

## Note

### The Equivalence of Two Problems on the Cube

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*Communicated by the Managing Editors*

Received November 5, 1990

Denote by  $Q_n$  the graph of the hypercube  $C^n = \{+1, -1\}^n$ . The following two seemingly unrelated questions are equivalent: 1. Let  $G$  be an induced subgraph of  $Q_n$  such that  $|V(G)| \neq 2^{n-1}$ . Denote  $\Delta(G) = \max_{x \in V(G)} \deg_G(x)$  and  $\Gamma(G) = \max(\Delta(G), \Delta(Q_n - G))$ . Can  $\Gamma(G)$  be bounded from below by a function of  $n$ ? 2. Let  $f: C^n \rightarrow \{+1, -1\}$  be a boolean function. The *sensitivity* of  $f$  at  $x$ , denoted  $s(f, x)$ , is the number of neighbors  $y$  of  $x$  in  $Q_n$  such that  $f(x) \neq f(y)$ . The sensitivity of  $f$  is  $s(f) = \max_{x \in C^n} s(f, x)$ . Denote by  $d(f)$  the degree of the unique representation of  $f$  as a real multilinear polynomial on  $C^n$ . Can  $d(f)$  be bounded from above by a function of  $s(f)$ ? © 1992 Academic Press, Inc.

#### 1. PRELIMINARIES

Denote by  $Q_n$  the graph on the  $n$ -dimensional cube  $C^n = \{+1, -1\}^n$ , where any two vertices are adjacent iff they differ in exactly one component. For an induced subgraph  $G$  of  $Q_n$ , denote the *maximal degree* of  $G$  by  $\Delta(G)$ , i.e.,

$$\Delta(G) = \max_{x \in V(G)} \deg_G(x).$$

In [1], it was shown that if  $G$  contains more than  $2^{n-1}$  vertices, then

$$\Delta(G) > \frac{1}{2}(\log n - \log \log n + 1)$$

\* Supported in part by an Eshkol doctoral fellowship, administered by the National Council for R&D, Israel Ministry of Science.

† Research supported in part by US-Israel Binational Science Foundation Grant #0378114 and a grant from the Israel Academy of Sciences.

and there exists a  $G$  such that

$$\Delta(G) < \sqrt{n} + 1.$$

The value  $\sqrt{n}$  is also conjectured to be the correct order of magnitude for a lower bound on  $\Delta(G)$ . Denote  $\Gamma(G) = \max(\Delta(G), \Delta(Q_n - G))$ .

Let  $f: C^n \rightarrow \{+1, -1\}$  be a boolean function. The *sensitivity* of  $f$  at  $x$ , denoted by  $s(f, x)$ , is the number of neighbors  $y$  of  $x$  for which  $f(x) \neq f(y)$ . The sensitivity of  $f$  is

$$s(f) = \max_{x \in C^n} s(f, x)$$

The sensitivity of  $f$  is sometimes called the *critical complexity* of  $f$ .

In theoretical computer science, much effort has been expended in the definition of various measures of complexity of boolean functions. Some are derived from an underlying computational model, such as *decision tree depth*. Here the function is computed by repeatedly reading input bits, until the function can be determined from the bits accessed. The *cost* of an algorithm is the number of bits read on the worst case input, and the complexity of a function is the cost of the best algorithm for this function. A similar measure is the *certificate complexity*. A 1-certificate (0-certificate) for  $f$  is an assignment to some subset of the variables that forces the value of  $f$  to 1 (0). The certificate complexity of  $f$  on  $x$ , denoted  $C(f, x)$ , is the size of the smallest certificate that agrees with  $x$ . The certificate complexity of  $f$  is

$$C(f) = \max_{x \in C^n} C(f, x).$$

Other measures of complexity are of a combinatorial nature, e.g., sensitivity. A related measure is *block-sensitivity*, defined: Denote  $[n] = \{1, \dots, n\}$  and let  $R \subset [n]$ . If  $x$  is the vector  $(x_1, \dots, x_n)$ , then  $x^{(R)}$  is defined as the vector with coordinates:

$$x_i^{(R)} = \begin{cases} x_i, & i \notin R \\ -x_i, & i \in R. \end{cases}$$

The block sensitivity of  $f$  at  $x$ , denoted  $bs(f, x)$ , is the largest number  $t$  such that there exist  $t$  *disjoint* sets  $R_1, \dots, R_t$  such that for all  $1 \leq i \leq t$ ,  $R_i \subset [n]$ , and  $f(x) \neq f(x^{(R_i)})$ . The block-sensitivity of  $f$  is

$$bs(f) = \max_{x \in C^n} bs(f, x).$$

A central activity in this field is determining the relation between various

measures. The measures of complexity  $s_1$  and  $s_2$  are *equivalent* if they are polynomially related; i.e., there exist polynomials  $p_1(x)$  and  $p_2(x)$  such that

$$\forall f, \quad s_1(f) \leq p_2(s_2(f)), \quad s_2(f) \leq p_1(s_1(f)).$$

Nisan [3] showed that decision tree depth, certificate complexity, and block-sensitivity are equivalent. Nisan first considered the more natural measure of sensitivity (which is block-sensitivity restricted to singletons), but was unable to prove equivalence to decision tree depth and certificate complexity. However, only after introducing block-sensitivity was equivalence obtained.

Yet another complexity measure is obtained from the unique representation of the boolean function  $f$  as a real multilinear polynomial over the cube:

$$f(x) = \sum_{I \subset [n]} \left[ \alpha_I \prod_{i \in I} x_i \right].$$

The coefficient  $\alpha_I$  (which satisfies  $-1 \leq \alpha_I \leq 1$  for all  $I \subset [n]$ ) is also called  $\hat{f}(I)$ , the *Fourier transform* of  $f$  at  $I$ . Denote by  $d(f)$  the *degree* of this polynomial, i.e.,

$$d(f) = \max_{I \subset [n]} \{ |I| : \alpha_I \neq 0 \}.$$

Nisan and Szegedy [4] show that  $d(f)$  is also equivalent to the three complexity measures mentioned above. As for the relation between sensitivity and degree, Szegedy [6] showed that

$$d(f) \geq \sqrt{s(f)}.$$

This can easily be shown to be tight. Whether  $s(f)$  is also equivalent to all of the above is still unknown. In particular, an upper bound on  $d(f)$  in terms of  $s(f)$  is sought and is conjectured to be  $s^2(f)$ . Such a bound would mean that sensitivity is equivalent to all the previously mentioned quantities. In the next section we show that this upper bound is equivalent to a lower bound on  $\Gamma(\cdot)$ .

## 2. THE EQUIVALENCE THEOREM

**THEOREM 2.1.** *The following are equivalent for any function  $h: N \rightarrow R$ :*

1. *For any induced subgraph  $G$  of  $Q_n$  such that  $|V(G)| \neq 2^{n-1}$ ,  $\Gamma(G) \geq h(n)$ .*
2. *For any boolean function  $f$ ,  $d(f) < h^{-1}(s(f))$ .*

*Proof.* We first transform 1 into a statement concerning boolean functions: Associate with the subgraph  $G$  a boolean function  $g$  such that  $g(x) = 1$  iff  $x \in V(G)$ . Note that  $\deg_G(x) = n - s(g, x)$  for  $x \in V(G)$  and the same holds in  $Q_n - G$  for  $x \notin V(G)$ . Denote by  $E(g)$  the average value of  $g$  on  $C^n$ . Now 1 and 2 are clearly equivalent to the following:

1'. For any boolean function  $g$ ,  $E(g) \neq 0$  implies  $\exists x: s(g, x) \leq n - h(n)$ .

2'. For any boolean function  $f$ ,  $s(f) < h(n)$  implies  $d(f) < n$ .

To see the equivalence of 1' and 2', define

$$g(x) = f(x) p(x),$$

where  $p(x)$  is the parity function of  $x$ :  $p(x) = \prod_{i=1}^n x_i$ . Note that for all  $x \in C^n$ ,  $s(g, x) = n - s(f, x)$  and for all  $I \subset [n]$ ,  $\hat{g}(I) = \hat{f}([n] - I)$ , therefore  $E(g) = \hat{g}(\emptyset) = \hat{f}([n])$ , where  $\hat{f}([n])$  is the Fourier transform of  $f$  at  $[n]$ , i.e., the highest order coefficient in the representation of  $f$  as a polynomial.

1'  $\rightarrow$  2'. Assume that  $d(f) = n$ , i.e.,  $\hat{f}([n]) \neq 0$ . This is equivalent to  $E(g) \neq 0$ . By 1',  $\exists x: s(g, x) \leq n - h(n)$ ; therefore  $\exists x: s(f, x) \geq h(n)$ , contradicting the premise.

2'  $\rightarrow$  1'. Assume that  $\forall x, s(g, x) > n - h(n)$ . This implies that  $s(f) < h(n)$ . By 2',  $d(f) < n$ , which is equivalent to  $\hat{f}([n]) = \hat{g}(\emptyset) = E(g) = 0$ , contradicting the premise. ■

### 3. CONCLUSION

Substituting  $h(x) = \sqrt{x}$  in Theorem 2.1 shows that the two bounds  $\Delta(G) \geq \sqrt{n}$  and  $d(f) < s^2(f)$  are equivalent. The example from [1] which shows that there exists  $G$  such that  $\Delta(G) \leq \sqrt{n} + 1$  can be used to show that the upper bound on  $d(f)$  would be tight if it were true. All of this means that a proof of  $\Delta(G) \geq \sqrt{n}$  would imply that boolean function sensitivity is equivalent to all other complexity measures mentioned in [3].

The sensitivity complexity measure  $s(f)$  is especially important, since it also lower bounds  $T(f)$ —the time needed by a parallel RAM to compute  $f$  (a parallel RAM is a collection of synchronous parallel processors sharing a global memory with no write-conflicts allowed). Cook and Dwork [2] have shown that  $T(f) \geq \log s(f)$ . In fact, Nisan [3] later improved this to  $T(f) \geq \log bs(f)$  (this is a stronger inequality, since for any  $f$ ,  $bs(f) \geq s(f)$ ). Simon [5] has also shown that a  $n$ -variable boolean function which depends on all its variables must have sensitivity at least  $\Omega(\log n)$ .

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