# Quantum finite multitape automata<sup>\*</sup>

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Abstract. Quantum finite automata were introduced by C. Moore, J. P. Crutchfield [MC 97], and by A. Kondacs and J. Watrous [KW 97]. This notion is not a generalization of the deterministic finite automata. Moreover, in [KW 97] it was proved that not all regular languages can be recognized by quantum finite automata. A. Ambainis and R. Freivalds [AF 98] proved that for some languages quantum finite automata may be exponentially more concise rather than both deterministic and probabilistic finite automata. In this paper we introduce the notion of quantum finite multitape automata and prove that there is a language recognized by a quantum finite automaton but not by deterministic or probabilistic finite automata. This is the first result on a problem which can be solved by a quantum computer but not by a deterministic or probabilistic computer. Additionally we discover unexpected probabilistic automata recognizing complicated languages.

## 1 Introduction

Recently a new type of algorithms has appeared, namely, *quantum* algorithms. Nobel prize winner physicist Richard Feynman asked in [Fe 82] what effects can have the principles of quantum mechanics, especially, the *principle of superposition* on computation. He gave arguments showing that it might be computationally expensive to simulate quantum mechanics on classical computers. This observation immediately lead to a conjecture predicting enormous advantages to quantum computers versus classical ones. D. Deutsch [De 89] introduced the commonly used notion of the quantum Turing machine and proved that quantum Turing machines compute exactly the same recursive functions as ordinary

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<sup>&</sup>lt;sup>‡</sup> Supported by Berkeley Fellowship for Graduate Studies.

<sup>&</sup>lt;sup>§</sup> Research supported by Grant No.96.0282 from the Latvian Council of Science

Research partially supported by the International Computer Science Institute, Berkeley, California, by the DFG grant KA 673/4-1, and by the ESPRIT BR Grants 7079 and ECUS030

deterministic Turing machines do. When Peter Shor [Sh 94] proved that quantum algorithms can factorize large integers and compute discrete logarithms in a polynomial time practical construction of quantum computers became a problem that involves many people and huge funding. Indeed, building a quantum computer would be equivalent to building a universal code-breaking machine since the intractability of the above-mentioned problems is the fundamental of the public-key cryptography.

Quantum mechanics differs from the classical physics very much. It suffices to mention *Heisenberg's uncertainty principle* asserting that one cannot measure both the position and the impulse of a particle simultaneously precisely. There is a certain trade-off between the accuracy of the two measurements. Another well-known distinction of quantum mechanics from the classical physics is the impossibility to measure any object without changing the object.

The fundamental atom of information is the quantum bit, henceforth abbreviated by the term 'qbit'.

Classical information theory is based on the classical bit as fundamental atom. This classical bit, henceforth called *cbit*, is in one of two classical states t (often interpreted as "true") and f (often interpreted as "false"). In quantum information theory the most elementary unit of information is the *quantum bit*, henceforth called *qbit*. To explain it, we first discuss a *probabilistic* counterpart of the classical bit, which we call here *pbit*. It can be t with a probability  $\alpha$  and f with probability  $\beta$ , where  $\alpha + \beta = 1$ . A *qbit* is very much like to *pbit* with the following distinction. For a *qbit*  $\alpha$  and  $\beta$  are not real but complex numbers with the property  $\|\alpha\|^2 + \|\beta\|^2 = 1$ .

Every computation done on qbit s is performed by means of unitary operators. One of the simplest properties of these operators shows that such a computation is reversible. The result always determines the input uniquely. It may seem to be a very strong limitation for such computations. Luckily this is not so. It is possible to embed any irreversible computation in an appropriate environment which makes it reversible. For instance, the computing agent could keep the inputs of previous calculations in successive order.

The following features of quantum computers are important (but far from the only characteristic features of them).

- Input, output, program and memory are represented by qbits.
- Any computation (step) can be represented by a unitary transformation of the computer as a whole.
- Any computation is reversible. Because of the unitarity of the quantum evolution operator, a deterministic computation can be performed by a quantum computer if and only if it is reversible.
- No qbit can be copied. After the qbit is processed, the original form of it is no more available.
- Measurements may be carried out on any qbit at any stage of the computation. However any measurement destroys the information. More precisely, the measurement turns a qbit into a classical bit with probabilities dependent on the qbit.

 Quantum parallelism: during a computation, a quantum computer proceeds down all coherent paths at once.

Quantum finite automata were introduced twice. First this was done by C. Moore and J.P.Crutchfield [MC 97]. Later in a different and non-equivalent way these automata were introduced by A. Kondacs and J. Watrous [KW 97].

The first definition just mimics the definition of 1-way finite probabilistic only substituting *stochastic* matrices by *unitary* ones. We use a more elaborated definition [KW 97].

QFA is a tuple  $M = (Q; \Sigma; \delta; q_0; Q_{acc}; Q_{rej})$  where Q is a finite set of states,  $\Sigma$  is an input alphabet,  $\delta$  is a transition function,  $q_0 \in Q$  is a starting state, and  $Q_{acc} \subset Q$  and  $Q_{rej} \subset Q$  are sets of accepting and rejecting states. The states in  $Q_{acc}$  and  $Q_{rej}$  are called *halting states* and the states in  $Q_{non} = Q - (Q_{acc} \cup Q_{rej})$ are called *non halting states*.  $\kappa$  and \$ are symbols that do not belong to  $\Sigma$ . We use  $\kappa$  and \$ as the left and the right endmarker, respectively. The *working alphabet* of M is  $\Gamma = \Sigma \cup \{\kappa; \$\}$ .

A superposition of M is any element of  $l_2(Q)$  (the space of mappings from Q to C with  $l_2$  norm). For  $q \in Q$ ,  $|q\rangle$  denotes the unit vector which takes value 1 at q and 0 elsewhere. All elements of  $l_2(Q)$  can be expressed as linear combinations of vectors  $|q\rangle$ . We will use  $\psi$  to denote elements of  $l_2(Q)$ .

The transition function  $\delta$  maps  $Q \times \Gamma \times Q$  to C. The value  $\delta(q_1; a; q_2)$  is the amplitude of  $|q_2\rangle$  in the superposition of states to which M goes from  $|q_1\rangle$  after reading a. For  $a \in \Gamma$ ,  $V_a$  is a linear transformation on  $l_2(Q)$  defined by

$$V_a(|q_1\rangle) = \sum_{q_2 \in Q} \delta(q_1; a; q_2) |q_2\rangle.$$

We require all  $V_a$  to be unitary.

The computation of a QFA starts in the superposition  $|q_0\rangle$ . Then transformations corresponding to the left endmarker  $\kappa$ , the letters of the input word x and the right endmarker \$ are applied. The transformation corresponding to  $a \in \Gamma$  consists of two steps.

1. First,  $V_a$  is applied. The new superposition  $\psi'$  is  $V_a(\psi)$  where  $\psi$  is the superposition before this step.

2. Then,  $\psi'$  is observed with respect to the observable  $E_{acc} \oplus E_{rej} \oplus E_{non}$ where  $E_{acc} = span\{|q\rangle : q \in Q_{acc}\}, E_{rej} = span\{|q\rangle : q \in Q_{rej}\}, E_{non} = span\{|q\rangle : q \in Q_{non}\}$ . This observation gives  $x \in E_i$  with the probability equal to the amplitude of the projection of  $\psi'$ . After that, the superposition collapses to this projection.

If we get  $\psi' \in E_{acc}$ , the input is accepted. If we get  $\psi' \in E_{rej}$ , the input is rejected. If we get  $\psi' \in E_{non}$ , the next transformation is applied.

We regard these two transformations as reading a letter a.

For probabilistic computation, the property that the probability of correct answer can be increased arbitrarily is considered evident. Hence, it is not surprising that [KW 97] wrote "with error probability bounded away from 1/2", thinking that all such probabilities are equivalent. However, mixing reversible (quantum computation) and non-reversible (measurements after each step) components in one model makes it impossible. This problem was first considered in the paper [AF 98] by A. Ambainis and R. Freivalds. The following theorems were proved there:

Let p be a prime. We consider the language  $L_p = \{a^i | i \text{ is divisible by } p\}$ . It is easy to see that any deterministic 1-way finite automaton recognizing  $L_p$  has at least p states.

**Theorem 1.1** For any  $\epsilon > 0$ , there is a QFA with  $O(\log p)$  states recognizing  $L_p$  with probability  $1 - \epsilon$ .

**Theorem 1.2** Any 1-way probabilistic finite automaton recognizing  $L_p$  with probability  $1/2 + \epsilon$ , for a fixed  $\epsilon > 0$ , has at least p states.

**Theorem 1.3** There is a language that can be recognized by a 1-QFA with probability 0.68... but not with probability  $7/9 + \epsilon$ .

We consider only multitape finite automata in this paper. A quantum automaton is defined in the natural way, demanding that the transformation ( the state, the information on the first head having or not having moved, the information on the second head having or not having moved, ...,  $\rightarrow$  the state, the information on the first head having or not having moved, the information on the second head having or not having moved, the information on the second head having or not having moved, the information on the second head having or not having moved, the information on the second head having or not having moved, ...) is unitary for arbitrary tuple (the symbol observed by the first head, the symbol observed by the first head,...).

**Definition 1.1** A quantum finite multitape automaton (QFMA)  $A = (Q; \Sigma; \delta; q_0; Q_{acc}; Q_{rej})$  is specified by the finite input alphabet  $\Sigma$ , the finite set of states Q, the initial state  $q_0 \in Q$ , the sets  $Q_{acc} \subset Q$ ,  $Q_{rej} \subset Q$  of accepting and rejecting states, respectively, with  $Q_{acc} \cap Q_{rej} = \emptyset$ , and the transition

 $\delta: Q \times \Gamma^m \times \{\downarrow, \rightarrow\}^m \longrightarrow C_{[0,1]},$ 

where m is the number of input tapes,  $\Gamma = \Sigma \cup \{\kappa, \$\}$  is the tape alphabet of A and  $\kappa,\$$  are endmarkers not in  $\Sigma$ , which satisfies the following conditions (of well-formedness):

1. Local probability condition

function

$$\forall (q_1, \sigma) \in Q \times \Gamma^m \sum_{(q,d) \in Q \times \{\downarrow, \to\}^m} |\delta(q_1, \sigma, q, d)| = 1.$$

2. Orthogonality of column vectors condition.

$$\forall q_1, q_2 \in Q, q_1 \neq q_2, \forall \sigma \in \Gamma^m \qquad \sum_{(q,d) \in Q \times \{\downarrow, \rightarrow\}^m} \delta^*(q_1, \sigma, q, d) \delta(q_2, \sigma, q, d) = 0.$$

#### 3. Separability condition.

 $M =_{def} \{1, 2, \dots, m\}$ . The k-th component of an arbitrary vector s will be defined as  $s^k$ . We shall understand by I an arbitrary element from the set  $P(M) - \{\emptyset\}$ .

$$R_{I} =_{def} A_{1} \times A_{2} \times \ldots \times A_{m}, where A_{i} = \begin{cases} \{\downarrow, \rightarrow\}, & if \ i \notin I \\ \{"nothing"\}, & if \ i \in I. \end{cases}$$

$$T_{I} =_{def} B_{1} \times B_{2} \times \ldots \times B_{m}, where B_{i} = \begin{cases} \{\downarrow, \rightarrow\}, & if \ i \in I \\ \{"nothing"\}, & if \ i \notin I. \end{cases}$$

The function  $R_i \times T_i \xrightarrow{d_I} \{\downarrow, \rightarrow\}^m$  is defined as follows:

$$d_I^i(r,t) =_{def} \begin{cases} r^i, & if \ i \notin I \\ t^i, & if \ i \in I. \end{cases}$$

$$d_I(r,t) =_{def} (d_I^1(r,t), d_I^2(r,t), \dots, d_I^m(r,t)).$$

 $\begin{array}{l} \forall I \in P(M) - \{ \emptyset \} \; \forall \sigma_1 \sigma_2 \in \Gamma^m \; \forall q_1, q_2 \in Q \; \forall t_1, t_2 \in T_I; \\ if \; \forall i \notin I \; \sigma_1^i = \sigma_2^i, \; \forall j \in I \; t_1^j \neq t_2^j \; then \end{array}$ 

$$\sum_{(q,r)\in Q\times R_I} \delta^*(q_1,\sigma_1,q,d_I(r,t_1))\delta(q_2,\sigma_2,q,d_I(r,t_2)) = 0.$$

States from  $Q_{acc} \cup Q_{rej}$  are called halting states and states from  $Q_{non} = Q - (Q_{acc} \cup Q_{rej})$  are called non halting states.

To process an input word vector  $x \in (\Sigma^*)^m$  by A it is assumed that the input is written on every tape k with the endmarkers in the form  $w_x^k = \kappa x^k$ \$ and that every such a tape, of length  $|x^k| + 2$ , is circular, i. e., the symbol to the right of \$ is  $\kappa$ .

For the fixed input word vector x we can define n to be an integer vector which determines the length of input word on every tape. So for every n we can define  $C_n$  to be the set of all possible configurations of A where  $|x^i| = n^i$ .  $|C_n| = |Q| \prod_{i=1}^m (n^i + 2)$ . Every such a configuration is uniquely determined by a pair  $|q, s\rangle$ , where  $q \in Q$  and  $0 \le s^i \le |x^i| + 1$  specifies the position of head on the *i*-th tape.

Every computation of A on an input x,  $|x^i| = n^i$ , is specified by a unitary evolution in the Hilbert space  $H_{A,n} = l_2(C_n)$ . Each configuration  $c \in C_n$  corresponds to the basis vector in  $H_{A,n}$ . Therefore a global state of A in the space  $H_{A,n}$  has a form  $\sum_{c \in C_n} \alpha_c |c\rangle$ , where  $\sum_{c \in C_n} |\alpha_c|^2 = 1$ . If the input word vector is x and the automaton A is in its global state  $|\psi\rangle = \sum_{c \in C_n} \alpha_c |c\rangle$ , then its further step is equivalent to the application of a linear operator  $U_x^{\delta}$  over Hilbert space  $l_2(C_n)$ .

#### Definition 1.2

$$U_x^{\delta} |\psi\rangle = \sum_{c \in C_n} \alpha_c U_x^{\delta} |c\rangle.$$

If a configuration  $c = |q', s\rangle$ , then  $U_x^{\delta}|c\rangle = \sum_{(q,d)\in Q\times\{\downarrow,\to\}^m} \delta(q', \sigma(s), q, d)|q, \tau(s, d)\rangle$ , where  $\sigma(s) = (\sigma^1(s), \dots, \sigma^m(s))$ ,  $\sigma^i(s)$  specifies the  $s^i$ -th symbol on the *i*-th tape,

where  $\sigma(s) = (\sigma^{1}(s), \dots, \sigma^{m}(s)), \sigma^{i}(s)$  specifies the s<sup>i</sup>-th symbol on the i-th tape, and

$$\tau(s,d) = (\tau^{1}(s,d), \dots, \tau^{m}(s,d)), \ \tau^{i}(s,d) = \begin{cases} (s^{i}+1) \bmod (n^{i}+2), \ if \ d^{i}=' \to ' \\ s^{i}, \ if \ d^{i}=' \downarrow '. \end{cases}$$

**Lemma 1.1** The well-formedness conditions are satisfied iff for any input x the mapping  $U_x^{\delta}$  is unitary.

**Definition 1.3** A QFMA  $A = (Q; \Sigma; \delta; q_0; Q_{acc}; Q_{rej})$  is simple if for each  $\sigma \in \Gamma^m$  there is a linear unitary operator  $V_{\sigma}$  over the inner-product space  $l_2(Q)$  and a function  $D: Q \longrightarrow \{\downarrow, \rightarrow\}^m$ , such that

$$\forall q_1 \in Q \ \forall \sigma \in \Gamma^m \ \delta(q_1, \sigma, q, d) = \begin{cases} \langle q | V_\sigma | q_1 \rangle, \ if \ D(q) = d \\ 0, \ otherwise. \end{cases}$$

**Lemma 1.2** If the automaton A is simple, then conditions of well-formedness are satisfied iff for every  $\sigma V_{\sigma}$  is unitary.

As in the case of single-tape quantum finite automata it is presumed that all the states are divided into *halting* and *nonhalting*, and whenever, the automaton comes into a halting state, the automaton stops, and accepts or rejects the input with a probability equal to the square of the modulo of the amplitude.

# 2 Reversible automata

A 1-way reversible finite automaton (RFA) is a QFA with  $\delta(q_1, a, q_2) \in \{0, 1\}$ for all  $q_1, a, q_2$ . Alternatively, RFA can be defined as a deterministic automaton where, for any  $q_2, a$ , there is at most one state  $q_1$  such that reading a in  $q_1$ leads to  $q_2$ . We use the same definitions of acceptance and rejection. States are partitioned into accepting, rejecting and non-halting states and a word is accepted (rejected) whenever the RFA enters an accepting (rejecting) state. After that, the computation is terminated. Similarly to quantum case, endmarkers are added to the input word. The starting state is one, accepting (rejecting) states can be multiple. This makes our model different from both [An 82] (where only one accepting state was allowed) and [Pi 92] (where multiple starting states with a non-deterministic choice between them at the beginning were allowed). We define our model so because we want it to be as close to our model of QFAs as possible. Generally, it's hard to introduce probabilism into finite automata without losing reversibility. However, there are some types of probabilistic choices that are consistent with reversibility. For example, it was proved by A. Ambainis and R. Freivalds that for the language  $L = \{a^{2n+3} | n \in \mathbb{N}\}$  not recognizable by a 1-way RFA, there are 3 1-way RFAs such that each word in the language is accepted by 2 of them and each word not in the language is rejected by 2 out of 3.

#### 3 Quantum vs. probabilistic automata

**Definition 3.1** We say that a language L is [m,n]-deterministically recognizable if there are n deterministic automata  $A_1$ ,  $A_2$ ,  $A_n$  such that:

a) if the input is in the language L, then all n automata  $A_1, \ldots, A_n$  accept the input;

b) if the input is not in the language L, then at most m of the automata  $A_1, \ldots, A_n$  accept the input.

**Definition 3.2** We say that a language L is [m,n]-reversibly recognizable if there are n deterministic reversible automata  $A_1$ ,  $A_2$ ,  $A_n$  such that:

a) if the input is in the language L, then all n automata  $A_1, \ldots, A_n$  accept the input;

b) if the input is not in the language L, then at most m of the automata  $A_1, \ldots, A_n$  accept the input.

**Lemma 3.1** If a language L is [1,n]-deterministically recognizable by 2-tape finite automata, then L is recognizable by a probabilistic 2-tape finite automaton with probability  $\frac{n}{n+1}$ .

**Proof.** The probabilistic automaton starts by choosing a random integer  $1 \leq r \leq (n+1)$ . After that , if  $r \leq n$ , then the automaton goes on simulating the deterministic automaton  $A_r$ , and, if r = n + 1, then the automaton rejects the input. The inputs in L are accepted with probability  $\frac{n}{n+1}$ , and the inputs not in the language are rejected with a probability no less than  $\frac{n}{n+1}$ .  $\Box$ 

**Lemma 3.2** If a language L is [1,n]-reversibly recognizable by 2-tape finite automata, then L is recognizable by a quantum 2-tape finite automaton with probability  $\frac{n}{n+1}$ .

**Proof.** In essence the algorithm is the same as in Lemma 3.1. The automaton starts by choosing a random integer  $1 \leq r \leq (n+1)$ . This is done by taking 3 different actions with amplitudes  $\frac{1}{\sqrt{3}}$  (the possibility to make such a choice is asserted in Lemma 4.6). After that , if  $r \leq n$ , then the automaton goes on simulating the deterministic automaton  $A_r$ , and, if r = n+1, then the automaton rejects the input. Acceptance and rejecting are made by entering the states where

measurement is made immediately. (Hence the probabilities are totaled, not the amplitudes.)  $\square$ 

First, we discuss the following 2-tape language

$$L_1 = \{ (x_1 \nabla x_2, y) \| x_1 = x_2 = y \}$$

where the words  $x_1, x_2, y$  are unary.

**Lemma 3.3** For arbitrary natural n, the language  $L_1$  is [1,n]-deterministically recognizable.

**Proof.** See Appendix. R. Freivalds [Fr 79] proved

**Theorem 3.1** The language  $L_1$  can be recognized with arbitrary probability  $1-\epsilon$  by a probabilistic 2-tape finite automaton but this language cannot be recognized by a deterministic 2-tape finite automaton.

**Proof.** By Lemma 5.1 *L* is [1,n]-deterministically recognizable for arbitrary *n*.By Lemma 3.1, the language is recognizable with probability  $\frac{n}{n+1}$ .  $\Box$ 

**Theorem 3.2** The language  $L_1$  can be recognized with arbitrary probability  $1-\epsilon$  by a quantum 2-tape finite automaton.

**Proof.** By Lemma 3.2.  $\Box$ 

We wish to prove a quantum counterpart of Theorem 3.1. We need some lemmas to this goal.

In an attempt to construct a 2-tape language recognizable by a quantum 2tape finite automaton but not by probabilistic 2-tape finite automata we consider a similar language

 $L_2 = \{ (x_1 \nabla x_2 \nabla x_3, y) \| \text{there are exactly 2 values of } x_1, x_2, x_3 \text{such that they equal} y \},\$ 

where the words  $x_1, x_2, x_3, y$  are unary.

**Theorem 3.3** A quantum automaton exists which recognizes the language  $L_2$  with a probability  $\frac{3}{5} - \epsilon$  for arbitrary positive  $\epsilon$ .

**Proof.** This automaton with amplitudes:

a) 
$$\frac{1}{\sqrt{5}} \times 1$$
  
b)  $\frac{1}{\sqrt{5}} \times (\cos \frac{2\pi}{3} + i \sin \frac{2\pi}{3})$   
c)  $\frac{1}{\sqrt{5}} \times (\cos \frac{4\pi}{3} + i \sin \frac{4\pi}{3})$   
d)  $\sqrt{\frac{2}{5}}$   
takes actions:  
a) compare  $x_1 = x_2 = y$ ,  
b) compare  $x_1 = x_3 = y$ ,  
c) compare  $x_1 = x_3 = y$ ,

d) says "accept".

If y equals all 3 words  $x_1, x_2, x_3$ , then the input is accepted with probability  $\frac{2}{5}$  (since the amplitudes of the actions a), b), c) total to 0). If y equals 2 out of 3 words  $x_1, x_2, x_3$ , then the input is accepted with probability  $\frac{3}{5}$ . If y equals at most one of the words  $x_1, x_2, x_3$ , then the input is accepted with probability  $\frac{2}{5}$  (only if the action d) is taken).  $\Box$ 

Unfortunately, the following theorem holds.

**Theorem 3.4** A probabilistic automaton exists which recognizes the language  $L_2$  with a probability  $\frac{21}{40}$ 

**Proof.** The probabilistic automaton with probability  $\frac{1}{2}$  takes an action A or B:

A) Choose a random j and compare  $x_j = y$ . If yes, accept with probability  $\frac{19}{20}$ . If no, accept with probability  $\frac{1}{20}$ .

B) Choose a random pair j, k and compare  $x_j = x_k = y$ . If yes, reject. If no, accept with probability  $\frac{12}{20}$ .

If y equals all 3 words  $x_1, x_2, x_3$  and the action A is taken, then the input is accepted with relative probability  $\frac{19}{20}$ . If y equals all 3 words  $x_1, x_2, x_3$ , then and the action A is taken, then the input is accepted with relative probability 0. This gives the acceptance probability in the case if y equals all 3 words  $x_1, x_2, x_3$ , to be  $\frac{19}{40}$  and the probability of the correct result "no" to be  $\frac{21}{40}$ .

If y equals 2 words out of  $x_1, x_2, x_3$  and the action A is taken, then the input is accepted with relative probability  $\frac{13}{20}$ . If y equals 2 words out of  $x_1, x_2, x_3$ and the action B is taken, then the input is accepted with relative probability  $\frac{8}{20}$ . This gives the acceptance probability in the case if y equals 2 words out of  $x_1, x_2, x_3$ , to be  $\frac{21}{40}$ .

If y equals only 1 word out of  $x_1, x_2, x_3$  and the action A is taken, then the input is accepted with relative probability  $\frac{7}{20}$ . If y equals only 1 word out of  $x_1, x_2, x_3$  and the action B is taken, then the input is accepted with relative probability  $\frac{12}{20}$ . This gives the acceptance probability in the case if y equals only 1 word out of  $x_1, x_2, x_3$ , to be  $\frac{19}{40}$  and the probability of the correct result "no" to be  $\frac{21}{40}$ .

If y equals no word of  $x_1, x_2, x_3$  and the action A is taken, then the input is accepted with relative probability  $\frac{1}{20}$ . If y equals no word of  $x_1, x_2, x_3$  and the action B is taken, then the input is accepted with relative probability  $\frac{12}{20}$ . This gives the acceptance probability in the case if y equals no word of  $x_1, x_2, x_3$ , to be  $\frac{13}{40}$  and the probability of the correct result "no" to be  $\frac{27}{40}$ .  $\Box$ 

Now we consider a modification of the language  $L_2$  which might be more difficult for a probabilistic recognition:

 $L_3 = \{(x_1 \nabla x_2 \nabla x_3, y_1 \nabla y_2) \| \text{there is exactly one value} \\ \text{such that there are exactly two values} j \text{such that} x_j = y_k. \}$ 

**Theorem 3.5** A quantum finite 2-tape automaton exists which recognizes the language  $L_3$  with a probability  $\frac{6}{11} - \epsilon$  for arbitrary positive  $\epsilon$ .

**Proof** is moved to Appendix. It is provided for the referees only, and it will not be included in the final text.

However this language also can be recognized by a probabilistic 2-tape finite automaton.

**Theorem 3.6** A probabilistic finite 2-tape automaton exists which recognizes the language  $L_3$  with a probability  $\frac{13}{25} - \epsilon$  for arbitrary positive  $\epsilon$ .

**Proof.** The probabilistic automaton with probability  $\frac{6}{25}$  takes action A or B or C or with probability  $\frac{7}{25}$  takes action D:

A) Choose a random k and two values of j. Then compare  $x_j = y_k$ . If yes, accept. If no, reject.

B) Chose a random k and compare  $x_1 = x_2 = x_3 = y_k$ . If yes, reject. If no, accept. C) Choose two values j and m. Then compare  $x_j = x_m = y_1 = y_2$ . If yes, reject. If no, accept.

D) Says "reject".

Notice that the actions A, B, C are probabilistic, and they can be performed only with probability  $1-\epsilon$  (actions A and B are described in the proof of Theorem 3.1 and action C is similar).

The acceptance	probabilities	equal:
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	А	В	С	total	
no $y_k$ equals 2 or 3 $x_j$	0	1	1	$\frac{12}{25}$	
one $y_k$ equals 2 $x_j$	$\frac{1}{6}$	1	1	$\frac{13}{25}$	
one $y_k$ equals 3 $x_j$	$\frac{1}{2}$	$\frac{1}{2}$	1	$\frac{12}{25}$	
two $y_k$ equal 2 $x_j$	$\frac{1}{3}$	1	$\frac{2}{3}$	$\frac{12}{25}$	
all $y_k$ equal all $x_j$	1	0	0	$\frac{6}{25}$	

Finally we consider a modification of the languages above which indeed is difficult for a probabilistic recognition:

 $L_4 = \{(x_1 \nabla x_2, y) || \text{ there is exactly one value } j \text{ such that } x_j = y. \}$ 

where the words  $x_1, x_2, y$  are binary.

**Theorem 3.7** A quantum finite 2-tape automaton exists which recognizes the language  $L_4$  with a probability  $\frac{2}{3} - \epsilon$  for arbitrary positive  $\epsilon$ .

Idea of the proof. The computations corresponding to the checks whether or not  $x_1 = y$  and  $x_2 = y$ , are performed with opposite amplitudes. If these two computations are successful, the amplitudes annihilate.

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## 4 Unitary matrices

**Lemma 4.1** For arbitrary real values  $\phi, \psi, \eta$ , the matrix

$$\begin{pmatrix} \cos\phi(\cos\eta + i\sin\eta) & \sin\phi(\cos\eta + i\sin\eta)\\ \sin\phi(\cos\psi + i\sin\psi) - \cos\phi(\cos\psi + i\sin\psi) \end{pmatrix}$$

is unitary.

**Corollary 4.1** The matrix 
$$\begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}$$
 is unitary.

**Corollary 4.2** The matrix  $\begin{pmatrix} \cos \phi \ i \sin \phi \\ i \sin \phi \ \cos \phi \end{pmatrix}$  is unitary.

**Corollary 4.3** The matrix  $\begin{pmatrix} \cos \phi & \sin \phi \\ \sin \phi - \cos \phi \end{pmatrix}$  is unitary.

This corollary is crucially important for the sequel. We will use it to prove that quantum automata (in contrast with deterministic or probabilistic automata) can do the counting modulo arbitrarily large prime numbers using only two states.

#### **Lemma 4.2** For arbitrary real values $\phi, \psi$ , the matrix

 $\begin{pmatrix} \cos\phi\cos\psi & i\sin\phi\cos\psi & i\cos\phi\sin\psi - \sin\phi\sin\psi\\ i\sin\phi\cos\psi & \cos\phi\cos\psi - \sin\phi\sin\psi & i\cos\phi\sin\psi\\ i\cos\phi\sin\psi - \sin\phi\sin\psi & \cos\phi\cos\psi & i\sin\phi\cos\psi\\ -\sin\phi\sin\psi & i\cos\phi\sin\psi & i\sin\phi\cos\psi & \cos\phi\cos\psi \end{pmatrix}$ 

is unitary.

Corollary 4.4 The matrix

$$\begin{pmatrix} \frac{1}{2} & \frac{i}{2} & -\frac{1}{2} \\ \frac{i}{2} & \frac{1}{2} & -\frac{1}{2} & \frac{i}{2} \\ \frac{i}{2} & -\frac{1}{2} & -\frac{1}{2} & \frac{i}{2} \\ \frac{i}{2} & -\frac{1}{2} & \frac{1}{2} & \frac{i}{2} & \frac{1}{2} \\ -\frac{1}{2} & \frac{i}{2} & \frac{i}{2} & \frac{1}{2} \end{pmatrix}$$

/

is unitary.

**Definition 4.1** We call the matrix

$$C = \begin{pmatrix} c_{11} & c_{12} \dots & c_{1\,kn} \\ c_{21} & c_{22} \dots & c_{2\,kn} \\ \dots & \dots & \dots \\ c_{kn\,1} & c_{kn\,2} \dots & c_{kn\,kn} \end{pmatrix}$$
  
a block-product of the matrices  $A = \begin{pmatrix} a_{11} & a_{12} \dots & a_{1\,k} \\ a_{21} & a_{22} \dots & a_{2\,k} \\ \dots & \dots & \dots \\ a_{k\,1} & a_{k\,2} \dots & a_{k\,k} \end{pmatrix}$  and  $B = \begin{pmatrix} b_{11} & b_{12} \dots & b_{1\,n} \\ b_{21} & b_{22} \dots & b_{2\,n} \\ \dots & \dots & \dots \\ b_{n\,1} & b_{n\,2} \dots & b_{n\,k} \end{pmatrix}$   
if  $C_{(m-1)k+i\,(l-1)k+j} = a_{i\,j}b_{m\,l}$ .

Lemma 4.3 If the matrices A and B are unitary, then their block-product is also a unitary matrix.

**Lemma 4.4** For arbitrary prime p, the matrix

$$\begin{pmatrix} \frac{1}{\sqrt{p}}(e^0) & \frac{1}{\sqrt{p}}(e^0) & \frac{1}{\sqrt{p}}(e^0) & \dots & \frac{1}{\sqrt{p}}(e^0) \\ \frac{2p\pi}{\sqrt{p}}(e^{\frac{2p\pi}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{2(p-1)\pi}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{2(p-2)\pi}{p}}) & \dots & \frac{1}{\sqrt{p}}(e^{\frac{2\pi}{p}}) \\ \frac{1}{\sqrt{p}}(e^{\frac{4p\pi}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{4(p-1)\pi}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{2(p-2)\pi}{p}}) & \dots & \frac{1}{\sqrt{p}}(e^{\frac{4\pi}{p}}) \\ \frac{1}{\sqrt{p}}(e^{\frac{5p\pi}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{5(p-1)\pi}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{5(p-2)\pi}{p}}) & \dots & \frac{1}{\sqrt{p}}(e^{\frac{5\pi}{p}}) \\ \frac{1}{\sqrt{p}}(e^{\frac{p-1}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{p-1}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{p-1}{p}}) & \dots & \frac{1}{\sqrt{p}}(e^{\frac{p-1}{p}}) \\ \dots & \dots & \dots & \dots \\ \frac{1}{\sqrt{p}}(e^{\frac{p-1}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{p-1}{p}}) & \frac{1}{\sqrt{p}}(e^{\frac{p-1}{p}}) & \dots & \frac{1}{\sqrt{p}}(e^{\frac{p-1}{p}}) \end{pmatrix} \end{pmatrix}$$

is unitary.

**Corollary 4.5** For arbitrary prime p, there is a unitary matrix  $C_p$  of size  $p \times p$  such that all the elements  $c_{1j}$  of this matrix equal  $\frac{1}{\sqrt{p}}$ .

**Corollary 4.6** For arbitrary natural number n, there is a unitary matrix  $C_n$  of size  $n \times n$  such that all the elements  $c_{1j}$  of this matrix equal  $\frac{1}{\sqrt{n}}$ .

**Corollary 4.7** For arbitrary natural number n, there is a unitary matrix  $C_n$  of size  $n \times n$  such that all the elements  $c_{i1}$  of this matrix equal  $\frac{1}{\sqrt{n}}$ .

These corollaries are used as a tool to perform an equiprobable choice among a finite number of possibilities.

#### 5 Proof of Lemma 3.3

**Lemma 5.1** For arbitrary natural n, the language  $L_1$  is [1,n]-deterministically recognizable.

The language L can be recognized by the following team of deterministic 1-way 2-tape finite automata  $\{A_1, A_2, \dots, A_n\}$ .

The automaton  $A_r$  performs cycles, each one consisting in reading n + 1 digits from  $x_1$  and r digits from y. When the symbol  $\nabla$  is met, the automaton memorizes the remainder of  $x_1$  modulo n and goes on (in cycles) reading n + 1 digits from  $x_2$  and n + 1 - r digits from y. If the input pair of words is in the language, the processing of the two tapes takes the same time. In this case the automaton accepts the pair, otherwise the automaton rejects it. This way, the automaton accepts the pair of words if and only if there are nonnegative integers u, v such that:

$$(n+1)u \le x_1$$

$$(n+1)(u+1) > x_1$$

$$(n+1)v \le x_2$$

$$(n+1)(v+1) > x_2$$

$$x_1 - (n+1)u = x_2 - (n+1)v = y - ru - (n+1-r)v$$

If  $x_1 = x_2$ , then the number -ru - (n+1-r)v does not depend on the choice of r. Either all  $x_i$  match the y, or no one does. If  $x_1 \neq x_2$ , then the numbers -ru - (n+1-r)v are all different for different values of r. Hence at most one of them can match y.  $\Box$ 

#### Proof of Theorem 3.5 6

**Theorem 3.5**. A quantum finite 2-tape automaton exists which recognizes the language  $L_3$  with a probability  $\frac{3}{5} - \epsilon$  for arbitrary positive  $\epsilon$ .

- This automaton takes the following actions with the following amplitudes: a) With amplitude  $\frac{1}{\sqrt{11}} \times (\cos \frac{0\pi}{6} + i \sin \frac{0\pi}{6})$  compares whether  $x_1 = x_2 = y_1$ ;

- b) With amplitude  $\frac{1}{\sqrt{11}} \times (\cos \frac{4\pi}{6} + i \sin \frac{4\pi}{6})$  compares whether  $x_1 = x_2 = y_1$ ; c) With amplitude  $\frac{1}{\sqrt{11}} \times (\cos \frac{4\pi}{6} + i \sin \frac{4\pi}{6})$  compares whether  $x_1 = x_3 = y_1$ ; d) With amplitude  $\frac{1}{\sqrt{11}} \times (\cos \frac{6\pi}{6} + i \sin \frac{6\pi}{6})$  compares whether  $x_1 = x_2 = y_2$ ; e) With amplitude  $\frac{1}{\sqrt{11}} \times (\cos \frac{6\pi}{6} + i \sin \frac{10\pi}{6})$  compares whether  $x_2 = x_3 = y_2$ ; f) With amplitude  $\frac{1}{\sqrt{11}} \times (\cos \frac{2\pi}{6} + i \sin \frac{2\pi}{6})$  compares whether  $x_1 = x_3 = y_2$ ;
- g) With amplitude  $\sqrt{\frac{5}{11}}$  says "accept".

These comparisons are probabilistic actions (as in Theorem 3.1; recall that the words  $x_i, y_k$  are unary) but they are simulated by a quantum automaton. This way, every action is replaced by several actions the number of which depends on

 $\epsilon$ . For instance, if  $\epsilon = \frac{1}{n}$  then the action a) is replaced by n actions: a1) With amplitude  $\frac{1}{\sqrt{11n}} \times (\cos \frac{0\pi}{6} + i \sin \frac{0\pi}{6})$  compares whether there are nonnegative integers u, v such that:

$$(n+1)u \le x_1$$

$$(n+1)(u+1) > x_1$$

$$(n+1)v \le x_2$$

$$(n+1)(v+1) > x_2$$

$$x_1 - (n+1)u = x_2 - (n+1)v = y_1 - u - nv$$

a2) With amplitude  $\frac{1}{\sqrt{11n}} \times (\cos \frac{0\pi}{6} + i \sin \frac{0\pi}{6})$  compares whether there are nonnegative integers u, v such that:

$$(n+1)u \le x_1$$

$$(n+1)(u+1) > x_1$$

$$(n+1)v \le x_2$$

$$(n+1)(v+1) > x_2$$

$$x_1 - (n+1)u = x_2 - (n+1)v = y_1 - 2u - (n-1)^2$$

 $x_1 - (n+1)u = x_2 - (n+1)v = y_1 - 2u - (n-1)v$ a3) With amplitude  $\frac{1}{\sqrt{11n}} \times (\cos \frac{0\pi}{6} + i \sin \frac{0\pi}{6})$  compares whether there are nonnegative integers u, v such that:

$$(n+1)u \le x_1$$
$$(n+1)(u+1) > x_1$$
$$(n+1)v \le x_2$$

$$(n+1)(v+1) > x_2$$
  
$$x_1 - (n+1)u = x_2 - (n+1)v = y_1 - 3u - (n-2)v$$

an) With amplitude  $\frac{1}{\sqrt{11n}} \times (\cos \frac{0\pi}{6} + i \sin \frac{0\pi}{6})$  compares whether there are nonnegative integers u, v such that:

$$(n+1)u \le x_1$$

$$(n+1)(u+1) > x_1$$

$$(n+1)v \le x_2$$

$$(n+1)(v+1) > x_2$$

$$x_1 - (n+1)u = x_2 - (n+1)v = y_1 - nu - v$$

If  $y_1 = y_2$ , then the total of amplitudes for the acceptance is 0 since the amplitude for comparison of  $y_1$  with arbitrary pair  $x_i, x_j$  is (minus 1) times the amplitude for the comparison of  $y_2$  with the same pair  $x_i, x_j$ .

If  $y_1 \neq y_2$ , and  $y_1 = x_1 = x_2$ , then  $y_2$  cannot equal more than one of the  $x_j$ , namely,  $x_3$ . In this case, all the actions am) [m = 1, 2, ..., n] end in acception and so do also no more than one of the actions bm), no more than one of the actions cm), no more than one of the actions dm), no more than one of the actions em), and no more than one of the actions fm). The total of the amplitudes for the accepting actions am) is  $\frac{n}{\sqrt{11n}} \times (\cos \frac{0\pi}{6} + i \sin \frac{0\pi}{6})$ 

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