

# The sum of $d$ small-bias generators fools polynomials of degree $d$

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## Abstract

We prove that the sum of  $d$  small-bias generators  $L : \mathbb{F}^s \rightarrow \mathbb{F}^n$  fools degree- $d$  polynomials in  $n$  variables over a prime field  $\mathbb{F}$ , for any fixed degree  $d$  and field  $\mathbb{F}$ , including  $\mathbb{F} = \mathbb{F}_2 = \{0, 1\}$ . Our result builds on, simplifies, and improves on both the work by Bogdanov and Viola (FOCS '07) and the beautiful follow-up by Lovett (STOC '08). The first relies on a conjecture that turned out to be true only for some degrees and fields, while the latter considers the sum of  $2^d$  small-bias generators (as opposed to  $d$  in our result).

## 1 Introduction

A *pseudorandom generator*  $G : \mathbb{F}^s \rightarrow \mathbb{F}^n$  for polynomials of degree  $d$  over a prime field  $\mathbb{F}$  is an efficient procedure that stretches  $s$  field elements into  $n \gg s$  field elements that *fool* any polynomial of degree  $d$  in  $n$  variables over  $\mathbb{F}$ : For every such polynomial  $p$ , the statistical distance between  $p(U)$ , for uniform  $U \in \mathbb{F}^n$ , and  $p(G(S))$ , for uniform  $S \in \mathbb{F}^s$ , is at most a small  $\epsilon$ .

The fundamental case of linear, i.e. degree-1, polynomials is first studied by Naor and Naor [NN] who give a generator with seed length  $s = O(\log_{|\mathbb{F}|} n)$  (for error  $\epsilon = 1/n$ ), which is optimal up to constant factors (cf. [AGHP]).<sup>1</sup> This generator is known as *small-bias generator*, and is one of the most celebrated results in pseudorandomness, with a myriad of applications (see, e.g., the references in [BV]).

The case of higher degree is first addressed by Luby, Veličković, and Wigderson [LVW], and a decade later by Bogdanov [Bog]. However, the generators in [LVW, Bog] have poor seed length or only work over very large fields.

Recently, Bogdanov and the author [BV] introduce a new approach to attack this problem over small fields, which we now describe. The work considers the generator  $G_k : \mathbb{F}^s \rightarrow \mathbb{F}^n$

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<sup>1</sup>Naor and Naor [NN] only consider the case  $\mathbb{F} = \mathbb{F}_2$ . However, it has been observed by several researchers that their result extends to any prime field.

that is obtained by summing  $k$  copies of a small-bias generator  $L : \mathbb{F}^{s'} \rightarrow \mathbb{F}^n$  by Naor and Naor [NN], which fools linear (i.e., degree-1) polynomials:

$$G_k(s_1, \dots, s_k) := L(s_1) + \dots + L(s_k),$$

where the sum is element-wise. [BV] shows that such a generator can be analyzed using the so-called *Gowers norms*. It unconditionally shows that  $G_d$  fools polynomials of degree  $d$  for  $d \leq 3$ . For larger  $d > 3$ , the work proves a conditional result. Specifically, it introduces a special case of a conjecture known as the Gowers inverse conjecture [GT1, Sam]. This special case is called the “ $d$  vs.  $d - 1$  Gowers inverse conjecture” and we subsequently refer to it as “d-GIC.” Under d-GIC, [BV] shows that  $G_d$  fools polynomials of degree  $d$  for every  $d$ . Moreover, a counting argument shows that  $G_d$  achieves the optimal dependence of the seed length  $s$  on the number of variables  $n$ , up to additive terms. (In particular,  $G_{d-1}$  does not fool polynomials of degree  $d$ .)

Subsequently, Lovett [Lov] unconditionally shows that  $G_{2^d}$  fools polynomials of degree  $d$ , for every  $d$ . Lovett’s proof does not use the theory of Gowers norms, but it applies to the sum of an exponential number  $2^d$  of small-bias generators, as opposed to  $d$  in [BV].

Very recently, Green and Tao [GT2] prove that d-GIC is true *when the field size  $|\mathbb{F}|$  is bigger than the degree  $d$  of the polynomial*. Thus, in this case, the approach in [BV] works and in particular one has that  $G_d$  fools polynomials of degree  $d$ . On the negative side, Green and Tao [GT2], and independently Lovett, Meshulam, and Samorodnitsky [LMS], show that d-GIC is *false* when the field size is much smaller than the degree of the polynomial (which in particular falsifies the more general Gowers inverse conjecture [GT1, Sam]). This falsity prevents the analysis in [BV] to go through for small fields, notably over  $\mathbb{F}_2 = \{0, 1\}$ . Still, it was left open to understand whether, regardless of the Gowers inverse conjecture, the generator  $G_d$  in [BV] fools polynomials of degree  $d$  over small fields such as  $\mathbb{F}_2$ . In this work we answer this question in the affirmative.

## 1.1 Our results

In this section we state our results. We state them over  $\mathbb{F}_2 = \{0, 1\}$  for simplicity, though they hold over any prime field (the necessary details appear in [BV]). Also, we state them for distributions rather than generators; the translation into the language of generators is immediate. Let us start by formalizing the standard notion of *fooling*.

**Definition 1** (Fool). *We say that a distribution  $W$  on  $\{0, 1\}^n$   $\epsilon$ -fools degree- $d$  polynomials in  $n$  variables over  $\mathbb{F}_2$  if for every such polynomial  $p$  we have:*

$$|\mathbb{E}_W e[p(W)] - \mathbb{E}_U e[p(U)]| \leq \epsilon,$$

where  $U$  is the uniform distribution over  $\{0, 1\}^n$  and  $e[x] := (-1)^x$ .

The requirement in Definition 1 informally means that degree- $d$  polynomials have advantage at most  $\epsilon$  in distinguishing a pseudorandom input  $W$  from a truly random input  $U$ .

This requirement can be equivalently expressed in terms of statistical distance (cf. [BV]), but the above formulation is more convenient for our purposes.

The following is our main theorem.

**Theorem 2** (The sum of  $d$  small-bias generators fools degree- $d$  polynomials). *Let  $Y_1, \dots, Y_d \in \{0, 1\}^n$  be  $d$  independent distributions that  $\epsilon$ -fool degree-1 polynomials in  $n$  variables over  $\mathbb{F}_2 = \{0, 1\}$ . Then the distribution  $W := Y_1 + \dots + Y_d$   $\epsilon_d$ -fools degree- $d$  polynomials in  $n$  variables over  $\mathbb{F}_2$  where*

$$\epsilon_d := 16 \cdot \epsilon^{1/2^{d-1}}.$$

Standard constructions of small-bias generators [NN, AGHP] has seed length  $O(\log n/\epsilon)$ . Plugging these in Theorem 2 gives an explicit generator  $\mathbb{F}_2^s \rightarrow \mathbb{F}_2^n$  whose output distribution (over random input)  $\epsilon$ -fools degree- $d$  polynomials with seed length  $s = O(d \cdot \log n + d \cdot 2^d \cdot \log(1/\epsilon))$ . Folklore constructions of small-bias generators have the more refined seed length  $\log n + O(\log(1/\epsilon))$ . Plugging these in Theorem 2 gives a generator whose output distribution  $\epsilon$ -fools degree- $d$  polynomials with seed length  $s = d \cdot \log n + O(d \cdot 2^d \cdot \log(1/\epsilon))$ , which for fixed  $d$  and  $\epsilon$  is optimal in  $n$  up to an additive constant [BV].

Although Theorem 2 improves on previous work [BV, Lov], it still gives nothing for degree  $d = \log_2 n$ . Whether this barrier can be broken is an interesting open problem that is reminiscent of the analogous open problem in the literature on correlation bounds (cf. [VW]).

## 2 Proof of Theorem 2

The proof of Theorem 2 builds on and somewhat simplifies [BV, Lov]. Following [BV, Lov], the proofs goes by induction on  $d$ . However, it differs in the inductive step. The inductive step in [BV] is a case analysis based on the *Gowers norm* of the polynomial  $p$  to be fooled, while the one in [Lov] is a case analysis based on the *Fourier coefficients* of  $p$ . The inductive step in this work is in hindsight natural: It is a case analysis based on the *bias* of  $p$ , which is the quantity

$$\mathbb{E}_{U \in \{0,1\}^n} e[p(U)] \in [-1, 1].$$

The next Lemma 3 deals with polynomials whose bias is close to 0, whereas Lemma 4 deals with polynomials whose bias is far from 0. The analysis in the case of bias close to 0 (Lemma 3) is the main contribution of this work and departure from [BV, Lov]. The simplification of the inductive step, mentioned above, is less crucial in the sense that one could plug Lemma 3 in the analysis in [Lov] to obtain Theorem 2 with a slightly worse error bound.

**Lemma 3** (Fooling polynomials with bias close to 0). *Let  $W \in \{0, 1\}^n$  be a distribution that  $\epsilon_d$ -fools degree- $d$  polynomials, and let  $Y \in \{0, 1\}^n$  be a distribution that  $\epsilon_1$ -fools degree-1 polynomials. Let  $p$  be a polynomial of degree  $d + 1$  in  $n$  variables over  $\mathbb{F}_2$ . Then*

$$|\mathbb{E}_{W,Y} e[p(W + Y)] - \mathbb{E}_U e[p(U)]| \leq 2 \cdot |\mathbb{E}_U e[p(U)]| + \epsilon_1 + \sqrt{\epsilon_d}.$$

*Proof of Lemma 3.* We start by an application of the Cauchy-Schwarz inequality which gives

$$\mathbb{E}_{W,Y} e [p(W + Y)]^2 \leq \mathbb{E}_W [\mathbb{E}_Y e [p(W + Y)]^2] = \mathbb{E}_{W,Y,Y'} e [p(W + Y) + p(W + Y')], \quad (1)$$

where  $Y'$  is independent from and identically distributed to  $Y$ . Now we observe that for every fixed  $Y$  and  $Y'$ , the polynomial  $p(U + Y) + p(U + Y')$  has degree  $d$  in  $U$ , though  $p$  has degree  $d + 1$ . Since  $W$   $\epsilon_d$ -fools degree- $d$  polynomials, we can replace  $W$  with the uniform distribution  $U \in \{0, 1\}^n$ :

$$\mathbb{E}_{W,Y,Y'} e [p(W + Y) + p(W + Y')] \leq \mathbb{E}_{U,Y,Y'} e [p(U + Y) + p(U + Y')] + \epsilon_d. \quad (2)$$

At this point, a standard argument shows that

$$\mathbb{E}_{U,Y,Y'} e [p(U + Y) + p(U + Y')] \leq \mathbb{E}_{U,U'} e [p(U) + p(U')] + \epsilon_1^2 = \mathbb{E}_U e [p(U)]^2 + \epsilon_1^2. \quad (3)$$

Therefore, chaining Equations (1), (2), and (3), we have that

$$\begin{aligned} |\mathbb{E}_{W,Y} e [p(W + Y)] - \mathbb{E}_U e [p(U)]| &\leq |\mathbb{E}_{W,Y} e [p(W + Y)]| + |\mathbb{E}_U e [p(U)]| \leq \\ &\sqrt{\mathbb{E}_U e [p(U)]^2 + \epsilon_1^2 + \epsilon_d} + |\mathbb{E}_U e [p(U)]| \leq 2 \cdot |\mathbb{E}_U e [p(U)]| + \epsilon_1 + \sqrt{\epsilon_d}, \end{aligned}$$

which concludes the proof of the lemma.

For completeness, we include a derivation of Equation (3) next. This equation makes no assumption on  $p$  and can be thought of as a form of the so-called expander mixing lemma. The derivation we present uses the Fourier expansion of  $p$ :  $e(p(x)) = \sum_{\alpha \in \{0,1\}^n} \hat{p}_\alpha \cdot \chi_\alpha(x)$ , where  $\chi_\alpha(x) := e(\sum_i \alpha_i \cdot x_i)$  is the inner product between  $\alpha$  and  $x$ . We have:

$$\begin{aligned} &\mathbb{E}_{U,Y,Y'} e [p(U + Y) + p(U + Y')] \\ &= \mathbb{E}_{U,Y,Y'} \left[ \left( \sum_{\alpha \in \{0,1\}^n} \hat{p}_\alpha \cdot \chi_\alpha(U + Y) \right) \left( \sum_{\beta \in \{0,1\}^n} \hat{p}_\beta \cdot \chi_\beta(U + Y') \right) \right] \\ &= \mathbb{E}_{U,Y,Y'} \left[ \sum_{\alpha,\beta} \hat{p}_\alpha \cdot \hat{p}_\beta \cdot \chi_{\alpha+\beta}(U) \cdot \chi_\alpha(Y) \cdot \chi_\beta(Y') \right] \\ &\quad \text{Here we use standard manipulations, e.g. } \chi_\alpha(U + Y) = \chi_\alpha(U) \cdot \chi_\alpha(Y). \\ &= \mathbb{E}_{Y,Y'} \left[ \sum_{\gamma=\alpha+\beta} \hat{p}_\gamma^2 \cdot \chi_\gamma(Y) \cdot \chi_\gamma(Y') \right] \\ &\quad \text{Because } \mathbb{E}_U e [\chi_{\alpha+\beta}(U)] \text{ equals 0 when } \alpha \neq \beta, \text{ and 1 otherwise.} \\ &= \mathbb{E}_U e [p(U)]^2 + \sum_{\gamma \neq 0} \hat{p}_\gamma^2 \cdot (E_Y [\chi_\gamma(Y)])^2 \\ &\quad \text{Because } \hat{p}_0 = \mathbb{E}_U e [p(U)], \text{ and } \chi_0(Y) \equiv 1. \\ &\leq \mathbb{E}_U e [p(U)]^2 + \epsilon_1^2 \cdot \sum_{\gamma \neq 0} \hat{p}_\gamma^2 \\ &\quad \text{Because } Y \text{ } \epsilon_1\text{-fools degree-1 polynomials such as } \sum_i \gamma_i \cdot Y_i. \\ &\leq \mathbb{E}_U e [p(U)]^2 + \epsilon_1^2. \\ &\quad \text{Because } \sum_{\gamma \neq 0} \hat{p}_\gamma^2 \leq \sum_\gamma \hat{p}_\gamma^2 = 1 \text{ by Parseval's identity.} \quad \square \end{aligned}$$

We now move to the case of bias far from 0. This case was solved both in [BV] and more compactly in [Lov]. We present a stripped-down version of the solution in [Lov] which is sufficient for our purposes and achieves slightly better parameters.

**Lemma 4** (Fooling polynomials with bias far from 0). *Let  $W$  be a distribution that  $\epsilon_d$ -fools degree- $d$  polynomials. Let  $p$  be a polynomial of degree  $d + 1$ . Then*

$$|\mathbb{E}_W e[p(W)] - \mathbb{E}_U e[p(U)]| \leq \frac{\epsilon_d}{|\mathbb{E}_U e[p(U)]|}.$$

*Proof of Lemma 4.* We have the following derivation

$$\begin{aligned} & |\mathbb{E}_W e[p(W)] - \mathbb{E}_U e[p(U)]| \cdot |\mathbb{E}_U e[p(U)]| \\ &= |\mathbb{E}_{W,U'} e[p(W) + p(U')] - \mathbb{E}_{U,U'} e[p(U) + p(U')]| \\ &= |\mathbb{E}_{W,U'} e[p(W) + p(W + U')] - \mathbb{E}_{U,U'} e[p(U) + p(U + U')]| \\ &\quad \text{Because } U' \text{ is uniformly distributed over } \{0, 1\}^n. \\ &\leq \mathbb{E}_{U'} |\mathbb{E}_W e[p(W) + p(W + U')] - \mathbb{E}_U e[p(U) + p(U + U')]| \leq \epsilon_d, \end{aligned}$$

where in the last inequality we use that for every fixed  $U'$  the polynomial  $p(x) + p(x + U')$  has degree  $d$  in  $x$ , though  $p$  has degree  $d + 1$ , and that  $W$   $\epsilon_d$ -fools degree- $d$  polynomials.  $\square$

To conclude, we work out the parameters for the proof of Theorem 2.

*Proof of Theorem 2.* Let  $\epsilon_d$  be the error for polynomials of degree  $d$ , i.e. the maximum over polynomials  $p$  of degree  $d$  of the quantity

$$|\mathbb{E}_W e[p(W)] - \mathbb{E}_U e[p(U)]|.$$

We claim that for every  $d > 0$  we have

$$\epsilon_{d+1} \leq 4 \cdot \sqrt{\epsilon_d}. \quad (\star)$$

Indeed, let  $p$  be an arbitrary polynomial of degree  $d + 1$ . If  $|\mathbb{E}_U e[p(U)]| \leq \sqrt{\epsilon_d}$  we have by Lemma 3 that

$$|\mathbb{E}_W e[p(W)] - \mathbb{E}_U e[p(U)]| \leq 2 \cdot \sqrt{\epsilon_d} + \epsilon + \sqrt{\epsilon_d} \leq 4 \cdot \sqrt{\epsilon_d},$$

which confirms  $(\star)$  in this case. Otherwise, if  $|\mathbb{E}_U e[p(U)]| \geq \sqrt{\epsilon_d}$  we have by Lemma 4 that

$$|\mathbb{E}_W e[p(W)] - \mathbb{E}_U e[p(U)]| \leq \frac{\epsilon_d}{\sqrt{\epsilon_d}} = \sqrt{\epsilon_d} \leq 4 \cdot \sqrt{\epsilon_d},$$

which again confirms  $(\star)$  in this case.

Finally, from  $(\star)$  it follows that

$$\epsilon_d \leq 4^{\sum_{i=0}^{d-2} 2^{-i}} \cdot \epsilon^{1/2^{d-1}} \leq 16 \cdot \epsilon^{1/2^{d-1}}$$

for every  $d$ , and thus the theorem is proved.  $\square$

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## References

- [AGHP] N. Alon, O. Goldreich, J. Håstad, and R. Peralta. Simple constructions of almost  $k$ -wise independent random variables. *Random Structures & Algorithms*, 3(3):289–304, 1992. 1, 3
- [Bog] A. Bogdanov. Pseudorandom generators for low degree polynomials. In *STOC'05: Proceedings of the 37th Annual ACM Symposium on Theory of Computing*, pages 21–30, New York, 2005. ACM. 1
- [BV] A. Bogdanov and E. Viola. Pseudorandom bits for polynomials. In *48th Annual Symposium on Foundations of Computer Science*. IEEE, Oct. 2007. 1, 2, 3, 5
- [GT1] B. Green and T. Tao. An inverse theorem for the Gowers  $U^3$  norm, 2005. arXiv.org:math/0503014. 2
- [GT2] B. Green and T. Tao. The distribution of polynomials over finite fields, with applications to the Gowers norms, 2007. arXiv:0711.3191v1. 2
- [Lov] S. Lovett. Pseudorandom generators for low degree polynomials. In *Proceedings of the 40th Annual ACM Symposium on the Theory of Computing (STOC)*, Victoria, Canada, 17–20 May 2008. 2, 3, 5
- [LMS] S. Lovett, R. Meshulam, and A. Samorodnitsky. Inverse Conjecture for the Gowers norm is false. In *Proceedings of the 40th Annual ACM Symposium on the Theory of Computing (STOC)*, Victoria, Canada, 17–20 May 2008. 2
- [LVW] M. Luby, B. Velickovic, and A. Wigderson. Deterministic Approximate Counting of Depth-2 Circuits. In *Proceedings of the 2nd Israeli Symposium on Theoretical Computer Science (ISTCS)*, pages 18–24, 1993. 1
- [NN] J. Naor and M. Naor. Small-bias probability spaces: efficient constructions and applications. In *Proceedings of the 22nd Annual ACM Symposium on the Theory of Computing*, pages 213–223, 1990. 1, 2, 3
- [Sam] A. Samorodnitsky. Low-degree tests at large distances. In *Proceedings of the 39th Annual ACM Symposium on Theory of Computing, San Diego, CA USA*, 2007. 2
- [VW] E. Viola and A. Wigderson. Norms, XOR lemmas, and lower bounds for GF(2) polynomials and multiparty protocols. In *Proceedings of the 22nd Annual Conference on Computational Complexity*. IEEE, June 13–16 2007. To appear in the journal *Theory of Computing*. 3