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# Solving DQBF Through Quantifier Elimination

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**Abstract.** We show how to solve dependency quantified Boolean formulas (DQBF) using a quantifier elimination strategy which yields an equivalent QBF that can be decided using any standard QBF solver. The elimination is accompanied by a number of optimizations which help reduce memory consumption and computation time.

We apply our solver HQS to problems from the domain of verification of incomplete combinational circuits and demonstrate the effectiveness of the proposed algorithm. The results show enormous improvements both in the number of solved instances and in the computation times compared to existing work on validating DQBF.

## 1 Introduction

The last two decades have brought enormous progress in the solution of quantifier-free Boolean formulas—the famous NP-complete SAT problem. While first solvers for SAT were developed in the early 1960s, the actual breakthrough began in the mid-1990s with techniques like conflict-driven clause learning, non-chronological backtracking, watched-literal schemes and many more [1]. They reduced the computational effort to decide the satisfiability of a given formula by many orders of magnitude. Nowadays, formulas with millions of clauses and hundred thousands of variables can be solved.

These algorithmic advances paved the way to a successful adoption of SAT-based techniques in a wide range of applications, among others: hard- and software verification (e. g., bounded model checking [2, 3]), test pattern generation [4, 5], and planning [6].

At the same time it was realized that other applications are hard to model using quantifier-free formulas, but are expressible in a more natural and compact way using quantified Boolean formulas (QBF). Many lessons learned from the development of efficient SAT-solvers could be transferred to the quantified case. Together with further improvements specific to QBF, QBF solvers are becoming a serious possibility to tackle computationally hard problems [7].

The success of SAT- and QBF-based methods encourages us to investigate even more complex Boolean decision problems. Deciding so-called dependency quantified Boolean formulas (DQBF), which are a generalization of QBF, is

NEXPTIME-complete [8]. In contrast to QBF, where variable dependencies always follow a linear order (each existential variable depends on all universal variables left of its position in the quantifier prefix), the dependencies in DQBF are explicitly stated by Henkin-quantifiers [9] allowing also non-linear dependencies which are not expressible with QBF. For many relevant problems from this complexity class there are natural transformations into equivalent DQBF problems. Examples are realizability of incomplete digital circuits [10], the analysis of non-cooperative games with incomplete information [8], and certain bit-vector logics [11, 12]. We expect—as it was the case with SAT and QBF—even more applications to come with the availability of efficient DQBF solvers.

Currently the lack of efficient DQBF solvers severely limits its applicability to practical problems. While Balabanov et al. [13] investigate the theory of DQBF, a first extension of the QDPLL algorithm for deciding QBF to DQBF was proposed by Fröhlich et al. [14], but without experimental evaluation. Finkbeiner and Tentrup [15] gave a fast, but incomplete algorithm for the refutation of DQBF, which is not able to give conclusive answers on satisfied instances. This can easily be combined with complete DQBF solvers as a preprocessing step. A basic variable elimination strategy is presented in [10] demonstrating the need for DQBF formulations experimentally on a series of small incomplete circuits. The DQBF formulations used in [10] solve the so-called realizability problem for incomplete digital designs, i. e., they investigate the question whether missing parts in a design can be implemented in a way such that the complete design is equivalent to a given specification. For most instances, (approximate) QBF formulations returned inconclusive answers, and only DQBF was able to prove unrealizability of the incomplete design. Recently, Fröhlich et al. [16] presented the DQBF solver IDQ, which applies similar instantiation-based techniques as used in state-of-the-art decision procedures for effectively propositional logics [17]; it is the first publically available DQBF solver.

In this paper we present an elimination-based strategy for solving DQBF, accompanied by several optimizations. Our strategy eliminates a minimum set of variables that cause the non-linear dependencies. Hence, the elimination routine is guided in order to obtain an easier-to-solve QBF instance. Once there are only linear dependencies left, we employ a QBF solver for the remaining instance. The elimination routine for both QBF and DQBF is based on And-Inverter-Graphs (AIGs) [18, 19]. For this data structure we introduce an efficient algorithm for pure and unit literal detection, which is employed between the elimination routines.

As reference application we use the analysis of incomplete digital circuits as in [10, 15, 16]. These circuits contain parts which are not yet implemented: so-called black boxes. An important question for an incomplete design is its realizability, which is also known as the partial equivalence checking problem (PEC) [20]. Incomplete designs with black boxes come into play, if already in an early stage of development, when not all modules are available yet, verification techniques are to be employed to find errors in the finished parts. Circuits can also be incomplete because parts have been removed that are notoriously hard to verify like multipliers or large memories, but that are expected not to influence the

property to be checked. Also for diagnostic purposes, the removal of parts of a circuit can be beneficial. Previous SAT- and QBF-based approaches to the PEC problem fail to provide accurate answers in case the design contains more than one black box: Since the quantifiers in QBF are linearly ordered, the exact dependencies of the black boxes on different subsets of the circuit’s signals cannot be expressed [10], such that QBF can only serve as an approximate decision method (like SAT as well).

In our experiments we show that our elimination-based approach improves existing ones significantly: Our solver HQS (“Henkin Quantified Solver”) is able to validate 50 % more benchmark instances than IDQ on the given benchmark set and speeds up the computation time by up to four orders of magnitude.

This technical report is an extended version of [21], containing proofs of the main theorems and additional experimental results in the appendix.

*Organization of the paper* In the following section we briefly introduce the necessary foundations. In Section 3 we present our elimination strategy and the complete algorithm for validating DQBF. An experimental evaluation and comparison with IDQ follows in Section 4. The final Section 5 concludes the paper and gives directions for future research.

## 2 Foundations

In this section, we briefly review the foundations of DQBF, QBF, and And-Inverter Graphs (AIGs).

### 2.1 Dependency quantified Boolean formulas

**Definition 1 (DQBF syntax).** *Let  $V = \{x_1, \dots, x_n, y_1, \dots, y_m\}$  be a set of Boolean variables. A dependency quantified Boolean formula (DQBF)  $\psi$  over the variables  $V$  has the following form:*

$$\forall x_1 \forall x_2 \dots \forall x_n \exists y_1(D_{y_1}) \exists y_2(D_{y_2}) \dots \exists y_m(D_{y_m}) : \varphi,$$

where  $D_{y_i} \subseteq \{x_1, \dots, x_n\}$  for  $i = 1, \dots, m$  and  $\varphi$  is a Boolean formula over  $V$ . The set  $D_{y_i}$  is called the dependency set of  $y_i$ , and  $\exists y_i(D_{y_i})$  is also called a Henkin quantifier [9].

The part  $\forall x_1 \dots \exists y_m(D_{y_m})$  is the *quantifier prefix* of  $\psi$  and  $\varphi$  its *matrix*. Often it is assumed w. l. o. g. that  $\varphi$  is given in conjunctive normal form (CNF). A CNF is a conjunction of clauses, which are disjunctions of variables or their negations (literals). We fix the sets  $V_\psi^\exists = \{y_1, \dots, y_m\}$  of existential and  $V_\psi^\forall = \{x_1, \dots, x_n\}$  of universal variables of  $\psi$ .

For a set  $V' \subseteq V$  of variables, let  $\mathcal{A}(V') = \{v : V' \rightarrow \{0, 1\}\}$  denote the set of assignments of the variables in  $V'$ .

**Definition 2 (DQBF semantics).** Let  $\psi = \forall x_1 \dots \forall x_n \exists y_1(D_{y_1}) \dots \exists y_m(D_{y_m}) : \varphi$  be a DQBF. It is satisfied iff for all  $i = 1, \dots, m$  there are Boolean functions  $s_{y_i} : \mathcal{A}(D_{y_i}) \rightarrow \{0, 1\}$  such that replacing each  $y_i$  in  $\varphi$  by (a Boolean expression for)  $s_{y_i}$  yields a tautology. The function  $s_{y_i}$  is called a Skolem function for  $y_i$ .

Currently there are three solving techniques known in the literature for validating DQBF: 1) search-based [14], 2) elimination-based [10], and 3) instantiation-based [16]. In this paper, we utilize an elimination-based algorithm using AIGs (cf. Section 2.3) as in [10].

An well considered special case of a DQBF is a QBF, in which the dependencies of the existential variables on the universal ones induce a linear ordering. Its syntax is typically defined as follows:

**Definition 3 (QBF syntax).** Let  $V = \{x_1, \dots, x_n, y_1, \dots, y_m\}$  be a set of Boolean variables. A quantified Boolean formula (QBF)  $\Psi$  over the variables  $V$  has the following form:

$$\Psi = \forall X_1 \exists Y_1 \forall X_2 \exists Y_2 \dots \forall X_k \exists Y_k : \varphi,$$

where  $X_1, \dots, X_k$  is a partition of  $\{x_1, \dots, x_n\}$  and  $Y_1, \dots, Y_k$  is a partition of  $\{y_1, \dots, y_m\}$  with  $X_i \neq \emptyset$  for  $i = 2, \dots, k$  and  $Y_j \neq \emptyset$  for  $j = 1, \dots, k - 1$ . The matrix  $\varphi$  is a Boolean formula over  $V$ .

Let  $\Psi$  be a QBF as in Definition 3. It is equivalent to the DQBF  $\psi = \forall x_1 \dots \forall x_n \exists y_1(D_{y_1}) \dots \exists y_m(D_{y_m}) : \varphi$ , where  $D_{y_i} = \bigcup_{k=1}^{\ell} X_k$  with  $Y_\ell$  being the unique set with  $y_i \in Y_\ell$ .

*Example 1.* Consider the following DQBF:

$$\psi = \forall x_1 \forall x_2 \exists y_1(x_1) \exists y_2(x_2) : \varphi.$$

The existential variables  $y_1$  and  $y_2$  depend only on one universal variable. There is no QBF prefix exactly representing these dependencies.

Deciding whether a DQBF is satisfied is NEXPTIME-complete [8], whereas deciding a QBF is “only” PSPACE-complete [22].

## 2.2 Variable elimination for DQBF

Elimination-based methods for propositional formulas are based on the elimination of certain variables by combining cofactors of the formula in an appropriate way. The authors in [10] present a DQBF algorithm based on variable elimination of universal literals.

For a variable  $v \in V$  and a Boolean expression  $\psi$  over  $V \setminus \{v\}$ , let  $\varphi[\psi/v]$  denote the expression which results from replacing all occurrences of  $v$  in  $\varphi$  by  $\psi$ .

**Theorem 1 (Elimination of universal variables, [10]).** Let  $E_{x_i} = \{y_j \in V_\psi^\exists \mid x_i \in D_{y_j}\}$  be the set of existential variables which depend on the universal variable  $x_i$ . Then  $\psi$  is equivalent to the following DQBF:

$$\begin{aligned} \psi' := & \forall x_1 \dots \forall x_{i-1} \forall x_{i+1} \dots \forall x_n \\ & \exists y_1(D_{y_1} \setminus \{x_i\}) \dots \exists y_m(D_{y_m} \setminus \{x_i\}) \underbrace{\exists y'_j(D_{y'_j} \setminus \{x_i\})}_{\text{for all } y_j \in E_{x_i}} : \\ & \varphi[0/x_i] \wedge \varphi[1/x_i][y'_j/y_j \ \forall y_j \in E_{x_i}]. \end{aligned}$$

*Proof.* To simplify notation, w. l. o. g. assume  $i = 1$ , i. e., we eliminate  $x_1$ . Then we have:

$$\begin{aligned} & \models \psi \\ \Leftrightarrow & \exists s_{y_1, D_1}, \dots, s_{y_m, D_m} \text{ with } \models \forall x_1 \dots \forall x_n : \varphi[s_{y_j, D_j}/y_j \ \forall y_j \in V_\psi^\exists] \\ \Leftrightarrow & \exists s_{y_1, D_1}, \dots, s_{y_m, D_m} \text{ with} \\ & \models \forall x_2 \dots \forall x_n : \varphi[s_{y_j, D_j}/y_j \ \forall y_j \in V_\psi^\exists][0/x_1] \wedge \varphi[s_{y_j, D_j}/y_j \ \forall y_j \in V_\psi^\exists][1/x_1] \\ \Leftrightarrow & \exists s_{y_1, D_1}, \dots, s_{y_m, D_m} \text{ with } \models \forall x_2 \dots \forall x_n : \\ & \varphi[0/x_1][s_{y_k, D_k}/y_k \ \forall y_k \in V_\psi^\exists \setminus E_{x_1}][s_{y_j, D_j|_{x_1=0}}/y_j \ \forall y_j \in E_{x_1}] \\ & \wedge \varphi[1/x_1][s_{y_k, D_k}/y_k \ \forall y_k \in V_\psi^\exists \setminus E_{x_1}][s_{y_j, D_j|_{x_1=1}}/y_j \ \forall y_j \in E_{x_1}] \\ \Leftrightarrow & \exists s_{y_1, D_1}, \dots, s_{y_m, D_m} \text{ with} \\ & \models \forall x_2 \dots \forall x_n : \left( \varphi[0/x_1] \wedge \varphi[1/x_1][y'_j/y_j \ \forall y_j \in E_{x_1}] \right) \\ & [s_{y_k, D_k}/y_k \ \forall y_k \in V_\psi^\exists \setminus E_{x_1}][s_{y_j, D_j|_{x_1=0}}/y_j \ \forall y_j \in E_{x_1}][s_{y_j, D_j|_{x_1=1}}/y'_j \ \forall y_j \in E_{x_1}] \\ \Leftrightarrow & \models \forall x_2 \dots \forall x_n \underbrace{\exists y_k(D_k)}_{\text{for all } y_k \notin E_{x_1}} \underbrace{\exists y_j(D_j \setminus \{x_1\}) \exists y'_j(D_j \setminus \{x_1\})}_{\text{for all } y_j \in E_{x_1}} : \\ & \varphi[0/x_1] \wedge \varphi[1/x_1][y'_j/y_j \ \forall y_j \in E_{x_1}] \\ \Leftrightarrow & \models \forall x_2 \dots \forall x_n \exists y_1(D_1 \setminus \{x_1\}) \dots \exists y_m(D_m \setminus \{x_1\}) \underbrace{\exists y'_j(D_j \setminus \{x_1\})}_{\text{for all } y_j \in E_{x_1}} : \\ & \varphi[0/x_1] \wedge \varphi[1/x_1][y'_j/y_j \ \forall y_j \in E_{x_1}]. \end{aligned}$$

□

This proof was originally published in [10].

This elimination rule allows us to replace the DQBF at hand by an equivalent one which contains one universal variable less—until a pure SAT formula is obtained.

In general, the universal elimination comes at the price of additional existential variables, which can be eliminated (like in QBF) if they depend on all universal variables occurring in the formula:

**Theorem 2 (Elimination of existential variables, [10]).** *Consider the following DQBF:*

$$\psi := \forall x_1 \dots \forall x_n \exists y_1(D_{y_1}) \dots \exists y_m(D_{y_m}) : \varphi$$

*If  $D_{y_i} = V_\psi^\forall$ , i. e., if  $y_i$  depends on all universal variables,  $\psi$  is equivalent to:*

$$\forall x_1 \dots \forall x_n \exists y_1(D_{y_1}) \dots \exists y_{i-1}(D_{y_{i-1}}) \exists y_{i+1}(D_{y_{i+1}}) \dots \exists y_m(D_{y_m}) : \varphi[0/y_i] \vee \varphi[1/y_i].$$

The algorithm in [10] uses the elimination of existential variables whenever possible. Otherwise, all universal variables are eliminated until a SAT instance remains, which is validated by a standard SAT-solver.

### 2.3 And-Inverter Graphs

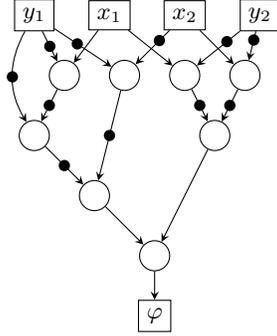
*And-Inverter Graphs* (AIGs) [23, 18] are Boolean circuits composed solely of AND gates with two inputs and inverters. They can be used for representing arbitrary Boolean formulas. In contrast to BDDs [24], AIGs are not canonical—for each propositional formula several structurally different AIG representations can exist—allowing them to be potentially more compact than BDDs.

*Example 2.* In Fig. 1 a small AIG is shown. The white nodes represent the AND gates and the small black nodes the inverter gates. The AIG represents the function:

$$\varphi = \left( \left( \left( \overline{\overline{y_1} \wedge x_1} \wedge \overline{y_1} \right) \wedge \overline{\overline{y_1} \wedge x_2} \right) \wedge \left( \overline{x_1 \wedge y_2} \wedge \overline{x_2 \wedge y_2} \right) \right)$$

For readability reasons we write  $\bar{\chi}$  instead of  $\neg\chi$ . By simple transformations one can see that  $\varphi$  is equivalent to the CNF  $(y_1 \vee x_1) \wedge (y_1 \vee x_2) \wedge (y_2 \vee \neg x_1) \wedge (y_2 \vee \neg x_2)$ .

A special case of AIGs are so-called Functionally Reduced And-Inverter Graphs (FRAIGs) [25]. In contrast to AIGs a FRAIG is “pseudo-canonical”, i. e., there are no two distinct gates in the FRAIG computing the same (or inverse) function. However, FRAIGs still allow multiple structurally different representations of the same function. FRAIGs efficiently support conjunction, disjunction and composition of Boolean functions as well as existential and universal quantification of a single Boolean variable. Hence, FRAIGs have successfully been employed in decision procedures for QBF [19] and DQBF [10]. In our implementation we usually employ AIGs, which are converted into FRAIGs from time to time. Since the distinction is not mandatory for our algorithms, we will always refer in the following to AIGs in the meaning of “AIG or FRAIG”,



**Fig. 1.** Example of an AIQ for the function  $\varphi = (y_1 \vee x_1) \wedge (y_1 \vee x_2) \wedge (y_2 \vee \neg x_1) \wedge (y_2 \vee \neg x_2)$

### 3 Improving DQBF solving

In this section we describe our improved method for elimination-based DQBF solving, based on two cornerstones: 1) a sophisticated heuristics to determine the next variable to be eliminated based on dependency graphs (cf. Section 3.1), and 2) pure literal detection on AIQs (cf. Section 3.2). Furthermore, we utilize techniques known from QBF to improve the scalability. The complete algorithm is summarized in Section 3.3.

#### 3.1 From DQBF to QBF

For QBFs, which exhibit a linearly ordered quantifier prefix, many efficient solvers exist, both search-based (e.g., DepQBF [26]) as well as based on variable elimination (e.g., AIGSOLVE [19, 27]). In order to benefit from the progress these solvers made, we strive for turning the DQBF at hand into an equivalent QBF by eliminating a small (or even minimum) set of universal variables which are responsible for the DQBF's non-linear dependencies. To do so, we define a dependency graph representing the variable dependencies in a DQBF.

**Definition 4.** Let  $\psi = \forall x_1 \dots \forall x_n \exists y_1(D_{y_1}) \dots \exists y_m(D_{y_m}) : \varphi$  be a DQBF. The dependency graph of  $\psi$  is the directed graph  $G_\psi = (V, E)$  with  $V = \{y_1, \dots, y_m\}$  and  $E = \{(y_i, y_\ell) \in V \times V \mid D_{y_i} \not\subseteq D_{y_\ell}\}$ .

The intuition is that there is an edge from the node  $y_i$  to  $y_\ell$  in  $G_\psi$  iff the variable  $y_i$  in  $\psi$  depends on some universal variables on which  $y_\ell$  does not. This implies that if there is an equivalent QBF prefix, then  $y_i$  has to be placed right of  $y_\ell$  with the variables from  $D_{y_i} \setminus D_{y_\ell}$  in between.

Based on dependency graphs and this observation we can detect whether there is an equivalent QBF prefix for a given DQBF.

**Theorem 3.** A DQBF  $\psi$  has an equivalent QBF prefix iff  $G_\psi$  is acyclic.



**Fig. 2.** Example of a dependency graph for  $\forall x_1 \forall x_2 \exists y_1(x_1) \exists y_2(x_2) : \varphi$

*Proof.* First let  $\Psi = \forall X_1 \exists Y_1 \forall X_2 \dots \exists Y_k : \varphi$  be a QBF that is equivalent to a DQBF  $\psi$ , where  $X_1, \dots, X_k$  is a partitioning of  $V_\psi^\forall$  and  $Y_1, \dots, Y_k$  a partitioning of  $V_\psi^\exists$ . We have that  $(y_i, y_\ell) \in E$  iff there are  $j, j' \in \{1, \dots, k\}$  with  $j > j'$  and  $y_i \in Y_j$  and  $y_\ell \in Y_{j'}$ , i. e., there is only an edge from  $y_i$  to  $y_k$  if  $y_k$ 's position in the QBF-prefix is to left of the position of  $y_i$ . Therefore  $G_\psi$  is acyclic.

Now let  $G_\psi$  be acyclic. We inductively construct an equivalent QBF-prefix for  $\psi$ . Set  $G_1 := G_\psi$ . Since  $G_1$  is acyclic, there is a set  $Y_1 \neq \emptyset$  of nodes (existential variables) that do not have an out-going edge. All these variables have the same dependency set. Assume the converse, then there are  $y, y' \in V$ ,  $y \neq y'$ , with  $D_y \neq D_{y'}$ . This implies  $D_y \setminus D_{y'} \neq \emptyset$  or  $D_{y'} \setminus D_y \neq \emptyset$ , i. e.,  $y$  or  $y'$  has an out-going edge, which contradicts the definition of  $Y_1$ . We set  $X_1 := D_y$  for arbitrary  $y \in Y_1$ .

Now assume, for some  $i \geq 1$ , we have defined  $G_j$ ,  $X_j$ , and  $Y_j$  for all  $1 \leq j \leq i$ . We remove from  $G_i$  all nodes in  $Y_i$  together with their incident edges to obtain  $G_{i+1}$ . Since  $G_i$  is acyclic, so is  $G_{i+1}$ . If  $G_{i+1}$  is empty, we set  $\ell := i + 1$  and terminate. Otherwise let  $X_{i+1} \neq \emptyset$  be the set of nodes without out-going edges. With the same argument as before, we can conclude that all nodes in  $X_{i+1}$  have the same dependency set. We set  $X_{i+1} := D_y \setminus (X_1 \cup \dots \cup X_i)$ .

By this procedure we have defined the sets  $X_1, \dots, X_{k-1}$  and  $Y_1, \dots, Y_{k-1}$ . We set  $X_\ell := V_\psi^\forall \setminus \bigcup_{i=1}^{k-1} X_i$ . The QBF-prefix  $\forall X_1 \exists Y_1 \dots \forall X_{k-1} \exists Y_{k-1} \forall X_k$  expresses the same dependencies as the prefix of  $\psi$ .  $\square$

*Example 3.* Consider the DQBF from Example 1:  $\psi = \forall x_1 \forall x_2 \exists y_1(x_1) \exists y_2(x_2) : \varphi$ . As discussed in the example this DQBF has no equivalent QBF prefix. The corresponding dependency graph is given in Fig. 2. From Theorem 3 we see that there is indeed no equivalent QBF prefix due to the cycle.

The following lemma and theorem lead us to a mechanism for obtaining a QBF, given an arbitrary DQBF.

**Lemma 1.** *Let  $\psi$  be a DQBF as before and  $G_\psi = (V, E)$  its dependency graph. Let  $y_1 \rightarrow y_2 \rightarrow \dots \rightarrow y_k \rightarrow y_{k+1} = y_1$  be a cycle in  $G_\psi$  of length  $k$ . Then there is  $i \in \{1, \dots, k\}$  such that  $y_{i+1} \rightarrow y_i$ .*

*Proof.* Let the cycle be given as above. Assume  $(y_{i+1}, y_i) \notin E$  for all  $1 \leq i \leq k$ . That means,  $D_{y_{i+1}} \subseteq D_{y_i}$ . Due to the transitivity of the subset relation we have

$$D_{y_1} = D_{y_{k+1}} \subseteq D_{y_k} \subseteq D_{y_{k-1}} \subseteq \dots \subseteq D_{y_1}.$$

This implies  $D_{y_1} = D_{y_2} = \dots = D_{y_k}$ .

On the other hand,  $y_i \rightarrow y_{i+1}$  implies  $D_{y_i} \not\subseteq D_{y_{i+1}}$  for all  $i = 1, \dots, k$ , i. e.,  $D_{y_i} \neq D_{y_{i+1}}$ , which is a contradiction.  $\square$

**Theorem 4.** For a DQBF  $\psi$ , its dependency graph  $G_\psi$  is cyclic iff there are  $y_i, y_k$  such that  $D_{y_i} \not\subseteq D_{y_k}$  and  $D_{y_k} \not\subseteq D_{y_i}$ .

*Proof.* Lemma 1 implies that if  $G_\psi$  is cyclic, then there are  $y, y'$  with  $y \rightarrow y'$  and  $y' \rightarrow y$ . The other direction is trivial.  $\square$

To check whether a DQBF is actually expressible as QBF, one only has to consider the dependencies between pairs of variables. This property can be used to determine a minimum set of universal variables that have to be eliminated to obtain an equivalent QBF, which can be solved by an arbitrary QBF solver.

We express the problem to determine such a minimum set as a partial MaxSAT problem. MaxSAT is an extension to SAT, determining the maximum number of simultaneously satisfied clauses of a Boolean formula in CNF. Partial MaxSAT is a variation where clauses can be defined as hard or soft. Hard clauses *must* be satisfied, otherwise the MaxSAT instance is unsatisfiable (as in pure SAT), and soft clauses *may* be satisfied (as in pure MaxSAT). Many graph theoretic problems, such as the Max-flow-min-cut problem can be expressed by MaxSAT or its variations. The interested reader is referred to [1] for more details about MaxSAT.

Our partial MaxSAT instance to determine the minimum dependency set to be eliminated is built as follows: For each universal variable  $x \in V_\psi^\forall$  we introduce a variable  $\hat{x}$  of the MaxSAT problem such that for an optimal solution  $\hat{x} = 1$  means that  $x$  belongs to the set of variables to be eliminated.

For each binary loop consisting of existential variables  $y, y' \in V_\psi^\exists$  with dependency sets  $D_y, D_{y'}$  we have to eliminate universal variables such that  $D_y$  becomes a subset of  $D_{y'}$  or vice versa. That means we either have to eliminate all variables in  $D_y \setminus D_{y'}$  or all in  $D_{y'} \setminus D_y$ . This can be expressed as partial MaxSAT as follows:

Let  $C_\psi := \{\{y, y'\} \subseteq V_\psi^\exists \mid D_y \not\subseteq D_{y'} \wedge D_{y'} \not\subseteq D_y\}$  be the set of binary cycles representing the hard constraint:

$$\zeta_\psi^{\text{hard}} := \bigwedge_{\{y, y'\} \in C_\psi} \left( \bigwedge_{x \in D_y \setminus D_{y'}} \hat{x} \vee \bigwedge_{x \in D_{y'} \setminus D_y} \hat{x} \right). \quad (1)$$

The soft constraint for optimization is given by

$$\zeta_\psi^{\text{soft}} := \bigwedge_{x \in V_\psi^\forall} \neg \hat{x}. \quad (2)$$

The MaxSAT solver determines an assignment  $\nu$  of the variables such that  $\zeta_\psi^{\text{hard}}$  is satisfied and a maximum number of variables  $x \in V_\psi^\forall$  is assigned to 0. Hence, we obtain a minimal set of universal variables  $\{x \in V_\psi^\forall \mid \nu(\hat{x}) = 1\}$  that has to be eliminated in the DQBF  $\psi$  such that  $\psi$  can be expressed as QBF.

### 3.2 Unit and Pure Variables

The detection of unit and pure literals is a fundamental technique for search-based QBF solvers which work on a formula in CNF.

The following definition holds for any propositional formula and defines a semantic criterion for a unit or pure variable.

**Definition 5 (Unit and pure variables).** *Let  $\varphi$  be a propositional formula over the variable set  $V$ . A variable  $v \in V$  is positive (negative) unit in  $\psi$ , if  $\varphi[0/v]$  ( $\varphi[1/v]$ , resp.) is unsatisfiable.*

*The variable  $v \in V$  is positive (negative) pure in  $\psi$ , if  $\varphi[0/v] \wedge \neg\varphi[1/v]$  ( $\varphi[1/v] \wedge \neg\varphi[0/v]$ , resp.) is unsatisfiable.*

In case we have detected a unit or pure variable, it can be eliminated as given in the following theorem (cf. also [16]):

**Theorem 5 (Elimination of unit and pure variables).** *Let  $\psi = Q : \varphi$  be a DQBF over  $V$  and  $v \in V$ . We denote the quantifier prefix which results from  $Q$  by removing all occurrences of  $v$  by  $Q \setminus \{v\}$ .*

- *If  $v$  is existentially quantified and positive (negative) unit, then  $\psi$  is equivalent to  $Q \setminus \{v\} : \varphi_{|v=1}$  ( $=0$ ).*
- *If  $v$  is universally quantified and positive or negative unit, then  $\psi$  is unsatisfied.*
- *If  $v$  is existentially quantified and positive (negative) pure, then  $\psi$  is equivalent to  $Q \setminus \{v\} : \varphi_{v=1}$  ( $=0$ ).*
- *If  $v$  is universally quantified and positive (negative) pure, then  $\psi$  is equivalent to  $Q \setminus \{v\} : \varphi_{v=0}$  ( $=1$ ).*

*Proof.* Let  $Q : \varphi$  be the DQBF

$$\psi = \forall x_1 \forall x_2 \dots \forall x_n \exists y_1(D_{y_1}) \exists y_2(D_{y_2}) \dots \exists y_m(D_{y_m}) : \varphi.$$

- First assume,  $v$  is *existentially quantified and positive unit*. W.l.o.g. we assume  $v = y_1$ .  $\psi$  is satisfied iff there are Skolem functions  $s_{y_1}, \dots, s_{y_m}$  such that

$$\varphi(x_1, \dots, x_n, s_{y_1}(D_{y_1}), \dots, s_{y_m}(D_{y_m}))$$

is a tautology. We prove the theorem by contradiction and assume that  $s_{y_1}$  is the constant 0 function. Then there exists an assignment  $\nu$  of the universal variables such that  $s_{y_1}(\nu(D_{y_1})) = 0$  hold. We have:

$$\begin{aligned} & \varphi(\nu(x_1), \dots, \nu(x_n), s_{y_1}(\nu(D_{y_1})), \dots, s_{y_m}(\nu(D_{y_m}))) \\ & \equiv \varphi(\nu(x_1), \dots, \nu(x_n), 0, s_{y_2}(\nu(D_{y_2})), \dots, s_{y_m}(\nu(D_{y_m}))) \\ & \equiv \varphi_{|y_1=0}(\nu(x_1), \dots, \nu(x_n), s_{y_2}(\nu(D_{y_2})), \dots, s_{y_m}(\nu(D_{y_m}))) \\ & \equiv 0. \end{aligned}$$

The last step holds due to the assumption that  $v = y_1$  is positive unit. By applying a constant 0 function,  $\psi$  is always unsatisfiable, which is a contradiction to the assumption. Therefore  $s_{y_1}$  is the constant 1 function, or  $Q \setminus \{v\} : \varphi_{|v=1}$  is equivalent to  $\psi$  respectively. The proof in case that  $v$  is existentially quantified and negative unit can be executed analogously.

- Now let  $v = x_1$  be *universally quantified and positive unit*.  $Q : \varphi$  is satisfied iff there are Skolem functions  $s_{y_1}, \dots, s_{y_m}$  such that  $\varphi(x_1, \dots, x_n, s_{y_1}(D_{y_1}), \dots, s_{y_m}(D_{y_m}))$  is a tautology. Now choose an arbitrary assignment  $\nu$  of the universal variables with  $\nu(x_1) = 0$ . We have

$$\begin{aligned}
& \varphi(\nu(x_1), \dots, \nu(x_n), s_{y_1}(\nu(D_{y_1})), \dots, s_{y_m}(\nu(D_{y_m}))) \\
& \equiv \varphi(0, \dots, \nu(x_n), s_{y_1}(\nu(D_{y_1})), \dots, s_{y_m}(\nu(D_{y_m}))) \\
& \equiv \varphi_{|x_1=0}(\nu(x_2), \dots, \nu(x_n), s_{y_1}(\nu(D_{y_1})), \dots, s_{y_m}(\nu(D_{y_m}))) \\
& \equiv 0.
\end{aligned}$$

Again, the last step holds due to the assumption that  $v = x_1$  is positive unit. Since  $\varphi(x_1, \dots, x_n, s_{y_1}(D_{y_1}), \dots, s_{y_m}(D_{y_m}))$  must be 1 for each assignment of  $x_1, \dots, x_n$ ,  $Q : \varphi$  is not satisfied. The case that  $v$  is universally quantified and negatively unit, is analogous.

- Let  $v = x_1$  be *universally quantified and positive pure*, i. e.,  $\not\models \varphi_{|x_1=0} \wedge \neg \varphi_{|x_1=1}$ . This is equivalent to  $\models \varphi_{|x_1=0} \Rightarrow \varphi_{|x_1=1}$  and to every satisfying assignment of  $\varphi_{|x_1=0}$  being also a satisfying assignment of  $\varphi_{|x_1=1}$ . If  $Q \setminus \{x_1\} : \varphi_{|x_1=0}$  is unsatisfied, then also  $Q : \varphi$ . So let  $Q \setminus \{x_1\} : \varphi_{|x_1=0}$  be satisfied. Then there are Skolem functions  $s_{y_1}, \dots, s_{y_m}$  for the existential variables  $y_1, \dots, y_m$  such that  $\varphi_{|x_1=0}(x_2, \dots, x_n, s_{y_1}(D'_{y_1}), \dots, s_{y_m}(D'_{y_m}))$  is a tautology (with  $D'_{y_i} = D_{y_i} \setminus \{x_1\}$ ). By assumption,  $\varphi_{|x_1=1}(x_2, \dots, x_n, s_{y_1}(D'_{y_1}), \dots, s_{y_m}(D'_{y_m}))$  is a tautology, too, and therefore also

$$\begin{aligned}
& (x_1 \wedge \varphi_{|x_1=1}(x_2, \dots, x_n, s_{y_1}(D'_{y_1}), \dots, s_{y_m}(D'_{y_m}))) \vee \\
& (\neg x_1 \wedge \varphi_{|x_1=0}(x_2, \dots, x_n, s_{y_1}(D'_{y_1}), \dots, s_{y_m}(D'_{y_m}))) \\
& \equiv \varphi(x_1, \dots, x_n, s_{y_1}(D'_{y_1}), \dots, s_{y_m}(D'_{y_m}))
\end{aligned}$$

holds. Therefore, if  $Q \setminus \{x_1\} : \varphi_{|x_1=0}$  is satisfied, then also  $Q : \varphi$ . The proof for negative pure universal variables can be carried out analogously.

- Finally let  $v = y_1$  be *existentially quantified and positive pure*. That means, as for universal positive pure variables that every satisfying assignment of  $\varphi_{|y_1=0}$  is also a satisfying assignment of  $\varphi_{|y_1=1}$ .

Assume that  $Q \setminus \{\exists y_1\} : \varphi_{|y_1=1}$  is satisfied. So there are Skolem functions  $s_{y_2}(D_{y_2}), \dots, s_{y_m}(D_{y_m})$  such that  $\varphi_{|y_1=1}(x_1, \dots, x_n, s_{y_2}(D_{y_2}), \dots, s_{y_m}(D_{y_m}))$  is a tautology. This is equivalent to  $\varphi(x_1, \dots, x_n, 1, s_{y_2}(D_{y_2}), \dots, s_{y_m}(D_{y_m}))$  being a tautology. Therefore  $s_{y_1}(D_{y_1}) = 1$  is a Skolem function for  $y_1$  in  $\psi$ .

Now let  $\psi := Q : \varphi$  be satisfied. We have to show that  $s_{y_1}(D_{y_1}) = 1$  is a Skolem function for  $y_1$  in  $\psi$ . If  $\psi$  is satisfied, there are Skolem functions  $s_{y_1}(D_{y_1}), \dots, s_{y_m}(D_{y_m})$  with  $\varphi(x_1, \dots, x_n, s_{y_1}(D_{y_1}), \dots, s_{y_m}(D_{y_m}))$  being a tautology. Let  $\nu : V_\psi^\forall \rightarrow \{0, 1\}$  be an arbitrary assignment of the universal variables. Then we have

$$\models \varphi(\nu(x_1), \dots, \nu(x_n), s_{y_1}(\nu(D_{y_1})), \dots, s_{y_m}(\nu(D_{y_m})))$$

If  $s_{y_1}(\nu(D_{y_1})) = 0$  holds, then  $\nu$  is a satisfying assignment of  $\varphi_{|y_1=0}(x_1, \dots, x_n, s_{y_2}(D_{y_2}), \dots, s_{y_m}(D_{y_m}))$  and therefore by assumption also of  $\varphi_{|y_1=1}(x_1, \dots, x_n, s_{y_2}(D_{y_2}), \dots, s_{y_m}(D_{y_m}))$ . If  $s_{y_1}(\nu(D_1)) = 1$  holds, then  $\nu$  is a satisfying assignment of  $\varphi_{|y_1=1}(x_1, \dots, x_n, s_{y_2}(D_{y_2}), \dots, s_{y_m}(D_{y_m}))$ .

Hence every assignment  $\nu$  satisfies  $\varphi_{|y_1=1}(x_1, \dots, x_n, s_{y_2}(D_{y_2}), \dots, s_{y_m}(D_{y_m}))$  and  $s_{y_2}(D_{y_2}), \dots, s_{y_m}(D_{y_m})$  are Skolem functions of  $Q \setminus \{y_1\} : \varphi_{|y_1=1}$ .

The proof for existential negative pure variables is similar.

□

QBF solvers employ a sufficient (but not necessary) syntactic criterion to determine unit and pure variables efficiently. A sufficient *and* necessary check is usually too expensive, as in general it requires a check for satisfiability for each variable.

**Lemma 2 (Unit and pure variables in CNF).** *Given a Boolean formula  $\varphi$  over variables  $V$  in CNF and  $v \in V$ , the variable  $v$  is positive (negative) unit, if there exists a clause in  $\varphi$  consisting only of  $v$  ( $\neg v$ ). The variable  $v$  is positive (negative) pure, if  $v$  occurs only positive (negative) in the whole CNF.*

In search-based QBF solvers this check is performed dynamically by considering the CNF resulting from temporarily assigned variables. Furthermore, detection of syntactic pure variables is a standard technique for SAT- and QBF-preprocessors. There exists also semantic checks in preprocessors [27] which require a SAT-check for each variable.

The elimination of unit and pure variables from a formula is particularly beneficial for DQBF because it does not require the duplication of any variables, but rather reduces both the number of variables and the size of the AIG.

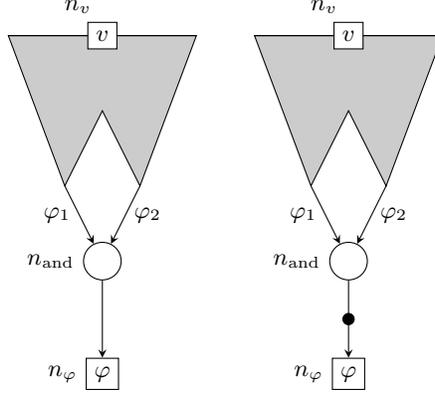
Therefore we want to detect pure literals whenever possible between two elimination steps. Unfortunately semantic checks are quite costly for large AIG structures, and hence we prefer a syntactic check. However, the clause structure is destroyed when using AIGs for representing the matrix, and thus these rather cheap syntactic checks using clauses cannot be applied within our framework. Instead we exploit the following syntactic criterion for AIGs:

**Theorem 6 (Unit and pure variables in AIGs).** *Let  $\psi$  be a DQBF over variables  $V$  with matrix  $\varphi$  and assume that  $\varphi$  is represented by an AIG. Let  $n_v$  be the input node corresponding to  $v \in V$  and  $n_\varphi$  the output node corresponding to  $\varphi$ .*

*If there is a path from  $n_v$  to  $n_\varphi$  without negation then  $v$  is positive unit. If there is a path from  $n_v$  to  $n_\varphi$  such that the only negation on this path appears between  $n_v$  and the first AND node, then  $v$  is negative unit.*

*If the number of negations on all paths from  $n_v$  to  $n_\varphi$  is even, then  $v$  is positive pure. If the number of negations is odd on all paths,  $v$  is negative pure.*

*Proof.* First, we consider the case of a syntactically pure variable. We prove the theorem by induction on the number of AND nodes in the AIG representation of  $\varphi$ . We thereby restrict ourselves to the case of an even number of negations. The other case of an odd number can be shown analogously.



**Fig. 3.** AIG representation of  $\varphi$  with non-negated (left) or negated (right) output edge

First we assume that the AIG does not contain any AND nodes, i. e., the root node  $n_\varphi$ , which represents  $\varphi$ , is directly connected with the input node  $n_v$ . This edge is either negated or not. Therefore we have  $\varphi = v$  or  $\varphi = \neg v$ . In the former case, the number of negations is even on all paths, and according to Theorem 5 we have  $\varphi|_{v=0} \wedge \neg\varphi|_{v=1} = 0 \wedge \neg 1 = 0$  and  $v$  is positive pure. Analogously in the latter case the number of negations is odd on all paths and  $\varphi|_{v=1} \wedge \neg\varphi|_{v=0} = 0 \wedge \neg 1 = 0$  hold. Hence,  $v$  is negative pure.

Now assume that we have shown the claim for all AIGs with up to  $n$  AND nodes and that the AIG for  $\varphi$  consists of  $n + 1$  AND nodes. Then we have the situation shown in Fig. 3 (left or right, depending on whether the output edge is negated or not).

In case the output edge is not negated (Fig. 3 (left)) we have  $\varphi = \varphi_1 \wedge \varphi_2$  or some expressions  $\varphi_1$  and  $\varphi_2$ , corresponding to the inputs of node  $n_{\text{and}}$ . As the edge to the output node  $n_\varphi$  is not negated and all paths to  $n_\varphi$  contain an even number of negations by assumption, all paths leading to  $n_{\text{and}}$  also have to exhibit an even number of negations. As the two sub-AIGs leading to the inputs of  $n_{\text{and}}$  contain at most  $n$  AND nodes, the induction hypothesis holds for them. Therefore we have:

$$\begin{aligned}
\varphi|_{v=0} \wedge \neg\varphi|_{v=1} &\equiv (\varphi_1 \wedge \varphi_2)|_{v=0} \wedge \neg(\varphi_1 \wedge \varphi_2)|_{v=1} \\
&\equiv \varphi_1|_{v=0} \wedge \varphi_2|_{v=0} \wedge (\neg\varphi_1|_{v=1} \vee \neg\varphi_2|_{v=1}) \\
&\equiv (\varphi_1|_{v=0} \wedge \varphi_2|_{v=0} \wedge \neg\varphi_1|_{v=1}) \vee (\varphi_1|_{v=0} \wedge \varphi_2|_{v=0} \wedge \neg\varphi_2|_{v=1}) \\
&\equiv (0 \wedge \varphi_2|_{v=0}) \vee (0 \wedge \varphi_1|_{v=0}) \\
&\equiv 0.
\end{aligned}$$

So  $v$  is positive pure.

Accordingly,  $\varphi = \neg(\varphi_1 \wedge \varphi_2)$  holds in case that the output edge is negated (cf. Fig. 3 (right)). If all paths from  $n_v$  to  $n_\varphi$  contain an even number of negations, then all paths to  $n_{\text{and}}$  contain an odd number of negations. For the sub-AIGs

corresponding to the inputs of  $n_{\text{and}}$  the induction hypothesis holds and we obtain:

$$\begin{aligned}
\varphi|_{v=0} \wedge \neg\varphi|_{v=1} &\equiv \neg(\varphi_1 \wedge \varphi_2)|_{v=0} \wedge \neg\neg(\varphi_1 \wedge \varphi_2)|_{v=1} \\
&\equiv (\neg\varphi_1|_{v=0} \vee \neg\varphi_2|_{v=0}) \wedge (\varphi_1|_{v=1} \wedge \varphi_2|_{v=1}) \\
&\equiv (\neg\varphi_1|_{v=0} \wedge \varphi_1|_{v=1} \wedge \varphi_2|_{v=1})(\neg\varphi_2|_{v=0} \wedge \varphi_1|_{v=1} \wedge \varphi_2|_{v=1}) \\
&\equiv (0 \wedge \varphi_2|_{v=1}) \vee (0 \wedge \varphi_1|_{v=1}) \\
&\equiv 0 .
\end{aligned}$$

This shows that also in this case  $v$  is positive pure.

The proof for a syntactic unit variable is done in an analogous manner:

Again, we assume that the AIG does not contain any AND nodes, with root node  $n_\varphi$  and input node  $n_v$ , which is either negated or not. Therefore we have  $\varphi = v$  or  $\varphi = \neg v$ . In the former case, there exists a path without any negation and according to Theorem 5  $\varphi|_{v=0} = 0$  holds, hence  $v$  is positive unit. If we assume  $\varphi = \neg v$  there exists only one negation between  $n_v$  and the first AND node. Analogously we have  $\varphi|_{v=1} = \neg 1 = 0$  and therefore a negative unit.

Now assume that we have shown this claim for all AIGs with up to  $n$  AND nodes and that the AIG for  $\varphi$  consists of  $n + 1$  AND nodes as in Fig. 3 (left), where w.l.o.g. we assume that there is a path without any negation through the AIG representing  $\varphi_1$ . Therefore  $v$  is positive unit in  $\varphi_1$  and hence  $\varphi_1|_{v=0} = 0$  holds.

Then we obtain that  $v$  is positive unit in  $\varphi$ :

$$\begin{aligned}
\varphi|_{v=0} &\equiv (\varphi_1 \wedge \varphi_2)|_{v=0} \\
&\equiv (\varphi_1|_{v=0} \wedge \varphi_2|_{v=0}) \\
&\equiv (0 \wedge \varphi_2|_{v=0}) \\
&\equiv 0 .
\end{aligned}$$

Analogously, assume a path in  $\varphi_1$ , where there is exactly one negation between  $n_v$  and the first AND node.  $v$  is negative unit in  $\varphi_1$  and  $\varphi_1|_{v=1} = 0$  holds.

Finally, according to Theorem 5 we obtain that  $v$  is negative unit in  $\varphi$ :

$$\begin{aligned}
\varphi|_{v=1} &\equiv (\varphi_1 \wedge \varphi_2)|_{v=1} \\
&\equiv (\varphi_1|_{v=1} \wedge \varphi_2|_{v=1}) \\
&\equiv (0 \wedge \varphi_2|_{v=1}) \\
&\equiv 0 .
\end{aligned}$$

□

By a recursive traversal of the AIG we can determine in  $O(|\varphi| + |V|)$  variables which are unit or pure according to Theorem 6. Here  $|\varphi|$  denotes the number of AND-nodes in the AIG-representation of  $\varphi$  and  $|V|$  the number of variables it depends on. Unit and pure variables are eliminated using Theorem 5.

*Example 4.* Consider again the AIG in Fig. 1. The syntactic check for purity identifies variable  $y_2$  as positive pure because both paths from  $y_2$  to  $\varphi$  contain

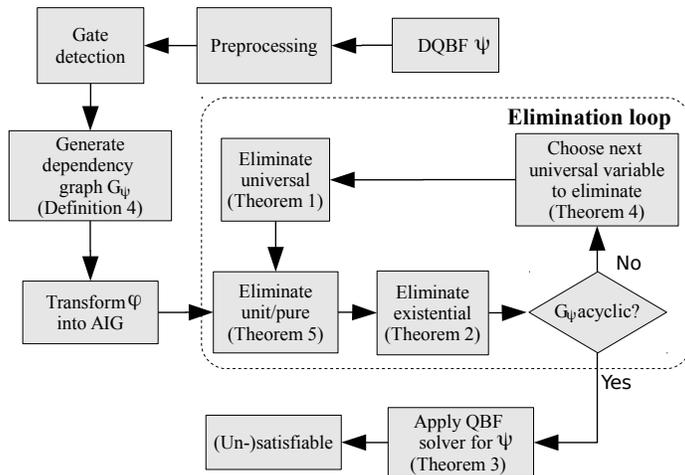


Fig. 4. Algorithmic Flow

an even number of inverters (2 inverters each). The syntactic check fails for the other three variables.

However, as the AIG represents the function  $(y_1 \vee x_1) \wedge (y_1 \vee x_2) \wedge (y_2 \vee \neg x_1) \wedge (y_2 \vee \neg x_2)$ , it is easy to see that  $y_1$  is positive pure according to Lemma 2, which is not detected by the syntactic AIG check.

Like on CNFs the syntactic check on AIGs is incomplete, but—as experiments show—very fast and still detects the majority of all unit and pure variables.

### 3.3 Elimination-based algorithm

The elimination procedure including all improvements presented in this section is illustrated in Fig. 4.

First, we utilize some basic preprocessing steps on the CNF, which are known from QBF preprocessing, but have been adapted to the DQBF setting, namely universal reduction [28], removal of equivalent variables, and unit literal propagation.

After propagating all unit literals, we use the generalized universal reduction rule for clauses [1, Section 23.5]. In QBF, a universal literal  $u$  can be deleted from a clause if no existential literal in the clause depends on  $u$ , which can be naturally extended to DQBF [13, Section 4], [16]. Additionally we detect equivalent variables by analyzing the binary clauses of the formula and perform corresponding replacements. We apply these techniques in alternation until the CNF does not change anymore.

As a last preprocessing step we apply gate detection [19]: Tseitin-encoding [29] introduces an auxiliary variable for each gate output and adds clauses which encode the relationship between gate inputs and the gate output. We detect

these clauses for AND, OR, and XOR gates (with arbitrarily negated inputs), remove the clauses for encoding the relationship between gate inputs and output, and store the relationship directly. The result is a CNF plus a list of gates. After preprocessing we create an AIG representation from the CNF. In this AIG, we replace all literals representing a gate output by the function computed by its gate using the `compose` operation on AIGs. In particular, there is no need for explicit elimination of the internal auxiliary variables resulting from the Tsetin encoding. The interested reader is referred to [19] for further details.

Before the main solving loop starts, we determine a dependency graph  $G_\psi$  (Definition 4), as described in Section 3.1. Based on  $G_\psi$  we use the MaxSAT formulation of Equations 1 and 2 to compute a minimal set of universal variables whose elimination leads to a QBF problem. These universal variables are ordered according to the number of copies of existential variables which would be introduced by an elimination according to Theorem 1. However, we do not perform these eliminations immediately. Rather, in the main solving loop we first check for pure and unit literals of the current AIG (cf. Theorems 5 and 6). Additionally we eliminate all existential variables depending on all universal variables using Theorem 2.

In case  $G_\psi$  is still cyclic, we then choose the next universal variable from the ordered list computed above and eliminate it using Theorem 1.

Finally, when  $G_\psi$  becomes acyclic, we build a QBF upon the linearized dependency structure of the DQBF and employ a QBF solver. Here, we utilize AIGSOLVE [27], which also uses AIGs, thus we can feed the remaining AIG directly into this solver instead of transforming the AIG back into CNF.

Note, we also can stop the main loop if at any time an AIG representation of the constant 0 (or 1) function is obtained (see also [10]). Variables which do not occur in the support of the matrix can be removed from the quantifier prefix.

## 4 Experimental evaluation

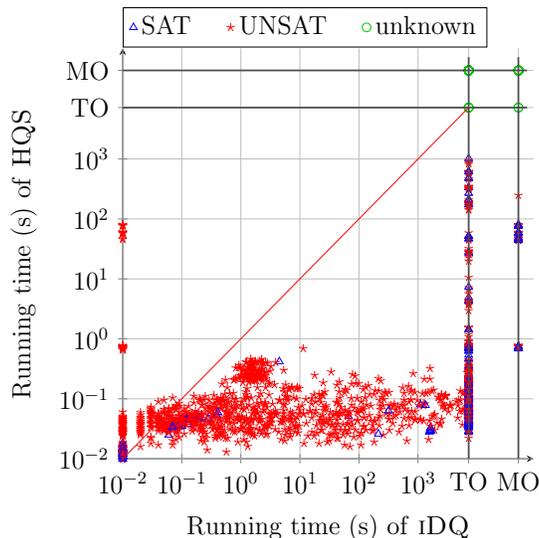
We created a prototypic implementation in C++ of the previously described techniques, called HQS. We used the library `aigpp` [18] for the manipulation of AIGs and AIGSOLVE [27] as a QBF solver. `antom` [30] serves as the solver for partial MaxSAT. For comparison, we took the only available DQBF solver `iDQ` by Fröhlich et al. [16].

For evaluation we used the same 1100 PEC instances as [16].<sup>1</sup> These instances are PEC problems encompassing *adders*, arbiter implementations *lookahead* and *bitcell* described in [31] as well as the circuit family *pec\_xor* from [15].<sup>2</sup> The additional  $3 \cdot 240$  benchmarks *Z4*, *comp*, and *C432* consist also of PEC problems on circuits from the ISCAS 85 circuit library (cf. [32]).

All experiments were run on one Intel Xeon E5-2650v2 core at 2.60 GHz, with 64 GB of main memory and Ubuntu Linux 12.04 in 64-bit mode as operating

<sup>1</sup> Note that in spite of similar names, they differ from the instances used in [15].

<sup>2</sup> In our presentation of the results we omit the 100 *pec\_xor-2* instances as both `iDQ` and HQS solved each instance in less than 0.05 seconds.



**Fig. 5.** Comparison of IDQ and HQS on all 1820 instances

system. We aborted all experiments whose computation time exceeded two hours or which required more than 8 GB of memory.

Figure 5 compares the running times of IDQ and HQS on the considered 1820 benchmark instances. Note that the axes are in logarithmic scale. A marker below the diagonal means that HQS was faster than IDQ, a marker above it that IDQ was faster. A marker on the vertical (horizontal) lines denoted “TO” and “MO” indicate that IDQ (HQS) ran out of time (memory, resp.). The runtimes of HQS are clearly superior for almost all instances, on some instances by four orders of magnitude. HQS solves all instances solved by IDQ and 520 additional ones. An overview on the results for the different benchmark classes is given in Table 1. There we give the number of instances in a class and for both solvers the numbers of solved and unsolved instances, additionally separated into SAT/UNSAT instances and timeout/memout. The columns “total time” give the accumulated running time (in seconds) of the respective solver on those instances which were solved by both solvers. It should be noted that HQS evaluates 1413 of 1555 solved instances ( $\approx 90\%$ ) in less than 1 second (IDQ only 507 of 1035 instances). The time for solving the MaxSAT problem for choosing the variables to eliminate was below 0.06 seconds for all instances. The time for syntactic unit/pure checks was less than 4% of the runtime of each instance.

There are a few instances on which IDQ is faster, in particular among the *z4*, *comp*, and *C432* instances. For these instances, IDQ needs only a single SAT solver call to detect unsatisfiability—and therefore very little time. We could integrate such a SAT solver call into our preprocessing routine; this would reduce the running times for some instances without increasing it measurably for other instances. A more detailed evaluation can be found in Appendix A.

**Table 1.** Experimental results

<b>HQS</b>						
Benchmark	#inst.	solved	(SAT/UNSAT)	unsolved	(TO/MO)	total time
adder	300	300	(42/258)	0	(0/0)	9.72
bitcell	300	300	(7/293)	0	(0/0)	11.27
lookahead	300	300	(10/290)	0	(0/0)	23.17
pec_xor	200	200	(24/176)	0	(0/0)	33.60
z4	240	240	(72/168)	0	(0/0)	4.86
comp	240	155	(39/116)	85	(9/76)	17.82
C432	240	60	(19/41)	180	(0/180)	1,332.58
total	1,820	1,555	(213/1,342)	265	(9/256)	1,433.02

<b>iDQ</b>						
Benchmark	#inst.	solved	(SAT/UNSAT)	unsolved	(TO/MO)	total time
adder	300	216	(3/213)	84	(84/0)	89,827.94
bitcell	300	190	(2/188)	110	(110/0)	78,106.86
lookahead	300	273	(4/269)	27	(27/0)	39,540.15
pec_xor	200	200	(24/176)	0	(0/0)	181.58
z4	240	111	(8/103)	129	(129/0)	41,626.30
comp	240	25	(0/25)	215	(180/35)	11.60
C432	240	20	(0/20)	220	(85/135)	0.20
total	1,820	1,035	(41/994)	785	(615/170)	249,294.63

## 5 Conclusion

We presented a quantifier-elimination strategy based on dependency graphs which can be used to solve DQBF. We developed an efficient detection of unit and pure variables on AIGs, and state an approach turning the DQBF into a QBF by eliminating a minimum set of universal variables. Together with optimizations like simple preprocessing steps, this constitutes a decision procedure which is clearly superior to the available methods both in computation time and the number of solved instances.

Future work will concentrate on more sophisticated preprocessing techniques and improvements on the choice and order of variables to be eliminated in order to increase the scalability. Currently we are working on the integration of Finkbeiner and Tentrup’s unsatisfiability filter [15] into the preprocessor of HQS. First experiments show a significant speed-up on many hard unsatisfiable instances. It seems desirable to have similar filters also for satisfiable instances.

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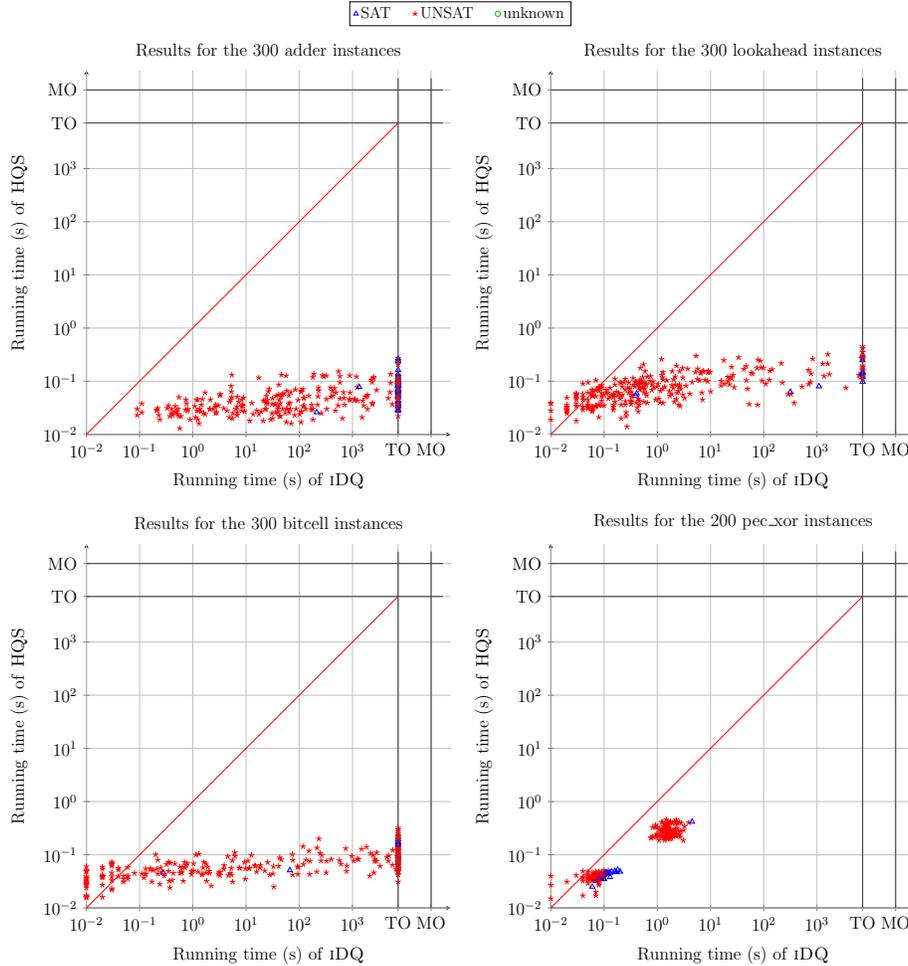
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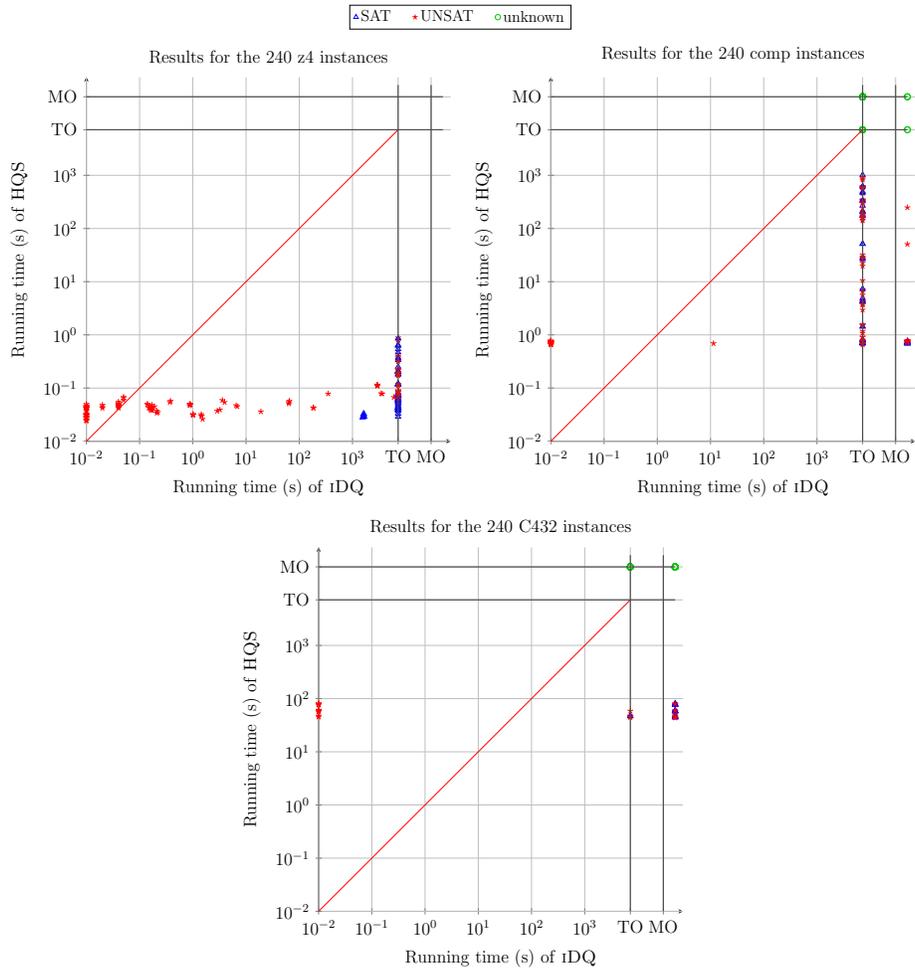
## A Detailed experimental results

Figures 6 and 7 give a detailed comparison of the computation times of iDQ and HQS for each group of instances. We can observe that HQS is in general faster on all classes of benchmarks and solves more instances.



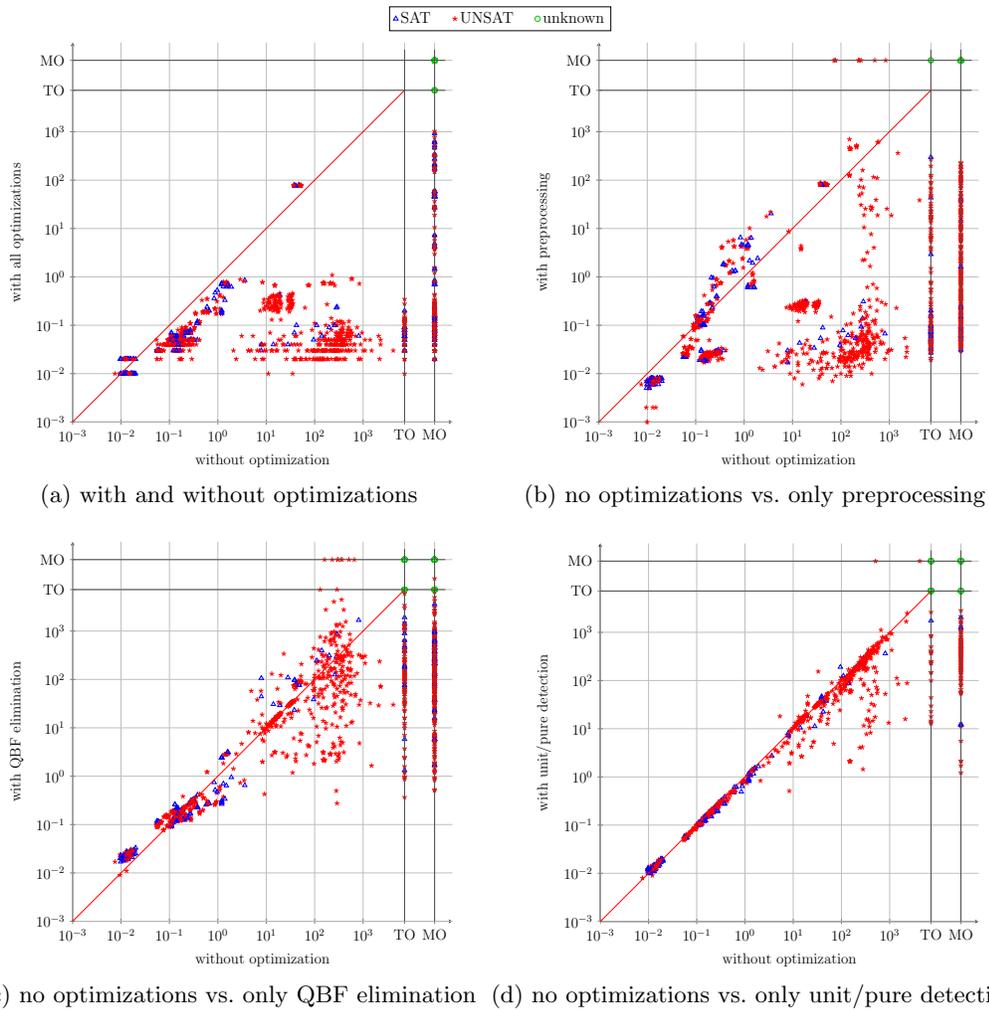
**Fig. 6.** Results for *adder*, *lookahead*, *bitcell*, and *pec\_xor*

Figure 8 demonstrates the effectiveness of the different optimizations. We consider three optimizations: (1) preprocessing, (2) eliminating a minimum set of variables in order to obtain a QBF and order the variables according to the size of their dependency sets, (3) the detection of unit and pure variables on the AIG using a syntactic check.



**Fig. 7.** Results for *z4*, *comp*, and *C432*

Figure 8(a) compares the running times and the number of solved instances with and without all optimizations. In the other three Figures 8(b)–(d) we enable a single optimization and compare it to the unoptimized algorithms. Preprocessing (Figure 8(b)) obviously has a strong effect on both: the computation times and the number of solved instances. The other two optimizations (Figure 8(c)–(d)) mainly increase the number of solved instances. By applying our elimination strategy the formula can be converted faster (or at all) into an equivalent QBF, which is in general solved very efficient by the QBF back-end solver. Finally, every detected pure variable avoids the expansion of the whole AIG or the duplication of the variable reducing the overall memory requirements.



**Fig. 8.** Effectiveness of the optimizations