# Poly-logarithmic independence fools $A C^{0}$ circuits 

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#### Abstract

We prove that poly-sized $A C^{0}$ circuits cannot distinguish a poly-logarithmically independent distribution from the uniform one. This settles the 1990 conjecture by Linial and Nisan [LN90]. The only prior progress on the problem was by Bazzi [Baz07], who showed that $O\left(\log ^{2} n\right)$ independent distributions fool poly-size DNF formulas. Razborov [Raz08] has later given a much simpler proof for Bazzi's theorem.


## I. Introduction

## A. The problem

The main problem we consider is on the power of $r$-independence to fool $A C^{0}$ circuits. For a distribution $\mu$ on the finite support $\{0,1\}^{n}$, we denote by $\mathbf{E}_{\mu}[F]$ the expected value of $F$ on inputs drawn according to $\mu$. For an event $X$, we denote by $\mathbf{P}_{\mu}[X]$ its probability under $\mu$. When the distribution under consideration is the uniform distribution on $\{0,1\}^{n}$, we suppress the subscript and let $\mathbf{E}[F]$ denote the expected value of $F$, and $\mathbf{P}[X]$ the probability of $X$. A distribution $\mu$ is said to $\varepsilon$-fool a function $F$ if

$$
\left|\mathbf{E}_{\mu}[F]-\mathbf{E}[F]\right|<\varepsilon .
$$

The distribution $\mu$ on $\{0,1\}^{n}$ is $r$-independent if every restriction of $\mu$ to $r$ coordinates is uniform on $\{0,1\}^{r} . A C^{0}$ circuits are circuits with $A N D, O R$ and NOT gates, where the fan-in of the gates is unbounded. The depth of a circuit $C$ is the maximum number of $A N D / O R$ gates between an input of $C$ and its output. The problem we study is

Main Problem. How large does $r=r(m, d, \varepsilon)$ have to be in order for every r-independent distribution $\mu$ on $\{0,1\}^{n}$ to $\varepsilon$-fool every function $F$ that is computed by a depth-d $A C^{0}$ circuit of size $\leq m$ ?

Prior to our work, Bazzi [Baz07], [Baz09], in a proof that was later simplified by Razborov [Raz08], showed that a poly-logarithmic $r$ is sufficient for $d=2$ (i.e. when the $F$ 's are DNF or CNF formulas):

Theorem 1. [Baz07], [Raz08] r(m,2, $\varepsilon$ )-independence $\varepsilon$-fools depth- 2 circuits, where

$$
r(m, 2, \varepsilon)=O\left(\log ^{2} \frac{m}{\varepsilon}\right)
$$

Our main result is that for any constant $d, r(m, d, \varepsilon)$ is poly-logarithmic in $m / \varepsilon$. This gives a huge class of distributions that look random to $A C^{0}$ circuits. For example, as in [Baz09], it implies that linear codes with poly-logarithmic seed length can be PRGs for $A C^{0}$.

## B. Main results

We prove the following:
Main Theorem. Let $s \geq \log m$ be any parameter. Let $F$ be a boolean function computed by a circuit of depth $d$ and size $m$. Let $\mu$ be an r-independent distribution where

$$
r \geq r(s, d)=3 \cdot 60^{d+3} \cdot(\log m)^{(d+1)(d+3)} \cdot s^{d(d+3)}
$$

then

$$
\left|\mathbf{E}_{\mu}[F]-\mathbf{E}[F]\right|<\varepsilon(s, d),
$$

where $\varepsilon(s, d)=0.82^{s} \cdot(15 m)$.
In particular, by taking $s=5 \log \frac{15 m}{\varepsilon}$, we get the following:

Corollary 2. $r(m, d, \varepsilon)$-independence $\varepsilon$-fools depth- $d$ $A C^{0}$ circuits of size $m$, where

$$
\begin{aligned}
& r(m, d, \varepsilon)= \\
& 3 \cdot 60^{d+3} \cdot(\log m)^{(d+1)(d+3)} \cdot\left(5 \log \frac{15 m}{\varepsilon}\right)^{d(d+3)}= \\
& \left(\log \frac{m}{\varepsilon}\right)^{O\left(d^{2}\right)}
\end{aligned}
$$

Note that by choosing $\varepsilon=2^{-n^{\delta}}$ for a small $\delta=$ $\delta(d)$, one sees that polynomial independence fools $A C^{0}$ circuits up to an exponentially small error. The results carry some meaning for super-constant $d$ 's up to $d=$ $\tilde{O}(\sqrt{\log m})$.

The original conjecture by [LN90] was that for constant $\varepsilon, r(m, d, \varepsilon)=O\left((\log m)^{d-1}\right)$. Thus our results
leave a gap between $O(d)$ and $O\left(d^{2}\right)$ in the exponent. We believe that the conjecture is true with $O(d)$.

As in [Baz09], we can use [AGM02] to show that almost $r$-independent distributions also fool $A C^{0}$. A distribution $\mu$ is called a $(\delta, r)$-approximation, if $\mu$ is $\delta$ close to uniform for every $r$ (distinct) coordinates. Thus an $r$-independent distribution is a $(0, r)$-approximation. We use the following theorem.

Theorem 3. [AGM02] Let $\mu$ be a $(\delta, r)$-approximation over $n$ variables. Then $\mu$ is $n^{r} \cdot \delta$-close to an $r$ independent distribution $\mu^{\prime}$.

Theorem 3 and the Main Theorem immediately imply:

Corollary 4. For every boolean circuit $F$ of depth $d$ and size $m$ over $n$ variables and any $s \geq \log m$, let

$$
r \geq r(s, d)=3 \cdot 60^{d+3} \cdot(\log m)^{(d+1)(d+3)} \cdot s^{d(d+3)}
$$

Then for any $(\delta, r)$-approximation $\mu$,

$$
\left|\mathbf{E}_{\mu}[F]-\mathbf{E}[F]\right|<\varepsilon(s, d)+n^{r} \cdot \delta=0.82^{s} \cdot(15 m)+n^{r} \cdot \delta .
$$

Corollary 4 in turn implies:
Corollary 5. $(\delta, r(m, d, \varepsilon))$-approximations $\varepsilon$-fool depth-d $A C^{0}$ circuits of size $m$, where

$$
\begin{aligned}
& r(m, d, \varepsilon)= \\
& 4 \cdot 60^{d+3} \cdot(\log m)^{(d+1)(d+3)} \cdot\left(5 \log \frac{15 m}{\varepsilon}\right)^{d(d+3)}
\end{aligned}
$$

as long as $\delta$ is sufficiently small so that

$$
\frac{\varepsilon}{\delta}>2 n^{r(m, d, \varepsilon)} .
$$

## C. Techniques and proof outline

As in [Baz09], our strategy is to approximate $F$ with low degree polynomials over $\mathbb{R}$. The reason being that degree- $r$ polynomials are completely fooled by $r$ independence.
Proposition 6. Let $f: \mathbb{R}^{n} \rightarrow \mathbb{R}$ be a degree-r polynomial, and let $\mu$ be an r-independent distribution.
Then $f$ is completely fooled by $\mu$ :

$$
\mathbf{E}_{\mu}[f]=\mathbf{E}[f] .
$$

Proposition 6 is true by linearity of expectation, since every term of $f$ is a product of $\leq r$ variables, whose distribution is uniform under $\mu$.

In our construction we combine two types of approximations of $A C^{0}$ circuits by low degree polynomials over $\mathbb{R}$. The first one is combinatorial in the spirit of [Raz87], [Smo87], [BRS91], [Tar93] (for a comprehensive survey on polynomials in circuit complexity see e.g.
[Bei93]). These approximating polynomials agree with $F$ on all but a small fraction of inputs. Thus for such a polynomial $f, \mathbf{P}[f=F]$ is very close to 1 . While essentially using the same construction as [BRS91], [Tar93], utilizing tools from [VV85], we repeat the construction from scratch in Lemma 8, since we want to reason about details of the construction. We believe that any construction in this spirit would fit in our proof.

The second approximation is based on Fourier analysis and uses [LMN93] where it is shown that any $A C^{0}$ function $G$ can be approximated by a low degree polynomial $g$ so that the $\ell_{2}$ norm $\|G-g\|_{2}^{2}$ is small. There is no guarantee, however, that $g$ agrees with $G$ on any input (most likely, it doesn't).

We use an approximation $f$ of $F$ of the first type as the starting point of our construction. Thus $\mathbf{P}[f \neq F]$ is very small. If we knew that $\|F-f\|_{2}^{2}$ is small we would be done by a simple argument similar to one that appeared in [Baz09]. Unfortunately, there are no guarantees, that $f$ is close to $F$ on average, since $f$ may deviate wildly on points where $f \neq F$ (in fact, it is likely untrue that $\|F-f\|_{2}^{2}$ is small).

Our key insight is that in the construction of $f$, the indicator function $\mathcal{E}$ of where $f$ fails to agree with $F$ is an $A C^{0}$ function itself. Thus $\mathcal{E}=1$ whenever $f \neq F$, and $\mathbf{P}[\mathcal{E}=1]$ is very small (since $f=F$ most of the time). We then use a low-degree approximation $\tilde{\mathcal{E}}$ of $\mathcal{E}$ of the second type so that $\|\tilde{\mathcal{E}}-\mathcal{E}\|_{2}^{2}$ is very small. We then take $f^{\prime}=f \cdot(1-\tilde{\mathcal{E}})$. The idea is that $1-\tilde{\mathcal{E}} \approx$ $1-\mathcal{E}$ will kill the values of $f$ where it misbehaves (and thus $\mathcal{E}=1$ ), while leaving other values (where $\mathcal{E}=0$ ) almost unchanged. Note that the values where $f=0$ remain completely unchanged, and thus $f^{\prime}$ is a semiexact approximation of $F$. In Lemma 10 we show that $\left\|F-f^{\prime}\right\|_{2}^{2}$ is small. We choose $f^{\prime}$ to "almost agree" with $F$ against both the uniform distribution and the distribution $\mu$, a property we use to finish the proof.

It should be noted that while an inductive proof on the depth $d$ of $F$ is a natural approach to the problem, a non-inductive construction appears to yield much better parameters for the theorem.

## D. Paper organization

The rest of the paper is organized as follows. In Section II we repeat the [LMN93] theorem on lowdegree $\ell_{2}$-approximation, and develop low-degree approximation tools that are used in the proof of the main theorem. In Section III we prove the main theorem.

## Acknowledgments

I am grateful to Louay Bazzi, Marek Karpinski, and Alex Samorodnitsky for stimulating discussions at
the early stages of this work. I would like to thank Nina Balcan, Henry Cohn, Ehud Friedgut, Nick Harvey, Toni Pitassi, Madhu Sudan, and many others for the many useful comments on the earlier versions of this manuscript.

## II. SEmi-EXACT Approximations and ERror FUNCTIONS

We will make use of the [LMN93] bound:
Lemma 7. ([LMN93]) If $F:\{0,1\}^{n} \rightarrow\{0,1\}$ is a boolean function computable by a depth-d circuit of size $m$, then for every there is a degree t polynomial $\tilde{f}$ with
$\|F-\tilde{f}\|_{2}^{2}=\frac{1}{2^{n}} \sum_{x \in\{0,1\}^{n}}|F(x)-\tilde{f}(x)|^{2} \leq 2 m \cdot 2^{-t^{1 / d} / 20}$.
As a first step, we prove the following lemma:
Lemma 8. Let $\nu$ be any probability distribution on $\{0,1\}^{n}$. For a circuit of depth $d$ and size $m$ computing a function $F$, for any $s$, there is a degree $r=(s \cdot \log m)^{d}$ polynomial $f$ and a boolean function $\mathcal{E}_{\nu}$ computable by a circuit of depth $<d+3$ and size $O\left(m^{2} r\right)$ such that

- $\mathbf{P}_{\nu}\left[\mathcal{E}_{\nu}(x)=1\right]<0.82^{s} m$, and
- whenever $\mathcal{E}_{\nu}(x)=0, f(x)=F(x)$.

Thus $\mathcal{E}_{\nu}$ tells us whether there is a mistake in $f$, and the weight of the mistakes as measured by $\nu$ is very small. Note that when there is a mistake, $f$ does not have to be equal to $1-F$, and can actually be quite large in absolute value. The functions $f$ and $\mathcal{E}_{\nu}$ are illustrated on Fig. 1 (b) and (c).

Proof: We construct the polynomial $f$ by induction on $d$, and show that w.h.p. $f=F$. The function $\mathcal{E}_{\nu}$ follows from the construction. Note that we do not know anything about the measure $\nu$ and thus cannot give an explicit construction for $f$. Instead, we will construct a distribution on polynomials $f$ that succeeds with high probability on any given input. Thus the distribution is expected to have a low error with respect to $\nu$, which implies that there is a specific $f$ that has a small error with respect to $\nu$.

We will show how to make a step with an $A N D$ gate. Since the whole construction is symmetric with respect to 0 and 1, the step also holds for an $O R$ gate. Let

$$
F=G_{1} \wedge G_{2} \wedge \ldots \wedge G_{k}
$$

where $k<m$. For convenience, let us assume that $k=$ $2^{\ell}$ is a power of 2 . We take a collection of

$$
t:=s \log m
$$

random subsets of $\{1,2, \ldots, k\}$ where each element is included with probability $p$ independently of the
others: at least $s$ subsets for each of the $p=$ $2^{-1}, 2^{-2}, \ldots, 2^{-\ell}=1 / k$. Denote the sets by $S_{1}, \ldots, S_{t}$ - we ignore empty sets. In addition, we make sure to include $\{1, \ldots, k\}$ as one of the sets. Let $g_{1}, \ldots, g_{k}$ be the approximating polynomials for $G_{1}, \ldots, G_{k}$. We set

$$
f:=\prod_{i=1}^{t}\left(\sum_{j \in S_{i}} g_{j}-\left|S_{i}\right|+1\right)
$$

By the induction assumption, the degrees of $g_{j}$ are $d^{\prime} \leq$ $(s \cdot \log m)^{d-1}$, hence the degree of $f$ is bounded by $t \cdot d^{\prime} \leq(s \cdot \log m)^{d}$. Next we bound the error $\mathbf{P}[f \neq F]$. It consists of two terms:

$$
\begin{align*}
& \mathbf{P}_{\nu}[f \neq F] \leq \mathbf{P}_{\nu}\left[g_{j} \neq G_{j} \text { for some } j\right]+ \\
& \quad \mathbf{P}_{\nu}\left[\prod_{i=1}^{t}\left(\sum_{j \in S_{i}} G_{j}-\left|S_{i}\right|+1\right) \neq \prod_{j=1}^{k} G_{j}\right] . \tag{1}
\end{align*}
$$

In other words, to make a mistake, either one of the inputs has to be dirty, or the approximating function for the AND has to make a mistake. We will focus on the second term. The first term is bounded by union bound. We fix a vector of specific values $G_{1}(x), \ldots, G_{k}(x)$, and calculate the probability of an error over the possible choices of the random sets $S_{i}$.

Note that if all the $G_{j}(x)$ 's are 1 then the value of $F(x)=1$ is calculated correctly with probability 1. Suppose that $F(x)=0$ (and thus at least one of the $G_{j}(x)$ 's is 0 ). Let $1 \leq z \leq k$ be the number of zeros among $G_{1}(x), \ldots, G_{k}(x)$, and $\alpha$ be such that $2^{\alpha} \leq z<2^{\alpha+1}$. Let $S$ be a random set as above with $p=2^{-\alpha-1}$. Our formula will work correctly if $S$ hits exactly one 0 among the $z$ zeros of $G_{1}(x), \ldots, G_{k}(x)$. The probability of this event is exactly

$$
z \cdot p \cdot(1-p)^{z-1} \geq \frac{1}{2} \cdot(1-p)^{1 / p-1}>\frac{1}{2 e}>0.18
$$

Hence the probability of being wrong after $s$ such sets is bounded by $0.82^{s}$. Since this is true for any value of $x$, we can find a collection of sets $S_{i}$ such that the probability of error as measured by $\nu$ is at most $0.82^{s}$ according to $\nu$. By making the same probabilistic argument at every node and applying union bound, we get that the condition "if the inputs are correct then the output is correct" is satisfied by all nodes except with probability $<0.82^{s} \mathrm{~m}$. Thus the error of the polynomial is $<0.82^{s} \mathrm{~m}$.

Finally, if we know the sets $S_{i}$ at every node, it is easy to check whether there is a mistake by checking that no set contains exactly one 0 , thus yielding the depth $<(d+3)$ function $\mathcal{E}_{\nu}$. The blowup in size is at most $O(m r)$ since at each node we take a disjunction


Figure 1. An illustration depicting the various functions constructed in the paper. The boolean cube $\{0,1\}^{n}$ is represented by the $x$ axis in these figures. Graph (a) depicts an $A C^{0}$ boolean function $F$; graphs (b) and (c) show the functions $f$ and $\mathcal{E}_{\nu}$ from Lemma 8: the support of $\mathcal{E}_{\nu}$ is small w.r.t. to the measure $\nu, f=F$ outside the support of $\mathcal{E}_{\nu}$ while behaving wildly inside; graphs (d)-(f) depict the functions from Lemma 10, in particular we see on graph (f) that $F^{\prime}=0$ implies $f^{\prime}=0$
over all the possible pairs of $\left(S_{i}, G_{j} \in S_{i}\right)$ of whether $G_{j}$ is the only 0 in the set $S_{i}$.
Recall that $t=s \log m$ as in the proof of Lemma 8; we have:

Proposition 9. In Lemma 8, for $s \geq \log m,\|f\|_{\infty}<$ $(2 m)^{\operatorname{deg}(f)-2}=(2 m)^{t^{d}-2}$.

Proof: We prove the statement by induction on $d$. For $d=1, \operatorname{deg}(f)=t$ and the functions $g_{j}$ are just $0 / 1$-valued literals. Since $\left|S_{i}\right| \leq m$ for all $i$, we have for every $x$ :

$$
|f(x)|=\prod_{i=1}^{t}\left|\sum_{j \in S_{i}} g_{j}-\left|S_{i}\right|+1\right| \leq m^{t}<(2 m)^{t-2}
$$

For the step, assuming the statement is true for $d-1 \geq$ 1 , we have

$$
\begin{aligned}
\|f\|_{\infty} \leq & \prod_{i=1}^{t}\left|\sum_{j \in S_{i}}\left\|g_{j}\right\|_{\infty}+\left|S_{i}\right|-1\right| \\
\prod_{i=1}^{t}\left|m \cdot(2 m)^{t^{d-1}-2}+m\right| & < \\
\left((2 m)^{t^{d-1}-1}\right)^{t} & <(2 m)^{t^{d}-2}
\end{aligned}
$$

Applying results from [LMN93] we can now take any shallow function $F$ and modify it a little bit, so that the modified function would have a good one-sidederror approximation. The ingredients of the proof are illustrated on Fig. 1 (d)-(f).

Lemma 10. Let $F$ be computed by a circuit of depth $d$ and size $m$. Let $s_{1}, s_{2}$ be two parameters with $s_{1} \geq$ $\log m$. Let $\mu$ be any probability distribution on $\{0,1\}^{n}$. Set

$$
\nu:=\frac{1}{2}\left(\mu+U_{\{0,1\}^{n}}\right) .
$$

Let $\mathcal{E}_{\nu}$ be the function from Lemma 8 with $s=s_{1}$. Set $F^{\prime}=F \vee \mathcal{E}_{\nu}$. Then there is a polynomial $f^{\prime}$ of degree $r_{f} \leq\left(s_{1} \cdot \log m\right)^{d}+s_{2}$, such that

- $\mathbf{P}\left[F \neq F^{\prime}\right]<2 \cdot 0.82^{s_{1}} m$;
- $\mathbf{P}_{\mu}\left[F \neq F^{\prime}\right]<2 \cdot 0.82^{s_{1}} \mathrm{~m}$;
- $\left\|F^{\prime}-f^{\prime}\right\|_{2}^{2}<0.82^{s_{1}} \cdot(4 m)+$

$$
2^{2.9\left(s_{1} \cdot \log m\right)^{d} \log m-s_{2}^{1 /(d+3)} / 20} \text {, and }
$$

- $f^{\prime}(x)=0$ whenever $F^{\prime}(x)=0$.

Proof: The first two properties follow from Lemma 8 directly, since
$\mathbf{P}\left[\mathcal{E}_{\nu}=1\right], \mathbf{P}_{\mu}\left[\mathcal{E}_{\nu}=1\right] \leq 2 \cdot \mathbf{P}_{\nu}\left[\mathcal{E}_{\nu}=1\right]<2 \cdot 0.82^{s_{1}} m$.

Let $f$ be the approximating polynomial for $F$ from that lemma, so that $F=F^{\prime}=f$ whenever $\mathcal{E}_{\nu}=0$, and thus $f=0$ whenever $F^{\prime}=0$. By Proposition 9 we have

$$
\|f\|_{\infty}<(2 m)^{\left(s_{1} \cdot \log m\right)^{d}}<2^{1 \cdot 4\left(s_{1} \cdot \log m\right)^{d} \log m}
$$

We let $\tilde{\mathcal{E}_{\nu}}$ be the low degree approximation of $\mathcal{E}_{\nu}$ of degree $s_{2}$. By [LMN93] (Lemma 7), we have

$$
\left\|\mathcal{E}_{\nu}-\tilde{\mathcal{E}_{\nu}}\right\|_{2}^{2}<O\left(m^{3}\right) \cdot 2^{-s_{2}^{1 /(d+3)} / 20}
$$

Let

$$
f^{\prime}:=f \cdot\left(1-\tilde{\mathcal{E}_{\nu}}\right)
$$

Then $f^{\prime}=0$ whenever $F^{\prime}=0$. It remains to estimate $\left\|F^{\prime}-f\right\|_{2}^{2}$ :

$$
\begin{array}{r}
\left\|F^{\prime}-f^{\prime}\right\|_{2}^{2} \leq \\
2 \cdot\left\|F^{\prime}-f \cdot\left(1-\mathcal{E}_{\nu}\right)\right\|_{2}^{2}+2 \cdot\left\|f \cdot\left(1-\mathcal{E}_{\nu}\right)-f^{\prime}\right\|_{2}^{2}= \\
2 \cdot\left\|\mathcal{E}_{\nu}\right\|_{2}^{2}+2 \cdot\left\|f \cdot\left(\mathcal{E}_{\nu}-\widetilde{\mathcal{E}_{\nu}}\right)\right\|_{2}^{2} \leq \\
2 \cdot \mathbf{P}\left[\mathcal{E}_{\nu}=1\right]+2 \cdot\|f\|_{\infty}^{2} \cdot\left\|\mathcal{E}_{\nu}-\widetilde{\mathcal{E}_{\nu}}\right\|_{2}^{2}< \\
0.82^{s_{1}}(4 m)+2^{2.9\left(s_{1} \cdot \log m\right)^{d} \log m-s_{2}^{1 /(d+3)} / 20},
\end{array}
$$

which completes the proof.

## III. MAIN THEOREM

As in Bazzi's proof ([Baz07], Lemma 3.3) we can now use Lemma 10 to prove the following:
Lemma 11. For every boolean circuit $F$ of depth $d$ and size $m$ and any $s \geq \log m$, and for any probability distribution $\mu$ on $\{0,1\}$ there is a boolean function $F^{\prime}$ and a polynomial $f_{l}^{\prime}$ of degree less than

$$
r_{f}=3 \cdot 60^{d+3} \cdot(\log m)^{(d+1)(d+3)} \cdot s^{d(d+3)}
$$

such that

- $\mathbf{P}_{\mu}\left[F \neq F^{\prime}\right]<\varepsilon(s, d) / 3$,
- $\mathbf{P}\left[F \neq F^{\prime}\right]<\varepsilon(s, d) / 3$,
- $f_{l}^{\prime} \leq F^{\prime}$ on $\{0,1\}^{n}$, and
- $\mathbf{E}\left[F^{\prime}-f_{l}^{\prime}\right]<\varepsilon(s, d) / 3$,
where $\varepsilon(s, d)=0.82^{s} \cdot(15 m)$.
Proof: Let $F^{\prime}$ be the boolean function and let $f^{\prime}$ be the polynomial from Lemma 10 with $s_{1}=s$ and $s_{2} \approx 60^{d+3} \cdot(\log m)^{(d+1)(d+3)} \cdot s^{d(d+3)}$. The first two properties follow directly from the lemma. Set

$$
f_{l}^{\prime}:=1-\left(1-f^{\prime}\right)^{2}
$$

It is clear that $f_{l}^{\prime} \leq 1$ and moreover $f_{l}^{\prime}=0$ whenever $F^{\prime}=0$, hence $f_{l}^{\prime} \leq F^{\prime}$. Finally, $F^{\prime}(x)-f_{l}^{\prime}(x)=0$ when $F^{\prime}(x)=0$, and is equal to

$$
F^{\prime}(x)-f_{l}^{\prime}(x)=\left(1-f^{\prime}(x)\right)^{2}=\left(F^{\prime}(x)-f^{\prime}(x)\right)^{2}
$$

when $F^{\prime}(x)=1$, thus

$$
\mathbf{E}\left[F^{\prime}-f_{l}^{\prime}\right] \leq\left\|F^{\prime}-f^{\prime}\right\|_{2}^{2}<0.82^{s} \cdot(5 m)=\varepsilon(s, d) / 3
$$

by Lemma 10. To finish the proof we note that the degree of $f_{l}^{\prime}$ is bounded by

$$
2 \cdot\left(\left(s_{1} \cdot \log m\right)^{d}+s_{2}\right)<2.5 \cdot s_{2}<r_{f}
$$

Lemma 11 implies the following:
Lemma 12. Let $s \geq \log m$ be any parameter. Let $F$ be a boolean function computed by a circuit of depth $d$ and size $m$. Let $\mu$ be an r-independent distribution where

$$
r \geq 3 \cdot 60^{d+3} \cdot(\log m)^{(d+1)(d+3)} \cdot s^{d(d+3)}
$$

then

$$
\mathbf{E}_{\mu}[F]>\mathbf{E}[F]-\varepsilon(s, d)
$$

where $\varepsilon(s, d)=0.82^{s} \cdot(15 m)$.
Proof: Let $F^{\prime}$ be the boolean function and let $f_{l}^{\prime}$ be the polynomial from Lemma 11. The degree of $f_{l}^{\prime}$ is $<r$. We use the fact that since $\mu$ is $r$-independent, $\mathbf{E}_{\mu}\left[f_{l}^{\prime}\right]=\mathbf{E}\left[f_{l}^{\prime}\right]$ (see Proposition 6 above):

$$
\begin{gathered}
\mathbf{E}_{\mu}[F] \geq \mathbf{E}_{\mu}\left[F^{\prime}\right]-\mathbf{P}_{\mu}\left[F \neq F^{\prime}\right] \geq \\
\mathbf{E}_{\mu}\left[f_{l}^{\prime}\right]-\varepsilon(s, d) / 3=\mathbf{E}\left[f_{l}^{\prime}\right]-\varepsilon(s, d) / 3= \\
\mathbf{E}\left[F^{\prime}\right]-\mathbf{E}\left[F^{\prime}-f_{l}^{\prime}\right]-\varepsilon(s, d) / 3>\mathbf{E}\left[F^{\prime}\right]-2 \varepsilon(s, d) / 3 \geq \\
\mathbf{E}[F]-\mathbf{P}\left[F^{\prime} \neq F\right]-2 \varepsilon(s, d) / 3>\mathbf{E}[F]-\varepsilon(s, d) .
\end{gathered}
$$

The dual inequality to Lemma 12 follows immediately by applying the lemma to the negation $\bar{F}=1-F$ of $F$. We have $\mathbf{E}_{\mu}[\bar{F}]>\mathbf{E}[\bar{F}]-\varepsilon(s, d)$, and thus
$\mathbf{E}_{\mu}[F]=1-\mathbf{E}_{\mu}[\bar{F}]<1-\mathbf{E}[\bar{F}]+\varepsilon(s, d)=\mathbf{E}[F]+\varepsilon(s, d)$.
Together, these two statements yield the main theorem:
Main Theorem. Let $s \geq \log m$ be any parameter. Let $F$ be a boolean function computed by a circuit of depth $d$ and size $m$. Let $\mu$ be an r-independent distribution where

$$
r \geq r(s, d)=3 \cdot 60^{d+3} \cdot(\log m)^{(d+1)(d+3)} \cdot s^{d(d+3)},
$$

then

$$
\left|\mathbf{E}_{\mu}[F]-\mathbf{E}[F]\right|<\varepsilon(s, d),
$$

where $\varepsilon(s, d)=0.82^{s} \cdot(15 m)$.

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