Quantum associative memory with distributed queries

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Abstract

This paper discusses a model of quantum associative memory which generalizes the completing associative memory proposed by Ventura and Martinez. Similar to this model, our system is based on Grover’s well known algorithm for searching an unsorted quantum database. However, the model presented in this paper suggests the use of a distributed query of general form. It is demonstrated that spurious memories form an unavoidable part of the quantum associative memory model; however, the very presence of these spurious states provides the possibility of organizing a controlled process of data retrieval using a specially formed initial state of the quantum database and also of the transformation performed upon it. Concrete examples illustrating the properties of the proposed model are also presented.

Keywords: quantum associative memory, distributed query, spurious memory

1 Introduction

Development of efficient and/or biologically plausible models of associative memory forms a very important part of neurocomputing research [1]. Among the various approaches to this problem Hopfield’s model of content-addressable memory [2], Kosko’s bidirectional autoassociative memory (BAM) [3] and Pollack’s Recursive Auto-Associative Memory (RAAM) [4] are most widely known and comprehensively studied. One of the most important characteristics of neural models of associative memory is capacity. Unfortunately, the capacity of existing neural memories grows only linearly with the number of neurons in a network (bits in a pattern) [1].

Recently, studies in quantum computation have been enriched by new works addressing the idea of developing quantum neural networks [5-8]. These quantum neural networks have many promising characteristics, both in the case of supervised and unsupervised learning. In particular, an associative memory based on the use of Grover’s quantum search algorithm [9] has been proposed by Ventura and Martinez [10-12]. This network solves the completion problem; that is, it can restore the full pattern when initially presented with just a part of that pattern. One of the most attractive properties of this memory is its exponential capacity.

This paper presents a generalization of this model which is able to retrieve a memory state when presented with a corrupted (noisy) version of the pattern. So, the model presented here solves the problem of associative search for which no part of the input stimulus is guaranteed to be noise free.

The second section briefly outlines Grover’s algorithm and discusses its main features and possible interpretations. It is argued that this algorithm in some sense already solves the completion problem for the simplest case of a database having a full set of coding fragments in its stored patterns. The third section gives a short review of the model of quantum associative memory proposed by Ventura and Martinez and discuss its peculiarities. The fourth section generalizes Grover’s algorithm to the case of distributed query and derives an analytical solution for the iterated amplitudes of the basis states whose
superposition describes the state of a quantum database. The fifth section describes a generalized quantum associative memory which utilizes Grover’s algorithm and compares two distinct approaches to its implementation. The second method, which uses the trick of exchanging stored memories for spurious ones, provides controlled iteration of the system and reaches a quantum state for which the probabilities of retrieving a given memory state take prescribed values defined by the form of the distributed query. Analytical solutions for the iterated state and its different averages are used to analyze these variants of the presented model. The conclusion summarizes the main results presented in the article.

2 Grover’s algorithm

This algorithm, proposed by Grover [9], effectively searches for an entry in an unsorted quantum database of \( N \) entries, requiring only \( O(\sqrt{N}) \) steps to produce the appropriate item with high probability. This compares favorably with the classical lower bound of \( N/2 \) for the number of queries needed to find an item in an unsorted classical database containing \( N \) records. Further, it has been proved that the efficiency of Grover’s algorithm is optimal for quantum search [13].

Here, the algorithm is briefly described so that some kind of equivalence between Grover’s approach and the quantum associative memory proposed in [10,11] may be established. Consider the database (phone book) containing the four records shown in Table 1. Each record consists of two parts – name and phone number. The records are ordered by name and unordered with respect to the phone numbers. The direct problem of searching for the associated phone number given a name is obviously very easy. On the contrary, the inverse problem of searching for the associated name given a phone number is difficult. The most efficient classical solution is random search which demands in the worst case \( N-1 \) queries and on average \( N/2 \) queries.

### Table 1.
The records in a phone book are ordered by names coded using five bits. Phone numbers (unordered) are coded using two bits. In the quantum analog of a phone book bits are replaced by qubits and the states of qubits corresponding to names and numbers are entangled.

<table>
<thead>
<tr>
<th>Name</th>
<th>Code</th>
<th>Number</th>
<th>Code</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>00001</td>
<td>3</td>
<td>11</td>
</tr>
<tr>
<td>B</td>
<td>00010</td>
<td>0</td>
<td>00</td>
</tr>
<tr>
<td>C</td>
<td>00011</td>
<td>2</td>
<td>10</td>
</tr>
<tr>
<td>D</td>
<td>01010</td>
<td>1</td>
<td>01</td>
</tr>
</tbody>
</table>

This classical random search can be thought of as testing each entry with probability \( N^{-1} \). This situation may be described in a quasi-quantum manner. Before testing every entry, NameFinder is in the indefinite state \( |\psi\rangle \), which can be represented as a superposition of all possibilities:

\[
|\psi\rangle = \sum_{\text{name, number}} \frac{1}{N} |\text{name, number}\rangle
\]

Each testing corresponds to the collapse of the superposition (1) to one of its items with equal probability \( N^{-1} \) (this collapsing just means that NameFinder makes a decision to explore a particular entry in the phone book). In a classical statement of the problem, the coefficients of each entry in (1) are real numbers (probabilities) and the entries do not interfere. Also the state of NameFinder \( |\psi\rangle \) does not evolve during the search process.

In a truly quantum approach to the problem, the state of NameFinder can be described as a superposition of quantum basis states (wavefunction).
where \( c_k \) are complex numbers, and the probability of the system to be found after the measurement in a given basis state \( k \) (probability to choose entry \( k \)) is given by \( |c_k|^2 \). The full database record \( |\text{name, number}\rangle \) is represented by two parts (sub-records) consisting of sets of qubits which code the name and phone number respectively. For example, in Table 1 each phone number can be coded by two qubits and each name (one character) by five qubits (sufficient to represent any character in the English alphabet). The states of these two parts of the record are entangled. This means that if the wavefunction collapses to the basis state having a given phone number value, the measurement of the state of the qubits which encode the name will produce the name corresponding to that phone number.

Typically, the first sub-record of qubits corresponding to the name is omitted and the problem is formulated as the search for the entry with a given (marked) second record \( |\text{number}\rangle \) value known to Oracle. In other words, Oracle knows the number and asks NameFinder to find the corresponding name in his phone book. So in the language of associative memory, Oracle presents an external stimulus and NameFinder retrieves a memory item. Further, the wavefunction \( |\psi\rangle \) will be referred to as describing the state of NameFinder in a database or in a memory. These two cases are differentiated only for convenience in order to specify correspondingly the search for a record in a database having an enormous number (all \( 2^d \) patterns consisting of \( d \) qubits) of data items, for which Grover’s algorithm was originally developed, and in a memory, which ordinarily includes a rather more restricted number of patterns (\( p \ll 2^d \)).

The essence of the quantum approach to the search is constructing an iteration scheme consisting of transformations performed by Oracle and NameFinder which change the NameFinder wavefunction in such a manner that the amplitude of the marked state (stimulus) increases and the amplitudes of other states decreases. Then the collapse of NameFinder’s wave function (choice of entry in the database) will give, with high probability, the desired marked entry. Finally, the entanglement of the states of the two parts of the record will permit the retrieval of the desired associated name. Formally, this algorithm can be described as follows:

- Initialize the state of NameFinder-in-Database \( |\psi\rangle = \sum_{x=0}^{2^d-1} a_x |x\rangle \) as an equiprobable superposition \( |s\rangle \) of all entries:

\[
|\psi^{(0)}\rangle = |s\rangle = \sum_{x=0}^{2^d-1} \frac{1}{\sqrt{2^d}} |x\rangle
\]

where \( d \) denotes the number of qubits used to represent a phone number and \( |x\rangle \) denotes a basis state whose binary form corresponds to the phone number \( x \).

- Iterate the state \( |\psi\rangle \) using the sequence of two transformations \( U_p \) and \( U_s \):
  1. \( U_p \): Oracle inverts the phase of marked state \( |p\rangle \):

\[
a_x \rightarrow \begin{cases} 
-a_x, & \text{if } x = p \\
 a_x, & \text{if } x \neq p 
\end{cases}
\]

2. \( U_s \): NameFinder-in-Database inverts the amplitudes of all states around their average value:

\[
a_x \rightarrow 2\langle a \rangle - a_x
\]
where
\[ \langle a \rangle = 2^{-d} \sum_{x=0}^{2^d} a_x \]

It has been shown [15] that after \( T = O(\sqrt{N}) \) \( (N = 2^d) \) iterations the amplitude \( a_p \) becomes very close to unity while the amplitudes of the other states almost vanish. At this point, measuring the state of \( \text{NameFinder-in-Database} \) will, with high probability, produce that basis state \( | p \rangle \) and the entangled name coded in the remaining qubits can be directly retrieved.

Note that the database considered here is complete: it contains all possible \( 2^d \) phone numbers. So, the collapse of the system into any basis state will provide a valid database record. This situation corresponds to the absence of spurious memories in quantum models of associative memory.

### 3 Completing quantum associative memory

A quantum associative memory with a capacity exponential in the number of qubits and based on Grover’s algorithm has been proposed by Ventura and Martinez [10-12]. This kind of memory solves the problem of pattern completion. It can restore the full pattern when presented with only a partial one. It is crucial that this partial pattern would exactly coincide with some part of a valid full pattern. Then, in the recall phase only the remainder of the pattern is reconstructed while the initially presented partial pattern remains intact. This kind of associative memory clearly differs from the general statement of associative search. Indeed, general associative memory should also retrieve valid memory items when presented with noisy versions of these patterns. In fact, the original Grover’s algorithm also solves the completion problem – in the interpretation described above it retrieves a full pattern \( (\text{name, number}) \) when presented with a partial pattern \( (?, \text{number}) \).

The main difference between the quantum associative memory developed in [10-12] and Grover’s search algorithm of an unsorted database is that for the quantum associative memory the number of entries is smaller than \( 2^d \) and they form a set \( M \) of so called memory states. This is an important distinction for at least two reasons. First, while the creation of a quantum superposition containing all \( 2^d \) possible basis states is a straightforward operation [9], the creation of a quantum superposition containing only those basis states that correspond to valid memories is a nontrivial task, the solution to which is detailed in [14]. Second, the superposition of all \( 2^d \) basis states is a special case of the general wave function and researchers have to this point had difficulty discovering a practical use for Grover’s algorithm, theoretically spectacular as it is. The quantum associative memory of [10-12] provides perhaps the first such practical application.

Hence in the case of the quantum associative memory, the initial state of \( \text{NameFinder-in-Memory} \) is described as a superposition of these states

\[ |\psi^{(0)}\rangle = |m\rangle = \sum_{x \in M} \frac{1}{\sqrt{P}} |x\rangle \]  \hspace{1cm} (6)

where \( P \) denotes the number of memorized patterns.

Also, the transformation performed by \( \text{NameFinder-in-Memory} \) now suggests the inversion of the amplitudes around the average of only the amplitudes of memorized patterns (zero amplitudes of the other ones play no role) [10]:

\[ 2'. \ U_s \Rightarrow U_m': \ \text{NameFinder-in-Memory} \ inverts \ amplitudes \ of \ all \ states \ around \ the \ average \ value \ for \ memorized \ patterns. \]

\[ a_x \rightarrow 2\langle a \rangle_M - a_x \]  \hspace{1cm} (7)
\[ \langle a \rangle_M = \frac{1}{P} \sum_{x \in M} a_x \]

(Note, however, that in other versions of the model the original transformation \( U_s \) is used [11-12]). Further, it is assumed that Oracle knows a part of one of the memorized patterns and no other pattern has the same part. Then Grover’s algorithm is used to find the pattern having this part and entanglement permits the restoration of the remainder of this pattern.

3.1 Completing versus general associative memory

In general, associative search suggests the possibility of retrieving valid memory items when presented with any possible external stimulus, including noisy stimuli. It is desirable to retrieve the memory state which is most similar to the given stimulus, i.e. that memory state which differs from the stimulus in the minimal number of bits (qubits). In the binary case, this corresponds to the minimizing the Hamming distance between query and memory state. So, this kind of memory is a correcting memory rather than a completing one. However, the quantum associative memory proposed in [10-12] does not actually take into account the distance between states but only uses information about the presence of some prescribed bit values in the memory state. Moreover, the quantum state describing the simplest form of completing memory [10] will not evolve while performing iterations if the query (part of memory item) presented by Oracle does not coincide with some part of a valid memory state. In order to overcome these difficulties, it is possible to introduce a metric into the quantum search algorithm in the form of distributed queries. Further, this complication of a query permits a quantum state to evolve in the case of the use of the transformation \( U_m \); however it also leads to the appearance of spurious memories. Hence, while the completing associative memory proposed by Ventura and Martinez can be free from spurious memories (and can also contain them in other model variants [11-12]), general correcting associative memory should contain spurious states. Really, the appearance of spurious states in the model proposed in [12] is due to the traditional form of the transformation \( U_s \). In a quantum associative memory of the general form discussed here, spurious memories arise even in the case when the transformation \( U_m \) is used. (Tables 2 and 3 illustrate the difference between these two situations.) The following discussion will be restricted to considering the case of the use of the transformation \( U_m \) in constructing a scheme for a general quantum associative memory. More complex cases will be considered in the future.

Distributed query means a query, or stimulus to the system, has the form of a superposition, just as a quantum memory does,

\[ |b^p\rangle = \sum_{x=0}^{2^d-1} b^p_x |x\rangle \tag{8} \]

which in general includes all basis states. (Index \( p \) marks one of these states, \( |p\rangle \), which plays the role of the center of the distribution). However, the introduction of distributed query demands the modification of the memory (phone book) in such a manner that it has every possible phone number (basis state), despite the fact that most of them have no corresponding name.

Table 2. A completing associative memory can use a database (phone book) free from spurious memory (it contains only some of the possible numbers) and any query must represent an exact part of a phone number in the database, for example (*11), (01*) etc. Any query of a form similar to (10*) or (001) etc. will not cause any evolution of the quantum state describing a completing memory if transformation \( U_m \) is used.

<table>
<thead>
<tr>
<th>Record</th>
<th>Name</th>
<th>Phone</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Alice</td>
<td>010</td>
</tr>
</tbody>
</table>
Table 3. The general (correcting) form of associative memory which uses distributed queries suggests that the database (phone book) includes a full set of numbers (8) but that some of them are not used, i.e. correspond to spurious memories. Collapse of the wave function into one of the basis states corresponding to such a spurious memory does not provide useful information.

<table>
<thead>
<tr>
<th>Record</th>
<th>Name</th>
<th>Phone</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>not used</td>
<td>000</td>
</tr>
<tr>
<td>2</td>
<td>not used</td>
<td>001</td>
</tr>
<tr>
<td>3</td>
<td>Alice</td>
<td>010</td>
</tr>
<tr>
<td>4</td>
<td>not used</td>
<td>011</td>
</tr>
<tr>
<td>5</td>
<td>not used</td>
<td>100</td>
</tr>
<tr>
<td>6</td>
<td>not used</td>
<td>101</td>
</tr>
<tr>
<td>7</td>
<td>not used</td>
<td>110</td>
</tr>
<tr>
<td>8</td>
<td>Bob</td>
<td>111</td>
</tr>
</tbody>
</table>

This modification introduces the possibility of the quantum state collapsing into a basis state corresponding to an entry having the code “not used” in the Name field. In other words, such a memory will have so-called spurious memory states. However, before considering such a generalized associative memory, the application of Grover’ algorithm to the case when Oracle defines not a single query (marked state) or finite set of such states [15] but rather defines a distributed fuzzy query.

First, suppose that in distributed query (8) real amplitudes are distributed such that the maximal value occurs for some definite state $|x\rangle = |p\rangle$ $\| b^p_x \|^2 = \max_x \| b^p_x \|^2$, and the amplitudes of the other basis states decrease monotonically with Hamming distance $|x - p |$. From here on, $|p\rangle$ shall be referred to as the query center.

One way to satisfy these conditions follows from the binomial distribution

$$
|b^p_x|^2 = q^{p-x}(1-q)^{d-|p-x|}
$$

where $|p-x|$ denotes Hamming distance between $|p\rangle$ and $|x\rangle$, $d$ is the number of qubits needed to code a phone number and $0 < q < 1/2$ is an arbitrary value which tunes the width of the distribution. For example, $d = 2; q = ¼; |p\rangle = |11\rangle$, produces the following distributed query

$$
|b^p\rangle = \frac{3}{4}|11\rangle + \frac{\sqrt{3}}{4}|01\rangle + \frac{\sqrt{3}}{4}|10\rangle + \frac{1}{4}|00\rangle
$$

It is important to note once more that introducing a distributed query with Hamming distance-dependent amplitudes for the basis states incorporates a metric into the model which permits comparison of the similarity of the stimulus and the retrieved memory. This is a necessary condition for associative searching. For this type of query the transformation performed by Oracle will have the form

$$
U_b = 1 - 2\langle b^p | b^p \rangle
$$

It is well known that in the case of the traditional Grover’s algorithm the transformation performed by NameFinder-in-Database inverts the amplitudes of the basis states around their mean value.
\[ a_x \rightarrow 2\langle a \rangle - a_x \]  

(11)

In the case of the simple *completing* associative memory [10], *only memory states* are used in building the transformation performed by a network

\[ U_m = 2 |m\rangle\langle m| - 1 \]  

(12)

Correspondingly, this transformation inverts the amplitudes of the memory states (their equiprobable superposition ordinarily forms the initial quantum state) around the average value of *only these memories* (remember, no additional spurious memories arise).

In the case of a distributed query, the *Oracle* transformation is defined by

\[ U_b = 1 - 2 |b^p\rangle\langle b^p| \]

\[ U_b : \psi \rightarrow \psi - 2 |b^p\rangle\langle b^p| \psi \]  

(13)

Since

\[ \langle b^p | \psi \rangle = \sum_{x} b_x^p \langle x| \sum_{y} a_y \rangle y \]

\[ = \sum_{x} b_x^p \sum_{y} a_y \langle x| y \rangle = \sum_{x} b_x^p \sum_{y} a_y \delta_{xy} = \sum_{x} a_x b_x^p \equiv \langle a | b^p \rangle \]  

(14)

where \( \delta \) is the Kronecker delta function and \( \langle a | b^p \rangle \) represents an overlapping of the current quantum state with the query state,

\[ \psi - 2 |b^p\rangle\langle b^p| \psi \]  

\[ = \sum_{x} a_x |x\rangle - 2 |b^p\rangle \langle a | b^p \rangle = \sum_{x} \left( a_x - 2 \langle a | b^p \rangle b_x^p \right) |x\rangle \]

or

\[ a_x \rightarrow a_x - 2\langle a | b^p \rangle b_x^p \]  

(15)

Note, that expression (15) can be transformed into the classical Grover’s transformation \( U_s \) if \( b_x^p = \delta_{xp} \).

In this case \( \langle a | b^p \rangle = \sum_x a_x b_x^p = \sum_x a_x \delta_{xp} = a_p \) and (15) takes a familiar form:

\[ a_x \rightarrow a_x - 2a_p \delta_{xp} \]  

i.e. \( a_x = \begin{cases} a_x & \text{if } x \neq p \\ -a_x & \text{otherwise} \end{cases} \)

4 Grover’s search in unsorted database with distributed query

Before considering the quantum associative memory with distributed queries, a generalization of Grover’s original algorithm in the context of distributed queries will be presented. In general, a quantum database can have an arbitrary initial quantum state. In fact, the algorithm proposed in [14] is used to create database states representing quantum memory of arbitrary sets of patterns. How will Grover’s original algorithm perform with such quantum states using a distributed query model?

4.1 Deriving the equations for averages
If for some iteration $\tau$ the state of the system is described by the superposition

$$\psi^{(r)} = \sum_x a_x^{(r)} | x >$$

(16)

then after the transformation $U_b$ (first sub-step of an iteration), the superposition becomes

$$\psi^{(r+1/2)} = \sum_x a_x^{(r+1/2)} | x > = \sum_x \left( a_x^{(r)} - 2a | b^p \rangle^{(r)} b^p_x \right) | x >$$

(17)

After $U_s$, the transformation performed by the NameFinder-in-Memory (second sub-step of an iteration), which inverts amplitudes around their average value,

$$\psi^{(r+1)} = \sum_x (2a^{(r+1/2)} - a_x^{(r+1/2)}) | x > = \sum_x \left( a_x^{(r)} - 2a | b^p \rangle^{(r)} b^p_x - a_x^{(r)} + 2a | b^p \rangle^{(r)} b^p_x \right) | x >$$

(18)

Thus, one iteration causes the following change of amplitudes:

$$a_x^{(r+1)} = 2a^{(r)} - 2a | b^p \rangle^{(r)} b^p_x - a_x^{(r)} + 2a | b^p \rangle^{(r)} b^p_x$$

(19)

or, equivalently,

$$a_x^{(r+1)} = 2a^{(r)} - 4a | b^p \rangle^{(r)} b^p_x - a_x^{(r)} + 2a | b^p \rangle^{(r)} b^p_x$$

(20)

Note, that Grover’s original iteration scheme follows from (20) if

$$\langle a | b^p \rangle = \sum_x a_x b^p_x = \sum_x a_x \delta_{xp} = a_p$$

$$\langle b^p \rangle = \frac{1}{d} \sum_x b^p_x = \frac{1}{d} \sum_x \delta_{xp} = \frac{1}{d}$$

(21)

Then, $a_x^{(r+1)} = 2\langle a \rangle^{(r)} - 4N^{-1}a_p^{(r)} - a_x^{(r)} + 2a_p^{(r)} \delta_{xp}$

(22)

where $N = 2^d$ denotes all possible states for a register consisting of $d$ qubits. The particular case of initial state $|\psi\rangle = \frac{1}{2}(|00\rangle + |01\rangle + |10\rangle + |11\rangle)$, means that $\langle a \rangle^{(0)} = 1/2$ and for any $p$

$$a_p^{(1)} = \begin{cases} 2 \cdot \frac{1}{2} - (1 + 4N^{-1}) \cdot \frac{1}{2} = 0 & \text{if } x \neq p \\ 2 \cdot \frac{1}{2} + (1 - 4N^{-1}) \cdot \frac{1}{2} = 1 & \text{otherwise} \end{cases}$$
and so follows the well-known result that for these conditions any query will transform the state of the system to that of the marked state (find the requested number in the phone book) after only one iteration.

Now, starting from Equation (20) it is possible to obtain a closed system of two equations for the average values \( \langle a \rangle \) and \( \langle a | b^p \rangle \) (the approach is analogous to one used in [16]). Multiplying this equation by \( N^{-1} \) and adding the terms corresponding to all basis states results in

\[
\langle a \rangle^{(r+1)} = 2\langle a \rangle^{(r)} - 4\langle a | b^p \rangle^{(r)} \langle b^p \rangle - \langle a \rangle^{(r)} + 2\langle a | b^p \rangle^{(r)} \langle b^p \rangle
\]

or, equivalently,

\[
\langle a \rangle^{(r+1)} = \langle a \rangle^{(r)} - 2\langle a | b^p \rangle^{(r)} \langle b^p \rangle
\]  

(23)

Multiplying each of the Equations (20) by its corresponding \( b^p \) value and summing over \( x \) produces

\[
\langle a | b^p \rangle^{(r+1)} = 2\langle a \rangle^{(r)} N\langle b^p \rangle - 4\langle a | b^p \rangle^{(r)} N\langle b^p \rangle^2 - \langle a | b^p \rangle^{(r)} + 2\langle a | b^p \rangle^{(r)} \sum_x (b^p_x)^2
\]

Taking into account that vector \( b^p = (b^p_0, ..., b^p_N) \) has a unity norm simplifies the last expression as

\[
\langle a | b^p \rangle^{(r+1)} = 2N\langle b^p \rangle\langle a \rangle^{(r)} + (1 - 4N\langle b^p \rangle^2)(\langle a | b^p \rangle^{(r)})
\]  

(24)

Finally, a closed system for the averages can be written as follows:

\[
\begin{aligned}
\langle a \rangle^{(r+1)} &= \langle a \rangle^{(r)} - 2\langle a | b^p \rangle^{(r)} \\
\langle a | b^p \rangle^{(r+1)} &= 2N\langle b^p \rangle\langle a \rangle^{(r)} + (1 - 4N\langle b^p \rangle^2)(\langle a | b^p \rangle^{(r)})
\end{aligned}
\]  

(25)

In order to reduce this to Grover’s original scheme, assign \( \langle a | b^p \rangle = a_p \) and \( \langle b^p \rangle = N^{-1} \). Then

\[
\begin{aligned}
\langle a \rangle^{(r+1)} &= \langle a \rangle^{(r)} - 2a_p^{(r)} / N \\
a_p^{(r+1)} &= 2\langle a \rangle^{(r)} + (1 - 4/N)a_p^{(r)}
\end{aligned}
\]  

(27)

Thus, for the case of Grover’s iterations the second equation in (25) is transformed into the equation for the amplitude of the marked state (Oracle’s query). In what follows, the convenient and more compact notation below will be used to represent the average values of system (25)

\[
\begin{aligned}
\alpha^{(r)} &= \langle a \rangle^{(r)} \\
\beta^{(r)} &= \langle a | b^p \rangle^{(r)}
\end{aligned}
\]

4.2 Solving the equations for averages

An analytical solution of the system for average values will now be derived. Rewriting (25) using the notation introduced above results in

\[
\begin{aligned}
\alpha^{(r+1)} &= \alpha^{(r)} - 2\langle b^p \rangle \beta^{(r)} \\
\beta^{(r+1)} &= 2N\langle b^p \rangle \alpha^{(r)} + (1 - 4N\langle b^p \rangle^2)\beta^{(r)}
\end{aligned}
\]  

(28)
Rewriting the second equation in (28) as

\[ \beta^{(r+1)} = 2N\langle b^p \rangle \{ \alpha^{(r)} - 2\langle b^p \rangle \beta^{(r)} \} + \beta^{(r)} \]  

(29)

it may be seen that the expression in curly braces is equivalent to the right hand side of the first equation of (28). Hence,

\[ \beta^{(r+1)} = 2N\langle b^p \rangle \alpha^{(r+1)} + \beta^{(r)} \]  

(30)

Manipulating the last equation gives

\[ \alpha^{(r+1)} = \frac{\beta^{(r+1)} - \beta^{(r)}}{2N\langle b^p \rangle} \]  

(31)

Substituting expression (31) into the first equation of system (28) and using (30) and some algebra gives

\[ \beta^{(r+1)} + \beta^{(r-1)} = 2\beta^{(r)} (1 - 2N\langle b^p \rangle^2) \]  

(32)

Suppose that the solution of Equation (32) is of the form

\[ \beta^{(r)} = B \cos(\omega \tau + \varphi) \]  

(33)

Inserting this expression into Equation (32) and using some trigonometry, an expression for the frequency may be obtained

\[ \cos \omega = 1 - 2N\langle b^p \rangle^2 \]  

(35)

or, because \( \cos \omega = 1 - 2\sin^2(\omega / 2) \),

\[ \omega = 2 \arcsin(N\langle b^p \rangle) \]  

(36)

Now using Equation (31) produces the analytical form of \( \alpha^{(r)} \)

\[ \alpha^{(r)} = \frac{\beta^{(r)} - \beta^{(r-1)}}{2N\langle b^p \rangle} = B \frac{\cos(\omega \tau + \varphi) - \cos(\omega(\tau - 1) + \varphi)}{2N\langle b^p \rangle} \]  

(37)

The values of the constants \( B \) and \( \varphi \) can be found from the initial conditions:

\[ \alpha^{(0)} = B \frac{\cos \varphi - \cos(\omega - \varphi)}{2N\langle b^p \rangle} \]  

(38)

\[ \beta^{(0)} = B \cos \varphi \]  

(39)

It follows from (38) and (39), that
\[ \alpha^{(0)} = \frac{\beta^{(0)}}{\cos \varphi} \cos \varphi - \cos(\omega \varphi) \]

(40)

and after some transformation

\[ \tan \varphi = \frac{1}{\sin \omega} \left\{ 1 - \cos \omega - 2N \frac{\alpha^{(0)}}{\beta^{(0)}(b^p)} \right\} \]

(41)

The other constant is expressible in terms of the phase value

\[ B = \frac{\beta^{(0)}}{\cos \varphi} \]

(42)

Note that the amplitude of every basis state, including those that have a zero amplitude in the initial state, can take non-zero values during the iteration process. This corresponds to the development of spurious memories.

**Example 1.** Consider an unsorted database with phone numbers encoded with two qubits \(d=2\) and the distributed query

\[ |b^p\rangle = \frac{4}{5} |00\rangle + \frac{2}{5} |01\rangle + \frac{2}{5} |10\rangle + \frac{1}{5} |11\rangle \]

This distribution of basis state amplitudes is visually represented by the histogram in the top right corner of Figure 1. Grover’s iterations should be continued until the average overlap \(\beta^{(\tau)} = \langle a \mid b^p \rangle^{(\tau)}\) reaches one of the values \{\pm 1\}. After this it is necessary to perform a measurement and the probability for the system to be found in a given basis state becomes the prescribed function of Hamming distance from this state to the query center. In this example, if the initial state of the database is an equiprobable superposition of all basis states, the value +1 is reached after three iterations.

![Figure 1](image-url)

**Figure 1.** Right top corner: histogram of query amplitude distribution. Left top corner: initial equally weighted state, describing unsorted database. Histograms of the iterated state amplitudes, with their corresponding \(\beta\) values, are placed along the diagonal. For the third iteration the distribution (lower right...
corner) of the basis state amplitudes coincides with the distribution of amplitudes in the query (right top corner).

For the previous example the period of state oscillations can be found using expression (35). Since $N = 4$, and $\langle b^P \rangle = 1/4(0.8 + 0.4 + 0.4 + 0.2) = 0.45$, then

$$\cos \omega = 1 - 8 \cdot (0.45 \cdot 0.45) \geq - 0.6; \; \omega \geq 0.7\pi; \; \frac{2\pi}{T} \geq 0.7\pi; \; T \geq 3$$

So, it may be concluded that the state of the system oscillates very quickly.

5 Quantum associative memory with distributed query

Now the general form for a quantum associative memory may be considered. Recall that if binary patterns are considered this kind of memory suggests the retrieval of a memory state whose Hamming distance from the presented stimulus is minimal. Also, recall that for the case of memory which includes only a restricted number of memory states it is necessary to change the transformation $U_s = 2|s\rangle\langle s| - 1$ to the transformation $U_m = 2|m\rangle\langle m| - 1$.

Of course, it would be desirable if any quantum state $|\psi\rangle$ describing our memory would not contain nonzero amplitudes for any basis state not corresponding to one of the memory patterns (the absence of spurious memories). But while this may be possible for a completing associative memory it is impossible in the case of distributed queries because after the transformation performed by Oracle

$$a_x \rightarrow a_x - 2\langle a | b^P \rangle b^P_x$$

so that in general all amplitudes take non-zero values during the course of the algorithm’s iteration.

5.1 Model description

Hence, the state of our system will be described by an arbitrary wavefunction $|\psi\rangle$ and the transformation performed by the NameFinder-in-Memory will have the form

$$U_m : \psi \rightarrow 2|m\rangle\langle m|\psi\rangle - |\psi\rangle$$  \hspace{1cm} (44)

Since

$$\langle m |\psi\rangle = \frac{1}{\sqrt{P}} \sum_{x \in M} \left( \langle x | \sum_{y} a_y | y \rangle \right) = \frac{1}{\sqrt{P}} \sum_{x \in M} \sum_{y} a_y \langle x | y \rangle = \frac{1}{\sqrt{P}} \sum_{x \in M} \sum_{y} a_y \delta_{xy} = \frac{1}{\sqrt{P}} \sum_{x \in M} a_x = \sqrt{P} \langle a \rangle_m$$

where $P$ is the number of patterns in memory and $\langle a \rangle_m = \frac{1}{P} \sum_{x \in M} a_x$, it can be established that

$$2|m\rangle\langle m|\psi\rangle - |\psi\rangle = \sum_{x \in M} 2\langle a \rangle_m |x\rangle - \sum_{x} a_x |x\rangle$$

and the NameFinder-in-Memory transformation will be defined as

$$a_x \rightarrow \begin{cases} 2\langle a \rangle_m - a_x & \text{if } x \in M \\ -a_x & \text{otherwise} \end{cases}$$ \hspace{1cm} (45)$$

Hence, the amplitude transformation performed by NameFinder-in-Memory will have a form similar to the NameFinder-in-Database transformation in Grover’s algorithm, but this transformation will be
applied only to the basis states corresponding to valid memories. For the rest of the basis states this transformation resembles the Oracle transformation (performing phase inversion). Finally, the generalized search algorithm described in the previous section can be adapted to the case in point, taking the following “anti-symmetrical” form:

**Oracle transformation:**

\[ a_x \rightarrow a_x - 2\langle a | b^P \rangle b^P_x \]

**NameFinder-in-Memory transformation:**

\[ a_x \rightarrow 2\langle a | m \rangle m_x - a_x \] (46)

**Example 2.** Consider the case of a memory state containing a single valid pattern \( |m\rangle = |01\rangle \) and the distributed query centered on the basis state \( |11\rangle \),

\[ |b\rangle = \frac{9}{10} |11\rangle + \frac{3}{10} |10\rangle + \frac{3}{10} |01\rangle + \frac{1}{10} |00\rangle \]

The Oracle transformation converts the initial state of the memory \( |m\rangle^{(0)} = |01\rangle \) into

\[ |m\rangle^{(1/2)} = -0.06 |00\rangle + 0.82 |01\rangle - 0.18 |10\rangle - 0.54 |11\rangle \]

and the NameFinder-in-Memory transformation completes a single iteration as

\[ |m\rangle^{(1)} = 0.06 |00\rangle + 0.82 |01\rangle + 0.18 |10\rangle + 0.54 |11\rangle \]

Thus, after the first iteration the probability of measuring the system and finding the memory state \( |01\rangle \) (which is a Hamming distance of one from the query center \( |11\rangle \) ) takes a value 0.82 \( \approx 0.67 \). The probability of collapsing into the spurious state \( |11\rangle \) is 0.54 \( \approx 0.29 \), and the probabilities for the system to be found in other spurious states are considerably lower.

**Example 3.** Consider the case of a memory containing two states \( |m\rangle = \frac{1}{\sqrt{2}} |00\rangle + \frac{1}{\sqrt{2}} |01\rangle \), and suppose that the query has the same form as for the previous example. Let the initial state of our memory be \( |a\rangle^{(0)} = |m\rangle \). In this case after one iteration this state will take the form

\[ |a\rangle^{(1)} = 0.54 |00\rangle + 0.65 |01\rangle + 0.17 |10\rangle + 0.51 |11\rangle \]

The probability for the system to be found after a measurement in the basis state \( |01\rangle \), which is nearest to the query center (in the sense of Hamming distance) takes a value 0.65 \( \approx 0.42 \). The probability for the system to be found in the memory state \( |00\rangle \), for which query amplitude is minimal is not small: 0.54 \( \approx 0.29 \). Also, the probability for the system to be observed in the spurious state \( |11\rangle \), which corresponds to the query center, is fairly large. Both of these examples demonstrate that the properties of a quantum associative memory with distributed query seem to be reasonable.

In order to transform the expressions obtained earlier for Grover’s algorithm with distributed query to the case of a general quantum associative memory for which the transformation NameFinder-in-Memory is used, it is necessary to make changes

\[ \alpha = \langle a \rangle \Rightarrow \alpha_m = \langle a | m \rangle \]

\[ \langle b^P \rangle \Rightarrow b_m = \langle b | m \rangle \]

\[ 2N \Rightarrow 2 \] (47)

For example, the expressions for the state frequency (35-36) take the form

\[ \omega = \arccos(1 - 2b^2_m) \] (48)
Now, using numbers from Example 3 gives \( b_m = (0.1 + 0.3)/\sqrt{2} \approx 0.283 \) and \( \omega = \arccos(1 - 2 \cdot 0.283^2) \approx 0.57 \). Therefore \( T = \frac{2\pi}{\omega} \approx 11 \), and the memory recall will require 11 iterations of the algorithm to maximize the likelihood of obtaining the correct result (in this case observing the basis state \( |01\rangle \)).

### 5.2 Analytical solution for amplitudes

The third example will become clearer after deriving an analytical solution for the amplitudes. From Equations (46) it may be shown that

\[
a^{(r+1)}_x = -a^{(r)}_x + 2\alpha m_x + 2\beta_m (b^p_x - 2b_m m_x) \tag{50}
\]

Taking into account Equations (47) the following modifications of the expressions for the averages \( \alpha_m \) and \( \beta_m \) originally written in expressions (33) and (37) may be obtained:

\[
\alpha_m = \frac{B}{2b_m} (\cos(\omega \tau + \phi) - \cos(\omega (\tau - 1) + \phi)) = -\frac{B}{b_m} \sin(\omega \tau + \phi - \omega / 2) \sin(\omega / 2) \tag{51}
\]

\[
\beta_m = B \cos(\omega \tau + \phi) \tag{52}
\]

Inserting the last expressions into Equation (50) results in

\[
a^{(r+1)}_x = -a^{(r)}_x + 2B \left( (b^p_x - 2b_m m_x) \cos(\omega \tau + \phi) - \frac{m_x}{b_m} \sin(\omega / 2) \sin(\omega \tau + \phi - \omega / 2) \right) \tag{53}
\]

Suppose the solution of the last equation can be cast in the form

\[
a_x = A_x \cos(\omega \tau + \delta_x) \tag{54}
\]

Inserting (54) in (53) it is possible to derive

\[
A_x (\cos(\omega \tau + \omega + \delta_x) + \cos(\omega \tau + \delta_x)) \equiv 2A_x \cos(\omega / 2) \cos(\omega \tau + \delta_x + \omega / 2) =
\]

\[
2B(b^p_x - 2b_m m_x) \cos(\omega \tau + \phi) - \frac{2Bm_x}{b_m} \sin(\omega / 2) \sin(\omega \tau + \phi - \omega / 2) \equiv
\]

\[
2B(f_x \cos(\omega \tau + \phi) + g_x \sin(\omega \tau + \phi)) \tag{55}
\]

where

\[
f_x \overset{\text{def}}{=} \frac{m_x}{b_m} \sin^2(\omega / 2) + b^p_x - 2b_m m_x = b^p_x - m_x \sin(\omega / 2) \tag{56}
\]
\[
g_x = \frac{m_x}{2b_m} \sin \omega \quad (57)
\]

Introducing a variable

\[
\zeta_x = \arccos \left( \frac{f_x}{\sqrt{f_x^2 + g_x^2}} \right) \quad (58)
\]

Equation (55) can be transformed to

\[
a_x \cos(\omega/2) \cos(\omega \tau + \delta_x + \omega/2) = B \sqrt{f_x^2 + g_x^2} \cos(\omega \tau + \varphi + \zeta_x) \quad (59)
\]

From the last equation it follows immediately that the expressions for coefficients \(A_x\) and phases \(\delta_x\) are

\[
A_x = \frac{B \sqrt{f_x^2 + g_x^2}}{\cos(\omega/2)} \quad (60)
\]

\[
\delta_x = \zeta_x + \varphi - \omega/2 \quad (61)
\]

Expressions (54, 60-61) give the analytical form of the amplitudes of the basis states for a general quantum associative memory based on the use of Grover’s algorithm with distributed query. Figure 2 shows the analytical form of the amplitudes for the memory defined in the third example.
Figure 2. Dependence of basis state amplitudes on Grover’s iterations for the memory defined in Example 3.

It can be seen in Figure 2 that the state |01⟩ has a phase delay compared to the other valid memory state, |00⟩. The observed delay is connected with the greater amplitude which basis state |01⟩ has in the query. The change of phase $\delta$ due to the change of basis state amplitude $b^p_x$ in a query is dependent only upon the variation of $\zeta$; considering the derivative and applying some algebra

$$\frac{\partial}{\partial b^p_x} \cos \zeta_x = \frac{1}{\sqrt{(b^p_x)^2 - 2b_m b^p_x m_x + m_x^2}} \left( \frac{(b^p_x - m_x b_m)^2}{(b^p_x)^2 - 2b_m b^p_x m_x + m_x^2} \right) = \frac{\sin^2 \zeta_x}{\sqrt{(b^p_x)^2 - 2b_m b^p_x m_x + m_x^2}} \geq 0$$

Hence,

$$-\sin \zeta_x \frac{\partial \zeta_x}{\partial b^p_x} = -\frac{m_x \sin \omega}{2b_m} \frac{\zeta_x}{\partial b^p_x} \geq 0$$

or

$$\frac{\partial \zeta_x}{\partial b^p_x} \leq 0$$

Thus, the state |01⟩ which has a greater amplitude in the query than does the state |00⟩ will also have a lower phase value $\zeta$, and, consequently, lower value of $\delta$. In query formation, amplitudes of basis states monotonously decrease with Hamming distance from the query center; therefore, the memory state nearest to this center will have maximal amplitude and, consequently, minimal phase value. It may also be seen in Figure 2 that those basis states which do not belong to the set of valid memories (spurious memories) all have the same phase value.

As will be seen shortly, this fact is very important, and it can be verified using expression (58). Indeed, for any spurious memory, from (56-58) it follows that $m_x = 0 \Rightarrow g_x = 0 \Rightarrow \cos \zeta_x = 1 \Rightarrow \zeta_x = 0$. Then, from expression (61) the fact that $\delta_x = \varphi - \omega / 2 = \text{const}$ may be derived.

Consider, once again, Figure 2. Maximal amplitude belongs to the state |11⟩ for which the amplitude of the query is also maximal. On the other hand, the state |00⟩ has both minimal amplitude and minimal query amplitude. However, in general, this relation is not valid. Indeed, the dependence of the basis state amplitude on query amplitude

$$A_x = \frac{B \sqrt{f^2_x + g^2_x}}{\cos(\omega / 2)}$$

can be rewritten using explicit expressions for the parameters $f_x$ and $g_x$ given in (56) and (57) as follows

$$A_x = \frac{B}{\cos(\omega / 2)} \sqrt{(b^p_x)^2 - 2b_m b^p_x m_x + m_x^2}$$
The expression on the right hand side is not a monotonic function of $b^p_x$ and takes a minimal value on an internal point on the interval $[0,1]$. Therefore, basis states with lower query amplitude can have greater amplitude in memory.

Note, that in principle, in a distributed query some amplitudes can have zero values. Then, corresponding basis states will never arise as spurious memories. Thus, spurious memories are generated by the query itself demonstrating the principle: arise if suggested.

The evolution of the amplitudes of memory states is rather complicated. In general there exists no a priori knowledge about the structure of memory nor about the correspondence between the location of memories and the query center in configuration space. It is therefore difficult to obtain an analytical expression for the number of iterations needed to reach the maximal values of amplitudes of memories in the vicinity of this center. It is clear that the difficulty in deriving the necessary estimate is connected with the probabilistic character of the parameter $b_m$. But the situation can be improved considerably by changing the way the memory is structured, taking advantage of the fact that oscillating spurious memories all have the same phase value.

5.3 Memories become easily retrieved when they become spurious

The trick is simply to exchange the valid states to be memorized with other (spurious) states and vice versa. Namely, create a memory whose initial state is

$$|\psi^{(0)}\rangle = |\tilde{m}\rangle = \sum_{x \in M} \frac{1}{\sqrt{P}} |x\rangle$$

and correspondingly, the transformation performed by this memory is defined as

$$U_{\tilde{m}}: \psi = 2 |\tilde{m}\rangle\langle\tilde{m}|\psi\rangle - |\psi\rangle$$

In some sense this kind of memory is ideologically similar to the immune system, which includes antibodies corresponding to antigens that do not belong to the host organism and has almost no antibodies to its own proteins. Analogically, the associative memory is formed by memorizing patterns which should not be recalled.

In this system, all valid memories be treated as spurious ones. All these memories will be in phase with a phase value of $-\pi/2$ and have initial amplitudes of zero. Therefore, the amplitudes of these states will evolve according to

$$a^x = A_x \sin \omega \tau$$

and because for these states $m_x = 0$,

$$A_x = \frac{B}{\cos(\omega/2)} \sqrt{(b^p_x)^2 - 2b_m b^p_x + m_x^2} = \frac{Bb^p_x}{\cos(\omega/2)}$$

based upon the query amplitudes. Applying this trick to Example 3, the initial state of memory is

$$b_m = 2^{-d/2} \sum_{x \in M} b^p_x \equiv \tilde{b}_m = 2^{-d/2} \sum_{x=0}^{2^d-1} b^p_x$$
Then, \( \langle b | m \rangle = (0.3 + 0.9)/\sqrt{2} \approx 0.85 \) and \( \omega \approx 2.03 \), and now the states of the system will oscillate with considerably lower (compared to the original 11) period \( T = 2\pi/\omega \approx 3.1 \).

Since the amplitudes of all spurious memories achieve their maximal values at \( \tau = T/4 \), the system will be in a state for which the amplitudes of these “spurious” memories become proportional to their amplitudes in the query for \( \tau \approx 0.77 \). Of course, the number of iterations must be integer valued so it is necessary to add some periods for \( \tau + nT \) to be as near as possible to an integer value. In this case, for \( n = 2, 0.77 + 6.2 = 6.97 \approx 7 \), so it will suffice to perform seven iterations of the algorithm before performing a measurement. After these iterations the state of system will be

\[
\psi = 0.19|00\rangle + 0.57|01\rangle + 0.6|10\rangle + 0.53|11\rangle
\]

It is evident that the amplitudes of the “spurious” memories (really the valid memories) in this state are in the same proportion as in the query (a ratio of 1/3). Nevertheless, in general it seems that the difficulty inherent in the previous scheme still remains: a dependence of the frequency \( \omega \) on the set of valid memory patterns.

However, taking into account that now the number of “valid” patterns is very near \( 2^d \) (because \( p << 2^d \)), \( b_m = 2^{-d/2} \sum_{x \in M} b_x^p \approx \tilde{b}_m = 2^{-d/2} \sum_{x=0}^{2^d-1} b_x^p \). Therefore, in contrast to the more intuitive approach (memorizing the valid patterns) the value of \( b_m \) can be approximated as \textit{a priori} knowledge, dependent only on the form of the query, not on the form of the set of memory states. Therefore, using this approximated value of \( b_m \) and the corresponding approximation of the frequency \( \tilde{\omega} = 2 \arcsin \tilde{b}_m \), the number of iterations \( T_{\text{max}} \) needed to transform the system into a state such that the amplitudes of spurious (actually valid) memories become maximal and proportional to their amplitudes in the distributed query is

\[
\tilde{\omega}T_{\text{max}} = \frac{\pi}{2} \Rightarrow T_{\text{max}} \approx \frac{\pi}{2\tilde{\omega}}
\]

### 4.4. A note on initializing quantum states

It has been recently pointed out that there is potentially an inherent problem with current quantum computational algorithms [17]. Quantum computers and quantum algorithms rely heavily on the phase information of quantum states -- if the relative phases of the various states in a system are not correct, the computation will not work. Kak discusses the fact that quantum systems can possess random initial phases, whereas quantum algorithms implicitly assume some known initial phase conditions from which to begin the computation. The consensus seems to be that it is possible that this initial variability in the state phases may be compensated for by quantum error correction schemes [18][19]. However, these schemes may also be flawed. Classical error correction is based upon the fact that errors in classical systems are discrete -- a bit is flipped with some small probability. However, because quantum computational systems contain phase information, they are susceptible to a continuum of possible errors, and quantum error correction schemes developed to date address only a small number of special cases. Therefore, the issue to be resolved is whether or not in practice (that is in constructing a quantum computer) we will encounter mostly those few cases of error which have been treated in the literature or we will see the many other possibilities that Kak points out.

### 6 Conclusion
A model of quantum associative memory which is able to retrieve memory states with probability proportional to the amplitudes these states have in a query has been presented. This quantum memory can retrieve valid stored patterns from arbitrary stimuli represented by a distributed query of general form (fuzzy stimulus). Further investigation of the model is needed to estimate its other possible merits and limitations.

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