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### THRESHOLD CIRCUITS FOR ITERATED MATRIX PRODUCT AND POWERING\*

#### CARLO MEREGHETTI<sup>1</sup> AND BEATRICE PALANO<sup>2</sup>

**Abstract**. The complexity of computing, via threshold circuits, the *iterated product* and *powering* of fixed-dimension  $k \times k$  matrices with integer or rational entries is studied. We call these two problems  $\mathsf{IMP}_k$  and  $\mathsf{MPOW}_k$ , respectively, for short. We prove that:

- (i) For  $k \ge 2$ ,  $\mathsf{IMP}_k$  does not belong to  $\mathsf{TC}^0$ , unless  $\mathsf{TC}^0 = \mathsf{NC}^1$ .
- (ii) For stochastic matrices:  $IMP_2$  belongs to  $TC^0$  while, for  $k \geq 3$ ,  $IMP_k$  does not belong to  $TC^0$ , unless  $TC^0 = NC^1$ .
- (iii) For any k, MPOW<sub>k</sub> belongs to  $TC^0$ .

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#### Introduction

In this work, we study the parallel complexity of performing some matrix operations. As computational model, we use threshold circuits [12]. We are interested in solving problems by using threshold circuits of constant depth. In this regard, we focus on the class  $TC^0$  [7] of problems solvable by constant depth families of (unbounded fan-in) threshold circuits of polynomial size. Several arithmetic and linear algebra operations lie in  $TC^0$ : the iterated sum and product of integers and rationals, integer division, matrix multiplication, etc. (see [7,8,16]). We know that  $TC^0$  is contained in  $NC^1$ , the class of problems solvable by families of (bounded

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fan-in) AND/OR/NOT-circuits of polynomial size and logarithmic depth [2]. (We do not impose any uniformity condition on circuit families.) It is widely accepted, although still unproved, that  $TC^0 \neq NC^1$ , as the opposite would cause the unlikely collapse of the hierarchy  $\{TC_d^0\}_{d\geq 1}$  (see Sect. 1 for a brief discussion).

The first problem we shall be dealing with is the computation of the iterated product of fixed-dimension  $k \times k$  matrices with integer or rational entries. We call this problem  $IMP_k$ , for short. It can be easily seen that  $IMP_k$  is in  $NC^1$ . Here, we investigate whether  $IMP_k$  can be solved in  $TC^0$  as well. By considering the algebraic characterization of regular languages in  $TC^0$  proposed by Barrington et al. in [1], we show that: if  $TC^0 \neq NC^1$  then, for any  $k \geq 2$ ,  $IMP_k$  does not belong to  $TC^0$ .

We then focus on studying the parallel complexity of  $IMP_k$  on the relevant class of stochastic matrices. The interest in such a class of matrices is related to the study of fast parallel algorithms for recognizing probabilistic languages [9, 10]. We prove a slightly better situation:  $IMP_2$  for stochastic matrices belongs to  $TC^0$ . On the other hand, if  $TC^0 \neq NC^1$  then, for any  $k \geq 3$ ,  $IMP_k$  for stochastic matrices does not belong to  $TC^0$ .

The second problem we shall consider is powering fixed-dimension  $k \times k$  matrices with integer or rational entries. We call this problem MPOW<sub>k</sub>, for short. By using notions from linear algebra plus fast mathematics on polynomials, we are able to show that, for any k, MPOW<sub>k</sub> belongs to  $TC^0$ . So, in sharp contrast with IMP<sub>k</sub>, fixing matrix dimension leads to a full feasibility result for MPOW<sub>k</sub>.

The paper is organized as follows: Section 1 contains basic definitions and results concerning circuits and algebraic automata theory. In Section 2, we evaluate the hardness of computing the iterated product of fixed-dimension matrices, both in the case of integer or general rational matrices, and in the particular case of stochastic matrices. In Section 3, we exhibit an algorithm to power in TC<sup>0</sup> fixed-dimension integer or rational matrices.

The results presented here are contained in a preliminary form in [9], where related issues concerning the possibility of accepting in TC<sup>0</sup> regular and probabilistic languages are addressed, too (see also [10] for this latter topic).

#### 1. Preliminaries

We assume some familiarity with the complexity classes defined via traditional and threshold circuits [2,7,12,15]. We recall that  $\mathbf{TC}_d^0$  is the class of problems solvable by families of (unbounded fan-in) threshold circuits of polynomial weights and size, and constant depth d. Then, the class  $\mathbf{TC}^0 = \bigcup_d \mathbf{TC}_d^0$ , introduced in [7], contains the problems solvable by families of threshold circuits of polynomial size and constant depth. Typical problems in  $\mathbf{TC}^0$  are: the iterated sum of integers (in  $\mathbf{TC}_2^0$ ), integer division (in  $\mathbf{TC}_3^0$ ), iterated product of integers (in  $\mathbf{TC}_4^0$ ), iterated sum and product of rationals, matrix multiplication, and modular arithmetics (the

reader is referred to [7,8,16] where he can find a deep study on the exact number of layers in threshold circuits for several tasks).

It is easy to see that  $TC^0 \subseteq NC^1$ , where  $NC^1$  is the class of problems that can be solved in logarithmic depth by families of (bounded fan-in) AND/OR/NOT-circuits of polynomial size [2]. It is still an open problem to decide whether such an inclusion is proper. Indeed,  $TC^0 = NC^1$  would imply the collapse, from a certain level on, of the hierarchy  $TC^0_1 \subset TC^0_2 \subset TC^0_3 \subseteq TC^0_4 \subseteq \dots$  (see, e.g. [14], Th. IX.1.6) which is a hardly believed event (notice that the first three levels of the hierarchy have already been separated [7]). Thus, it is customarily and reasonably assumed that  $TC^0 \neq NC^1$ .

A problem f is  $TC^0$ -reducible to a problem g whenever f can be solved by a family of  $TC^0$  circuits with oracle gates for g. It is easy to see that  $g \in TC^0$  implies  $f \in TC^0$  as well.

Here, we impose no uniformity condition on our circuit families. As it turns out, this makes no difference to our conclusions (see [1] and [14], Sect. VIII.2 for a discussion).

Let us now briefly review some elementary notions from algebraic automata theory. For more details, we refer the reader to [4,6]. Given an alphabet  $\Sigma$ ,  $\Sigma^*$  denotes the free monoid of all strings on  $\Sigma$ . Given a language  $L \subseteq \Sigma^*$ , the syntactic monoid  $\mathcal{M}(L)$  is the quotient monoid  $\Sigma^*/\sim_L$ , where  $\sim_L \subseteq \Sigma^* \times \Sigma^*$  is the congruence defined as:  $x \sim_L y$  whenever  $vxw \in L$  if and only if  $vyw \in L$ , for any  $v, w \in \Sigma^*$ .

A deterministic (this attribute will always be understood) automaton  $A=(Q,\Sigma,\delta,q_0,F)$  consists of the finite set Q of states, the input alphabet  $\Sigma$ , the initial state  $q_0$ , the set  $F\subseteq Q$  of final states, and the transition function  $\delta:Q\times\Sigma\to Q$  that extends to strings as usual. The language recognized by A is the set  $L(A)=\{x\in\Sigma^*\mid \delta(q_0,x)\in F\}$ .

The recognizing group-like automaton on a group<sup>1</sup>  $(G, \cdot)$  is defined as  $\mathfrak{G}_i = (G, G, \cdot, i, \{i\})$  with i, the identity of  $(G, \cdot)$ , being both the initial and the unique final state. It is easy to see the relation

$$\mathcal{M}(L(\mathfrak{G}_{\mathfrak{i}})) \cong G,$$
 (1)

where "≅" stands for "isomorphic to".

A seminal result in [1] relates the possibility for a regular language to be recognized in  $TC^0$  with the "group structure" of its syntactic monoid. To see this, we first need some terminology. Given a class  $\mathscr K$  of algebras, we let the classes:  $\mathbf H(\mathscr K)$  of all homomorphic images, and  $\mathbf S(\mathscr K)$  of all subalgebras, of algebras in  $\mathscr K$ . A group is called *simple* whenever it has no other normal subgroups but the trivial ones. Thus, we have:

**Theorem 1.1.** ([1], Th. 8 (b)) Let L be a regular language. If  $TC^0 \neq NC^1$  then  $L \in TC^0$  if and only if each simple group in  $HS(\mathcal{M}(L))$  is Abelian.

<sup>&</sup>lt;sup>1</sup>We will always be considering *finite* groups.

This theorem can be rewritten for recognizing group-like automata as:

**Proposition 1.2.** Let  $\mathfrak{G}_i$  be the recognizing group-like automaton on a group  $(G,\cdot)$ . If  $TC^0 \neq NC^1$  then  $L(\mathfrak{G}_i) \in TC^0$  if and only if each simple group in HS(G) is Abelian.

*Proof.* Just observe that  $\mathcal{M}(L(\mathfrak{G}_i)) \cong G$ , as Relation (1) shows. Thus, the claimed result follows at once from Theorem 1.1.

Proposition 1.2 will be our main tool in the next section, where we inspect the possibility of performing iterated matrix multiplications with constant depth threshold circuits.

#### 2. The complexity of iterated matrix multiplication

We begin by studying the complexity of computing the iterated product of fixed-dimension integer matrices. Formally, this problem can be stated as

• ITERATED  $k \times k$  MATRIX PRODUCT (IMP<sub>k</sub>) INPUT: Integer matrices  $M_1, M_2, \ldots, M_n$  of dimension  $k \times k$ , with *n*-bit entries.

OUTPUT: The iterated product  $M_1 \cdot M_2 \cdot \ldots \cdot M_n$ .

We are going to give a fine – in terms of k – evaluation of the difficulty of  $\mathsf{IMP}_k$ .

To this purpose, for each prime power  $p^m > 3$ , consider the set LF $(2, p^m)$  of  $2 \times 2$  matrices of determinant unity, with entries in the Galois field GF $(p^m)$ . Moreover, let "·" be the usual row-column product with arithmetics performed in GF $(p^m)$ . It is a very well-known fact in group theory (see, e.g. [3], Chap. I (Second Part)) that:

**Theorem 2.1.** (LF(2,  $p^m$ ), ·) is a simple nonabelian group of order  $\frac{p^m(p^{2m}-1)}{2;1}$  (2;1 depending on p > 2; p = 2).

Now, let us consider the group  $(LF(2,5),\cdot)$  whose matrices have entries in GF(5) which actually is  $\mathbf{Z}_5$ . We can show that

**Theorem 2.2.** If  $TC^0 \neq NC^1$  then  $IMP_2$  on LF(2,5) does not belong to  $TC^0$ .

*Proof.* Let  $\mathfrak{L}\mathfrak{F}_i$  be the recognizing group-like automaton on the group (LF(2,5),·) in which arithmetics is performed "mod 5" (i denotes the  $2 \times 2$  identity matrix). If IMP<sub>2</sub> on LF(2,5) was in TC<sup>0</sup>, then membership in  $L(\mathfrak{L}\mathfrak{F}_i)$  could be checked in TC<sup>0</sup> as well.

In fact, to decide whether a string (of LF(2,5) matrices)  $\mu_1\mu_2\cdots\mu_n$  belongs to  $L(\mathfrak{LF}_i)$ , we could compute in  $TC^0$  the iterated product  $\mu_1\cdot\mu_2\cdot\ldots\cdot\mu_n$ , and accept if and only if the resulting matrix is i.

So, we would get that  $L(\mathfrak{L}_{i}) \in \mathrm{TC}^{0}$  but this, under the assumption  $\mathrm{TC}^{0} \neq \mathrm{NC}^{1}$ , would contradict Proposition 1.2, since  $\mathrm{HS}(\mathrm{LF}(2,5))$  contains  $\mathrm{LF}(2,5)$  itself which is a *simple nonabelian group*, as pointed out in Theorem 2.1.

Thus,  $IMP_2$  turns out to be hard for a *finite* group (namely, LF(2, 5) which has exactly 60 elements) of *integer* matrices with very small entries. Hence, a fortiori, the general  $IMP_2$  is hard, since otherwise we could apply (in  $TC^0$ , see Sect. 1) the "mod 5" transformation and solve  $IMP_2$  on LF(2, 5) in  $TC^0$ . The unique easy instance of  $IMP_k$  is then the trivial  $IMP_1$ , i.e., the iterated product of integers which is in  $TC^0$ , as observed in Section 1.

A brief remark is in order. It is quite obvious that the complexity analysis so far exhibited would remain unchanged if  $\mathsf{IMP}_k$  referred to  $k \times k$  matrices with rational entries expressed as pairs of n-bit integers (numerator, denominator). Thus, it is fair to use  $\mathsf{IMP}_k$  even to denote the problem of performing iterated multiplications of fixed-dimension rational matrices.

Summing up, we have that: if  $TC^0 \neq NC^1$  then  $IMP_k$  for rational matrices belongs to  $TC^0$  if and only if k = 1.

Let us now focus on a relevant subclass of rational matrices: the *stochastic matrices*, *i.e.*, matrices whose entries are rational numbers in the interval [0, 1], and where each row sum equals 1. Our interest in stochastic matrices comes also from the fact that fast algorithms for computing their iterated product would imply fast recognition of *probabilistic languages*, a topic that is investigated in [9, 10].

We soon discover that  $\mathsf{IMP}_k$  for stochastic matrices turns out to be a slightly more feasible problem. In fact, contrary to Theorem 2.2, we can show that

**Theorem 2.3.** For stochastic matrices,  $IMP_2$  belongs to  $TC^0$ .

*Proof.* First, notice that any  $2 \times 2$  stochastic matrix can be written as  $\begin{pmatrix} a & 1-a \\ b & 1-b \end{pmatrix}$ , where a and b are rational numbers in [0,1]. Thus, it is not hard to see that the iterated product  $P_1 \cdot P_2 \cdot \ldots \cdot P_n$  of  $2 \times 2$  stochastic matrices, with  $P_i = \begin{pmatrix} a_i & 1-a_i \\ b_i & 1-b_i \end{pmatrix}$ , yields the  $2 \times 2$  stochastic matrix  $P = \begin{pmatrix} \alpha_1 & 1-\alpha_1 \\ \alpha_2 & 1-\alpha_2 \end{pmatrix}$  where, for  $\ell = 1, 2$ , we have

$$\alpha_{\ell} = \sum_{i=1}^{n} \sigma_{i\ell} \prod_{j=i+1}^{n} (a_j - b_j)$$
 with  $\sigma_{i\ell} = \begin{cases} a_1 & \text{if } i = \ell = 1 \\ b_i & \text{otherwise.} \end{cases}$ 

We stipulate that the inner product yields 1 whenever the lower index exceedes the upper one. Hence, computing the entries of P reduces to sum a linear amount of products each one involving a linear amount of rationals. All this can be done in  $TC^0$ , as seen in Section 1.

Unfortunately, this is the only case of feasible iterated product for stochastic matrices, as witnessed by the following:

**Theorem 2.4.** There exists a finite set  $\mathscr{B}$  of  $3 \times 3$  stochastic matrices for which the iterated product does not belong to  $TC^0$ , unless  $TC^0 = NC^1$ .

*Proof.* We are to show that  $IMP_2$  on the *finite* group LF(2,5) is  $TC^0$ -reducible to the iterated product on a set  $\mathscr{B}$  of  $3 \times 3$  stochastic matrices, having the same cardinality as LF(2,5). Then, by Theorem 2.2, we get the claimed result.

We make use of the following transformation  $\Gamma$ , easily seen to be implemented in  $TC^0$ : let  $P=\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$  be a  $2\times 2$  matrix in LF(2,5), and hence with entries in  ${\bf Z}_5$ ; we define the  $3\times 3$  matrix

$$\Gamma(P) = \begin{pmatrix} \frac{a_{11}}{2^3} & \frac{a_{12}}{2^3} & 1 - \frac{a_{11} + a_{12}}{2^3} \\ \frac{a_{21}}{2^3} & \frac{a_{22}}{2^3} & 1 - \frac{a_{21} + a_{22}}{2^3} \\ 0 & 0 & 1 \end{pmatrix}.$$

Next, we let  $\mathscr{B} = \{\Gamma(P) \mid P \in \mathrm{LF}(2,5)\}$ . Obviously,  $\mathscr{B}$  is a set of  $3 \times 3$  stochastic matrices, with the same cardinality as  $\mathrm{LF}(2,5)$ . At this point, it is easy to see that, for any given n-tuple  $P_1, P_2, \ldots, P_n$  of matrices in  $\mathrm{LF}(2,5)$ , we have

$$\Gamma(P_1)\cdot\Gamma(P_2)\cdot\ldots\cdot\Gamma(P_n)=\left(egin{array}{ccc} rac{P_1\cdot P_2\cdot\ldots\cdot P_n}{2^{3n}} & 1-rac{lpha_1}{2^{3n}} \ 0 & 0 & 1 \end{array}
ight),$$

which is a  $3 \times 3$  stochastic matrix, where  $\alpha_1$  (resp.  $\alpha_2$ ) equals the sum of the entries in the first (resp. second) row of  $P_1 \cdot P_2 \cdot \ldots \cdot P_n$ . Hence, to compute  $P_1 \cdot P_2 \cdot \ldots \cdot P_n$  in LF(2, 5), we first compute  $\Gamma(P_i)$  in TC<sup>0</sup>, and then use an oracle to evaluate the iterated product  $\Gamma(P_1) \cdot \Gamma(P_2) \cdot \ldots \cdot \Gamma(P_n)$ . Finally, we read off  $P_1 \cdot P_2 \cdot \ldots \cdot P_n$  from the resulting matrix, and transform (in TC<sup>0</sup>) the entries "mod 5".

Indeed, the hardness of  $IMP_3$  on  $\mathcal{B}$  implies the hardness of  $IMP_k$  for general stochastic matrices, for any  $k \geq 3$ .

#### 3. The complexity of matrix powering

Let us now turn to study the parallel complexity of powering fixed-dimension integer matrices. The problem formalizes as:

•  $k \times k$  MATRIX POWERING (MPOW<sub>k</sub>) INPUT: An integer matrix M of dimension  $k \times k$ , with n-bit entries. OUTPUT: The n-th power  $M^n$ .

We are going to show that  $MPOW_k$  is in  $TC^0$ , for any k.

To this aim, we need to recall a few elementary notions from linear algebra (see, e.g. [13]). Let M be a  $k \times k$  integer matrix. Its characteristic polynomial is defined as  $p_M(x) = \det(M - xI) = (-1)^k x^k + c_{k-1} x^{k-1} + \cdots + c_1 x + c_0$ , where I is the  $k \times k$  identity matrix, while each  $c_i$  is known to be the sum of all the principal

minors of order (k-i), taken with the sign  $(-1)^i$ . The Cayley-Hamilton theorem states that

$$p_M(M) = \mathbf{0},\tag{2}$$

with **0** being the  $k \times k$  zero matrix. Let us see how to use this fact to efficiently compute  $M^n$ . If we divide  $x^n$  by  $p_M(x)$ , we obtain the equation

$$x^n = q(x) p_M(x) + r_M(x),$$
 (3)

where the remainder  $r_M(x)$  is a polynomial of degree not exceeding k-1. Evaluating equation (3) in M yields  $M^n = q(M) p_M(M) + r_M(M)$  and, by equation (2), we get

$$M^n = r_M(M)$$
.

This leads to the following algorithm to compute  $M^n$ :

- (i) compute  $p_M(x)$ ;
- (ii) compute  $r_M(x) = x^n \mod p_M(x)$ ;
- (iii) evaluate  $r_M(M)$ .

Such an algorithm can be implemented in  $TC^0$ .

Here, we will not examine the technical details of the implementation for which we refer the reader to [11] (where the number of neuron layers is also investigated). However, we would briefly argue that each step of the algorithm is in TC<sup>0</sup>.

- STEP (i): To get the *i*-th coefficient of  $p_M(x)$ , for  $0 \le i \le k-1$ , we basically have to sum  $\binom{k}{k-i}$  determinants of  $(k-i) \times (k-i)$  submatrices of M; this can be clearly done in constant depth. Hence, by computing in parallel all such coefficients, we obtain  $p_M(x)$  in  $TC^0$ .
- STEP (ii): We can refer to fast parallel algorithms for dividing polynomials presented, e.g., in [5]. This would suffice to show that  $r_M(x)$  can be computed in  $TC^0$ . In [11], we have preferred to suitably transform polynomials into integers, and then to operate with such integers.
- STEP (iii): The polynomial  $r_M(x)$  has degree at most k-1. Hence, computing  $r_M(M)$  amounts to computing a linear combination of powers  $M^i$ , with  $i \leq k-1$ . Even this task is easily seen to be in  $TC^0$ .

Thus, we can conclude that:

**Theorem 3.1.** For any k, MPOW<sub>k</sub> belongs to  $TC^0$ .

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