

# A Lower Bound for Nonadaptive, One-Sided Error Testing of Unateness of Boolean Functions over the Hypercube

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## Abstract

A Boolean function  $f : \{0, 1\}^d \mapsto \{0, 1\}$  is unate if, along each coordinate, the function is either nondecreasing or nonincreasing. In this note, we prove that any nonadaptive, one-sided error unateness tester must make  $\Omega(\frac{d}{\log d})$  queries. This result improves upon the  $\Omega(\frac{d}{\log^2 d})$  lower bound for the same class of testers due to Chen et al. (STOC, 2017).

## 1 Introduction

We study the problem of deciding whether a Boolean function  $f : \{0, 1\}^d \mapsto \{0, 1\}$  is *unate* in the property testing model [7, 5]. A function is unate if, for each dimension  $i \in [d]$ , the function is either nondecreasing along the  $i^{\text{th}}$  coordinate or nonincreasing along the  $i^{\text{th}}$  coordinate. A property tester for unateness is a randomized algorithm that takes as input a proximity parameter  $\varepsilon \in (0, 1)$  and has query access to a function  $f$ . If  $f$  is unate, it must accept with probability at least  $2/3$ . If  $f$  is  $\varepsilon$ -far from unate, it must reject with probability at least  $2/3$ . A tester has *one-sided error* if it always accepts unate functions. A tester is *nonadaptive* if it chooses all of its queries in advance; it is *adaptive* otherwise.

The problem of testing unateness was introduced by Goldreich et al. [4]. Following a result of Khot and Shinkar [6], Baleshzar et al. [1] settled the complexity of unateness testing for *real-valued functions*. Unateness can be tested with  $O(\frac{d}{\varepsilon})$  queries adaptively and with  $O(\frac{d \log d}{\varepsilon})$  queries nonadaptively. For constant  $\varepsilon$ , these complexities are optimal.

On the other hand, for the Boolean range, the complexity is far from settled. Baleshzar et al. [2] proved that  $\Omega(\sqrt{d})$  queries are necessary for nonadaptive, one-sided error testers. Chen et al. [3] improved the lower bound for this class of testers to  $\Omega(\frac{d}{\log^2 d})$ . They also proved a lower bound of  $\Omega(\frac{\sqrt{d}}{\log^2 d})$  for adaptive, two-sided error unateness testers.

In this note, we use a construction similar to the one used by Chen et al. [3] to get an  $\Omega(\frac{d}{\log d})$  for nonadaptive, one-sided error unateness testers of Boolean functions over the hypercube. Our analysis of the lower bound construction is simpler and gives a better dependence on  $d$ . There is

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still a gap of  $\log^2 d$  between the query complexity of the best known algorithm for this problem (from [1]) and our lower bound.

## 2 The Lower Bound

In this section, we prove the following theorem.

**Theorem 2.1.** *Any nonadaptive, one-sided error unateness tester for functions  $f : \{0, 1\}^d \mapsto \{0, 1\}$  with the distance parameter  $\varepsilon \leq \frac{1}{8}$  must make  $\Omega(\frac{d}{\log d})$  queries.*

*Proof.* We first define a hard distribution consisting of Boolean functions that are  $\frac{1}{8}$ -far from unate. By Yao's minimax principle [8], it is sufficient to give a distribution on functions for which every deterministic tester fails with high probability. A deterministic nonadaptive tester is determined by a set of query points  $Q \subseteq \{0, 1\}^d$ . We prove that if  $|Q| \leq \frac{d}{30 \log d}$ , then the tester fails with probability more than  $2/3$  over the hard distribution.

The hard distribution  $\mathcal{D}$  is defined as follows: pick 3 dimensions  $a, b, c \in [d]$  uniformly at random and define  $f_{a,b,c}(x) = x_a \cdot x_b + (1 - x_a) \cdot x_c$ . We call  $a, b, c$  the *influential dimensions*, since the value of the function depends only on them. The coordinate  $x_a$  determines if  $f_{a,b,c}(x)$  should be set to  $x_b$  or  $x_c$ . If  $x_a = 1$ , then  $f_{a,b,c}(x) = x_b$ , otherwise,  $f_{a,b,c}(x) = x_c$ .

There are  $\binom{d}{3}$  functions in the support of  $\mathcal{D}$ . The next claim states that all of them are far from unate.

**Claim 2.2.** *Every function  $f_{a,b,c}$  in the support of  $\mathcal{D}$  is  $\frac{1}{8}$ -far from unate.*

*Proof.* Consider an edge  $(x, y)$  along the dimension  $a$ . We have  $x_a = 0$  and  $y_a = 1$ , and  $x_i = y_i$  for all  $i \in [d] \setminus \{a\}$ . By definition,  $f_{a,b,c}(x) = x_c$  and  $f_{a,b,c}(y) = y_b$ . If  $x_b = y_b = 1$  and  $x_c = y_c = 0$ , then  $f_{a,b,c}$  is increasing along the edge  $(x, y)$ . On the other hand, if  $x_b = y_b = 0$  and  $x_c = y_c = 1$ , then  $f_{a,b,c}$  is decreasing along  $(x, y)$ . Thus, with respect to  $f_{a,b,c}$ , at least  $2^{d-3}$  edges along the dimension  $a$  are decreasing and at least  $2^{d-3}$  edges along the dimension  $a$  are increasing. Hence, at least  $2^{d-3}$  function values of  $f_{a,b,c}$  need to be changed to make it unate. Consequently,  $f_{a,b,c}$  is  $\frac{1}{8}$ -far from unate.  $\square$

Note that any one-sided error tester for unateness must accept if the query answers are consistent with a unate function. Let  $f|_Q$  denote the restriction of the function  $f$  to the points in  $Q$ . We say that  $f|_Q$  is *extendable* to a unate function if there exists a unate function  $g$  such that  $g|_Q = f|_Q$ . For  $f \sim \mathcal{D}$ , we show that if  $|Q| \leq \frac{d}{30 \log d}$ , then, with high probability,  $f|_Q$  is extendable to a unate function. Consequently, the tester accepts with high probability.

Next, we define a conjunctive normal form (CNF) formula  $\phi(f|_Q)$ . Intuitively, each pair  $(x, y)$  of domain points on which  $f$  differs imposes a constraint on  $f$  (assuming that  $f$  is unate). Specifically, at least one of the dimensions on which  $x$  and  $y$  differ must be consistent (i.e., nondecreasing or nonincreasing) with the change of the function value between  $x$  and  $y$ . This constraint is formalized in the definition of  $\phi(f|_Q)$  as follows. For each dimension  $i$ , we have a variable  $z_i$  which is true if  $f$  is nondecreasing along the dimension  $i$ , and false if it is nonincreasing along that dimension. For each  $x, y \in Q$  such that  $f(x) = 1$  and  $f(y) = 0$ , create a clause (think of  $x, y$  as sets where  $i \in x$  iff  $x_i = 1$ )

$$c_{x,y} = \bigvee_{i \in x \setminus y} z_i \vee \bigvee_{i \in y \setminus x} \bar{z}_i.$$

Set  $\phi(f|_Q) = \bigwedge_{x,y \in Q: f(x)=1, f(y)=0} c_{x,y}$ .

**Observation 2.3.** *The restriction  $f|_Q$  is a certificate for non-unateness iff  $\phi(f|_Q)$  is unsatisfiable.*

Now we need to show that, with probability greater than  $2/3$  over  $f \sim \mathcal{D}$ , the CNF formula  $\phi(f|_Q)$  is satisfiable. This follows from Claims 2.4 and 2.5.

The width of a clause is the number of literals in it; the width of a CNF formula is the minimum width of a clause in it.

**Claim 2.4.** *With probability at least  $2/3$  over  $f \sim \mathcal{D}$ , the width of  $\phi(f|_Q)$  is at least  $3 \log d$ .*

*Proof.* Consider a graph  $G$  with vertex set  $Q$ , and an edge between  $x, y \in Q$  if  $|x \Delta y| \leq 3 \log d$  (Here,  $x \Delta y$  is the symmetric difference between the sets  $x$  and  $y$ ). Take an arbitrary spanning forest  $F$  of  $G$ . Observe that for any edge  $(u, v)$  of  $G$ , we have  $u \Delta v \subseteq \bigcup_{(x,y) \in F} x \Delta y$ . Note that  $F$  has at most  $\frac{d}{30 \log d}$  edges. Let  $C = \bigcup_{(x,y) \in F} x \Delta y$ , the set of dimensions captured by  $Q$ . We have  $|C| \leq \sum_{(x,y) \in F} |x \Delta y| \leq \frac{d}{30 \log d} \cdot 3 \log d \leq \frac{d}{10}$ . Over the distribution  $\mathcal{D}$ , the probability that at least one of the influential dimensions,  $\{a, b, c\}$ , is in  $C$  is at most  $3/10$  which is less than  $1/3$ . Hence, with probability at least  $2/3$ , no  $(u, v) \in G$  contributes a clause to  $\phi(f|_Q)$ . Therefore, the width of  $\phi(f|_Q)$  is at least  $3 \log d$ .  $\square$

**Claim 2.5.** *Any CNF that has width at least  $3 \log d$  and at most  $d^2$  clauses is satisfiable.*

*Proof.* Apply the probabilistic method. A clause is not satisfied by a random assignment with probability at most  $1/d^3$ . Hence, the expected number of unsatisfied clauses is at most  $\frac{d^2}{d^3} < 1$ .  $\square$

Thus,  $f|_Q$  is a certificate for non-unateness with probability at most  $1/3$  when  $|Q| \leq \frac{d}{30 \log d}$ , which completes the proof of Theorem 2.1.  $\square$

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