

### Bounded-Width Probabilistic OBDDs and Read-Once Branching Programs are Incomparable

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#### Abstract

Restricted branching programs are considered by the investigation of relationships between complexity classes of Boolean functions. Read-once ordered branching programs (or OBDDs) form the most restricted class of this computation model. Since the problem of proving exponential lower bounds on the complexity for general probabilistic OBDDs is open so far, it is interesting to study this problem in a restricted setting. For this reason we deal in this work with probabilistic OBDDs whose width is bounded.

We prove in this work that probabilistic OBDDs of width bounded by a constant can be more powerful than even non-deterministic readonce branching programs. To do it we present a probabilistic OBDD of constant width computing the known function PERM. We prove for several known functions that they cannot be computed by probabilistic OBDDs of constant width. To show it we present a new method allowing to obtain lower bound  $\Omega(n)$  on the width of corresponding OBDDs (n is the number of variables).

#### 1 Introduction

In order to study the relationship between different complexity classes restricted models of computation are considered. Branching programs are one of the most investigated computation models (see [17], [5] for a lot of references) during the last years. In particular, read-once ordered branching programs (*OBDDs*) determine complexity classes whose relationships

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are successfully proven [2], [11], [12]. The same time OBDDs are convenient tools to represent Boolean functions because of the possibility to manipulate them efficiently [7]. Probabilistic OBDDs are the most general OBDDs.

We recall basic definitions. A deterministic branching program P is a directed acyclic multi-graph with a source node and two distinguished sink nodes: accepting and rejecting. The out-degree of each non-sink node is exactly 2. Each node is labeled by some variable  $x_i$ , and two arcs outgoing from  $x_i$ -node are labeled by 0 and 1. The label "a" indicates that only inputs satisfying  $x_i = a$  may follow this arc in the computation. A branching program P computes a function  $h_n$  in the obvious way: for each  $\mathbf{x} \in \{0,1\}^n$ ,  $h_n(\mathbf{x}) = 1$  if and only if there is a directed path starting in the source and leading to the accepting node such that all labels along this path are consistent with  $\mathbf{x} = x_1 x_2 \dots x_n$ . The branching program becomes nondeterministic if we allow guessing nodes, that are nodes with two outgoing arcs being unlabeled. A non-deterministic branching program P outputs 1 on an input x if and only if there exists (at least one) computation on x starting in the source node and leading to the accepting node. A probabilistic branching program has in addition to its standard (deterministic) nodes specially designated nodes called random nodes. Each such node corresponds to a random input  $y_i$  having values from  $\{0,1\}$  each with probability 1/2. We say that a probabilistic branching program *computes* a function h if it outputs  $h(\mathbf{x})$  with probability greater than 1/2 for any input  $\mathbf{x}$ . If this probability is at least  $1/2 + \epsilon$  for some  $\epsilon > 0$  one says that the computation has bounded error  $\epsilon$ . A probabilistic branching program P on n variables determines a function  $c_P: \{0,1\}^n \to [0,1]; c_P(\mathbf{x})$  is the probability that P reaches the accepting sink on the input  $\mathbf{x}$ . We call this function the *characteristic* function of the branching program P.

We define the *complexity* of a branching program P as the number of its nodes. We denote the class of Boolean functions computable by polynomial size nondeterministic branching programs as NP-BP. We say that a function belongs to the class PP-BP if and only if there is a polynomial size probabilistic branching program computing this function. For a probabilistic computation with bounded error, we use another notation for the complexity class. Let  $BPP_{\epsilon}$ -BP be the class of functions computable with error  $\epsilon > 0$  by polynomial size probabilistic branching programs. Furthermore, let BPP- $BP := \bigcup_{0 < \epsilon \le 1/2} BPP_{\epsilon}$ -BP. For a restricted class of branching programs Q, we define analogous complexity classes using "-Q" as a suffix to their notations.

A read-once branching program (BP1) is a branching program in which every variable is tested at most once on each path. A BP1 is called ordered

(or OBDD) if the variables have to be tested according to some fixed ordering  $\pi$ . An OBDD is called *oblivious* if it can be leveled, i.e. arcs lead only to nodes of the neighboring level, and each level contains only x-nodes for some fixed variable x. The width of an OBDD is the maximum number of nodes belonging to a level. OBDDs having the width bounded by a constant (bwOBDD for short) are studied in this work. The restriction on the width was studied earlier for general deterministic branching programs [3]. Recently, Newman [16] showed that functions computable by deterministic bwOBDDs can be computed probabilistically with bounded error with constant number of queries.

Although there are results concerning incomparability of probabilistic OBDDs with bounded error on the one hand and non-deterministic OBDDs [1] or even non-deterministic BP1s [19] on the other hand, the power of probabilistic OBDDs without bounded error was not studied yet. In this paper we present a new method helping to find probabilistic OBDDs computing certain functions and a new technique to find lower bounds on the width of probabilistic OBDDs. We show in this work that probabilistic bwOBDD can be more powerful than non-deterministic BP1. We present basic lemmas helping to obtain different characteristic functions of some probabilistic bwOBDDs. Using these lemmas, we prove that the function PERM known to be hard for non-deterministic read-once branching programs can be computed by probabilistic bwOBDD.

There is no known exponential lower bound of the complexity of probabilistic OBDDs without bounded error. We prove for several known functions that they can not be computed by probabilistic OBDDs of a constant width. To show it we present a new method allowing to give lower bound  $\Omega(n)$  on the width of corresponding OBDDs (n is the number of variables).

### 2 Characteristic functions of probabilistic bwOBDDs

What kind of functions can be the characteristic functions of OBDDs or of bwOBDDs? We need the following modifications of simple lemmas from [14] (see also [15]).

**Lemma 1** For any constant  $\alpha$ ,  $0 \le \alpha \le 1$ , if the binary representation of  $\alpha$  has t positions then there exists a bwOBDD  $B(\alpha)$  of width 2 with t levels consisting only of random nodes such that  $c_{B(\alpha)} = \alpha$ .

**Lemma 2** Let  $c_{B_1}$  and  $c_{B_2}$  be the characteristic functions of bwOBDDs  $B_1$ 

of width  $w_1$  and  $B_2$  of width  $w_2$ , respectively, reading deterministic variables of the same set X in the same order. Then the following functions are the characteristic functions of some bwOBDDs with the same variable order:  $1 - c_{B_1}(\mathbf{x})$ ,  $1/2(c_{B_1}(\mathbf{x}) + c_{B_2}(\mathbf{x}))$ ,  $c_{B_1}(\mathbf{x})c_{B_2}(\mathbf{x})$ . These OBDDs that we denote as  $1 - B_1$ ,  $1/2(B_1 + B_2)$ , and  $B_1B_2$  have the width  $w_1$ ,  $w_1 + w_2$ , and  $w_1w_2$  respectively.

If bwOBDDs  $B_1$  and  $B_2$  have disjoint sets of variables then there are bwOBDDs  $1/2(B_1+B_2)$  and  $B_1B_2$  with characteristic functions  $1/2(c_{B_1}(\mathbf{x})+c_{B_2}(\mathbf{x}))$  and  $c_{B_1}(\mathbf{x})c_{B_2}(\mathbf{x})$  of width  $max(w_1,w_2)+1-sg(|w_1-w_2|)$  and  $max(w_1,w_2)$  respectively.

Using these lemmas it is easy to construct a bwOBDD computing some function  $f_n$  if a bwOBDD B with the following property is known. There is some number  $p_n$  such that  $c_B(\mathbf{x}) > p_n$  if and only if  $f_n(\mathbf{x}) = 1$ . The following theorem presents a method to produce a desired bwOBDD if for a known bwOBDD B and some number  $p_n$ ,  $c_B(\mathbf{x}) = p_n$  if and only if  $f_n(\mathbf{x}) = 1$ .

**Theorem 1** Let  $f_n$  be a function on n variables. Let  $p_n$  be a real and B be a probabilistic OBDD of width c with the following property. For every word  $\mathbf{x}$ ,  $|\mathbf{x}| = n$ ,  $f_n(\mathbf{x}) = 1$  if and only if  $c_B(\mathbf{x}) = p_n$ . Then  $f_n$  is computable by a probabilistic OBDD of width  $c^2 + 2c + 2$ .

*Proof.* Let B have n' levels. Then there is an  $\epsilon_n \geq (1/2)^{n'}$  such that if  $f(\mathbf{x}) = 0$  then  $|c_B(\mathbf{x}) - p_n| \geq \epsilon_n$ . Let  $s_n$  be a real with binary representation equal to the prefix of n' + 1 bits of the binary representation of  $p_n$ : i.e.  $p_n = 0.p_n^{(1)} \dots p_n^{(n'+1)} \dots$  and  $s_n = 0.p_n^{(1)} \dots p_n^{(n'+1)}$ .

Consider  $s_n > 1/2$ . Because of Lemmas 1 and 2, there exists a bwOBDD  $B_1 = B(\frac{1}{2}(1-B+B(p')))$  of width c(c+2) with the characteristic function  $c_{B_1} = c_B(\frac{1}{2}(1-c_B+p')), \ p'=2s_n-1$ . The function  $c_{B_1}$  has the maximum equal to  $p'' = s_n^2/2$  if  $c_B = s_n$ .

If  $s_n \leq 1/2$  then there exists a bwOBDD  $B_1 = (1-B)(\frac{1}{2}(B+B(p')))$  of width c(c+2) with the characteristic function  $c_{B_1} = (1-c_B)(\frac{1}{2}(c_B+p'))$ ,  $p'=1-2s_n$ . The function  $c_{B_1}$  has the maximum equal to  $p''=(1-s_n)^2/2$  if  $c_B=s_n$ .

For both cases if  $c_B(\mathbf{x}) \neq p_n$ , i.e.  $f_n(\mathbf{x}) = 0$ , then  $c_{B_1} \leq p'' - \epsilon_n^2/2$  otherwise  $f_n(\mathbf{x}) = 1$  and  $c_{B_1} \geq p'' - \frac{1}{2}^{(2(n'+1)+1)}$ . Let p''' be a real with binary representation equal to the prefix of 2n' + 3 bits of the binary representation of  $1 - p'' + \epsilon_n^2/4$ . The  $bwOBDD \ 1/2(B_1 + B(p'''))$  of width c(c+2) + 2 with the characteristic function  $\frac{1}{2}(c_{B_1} + p''')$  computes the function  $f_n$ .

We studied in this section *OBDDs* with a fixed variable ordering. Although these computation seem to be somewhat poor they are sufficiently powerful for our purpose.

# 3 Probabilistic OBDD of constant width can do more than polynomial nondeterministic BP1s

The author presented in [14] a function belonging to the class  $Q = PP-bwOBDD \setminus (BPP-OBDD \cup NP-OBDD)$ . Now we investigate the even harder function  $PERM_n$  corresponding to the set of permutation matrixes. It is known ([9], [13]) that this function is hard for nondeterministic read-once branching programs. Recall that  $PERM_n : \{0,1\}^{m^2} \to \{0,1\}, n = m^2$ , and  $PERM_n(\mathbf{x}) = 1$  if and only if every row and every column of the Boolean  $m \times m$ -matrix  $x = (x_{1,1}, x_{1,2}, \dots, x_{m,m})$  contains exactly one 1. The function  $PERM_n$  can be computed by a polynomial size probabilistic OBDD with bounded error [18]. Due to [14] probabilistic bwOBDDs with bounded error are not more powerful than deterministic bwOBDDs. Therefore,  $PERM_n$  does not belong to BPP-bwOBDD. We show that this function can be computed by a probabilistic bwOBDD if there is no bound on the error.

**Theorem 2** The function  $PERM_n$  can be computed by a probabilistic OBDD  $B(PERM_n)$  of constant width.

Proof. We describe the OBDD  $B(PERM_n)$  computing  $PERM_n$ . This OBDD reads variables in the order  $x_{1,1}, x_{1,2}, \ldots, x_{m,m}$ .  $B(PERM_n)$  has the following parts. For any  $i, 1 \leq i \leq m$ , a deterministic OBDD  $P_1^{(i)}$  and a probabilistic OBDD  $P_2^{(i)}$  read the i-th row of the matrix and check whether this row contains exactly one 1. If this is the case then  $P_2^{(i)}$  reaches the accepting sink with probability  $(\frac{1}{2})^j$  for  $x_{i,j} = 1$ . There exist such OBDDs having on each level 3 nodes one of which will be called rejecting node. All paths from this rejecting node lead to the rejecting sink.

We firstly describe a probabilistic OBDD P for which reals  $p_m$  and  $\epsilon_m$  exist such that  $c_P(\mathbf{x}) = p_m$  if  $PERM_n(\mathbf{x}) = 1$ , and  $|p_m - c_P(\mathbf{x})| \geq \epsilon_m$  otherwise. The source of P is a random node selecting  $P_2^{(1)}$  or  $P_1^{(1)}$ .

The accepting sink of  $P_1^{(i)}$ ,  $1 \le i < m-1$ , is identified with a random node leading to  $P_1^{(i+1)}$  or to  $P_2^{(i+1)}$ . This part can be written as  $P_1^{(i)}(\frac{1}{2}(P_1^{(i+1)} + P_2^{(i+1)}))$ . The accepting sink of  $P_1^{(m-1)}$  is identified with

a random node leading to  $P_2^{(m)}$  and to a rejecting node (the subprogram  $P_1^{(m-1)} \frac{1}{2} P_2^{(m)}$ ).

The accepting sink of  $P_2^{(i)}$ ,  $1 \le i < m$ , is identified with a random node leading to a rejecting node and to  $P_3^{(i+1)}$  (the subprogram  $P_2^{(i)} \frac{1}{2} P_3^{(i+1)}$ ).

The later program  $P_3^{(i)}$ ,  $2 \le i \le m$ , is the copy of  $P_1^{(i)}$  which accepting sink, for i < m, is identified with a random node leading to a rejecting node and to  $P_3^{(i+1)}$  (the subprogram  $P_1^{(i)} \frac{1}{2} P_3^{(i+1)}$ ). Note that  $P_3^{(i+1)}$  is reachable from  $P_2^{(i)}$  too.

P reaches the accepting sink only if each i-th row of X contains exactly one non-zero element  $x_{i,j_i}$  and  $c_P(\mathbf{x}) = (\frac{1}{2})^m \sum_{i=1}^m (\frac{1}{2})^{j_i}$  in this case. It is easy to see that the function  $c_P(\mathbf{x})$  is equal to

$$\left(\frac{1}{2}\right)^m \sum_{i=1}^m \left(\frac{1}{2}\right)^i = \left(\frac{1}{2}\right)^{m-1} - \left(\frac{1}{2}\right)^{2m} = p_m$$

if and only if  $\{j_i|1\leq j_i\leq m, 1\leq i\leq m\}=\{i|1\leq i\leq m\}$  i.e. x corresponds to a permutation matrix. Otherwise  $|p_m-c_P(\mathbf{x})|\geq (\frac{1}{2})^{2m}$ . Therefore there exists the desired probabilistic  $OBDD\ B(PERM_n)$  of constant width (Theorem 1).

**Corollary 1** The function  $PERM_n$  is computable by a probabilistic OBDD of width 59.

Indeed using the construction in the proof of Theorem 1 we transform the bwOBDD P having on each level 7 nodes with a rejecting node. Each level of the probabilistic OBDD  $B(PERM_n)$  has the unique rejecting node. Therefore  $B(PERM_n)$  has the width 7(6+2)+2+1=59.

## 4 Lower bounds on the width of probabilistic OBDDs

It is known that lower bounds on the complexity of OBDDs can be obtained from lower bounds for one-round communication complexity. One considers in such an investigation the following matrix. Let  $X_1$  be a subset of variables set X. Then for a function f on X denote as  $CM_f^{X_1}$  the communication matrix. Each row of this matrix corresponds to an assignment a to  $X_1$  and each column to an assignment b to  $X \setminus X_1$ . The element  $CM_f^{X_1}(a;b)$  of  $CM_f^{X_1}$  on the intersection of this row and this column is equal to f(a;b):

(a; b) denotes in our paper the complete assignment to X where, for example, variables in  $X_1$  obtain assignments corresponding to a.

We shall show that there are functions that can not be computed by probabilistic bwOBDDs. These functions are functions known to be hard for randomized OBDDs too ([11], [20]). Some of these functions are so-called k-stable functions. But k-stable functions are hard for read-once branching programs too (see e.g., [10]). Because there are functions in the class P- $BP1 \setminus PP$ -bwOBDD weaker property than k-stability can be sufficient for our purpose. The same statement holds if one considers BPP-OBDD: not only k-stable functions do not belong to BPP-OBDD. Therefore, for example, Sauerhoff [20] examined by an investigation of the complexity of a function in the context of probabilistic OBDD-computations whether this function can be reduced from the known function  $INDEX_m$  being hard for one-round communication games. We show directly that functions satisfying a generalized version of k-stability are hard for probabilistic bwOBDDs without bounded error.

A Boolean  $m \times n$  matrix is full if its rows contain every vector from  $\{0,1\}^n$ . A Boolean matrix is k-full if it has a  $2^k \times k$  submatrix being full. For example the communication matrix of the function  $INDEX_m$  is m-full. If the communication matrix  $CM_f^{X_1}$  is k-full then assignments corresponding to the rows and the columns of a  $2^k \times k$  full submatrix of  $CM_f^{X_1}$  are called critical.

**Definition 1** Let f be a function on a set X of n variables. We call this function  $(k_1, k_2)$ -indefinite if for any  $X_1 \subseteq X$ ,  $|X_1| = k_1$ , the communication matrix  $CM_f^{X_1}$  is  $k_2$ -full.

To understand the relationship between our definition and the definition of k-stability we just note that if a function is k-stable then it is (k, k)-indefinite.

**Theorem 3** A probabilistic oblivious OBDD computing a  $(k_1, k_2)$ -indefinite function f has width at least  $k_2$ .

*Proof.* Let f be computable by some probabilistic oblivious OBDD P. Consider the level L of P when exactly  $k_1$  variables from X are read. Let this set of read variables be  $X_1$ . Let L have  $k_2 - 1$  nodes.

Let  $\{b_i|1 \leq i \leq k_2\}$  and  $A = \{a_i|1 \leq i \leq 2^{k_2}\}$  be sets of critical assignments to  $X \setminus X_1$  and to  $X_1$  respectively. For each  $a_i \in A$ , we consider the probabilistic distribution  $\mu(a_i) = (\mu^{(1)}(a_i) \dots \mu^{(k_2-1)}(a_i))$  to reach the

nodes of L:  $\mu^{(j)}(a_i)$  is the probability to reach the j-th node of L from the source of P if the variables from  $X_1$  have values  $a_i$ . We also consider for each  $b_i$ ,  $1 \le i \le k_2$ , a column vector  $\nu(b_i) = (\nu^{(1)}(b_i) \dots \nu^{(k_2-1)}(b_i))$ :  $\nu^{(j)}(b_i)$  is the probability to reach the accepting sink of P from the j-th node of L if the values of variables from  $X \setminus X_1$  correspond to  $b_i$ . Then

$$CM_f^{X_1}(a_i;b_j) = 1 \leftrightarrow c_P(a_i;b_j) = \mu(a_i)\nu(b_j) > 1/2.$$

There are coefficients  $\alpha_i$  not all equal to 0 such that  $\sum_{i=1}^{k_2} (\alpha_i \nu(b_i)) = 0$ . Without loss of generality, assume that  $\alpha_i \neq 0$  for all  $i, 1 \leq i \leq k_2$ , and assume

$$\nu(b_1)=\sum_{i=2}^{k_2}(\alpha_i\nu(b_i)).$$

We consider two cases:  $\sum_{i=2}^{k_2} \alpha_i \leq 1$  and  $\sum_{i=2}^{k_2} \alpha_i > 1$ . We take  $a \in A$  such that for the first case,  $CM_f^{X_1}(a;b_i) = 1$  if  $\alpha_i < 0$ ,  $CM_f^{X_1}(a;b_i) = 0$  if  $\alpha_i > 0$ , where  $2 \leq i \leq k_2$ , and  $CM_f^{X_1}(a;b_1) = 1$ . Then

$$1/2 < c_P(a;b_1) = \mu(a)\nu(b_1) = \sum_{i=2}^{k_2} (\alpha_i \mu(a)\nu(b_i)) < 1/2 \sum_{i=2}^{k_2} (\alpha_i) \le 1/2.$$

If  $\sum_{i=2}^{k_2} (\alpha_i) > 1$  then we take  $a \in A$  such that  $CM_f^{X_1}(a; b_i) = 1$  if  $\alpha_i > 0$ ,  $CM_f^{X_1}(a; b_i) = 0$  if  $\alpha_i < 0$ ,  $2 \le i \le k_2$ , and  $CM_f^{X_1}(a; b_1) = 0$ . Then

$$1/2 > c_P(a;b_1) = \mu(a)\nu(b_1) = \sum_{i=2}^{k_2} (\alpha_i \mu(a)\nu(b_i)) > 1/2 \sum_{i=2}^{k_2} (\alpha_i) > 1/2.$$

It is possible to find this a because  $\{b_i|i=1,k_2\}$  and  $A=\{a_i|i=1,2^{k_2}\}$  are the sets of critical assignments to  $X\setminus X_1$  and to  $X_1$  respectively. The obtained contradictions prove the theorem.

Corollary 2 The k-stable functions do not belong to PP-bwOBDD.

The following functions are well investigated:  $ISA_n$ ,  $ACH_n$ , and  $HWB_n$  (see for the definitions [6], [4], and [8] respectively). It is known that these functions belong to P-BP1 and to NP- $OBDD \subset PP$ -OBDD ([2]).

**Corollary 3** Let  $n = 2^r + r$  and  $n = 2m + \log m$ . The functions on n variables  $ISA_n$ ,  $ACH_n$ , and  $HWB_n$  require a probabilistic OBDDs without bounded error of width  $k = \frac{2^r}{r} - 1$ , m/4, and 0.1n respectively.

We omit here simple proofs following from the original proofs of [6], [4], and [8] respectively that  $ISA_n$  is the (k, k)-indefinite function,  $ACH_n$  is the (m, m/4)-indefinite function, and  $HWB_n$  is the (0.6n, 0.1n)-indefinite function.

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