Sensitivity, Affine Transforms and Quantum Communication Complexity

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Abstract

In this paper, we study the Boolean function parameters sensitivity (s), block sensitivity (bs), and alternation (alt) under specially designed affine transforms and show several applications. For a function $f : \mathbb{F}_2^n \to \{-1, 1\}$, and $A = Mx + b$ for $M \in \mathbb{F}_2^{n \times n}$ and $b \in \mathbb{F}_2^n$, the result of the transformation g is defined as $\forall x \in \mathbb{F}_2^n, g(x) = f(Mx + b)$.

As a warm up, we study alternation under linear shifts (when M is restricted to be the identity matrix) called the *shift invariant alternation* (the smallest alternation that can be achieved for the Boolean function f by shifts, denoted by $\mathsf{salt}(f)$). By a result of Lin and Zhang [ICALP 2017], it follows that $\mathsf{bs}(f) \leq O(\mathsf{salt}(f)^2 \mathsf{s}(f))$. Thus, to settle the Sensitivity Conjecture ($\forall\ f, \mathsf{bs}(f) \leq \mathsf{poly}(\mathsf{s}(f))$), it suffices to argue that $\forall\ f, \mathsf{salt}(f) \leq \mathsf{poly}(\mathsf{s}(f))$. However, we exhibit an explicit family of Boolean functions for which $\mathsf{salt}(f)$ is $2^{\Omega(\mathsf{s}(f))}$.

Going further, we use an affine transform A, such that the corresponding function g satisfies $\mathsf{bs}(f, 0^n) \leq \mathsf{s}(g)$. We apply this in the setting of quantum communication complexity to prove that for $F(x, y) \overset{\text{def}}{=} f(x \wedge y)$, the bounded error quantum communication complexity of F with prior entanglement, $Q_{1/3}^*(F)$ is $\Omega(\sqrt{\mathsf{bs}(f, 0^n)})$. Our proof builds on ideas from Sherstov [Quantum Information and Computation, 10:435455, 2010] where we use specific properties of the above affine transformation. Using this, we show the following.

(a) For a fixed prime p and an ϵ, $0 < \epsilon < 1$, any Boolean function f that depends on all its inputs with $\mathsf{deg}_p(f) \leq (1 - \epsilon) \log n$ must satisfy $Q_{1/3}^*(F) = \Omega\left(\frac{n^{\epsilon/2}}{\log n}\right)$. Here, $\mathsf{deg}_p(f)$ denotes the degree of the multilinear polynomial over $\mathbb{F}_p$ which agrees with f on Boolean inputs.

(b) For Boolean function f such that there exists primes p and q with $\mathsf{deg}_q(f) \geq \Omega(\mathsf{deg}_p(f)^\delta)$ for $\delta > 2$, the deterministic communication complexity - $\mathsf{D}(F)$ and $Q_{1/3}^*(F)$ are polynomially related. In particular, this holds when $\mathsf{deg}_p(f) = O(1)$. Thus, for this class of functions, this answers an open question (see Buhrman and de Wolf [CCC 2001]) about the relation between the two measures.

Restricting back to the linear setting, we construct linear transformation A, such that the corresponding function g satisfies, $\mathsf{alt}(f) \leq 2\mathsf{s}(g) + 1$. Using this new relation, we exhibit Boolean functions f (other than the parity function) such that $\mathsf{s}(f)$ is $\Omega(\sqrt{\mathsf{sparsity}(f)})$ where $\mathsf{sparsity}(f)$ is the number of non-zero coefficients in the Fourier representation of f. This family of Boolean functions also rule out a potential approach to settle the XOR Log-Rank conjecture via the recently settled Sensitivity conjecture [Hao Huang, Annals of Mathematics, 190(3): 949-955, 2019].

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1   Introduction

For a Boolean function $f : \{0,1\}^n \to \{-1,1\}$, *sensitivity* of f on $x \in \{0,1\}^n$, is the maximum number of indices $i \in [n]$, such that $f(x \oplus e_i) \neq f(x)$ where $e_i \in \{0,1\}^n$ with exactly the ith bit as 1. The *sensitivity* of f (denoted by $\mathsf{s}(f)$) is the maximum sensitivity of f over all inputs. A related parameter is the *block sensitivity* of f (denoted by $\mathsf{bs}(f)$), where we allow disjoint blocks of indices to be flipped instead of a single bit. Another parameter is the deterministic *decision tree complexity* (denoted by $\mathsf{DT}(f)$) which is the depth of an optimal decision tree computing the function f. The *certificate complexity* of f (denoted by $\mathsf{C}(f)$) is the non-deterministic variant of the decision tree complexity. The parameter $\mathsf{s}(f)$ was originally studied by Cook *et al.* [CDR86] in connection with the CREW-PRAM model of computation. Subsequently, Nisan and Szegedy [NS94] (see also [Nis91]) introduced the parameters $\mathsf{bs}(f)$ and $\mathsf{C}(f)$ and conjectured that for any function $f : \{0,1\}^n \to \{-1,1\}$, $\mathsf{bs}(f) \leq \mathsf{poly}(\mathsf{s}(f))$ - known as the Sensitivity Conjecture. Later developments, which revealed several connections between sensitivity, block sensitivity and the other Boolean function parameters, demonstrated the fundamental nature of the conjecture (see [HKP11] for a survey and several equivalent formulations of the conjecture). This conjecture has recently been resolved in [Hua19] by showing the following which implies that $\mathsf{bs}(f) = O(\mathsf{s}(f)^4)$.

Theorem 1.1 (Sensitivity Theorem [Hua19])**.** *For every Boolean function f, $\mathsf{deg}(f) \leq \mathsf{s}(f)^2$.*

Shi and Zhang [ZS10] studied the parity complexity variants of $\mathsf{bs}(f), \mathsf{C}(f)$ and $\mathsf{DT}(f)$ and observed that such variants have the property that they are invariant under arbitrary invertible linear transforms (over $\mathbb{F}_2^n$). They also showed existence of Boolean functions where under *all* invertible linear transforms of the function, the decision tree depth is linear while their parity variant of decision tree complexity is at most logarithmic in the input length.

Our Results : While the existing studies focus on understanding the Boolean function parameters under the effect of arbitrary invertible affine transforms, in this work, we study the relationship between the above parameters of Boolean functions $f : \mathbb{F}_2^n \to \{-1,1\}$, under specific affine transformations over $\mathbb{F}_2^n$. More precisely, we explore the relationship of the above parameters for the function $g : \mathbb{F}_2^n \to \{-1,1\}$ and f, where g is defined as $g(x) = f(Mx + b)$ for specific $M \in \mathbb{F}_2^{n \times n}$ and $b \in \mathbb{F}_2^n$ (where is M not necessarily invertible). We show the following results, and their corresponding applications, which we explain along with the context in which they are relevant.

Alternation under shifts : We study the parameters when the transformation is very structured - namely the matrix M is the identity matrix and $b \in \mathbb{F}_2^n$ is a linear shift. More precisely, we study $f_b(x) \stackrel{\text{def}}{=} f(x+b)$ where b is the shift. Observe that all the parameters mentioned above are invariant under shifts. A Boolean function parameter which is neither shift invariant nor invariant under invertible linear transforms is the *alternation*, a measure of non-monotonicity of Boolean function (see Section 2 for a formal definition). To see this for the case of shifts, if we take f as the majority function on n bits, then there exists shifts $b \in \{0,1\}^n$ where $\mathsf{alt}(f_b) = \Omega(n)$ while $\mathsf{alt}(f) = 1$.

A result related to Sensitivity Conjecture by Lin and Zhang [LZ17] shows that $\mathsf{bs}(f) \leq O(\mathsf{s}(f)\mathsf{alt}(f)^2)$. This bound for $\mathsf{bs}(f)$, implies that to settle the Sensitivity Conjecture, it suffices to show that $\mathsf{alt}(f)$ is upper bounded by $\mathsf{poly}(\mathsf{s}(f))$ for all Boolean functions f. However, the authors [DS19] ruled this out, by exhibiting a family of functions where $\mathsf{alt}(f)$ is at least $2^{\Omega(\mathsf{s}(f))}$.

Observing that the parameters $\mathsf{s}(f), \mathsf{bs}(f)$ are invariant under shifts, we define a new quantity *shift-invariant alternation*, $\mathsf{salt}(f)$, which is the minimum alternation of any function g obtained from f upon shifting by a vector $b \in \{0,1\}^n$ (Definition 3.1). By the aforementioned bound on

$\mathsf{bs}(f)$ of [LZ17], it is easy to observe that $\mathsf{bs}(f) \leq O(\mathsf{s}(f)\mathsf{salt}(f)^2)$. We also show that there exists a family of Boolean functions f with $\mathsf{bs}(f) = \Omega(\mathsf{s}(f)\mathsf{salt}(f))$ (Proposition 3.5).

It is conceivable that $\mathsf{salt}(f)$ is much smaller compared to $\mathsf{alt}(f)$ for a Boolean function f and hence that $\mathsf{salt}(f)$ can potentially be upper bounded by $\mathsf{poly}(\mathsf{s}(f))$ thereby settling the Sensitivity Conjecture. However, we rule this out by showing the following stronger gap, about the same family of functions demonstrated in [DS19] (see also [GSW16]).

Proposition 1.2. *There exists an explicit family of Boolean functions for which* $\mathsf{salt}(f)$ *is* $2^{\Omega(\mathsf{s}(f))}$.

Block Sensitivity under Affine Transformations : We now generalize our theme of study to the affine transforms over $\mathbb{F}_2^n$. In particular, we explore how to design affine transformations in such a way that block sensitivity of the original function (f) is upper bounded by the sensitivity of the new function (g). We use $\mathsf{bs}(f, a)$ to denote the number of sensitive blocks of f on the input a.

Lemma 1.3. *For any* $f : \mathbb{F}_2^n \to \{-1, 1\}$ *and* $a \in \{0, 1\}^n$, *there exists an affine transform* $A : \mathbb{F}_2^n \to \mathbb{F}_2^n$ *such that for* $g(x) = f(A(x))$,

(a) $\mathsf{bs}(f, a) \leq \mathsf{s}(g, 0^n)$, *and*

(b) $g(x) = f((x_{i_1}, x_{i_2}, \ldots, x_{i_n}) \oplus a)$ *where* $i_1, \ldots, i_n \in [n]$ *are not necessarily distinct.*

The above transformation is used in Nisan and Szegedy (see Lemma 7 of [NS94]) to show that $\mathsf{bs}(f) \leq 2\mathsf{deg}(f)^2$. Here, $\mathsf{deg}(f)$ is the degree of the multilinear polynomial over reals that agrees with f on Boolean inputs. We show another application of Lemma 1.3 in the context of quantum communication complexity, a model for which was introduced by Yao [Yao93]. In this model, two parties Alice and Bob have to compute a function $F : \{0,1\}^n \times \{0,1\}^n \to \{-1,1\}$, where Alice is given an $x \in \{0,1\}^n$ and Bob is given a $y \in \{0,1\}^n$. Both the parties have to come up with a *quantum protocol* where they communicate qubits via a quantum channel and compute f while minimizing the number of qubits exchanged (which is the *cost* of the quantum protocol) in the process. In this model, we allow protocols to have prior entanglement. We define $Q_{1/3}^*(F)$ as the minimum cost quantum protocol computing F with prior entanglement. For more details on this model, see [Raz03]. The corresponding analog in the classical setting is the bounded error randomized communication model where the parties communicate with $0, 1$ bits and share an unbiased random source. We define $\mathsf{R}_{1/3}(F)$ as the minimum cost randomized protocol computing F with error at most $1/3$. It can be shown that $Q_{1/3}^*(F) \leq \mathsf{R}_{1/3}(F) \leq \mathsf{D}(F)$.

One of the fundamental goals in quantum communication complexity is to see if there are functions where their randomized communication complexity is significantly larger than their quantum communication complexity. It has been the conjectured by Shi and Zhu [SZ09] that this is not the case in general (which they called the Log-Equivalence Conjecture). In this work, we are interested in the case when $F(x, y)$ is of the form $f(x \wedge y)$ where $f : \{0,1\}^n \to \{-1,1\}$ and $x \wedge y$ is the string obtained by bitwise AND of x and y.

Question 1.4. *For* $f : \{0,1\}^n \to \{-1,1\}$, *let* $F : \{0,1\}^n \times \{0,1\}^n \to \{-1,1\}$ *be defined as* $F(x, y) = f(x \wedge y)$. *Is it true that for any such* F, $\mathsf{D}(F) \leq \mathsf{poly}(Q_{1/3}^*(F))$?

Since $\mathsf{R}_{1/3}(F) \leq \mathsf{D}(F)$, answering the above question in positive would show that the classical randomized communication model is as powerful as the quantum communication model for the class of functions $F(x, y) = f(x \wedge y)$. This question for such restricted F has also been proposed

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by Klauck [Kla07] as a first step towards answering the general question (see also [BdW01]). In this direction, Razborov [Raz03] showed that for the special case when f is symmetric, $F(x,y) = f(x \wedge y)$ satisfy $\mathsf{D}(F) \leq O(Q_{1/3}^*(F)^2)$. In the process, Razborov developed powerful techniques to obtain lower bounds on $Q_{1/3}^*(F)$ which were subsequently generalized by Sherstov [She08], Shi and Zhu [SZ09]. Subsequently, in a slightly different direction, Sherstov [She10] showed that instead of computing $F(x,y) = f(x \wedge y)$ alone, if we consider F to be the problem of computing both of $F_1(x,y) = f(x \wedge y)$ and $F_2(x,y) = f(x \vee y)$, then $\mathsf{D}(F) = O(Q_{1/3}^*(F)^{12})$ for all Boolean functions f where $Q_{1/3}^*(F) = \max \left\{ Q_{1/3}^*(F_1), Q_{1/3}^*(F_2) \right\}$ and $\mathsf{D}(F) = \max \left\{ \mathsf{D}(F_1), \mathsf{D}(F_2) \right\}$. Using Lemma 1.3, we build on the ideas of Sherstov [She10] and obtain a lower bound for $Q_{1/3}^*(F)$ where $F(x,y) = F_1(x,y) = f(x \wedge y)$.

Theorem 1.5. *Let* $f : \{0,1\}^n \to \{-1,1\}$ *and* $F(x,y) = f(x \wedge y)$, *then,*

$$Q_{1/3}^*(F) = \Omega\left(\sqrt{\mathsf{bs}(f, 0^n)} \right).$$

In this context, we make an important comparison[1] with a result of Sherstov [She10]. He proved that for $F'(x,y) = f_b(x \wedge y)$, where $b \in \{0,1\}^n$ is the input on which $\mathsf{bs}(f,x)$ is maximum, $Q_{1/3}^*(F') = \Omega(\sqrt{\mathsf{bs}(f)}) \geq \Omega(\sqrt{\mathsf{bs}(f,0^n)})$ (Corollary 4.5 of [She10]). Notice that F and F' differ by a linear shift of f with b.[2] Moreover, $Q_{1/3}^*(F)$ can change drastically even under such (special) linear shifts of f. For example, consider $f = \wedge_n$. Since $\mathsf{bs}(f)$ is maximized at 1^n, $b = 1^n$. Hence, the function F' is the disjointness function for which $Q_{1/3}^*(F') = \Omega(\sqrt{n})$ [Raz03] whereas, $Q_{1/3}^*(F) = O(1)$. The same counterexample also shows that $Q_{1/3}^*(F) = \Omega(\sqrt{\mathsf{bs}(f)})$ cannot hold for all f (see Remark 4.2). Since the lower bounds shown on quantum communication complexity are on different functions, Theorem 1.5 is incomparable with the result of Sherstov (Corollary 4.5 of [She10]).

Using the above result, for a prime p, we show that if f has small degree when expressed as a polynomial over $\mathbb{F}_p$ (denoted by $\deg_p(f)$), the quantum communication complexity of F is large.

Theorem 1.6. *Fix a prime* p. *Let* $f : \{0,1\}^n \to \{-1,1\}$ *where* f *depends on all the variables. Let* $F(x,y) = f(x \wedge y)$. *For any* $0 < \epsilon < 1$ *such that* $\deg_p(f) \leq (1 - \epsilon) \log n$, *we have*

$$Q_{1/3}^*(F) = \Omega\left(\frac{n^{\epsilon/2}}{\log n} \right).$$

Observe that, though Theorem 1.5 does not answer Question 1.4 in positive for all functions, we could show a class of Boolean function for which $\mathsf{D}(F)$ and $Q_{1/3}^*(F)$ are polynomially related. More specifically, we show this for the set of all Boolean functions f such that there exists two distinct primes p, q with $\deg_p(f)$ and $\deg_q(f)$ are sufficiently far apart (Theorem 1.7).

Theorem 1.7. *Let* $f : \{0,1\}^n \to \{-1,1\}$ *with* $F(x,y) = f(x \wedge y)$. *Fix* $0 < \epsilon < 1$. *If there exists distinct primes* p, q *such that* $\deg_q(f) = \Omega(\deg_p(f)^{\frac{2}{1-\epsilon}})$, *then* $\mathsf{D}(F) = O(Q_{1/3}^*(F)^{2/\epsilon})$.

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[1] Recently, it was noticed that Theorem 1.5 had already appeared in arXiv version 1 of [She09] but did not appear in later versions.

[2] More importantly, this b in Corollary 4.5 of [She10] cannot be fixed to 0^n for all Boolean functions to conclude Theorem 1.5. See Section A for details.

By the result of Gopalan *et al.* (Theorem 1.2, [GLS09]), any Boolean function f with $\deg_p(f) = o(\log n)$ must have $\deg_q(f) = \Omega(n^{1-o(1)})$ thereby satisfying the condition of Theorem 1.7. Hence for all such functions, Theorem 1.7 answers Question 1.4 in positive. Observe that the same can also be derived from Theorem 1.6.

Alternation under Linear Transforms : We now restrict our study to linear transforms. Again, in this context, the aim is to design special linear transforms for the parameters of interest. In particular, in this case, we show linear transforms for which we can upper bound the alternation of the original function in terms of the sensitivity of the resulting function. More precisely, we prove the following lemma:

Lemma 1.8. *For any $f : \mathbb{F}_2^n \to \{-1, 1\}$, there exists an invertible linear transform $L : \mathbb{F}_2^n \to \mathbb{F}_2^n$ such that for $g(x) = f(L(x))$,*

$$\mathsf{alt}(f) \leq 2\mathsf{s}(g) + 1.$$

We show an application of the above result in the context of the parameter sensitivity. Nisan and Szegedy [NS94] showed that for any Boolean function f, $\mathsf{s}(f) \leq 2\deg(f)^2$. However, the situation is quite different for $\deg_2(f)$ - noticing that for f being parity on n variables, $\deg_2(f) = 1$ and $\mathsf{s}(f) = n$ - the gap can even be unbounded. Though parity may appear as a corner case, there are other functions like the Boolean inner product function[3] IP_n whose $\mathbb{F}_2$-degree is constant while sensitivity is $\Omega(n)$ thereby ruling out the possibility that $\mathsf{s}(f) \leq \deg_2(f)^2$. It is known that if f is not the parity on n variables (or its negation), $\deg_2(f) \leq \log \mathsf{sparsity}(f)$ [BC99, GOS$^+$11]. Hence, as a structural question about the two parameters, we ask : for f other than the parity function, is it true that $\mathsf{s}(f) \leq \mathsf{poly}(\log \mathsf{sparsity}(f))$.[4] In fact, the Sensitivity Theorem (Theorem 1.1) by [Hua19] implies that for every Boolean function f, $\log \mathsf{sparsity}(f) = O(\mathsf{s}(f)^2)$. Hence, if we could answer our question in affirmative, it would imply that $\mathsf{s}(f)$ and $\log \mathsf{sparsity}(f)$ are polynomially related. We use Lemma 1.8, which is in the theme of studying alternation and sensitivity in the context of linear transformations, to show that this is not the case, by exhibiting a family of functions where the gap is exponential.

Theorem 1.9. *There exists a family of functions $\{g_k \mid k \in \mathbb{N}\}$ such that*

$$\mathsf{s}(g_k) \geq \frac{\sqrt{\mathsf{sparsity}(g_k)}}{2} - 1.$$

This family of Boolean functions also rules out a potential approach to settle the XOR Log-Rank conjecture via the recently settled Sensitivity conjecture [Hua19]. We elaborate on this approach and how our function family rules it out in Section 5.

2 Preliminaries

In this section, we define the notations used. Define $[n] = \{1, 2, \ldots, n\}$. For $S \subseteq [n]$, define $e_S \in \{0,1\}^n$ to be the indicator vector of the set S. For $x, y \in \{0,1\}^n$, we denote $x \wedge y$ (resp. $x \oplus y$) $\in \{0,1\}^n$ as the string obtained by bitwise AND (resp. XOR) of x and y. We use x_i to denote the ith bit of x.

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[3]$\mathsf{IP}_n(x_1, x_2, \ldots, x_n, y_1, y_2, \ldots, y_n) = \sum_i x_i y_i \mod 2$

[4]Observe that functions like IP_n though have low $\mathbb{F}_2$-degree similar to parity however have high sparsity and hence does not rule this out.

We now define the Boolean function parameters we use. Let $f : \{0,1\}^n \to \{-1,1\}$ and $a \in \{0,1\}^n$, we define, 1) the *sensitivity* of f on a as $\mathsf{s}(f, a) = |\{i \mid f(a \oplus e_i) \neq f(a), i \in [n]\}|$, 2) the *block sensitivity* of f on a, $\mathsf{bs}(f, a)$ to be the maximum number of disjoint blocks $\{B_i \mid B_i \subseteq [n]\}$ such that $f(a \oplus e_{B_i}) \neq f(a)$ and 3) the *certificate complexity* of f on a, $\mathsf{C}(f, a)$ to be the size of the smallest set $S \subseteq [n]$ such that fixing f according to a on the location indexed by S causes the function to become constant. For $\phi \in \{\mathsf{s}, \mathsf{bs}, \mathsf{C}\}$, we define $\phi(f) = \max_{a \in \{0,1\}^n} \phi(f, a)$ and are respectively called the sensitivity, the block sensitivity and the certificate complexity of f. By definition, the three parameters are shift invariant, by which we mean $\forall\, b \in \{0,1\}^n$, $\phi(f_b) = \phi(f)$ for $\phi \in \{\mathsf{s}, \mathsf{bs}, \mathsf{C}\}$ where $f_b(x) \overset{\text{def}}{=} f(x \oplus b)$. Also, it can be shown that $\mathsf{s}(f) \leq \mathsf{bs}(f) \leq \mathsf{C}(f)$.

For $x, y \in \{0,1\}^n$, define $x \prec y$ if $\forall i \in [n]$, $x_i \leq y_i$. We define a *chain* $\mathcal{C}$ on $\{0,1\}^n$ as $(0^n = x^{(0)}, x^{(1)}, \ldots, x^{(n-1)}, x^{(n)} = 1^n)$ such that for all $i \in [n]$, $x^{(i)} \in \{0,1\}^n$ and $x^{(i-1)} \prec x^{(i)}$. We define *alternation* of f for a chain $\mathcal{C}$, denoted $\mathsf{alt}(f, \mathcal{C})$ as the number of times the value of f changes in the chain. We define alternation of a function $\mathsf{alt}(f)$ as $\max_{\text{chain } \mathcal{C}} \mathsf{alt}(f, \mathcal{C})$.

Every Boolean function f can be expressed uniquely as a multilinear polynomial $p(x)$ in $\mathbb{F}[x_1, \ldots, x_n]$ over any field $\mathbb{F}$ such that $p(x) = f(x)\ \forall x \in \{0,1\}^n$. Fix a prime p. We denote $\deg(f)$ (resp. $\deg_p(f)$) to be the degree of the multilinear polynomial computing f over reals (resp. $\mathbb{F}_p$). We define $\mathsf{DT}(f)$ as the depth of an optimal decision tree computing f. It is known that for all Boolean functions f, $\deg_p(f) \leq \deg(f) \leq \mathsf{DT}(f) \leq \mathsf{bs}(f)^3$.

Sparsity of a Boolean function $f : \{0,1\}^n \to \{-1,1\}$ (denoted by $\mathsf{sparsity}(f)$) is the number of non-zero Fourier coefficients in the Fourier representation of f. For more details on this parameter, see [O'D14]. For more details on $\mathsf{DT}(f)$ and other related parameters, see the survey by Buhrman, de Wolf [BdW02] and Hatami *et al.* [HKP11].

We consider the two party classical communication model. Given a function $f : \{0,1\}^n \times \{0,1\}^n \to \{-1,1\}$, Alice is given an $x \in \{0,1\}^n$ and Bob is given $y \in \{0,1\}^n$. They can communicate with each other and their aim is to compute $f(x,y)$ while communicating minimum number of bits. We call the procedure employed by Alice and Bob to computing f as the *protocol*. We define $\mathsf{D}(f)$ as the minimum cost of a deterministic protocol computing f. For functions of the form $F(x, y) = f(x \wedge y)$, it is known that $\mathsf{D}(F) \leq 2\mathsf{DT}(f)$ [MO09]. For more details on communication complexity of Boolean functions, refer [KN06].

3   Warm up: Alternation under Shifts

In this section, as a warm-up, we study sensitivity and alternation under linear shifts (when the matrix M is the identity matrix). We introduce a parameter, *shift-invariant alternation* (salt). We then show the existence of Boolean functions whose shift-invariant alternation is exponential in its sensitivity (see Proposition 1.2) thereby ruling out the possibility that $\mathsf{salt}(f)$ can be upper bounded by a polynomial in $\mathsf{s}(f)$ for all Boolean functions f.

Recall from the introduction that the parameters s, bs and C are shift invariant while alt is not. To see that alt is not shift-invariant, for an even number n, consider the Boolean function defined as $\mathsf{Maj}_n(x) = 1 \iff \sum_i x_i > n/2$. For an even n, define $\mathsf{ShiftMaj}_n(x) = \mathsf{Maj}_n(x \oplus 1^{n/2}0^{n/2})$. It is possible to exhibit a chain σ such that $\mathsf{alt}(\mathsf{ShiftMaj}_n, \sigma) = n$, while $\mathsf{alt}(\mathsf{ShiftMaj}_n(x \oplus 1^{n/2}0^{n/2})) = \mathsf{alt}(\mathsf{Maj}_n) = 1$.

We define a variant of alternation which is invariant under shifts.

Definition 3.1 (Shift-invariant Alternation). For $f : \{0,1\}^n \to \{-1,1\}$, the *shift-invariant alternation* (denoted by $\mathsf{salt}(f)$) is defined as $\min_{b \in \{0,1\}^n} \mathsf{alt}(f_b)$.

We remark that $\mathsf{salt}(\mathsf{ShiftMaj}_n) = 1$. Hence the gap between measures alt and salt can be unbounded.

A family of functions with $\mathsf{salt}(f) = \Omega(2^{\mathsf{s}(f)})$: We now exhibit a family of functions $\mathcal{F}$ where for all $f \in \mathcal{F}$, $\mathsf{salt}(f) \geq 2^{\mathsf{s}(f)}$ thereby ruling out the possibility that $\mathsf{salt}(f)$ can be upper bounded by a polynomial in $\mathsf{s}(f)$. The family $\mathcal{F}$ is the same class of Boolean functions for which alternation is at least exponential in sensitivity due to [DS19].

Definition 3.2 (Definition 1 from [DS19]. See also Proof of Lemma A.1 of [GSW16])**.** Consider the family defined as follows.

$$\mathcal{F} = \left\{ f_k \mid f_k : \{0,1\}^{2^k-1} \to \{-1,1\}, k \in \mathbb{N} \right\}$$

The Boolean function f_k is computed by a decision tree which is a full binary tree of depth k with 2^k leaves. A leaf node is labeled as 0 (resp. 1) if it is the left (resp. right) child of its parent. All the nodes (except the leaves) are labeled by a distinct variable.

We remark that Gopalan *et al.* [GSW16] demonstrates an exponential lower bound on tree sensitivity (introduced by them as a generalization of the parameter sensitivity) in terms of decision tree depth for the same family of functions in Definition 3.2. We remark that, in general, lower bound on tree sensitivity need not implies a lower bound on alternation. For instance, if we consider the Majority function Maj_n, the tree sensitivity can be shown to be $\Omega(n)$ while alternation is 1.

The authors [DS19] have shown that for any $f \in \mathcal{F}$, there exists of a chain of large alternation in f. However, this is not sufficient to argue existence of a chain of large alternation under *every* linear shift. We now proceed to prove an exponential lower bound on $\mathsf{salt}(f)$ in terms of $s(f)$ for all $f \in \mathcal{F}$.

Proposition 1.2. For $f_k \in \mathcal{F}$, $\mathsf{salt}(f_k) \geq 2^{\Omega(\mathsf{s}(f_k))}$.

Proof. We show[5] that for $f_k \in \mathcal{F}$ and $n = 2^k - 1$, for all $c \in \{0,1\}^n$, $\mathsf{alt}(f_k(x \oplus c)) \geq 2^{k-2}$. Since $\mathsf{s}(f_k) \leq k$ by construction of f_k, the result follows.

Proof is by induction on k. For $k = 2$, f is a function on 3 variables and it can be verified that for all $c \in \{-1,1\}^3$, $\mathsf{alt}(f(x \oplus c)) \geq 1$. Now consider an $f_{k+1} \in \mathcal{F}$ computed by a decision tree T with the variable x_t as its root. Let h_1 and h_2 be the left and right subtrees of x_t in T. Note that $h_1(z')$ and $h_2(z'')$ depends on $n = 2^k - 1$ variables and belongs to $\mathcal{F}$ by construction. Hence, by induction, for all $c \in \{-1,1\}^n$, $\mathsf{alt}(h_1(z' \oplus c))$ and $\mathsf{alt}(h_2(z'' \oplus c))$ is at least 2^{k-2}. For $m = 2^{k+1} - 1$, consider any $c = (c', b, c'') \in \{-1,1\}^m$ where $c', c'' \in \{0,1\}^n$ and $b \in \{-1,1\}$. Since h_1 and h_2 are variable disjoint, $\mathsf{alt}(f(x \oplus c)) \geq \mathsf{alt}(h_1(z' \oplus c')) + \mathsf{alt}(h_2(z'' \oplus c'')) \geq 2^{k-2} + 2^{k-2} = 2^{k-1}$ completing the induction. $\square$

A family of functions with $\mathsf{bs}(f) = \Omega(\mathsf{s}(f)\mathsf{salt}(f))$: Lin and Zhang [LZ17] showed that for any Boolean function $f : \{0,1\}^n \to \{-1,1\}$,

$$\mathsf{bs}(f) = O(\mathsf{alt}(f)^2 \mathsf{s}(f)) \tag{1}$$

The fact that the measures bs and s are invariant under shifts implies the following proposition.

Proposition 3.3. *For any $f : \{0,1\}^n \to \{-1,1\}$, $\mathsf{bs}(f) \leq O(\mathsf{salt}(f)^2\mathsf{s}(f))$.*

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[5]In this proof, for simplicity, we abuse the notation $f_k(x \oplus c)$ to denote the function obtained by shifting f_k by c.

Proof. For any $b \in \{0, 1\}^n$, recall that $f_b(x)$ is defined to be $f(x \oplus b)$. Applying Eq. (1) to f_b, we get that $\mathsf{bs}(f_b) = O(\mathsf{alt}(f)^2\mathsf{s}(f_b))$. Since, bs and s are invariant under shifts, for any b, $\mathsf{bs}(f) = \mathsf{bs}(f_b) = O(\mathsf{alt}(f_b)^2\mathsf{s}(f_b)) = O(\mathsf{alt}(f_b)^2\mathsf{s}(f))$. Choosing b to be a shift that minimizes the alternation of f_b completes the proof. $\qquad\square$

We now exhibit a family of functions for which $\mathsf{bs}(f)$ is at least $\frac{\mathsf{s}(f) \cdot \mathsf{salt}(f)}{4}$.

Before proceeding, we show a tight composition result for alternation of Boolean functions when composed with OR_k (which is the k bit Boolean OR function).

For functions $f_1, \ldots, f_k$ where each $f_i : \{0, 1\}^n \to \{-1, 1\}$, define the function $OR_k \circ \overline{f} : \{-1, 1\}^{nk} \to \{-1, 1\}$ as $\vee_{i=1}^k f_i(x^{(i)})$ where for each $i \in [k]$, $x^{(i)} = (x_1^{(i)}, \ldots, x_n^{(i)}) \in \{0, 1\}^n$ is input to the function f_i.

Lemma 3.4. *Consider k Boolean functions $f_1, \ldots, f_k$ where each $f_i : \{0, 1\}^n \to \{-1, 1\}$ satisfy, $f_i(0^n) = f_i(1^n) = 0$. Then,*

$$\mathsf{alt}(OR_k \circ \overline{f}) = \sum_{i=1}^k \mathsf{alt}(f_i).$$

Proof. Let $f = OR_k \circ \overline{f}$ and $\mathcal{C}$ be a chain in $\{-1, 1\}^{nk}$ for which $\mathsf{alt}(f, \mathcal{C})$ is maximized. Without loss of generality, let all the functions be non-constant. Let $\mathcal{C}_i$ be the chain in $\{0, 1\}^n$ obtained by restricting $\mathcal{C}$ to variables $x_1^{(i)}, \ldots, x_n^{(i)}$ of f_i. Observe that if f changes it value, it must be that at least one of the f_i's have changed their evaluation along the chain $\mathcal{C}$. Since the functions are variable disjoint, such a change must be witnessed in the chain $\mathcal{C}_i$ for some i. Hence

$$\mathsf{alt}(f) = \mathsf{alt}(f, \mathcal{C}) \leq \sum_{i=1}^k \mathsf{alt}(f_i, \mathcal{C}_i) \leq \sum_{i=1}^k \mathsf{alt}(f_i)$$

To show that $\mathsf{alt}(f) \geq \sum_{i=1}^k \mathsf{alt}(f_i)$, we exhibit a chain $\mathcal{C}$ in $\{-1, 1\}^{nk}$ of alternation $\sum_{i=1}^k \mathsf{alt}(f_i)$. Let $\mathcal{C}_i = (0^n = z^{(i0)} \prec z^{(i1)} \prec z^{(i2)} \prec \ldots \prec z^{(in)} = 1^n)$ be a chain in $\{0, 1\}^n$ for which f_i achieves maximum alternation. We construct a chain $\mathcal{C}$ by "gluing" together these k chains. More precisely, let $\mathcal{C}$ by the chain such that for all $i \in [k]$, when restricted to the variables $x_1^{(i)}, \ldots, x_n^{(i)}$, we get a chain given by,

$$\overbrace{0^n \prec \ldots \prec 0^n}^{n(i-1) \text{ times}} \prec z^{(i0)} \prec z^{(i1)} \prec z^{(i2)} \prec \ldots \prec z^{(in)} \prec \overbrace{1^n \prec \ldots \prec 1^n}^{n(k-i) \text{ times}}$$

By construction of $\mathcal{C}$, since $f_j(0^n) = f_j(1^n) = 0$ for all $j \in [k]$, at any input of the chain $\mathcal{C}$, there is exactly one f_i that causes f to alternate. Hence,

$$\mathsf{alt}(f, \mathcal{C}) \geq \sum_{i=1}^k \mathsf{alt}(f_i, \mathcal{C}_i) = \sum_{i=1}^k \mathsf{alt}(f_i)$$

$\qquad\square$

Proposition 3.5. *There exists a family of Boolean functions for which $\mathsf{bs}(f) \geq \frac{\mathsf{s}(f) \cdot \mathsf{salt}(f)}{4}$.*

Proof. We consider the Rubinstein's function $f_R : \{0,1\}^{n^2} \to \{0,1\}$ [Rub95] where the input is treated as $n \times n$ matrix which evaluates to 1 iff there is a row with two consecutive ones starting at the odd position and rest of the entries being zero. Alternatively, we can view f_R as $OR_n \circ \overline{h}$ with $h : \{0,1\}^n \to \{-1,1\}$ where $h(a) = 1$ iff there are two consecutive ones starting at the odd position with rest of the entries as zero in $a \in \{0,1\}^n$. It can be verified that $\mathsf{alt}(h) = 2$. Since $h(0^n) = h(1^n) = 0$, applying Lemma 3.4 with $f_i = h$ for all $i \in [n]$, we get that $\mathsf{alt}(f_R) = \mathsf{alt}(h) \cdot n = 2n$. It is known that $\mathsf{bs}(f_R) \geq \frac{n^2}{2}$ while $\mathsf{s}(f_R) \leq n$ [Rub95], thereby showing that $\mathsf{bs}(f_R) \geq \frac{\mathsf{s}(f_R) \cdot \mathsf{alt}(f_R)}{4} \geq \frac{\mathsf{s}(f_R) \cdot \mathsf{salt}(f_R)}{4}$. $\qquad\square$

We remark that the above bound is stronger than what is needed in the context because, $\mathsf{bs}(f_R) \geq \frac{\mathsf{s}(f_R) \cdot \mathsf{alt}(f_R)}{4}$.

Lower bounding salt **:** By definition, $\mathsf{salt}(f) \leq \mathsf{alt}(f)$ and in addition, we have seen a Boolean function f for which $\mathsf{salt}(f) = 1$ while $\mathsf{alt}(f) = \Omega(n)$. This makes $\mathsf{alt}(f)$ particularly unsuitable in obtaining lower bounds on $\mathsf{salt}(f)$. We define a modified variant of the measure alternation called as *subcube alternation* and show that this new measure is always a lower bound on $\mathsf{salt}(f)$.

To define this variant, we define the following notion of restrictions. For any $S \subseteq [n]$, define $f|_S$ as the function f defined on the domain $\{x | x \leq e_S\}$ and $f|_{\overline{S}}(x)$ as $f(x \oplus e_S)$ for $\{x | x \geq e_S\}$.

Definition 3.6 (Subcube alternation)**.** For a Boolean function f, define the subcube alternation scalt of f as $\mathsf{scalt}(f) = \min_{B \subseteq [n]}(\mathsf{alt}(f|_B) + \mathsf{alt}(f|_{\overline{B}}))$.

More precisely (in Lemma 3.8), we show that $\forall f$, $\mathsf{salt}(f) \geq \mathsf{scalt}(f)$. In arguing the same, we use the following claim which gives an exact expression for maximum alternation of a shifted functions over all chains that contain the shift.

Lemma 3.7. *For $f : \{0,1\}^n \to \{-1,1\}$ and any $B \subseteq [n]$, and let $\mathcal{C}_B$ be the collection of maximal chains containing e_B. Then,*

$$\max_{\sigma \in \mathcal{C}_B} \mathsf{alt}(f(x \oplus e_B), \sigma) = \mathsf{alt}(f|_B) + \mathsf{alt}(f|_{\overline{B}}).$$

Proof. Let $g(x) = f(x \oplus e_B)$. Denote by $\overline{x}$ the bitwise complement of x. We claim that,

$$\forall x : x \leq e_B, \ g(x) = f|_B(\overline{x}) \tag{2}$$
$$\forall x : x \geq e_B, \ g(x) = f|_{\overline{B}}(x) \tag{3}$$

Fig. 1 illustrates the subcubes of interest in the original function and how they change for the function under shift. Now for any chain σ containing e_B in the Boolean hypercube, $\mathsf{alt}(g, \sigma) = \mathsf{alt}(f|_B) + \mathsf{alt}(f_{\overline{B}})$.

To see Eq. (3) observe that for any $x \geq e_B$, $x = y \oplus e_B$ with $y \leq e_{\overline{B}}$. Hence $g(x) = f(y) = f|_{\overline{B}}(x)$. For Eq. (2), since $x \leq e_B$, $g(x) = f(x \oplus e_B) = f|_B(\overline{x})$ as restricted to B, $x \oplus B$ complements x (with locations outside B set to 0).

Any maximal chain σ containing e_B must completely lie in the subcubes $\{x \mid x \leq e_B\}$ and $\{x \mid x \geq e_B\}$. Hence, $\max_{\sigma \in \mathcal{C}_B} \mathsf{alt}(f(x \oplus e_B), \sigma) \leq \mathsf{alt}(f|_B) + \mathsf{alt}(f|_{\overline{B}})$. Also, any maximal chain in the subcubes mentioned can be combined in the natural way to get a maximal chain for the whole subcube which contains e_B. Hence $\max_{\sigma \in \mathcal{C}_B} \mathsf{alt}(f(x \oplus e_B), \sigma) \geq \mathsf{alt}(f|_B) + \mathsf{alt}(f|_{\overline{B}})$.
$\qquad\square$

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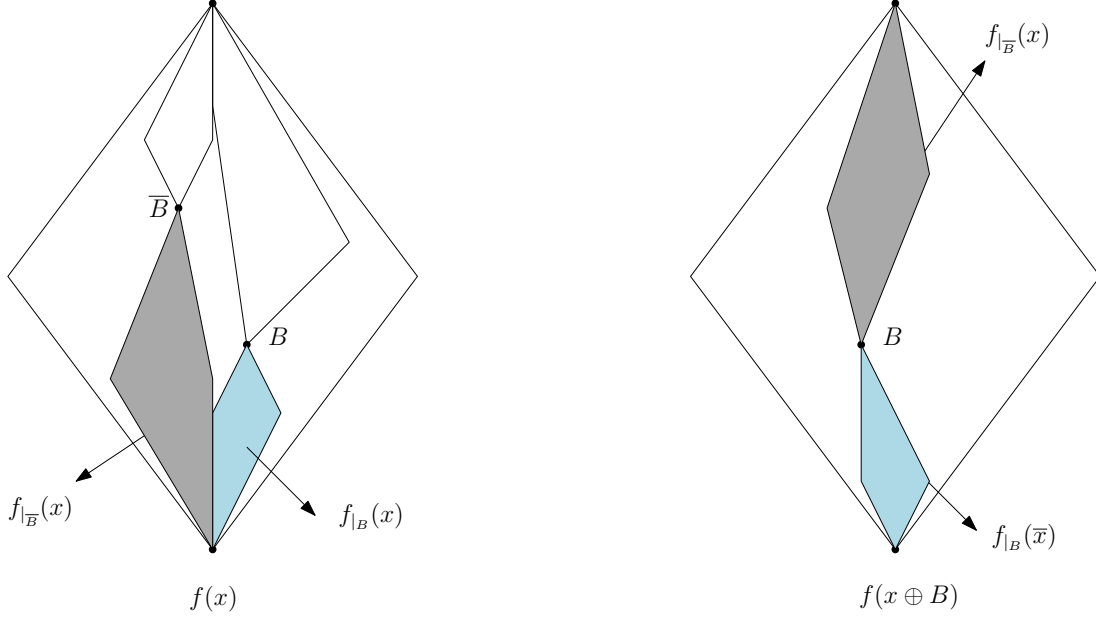
Figure 1: Boolean function f under shift

We can now conclude the lower bound on salt using Lemma 3.7.

Lemma 3.8. *For any $f : \{0,1\}^n \to \{-1,1\}$, $\mathsf{salt}(f) \geq \mathsf{scalt}(f)$.*

Proof. Let $S \subseteq [n]$ be a shift for which $\mathsf{alt}(f(x \oplus S))$ is minimum and $\mathcal{C}_S$ denotes the maximal chains containing e_S. Hence,

$$\mathsf{salt}(f) = \mathsf{alt}(f(x \oplus e_S)) \geq \max_{\sigma \in \mathcal{C}_S} \mathsf{alt}(f(x \oplus e_S)) \tag{4}$$

Combining with Lemma 3.7, we have $\mathsf{salt}(f) \geq \mathsf{alt}(f|_S) + \mathsf{alt}(f|_{\overline{S}})$ which is at least $\min_{B \subseteq [n]}(\mathsf{alt}(f|_B) + \mathsf{alt}(f|_{\overline{B}}))$ □

4 Affine Transforms : Lower Bounds on Quantum Communication Complexity

In this section, we study the affine transformation in its full generality applied to block sensitivity and sensitivity, and use it to prove Theorem 1.6 and Theorem 1.7 from the introduction. We achieve this using affine transforms as our tool (Section 4.1), by which we derive a new lower bound for $Q_{1/3}^*(F)$ in terms of $\mathsf{bs}(f, 0^n)$ (Section 4.2). Using this and a lower bound on $\mathsf{bs}(f, 0^n)$ (Proposition 4.4), we show that for any Boolean function f, and any prime p, $Q_{1/3}^*(F) \geq \Omega\left(\frac{\sqrt{\mathsf{DT}(f)}}{\deg_p(f)}\right)$. This immediately implies that if there is a p such that $\deg_p(f)$ is constant, then $\mathsf{D}(F) \leq 2\mathsf{DT}(f) \leq O(Q_{1/3}^*(F)^2)$ thereby answering Question 1.4 in positive for such functions. We relax this requirement and show that if there exists distinct primes p and q for which $\deg_p(f)$ and $\deg_q(f)$ are not very close, then $\mathsf{D}(F) \leq \mathsf{poly}(Q_{1/3}^*(F))$ (Theorem 1.7).

4.1 Upper Bound for Block Sensitivity via Affine Transforms

In this section, we describe our main tool. Given an $f : \{0,1\}^n \to \{-1,1\}$ and any $a \in \{0,1\}^n$, we exhibit an affine transform $A : \mathbb{F}_2^n \to \mathbb{F}_2^n$ such that for $g(x) = f(Ax)$, $\mathsf{bs}(f,a) \leq \mathsf{s}(g,0^n)$.

Before describing the affine transform, we note that a linear transform is already known to achieve a weaker bound of $\mathsf{bs}(f) \leq O(\mathsf{s}(g)^2)$ due to Sherstov [She10].

Proposition 4.1 (Lemma 3.3 of [She10]). *For any $f : \mathbb{F}_2^n \to \{-1,1\}$, there exists a linear transform $L : \mathbb{F}_2^n \to \mathbb{F}_2^n$ such that for $g(x) = f(Lx)$, $\mathsf{bs}(f) = O(\mathsf{s}(g)^2)$.*

See Observation A.3 in Section A for an explicit description of the linear transform achieving the bounds in the above proposition.

Now we describe an affine transform which improves the bound on $\mathsf{bs}(f)$ in the above proposition to linear in $\mathsf{s}(g)$. This affine transform has already been used in Nisan and Szegedy (see Lemma 7 of [NS94]) to show that $\mathsf{bs}(f) \leq 2\mathsf{deg}(f)^2$. Since the exact form of g is relevant in the subsequent arguments, we explicitly prove it here bringing out the structure of the affine transform that we require.

Lemma 1.3. For any $f : \mathbb{F}_2^n \to \{-1,1\}$ and $a \in \{0,1\}^n$, there exists an affine transform $A : \mathbb{F}_2^n \to \mathbb{F}_2^n$ such that for $g(x) = f(A(x))$,

(a) $\mathsf{bs}(f,a) \leq \mathsf{s}(g,0^n)$, and

(b) $g(x) = f((x_{i_1}, x_{i_2}, \ldots, x_{i_n}) \oplus a)$ where $i_1, \ldots, i_n \in [n]$ are not necessarily distinct.

Proof. Let $\mathsf{bs}(f,a) = k$ and $\{B_1, \ldots, B_k\}$ be the sensitive blocks on a. Since the blocks are disjoint, $\{B_i \mid i \in [k]\}$ viewed as vectors over $\mathbb{F}_2^n$ are linearly independent. Hence, there is a linear transform $L : \mathbb{F}_2^n \to \mathbb{F}_2^n$ such that $L(e_i) = B_i$ for $i \in [k]$.[6] Define $A(x) = L(x) \oplus a$. For $g(x) = f(A(x))$,

$$
\begin{aligned}
\mathsf{s}(g,0^n) &= |\{i \mid g(0^n) \neq g(0^n \oplus e_i), i \in [n]\}| \\
&= |\{i \mid f(a) \neq f(a \oplus L(e_i)), i \in [n]\}| = \mathsf{bs}(f,a)
\end{aligned}
$$

which completes the proof of main statement and Item a. Item b holds as the sensitive blocks are disjoint. $\square$

4.2 From Block Sensitivity Lower Bound at 0^n to Quantum Communication Lower Bounds

We now prove a lower bound for $Q_{1/3}^*(F)$ in terms of $\mathsf{bs}(f,0^n)$.

Theorem 1.5. Let $f : \{0,1\}^n \to \{-1,1\}$ and $F(x,y) = f(x \wedge y)$, then,

$$
Q_{1/3}^*(F) = \Omega\left(\sqrt{\mathsf{bs}(f,0^n)}\right).
$$

Proof. We first state a weaker version of this result which follows from Theorem 4.2 of Sherstov [She10]. The result, which is based on a powerful method of proving quantum communication lower bounds due to Razborov [Raz03] and Klauck [Kla07], says that for a Boolean function $g : \{0,1\}^n \to \{-1,1\}$ with $G(x,y) = g(x \wedge y)$, if there exists an $z \in \{0,1\}^n$ such that $z_i = 0$ for

---

[6]For completeness of definition of L, for $i \notin [k]$, we define $L(e_i) = 0^n$.

all $i \in [k]$ and $g(z \oplus e_1) = g(z \oplus e_2) = \ldots = g(z \oplus e_k) \neq g(z)$, then $Q_{1/3}^*(G) = \Omega(\sqrt{k})$. This immediately implies that for any $g : \{0,1\}^n \to \{-1,1\}$,

$$Q_{1/3}^*(G) = \Omega\left(\sqrt{\mathsf{s}(g, 0^n)}\right) \tag{5}$$

Given an f, we now describe a $g : \{0,1\}^n \to \{-1,1\}$ such that $Q_{1/3}^*(F) \geq Q_{1/3}^*(G)$ and $Q_{1/3}^*(G) = \Omega(\sqrt{\mathsf{bs}(f, 0^n)})$ as follows thereby completing the proof.

Applying Lemma 1.3 with $a = 0^n$ to f, we obtain $g(x) = f(x_{i_1}, x_{i_2}, \ldots, x_{i_n})$. We note that F and G can be viewed as a $2^n \times 2^n$ matrix with (x, y)th entry being $f(x \wedge y)$ and $g(x \wedge y)$ respectively. By construction of g, using the observation that the matrix G appears as a submatrix of F, $Q_{1/3}^*(F) \geq Q_{1/3}^*(G)$. This observation is used in Sherstov (for instance, see proof of Theorem 5.1 of [She10]) without giving details. For completeness, we give the details here. Let $S = \{i_1, \ldots i_n\} \subseteq [n]$ of size k. For $j \in S$, let $B_j = \{t \mid i_t = j\}$. Hence g depends only on these k input variables of S and all the variables with indices in B_j are assigned the variable x_j. This implies that

$$g(x) = f(\oplus_{j \in S} x_j e_{B_j}) \tag{6}$$

We now exhibit a submatrix of F containing G. Consider the submatrix of F with rows and columns restricted to

$$W = \left\{ a_1 e_{B_1} \oplus a_2 e_{B_2} \oplus \ldots a_k e_{B_k} \mid (a_1, a_2 \ldots, a_k) \in \{-1,1\}^k \right\}.$$

For $u, y \in W$,

$$
\begin{aligned}
F(u, y) &= f(u \wedge y) \\
&= f((u_1 e_{B_1} \oplus \ldots \oplus u_k e_{B_k}) \wedge (y_1 e_{B_1} \oplus \ldots \oplus y_k e_{B_k})) \\
&= f(u_1 \wedge y_1 e_{B_1} \oplus \ldots \oplus u_k \wedge y_k e_{B_k}) && [B_j\text{s are disjoint}] \\
&= g(u \wedge y) && [\text{By Eq. (6)}]
\end{aligned}
$$

Applying Eq. (5) to the g obtained, we have $Q_{1/3}^*(G) \geq \Omega(\sqrt{\mathsf{s}(g, 0^n)})$. Hence, by Item a of Lemma 1.3, as $a = 0^n$, we have $Q_{1/3}^*(G) \geq \Omega(\sqrt{\mathsf{bs}(f, 0^n)})$. $\qquad \square$

Remark 4.2. *Observe that for an arbitrary $a \in \{0,1\}^n$ for $g(x) = f(x \oplus a)$, the statement $Q_{1/3}^*(G) \leq Q_{1/3}^*(F)$ does not hold. Otherwise, we would have $Q_{1/3}^*(F) = \Omega(\sqrt{\mathsf{bs}(f)})$ for all f which is not true (see the discussion after Theorem 1.5 in the Introduction).*

4.3 Putting Them Together

We are now ready to prove Theorem 1.6 and Theorem 1.7. A critical component of our proof is the following stronger connection between $\mathsf{DT}(f)$ and $\mathsf{bs}(f, 0^n)$. Buhrman and de Wolf, in their survey [BdW02], showed the following with the proof attributed to Noam Nisan and Roman Smolensky.

Lemma 4.3 ([BdW02]). *For any Boolean function $f : \{0,1\}^n \to \{-1,1\}$, $\mathsf{DT}(f) \leq \mathsf{bs}(f) \cdot \deg(f)^2$*

The same proof can be adapted to show the following strengthening of their result.

Proposition 4.4. *For any $f : \{0,1\}^n \to \{-1,1\}$, and any prime p,*

$$\mathsf{DT}(f) \leq \mathsf{bs}(f, 0^n) \cdot \deg_p(f)^2.$$

Proof. We observe that the arguments of Buhrman and de Wolf (more specifically, Lemma 5, Lemma 6 and Theorem 12 of [BdW02]), can give a stronger upper bound than $\mathsf{bs}(f) \cdot \deg(f)^2$, namely $\mathsf{bs}(f, 0^n) \cdot \deg_p(f)^2$. This is important in our context since we are able to bound $Q^*_{1/3}(F)$ only by $\mathsf{bs}(f, 0^n)$.

Let $p_f(x) \in \mathbb{F}_p[x_1, \ldots, x_n]$ be an $\mathbb{F}_p$ polynomial representation of f. As p_f is a multilinear, we view monomials as subsets of variables. We define *size* of a monomial as the number of variables in it. Let S_f be the collection of all monomials of maximal size in p_f. We show that,

Claim 4.5. *For any Boolean function f, there is a set of variables of size at most $\mathsf{bs}(f, 0^n) \cdot \deg_p(f)$ which has a non-empty intersection with all the monomials in S_f.*

We call this set as a hitting set for S_f. We now assume this claim. Hence, querying these variables fixes them and results in a function whose $\mathbb{F}_p$-degree is at most $\deg_p(f) - 1$. We repeat this on the resulting function to obtain the desired decision tree where at most $\mathsf{bs}(f, 0^n) \cdot \deg_p(f)^2$ variables gets queried.

Proof of Claim 4.5 We now argue the existence of a hitting set, which has a non-empty intersection with all the monomials in S_f, of size at most $\mathsf{bs}(f, 0^n) \cdot \deg_p(f)$.

Firstly, observe that every monomial m in S_f must have a non-empty set B of indices of variables in m such that $f(0^n) \neq f(0^n \oplus e_B)$. To see this, restrict f to indices in the monomial m by setting all variables not in the monomial to 0. Let g be the resulting function. By construction, g is non-constant as the monomial m appears in the $\mathbb{F}_p$ representation of g. Hence there must be some setting of the input to g such that its evaluation differs from that of the all zero input.

We construct a hitting set H as follows: for each monomial m in S_f, if no variable in H appear in m, add all the variables in it to H. Since, each such monomial contains a sensitive block on the input 0^n, the number of monomials that gets added to H is at most $\mathsf{bs}(f, 0^n)$. Since each monomial is of size at most $\deg_p(f)$, total size of the hitting set is at most $\mathsf{bs}(f, 0^n) \cdot \deg_p(f)$. $\qquad\square$

We now give a proof of Theorem 1.6 and Theorem 1.7.

Theorem 1.6. Fix a prime p. Let $f : \{0,1\}^n \to \{-1,1\}$ where f depends on all the inputs. Let $F(x,y) = f(x \wedge y)$. For any $0 < \epsilon < 1$ such that $\deg_p(f) \leq (1 - \epsilon) \log n$, we have

$$Q^*_{1/3}(F) = \Omega\left(\frac{n^{\epsilon/2}}{\log n}\right).$$

Proof. Applying Theorem 1.5 and Proposition 4.4, we have

$$Q^*_{1/3}(F) \geq \Omega\left(\frac{\sqrt{\mathsf{DT}(f)}}{\deg_p(f)}\right) \tag{7}$$

As observed in Gopalan *et al.* [GLS09], by a modification to an argument in the proof of Nisan and Szegedy (Theorem 1 of [NS94]), it can be shown that $\deg(f) \geq \frac{n}{2^{\deg_p(f)}}$. Since, $\mathsf{DT}(f) \geq \deg(f)$, we

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have $\mathsf{DT}(f) \geq \frac{n}{2^{\deg_p(f)}}$. Hence Eq. (7) gives,

$$Q^*_{1/3}(F) = \Omega\left(\frac{\sqrt{n}}{\deg_p(f)2^{\deg_p(f)/2}}\right) = \Omega\left(\frac{n^{\epsilon/2}}{(1-\epsilon)\log n}\right)$$

where the last lower bound follows upon applying the bound on $\deg_p(f)$. $\qquad\square$

As a demonstrative example, we show a weaker lower bound on quantum communication complexity with prior entanglement for the generalized inner product function $\mathsf{GIP}_{n,k}(x,y) \overset{\text{def}}{=} \oplus_{i=1}^n \bigwedge_{j=1}^k (x_{ij} \wedge y_{ij})$ when $k = \frac{1}{2}\log n$. We remark that a lower bound of $\Omega(n)$ is known for the inner product function [CvDNT99].

Note that $\mathsf{GIP}_{n,k}$ can be expressed as $f \circ \wedge$, where $f(z) \overset{\text{def}}{=} \oplus_{i=1}^n \bigwedge_{j=1}^k z_{ij}$, with $\deg_2(f) = k$. Applying Theorem 1.6 with $\epsilon = 1/2$ and $p = 2$, we have $Q^*_{1/3}(\mathsf{GIP}_{n,\frac{1}{2}\log n}) = \Omega\left(\frac{n^{1/4}}{\log n}\right)$. Though this bound is arguably weak, Theorem 1.6 gives a non-trivial lower bound for a all those Boolean functions f with small $\deg_p(f)$ for some prime p.

Theorem 1.7. *Let* $f : \{0,1\}^n \to \{-1,1\}$ *with* $F(x,y) = f(x \wedge y)$. *Fix* $0 < \epsilon < 1$. *If there exists distinct primes* p, q *such that* $\deg_q(f) = \Omega(\deg_p(f)^{\frac{2}{1-\epsilon}})$, *then* $\mathsf{D}(F) = O(Q^*_{1/3}(F)^{2/\epsilon})$.

Proof. Applying, Theorem 1.5 and Proposition 4.4, for any prime t, $Q^*_{1/3}(F) \geq \Omega\left(\frac{\sqrt{\mathsf{DT}(f)}}{\deg_t(f)}\right)$. By hypothesis, $\deg_p(f) \leq O(\deg_q(f)^{\frac{1-\epsilon}{2}}) \leq O(\mathsf{DT}(f)^{\frac{1-\epsilon}{2}})$ implying that for $t = p$, $\mathsf{D}(F) \leq 2\mathsf{DT}(f) \leq O(Q^*_{1/3}(F)^{2/\epsilon})$. $\qquad\square$

Remark 4.6. *For any Boolean function* f, *if there exists a prime* p *with* $\deg_p(f) \leq c\log n$ *for some* $c < 1/2$, *then by main result of [GLS09] relating degree of Boolean functions under different field characteristics, for any prime* $q \neq p$, $\deg_q(f) = \Omega(\frac{n^{1-2c}}{c\log p\log n}) = \Omega((\log n)^2)$. *Hence any such* f *satisfies the condition that* $\deg_q(f) = \Omega(\deg_p(f)^{\frac{2}{1-\epsilon}})$ *for some constant* ϵ *and by Theorem 1.7,* $\mathsf{D}(F) = O(Q^*_{1/3}(F)^{2/\epsilon})$.

5 Linear Transforms : Sensitivity versus Sparsity

Continuing in the theme of affine transforms, in this section, we first establish an upper bound on alternation of a function in terms of sensitivity of the function after application of a suitable linear transform. Using this, we show the existence of a function whose sensitivity is asymptotically as large as square root of sparsity (see introduction for a motivation and discussion).

Lemma 1.8. *For any* $f : \{0,1\}^n \to \{-1,1\}$, *there exists an invertible linear transform* $L : \mathbb{F}_2^n \to \mathbb{F}_2^n$ *such that for* $g(x) = f(L(x))$, $\mathsf{alt}(f) \leq 2\mathsf{s}(g) + 1$.

Proof. Let $0^n \prec x_1 \prec x_2 \ldots \prec x_n = 1^n$ be a chain $\mathcal{C}$ of maximum alternation in the Boolean hypercube of f. Since chain $\mathcal{C}$ has maximum alternation, there must be at least $(\mathsf{alt}(f) - 1)/2$ many zeros and $(\mathsf{alt}(f) - 1)/2$ many ones when the x_is are evaluated on f. Note that the set of n distinct inputs $x_1, x_2, \ldots, x_n$ seen as vectors in $\mathbb{F}_2^n$ are linearly independent and hence is a basis of

$\mathbb{F}_2^n$. Hence there exists an invertible[7] linear transform $L : \mathbb{F}_2^n \to \mathbb{F}_2^n$ taking standard basis vectors to the these vectors, i,e. $L(e_i) = x_i$ for $i \in [n]$.

To prove the result, we now show that $\mathsf{s}(g, 0^n) \geq \frac{\mathsf{alt}(f)-1}{2}$. The neighbors of 0^n in the hypercube of g are $\{e_i \mid i \in [n]\}$ and each of them evaluates to $g(e_i) = f(L(e_i)) = f(x_i)$ for $i \in [n]$. Since there are at least $(\mathsf{alt}(f)-1)/2$ many zero and at least those many ones among x_is when evaluated by f, there must be at least $(\mathsf{alt}(f)-1)/2$ many neighbors of 0^n which differ in evaluation with $g(0^n)$ (independent of the value of $g(0^n)$). Hence $\mathsf{s}(g) \geq s(g, 0^n) \geq \frac{\mathsf{alt}(f)-1}{2}$ which completes the proof. $\qquad\square$

We now describe the family of functions and argue an exponential gap between sensitivity and logarithm of sparsity, as stated in the following Theorem.

Theorem 1.9. There exists a family of functions $\{g_k \mid k \in \mathbb{N}\}$ such that

$$\mathsf{s}(g_k) \geq \frac{\sqrt{\mathsf{sparsity}(g_k)}}{2} - 1.$$

Proof. For the family of functions $f_k \in \mathcal{F}$ (Definition 3.2), $\mathsf{alt}(f_k) \geq 2^{(\log \mathsf{sparsity}(f_k))/2} - 1$ [DS19].

We now use this family $\mathcal{F}$ to describe the family of functions g_k. For every $f_k \in \mathcal{F}$, let $g_k(x) = f_k(L(x))$ such that $\mathsf{alt}(f_k) \leq 2\mathsf{s}(g_k) + 1$ as guaranteed by Lemma 1.8. Since, we have $\mathsf{alt}(f_k) \geq 2^{(\log \mathsf{sparsity}(f_k))/2} - 1$, it must be that

$$\mathsf{s}(g_k) \geq \frac{1}{2}(\mathsf{alt}(f_k) - 1) \geq \frac{1}{2}(2^{(\log \mathsf{sparsity}(f_k))/2} - 2) \geq \frac{\sqrt{\mathsf{sparsity}(f_k)}}{2} - 1$$

As the parameter $\mathsf{sparsity}$ does not change under invertible linear transforms (Ex 3.1 [O'D14]), $\mathsf{s}(g_k) \geq 0.5\sqrt{\mathsf{sparsity}(f_k)} - 1 = 0.5\sqrt{\mathsf{sparsity}(g_k)} - 1$. $\qquad\square$

We now describe how the family of Boolean functions in Theorem 1.9 rule out a possibility of settling XOR Log-Rank conjecture, a conjecture in classical communication complexity, using a recent proof of Sensitivity Conjecture. First, we describe the XOR Log-Rank conjecture and then give a potential way to prove the XOR Log-Rank conjecture using the recent resolution of Sensitivity Conjecture [Hua19]. Following this, we argue how the family of Boolean functions in Theorem 1.9 rules out this possibility.

For an $f : \{0,1\}^n \to \{-1,1\}$, define $F_\oplus : \{0,1\}^n \times \{0,1\}^n \to \{-1,1\}$ as $F_\oplus(x,y) = f(x \oplus y)$. The XOR Log-Rank conjecture says that, for every f, the deterministic communication cost of computing the corresponding $F_\oplus$ must satisfy $\mathsf{D}(F_\oplus) = \mathsf{poly}(\log \mathsf{sparsity}(f))$. An equivalent formulation of the Sensitivity Conjecture due to Hatami *et al.* (Proposition 5.10, [HKP11]) says that for every f, $\mathsf{D}(F_\oplus) = \mathsf{poly}(\mathsf{s}(f))$. With the Sensitivity conjecture now proven [Hua19], one way to prove the XOR Log-Rank conjecture is to show that for all Boolean functions f, $\mathsf{s}(f) \leq \mathsf{poly}(\log \mathsf{sparsity}(f))$. Unfortunately, the existence of a family of Boolean functions in Theorem 1.9 rules out this possibility.

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[7] L is actually the change of basis transform from standard basis vectors to x_is and hence is bijective.

6 Conclusion and Future directions

In this paper, we study the Boolean function complexity measures, namely sensitivity, block sensitivity, and alternation under affine transforms. We showed design of special transforms which achieves structurally revealing statements about the resulting function. We used their properties to show lower bounds on the bounded error quantum communication complexity of Boolean function whose $\mathbb{F}_p$-degree is small. We showed that classical and quantum communication complexity are polynomially related for certain special class of functions. We also demonstrated Boolean functions where sensitivity of the function is as large as the square root of its sparsity.

The main open question is to see if the tools developed here can be pushed to remove the restriction on $\deg_p$ and $\deg_q$ of Boolean functions in Theorem 1.7 thereby proving the Quantum Classical equivalence (Question 1.4).

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References

[BC99]   Anna Bernasconi and Bruno Codenotti. Spectral analysis of boolean functions as a graph eigenvalue problem. *IEEE Trans. Computers*, 48(3):345–351, 1999.

[BdW01]   Harry Buhrman and Ronald de Wolf. Communication complexity lower bounds by polynomials. In *Proceedings of the 16th Annual IEEE Conference on Computational Complexity, Chicago, Illinois, USA, June 18-21, 2001*, pages 120–130, 2001.

[BdW02]   Harry Buhrman and Ronald de Wolf. Complexity measures and decision tree complexity: a survey. *Theor. Comput. Sci.*, 288(1):21–43, 2002.

[CDR86]   Stephen A. Cook, Cynthia Dwork, and Rüdiger Reischuk. Upper and lower time bounds for parallel random access machines without simultaneous writes. *SIAM J. Comput.*, 15(1):87–97, 1986.

[CvDNT99] Richard Cleve, Wim van Dam, Michael Nielsen, and Alain Tapp. Quantum entanglement and the communication complexity of the inner product function. In Colin P. Williams, editor, *Quantum Computing and Quantum Communications*, pages 61–74, Berlin, Heidelberg, 1999. Springer Berlin Heidelberg.

[DS19]   Krishnamoorthy Dinesh and Jayalal Sarma. Alternation, sparsity and sensitivity: Bounds and exponential gaps. *Theor. Comput. Sci.*, 771:71–82, 2019. A preliminary version appeared in CALDAM 2018.

[GLS09]   Parikshit Gopalan, Shachar Lovett, and Amir Shpilka. On the complexity of boolean functions in different characteristics. In *Proceedings of the 24th Annual IEEE Conference on Computational Complexity, CCC 2009, Paris, France, 15-18 July 2009*, pages 173–183, 2009.

[GOS⁺11]   Parikshit Gopalan, Ryan O'Donnell, Rocco A. Servedio, Amir Shpilka, and Karl Wimmer. Testing fourier dimensionality and sparsity. *SIAM J. Comput.*, 40(4):1075–1100, 2011. A preliminary version appeared in ICALP 2009.

[GSW16]   Parikshit Gopalan, Rocco A. Servedio, and Avi Wigderson. Degree and sensitivity: Tails of two distributions. In *31st Conference on Computational Complexity, CCC 2016, May 29 to June 1, 2016, Tokyo, Japan*, pages 13:1–13:23, 2016.

[HKP11]   Pooya Hatami, Raghav Kulkarni, and Denis Pankratov. *Variations on the Sensitivity Conjecture.* Number 4 in Graduate Surveys. Theory of Computing Library, 2011.

[Hua19]   Hao Huang. Induced subgraphs of hypercubes and a proof of the Sensitivity Conjecture. *Annals of Mathematics*, 190(3):949–955, 2019.

[Kla07]   Hartmut Klauck. Lower bounds for quantum communication complexity. *SIAM J. Comput.*, 37(1):20–46, 2007.

[KN06]   Eyal Kushilevitz and Noam Nisan. *Communication complexity.* Cambridge University Press, 2nd edition, 2006.

[LZ17]   Chengyu Lin and Shengyu Zhang. Sensitivity conjecture and log-rank conjecture for functions with small alternating numbers. In *44th International Colloquium on Automata, Languages, and Programming, ICALP 2017, July 10-14, 2017, Warsaw, Poland*, pages 51:1–51:13, 2017.

[MO09]   Ashley Montanaro and Tobias Osborne. On the communication complexity of XOR functions. *CoRR*, abs/0909.3392, 2009.

[Nis91]   Noam Nisan. CREW PRAMs and decision trees. *SIAM J. Comput.*, 20(6):999–1007, 1991.

[NS94]   Noam Nisan and Mario Szegedy. On the degree of boolean functions as real polynomials. *Comput. Complex.*, 4:301–313, 1994. A preliminary version appeared in STOC 1992.

[O'D14]   Ryan O'Donnell. *Analysis of Boolean Functions.* Cambridge University Press, 2014.

[Raz03]   A A Razborov. Quantum communication complexity of symmetric predicates. *Izvestiya: Mathematics*, 67(1):145, 2003.

[Rub95]   David Rubinstein. Sensitivity vs. block sensitivity of Boolean functions. *Combinatorica*, 15(2):297–299, 1995.

[She08]   Alexander A. Sherstov. The pattern matrix method for lower bounds on quantum communication. In *Proceedings of the 40th Annual ACM Symposium on Theory of Computing, Victoria, British Columbia, Canada, May 17-20, 2008*, pages 85–94, 2008.

[She09]   Alexander A. Sherstov. On quantum-classical equivalence for composed communication problems. *CoRR*, abs/0906.1399v1, 2009.

[She10]   Alexander A. Sherstov. On quantum-classical equivalence for composed communica-
             tion problems. *Quantum Information & Computation*, 10(5&6):435–455, 2010.

[SZ09]   Yaoyun Shi and Yufan Zhu. Quantum communication complexity of block-composed
             functions. *Quantum Information & Computation*, 9(5):444–460, 2009.

[Yao93]   Andrew Chi-Chih Yao. Quantum circuit complexity. In *34th Annual Symposium on
             Foundations of Computer Science, Palo Alto, California, USA, 3-5 November 1993*,
             pages 352–361, 1993.

[ZS10]   Zhiqiang Zhang and Yaoyun Shi. On the parity complexity measures of Boolean
             functions. *Theor. Comput. Sci.*, 411(26-28):2612–2618, 2010.

A   Quantum communication lower bound from block sensitivity

Sherstov in [She10] showed the following lower bound on quantum communication cost of an affine
shift of a Boolean function in terms of its block sensitivity.

Corollary A.1 (Corollary 4.5 of [She10])**.** *Let $f : \{0,1\}^n \to \{-1,1\}$ be given. Then for some
$z \in \{0,1\}^n$, the matrix $F' = [f_z(x \wedge y)]_{x,y} = [f(\ldots, (x_i \wedge y_i) \oplus z_i, \ldots)]_{x,y}$ obeys*

$$Q_{1/3}^*(F') = \Omega(\sqrt{\mathsf{bs}(f)})$$

In this section, we elaborate on why one
cannot set $z = 0^n$ for all Boolean functions and obtain Theorem 1.5. The above corollary
crucially uses two results. The first one is Lemma 3.3 of [She10] which shows that there exists
a Boolean function $g : \{0,1\}^n \to \{-1,1\}$ such that $\mathsf{bs}(f) \leq O(\mathsf{s}(g)^2)$ which is similar in spirit
to Lemma 1.3. The second one is Theorem 4.2 of [She10] which shows a lower bound for $Q_{1/3}^*(G)$
in terms of sensitivity of g (where $G(x,y) = g(x \wedge y)$). We reproduce the respective statements of
both below.

Lemma A.2 (Lemma 3.3 of [She10])**.** *Let $f : \{0,1\}^n \to \{-1,1\}$. Then there exists a $g : \{0,1\}^n \to
\{-1,1\}$ such that $s(g) = \Omega(\sqrt{\mathsf{bs}(f)})$ and $g(x) = f(x_{i_1}, \ldots, x_{i_n})$ for some $i_1, \ldots, i_n \in [n]$*

The function g is defined as follows.
Let z be the input on which $\mathsf{bs}(f, z)$ is maximum and $f(z) = 0$. Let $S_1, \ldots, S_k \subseteq [n]$ be the
sensitive blocks on z. Define $A_i = \{j \in S_i \mid z_j = 0\}$ and $B_i = \{j \in S_i \mid z_j = 1\}$. Let I be the indices
$i \in [k]$ such that both A_i and B_i are both non-empty.
Then

$$g(x) = f\left(\bigoplus_{i \in I} x_{minA_i} e_{A_i} \oplus \bigoplus_{i \in I} x_{minB_i} e_{B_i} \oplus \bigoplus_{i \in [k] \setminus I} x_{minS_i} e_{S_i} \oplus \bigoplus_{i \notin S_1 \cup \ldots \cup S_k} x_i e_i\right)$$

Observation A.3. *We observe that the above result of Sherstov (Lemma 3.3 of [She10]) can be
seen as applying a suitable linear transform to the Boolean function f to bound the block sensitivity
of f which is similar in spirit to Lemma 1.3.*

More precisely, the g obtained in Lemma 3.3 of [She10] can be described as $f(L(x))$ where L is defined as, for $j \in [n]$,

$$L(e_j) = \begin{cases} e_j \ \textit{if } j \notin S_1 \cup \ldots \cup S_k \\ e_{A_i} \ \textit{if } \exists i \in [k], \ \textit{such that } j = \min\{A_i\} \\ e_{B_i} \ \textit{if } \exists i \in [k], \ \textit{such that } j = \min\{B_i\} \\ 0^n \ \textit{otherwise} \end{cases}$$

By definition g as above, Sherstov showed that $\mathsf{s}(g, z) = \Omega(\sqrt{\mathsf{bs}(f)})$.

Theorem A.4 (Theorem 4.2 of [She10])**.** *For a Boolean function $g : \{0,1\}^n \to \{-1,1\}$ with $G(x,y) = g(x \wedge y)$, if there exists an $w \in \{0,1\}^n$ such that $w_i = 0$ for $i \in [k]$ and $g(w \oplus e_1) = g(w \oplus e_2) = \ldots = g(w \oplus e_k) \neq g(w)$, then $Q^*_{1/3}(G) = \Omega(\sqrt{k})$.*

To use the above result, one way is to start with a function g for which sensitivity is large at 0^n. To achieve, consider the shifted function f_z where z is the same input on which block sensitivity is maximized as before. This is because, by the choice of z, f_z will have maximum block sensitivity at 0^n which upon applying Lemma 3.3 of [She10] ensures that the function g obtained has a large k (i.e. sensitivity) at 0^n. This is exactly what is achieved in the proof of Corollary 4.5 of [She10].

Hence the choice is z is tied up with the block sensitivity of function f.