arcsin\sqrt{\lambda}$. After R Grover iterations, the initial state is taken to

$$G^R |\varphi\rangle = \cos((2R+1)\arcsin\sqrt{\lambda})|\alpha\rangle + \sin((2R+1)\arcsin\sqrt{\lambda})|\beta\rangle. \quad (6)$$

Hence, the success probability is

$$P = \sin^2((2R+1)\arcsin\sqrt{\lambda}). \quad (7)$$

The curve of P is shown in **Figure 1**.

2.3. The Drawback of Grover's Algorithm

It is easy to deduce from Equations (4) and (7) that when $(3-\sqrt{5})/8 < \lambda < 0.14645$, P_λ decreases rapidly. When $0.14645 \leq \lambda < 1/4$, P_λ increases rapidly. When

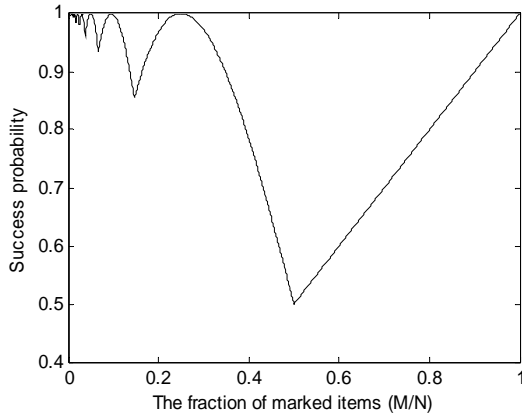


Figure 1. The success probability of Grover's algorithm.

$1/4 \leq \lambda < 1/2$, P_λ decreases rapidly. When $\lambda \geq 1/2$, there is $R=0$, $P_\lambda = \lambda$, and the algorithm is disabled. Hence, the Grover's algorithm is no longer useful when $\lambda > 1/4$.

The reason for the problem is that the two phase rotations in Grover iteration are fully equivalent in both amplitude and direction, namely π . According to [12], the result of such phase rotations is that, for the one Grover iteration, the phase of the $|\varphi\rangle$ increases $2 \arcsin \sqrt{\lambda}$ radians, and that rotating through at least $\arccos \sqrt{\lambda}$ radians takes the $|\varphi\rangle$ to $|\beta\rangle$. When $\lambda \in [3 - \sqrt{5}/8, 1)$, there are only $\lambda_1 = 1/4$ and $\lambda_2 = 3 - \sqrt{5}/8$ make R an integer, namely,

$$R_1 = \frac{\arccos \sqrt{\lambda_1}}{2 \arcsin \sqrt{\lambda_1}} = 1, \quad R_2 = \frac{\arccos \sqrt{\lambda_2}}{2 \arcsin \sqrt{\lambda_2}} = 2.$$

3. The Grover Algorithm with Fixed-Phase Shifts

3.1. The Searching Engine Description

Assume that $|s_1\rangle, |s_2\rangle, \dots, |s_M\rangle$ denote the marked states, and that $|t_1\rangle, |t_2\rangle, \dots, |t_{N-M}\rangle$ denote the non-marked states. Let

$$|s\rangle = \frac{1}{\sqrt{M}} \sum_{i=1}^M |s_i\rangle, |t\rangle = \frac{1}{\sqrt{N-M}} \sum_{i=1}^{N-M} |t_i\rangle,$$

$$\sin \theta = \sqrt{M/N},$$

and $0 < \theta \leq \pi/2$, then the initial system can be written as follows,

$$\begin{aligned} |\varphi^{(0)}\rangle &= \frac{1}{\sqrt{N}} \sum_{i=0}^{N-1} |i\rangle \\ &= \frac{\cos \theta}{\sqrt{N-M}} \sum_{i=1}^{N-M} |t_i\rangle + \frac{\sin \theta}{\sqrt{M}} \sum_{i=1}^M |s_i\rangle. \quad (8) \\ &= \cos \theta |t\rangle + \sin \theta |s\rangle \end{aligned}$$

Two phase shifting operator in Grover's algorithm may be generally expressed as follows,

$$I_s = I - (1 - e^{i\alpha}) |s\rangle \langle s|, \quad (9)$$

$$I_0 = I - (1 - e^{i\alpha}) |0\rangle \langle 0|. \quad (10)$$

Hence, the searching engine is described as follows,

$$\begin{aligned} G &= -HI_0H^+I_s \\ &= \begin{pmatrix} (1 - e^{i\alpha}) \cos^2 \theta - 1 & e^{i\alpha} (1 - e^{i\alpha}) \sin \theta \cos \theta \\ (1 - e^{i\alpha}) \sin \theta \cos \theta & e^{i\alpha} [(1 - e^{i\alpha}) \sin^2 \theta - 1] \end{pmatrix}, \quad (11) \end{aligned}$$

where H denotes Walsh-Hadamard operator.

3.2. The Searching Process Description

Quantum search is such a process of repeatedly applying the search engine G to the initial state $|\varphi^{(0)}\rangle$, which increase the probability amplitude of the marked state. After diagonalization, the successive operations of searching engine G can be written as follows [8],

$$G^n = Q\Lambda^n Q^+ = \frac{1}{C} \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}, \quad (12)$$

where

$$Q = \frac{1}{\sqrt{C}} \begin{pmatrix} (\cos(\alpha/2) \sin \theta + \cos \theta) e^{i(\alpha/2)} & \cos \theta \\ -\cos \theta & (\cos(\alpha/2) \sin \theta + \cos \theta) e^{-i(\alpha/2)} \end{pmatrix}, \quad (13)$$

$$\Lambda = \begin{pmatrix} -e^{i(\alpha-2\theta)} & 0 \\ 0 & -e^{i(\alpha+2\theta)} \end{pmatrix}, \quad (14)$$

$$C = \cos^2 \theta + (\cos(\alpha/2) \sin \theta + \cos \theta)^2, \quad (15)$$

$$\theta = \arcsin(\sin(\alpha/2) \sin \theta), \quad (16)$$

$$\begin{aligned} a_{11} &= -\cos^2(\theta) e^{i(n\alpha+2n\theta)} \\ &\quad - (\cos(\alpha/2) \sin \theta + \cos \theta)^2 e^{i(n\alpha-2n\theta)}, \quad (17) \end{aligned}$$

$$\begin{aligned} a_{12} &= \cos \theta (\cos(\alpha/2) \sin \theta + \cos \theta) \\ &\quad \cdot (e^{i(n\alpha+\alpha/2-2n\theta)} - e^{i(n\alpha+\alpha/2+2n\theta)}), \quad (18) \end{aligned}$$

$$\begin{aligned} a_{21} &= \cos \theta (\cos(\alpha/2) \sin \theta + \cos \theta) \\ &\quad \cdot (e^{i(n\alpha-\alpha/2-2n\theta)} - e^{i(n\alpha-\alpha/2+2n\theta)}), \quad (19) \end{aligned}$$

$$\begin{aligned} a_{22} &= -\cos^2(\theta) e^{i(n\alpha-2n\theta)} \\ &\quad - (\cos(\alpha/2) \sin \theta + \cos \theta)^2 e^{i(n\alpha+2n\theta)}. \quad (20) \end{aligned}$$

After n applications of G on $|\varphi^{(0)}\rangle$ we get,

$$|\varphi^{(n)}\rangle = G^n |\varphi^{(0)}\rangle = a^n |t\rangle + b^n |s\rangle. \quad (21)$$

such that,

$$\begin{aligned}
 & a^n \\
 &= \frac{1}{C} \left(\left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right) \right. \\
 &\quad \cdot \sin(2\theta) \sin(2n\theta) e^{i\left(n\alpha + \frac{\alpha}{2} \frac{\pi}{2}\right)} \\
 &\quad - \cos^3(\theta) e^{i(n\alpha + 2n\theta)} \\
 &\quad \left. - \cos \theta \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^2 e^{i(n\alpha - 2n\theta)} \right)
 \end{aligned} \tag{22}$$

$$\begin{aligned}
 & b^n \\
 &= \frac{1}{C} \left(2 \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right) \right. \\
 &\quad \cdot \cos^2 \theta \sin(2n\theta) e^{i\left(n\alpha - \frac{\alpha}{2} \frac{\pi}{2}\right)} \\
 &\quad - \sin \theta \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^2 e^{i(n\alpha + 2n\theta)} \\
 &\quad \left. - \sin \theta \cos^2(\theta) e^{i(n\alpha - 2n\theta)} \right)
 \end{aligned} \tag{23}$$

3.3. The Success Probability of the Proposed Algorithm

From Equations (21)-(23), after n iterations, the success probability P_s and the failure probability P_t can be respectively written as follows,

$$\begin{aligned}
 P_s &= |b^n|^2 \\
 &= \frac{1}{C^2} \left(4 \cos^4(\theta) \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^2 \sin^2(2n\theta) \right. \\
 &\quad + \sin^2(\theta) \left(\cos^4 \theta + \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^4 \right) \\
 &\quad + 4 \cos^2 \theta \sin \theta \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^3 \\
 &\quad \cdot \sin(2n\theta) \sin\left(2n\theta + \frac{\alpha}{2}\right) \\
 &\quad - 4 \cos^4 \theta \sin \theta \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right) \\
 &\quad \cdot \sin(2n\theta) \sin(2n\theta - \alpha/2) \\
 &\quad \left. + 2 \sin^2 \theta \cos^2 \theta \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^2 \cos(4n\theta) \right)
 \end{aligned} \tag{24}$$

$$\begin{aligned}
 P_t &= |a^n|^2 \\
 &= \frac{1}{C^2} \left(4 \sin^2 \theta \cos^2 \theta \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^2 \sin^2(2n\theta) \right. \\
 &\quad + \cos^2(\theta) \left(\cos^4 \theta + \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^4 \right) \\
 &\quad + 4 \cos^4 \theta \sin \theta \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right) \\
 &\quad \cdot \sin(2n\theta) \sin(2n\theta - \alpha/2) \\
 &\quad - 4 \cos^2 \theta \sin \theta \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^3 \\
 &\quad \cdot \sin(2n\theta) \sin(2n\theta + \alpha/2) \\
 &\quad \left. + 2 \cos^4 \theta \left(\cos\left(\frac{\alpha}{2}\right) \sin \theta + \cos \theta \right)^2 \cos(4n\theta) \right)
 \end{aligned} \tag{25}$$

and meet the following relationship $P_s + P_t = 1$.

3.4. The Required Number of Iterations of the Proposed Algorithm

According to [4], the searching engine G can be thought as a rotation in a 3-dimensional space, and the polarization vector of marked state is $(0,0,1)$. For the initial state, the polarization vector is about $(0,0,-1)$, each searching iteration is a rotation of the polarization vector through angle $4 \arcsin(\sin(\alpha/2) \sin \theta)$. Therefore, the required number of iterations to get a match with the highest possible probability can be calculated as follows,

$$R = \left\lfloor \frac{\pi}{4 \arcsin(\sin(\alpha/2) \sin \theta)} \right\rfloor, \tag{26}$$

where $\lfloor \cdot \rfloor$ is the floor operation. When $M \gg N$,

$$R \approx \left\lfloor \frac{\pi}{4 \sin(\alpha/2) \sqrt{M/N}} \right\rfloor = O(\sqrt{N/M}). \tag{27}$$

3.5. The Relation between the Proposed Algorithm and the Other Algorithms

According to the different values of α in Equation (26), the proposed algorithm can be transformed into the different algorithms. For instance, when $\alpha = \pi$,

$$R = \left\lfloor \frac{\pi}{4 \arcsin(\sin(\alpha/2) \sin \theta)} \right\rfloor \approx \left\lfloor \frac{\pi \sqrt{N}}{4 \sqrt{M}} \right\rfloor, \tag{28}$$

in this case, the proposed algorithm is equivalent to the Grover's original algorithm.

When $\alpha = \pi/2$ or $\alpha = 3\pi/2$,

$$R = \left\lceil \frac{\pi}{4 \arcsin(\sin(\alpha/2) \sin \theta)} \right\rceil \approx \left\lceil \frac{\pi}{2\sqrt{2}} \sqrt{\frac{N}{M}} \right\rceil, \quad (29)$$

in this case, the proposed algorithm is equivalent to the algorithm in [7].

When $\alpha = 1.825\pi$ or $\alpha = 0.175\pi$,

$$R = \left\lceil \frac{\pi}{4 \arcsin(\sin(\alpha/2) \sin \theta)} \right\rceil \approx \left\lceil \frac{1.825\pi}{2 \sin \theta} \right\rceil, \quad (30)$$

in this case, the proposed algorithm is equivalent to the algorithm in [9].

It can be seen from above depiction that the different algorithms can be achieved from the proposed algorithm according the different values of α . The above result shows that the proposed algorithm is the promotion of other search algorithms, it is the more flexible algorithm, and other algorithms can be seen as a special case of this algorithm.

In the proposed algorithm, if two phase shifts are fixed at 0.1π , then the required number of iterations is

$$R = \left\lceil \frac{\pi}{4 \arcsin(\sin(\alpha/2) \sin \theta)} \right\rceil \approx \left\lceil 5\sqrt{\frac{N}{M}} \right\rceil. \quad (31)$$

In this case, examine the system after $R \approx \left\lceil 5\sqrt{M/N} \right\rceil$, the success probability will be at least 99.38% compared with 98% for Younes [9], 84.72% for Younes *et al.* [7], and 50% for the Grover's original algorithm, which is better than any know fixed operator quantum search algorithms.

When $\alpha = 0.01\pi$ or $\alpha = 1.99\pi$,

$$R = \left\lceil \frac{\pi}{4 \arcsin(\sin(\alpha/2) \sin \theta)} \right\rceil \approx \left\lceil 50\sqrt{\frac{N}{M}} \right\rceil, \quad (32)$$

Figure 2 shows the success probability as a function of the ratio M/N for the proposed algorithm after inserting the calculated number of iterations R shown in Equation (26) in P_s shown in Equation (24), where the phase shifts are set to $\pi, \pi/2, 1.825\pi$, and 0.1π , respectively.

4. Performance Comparison

When $N = 10^2, 10^3, 10^4, 10^5, 10^6$ and $0 < M/N \leq 1$, the performance comparison of the proposed algorithm with Younes algorithm [9], Younes algorithm [7] and the Grover's original algorithm is shown in **Tables 1-5**.

From **Tables 1-5**, we can see that the success probability of proposed algorithm is the highest among four algorithms; however, the inquired number of iterations is the greatest. Despite this, from Equation (27), it can still achieve quadratic speed up over classical search algo-

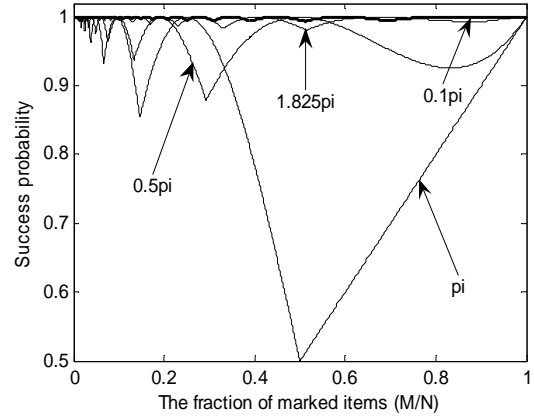


Figure 2. The success probability curves of the proposed algorithm with the different fixed-phases after the required number of iterations.

Table 1. The performance comparison of four algorithms ($N = 10^2, \alpha = 0.1\pi$).

	Prop.Alg	Younes [9]	Younes [7]	Grover's Alg
P_{\min}	0.9942	0.9819	0.8645	0.5000
$(M/N)_{P=P_{\min}}$	0.5100	0.5200	0.3000	0.5000
R_{\max}	50	28	11	7

Table 2 The performance comparison of four algorithms ($N = 10^3, \alpha = 0.1\pi$).

	Prop.Alg	Younes [9]	Younes [7]	Grover's Alg
P_{\min}	0.9939	0.9801	0.8481	0.5000
$(M/N)_{P=P_{\min}}$	0.5120	0.5140	0.3080	0.5000
R_{\max}	158	90	35	24

Table 3. The performance comparison of four algorithms ($N = 10^4, \alpha = 0.1\pi$).

	Prop.Alg	Younes [9]	Younes [7]	Grover's Alg
P_{\min}	0.9938	0.9800	0.8473	0.5000
$(M/N)_{P=P_{\min}}$	0.5123	0.5137	0.3084	0.5000
R_{\max}	502	286	111	78

Table 4. The performance comparison of four algorithms ($N = 10^5, \alpha = 0.01\pi$).

	Prop.Alg	Younes [9]	Younes [7]	Grover's Alg
P_{\min}	0.9999	0.9799	0.8472	0.5000
$(M/N)_{P=P_{\min}}$	0.4960	0.5136	0.3084	0.5000
R_{\max}	15812	906	351	248

Table 5. The performance comparison of four algorithms ($N = 10^6$, $\alpha = 0.01\pi$).

	Prop.Alg	Younes [9]	Younes [7]	Grover's Alg
P_{\min}	0.9999	0.9799	0.8472	0.5000
$(M/N)_{P=P_{\min}}$	0.4960	0.5136	0.3084	0.5000
R_{\max}	50002	2866	1110	785

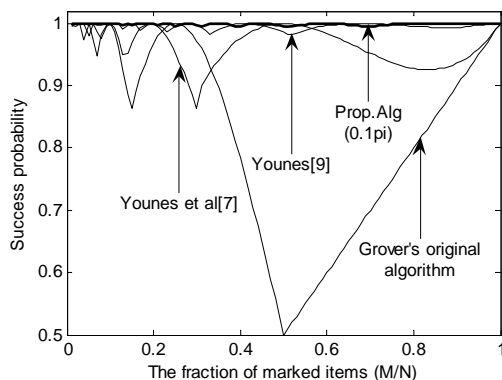
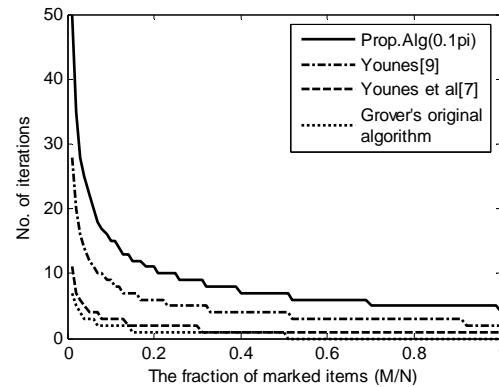
rithms. When $N=10^2$ and $\alpha=0.1\pi$, from Equation (31), $R_{\max} \approx 5\sqrt{N} = N/2 = 50$, this algorithm does not seem to have achieved the quadratic speed up of Grover's original algorithm. However, when $N=10^6$ and $\alpha=0.01\pi$, $R_{\max} \approx 50\sqrt{N} = N/20 \approx 50,002$, this algorithm has demonstrated good acceleration role, and with the increase of N , this kind of speed up will be increasingly significant.

These results can be analyzed as follows. As the saying goes, there is no free lunch. The proposed scheme obtains high successful probability at the expense of the number of iterations. When the phase shifts are fixed at 0.1π , the success probability of at least 99.38% can be obtained, and when the phase shifts are fixed at 0.01π , the success probability of at least 99.99% can be obtained after more iteration. Since the phase shifts α is set to a adjustable parameter in Equation (11), the proposed scheme is more flexible, and it can obtain a balance between speed and accuracy.

When $N=10^2$, the probability curves of four search algorithms are shown in **Figure 3**, and the inquired iterations number curves are shown in **Figure 4**.

5. Conclusion

In the future quantum computers, the search engine should be constructed from fixed operators that can handle the whole possible range of the search problem, *i.e.* whether a single match or multiple matches exist in the search space. Based on this consideration, in this letter, we pre-

**Figure 3. The success probability distribution of four quantum search algorithms after the required number of iterations.****Figure 4. The required number of iterations for the proposed algorithm, Younes algorithm [9], Younes *et al.* algorithm [7] and Grover's original algorithm.**

sent a fixed-phase quantum search algorithm with more flexible behavior. The phase shifts can be fixed at the different values to meet the needs of different practical problems. If research requires a relatively rapid speed, the value of the phase shifts should be appropriately increased, if search requires a higher success probability, the value of the phase shifts should be appropriately decreased. When the phases shifts are fixed at 0.1π , the proposed algorithm can get a solution with probability at least 99.38% in $\lfloor 5\sqrt{N/M} \rfloor$ iterations.

6. Acknowledgements

This paper was supported by Scientific Research Foundations of Heilongjiang Provincial Education Department (Grant No. 12511009).

REFERENCES

- [1] L. K. Grover, "A Fast Quantum Mechanical Algorithm for Database Search," *Proceedings of the 28th Annual ACM Symposium on the Theory of Computing*, Philadelphia, 22-24 May 1996, pp. 212-221.
- [2] L. K. Grover, "Quantum Computers Can Search Rapidly by Using Any Transformation," *Physical Review Letters*, Vol. 80, No. 19, 1998, pp. 4329-4332. [doi:10.1103/PhysRevLett.80.4329](https://doi.org/10.1103/PhysRevLett.80.4329)
- [3] G. L. Long, Y. S. Li and W. L. Zhang, "Phase Matching in Quantum Searching," *Physics Letters A*, Vol. 262, No. 1, 1999, pp. 27-34. [doi:10.1016/S0375-9601\(99\)00631-3](https://doi.org/10.1016/S0375-9601(99)00631-3)
- [4] G. L. Long, X. Li and Y. Sun, "Phase Matching Condition for Quantum Search with a Generalized Initial State," *Physics Letters A*, Vol. 294, No. 3-4, 2002, pp. 143-152. [doi:10.1016/S0375-9601\(02\)00055-5](https://doi.org/10.1016/S0375-9601(02)00055-5)
- [5] L. K. Grover, "Fixed-Point Quantum Search," *Physical Review Letters*, Vol. 95, 2005, Article ID: 150501.
- [6] D. F. Li, X. X. Li and H. T. Huang, "Phase Condition for the Grover Algorithm," *Theoretical and Mathematical Physics*, Vol. 144, No. 3, 2005, pp. 1279-1287. [doi:10.1007/s11232-005-0159-x](https://doi.org/10.1007/s11232-005-0159-x)

- [7] A. Younes, *et al.*, “Quantum Search Algorithm with More Reliable Behavior Using Partial Diffusion,” *Quantum Communication*, Vol. 734, 2006, pp. 171-174.
- [8] G. L. Long, “Grover Algorithm with Zero Theoretical Failure Rate,” *Physical Review A*, Vol. 64, No. 2, 2001, Article ID: 022307. [doi:10.1103/PhysRevA.64.022307](https://doi.org/10.1103/PhysRevA.64.022307)
- [9] A. Younes, “Fixed Phase Quantum Search Algorithm,” 2007. <http://arxiv.org/pdf/0704.1585.pdf>
- [10] C. Cafaro and S. Mancini, “On Grover’s Search Algorithm from a Quantum Information Geometry Viewpoint”. *Physica A*, Vol. 391, No. 4, 2012, pp. 1610-1625.
- [11] H. Y. Li, C. W. Wu, W. T. Liu, *et al.*, “Fast Quantum Search Algorithm for Databases of Arbitrary Size and Its Implementation in a Cavity QED System,” *Physics Letters A*, Vol. 375, No. 48, 2011, pp. 4249-4254. [doi:10.1016/j.physleta.2011.10.016](https://doi.org/10.1016/j.physleta.2011.10.016)
- [12] M. A. Nielsen and I. L. Chuang, “Quantum Computation and Quantum Information,” Cambridge University Press, Cambridge, 2000.