

# Notes on Validated Model Counting

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### 1 Notation

Let  $X = \{x_1, x_2, \dots, x_n\}$  be a set of Boolean variables. An *assignment* is a function  $\alpha$  assigning Boolean values to the variables:  $\alpha : X \rightarrow \{0, 1\}$ . We can also view an assignment as a set of *literals*  $\{l_1, l_2, \dots, l_n\}$ , where each literal  $l_i$  is either  $x_i$  or  $\bar{x}_i$ , corresponding to the assignments  $\alpha(x_i) = 1$  or 0, respectively.

#### 1.1 Boolean Functions

A *Boolean function*  $f : 2^X \rightarrow \{0, 1\}$  can be characterized by the set of assignments for which the function evaluates to 1:  $\mathcal{M}(f) = \{\alpha \mid f(\alpha) = 1\}$ . Let **1** denote the Boolean function that assigns value 1 to every assignment, and **0** denote the assignment that assigns value 0 to every assignment. These are characterized by the universal and empty assignment sets, respectively.

From this we can define the *negation* of function  $f$  as the function  $\neg f$  such that  $\mathcal{M}(\neg f) = \{\alpha \mid f(\alpha) = 0\}$ . We can also define the conjunction and disjunction operations over functions  $f_1$  and  $f_2$  as characterized by the sets  $\mathcal{M}(f_1 \wedge f_2) = \mathcal{M}(f_1) \cap \mathcal{M}(f_2)$  and  $\mathcal{M}(f_1 \vee f_2) = \mathcal{M}(f_1) \cup \mathcal{M}(f_2)$ .

For assignment  $\alpha$  and a Boolean formula  $E$  over  $X$ , we use the notation  $\alpha[E/x_i]$  to denote the assignment  $\alpha'$ , such that  $\alpha'(x_j) = \alpha(x_j)$  for all  $j \neq i$  and  $\alpha'(x_i) = \alpha(E)$ , where  $\alpha(E)$  is the value obtained by evaluating formula  $E$  with each variable assigned the value given by  $\alpha$ . In particular, the notation  $\alpha[\bar{x}_i/x_i]$  indicates the assignment in which the value assigned to  $x_i$  is complemented, while others remain unchanged.

A Boolean function  $f$  is said to be *independent* of variable  $x_i$  if every  $\alpha \in \mathcal{M}(f)$  has  $\alpha[\bar{x}_i/x_i] \in \mathcal{M}(f)$ . The *dependency set* of  $f$ , denoted  $D(f)$  consists of all variables  $x_i$  for which  $f$  is *not* independent.

#### 1.2 Separable Cost Functions

Let  $\mathcal{Z}$  denote the elements of a commutative ring. A *separable cost function*  $\sigma : X \rightarrow \mathcal{Z}$  assigns a value from the ring to each variable. We extend this function by defining the cost of literal as  $\sigma(\bar{x}_i) = 1 - \sigma(x_i)$ , the cost of an assignment as  $\sigma(\alpha) = \prod_{l_i \in \alpha} \sigma(l_i)$ , and the cost of a function  $f$  as  $\sigma(f) = \sum_{\alpha \in \mathcal{M}(f)} \sigma(\alpha)$ .

**Example 1:** Let  $\mathcal{Z}$  be the set of rational numbers and  $\sigma(x_i) = 1/2$  for all variables  $x_i$ . Then cost of every assignment will be  $1/2^n$ , and the cost of a function will be its

density, denoted  $\rho(f)$ . That is  $\rho(f) \in [0, 1]$  will be the fraction of assignments for which the function evaluates to 1. Given an ability to compute the density of function  $f$ , we can scale this by  $2^n$  to compute  $|\mathcal{M}(f)|$ , the core task of model counting. Using density as our metric has the advantage that it does not vary when the function is embedded in a larger domain  $X' \supseteq X$ .

**Example 2:** Let  $\mathcal{Z}$  be a field with  $|\mathcal{Z}| > 2n$ , and let  $\mathcal{H}$  be the set functions mapping elements of  $X$  to elements of  $\mathcal{Z}$ . For two distinct functions  $f_1$  and  $f_2$  and a randomly chosen  $h \in \mathcal{H}$ , the probability that  $h(f_1) = h(f_2)$  will be at most  $2^n/|\mathcal{Z}| < 1/2$ . Therefore, these functions can be used as part of a randomized algorithm for equivalence testing [1].

### 1.3 Computing Cost Functions

Three key properties of separable cost functions makes it possible, in some cases, to compute the cost of a Boolean formula without enumerating all of its satisfying solutions.

**Lemma 1 (Complementation).** *For separable cost function  $\sigma$  and Boolean function  $f$ :  $\sigma(\neg f) = 1 - \sigma(f)$ .*

**Lemma 2 (Conjunction over Independent Domains).** *For separable cost function  $\sigma$  and Boolean functions  $f_1$  and  $f_2$  such that  $D(f_1) \cap D(f_2) = \emptyset$ :  $\sigma(f_1 \wedge f_2) = \sigma(f_1) \cdot \sigma(f_2)$ .*

**Lemma 3 (Disjunction over Disjoint Domains).** *For separable cost function  $\sigma$  and Boolean functions  $f_1$  and  $f_2$  such that  $\mathcal{M}(f_1) \cap \mathcal{M}(f_2) = \emptyset$ :  $\sigma(f_1 \vee f_2) = \sigma(f_1) + \sigma(f_2)$ .*

## References

1. Blum, M., Chandra, A.K., Wegman, M.N.: Equivalence of free Boolean graphs can be decided probabilistically in polynomial time. *Information Processing Letters* **10**(2), 80–82 (18 March 1980)