# Substitutions into propositional tautologies 

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#### Abstract

We prove that there is a polynomial time substitution $\left(y_{1}, \ldots, y_{n}\right):=$ $g\left(x_{1}, \ldots, x_{k}\right)$ with $k \ll n$ such that whenever the substitution instance $A\left(g\left(x_{1}, \ldots, x_{k}\right)\right)$ of a 3DNF formula $A\left(y_{1}, \ldots, y_{n}\right)$ has a short resolution proof it follows that $A\left(y_{1}, \ldots, y_{n}\right)$ is a tautology. The qualification "short" depends on the parameters $k$ and $n$.


Let $A(y)$ be a 3DNF propositional formula in $n$ variables $y=\left(y_{1}, \ldots, y_{n}\right)$ and assume that we want to prove that $A(y)$ is a tautology. By substituting $y:=g(x)$ with $x=\left(x_{1}, \ldots, x_{k}\right)$ we get formula $A(g(x))$ which is, as long as $g$ is computable in (non-uniform) time $n^{O(1)}$, expressible as 3DNF of size $n^{O(1)}$. The formula uses $n^{O(1)}$ auxiliary variables $z$ besides variables $x$ but only $x$ are essential: We know apriori (and can witness by a polynomial time constructible resolution proof) that any truth assignment satisfying $\neg A\left(g\left(x_{1}, \ldots, x_{k}\right)\right)$ would be determined already by its values at $x_{1}, \ldots, x_{k}$.

If $A(y)$ is a tautology, so is $A(g(x))$. In this paper we note that the emerging theory of proof complexity generators (Section 1) provides a function $g$ with $k \ll n$ for which a form of inverse also holds (the precise statement is in Section 2):

For the following choices of parameters:

- $k=n^{\delta}$ and $s=2^{n^{\epsilon}}$, for any $\delta>0$ there is $\epsilon=\epsilon(\delta)>0$, or

[^0]- $k=\log (n)^{c}$ and $s=n^{\log (n)^{\mu}}$, for $c>1, \mu>0$ specific constants,
it holds:
There is a function $g$ computable in time $n^{O(1)}$ extending $k$ bits to $n$ bits such that whenever $A(g(x))$ is a tautology and provable by a resolution proof of size at most s then $A(y)$ is a tautology too.

Unless you are an ardent optimist you cannot hope to improve the bound to $s$ so that it would allow an exhaustive search over $\{0,1\}^{k}$. In fact, it follows that unless $\mathcal{P}=\mathcal{N} \mathcal{P}$ no automated provers (or SAT solvers) that are based on DPLL procedure [4, 5], even augmented by clause learning [16] or restarts of the procedure [6] can run in time subexponential $\left(2^{k^{0(1)}}\right)$ in the number of essential variables, as their computations yield resolution proofs of size polynomial in the time [2], cf. Section 3. However, for the particular function $g$ we use, the exhaustive search yields something (assuming the existence of strong one-way functions): If $A(g(x))$ is a tautology then there are at most $2^{n} / n^{\omega(1))}$ falsifying truth assignments to $\left.A(y)\right)$ (Section 3). This is a consequence of results of Razborov and Rudich [15].

Notation: $x, y, z, \ldots$ and $a, b, \ldots$ are tuples of variables and of bits respectively, the individual variables or bits being denoted $x_{i}, y_{j}, \ldots$ and $a_{i}, b_{j}, \ldots$, respectively. $[n]$ is $\{1, \ldots, n\}$.

## 1 Proof complexity generators

A proof complexity generator is any function $g:\{0,1\}^{*} \rightarrow\{0,1\}^{*}$ given by a family of circuits ${ }^{1}\left\{C_{k}\right\}_{k}$, each $C_{k}$ computing function $g_{k}:\{0,1\}^{k} \rightarrow$ $\{0,1\}^{n(k)}$ for some injective function $n(k)>k$. (We want injectivity of $n(k)$ so that any string is in the range of at most one $g_{k}$.) We assume that circuits $C_{k}$ have size $n(k)^{O(1)}$. Functions $g$ of interest are those for which it is hard to prove that any particular string from $\{0,1\}^{n(k)}$ is outside of the range of $g_{k}$. This can be formalized as follows.

Assume $m(k)$ is the size of $C_{k}$. The set of $\tau$-formulas corresponding to $C_{k}$ is parameterized by $b \in\{0,1\}^{n(k)} \backslash \operatorname{Rng}\left(g_{k}\right)$. Given such a $b$, construct propositional formula $\tau\left(C_{k}\right)_{b}$ (denoted simply $\tau(g)_{b}$ when $C_{k}$ s are canonical) as follows: The atoms of $\tau\left(C_{k}\right)_{b}$ are $x_{1}, \ldots, x_{k}$ for bits of an input $x \in$ $\{0,1\}^{k}$ and auxiliary atoms $z_{1}, \ldots, z_{m(k)}$ for bit values of subcircuits of $C_{k}$ determined by the computation of $C_{k}$ on $x$. The formula expresses in a

[^1]DNF that if $z_{j}$ 's are correctly computed as in $C_{k}$ with input $x$ then the output $C_{k}(x)$ differs from $b$. The size of $\tau\left(C_{k}\right)_{b}$ is proportional to $m(k)$. The formula is a tautology as $b \notin R n g(g)$.

The $\tau$-formulas have been defined in [8] and independently in [1], and their theory is being developed ${ }^{2}$. We now recall only few facts we shall use later.

The next definition formalizes the concept of "hard to prove" in two ways; the first one follows [14], the second one is from [10]. We apply these concepts only to resolution but they are well-defined for an arbitrary propositional proof system in the sense of [3].

Definition 1.1 Let $s(k) \geq 1$ be a function, and let $g=\left\{g_{k}\right\}_{k}$ be a function as above.

- Function $g$ is $s(k)$-hard for resolution if any formula $\tau\left(C_{k}\right)_{b}, b \in$ $\{0,1\}^{n(k)} \backslash \operatorname{Rng}(g)$, requires resolution proofs of size at least $s(k)$.
- $g$ is $s(k)$-iterable for resolution iff all disjunctions of the form

$$
\tau\left(C_{k}\right)_{B_{1}}\left(x^{1}\right) \vee \ldots \vee \tau\left(C_{k}\right)_{B_{t}}\left(x^{1}, \ldots, x^{t}\right)
$$

require resolution proofs of size at least $s(k)$. Here $t \geq 1$ is arbitrary, and $B_{1}, \ldots, B_{t}$ are circuits such that:
$-x^{i}$ are disjoint $n(k)$-tuples of atoms, for $i \leq t$.

- $B_{1}$ has no inputs, and inputs to $B_{i}$ are among $x^{1}, \ldots, x^{i-1}$, for $i \leq t$.
- Circuits $B_{1}, \ldots, B_{t}$ are just substitutions of variables and constants for variables.

Note that the $s(k)$-iterability implies the $s(k)$-hardness. (The proof of Theorem 2.1 uses only hardness of the function but we need iterability to get a hard function computable in uniform polynomial time in Corollary 1.5.)

Exponentially hard functions for resolution do exists. A $\mathcal{P} / p o l y$-function, a linear map over $\mathbf{F}_{2}$ defined by a sparse matrix with a suitable "expansion" property, $2^{k^{\Omega(1)}}$-hard for resolution was constructed in [10, Thm.4.2]. Razborov [14, Thms.2.10,2.20] gave an independent construction and he noticed that any proof of hardness utilising only the expansion property of a matrix implies, in fact, $2^{k^{\Omega(1)}}$-iterability as well. We use a weaker statement than what is actually proved in [14].

[^2]Theorem 1.2 (Razborov[14]) There exists a function $g=\left\{g_{w}\right\}_{w}$, with $g_{w}:\{0,1\}^{w} \rightarrow\{0,1\}^{w^{2}}$, computed by size $O\left(w^{3}\right)$ circuits, that is $2^{w^{\Omega(1)}}$ iterable for resolution.

However, what we want is a function computed by a uniform algorithm (it is not known at present how to construct explicitly the matrices used in $[10,14])$ in order that our substitution is polynomial time computable too. Fortunately, we can get a uniform function from Theorem 1.2, using a result from [10].

Definition 1.3 Let $m \geq \ell \geq 1$. The truth table function $\mathbf{t t}_{m, \ell}$ takes as input $m^{2}$ bits describing ${ }^{3}$ a size $\leq m$ circuit $C$ with $\ell$ inputs, and outputs $2^{\ell}$ bits: the truth table of the function computed by $C$.
$\mathbf{t t}_{m, \ell}$ is, by definition, equal to zero at inputs that do not encode a size $\leq m$ circuit with $\ell$ inputs.

Theorem 1.4 (Krajíček[10]) Assume that there exists a $\mathcal{P} /$ poly-function $g=\left\{g_{w}\right\}_{w}$, with $g_{w}:\{0,1\}^{w} \rightarrow\{0,1\}^{w^{2}}$, that is $2^{w^{\Omega(1)}}$-iterable for resolution.

Then:

1. For any $1>\delta>0$, the truth table function $\mathbf{t t}_{2^{\delta \ell}, \ell}$ is $2^{2^{\Omega(\delta \ell)}}$-iterable for resolution.
2. There is a constant $c \geq 1$ such that the truth table function $\mathbf{t t}_{\ell^{c}, \ell}$ is $2^{\ell^{1+\Omega(1)}}$-iterable for resolution.

The theorem (see [10, Thm.4.2]) is proved by iterating the circuit computing $g_{w}$ along an $w$-ary tree of depth $t$, suitable $t$. The two statements stated explicitly are just two extreme choices of parameters, but the proof yields an explicit trade-off for a range of parameters. We state this without repeating the construction from [10].

Let $c \geq 1$ and $\epsilon>0$ be arbitrary constants. Assume that there is a function $g=\left\{g_{w}\right\}_{w}$, with $g_{w}:\{0,\}^{w} \rightarrow\{0,1\}^{w^{2}}$, computed by size $w^{c}$ circuits and that is $2^{w^{\epsilon}}$-iterable for resolution.

Then the truth function $\mathbf{t t}_{m, \ell}$ is $s$-iterable for the following choices of parameters, with $t \geq 1$ arbitrary:

1. $m:=w^{c} \cdot t$,

[^3]2. $\ell:=t \cdot \log (w)$,
3. $s:=2^{w^{\epsilon}-t \log (w)}$.

Corollary 1.5 1. For every $c>1$ there are $\epsilon>0$ and a polynomial time computable function $g=\left\{g_{k}\right\}_{k}$,

$$
g_{k}:\{0,1\}^{k} \rightarrow\{0,1\}^{k^{c}}
$$

that is $2^{k^{\epsilon}}$-hard for resolution.
2. There are $\epsilon>\delta>0$ and a polynomial time computable function $g=$ $\left\{g_{k}\right\}_{k}$,

$$
g_{k}:\{0,1\}^{k} \rightarrow\{0,1\}^{2^{k^{\delta}}}
$$

that is $2^{k^{\epsilon}}$-hard for resolution.

## 2 The substitution

Theorem 2.1 1. For any $\delta>0$ there are $\mu>0$ and a polynomial time computable function $g=\left\{g_{k}\right\}_{k}$, extending $k=n^{\delta}$ bits to $n=n(k)$ bits such that for any 3DNF formula $A(y), y=\left(y_{1}, \ldots, y_{n}\right)$, it holds:

- If $A\left(g_{k}(x)\right)$ has a resolution proof of size at most $2^{n^{\mu}}$ then $A(y)$ is a tautology.

2. There are $c>1, \mu>0$ and a polynomial time computable function $g=\left\{g_{k}\right\}_{k}$, extending $k=\log (n)^{c}$ bits to $n=n(k)$ bits such that for any 3DNF formula $A(y), y=\left(y_{1}, \ldots, y_{n}\right)$, it holds:

- If $A\left(g_{k}(x)\right)$ has a resolution proof of size at most $n^{\log (n)^{\mu}}$ then $A(y)$ is a tautology.


## Proof :

For Part 1. let $\delta>0$ be arbitrary. Put $c:=\delta^{-1}$, and take $\epsilon>0$ and the polynomial time function $g=\left\{g_{k}\right\}_{k}$ guaranteed by Corollary 1.5 (Part 1). Hence $g_{k}:\{0,1\}^{n^{\delta}} \rightarrow\{0,1\}^{n}$, for $k=n^{\delta}$.

Assume $A(y)$ is not a tautology and let $b \in\{0,1\}^{n}$ is a falsifying assignment. Then $\tau(g)_{b}$ can be proved in resolution by combining a size $s$ proof of $A(g(x))$ with a size $n^{O(1)}$ proof of $\neg A(b)$. By the $2^{k^{\epsilon}}$-hardness of $g$ it must hold that

$$
s+n^{O(1)} \geq 2^{n^{\delta \epsilon}}
$$

Hence $s$ must be at least $2^{n^{\mu}}$, for suitable $\mu<\delta \epsilon$.
Part 2 is proved analogously, using Corollary 1.5 (Part 2).
q.e.d.

Note that if $g(x)$ is a hard proof complexity generator, so is function $(x, z) \rightarrow(g(x), z)$. Hence we may apply the substitutions from the theorem only to some variables $y_{i}$.

## 3 Remarks

We conclude by some remarks. First we substantiate the comment about automated theorem provers and SAT-solvers from the introduction.

Let $B(x, z)$ be the formula $A(g(x))$ with the auxiliary variables $z$ also displayed. The $k$ variables $x$ are essential in $B$ in the sense that there is a $O(|B|)$ size resolution proof of

$$
B(x, z) \vee B(x, w) \vee z_{j} \equiv w_{j}
$$

for all $j$. (In fact, such a proof is easily constructible once we have the algorithm for $g$.) Assume that it would be always possible to find a resolution proof of a formula whose size would be subexponential in the minimal number of essential variables and polynomial in the size of the formula; in our case $2^{k^{o(1)}}|A(g(x))|^{O(1)}$.

Taking $g$ from Theorem 2.1 (part 2) this would get a size $|A(g)|^{O(1)}$ proof of $A(g(x))$, which is bellow the required upper bound $n^{\log (n)^{\mu}}$. Hence we could interpret this as a new proof system $R_{g}$ in the sense of CookReckhow [3]: A proof in $R_{g}$ of $A(y)$ is either a resolution proof or a size $|A(g(x))|^{c}$ (specific $c$ ) proof of $A(g(x))$. This proof system would allow for polynomial size proofs of all tautologies, hence $\mathcal{N P}=\operatorname{coN} \mathcal{P}$.

The equality $\mathcal{N P}=\cos \mathcal{P}$ followed only from assuming the existence of short resolution proofs. But automated provers (SAT-solvers) actually construct the proofs, or a proof can be constructed by a polynomial time algorithm from the description of any particular successful computation. Hence the existence of automated provers (SAT-solvers) running in time subexponential in the number of essential variables implies even $\mathcal{P}=\mathcal{N} \mathcal{P}$ (or $\mathcal{N P} \subseteq \mathcal{B P P}$ if the prover is randomised).

Our second remark concerns the exhaustive search; in other words, what do we know about $A(y)$ if we only know that $A(g(x))$ is a tautology but we do not have a short proof of that fact.

Take for $g$ the function from Theorem 2.1 (Part 1.), or any $\mathbf{t t}_{m(\ell), \ell}$ with $m(\ell)=\ell^{\omega(1)}$. Let $n:=2^{\ell}$, and interpret strings $b \in\{0,1\}^{n}$ as truth tables of boolean functions in $\ell$ variables. Hence $b \notin R n g(g)$ implies that $b$ is not computable by a circuit of size $\ell^{O(1)}$.

Assume $A(g(x))$ is a tautology while $A(y)$ is not. Define set $C \subseteq\{0,1\}^{n}$ by:

$$
C:=\left\{b \in\{0,1\}^{n} \mid \neg A(b)\right\} .
$$

Then it satisfies:
(1) $C$ is in $\mathcal{P} /$ poly.
(2) $b \in C$ implies that $b$ is not computable by a size $\ell^{O(1)}$ circuit (i.e. $b$ is not in $\mathcal{P} /$ poly $)$.

Razborov and Rudich [15] defined the concept of a $\mathcal{P} /$ poly-natural proof against $\mathcal{P} /$ poly. It is a $\mathcal{P} /$ poly subset $C$ of $\{0,1\}^{n}$ satisfying condition (2) above, and also condition
(3) The cardinality of $C$ is at least $2^{n} / n^{c}$, some $c \geq 1$.

They proved a remarkable theorem (see [15]) that no such set exists, unless strong pseudo-random number generators do not exists (or, equivalently, strong one-way function do not exists).

In our situation this implies that (under the same assumption) there can be at most $2^{n} / n^{\omega(1)}$ assignments falsifying $A(y)$.

Let me conclude with an open problem: Can the substitution speed-up proofs more than polynomially? That is, are there formulas $A(y)$ having long resolution proofs but $A(g(x))$ having short resolution proofs? In yet another words, does $R$ simulate the system $R_{g}$ defined earlier?

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[^1]:    ${ }^{1}$ In general we could allow functions computable in $\operatorname{NTime}\left(n(k)^{O(1)}\right) /$ poly $\cap$ coNTime $\left(n(k)^{O(1)}\right) /$ poly.

[^2]:    ${ }^{2}[9,13,10,14,11,12]$; the reader may want to read the introductions to [10] or [14], to learn about the main ideas.

[^3]:    ${ }^{3} O(m \log (m))$ bits would suffice but we want simple formulas.

