

A polynomial time algorithm for a class of Quantified Integer Programs

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Abstract. It is well known that the Quantified Satisfiability problem (QSAT) is PSPACE-complete. It follows that the problem of deciding the language of 0/1 Quantified Integer Programs (QIPs) i.e., testing whether a linear system of inequalities has a quantified lattice point is PSPACE-complete. One aspect of research is to focus on designing polynomial time procedures for interesting special cases. In this paper, we show that if the constraint matrix defining a 0/1 QIP is totally unimodular (TUM), then the QIP can be decided in polynomial time .

1 Introduction

Quantified decision problems are useful in modeling situations, wherein a policy (action) can depend upon the effect of imposed stimuli. A typical such situation is a 2– person game. Consider a board game comprised of an initial configuration and two players A and B each having a finite set of moves. A can win the game if the decision problem: *Given the initial configuration, does A have a first move (policy), such that for all possible first moves of B (imposed stimulus), A has a second move, such that for all possible second moves of B, \dots , A eventually wins?* can be answered affirmatively. The board configuration can be represented as a boolean expression or a constraint matrix; the effort involved in representing the board configuration typically determines the tractability of the decision problem.

Definition 1. Let $\{x_1, x_2, \dots, x_n\}$ be a set of n boolean variables. A disjunction of literals (a literal is either x_i or its complement \bar{x}_i) is called a clause, represented by C_i . A satisfiability problem of the form:

$$Q_1 x_1 Q_2 x_2 \dots Q_n x_n C \quad (1)$$

where each Q_i is either a \exists or \forall and $C = C_1 \wedge C_2 \dots \wedge C_m$, is called a Quantified Satisfiability (QSAT) problem.

QSAT has been shown to be PSPACE-complete, even when there are at most 3 literals per clause (Q3SAT) [Pap94], although polynomial time algorithms exist for the case in which there are at most two literals per clause [APT79, Gav93].

Definition 2. Let x_1, x_2, \dots, x_n be a set of n 0/1 variables. An integer program of the form

$$Q_1 x_1 \in \{0, 1\} Q_2 x_2 \in \{0, 1\}, \dots Q_n x_n \in \{0, 1\} \mathbf{A} \cdot \vec{x} \leq \vec{\mathbf{b}}? \quad (2)$$

where each Q_i is either \exists or \forall is called a 0/1 Quantified Integer Program (QIP).

The PSPACE-completeness of QIPs follows directly from the PSPACE-completeness of QSAT; in fact the reduction from QSAT to QIP is identical to the one from SAT to 0/1 Integer Programming. The matrix \mathbf{A} is called the *constraint matrix* of the QIP. Without loss of generality, we assume that the quantifiers are strictly alternating, $Q_1 = \exists$; further we denote the existentially quantified variables using $x_i, i = 1, 2, \dots, n$ and the universally quantified variables using $y_i, i = 1, 2, \dots, n$. Thus we can write an arbitrary 0/1 QIP as :

$$\exists x_1 \in \{0, 1\} \forall y_1 \in \{0, 1\} \exists x_2 \in \{0, 1\} \forall y_2 \in \{0, 1\} \dots \exists x_n \in \{0, 1\} \forall y_n \in \{0, 1\} \mathbf{A} \cdot [\vec{x} \ \vec{y}]^T \leq \vec{\mathbf{b}}? \quad (3)$$

for suitably chosen $\vec{x}, \vec{y}, \mathbf{A}, \vec{\mathbf{b}}, n$

Definition 3. A TQIP is a QIP in which the constraint matrix is totally unimodular.

Definition 4. A linear program of the form

$$\exists x_1 \in [0, 1] \forall y_1 \in [0, 1] \exists x_2 \in [0, 1] \forall y_2 \in [0, 1] \dots \exists x_n \in [0, 1] \forall y_n \in [0, 1] \mathbf{A} \cdot [\vec{x} \ \vec{y}]^T \leq \vec{\mathbf{b}} \quad (4)$$

is called a 0/1 Quantified Linear Program (QLP).

Definition 5. A TQLP is a QLP in which the constraint matrix is totally unimodular.

The complexity of QLPs (0/1 or otherwise) is not known [Joh], although the class of TQLPs can be decided in polynomial time [Sub01a] (See §A).

2 Algorithms and Complexity

Lemma 1.

$$\begin{aligned} \mathbf{L} : \exists x_1 \in \{0, 1\} \forall y_1 \in \{0, 1\} \dots \exists x_n \in \{0, 1\} \forall y_n \in \{0, 1\} \mathbf{A} \cdot [\vec{x} \ \vec{y}]^T \leq \vec{\mathbf{b}} \\ \Leftrightarrow \mathbf{R} : \exists x_1 \in \{0, 1\} \forall y_1 \in [0, 1] \dots \exists x_n \in [0, 1] \forall y_n \in [0, 1] \mathbf{A} \cdot [\vec{x} \ \vec{y}]^T \leq \vec{\mathbf{b}} \end{aligned} \quad (5)$$

Proof: $\mathbf{R} \Rightarrow \mathbf{L}$ is trivial. We focus on $\mathbf{L} \Rightarrow \mathbf{R}$. Pick some vector $\vec{y}^j \in \{0, 1\}^n$; let $\vec{x}^j = [x'_1, x'_2, \dots, x'_n]^T = [c_0, f_1(y'_1), f_2(y'_1, y'_2), \dots, f_{n-1}(y'_1, y'_2, \dots, y'_{n-1})]$ be such that $\mathbf{A} \cdot [\vec{x}^j \ \vec{y}^j]^T \leq \vec{\mathbf{b}}$ (where the f_i are the Skolem functions capturing the dependence of x_i on $y'_1, y'_2, \dots, y'_{i-1}$ and c_0 is a constant in $[0, 1]$). Likewise, pick a second vector $\vec{y}^i \in \{0, 1\}^n$ and let $\vec{x}^i = [x''_1, x''_2, \dots, x''_n]^T = f_{n-1}(y''_1, y''_2, \dots, y''_{n-1})$, such that $\mathbf{A} \cdot [\vec{x}^i \ \vec{y}^i]^T \leq \vec{\mathbf{b}}$. Now consider the parametric point

$\vec{y}^i = \lambda \cdot \vec{y}^j + (1 - \lambda) \cdot \vec{y}^i, 0 \leq \lambda \leq 1$. We shall show that the parametric point $\vec{x}^i = \lambda \cdot \vec{x}^j + (1 - \lambda) \cdot \vec{x}^i, 0 \leq \lambda \leq 1$ is such that $\mathbf{A} \cdot [\vec{x}^i \ \vec{y}^i]^T \leq \vec{\mathbf{b}}$. Observe that $\mathbf{A} \cdot [\vec{x}^i \ \vec{y}^i]^T = \mathbf{A} \cdot [\lambda \cdot \vec{x}^j + (1 - \lambda) \cdot \vec{x}^i \ \lambda \cdot \vec{y}^j + (1 - \lambda) \cdot \vec{y}^i]^T = \mathbf{A} \cdot [\lambda \cdot \vec{x}^j \ \lambda \cdot \vec{y}^j]^T + \mathbf{A} \cdot [(1 - \lambda) \cdot \vec{x}^i \ (1 - \lambda) \cdot \vec{y}^i]^T = \lambda \cdot \mathbf{A} \cdot [\vec{x}^j \ \vec{y}^j]^T + (1 - \lambda) \cdot \mathbf{A} \cdot [\vec{x}^i \ \vec{y}^i]^T \leq \lambda \cdot \vec{\mathbf{b}} + (1 - \lambda) \cdot \vec{\mathbf{b}} \leq \vec{\mathbf{b}}$, since $0 \leq \lambda \leq 1$. Thus the feasible solution space of a Quantified Linear Program is convex and the lemma is proven. \square

Lemma 2.

$$\begin{aligned} \mathbf{L} : \exists x_1 \in \{0, 1\} \forall y_1 \in \{0, 1\} \dots \exists x_n \in \{0, 1\} \forall y_n \in \{0, 1\} \mathbf{A} \cdot [\vec{x} \ \vec{y}]^T \leq \vec{\mathbf{b}} \\ \Leftrightarrow \mathbf{R} : \exists x_1 \in [0, 1] \forall y_1 \in \{0, 1\} \dots \exists x_n \in [0, 1] \forall y_n \in \{0, 1\} \mathbf{A} \cdot [\vec{x} \ \vec{y}]^T \leq \vec{\mathbf{b}} \end{aligned} \quad (6)$$

Proof: Consider any vector $\vec{y} = \{0, 1\}^n$. Substituting this vector in System (3) results in a standard integer program of the form $\exists \vec{x} = \{0, 1\}^n \mathbf{G} \cdot \vec{x} \leq \vec{\mathbf{d}}$, where \mathbf{G} is totally unimodular. Consequently, this system has a solution if and only if the system $\exists \vec{x} = [0, 1]^n \mathbf{G} \cdot \vec{x} \leq \vec{\mathbf{d}}$ is feasible and Lemma (2) follows. \square

Theorem 1. TQIPs can be relaxed to TQLPs, while preserving the integrality of the solution space and hence can be decided in polynomial time.

Proof: Use Lemma (1) to relax the universally quantified variables and Lemma (2) to relax the existentially quantified variables to get a TQLP; then use Algorithm (A.1) in Appendix §A to decide the TQLP in polynomial time. \square

3 Conclusion

The technique used in this paper is different from the one used in [Sub01b] to provide a polyhedral projection procedure to decide Quantified 2-SAT problems.

A Deciding Quantified Linear Programs

In this section, we outline the strategy used in [Sub01a] to solve QLPs. The principal idea underlying Algorithm (A.1) is the elimination of the quantified variables while preserving the solution space. Elimination of a universally quantified variable leaves the number of constraints unchanged, whereas the elimination of an existentially quantified variable using a strategy such as Fourier-Motzkin elimination could lead to a quadratic increase in the number of constraints (see [Sch87]); consequently Algorithm (A.1) could take exponential time in the worst case. In the case of TQLPs though, it runs in time $O(n^5 \cdot \log n)$, where n represents the number of variables in the QLP.

Fast convergence in TQLPs is guaranteed by the following lemma

Lemma 3. *Given a totally unimodular matrix \mathbf{A} of dimensions $m \times n$, for a fixed n , $m = O(n^2)$, if each row is unique.*

Proof: *The above lemma was proved for a superset of totally unimodular matrices viz. totally balanced matrices in [Ans80, AF84]. It therefore follows that Lemma (3) is true. \square*

The import of Lemma (3) is that a totally unimodular constraint matrix cannot have more than $O(n^2)$ non-redundant constraints. The elimination of an existentially quantified variable through Fourier-Motzkin elimination could potentially result in $O(n^4)$ constraints. Eliminating the redundant constraints is a sort operation, that can be implemented in time $O(n^5 \cdot \log n)$ time¹.

Function QLP-DECIDE ($\mathbf{A}, \vec{\mathbf{b}}, \mathbf{Q}$)

- 1: {The array \mathbf{Q} stores the quantifiers i.e. $\mathbf{Q}[i] = Q_i$ }
- 2: **for** ($i = n$ **down to** 1) **do**
- 3: **if** ($\mathbf{Q}[i] = \exists$) **then**
- 4: ELIM-UNIV-VARIABLE(y_i)
- 5: **if** (CHECK-INCONSISTENCY()) **then**
- 6: **return** (**false**)
- 7: **end if**
- 8: PRUNE-CONSTRAINTS()
- 9: **else**
- 10: ELIM-EXIST-VARIABLE(x_i)
- 11: **if** (CHECK-INCONSISTENCY()) **then**
- 12: **return** (**false**)
- 13: **end if**
- 14: **end if**
- 15: **end for**
- 16: System is feasible
- 17: **return**

Algorithm A.1: A Quantifier Elimination Algorithm for deciding Query **E**

Function ELIM-UNIV-VARIABLE ($\mathbf{A}, \vec{\mathbf{b}}, i$)

- 1: Substitute $x_i = 0$ in each constraint that can be written in the form $x_i \geq ()$
- 2: Substitute $x_i = 1$ in each constraint that can be written in the form $x_i \leq ()$

Algorithm A.2: Eliminating Universally Quantified variable $x_i \in [0, 1]$

The procedure ELIM-EXIST-VARIABLE is implemented through the polyhedral projection algorithm known as the Fourier-Motzkin elimination procedure [Sch87] as discussed above.

¹ $O(n^4)$ row vectors can be sorted in time $n^4 \cdot \log n^4$; each comparison takes $O(n)$ time.

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