INITIALIZING THE AMPLITUDE DISTRIBUTION OF A QUANTUM STATE

Dan Ventura and Tony Martinez

Neural Networks and Machine Learning Laboratory Department of Computer Science Brigham Young University, Provo, Utah 84602 dan@axon.cs.byu.edu, martinez@cs.byu.edu

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To date, quantum computational algorithms have operated on a superposition of all basis states of a quantum system. Typically, this is because it is assumed that some function f is known and implementable as a unitary evolution. However, what if only some points of the function f are known? It then becomes important to be able to encode only the knowledge that we have about f. This paper presents an algorithm that requires a polynomial number of elementary operations for initializing a quantum system to represent only the m known points of a function f.

Key words: quantum computation, quantum algorithms, amplitude manipulation.

1. INTRODUCTION

The study of computation as a physical process has produced fascinating results in the form of new approaches to problem solving and information processing that exhibit impressive speed-up over classical approaches for some problems [1,2,3,4,5,6]. All of these algorithms operate on a superposition of all basis states, usually assuming prior knowledge of some function f that can be effected as a unitary evolution of the system. However, it is often the case in computational problem solving that a function f is not known but rather that some example points of the function are all the knowledge available. Such situations are particularly common in the field of computational learn-

ing, which typically deals with learning a function from a set of examples by discovering correlations between the functional inputs and the function's value. If quantum computation is to become a broadly applicable paradigm, it must be possible for quantum algorithms to function given only a limited set of functional points. The first major step toward this goal is developing the ability to encode a set of examples of a function as the state of a quantum system. One solution to this problem is to associate an index with each functional example and then to combine a classical lookup table of functional points with a quantum superposition of only the indices. This is easily achieved using standard quantum computational techniques and may be suitable for some applications. However, in the case of quantum computational learning, this is not adequate. Consider a series of unitary operations as a learning algorithm for discovering correlations between functional inputs and outputs. The information available in the functional examples (both input and output) must be encoded in the quantum system for the learning algorithm to have a chance of succeeding. If this can be achieved, then the possibility of quantum computational learning algorithms presents itself. The task then would be to produce unitary operators that effect the desired learning. For example, fourier-based machine learning algorithms such as those by Kushilevitz and Mansour [7] and Jackson [8] are tempting candidates for generalization into the quantum realm. In fact, Bshouty and Jackson investigated a quantum computational approach to learning [9]. However, their algorithm lacks the capability proposed here and therefore, though theoretically interesting, is not a plausible quantum algorithm. Another possibility is the implementation of a quantum associative memory. The algorithm presented here could be used to store memory patterns and Grover's algorithm [2] could be used for their recall. This paper presents an algorithm for encoding, in polynomial time, a set of examples of a function as the state of a quantum system.

2. DEFINITIONS

Given a set T of m examples of a function f, the goal is to produce

$$|\tilde{f}\rangle = \frac{1}{\sqrt{m}} \sum_{\bar{z} \in \mathcal{T}} f(\bar{z}) |\bar{f}\rangle$$
 (1)

as the quantum state of n qubits. It will be shown that the state $|\tilde{f}\rangle$ can be constructed using a polynomial number (in n and m) of elementary operations on one, two, or three qubits. In what follows, the specific qubits to which an operator is to be applied are indicated as subscripts on that operator. For simplicity, consideration will first be restricted to the case of $f: \bar{z} \to s$ with $\bar{z} \in \{0,1\}^n$ and $s \in \{-1,1\}$ (and we

will in general restrict $f(\bar{z})$ to points on the unit circle in the complex plane to facilitate normalization requirements). First, define the set of 2-qubit operators

$$\hat{S}^{s,p} = \begin{bmatrix} 1 & 0 & 0 & 0\\ 0 & 1 & 0 & 0\\ 0 & 0 & \sqrt{\frac{p-1}{p}} & \frac{-s}{\sqrt{p}}\\ 0 & 0 & \frac{s}{\sqrt{p}} & \sqrt{\frac{p-1}{p}} \end{bmatrix}, \tag{2}$$

where $s \in \{-1,1\}$ and $m \ge p \ge 1$. These operators form a set of conditional Hadamard-like transforms that will be used to incorporate the example set into a coherent quantum state. There will be a different $\hat{S}^{s,p}$ operator associated with each example in the set \mathcal{T} . Next, define

$$\hat{F} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \tag{3}$$

which flips the state of a qubit (and is perhaps more familiar as the Pauli spin matrix σ_x). Now, define the 2-qubit Control-NOT operator

$$\hat{F}^0 = \begin{bmatrix} \hat{F} & \hat{0} \\ \hat{0} & \hat{I}_2 \end{bmatrix}, \tag{4}$$

where $\hat{0}$ and \hat{I}_2 are the 2×2 zero and identity matrices, respectively, which conditionally flips the second qubit if and only if the first qubit is in the $|0\rangle$ state. Similarly define \hat{F}^1 which conditionally flips the second qubit if and only if the first qubit is in the $|1\rangle$ state [the matrix representing \hat{F}^1 is the same as in Eq. (4) with \hat{I}_2 and \hat{F} exchanged]. Finally, introduce four 3-qubit operators (really just four versions of the Fredkin gate [10]). These operators are used to identify specific states in a superposition, similar to Grover's identification of which state should be phase-rotated in his search algorithm [2]. The first of these,

$$\hat{A}^{00} = \begin{bmatrix} \hat{F} & \hat{0} \\ \hat{0} & \hat{I}_{6} \end{bmatrix}, \tag{5}$$

where the $\hat{0}$ are 6×2 and 2×6 zero matrices and \hat{I}_6 is the 6×6 identity matrix, conditionally flips the third bit if and only if the first two are in the state $|00\rangle$. Note that this is actually equivalent to \hat{F}^{00} , where \hat{F}^{00} is the 3-qubit generalization of Eq. (4); however, this operator will be used with the third qubit always in the $|0\rangle$ state and thus can be thought of as performing a logical AND of the negation of the first

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Generate |\tilde{f}\rangle = |x_1 \dots x_n, g_1 \dots g_{n-1}, c_1 c_2\rangle = |\bar{0}\rangle
 2.
             for m \ge p \ge 1
                            for 1 \le j \le n
 3.
                                          if z_{pj} \neq z_{p+1j} (where z_{m+1} = \{0\}^n)
\hat{F}_{c_2x_j}^0 |\tilde{f}\rangle
 4.
 5.
 6.
                            \hat{S}^{s_p,p}_{c_1c_2}|\tilde{f}\rangle
 7.
                            \hat{A}_{x_1 x_2 g_1}^{z_1 z_2} | \tilde{f} \rangle
for 3 \le k \le n
 8.
 9.
                                          \hat{A}_{x_k g_{k-2} g_{k-1}}^{z_k 1} | \tilde{f} \rangle
 10.
                           \begin{array}{c} \hat{F}_{g_{n-1}c_1}^1 | \tilde{f} \rangle \\ \text{for } n \geq k \geq 3 \\ \hat{A}_{x_k g_{k-2} g_{k-1}}^{z_k 1} | \tilde{f} \rangle \\ \hat{A}_{x_1 x_2 g_1}^{z_1 z_2} | \tilde{f} \rangle \end{array}
 11.
 12.
 13.
 14.
15. \hat{F}_{c_2}|f\rangle
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Fig. 1. Initializing amplitude distribution in a quantum system.

two qubits, setting the third to $|1\rangle$ if and only if the first two are $|00\rangle$. The other 3-qubit operators, \hat{A}^{01} , \hat{A}^{10} and \hat{A}^{11} , are variations of \hat{A}^{00} in which \hat{F} occurs in the other three possible locations along the main diagonal.

3. ALGORITHM

The algorithm for encoding \mathcal{T} into a quantum system requires n+(n-1)+2 qubits, arranged in three quantum registers labeled x,g, and c. The quantum state of all three registers together is represented as $|x,g,c\rangle$, and the algorithm proceeds as in Fig. 1. The x register will hold a superposition of the examples in the set \mathcal{T} - there are n qubits in the register, and the value of f for an example will be used as the coefficient for the state corresponding to that example. The g and c registers contain only ancillary qubits and are restored to the state $|\bar{0}\rangle$ by the end of the algorithm. A high-level intuitive description of the algorithm is as follows. The system is initially in the state $|\bar{0}\rangle$. The qubits in the x register are conditionally flipped so that their states correspond to the first example. The $\hat{S}^{s,p}$ operator corresponding to that example then changes the basis of the system such that one of the states has a coefficient that matches the example's value for f. This

same state then is changed to one that exists outside of the subspace affected by the $\hat{S}^{s,p}$ operators, in effect making it permanent, and the process is repeated for each example. When all the examples have been processed, the result is a coherent superposition of states corresponding to the sample points, where the amplitudes of the states all have the same magnitude but have different phases according to the values of their corresponding examples. This is somewhat reminiscent of the first part of Bernstein and Vazirani's Fourier sampling mechanism [11]. However, they assume that the function f is known and calculable in polynomial time, whereas here we do not know the function f but only a small set of examples drawn from f.

In analyzing the complexity of the algorithm it is assumed that line 1 can be done trivially. The loop of line 2 is repeated m times and consists of n (lines 3-5) + 3 (6-8) + n-2 (9-10) + 1 (11) + n-2 (12-13) + 1 (14) operations, and line 15 requires one more operation. Thus, the entire algorithm requires m(n+3+n-2+1+n-2+1)+1=m(3n+1)+1 operations and is O(mn). This is optimal in the sense that just reading each instance once cannot be done in any fewer than mn steps.

A concrete example of a simple 2-input function will help clarify the preceding discussion. Suppose that we are given the set $T = \{f(01) = -1, f(10) = 1, f(11) = -1\}$. The initial state is $|\bar{0}\rangle$, and the algorithm evolves the quantum state through the series of unitary operations described in Fig. 1. For convenience lines 3-6, 8-10, and 12-14 of the algorithm are agglomerated as the compound operators FLIP, AND, and AND^{\dagger} respectively and are treated in more detail in Eq. (6-8) and Figs. 2 and 3. First the qubit states of the x register are flipped to match the first example, and the c register is marked so the state will be affected by the $\hat{S}^{s,p}$ operator.

$$|00,0,00\rangle \xrightarrow{FLIP} |01,0,10\rangle.$$

Next, the appropriate $\hat{S}^{s,p}$ operator (with s equal to the functional value of the example being processed, here -1, and p equal to the number of the instances, including the current one, yet to be processed, here 3) is applied, creating a new state in the superposition. This process will be referred to as state generation and corresponds to line 7 of Fig. 1.

$$\stackrel{\hat{S}^{-1,3}}{\longrightarrow} -\frac{1}{\sqrt{3}}|01,0,11\rangle + \sqrt{\frac{2}{3}}|01,0,10\rangle.$$

The two states just affected by the $\hat{S}^{s,p}$ operator are marked in their g registers.

$$\stackrel{AND}{\longrightarrow} -\frac{1}{\sqrt{3}}|01,1,11\rangle + \sqrt{\frac{2}{3}}|01,1,10\rangle.$$

One of the marked states is made permanent by setting its c register to $|01\rangle$ and this state now represents the first example. The c register of the other state is returned to $|00\rangle$, and it is ready to generate a new state. This is performed by line 11 of the algorithm.

$$\stackrel{\hat{F}^1_{g_{n-1}c_1}}{\longrightarrow} -\frac{1}{\sqrt{3}}|01,1,01\rangle + \sqrt{\frac{2}{3}}|01,1,00\rangle.$$

Finally the work done in the g register is undone and at this point one pass through the loop of line 2 of the algorithm has been performed.

$$\stackrel{AND^{\dagger}}{\longrightarrow} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \sqrt{\frac{2}{3}}|01,0,00\rangle.$$

Now, the entire process is repeated for the second example. Again, the x register of the appropriate state (that state whose c_2 qubit is in the $|0\rangle$ state) is selectively flipped to match the new example, and its c register is again marked. This time the selective qubit state flipping occurs for those qubits that correspond to bits in which the first and second examples differ – both in this case. Notice that the state representing the first example is not affected by this operation, nor any of those that follow.

$$\stackrel{FLIP}{\longrightarrow} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \sqrt{\frac{2}{3}}|10,0,10\rangle.$$

Next, another $\hat{S}^{s,p}$ operator is applied, this time with s=1 and p=2.

$$\xrightarrow{\hat{S}^{1,2}} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \frac{1}{\sqrt{2}}\sqrt{\frac{2}{3}}|10,0,11\rangle + \sqrt{\frac{1}{2}}\sqrt{\frac{2}{3}}|10,0,10\rangle.$$

The two states just affected by the $\hat{S}^{s,p}$ operator are marked in their g registers.

$$\xrightarrow{AND} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \frac{1}{\sqrt{3}}|10,1,11\rangle + \sqrt{\frac{1}{3}}|10,1,10\rangle.$$

The c register is used to save another state (this one representing the second example).

$$\stackrel{\hat{F}^1_{g_{n-1}c_1}}{\longrightarrow} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \frac{1}{\sqrt{3}}|10,1,01\rangle + \sqrt{\frac{1}{3}}|10,1,00\rangle.$$

A final bit of house keeping resets the g register, preparing for the process to be repeated again.

$$\stackrel{AND^{\dagger}}{\longrightarrow} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \frac{1}{\sqrt{3}}|10,0,01\rangle + \sqrt{\frac{1}{3}}|10,0,00\rangle.$$

Finally, the third example is considered. The x register of the generator state is again selectively flipped and its c register is marked. This time, only those qubits corresponding to bits that differ in the second and third examples are flipped, in this case just qubit x_2 .

$$\xrightarrow{FLIP} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \frac{1}{\sqrt{3}}|10,0,01\rangle + \sqrt{\frac{1}{3}}|11,0,10\rangle.$$

Again a new state is generated to represent this third example. Notice that the generator state is now left with a magnitude of 0, because there are no more examples to process.

$$\stackrel{\hat{S}^{-1,1}}{\longrightarrow} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \frac{1}{\sqrt{3}}|10,0,01\rangle - \frac{1}{\sqrt{1}}\sqrt{\frac{1}{3}}|11,0,11\rangle + \sqrt{\frac{0}{1}}\sqrt{\frac{1}{3}}|11,0,10\rangle$$

Once again the house keeping steps are performed.

$$\xrightarrow{AND} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \frac{1}{\sqrt{3}}|10,0,01\rangle - \frac{1}{\sqrt{3}}|11,1,11\rangle$$

$$\xrightarrow{\hat{F}_{g_{n-1}e_1}^1} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \frac{1}{\sqrt{3}}|10,0,01\rangle - \frac{1}{\sqrt{3}}|11,1,01\rangle$$

$$\xrightarrow{AND^{\dagger}} -\frac{1}{\sqrt{3}}|01,0,01\rangle + \frac{1}{\sqrt{3}}|10,0,01\rangle - \frac{1}{\sqrt{3}}|11,0,01\rangle.$$

Finally, line 15 performs one last operation, restoring all the ancillary qubits to their initial state.

$$\xrightarrow{\hat{F}_{c_2}} -\frac{1}{\sqrt{3}} |01,0,00\rangle + \frac{1}{\sqrt{3}} |10,0,00\rangle - \frac{1}{\sqrt{3}} |11,0,00\rangle.$$

At this point, the g and c registers are not entangled with the x register, and therefore the system can be simplified as

$$-\frac{1}{\sqrt{3}}|01\rangle + \frac{1}{\sqrt{3}}|10\rangle - \frac{1}{\sqrt{3}}|11\rangle$$

and it may be seen that the partial function defined by the set T is now represented as a quantum superposition in the x register.

Since this is a fairly complicated algorithm, perhaps the following analogy will help the reader gain an intuitive idea of how the algorithm works. Imagine that you have a sphere that you would like to divide into (equal sized) sections, each of which is to be painted a different color. The catch is that the only painting mechanism available requires that you dip the entire sphere into a vat of paint. You do

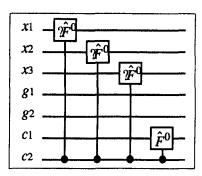


Fig. 2. Quantum network for implementing FLIP.

however, have a "fixer" or varnish-type solution available to you that may be applied by hand to select areas of the sphere. The solution is to dip the entire sphere into the vat containing the first color and then apply the fixer over the portion of the sphere that is to remain that color. Then the sphere is dipped into the vat with the second color. The fixer maintains the first color in its small area of the sphere, but the rest of the sphere changes to the new color. Now some fixer is applied to the area of this new color that is to remain permanent, and the process is repeated until the sphere is completely painted. In the analogy, the sphere is the quantum system, the colors are the basis states corresponding to the functional points to be incorporated into the system, the FLIP operation corresponds to selecting a vat of a certain color, the $\hat{S}^{s,p}$ operator is dipping the sphere into the vat and the AND, $\hat{F}^1_{g_{n-1}c_1}$ and AND^\dagger operations correspond to applying the fixer. Of course, this is a fairly abstract analogy, and like most, should not be taken too literally.

Here the complex operators FLIP, AND, and AND^{\dagger} are specified in more detail. In operator notation,

$$FLIP = \hat{F}_{c_2c_1}^0 \hat{F}_{c_2x_j}^0 (n \ge j \ge 1, z_{pj} \ne z_{p+1j}).$$
 (6)

Alternatively, FLIP may be represented as a quantum network as in [12]. Fig. 2 shows one such network for a 3-input function. In the figure, a '?' indicates that the \hat{F}^0 operator is applied only if the value of the bit in question differs from the value of that bit for the previous example. Also, in operator notation,

$$AND = \hat{A}_{x_n g_{n-2} g_{n-1}}^{z_n 1} \cdots \hat{A}_{x_3 g_1 g_2}^{z_3 1} \hat{A}_{x_1 x_2 g_1}^{z_1 z_2}$$
(7)

and

$$AND^{\dagger} = \hat{A}_{x_1 x_2 g_1}^{z_1 z_2} \hat{A}_{x_3 g_1 g_2}^{z_3 1} \cdots \hat{A}_{x_n g_{n-2} g_{n-1}}^{z_n 1}$$
 (8)

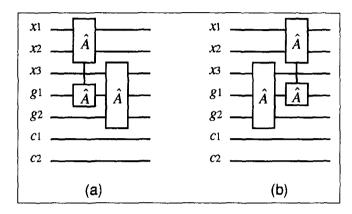


Fig. 3. Quantum networks for implementing AND and AND^{\dagger} .

AND and AND[†] also may be represented as quantum networks, and corresponding example diagrams appear in Figs. 3(a) and 3(b) respectively. Note that the only difference between the two is the order in which the qubits are operated upon, since all the operators are their own adjoints.

The quantum network representation of the entire algorithm requires the addition of gates for performing the \hat{S} and \hat{F} operations, and the complete network is shown in Fig. 4. The network shown will be repeated several times – once for each instance in the training set. Each such repetition will make use of different \hat{A} operators and flip different qubits using the \hat{F} gates, according to the particular example being processed.

The trick that makes the algorithm work is the fact that whenever an $\hat{S}^{s,p}$ operator is applied, there are no states with the c register in the $|11\rangle$ state. This is crucial to the construction of $|\tilde{f}\rangle$ because it allows the operations to be unitary without generating "extraneous" states. Another key is the ability to identify a specific state in the superposition (the one just generated) using the AND and AND^{\dagger} operators; this provides the ability to mark that particular state and only that state so as not to operate on it again. And this is done reversibly so that no disruptive entanglement results.

4. GENERALIZING THE ALGORITHM

The algorithm of Fig. 1 can handle only binary functions. However,

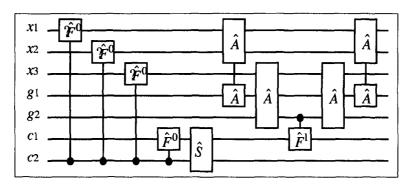


Fig. 4. Quantum network for encoding a single function example.

generalization to N-valued functions is straightforward. The goal now is to produce

$$|\tilde{f}\rangle = \frac{1}{\sqrt{m}} \sum_{\bar{z} \in \mathcal{I}} f(\bar{z}) |\bar{f}\rangle$$
 (9)

with $\bar{z} \in \{0, ..., N-1\}^n$ and $f(\bar{z}) \in \{e^{(2s\pi i)/N}\}, 0 \le s \le N-1$. The only necessary changes are a generalization of the x register (and only the x register) to include systems with more than two states and generalization of some of the operators. The operator generalizations are as follows:

$$\hat{S}^{s,p} = \begin{bmatrix} 1 & 0 & 0 & 0\\ 0 & 1 & 0 & 0\\ 0 & 0 & \sqrt{\frac{p-1}{p}} & \frac{-e^{(2s\pi i)/N}}{\sqrt{p}}\\ 0 & 0 & \frac{e^{(2s\pi i)/N}}{\sqrt{p}} & \sqrt{\frac{p-1}{p}} \end{bmatrix}$$
(10)

where now $s \in \{0, ..., N-1\}$ and $m \ge p \ge 1$. As mentioned earlier, the function values are restricted to equally spaced points on the unit circle in the complex plane:

$${}^{N}\hat{F}^{0} = \begin{bmatrix} \hat{C} & \hat{0} \\ \hat{0} & \hat{I}_{N} \end{bmatrix} \tag{11}$$

where $\hat{0}$ and \hat{I}_N are the $N \times N$ zero and identity matrices respectively, and \hat{C} is the $N \times N$ circulant matrix obtained by shifting all rows of \hat{I}_N down one, with the Nth row becoming the first (\hat{C} performs the functional equivalent of incrementing the binary label associated with each state and throwing away the final carry). There is now a set of

 \hat{F}^0 operators, one for 2-state qubits, one for 3-state qubits, etc., and the \hat{F}^0 operator defined in Eq. (4) is the special case for N=2. These operators act on a pair of qubits, the first of which is in the c register and thus is still just a 2-state system, and the second of which is in the x register and thus now has N states, as $\hat{F}^0|0q\rangle = |0(q+1) \bmod N\rangle$ and $\hat{F}^0|1q\rangle = |1q\rangle$, conditionally rotating the second qubit (labeled q) to the next highest value (but rotating the maximum value back to 0) if the first qubit is in the $|0\rangle$ state. Note that \hat{F}^1 does not need to be generalized since it only acts on qubits in the q and q registers, which are still only 2-state qubits

$${}^{N}\hat{A}^{00} = \begin{bmatrix} \hat{F} & \hat{0} \\ \hat{0} & \hat{I}_{a} \end{bmatrix} \tag{12}$$

where $a = 2N^2 - 2$, the $\hat{0}$ are $2 \times a$ and $a \times 2$ zero matrices, \hat{I}_a is the $a \times a$ identity matrix, \hat{F} is as defined in Eq. (3) and the entire matrix is $2N^2 \times 2N^2$. Again there is now a set of \hat{A}^{00} operators and the \hat{A}^{00} operator defined in Eq. (5) is the special case for N=2. These operators act on three qubits, as ${}^{N}\hat{A}^{00}|00r\rangle = |00(r+1) \bmod 2\rangle$ and ${}^{N}\hat{A}^{00}|pqr\rangle = |pqr\rangle$ for $p,q \neq 0$, conditionally flipping the third qubit (labeled r), which will still be a 2-state qubit as it is in the g register, if the first two qubits are in the $|00\rangle$ state. Note that whereas the operator \hat{A}^{00} defined in Eq. (5) was really equivalent to a 3-qubit version of Eq. (4), this is *not* the case with these generalized operators. In other words, Eq. (12) is not equivalent to a 3-qubit generalization of Eq. (11). This is because ${}^{N}\hat{F}^{0}$ now affects qubits (in the x register) with more than 2 states, therefore requiring a rotation through the various states rather than a simple flip between 2 possible states; whereas ${}^{N}\hat{A}^{00}$ still affects qubits with only 2 states and thus simply toggles between states. Since this operator will be used with the third bit always in the |0) state, it can still be thought of as performing a sort of logical AND (actually a generalized AND where the inputs can be greater than binary) on the values of the first two qubits, setting the third to $|1\rangle$ if and only if the first two are $|00\rangle$. The other \hat{A} operators must also be generalized, a set for each possible pair of values that any of the qubits in the x and g registers can assume. In other words we need all the sets of operators ${}^{N}\hat{A}^{ij}$ for $0 \le i, j \le N-1$. However, these sets of operators are simple variations of the ${}^{N}\hat{A}^{00}$ set given in Eq. (12) with the sub-matrix \hat{F} occurring in different locations along the diagonal according to the values for i and j.

The algorithm for the more general case is the same as in Fig. 1, except for two changes. First, the operators must be replaced with their

generalizations defined in Eqs. (10-12), and second, the IF statement of line 4 must be changed to a WHILE statement. This is again because instead of simply flipping a qubit's state between two possibilities as the binary operator \hat{F}^0 defined in Eq. (4) does, the generalized operator $^N\hat{F}^0$ defined in Eq. (11) rotates a qubit's state through N possibilities in a specific order one at a time. Thus, the compound FLIP operation defined by lines 3-6 of the algorithm now may require up to (N-1)n operations instead of the O(n) required for the simpler binary case. This brings the total number of operations required for the algorithm up to (N+1)mn+m+1, which is O(mnN).

5. CONCLUSION

In summary, this paper presents a polynomial-time algorithm for initializing a quantum system to represent m known points of a function f. The result is a quantum superposition with m non-zero coefficients - the creation of which is a nontrivial task compared to creating a superposition of all basis states. It may be appropriate to mention here that recently work as been done to analyze Grover's algorithm [2] for the case of arbitrary initial amplitude distributions [13], such as would result from the application of the algorithm described here. Further, the paper suggests a new field to which quantum computation may be applied to advantage - that of computational learning. In fact, it is the authors' opinion that this application of quantum computation will, in general, demonstrate much greater returns than its application to more traditional computational tasks (though Shor's algorithm is an obvious exception). We make this conjecture because results in both quantum computation and computational learning are by nature probabilistic and inexact, whereas most traditional computational tasks require precise and deterministic outcomes. Also, progress in (classical) randomized algorithms for many computational (learning) problems may be taken as further evidence that quantum approaches may be fruitful in this area as well.

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