The Computational Complexity of (XOR, AND)-Counting Problems

Andrzej Ehrenfeucht¹ and Marek Karpinski²
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Abstract

We characterize the computational complexity of counting the exact number of satisfying assignments of (XOR, AND)-formulas in their RSE-representation (i.e. equivalently, polynomials in GF[2] $[x_1, ..., x_n]$). This problem refrained for some time efforts to find a polynomial time solution and efforts to prove the problem to be #P-complete. Both main results can be generalized to the arbitrary finite fields GF[q]. Because counting the number of solutions of polynomials over finite fields is generic for many other algebraic counting problems, the results of this paper settle a border line for the algebraic problems with a polynomial time counting algorithms and for problems which are #P-complete. In [Karpinski, Luby 89] the counting problem for arbitrary multivariate polynomials over GF[2] has been proved to have randomized polynomial time approximation algorithms.

¹Dept. of Computer Science, University of Colorado, Boulder, Colorado 80309

²International Computer Science Institute, Berkeley, California 94704. On leave from the University of Bonn. Research partially supported by the Leibniz Center for Research in Computer Science, by the DFG Grant KA 673/2-1, and by the SERC Grant GR-E 68297.

1 Introduction

Let us denote by kXOR the class of all formulas f of the form $f = \bigoplus a_A \land \bigwedge_{i \in A} x_i$, for a 0-1 vector $(a_A)_{A \subseteq \{1,\ldots,n\}}$ such that $|A| \leq k$ (or equivalently, kXOR-formulas f are Galois polynomials $f \in GF[2][x_1,\ldots,x_n]$ of degree at most k). XOR $= \bigcup_k k$ XOR. For a formula $f \in X$ OR with n variables, denote $\#f = |\{(x_1,\ldots,x_n)|f(x_1,\ldots,x_n)=1\}|$. The counting problem for kXOR is the problem of computing #f for any given formula $f \in k$ XOR.

In this paper we prove that the problem of exact counting the number of satisfying arguments of 3XOR-formulas (polynomials of degree 3 over GF[2]) is #P-complete. We design also an $O(n^3)$ -time algorithm for the 2XOR-counting problem.

2 Some Auxiliary Lemmas on Polynomials over GF[2]

Suppose $w_i \in GF[2][x_i, \ldots, x_n]$, $i = 1, \ldots, m$, define a polynomial $u = \bigoplus_{i=1}^m w_i z_i$ for new variables $z_i \notin \{x_1, \ldots, x_n\}$. Define by $\#s(\{w_i\})$ the number of solutions of the system $\{w_i = 0\}_{i=1,\ldots,m}$. For a single polynomial u, #s(u) denotes the number of solutions of u, i.e. $\#s(u) = \#\{\bar{x} | u(\bar{x}) = 0\}$. With this notation we formulate the following

Lemma 1.

$$#s(u) = #s({w_i})2^m + (2^n - #s({w_i}))2^{m-1}$$

Proof:

Suppose $x \in s(\{w_i\})$, then all the vectors $z \in \{0,1\}^m$ are solutions of $u = \bigoplus_{i=1}^m w_i z_i$. There are $\#s(\{w_i\})2^m$ of them. Suppose now that $x \notin s(\{w_i\})$. Denote by $K_x = \{i|w_i(x) \neq 0\}$ the set of all indices of polynomials w_i so that $w_i(x) \neq 0$.

Let us characterize the vectors $y \in \{0,1\}^m$ such that xy is a solution of u. y could be 0 or 1 everywhere besides the coordinates in K_x . On the coordinates of K_x , the number of 1's must add up to 0 (mod 2). There are therefore

$$2^{m-|K_x|} \sum_{r=0}^{|K_x|/2} \binom{|K_x|}{2r} = 2^{m-|K_x|} 2^{|K_x|-1} = 2^{m-1}$$

vectors y such that xy is a solution of u. We note that this number now is independent of the particular form of K_x . This gives for different $x \notin s(\{w_i\})$ different solutions of u, and results in $(2^n - \#s(\{w_i\})2^{m-1})$ additional solutions of u.

We derive some corollaries from Lemma 1.

Lemma 2. The system $\{w_i = 0\}_{i=1,...m}$ has a solution iff $\#s(u) > 2^{n+m-1}$.

 $(\#s(u) \ge 2^{n+m-1} \text{ always holds.})$

Lemma 3.

$$\#s(\{w_i\}) = \frac{\#s(u) - 2^{n+m-1}}{2^{m-1}}$$

In the next section we shall make use of the Lemmas above.

3 3XOR-Counting and -Majority Problems are Hard to Compute

We state now our main hardness result.

Theorem 1. Given an arbitrary 3XOR formula $f \in GF[2][x_1, \ldots, x_n]$, the problem of computing #f is #P-complete.

Proof:

Let us take a monotone 2DNF formula $f = c_1 \lor c_2 \lor ... \lor c_m$ where $c_i = (a_i \land b_i)$ and a_i, b_i are variables. The problem of computing #f for any given monotone 2DNF formula is #P-complete (cf., e.g. [V 79]). We define the system w_i of polynomials by $w_i = a_i b_i$, i = 1, ..., m and construct the polynomial $u = \bigoplus_{i=1}^m w_i z_i$ as in section 2.

By Lemma 3

$$#f = 2^n - \frac{#s(u) - 2^{n+m-1}}{2^{m-1}}.$$

Therefore computing #f for monotone 2DNF formulas f is polynomial time reducible to computing #s(u) for 3XOR formulas.

We characterize also Majority and Solutions' Equilibrium Problems for 4XOR-formulas. (SAT for polynomials f is equivalent with checking whether $f \equiv 0$, trivially doable for explicitly given f.)

For the corresponding results for the (\land, \lor, \neg) -basis see [GJ 79].

Theorem 2. Given any 4XOR formula $f \in GF[2][x_1, \ldots, x_n]$, the problems of deciding whether $\#f > 2^{n-1}$ and $\#f = 2^{n-1}$ are both NP-hard.

Proof:

Let us take 3CNF formula $f = \bigwedge_{i=1}^m (a_i \vee b_i \vee c_i)$ over n variables x_1, \ldots, x_n where a_i, b_i, c_i are literals (nonnegated and negated variables). We shall rewrite f into the system of m equations $\{w_i = (a_i \vee b_i \vee c_i) \oplus 1\}_{i=1,\ldots,m}$ in (XOR, AND) basis by writing

$$\neg x = 1 \oplus x$$

and

$$(a_i \vee b_i \vee c_i) = a_i \oplus b_i \oplus c_i \oplus a_i b_i \oplus a_i c_i \oplus b_i c_i \oplus a_i b_i c_i.$$

Let us construct a polynomial $u \in GF[2][x_1, \ldots, x_n, x_{n+1}, \ldots, x_{n+m}]$ as in Section 2. For k = n+m, the problem of deciding 3CNF SAT is polynomial time reducible to the problem of checking whether $\#s(u) > 2^{k-1}$ or $\#s(u) = 2^{k-1}$.

Remark: Using Valiant's result (cf. [GJ 79], p. 251) on systems of algebraic equations over GF[2], we can analogously prove that the Majority and Equilibrium Problems are NP-hard already for 3XOR-formulas.

4 2XOR-Counting Problem

We are going to design an algorithm to count the number #f for arbitrary $f \in 2XOR$ $(f \in F[2][x_1, ..., x_n], f$ is polynomial of degree 2).

Theorem 3. Given arbitrary 2XOR-formula f, there exists an algorithm working in $O(n^3)$ time for computing #f.

We shall call $f \in GF[2][x_1, \ldots, x_n]$ read-once if every variable x_i in f appears in f at most once.

The proof of Theorem 3 will be based on the following sequence of results.

Lemma 4. Given arbitrary 2XOR-formula $f, f \in GF[2][x_1, \ldots, x_n]$, there exists a read-once 2XOR-formula $g \in GF[2][y_0, \ldots, y_m]$, $m \leq n$, a nonsingular $m \times n$ matrix $T = (t_{ij})$ and an m vector $C = (c_i)$ such that

$$g(\bigoplus_{j=1}^n t_{0j}x_j + c_0, \bigoplus_{j=1}^n t_{1j}x_j + c_1, \dots, \bigoplus_{j=1}^n t_{m-1,j}x_j + c_{m-1}) = f(x_1, \dots, x_n).$$

There exists an algorithm for computing matrix $T = (t_{ij})$ and vector $C = (c_1)$ for arbitrary 2XOR-formulas f working in $O(n^3)$ time. The form of g can be chosen to be

$$g = y_0 \oplus y_1 y_2 \oplus y_3 y_4 \oplus \ldots \oplus y_{m-2} y_{m-1} \oplus z$$
 or $g = y_0 y_1 \oplus y_2 y_3 \oplus \ldots \oplus y_{m-2} y_{m-1} \oplus z$

where $z \in \{0, 1\}$.

Proof:

We shall describe an algorithm for computing matrix $T = (t_{ij})$, vector $C = (c_i)$ and constant z. The algorithm will be by recursion on the set of variables $Var(f) = \{x_1, \ldots, x_n\}$.

Recursion Stage x_i :

Let x := x,

Rewrite f as $f = x\alpha \oplus \beta$ where α is a linear form, and β is the rest of f.

Represent (recursively)

$$\beta = y_0 \oplus y_1 y_2 \oplus y_3 y_4 \oplus \ldots \oplus y_{k-2} y_{k-1} \oplus z$$
 type I

or

$$\beta = y_0 y_1 \oplus y_2 y_3 \oplus \ldots \oplus y_{m-2} y_{m-1} \oplus z$$
 type II

where $z \in GF[2]$ and corresponding nonsingular $k \times (n-i)$ matrix T_{β} and vector C_{β} . Note that $k \leq n-i$.

Consider the following cases:

Case 1. $\alpha = 1$.

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$$\beta$$
 is of type I.

Construct new variables

$$y'_0 := y_0 \oplus x$$

 $y'_i := y_i \quad i = 1, ..., k-1$

- β is of type II.

Construct new variables

$$y'_0 := x$$

 $y'_{i+1} := y_i i = 0, ..., k-1$

Case 2. α is linear independent of the variables of β (α cannot be expressed as a linear combination of the rows of matrix T_{β}). Note that in this case, k < n - i.

Construct new variables

$$y_i' := y_i i = 0, \dots, k-1$$
 $y_k' := x$
 $y_{k+1}' := \alpha$

Case 3. α is linear dependent on the variables of β .

Let
$$\alpha = y_{i_1} \oplus \ldots \oplus y_{i_s} \oplus \left\{ \begin{array}{c} 0 \\ 1 \end{array} \right.$$

3.a. y_s and y_t in α form a term of β .

$$\ldots \oplus xy_s \oplus xy_t \oplus y_sy_t = \ldots(x \oplus y_s)(x \oplus y_t) \oplus \underbrace{x}_{\text{an 'extra' } x}$$

Construct new variables

$$y'_s := y_s \oplus x y'_t := y_t \oplus x$$

3.b. y_s is in α but its 'partner' y_t in a term of β is not in α .

$$\ldots \oplus xy_s \oplus y_sy_t = \ldots y_s(x \oplus y_t)$$

Construct new variables

$$\begin{array}{ccc} y'_s & := & y_s \\ y'_t & := & y_t \oplus x \end{array}$$

3.c. β is of type I.

- α is independent of y_0 and the number of 'free' x is odd. Construct new variable

$$y_0' := y_0 \oplus x$$

- α is dependent of y_0 and the number of 'free' x is odd.

$$\dots \oplus xy_0 \oplus y_0 \oplus x = \dots (x \oplus 1)(y_0 \oplus 1) \oplus 1$$

Construct new variables

$$z' := z \oplus 1$$
 $y'_i := y_{i+1}$ $i = 0, ..., k-2$
 $c'_i := c_{i+1}$ $i = 0, ..., k-2$
 $y'_{k-1} := y_0$
 $c'_{k-1} := c_0 + 1$
 $y'_k := x$
 $c'_k := 1$
 $g \text{ is of type II.}$

- α is dependent of y_0 and the number of 'free' x is even.

$$\ldots \oplus xy_0 \oplus y_0 = \ldots (x \oplus 1)y_0$$

Construct new variables

$$z' := z$$
 $y'_i := y_{i+1}$ $i = 0, ..., k-2$
 $c'_i := c_{i+1}$ $i = 0, ..., k-2$
 $y'_{k-1} := y_0$
 $c'_{k-1} := c_0$
 $y'_k := x$
 $c'_k := 1$
 $g \text{ is of type II.}$

3.d. β is of type II and the number of 'free' x is odd.

Construct new variables

$$y_0' := x$$
 $y_{i+1}' := y_i$ $i = 0, ..., k-1$
 g is of type I.

It is not difficult to check that the algorithm produces the *substitution* matrix $T = [t_{ij}]$ as defined in Lemma 4.

The algorithm works in n recursive steps and each step runs in $O(n^2)$ time.

We complete the proof of Theorem 3.

Lemma 5.

$$\#f = \#g2^{n-m}$$

Proof: Obvious from linear algebra.

Finally, the direct counting arguments give us the following.

Lemma 6.

1. Given a 2XOR-formula $g \in GF[2][x_1, \ldots, x_n]$, $g = x_1x_2 \oplus x_3x_4 \oplus \ldots \oplus x_{n-2}x_{n-1} \oplus x_n$,

$$\#g=2^{n-1}.$$

2. Given a 2XOR-formula $g \in GF[2][x_1, \ldots, x_n]$, $g = x_1x_2 \oplus x_3x_4 \oplus \ldots \oplus x_{n-1}x_n$

$$\#g = 2^{n-1} - 2^{\frac{n-2}{2}}.$$

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