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Information Processing Letters 101 (2007) 163-167

Information Processing Letters

www.elsevier.com/locate/ipl

# Substitutions into propositional tautologies

Jan Krajíček<sup>1,2</sup>

Isaac Newton Institute, Cambridge, CB3 OEH, UK

Received 5 April 2006; received in revised form 1 September 2006; accepted 8 September 2006 Available online 9 October 2006

Communicated by K. Iwama

#### Abstract

We prove that there is a polynomial time substitution  $(y_1, \ldots, y_n) := g(x_1, \ldots, x_k)$  with  $k \ll n$  such that whenever the substitution instance  $A(g(x_1, \ldots, x_k))$  of a 3DNF formula  $A(y_1, \ldots, y_n)$  has a short resolution proof it follows that  $A(y_1, \ldots, y_n)$  is a tautology. The qualification "short" depends on the parameters k and n. © 2006 Elsevier B.V. All rights reserved.

Keywords: Computational complexity; Proof complexity; Automated theorem proving

Let A(y) be a 3DNF propositional formula in *n* variables  $y = (y_1, ..., y_n)$  and assume that we want to prove that A(y) is a tautology. By substituting y := g(x) with  $x = (x_1, ..., x_k)$  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 a priori (and can witness by a polynomial time constructible resolution proof) that any truth assignment satisfying  $\neg A(g(x_1, ..., x_k))$  would be determined already by its values at  $x_1, ..., x_k$ .

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  $s = 2^{n^{\varepsilon}}$ , for any  $\delta > 0$  there is  $\varepsilon = \varepsilon(\delta) > 0$ , or
- $k = \log(n)^c$  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  $\mathcal{P} = \mathcal{NP}$  no automated provers (or SAT solvers) that are based on DPLL procedure [4,5], even augmented by clause learning [15] or restarts of the procedure [6]

E-mail address: krajicek@maths.ox.ac.uk (J. Krajíček).

<sup>&</sup>lt;sup>1</sup> On leave from Mathematical Institute, Academy of Sciences and Faculty of Mathematics and Physics, Charles University, Prague.

<sup>&</sup>lt;sup>2</sup> The paper was written while I was at the Isaac Newton Institute in Cambridge (program Logic and Algorithms), supported by an EP-SRC grant # N09176. Also supported in part by grants A1019401, AV0Z10190503, MSM0021620839, 201/05/0124, and LC505.

<sup>0020-0190/\$ –</sup> see front matter @ 2006 Elsevier B.V. All rights reserved. doi:10.1016/j.ipl.2006.09.002

can run in time subexponential  $(2^{k^{o(1)}})$  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 A(y) (Section 3). This is a consequence of results of Razborov and Rudich [14].

**Notation.** x, y, z, ... and a, b, ... are tuples of variables and of bits, respectively, the individual variables or bits being denoted  $x_i, y_j, ...$  and  $a_i, b_j, ...$ , respectively. [*n*] is  $\{1, ..., n\}$ .

# 1. Proof complexity generators

A proof complexity generator is any function  $g: \{0, 1\}^* \rightarrow \{0, 1\}^*$  given by a family of circuits<sup>3</sup>  $\{C_k\}_k$ , 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  $b \in \{0, 1\}^{n(k)} \setminus$  $Rng(g_k)$ . Given such a *b*, construct propositional formula  $\tau(C_k)_b$  (denoted simply  $\tau(g)_b$  when  $C_k$ s are canonical) as follows: The atoms of  $\tau(C_k)_b$  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 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  $\tau(C_k)_b$  is proportional to m(k). The formula is a tautology as  $b \notin Rng(g)$ .

The  $\tau$ -formulas have been defined in [7] and independently in [1], and their theory is being developed.<sup>4</sup> 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 [13], the second one is from [9]. 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  $s(k) \ge 1$  be a function, and let  $g = \{g_k\}_k$  be a function as above.

- Function g is s(k)-hard for resolution if any formula τ(C<sub>k</sub>)<sub>b</sub>, b ∈ {0, 1}<sup>n(k)</sup> \ Rng(g), requires resolution proofs of size at least s(k).
- *g* is *s*(*k*)-iterable for resolution iff all disjunctions of the form

$$\tau(C_k)_{B_1}(x^1) \vee \cdots \vee \tau(C_k)_{B_t}(x^1, \dots, x^t)$$

require resolution proofs of size at least s(k). Here  $t \ge 1$  is arbitrary, and  $B_1, \ldots, B_t$  are circuits with n(k) output bits such that:

- $-x^i$  are disjoint k-tuples of atoms, for  $i \leq t$ .
- $B_1$  has no inputs, and inputs to  $B_i$  are among  $x^1$ , ...,  $x^{i-1}$ , 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 latter being the iterability condition with t = 1. (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.)

The disjunction from the definition of the iterability can be informally interpreted as follows. Assume that it is a tautology. Then it may be that already the first disjunct  $\tau(C_k)_{B_1}(x^1)$  is a tautology, meaning that the string  $B_1$  is outside of the range of  $g_k$ . If not, and  $a^1 \in \{0, 1\}^k$  is such that  $g_k(a^1) = B_1$ , then  $B_2(a^1)$  is the next candidate for a string being outside of the range of  $g_k$ . If that fails (and  $a^2$  is a witness) then we move on to  $B_3(a^1, a^2)$ , etc. The fact that the disjunction is a tautology means that in this process we find a string outside of the range of  $g_k$  in at most t rounds.

Exponentially hard functions for resolution do exists. A  $\mathcal{P}/poly$ -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 [9, Theorem 4.2]. Razborov [13, Theorems 2.10, 2.20] gave an independent construction and he noticed that any proof of hardness utilizing 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 [13].

**Theorem 1.2.** (Razborov [13].) *There exists a function*  $g = \{g_w\}_w$ , with  $g_w : \{0, 1\}^w \to \{0, 1\}^{w^2}$ , computed by size  $O(w^3)$  circuits, that is  $2^{w^{\Omega(1)}}$ -iterable for resolution.

<sup>&</sup>lt;sup>3</sup> In general we could allow functions computable in  $NTime(n(k)^{O(1)})/poly \cap coNTime(n(k)^{O(1)})/poly$ .

<sup>&</sup>lt;sup>4</sup> [8,12,9,13,10,11]; the reader may want to read the introductions to [9] or [13], to learn about the main ideas.

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 [9,13]) 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 [9].

**Definition 1.3.** Let  $m \ge \ell \ge 1$ . The truth table function  $\mathbf{tt}_{m,\ell}$  takes as input  $m^2$  bits describing<sup>5</sup> a size  $\leq m$  circuit C with  $\ell$  inputs, and outputs  $2^{\ell}$  bits: the truth table of the function computed by C.

 $\mathbf{t}_{m,\ell}$  is, by definition, equal to zero at inputs that do not encode a size  $\leq m$  circuit with  $\ell$  inputs.

Theorem 1.4. (Krajíček [9].) Assume that there exists a  $\mathcal{P}$ /poly-function  $g = \{g_w\}_w$ , 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  $\mathbf{tt}_{2^{\delta \ell}} \ell$  is  $2^{2^{\Omega(\delta\ell)}}$ -iterable for resolution.
- (2) There is a constant  $c \ge 1$  such that the truth table function  $\mathbf{tt}_{\ell^c,\ell}$  is  $2^{\ell^{1+\Omega(1)}}$ -iterable for resolution.

The theorem (see [9, Theorem 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 [9].

Let  $c \ge 1$  and  $\varepsilon > 0$  be arbitrary constants. Assume that there is a function  $g = \{g_w\}_w$ , with  $g_w: \{0, \}^w \to$  $\{0, 1\}^{w^2}$ , computed by size  $w^c$  circuits and that is  $2^{w^c}$ iterable for resolution.

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

- 1.  $m := w^c \cdot t$ ,
- 2.  $\ell := t \cdot \log(w)$ , 3.  $s := 2^{w^{\varepsilon} t \log(w)}$ .

## Corollary 1.5.

(1) For every c > 1 there are  $\varepsilon > 0$  and a polynomial time computable function  $g = \{g_k\}_k$ ,

 $g_k: \{0, 1\}^k \to \{0, 1\}^{k^c},$ 

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

(2) There are  $\varepsilon > \delta > 0$  and a polynomial time computable function  $g = \{g_k\}_k$ ,  $g_k: \{0, 1\}^k \to \{0, 1\}^{2^{k^{\delta}}}.$ that is.  $2^{k^{\varepsilon}}$ -hard for resolution.

# 2. The substitution

#### Theorem 2.1.

- (1) For any  $\delta > 0$  there are  $\mu > 0$  and a polynomial time computable function  $g = \{g_k\}_k$ , extending k = $n^{\delta}$  bits to n = n(k) bits such that for any 3DNF formula A(y),  $y = (y_1, \ldots, y_n)$ , it holds:
  - If  $A(g_k(x))$  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  $g = \{g_k\}_k$ , extending  $k = \log(n)^c$ bits to n = n(k) bits such that for any 3DNF for*mula* A(y),  $y = (y_1, ..., y_n)$ , *it holds*:
  - If  $A(g_k(x))$  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  $\varepsilon > 0$  and the polynomial time function  $g = \{g_k\}_k$  guaranteed by Corollary 1.5 (Part 1). Hence  $g_k: \{0, 1\}^{n^{\delta}} \to \{0, 1\}^n$ , 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  $2^{k^{\varepsilon}}$ -hardness of g, it must hold that

 $s+n^{\mathcal{O}(1)} \ge 2^{n^{\delta\varepsilon}}.$ 

Hence s must be at least  $2^{n^{\mu}}$ , for suitable  $\mu < \delta \varepsilon$ .

Part 2 is proved analogously, using Corollary 1.5 (Part 2). □

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

 $B(x, z) \vee B(x, w) \vee z_i \equiv w_i$ 

<sup>&</sup>lt;sup>5</sup>  $O(m \log(m))$  bits would suffice but we want simple formulas.

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  $2^{k^{0(1)}} |A(g(x))|^{O(1)}$ .

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 Cook–Reckhow [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  $\mathcal{NP} = co\mathcal{NP}$ .

The equality  $\mathcal{NP} = co\mathcal{NP}$  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  $\mathcal{P} = \mathcal{NP}$  (or  $\mathcal{NP} \subseteq \mathcal{BPP}$  if the prover is randomized).

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  $\mathbf{tt}_{m(\ell),\ell}$  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  $b \notin Rng(g)$  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:

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

Then it satisfies:

- (1) C is in  $\mathcal{P}/poly$ .
- (2) b ∈ C implies that b is not computable by a size ℓ<sup>O(1)</sup> circuit (i.e. b is not in P/poly).

Razborov and Rudich [14] 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  $c \ge 1$ .

They proved a remarkable theorem (see [14]) that no such set exists, unless strong pseudo-random number

generators do not exists (or, equivalently, strong oneway 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?

# Acknowledgements

I am indebted to Antonina Kolokolova (Simon Fraser U.) for discussions on related topics. I thank Klas Markström (Umea) for explaining me a few facts about automated theorem provers and SAT-solvers, and to Pavel Pudlák (Prague) for comments on the draft of the paper.

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