# Pseudorandom generators hard for $k$-DNF resolution and polynomial calculus resolution 

By Alexander A. Razborov


#### Abstract

A pseudorandom generator $G_{n}:\{0,1\}^{n} \rightarrow\{0,1\}^{m}$ is hard for a propositional proof system $P$ if (roughly speaking) $P$ cannot efficiently prove the statement $G_{n}\left(x_{1}, \ldots, x_{n}\right) \neq b$ for any string $b \in\{0,1\}^{m}$. We present a function $\left(m \geq 2^{n^{\Omega(1)}}\right)$ generator which is hard for $\operatorname{Res}(\varepsilon \log n)$; here $\operatorname{Res}(k)$ is the propositional proof system that extends Resolution by allowing $k$-DNFs instead of clauses.

As a direct consequence of this result, we show that whenever $t \geq n^{2}$, every $\operatorname{Res}(\varepsilon \log t)$ proof of the principle $\neg \operatorname{Circuit}_{t}\left(f_{n}\right)$ (asserting that the circuit size of a Boolean function $f_{n}$ in $n$ variables is greater than $t$ ) must have size $\exp \left(t^{\Omega(1)}\right)$. In particular, $\operatorname{Res}(\log \log N)\left(N \sim 2^{n}\right.$ is the overall number of propositional variables) does not possess efficient proofs of NP $\nsubseteq$ $\mathbf{P} /$ poly. Similar results hold also for the system PCR (the natural common extension of Polynomial Calculus and Resolution) when the characteristic of the ground field is different from 2.

As a byproduct, we also improve on the small restriction switching lemma due to Segerlind, Buss and Impagliazzo by removing a square root from the final bound. This in particular implies that the (moderately) weak pigeonhole principle $\mathrm{PHP}_{n}^{2 n}$ is hard for $\operatorname{Res}(\varepsilon \log n / \log \log n)$.


## 1. Introduction

Propositional proof complexity is an area of study that has seen a rapid development over the last decade. It plays as important a role in the theory of feasible proofs as the role played by the complexity of Boolean circuits in the theory of efficient computations. And in most cases the basic question of propositional proof complexity boils down to this. Given a mathematical statement encoded as a propositional tautology $\phi$ and a class of admissible mathematical proofs formalized as a propositional proof system $P$, what is the minimal possible complexity of a $P$-proof of $\phi$ ?

[^0]1.1. General overview. For most "interesting" propositional proof systems $P$, one can easily define the accompanying (nonuniform) complexity class $\mathcal{C}_{P}$ typically consisting of functions computable by lines allowed in efficient $P$-proofs. This correspondence leads to a classification of propositional proof systems that is somewhat imprecise and potentially (and hopefully) timedependent but nonetheless very instructive. Namely, we call a propositional proof system $P$ weak if we (currently) know how to prove super-polynomial lower bounds for the accompanying circuit class $\mathcal{C}_{P}$ and strong otherwise.

There is a steady progress in studying the complexity of proofs in weak proof systems surveyed, e.g., in [Urq95], [Kra95], [Raz96], [BP01], [Pud98], [Raz02].

For strong proof systems the current situation is by far more miserable. Although there are no rigorous results along these lines (and, moreover, this feeling is not universal - see, e.g., [Kra04]), the empirical evidence strongly suggests that lower bounds for a proof system $P$ are even harder to attain than computational lower bounds for the companion class $\mathcal{C}_{P}$. Therefore, with our current understanding, we cannot apparently hope to show lower bounds for systems like Frege or Extended Frege without first making a major breakthrough in complexity theory.

A more accessible task that (in the author's opinion) is almost as interesting would be to show at least that proof complexity lower bounds are at most as hard as comparable problems in the computational world. Let us (informally) identify this task as

PROVING LOWER BOUNDS FOR STRONG PROOF SYSTEMS $P$ LIKE Frege or Extended Frege modulo any hardness AssumpTION IN THE PURELY COMPUTATIONAL WORLD, HOWEVER STRONG BUT STILL NATURAL AND BELIEVABLE

This task will be referred to as proving conditional lower bounds (for the proof system $P$ ). It should be remarked in this respect that $\mathbf{N P} \neq$ co-NP implies lower bounds for any propositional proof system whatsoever. Therefore, purely computational above refers to the demand that the assumption itself should speak only about computations and should not attempt to restrict the power of proofs even in a disguised form.

One extremely exciting and, in a sense, model approach to this task was gradually developed in the sequence of papers [Raz95b], [BPR97], [Kra97a], [Pud97] and finally became known as the Efficient Interpolation Property (EIP in what follows). EIP was shown to be true for some weak proof systems and it was also remarked that for every proof system (be it weak or strong) EIP implies conditional lower bounds. Unfortunately, it turned out rather soon
[KP98], [BPR00] that neither Frege nor Extended Frege have Efficient Interpolation modulo (somewhat ironically) hardness assumptions of the same sort that are needed to prove conditional lower bounds for proof systems with EIP.

This omnipresent hardness assumption is nothing other than the existence of pseudorandom generators (arbitrary or specific), which also turns out to be the main primitive of the modern cryptography. After the (apparent) failure of the efficient interpolation approach, it was independently proposed in [Kra01a], [ABSRW04] to employ pseudorandom generators for proving conditional lower bounds in a more direct manner. On the conceptual level, a mapping $G_{n}:\{0,1\}^{n} \rightarrow\{0,1\}^{m}$, where $m>n$, is called hard for a propositional proof system $P$ if $P$ cannot efficiently prove the (properly encoded) statement $G_{n}\left(x_{1}, \ldots, x_{n}\right) \neq b$ for $a n y^{1}$ string $b \in\{0,1\}^{m}$. Since $m>n$, for at least half of all $b$ 's, this statement is a tautology. Therefore, conditional lower bounds for a proof system $P$ follow from the following task:

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PROVE THAT FOR A REASONABLE CLASS OF MAPPINGS
G}\mp@subsup{|}{n}{}:{0,1\mp@subsup{}}{}{n}->{0,1\mp@subsup{}}{}{m}\mathrm{ WITH m}>n,\mathrm{ THEIR HARDNESS
IN ANY REASONABLE COMPUTATIONAL OR COMBINATO-
RIAL SENSE IMPLIES HARDNESS FOR P.
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This will be referred to as the generator approach (to conditional lower bounds for $P$ ).

The generator approach certainly does not work for arbitrary mappings $G_{n}$ believed to be pseudorandom generators in the standard sense of [Yao82], and specific counterexamples almost immediately follow from the results in [KP98] on the limitations of EIP. On the positive side, it was observed in [ABSRW04] that for proof systems $P$ with EIP, there is an easy and general way of converting any pseudorandom generator that is computationally hard (in the standard sense) into a pseudorandom generator that is hard for $P$. No such general transformation is known for a single nontrivial proof system without EIP.

Thus, it is vital for the generator approach that we somehow restrict the class of mappings for which one hopes to trade computational hardness for proof complexity hardness. Along these lines, [ABSRW04] specifically proposed to consider the class of Nisan-Wigderson generators. (Concrete results from that paper as well as from later improvements [AR03], [Kra04] will be reviewed after we are done with this general overview.) Some arguments advocating this choice (as opposed to other classical cryptographic constructions)

[^1]were presented in [ABSRW04] and in the introduction to [Raz04]: the principle expressing hardness of NW-generators is tightly related to such familiar personages in proof complexity as the pigeonhole principle and Tseitin tautologies.

Elaborating on these arguments, we can further remark that the NWgenerator is in a sense the quintessence of the very idea of "local consistency." Namely, the information contained in the output bits of a NW-generator is "local" (in the sense that every output bit depends only on a "small" subset of input bits that are "nearly independent" for different output bits), and it is locally consistent to such an extent that no interesting conclusion about the global behaviour of the generator can be obtained by an "easy" analysis of this local knowledge. This simple methodology is behind a great deal of lower bounds existing in proof complexity for weak proof systems, and it is also behind the efficient interpolation property.

This methodology also has a very clean mathematical meaning. A (nonexistent) falsifying assignment to the tautology $\phi$ corresponds to a manifold with given local properties. The proof system $P$ tries to argue that no such manifold may exist using tools at its disposal. And we (lower bounds provers) try to fool it by feeding into the potential proof something that looks like a (nonexisting) manifold to such an extent that $P$ cannot discern the difference. See [Raz98] for (apparently) the cleanest implementation of this intuitive scheme.

Anyway, the moving forces that make the Nisan-Wigderson generator work in the computational world are of so general a nature that we are ready to spell out the formal conjectures that the generator approach always works for NisanWigderson generators. More specifically (assuming that the constructions are based on combinatorial designs with the same parameters as in the seminal paper [NW94]),
Conjecture 1: Any NW-generator based on any poly-time function that is hard on average for $\mathbf{N C}^{\mathbf{1}}$ /poly is hard for the Frege proof system.
Conjecture 2: Any NW-generator based on any function in NP $\cap$ co-NP that is hard on average for $\mathbf{P} /$ poly (e.g., $B\left(f^{-1}(r)\right)$, where $f(x)$ is any one-way permutation and $B(x)$ its hard-core bit) is hard for Extended Frege.

The suggestion to use Nisan-Wigderson generators for lower bounds in proof complexity has been recently reiterated in [Kra04]. That paper also proposes a paradigm similar in spirit to the construction from [Gol11] in the context of computational complexity: hardness of the resulting mapping should depend on the randomness of the base functions rather than their complexity. So far all known results on the hardness of NW-generators for weak proof systems have not directly appealed to the randomness and used instead specific combinatorial properties of the base functions. But of course it remains to be
seen yet which of the two paradigms (if any) will turn out more fruitful in the long run.

The task of proving lower bounds (even conditional) for strong proof systems is, in the author's opinion, extremely interesting and well justified in its own right (whereas the popular motivation that this should be regarded as an intermediate step in approaching the NP vs. co-NP problem looks, again in the author's opinion, more of a speculation). One venerable way, however, to make this study even more interesting is to look at the proof complexity of statements whose validity is also not known and whose importance stretches well beyond any proof-theoretical studies.

To that end, [Raz95a] proposed ${ }^{2}$ to study the proof complexity of the principle $\neg \operatorname{Circuit}_{t}\left(f_{n}\right)$ expressing that the circuit size of the Boolean function $f_{n}$ in $n$ variables, given as its truth-table, is lower bounded by $t=t(n)$. (Thus, e.g., NP $\nsubseteq \mathbf{P} /$ poly is essentially equivalent to the validity of $\neg \operatorname{Circuit}_{t(n)}\left(\operatorname{Sat}_{n}\right)$ for some $t(n) \geq n^{\omega(1)}$.) [Raz95a] put forward the thesis (so far not refuted) that all existing proofs of lower bounds for restricted classes of circuits and for explicit functions translate to Extended Frege proofs (often to much weaker proof systems) of size $2^{O(n)}$. This makes the question of the efficient provability of the original principle $\neg \operatorname{Circuit}_{t}\left(f_{n}\right)$ for general circuits even more intriguing.

The connection of this question to the generator approach above is the same as in the context of Natural Proofs [RR97]. Namely, if we have a function pseudorandom generator $G_{n}:\{0,1\}^{t_{0}} \rightarrow\{0,1\}^{2^{n}}$ that is hard for a proof system $P$, such that the associated predicate $G(x)_{y}\left(x \in\{0,1\}^{t_{0}}, y \in\{0,1\}^{n}\right)$ can be computed by a size $t$ circuit, then for every fixed seed $x \in\{0,1\}^{t_{0}}$, the Boolean function with the truth-table $G(x)$ is also computed by a size $t$ circuit. Since for any given $f_{n}$ the system $P$ cannot efficiently refute that $f_{n}$ is different even from the functions $G(x)$ in the image of the generator $G$, it is not capable of efficient proofs of $\neg \operatorname{Circuit}_{t}\left(f_{n}\right)$. In plain words,

TO SHOW THAT $P$ DOES NOT HAVE EFFICIENT PROOFS OF THE FORMULA $\neg \operatorname{Circuit}_{t}\left(f_{n}\right)$, IT SUFFICES TO DESIGN A SUFFICIENTLY CONSTRUCTIVE PSEUDORANDOM GENERATOR HARD FOR $P$ AND SUCH THAT THE NUMBER OF OUTPUT BITS, AS A FUNCTION OF THE NUMBER OF INPUT BITS, IS AS LARGE AS POSSIBLE.

The larger number of output bits we can manage, the smaller are the parameters $t_{0}, t$ (relatively to $n$ ) and the stronger is the result. In particular, in

[^2]order to conclude the efficient unprovability of NP $\nsubseteq \mathbf{P} /$ poly (or, for that matter, that any function $f_{n}$ is not in $\mathbf{P} /$ poly), one needs a generator that stretches $n$ bits to $2^{n^{\varepsilon}}$ bits; such generators are commonly known as function pseudorandom generators.

The tight connection between pseudorandom generators and the tautologies $\neg \operatorname{Circuit}_{t}\left(f_{n}\right)$ has also been fruitfully exploited from more structural point of view. We already remarked above that NP $\neq$ co-NP implies the existence of hard tautologies for any propositional proof system, but this does not give any clue as to what these hard tautologies actually are. [IKW02, Th. 35] proved that under the assumption NEXP $\subset \mathbf{P} /$ poly, it is the specific tautologies $\neg \operatorname{Circuit}_{t}\left(f_{n}\right)$ (for any $f_{n}$ whatsoever) that are hard for any proof system. R. Impagliazzo (see a footnote in [Kra04, §1]), and independently M. Alekhnovich, recently observed that the same conclusion holds under the assumption BPP $\nsubseteq \mathbf{N P}$ (stronger than NP $\neq$ co-NP). Although none of these two assumptions looks particularly plausible (and none of them is "purely computational" either), this still serves as another indication of the "distinguished" character of the tautologies $\neg \operatorname{Circuit}_{t}\left(f_{n}\right)$.

The project of proving lower bounds for stronger and stronger classes of circuits until we arrive at $\mathbf{P} \neq \mathbf{N P}$ has met a solid obstacle in the form of Natural Proofs [RR97]. The project of proving lower bounds for stronger and stronger proof systems until we arrive at the (Extended) Frege proof system is restricted by the empirical observation that this task is even harder than the previous one. Given this gloomy background, fulfilling the generator approach for strong proof systems is certainly not an easy task, and the progress in this direction is much slower than originally hoped. However, it seems that we currently do not know of any general reasons (like those for the two previous projects formulated above) making us suspect that this task is unfeasible and/or requires an entirely different view of the subject.
1.2. Previous results and our contributions. In all known partial results along the generator approach, Nisan-Wigderson generators play the central role. Hardness results for generators of this kind are determined by combinatorial properties of the underlying set system, conditions imposed on the base functions, and by the specific way their computation is encoded as a propositional tautology. Disregarding for the moment all these technical issues, on the conceptual level [ABSRW04] proved that the NW-generator is hard for Resolution but only when the complexity is measured by width. Another result from [ABSRW04] says that the Nisan generator (that is, the partial case of the NW-generator in which all base functions must be linear mod 2) is hard for Polynomial Calculus ( PC in what follows) over fields $\mathbb{F}$ with char $(\mathbb{F}) \neq 2$.

The latter result was further extended and generalized in [AR03] to show that the Nisan-Wigderson generator is hard for PC over any field.

Results for resolution width and polynomial calculus degree are applicable to function generators stretching $n$ bits to $2^{n^{\varepsilon}}$ bits. As long as the size complexity measures are concerned, [ABSRW04] exhibited hard Nisan-Wigderson generators for the system PCR (that is, a natural common extension of Polynomial Calculus and Resolution) but only when $m \leq o\left(n^{2}\right)$. This poor input/output ratio hindered their potential application to proving that $\mathbf{N P} \nsubseteq \mathbf{P} /$ poly is hard even for Resolution, and this was established by somewhat different methods in [Raz04], [Raz04].

A prominent general way for enhancing the I/O performance of pseudorandom generators in proof complexity has been recently proposed in [Kra04]. Like the classical constructions in computational complexity [Yao82], [GGM86], it is very natural to try to achieve this goal by composing the given generator with itself. Unfortunately, it is far from clear whether hardness in the context of proof complexity is preserved under composition. [Kra04] proposed a way around this difficulty by showing that it is indeed the case if hardness is replaced by a stronger notion of $s$-iterability (the latter, in turn, being a variant of a similar notion of freeness earlier introduced in [Kra01b]). As a first application of this approach, [Kra04] showed that one particular construction of the Nisan generator from [ABSRW04] can be iterated with itself once, thus giving a pseudorandom generator with $m=n^{3-\varepsilon}$ output bits that is hard for Resolution.

In the current paper we continue this line of research. Let $\operatorname{Res}(k)$ be the propositional proof system that extends Resolution by allowing as its lines arbitrary $k$-DNFs. Our first main result (Theorem 2.7) exhibits Nisan generators that are hard for $\operatorname{Res}(k)$ and stretch $n$ input bits to as many as $n^{(\varepsilon \log n) / k}$ output bits. A relatively easy modification of this argument for $k=1$ shows that this generator (from $n$ to $n^{\varepsilon \log n}$ bits) is hard not only for Resolution but also for its extension PCR (Theorem 2.18). ${ }^{3}$ These results were proved independently of [Kra04]; the proof method uses the resolution width/PC degree bounds from [ABSRW04] cited above in combination with the machinery from the recent paper [SBI04] based upon the so-called small restriction switching lemma. In order to bring these two together, we also introduce a special kind of random restrictions specifically tailored to deal with Nisan generators.

Then, using a very simple reduction, we show that our generator is not only hard for $\operatorname{Res}(k)$ and PCR but it is in fact $\exp \left(n^{\Omega(1)}\right)$-iterable for these systems (Theorems 2.10 and 2.19). According to the paradigm from [Kra04], this implies that if we compose this generator with itself as many as $\exp \left(n^{\Omega(1)}\right)$

[^3]times, the resulting mapping will still be hard for $\operatorname{Res}(k) /$ PCR. Applying in particular the classical GGM-construction [GGM86], in this way we get a function generator $G_{n}:\{0,1\}^{n} \longrightarrow\{0,1\}^{2^{n^{\varepsilon}}}$ that is hard for $\operatorname{Res}(\varepsilon \log n), \varepsilon>0$ a sufficiently small constant (Theorem 2.12), and for PCR. As we discussed in Section 1.1, this implies that neither $\operatorname{Res}(\log \log N)$ (where $N \sim 2^{n}$ is the overall number of propositional variables) nor PCR possess efficient proofs of $\mathbf{N P} \nsubseteq \mathbf{P} /$ poly (Theorems 2.13 and 2.20 ), which are the first results of this kind for any propositional proof system that (most likely) does not have the Efficient Interpolation Property.

In the course of this work we were able to get a quadratic (in $k$ ) improvement of the small restriction switching lemma from [SBI04] based on Janson inequality (Lemma 4.4). This in particular implies that the moderately weak pigeonhole principle $\mathrm{PHP}_{n}^{2 n}$ is exponentially hard for $\operatorname{Res}(k)$ when $k \leq(\varepsilon \log n) / \log \log n$ (Theorem 2.15), as opposed to $k \leq \varepsilon \sqrt{\log n / \log \log n}$ in [SBI04] that in turn was an improvement on the previous papers [BT88] ( $k=1$, i.e., resolution) and [ABE02] $(k=2)$.

Finally we prove a miscellaneous result about the complexity of Nisan generators themselves that indicates the sensitivity of their proof complexity behaviour with respect to the choice of encoding. Namely, we show how to modify the (natural and reasonable) encoding of the Nisan generator used in the rest of the paper in such a way that this generator becomes hard for PCR even in the functional case $m=2^{n^{\varepsilon}}$ (Theorem 2.21). This encoding, however, is in a sense very bad: it does not seem to allow any reduction to $\neg \operatorname{Circuit}_{t}\left(f_{n}\right)$, and the proof method apparently fails already for $\operatorname{Res}(2)$.

## 2. Definitions and statements of our results

In this section we typically confine ourselves to defining only those notions that are needed for stating our main results. Auxiliary concepts needed for their proofs will normally appear in respective places.

Let $x$ be a propositional variable, i.e., a variable that ranges over the set $\{0,1\}$. A literal of $x$ is either $x$ (denoted sometimes as $x^{1}$ ) or $\bar{x}$ (denoted sometimes as $x^{0}$ ). A clause $[$ term $]$ is either a constant 0 or 1 (corresponding to FALSE and TRUE, respectively) or a disjunction [conjunction, respectively] of literals. A $C N F[D N F]$ is a conjunction of clauses [disjunction of terms], often specified as the set of all participating clauses [terms, respectively]. Accordingly, a clause/term/CNF/DNF is a sub-clause/sub-term/sub-CNF/sub-DNF of another clause/term/CNF/DNF if every literal/literal/clause/term appearing in the first, appears also in the second. A clause/term/CNF/DNF is monotone if it does not contain occurrences of negated literals $\bar{x}$.

For a Boolean function $f$ [a propositional formula $F]$, let $\operatorname{Vars}(f)[\operatorname{Vars}(F)]$ be the set of its essential variables [the set of variables explicitly occurring in
$F$, respectively]. The width of a clause $C$ [of a term $t$ ] is defined as $w(C) \stackrel{\text { def }}{=}$ $|\operatorname{Vars}(C)|[w(t) \stackrel{\text { def }}{=}|\operatorname{Vars}(t)|$, respectively]. A $k-C N F[k$-DNF] is a CNF [DNF] in which all clauses [terms, respectively] are of width at most $k .|F|$ is the number of terms in a DNF $F$.

An assignment to a Boolean function $f$ [a propositional formula $F$ ] is a mapping $\alpha: \operatorname{Vars}(f) \rightarrow\{0,1\}[\alpha: \operatorname{Vars}(F) \rightarrow\{0,1\}$, respectively]. A restriction of $f$ [of $F$ ] that, depending on the context, will be sometimes called $a$ partial assignment is a mapping $\rho: \operatorname{Vars}(f) \rightarrow\{0,1, \star\}[\rho: \operatorname{Vars}(F) \rightarrow\{0,1, \star\}$, respectively]. We let $\sup (\rho) \stackrel{\text { def }}{=} \rho^{-1}(\{0,1\})$ denote the set of assigned variables. The restriction of a function $f$ [of a formula $F]$ by $\rho$, denoted $\left.f\right|_{\rho}\left[\left.F\right|_{\rho}\right]$, is the Boolean function [propositional formula] obtained from $f$ [from $F$, respectively] by setting the value of each $x_{i} \in \sup (\rho)$ to $\rho\left(x_{i}\right)$ and leaving each $x_{i} \notin \sup (\rho)$ unassigned. In the case of propositional formulas we assume as usual that simplifications are performed only when a sub-formula has become explicitly constant. A variable substitution of variables in $V_{1}$ by variables in $V_{2}$ is a mapping $\rho$ that takes variables in $V_{1}$ to either propositional constants or literals of variables in $V_{2}$. (Thus, restrictions are viewed as a special case of variable substitutions with $V_{2}=V_{1}$.) Variable substitutions $\rho$ of variables in Vars $(f)$ or $\operatorname{Vars}(F)$ also naturally act on the Boolean function $f$ or propositional formula $F$ and, as before, we denote the result of this action by $\left.f\right|_{\rho},\left.F\right|_{\rho}$.

For an integer $n$, let $[n] \stackrel{\text { def }}{=}\{1,2, \ldots, n\}$. Whenever we use probabilistic methods, random variables will always appear in the bold face, including deterministic parameters they depend on, if any. We extensively use the $\Omega$-notation (customary in complexity theory) that is opposite to the ordinary $O$-notation. For example, given two functions $f, g$ with values in the set of nonnegative reals, $f \geq \Omega(g)$ means that there exists an absolute constant $\varepsilon>0$ such that $f \geq \varepsilon g$ for any specification of the parameters occurring in $f, g$.

Definition 2.1 ([Kra01a], [ABSRW04]). Let $m>n, C$ be a Boolean circuit with $n$ inputs $x_{1}, \ldots, x_{n}$ and $m$ outputs, and let $b \in\{0,1\}^{m}$ be an arbitrary Boolean vector. For every computational gate $v$ of the circuit $C$, we introduce a special extension variable $y_{v}$, and when $v$ is the $j$ th input gate, we identify $y_{v}$ with the corresponding propositional variable $x_{j}$. Let $\operatorname{Vars}_{C} \stackrel{\text { def }}{=}\left\{x_{1}, \ldots, x_{n}\right\} \cup$ $\left\{y_{v} \mid v\right.$ a computational gate of $\left.C\right\} .{ }^{4}$

By $\tau(C, b)$ we denote the CNF in the variables $\operatorname{Vars}_{C}$ that expresses the fact " $C\left(x_{1}, \ldots, x_{n}\right)=b$ " and consists of the following clauses:

[^4](1) $y_{v_{1}}^{\bar{\varepsilon}_{1}} \vee \cdots \vee y_{v_{d}}^{\bar{\varepsilon}_{d}} \vee y_{v}^{\pi\left(\varepsilon_{1}, \ldots, \varepsilon_{d}\right)}$ whenever $v:=\pi\left(v_{1}, \ldots, v_{d}\right)$ is an instruction of $C$ of arity $d$ and $\varepsilon \in\{0,1\}^{d}$ is an arbitrary vector;
(2) $y_{v_{i}}^{b_{i}}$ when $v_{i}$ is the $i$ th output gate of $C, i \in[m]$.

Remark 1. In this paper we will be mostly interested in the case when the circuit $C$ is linear; that is, it consists only of linear instructions $v:=v_{1} \oplus v_{2}$ of arity 2 . Moreover, the circuits considered will be read-once circuits in which variables are added one at a time. However, even with these restrictions there remains an ambiguity about the order in which the variables are introduced, and this does become important for weak proof systems like those considered in this paper. The next definition makes this formal.

For $A$ an $m \times n 0-1$ matrix and $i \in[m]$, let $J_{i}(A) \stackrel{\text { def }}{=}\left\{j \in[n] \mid a_{i j}=1\right\}$ and $X_{i}(A) \stackrel{\text { def }}{=}\left\{x_{j} \mid a_{i j}=1\right\}$ be the corresponding set of propositional variables. The set system $J_{i}(A)$ provides an alternative (and often more convenient) way to represent the matrix $A$.

Definition 2.2. An ordering $\leq$ of an $m \times n 0-1$ matrix $A$ is a tuple ( $\leq_{1}$, $\ldots, \leq_{m}$ ), where $\leq_{i}$ is a linear ordering of the set $J_{i}(A)$. Given any ordering $\leq$, let $C_{1}, \ldots, C_{m}$ be the (single-output) circuits naturally computing the parity functions $\bigoplus\left\{x_{j} \mid x_{j} \in X_{i}(A)\right\}$ according to this ordering. That is,

- gates of $C_{i}$ have the form $v_{\Sigma}^{i}$, where $\Sigma$ is a nonempty $\leq_{i}$-initial segment of $J_{i}(A)$;
- when $x_{j}$ is the minimal element of $J_{i}(A), v_{\left\{x_{j}\right\}}^{i}$ is the input gate $x_{j}$;
- instructions have the form $v_{\Sigma \cup\left\{x_{j}\right\}}^{i}:=v_{\Sigma}^{i} \oplus x_{j}$, where $\Sigma$ is a proper initial segment of $J_{i}(A)$ and $j$ is the minimal element in $J_{i}(A) \backslash \Sigma$;
- $v_{J_{i}(A)}^{i}$ is the output gate.

Denote by $C_{A, \leq}$ the $m$-output linear circuit which is a disconnected union of $C_{1}, \ldots, C_{m}$. (That is, these circuits do not have any gates in common except for input gates.) Let $\operatorname{Vars} \leq(A) \stackrel{\text { def }}{=} \operatorname{Vars}_{C_{A, \leq}}$ and $\tau \leq(A, b) \stackrel{\text { def }}{=} \tau\left(C_{A, \leq}, b\right)$.

Definition 2.3. For a set of rows $I \subseteq[m]$ in the matrix $A$, we define its boundary $\partial_{A}(I)$ as the set of all $j \in[n]$ (called boundary elements) such that $\left\{a_{i j} \mid i \in I\right\}$ contains exactly one 1 . We say that $A$ is an $(r, d)$-lossless expander if

$$
\begin{equation*}
\forall I \subseteq[m]\left(|I| \leq r \Rightarrow \sum_{i \in I}\left|J_{i}(A)\right|-\left|\partial_{A}(I)\right| \leq d \cdot|I|\right) . \tag{1}
\end{equation*}
$$

Remark 2. Another useful way to interpret the expansion property (1) in this definition is to say that rows in $I$ have at most $d$ nonboundary elements on average. In the regular case (that is, when all $J_{i}(A)$ have the same cardinality $s),(r, d)$-lossless expanders are exactly $(r, s, c)$-expanders in the terminology of
[ABSRW04] for $c=s-d$. We introduce this new definition mainly to stress that it is in fact the difference between $s$ and $c$ that matters, and for most applications, we need not know $s$ (and in fact even do not need regularity). Also, the "ordinary" lossless expanders recently constructed in [CRVW02] correspond to the case $d=\varepsilon s$ for a small constant $\varepsilon>0$, whence our choice of the name.

Despite recent progress, no explicit constructions of $(r, d)$-lossless expanders with the parameters sufficient for our purposes are currently known. Fortunately, we will be satisfied with the following simple nonconstructive bound.

Definition 2.4. $\boldsymbol{A}_{\boldsymbol{m}, \boldsymbol{n}}$ is a random $m \times n 0-1$ matrix in which all entries are independent and $\mathbf{P}\left[\boldsymbol{a}_{i j}=1\right]=n^{-2 / 3}$.

Theorem 2.5. Let $m \leq 2^{n^{\varepsilon}}$, where $\varepsilon>0$ is a sufficiently small constant. Then $\boldsymbol{A}_{\boldsymbol{m}, \boldsymbol{n}}$ is an $\left(n^{\Omega(1)}, O\left(\frac{\log m}{\log n}\right)\right)$-lossless expander with probability $\geq 1$ $O(1 / m)$.

The proof of Theorem 2.5 is straightforward; it is deferred to Section 7.
Definition 2.6. $\operatorname{Res}(k)$ is the propositional proof system whose lines are $k$-DNFs, whose only axioms are $\ell \vee \bar{\ell}(\ell$ a literal) and whose inference rules are given below. ( $F, G$ are $k$-DNFs, $1 \leq w \leq k$ and $\ell, \ell_{i}$ are literals.)

$$
\begin{array}{ll}
\frac{F}{F \vee \ell}(\text { WEAKENING }), & \frac{F \vee \ell_{1} \cdots F \vee \ell_{w}}{F \vee\left(\bigwedge_{i=1}^{w} \ell_{i}\right)} \text { (AND-INTRODUCTION), } \\
\frac{F \vee\left(\bigwedge_{i=1}^{w} \ell_{i}\right)}{F \vee \ell_{i}} \text { (AND-ELIMINATION), } & \frac{F \vee\left(\bigwedge_{i=1}^{w} \ell_{i}\right) \quad G \vee \bigvee_{i=1}^{w} \bar{\ell}_{i}}{F \vee G} \text { (CUT). }
\end{array}
$$

A $\operatorname{Res}(k)$ refutation of a set of $k$-DNFs is a $\operatorname{Res}(k)$ proof of 0 from this set. In particular, a $\operatorname{Res}(k)$ refutation of $a \operatorname{CNF} \tau$ is a $\operatorname{Res}(k)$ refutation of the set of clauses $\tau$ consists of.

We define the size of a $\operatorname{Res}(k)$ proof as the number of lines in it. Note that since in this paper we deal exclusively with lower bounds, this makes our results only stronger (as opposed to measuring the complexity by the number of bits).

Note that any variable substitution $\rho$ takes a $\operatorname{Res}(k)$ refutation of a CNF $\tau$ into a $\operatorname{Res}(k)$ refutation of $\left.\tau\right|_{\rho}$. This in particular implies that the minimal size of a $\operatorname{Res}(k)$ refutation of $\tau$ is at least the same size for $\left.\tau\right|_{\rho}$.

Finally, the case $k=1$ corresponds to Resolution: this system operates with clauses and has the only inference rule

$$
\frac{F \vee x \quad G \vee \bar{x}}{F \vee G}
$$

called resolution rule. (The weakening rule does not change the power of Resolution and is usually, often implicitly, also assumed for convenience.) The width of a resolution proof is the maximal width of a clause occurring in this proof.

Now we are ready to formulate our main result for Nisan generators. For the sake of definiteness, all algorithms in this paper are assumed to be base 2 .

Theorem 2.7. Let $A$ be an $m \times n(r, d)$-lossless expander, and assume that

$$
\begin{equation*}
\min _{i \in[m]}\left|J_{i}(A)\right| \geq C d(k+\log m) \tag{2}
\end{equation*}
$$

for a sufficiently large constant $C>0$. Let $\leq$ be an arbitrary ordering of $A$ and $b \in\{0,1\}^{m}$ be an arbitrary vector. Then every $\operatorname{Res}(k)$ refutation of $\tau_{\leq}(A, b)$ must be of size $\geq \exp \left(r / 2^{O(k d)}\right)$.

Combining Theorems 2.7 and 2.5 we get, in particular,
Corollary 2.8. Let $m, n, k$ be parameters such that

$$
\begin{equation*}
m \leq n^{(\varepsilon \log n) / k} \tag{3}
\end{equation*}
$$

where $\varepsilon>0$ is a sufficiently small constant. Then with probability $1-O(1 / m)$, the following holds. For every ordering $\leq$ of $\boldsymbol{A}_{\boldsymbol{m}, \boldsymbol{n}}$ and every $b \in\{0,1\}^{m}$, every $\operatorname{Res}(k)$-refutation of $\tau_{\leq}\left(\boldsymbol{A}_{\boldsymbol{m}, \boldsymbol{n}}, b\right)$ must have size $\geq \exp \left(n^{\Omega(1)}\right)$.

Definition 2.9 ([Kra04]). For an $n$-input $m$-output circuit $C$ and a vector $z_{1}, \ldots, z_{m}$ of propositional variables, let $\tau_{C}\left(x_{1}, \ldots, x_{n}, \vec{y}, z_{1}, \ldots, z_{m}\right)$ be defined in the same way as $\tau(C, b)$ (see Definition 2.1) with the difference that the "output axioms" $y_{v_{i}}^{b_{i}}$ get replaced by the two clauses $y_{v_{i}} \vee \bar{z}_{i}, \bar{y}_{v_{i}} \vee z_{i}$. (Thus, $\tau(C, b)$ is the same as $\tau_{C}(\vec{x}, \vec{y}, b)$.)

For a proof system $P$ and an integer $S$, the circuit $C$ is $S$-iterable in $P$ if for every CNF of the form

$$
\begin{equation*}
\bigwedge_{\nu=1}^{H} \tau_{C}\left(x_{1}^{(\nu)}, \ldots, x_{n}^{(\nu)}, \vec{y}^{(\nu)}, q_{1}^{(\nu)}, \ldots, q_{m}^{(\nu)}\right), \tag{4}
\end{equation*}
$$

every refutation of this CNF in the system $P$ must have size $\geq S$. Here $x_{j}^{(\nu)}, \vec{y}^{(\nu)}$ are pairwise disjoint tuples of variables, and every $q_{i}^{(\nu)}$ is either a Boolean constant or belongs to the set $\left\{x_{j}^{(\mu)} \mid j \in[n], \mu<\nu\right\}$.

Since $\tau(C, b)$ itself has the form (4) (with $H=1$ ), iterability is stronger than just hardness. As the proof of the following theorem will show, for the particular case of NW-generators, they are not very much apart from each other.

Theorem 2.10. Let $A$ be an $m \times n(r, d)$-lossless expander, and assume that

$$
\begin{equation*}
s \stackrel{\text { def }}{=} \min _{i \in[m]}\left|J_{i}(A)\right| \geq C d(k+\log m) \tag{5}
\end{equation*}
$$

for a sufficiently large constant $C>0$. Then for every ordering $\leq$ of the matrix $A$, the circuit $C_{A, \leq}$ from Definition 2.2 is $S$-iterable in in $\operatorname{Res}(k)$, where $S \geq \exp \left(\min \left\{\Omega(s / d), r / 2^{O(d)}\right\}\right)$.

As in [Kra04], this implies in particular that composing the generator $C_{A, \leq}$ with itself preserves hardness. We will note here only one particular iteration protocol corresponding to the classical construction from [GGM86].

Definition 2.11. Let $C$ be a Boolean circuit (over an arbitrary basis) with $n$ inputs and $2 n$ outputs, and let $h \geq 1$ be an integer. The circuit $C^{h}$ with $n$ inputs and $2^{h} n$ outputs is constructed as follows. For every binary string $u$ with $|u| \leq h-1$, we prepare an isomorphic copy $C_{u}$ of $C$. The inputs of $C^{h}$ are the inputs of $C_{\Lambda}$ ( $\Lambda$ the empty string), and the outputs of $C^{h}$ are the outputs of the circuits $C_{u}$ with $|u|=h-1$. Finally, for every $u$ with $|u|<h-1$, the first $n$ outputs of $C_{u}$ are identified with the inputs of $C_{u * 0}$, and the last $n$ output bits are identified with inputs of $C_{u * 1}$. (In everything else these circuits are completely independent.)

Theorem 2.12. Let $\leq$ be an arbitrary ordering of an $(2 n \times n)(r, d)$ lossless expander $A$, and assume that

$$
s \stackrel{\text { def }}{=} \min _{i \in[2 n]}\left|J_{i}(A)\right| \geq C d(k+\log n)
$$

for a sufficiently large constant $C>0$. Let $h \geq 1$ be an arbitrary integer, and let $b \in\{0,1\}^{2^{h} \cdot n}$ be an arbitrary vector. Then every $\operatorname{Res}(k)$ refutation of $\tau\left(C_{A, \leq}^{h}, b\right)$ must have size $\geq \exp \left(\min \left\{\Omega(s / d), r / 2^{O(k d)}\right\}\right)$.

For a Boolean function $f_{n}$ in $n$ variables $^{5}$ and $t \leq 2^{n}$, denote by $\operatorname{Circuit}_{t}\left(f_{n}\right)$ an $O(1)$-CNF of size $2^{O(n)}$ encoding the description of a size- $t$ fan-in 2 Boolean circuit over the standard basis $\{\neg, \wedge, \vee\}$ for computing $f_{n}$. We will recall an exact definition in Section 8; for the time being let us just observe that proving that the circuit size of $f_{n}$ is greater than $t$ is tantamount to showing that $\operatorname{Circuit}_{t}\left(f_{n}\right)$ is unsatisfiable.

[^5]Theorem 2.13. Let $f_{n}$ be any Boolean function in $n$ variables, and let $n^{2} \leq t \leq 2^{n}$. Then every $\operatorname{Res}(\varepsilon \log t)$ refutation of $\operatorname{Circuit}_{t}\left(f_{n}\right)(\varepsilon>0 a$ sufficiently small constant) must have size $\exp \left(t^{\Omega(1)}\right)$.

Let us now consider the pigeonhole principle.
Definition 2.14. Assume $m>n$. The (negation of the) onto pigeonhole principle is the unsatisfiable CNF in the variables $\left\{x_{i j} \mid i \in[m], j \in[n]\right\}$ denoted by $\neg$ onto- $\mathrm{PHP}_{n}^{m}$ that is the conjunction of the following clauses:

$$
\begin{aligned}
Q_{i} & \stackrel{\text { def }}{=} \bigvee_{j=1}^{n} x_{i j}(i \in[m]), \\
Q_{i_{1}, i_{2} ; j} & \stackrel{\text { def }}{=}\left(\bar{x}_{i_{1} j} \vee \bar{x}_{i_{2} j}\right)\left(i_{1} \neq i_{2} \in[m], j \in[n]\right), \\
Q_{j} & \stackrel{\text { def }}{=} \bigvee_{i=1}^{m} x_{i j}(j \in[n]) .
\end{aligned}
$$

(The prefix "onto" refers to the presence of the last group of axioms.)
[SBI04] proved that $\neg$ onto- $\mathrm{PHP}_{n}^{2 n}$ is exponentially hard to refute in $\operatorname{Res}(k)$ as long as $k \leq \sqrt{\log n / \log \log n}$. We improve this as follows.

Theorem 2.15. Every $\operatorname{Res}(k)$ refutation of $\neg$ onto- $\mathrm{PHP}_{n}^{2 n}$ must have size $\exp \left(n /(\log n)^{O(k)}\right)$.

This bound is exponential up to $k=(\varepsilon \log n) / \log \log n$.
Let us now recall another extension of Resolution that is of more algebraic flavour.

Definition 2.16 ([CEI96]). Let $\mathbb{F}$ be a fixed field. Polynomial Calculus (PC for short) is the proof system whose lines are polynomials $f \in \mathbb{F}\left[x_{1}, \ldots, x_{n}\right]$. (A polynomial $f$ is interpreted as the polynomial equation $f=0$.) It has polynomials $x_{i}^{2}-x_{i}(i \in[n])$ as default axioms and has two inference rules:

$$
\begin{aligned}
& \frac{f \quad g}{\alpha f+\beta g} ; \alpha, \beta \in \mathbb{F}(\text { Scalar Addition }), \\
& \frac{f}{x_{i} \cdot f}(\text { Variable Multiplication }) .
\end{aligned}
$$

Definition 2.17 ([ABSRW02]). Again let $\mathbb{F}$ be a fixed field. Polynomial Calculus with Resolution (PCR) is the proof system whose lines are polynomials from $F\left[x_{1}, \ldots, x_{n}, \bar{x}_{1}, \ldots, \bar{x}_{n}\right]$, where $\bar{x}_{1}, \ldots, \bar{x}_{n}$ are treated as new formal variables. PCR has all default axioms and inference rules of PC (including, of course, those that involve new variables $\bar{x}_{i}$ ), plus additional default axioms $x_{i}+\bar{x}_{i}=1(i \in[n])$.

For a clause $C$, denote by $\Gamma_{C}$ the monomial

$$
\begin{equation*}
\Gamma_{C} \stackrel{\text { def }}{=} \prod_{\bar{x} \in C} x \cdot \prod_{x \in C} \bar{x} \tag{6}
\end{equation*}
$$

and for a CNF $\tau$, let $\Gamma_{\tau} \stackrel{\text { def }}{=}\left\{\Gamma_{C} \mid C \in \tau\right\}$. (Note that $\tau$ is unsatisfiable if and only if the polynomials $\Gamma_{\tau}$ have no common root in $\mathbb{F}$ satisfying all default axioms of PCR.) A PCR refutation of a CNF $\tau$ is a PCR proof of the contradiction $1=0$ from $\Gamma_{\tau}$.

The degree of a PCR proof is defined as the maximal degree of a polynomial appearing in it, and its size is the number of different monomials in this proof.

Remark 3. PC and PCR are equivalent with respect to the degree measure (via the linear transformation $\bar{x}_{i} \mapsto 1-x_{i}$ ). Also note that we measure the size of PCR proofs differently from Definition 2.6; namely, by the number of monomials. We do not know if our results still hold if the size is measured by the number of lines.

All our lower bounds for $\operatorname{Res}(1)$ (= Resolution) generalize to PCR over any field $\mathbb{F}$ with $\operatorname{char}(\mathbb{F}) \neq 2$. That is,

Theorem 2.18. Let $A$ be an $m \times n(r, d)$-lossless expander, and assume that

$$
\begin{equation*}
\min _{i \in[m]}\left|J_{i}(A)\right| \geq C d \log m \tag{7}
\end{equation*}
$$

for a sufficiently large constant $C>0$. Let $\leq$ be an arbitrary ordering of $A$, $b \in\{0,1\}^{m}$ be an arbitrary vector, and $\mathbb{F}$ be an arbitrary field with $\operatorname{char}(\mathbb{F}) \neq 2$. Then every PCR refutation of $\tau_{\leq}(A, b)$ over the field $\mathbb{F}$ must be of size $\geq \exp \left(r / 2^{O(d)}\right)$.

Theorem 2.19. Let $A$ be an $m \times n(r, d)$-lossless expander such that (7) holds, and let $s \stackrel{\text { def }}{=} \min _{i \in[m]}\left|J_{i}(A)\right|$. Then for every ordering $\leq$ of the matrix $A$ and for every field $\mathbb{F}$ with $\operatorname{char}(\mathbb{F}) \neq 2$, the circuit $C_{A, \leq}$ is $S$-iterable in PCR over $F$, where $S \geq \exp \left(\min \left\{\Omega(s / d), r / 2^{O(d)}\right\}\right)$.

Let $\operatorname{Circuit}_{t}^{\oplus}\left(f_{n}\right)$ be defined in the same way as $\operatorname{Circuit}_{t}\left(f_{n}\right)$, with the exception that besides the standard connectives $\{\neg, \wedge, \vee\}$, the encoded circuit also allows PARITY gates of fan-in 2.

Theorem 2.20. Let $f_{n}$ be any Boolean function in $n$ variables, $n^{2} \leq$ $t \leq 2^{n}$, and $\mathbb{F}$ be any field with $\operatorname{char}(\mathbb{F}) \neq 2$. Then every PCR refutation of Circuit $_{t}^{\oplus}\left(f_{n}\right)$ over $\mathbb{F}$ must have size $\exp \left(t^{\Omega(1)}\right)$.

We conclude with one miscellaneous result about the hardness of Nisan generators that is specific to PCR. For these generators we still do not know
how to get more than $n^{\varepsilon \log n}$ output bits even in the case of Resolution. We now show how to make a function (that is, with $m=2^{n^{\varepsilon}}$ bits) Nisan generator hard for PCR at the expense of spoiling its encoding.

Every ordering $j_{1}, \ldots, j_{d}$ of a finite set $J$ can be also viewed as a cyclic order - that is, as an injective mapping $\alpha: J \longrightarrow S^{1}$ into the unit circle $S^{1}$ given by $\alpha\left(j_{\nu}\right)=\nu / d$. A cyclic interval is a set of the form $\alpha^{-1}(A)$, where $A \subseteq S^{1}$ is an arc.

Fix a tuple $\leq=\left(\leq_{1}, \ldots, \leq_{m}\right)$ of orderings of the sets $\left(J_{1}(A), \ldots, J_{m}(A)\right)$, and let

$$
\begin{aligned}
\operatorname{Vars}_{\leq}^{\mathrm{Cycl}}(A) \stackrel{\text { def }}{=} & \left\{x_{1}, \ldots, x_{n}\right\} \cup\left\{y_{\Delta}^{i} \mid i \in[m], \Delta \text { a cyclic interval in } J_{i}(A)\right. \\
& \text { such that } \left.\Delta=J_{i}(A) \text { or } \frac{\left|J_{i}(A)\right|}{3}-1 \leq|\Delta| \leq \frac{2\left|J_{i}(A)\right|}{3}+1\right\}
\end{aligned}
$$

Let $\tau_{\leq}^{\text {Cycl }}(A, b)$ be the 3-CNF consisting of the clauses that result from expanding those constraints of the form

$$
\begin{align*}
y_{\Delta \cup\{j\}}^{i} & \equiv y_{\Delta}^{i} \oplus x_{j}(i \in[m], j \text { adjacent to } \Delta),  \tag{8}\\
y_{\Delta_{1} \cup \Delta_{2}}^{i} & \equiv y_{\Delta_{1}}^{i} \oplus y_{\Delta_{2}}^{i}\left(i \in[m], \Delta_{1}, \Delta_{2} \text { disjoint and adjacent }\right),  \tag{9}\\
y_{J_{i}(A)}^{i} & \equiv b_{i}, \tag{10}
\end{align*}
$$

in which all variables belong to $\operatorname{Vars}_{\leq}^{\mathrm{Cycl}}(A)$. (That is, $\Delta, \Delta \cup\{j\}, \Delta_{1}, \Delta_{2}$ and $\Delta_{1} \cup \Delta_{2}$ must obey the size bound in its definition.)

Let us verify that $\tau_{\leq}^{\text {Cycl }}(A, b)$ is indeed a complete encoding of the linear system $A X=b$. (This is not immediately clear due to the presence of the constraints on $|\Delta|$. )

FACT 1. The system of $\mathbb{F}_{2}$-linear equations $A X=b$ is consistent if and only if $\tau_{\leq}^{\mathrm{Cycl}}(A, b)$ is satisfiable.

Proof. The "only if" part is obvious. For the opposite direction, we only have to show that the $x$-part of every satisfying assignment for $\tau_{\leq}^{\mathrm{Cycl}}(A, b)$ also satisfies the system $A X=b$.

For any given $i \in[m]$, fix an arbitrary partition $J_{i}(A)=\Delta_{i, 1} \dot{\cup} \Delta_{i, 2} \dot{\cup} \Delta_{i, 3}$ into three cyclic intervals of almost equal sizes: $\frac{\left|J_{i}(A)\right|}{3}-1 \leq\left|\Delta_{i \nu}\right| \leq \frac{\left|J_{i}(A)\right|}{3}+1$. Then (9) (applied twice) and (10) imply that in every satisfying assignment we have $y_{\Delta_{i, 1}}^{i} \oplus y_{\Delta_{i, 2}}^{i} \oplus y_{\Delta_{i, 3}}^{i}=b_{i}$. On the other hand, by summing up appropriate axioms (8) (with $j \in \Delta_{i \nu}$ ), we obtain $y_{\Delta_{i \nu^{\prime}}}^{i} \oplus \bigoplus_{j \in \Delta_{i \nu}} x_{j}=y_{\Delta_{i \nu} \cup \Delta_{i \nu^{\prime}}}^{i}\left(\nu^{\prime} \neq \nu\right)$ and then, using again (9), $\oplus_{j \in \Delta_{i \nu}} x_{j}=y_{\Delta_{i \nu}}^{i}$ for every $\nu \in[3]$. The statement follows.

THEOREM 2.21. Let $A$ be an $m \times n(r, d)$-lossless expander such that $\min _{i \in[m]}\left|J_{i}(A)\right| \geq C d \log m$ for a sufficiently large constant $C>0, \leq$ be an arbitrary ordering of $A, b \in\{0,1\}^{m}$ be an arbitrary vector, and $\mathbb{F}$ any field with $\operatorname{char}(\mathbb{F}) \neq 2$. Then every PCR refutation of $\tau_{\leq}^{\mathrm{Cycl}}(A, b)$ over the field $\mathbb{F}$ must have size $\geq \exp \left(r / d^{O(1)}\right)$.

The bound of this theorem is as good as we might hope. The encoding itself, however, is not very useful: it does not correspond to any circuit and, moreover, the proof of this theorem seems to completely break apart for $\operatorname{Res}(2)$.

## 3. Preliminaries

In this section we begin the proof of Theorem 2.7 by presenting some known results in the form adapted to our purposes. Recall several definitions from [ABSRW04].

Definition 3.1. A Boolean function $f$ is $\ell$-robust if every restriction $\rho$ such that $\left.f\right|_{\rho}=$ const satisfies $|\sup (\rho)| \geq \ell$.

Definition 3.2. Let $A$ be an $m \times n 0-1$ matrix. For every Boolean function $f$ with the property $\exists i \in[m]\left(\operatorname{Vars}(f) \subseteq X_{i}(A)\right)$, we introduce a new extension variable $y_{f}$. Let $\operatorname{Vars}(A)$ be the set of all these variables.

Given Boolean functions $\vec{g}=\left(g_{1}, \ldots, g_{m}\right)$ such that $\operatorname{Vars}\left(g_{i}\right) \subseteq X_{i}(A)$, we denote by $\tau(A, \vec{g})$ the CNF in the variables $\operatorname{Vars}(A)$ that consists of those clauses $y_{f_{1}}^{\varepsilon_{1}} \vee \cdots \vee y_{f_{w}}^{\varepsilon_{w}}$ for which there exists $i \in[m]$ such that

$$
\operatorname{Vars}\left(f_{1}\right) \cup \cdots \cup \operatorname{Vars}\left(f_{w}\right) \subseteq X_{i}(A)
$$

and

$$
g_{i} \leq f_{1}^{\varepsilon_{1}} \vee \cdots \vee f_{w}^{\varepsilon_{w}}
$$

We now have the following (minor) generalization of [ABSRW04, Th. 3.1].
Theorem 3.3. Let $A$ be an $(r, d)$-lossless expander of size $m \times n$, and let $g_{1}, \ldots, g_{m}$ be $\ell$-robust functions with $\operatorname{Vars}\left(g_{i}\right) \subseteq X_{i}(A)$, where $\ell \geq d+1$. Then every resolution refutation of $\tau(A, \vec{g})$ must have width $>\frac{r(\ell-d)}{2 \ell}$.

Proof. The only difference from [ABSRW04, Th. 3.1] is that we now use $(r, d)$-lossless expanders instead of $(r, s, c)$-expanders with $s-c=d$. The proof goes the same way, and we only remark on the changes to be made to its text. All these changes pertain to the proof of [ABSRW04, Claim 3.3].

First, the bound on the size of $\partial_{A}(I)$ now becomes $\left|\partial_{A}(I)\right| \geq \sum_{i \in I}\left|J_{i}(A)\right|-$ $d \cdot|I|$ (as opposed to $\left|\partial_{A}(I)\right| \geq c \cdot|I|$ in [ABSRW04]). Next, the proof of the crucial inequality $\left|J_{i_{1}}(A) \cap \partial_{A}(I)\right| \leq s-\ell$ in [ABSRW04] actually shows
$\left|J_{i_{1}}(A) \cap \partial_{A}(I)\right| \leq\left|J_{i_{1}}(A)\right|-\ell$. Last, the final calculation now looks like

$$
\begin{aligned}
\sum_{i \in I}\left|J_{i}(A)\right|-d \cdot|I| & \leq\left|\partial_{A}(I)\right| \leq \sum_{i \in I_{0}}\left|J_{i}(A)\right|+\sum_{i \in I_{1}}\left(\left|J_{i}(A)\right|-\ell\right) \\
& =\sum_{i \in I}\left|J_{i}(A)\right|-\ell \cdot\left|I_{1}\right| \leq \sum_{i \in I}\left|J_{i}(A)\right|-\ell \cdot(|I|-w(C))
\end{aligned}
$$

which implies our bound $w(C)>\frac{r(\ell-d)}{2 \ell}$ since $|I|>r / 2$.
We will only need the following special case of this.
Corollary 3.4. Let $A$ be an $(r, d)$-lossless expander of size $m \times n$ such that $\left|J_{i}(A)\right| \geq 2 d$ for all $i \in[m]$. Then for every ordering $\leq$ and every $b \in$ $\{0,1\}^{m}$, every resolution refutation of $\tau_{\leq}(A, b)$ must have width $>r / 4$.

Proof. In Theorem 3.3, let $g_{i}$ be the function $\oplus_{j \in J_{i}(A)} x_{j}=b_{i}$ and $\ell:=$ $2 d$. Since $\tau_{\leq}(A, b)$ is a sub-CNF of $\tau(A, \vec{g})$ (for any ordering $\leq$ ), the result follows.

Definition 3.5 ([SBI04]). A decision tree is a rooted binary tree such that every internal node is labelled with a variable, the edges leaving this node correspond to whether the variable is set to 0 or 1 , and the leaves are labelled with either 0 or 1 . As usual, we assume that on every given path no variable appears more than once. Then every path from the root to a leaf may be viewed as a partial assignment, and this assignment, in turn, will sometimes be identified with the corresponding leaf. For a decision tree $T$ and $\varepsilon \in\{0,1\}$, we write the set of paths (partial assignments) that lead from the root to a leaf labelled $\varepsilon$ as $B r_{\varepsilon}(T)$. We say that a decision tree $T$ strongly represents a DNF $F$ if for every $\pi \in B r_{0}(T)$ and for all $t \in F,\left.t\right|_{\pi}=0$ and for every $\pi \in B r_{1}(T)$, there exists $t \in F$ such that $\left.t\right|_{\pi}=1$. Let the representation height of $F, h(F)$ be the minimum height of a decision tree strongly representing $F$.

Proposition 3.6 ([SBI04]). Let $\tau$ be an $h$-CNF, $P$ be a $\operatorname{Res}(k)$ refutation of $\tau$, and let $\rho$ be a partial assignment so that for every line $F$ of $P, h\left(\left.F\right|_{\rho}\right) \leq h$. Then $\left.\tau\right|_{\rho}$ has a resolution refutation of width $\leq k h$.

Remark 4. One small technicality is that the system $\operatorname{Res}(k)$, as defined in [SBI04], does not automatically include the axioms $\ell \vee \bar{\ell}$. Therefore, formally speaking, an application of their result only implies that $\left.\tau^{*}\right|_{\rho}$ has a resolution refutation of the required width, where $\tau^{*}$ is obtained from $\tau$ by adding these trivial axioms. It is, however, obvious that the latter can be eliminated from any resolution proof without increasing its width.

Finally, we recall a combinatorial inequality originally proved in [Jan90] and further generalized in [AS08, §8.1].

Definition 3.7. For propositional variables $x_{1}, \ldots, x_{n}$ and probabilities $p_{1}, \ldots, p_{n} \in[0,1]$, denote by $\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}$ a random assignment that independently assigns every variable $x_{i}$ to 1 with probability $p_{i}$ and to 0 with probability ( $1-p_{i}$ ).

Proposition 3.8. Let $t_{1}, \ldots, t_{H}$ be monotone terms (not necessarily distinct) in the variables $x_{1}, \ldots, x_{n}$, and let $p_{1}, \ldots, p_{n} \in[0,1]$. Let

$$
\begin{equation*}
\kappa \stackrel{\text { def }}{=} \sum_{\alpha \in[H]} \mathbf{P}\left[t_{\alpha}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=1\right] \tag{11}
\end{equation*}
$$

be the expectation of the number of terms satisfied by $\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}$, and let

$$
\Delta \stackrel{\text { def }}{=} \sum_{\substack{\alpha \neq \beta \in[H] \\ \operatorname{Vars}\left(t_{\alpha}\right) \cap \operatorname{Vars}\left(t_{\beta}\right) \neq \emptyset}} \mathbf{P}\left[t_{\alpha}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=t_{\beta}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=1\right] .
$$

Then

$$
\mathbf{P}\left[\bigwedge_{\alpha \in[H]} t_{\alpha}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=0\right] \leq e^{-\kappa+\frac{\Delta}{2}} .
$$

Proof. The notation of $[\mathrm{AS} 08, \S 8.1]$ corresponds to ours as follows: $\Omega$ is $\left\{x_{1}, \ldots, x_{n}\right\}, I$ is $[H], A_{i}:=\operatorname{Vars}\left(t_{i}\right)(i \in[H])$, and $B_{i}$ is the event $t_{i}\left(\boldsymbol{a}_{\vec{p}}\right)=1$; we also renamed $\mu$ to $\kappa$. After this translation, our proposition is exactly the second inequality in [AS08, Th. 8.1.1].

Remark 5. We will actually need another version of this inequality ([AS08, Th. 8.1.2]), useful when $\Delta \gg \kappa$. However, we will need to dig into the proof of that theorem rather deeply. For this reason we do not formulate the corresponding general statement here, but rather we incorporate it as an ad hoc argument in the proof of Lemma 4.4.

## 4. Small restriction switching lemma

In this section we give a quadratic improvement on the original version of this lemma from [SBI04]. In our main application, the underlying random restriction will not act totally independently on different variables, but at least it will have some "weak local independence" property. We will be able to capture this property in the main statement so that the proof will not become more complicated than for truly independent restrictions, but this will require several auxiliary definitions.

Definition 4.1. We say that a (deterministic) restriction $\rho^{\prime}$ is a sub-restriction of another restriction $\rho$ if $\sup \left(\rho^{\prime}\right) \subseteq \sup (\rho)$ and $\rho, \rho^{\prime}$ coincide on $\sup \left(\rho^{\prime}\right)$. For random restrictions $\boldsymbol{\rho}, \boldsymbol{\rho}^{\prime}, \boldsymbol{\rho}^{\prime}$ is a sub-restriction of $\boldsymbol{\rho}$ if there exists a set $\Omega$, a random variable $\omega \in \Omega$, and functions $\pi, \pi^{\prime}$ from $\Omega$ to the set of all restrictions such that
(1) $\boldsymbol{\rho}$ has the same distribution as $\pi(\boldsymbol{\omega})$, and $\boldsymbol{\rho}^{\prime}$ has the same distribution as $\pi^{\prime}(\boldsymbol{\omega})$;
(2) for every individual $\omega \in \Omega, \pi^{\prime}(\omega)$ is a sub-restriction of $\pi(\omega)$.

Clearly, if $\rho^{\prime}$ is a sub-restriction of $\boldsymbol{\rho}$, then for any terms $t_{1}, \ldots, t_{H}$ we have the inequality $\mathbf{P}\left[\mathrm{V}_{\alpha \in[H]}\left(\left.t_{\alpha}\right|_{\rho^{\prime}} \equiv 1\right)\right] \leq \mathbf{P}\left[\mathrm{V}_{\alpha \in[H]}\left(\left.t_{\alpha}\right|_{\rho} \equiv 1\right)\right]$.

Definition 4.2. A weight function is any function $\mu:\left\{x_{1}, \ldots, x_{n}\right\} \longrightarrow \mathbb{Z}^{+}$ from variables to strictly positive integers. The weight $\mu(V)$ of a set of variables $V$ is defined as $\mu(V) \stackrel{\text { def }}{=} \sum_{x \in V} \mu(x)$, and the weight of a term $t$ is $\mu(t) \stackrel{\text { def }}{=}$ $\mu(\operatorname{Vars}(t))$. We will denote by $\mu_{\text {triv }}$ the trivial weight function identically equal to 1 ; in this case the weight of a term is equal to its width. For arbitrary weight function $\mu$, we have this in one direction: $\mu(t) \geq w(t)$. A DNF $F$ is a weighted $K-D N F$ ( with respect to a weight function $\mu$ ) if all terms $t \in F$ have weight $\leq K$.

Finally, we define the amount of "weak local independence" needed to carry out the proof of our switching lemma. It is similar to the ordinary $r$-wise independence with one important change. Namely, we do not demand that on small sets of variables our random restriction behaves exactly as the genuine independent and identically distributed restriction. We only require that it can be obtained from the latter by assigning more variables if necessary.

Definition 4.3. Let $\mu$ be a weight function, and let $p \in[0,1]$. For a set of variables $X$, let $\boldsymbol{\rho}_{\mu, \boldsymbol{X}, \boldsymbol{p}}$ be the random restriction of these variables that independently assigns every $x \in X$ to 0,1 with probability $p^{\mu(x)} / 2$, and leaves it unassigned with probability $\left(1-p^{\mu(x)}\right)$. Given an integer $r$, say that a random restriction $\boldsymbol{\rho}$ is $(r, \mu, p)$-independent if for every subset $X$ of variables with $|X| \leq r, \boldsymbol{\rho}_{\mu, \boldsymbol{X}, \boldsymbol{p}}$ is a sub-restriction of $\left.\boldsymbol{\rho}\right|_{X}$.

Thus, if $\boldsymbol{\rho}$ is $(r, \mu, p)$-independent, then for any set of terms $t_{1}, \ldots, t_{H}$ such that $X \stackrel{\text { def }}{=} \bigcup_{i=1}^{H} \operatorname{Vars}\left(t_{i}\right)$ has size at most $r$, we have

$$
\begin{equation*}
\mathbf{P}\left[\bigvee_{\alpha \in[H]}\left(t_{\alpha} \mid \rho_{\mu, X, \boldsymbol{p}} \equiv 1\right)\right] \leq \mathbf{P}\left[\bigvee_{\alpha \in[H]}\left(\left.t_{\alpha}\right|_{\boldsymbol{\rho}} \equiv 1\right)\right] \tag{12}
\end{equation*}
$$

(And, as a matter of fact, this is the only property of $(r, \mu, p)$-independent restrictions we will need.)

The rest of this section is devoted to the proof of the following lemma.
Lemma 4.4. Let $\mu$ be a weight function, and let $F$ be a weighted $K$-DNF with respect to $\mu$. Suppose that $\boldsymbol{\rho}$ is a random ( $r, \mu, p$ )-independent restriction. Then for every $h \leq r$,

$$
\begin{equation*}
\mathbf{P}\left[h\left(\left.F\right|_{\boldsymbol{\rho}}\right)>h\right] \leq \exp \left(-h(p / 2)^{O(K)}\right) . \tag{13}
\end{equation*}
$$

Proof. We begin with the case when $F$ is monotone, and in that case we are going to prove by induction on $K$ that

$$
\begin{equation*}
\mathbf{P}\left[h\left(\left.F\right|_{\boldsymbol{\rho}}\right)>h\right] \leq e^{-h(p \varepsilon)^{2 K}} \tag{14}
\end{equation*}
$$

where $\varepsilon>0$ is a constant chosen sufficiently small so that the arguments for Cases 1 and 2 below work. ${ }^{6}$ This clearly implies (13).

Base $K=0$ is obvious since $F$ is a constant and $h\left(\left.F\right|_{\rho}\right)=0$ with probability 1.

Inductive step. Let $F$ be a nontrivial monotone weighted $K$-DNF, and assume that (14) is established for all weighted $K^{\prime}$-DNF with $K^{\prime}<K$ (and for all integers $h$ ). We define a numerical invariant $\delta(F)$ that represents, up to a scaling factor, the optimal value of the parameter $\Delta$ in Proposition 3.8 when we attempt to apply it to the formula $F$. Further analysis will be sharply divided according to whether $\delta(F)$ is small or large; cf. [SBI04, Th. 3].

First, for a set of variables $V$, we let

$$
\delta(V) \stackrel{\text { def }}{=} \begin{cases}0 & \text { if } V=\emptyset \\ (2 / p)^{\mu(V)} & \text { otherwise }\end{cases}
$$

Next, given a random term $\boldsymbol{t} \in F$ (viewed, for the duration of this proof, as a probability distribution on the set of all terms of $F$ ), we define

$$
\delta(\boldsymbol{t}) \stackrel{\text { def }}{=} \mathbf{E}\left[\delta\left(\operatorname{Vars}(\boldsymbol{t}) \cap \operatorname{Vars}\left(\boldsymbol{t}^{\prime}\right)\right)\right]
$$

where $\boldsymbol{t}^{\prime}$ is an independent copy of $\boldsymbol{t}$. Finally, let

$$
\delta(F) \stackrel{\text { def }}{=} \min _{\boldsymbol{t}} \delta(\boldsymbol{t})
$$

where the minimum is taken over all random terms $\boldsymbol{t} \in F$. (This minimum exists since the space of all probability distributions on terms of $F$ is compact, and the function $\delta(\boldsymbol{t})$ is continuous.) Let

$$
\begin{equation*}
s \stackrel{\text { def }}{=}\left\lceil 2 h(p \varepsilon)^{2 K}\right\rceil \tag{15}
\end{equation*}
$$

(so that the right-hand side of (14) is roughly $e^{-s / 2}$ ). Consider two cases.
Case 1: $\delta(F) \leq s^{-1}$. Let $\boldsymbol{t} \in F$ be the random term for which $\delta(\boldsymbol{t}) \leq s^{-1}$. Arguing as in the proof of [AS08, Th. 8.1.2], let $\boldsymbol{t}_{\boldsymbol{1}}, \ldots, \boldsymbol{t}_{\boldsymbol{H}}$ be independent samples from $F$ according to the distribution of $\boldsymbol{t}$, where

$$
\begin{equation*}
H \stackrel{\text { def }}{=}\left\lceil s(2 / p)^{K}\right\rceil \tag{16}
\end{equation*}
$$

[^6]Then

$$
\mathbf{E}\left[\sum_{\alpha \neq \beta \in[H]} \delta\left(\operatorname{Vars}\left(\boldsymbol{t}_{\boldsymbol{\alpha}}\right) \cap \operatorname{Vars}\left(\boldsymbol{t}_{\boldsymbol{\beta}}\right)\right)\right]=H(H-1) \delta(\boldsymbol{t}) \leq H(H-1) s^{-1},
$$

and we fix any particular sampling $t_{1}, \ldots, t_{H}$ for which

$$
\begin{equation*}
\sum_{\alpha \neq \beta \in[H]} \delta\left(\operatorname{Vars}\left(t_{\alpha}\right) \cap \operatorname{Vars}\left(t_{\beta}\right)\right) \leq H(H-1) s^{-1} . \tag{17}
\end{equation*}
$$

The terms $t_{1}, \ldots, t_{H}$ altogether contain at most $H K$ variables, and $H K \leq$ $h \leq r$ as long as the constant $\varepsilon$ is small enough. Therefore, since $\boldsymbol{\rho}$ is $(r, \mu, p)$ independent, (12) implies

$$
\begin{equation*}
\mathbf{P}\left[h\left(\left.F\right|_{\boldsymbol{\rho}}\right)>h\right] \leq \mathbf{P}\left[\bigwedge_{\alpha \in[H]}\left(\left.t_{\alpha}\right|_{\boldsymbol{\rho}} \not \equiv 1\right)\right] \leq \mathbf{P}\left[\bigwedge_{\alpha \in[H]}\left(\left.t_{\alpha}\right|_{\boldsymbol{\rho}_{\mu, \boldsymbol{p}}} \not \equiv 1\right)\right] . \tag{18}
\end{equation*}
$$

Furthermore, since the terms $t_{\alpha}$ are monotone, $\left.t_{\alpha}\right|_{\rho_{\mu, p}} \equiv 1$ if and only if $t_{\alpha}(\boldsymbol{a})=1$, where $\boldsymbol{a}$ is the random (total) assignment obtained from $\boldsymbol{\rho}_{\mu, \boldsymbol{p}}$ by additionally assigning to 0 all unassigned variables. Now, $\boldsymbol{a}$ has the same distribution as the assignment $\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}$ from Definition 3.7 for the vector of probabilities $\tilde{p}$ given by $\tilde{p_{i}} \stackrel{\text { def }}{=} p^{\mu\left(x_{i}\right)} / 2$. Let $p_{i} \stackrel{\text { def }}{=}(p / 2)^{\mu\left(x_{i}\right)} ;$ then $p_{i} \leq \tilde{p}_{i}$ and, since the event $\bigvee_{\alpha \in[H]} t_{\alpha}(a)=0$ is anti-monotone in $a$, we get

$$
\begin{equation*}
\mathbf{P}\left[\bigwedge_{\alpha \in[H]}\left(t_{\alpha} \mid \rho_{\mu, \boldsymbol{p}} \not \equiv 1\right)\right]=\mathbf{P}\left[\bigwedge_{\alpha \in[H]} t_{\alpha}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=0\right] \leq \mathbf{P}\left[\bigwedge_{\alpha \in[H]} t_{\alpha}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=0\right] . \tag{19}
\end{equation*}
$$

We are going to upper bound $\mathbf{P}\left[\bigwedge_{\alpha \in[H]} t_{\alpha}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=0\right]$, given (17).
For every $\alpha \in[H]$ such that $\mu\left(t_{\alpha}\right)<K$, introduce a new auxiliary variable $y_{\alpha}$ with weight $K-\mu\left(t_{\alpha}\right)$, and replace $t_{\alpha}$ with $t_{\alpha} \wedge y_{\alpha}$. This operation does not change the value of the sum $\sum_{\alpha \neq \beta \in[H]} \delta\left(\operatorname{Vars}\left(t_{\alpha}\right) \cap \operatorname{Vars}\left(t_{\beta}\right)\right)$ in (17), and $\mathbf{P}\left[\bigwedge_{\alpha \in[H]} t_{\alpha}\left(\boldsymbol{a}_{\vec{p}}\right)=0\right]$ may only increase. Therefore, we may assume without loss of generality that all terms in $F$ have weight exactly $K$.

And now we apply Proposition 3.8. All events $\mathbf{P}\left[t_{\alpha}\left(\boldsymbol{a}_{\vec{p}}\right)=1\right]$ have the same probability $(p / 2)^{K}$; therefore, the quantity $\kappa$ given by (11) is equal to $H(p / 2)^{K}$. Also, whenever $\operatorname{Vars}\left(t_{\alpha}\right) \cap \operatorname{Vars}\left(t_{\beta}\right) \neq \emptyset$, we have

$$
\mathbf{P}\left[t_{\alpha}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=t_{\beta}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=1\right]=(p / 2)^{\mu\left(t_{\alpha} \wedge t_{\beta}\right)}=(p / 2)^{2 K} \delta\left(\operatorname{Vars}\left(t_{\alpha}\right) \cap \operatorname{Vars}\left(t_{\beta}\right)\right) ;
$$

therefore,

$$
\begin{aligned}
\Delta=(p / 2)^{2 K} \sum_{\alpha \neq \beta \in[H]} \delta\left(\operatorname{Vars}\left(t_{\alpha}\right) \cap \operatorname{Vars}\left(t_{\beta}\right)\right) & \leq(p / 2)^{2 K} H(H-1) s^{-1} \\
& \leq H(p / 2)^{K}=\kappa
\end{aligned}
$$

by (17) and (16). Applying Proposition 3.8, we get

$$
\mathbf{P}\left[\bigwedge_{\alpha \in[H]} t_{\alpha}\left(\boldsymbol{a}_{\overrightarrow{\boldsymbol{p}}}\right)=0\right] \leq e^{-\frac{H(p / 2)^{K}}{2}} \leq e^{-s / 2}
$$

which, along with (18) and (19), implies (14).
Case 2: $\delta(F) \geq s^{-1}$. Note first that

$$
\begin{aligned}
\delta(\boldsymbol{t}) & =\mathbf{E}\left[\delta\left(\operatorname{Vars}(\boldsymbol{t}) \cap \operatorname{Vars}\left(\boldsymbol{t}^{\prime}\right)\right)\right]=\sum_{V} \delta(V) \cdot \mathbf{P}\left[\operatorname{Vars}(\boldsymbol{t}) \cap \operatorname{Vars}\left(\boldsymbol{t}^{\prime}\right)=V\right] \\
& \leq \sum_{V} \delta(V) \cdot \mathbf{P}\left[V \subseteq \operatorname{Vars}(\boldsymbol{t}) \wedge V \subseteq \operatorname{Vars}\left(\boldsymbol{t}^{\prime}\right)\right] \\
& =\sum_{V \neq \emptyset}(2 / p)^{\mu(V)} \mathbf{P}[V \subseteq \operatorname{Vars}(\boldsymbol{t})]^{2} ;
\end{aligned}
$$

therefore, we also have

$$
\min _{t} \sum_{V \neq \emptyset}(2 / p)^{\mu(V)} \mathbf{P}[V \subseteq \operatorname{Vars}(\boldsymbol{t})]^{2} \geq s^{-1}
$$

Let $\boldsymbol{t}$ be the random term on which the quadratic form

$$
\sum_{V \neq \emptyset}(2 / p)^{\mu(V)} \mathbf{P}[V \subseteq \operatorname{Vars}(\boldsymbol{t})]^{2}
$$

in the variables $p_{t} \stackrel{\text { def }}{=} \mathbf{P}[\boldsymbol{t}=t]$ describing the density function of the associated distribution attains its minimal value $\delta \geq s^{-1}$. Denoting further $\mathbf{P}[V \subseteq \operatorname{Vars}(\boldsymbol{t})]$ (viewed as a linear form in the variables $p_{t}$ ) by $p_{V}$ we have

$$
\sum_{\substack{t \in F \\ p_{t} \neq 0}} \frac{\partial\left(\sum_{V \neq \emptyset}(2 / p)^{\mu(V)} p_{V}^{2}\right)}{\partial p_{t}} p_{t}=2 \sum_{V \neq \emptyset}(2 / p)^{\mu(V)} p_{V}^{2}=2 \delta .
$$

Since $\sum_{\substack{t \in F \\ p_{t} \neq 0}} p_{t}=1$, this implies the existence of some $t_{0} \in F$ with $p_{t_{0}} \neq 0$ and such that

$$
\frac{\partial\left(\sum_{V \neq \emptyset}(2 / p)^{\mu(V)} p_{V}^{2}\right)}{\partial p_{t_{0}}}=2 \sum_{\substack{V \neq \emptyset \\ V \subseteq \operatorname{Vars}\left(t_{0}\right)}}(2 / p)^{\mu(V)} p_{V} \geq 2 \delta .
$$

Therefore, for any term $t \in F$ (even when $p_{t}=0$ !), we also have

$$
\sum_{\substack{V \neq \varnothing \\ V \subseteq \operatorname{Vars}(t)}}(2 / p)^{\mu(V)} p_{V} \geq \delta
$$

since otherwise, by setting $p_{t_{0}}$ to 0 and $p_{t}$ to $p_{t}+p_{t_{0}}$, we would have obtained contradiction with the assumption that $t$ minimizes the quadratic form $\sum_{V \neq \emptyset}(2 / p)^{\mu(V)} \mathbf{P}[V \subseteq \operatorname{Vars}(\boldsymbol{t})]^{2}$.

Summarizing the above argument, we have found coefficients $p_{V} \geq 0$ ( $V$ a nonempty set of variables) and $\delta \geq s^{-1}$ such that

$$
\begin{align*}
& \sum_{V \neq \emptyset}(2 / p)^{\mu(V)} p_{V}^{2}=\delta,  \tag{20}\\
& \forall t \in F\left(\sum_{\substack{V \neq \emptyset \\
V \subseteq \operatorname{Vars}(t)}}(2 / p)^{\mu(V)} p_{V} \geq \delta\right) . \tag{21}
\end{align*}
$$

These are the only properties of $p_{V}$ S we will need in the rest of the proof; in particular, at this point we can forget their interpretation as certain probabilities.

Let us order all nonempty sets of variables $V$ in decreasing order with respect to the coefficient $p_{V}: p_{V_{1}} \geq p_{V_{2}} \geq p_{V_{3}} \geq \cdots$. Assume without loss of generality that the variables $x_{1}, \ldots, x_{n}$ are enumerated in the order in which they appear for the first time in the sequence $V_{1}, V_{2}, V_{3}, \ldots$. (equivalently, we require that every union $V_{1} \cup V_{2} \cup \cdots \cup V_{H}$ should be an initial segment in the set of variables ordered according to their indices). For a nonempty $V$, denote by $i(V)$ the maximal index $i$ for which $x_{i} \in V$.

Now we classify terms $t \in F$ as follows. Represent a given term $t \in F$ in the form $t=x_{i_{1}} \wedge x_{i_{2}} \wedge \cdots \wedge x_{i_{w}}\left(i_{1}<i_{2}<\cdots<i_{w}\right)$. For a nonempty $V \subseteq \operatorname{Vars}(t)$, let $\mu_{t}(V) \stackrel{\text { def }}{=} \mu\left(x_{i_{1}} \wedge x_{i_{2}} \wedge \cdots \wedge x_{i(V)}\right)$. (In other words, this is the weight of the minimal initial segment in $\operatorname{Vars}(t)$ containing $V$.) We split the sum in (21) with respect to the value of $\mu_{t}(V)$, and we classify $t$ according to which part is large enough. Formally, let

$$
F_{\mu} \stackrel{\text { def }}{=}\left\{t \in F \mid \sum_{\substack{\mu_{t}(V)=\mu \\ V \subseteq \text { Vars }(t)}} p_{V} \geq \delta(p / 4)^{\mu}\right\} ;
$$

let us see that $F=\bigvee_{\mu=1}^{K} F_{\mu}$. Assume for the sake of contradiction that for some $t \in F$ and all real $\mu$, we have

$$
\sum_{\substack{\mu_{t}(V)=\mu \\ V \subseteq \operatorname{Vars}(t)}} p_{V}<\delta(p / 4)^{\mu}<\delta(p / 2)^{\mu}
$$

Summing these inequalities with coefficients $(2 / p)^{\mu}$, we would get

$$
\sum_{V \subseteq \operatorname{Vars}(t)} p_{V}(2 / p)^{\mu_{t}(V)}<\delta,
$$

which would be in contradiction with (21) since clearly $\mu_{t}(V) \geq \mu(V)$ for all $t, V$. This contradiction shows that $t \in F_{\mu}$ for at least one $\mu$ that is indeed $F=\bigvee_{\mu=1}^{K} F_{\mu}$.

Then, noting that $\mu_{t}(V) \geq \mu(V)$ for all $t, V$, we have $F=\bigvee_{\mu=1}^{K} F_{\mu}$. This clearly implies $h\left(\left.F\right|_{\rho}\right) \leq \sum_{\mu=1}^{K} h\left(\left.F_{\mu}\right|_{\rho}\right)$ for every restriction $\rho$. (The concatenation of decision trees strongly representing $\left.F_{\mu}\right|_{\rho}$ for $\mu=1, \ldots, K$, strongly represents $\left.F\right|_{\rho}$.) Thus, we obtain

$$
\begin{equation*}
\mathbf{P}\left[h\left(\left.F\right|_{\rho}\right)>h\right] \leq \sum_{\mu=1}^{K} \mathbf{P}\left[h\left(\left.F_{\mu}\right|_{\rho}\right)>h 2^{-\mu}\right], \tag{22}
\end{equation*}
$$

and we treat every term in the right-hand side separately.
So, let us fix $\mu, 1 \leq \mu \leq K$, and let us fix $t \in F_{\mu} ; t=x_{i_{1}} \wedge x_{i_{2}} \wedge \cdots \wedge$ $x_{i_{w}}\left(i_{1}<i_{2}<\cdots<i_{w}\right)$. Let $d$ be the index for which $\mu\left(x_{i_{1}} \wedge x_{i_{2}} \wedge \cdots \wedge x_{i_{d}}\right)=\mu$. Then $d \leq \mu$, which implies that there are at most $2^{\mu}$ subsets $V \subseteq t$ with $\mu_{t}(V)=\mu$. Consulting the definition of $F_{\mu}$, we see that there exists a particular $V \subseteq \operatorname{Vars}(t)$ with $\mu_{t}(V)=\mu$ such that $p_{V} \geq \delta(p / 8)^{\mu}$.

Let this particular $V$ have rank $\ell$ in our enumeration of all nonempty subsets. We are going to upper bound $\sum_{\nu=1}^{\ell}\left|V_{\nu}\right|$. Notice that, according to the choice of this enumeration, $p_{V_{\nu}} \geq p_{V}$ for all $\nu \in[\ell]$. Therefore,

$$
\begin{aligned}
\sum_{\nu=1}^{\ell}(2 / p)^{\mu\left(V_{\nu}\right)} p_{V_{\nu}}^{2} & \geq p_{V}^{2} \cdot \sum_{\nu=1}^{\ell}(2 / p)^{\mu\left(V_{\nu}\right)} \\
& \geq \delta^{2}\left(p^{2} / 64\right)^{\mu} \sum_{\nu=1}^{\ell} \mu\left(V_{\nu}\right) \geq \delta^{2}\left(p^{2} / 64\right)^{\mu} \sum_{\nu=1}^{\ell}\left|V_{\nu}\right|
\end{aligned}
$$

Comparing this with (20), we get

$$
\begin{equation*}
\sum_{\nu=1}^{\ell}\left|V_{\nu}\right| \leq \delta^{-1}\left(64 / p^{2}\right)^{\mu} \leq s\left(64 / p^{2}\right)^{\mu} \tag{23}
\end{equation*}
$$

The conclusion (23) holds for every $t \in F_{\mu}$. Therefore, if we define $\ell_{\mu}$ as the maximal value $\ell$ for which the bound (23) holds, then for every $t \in F_{\mu}$, $V_{1} \cup \cdots \cup V_{\ell_{\mu}}$ will contain some $V \subseteq \operatorname{Vars}(t)$ with $\mu_{t}(V)=\mu$. Since $V_{1} \cup \cdots \cup V_{\ell_{\mu}}$ is an initial segment in $\left\{x_{1}, \ldots, x_{n}\right\}$, this implies that

$$
\mu\left(\left(V_{1} \cup \cdots \cup V_{\ell_{\mu}}\right) \cap \operatorname{Vars}(t)\right) \geq \mu .
$$

Now let $T_{\mu}$ be the (oblivious) decision tree that queries all variables in $V_{1} \cup \cdots \cup V_{\ell_{\mu}}$ (in an arbitrary order). Our analysis implies that for every leaf $\pi$ of $T_{\mu},\left.\left(F_{\mu}\right)\right|_{\pi}$ is a monotone weighted $(K-\mu)$-DNF. Therefore, we may apply the inductive assumption (14) and conclude that for each leaf $\pi$,

$$
\mathbf{P}\left[h\left(\left.\left(\left.F_{\mu}\right|_{\pi}\right)\right|_{\rho}\right)>\frac{1}{2} h 2^{-\mu}\right] \leq e^{-\frac{1}{2} h 2^{-\mu}(p \varepsilon)^{2(K-\mu)}}
$$

By the definition of $\ell_{\mu}$ and (15), the height of $T_{\mu}$ does not exceed $h(p \varepsilon)^{2 K}$. $\left\{O\left(1 / p^{2}\right)\right\}^{\mu}$, which is at most $\frac{1}{4} h 2^{-\mu}(p \varepsilon)^{2(K-\mu)}$, as long as $\varepsilon$ is small enough.

Summing over all leaves of $T_{\mu}$,

$$
\mathbf{P}\left[\exists \pi \in\{0,1\}^{V_{1} \cup \ldots \cup V_{\ell_{\mu}}}\left(h\left(\left.\left(\left.F_{\mu}\right|_{\pi}\right)\right|_{\rho}\right)>\frac{1}{2} h 2^{-\mu}\right)\right] \leq e^{-\frac{1}{4} h 2^{-\mu}(p \varepsilon)^{2(K-\mu)}}
$$

Noting that the height of the tree $\left(\left.T_{\mu}\right|_{\rho}\right)$ never exceeds $\frac{1}{2} h 2^{-\mu}$, the latter event is logically implied by $h\left(\left.F_{\mu}\right|_{\rho}\right)>h 2^{-\mu}$. Comparing with (22),
which is at most $e^{-h(p \varepsilon)^{2 K}}$, as long as $\varepsilon$ is small enough.
This proves (14) in Case 2, completes the inductive step, and completes the proof of Lemma 4.4 for monotone $F$.

In order to extend this lemma to the general case, let us note first that, by symmetry, we automatically have it when the DNF $F$ is pseudo-monotone, defined as those DNF that can be turned into monotone by negating some variables. Now, every DNF $F$ (interpreted as a set of terms) has a straightforward fractional cover by pseudo-monotone DNFs of acceptable size; that is, it possesses a random pseudo-monotone sub-DNF $\boldsymbol{G}$ that contains every individual term $t \in F$ with sufficiently large probability. This $\boldsymbol{G}$ is generated simply by picking a (total) assignment $\boldsymbol{a}$ at random and including those terms that satisfy it. If it were an ordinary cover, we would have been trivially done. The following general lemma (which will also be repeatedly used in both our applications of Lemma 4.4) shows how to handle the more general fractional case.

Lemma 4.5. Let $F$ be a $k$-DNF, and let $\boldsymbol{G}$ be a random sub-DNF such that

$$
\begin{equation*}
\forall t \in F(\mathbf{P}[t \in \boldsymbol{G}] \geq \varepsilon) \tag{24}
\end{equation*}
$$

where $\varepsilon$ is an arbitrary parameter. Let $\boldsymbol{\rho}$ be a random restriction such that for every fixed $G$ from the support of $\boldsymbol{G}$,

$$
\mathbf{P}\left[h\left(\left.G\right|_{\rho}\right)>h\right] \leq \delta,
$$

where $\delta$ is another parameter. Then

$$
\mathbf{P}\left[h\left(\left.F\right|_{\rho}\right)>h\left(\frac{2 k}{\varepsilon}+k+1\right)\right] \leq 2 \delta / \varepsilon .
$$

Proof of Lemma 4.5. Arguing by averaging over the distribution of $\boldsymbol{G}$, we get $\mathbf{P}[h(\boldsymbol{G} \mid \boldsymbol{\rho})>h] \leq \delta$, where $\boldsymbol{G}$ and $\boldsymbol{\rho}$ are assumed independent. By Markov's inequality,

$$
\mathbf{P}_{\boldsymbol{\rho}}\left[\mathbf{P}_{\boldsymbol{G}}\left[h\left(\left.\boldsymbol{G}\right|_{\boldsymbol{\rho}}\right)>h\right] \geq \varepsilon / 2\right] \leq 2 \delta / \varepsilon
$$

Therefore, we only have to show that for every individual restriction $\rho$,

$$
\begin{equation*}
\mathbf{P}\left[h\left(\left.\boldsymbol{G}\right|_{\rho}\right)>h\right]<\varepsilon / 2 \tag{25}
\end{equation*}
$$

logically implies $h\left(\left.F\right|_{\rho}\right) \leq h\left(\frac{2 k}{\varepsilon}+k+1\right)$.

This is done by an adaptation of the "block sensitivity" method [Nis91], $\left[\mathrm{BBC}^{+} 01\right]$ to our setting. Assume that (25) holds. We want to construct a decision tree of height $\leq h\left(\frac{2 k}{\varepsilon}+k+1\right)$ strongly representing $\left.F\right|_{\rho}$. We begin with a recursive construction of a sequence of decision trees $T_{0}, T_{1}, \ldots, T_{\ell}, \ldots$ that goes as follows.
$T_{0}$ is the trivial tree of height 0 . In order to construct $T_{\ell+1}$ from $T_{\ell}$, examine one-by-one all leaves $\pi$ of $T_{\ell}$. If either $\left.t\right|_{\pi}=0$ for all $\left.t \in F\right|_{\rho}$, or $\left.t\right|_{\pi}=1$ for some $\left.t \in F\right|_{\rho}$, we leave $\pi$ alone. Otherwise, pick up an arbitrary nontrivial term $\left.t_{\ell+1} \in\left(\left.F\right|_{\rho}\right)\right|_{\pi}$, and append to the leaf $\pi$ the oblivious decision tree querying all variables in $\operatorname{Vars}\left(t_{\ell+1}\right) \backslash(\sup (\rho) \cup \sup (\pi))$. Repeating this procedure for all leaves $\pi$ of $T_{\ell}$, we get $T_{\ell+1}$.

We terminate this construction after $s \stackrel{\text { def }}{=}\lceil(2 h / \varepsilon)\rceil$ steps. The only leaves $\pi$ of $T_{s}$ that still may violate Definition 3.5 are those for which the procedure of appending a new tree was repeated all $s$ times before we arrived at $\pi$. Let us fix any such leaf $\pi$, and let $t_{1}, t_{2}, \ldots, t_{s}$ be the terms picked by our algorithm along the path to $\pi$.

By (24),

$$
\mathbf{E}\left[\left|\left\{t_{1}, t_{2}, \ldots, t_{s}\right\} \cap \boldsymbol{G}\right|\right] \geq s \cdot \varepsilon
$$

On the other hand, denoting by Bad the indicator function of the event $h\left(\left.\boldsymbol{G}\right|_{\rho}\right)>h,(25)$ implies

$$
\mathbf{E}\left[\left|\left\{t_{1}, t_{2}, \ldots, t_{s}\right\} \cap \boldsymbol{G}\right| \cdot \boldsymbol{B a d}\right] \leq s \cdot \mathbf{E}[\boldsymbol{B a d}]<s \cdot \frac{\varepsilon}{2}
$$

Therefore, $\mathbf{E}\left[\left|\left\{t_{1}, t_{2}, \ldots, t_{s}\right\} \cap \boldsymbol{G}\right| \cdot(1-\boldsymbol{B a d})\right]>\frac{s \varepsilon}{2} \geq h$; pick up a particular sub-DNF $G$ that contains $(h+1)$ terms $t_{\alpha_{1}}, \ldots, t_{\alpha_{h+1}}$ from the list $\left\{t_{i} \mid i \in[s]\right\}$ (and possibly some other terms) such that $h\left(\left.G\right|_{\rho}\right) \leq h$. Let $T_{\pi}$ be the decision tree of height $\leq h$ strongly representing $G \mid{ }_{\rho}$. We complete out construction by appending the tree $T_{\pi}$ to the leaf $\pi$ (all vertices asking questions from $\sup (\pi)$ are contracted in the process), and repeating this for all leaves $\pi$ that still violate Definition 3.5.

Clearly, the final tree constructed in this way has height at most $s k+h \leq$ $h\left(\frac{2 k}{\varepsilon}+k+1\right)$, and we claim that it strongly represents $\left.F\right|_{\rho}$. For that we only have to check the leaves of the form $\pi * \sigma$, where $\pi$ is a problematic leaf of $T_{s}$ and $\sigma$ is a leaf of $T_{\pi}$ such that $\pi$ and $\sigma$ are consistent.

For every $\nu \in[h+1]$, let $V_{\nu} \stackrel{\text { def }}{=} \operatorname{Vars}\left(t_{\alpha_{\nu}}\right) \backslash \bigcup_{\alpha<\alpha_{\nu}} \operatorname{Vars}\left(t_{\alpha}\right)$. Then $V_{\nu}$ are disjoint and nonempty; therefore, since $|\sup (\sigma)| \leq h$, there exists at least one $\nu_{0} \in[h+1]$ such that $V_{\nu_{0}} \cap|\sup (\sigma)|=\emptyset$. Now, the values of $\pi$ at the variables $V_{\nu_{0}}$ can be changed in such a way that for the resulting partial assignment $\pi^{\prime}$, we have $t_{\alpha_{\nu_{0}}}\left|\pi^{\prime}=1 . V_{\nu_{0}} \cap\right| \sup (\sigma) \mid=\emptyset$ implies that $\pi^{\prime}$ and $\sigma$ are still consistent, hence $\left.t_{\alpha_{\nu_{0}}}\right|_{\sigma} \not \equiv 0$ and, since $t_{\alpha_{\nu_{0}}} \in G \mid \rho, \sigma \in B r_{0}\left(T_{\pi}\right)$ is impossible.

Thus, $\sigma \in B r_{1}\left(T_{\pi}\right)$, which implies that there exists a term $\left.\left.t \in G\right|_{\rho} \subseteq F\right|_{\rho}$ such that $\left.t\right|_{\sigma}=1$, and that implies $\left.t\right|_{\pi * \sigma}=1$.

We have constructed (under the assumption (25)) a tree of height $\leq$ $h\left(\frac{2 k}{\varepsilon}+k+1\right)$ strongly representing $\left.F\right|_{\rho}$. The proof of Lemma 4.5 is completed.

Let us now finish the proof of Lemma 4.4 for arbitrary weighted $K$-DNF $F$. For every (total) assignment $a$ to $F$, let $F_{a}$ be the set of all terms in $F$ satisfied by $a$. Then $F_{a}$ is a pseudo-monotone sub-DNF of $F$; therefore, we already have for it the required bound (13). Next, if $\boldsymbol{a}$ is picked completely at random, then for every $t \in F, \mathbf{P}\left[t \in F_{\boldsymbol{a}}\right]=2^{-w(t)} \geq 2^{-K}$. Thus, we may apply Lemma 4.5 with $k:=K, \boldsymbol{G}:=F_{\boldsymbol{a}}, \varepsilon:=2^{-K}, h:=\frac{h}{2 K 2^{K}+K+1}$ and $\delta:=\exp \left(-\frac{h}{2 K 2^{K}+K+1}(p / 2)^{O(K)}\right) \leq \exp \left(-h(p / 2)^{O(K)}\right)$, and this completes the proof of Lemma 4.4 in the general situation.

## 5. Hardness of Nisan generator for $\operatorname{Res}(k)$

In this section we prove Theorem 2.7. The proof is technically involved, and for that reason, it is split into a chain of auxiliary claims. They will be assembled together only at the end of the section, although some informal intuition as to what we are doing will be provided as we go along.

For the rest of the section, let us fix an $m \times n(r, d)$-lossless expander $A$, its ordering $\leq, b \in\{0,1\}^{m}$, and an integer $k \geq 1$ such that (2) holds. Without loss of generality we can assume that $r \geq k$ (since otherwise the bound is trivial) and that $d \geq 1$ (since otherwise $\tau_{\leq}(A, b)$ is consistent and hence does not possess any refutations whatsoever). The overall strategy of our proof appeared for the first time in [BP96] and has since become a standard tool in proof complexity. Namely, we want to design a random partial assignment $\rho$ of the variables Vars $\leq(A)$ that has the following two properties:
Height-reduction: for every fixed $k$-DNF $F, h\left(\left.F\right|_{\rho}\right)$ is small with high probability;
Width-preservation: with high probability every resolution refutation of $\left.\tau_{\leq}(A, b)\right|_{\rho}$ must have large width.

Now, if a small size $\operatorname{Res}(k)$ refutation $P$ of $\tau_{\leq}(A, b)$ existed, then with high probability $h\left(\left.F\right|_{\rho}\right)$ would be small for every $F \in P$ by the Height-reduction property, and we could apply Proposition 3.6. Its conclusion, however, would be in immediate contradiction with the Width-preservation property.

We begin realizing this plan with Width-preservation (mostly because this part is by far easier), and we will show that large width of resolution refutations of $\left.\tau_{\leq}(A, b)\right|_{\rho}$ is implied by a simple combinatorial property of $\rho$.

Some conventions on notation. In the rest of the paper we will be abbreviating the extension variables $y_{v_{\Sigma}^{i}}^{i}$ (where $\Sigma$ is a nonempty $\leq_{i}$-initial segment of $\left.J_{i}(A)\right)$ by $y_{\Sigma}^{i}$. For technical reasons it will be also convenient to introduce the variables $y_{\emptyset}^{i}$, along with the axioms $\bar{y}_{\emptyset}^{i}$. Likewise, it will be convenient not to identify $y_{\{j\}}^{i}$ with $x_{j}$ and with each other (as required in the general Definition 2.1), but to introduce instead new axioms $y_{\{j\}}^{i} \vee \bar{x}_{j}, \bar{y}_{\{j\}}^{i} \vee x_{j}$. These conventions imply that all variables in the vectors $\vec{x}, \vec{y}^{2}(i \in[m])$ are pairwise distinct, and the axioms of $\tau_{\leq}(A, b)$ become of particularly symmetric form

$$
\begin{equation*}
y_{\emptyset}^{i}=0, \quad y_{\Sigma \cup\{j\}}^{i}=y_{\Sigma}^{i} \oplus x_{j}, \quad y_{J_{i}(A)}^{i}=b_{i} \quad(i \in[m]) \tag{26}
\end{equation*}
$$

(More precisely, they are clauses of width $\leq 3$ resulting from the straightforward CNF expansion of these linear equations mod 2.)

Definition 5.1. A restriction $\rho$ of the variables $\operatorname{Vars} \leq(A)$ is closed if $\rho\left(y_{\emptyset}^{i}\right)$ $=0$ and $\rho\left(y_{J_{i}(A)}^{i}\right)=b_{i}$ for all $i \in[m]$. Let $J_{x}(\rho) \stackrel{\text { def }}{=}\left\{j \in[n] \mid x_{j} \in \sup (\rho)\right\} . \rho$ is sparse if it is closed and for every $i \in[m]$ and every two different initial segments $\Sigma \subset \Sigma^{\prime}$ in $J_{i}(A)$ such that $y_{\Sigma}^{i}, y_{\Sigma^{\prime}}^{i} \in \sup (\rho)$, we have $\left|\left(\Sigma^{\prime} \backslash \Sigma\right) \backslash J_{x}(\rho)\right| \geq 2 d$.

CLAIM 5.2. If $\rho$ is sparse, then every resolution refutation of $\left.\tau_{\leq}(A, b)\right|_{\rho}$ must be of width $>r / 4$.

Proof. Given a closed restriction $\rho$ of the variables Vars $\leq(A)$, define the $\left.\operatorname{matrix} A\right|_{\rho}$ as follows. For every row $i$ of the original matrix $A$, let $\emptyset=$ $\Sigma_{0}^{i} \subset \Sigma_{1}^{i} \subset \cdots \subset \Sigma_{s_{i}}^{i}=J_{i}(A)$ be the complete list of those initial segments $\Sigma$ in $J_{i}(A)$ for which $y_{\Sigma}^{i} \in \sup (\rho)$. Then the rows of the matrix $\left.A\right|_{\rho}$ are, by definition, indexed by the pairs $(i, \nu)\left(i \in[m], \nu \in\left[s_{i}\right]\right)$, its columns are indexed by $[n] \backslash J_{x}(\rho)$ (i.e., by those $x$-variables that are left unassigned by $\rho$ ), and the underlying set system is described as $J_{i, \nu}\left(\left.A\right|_{\rho}\right) \stackrel{\text { def }}{=}\left(\Sigma_{\nu}^{i} \backslash \Sigma_{\nu-1}^{i}\right) \backslash J_{x}(\rho)$.

Note that $\left.A\right|_{\rho}$ satisfies all assumptions of Corollary 3.4. Indeed, the bound $\left|J_{i, \nu}\left(\left.A\right|_{\rho}\right)\right| \geq 2 d$ is exactly the definition of sparseness. Next, given any set $\left\{\left(i_{1}, \nu_{1}\right), \ldots,\left(i_{\ell}, \nu_{\ell}\right)\right\}$ of rows in the matrix $\left.A\right|_{\rho}$ with $\ell \leq r$, applying the expansion property (1) for the original matrix $A$ to the set $\left\{i_{1}, \ldots, i_{\ell}\right\}$ gives us

$$
\begin{equation*}
\sum_{i \in\left\{i_{1}, \ldots, i_{\ell}\right\}}\left|J_{i}(A)\right|-\left|\partial_{A}\left(\left\{i_{1}, \ldots, i_{\ell}\right\}\right)\right| \leq d \cdot\left|\left\{i_{1}, \ldots, i_{\ell}\right\}\right| \leq d \cdot \ell \tag{27}
\end{equation*}
$$

On the other hand, it is easy to see that

$$
\begin{gather*}
\sum_{\alpha=1}^{\ell}\left|J_{i_{\alpha}, \nu_{\alpha}}\left(\left.A\right|_{\rho}\right)\right|-\left|\partial_{\left.A\right|_{\rho}}\left(\left\{\left(i_{1}, \nu_{1}\right), \ldots,\left(i_{\ell}, \nu_{\ell}\right)\right\}\right)\right|  \tag{28}\\
\quad \leq \sum_{i \in\left\{i_{1}, \ldots, i_{\ell}\right\}}\left|J_{i}(A)\right|-\left|\partial_{A}\left(\left\{i_{1}, \ldots, i_{\ell}\right\}\right)\right|
\end{gather*}
$$

Indeed, let $\ell_{j} \stackrel{\text { def }}{=}\left|\left\{\alpha \in[\ell] \mid j \in J_{i_{\alpha}, \nu_{\alpha}}\left(\left.A\right|_{\rho}\right)\right\}\right|$; then the left-hand side of (28) can be equivalently rewritten as $\sum_{j=1}^{n} f\left(\ell_{j}\right)$, where

$$
f(x) \stackrel{\text { def }}{=} \begin{cases}0, & x \leq 1, \\ x, & x \geq 2\end{cases}
$$

is a nondecreasing function. Likewise, the right-hand side is equal to $\sum_{j=1}^{n} f\left(\widehat{\ell}_{j}\right)$, where $\widehat{\ell}_{j} \stackrel{\text { def }}{=}\left|\left\{i \in\left\{i_{1}, \ldots, i_{\ell}\right\} \mid j \in J_{i}(A)\right\}\right|$. But $\ell_{j} \leq \widehat{\ell}_{j}$ since for every fixed $i$, the sets $\left\{J_{i_{\alpha}, \nu_{\alpha}}\left(\left.A\right|_{\rho}\right) \mid i_{\alpha}=i\right\}$ are mutually disjoint, whence (28) follows.

Now, (27) and (28) imply that $\left.A\right|_{\rho}$ is an $(r, d)$-lossless expander, and therefore all assumptions of Corollary 3.4 are fulfilled.

We now define (in a natural way) the assignment $\left.b\right|_{\rho}$ of the rows of the matrix $\left.A\right|_{\rho}$ by letting

$$
\left.\left(\left.b\right|_{\rho}\right)\right|_{i, \nu} \stackrel{\text { def }}{=} \rho\left(y_{\Sigma_{\nu-1}^{i}}^{i}\right) \oplus \rho\left(y_{\Sigma_{\nu}^{i}}^{i}\right) \oplus \bigoplus_{j \in\left(\Sigma_{\nu}^{i} \backslash \Sigma_{\nu-1}^{i}\right) \cap J_{x}(\rho)} \rho\left(x_{j}\right) .
$$

Finally, let $\leq\left.\right|_{\rho}$ be the ordering of the matrix $\left.A\right|_{\rho}$, where $\left.\left(\leq\left.\right|_{\rho}\right)\right|_{i, \nu}$ is the restriction of $\leq_{i}$ onto $J_{i, \nu}\left(\left.A\right|_{\rho}\right)$.

We extend the restriction $\rho$ to a variable substitution $\rho^{\prime}$ of variables in $\operatorname{Vars} \leq(A)$ by variables in $\operatorname{Vars}_{\leq\left.\right|_{\rho}}\left(\left.A\right|_{\rho}\right)$ defining it outside of $\sup (\rho)$ as follows. All $x_{j} \notin \sup (\rho)$ are simply left alone: $\rho^{\prime}\left(x_{j}\right) \stackrel{\text { def }}{=} x_{j}$. For every $y_{\Sigma}^{i} \notin \sup (\rho)$, we identify the index $\nu$ such that $\Sigma_{\nu-1}^{i} \subset \Sigma \subset \Sigma_{\nu}^{i}$, and we let

$$
\rho^{\prime}\left(y_{\Sigma}^{i}\right) \stackrel{\text { def }}{=} y_{\left(\Sigma \backslash \Sigma_{\nu-1}^{i}\right) \backslash J_{x}(\rho)}^{(i, \nu)} \oplus \rho\left(y_{\Sigma_{\nu-1}^{i}}^{i}\right) \oplus \bigoplus_{j \in\left(\Sigma \backslash \Sigma_{\nu-1}^{i}\right) \cap J_{x}(\rho)} \rho\left(x_{j}\right) ;
$$

cf. the definition of $\left.b\right|_{\rho}$. It is straightforward to check that this extension of the original restriction $\rho$ takes every equation in (26) either to $0=0$ or to an equation of the same form corresponding to $\tau_{\leq\left.\right|_{\rho^{\prime}}}\left(\left.A\right|_{\rho^{\prime}},\left.b\right|_{\rho^{\prime}}\right)$ and, therefore, maps every resolution refutation $P$ of $\left.\tau_{\leq}(A, b)\right|_{\rho}$ to a resolution refutation $\left.P\right|_{\rho^{\prime}}$ of $\tau_{\leq\left.\right|_{\rho^{\prime}}}\left(\left.A\right|_{\rho^{\prime}},\left.b\right|_{\rho^{\prime}}\right)$. By Corollary 3.4 (applied to $\left.\left(\left.A\right|_{\rho^{\prime}}, \leq\left|\left.\right|_{\rho^{\prime}}, b\right|_{\rho^{\prime}}\right)\right)$, $\left.P\right|_{\rho^{\prime}}$ must contain a clause of width $>r / 4$. Hence its preimage in $P$ also has width $>r / 4$. Claim 5.2 is proved.

Our second (and much more complicated) goal is to design a distribution on sparse restrictions that fullfils the Height-reduction property. After some consideration it becomes clear that in full generality this is impossible. Namely, if a term $t$ contains a couple of variables $y_{\Sigma}^{i}, y_{\Sigma^{\prime}}^{i}$ with $\Sigma, \Sigma^{\prime}$ close to each other, then no sparse restriction whatsoever can set $t$ to 1 , which completely ruins the whole argument. Thus, our most immediate task is to get rid of such "nasty" terms.

More notation. For a term $t$ in the variables Vars $\leq(A)$, by $J_{x}(t)$ we denote the set $\left\{j \in[n] \mid x_{j} \in \operatorname{Vars}(t)\right\}$, and we let $\operatorname{dom}(t) \stackrel{\text { def }}{=}\left\{i \in[m] \mid \exists \Sigma\left(y_{\Sigma}^{i} \in \operatorname{Vars}(t)\right)\right\}$. For a DNF $F, \operatorname{dom}(F) \stackrel{\text { def }}{=} \bigcup_{t \in F} \operatorname{dom}(t)$. For a nonempty $\leq_{i}$-initial segment $\Sigma$, let $r(\Sigma)$ be its right end (maximal element). For uniformity of notation, we let $r(\emptyset) \stackrel{\text { def }}{=} \operatorname{nil}_{i}$, where $\operatorname{nil}_{i}$ is an imaginary element with nil $i_{i}<_{i} j$ for all $j \in J_{i}(A)$. In this notation, the difference $\Sigma^{\prime} \backslash \Sigma$ of two initial segments $\Sigma \subseteq \Sigma^{\prime}$ coincides with the interval $\left(r(\Sigma), r\left(\Sigma^{\prime}\right)\right]$. For subsets $L, R \subseteq[n], \operatorname{Conv}(L, R)$ is the minimal interval containing them both.

Definition 5.3. Let $t$ be a term in the variables $\operatorname{Vars} \leq(A)$. For every $i \in \operatorname{dom}(t)$, list in the form $\Sigma_{1}^{i} \subset \Sigma_{2}^{i} \subset \cdots \subset \Sigma_{k_{i}}^{i}$ all initial segments $\Sigma$ such that $y_{\Sigma}^{i} \in \operatorname{Vars}(t)$. Say that $t$ is protected if there exists a system of subsets $L_{i \nu}, R_{i \nu} \subseteq J_{i}(A)\left(i \in \operatorname{dom}(t), \nu \in\left[k_{i}\right]\right)$ such that
(1) $L_{i, 1}<_{i} r\left(\Sigma_{1}^{i}\right)<_{i} R_{i, 1}<_{i} L_{i, 2}<_{i} r\left(\Sigma_{2}^{i}\right)<_{i} \ldots<_{i} r\left(\Sigma_{k_{i}}^{i}\right)<_{i} R_{i, k_{i}}$;
(2) $\left|L_{i \nu}\right|,\left|R_{i \nu}\right|=3 d$;
(3) $\sum_{i \in \operatorname{dom}(t)} \sum_{\nu \in\left[k_{i}\right]}\left|\operatorname{Conv}\left(L_{i \nu}, R_{i \nu}\right)\right| \leq 20 k d$;
(4) all $L_{i \nu}, R_{i \nu}$ are disjoint with $J_{x}(t)$.

The pair ( $L_{i \nu}, R_{i \nu}$ ) (often also written in the form ( $L_{i \Sigma}, R_{i \Sigma}$ ), where $\Sigma=\Sigma_{\nu}^{i}$ ), will be called a protection of the variable $y_{\Sigma_{\nu}^{i}}^{i} \in \operatorname{Vars}(t)$, and the entire set system $\left\{L_{i \nu}, R_{i \nu}\right\}$ will be called a protection of the term $t$. (Protections are, of course, not necessarily uniquely defined.) A DNF $F$ is protected if all terms $t \in F$ have this property.

We now show that every term $t$ with $w(t) \leq k$ can be replaced by a protected DNF that will be denoted by $R(t)$. For that purpose we will pick in a special way several "anchor" variables $y_{\Sigma}^{i}$; those will be left intact. All other $y$-variables will be expanded as linear forms in $x$-variables modulo our knowledge of the value of a nearby anchor variable (see Definition 5.6 for details), and $R(t)$ will then be the naive DNF-expansion of the resulting $\Lambda \oplus$-circuit.

Definition 5.4. Let $t$ be a term of width $\leq k$ in the variables $\operatorname{Vars} \leq(A)$. Denote by $t^{\prime}$ the term obtained from $t$ by appending to it the literals $\bar{y}_{\emptyset}^{i} \wedge$ $\left(y_{J_{i}(A)}^{i}\right)^{b_{i}}$ for every $i \in \operatorname{dom}(t)$. (If $t$ contains either $y_{\emptyset}^{i}$ or $\left(y_{J_{i}(A)}^{i}\right)^{\bar{b}_{i}}$ for some $i \in[m]$, we immediately abort the construction and let $R(t) \stackrel{\text { def }}{=} 0$.)

For a fixed $i \in \operatorname{dom}(t)$, list all initial segments $\Sigma$ with $y_{\Sigma}^{i} \in \operatorname{Vars}\left(t^{\prime}\right)$ : $\emptyset=\Sigma_{0}^{i} \subset \Sigma_{1}^{i} \subset \cdots \subset \Sigma_{k_{i}+1}^{i}=J_{i}(A)$, and let $r_{i \nu} \stackrel{\text { def }}{=} r\left(\Sigma_{\nu}^{i}\right)$. Let

$$
\begin{equation*}
\operatorname{Ker}_{i}(t) \stackrel{\text { def }}{=} J_{i}(A) \cap\left(J_{x}(t) \cup \bigcup_{i^{\prime} \in \operatorname{dom}(t) \backslash\{i\}} J_{i^{\prime}}(A)\right) . \tag{29}
\end{equation*}
$$

Construct the graph $G_{i}$ on $\left\{0,1, \ldots, k_{i}+1\right\}$ by connecting ( $\nu-1$ ) with $\nu$ if $\left|\left(r_{i, \nu-1}, r_{i \nu}\right] \backslash \operatorname{Ker}_{i}(t)\right|<6 d$, and let $\Gamma_{i, 0}, \Gamma_{i, 1}, \ldots, \Gamma_{i i_{i}}$ be the connected
components of this graph. $\Gamma_{i \alpha}$ is an interval in $\left\{0,1, \ldots, k_{i}+1\right\}$; let $\Gamma_{i \alpha}=$ $\left[\nu_{i \alpha}^{\ell}, \nu_{i \alpha}^{r}\right]$, and let $\hat{\Gamma}_{i \alpha} \stackrel{\text { def }}{=}\left(r_{i, \nu_{i \alpha}^{\ell}}, r_{i, \nu_{i \alpha}^{r}}\right]$ be the corresponding interval in $J_{i}(A)$.

Claim 5.5.
(a)

$$
\begin{equation*}
\sum_{i \in \operatorname{dom}(t)}\left|\operatorname{Ker}_{i}(t) \cup \bigcup_{\alpha=0}^{\ell_{i}} \hat{\Gamma}_{i \alpha}\right| \leq 14 k d ; \tag{30}
\end{equation*}
$$

(b) $\ell_{i} \geq 1$ for every $i \in[m]$. (That is, $y_{\emptyset}^{i}$ and $y_{J_{i}(A)}^{i}$ are not in the same connected component.)

Proof. (a) We have

$$
\begin{aligned}
\operatorname{Ker}_{i}(t) \cup \bigcup_{\alpha=0}^{\ell_{i}} \hat{\Gamma}_{i \alpha}= & \operatorname{Ker}_{i}(t) \cup \bigcup_{(\nu-1, \nu) \in G_{i}}^{\bigcup}\left(r_{i, \nu-1}, r_{i \nu}\right] \\
= & \operatorname{Ker}_{i}(t) \cup \bigcup_{(\nu-1, \nu) \in G_{i}}\left(\left(r_{i, \nu-1}, r_{i \nu}\right] \backslash \operatorname{Ker}_{i}(t)\right) \\
= & \left(J_{i}(A) \cap \bigcup_{i^{\prime} \in \operatorname{dom}(t) \backslash\{i\}} J_{i^{\prime}}(A)\right) \\
& \cup\left(J_{i}(A) \cap\left(J_{x}(t) \backslash \bigcup_{i^{\prime} \in \operatorname{dom}(t) \backslash\{i\}} J_{i^{\prime}}(A)\right)\right) \\
& \cup \bigcup_{(\nu-1, \nu) \in G_{i}}\left(\left(r_{i, \nu-1}, r_{i \nu}\right] \backslash \operatorname{Ker}_{i}(t)\right) .
\end{aligned}
$$

Accordingly,

$$
\begin{align*}
\sum_{i \in \operatorname{dom}(t)} & \left|\operatorname{Ker}_{i}(t) \cup \bigcup_{\alpha=0}^{\ell_{i}} \hat{\Gamma}_{i \alpha}\right| \\
\leq & \sum_{i \in \operatorname{dom}(t)}\left|J_{i}(A) \cap \bigcup_{i^{\prime} \in \operatorname{dom}(t) \backslash\{i\}} J_{i^{\prime}}(A)\right|  \tag{31}\\
& +\sum_{i \in \operatorname{dom}(t)}\left|J_{i}(A) \cap\left(J_{x}(t) \backslash \bigcup_{i^{\prime} \in \operatorname{dom}(t) \backslash\{i\}} J_{i^{\prime}}(A)\right)\right| \\
& +\sum_{i \in \operatorname{dom}(t)} \sum_{(\nu-1, \nu) \in G_{i}}\left|\left(r_{i, \nu-1}, r_{i \nu}\right] \backslash \operatorname{Ker}_{i}(t)\right| .
\end{align*}
$$

The first summand in the right-hand side is bounded by $k d$ due to the expansion property (1). Sets appearing in the second summand are disjoint, therefore, it is bounded by $\left|J_{x}(t)\right| \leq k$. Finally, $\left|\left(r_{i, \nu-1}, r_{i \nu}\right] \backslash \operatorname{Ker}_{i}(t)\right|<6 d$ for every $i \in \operatorname{dom}(t)$ and $(\nu-1, \nu) \in G_{i}$, and also $\sum_{i \in \operatorname{dom}(t)}\left|G_{i}\right| \leq \sum_{i \in \operatorname{dom}(t)}\left(k_{i}+1\right) \leq 2 k$. Therefore, the third summand is bounded by $12 k d$. (30) follows.

Part (b) of this claim immediately follows from part (a) and the bound (2).

Definition 5.6 (construction of $R(t)$ continued). Now fix one representative $\nu_{i \alpha}$ in every connected component $\Gamma_{i \alpha}$ such that $\nu_{i, 0}=0$ and $\nu_{i i_{i}}=k_{i}+1$. (Claim 5.5(b) guarantees that this is possible.) These will be the "anchor" variables mentioned before Definition 5.4, and they will serve as reference points for variables $y_{\Sigma_{\nu}^{i}}^{i}$ with $\nu \in \Gamma_{\alpha}$. Let $\varepsilon_{\alpha}^{i}$ be the sign with which the anchor variable $y_{\sum_{\nu_{i \alpha}}^{i}}^{i}$ appears in $t^{\prime}$. Then, since both $t^{\prime}$ and $R(t)$ assert that $y_{\sum_{\nu_{i \alpha}}^{i}}^{i}=\varepsilon_{\alpha}^{i}$, modulo this fact we can replace $y$-variables corresponding to all others $\nu \neq \nu_{i \alpha}$ in $\Gamma_{i \alpha}$ as a linear form in $x$-variables only.

Formally, we define the expression $\tilde{R}(t)$ as the result of replacing in the term $t^{\prime}$ all variables $y_{\Sigma_{\nu}^{i}}^{i}$ with $\nu \in \Gamma_{i \alpha}$ and $\nu \neq \nu_{i \alpha}$ by the following linear forms:

$$
\begin{equation*}
y_{\Sigma_{\nu}^{i}}^{i} \mapsto \bigoplus\left\{x_{j} \mid j \in \Sigma_{\nu}^{i} \Delta \Sigma_{\nu_{i \alpha}}^{i}\right\} \oplus \varepsilon_{\alpha}^{i} . \tag{32}
\end{equation*}
$$

Finally, we let $R(t)$ be the straightforward DNF expansion of $\tilde{R}(t)$, in which we also remove all "cosmetic" literals $\bar{y}_{\emptyset}^{i},\left(y_{J_{i}(A)}^{i}\right)^{b_{i}}$ inserted there at the beginning of the construction.

Claim 5.7.
(a) $R(t)$ is a protected DNF such that $|\operatorname{Vars}(R(t))| \leq O(k d)$.
(b) There exist $\operatorname{Res}(O(k d))$ inferences of $R(t)$ from $t, \tau_{\leq}(A, b)$ and, vice versa, of $t$ from $R(t), \tau_{\leq}(A, b)$ that have size $\exp (O(k d))$ and contain at most $O(k d)$ variables.

Proof. (a) The bound on $|\operatorname{Vars}(R(t))|$ follows from the construction and Claim 5.5(a). We protect terms in $R(t)$ as follows. Let $y_{\Sigma_{\nu_{i \alpha}}^{i}}^{i} \in \operatorname{Vars}(R(t))$. (Note that $\Sigma_{\nu_{i \alpha}}^{i} \notin\left\{\emptyset, J_{i}(A)\right\}$.) Protect this variable by the sets $L_{i \alpha}, R_{i \alpha}$, where $L_{i \alpha}$ consists of ( $3 d$ ) right-most points in $\left(r_{\nu_{i \alpha}^{\ell}-1}, r_{\nu_{i \alpha}^{\ell}}\right] \backslash \operatorname{Ker}_{i}(t)$, and $R_{i \alpha}$ consists of (3d) left-most points in $\left(r_{\nu_{i \alpha}^{r}}, r_{\nu_{i \alpha}^{r}+1}\right] \backslash \operatorname{Ker}_{i}(t)$. Let us check that all requirements of Definition 5.3 are fulfilled.

Since $\left(\nu_{i \alpha}^{\ell}-1, \nu_{i \alpha}^{\ell}\right),\left(\nu_{i \alpha}^{r}, \nu_{i \alpha}^{r}+1\right) \notin G_{i}$, the sets $\left(r_{\nu_{i \alpha}^{\ell}-1}, r_{\nu_{i \alpha}^{\ell}}\right] \backslash \operatorname{Ker}_{i}(t)$ and $\left(r_{\nu_{i \alpha}^{r}}, r_{\nu_{i \alpha}^{r}+1}\right] \backslash \operatorname{Ker}_{i}(t)$ have cardinality $\geq 6 d$ each, which implies (1). (2) is obvious.

For (3) note that the intervals $\operatorname{Conv}\left(L_{i \alpha}, R_{i \alpha}\right)$ are disjoint for every fixed $i$ and that $\operatorname{Conv}\left(L_{i \alpha}, R_{i \alpha}\right) \subseteq \operatorname{Ker}_{i}(t) \cup L_{i \alpha} \cup \hat{\Gamma}_{i \alpha} \cup R_{i \alpha}$. Hence

$$
\sum_{i \in \operatorname{dom}(t)} \sum_{\alpha}\left|\operatorname{Conv}\left(L_{i \alpha}, R_{i \alpha}\right)\right|=\sum_{i \in \operatorname{dom}(t)}\left|\bigcup_{\alpha} \operatorname{Conv}\left(L_{i \alpha}, R_{i \alpha}\right)\right| ;
$$

applying Claim 5.5(a) as well as the obvious estimate $\sum_{i \in \operatorname{dom}(t)}\left|\bigcup_{\alpha}\left(L_{i \alpha} \cup R_{i \alpha}\right)\right|$ $\leq 6 k d$, we get the required bound.

Finally, (4) immediately follows from the construction. Indeed, given any fixed protection ( $L_{i_{0} \alpha_{0}}, R_{i_{0} \alpha_{0}}$ ), all new $x$-variables introduced in $J_{i_{0}}(A)$ by the
substitution (32) either belong to $\operatorname{Ker}_{i_{0}}(t)$ or belong to $\bigcup_{\alpha} \hat{\Gamma}_{i_{0} \alpha}$. Both sets are disjoint with $L_{i_{0} \alpha_{0}}, R_{i_{0} \alpha_{0}}$.
(b) As we explained inside Definition 5.6, the axioms from (26) pertaining to the same connected component of $G_{i}$ semantically imply the equivalence of both sides in (32) in the presence of the literal $\left(y_{\sum_{\nu_{i \alpha}}^{i}}^{i}\right)^{\varepsilon_{\alpha}^{i}}$, contain the axioms $\bar{y}_{\emptyset}^{i},\left(y_{J_{i}(A)}^{i}\right)^{b_{i}}$, and thus semantically imply $t \equiv R(t)$. Now we only have to refer to the implicational completeness of $\operatorname{Res}(k)$ and to the well-known fact that everything provable in this theory (and, in fact, even in Resolution) also has a proof of size at most exponential in the number of variables.

Remark 6. It is worth noting that the construction of $R(t)$ in fact gives yet another property crucial for the argument: the protections $L_{i \alpha}, R_{i \alpha}$ are pairwise disjoint (even when $i$ varies). This property, however, will be reestablished in the proof of Claim 5.13 anyway and, for this reason, is not included in Definition 5.3.

Claim 5.8. Assume that $\tau_{\leq}(A, b)$ has a $\operatorname{Res}(k)$ refutation of size $S$. Then it also has an $\operatorname{Res}(O(k d))$ refutation of size $S \cdot \exp (O(k d))$ in which all lines are of the form $F \vee F^{\prime}$, where $F$ is a protected $O(k d)$-DNF and $\left|\operatorname{Vars}\left(F^{\prime}\right)\right| \leq O(k d)$.

Proof. For a $k$-DNF $F$, let $R(F) \stackrel{\text { def }}{=} \bigvee_{t \in F} R(t)$. Claim 5.7(b) implies that every axiom $C \in \tau_{\leq}(A, b)$ (of width $\leq 3$ ) gets converted to a DNF $R(C)$ that has an inference $P$ from $\tau_{\leq}(A, b)$ of size $\exp (O(k d))$ and with $|\operatorname{Vars}(P)| \leq$ $O(k d)$. Also, the image of any inference rule in Definition 2.6 is admissible, which can be seen by decoding and then encoding back the principal formula. For example, we simulate the $R$-image of AND-introduction as follows:

$$
\frac{\frac{R(F) \vee R\left(\ell_{1}\right)}{R(F) \vee \ell_{1}} \quad \cdots \quad \frac{R(F) \vee R\left(\ell_{w}\right)}{R(F) \vee \ell_{w}}}{\frac{R(F) \vee \bigwedge_{\nu=1}^{w} \ell_{\nu}}{R(F) \vee R\left(\bigwedge_{\nu=1}^{w} \ell_{\nu}\right)}}
$$

In order to do this encoding/decoding, we simply use the inferences from Claim 5.7(b) weakened by $R(F)$. Since weakening does not change the number of lines, this induced inference has all the required properties.

The claims proved so far will allow us to concentrate for the purpose of Height-reduction on protected DNFs, and we now define a distribution on sparse restrictions that, with an overwhelming probability, will reduce the representation height of any such DNF.

Definition 5.9. The random restriction $\boldsymbol{\rho}$ of the variables $\operatorname{Vars} \leq(A)$ is constructed as follows. Pick a subset $\boldsymbol{J} \subseteq[n]$ completely at random. Then for every $i \in[m]$, independently apply the following construction.

Pick a random set of variables $\tilde{\boldsymbol{Y}}^{i} \subseteq\left\{y_{\Sigma}^{i} \mid \Sigma\right.$ an initial segment of $\left.J_{i}(A)\right\}$ by including there the cosmetic variables $y_{\emptyset}^{i}$ and $y_{J_{i}(A)}^{i}$ with probability 1 , and every other $y_{\Sigma}^{i}$ with probability $1 /(2 d)$, independently of each other. Say that $y_{\Sigma}^{i}, y_{\Sigma^{\prime}}^{i} \in \tilde{\boldsymbol{Y}}^{i}$ with $\Sigma \subset \Sigma^{\prime}$ collide if $\left|\left(\Sigma^{\prime} \backslash \Sigma\right) \backslash \boldsymbol{J}\right|<2 d$. Remove from $\tilde{\boldsymbol{Y}}^{\boldsymbol{i}}$ all variables $y_{\Sigma}^{i} \notin\left\{y_{\emptyset}^{i}, y_{J_{i}(A)}^{i}\right\}$ that collide with at least one other variable in $\tilde{\boldsymbol{Y}}^{i}$ (possibly, with $y_{\emptyset}^{i}$ or $y_{J_{i}(A)}^{i}$ ). Let $\boldsymbol{Y}^{\boldsymbol{i}} \subseteq \tilde{\boldsymbol{Y}}^{i}$ be the resulting set.

By definition, $\boldsymbol{\rho}$ assigns $y_{\emptyset}^{i}$ to 0 , assigns $y_{J_{i}(A)}^{i}$ to $b_{i}$, and assigns all other variables in $\left\{x_{j} \mid j \in \boldsymbol{J}\right\} \cup \bigcup_{i=1}^{m} \boldsymbol{Y}^{\boldsymbol{i}}$ at random (and independently of each other).

Claim 5.10. $\mathbf{P}[\boldsymbol{\rho}$ is sparse $] \geq 1 / 2$.
Proof. By inspecting definitions, $\boldsymbol{\rho}$ may be not sparse only in the "pathological" case when for some $i \in[m]$, we have $\boldsymbol{Y}^{\boldsymbol{i}}=\left\{y_{\emptyset}^{i}, y_{J_{i}(A)}^{i}\right\}$. We bound the probability of this bad event separately for every $i \in[m]$.

By the bound (2), we may choose $s \geq \frac{C \log m}{9}$ disjoint intervals $\Delta_{1}, \ldots, \Delta_{s}$ in $J_{i}(A)$, of length $9 d$ each. Subdivide every $\Delta_{\nu}$ into three sub-intervals $\Delta_{\nu}^{\ell}, \Delta_{\nu}^{c}, \Delta_{\nu}^{r}$, where $\left|\Delta_{\nu}^{\ell}\right|=\left|\Delta_{\nu}^{r}\right|=4 d$ and $\left|\Delta_{\nu}^{c}\right|=d$. Then the following events:

- $\left|\Delta_{\nu}^{\ell} \cap \boldsymbol{J}\right| \leq 2 d,\left|\Delta_{\nu}^{r} \cap \boldsymbol{J}\right| \leq 2 d$;
- $\tilde{\boldsymbol{Y}}^{i}$ contains exactly one variable $y_{\Sigma}^{i}$ with $r(\Sigma) \in \Delta_{\nu}^{c}$ and no variables with $r(\Sigma) \in \Delta_{\nu}^{\ell} \cup \Delta_{\nu}^{r}$
have probability $\Omega(1)$ each, are independent, and logically imply that the variable $y_{\Sigma}^{i} \in \tilde{\boldsymbol{Y}}^{i}$ with $r(\Sigma) \in \Delta_{\nu}^{c}$ does not collide with any other variable and hence stays in $\boldsymbol{Y}^{i}$. Therefore, the probability that this happens for at least one $\nu \in[s]$ (and in particular $\boldsymbol{Y}^{i} \neq\left\{y_{\emptyset}^{i}, y_{J_{i}(A)}^{i}\right\}$ ) is at least $1-1 /(2 m)$, provided the constant $C$ in (2) is large enough. By the union bound, this implies $\mathbf{P}\left[\forall i \in[m]\left(\boldsymbol{Y}^{\boldsymbol{i}} \neq\left\{y_{\emptyset}^{i}, y_{J_{i}(A)}^{i}\right\}\right)\right] \geq 1 / 2$.

We are going to apply Lemma 4.4 to show that for any protected DNF, $h\left(\left.F\right|_{\rho}\right)$ is small with high probability. For that we need to know that the restriction $\boldsymbol{\rho}$, when restricted to the set of variables $\operatorname{Vars}(F)$, satisfies the weak independence property from Definition 4.3. The intuitive reason why it should be the case is already suggested by the proof of Claim 5.10. (The role of the "wings" $\Delta_{\nu}^{\ell}, \Delta_{\nu}^{r}$ in that proof will be played by protections.) The major problem is, of course, that protections need not be disjoint, may be inconsistent for different occurrences of the same variable etc. We circumvent this by showing that $F$ has a relatively small fractional cover by sub-DNFs for which these problems already do not occur, whereupon we will apply Lemma 4.5.

More notation. Given a protected DNF $F$, we fix once and for all protections $\left\{L_{i \Sigma}(t), R_{i \Sigma}(t)\right\}$ for every $t \in F$. Let also $P_{i \Sigma}(t) \stackrel{\text { def }}{=} L_{i \Sigma}(t) \cup R_{i \Sigma}(t)$.

Definition 5.11. A protected DNF $F$ is weakly regular if the following three properties hold:
(1) the sets $\bigcup_{t \in F} J_{x}(t)$ and $\bigcup_{t \in F} \bigcup_{i, \Sigma} P_{i \Sigma}(t)$ are disjoint;
(2) for every $y_{\Sigma}^{i} \in \operatorname{Vars}(F)$, the protection $\left(L_{i \Sigma}(t), R_{i \Sigma}(t)\right)$ does not actually depend on the term $t$ (and henceforth will be denoted simply by ( $\left.L_{i \Sigma}, R_{i \Sigma}\right)$ );
(3) for every fixed $i \in[m]$, intervals $\operatorname{Conv}\left(L_{i \Sigma}, R_{i \Sigma}\right)$, where $\Sigma$ runs over all $\leq_{i}$-initial segments with $y_{\Sigma}^{i} \in \operatorname{Vars}(F)$, are pairwise disjoint.

Remark 7. The adjective "weakly" refers to the fact that we do not require the protections $P_{i \Sigma}$ to be disjoint for different i. Recall (Remark 6) that this disjointness property automatically follows from our construction for protections in any fixed term $t \in F$. This, however, seems to be of absolutely no help whatsoever for the uniform version, when $t$ varies. The only way to enforce the disjointness of protections uniformly that we know of additionally requires $|F|$ to be small, and for that reason it will be incorporated in the proof of Claim 5.13.

Claim 5.12. For every protected $O(k d)$-DNF $F$, there exists a random weakly regular sub-DNF $\boldsymbol{G} \subseteq F$ such that $\min _{t \in F} \mathbf{P}[t \in \boldsymbol{G}] \geq \exp (-O(k d))$.

Proof. The proof consists of three independent steps, and at every step we enforce one property required in Definition 5.11.

Step 1. For a colouring $\chi:[n] \longrightarrow\{0,1\}$, let

$$
G_{\chi} \stackrel{\text { def }}{=}\left\{t \in F \mid \chi\left(J_{x}(t)\right) \equiv 0 \wedge \chi\left(\bigcup_{i, \Sigma} P_{i \Sigma}(t)\right) \equiv 1\right\}
$$

Then for every $\chi, G_{\chi}$ has property (1) in Definition 5.11. Let $\chi:[n] \longrightarrow\{0,1\}$ be picked completely at random. Then, due to property (3) in Definition 5.3, $\mathbf{P}\left[t \in G_{\chi}\right] \geq \exp (-O(k d))$ for every particular $t \in F$.

Step 2. The idea behind enforcing property (2) in Definition 5.11 is similar to Step 1, but calculations get substantially more involved. For any system of protections $\vec{P}=\left(P_{i \Sigma} \mid y_{\Sigma}^{i} \in \operatorname{Vars}(F)\right)$, the sub-DNF

$$
G_{\vec{P}} \stackrel{\text { def }}{=}\left\{t \in F \mid \forall y_{\Sigma}^{i} \in \operatorname{Vars}(t)\left(P_{i \Sigma}(t)=P_{i \Sigma}\right)\right\}
$$

has the required property (2). We only need to construct a random system $\overrightarrow{\boldsymbol{P}}$ in such a way that

$$
\begin{equation*}
\min _{t \in F} \mathbf{P}\left[t \in G_{\overrightarrow{\boldsymbol{P}}}\right] \geq \exp (-O(k d)) \tag{33}
\end{equation*}
$$

All $\boldsymbol{P}_{\boldsymbol{i} \boldsymbol{\Sigma}}$ will be mutually independent. In order to construct $\boldsymbol{P}_{\boldsymbol{i} \boldsymbol{\Sigma}}$ for a fixed variable $y_{\Sigma}^{i} \in \operatorname{Vars}(F)$, first pick an integer $\boldsymbol{\mu}_{\boldsymbol{i} \boldsymbol{\Sigma}} \geq 6 d$ according to the distribution $\mathbf{P}\left[\boldsymbol{\mu}_{i \boldsymbol{\Sigma}}=\mu\right]=2^{6 d-\mu-1}(\mu \geq 6 d)$. Next, let $\boldsymbol{\Delta}_{i \boldsymbol{\Sigma}}$ be the interval of length $2 \boldsymbol{\mu}_{\boldsymbol{i} \boldsymbol{\Sigma}}$ centered at $r(\Sigma)$. (If $|\Sigma|<\boldsymbol{\mu}_{\boldsymbol{i} \boldsymbol{\Sigma}}$ or $\left|J_{i}(A) \backslash \Sigma\right|<\boldsymbol{\mu}_{\boldsymbol{i} \boldsymbol{\Sigma}}$, then we abort the construction and output $\boldsymbol{P}_{\boldsymbol{i \Sigma}}$ arbitrarily.) Finally, pick $\boldsymbol{P}_{\boldsymbol{i \Sigma}}$ as a random subset of $\boldsymbol{\Delta}_{\boldsymbol{i} \boldsymbol{\Sigma}}$ that has cardinality $6 d$.

Consider an arbitrary $t \in F$, and let $\mu_{i \Sigma}(t) \stackrel{\text { def }}{=}\left|\operatorname{Conv}\left(L_{i \Sigma}(t), R_{i \Sigma}(t)\right)\right|$. Then $\mathbf{P}\left[\boldsymbol{\mu}_{\boldsymbol{i} \boldsymbol{\Sigma}}=\mu_{i \Sigma}(t)\right] \geq \exp \left(-O\left(\mu_{i \Sigma}(t)\right)\right)$ and $\mathbf{P}\left[\boldsymbol{P}_{\boldsymbol{i} \boldsymbol{\Sigma}}=P_{i \Sigma}(t) \mid \boldsymbol{\mu}_{\boldsymbol{i} \boldsymbol{\Sigma}}=\mu_{i \Sigma}(t)\right]$ $\geq\binom{ 2 \mu_{i \Sigma}(t)}{6 d}^{-1} \geq\left(\frac{\mu_{i \Sigma}(t)}{d}\right)^{-O(d)}$. Combining these bounds together, we get $\mathbf{P}\left[\boldsymbol{P}_{\boldsymbol{i} \boldsymbol{\Sigma}}=P_{i \Sigma}(t)\right] \geq \exp \left(-O\left(\mu_{i \Sigma}(t)\right)\right) \cdot\left(\frac{\mu_{i \Sigma}(t)}{d}\right)^{-O(d)}$. Multiplying over all $y_{\Sigma}^{i} \in$ $\operatorname{Vars}(t)$,

$$
\mathbf{P}\left[t \in G_{\overrightarrow{\boldsymbol{P}}}\right] \geq \prod_{y_{\Sigma}^{i} \in \operatorname{Vars}(t)}\left\{\exp \left(-O\left(\mu_{i \Sigma}(t)\right)\right) \cdot\left(\frac{\mu_{i \Sigma}(t)}{d}\right)^{-O(d)}\right\}
$$

Since $\sum_{y_{\Sigma}^{i} \in \operatorname{Vars}(t)} \mu_{i \Sigma}(t) \leq O(k d)$ by property (3) in Definition 5.3, the first term in this product is bounded from below by $\exp (-O(k d))$. In order to bound the second term, we apply the arithmetic-geometric mean inequality. Denoting the number of $y$-variables in $t$ by $w$ (which, due to properties (2) and (3) in Definition 5.3, is bounded by $O(k)$ ), we have the calculation

$$
\begin{aligned}
\prod_{y_{\Sigma}^{i} \in \operatorname{Vars}(t)}\left(\frac{\mu_{i \Sigma}(t)}{d}\right)^{-O(d)} & \geq\left(\frac{\sum_{y_{\Sigma}^{i} \in \operatorname{Vars}(t)} \mu_{i \Sigma}(t)}{d w}\right)^{-O(d w)} \\
& \geq(O(k / w))^{-O(d w)} \geq \exp (-O(k d))
\end{aligned}
$$

This proves (33) and concludes the analysis at Step 2.
Step 3 . For this step we additionally assume that $F$ already satisfies the consistency property (2) in Definition 5.11 and let $\vec{P}$ be the corresponding system of protections of the variables $y_{\Sigma}^{i} \in \operatorname{Vars}(F)$. Choose $\tilde{\boldsymbol{Y}}^{i}$ according to Definition 5.9. Say that $y_{\Sigma}^{i}, y_{\Sigma^{\prime}}^{i} \in \operatorname{Vars}(F) \cap \tilde{\boldsymbol{Y}}^{i} \vec{P}$-collide if $\operatorname{Conv}\left(L_{i \Sigma}, R_{i \Sigma}\right) \cap$ $\operatorname{Conv}\left(L_{i \Sigma^{\prime}}, R_{i \Sigma^{\prime}}\right) \neq \emptyset$. For every $\vec{P}$-colliding pair $y_{\Sigma}^{i}, y_{\Sigma^{\prime}}^{i}$ identify arbitrarily any one point $j$ in the above intersection, and remove from $\tilde{\boldsymbol{Y}}^{i}$ that variable $y_{\Sigma_{0}}^{i}\left(\Sigma_{0} \in\left\{\Sigma, \Sigma^{\prime}\right\}\right)$ for which the $\leq_{i}$-distance between $r\left(\Sigma_{0}\right)$ and $j$ is larger (both $y_{\Sigma}^{i}$ and $y_{\Sigma^{\prime}}^{i}$ if these distances are equal). Let $\boldsymbol{Y}^{i}(\vec{P})$ be the set of remaining variables.

Clearly, the sub-DNF

$$
G_{\boldsymbol{Y}(\vec{P})} \stackrel{\text { def }}{=}\left\{t \in F \mid \operatorname{Vars}(t) \subseteq\left\{x_{1}, \ldots, x_{n}\right\} \cup \bigcup_{i \in \operatorname{dom}(t)}^{\bigcup} \boldsymbol{Y}^{i}(\vec{P})\right\}
$$

has the required property (3). Further, every variable $y_{\Sigma}^{i}$ may be removed from $\tilde{\boldsymbol{Y}}^{i}$ only if it $\vec{P}$-collides with some $y_{\Sigma^{\prime}}^{i}$ such that the $\leq_{i}$-distance between $r(\Sigma)$ and $r\left(\Sigma^{\prime}\right)$ does not exceed $2\left|\operatorname{Conv}\left(L_{i \Sigma}, R_{i \Sigma}\right)\right|$.

Now, for every particular term $t \in F$, mark all those variables $y_{\Sigma^{\prime}}^{i}$ in $\operatorname{Vars}(F)$ for which the above bound on the $\leq_{i}$-distance holds for at least one $y_{\Sigma}^{i} \in \operatorname{Vars}(t)$. Property (3) in Definition 5.3 implies that altogether we have marked at most $O(k d)$ variables. Therefore, the event "for every marked variable $y_{\Sigma}^{i}, y_{\Sigma}^{i} \in \bigcup_{i \in \operatorname{dom}(t)} \tilde{\boldsymbol{Y}}^{i}$ if and only if $y_{\Sigma}^{i} \in \operatorname{Vars}(t)$ " has probability $\geq \exp (-O(k d))$. On the other hand, since variables in $\operatorname{Vars}(t)$ never $\vec{P}$-collide with each other (due to property (1) in Definition 5.3), this event logically implies that $\operatorname{Vars}(t) \subseteq\left\{x_{1}, \ldots, x_{n}\right\} \cup \bigcup_{i \in \operatorname{dom}(t)} \boldsymbol{Y}^{\boldsymbol{i}}(\vec{P})$. Thus, $\mathbf{P}\left[t \in G_{\boldsymbol{Y}(\vec{P})}\right] \geq \exp (-O(k d))$, which completes the analysis of Step 3 .

Combining things together, the random sub-DNF $G_{\boldsymbol{\chi}} \cap G_{\overrightarrow{\boldsymbol{P}}} \cap G_{\boldsymbol{Y}(\overrightarrow{\boldsymbol{P}})}$ (where $\chi, \overrightarrow{\boldsymbol{P}}$ and the auxiliary random variables $\tilde{\boldsymbol{Y}}^{i}$ are independent) has all the required properties. This completes the proof of Claim 5.12.

We assume that any weakly regular DNF $F$ is weighted according to the following weight function $\mu_{F}: \mu_{F}\left(x_{j}\right) \stackrel{\text { def }}{=} 1$ and $\mu_{F}\left(y_{\Sigma}^{i}\right) \stackrel{\text { def }}{=}\left|\operatorname{Conv}\left(L_{i \Sigma}, R_{i \Sigma}\right)\right|$.

Claim 5.13. There exists an absolute constant $p>0$ such that for every weakly regular protected DNF $F$, the random restriction $\rho$ restricted to the variables $\operatorname{Vars}(F)$ is $\left(r, \mu_{F}, p\right)$-independent.

Proof. Let $Z \subseteq \operatorname{Vars}(F)$ be a set of variables with $|Z| \leq r$. Denote by $I$ the set of all $i \in[m]$ for which $Z$ contains at least one variable $y_{\Sigma}^{i}$. Clearly, $|I| \leq r$. Therefore, by the expansion property (1), there exists $i_{0} \in I$ such that

$$
\begin{equation*}
\left|J_{i_{0}}(A) \cap \bigcup_{i \in I \backslash\left\{i_{0}\right\}} J_{i}(A)\right| \leq d \tag{34}
\end{equation*}
$$

This implies that for every $\Sigma$ with $y_{\Sigma}^{i_{0}} \in \operatorname{Vars}(F)$, there exist $L_{i_{0} \Sigma}^{\prime} \subseteq L_{i_{0} \Sigma}$ and $R_{i_{0} \Sigma}^{\prime} \subseteq R_{i_{0} \Sigma}$ of size $2 d$ each that are disjoint with $\bigcup_{i \in I \backslash\left\{i_{0}\right\}} J_{i}(A)$. In particular, these $L_{i_{0} \Sigma}^{\prime}, R_{i_{0} \Sigma}^{\prime}$ are disjoint not only between themselves (by property (3) in Definition 5.11), but also with any other $P_{i \Sigma}$ when $i \neq i_{0}$. Next, choose $i_{1} \in I \backslash\left\{i_{0}\right\}$ such that $\left|J_{i_{1}}(A) \cap \bigcup_{i \in I \backslash\left\{i_{0}, i_{1}\right\}} J_{i}(A)\right| \leq d$ and repeat this procedure until we find pairwise disjoint subsets $L_{i \Sigma}^{\prime} \subseteq L_{i \Sigma}, R_{i \Sigma}^{\prime} \subseteq R_{i \Sigma}$ for all $y_{\Sigma}^{i} \in Z$. As before, let $P_{i \Sigma}^{\prime} \stackrel{\text { def }}{=} L_{i \Sigma}^{\prime} \cup R_{i \Sigma}^{\prime}$.

Now, the space $\Omega$ required in Definition 4.1 consists of three parts: $\Omega=$ $\Omega_{1} \times \Omega_{2} \times \Omega_{3}$, and $\boldsymbol{\omega}=\left(\boldsymbol{\omega}_{\mathbf{1}}, \boldsymbol{\omega}_{\mathbf{2}}, \boldsymbol{\omega}_{\mathbf{3}}\right)$, where $\boldsymbol{\omega}_{\boldsymbol{i}} \in \Omega_{i}$ are independent. The role of $\boldsymbol{\omega}_{\mathbf{1}}$ is played by the random set $\left\{x_{j} \mid j \in \boldsymbol{J}\right\} \cup \bigcup_{i=1}^{m} \tilde{\boldsymbol{Y}}^{i}$ from Definition 5.9, and $\omega_{2}$ consists of additional independent Bernoulli variables used for assigning $\left\{x_{j} \mid j \in \boldsymbol{J}\right\} \cup \bigcup_{i=1}^{m} \boldsymbol{Y}^{\boldsymbol{i}} . \pi$ depends only on $\omega_{1}, \omega_{2}$ and represents the construction of the random variable $\boldsymbol{\rho}$ from that definition.

Next, we construct the second required mapping $\pi^{\prime}\left(\omega_{1}, \omega_{2}, \omega_{3}\right)$ as follows. Let $J \stackrel{\text { def }}{=}\left\{j \mid x_{j} \in \omega_{1}\right\}$, and let $\tilde{Y}^{i}$ consist of all $y_{\Sigma}^{i} \in \omega_{1}$. First, define from $J$ and $\tilde{Y}^{i}$ random subsets $J^{\prime} \subseteq J$ and $\left(\boldsymbol{Y}^{i}\right)^{\prime} \subseteq \tilde{Y}^{i} \cap Z$. (The third random variable $\omega_{\mathbf{3}}$ accounts for the extra randomness used in this construction.) Let $j \in J^{\prime}$ with probability $2 p$, independently for all $j \in J \cap Z$. If $y_{\Sigma}^{i} \in \tilde{Y}^{i} \cap Z$ and either $J \cap P_{i \Sigma}^{\prime} \neq \emptyset$ or $\tilde{Y}^{i}$ contains any other $y_{\Sigma^{\prime}}^{i}$ with $r\left(\Sigma^{\prime}\right) \in \operatorname{Conv}\left(L_{i \Sigma}^{\prime}, R_{i \Sigma}^{\prime}\right)$, then $y_{\Sigma}^{i} \notin\left(\boldsymbol{Y}^{i}\right)^{\prime}$. Otherwise, $y_{\Sigma}^{i}$ is included into $\left(\boldsymbol{Y}^{i}\right)^{\prime}$ with some probability $p_{i \Sigma}$ to be specified later, independently for all variables $y_{\Sigma}^{i}$ (and independently of $\left.\boldsymbol{J}^{\prime}\right)$. Finally, $\pi^{\prime}\left(\omega_{1}, \omega_{2}, \omega_{3}\right)$ assigns variables in $\left\{x_{j} \in Z \mid j \in \boldsymbol{J}^{\prime}\right\} \cup \bigcup_{i \in I}\left(\boldsymbol{Y}^{i}\right)^{\prime}$ using $\omega_{2}$.
$y_{\Sigma}^{i} \in\left(\boldsymbol{Y}^{i}\right)^{\prime}$ can take place only if $J \cap P_{i \Sigma}^{\prime}=\emptyset$ and $\tilde{Y}^{i}$ does not contain any $y_{\Sigma^{\prime}}^{i}$ with $r\left(\Sigma^{\prime}\right) \in \operatorname{Conv}\left(L_{i \Sigma}^{\prime}, R_{i \Sigma}^{\prime}\right)$ other than $y_{\Sigma}^{i}$. This implies that $y_{\Sigma}^{i}$ cannot collide with any other variable in $\tilde{Y}^{i}$. Therefore, $\left(\boldsymbol{Y}^{i}\right)^{\prime} \subseteq \boldsymbol{Y}^{i}$, and $\pi^{\prime}\left(\omega_{1}, \omega_{\mathbf{2}}, \omega_{\mathbf{3}}\right)$ is a sub-restriction of $\pi\left(\omega_{\mathbf{1}}, \omega_{\mathbf{2}}\right) \approx \rho$.

We are only left to show that for a suitable choice of the probabilities $p_{i \Sigma}$, $\pi^{\prime}\left(\boldsymbol{\omega}_{\mathbf{1}}, \boldsymbol{\omega}_{\mathbf{2}}, \boldsymbol{\omega}_{\mathbf{3}}\right)$ will be equidistributed with $\boldsymbol{\rho}_{\mu_{\boldsymbol{F}}, Z, \boldsymbol{p}}$.

The properties listed in Definition 5.11, along with the disjointness property of $L_{i \Sigma}^{\prime}, R_{i \Sigma}^{\prime}$ ensured at the beginning of our proof, imply that the facts $j \in \boldsymbol{J}^{\prime}, y_{\Sigma}^{i} \in\left(\boldsymbol{Y}^{i}\right)^{\prime}\left(x_{j}, y_{\Sigma}^{i} \in Z\right)$ depend on the behaviour of $J, \tilde{Y}^{i}$ on pairwise different variables. Therefore, if we also randomize over $\boldsymbol{J}, \tilde{\boldsymbol{Y}}^{i}$, all these events are independent of each other. Denote by $A$ the following event: " $\boldsymbol{J} \cap P_{i \Sigma}^{\prime}=\emptyset$ and $y_{\Sigma}^{i} \in \tilde{\boldsymbol{Y}}^{i}$ and $\tilde{\boldsymbol{Y}}^{i}$ does not contain any $y_{\Sigma^{\prime}}^{i}$ with $r\left(\Sigma^{\prime}\right) \in \operatorname{Conv}\left(L_{i \Sigma}^{\prime}, R_{i \Sigma}^{\prime}\right)$ other than $y_{\Sigma}^{i}$ ": this is exactly the prerequisite for including $y_{\Sigma}^{i}$ into $\left(\boldsymbol{Y}^{i}\right)^{\prime}$. The last remark implies that the three parts of this event are independent, and hence we conclude

$$
\begin{aligned}
\mathbf{P}[A]=2^{-4 d} \frac{1}{2 d}\left(1-\frac{1}{2 d}\right)^{\left|\operatorname{Conv}\left(L_{i \Sigma}^{\prime}, R_{i \Sigma}^{\prime}\right)\right|-1} & \geq 2^{-4 d} \cdot \frac{1}{2 d}\left(1-\frac{1}{2 d}\right)^{\mu\left(y_{\Sigma}^{i}\right)-1} \\
& \geq \exp \left(-O\left(\mu\left(y_{\Sigma}^{i}\right)\right)\right),
\end{aligned}
$$

where for the last estimate we used the obvious bound $\mu\left(y_{\Sigma}^{i}\right) \geq 6 d$. Hence,

$$
\begin{equation*}
\mathbf{P}\left[y_{\Sigma}^{i} \in\left(\boldsymbol{Y}^{i}\right)^{\prime}\right]=\mathbf{P}[A] \cdot \mathbf{P}\left[y_{\Sigma}^{i} \in\left(\boldsymbol{Y}^{i}\right)^{\prime} \mid A\right] \geq \exp \left(-O\left(\mu\left(y_{\Sigma}^{i}\right)\right)\right) p_{i \Sigma} . \tag{35}
\end{equation*}
$$

Therefore, if $p>0$ is small enough, the probabilities $p_{i \Sigma} \leq 1$ can be chosen in such a way that $\mathbf{P}\left[y_{\Sigma}^{i} \in\left(\boldsymbol{Y}^{i}\right)^{\prime}\right]=p^{\mu\left(y_{\Sigma}^{i}\right)}$ and, clearly, $\mathbf{P}\left[j \in \boldsymbol{J}^{\prime}\right]=p$ for $x_{j} \in Z$. Thus, $\pi^{\prime}\left(\boldsymbol{\omega}_{\mathbf{1}}, \boldsymbol{\omega}_{\mathbf{2}}, \boldsymbol{\omega}_{\mathbf{3}}\right)$ has the same distribution as $\boldsymbol{\rho}_{\boldsymbol{\mu}_{F}, \boldsymbol{Z}, \boldsymbol{p}}$, which completes the proof of Claim 5.13.

Now we have at our disposal all tools necessary for proving the following switching lemma for protected DNFs.

Claim 5.14. Let $F$ be a protected $O(k d)$-DNF. Then for every $h \leq r$,

$$
\mathbf{P}\left[h\left(\left.F\right|_{\rho}\right) \geq h\right] \leq \exp \left(-h / 2^{O(k d)}\right) .
$$

Proof. Assume first that $F$ is weakly regular, and let $\mu=\mu_{F}$. Then property (3) in Definition 5.3 implies that $F$ is even a weighted $O(k d)$-DNF. With this remark, our claim for weakly regular $F$ immediately follows from Lemma 4.4 and Claim 5.13.

In order to generalize this to the case of arbitrary protected $O(k d)$-DNF, we use Lemma 4.5 in combination with Claim 5.12, just in the same way as at the end of Section 4.

Now we can easily finish the proof of Theorem 2.7. Let $P$ be a $\operatorname{Res}(k)$ refutation of $\tau_{\leq}(A, b)$ that has size $S$. We need to prove that $S \geq \exp \left(r / 2^{O(k d)}\right)$.

Applying Claim 5.8, for some $K=O(k d)$ we get a $\operatorname{Res}(K)$ refutation $P^{\prime}$ of $\tau_{\leq}(A, b)$ that has size $\leq S 2^{K}$, and in which every line has the form $F \vee F^{\prime}$, where $F$ is a protected $K$-DNF and $\left|\operatorname{Vars}\left(F^{\prime}\right)\right| \leq K$. By Claims 5.10 and 5.2,
(36) $\mathbf{P}\left[\right.$ every resolution refutation of $\left.\tau_{\leq}(A, b)\right|_{\rho}$ has width $\left.>r / 4\right] \geq 1 / 2$.

Comparing this with Proposition $3.6(k:=K$ and $h:=r /(4 K))$, we get

$$
\begin{equation*}
\mathbf{P}\left[\exists G \in P^{\prime}\left(h\left(\left.G\right|_{\rho}\right)>r /(4 K)\right)\right] \geq 1 / 2 . \tag{37}
\end{equation*}
$$

On the other hand,

$$
\begin{equation*}
\mathbf{P}\left[\exists G \in P^{\prime}\left(h\left(\left.G\right|_{\rho}\right)>r /(4 K)\right)\right] \leq S 2^{K} \cdot \max _{G \in P^{\prime}} \mathbf{P}\left[h\left(\left.G\right|_{\rho}\right)>r /(4 K)\right], \tag{38}
\end{equation*}
$$

and we treat every line $G \in P^{\prime}$ individually. Let $G=F \vee F^{\prime}$, where $F$ is a protected $K$-DNF and $\left|\operatorname{Vars}\left(F^{\prime}\right)\right| \leq K$. Obviously, $h\left(\left.F^{\prime}\right|_{\rho}\right) \leq K$, and we can assume without loss of generality that $K<r /(8 K)$. (Otherwise, the bound we are proving becomes trivial.) Thus,

$$
\begin{equation*}
\mathbf{P}\left[h\left(\left.G\right|_{\rho}\right)>r /(4 K)\right] \leq \mathbf{P}\left[h\left(\left.F\right|_{\rho}\right)>r /(8 K)\right] . \tag{39}
\end{equation*}
$$

Applying Claim 5.14 (with $h:=r /(8 K)-1$ ), we find

$$
\begin{equation*}
\mathbf{P}\left[h\left(\left.F\right|_{\boldsymbol{\rho}}\right)>r /(8 K)\right] \leq \exp \left(-r / 2^{O(k d)}\right) \tag{40}
\end{equation*}
$$

Theorem 2.7 follows by comparing (40), (39) and (38) with (37).

## 6. Stretching the number of output bits

In this section we prove Theorems 2.10 and 2.12.
Proof of Theorem 2.10. Let $A$ be an $m \times n(r, d)$-lossless expander such that the bound (5) holds, and let $\leq$ be an arbitrary ordering of $A$. Let $S$ be the minimal size of a $\operatorname{Res}(k)$ refutation $P$ of any CNF having the form

$$
\begin{equation*}
\bigwedge_{\nu=1}^{H} \tau_{C_{\leq, A}}\left(x_{1}^{(\nu)}, \ldots, x_{n}^{(\nu)}, \vec{y}^{(\nu)}, q_{1}^{(\nu)}, \ldots, q_{m}^{(\nu)}\right), \tag{41}
\end{equation*}
$$

from Definition 2.9. First we remark that we may assume without loss of generality that $H \leq S$.

Indeed, $P$ may contain at most $S$ axioms; therefore, there exist at most $S$ indices $\nu \in[H]$ for which $P$ contains an axiom from

$$
\tau_{C_{\leq, A}}\left(x_{1}^{(\nu)}, \ldots, x_{n}^{(\nu)}, \vec{y}^{(\nu)}, q_{1}^{(\nu)}, \ldots, q_{m}^{(\nu)}\right)
$$

Remove from (41) all other conjunctive terms; then $P$ will still be a $\operatorname{Res}(k)$ refutation of the resulting sub-CNF. This sub-CNF itself has the form (41), with the only exception that some $q_{i}^{(\nu)}$ may be equal to variables not appearing in the lists $\vec{x}^{(\nu)}, \vec{y}^{(\nu)}$. Substituting them arbitrarily with Boolean constants 0,1 , we get a CNF of the form (41) with $H \leq S$ that still has a $\operatorname{Res}(k)$ refutation of size $S$.

Note that (41) encodes the system of linear equations

$$
\bigoplus_{j \in J_{i}(A)} x_{j}^{(\nu)}=q_{i}^{(\nu)}(\nu \in[H], i \in[m]) .
$$

Let us transfer those $q_{i}^{(\nu)}$ that actually appear in the list $\left\{x_{j}^{(\mu)} \mid j \in[n], \mu<\nu\right\}$ to the left-hand side. Then we get a linear system with constant right-hand side, and it also has the form $\hat{A} X=\hat{b}$ for some matrix $\hat{A}$ and vector $\hat{b}$. It will turn out that $\hat{A}$ has almost as good expansion properties as $A$ itself, and we will apply to it Theorem 2.7.

In order to formalize this intuition, let $A^{H}$ be the direct sum of $H$ copies of $A$. That is, rows of $A^{H}$ are indexed by $[H] \times[m]$, columns are indexed by $[H] \times[n]$, and

$$
a_{(\nu, i),(\mu, j)}^{H} \stackrel{\text { def }}{=} \begin{cases}a_{i j} & \text { if } \nu=\mu, \\ 0 & \text { if } \nu \neq \mu .\end{cases}
$$

Let $\hat{A}$ be obtained from $A^{H}$ by additionally setting $\hat{a}_{(\nu, i),(\mu, j)}:=1$ whenever $q_{i}^{(\nu)}=x_{j}^{(\mu)}$. Also let

$$
\hat{b}_{(\nu, i)} \stackrel{\text { def }}{=} \begin{cases}q_{i}^{(\nu)} & \text { if } q_{i}^{(\nu)} \text { is a Boolean constant }, \\ 0 & \text { if } q_{i}^{(\nu)} \text { is a variable }\end{cases}
$$

Finally, we order the rows of $A^{H}$ according to the ordering $\leq$ of the matrix $A$. Whenever we add in $\hat{A}$ the new 1 -entry $(\mu, j)$ to the row $(\nu, i)$ (that is, when $q_{i}^{(\nu)}=x_{j}^{(\mu)}$ ), we declare it to be the largest element in $J_{(i, \nu)}(\hat{A})$. Denote the resulting ordering of $\hat{A}$ by $\hat{\leq}$.

Given the above intuition, it is straightforward to check that there exists a (naturally defined) variable substitution that takes every clause in (41) into a clause having a resolution inference from $\tau_{\hat{\jmath}}(\hat{A}, \hat{b})$ of size $O(1)$. This implies that $\tau_{\hat{<}}(\hat{A}, \hat{b})$ has a $\operatorname{Res}(k)$ refutation of size $O(S)$.

On the other hand, it is easy to see that the direct sum of $(r, d)$-lossless expanders is still an $(r, d)$-lossless expander. Also, if we append at most one

1-entry per row to an $(r, d)$-lossless expander, we come up with an $(r, d+1)$ lossless expander. Applying Theorem 2.7, we see that either condition (2) is violated for $\hat{A}$ or we have $S \geq \exp \left(r / 2^{O(k d)}\right)$ (in which case we are done). It remains to note that in the first case we have $s \leq C d(k+\log (m H))$, which implies (if the constant in (5) is twice as large as the constant in (2)) that $H \geq \exp (\Omega(s / d))$. Since $S \geq H$, Theorem 2.10 follows in this case as well.

Proof of Theorem 2.12 (cf. [Kra04]). Given any Boolean circuit $C$ with $n$ inputs and $2 n$ outputs and any $b \in\{0,1\}^{2^{h \cdot n}}, \tau\left(C^{h}, b\right)$ can be expressed in the form (4) as $\bigwedge_{|u| \leq h-1} \tau_{C}\left(x_{1}^{(u)}, \ldots, x_{n}^{(u)}, \vec{y}^{(u)}, q_{1}^{(u)}, \ldots, q_{2 n}^{(u)}\right)$, where

$$
q_{i}^{(u)} \stackrel{\text { def }}{=} \begin{cases}x_{i}^{(u * 0)} & \text { if }|u|<h-1, i \leq n, \\ x_{n-i}^{(u * 1)} & \text { if }|u|<h-1, i \geq n+1, \\ b_{(u * 0), i} & \text { if }|u|=h-1, i \leq n, \\ b_{(u * 1), i} & \text { if }|u|=h-1, i \geq n+1 .\end{cases}
$$

Now Theorem 2.12 immediately follows from Theorem 2.10.

## 7. Random matrices have good expansion properties

Statements of this sort have been reappearing in the literature at a steady rate beginning from [Pin73]. However, we have not been able to find any particular source handling the matter in the generality needed for our purposes. Thus, we prove Theorem 2.5 from scratch. (This is not hard anyway.)

Let

$$
r \stackrel{\text { def }}{=} n^{\delta}, \quad d \stackrel{\text { def }}{=} C \frac{\log m}{\log n}
$$

where $\delta$ is a sufficiently small and $C$ a sufficiently large constant; assume for simplicity that $d$ is even. Note that $d \leq n^{\varepsilon}$, where $\varepsilon$ is the constant from the statement of the theorem. We claim that if a set $I$ of rows with $|I|=\ell \leq r$ in the matrix $\boldsymbol{A}_{\boldsymbol{m}, \boldsymbol{n}}$ violates the expansion property (1), then there exists a set of $\ell d / 2$ columns $J$ such that $\boldsymbol{A}_{\boldsymbol{m}, \boldsymbol{n}}$ contains at least $\ell d$ ones in the rectangle $I \times J$. Indeed, let $J \stackrel{\text { def }}{=}\left\{j \in[n]\left|\left|\left\{i \in I \mid j \in J_{i}(A)\right\}\right| \geq 2\right\}\left(=\bigcup_{i \in I} J_{i}(A) \backslash \partial_{A}(I)\right)\right.$. The bound $\ell d$ on the number of ones in $I \times J$ follows from our assumption that $I$ violates (1). Hence, if $|J| \leq \ell d / 2$, we are done. (Add $\ell d / 2-|J|$ columns to $J$ arbitrarily.) Otherwise, its arbitrary subset of cardinality exactly $\ell d / 2$ would do. Calling a rectangle $I \times J$ such that $|I| \leq r,|J|=\frac{d}{2}|I|$ and $I \times J$ contains at least $d|I|$ ones dense, it remains to show that the probability of existence of at least one dense rectangle is $O(1 / m)$.

For any fixed value of $\ell$, there exist at most $m^{\ell}$ choices of $I$ with $|I|=\ell$, at most $n^{\ell d / 2}$ choices of $J$ with $|J|=\ell d / 2$, and at most $(\ell d)^{O(\ell d)}$ choices of $\ell d$
positions in $I \times J$. Therefore, the probability that for given $\ell$ there exists at least one dense $\ell \times(\ell d / 2)$ rectangle does not exceed

$$
m^{\ell} n^{\ell d / 2}(\ell d)^{O(\ell d)}\left(n^{-2 / 3}\right)^{\ell d} \leq m^{\ell}\left(\frac{(\ell d)^{O(1)}}{n}\right)^{\ell d / 6}
$$

Next, this is $\leq m^{\ell} n^{-\Omega(\ell d)}$ provided the constants $\varepsilon, \delta$ are small enough which, in turn, is $\leq m^{-2 \ell}$ provided the constant $C$ in the definition of $d$ is large enough.

This is the bound on the probability that $\boldsymbol{A}_{\boldsymbol{m}, \boldsymbol{n}}$ contains a dense rectangle $I \times J$ with $|I|=\ell$. Therefore, the probability that $\boldsymbol{A}_{\boldsymbol{m}, n}$ is not an $(r, d)$-lossless expander is bounded by $\sum_{\ell=1}^{\infty} m^{-2 \ell} \leq O(1 / m)$. Theorem 2.5 is proved.

## 8. Unprovability of circuit lower bounds by small $\operatorname{Res}(k)$ proofs

The general idea toward extracting Theorem 2.13 from a pseudorandom function generator was extensively discussed in the introduction. We are, however, dealing with a rather weak proof system and, moreover, Definition 2.2 severely restricts the choice of the encoding for the circuit $C_{A, \leq}^{h}$. Thus, we should be careful in checking that the natural reduction can be indeed carried over with the limited tools at our disposal; cf. the previous arguments of this sort in [Raz98], [Raz04], [Raz04], [Kra04].

Let $f_{n}$ be a Boolean function in $n$ variables and $n^{2} \leq t \leq 2^{n}$. We begin with reproducing the formal definition of the $\operatorname{CNF} \operatorname{Circuit}_{t}\left(f_{n}\right)$ from [Raz98], [Raz04].

First, we list all variables of $\operatorname{Circuit}_{t}\left(f_{n}\right)$ (some of them have peculiar long names like InputType ${ }_{\nu}^{\prime}(v)$ ), along with their intended meaning:

$$
y_{a v}\left(a \in\{0,1\}^{n}, v \in[t]\right) \text { - the Boolean value computed at }
$$

the computational node $v$ on
the input string $a$;
$y_{a \nu v}\left(a \in\{0,1\}^{n}, \nu \in\{1,2\}, v \in[t]\right)$ - the value on $a$ brought to $v$ by the $\nu$ 's input to $v$;
$\operatorname{Fanin}(v)$ - this is 0 if $v$ is NOT-gate and 1 if $v$ is AND-gate or OR-gate;
Type $(v)-$ when $\operatorname{Fanin}(v)=1$, this is 0 if $v$ is AND-gate and 1 if $v$ is OR-gate;
$\operatorname{InputType}_{\nu}(v)-\operatorname{this}$ is 0 if $\nu$ 's input to $v$ is a constant or a variable and 1 if it is one of the previous computational gates;

$$
\begin{aligned}
\operatorname{InputType}_{\nu}^{\prime}(v)- & \text { when } \operatorname{InputType}_{\nu}(v)=0 \text {, this is } 0 \\
& \text { if } \nu \text { 's input to } v \text { is a constant, } \\
& \text { and } 1 \text { if it is a variable; } \\
\operatorname{InputType}_{\nu}^{\prime \prime}(v)- & \text { when } \operatorname{InputType}_{\nu}(v)= \\
& \operatorname{InputType}_{\nu}^{\prime}(v)=0, \text { this equals the } \\
& \nu \text { 's input to } v ;
\end{aligned}
$$

$\operatorname{InputVar}_{\nu}(v, i)(i \in[n])-$ when $\operatorname{InputType}_{\nu}(v)=0$,
$\operatorname{InputType}{ }_{\nu}^{\prime}(v)=1$, this is 1 if and only
if $\nu$ 's input to $v$ is the $i$ th variable;
$\operatorname{INPUTVAR}_{\nu}(v, i)$ - equals $\bigvee_{i^{\prime} \leq i} \operatorname{InputVar}\left(v, i^{\prime}\right)$, introduced to keep bottom fan-in bounded;
$\operatorname{InputNode}_{\nu}\left(v, v^{\prime}\right)\left(v^{\prime}<v\right)-$ when $\operatorname{InputType}_{\nu}(v)=1$, this is 1
if and only if $\nu$ 's input to $v$ is the previous gate $v^{\prime}$;
$\operatorname{INPUTNODE}_{\nu}\left(v, v^{\prime}\right)$ - analogously to $\operatorname{INPUTVAR}_{\nu}(v, i)$.
$\operatorname{Circuit}_{t}\left(f_{n}\right)$ is the conjunction of (conjunctive normal forms equivalent to) the following axioms:

$$
\begin{aligned}
& \neg \operatorname{InputType}_{\nu}(v) \wedge \neg \operatorname{InputType}_{\nu}^{\prime}(v) \longrightarrow\left(y_{a \nu v} \equiv \operatorname{InputType}_{\nu}^{\prime \prime}(v)\right) ; \\
& \neg \operatorname{InputType}_{\nu}(v) \wedge \operatorname{InputType}_{\nu}^{\prime}(v) \longrightarrow \neg\left(\operatorname{InputVar}_{\nu}(v, i) \wedge\right. \\
& \left.\quad \operatorname{InputVar}_{\nu}\left(v, i^{\prime}\right)\right)\left(i \neq i^{\prime}\right) ; \\
& \neg \operatorname{InputType}_{\nu}(v) \wedge \operatorname{InputType}_{\nu}^{\prime}(v) \longrightarrow\left(\operatorname{INPUTVAR}_{\nu}(v, i) \equiv\right. \\
& \left.\quad\left(\operatorname{INPUTVAR}_{\nu}(v, i-1) \vee \operatorname{InputVar}_{\nu}(v, i)\right)\right) \\
& \quad\left(\operatorname{INPUTVAR}_{\nu}(v, 0) \stackrel{\text { def }}{=} 0\right) ; \\
& \neg \operatorname{InputType}_{\nu}(v) \wedge \operatorname{InputType}_{\nu}^{\prime}(v) \longrightarrow \operatorname{INPUTVAR}_{\nu}(v, n) ; \\
& \neg \operatorname{InputType}_{\nu}(v) \wedge \operatorname{InputType}_{\nu}^{\prime}(v) \wedge \operatorname{InputVar}_{\nu}(v, i) \longrightarrow\left(y_{a \nu v} \equiv a_{i}\right) ;
\end{aligned}
$$

the analogous group of axioms for InputNode;

$$
\begin{aligned}
& \neg \operatorname{Fanin}(v) \longrightarrow\left(y_{a v} \equiv \neg y_{a 1 v}\right) ; \\
& \operatorname{Fanin}(v) \wedge \neg \operatorname{Type}(v) \longrightarrow\left(y_{a v} \equiv\left(y_{a 1 v} \wedge y_{a 2 v}\right)\right) ; \\
& \operatorname{Fanin}(v) \wedge \operatorname{Type}(v) \longrightarrow\left(y_{a v} \equiv\left(y_{a 1 v} \vee y_{a 2 v}\right)\right) ; \\
& y_{a t} \equiv f(a) .
\end{aligned}
$$

Let $t_{0} \stackrel{\text { def }}{=} \delta \sqrt{t / n}, \delta$ be a sufficiently small constant. Note that $t_{0} \geq t^{\Omega(1)}$ (since $\left.t \geq n^{2}\right)$. Fix an arbitrary $\left(2 t_{0} \times t_{0}\right)\left(t^{\Omega(1)}, O(1)\right)$-lossless expander $A$
such that $\left|J_{i}(A)\right| \geq t^{\Omega(1)}$ for all $i \in\left[2 t_{0}\right]$. (Its existence follows from Theorem 2.5.) Fix an arbitrary ordering $\leq$ of $A$, and consider the iterated circuit $C_{A, \leq}^{n+1}$ from Section 6. Let $b \in\{0,1\}^{2^{n+1} t_{0}}$ be given by $b_{a j} \xlongequal{\text { def }} f_{n}\left(a_{1}, \ldots, a_{n}\right)(a \in$ $\left.\{0, \overline{1}\}^{n}, j \in\left[2 t_{0}\right]\right)$.

In order to prove Theorem 2.13, we are going to reduce $\tau\left(C_{A, \leq}^{n+1}, b\right)$ to $\operatorname{Circuit}_{t}\left(f_{n}\right)$. (That is, substitute variables of $\operatorname{Circuit}_{t}\left(f_{n}\right)$ by "simple" formulas in the variables of $\tau\left(C_{A, \leq}^{n+1}, b\right)$ so that "simple" (which in our context means $\operatorname{Res}(k))$ refutations of $\operatorname{Circuit}_{t}\left(f_{n}\right)$ get transformed into simple $(\operatorname{Res}(2 k))$ refutations of $\tau\left(C_{A, \leq}^{n+1}, b\right)$.) For that purpose we will convert the circuit $C_{A, \leq}^{n+1}$ (with $t_{0}$ inputs and $2^{n+1} t_{0}$ outputs, naturally split into $2^{n}$ groups with $2 t_{0}$ bits each, of which we will select one bit per group) to the (single-output) Boolean circuit $D_{n, \vec{x}}$ in $n$ Boolean variables $z_{1}, \ldots, z_{n}$ parametrized by $t_{0}$ variables $x_{1}, \ldots, x_{t_{0}}$. We will require that $D_{n, \vec{x}}\left(a_{1}, \ldots, a_{n}\right)$ is equal to the $a$ th selected bit in $C_{A, \leq}^{n+1}(x)$ and that the size of $D_{n, \vec{x}}$ is polynomial in $n$. We employ the same construction that was used for self-defeating Natural Proofs [RR97, Th. 4.1]. But since we need to check that this transformation can be carried over in a rather week proof system, we provide a few (tedious) technical details. The new parameter $\ell$ below corresponds to the iteration level.

The skeleton of $D_{n, \vec{x}}$ consists of the gates $v[\ell, i, \Sigma]$, where $0 \leq \ell \leq n$, $i \in\left[2 t_{0}\right]$, and $\Sigma$ is an initial segment in $J_{i}(A)$ computing Boolean functions $f_{\vec{x}}[\ell, i, \Sigma]$ in the variables $z_{1}, \ldots, z_{n}$. The values $f_{\vec{x}}[\ell, i, \Sigma]\left(a_{1}, \ldots, a_{n}\right)$ of these functions are defined as follows. We take the circuit $\left(C_{A, \leq}\right)_{a_{1} \ldots a_{\ell}}$ from Definition 2.11, look at the gate $v_{\Sigma}^{i}$ in this circuit (which we will denote in what follows by $v_{\Sigma}^{i}\left(a_{1} \ldots a_{\ell}\right)$ ), and output as $f_{\vec{x}}[\ell, i, \Sigma]\left(a_{1}, \ldots, a_{n}\right)$ its value when $\tau\left(C_{A, \leq}^{n+1}, b\right)$ is fed with $x_{1}, \ldots, x_{t_{0}}$ (We also naturally let $f_{\vec{x}}[\ell, i, \emptyset]\left(a_{1}, \ldots, a_{n}\right) \stackrel{\text { def }}{=}$ 0 .) Then these functions have the following recursive definition:

$$
\begin{align*}
f_{\vec{x}}[\ell, i, \emptyset]: & =0, \\
f_{\vec{x}}[0, i, \Sigma \cup\{j\}]: & =f_{\vec{x}}[0, i, \Sigma] \oplus x_{j}, \\
f_{\vec{x}}[\ell, i, \Sigma \cup\{j\}]: & =f_{\vec{x}}[\ell, i, \Sigma] \oplus\left\{\left(\bar{z}_{\ell} \wedge f_{\vec{x}}\left[\ell-1, j, J_{j}(A)\right]\right)\right.  \tag{42}\\
& \left.\vee\left(z_{\ell} \wedge f_{\vec{x}}\left[\ell-1, j+t_{0}, J_{j+t_{0}}(A)\right]\right)\right\}, \ell \geq 1 .
\end{align*}
$$

We define $D_{n, \vec{x}}$ as the circuit resulting from expanding these recursive definitions in the standard basis $\{\neg, \wedge, \vee\}$. We let the output gate of $D_{n, \vec{x}}$ be $v\left[n, 1, J_{1}(A)\right]$; cf. the definition of $b$ above.
$D_{n, \vec{x}}$ has size $O\left(t_{0}^{2} n\right)$, which is at most $t$ if the constant $\delta$ in the definition of $t_{0}$ is small enough. We begin constructing the required substitution $\rho$ by first assigning all structural variables $\operatorname{Fanin}(v)$, Type $(v)$ etc., except for InputType ${ }_{\nu}^{\prime \prime}(v)$, to appropriate Boolean constants describing the topology of $D_{n, \vec{x}}$. We also let $\rho\left(\operatorname{InputType}_{\nu}^{\prime \prime}(v)\right) \stackrel{\text { def }}{=} x_{j}$ whenever $v$ is the gate of $D_{n, \vec{x}}$ (necessarily resulting from an instruction in (42) of the second type) whose $\nu$ 's
input is $x_{j}$. In the case $\operatorname{InputType}_{\nu}(v)=1$, we let $\rho\left(y_{a \nu v}\right) \stackrel{\text { def }}{=} \rho\left(y_{a v^{\prime}}\right)\left(\rho\left(y_{a v^{\prime}}\right)\right.$ themselves are yet to be defined), where ( $v^{\prime}, v$ ) is the $\nu$ 's input leading to $v$ in $D_{n, \vec{x}}$. If $\neg \operatorname{InputType}_{\nu}(v)$, then $\nu$ 's input to $v$ is either one of the variables $z_{i}$ or a known Boolean constant $\varepsilon \in\{0,1\}$ or one of the unknown constants $x_{j}$, and we let $\rho\left(y_{a \nu v}\right)$ be $a_{i}, \varepsilon$ or $x_{j}$, respectively. Further, if the gate $v[\ell, i, \Sigma]$ explicitly appears in (42), then we let $\rho\left(y_{a, v[\ell, i, \Sigma]}\right) \stackrel{\text { def }}{=} y_{v_{\Sigma}^{i}\left(a_{1} \ldots a_{\ell}\right)}^{i}$ (which is the variable of $\left.\tau\left(C_{A, \leq}^{n+1}, b\right)\right)$.

However, we cannot extend $\rho$ as a variable substitution to the remaining variables $y_{a v}$, where $v$ is an auxiliary gate resulting from expanding instructions in (42) in the standard basis $\{\neg, \wedge, \vee\}$. But even in this case we still can let $\rho\left(y_{a v}\right)$ be a Boolean function of just two variables $y_{v_{\Sigma}^{i}\left(a_{1} \ldots a_{\ell}\right)}^{i}$ and $y_{v_{j_{j}(A)}^{j^{*}}}^{j^{*}}\left(a_{1} \ldots a_{\ell-1}\right)$, where $j^{*}=j$ (the right end of $\Sigma$ ) if $a_{\ell}=0$ and $j+t_{0}$ if $a_{\ell}=1$.

Summarizing the above argument, we have constructed a substitution $\rho$ that takes every variable of $\operatorname{Circuit}_{t}\left(f_{n}\right)$ to a Boolean function depending of at most two variables of $\tau\left(C_{A, \leq}^{n+1}, b\right)$. Let us extend this substitution to $k$-DNF formulas $F$ by letting $\rho(F) \stackrel{\text { def }}{=} \bigvee_{t \in F} \widehat{\rho(t)}$, where $\widehat{\rho(t)}$ is the straightforward DNF expansion of $\rho(t)$. Clearly, $\rho(F)$ is a $2 k$-DNF, and the $\rho$-image of any inference rule of $\operatorname{Res}(k)$ can be simulated in $\operatorname{Res}(2 k)$ by an inference of size $\exp (O(k))$. Further, given the "intended interpretation," it is easy to check by inspection that for every axiom $C$ of $\operatorname{Circuit}_{t}\left(f_{n}\right), \rho(C)$ has a Res(2) inference of size $O(1)$ from $\tau\left(C_{A, \leq}^{n+1}, b\right)$.

These remarks imply that every $\operatorname{Res}(k)$ refutation of $\operatorname{Circuit}_{t}\left(f_{n}\right)$ of size $S$ gives rise to a $\operatorname{Res}(2 k)$ refutation of $\tau\left(C_{A, \leq}^{n+1}, b\right)$ of size $S \cdot \exp (O(k))$. The proof of Theorem 2.13 is now completed by applying Theorem 2.12 (with $n:=$ $\left.t_{0} ; r, s \geq t^{\Omega(1)} ; d \leq O(1) ; h:=n+1\right)$.

## 9. Pigeonhole principle

In this section we prove Theorem 2.15. We first review some more material from [SBI04].

Let $m=2 n$. (The same proof works with $m=(1+\varepsilon) n$, for any absolute constant $\varepsilon>0$.) For a bipartite graph $G=([m] \cup[n], E)(E \subseteq$ $[m] \times[n])$, let $\neg \operatorname{onto}-\mathrm{PHP}(G)$ be the CNF in the variables $\left\{x_{i j} \mid(i, j) \in E\right\}$ that is obtained from $\neg$ onto- $\mathrm{PHP}_{n}^{m}$ by the restriction assigning to 0 all variables $\left\{x_{i j} \mid(i, j) \notin E\right\}$. Let $\boldsymbol{\rho}^{\boldsymbol{G}}$ be the random restriction constructed in the following way. Pick a random subset $\boldsymbol{J} \subseteq[n]$ by including there every $j \in[n]$ with probability $1 / 4$ independently of each other. For $j \in \boldsymbol{J}$, select uniformly (and independently for different $j$ ) one neighbour $\boldsymbol{i}_{j}$ of $j$ in G, assign $x_{\boldsymbol{i}_{j} j}$ to 1 , and assign to 0 all other $x_{i j}$ with $(i, j) \in E$ and $i \neq \boldsymbol{i}_{j}$.

Proposition 9.1 ([SBI04]). There exists a graph $G$ of maximal degree $O(\log n)$ such that
$\mathbf{P}\left[\right.$ every resolution refutation of $\neg$ onto- $\left.\operatorname{PHP}(G)\right|_{\rho^{G}}$ has width $\left.>n / 24\right] \geq 1 / 2$.
Proof. This is essentially [SBI04, Lemma 18], with the only difference that we have additionally included the group of onto axioms $Q_{j}$. This does not affect its proof in any way.

From now on, fix any particular graph $G$ with properties from Proposition 9.1, and let $\Delta \leq O(\log n)$ be its maximal degree. Since $\neg$ onto- $\operatorname{PHP}(G)$ is obtained from $\neg$ onto- $\mathrm{PHP}_{n}^{m}$ by a restriction, it is sufficient to prove the required bound for $\neg$ onto- $\operatorname{PHP}(G)$. The proof follows the pattern laid out in Section 5, although in the current case it is much simpler.

Definition 9.2. A term in the variables $\left\{x_{i j} \mid(i, j) \in E\right\}$ is reduced if it is monotone and does not contain any sub-term of the form $x_{i j} \wedge x_{i^{\prime} j}, i \neq i^{\prime}$. A DNF $F$ is reduced if all terms $t \in F$ have this property.

Reduced DNFs will play the role of protected ones in Section 5. Let us mention for the record (we will need this in the proof of Claim 9.4) that every reduced DNF that mentions at most $k$ pigeons contains at most $k \Delta$ variables and $O\left(\Delta^{k}\right)$ clauses.

Definition 9.3. For a term $t$, let us denote by $R(t)$ the reduced DNF that is constructed as follows. Let $\tilde{R}(t)$ be the result of replacing in the term $t$ all negative literals $\bar{x}_{i j}$ with $\bigvee\left\{x_{i^{\prime} j} \mid\left(i^{\prime}, j\right) \in E \wedge i^{\prime} \neq i\right\}$. Let $R(t)$ be the straightforward DNF expansion of $\tilde{R}(t)$, in which we remove all terms containing at least one sub-term of the form $x_{i j} \wedge x_{i^{\prime} j}\left(i^{\prime} \neq i\right)$.

Claim 9.4. Let $t$ be a term of width $\leq k$ in the variables $\left\{x_{i j} \mid(i, j) \in E\right\}$. (a) $R(t)$ is a reduced DNF.
(b) There exist $\operatorname{Res}(O(k))$-inferences of $R(t)$ from $t, \neg$ onto- $P H P(G)$ and, vice versa, of trom $R(t), \neg$ onto- $\operatorname{PHP}(G)$ that have size $\Delta^{O(k)}$ and contain at most $k \Delta$ variables.

Proof. (a) is obvious. For part (b) we could have used the same reasoning based on implicational completeness as in the proof of Claim 5.7, but this would have led to an inference of size $\exp (O(k \Delta))$, at least a priori. We circumvent this by the following ad hoc hybrid-type argument.

Let $\left\{i_{1}, \ldots, i_{\ell}\right\}(\ell \leq k)$ be an enumeration of all pigeons mentioned in $t$. For $0 \leq \nu \leq \ell$, split the term $t$ as $t=t_{\nu}^{\prime} \wedge t_{\nu}^{\prime \prime}$, where $t_{\nu}^{\prime}$ is the part corresponding to the pigeons $\left\{i_{1}, \ldots, i_{\nu}\right\}$ and $t_{\nu}^{\prime \prime}$ corresponds to the remaining pigeons $\left\{i_{\nu+1}, \ldots, i_{\ell}\right\}$. Let $R_{\nu}(t) \stackrel{\text { def }}{=} R\left(t_{\nu}^{\prime}\right) \wedge t_{\nu}^{\prime \prime}$, so that $R_{0}(t)=t$ and $R_{\ell}(t)=$ $R(t)$. We consequently infer in $\operatorname{Res}(k)$ all equivalences $R_{\nu}(t) \equiv R_{\nu+1}(t)$ and
then combine them together. Since by an earlier observation, every reduced DNF $R_{\nu}(t)$ contains only $O\left(\Delta^{k}\right)$ clauses, it can be done by an inference of size $\Delta^{O(k)}$.

Claim 9.5. Assume that $\neg$ onto- $P H P(G)$ has a $\operatorname{Res}(k)$ refutation of size $S$. Then it also has an $\operatorname{Res}(O(k))$ refutation of size $S \cdot \Delta^{O(k)}$ in which all lines are of the form $F \vee F^{\prime}$, where $F$ is a reduced $k-D N F$, and $\left|\operatorname{Vars}\left(F^{\prime}\right)\right| \leq k \Delta$.

Proof. As in the proof of Claim 5.8, for a DNF $F$, let $R(F) \stackrel{\text { def }}{=} \bigvee_{t \in F} R(t)$. Notice that $R$ does not change the monotone axioms $Q_{i}, Q_{j}$ and that $R\left(Q_{i_{1}, i_{2} ; j}\right)$ $=Q_{j}$. With these remarks in mind, the rest of the proof is identical to the proof of Claim 5.8.

Finally, let us prove the following PHP-oriented switching lemma.
Claim 9.6. For any reduced $k$-DNF $F$ and any parameter $h$,

$$
\mathbf{P}\left[h\left(\left.F\right|_{\rho}\right)>h\right] \leq \exp \left(-h / \Delta^{O(k)}\right) .
$$

Proof. Call a reduced DNF $F$ regular if $\operatorname{Vars}(F)$ does not contain any pair of variables $x_{i j}, x_{i^{\prime} j}$ with $i \neq i^{\prime}$. Then for every regular $F, \rho^{G}$ acts independently on the variables from $\operatorname{Vars}(F)$ and, moreover, $\mathbf{P}\left[\rho^{\boldsymbol{G}}\left(x_{i j}\right)=0\right]$, $\mathbf{P}\left[\boldsymbol{\rho}^{\boldsymbol{G}}\left(x_{i j}\right)=1\right] \geq \frac{1}{4 \Delta}$. This implies that the restriction of $\boldsymbol{\rho}^{\boldsymbol{G}}$ to the variables in $\operatorname{Vars}(F)$ is $\left(\infty, \mu_{\text {triv }}, 1 /(2 \Delta)\right)$-independent. Applying Lemma 4.4, we prove our claim in the case $F$ is regular.

For the general case, we apply the same trick as before. Let the function $\boldsymbol{\theta}:[n] \longrightarrow[2 n]$ be picked completely at random, and let

$$
G_{\theta} \stackrel{\text { def }}{=}\left\{t \in F \mid \forall x_{i j} \in \operatorname{Vars}(t)(i=\theta(j))\right\} .
$$

Then $G_{\theta}$ is regular, and $\forall t \in F\left(\mathbf{P}\left[t \in G_{\boldsymbol{\theta}}\right] \geq \Delta^{-k}\right)$. Now the proof is completed by applying Lemma 4.5.

The proof of Theorem 2.15 is completed in exactly the same way as the proof of Theorem 2.7 at the end of Section 5.

## 10. Polynomial Calculus with Resolution

Throughout this section we fix an arbitrary field $\mathbb{F}$ with $\operatorname{char}(\mathbb{F}) \neq 2$. First we need the following generalization of Corollary 3.4.

Corollary 10.1. Let $A$ be an $(r, d)$-lossless expander of size $m \times n$ such that $\left|J_{i}(A)\right| \geq 2 d$ for all $i \in[m]$. Then for every ordering $\leq$ and every $b \in$ $\{0,1\}^{m}$, every PCR refutation of $\tau_{\leq}(A, b)$ must have degree $>r / 8$.

Proof. Although this follows by the technique of [BGIP01] (cf. remark in [ABSRW04] before Theorem 3.10), it is easier to apply a more general result
from [BSI10], which for our purposes can be stated as follows. If a CNF $\tau$ results from expanding a set of $\mathbb{F}_{2}$-linear equations, then every PCR refutation of $\tau$ over $\mathbb{F}$ (remember that $\operatorname{char}(\mathbb{F}) \neq 2$ ) gives rise to a resolution refutation of $\tau$ whose width is at most twice as large as the degree of $P$.

Remark 8. Strictly speaking, [BSI10, Th. 2.7] is formulated in terms of so-called Gaussian refutations, but it is a well-known fact that for systems of linear equations, Gaussian width and resolution width coincide. It is also worth mentioning here (we will need this observation in the proof of Claim 10.4 below) that the overhead factor of two comes from the following stronger property that is a byproduct of their proof: for every clause $C$ appearing in the resulting resolution refutation, we have $\operatorname{Vars}(C) \subseteq \operatorname{Vars}\left(m_{1}\right) \cup \operatorname{Vars}\left(m_{2}\right)$, where $m_{1}$ and $m_{2}$ are some monomials in the original PCR refutation.

Since $\tau_{\leq}(A, b)$ always has this "linear" form, Corollary 10.1 follows from Corollary 3.4.

Now the proofs of Theorems 2.18, 2.19 and 2.20 are more or less straightforward adaptation of the corresponding results for Resolution $(=\operatorname{Res}(1))$.

Proof of Theorem 2.18. Monomials in the variables $x_{1}, \ldots, x_{n}, \bar{x}_{1}, \ldots, \bar{x}_{n}$ can be identified, via the transformation (6) and up to a multiplicative constant $\alpha \in \mathbb{F}^{*}$, with respective clauses. In this way, variable substitutions naturally act on polynomials from $\mathbb{F}\left[x_{1}, \ldots, x_{n}, \bar{x}_{1}, \ldots, \bar{x}_{n}\right]$ and take PCR inferences to PCR inferences of the same (or lesser) size. This remark, along with Corollary 10.1, implies the following analogue of Claim 5.2 (proved in exactly the same way).

CLAIM 10.2. If $\rho$ is sparse, then every PCR refutation of $\left.\tau_{\leq}(A, b)\right|_{\rho}$ must have degree $>r / 8$.

The analogue of (36) will thus be

$$
\begin{equation*}
\mathbf{P}\left[\text { every PCR refutation of }\left.\tau_{\leq}(A, b)\right|_{\rho} \text { has degree }>r / 8\right] \geq 1 / 2 \tag{43}
\end{equation*}
$$

Consider the mapping $R$ from Definitions 5.4 and 5.6 restricted to literals ( $R$, in fact, is almost always identical, except for literals of those variables $y_{\Sigma}^{i}$ for which either $\Sigma$ or $J_{i}(A) \backslash \Sigma$ is very small). Let $R_{\mathbb{F}}$ denote the corresponding polynomial homomorphism over the field $\mathbb{F}$. $R_{\mathbb{F}}$ takes any PCR refutation $P$ of $\tau_{\leq}(A, b)$ into another PCR refutation in which every line has the form $R_{\mathbb{F}}(f) \cdot f^{\prime}$, where $f \in P$ and $\operatorname{deg}\left(f^{\prime}\right) \leq O(d)$. Note that unlike Claim 5.8, we do not make any conclusions about the size of the refutation $R_{\mathbb{F}}(P)$. (In fact it may grow out of control.)

Next, we remark that the proof of Claim 5.14 actually allows a finer analysis for protected $O(d)$-DNF of the form $R(C), C$ a clause. Namely, assuming $k=1$, we get

Claim 10.3. For every clause $C$ in the variables $\operatorname{Vars} \leq(A)$, and every $h \leq r$, either $|\operatorname{Vars}(R(C))| \leq h$ or

$$
\mathbf{P}\left[\left.R(C)\right|_{\rho} \neq 1\right] \leq \exp \left(-h / 2^{O(d)}\right)
$$

(Unlike all previous claims of this sort, $\left.R(C)\right|_{\rho} \neq 1$ here simply means that $\left.R(C)\right|_{\rho}$ is not semantically equal to 1.)

Proof of Claim 10.3. Assume that $|\operatorname{Vars}(R(C))|>h$. Applying Claim 5.14, we get that with the required probability $1-\exp \left(-h / 2^{O(d)}\right)$, there exists a decision tree $T_{\boldsymbol{\rho}}$ of height $<h$ strongly representing $R(C) \mid \rho$. Inspecting the proof of Lemma 4.4 for monotone $F$, we see that the tree $T$ given by this construction actually has a stronger property. Namely, whenever $\pi \in B r_{0}(T)$, then not only $\left.\left(\left.t\right|_{\rho}\right)\right|_{\pi}=0$ for all $t \in F$ (as required by Definition 3.5) but, in fact, even $\left.t\right|_{\pi}=0$. The proof of Lemma 4.5 can be easily modified to preserve this property: we only have to go over terms $t \in F$ (rather than $\left.t \in F\right|_{\rho}$ ) in the construction of the sequence $T_{0}, T_{1}, \ldots, T_{\ell}, \ldots$. Therefore, we may also assume that the tree $\left.T\right|_{\rho}$ also has this stronger property.
$R(C)$, however, is the result of a DNF expansion of a disjunction of linear forms. Such DNFs can be set to 0 only by restrictions that assign all their variables. Therefore, since $|\operatorname{Vars}(R(C))|>h$ and the height of $\left.T\right|_{\rho}$ is at most $h$, we have $B r_{0}\left(\left.T\right|_{\rho}\right)=\emptyset .\left.R(C)\right|_{\rho}=1$ (in the semantical sense) follows.

We now finish the proof of Theorem 2.18. Let $P$ be a PCR refutation of $\tau_{\leq}(A, b)$. From (43), we get in particular

$$
\mathbf{P}\left[\exists g \in R_{\mathbb{F}}(P)\left(\operatorname{deg}\left(\left.g\right|_{\rho}\right)>r / 8\right)\right] \geq 1 / 2
$$

As we remarked above, every line in the refutation $R_{\mathbb{F}}(P)$ has the form $R_{\mathbb{F}}(f) \cdot f^{\prime}$ with $f \in P$ and $\operatorname{deg}\left(f^{\prime}\right) \leq O(d) \leq r / 16$. Next, $\operatorname{deg}\left(R_{\mathbb{F}}(f) \mid \rho\right)>r / 16$ implies that there exists a monomial $\alpha \cdot \Gamma_{C}$ in $f\left(\alpha \in \mathbb{F}^{*}\right)$ such that $|\operatorname{Vars}(R(C))|>$ $r / 16$ and $\left.R_{\mathbb{F}}\left(\Gamma_{C}\right)\right|_{\rho} \neq 0$. Applying Claim 10.3 (with $h:=r / 16$ ), we see that the probability of this event for every particular $\alpha \cdot \Gamma_{C} \in P$ is bounded by $\exp \left(-r / 2^{O(d)}\right)$. Theorem 2.18 now follows by the same calculation as at the end of Section 5.

All proofs in Section 6 hold for any proof system that is closed under variable substitutions. In particular, Theorem 2.19 (as well as the analogue of Theorem 2.12 for PCR not stated explicitly in Section 2) follows from Theorem 2.18 by the same proof.

The only problem with the proof of Theorem 2.13 is that the reduction $\rho$ constructed in Section 8 is not a variable substitution (and, as we already remarked above in the proof of Theorem 2.18, the size of PCR proofs may blow up exponentially from applying such $\rho$ ). The only variables that create this problem are $y_{a v}$, where $v$ is an auxiliary gate resulting from expanding
the instructions in (42). This time, however, our circuit $D_{n, \vec{x}}$ is also allowed to contain the parity gates; therefore, such auxiliary gates are confined to be one of $\bar{z}_{\ell} \wedge f\left[\ell-1, j, J_{j}(A)\right], z_{\ell} \wedge f\left[\ell-1, j+t_{0}, J_{j+t_{0}}(A)\right],\left(\bar{z}_{\ell} \wedge f\left[\ell-1, j, J_{j}(A)\right]\right) \vee\left(z_{\ell} \wedge\right.$ $\left.f\left[\ell-1, j+t_{0}, J_{j+t_{0}}(A)\right]\right)$ explicitly appearing in (42). It remains to notice that for every fixed $a \in\{0,1\}^{n}$, every one of these three gates computes either zero or one of the two functions $f\left[\ell-1, j, J_{j}(A)\right], f\left[\ell-1, j+t_{0}, J_{j+t_{0}}(A)\right]$. Thus, at the expense of allowing $\oplus$-gates, $\rho$ can be turned into a variable substitution, and the rest of the proof carries over to PCR without any further changes.

We now turn to the proof of our last result, Theorem 2.21, which does not seem to have any obvious analogue for $\operatorname{Res}(k)$.

Let $\rho$ be a restriction of the variables $\operatorname{Vars}_{\leq}^{\mathrm{Cycl}}(A)$. Denote by $E^{i}(\rho)$ the set of all endpoints of all cyclic intervals $\Delta \neq J_{i}(A)$ with $y_{\Delta}^{i} \in \sup (\rho) . E^{i}(\rho)$ defines a partition of $J_{i}(A)$ into cyclic intervals $\Delta_{1}^{i}(\rho), \ldots, \Delta_{\ell_{i}(\rho)}^{i}(\rho)$ such that whenever $y_{\Delta}^{i} \in \sup (\rho), \Delta$ is a disjoint union of some of these intervals. Say that $\rho$ is consistent if whenever $y_{\Delta(1)}^{i}, \ldots, y_{\Delta(w)}^{i} \in \sup (\rho)$, and $\Delta(1) \oplus \cdots \oplus$ $\Delta(w)=0$, we have $\rho(\Delta(1)) \oplus \cdots \oplus \rho(\Delta(w))=0$. Like in Section 5, let $J_{x}(\rho) \stackrel{\text { def }}{=}\left\{j \in[n] \mid x_{j} \in \sup (\rho)\right\}$, and say that $\rho$ is sparse if $\left|\Delta_{\nu}^{i}(\rho) \backslash J_{x}(\rho)\right| \geq 2 d$ for every $i \in[m], \nu \in\left[\ell_{i}(\rho)\right]$.

For a clause $C$ in the variables $\operatorname{Vars}_{\leq}^{\mathrm{Cycl}}(A)$, let

$$
J_{x}(C) \stackrel{\text { def }}{=}\left\{j \in[n] \mid x_{j} \in \operatorname{Vars}(C)\right\}
$$

and

$$
\operatorname{dom}(C) \stackrel{\text { def }}{=}\left\{i \in[m] \mid \exists \Delta \neq J_{i}(A)\left(y_{\Delta}^{i} \in \operatorname{Vars}(C)\right)\right\}
$$

Call the quantity $w_{A}(C) \stackrel{\text { def }}{=}\left|J_{x}(C)\right|+|\operatorname{dom}(C)|$ the $A$-width of the clause $C$. The $A$-degree of a monomial $\alpha \cdot \Gamma_{C}\left(\alpha \in \mathbb{F}^{*}\right)$ is defined as $\operatorname{deg}_{A}\left(\alpha \cdot \Gamma_{C}\right) \stackrel{\text { def }}{=} w_{A}(C)$, and the $A$-degree of a polynomial is the maximal $A$-degree of a monomial occurring in it. ${ }^{7}$ The $A$-width of a resolution proof $[A$-degree of a PCR proof] is the maximal $A$-width [ $A$-degree] of a clause [polynomial, respectively] occurring in it.

Claim 10.4. If $\rho$ is consistent and sparse, then every PCR refutation $P$ of $\tau_{<}^{\text {Cycl }}(A, b)$ must contain a monomial $m$ such that $\operatorname{deg}_{A}(m)>r / 8$ and $\left.m\right|_{\rho} \not \equiv \overline{0}$.

Proof. This is analogous to the proof of Claim 5.2, so we only remark the differences. First, in the cyclic case the reduction $\rho$ constructed in the proof of that claim is no longer a variable substitution: $\rho\left(y_{\Delta}^{i}\right)$ is in general the parity

[^7]of two variables in $\operatorname{Vars}_{\leq\left.\right|_{\rho}}^{\mathrm{Cycl}}\left(\left.A\right|_{\rho}\right)$. If, however, we extend in a natural way the notions of $A$-width and $A$-degree to clauses/monomials in the variables $\operatorname{Vars}_{\leq\left.\right|_{\rho}}^{\mathrm{Cycl}}\left(\left.A\right|_{\rho}\right)$ (simply by ignoring the second superscript $\nu$ in $y_{\Delta}^{(i, \nu)}$ ), then it turns out that $\rho$ still does not increase $A$-degree. Assuming (for the sake of contradiction) that no monomial $m$ with the required properties exist, we conclude that $\operatorname{deg}_{A}\left(\left.P\right|_{\rho}\right) \leq r / 8$.

Next, we use the same argument from [BSI10] as in the proof of Corollary 10.1 to convert $\left.P\right|_{\rho}$ into a resolution refutation of $\tau_{\leq\left.\right|_{\rho}}^{\mathrm{Cycl}}\left(\left.A\right|_{\rho},\left.b\right|_{\rho}\right)$. As we already noticed above (see Remark 8), for every clause $C$ in the resulting resolution refutation, $\operatorname{Vars}(C) \subseteq \operatorname{Vars}\left(m_{1}\right) \cup \operatorname{Vars}\left(m_{2}\right)$, where $m_{1}, m_{2}$ are some monomials in the original PC refutation. Hence, the $A$-width of this resulting refutation of $\tau_{\leq\left.\right|_{\rho}}^{\mathrm{Cycl}}\left(\left.A\right|_{\rho},\left.b\right|_{\rho}\right)$ is at most $r / 4$.

Finally, when Corollary 3.4 is applied to the triple $\left(\left.A\right|_{\rho}, \leq\left.\right|_{\rho},\left.b\right|_{\rho}\right)$, it can be generalized in two ways. First, $\tau_{\leq l_{\rho}}\left(\left.A\right|_{\rho},\left.b\right|_{\rho}\right)$ can be replaced by $\tau_{\leq\left.\right|_{\rho}}^{\mathrm{Cycl}}\left(\left.A\right|_{\rho},\left.b\right|_{\rho}\right)$ (since, like $\tau_{\leq\left.\right|_{\rho}}\left(\left.A\right|_{\rho},\left.b\right|_{\rho}\right)$, this is also a sub-CNF of $\tau\left(\left.A\right|_{\rho}, \vec{g}\right)$ for the same $\left.\vec{g}\right)$. Second, and this is more crucial, width can be replaced by $A$-width. This is also done by an easy adjustment of [ABSRW04, Th. 3.1] and Theorem 3.3. The only nontrivial thing to be remarked in this respect is that in the matrix $\left.A\right|_{\rho}$, the expansion property holds for every set of rows $\left\{\left(i_{1}, \nu_{1}\right), \ldots,\left(i_{\ell}, \nu_{\ell}\right)\right\}$ whose projection $\left\{i_{1}, \ldots, i_{\ell}\right\}$ onto the first coordinate has size $\leq r$ (cf. the bound (27) in the proof of Claim 5.2), that is, to every set of the form $\operatorname{dom}(C)$ with $w_{A}(C) \leq r$.

The generalization of Corollary 3.4 obtained in this way gives the required contradiction with $\operatorname{deg}_{A}\left(\left.P\right|_{\rho}\right) \leq r / 8$ and completes the proof of Claim 10.4.

Next, similarly to Definition 5.9 , we define a consistent random restriction $\boldsymbol{\rho}$ of the variables $\operatorname{Vars}_{\leq}^{\mathrm{Cycl}}(A)$. As before, $\boldsymbol{\rho}$ assigns $x$-variables completely at random (with $\mathbf{P}\left[\boldsymbol{\rho}\left(x_{j}\right)=0\right]=\mathbf{P}\left[\boldsymbol{\rho}\left(x_{j}\right)=1\right]=1 / 4$ ). $y$-variables are assigned as follows. Pick a random subset of endpoints $\tilde{\boldsymbol{E}}^{\boldsymbol{i}}$ by including there every endpoint with probability $(1 / 2 d)$, independently of each other. $\tilde{\boldsymbol{E}}^{i}$ induces a partition of $J_{i}(A)$ into cyclic intervals $\tilde{\Delta}_{1}^{i}, \ldots, \tilde{\Delta}_{\tilde{\ell}_{i}}^{i}$. We take all those $\tilde{\Delta}_{\nu}^{i}$ for which $\left|\tilde{\boldsymbol{\Delta}}_{\boldsymbol{\nu}}^{\boldsymbol{i}} \backslash J_{x}(\boldsymbol{\rho})\right|<2 d$, and we remove from $\tilde{\boldsymbol{E}}^{\boldsymbol{i}}$ both endpoints of these intervals. Let $\boldsymbol{E}^{i}$ be the resulting set of endpoints and $\boldsymbol{\Delta}_{1}^{i}, \ldots, \boldsymbol{\Delta}_{\ell_{i}}^{i}$ be the corresponding partition into cyclic intervals. We pick at random Boolean values $\boldsymbol{b}_{\boldsymbol{i}, \mathbf{1}}, \ldots, \boldsymbol{b}_{\boldsymbol{i}, \boldsymbol{\ell}_{\boldsymbol{i}}}$ subjected to the only linear constraint $\bigoplus_{\nu=1}^{\ell_{i}} \boldsymbol{b}_{\boldsymbol{i}, \boldsymbol{\nu}}=b_{i}$, and in the natural way we assign all those variables $y_{\Delta}^{i} \in \operatorname{Vars}_{\leq}^{\text {Cycl }}(A)$ for which both endpoints of $\Delta$ are in $\boldsymbol{E}^{\boldsymbol{i}}$. That is, if $\Delta=\bigcup_{\nu \in \Gamma} \boldsymbol{\Delta}_{\boldsymbol{\nu}}^{\boldsymbol{i}}$, then $\boldsymbol{\rho}\left(y_{\Delta}^{i}\right) \stackrel{\text { def }}{=} \bigoplus_{\nu \in \Gamma} \boldsymbol{b}_{\boldsymbol{i}, \boldsymbol{\nu}}$. Clearly, $\rho$ is consistent with probability 1.

Claim 10.5. $\mathbf{P}[\boldsymbol{\rho}$ is sparse $] \geq 1 / 2$.
Proof. By exactly the same analysis as in the proof of Claim 5.10.

Lastly, we need the following simple version of Claim 10.3 (Note the polynomial dependence on $d$ - this is where our choice of the encoding will be used.)

Claim 10.6. Let $C$ be any clause in the variables $\operatorname{Vars}_{\leq}^{\text {Cycl }}(A)$ with $w_{A}(C)$ $\leq r$. Then

$$
\mathbf{P}\left[\left.C\right|_{\rho} \not \equiv 1\right] \leq \exp \left(-w_{A}(C) / d^{O(1)}\right) .
$$

Proof. We may assume without loss of generality that $C$ consists either only of $x$-variables, or only of $y$-variables. In the first case the claim is obvious (since $\rho$ acts on $x$-variables completely at random). In the second, for every $i \in \operatorname{dom}(C)$, choose arbitrarily a cyclic interval $\Delta^{i}$ with $\frac{\left|J_{i}(A)\right|}{3}-1 \leq\left|\Delta^{i}\right| \leq$ $\frac{2\left|J_{i}(A)\right|}{3}+1$ such that $y_{\Delta^{i}}^{i} \in \operatorname{Vars}(C)$. Let $j_{1}^{i}, j_{2}^{i}$ be its endpoints, and for $\alpha=1,2$, let $L_{i \alpha}, R_{i \alpha}$ be the two cyclic intervals, of length $5 d$ each, with the endpoint $j_{\alpha}^{i}$. Note that, due to the constraints on $\left|\Delta^{i}\right|$, we have $\operatorname{Conv}\left(L_{i 1}, R_{i 1}\right) \cap$ $\operatorname{Conv}\left(L_{i 2}, R_{i 2}\right)=\emptyset$, and we use $L_{i \alpha}, R_{i \alpha}$ as we used protections in Section 5. The property of weak regularity is immediate ( $C$ does not contain $x$-variables, and for every $i \in[m]$ contains at most one $y_{\Delta}^{i}$ ), therefore we need neither the reduction operator $R$ nor any analogue of Claim 5.12.

In the proof of Claim 5.13, the sub-protections $L_{i \alpha}^{\prime}, R_{i \alpha}^{\prime}$ will now have cardinalities $4 d$ (as opposed to $2 d$ ), and we can save by relaxing the requirement $J \cap P_{i \alpha}^{\prime} \neq \emptyset$ in the definition of $\left(\boldsymbol{Y}^{i}\right)^{\prime}$ to $\left|J \cap L_{i \alpha}^{\prime}\right|,\left|J \cap R_{i \alpha}^{\prime}\right| \leq 2 d$. Then in (35) we have the better bound $\mathbf{P}\left[y_{\Sigma}^{i} \in\left(\boldsymbol{Y}^{i}\right)^{\prime}\right] \geq \Omega\left(p_{i \Sigma} / d\right)$, and with the same argument we get that $\boldsymbol{\rho}$ restricted to $\operatorname{Vars}(C)$ is $\left(r, \mu_{\text {triv }}, \Omega(1 / d)\right)$-independent. Claim 10.6 follows.

Theorem 2.21 is immediately implied by Claims 10.4, 10.5 and 10.6.

## 11. Open problems

The central open problem in this area is obvious: construct pseudorandom generators that would be hard for as strong proof systems as possible and get as many pseudo-random output bits as possible. We complement this with several other (more minor) questions.

Does there exist a function pseudorandom generator of Nisan-Wigderson type that is hard for Resolution? Say, do there exist any $A, \leq, b$ such that $m=2^{n^{\varepsilon}}$ and the minimal resolution refutation size of $\tau_{\leq}(A, b)$ is exponential in $n$ ? The importance of this problem is, of course, greatly undermined by the iterability trick that allowed us to turn around it in Theorem 2.13. Still, this problem might be interesting in its own right. Also, it is not clear at the moment how general this trick will turn out so, after all, a better understanding of the hardness of NW-generators themselves still may be useful in further research.

The next problem is of similar flavour. Even $n^{\Omega(\log n)}$ output bits we were able to get only for Nisan generators, that is, when the base functions are parity functions. Get more than quadratic number of output bits for a wider class of base functions. Suppose for example that $m=O\left(n^{2}\right)$, and the functions $\vec{g}$ are picked at random. Is $\tau\left(\boldsymbol{A}_{\boldsymbol{m} \boldsymbol{n}}, \vec{g}\right)$ hard for Resolution?

If the last problem is solved, then we might ask to extend the resulting bound to the system PCR over fields of characteristic 2. (This case is left completely open by the current paper.)

Does PCR possess efficient proofs of NP $\nsubseteq \mathbf{P} /$ poly when the latter class is defined by circuits over the standard basis $\{\neg, \wedge, \vee\}$ ? The natural attempt to simply ignore the difficulty occurred in the proof of Theorem 2.20 leads to the system $\operatorname{PCRes}(2)$, which is a natural hybrid of PC and $\operatorname{Res}(2)$. Now, lower bounds for this system are known (see the much more general result in [Kra97b]), but what we really need is a pseudo-random generator hard for it.

Last, but not the least, construct explicit lossless expanders (ideally, expanders with parameters close to those in Theorem 2.5). The importance of this last problem stretches of course well beyond proof complexity. (See, e.g., the impressive list of potential applications of expanders in [CRVW02].)

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(Received: March 19, 2003)
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Institute for Advanced Study, Princeton, NJ
on leave from Steklov Mathematical Institute, Moscow, Russia
E-mail: razborov@cs.uchicago.edu
Current address: University of Chicago, Chicago, IL


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[^1]:    ${ }^{1}$ The reader wondering whether it might be more natural to weaken here the condition "for any string $b$ " to "for some string $b$ " or perhaps to "for some explicit string $b$ " is referred to an extensive discussion of this issue in [ABSRW04, §1] where the concept was introduced.

[^2]:    ${ }^{2}$ [Raz95a] dealt with provability of $\Sigma_{0}^{b}$-statements in the theories of Bounded Arithmetic, which is the uniform counterpart of propositional proof complexity. At the suggestion of Jan Krajíček, in later papers it was recast in more convenient framework of propositional proof complexity.

[^3]:    ${ }^{3}$ Here and in the rest of introduction we implicitly assume that $\operatorname{char}(\mathbb{F}) \neq 2$.

[^4]:    ${ }^{4}$ Although $\operatorname{Vars}_{C}$ is in a one-to-one correspondence with the set of all gates, we prefer to draw a clear distinction between inputs and computational gates.

[^5]:    ${ }^{5}$ Note that the "intended meaning" of $n$ in Theorems 2.13 and 2.20 is completely different from all other results

[^6]:    ${ }^{6}$ The factor 2 in (14) can be removed by using a slightly more sophisticated analysis in Case 2 below, but we do not need that precision in what follows.

[^7]:    ${ }^{7}$ Our notion of $A$-degree is slightly different from the one used in [ABSRW04], mainly since the variables we consider here are automatically stratified with respect to rows $i \in[\mathrm{~m}]$.

