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 \(f : \mathbb{F}_2^n \to \{0, 1\}\), 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 \(salt(f)\)). By a result of Lin and Zhang [LZ17], it follows that \(bs(f) \leq O(salt(f)^2 s(f))\). Thus, to settle the SENSITIVITY CONJECTURE \((\forall f, bs(f) \leq poly(s(f)))\), it suffices to argue that \(\forall f, salt(f) \leq poly(s(f))\). However, we exhibit an explicit family of Boolean functions for which \(salt(f) = 2^{\Omega(n/f)}\).

Going further, we use an affine transform \(A\), such that the corresponding function \(g\) satisfies \(bs(f, 0^n) \leq s(g)\), to prove that for \(F(x, y) \eqdef f(x \& y)\), the bounded error quantum communication complexity of \(F\) with prior entanglement, \(Q^{1/3}_{1/3}(F)\) is \(\Omega(\sqrt{bs(f, 0^n)})\). Our proof builds on ideas from Sherstov [She10] where we use specific properties of the above affine transformation. Using this, we show the following.

(a) For a fixed prime \(p\) and an \(\epsilon, 0 < \epsilon < 1\), any Boolean function \(f\) that depends on all its inputs with \(deg_{p}(f) \leq (1 - \epsilon) \log n\) must satisfy \(Q^{1/3}_{1/3}(F) = \Omega\left(\frac{n^{1/2}}{\log n}\right)\). Here, \(deg_{p}(f)\) 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 \(deg_{p}(f) \geq \Omega(deg_{p}(f)^{\delta})\) for \(\delta > 2\), the deterministic communication complexity - \(D(F)\) and \(Q^{1/3}_{1/3}(F)\) are polynomially related. In particular, this holds when \(deg_{p}(f) = O(1)\). Thus, for this class of functions, this answers an open question (see [BdW01]) 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, \(alt(f) \leq 2s(g) + 1\). Using this new relation, we exhibit Boolean functions \(f\) (other than the parity function) such that \(s(f) = \Omega(\sqrt{sparsity(f)})\) where \(sparsity(f)\) is the number of non-zero coefficients in the Fourier representation of \(f\).

1 Introduction

For a Boolean function \(f : \{0, 1\}^n \to \{0, 1\}\), 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 \(i^{th}\) bit
as 1. The sensitivity of \( f \) (denoted by \( s(f) \)) is the maximum sensitivity of \( f \) over all inputs. A related parameter is the block sensitivity of \( f \) (denoted by \( bs(f) \)), 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 \( DT(f) \)) which is the depth of an optimal decision tree computing the function \( f \). The certificate complexity of \( f \) (denoted by \( C(f) \)) is the non-deterministic variant of the decision tree complexity. The parameter \( s(f) \) was originally studied by Cook et al. [CDR86] in connection with the CREW-PRAM model of computation. Subsequently, Nisan and Szegedy [NS92] (see also [Nis91]) introduced the parameters \( bs(f) \) and \( C(f) \) and conjectured that for any function \( f : \{0,1\}^n \to \{0,1\} \), \( \text{bs}(f) \leq \text{poly}(s(f)) \) - 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 [HLP11] for a survey and several equivalent formulations of the conjecture). The best known upper bound for \( \text{bs}(f) \) in terms of \( s(f) \) is \( \text{bs}(f) \leq \frac{s(f)(1 + o(1))s(f)}{2^{s(f) - 1}} \) due to He et al. [HLS17] improving a result of Ambainis et al. [And16].

Shi and Zhang [ZS10] studied the parity complexity variants of \( \text{bs}(f), C(f) \) and \( DT(f) \) 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 \( f : \mathbb{F}_2^n \to \{0,1\} \), under specific affine transformations over \( \mathbb{F}_2^n \). More precisely, we explore the relationship of the above parameters for the function \( g : \mathbb{F}_2^n \to \{0,1\} \) and \( f \), where \( g \) is defined as \( g(x) = f(Mx + b) \) for specific \( M \in \mathbb{F}_2^{n \times n} \) and \( b \in \mathbb{F}_2^n \). 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 \( f_b(x) \equiv f(x + b) \) 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 \( \text{alt}(f_b) = \Omega(n) \) while \( \text{alt}(f) = 1 \).

A recent result related to Sensitivity Conjecture by Lin and Zhang [LZ17] shows that \( \text{bs}(f) \leq O(s(f)\text{alt}(f)^2) \). This bound for \( \text{bs}(f) \), implies that to settle the Sensitivity Conjecture, it suffices to show that \( \text{alt}(f) \) is upper bounded by \( \text{poly}(s(f)) \) for all Boolean functions \( f \). However, the authors [DS18] ruled this out, by exhibiting a family of functions where \( \text{alt}(f) \) is at least \( 2^{\Omega(s(f))} \).

Observing that the parameters \( s(f), \text{bs}(f) \) are invariant under shifts, we define a new quantity shift-invariant alternation, \( \text{salt}(f) \), which is the minimum alternation of any function \( g \) obtained from \( f \) by shifting by a vector \( b \in \{0,1\}^n \) (see Definition 3.1). By the aforementioned bound on \( \text{bs}(f) \) of Lin and Zhang [LZ17], it is easy to observe that \( \text{bs}(f) \leq O(s(f)\text{alt}(f)^2) \). We also show that there exists a family of Boolean functions \( f \) with \( \text{bs}(f) = \Omega(s(f)\text{salt}(f)) \) (See Proposition 3.5).

It is conceivable that \( \text{salt}(f) \) is much smaller compared to \( \text{alt}(f) \) for a Boolean function \( f \) and hence that \( \text{salt}(f) \) can potentially be upper bounded by \( \text{poly}(s(f)) \) thereby settling the Sensitivity Conjecture. However, we rule this out by showing the following stronger gap, about the same
Proposition 1.1. There exists an explicit family of Boolean functions for which salt$(f)$ is $2^{\Omega(s(f))}$.

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 $(g)$.

Lemma 1.2. For any $f : \mathbb{F}_2^n \rightarrow \{-1, 1\}$ and $a \in \{0, 1\}^n$, there exists an affine transform $A : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n$ such that for $g(x) = f(A(x))$,

(a) $bs(f, a) \leq s(g, 0^n)$, 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 [NS92]) to show that $bs(f) \leq 2\deg(f)^2$. Here, $\deg(f)$ is the degree of the multilinear polynomial over reals that agrees with $f$ on Boolean inputs. We show another application of Lemma 1.2 in the context of quantum communication complexity, a model for which was introduced by Yao [Yao03]. In this model, two parties Alice and Bob have to compute a function $F : \{0, 1\}^n \times \{0, 1\}^n \rightarrow \{0, 1\}$, 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 $R_{1/3}(F)$ as the minimum cost randomized protocol computing $F$ with error at most 1/3. It can be shown that $Q_{1/3}(F) \leq R_{1/3}(F) \leq D(F)$.

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 as the Log-Equivalence Conjecture). In this work, we are interested in the case when $F(x, y)$ is of the form $f(x \land y)$ where $f : \{0, 1\}^n \rightarrow \{0, 1\}$ and $x \land y$ is the string obtained by bitwise AND of $x$ and $y$.

Question 1.3. For $f : \{0, 1\}^n \rightarrow \{0, 1\}$, let $F : \{0, 1\}^n \times \{0, 1\}^n \rightarrow \{0, 1\}$ be defined as $F(x, y) = f(x \land y)$. Is it true that for any such $F$, $D(F) \leq poly(Q_{1/3}(F))$?
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 $D(F) = O(Q_{1/3}^{*}(F)\sqrt{2})$ for all Boolean functions $f$ where $Q_{1/3}^{*}(F) = \max \{Q_{1/3}^{*}(F_1), Q_{1/3}^{*}(F_2)\}$ and $D(F) = \max \{D(F_1), D(F_2)\}$. Using Lemma 1.2, 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 1.4.** Let $f : \{0, 1\}^n \rightarrow \{-1, 1\}$ and $F(x, y) = f(x \wedge y)$, then,

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

In this context, we make an important comparison 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 bs$(f, x)$ is maximum, $Q_{1/3}^{*}(F') = \Omega(\sqrt{\text{bs}(f)}) \geq \Omega(\sqrt{\text{bs}(f, 0^n)})$ (Corollary 4.5 of [She10]). Notice that $F$ and $F'$ differ by a linear shift of $f$ with $b$. More precisely, $Q_{1/3}^{*}(F)$ can change drastically even under such (special) linear shifts of $f$. For example, consider $f = \wedge_n$. Since 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 $Q_{1/3}^{*}(F) = \Omega(\sqrt{\text{bs}(f)})$ cannot hold for all $f$ (see Remark 4.2). Since the lower bounds shown on quantum communication complexity are on different functions, Theorem 1.4 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 $\text{deg}_p(f)$), the quantum communication complexity of $F$ is large.

**Theorem 1.5.** Fix a prime $p$. Let $f : \{0, 1\}^n \rightarrow \{-1, 1\}$ where $f$ depends on all the variables. Let $F(x, y) = f(x \wedge y)$. For any $0 < \epsilon < 1$ such that $\text{deg}_p(f) \leq (1 - \epsilon) \log n$, we have

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

Observe that, though Theorem 1.4 does not answer Question 1.3 in positive for all functions, we could show a class of Boolean function for which $D(F)$ 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 $\text{deg}_p(f)$ and $\text{deg}_q(f)$ are sufficiently far apart (Theorem 1.6).

**Theorem 1.6.** Let $f : \{0, 1\}^n \rightarrow \{-1, 1\}$ with $F(x, y) = f(x \wedge y)$. Fix $0 < \epsilon < 1$. If there exists distinct primes $p, q$ such that $\text{deg}_q(f) = \Omega(\text{deg}_p(f)^{2/\epsilon})$, then $D(F) = O(Q_{1/3}^{*}(F)^{2/\epsilon})$.

By the result of Gopalan et al. (Theorem 1.2, [GLS09]), any Boolean function $f$ with $\text{deg}_p(f) = o(\log n)$ must have $\text{deg}_q(f) = \Omega(n^{1-o(1)})$ thereby satisfying the condition of Theorem 1.6. Hence for all such functions, Theorem 1.6 answers Question 1.3 in positive. Observe that the same can also be derived from Theorem 1.5.

**Alternation under Linear Transforms:** We now restrict our study to linear transforms. Again, the aim is to design special linear transforms which transforms the parameters of interest for us. In

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1 More importantly, this $b$ in Corollary 4.5 of [She10] cannot be fixed to $0^n$ for all Boolean functions to conclude Theorem 1.4. See Appendix A.1 for details.
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 1.7.** For any \( f : \mathbb{F}_2^n \to \{0,1\} \), there exists an invertible linear transform \( L : \mathbb{F}_2^n \to \mathbb{F}_2^n \) such that for \( g(x) = f(L(x)) \),

\[
\text{alt}(f) \leq 2s(g) + 1
\]

We show an application of the above result in the context of sensitivity. Nisan and Szegedy [NS92] showed that for any Boolean function \( f \), \( s(f) \leq 2\deg(f)^2 \). However, the situation is quite different for \( \deg_2(f) \) - noticing that for \( f \) being parity on \( n \) variables, \( \deg_2(f) = 1 \) and \( s(f) = n \) - the gap can even be unbounded. Though parity may appear as a corner case, there are other functions like the Boolean inner product function\(^2\) \( \text{IP}_n \) whose \( \mathbb{F}_2 \)-degree is constant while sensitivity is \( \Omega(n) \) thereby ruling out the possibility that \( s(f) \leq \deg_2(f)^2 \). It is known that if \( f \) is not the parity on \( n \) variables (or its negation), \( \deg_2(f) \leq \log \text{sparsity}(f) \) [BC99, GOS+09]. Hence, as a structural question about the two parameters, we ask : for \( f \) other than the parity function, is it true that \( s(f) \leq \text{poly}(\log \text{sparsity}(f)) \). Observe that \( \text{IP}_n \) has high sparsity and hence does not rule this out. We use Lemma 1.7, which is in the theme of studying alternation and sensitivity in the context of linear transformations, to improve this gap and show that there is a family of functions where this gap is exponential.

**Theorem 1.8.** There exists a family of functions \( \{g_k \mid k \in \mathbb{N}\} \) such that

\[
s(g_k) \geq \frac{\sqrt{\text{sparsity}(g_k)}}{2} - 1
\]

## 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 \( i \)th bit of \( x \).

We now define the Boolean function parameters we use. Let \( f : \{0,1\}^n \to \{0,1\} \) and \( a \in \{0,1\}^n \), we define, 1) the sensitivity of \( f \) on \( a \) as \( s(f,a) = |\{i \mid f(a \oplus e_i) \neq f(a), i \in [n]\}| \), 2) the block sensitivity of \( f \) on \( a \), \( bs(f,a) \) 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 \( a \), \( C(f,a) \) 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 \( \phi \in \{s,bs,C\} \), 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 \( \phi \in \{s,bs,C\} \) where \( f_b(x) \overset{\text{def}}{=} f(x \oplus b) \). Also, it can be shown that \( s(f) \leq bs(f) \leq C(f) \).

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

\(^2\) \( |\text{IP}_n(x_1, x_2, \ldots, x_n, y_1, y_2, \ldots, y_n) = \sum_i x_i y_i \mod 2 \)
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 $\text{deg}(f)$ (resp. $\text{deg}_p(f)$) to be the degree of the multilinear polynomial computing $f$ over reals (resp. $\mathbb{F}_p$). We define $\text{DT}(f)$ as the depth of an optimal decision tree computing $f$. It is known that for all Boolean functions $f$, $\text{deg}_p(f) \leq \text{deg}(f) \leq \text{DT}(f) \leq \text{bs}(f)^3$.

Sparsity of a Boolean function $f : \{0,1\}^n \rightarrow \{-1, 1\}$ (denoted by $\text{sparsity}(f)$) 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 $\text{DT}(f)$ 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 $f : \{0,1\}^n \times \{0,1\}^n \rightarrow \{0,1\}$, 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 $D(f)$ as the minimum cost of a deterministic protocol computing $f$. For functions of the form $F(x, y) = f(x \land y)$, it is known that $D(F) \leq 2\text{DT}(f)$ [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 1.1) thereby ruling out the possibility that $\text{salt}(f)$ can be upper bounded by a polynomial in $s(f)$ for all Boolean functions $f$.

Recall from the introduction that the parameters $s, \text{bs}$ and $C$ are shift invariant while $\text{alt}$ is not. We define a variant of alternation which is invariant under shifts.

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

A family of functions with $\text{salt}(f) = \Omega(2^{s(f)})$: We now exhibit a family of functions $\mathcal{F}$ where for all $f \in \mathcal{F}$, $\text{salt}(f) \geq 2^{s(f)}$ thereby ruling out the possibility that $\text{salt}(f)$ can be upper bounded by a polynomial in $s(f)$. The family $\mathcal{F}$ is the same class of Boolean functions for which alternation is at least exponential in sensitivity due to [DS18].

**Definition 3.2** (Definition 1 from [DS18]. See also Proof of Lemma A.1 of [GSW16]). Consider the family defined as follows.

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

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 of 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\(^3\) \(\text{Maj}_n\), the tree sensitivity can be shown to be \(\Omega(n)\) while alternation is 1.

The authors [DS18] have shown that for any \(f \in \mathcal{F}\), there exists 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 \(\text{salt}(f)\) in terms of \(s(f)\) for all \(f \in \mathcal{F}\).

**Proposition 1.1.** For \(f_k \in \mathcal{F}\), \(\text{salt}(f_k) \geq 2^{\Omega(s(f_k))}\).

**Proof.** We show\(^4\) that for \(f_k \in \mathcal{F}\) and \(n = 2^k - 1\), for all \(c \in \{0, 1\}^n\), \(\text{alt}(f_k(x \oplus c)) \geq 2^{k-2}\). Since \(s(f_k) \leq k\) 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 \(c \in \{0, 1\}^3\), \(\text{alt}(f(x \oplus c)) \geq 1\). 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 \(c \in \{0, 1\}^n\), \(\text{alt}(h_1(z' \oplus c))\) and \(\text{alt}(h_2(z'' \oplus c))\) is at least \(2^{k-2}\). For \(m = 2^{k+1} - 1\), consider any \(c = (c', b, c'') \in \{0, 1\}^m\) where \(c', c'' \in \{0, 1\}^n\) and \(b \in \{0, 1\}\). Since \(h_1\) and \(h_2\) are variable disjoint, \(\text{alt}(f(x \oplus c)) \geq \text{alt}(h_1(z' \oplus c')) + \text{alt}(h_2(z'' \oplus c'')) \geq 2^{k-2} + 2^{k-2} = 2^{k-1}\) completing the induction. \(\square\)

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

\[
\text{bs}(f) = O(\text{alt}(f)^2 s(f)) \tag{1}
\]

This immediately gives the following proposition.

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

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

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

For functions \(f_1, \ldots, f_k\) where each \(f_i : \{0, 1\}^n \rightarrow \{0, 1\}\), define the function \(\text{OR}_k \circ \mathcal{T} : \{0, 1\}^{nk} \rightarrow \{0, 1\}\) as \(\forall_{i=1}^k f_i(x^{(i)})\) 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 \(f_i : \{0, 1\}^n \rightarrow \{0, 1\}\) satisfy, \(f_i(0^n) = f_i(1^n) = 0\). Then,

\[
\text{alt}(\text{OR}_k \circ \mathcal{T}) = \sum_{i=1}^k \text{alt}(f_i)
\]

**Proof.** Let \(f = \text{OR}_k \circ \mathcal{T}\) and \(\mathcal{C}\) be a chain in \(\{0, 1\}^{nk}\) for which \(\text{alt}(f, \mathcal{C})\) 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

\[^3\text{Maj}_n(x) = 1 \iff \sum x_i \geq \lfloor n/2 \rfloor\]

\[^4\text{In this proof, for simplicity, we abuse the notation } f_k(x \oplus c) \text{ to denote the function obtained by shifting } f_k \text{ by } c.\]

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at least one of the $f_i$’s have changed their evaluation along the chain $C$. Since the functions are variable disjoint, such a change must be witnessed in the chain $C_i$ for some $i$. Hence

$$\text{alt}(f) = \text{alt}(f, C) \leq \sum_{i=1}^{k} \text{alt}(f_i, C_i) \leq \sum_{i=1}^{k} \text{alt}(f_i)$$

To show that $\text{alt}(f) \geq \sum_{i=1}^{k} \text{alt}(f_i)$, we exhibit a chain $C$ in $\{0,1\}^{nk}$ of alternation $\sum_{i=1}^{k} \text{alt}(f_i)$. Let $C_i = (0^n = z^{(i0)} < z^{(i1)} < z^{(i2)} < \ldots < z^{(im)} = 1^n)$ be a chain in $\{0,1\}^n$ for which $f_i$ achieves maximum alternation. We construct a chain $C$ by “gluing” together these $k$ chains. More precisely, let $C$ be a 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

$$0^n < \ldots < 0^n < z^{(i0)} < z^{(i1)} < z^{(i2)} < \ldots < z^{(im)} < 1^n < \ldots < 1^n$$

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

$$\text{alt}(f, C) \geq \sum_{i=1}^{k} \text{alt}(f_i, C_i) = \sum_{i=1}^{k} \text{alt}(f_i)$$

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

**Proof.** We consider the Rubinstein’s function $f_R : \{0,1\}^{n^2} \rightarrow \{0,1\}$ [Rub95] where the input is treated as $n \times n$ matrix which evaluates to 1 ifff 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 $OR_n \circ h$ with $h : \{0,1\}^n \rightarrow \{0,1\}$ where $h(a) = 1$ ifff 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 $\text{alt}(h) = 2$. Since $h(0^n) = h(1^n) = 0$, applying Lemma 3.4 with $f_i = h$ for all $i \in [n]$, we get that $\text{alt}(f_R) = \text{alt}(h) \cdot n = 2n$. It is known that $\text{bs}(f_R) \geq \frac{n^2}{2}$ while $s(f) \leq n$ [Rub95], thereby showing that $\text{bs}(f_R) \geq \frac{s(f_R) \cdot \text{alt}(f_R)}{4}$.

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

### 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 1.5 and Theorem 1.6 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 $\text{bs}(f, 0^n)$ (Section 4.2). Using this and a lower bound on $\text{bs}(f, 0^n)$ (Proposition 4.3), we show that for any Boolean function $f$, and any prime
\[ p, \ Q^*_{1/3}(F) \geq \Omega \left( \sqrt[3]{DT(f)} \right). \] This immediately implies that if there is a \( p \) such that \( \deg_p(f) \) is constant, then \( D(F) \leq 2DT(f) \leq O(Q^*_{1/3}(F))^2 \) thereby answering Question 1.3 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 \( D(F) \leq \poly(Q^*_{1/3}(F)) \) (Theorem 1.6).

### 4.1 Upper Bound for Block Sensitivity via Affine Transforms

In this section, we describe our main tool. Given an \( f : \{0,1\}^n \to \{0,1\} \) 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 \( g(x) = f(Ax) \), \( bs(f, a) \leq s(g, 0^n) \).

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

**Proposition 4.1** (Lemma 3.3 of [She10]). For any \( f : \mathbb{F}_2^n \to \{0,1\} \), there exists a linear transform \( L : \mathbb{F}_2^n \to \mathbb{F}_2^n \) such that for \( g(x) = f(Lx) \), \( bs(f) = O(s(g)^2) \).

See Observation A.3 in Appendix A.1 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 \( bs(f) \) in the above proposition to linear in \( s(g) \). This affine transform has already been used in Nisan and Szegedy (see Lemma 7 of [NS92]) to show that \( bs(f) \leq 2\deg(f)^2 \). 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 1.2.** 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 \( g(x) = f(Ax) \),

(a) \( bs(f, a) \leq s(g, 0^n) \), 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 \( bs(f, a) = k \) 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 \( i \in [k] \).\[^5\] Define \( A(x) = L(x) \oplus a \). For \( g(x) = f(A(x)) \),

\[
bs(g, 0^n) = | \{ i \mid g(0^n) \neq g(0^n \oplus e_i), i \in [n] \} | = | \{ i \mid f(a) \neq f(a \oplus L(e_i)), i \in [n] \} | = bs(f, a)
\]

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 \( bs(f, 0^n) \).

**Theorem 1.4.** Let \( f : \{0,1\}^n \to \{-1,1\} \) and \( F(x, y) = f(x \land y) \), then,

\[
Q^*_{1/3}(F) = \Omega \left( \sqrt{bs(f, 0^n)} \right)
\]

\[^5\]For completeness of definition of \( L \), for \( i \not\in [k] \), we define \( L(e_i) = 0^n \).
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 $g : \{0, 1\}^n \rightarrow \{-1, 1\}$ with $G(x, y) = g(x \land y)$, if there exists an $z \in \{0, 1\}^n$ such that $z_i = 0$ for $i \in [k]$ and $g(z \oplus e_1) = g(z \oplus e_2) = \ldots = g(z \oplus e_k) = g(z)$, then $Q_{1/3}^*(G) = \Omega(\sqrt{k})$. This immediately implies that for any $g : \{0, 1\}^n \rightarrow \{-1, 1\},$

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

(2)

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

Applying Lemma 1.2 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 $f(x \land y)$ and $g(x \land y)$ 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

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

(3)

We now exhibit a submatrix of $F$ containing $G$. Consider the submatrix of $F$ with rows and columns restricted to $W = \{a_1 e_{B_1} \oplus a_2 e_{B_2} \oplus \ldots a_k e_{B_k} \mid (a_1, a_2, \ldots, a_k) \in \{0, 1\}^k\}$. For $u, y \in W$,

$$F(u, y) = f(u \land y) = f((u_1 e_{B_1} \oplus \ldots \oplus u_k e_{B_k}) \land (y_1 e_{B_1} \oplus \ldots \oplus y_k e_{B_k}))$$

$$= f(u_1 \land y_1 e_{B_1} \oplus \ldots \oplus u_k \land y_k e_{B_k})$$

$[B_j$s are disjoint$]$  

$[\text{By Eq. (3)}]$

Applying Eq. (2) to the $g$ obtained, we have $Q_{1/3}^*(G) \geq \Omega(\sqrt{s(g, 0^n)})$. Hence, by Item a of Lemma 1.2, as $a = 0^n$, we have $Q_{1/3}^*(G) \geq \Omega(\sqrt{s(g, 0^n)})$. 

$\Box$

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 $Q_{1/3}^*(F) = \Omega(\sqrt{\text{bs}(f)})$ for all $f$ which is not true (see the discussion after Theorem 1.4 in the Introduction).

4.3 Putting Them Together

We are now ready to prove Theorem 1.5 and Theorem 1.6. A critical component of our proof is the following stronger connection between DT$(f)$ and bs$(f, 0^n)$. Buhrman and de Wolf, in their survey [BdW02], showed that DT$(f) \leq \text{bs}(f) \cdot \deg(f)^2$ where the proof is attributed to Noam Nisan and Roman Smolensky. The same proof can be adapted (see Appendix A.2 for details) to show the following strengthening of their result.

Proposition 4.3. For any $f : \{0, 1\}^n \rightarrow \{0, 1\}$, and any prime $p$, $\text{DT}(f) \leq \text{bs}(f, 0^n) \cdot \deg_p(f)^2$
We now give a proof of Theorem 1.5 and Theorem 1.6.

**Theorem 1.5.** Fix a prime $p$. Let $f : \{0, 1\}^n \to \{-1, 1\}$ where $f$ depends on all the inputs. Let $F(x, y) = f(x \land y)$. 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 1.4 and Proposition 4.3, we have $Q_{1/3}^*(F) \geq \Omega\left(\sqrt{\frac{\deg_{p}(f)}{\deg_p(f)}}\right)$. As observed in Gopalan et al. [GLS09], by a modification to an argument in the proof of Nisan and Szegedy (Theorem 1 of [NS92]), it can be shown that $\deg(f) \geq \frac{n}{2^{\deg_p(f)}}$. Since $\DT(f) \geq \deg(f)$, we have $\DT(f) \geq \frac{n}{2^{\deg_p(f)}}$. Hence,

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

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

As a demonstrative example, we show a weaker lower bound on quantum communication complexity with prior entanglement for the generalized inner product function $\GIP_{n,k}(x, y) \triangleq \oplus_{i=1}^n \land_{j=1}^k (x_{ij} \land y_{ij})$ when $k = \frac{1}{2} \log n$. We remark that a lower bound of $\Omega(n)$ is known for the inner product function [CvDNT13].

Note that $\GIP_{n,k}$ can be expressed as $f \circ \land$, where $f(z) \triangleq \oplus_{i=1}^n \land_{j=1}^k z_{ij}$, with $\deg_2(f) = k$. Applying Theorem 1.5 with $\epsilon = 1/2$ and $p = 2$, we have $\Q_{1/3}^*(\GIP_{n,\frac{1}{2} \log n}) = \Omega\left(n^{1/4} \log n\right)$. Though this bound is arguably weak, Theorem 1.5 gives a non-trivial lower bound for a all those Boolean functions $f$ with small $\deg_p(f)$ for some prime $p$.

**Theorem 1.6.** Let $f : \{0, 1\}^n \to \{-1, 1\}$ with $F(x, y) = f(x \land y)$. Fix $0 < \epsilon < 1$. If there exists distinct primes $p, q$ such that $\deg_q(f) = \Omega(\deg_p(f)^{2/\epsilon})$, then $D(F) = O(\Q_{1/3}^*(F)^{2/\epsilon})$.

**Proof.** Applying, Theorem 1.4 and Proposition 4.3, we get that for any prime $t$, $\Q_{1/3}^*(F) \geq \Omega\left(\sqrt{\frac{\deg_{t}(f)}{\deg_t(f)}}\right)$. By hypothesis, $\deg_p(f) \leq O(\deg_q(f)^{1+\epsilon}) \leq O(\DT(f)^{1+\epsilon})$ implying that for $t = p$, $D(F) \leq 2\DT(f) \leq O(\Q_{1/3}^*(F)^{2/\epsilon})$. \qed

## 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 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 1.7.** For any $f : \{0, 1\}^n \to \{0, 1\}$, there exists an invertible linear transform $L : F_2^n \to F_2^n$ such that for $g(x) = f(L(x))$, $\alt(f) \leq 2s(g) + 1$.
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 $(\text{alt}(f) - 1)/2$ many zeros and $(\text{alt}(f) - 1)/2$ many ones when the $x_i$s 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 form a basis of $\mathbb{F}_2^n$. Hence there exists an invertible\(^6\) linear transform $L : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n$ 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 $s(g, 0^n) \geq \frac{\text{alt}(f) - 1}{2}$. 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 $(\text{alt}(f) - 1)/2$ many zeros and at least these ones among $x_i$s when evaluated by $f$, there must be at least $(\text{alt}(f) - 1)/2$ many neighbors of $0^n$ which differ in evaluation with $g(0^n)$ (independent of the value of $g(0^n)$). Hence $s(g) \geq s(g, 0^n) \geq \frac{\text{alt}(f) - 1}{2}$ which completes the proof.

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

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

**Proof.** We remark that for the family of functions $f_k \in \mathcal{F}$ (Definition 3.2), $\text{alt}(f_k) \geq 2^{(\log \text{sparsity}(f_k))/2} - 1$ [DS18].

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 $\text{alt}(f_k) \leq 2s(g_k) + 1$ as guaranteed by Lemma 1.7. Since, we have $\text{alt}(f_k) \geq 2^{(\log \text{sparsity}(f_k))/2} - 1$, it must be that

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

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

### 6 Discussion and Open Problems

In this paper, we study the Boolean function parameters, 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.

There are several questions that are opened up in this line of work. Given an $f$, Observation A.3 says that Sherstov’s result exhibits a linear transform $L$ such that for $g(x) = f(Lx)$, $\text{bs}(f) = O(s(g)^2)$. We obtain an affine transform (in Lemma 1.2) where the corresponding $g$ satisfy, $\text{bs}(f) = O(s(g))$. In this context, a prominent direction is to use the structure of our linear transformation

\(^6\) $L$ is actually the change of basis transform from standard basis vectors to $x_i$s and hence is bijective.
in Lemma 1.2 to establish better upper bounds for $s(g)$ in terms of $s(f)$. Upper bounds better than $2^{s(f)}$ would improve the best known upper bound of $bs(f)$ in terms of $s(f)$. It is even more interesting to restrict to the case of $bs_3(f)$ and upper bound the corresponding $s(g)$ by $s(f)^{3-\epsilon}$ for some $\epsilon > 0$. By a result of Tal [Tal16], this suffices to improve the best known upper bound for block sensitivity in terms of sensitivity.

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A Appendix

A.1 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 $f : \{0,1\}^n \rightarrow \{-1,1\}$ 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

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

In this section, we elaborate on why one cannot set $z = 0^n$ for all Boolean functions and obtain Theorem 1.4. The above corollary crucially uses two results. The first one is Lemma 3.3 of [She10] which shows that there exists a Boolean function $g : \{0,1\}^n \rightarrow \{0,1\}$ such that $\text{bs}(f) \leq O(s(g)^2)$ which is similar in spirit to Lemma 1.2. 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 $f : \{0,1\}^n \rightarrow \{-1,1\}$. Then there exists a $g : \{0,1\}^n \rightarrow \{-1,1\}$ such that $s(g) = \Omega(\sqrt{bs(f)})$ 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 $\text{bs}(f,z)$ 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

$$g(x) = f \left( \bigoplus_{i \in I} x_{\min A_i} e_{A_i} \oplus \bigoplus_{i \in I} x_{\min B_i} e_{B_i} \oplus \bigoplus_{i \in [k]\setminus I} x_{\min S_i} e_{S_i} \oplus \bigoplus_{i \not\in S_1 \cup \ldots \cup S_k} x_i e_i \right)$$

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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 1.2.

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

$$L(e_j) = \begin{cases} e_j & \text{if } j \not\in S_1 \cup \ldots \cup S_k \\ e_{A_i} & \text{if } \exists i \in [k], \text{ such that } j = \min\{A_i\} \\ e_{B_i} & \text{if } \exists i \in [k], \text{ such that } j = \min\{B_i\} \\ 0^n & \text{otherwise} \end{cases}$$

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

Theorem A.4 (Theorem 4.2 of [She10]). For a Boolean function $g : \{0,1\}^n \to \{-1,1\}$ with $G(x, y) = g(x \land y)$, 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 $f$.

A.2 Proof of Proposition 4.3

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

Let $p_f(x) \in \mathbb{F}_p[x_1, \ldots, x_n]$ be an $\mathbb{F}_p$ polynomial representation of $f$. Let $S_f$ be the collection of all monomials of maximal size in $p_f$. We show that for any Boolean function $f$, there are most $bs(f, 0^n) \cdot \deg(f)$ many variables which has a non-empty intersection with all the monomials in $S_f$. Hence querying these variables 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 $bs(f, 0^n) \cdot \deg_p(f)^2$ variables gets queried. 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 $bs(f, 0^n) \cdot \deg(f)$.

Firstly, observe that every monomial $m$ in $S_f$ must have a non-empty set of indices $B$ 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 $bs(f, 0^n)$. Since each monomial is of size at most $\deg_p(f)$, total size of the hitting set is at most $bs(f, 0^n) \cdot \deg_p(f)$. 

\qed