# New Constructions with Quadratic Separation between Sensitivity and Block Sensitivity * 

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#### Abstract

Nisan and Szegedy [15] conjectured that block sensitivity is at most polynomial in sensitivity for any Boolean function. There is a huge gap between the best known upper bound on block sensitivity in terms of sensitivity - which is exponential, and the best known separating examples - which give only a quadratic separation between block sensitivity and sensitivity.

In this paper we give various new constructions of families of Boolean functions that exhibit quadratic separation between sensitivity and block sensitivity. Some of our constructions match the current largest separation between sensitivity and block sensitivity by Ambainis and Sun [5]. Our constructions have several novel aspects. We use more general function compositions instead of just OR-composition, and give constructions based on algebraic operations. In addition, we give the first direct constructions of families of Boolean functions that have both 0 -block sensitivity and 1-block sensitivity quadratically larger than sensitivity.


## 1 Introduction

The Sensitivity Conjecture posed by Nisan and Szegedy [15] is one of the most intriguing, yet elusive problems in computational complexity theory.

[^0]The sensitivity $s(f)$ of a Boolean function $f$ is the maximum over all inputs $x$ of the number of coordinate positions $i$ such that changing the value of the $i$-th bit of $x$ changes the value of the function. The block sensitivity $b s(f)$ of a Boolean function $f$ is the maximum over all inputs $x$ of the number of disjoint blocks of bits such that changing the value of all bits of $x$ in any given block changes the value of the function. (See Section 2 for more formal definitions.) Sensitivity was introduced by Cook, Dwork and Reischuk [8] as a measure to prove lower bounds on the parallel complexity of Boolean functions in the CREW PRAM model. Nisan [14] defined the more general block sensitivity measure, and showed that the CREW PRAM complexity of any Boolean function $f$ is characterized by its block sensitivity up to constant factors as $\Theta(\log b s(f))$. Nisan also showed that several other complexity measures, including certificate complexity and decision tree depth are polynomially related to block sensitivity. Nisan and Szegedy [15] showed that the degree of real polynomials representing a Boolean function $f$ is also polynomially related to its block sensitivity. These relations extend to approximate representation by real polynomials and to randomized and quantum decision tree depth. Thus, a number of important complexity measures are polynomially related to block sensitivity. See $[6,11]$ for a survey.

However, it remains open to fully understand the relationship between sensitivity and block sensitivity. Of course for any Boolean function $f, s(f) \leq$ $b s(f)$. Nisan and Szegedy [15] conjectured that block sensitivity is at most polynomial in sensitivity for any Boolean function $f$. They even raised the possibility that $b s(f)=O\left(s(f)^{2}\right)$. This possibility is still not ruled out - the best separation so far remains quadratic. The current best upper bound on block sensitivity in terms of sensitivity by Ambainis et al. [2, 4] is exponential: $b s(f) \leq s(f) 2^{s(f)-1}$. (More precisely, $b s(f) \leq \max \left\{2^{s(f)-1}\left(s(f)-\frac{1}{3}\right), s(f)\right\}$ [4].) This improves the earlier upper bounds of Kenyon and Kutin and Simon $[13,17]$.

The first example of a function with quadratic separation between its sensitivity and block sensitivity was given by Rubinstein [16] who constructed a function $f$ with $b s(f)=\frac{1}{2} s(f)^{2}$. Other constructions with quadratic separation were given in $[20,7,10,5]$. The largest separation so far is achieved by the construction of Ambainis and Sun [5] who gave a function $f$ with $b s(f)=\frac{2}{3} s(f)^{2}-\frac{1}{3} s(f)$.

Improving the constant $\frac{2}{3}$ in the separation would be interesting, since a function $f$ with $b s(f)>c s(f)^{2}$ for a constant $c>1$ would imply a construc-
tion with superquadratic separation by iterated composition of the function $f$ [3].

In order to better understand the relationship between sensitivity and block sensitivity, the one-sided versions of the measures 0 -sensitivity $s_{0}(f)$, 1 -sensitivity $s_{1}(f), 0$-block sensitivity $b s_{0}(f)$ and 1-block sensitivity $b s_{1}(f)$ have also been extensively studied. These measures are obtained by restricting attention to inputs $x \in f^{-1}(0)$ for defining 0 -sensitivity and 0 block sensitivity and to inputs $x \in f^{-1}(1)$ for defining 1 -sensitivity and 1-block sensitivity, respectively. (See Section 2 for formal definitions.) Then $s(f)=\max \left\{s_{0}(f), s_{1}(f)\right\}$ and $b s(f)=\max \left\{b s_{0}(f), b s_{1}(f)\right\}$.

Ambainis and Prusis [3] (improving the constant in a result of Kenyon and Kutin [13]) proved that $b s_{0}(f) \leq \frac{2}{3} s_{0}(f) C_{1}(f)$ where $C_{1}(f)$ denotes the 1-certificate complexity of $f$. See Section 2 for the definition of certificate complexity. On the other hand, Nisan [14] proved that $C_{1}(f) \leq b s_{1}(f) s_{0}(f)$. The analogous statements also hold for upper bounding $b s_{1}$ and $C_{0}$, respectively. Combining these results implies that in order to obtain much stronger separation between sensitivity and block sensitivity it is necessary to construct functions $f$ such that both $b s_{0}(f)$ and $b s_{1}(f)$ are significantly larger than $s(f)$.

Avishay Tal [19] pointed out to us, that one can get such examples by the following trick. Let $g$ be any function with $b s(g)=\Omega\left(s(g)^{2}\right)$, then taking $f(x, y)=g(x) \vee \neg g(y)$ will give $\min \left\{b s_{0}(f), b s_{1}(f)\right\}=\Omega\left(s(f)^{2}\right)$. Notice however that in this example the function $f$ cannot give a significantly larger separation between its block sensitivity and sensitivity than what was already achieved by the function $g$. Thus, limitations on the separation that follow from properties of the function $g$ will be inherited by the function $f$. By direct constructions, the largest simultaneous separation has been $\min \left\{b s_{0}(f), b s_{1}(f)\right\}=\Omega\left(s(f)^{\log _{2} 3}\right)$ in [1]. On the other hand, all previous direct constructions with quadratic separation between $b s(f)$ and $s(f)$ had $\min \left\{b s_{0}(f), b s_{1}(f)\right\}=O(s(f))$.

### 1.1 Our Results

In this paper we give various new constructions of families of Boolean functions that exhibit quadratic separation between sensitivity and block sensitivity. Our constructions have several novel aspects.

All previous constructions - with the exception of Chakraborty's functions [7] - were of the form $f=O R_{m} \circ g_{k}$ that is $f:\{0,1\}^{m k} \rightarrow\{0,1\}$ was obtained
by composing the $m$-bit OR function with an appropriately chosen inner function $g$ on $k$ bits. Chakraborty [7] did not use function composition at all. As for the choice of the inner function, Gopalan, Servedio, Tal and Wigderson [10] defined the inner function $g$ based on codewords of a Hamming code. All other constructions (including Chakraborty [7]) used the presence of certain patterns in the input $x$ to set the function value $g(x)$ to 1 .

We observe that other function compositions instead of OR-composition can also yield quadratic separations. We define new functions, that could be used as inner or outer functions, based on algebraic criterions related to multiplication in finite fields or polynomial multiplication. We give two versions of our "building block" functions based on finite field multiplication, denoted $g_{F F}$ and $g_{F F}^{*}$, and two functions $g_{p o l y}$ and $g_{p o l y}^{*}$ based on polynomial multiplication. Using OR-composition, $g_{F F}$ yields constant $\frac{1}{4}, g_{F F}^{*}$ and $g_{p o l y}$ give constant $\frac{1}{2}$ in quadratic separations, while $g_{\text {poly }}^{*}$ gives constant $\frac{2}{3}$, matching the current largest separation between sensitivity and block sensitivity by Ambainis and Sun [5].

We present a general framework based on certificates that captures most previous constructions, and also highlights their limitations. We also give a general condition for achieving quadratic separations with the $\frac{2}{3}$ constant for functions defined by families of certificates. The function by Ambainis and Sun [5] fits into this framework.

In addition, we provide the first direct constructions of families of Boolean functions $f$ with $\min \left\{b s_{0}(f), b s_{1}(f)\right\}=\Omega\left(s(f)^{2}\right)$. Our simultaneous quadratic separation of both 0-block sensitivity and 1-block sensitivity from sensitivity is based on a more refined study of the effects of function composition on these measures. We also present sufficient conditions for achieving such simultaneous separations and give several examples of functions satisfying these conditions.

## 2 Preliminaries

Let $f:\{0,1\}^{n} \rightarrow\{0,1\}$ be a Boolean function. For $x \in\{0,1\}^{n}$ and $i \in[n]$ we denote by $x^{i}$ the input obtained by flipping the $i$-th bit of $x$. More generally, for $S \subseteq[n]$ we denote by $x^{S}$ the input obtained by flipping the bits of $x$ in all coordinates in the subset $S$.

Definition 1. Sensitivity The sensitivity $s(f, x)$ of a Boolean function $f$ on input $x$ is the number of coordinates $i \in[n]$ such that $f(x) \neq f\left(x^{i}\right)$.

The 0-sensitivity and 1-sensitivity of $f$ are defined as $s_{0}(f)=\max \{s(f, x)$ : $f(x)=0\}$ and $s_{1}(f)=\max \{s(f, x): f(x)=1\}$, respectively. The sensitivity of $f$ is defined as $s(f)=\max \left\{s(f, x): x \in\{0,1\}^{n}\right\}=\max \left\{s_{0}(f), s_{1}(f)\right\}$.

Definition 2. Block Sensitivity The block sensitivity bs $(f, x)$ of a Boolean function $f$ on input $x$ is the maximum number of pairwise disjoint subsets $S_{1}, \ldots, S_{k}$ of $[n]$ such that for each $i \in[k] f(x) \neq f\left(x^{S_{i}}\right)$. The 0 -block sensitivity and 1-block sensitivity of $f$ are defined as $b s_{0}(f)=\max \{b s(f, x)$ : $f(x)=0\}$ and $b s_{1}(f)=\max \{b s(f, x): f(x)=1\}$, respectively. The block sensitivity of $f$ is defined as $b s(f)=\max \left\{b s(f, x): x \in\{0,1\}^{n}\right\}=$ $\max \left\{b s_{0}(f), b s_{1}(f)\right\}$.

It is convenient to refer to coordinates $i \in[n]$ such that $f(x) \neq f\left(x^{i}\right)$ as sensitive bits for $f$ on $x$. Similarly, a subset $S \subseteq[n]$ is called a sensitive block for $f$ on $x$ if $f(x) \neq f\left(x^{S}\right)$.

Definition 3. Partial assignment Given an integer $n>0$, a partial assignment $\alpha$ is a function $\alpha:[n] \rightarrow\{0,1, \star\}$. A partial assignment $\alpha$ corresponds naturally to a setting of $n$ variables $\left(x_{1}, x_{2}, \ldots x_{n}\right)$ to $\{0,1, \star\}$ where $x_{i}$ is set to $\alpha(i)$. The variables set to $\star$ are called unassigned or free, and we say that the variables set to 0 or 1 are fixed.
We say that $x \in\{0,1\}^{n}$ agrees with $\alpha$ if $x_{i}=\alpha(i)$ for all $i$ such that $\alpha(i) \neq \star$. The size of a partial assignment $\alpha$ is defined as the number of fixed variables of $\alpha$.

Definition 4. Certificate For a function $f:\{0,1\}^{n} \rightarrow\{0,1\}$ and input $x \in\{0,1\}^{n}$ a partial assignment $\alpha$ is a certificate of $f$ on $x$ if $x$ agrees with $\alpha$ and any input $y$ agreeing with $\alpha$ satisfies $f(y)=f(x)$.
The size of a certificate $\alpha$ is defined as the size of the partial assignment $\alpha$.
Definition 5. Certificate Complexity The certificate complexity $C(f, x)$ of a Boolean function $f$ on input $x$ is the size of the smallest certificate of $f$ on $x$. The 0-certificate complexity and 1-certificate complexity of $f$ are defined as $C_{0}(f)=\max \{C(f, x): f(x)=0\}$ and $C_{1}(f)=\max \{C(f, x):$ $f(x)=1\}$, respectively. The certificate complexity of $f$ is defined as $C(f)=$ $\max \left\{C(f, x): x \in\{0,1\}^{n}\right\}=\max \left\{C_{0}(f), C_{1}(f)\right\}$.

The above measures and their relationship to other complexity measures have been extensively studied. See $[6,11]$ for a survey. Here we only mention a result by Nisan [14] relating these measures to each other.

Theorem 1. [14] For any Boolean function $f$ we have

$$
s(f) \leq b s(f) \leq C(f)
$$

Moreover, for $z \in\{0,1\}$ we have

$$
s_{z}(f) \leq b s_{z}(f) \leq C_{z}(f)
$$

## Definition 6. Function defined by a set of partial assignments Let

 $C=\left\{\gamma_{1}, \ldots \gamma_{t}\right\}$ be a set of partial assignments where $\gamma_{i}:[n] \rightarrow\{0,1, \star\}$.Then $C$ naturally defines a function $g_{C}:\{0,1\}^{n} \rightarrow\{0,1\}$ as $g_{C}(x)=1$ if and only if $x$ agrees with some partial assignment $\gamma_{i} \in C$.

Definition 7. Distances The distance between two inputs $x, y \in\{0,1\}^{n}$ is defined as the number of bits in which they differ.
The distance between an input $x \in\{0,1\}^{n}$ and a partial assignment $\alpha:[n] \rightarrow$ $\{0,1, \star\}$ is defined as the minimum distance between $x$ and any input $y$ agreeing with $\alpha$.
The distance between two partial assignments $\alpha, \beta:[n] \rightarrow\{0,1, \star\}$ is defined as the minimum distance between any input $x$ agreeing with $\alpha$ and any input $y$ agreeing with $\beta$.

Definition 8. Function Composition For Boolean functions $f:\{0,1\}^{m} \rightarrow$ $\{0,1\}$ and $g:\{0,1\}^{k} \rightarrow\{0,1\}$ the function $f \circ g:\{0,1\}^{m k} \rightarrow\{0,1\}$ is defined on $z \in\{0,1\}^{m k}$ as

$$
f \circ g(z)=f\left(g\left(z_{1}, \ldots z_{k}\right), g\left(z_{k+1}, \ldots, z_{2 k}\right), \ldots, g\left(z_{(m-1) k+1}, \ldots, z_{m k}\right)\right)
$$

Properties of function composition were formally studied with respect to sensitivity and block sensitivity (as well as other related measures) by Tal and Gilmer et al. [18, 9]. We note the following two properties, relevant for us.

Lemma 1. [18, 9] For any Boolean functions $f$ and $g$ we have $s(f \circ g) \leq$ $s(f) s(g)$.

Definition 9. [18] For $z \in\{0,1\}$ we say that $f:\{0,1\}^{n} \rightarrow\{0,1\}$ is in $z$-good form, if
(1) $f\left(z^{n}\right)=z$ and (2) $b s(f)=b s\left(f, z^{n}\right)$

Lemma 2. [18] If both $f$ and $g$ are in 0-good form, or if both $f$ and $g$ are in 1-good form, then $b s(f \circ g) \geq b s(f) b s(g)$.

### 2.1 Previous Constructions with Quadratic Separation

All previous constructions that achieve quadratic separation between sensitivity and block sensitivity - with the exception of Chakraborty's functions [7] - were based on the following "OR-composition Lemma" first used by Rubinstein [16].

Lemma 3. [16] For any function $g:\{0,1\}^{m} \rightarrow\{0,1\}$, we have

- $s_{0}\left(O R_{n} \circ g\right)=n \cdot s_{0}(g)$
- $s_{1}\left(O R_{n} \circ g\right)=s_{1}(g)$
- $b s_{0}\left(O R_{n} \circ g\right)=n \cdot b s_{0}(g)$
- $b s_{1}\left(O R_{n} \circ g\right)=b s_{1}(g)$

The quadratic separations of $[16,20,5,10]$ are based on using this lemma and considering functions of the form $f=O R_{n} \circ g$ for appropriately chosen inner functions $g$.

Next we briefly describe the previous constructions of functions with quadratic separation.

1. Rubinstein's function [16] Define $g:\{0,1\}^{2 m} \rightarrow\{0,1\}$ as $g(x)=1$ iff $x_{2 j-1}=x_{2 j}=1$ for some $j \in[m]$ and $x_{i}=0$ for $i \neq 2 j-1,2 j$.
This gives $s_{0}(g)=1, s_{1}(g)=b s_{1}(g)=2 m, b s_{0}(g)=m$.
Let $f=O R_{2 m} \circ g$. Then $s_{0}(f)=s_{1}(f)=b s_{1}(f)=2 m, b s_{0}(f)=2 m^{2}$, giving $b s(f)=\frac{1}{2} s(f)^{2}$.
2. Virza's function [20] Define $g:\{0,1\}^{2 m+1} \rightarrow\{0,1\}$ as $g(x)=1$ iff one of the following holds:
1) $\exists j \in[m]$ such that $\left(x_{2 j-1}=x_{2 j}=1\right)$ and ( $\left.x_{i}=0 \quad \forall i \neq 2 j-1,2 j\right)$.
2) $\left(x_{2 m+1}=1\right)$ and $\left(x_{i}=0 \forall i \neq 2 m+1\right)$.

This gives $s_{0}(g)=1, s_{1}(g)=b s_{1}(g)=2 m+1, b s_{0}(g)=m+1$.
Let $f=O R_{2 m+1} \circ g$. Then $s_{0}(f)=s_{1}(f)=b s_{1}(f)=2 m+1, b s_{0}(f)=$ $(m+1)(2 m+1)$. Therefore, $b s(f)=\frac{1}{2} s(f)^{2}+\frac{1}{2} s(f)$.
3. Ambainis and Sun's function [5] Define $g:\{0,1\}^{2(2 m+1)} \rightarrow\{0,1\}$ as $g(x)=1$ iff $\exists j \in[2 m+1]$ such that

1) $x_{2 j-1}=x_{2 j}=1$, and
2) For all $i \in[m], x_{2 j+2 i}=x_{2 j-2 i}=x_{2 j-2 i-1}=0$.

Here, the index of $x$ is taken modulo $(2(2 m+1))$ i.e. we index $x$ as if it were laid around a circle.

This gives $s_{0}(g)=1, s_{1}(g)=b s_{1}(g)=3 m+2, b s_{0}(g)=2 m+1$.
Let $f=O R_{3 m+2} \circ g$. Then $s_{0}(f)=s_{1}(f)=b s_{1}(f)=3 m+2, b s_{0}(f)=$ $(3 m+2)(2 m+1)$. Therefore, $b s(f)=\frac{2}{3} s(f)^{2}-\frac{1}{3} s(f)$.
4. Function based on Hamming code by Gopalan, Servedio, Tal and Wigderson (Section 7 in [10])
Consider the Hamming code on $m=2^{r}-1$ bits.
Define $g:\{0,1\}^{m} \rightarrow\{0,1\}$ as
$g(x)=1$ iff $x$ is a codeword of the Hamming code on $m$ bits.
This gives $s_{0}(g)=1, s_{1}(g)=b s_{1}(g)=m, b s_{0}(g)=\frac{m+1}{2}$.
Let $f=O R_{m} \circ g$. Then, $s_{0}(f)=s_{1}(f)=b s_{1}(f)=m, b s_{0}(f)=\frac{m(m+1)}{2}$. Thus $b s(f)=\frac{1}{2} s(f)^{2}+\frac{1}{2} s(f)$.

Finally, we describe a construction by Chakraborty that does not involve function composition. He also constructed another function in [7], which is similar to the one we describe.
5. Chakraborty's function [7] For integers $k, m$ such that $2<k<m$ and $2 k \mid m$, the function $g_{k}:\{0,1\}^{m} \rightarrow\{0,1\}$ is defined as follows.
For $x=\left(x_{0}, \ldots x_{m-1}\right), g_{k}(x)=1$ iff $\exists i \in\{0, \ldots m-1\}$ such that $x_{i}=x_{i+1(\bmod m)}=1$ and $x_{j}=0$ for all $j \in\{i+2(\bmod m), \ldots, i+k-1($ $\bmod m)\}$.
Then, $s_{0}\left(g_{k}\right)=\frac{2 m}{k}, s_{1}\left(g_{k}\right)=k, b s_{0}\left(g_{k}\right)=\frac{m}{2}$ and $b s_{1}\left(g_{k}\right)=k$.
Therefore, setting $k=\sqrt{2 m}$ gives $s\left(g_{\sqrt{2 m}}\right)=\sqrt{2 m}$ and $b s\left(g_{\sqrt{2 m}}\right)=\frac{m}{2}$. So we have $b s\left(g_{\sqrt{2 m}}\right)=\frac{1}{4} s\left(g_{\sqrt{2 m}}\right)^{2}$.

## 3 New Building Blocks for Quadratic Separation

Here we define several new functions that we will use as inner or outer functions in various function compositions to obtain quadratic separations.

### 3.1 A General Framework Based on Certificates

In this subsection we observe that several previous constructions fit into a common framework, that also explains the limitations of these constructions. Previously, Karthik and Tavenas [12] studied limitations of separations in a certificate framework under some special conditions.

We start with two lemmas that are also helpful for analyzing our new constructions presented in later subsections.

Lemma 4. Let $C$ be a set of partial assignments with $|C| \geq 2$ and consider the function $g_{C}$ defined by $C$.

1. If the distance between any two partial assignments $\gamma_{i}, \gamma_{j} \in C$ for $i \neq j$ is at least 2 then $s_{1}\left(g_{C}\right)=b s_{1}\left(g_{C}\right)=C_{1}\left(g_{C}\right)$.
2. If the distance between any two partial assignments $\gamma_{i}, \gamma_{j} \in C$ for $i \neq j$ is at least 3 then $s_{0}\left(g_{C}\right)=1$.

Proof. We first note that since the distance between any two partial assignments is at least 2 , the set of partial assignments $C$ also forms a set of 1-certificates for $g_{C}$, such that every 1-input agrees with exactly one partial assignment from $C$.

To see that $s_{1}\left(g_{C}\right)=b s_{1}\left(g_{C}\right)=C_{1}\left(g_{C}\right)$, for any 1-input $x$ consider the unique $\gamma_{i} \in C$ agreeing with $x$. The bits fixed by $\gamma_{i}$ form exactly the set of sensitive bits for $f$ on $x$, since any two partial assignments in $C$ are at distance at least 2 from each other. Note also that $C_{1}(f, x)$ is at most the number of bits fixed by $\gamma_{i}$. Thus, the statement follows by Theorem 1.

When the pairwise distance between the partial assignments in $C$ is at least 3 , then $s_{0}\left(g_{C}\right)=1$ since for any 0 -input $x$, there is at most one certificate $\gamma_{i} \in C$ such that $x$ is at distance 1 from it.

Lemma 5. Let $C$ be a set of partial assignments such that the function $g_{C}$ defined by $C$ is not constant. Then $b s_{0}\left(g_{C}\right) \leq|C|$.

Proof. Note that for a given function $g$ there may be several different sets $C$ such that $g=g_{C}$. The statement of the lemma holds for any such set of partial assignments $C$.

The proof is based on the following claim.

Claim 1. Let $f:\{0,1\}^{n} \rightarrow\{0,1\}$ be any non-constant Boolean function. For $x \in f^{-1}(0)$ let $A$ and $B$ be disjoint subsets of $[n]$ such that $f\left(x^{A}\right)=$ $f\left(x^{B}\right)=1$. Then there is no partial assignment $\alpha:[n] \rightarrow\{0,1, *\}$ that is a certificate of $f$ on both $x^{A}$ and $x^{B}$.
Proof of Claim: Suppose that some partial assignment $\alpha:[n] \rightarrow\{0,1, *\}$ is a certificate of $f$ on both $x^{A}$ and $x^{B}$. This means that both $x^{A}$ and $x^{B}$ agree with $\alpha$ in the bits fixed by $\alpha$. On the other hand, since $f(x)=0, x$ must differ from $\alpha$ in at least one bit fixed by $\alpha$. Since $x^{A}$ agrees with $\alpha$, the set $A$ must contain all the coordinates where $x$ and $\alpha$ disagree. Similarly, since $x^{B}$ agrees with $\alpha$, the set $B$ also must contain all the coordinates where $x$ and $\alpha$ disagree. Therefore the sets $A$ and $B$ cannot be disjoint.

To prove the lemma, let $x$ be any 0 -input of $g_{C}$. Let $B_{1}, \ldots, B_{k}$ be disjoint sensitive blocks for $g_{C}$ on $x$. Note that for every $i \in[k], x^{B_{i}}$ must agree with some partial assignment in $C$. By the above claim, for $i \neq j, x^{B_{i}}$ and $x^{B_{j}}$ must agree with different partial assignments from $C$. Therefore, the number of disjoint sensitive blocks is at most $|C|$.

Next we give a sufficient condition for achieving quadratic separations with constant $\frac{2}{3}$, and prove a lemma that we will use in some of our later constructions in Subsection 3.3.

Definition 10. For an odd integer $m>2$, consider a set of partial assignments $C=\left\{\gamma_{1}, \ldots \gamma_{m}\right\}$, with $\gamma_{i}:[2 m] \rightarrow\{0,1, *\}$. We say that the set of partial assignments $C$ is good if it satisfies the following two properties:
(a) The distance between any two partial assignments $\gamma_{i}, \gamma_{j} \in C$ for $i \neq j$ is at least 3
(b) Each partial assignment $\gamma_{i} \in C$ has exactly two bits set to $1, \frac{3}{2}(m-1)$ bits set to 0 and the remaining bits free.

We prove the following lemma for functions $g_{C}$ defined by a good set of partial assignments $C$.

Lemma 6. For an odd integer $m>2$, and any function $g_{C}:\{0,1\}^{2 m} \rightarrow$ $\{0,1\}$ defined by a good set of partial assignments $C$, we have

1. $s_{0}\left(g_{C}\right)=1$
2. $b s_{0}\left(g_{C}\right)=m$

$$
\text { 3. } s_{1}\left(g_{C}\right)=b s_{1}\left(g_{C}\right)=C_{1}\left(g_{C}\right)=\frac{3 m+1}{2}
$$

Proof. Lemma 4 directly implies the first and third statements, and Lemma 5 implies that $b s_{0}\left(g_{C}\right) \leq m$. It remains to prove that $b s_{0}\left(g_{C}\right) \geq m$.

Recall that any two certificates $\gamma_{i}, \gamma_{j}$ must be at a distance at least 3 from each other. But since each certificate only sets exactly 2 bits to 1 , this implies that the bits set to 1 by $\gamma_{i}$ must be disjoint from the bits set to 1 by $\gamma_{j}$, for any $\gamma_{i}, \gamma_{j} \in C$ and $i \neq j$.

Thus, for the 0 -input $0^{2 m}$, the pair of bits set to 1 by the certificate $\gamma_{i}$ gives a sensitive block for every $i \in[m]$. All these blocks are pairwise disjoint and therefore $b s_{0}\left(g_{C}\right) \geq m$.

Using OR-composition (Lemma 3) yields the following.
Theorem 2. Consider any function $g_{C}:\{0,1\}^{2 m} \rightarrow\{0,1\}$ defined by a good set of partial assignments $C$, for an odd integer $m>2$. Then the function $f=O R_{\frac{3 m+1}{2}} \circ g_{C}$ has

$$
b s(f)=\frac{2}{3} s(f)^{2}-\frac{1}{3} s(f)
$$

We note that the inner function defined by Ambainis and Sun [5] can be shown to fit into this framework. We will use Lemma 6 to analyze the functions defined in subsection 3.3.

Next we observe that Lemma 6 and Theorem 2 can be further generalized as follows. Recall that the size of a partial assignment is the number of variables fixed by it (see Section 2).

Theorem 3. Let $C$ be a set of partial assignments with $|C| \geq 2$ that satisfies the following two properties:
(a) The distance between any two $\gamma_{i}, \gamma_{j} \in C$ for $i \neq j$ is at least 3
(b) The sets of bits set to 1 by the partial assignments in $C$ are pairwise disjoint.
Let $\ell$ denote the size of the largest partial assignment in $C$, and let $m=|C|$ denote the number of partial assignments in $C$. Then for the function $g_{C}$ defined by the set of partial assignments $C$ we have

$$
\begin{aligned}
& \text { 1. } s_{0}\left(g_{C}\right)=1 \\
& \text { 2. } b s_{0}\left(g_{C}\right)=m
\end{aligned}
$$

3. $s_{1}\left(g_{C}\right)=b s_{1}\left(g_{C}\right)=C_{1}\left(g_{C}\right)=\ell$

Hence, the function $f=O R_{\ell} \circ g_{C}$ has

$$
b s(f)=\frac{m}{\ell} s(f)^{2}
$$

The proof of this theorem follows directly from Lemma 4, Lemma 5 and the OR-composition Lemma (Lemma 3).

However, this approach has the following limitation.
Lemma 7. Let $C$ be a set of partial assignments with $|C| \geq 2$ such that the sets of bits set to 1 by the partial assignments in $C$ are pairwise disjoint. Let $\ell$ denote the size of the largest partial assignment in $C$, and let $m=|C|$ denote the number of partial assignments in $C$. Let $D$ denote the average pairwise distance between the partial assignments in $C$. Then

$$
\ell \geq \frac{D}{2}(m-1)
$$

The proof of this Lemma is a straightforward generalization of an argument of Ambainis and Sun [5].

Proof. (Implicit in the proof of Theorem 2 in [5]) For $\gamma_{i} \in C$, let $A_{i}$ denote the set of bits fixed to 0 by $\gamma_{i}$, and let $B_{i}$ denote the set of bits fixed to 1 by $\gamma_{i}$. Then, for $i \neq j$ the distance between $\gamma_{i}$ and $\gamma_{j}$ is equal to $\left|A_{i} \cap B_{j}\right|+\left|A_{j} \cap B_{i}\right|$. Thus,

$$
\sum_{i, j: i \neq j}\left|A_{i} \cap B_{j}\right|=D \frac{m(m-1)}{2}
$$

This implies that for some $i$,

$$
\sum_{j: i \neq j}\left|A_{i} \cap B_{j}\right| \geq D \frac{(m-1)}{2}
$$

which in turn implies that $\left|A_{i}\right| \geq D \frac{(m-1)}{2}$, since the sets $B_{j}$ are pairwise disjoint. The statement of the lemma follows since $\ell \geq \max _{i}\left|A_{i}\right|$.

Lemma 7 implies that using a construction satisfying the conditions of Theorem 3 with average pairwise distance $D$ cannot give a constant greater than $\frac{2}{D}$ in quadratic separations. The construction of Ambainis and Sun fits
into this framework and matches the bound of Lemma 7 with $D=3$ yielding the current best quadratic separation with constant $2 / 3$. The constructions of Rubinstein and Virza also fit into this framework and match the bound of Lemma 7 with $D=4$, yielding quadratic separations with constant $1 / 2$. Note that $D \geq 3$ is part of the conditions in Theorem 3, thus this framework cannot give constants greater than $2 / 3$ in quadratic separations.

### 3.2 Using Finite Field Multiplication

In this subsection, we give constructions of families of functions based on Finite Field Multiplication, which achieve quadratic separation between block sensitivity and sensitivity. We will define two versions, $g_{F F}$ and $g_{F F}^{*}$ yielding quadratic separations with constants $\frac{1}{4}$ and $\frac{1}{2}$, respectively.

Fix an irreducible polynomial $p$ of degree $m$ in $\mathbb{F}_{2}[z]$ and consider the representation of the elements of $\mathbb{F}_{2^{m}}$ as univariate polynomials modulo $p$. For $a \in\{0,1\}^{m}$, we interpret $a=\left(a_{0}, \ldots a_{m-1}\right)$ as an element of $\mathbb{F}_{2^{m}}$ under this representation.

## Definition 11. Function based on Finite Field Multiplication

The function $g_{F F}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$ is defined as follows:
$g_{F F}(a, b)=1$ if and only if $a \cdot b=c$, where $c \in \mathbb{F}_{2^{m}}$ is the element represented as $(0, \ldots, 0,1)$ and multiplication is over the field $\mathbb{F}_{2^{m}}$.

We prove the following lemma listing the values of sensitivity and block sensitivity for the function $g_{F F}$.

Lemma 8. For the function $g_{F F}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$, we have

- $s_{0}\left(g_{F F}\right) \leq 2$ - $s_{1}\left(g_{F F}\right)=2 m$
- $m+1 \geq b s_{0}\left(g_{F F}\right) \geq m$
- $b s_{1}\left(g_{F F}\right)=2 m$

Proof.

- $s_{0}\left(g_{F F}\right) \leq 2$

For any non-zero $a \in \mathbb{F}_{2^{m}}$, there exists a unique $b \in \mathbb{F}_{2^{m}}$ such that $a \cdot b=(0, \ldots, 0,1)$ i.e. $g_{F F}(a, b)=1$. Therefore, for any input $(a, b) \in$ $g_{F F}^{-1}(0)$, at most 1 bit $j$ of $a$ may be flipped to get $a^{j} \cdot b=(0,0 \ldots 1)$ i.e. $a$ has at most 1 sensitive bit. Similarly, at most 1 bit of $b$ may be sensitive.

- $s_{1}\left(g_{F F}\right)=2 m$

Consider any input $(a, b) \in g_{F F}^{-1}(1)$. Flipping any bit of $a$ or $b$ changes the value of the product $a \cdot b$. Therefore every bit of $(a, b)$ is sensitive, giving $s_{1}\left(g_{F F}\right)=2 m$.

- $m+1 \geq b s_{0}\left(g_{F F}\right) \geq m$

Consider the 0 -input $a=(0, \ldots 0), b=(0, \ldots 0)$.
For each $j \in\{0, \ldots m-1\}$, we can flip the pair of bits $\left(a_{j}, b_{m-1-j}\right)$, so that their product becomes $c=(0, \ldots, 0,1)$. This gives $m$ disjoint sensitive blocks.

To see that $m+1 \geq b s_{0}\left(g_{F F}\right)$, note that since $s_{0}\left(g_{F F}\right) \leq 2$, on any 0 -input there are at most two sensitive blocks of size 1 , and all other sensitive blocks must have size at least 2 . Thus, $b s_{0}\left(g_{F F}\right) \leq 2+\frac{2 m-2}{2}=$ $m+1$.

- $b s_{1}\left(g_{F F}\right)=2 m$

This follows since $2 m \geq b s_{1}\left(g_{F F}\right) \geq s_{1}\left(g_{F F}\right)=2 m$

The following theorem follows from Lemma 8 and the OR-composition Lemma.

Theorem 4. The function $f=O R_{m} \circ g_{F F}$ has

$$
b s(f) \geq \frac{1}{4} s(f)^{2}
$$

We now modify the function $g_{F F}$ to improve the constant of separation from $\frac{1}{4}$ to $\frac{1}{2}$.

Definition 12. The function $g_{F F}^{*}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$ is defined as follows:
$g_{F F}^{*}(a, b)=1$ if and only if the following two conditions hold.

1. $a \cdot b=c$, where $c \in \mathbb{F}_{2^{m}}$ is the element represented as $(0, \ldots, 0,1)$ and multiplication is over the field $\mathbb{F}_{2^{m}}$
2. $a_{0} \oplus a_{1} \ldots \oplus a_{m-1}=1$

Lemma 9. For the function $g_{F F}^{*}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$, we have

- $s_{0}\left(g_{F F}^{*}\right)=1$
- $s_{1}\left(g_{F F}^{*}\right)=2 m$
- $b s_{0}\left(g_{F F}^{*}\right)=m$
- $b s_{1}\left(g_{F F}^{*}\right)=2 m$

Proof. Note that Conditions 1 and 2 of Definition 12 both have to hold for 1 inputs, and at least one is violated for 0 inputs.

- $s_{0}\left(g_{F F}^{*}\right)=1$

For a 0-input ( $a, b$ ) which satisfies condition 1, flipping any bit of $a$ or $b$ changes the product $a \cdot b$ and condition 1 is no longer satisfied. Therefore, such a 0 -input has no sensitive bit.
For any 0 -input ( $a, b$ ) which leaves condition 1 unsatisfied, both $a$ and $b$ can have at most one sensitive bit each as observed in the proof of Lemma 8.
We further note that for any given 0 -input $(a, b)$, only one of $a$ or $b$ can have a sensitive bit because condition 2 has to hold for 1-inputs. Therefore, $s_{0}\left(g_{F F}^{*}\right)=1$.

- $s_{1}\left(g_{F F}^{*}\right)=2 m$

Consider any 1 -input ( $a, b$ ). Flipping any bit of $a$ or $b$ changes the value of the product $a \cdot b$ and condition 1 is no longer satisfied. Therefore every bit of $(a, b)$ is sensitive, giving $s_{1}\left(g_{F F}^{*}\right)=2 m$.

- $b s_{0}\left(g_{F F}^{*}\right)=m$

Consider the 0 -input $a=(0, \ldots 0), b=(0, \ldots 0)$.
For each $j \in\{0, \ldots m-1\}$, we can flip the pair of bits $\left(a_{j}, b_{m-1-j}\right)$, so that their product becomes $c=(0,0 \ldots 1)$ to satisfy the first condition. Since $a^{j}$ has exactly one 1 , the second condition is satisfied as well, and $g_{F F}^{*}\left(a^{j}, b^{m-1-j}\right)=1$. This gives $m$ disjoint sensitive blocks and therefore, $b s_{0}\left(g_{F F}^{*}\right) \geq m$.
To see that $b s_{0}\left(g_{F F}^{*}\right) \leq m$, note that since $s_{0}\left(g_{F F}^{*}\right)=1$, on any 0 -input there is at most one sensitive block of size 1, and all other sensitive blocks must have size at least 2. Thus, $b s_{0}\left(g_{F F}^{*}\right) \leq\left\lfloor 1+\frac{2 m-1}{2}\right\rfloor=m$.

- $b s_{1}\left(g_{F F}^{*}\right)=2 m$

This follows since $2 m \geq b s_{1}\left(g_{F F}^{*}\right) \geq s_{1}\left(g_{F F}^{*}\right)=2 m$

Using Lemma 9 and the OR-composition Lemma gives the following theorem.

Theorem 5. The function $f=O R_{2 m} \circ g_{F F}^{*}$ has

$$
b s(f)=\frac{1}{2} s(f)^{2}
$$

Remark 1. We could replace $c=(0, \ldots, 0,1)$ in the above definitions by other field elements and still achieve quadratic separations. In fact using any $c \in \mathbb{F}_{2^{m}}$, we would get $s_{0} \leq 2$ and $s_{1}=2 m$ for the inner function. However, we need to choose c carefully to guarantee that $b s_{0}$ of the inner function is large enough.

### 3.3 Using Polynomial Multiplication

We now describe another family of functions similar in essence to the one involving finite field multiplication, but easier to analyze. We will define two versions, $g_{\text {poly }}$ and $g_{\text {poly }}^{*}$ yielding quadratic separations with constants $\frac{1}{2}$ and $\frac{2}{3}$, respectively.

Here we consider polynomials over the Integers. For $a \in\{0,1\}^{m}$, we interpret the bits of $a=\left(a_{0}, \ldots a_{m-1}\right)$ as the coefficients of a univariate polynomial $p_{a}$ that is $p_{a}(z)=a_{0}+a_{1} z+\ldots a_{m-1} z^{m-1}$.

## Definition 13. Function based on Polynomial Multiplication

Let $m \geq 2$ be an integer. The function $g_{\text {poly }}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$ is defined as follows:
$g_{p o l y}(a, b)=1$ if and only if $p_{a}(z) \cdot p_{b}(z)$ has a non-zero coefficient for $z^{m-1}$ and has coefficient 0 for $z^{j}$ for all $j<m-1$.

It is convenient to use the following equivalent definition.
Definition 14 (Alternative definition).
Let $m \geq 2$ be an integer. Consider the set of partial assignments $C=$ $\left\{\gamma_{0}, \gamma_{1} \ldots \gamma_{m-1}\right\}$ where $\gamma_{i}:\{0,1, \ldots, 2 m-1\} \rightarrow\{0,1, *\}$, defined as follows. For every $i \in\{0, \ldots m-1\}$,

$$
\gamma_{i}(j)= \begin{cases}1, & \text { if } j=i \\ 0, & \text { if } j<i \\ \star, & \text { if } m-1 \geq j>i\end{cases}
$$

$\gamma_{i}(j)= \begin{cases}1, & \text { if } j=2 m-1-i \\ 0, & \text { if } m \leq j<2 m-1-i \\ \star, & \text { if } j>2 m-1-i\end{cases}$
Now the function $g_{p o l y}$ is the function defined by the set of partial assignments $C$ i.e. $g_{C}$.

We now analyze this function for its sensitivity and block sensitivity.
Lemma 10. For $g_{\text {poly }}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$, we have

- $s_{0}\left(g_{\text {poly }}\right)=2$
- $s_{1}\left(g_{\text {poly }}\right)=m+1$
- $b s_{0}\left(g_{\text {poly }}\right)=m$
- $b s_{1}\left(g_{\text {poly }}\right)=m+1$

Proof.

- $s_{1}\left(g_{p o l y}\right)=b s_{1}\left(g_{p o l y}\right)=m+1$

First, we observe from the alternative definition of $g_{p o l y}$ that every 1input of $g_{\text {poly }}$ agrees with a certificate $\gamma_{i}$ from the set $C$. Each $\gamma_{i} \in C$ fixes $m+1$ bits. Also, note that any two certificates of $C$ are at a distance at least 2 from each other. Therefore, by Lemma $4 s_{1}\left(g_{p o l y}\right)=$ $b s_{1}\left(g_{\text {poly }}\right)=C_{1}\left(g_{\text {poly }}\right)=(m+1)$.

- $s_{0}\left(g_{\text {poly }}\right)=2$

First we prove $s_{0}\left(g_{p o l y}\right) \leq 2$.
Observe that if either $a$ or $b$ is an all 0 vector, then the sensitivity on such input $(a, b)$ is at most 1 . Let $(a, b)$ be any 0 -input of $g_{\text {poly }}$ such that neither $a$ nor $b$ is all 0 . Let $i \in\{0, \ldots m-1\}$ be the smallest index such that $a_{i}=1$ and $j$ be the smallest index such that $b_{j}=1$. Then, we have two cases.
Case 1: $i+j>m-1$. In this case, in the product $p_{a}(z) \cdot p_{b}(z)$ the coefficients are 0 for $z^{j}$ for all $j \leq m-1$. Thus, the only bits which can be flipped to change the value of $g_{p o l y}$ from 0 to 1 are $a_{m-1-j}$ and $b_{m-1-i}$.

Case 2: $i+j<m-1$. Now, the only way to flip a bit and possibly change the value of $g_{\text {poly }}$ to 1 is by flipping the bits $a_{i}$ or $b_{j}$.

Next we argue that $s_{0}\left(g_{p o l y}\right) \geq 2$.
The following 0 -input $(a, b)$ achieves $s_{0}\left(g_{\text {poly }},(a, b)\right)=2$.
Let $a_{m-1}=1, a_{i}=0 \forall i<(m-1)$.
Similarly, $b_{m-1}=1, b_{i}=0 \forall i<(m-1)$.
Notice that $(a, b)$ has 2 sensitive bits: $a_{0}$ and $b_{0}$.

- $b s_{0}\left(g_{\text {poly }}\right)=m$

Lemma 5 implies that $b s_{0}\left(g_{p o l y}\right) \leq m$.
It remains to prove that $b s_{0}\left(g_{\text {poly }}\right) \geq m$. Consider the 0 -input with $a_{i}=b_{i}=0$ for all $i \in\{0, \ldots m-1\}$. We can flip the pair of bits $a_{i}, b_{m-1-i}$ for $i \in\{0, \ldots m-1\}$ so that the function changes value from 0 to 1 . Therefore, $b s_{0}\left(g_{\text {poly }}\right) \geq m$.

Lemma 10 and the OR-composition Lemma imply the following theorem.
Theorem 6. Consider $g_{\text {poly }}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$ for any odd integer $m>2$.
Then the function $f=O R_{\frac{m+1}{2}} \circ g_{\text {poly }}$ has

$$
b s(f)=\frac{1}{2} s(f)^{2}-\frac{1}{2} s(f)
$$

We modify the above function to improve the constant of separation from $\frac{1}{2}$ to $\frac{2}{3}$.

Definition 15. Let $m \geq 2$ be an integer. The function $g_{\text {poly }}^{*}:\{0,1\}^{m} \times$ $\{0,1\}^{m} \rightarrow\{0,1\}$, is defined as $g_{\text {poly }}^{*}(a, b)=1$ if and only if all the following conditions are met.

1. $p_{a}(z) \cdot p_{b}(z)$ has a non-zero coefficient for $z^{m-1}$ and has coefficient 0 for $z^{j}$ for all $j<m-1$
2. If $j$ is the smallest index such that $a_{j}=1$, then $a_{i}=0$ for all $i$ such that $i>j$ and $i+j$ is odd
3. If $k$ is the smallest index such that $b_{k}=1$, then
$b_{i}=0$ for all $i$ such that $i>k$ and $i+k$ is even
It is again helpful to consider an equivalent definition based on certificates.
Definition 16 (Alternative definition).
Let $m \geq 2$ be an integer. Consider the set of partial assignments $C^{\prime}=$ $\left\{\gamma_{0}^{\prime}, \gamma_{1}^{\prime}, \ldots \gamma_{m-1}^{\prime}\right\}$, where $\gamma_{i}^{\prime}:\{0,1, \ldots, 2 m-1\} \rightarrow\{0,1, *\}$ defined as follows. For every $i \in\{0, \ldots m-1\}$,

$$
\begin{gathered}
\gamma_{i}^{\prime}(j)= \begin{cases}1, & \text { if } j=i \\
0, & \text { if } j<i \\
0, & \text { if } m-1 \geq j>i \text { and } i+j \text { is odd } \\
\star, & \text { if } m-1 \geq j>i \text { and } i+j \text { is even }\end{cases} \\
\gamma_{i}^{\prime}(j)= \begin{cases}1, & \text { if } j=2 m-1-i \\
0, & \text { if } m \leq j<2 m-1-i \\
0, & \text { if } j>2 m-1-i \text { and }(2 m-1-i)+j \text { is even } \\
\star, & \text { if } j>2 m-1-i \text { and }(2 m-1-i)+j \text { is odd }\end{cases}
\end{gathered}
$$

Now the function $g_{\text {poly }}^{*}$ is the function defined by the set of partial assignments $C^{\prime}$ i.e. $g_{C^{\prime}}$.
Lemma 11. Consider $g_{\text {poly }}^{*}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$ for any odd integer $m>2$. Then

- $s_{0}\left(g_{\text {poly }}^{*}\right)=1$
- $s_{1}\left(g_{\text {poly }}^{*}\right)=\frac{3 m+1}{2}$
- $b s_{0}\left(g_{\text {poly }}^{*}\right)=m$
- $b s_{1}\left(g_{\text {poly }}^{*}\right)=\frac{3 m+1}{2}$

Proof. It is clear from the alternative definition of $g_{\text {poly }}^{*}$ that it is defined by a set of good partial assignments as introduced in definition 10. The result then follows from Lemma 6.

The following theorem follows from Lemma 11 and the OR-composition Lemma.

Theorem 7. Consider $g_{\text {poly }}^{*}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$ for any odd integer $m>2$.
Then the function $f=O R_{\frac{3 m+1}{2}} \circ g_{\text {poly }}^{*}$ has

$$
b s(f)=\frac{2}{3} s(f)^{2}-\frac{1}{3} s(f)
$$

Note that this bound matches the current best quadratic separation of Ambainis and Sun [5].

## 4 Additional Properties of Function Composition

As we noted in Section 2, properties of function composition have been formally studied by Tal and Gilmer et al. [18, 9] in the context of separating sensitivity and block sensitivity. Here we take a closer look at the effect of function composition on the measures 0 -sensitivity, 1 -sensitivity, 0 -block sensitivity and 1-block sensitivity. These properties provide the tools we need to obtain quadratic separation of both 0-block sensitivity and 1-block sensitivity from sensitivity.

First we define measures to quantify the number of sensitive bits for $f$ on $x$ which are equal to 0 and those that are equal to 1 in $x$.

Definition 17. For a function $f:\{0,1\}^{n} \rightarrow\{0,1\}$ and input $x \in\{0,1\}^{n}$, we define
$\sigma_{1}(f, x)=\mid\left\{i \mid x_{i}=1 \quad\right.$ AND $\left.\quad f(x) \neq f\left(x^{i}\right)\right\} \mid$
$\sigma_{0}(f, x)=\mid\left\{i \mid x_{i}=0 \quad\right.$ AND $\left.\quad f(x) \neq f\left(x^{i}\right)\right\} \mid$.
We will use the following notation. We index the bits of the input $y \in$ $\{0,1\}^{m n}$ to $f \circ g$ as $y=\left(y_{11}, y_{12}, \ldots y_{1 m}, y_{21}, \ldots y_{2 m}, \ldots y_{n 1}, \ldots y_{n m}\right)$.
We denote by $y_{i}$ the $i$-th group of $m$ bits of $y$, that is $y_{i}=\left(y_{i 1}, y_{i 2}, \ldots y_{i m}\right)$.
Lemma 12. For any functions $f:\{0,1\}^{n} \rightarrow\{0,1\}$ and $g:\{0,1\}^{m} \rightarrow\{0,1\}$, we have

$$
\begin{aligned}
& s_{0}(f \circ g)=\max _{x \in f^{-1}(0)}\left\{\sigma_{0}(f, x) s_{0}(g)+\sigma_{1}(f, x) s_{1}(g)\right\} \\
& s_{1}(f \circ g)=\max _{x \in f^{-1}(1)}\left\{\sigma_{0}(f, x) s_{0}(g)+\sigma_{1}(f, x) s_{1}(g)\right\}
\end{aligned}
$$

Proof. We prove the first equation. The second equation has an analogous proof.

First we prove $s_{0}(f \circ g) \geq \max _{x \in f^{-1}(0)}\left\{\sigma_{0}(f, x) s_{0}(g)+\sigma_{1}(f, x) s_{1}(g)\right\}$. Consider the input $a \in f^{-1}(0)$ for which $\left(\sigma_{0}(f, a) s_{0}(g)+\sigma_{1}(f, a) s_{1}(g)\right)$ is maximized. Note that $s_{0}(g), s_{1}(g)$ don't change for different choices of $a \in f^{-1}(0)$. Now,
consider an input $y \in\{0,1\}^{m n}$ such that, $a=\left(g\left(y_{1}\right), \ldots g\left(y_{n}\right)\right)$ and for each $i \in[n]$, if $a_{i}=g\left(y_{i}\right)=0$, then $s\left(g, y_{i}\right)=s_{0}(g)$ and if $a_{i}=g\left(y_{i}\right)=1$, then $s\left(g, y_{i}\right)=s_{1}(g)$. So if $a_{i}=0$, we choose as $y_{i}$ a 0 -input of $g$ which achieves the 0 -sensitivity of $g$, and similarly, if $a_{i}=1$, we choose as $y_{i}$ a 1 -input of $g$ which achieves the 1 -sensitivity of $g$.

Since $a \in f^{-1}(0), y$ must be a 0 -input of $f \circ g$. Therefore, we have

$$
s_{0}(f \circ g) \geq s(f \circ g, y) \geq \sigma_{0}(f, a) s_{0}(g)+\sigma_{1}(f, a) s_{1}(g)
$$

Next we prove $s_{0}(f \circ g) \leq \max _{x \in f^{-1}(0)}\left\{\sigma_{0}(f, x) s_{0}(g)+\sigma_{1}(f, x) s_{1}(g)\right\}$. Consider an input $y \in\{0,1\}^{m n}$ which achieves the 0 -sensitivity of $f \circ g$ i.e. $s_{0}(f \circ g)=$ $s(f \circ g, y)$. Let $g\left(y_{1}\right)=x_{1}, g\left(y_{2}\right)=x_{2}$ and so on, and let $x=\left(x_{1}, x_{2} \ldots x_{n}\right)$. Consider the expression $\left(\sigma_{0}(f, x) s_{0}(g)+\sigma_{1}(f, x) s_{1}(g)\right)$. Now, if a bit $y_{i j}$ of $y$ is sensitive for $f \circ g$, then the bit $x_{i}=g\left(y_{i}\right)$ must be a sensitive bit for $f$ on $x$.
Now, consider the set $\mathcal{X}_{0}$ of indices $i \in[n]$ constructed the following way: $i$ is included in $\mathcal{X}_{0}$ if and only if $x_{i}=g\left(y_{i}\right)=0$ and there is a bit $y_{i j}$ sensitive for $f \circ g$ on $y$.
Similarly, we define the set $\mathcal{X}_{1}$ of indices $i \in[n]$ constructed the following way: $i$ is included in $\mathcal{X}_{1}$ if and only if $x_{i}=g\left(y_{i}\right)=1$ and there is a bit $y_{i j}$ sensitive for $f \circ g$ on $y$.
Note that for every $i \in \mathcal{X}_{0}$, the bit $x_{i}$ is a 0 -bit of $x$ and $f$ is sensitive to the $i$-th bit on $x$. So $\left|\mathcal{X}_{0}\right| \leq \sigma_{0}(f, x)$.
Similarly $\left|\mathcal{X}_{1}\right| \leq \sigma_{1}(f, x)$.
Now,

$$
\begin{aligned}
s(f \circ g, y) & =\sum_{i \in \mathcal{X}_{0}} s\left(g, y_{i}\right)+\sum_{j \in \mathcal{X}_{1}} s\left(g, y_{j}\right) \\
& \leq \sum_{i \in \mathcal{X}_{0}} s_{0}(g)+\sum_{j \in \mathcal{X}_{1}} s_{1}(g) \\
& \leq \sigma_{0}(f, x) s_{0}(g)+\sigma_{1}(f, x) s_{1}(g)
\end{aligned}
$$

Therefore,

$$
s_{0}(f \circ g)=s(f \circ g, y) \leq \sigma_{0}(f, x) s_{0}(g)+\sigma_{1}(f, x) s_{1}(g)
$$

Since $x \in f^{-1}(0)$, this concludes the proof.

To simplify the equations of Lemma 12 (at the cost of being less precise), we define

- $\sigma_{0}^{0}(f):=\max _{x \in f^{-1}(0)} \sigma_{0}(f, x)$
- $\sigma_{1}^{0}(f):=\max _{x \in f^{-1}(0)} \sigma_{1}(f, x)$
- $\sigma_{0}^{1}(f):=\max _{x \in f^{-1}(1)} \sigma_{0}(f, x)$
- $\sigma_{1}^{1}(f):=\max _{x \in f^{-1}(1)} \sigma_{1}(f, x)$

Finally, we define

$$
\begin{aligned}
\sigma_{0}(f) & :=\max \left\{\sigma_{0}^{0}(f), \sigma_{0}^{1}(f)\right\} \\
\sigma_{1}(f) & :=\max \left\{\sigma_{1}^{0}(f), \sigma_{1}^{1}(f)\right\}
\end{aligned}
$$

We can now use Lemma 12 to get the following bounds.
Corollary 1. For any functions $f:\{0,1\}^{n} \rightarrow\{0,1\}$ and $g:\{0,1\}^{m} \rightarrow\{0,1\}$

$$
\begin{aligned}
& s_{0}(f \circ g) \leq \sigma_{0}^{0}(f) s_{0}(g)+\sigma_{1}^{0}(f) s_{1}(g) \\
& s_{1}(f \circ g) \leq \sigma_{0}^{1}(f) s_{0}(g)+\sigma_{1}^{1}(f) s_{1}(g)
\end{aligned}
$$

Note that the equalities in Lemma 12 change to inequalities in Corollary 1 , since the max of $\sigma_{0}(f, x)$ and $\sigma_{1}(f, x)$ may be achieved on different inputs among $x \in f^{-1}(0)$ (or among $x \in f^{-1}(1)$, respectively).
We now state some simple observations for these measures.
Lemma 13. For any function $f:\{0,1\}^{n} \rightarrow\{0,1\}$, and any input $x$, we have

1. $s(f, x)=\sigma_{0}(f, x)+\sigma_{1}(f, x)$
2. $\sigma_{0}^{0}(f) \leq s_{0}(f) \quad$ 5. $\sigma_{0}^{1}(f) \leq s_{1}(f)$
3. $\sigma_{1}^{0}(f) \leq s_{0}(f)$
4. $\sigma_{1}^{1}(f) \leq s_{1}(f)$
5. $\sigma_{0}^{0}(f)+\sigma_{1}^{0}(f) \geq s_{0}(f)$
6. $\sigma_{0}^{1}(f)+\sigma_{1}^{1}(f) \geq s_{1}(f)$

The proof of Lemma 13 is straightforward from the definitions.

Now we present an observation about these measures for monotone functions.

Lemma 14. For any monotone function $f:\{0,1\}^{n} \rightarrow\{0,1\}$, we have

- $\sigma_{0}^{1}(f)=0$
- $\sigma_{1}^{0}(f)=0$

The proof follows from the definition of monotone functions.
We now consider the effects of function composition on 0 - block sensitivity and 1-block sensitivity.

Lemma 15. For any functions $f:\{0,1\}^{n} \rightarrow\{0,1\}$ and $g:\{0,1\}^{m} \rightarrow\{0,1\}$, we have

$$
\begin{aligned}
b s_{0}(f \circ g) & \geq b s_{0}(f) \cdot \min \left\{b s_{0}(g), b s_{1}(g)\right\} \\
b s_{1}(f \circ g) & \geq b s_{1}(f) \cdot \min \left\{b s_{0}(g), b s_{1}(g)\right\}
\end{aligned}
$$

Proof. Consider $x \in\{0,1\}^{n}$ such that $f(x)=0$ and $b s_{0}(f)=b s(f, x)$.
Now, consider input $y \in\{0,1\}^{m n}$ such that, $x=\left(g\left(y_{1}\right), \ldots g\left(y_{n}\right)\right)$ and for each $i \in[n]$, if $x_{i}=g\left(y_{i}\right)=0$, then $b s\left(g, y_{i}\right)=b s_{0}(g)$ and if $x_{i}=g\left(y_{i}\right)=1$, then $b s\left(g, y_{i}\right)=b s_{1}(g)$. So if $x_{i}=0$, we choose as $y_{i}$ a 0 -input of $g$ on which its 0 -block sensitivity is achieved, and similarly, if $x_{i}=1$, we choose as $y_{i}$ a 1 -input of $g$ on which its 1 -block sensitivity is achieved.
Now, we claim that $b s_{0}(f \circ g, y) \geq b s_{0}(f) \min \left\{b s_{0}(g), b s_{1}(g)\right\}$. To see this, let $\rho_{1}, \rho_{2}, \ldots, \rho_{k}$ be the disjoint sensitive blocks for $f$ on $x$ where $k=b s(f, x)$. For each of these sensitive blocks, there are at least $\min \left\{b s_{0}(g), b s_{1}(g)\right\}$ disjoint blocks of $y$ such that flipping any of them changes the value of $f \circ g$. This gives at least $b s_{0}(f) \cdot \min \left\{b s_{0}(g), b s_{1}(g)\right\}$ disjoint sensitive blocks for $f \circ g$ on the input $y$, and the first equation follows.
The second equation can be proved in an analogous way.
We get a stronger form of Lemma 15 if $f$ satisfies some additional conditions.

Lemma 16. For any functions $f:\{0,1\}^{n} \rightarrow\{0,1\}$ and $g:\{0,1\}^{m} \rightarrow\{0,1\}$ if $f$ satisfies (1) $f\left(0^{n}\right)=0$, (2) $b s_{0}(f)=b s\left(f, 0^{n}\right)$, then $b s_{0}(f \circ g) \geq b s_{0}(f)$. $b s_{0}(g)$
and if $f$ satisfies (1) $f\left(1^{n}\right)=1$, (2) $b s_{1}(f)=b s\left(f, 1^{n}\right)$, then $b s_{1}(f \circ g) \geq$ $b s_{1}(f) \cdot b s_{1}(g)$

Proof. Consider input $y \in\{0,1\}^{m n}$ such that, $0^{n}=\left(g\left(y_{1}\right), \ldots g\left(y_{n}\right)\right)$ and $b s\left(g, y_{i}\right)=b s_{0}(g)$ for each $i \in[n]$.

Now, we claim that $b s(f \circ g, y) \geq b s_{0}(f) b s_{0}(g)$. To see this, let $\rho_{1}, \rho_{2}, \ldots, \rho_{k}$ be the disjoint sensitive blocks for $f$ on input $0^{n}$, where $k=b s\left(f, 0^{n}\right)$. For each of these $k=b s\left(f, 0^{n}\right)$ sensitive blocks, there are $b s_{0}(g)$ disjoint blocks of $y$ that we can flip and change the value of $f \circ g$.
This gives $b s_{0}(f) \cdot b s_{0}(g)$ disjoint sensitive blocks for $f \circ g$ on the input $y$. The second inequality can be proved analogously.

Comparing the statement of Lemma 16 with Lemma 2 of Tal [18] we note that in the context of 0-block sensitivity and 1-block sensitivity it is enough to require an additional condition for the outer function. On the other hand the condition on the inner function in Lemma 2 of Tal [18] is necessary as illustrated by considering $f=O R_{n}$ and $g=A N D_{n}$.

Note that the conditions we require are similar to, but slightly different from Definition 9 by Tal [18] of being in $z$-good form. It follows from Definition 9 , that if $f$ is in $z$-good form, then $b s(f)=b s_{z}(f)$. Our conditions do not require that $b s(f)=b s_{z}(f)$ for a specific $z$.

## 5 Quadratic Separation of both $b s_{0}(f)$ and $b s_{1}(f)$ from $s(f)$

We obtain constructions of functions with quadratic separation of both 0 block sensitivity and 1 -block sensitivity from sensitivity by considering various compositions of our new building blocks as well as some of the inner functions used in previous quadratic separations.

Theorem 8. Consider $g_{\text {poly }}:\{0,1\}^{m} \times\{0,1\}^{m} \rightarrow\{0,1\}$.
Let $f:\{0,1\}^{4 m^{2}} \rightarrow\{0,1\}$ be defined as $f=g_{\text {poly }} \circ g_{\text {poly }}$.
Then, we have

- $s_{0}(f)=2(m+1)$
- $s_{1}(f)=4 m$
- $b s_{0}(f) \geq m^{2}$
- $b s_{1}(f) \geq m(m+1)$

Therefore, we have

$$
\min \left\{b s_{0}(f), b s_{1}(f)\right\}=\Omega\left(s(f)^{2}\right)
$$

Proof. In this proof, we refer to $g_{\text {poly }}$ by $g$, and we use the notation $b s_{\text {min }}(f)=$ $\min \left\{b s_{0}(f), b s_{1}(f)\right\}$.

We first prove the following claims about $\sigma$-values for $g$.
Claim 2. For any input $x \in g^{-1}(0)$ exactly one of the following must be true:

- $\sigma_{0}(g, x)=s(g, x)$ and $\sigma_{1}(g, x)=0$
- $\sigma_{1}(g, x)=s(g, x)$ and $\sigma_{0}(g, x)=0$

Proof of Claim:
For any 0 -input $x=(a, b)$ of $g,(1)$ of Lemma 13 states that
$\sigma_{0}(g, x)+\sigma_{1}(g, x)=s(g, x)$.
As in the definition of $g_{\text {poly }}$, consider the polynomials $p_{a}(z), p_{b}(z)$. If the lowest degree monomial of $p_{a}(z) p_{b}(z)$ with a non-zero coefficient is $z^{t}$ then, we have 2 cases:
Case 1: $t<m-1$. In this case, no 0 -bit of $a$ or $b$ can be sensitive. Therefore, $\sigma_{0}(g, x)=0$ and $\sigma_{1}(g, x)=s(g, x)$.
Case 2: $t>m-1$. In this case, no 1-bit of $a$ or $b$ can be sensitive. Therefore, $\sigma_{1}(g, x)=0$ and $\sigma_{0}(g, x)=s(g, x)$.
Claim 3. For an input $x \in g^{-1}(1)$,

- $\sigma_{0}(g, x)=m-1$
- $\sigma_{1}(g, x)=2$


## Proof of Claim:

Recall the alternative definition based on certificates. Any 1-input $x$ of $g$ belongs to a unique subcube given by a certificate $\gamma_{i} \in C$. Since the subcubes corresponding to different certificates in $C$ are disjoint and at distance at least 2 from each other, every bit of $x$ that is fixed by $\gamma_{i}$ is sensitive.
Since each certificate fixes exactly 2 bits to 1 and $(m-1)$ bits to 0 , we have $\sigma_{0}(g, x)=(m-1)$ and $\sigma_{1}(g, x)=2$.
We can now use Lemma 12 to compute the sensitivity of $f$ :
$s_{0}(f)=2 s_{1}(g)=2(m+1)$.
$s_{1}(f)=(m-1) \cdot 2+2 \cdot(m+1)=4 m$
Since $g\left(0^{2 m}\right)=0$ and $b s_{0}(g)=b s\left(g, 0^{2 m}\right)$, we can use Lemma 16 to get $b s_{0}(f) \geq b s_{0}(g)^{2}=m^{2}$.
We can use Lemma 15 to get
$b s_{1}(f) \geq b s_{1}(g) \cdot \min \left\{b s_{0}(g), b s_{1}(g)\right\}=m(m+1)$.
Therefore, we have $b s(f) \geq \frac{s(f)^{2}}{16}$ and $b s_{\min }(f)=\Omega\left(s(f)^{2}\right)$.

We prove the following general theorem.
Theorem 9. For functions $f:\{0,1\}^{n} \rightarrow\{0,1\}$ and $g:\{0,1\}^{n} \rightarrow\{0,1\}$ such that the following conditions hold:

1. $\sigma_{1}(f)=c_{1}$, where $c_{1}$ is some fixed constant
2. $s_{0}(g)=c_{2}$, where $c_{2}$ is some fixed constant
3. $b s_{0}(f), b s_{1}(f), b s_{0}(g), b s_{1}(g)=\theta(n)$

We have,

$$
\min \left\{b s_{0}(f \circ g), b s_{1}(f \circ g)\right\}=\Omega\left(s(f \circ g)^{2}\right)
$$

The proof is straightforward from Corollary 1 and Lemma 15.
This theorem allows us to use various compositions of our new building blocks and some of the inner functions of previous constructions to obtain other functions with both 0 -block sensitivity and 1-block sensitivity quadratically larger than sensitivity.
In particular, let $f, g$ be any two functions from the following list of functions: Rubinstein's inner function [16], Virza's inner function [20], Ambainis and Sun's inner function [5], $g_{\text {poly }}, g_{\text {poly }}^{*}$. In addition, we can also let $g$ be $g_{F F}, g_{F F}^{*}$, or the inner function of the function based on Hamming Code [10]. Then, $b s_{0}(f \circ g)$ and $b s_{1}(f \circ g)$ are both quadratically larger than $s(f \circ g)$.

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