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Deciding Boolean Algebra with Presburger Arithmetic

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Abstract We describe an algorithm for deciding the first-order multisorted theory BAPA, which combines Boolean algebras of sets of uninterpreted elements (BA) and Presburger arithmetic operations (PA). BAPA can express the relationship between integer variables and cardinalities of unbounded finite sets, and it supports arbitrary quantification over sets and integers. Our motivation for BAPA is deciding verification conditions that arise in the static analysis of data structure consistency properties. Data structures often use an integer variable to keep track of the number of elements they store; an invariant of such a data structure is that the value of the integer variable is equal to the number of elements stored in the data structure. When the data structure content is represented by a set, the resulting constraints can be captured in BAPA. BAPA formulas with quantifier alternations arise when verifying programs with annotations containing quantifiers or when proving simulation relation conditions for refinement and equivalence of program fragments. Furthermore, BAPA constraints can be used for proving the termination of programs that manipulate data structures, as well as in constraint database query evaluation and loop invariant inference. We give a formal description of an algorithm for deciding BAPA. We analyze our algorithm and show that it has optimal alternating time complexity and that the complexity of BAPA matches the complexity of PA. Because it works by a reduction to PA, our algorithm yields the decidability of a combination of sets of uninterpreted elements with any decidable extension of PA. When restricted to BA formulas, the algorithm can be used to decide BA in optimal alternating time. Furthermore, the algorithm can eliminate individual quantifiers from a formula with free variables and therefore perform projection onto a desirable

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set of variables. We have implemented our algorithm and used it to discharge verification conditions in the Jahob system for data structure consistency checking of Java programs; our experience suggest that a straightforward implementation of the algorithm is effective on nontrivial formulas as long as the number of set variables is small. We also report on a new algorithm for solving the quantifier-free fragment of BAPA.

Key words Boolean algebra · Presburger arithmetic · decision procedure · quantifier elimination · complexity · program verification

1 Introduction

Program analysis and verification tools can greatly contribute to software reliability, especially when used throughout the software development process. Such tools are even more valuable if their behavior is predictable, if they can be applied to partial programs, and if they allow the developer to communicate the design information in the form of specifications. Combining the basic idea of [22] with decidable logics leads to analysis tools that have these desirable properties. Such analyses are precise (because formulas represent loop-free code precisely) and predictable (because the checking of verification conditions terminates either with a counterexample or with a proof that there are no counterexamples).

A key challenge in this approach to program analysis and verification is to identify a logic that captures an interesting class of program properties but is nevertheless decidable. In [35], we identify the first-order theory of Boolean algebras (BA) as a useful language for reasoning about dynamically allocated objects: BA allows expressing generalized typestate properties and reasoning about data structures as dynamically changing sets of objects. Here we are interested in BA of all subsets of some set; this theory was shown decidable in [36, 53]; see [25] for the decidability and the complexity of all models of BA axioms.

The motivation for this paper is the fact that we often need to reason not only about the content of a data structure but also about the size of a data structure. For example, we may want to express the fact that the number of elements stored in a data structure is equal to the value of an integer variable that is used to cache the data structure size, or we may want to introduce a decreasing integer measure on the data structure to show program termination. These considerations lead to a natural generalization of the first-order theory of BA of sets, a generalization that allows integer variables in addition to set variables and allows stating relations of the form $|A| = k$, meaning that the cardinality of the set A is equal to the value of the integer variable k. Once we have integer variables, a natural question arises: Which relations and operations on integers should we allow? It turns out that, using only the BA operations and the cardinality operator, we can already define all operations of PA. This leads to the structure BAPA, which properly generalizes both BA and PA.

As we explain in Section 2, a version of BAPA was shown decidable in [18]. Recently, a decision procedure for a fragment of BAPA without quantification over sets was presented in [63], cast as a multisorted theory. Starting from [35] as our motivation, we used quantifier elimination in [30] to show the decidability of the full BAPA, which was initially stated as an open question in [63]. A quantifier-elimination

algorithm for a single-sorted version of BAPA was presented independently in [49] as a way of evaluating queries in constraint databases; that work, however, leaves open the complexity of the decision problem.

Our paper gives the first formal description of a decision procedure for the full first-order theory of BAPA. Furthermore, we analyze our decision procedure and show that it yields optimal computational complexity for BAPA, identical to the complexity of PA [5]. This analysis solves the question of the computational complexity of BAPA.[1] We have also implemented our decision procedure; we report on our initial experience in using the decision procedure in a system for automatic verification of imperative data structures.

1.1 Contributions

We summarize the contributions of our paper as follows.

1. As a *motivation* for BAPA, we show in Section 3 that BAPA constraints can be used for program analysis and verification by expressing (1) data structure invariants and the correctness of procedures with respect to their specifications,[2] (2) simulation relations between program fragments, (3) termination conditions for programs that manipulate data structures, and (4) projection of formulas onto a desired set of variables, with applications in static analysis, model checking, automated abstraction, and relational query evaluation.
2. We present an *algorithm* α (Section 4) that translates BAPA sentences into PA sentences by translating set quantifiers into integer quantifiers, and we show how it can be used to decide the truth value of PA formulas and to eliminate individual quantifiers (Section 6).
3. We analyze our algorithm α and show that its complexity matches the lower bound for PA and is therefore *optimal* (Section 5). This result solves the question of the complexity of the decision problem for BAPA and is the main technical contribution of this paper. Our analysis includes showing an alternating time upper bound for PA, parameterized by the number of quantifier alternations.
4. We discuss our initial experience in using our *implementation* of BAPA to discharge verification conditions generated in the Jahob verification system [32].
5. We observe the following additional results:
   a. PA sentences generated by translating BA sentences without cardinalities can be decided in optimal alternating time for BA (Section 5.4), which gives an alternative proof of upper *bound for* BA *of sets*;
   b. Our algorithm extends to *countable sets* with a predicate distinguishing finite and infinite sets (Section 8);
   c. In contrast to the undecidability of monadic second-order logic over strings (MSOL) when extended with equicardinality operator, we identify a decidable combination of *MSOL with* BA (Section 8).

[1] In [27], we have obtained only the corresponding space upper bound on BAPA; we thank Dexter Kozen for suggesting to use an alternating complexity class of [5] to establish the optimal bound.

[2] This motivation was presented first in [30] and was subsequently used in [64].

Figure 1 Formulas of Boolean algebra (BA).

$$F ::= A \mid F_1 \wedge F_2 \mid F_1 \vee F_2 \mid \neg F \mid \exists x.F \mid \forall x.F$$

$$A ::= B_1 = B_2 \mid B_1 \subseteq B_2 \mid |B| = K \mid |B| \geq K$$

$$B ::= x \mid \mathbf{0} \mid \mathbf{1} \mid B_1 \cup B_2 \mid B_1 \cap B_2 \mid B^c$$

$$K ::= 0 \mid 1 \mid 2 \mid \ldots$$

d. We outline recent results on solving quantifier-free fragment of BAPA using a technique that, for the first time, avoids explicitly constructing an exponentially large system of integer constraints. Our technique yields a polynomial space upper bound on *quantifier-free fragment of* BAPA.

A preliminary version of our results, including the algorithm and complexity analysis, appears in [27, 30] along with additional details on quantifier elimination. The recent results on quantifier-free fragment of BAPA are described in [37, Section 3].

2 The First-Order Theory BAPA

Figure 3 presents the syntax of Boolean algebra with Presburger arithmetic (BAPA), which is the focus of this paper. We next present some justification for the operations in Figure 3. Our initial motivation for BAPA was the use of BA to reason about data structures in terms of sets [34]. Our language for BA (Figure 1) allows cardinality constraints of the form $|A| = K$ where K is a *constant* integer. Such constant cardinality constraints enable quantifier elimination for the resulting language [36, 53]. However, they do not allow stating constraints such as $|A| = |B|$ for two sets A and B, and they cannot represent constraints on relationships between sizes of sets and integer variables. Consider therefore the equicardinality relation $A \sim B$ that holds iff $|A| = |B|$, and consider BA extended with relation $A \sim B$. Define the ternary relation $\mathsf{plus}(A, B, C) \iff (|A| + |B| = |C|)$ by the formula $\exists x_1.\ \exists x_2.\ x_1 \cap x_2 = \emptyset \ \wedge\ A \sim x_1 \ \wedge\ B \sim x_2 \ \wedge\ x_1 \cup x_2 = C$. The relation $\mathsf{plus}(A, B, C)$ allows us to express addition using arbitrary sets as representatives for natural numbers; $\emptyset$ can represent the natural number 0, and any singleton set can represent the natural number 1. (The property of A being a singleton is definable by using, for example, the first-order formula $A \neq \emptyset \wedge \forall B.A \cap B = B \Rightarrow (B = \emptyset \vee B = A)$.) Moreover, we can represent integers as equivalence classes of pairs of natural numbers under the equivalence relation $(x, y) \approx (u, v) \iff x + v = u + y$; this construction also allows us to express the unary predicate of being nonnegative. The quantification over pairs of sets represents quantification over integers, and quantification over integers with the addition operation and the predicate 'being nonnegative' can express all PA operations, presented in Figure 2. Therefore, a natural closure under definable

Figure 2 Formulas of Presburger arithmetic (PA).

$$F ::= A \mid F_1 \wedge F_2 \mid F_1 \vee F_2 \mid \neg F \mid \exists k.F \mid \forall k.F$$

$$A ::= T_1 = T_2 \mid T_1 < T_2 \mid K \mid T$$

$$T ::= K \mid T_1 + T_2 \mid K \cdot T$$

$$K ::= \ldots -2 \mid -1 \mid 0 \mid 1 \mid 2 \ldots$$

$$F ::= A \mid F_1 \wedge F_2 \mid F_1 \vee F_2 \mid \neg F \mid \exists x.F \mid \forall x.F \mid \exists k.F \mid \forall k.F$$

$$A ::= B_1 = B_2 \mid B_1 \subseteq B_2 \mid T_1 = T_2 \mid T_1 < T_2 \mid K \,\mathsf{dvd}\, T$$

$$B ::= x \mid \mathbf{0} \mid \mathbf{1} \mid B_1 \cup B_2 \mid B_1 \cap B_2 \mid B^c$$

$$T ::= k \mid K \mid \mathsf{MAXC} \mid T_1 + T_2 \mid K \cdot T \mid \ \mid B \mid$$

$$K ::= \ldots -2 \mid -1 \mid 0 \mid 1 \mid 2 \ldots$$

Figure 3 Formulas of Boolean algebra with Presburger arithmetic (BAPA).

operations leads to our formulation of the language BAPA in Figure 3, which contains both sets and integers.

The argument above also explains why we attribute the decidability of BAPA to [18, Section 8], which showed the decidability of BA over sets extended with the equicardinality relation $\sim$, using the decidability of the first-order theory of the addition of cardinal numbers.

The language BAPA has two kinds of quantifiers: quantifiers over integers and quantifiers over sets; we distinguish between these two kinds by denoting integer variables with symbols such as k, l and set variables with symbols such as x, y. We use the shorthand $\exists^+ k.F(k)$ to denote $\exists k.k \geq 0 \wedge F(k)$ and, similarly, $\forall^+ k.F(k)$ to denote $\forall k.k \geq 0 \Rightarrow F(k)$. In summary, the language of BAPA in Figure 3 subsumes the language of PA in Figure 2, subsumes the language of BA in Figure 1, and contains nontrivial combination of these two languages in the form of using the cardinality of a set expression as an integer value.

The semantics of operations in Figure 3 is the expected one. We interpret integer terms as integers and interpret set terms as elements of the powerset of a finite set. The MAXC constant denotes the size of the finite universe $\mathcal{U}$, so we require $\mathsf{MAXC} = |\mathcal{U}|$ in all models. Our results generalize to the Boolean algebra of powersets of a countable set; see Section 8.

3 Applications of BAPA

This section illustrates the importance of BAPA constraints. Section 3.1 shows the uses of BAPA constraints to express and verify data structure invariants as well as procedure preconditions and postconditions. Section 3.2 shows how a class of simulation relation conditions can be proved automatically by using a decision procedure for BAPA. Section 3.3 shows how BAPA can be used to express and prove termination conditions for a class of programs. Section 3.4 discusses the applications of quantifier elimination, which is relevant to BAPA because our decision procedure is based on quantifier elimination.

3.1 Verifying Data Structure Consistency

Figure 4 presents a procedure insert in a language that directly manipulates sets. Such languages either can be directly executed [17] or can arise as abstractions of programs in standard languages [35]. The program in Figure 4 manipulates a global set of objects content and an integer field size. The program maintains an invariant

Figure 4 An example procedure.

```
var content : set;
var size : integer;
invariant I ⟺ (size = |content|);

procedure insert(e : element)
maintains I
requires |e| = 1 ∧ |e ∩ content| = 0
ensures size′ > 0
{
  content := content ∪ e;
  size := size + 1;
}
```

I that the size of the set content is equal to the value of the variable size. The insert procedure inserts an element e into the set and correspondingly updates the integer variable. The requires clause (precondition) of the insert procedure is that the parameter e is a non-null reference to an object that is not stored in the set content. The ensures clause (postcondition) of the procedure is that the size variable after the insertion is positive. Note that we represent references to objects (such as the procedure parameter e) as sets with at most one element. An empty set represents a null reference; a singleton set $\{o\}$ represents a reference to object o. The value of a variable after procedure execution is indicated by marking the variable name with a prime.

The insert procedure maintains an invariant, I, that captures the relationship between the size of the set content and the integer variable size. The invariant I is implicitly conjoined with the requires and the ensures clauses of the procedure. The Hoare triple in Figure 5 summarizes the resulting correctness condition for the insert procedure. Figure 6 presents a verification condition corresponding to the Hoare triple in Figure 5. Note that the verification condition contains both set and integer variables, contains quantification over these variables, and relates the sizes of sets to the values of integer variables. Our small example leads to a formula without quantifier alternations; in general, formulas that arise in verification may contain alternations of existential and universal variables over both integers and sets. This paper shows the decidability of such formulas and presents the complexity of the decision procedure.

3.2 Proving Simulation Relation Conditions

BAPA constraints are also useful when proving that a given binary relation on states is a simulation relation between two program fragments. Figure 7 shows one such example. The concrete procedure start1 manipulates two sets: a set of running

Figure 5 Hoare triple for insert procedure.

$$\{|e| = 1 \wedge |e \cap \text{content}| = 0 \wedge \text{size} = |\text{content}|\}$$
$$\text{content} := \text{content} \cup e; \quad \text{size} := \text{size} + 1;$$
$$\{\text{size}' > 0 \wedge \text{size}' = |\text{content}'|\}$$

Figure 6 Verification condition for Figure 5.

$$\forall e.\ \forall \mathsf{content}.\ \forall \mathsf{content}'.\ \forall \mathsf{size}.\ \forall \mathsf{size}'.\\ (|e| = 1 \wedge |e \cap \mathsf{content}| = 0 \wedge \mathsf{size} = |\mathsf{content}| \wedge\\ \mathsf{content}' = \mathsf{content} \cup e \wedge \mathsf{size}' = \mathsf{size} + 1) \Rightarrow\\ \mathsf{size}' > 0 \wedge \mathsf{size}' = |\mathsf{content}'|$$

processes and a set of suspended processes in a process scheduler. The procedure start1 inserts a new process x into the set of running processes R, unless there are already too many running processes. The procedure start2 is a version of the procedure that operates in a more abstract state space: it maintains only the union P of all processes and the number k of running processes. Figure 7 shows a forward simulation relation r between the transition relations for start1 and start2. The standard simulation relation diagram condition is $\forall s_1.\forall s_1'.\forall s_2.(t_1(s_1, s_1') \wedge r(s_1, s_2)) \Rightarrow \exists s_2'.\ (t_2(s_2, s_2') \wedge r(s_1', s_2'))$. In the presence of preconditions, $t_1(s_1, s_1') = (\mathsf{pre}_1(s_1) \Rightarrow \mathsf{post}_1(s_1, s_1'))$ and $t_2(s_2, s_2') = (\mathsf{pre}_2(s_2) \Rightarrow \mathsf{post}_2(s_2, s_2'))$, and sufficient conditions for simulation relation are as follows.

1. $\forall s_1.\forall s_2.r(s_1, s_2) \wedge \mathsf{pre}_2(s_2) \Rightarrow \mathsf{pre}_1(s_1)$
2. $\forall s_1.\forall s_1'.\forall s_2.\exists s_2'.\ r(s_1, s_2) \wedge \mathsf{post}_1(s_1, s_1') \wedge \mathsf{pre}_2(s_2)$
   $\Rightarrow \mathsf{post}_2(s_2, s_2') \wedge r(s_1', s_2')$

Figure 7 shows BAPA formulas that correspond to the simulation relation conditions in this example. Note that the second BAPA formula has a quantifier alternation, which illustrates the relevance of quantifiers in BAPA.

var R : **set**;
var S : **set**;

procedure start1(x)
requires $x \notin \mathsf{R} \wedge |x| = 1 \wedge |\mathsf{R}| < \mathsf{MAXR}$
ensures $\mathsf{R}' = \mathsf{R} \cup x \wedge \mathsf{S}' = \mathsf{S}$
{
  $\mathsf{R} := \mathsf{R} \cup x$;
}

var P : **set**;
var k : **integer**;

procedure start2(x)
requires $x \notin \mathsf{P} \wedge |x| = 1 \wedge \mathsf{k} < \mathsf{MAXR}$
ensures $\mathsf{P}' = \mathsf{P} \cup x \wedge \mathsf{k}' = \mathsf{k} + 1$
{
  $\mathsf{P} := \mathsf{P} \cup x$;
  $\mathsf{k} := \mathsf{k} + 1$;
}

Simulation relation *r:*
$r((\mathsf{R}, \mathsf{S}), (\mathsf{P}, \mathsf{k})) = (\mathsf{P} = \mathsf{R} \cup \mathsf{S} \wedge \mathsf{k} = |\mathsf{R}|)$

Simulation relation conditions in BAPA:

1. $\forall x, \mathsf{R}, \mathsf{S}, \mathsf{P}, \mathsf{k}.(\mathsf{P} = \mathsf{R} \cup \mathsf{S} \wedge \mathsf{k} = |\mathsf{R}|) \wedge (x \notin \mathsf{P} \wedge |x| = 1 \wedge \mathsf{k} < \mathsf{MAXR}) \implies$
   $(x \notin \mathsf{R} \wedge |x| = 1 \wedge |\mathsf{R}| < \mathsf{MAXR})$
2. $\forall x, \mathsf{R}, \mathsf{S}, \mathsf{R}', \mathsf{S}', \mathsf{P}, \mathsf{k}.\exists \mathsf{P}', \mathsf{k}'.((\mathsf{P} = \mathsf{R} \cup \mathsf{S} \wedge \mathsf{k} = |\mathsf{R}|) \wedge (\mathsf{R}' = \mathsf{R} \cup x \wedge \mathsf{S}' = \mathsf{S}) \wedge$
   $(x \notin \mathsf{P} \wedge |x| = 1 \wedge \mathsf{k} < \mathsf{MAXR})) \implies$
   $(\mathsf{P}' = \mathsf{P} \cup x \wedge \mathsf{k}' = \mathsf{k} + 1) \wedge (\mathsf{P}' = \mathsf{R}' \cup \mathsf{S}' \wedge \mathsf{k}' = |\mathsf{R}'|)$

Figure 7 Proving simulation relation in BAPA.

Figure 8 Terminating program.

```
var iter : set;

procedure iterate()
{
  while iter ≠ ∅ do
    var e : set;
    e := choose iter;
    iter := iter \ e;
    process(e);
  done
}
```

3.3 Proving Program Termination

We next show that BAPA is useful for proving program termination. A standard technique for proving termination of a loop is to introduce a ranking function f that maps program states to nonnegative integers, then prove that the value of the function decreases at each loop iteration. In other words, if $t(s, s')$ denotes the relationship between the state at the beginning and the state at the end of each loop iteration, then the condition $\forall s.\forall s'.t(s, s') \Rightarrow f(s) > f(s')$ holds. Figure 8 shows an example program that processes each element of the initial value of set iter; this program can be viewed as manipulating an iterator over a data structure that implements a set. Using the ability to take cardinality of a set allows us to define a natural ranking function for this program. Figure 9 shows the termination proof based on such ranking function. The resulting termination condition can be expressed as a formula that belongs to BAPA and can be discharged by using our decision procedure. In general, we can reduce the termination problem of programs that manipulate both sets and integers to showing a simulation relation with a fragment of a terminating program that manipulates only integers, which can be proved terminating by using techniques of [45]. The simulation relation condition can be proved correct by using our BAPA decision procedure whenever the simulation relation is expressible with a BAPA formula. While one could, in principle, use finite sets directly to describe certain ranking functions, the ability to abstract sets into integers allows us to use existing tools and techniques developed for integer ranking functions.

Figure 9 Termination proof for Figure 8.

Ranking function:
$f(s) = |s|$

Transition relation:
$t(\mathsf{iter}, \mathsf{iter}') = (\exists e.\ |e| = 1 \wedge e \subseteq \mathsf{iter} \wedge \mathsf{iter}' = \mathsf{iter} \setminus e)$

Termination condition in BAPA:
$\forall \mathsf{iter}.\forall \mathsf{iter}'.\ (\exists e.|e| = 1 \wedge e \subseteq \mathsf{iter} \wedge \mathsf{iter}' = \mathsf{iter} \setminus e)$
$\Rightarrow |\mathsf{iter}'| < |\mathsf{iter}|$

3.4 Quantifier Elimination

The fact that BAPA admits quantifier elimination enables applications that involve hiding certain variables from a BAPA formula. Hiding a variable x in a formula means existentially quantifying over x and then eliminating the quantifier $\exists x$. This process can also be viewed as a projection of a formula onto variables other than x. As an example, Figure 10 shows the transition relation inspired by the procedure insert in Figure 5. The transition relation mentions a variable e that is local to the procedure and not meaningful outside it. In the public description of the transition relation, the variable e is existentially quantified. Our quantifier elimination algorithm (Sections 4 and 6) removes the quantifier from the formula and obtains an equivalent formula without quantifiers, such as the one shown in the lower part of Figure 10.

In general, variable hiding is useful in projecting state transitions and invariants onto a desired set of variables, computing relational composition of transition relations, and computing the image of a set under a transition relation. Such symbolic computation of transition relations, with appropriate widenings, can be used to generalize static analyses such as [35] and model-checking approaches such as [10]. The quantifier elimination process here ensures that the transition relation remains represented by a quantifier-free formula throughout the analysis.

Quantifier elimination is also useful for query evaluation in constraint databases [49] and loop invariant inference [23].

4 Decision Procedure for BAPA

This section presents our algorithm, denoted α, that decides the validity of BAPA sentences. The algorithm reduces a BAPA sentence to an equivalent PA sentence with the same number of quantifier alternations and an exponential increase in the total size of the formula. This algorithm has several desirable properties:

1. Given an optimal algorithm for deciding PA sentences, the algorithm α is optimal for deciding BAPA sentences and shows that the complexity of BAPA is the same as the complexity of PA (Section 5).
2. The algorithm α does not eliminate integer variables but instead produces an equivalent quantified PA sentence. The resulting PA sentence can therefore be decided by using *any* decision procedure for PA, including the decision procedures based on automata [6, 24].
3. The algorithm α can eliminate set quantifiers from any extension of PA. We thus obtain a technique for adding a particular form of set reasoning to every extension of PA, and the technique preserves the decidability of the extension. One example of decidable theory that extends PA is MSOL over strings; see Section 8.

$$\exists e.\ |e| = 1 \wedge \mathsf{content}' = \mathsf{content} \cup e$$

$$\Downarrow$$

$$\mathsf{content} \subseteq \mathsf{content}' \ \wedge\ |\mathsf{content}' \setminus \mathsf{content}| \leq 1 \ \wedge\ |\mathsf{content}'| \geq 1$$

Figure 10 Eliminating a local variable from a transition relation.

4. For simplicity we present the algorithm α as a decision procedure for formulas with no free variables, but the algorithm can be used to transform and simplify formulas with free variables as well, because it transforms one quantifier at a time starting from the innermost one. We explore this version of our algorithm in Section 6.

We next describe the algorithm α for transforming a BAPA sentence F_0 into a PA sentence. As the first step of the algorithm, transform F_0 into prenex form

$$Q_p v_p. \ldots Q_1 v_1.\ F(v_1, \ldots, v_p),$$

where F is quantifier-free and each quantifier $Q_i v_i$ is of one the forms $\exists k, \forall k, \exists y, \forall y$, where k denotes an integer variable and y denotes a set variable.

The next step of the algorithm is to separate F into a BA part and a PA part. To achieve this, replace each formula $b_1 = b_2$, where b_1 and b_2 are set expressions, with the conjunction $b_1 \subseteq b_2 \ \wedge\ b_2 \subseteq b_1$, and replace each formula $b_1 \subseteq b_2$ with the equivalent formula $|b_1 \cap b_2^c| = 0$. In the resulting formula, each set variable x occurs in some term $|t(x)|$. Next, use the same reasoning as when generating disjunctive normal form for propositional logic to write each set expression $t(x)$ as a union of cubes (regions in the Venn diagram). The cubes have the form $\bigcap_{i=1}^{n} x_i^{\alpha_i}$, where $x_i^{\alpha_i}$ is either x_i or x_i^c; there are $m = 2^n$ cubes $s_1, \ldots, s_m$. Suppose that $t(x) = s_{j_1} \cup \ldots \cup s_{j_a}$; then replace the term $|t(x)|$ with the term $\sum_{i=1}^{a} |s_{j_i}|$. In the resulting formula, each set x appears in an expression of the form $|s_i|$, where s_i is a cube. For each s_i introduce a new variable l_i. The resulting formula is then equivalent to

$$\begin{aligned} &Q_p v_p. \ldots Q_1 v_1. \\ &\qquad \exists^{+} l_1, \ldots, l_m.\ \bigwedge_{i=1}^{m} |s_i| = l_i \ \wedge\ G_1, \end{aligned} \tag{1}$$

where G_1 is a PA formula. Formula (1) is the starting point of the main phase of algorithm α. This phase successively eliminates quantifiers $Q_1 v_1, \ldots, Q_p v_p$ while maintaining a formula of the form

$$\begin{aligned} &Q_p v_p \ldots Q_r v_r. \\ &\qquad \exists^{+} l_1 \ldots l_q.\ \bigwedge_{i=1}^{q} |s_i| = l_i \ \wedge\ G_r, \end{aligned} \tag{2}$$

where G_r is a PA formula, r grows from 1 to $p+1$, and $q = 2^e$, where e for $0 \leq e \leq n$ is the number of set variables among $v_p, \ldots, v_r$. The list $s_1, \ldots, s_q$ is the list of all 2^e partitions formed from the set variables among $v_p, \ldots, v_r$.

We next show how to eliminate the innermost quantifier $Q_r v_r$ from the formula (2). During this process, the algorithm replaces the formula G_r with a formula G_{r+1}, which has more integer quantifiers. If v_r is an integer variable, then the number of sets q remains the same; and if v_r is a set variable, then q reduces from 2^e to 2^{e-1}. We next consider each of the four possibilities $\exists k, \forall k, \exists y, \forall y$ for the quantifier $Q_r v_r$. Case $\exists k$: Consider first the case $\exists k$. Because k does not occur in $\bigwedge_{i=1}^{q} |s_i| = l_i$, simply move the existential quantifier to G_r, and let $G_{r+1} = \exists k. G_r$, which completes the step.

Case $\forall k$: For universal quantifiers, it suffices to let $G_{r+1} = \forall k.G_r$, again because k does not occur in $\bigwedge_{i=1}^{q} |s_i| = l_i$.
Case $\exists y$: We next show how to eliminate an existential set quantifier $\exists y$ from

$$\exists y.\, \exists^{+} l_1 \ldots l_q.\, \bigwedge_{i=1}^{q} |s_i| = l_i \;\wedge\; G_r, \tag{3}$$

which is equivalent to $\exists^{+} l_1 \ldots l_q.\, (\exists y.\, \bigwedge_{i=1}^{q} |s_i| = l_i) \;\wedge\; G_r$. This is the key step of the algorithm and relies on the following lemma.

Lemma 1 Let $b_1, \ldots, b_n$ be finite disjoint sets and $l_1, \ldots, l_n, k_1, \ldots, k_n$ be natural numbers. Then the following two statements are equivalent:

1. There exists a finite set y such that

$$\bigwedge_{i=1}^{n} |b_i \cap y| = k_i \;\wedge\; |b_i \cap y^c| = l_i. \tag{4}$$

2.

$$\bigwedge_{i=1}^{n} |b_i| = k_i + l_i. \tag{5}$$

Proof ($\Rightarrow$) Suppose that there exists a set y satisfying Eq. (4). Because $b_i \cap y$ and $b_i \cap y^c$ are disjoint, $|b_i| = |b_i \cap y| + |b_i \cap y^c|$, so $|b_i| = k_i + l_i$.

($\Leftarrow$) Suppose that Eq. (5) holds, so $|b_i| = k_i + l_i$ for each of the pairwise disjoint sets $b_1, \ldots, b_n$. For each b_i choose a subset $y_i \subseteq b_i$ such that $|y_i| = k_i$. Because $|b_i| = k_i + l_i$, we have $|b_i \cap y_i^c| = l_i$. Having chosen $y_1, \ldots, y_n$, let $y = \bigcup_{i=1}^{n} y_i$. For $i \neq j$ we have $b_i \cap y_j = \emptyset$ and $b_i \cap y_j^c = b_i$, so $b_i \cap y = y_i$ and $b_i \cap y^c = b_i \cap y_i^c$. By the choice of y_i, we conclude that y is the desired set for which Eq. (4) holds. □

In the quantifier elimination step, assume without loss of generality that the set variables $s_1, \ldots, s_q$ are numbered such that $s_{2i-1} \equiv s_i' \cap y^c$ and $s_{2i} \equiv s_i' \cap y$ for some cube s_i'. Then apply Lemma 1, and replace each pair of conjuncts

$$|s_i' \cap y^c| = l_{2i-1} \;\wedge\; |s_i' \cap y| = l_{2i}$$

with the conjunct $|s_i'| = l_{2i-1} + l_{2i}$, yielding the formula

$$\exists^{+} l_1 \ldots l_q.\, \bigwedge_{i=1}^{q'} |s_i'| = l_{2i-1} + l_{2i} \;\wedge\; G_r \tag{6}$$

for $q' = 2^{e-1}$. Finally, to obtain a formula of the form Eq. (2) for $r+1$, introduce fresh variables l_i' constrained by $l_i' = l_{2i-1} + l_{2i}$, rewrite Eq. (6) as

$$\exists^{+} l_1' \ldots l_{q'}'.\, \bigwedge_{i=1}^{q'} |s_i'| = l_i' \;\wedge\; (\exists^{+} l_1 \ldots l_q.\, \bigwedge_{i=1}^{q'} l_i' = l_{2i-1} + l_{2i} \;\wedge\; G_r),$$

and let

$$G_{r+1} \equiv \exists^{+} l_1 \ldots l_q.\ \bigwedge_{i=1}^{q'} l'_i = l_{2i-1} + l_{2i}\ \wedge\ G_r. \qquad (\exists\text{-step})$$

This completes the description of the elimination of an existential set quantifier $\exists y$.

Case $\forall y$: To eliminate a set quantifier $\forall y$, observe that

$$\neg(\exists^{+} l_1 \ldots l_q.\ \bigwedge_{i=1}^{q} |s_i| = l_i\ \wedge\ G_r)$$

is equivalent to $\exists^{+} l_1 \ldots l_q.\ \bigwedge_{i=1}^{q} |s_i| = l_i\ \wedge\ \neg G_r$, because existential quantifiers over l_i together with the conjuncts $|s_i| = l_i$ act as definitions for l_i, so we may first substitute all values l_i into G_r, then perform the negation, and then extract back the definitions of all values l_i. By expressing $\forall y$ as $\neg\exists y\neg$, we can show that the elimination of $\forall y$ is analogous to elimination of $\exists y$: introduce fresh variables $l'_i = l_{2i-1} + l_{2i}$, and let

$$G_{r+1} \equiv \forall^{+} l_1 \ldots l_q.\ (\bigwedge_{i=1}^{q'} l'_i = l_{2i-1} + l_{2i})\ \Rightarrow\ G_r. \qquad (\forall\text{-step})$$

Final step: After eliminating all quantifiers by repeatedly applying the step of the algorithm, we obtain a formula of the form $\exists^{+} l.\ |1| = l \wedge G_{p+1}(l)$. Namely, in the step when we have only one set variable y and its negation y^c, we can write $|y|$ and $|y^c|$ as $|1 \cap y$ and $|1 \cap y^c|$ and apply the algorithm one more time. We then define the result of the algorithm, denoted $\alpha(F_0)$, to be the PA sentence $G_{p+1}(\mathsf{MAXC})$.

We summarize the algorithm in Figure 11. We use $f; g$ to denote the function composition $g \circ f$, and we use f^* to denote iterative application of function f. The prenex function transforms a formula into the prenex form, whereas the separate function transforms it into form Eq. (1). We have argued above that each of the individual steps of the algorithm is equivalence preserving, so we have the following lemma.

$$\begin{aligned}
&\alpha_1\,(\exists y.\ \exists^{+} l_1 \ldots l_{2q'}.\ \textstyle\bigwedge_{i=1}^{q'} |s_i \cap y^c| = l_{2i-1} \wedge |s_i \cap y| = l_{2i} \wedge G_r) = \\
&\qquad \exists^{+} l'_1 \ldots l'_{q'}.\ \textstyle\bigwedge_{i=1}^{q'} |s_i| = l'_i \wedge \exists^{+} l_1 \ldots l_{2q'}.\ \bigwedge_{i=1}^{q'} .l'_i = l_{2i-1} + l_{2i} \wedge G_r \\
&\alpha_1\,(\forall y.\ \exists^{+} l_1 \ldots l_{2q'}.\ \textstyle\bigwedge_{i=1}^{q'} |s_i \cap y^c| = l_{2i-1} \wedge |s_i \cap y| = l_{2i} \wedge G_r) = \\
&\qquad \exists^{+} l'_1 \ldots l'_{q'}.\ \textstyle\bigwedge_{i=1}^{q'} |s_i| = l'_i \wedge \forall^{+} l_1 \ldots l_{2q'}.\ \bigwedge_{i=1}^{q'} .l'_i = l_{2i-1} + l_{2i} \Rightarrow G_r \\
&\alpha_1\,(\exists k.\ \exists^{+} l_1 \ldots l_{2q'}.\ \textstyle\bigwedge_{i=1}^{q'} |s_i \cap y^c| = l_{2i-1} \wedge |s_i \cap y| = l_{2i} \wedge G_r) = \\
&\qquad \exists^{+} l_1 \ldots l_{2q'}.\ \textstyle\bigwedge_{i=1}^{q'} |s_i \cap y^c| = l_{2i-1} \wedge |s_i \cap y| = l_{2i} \wedge \exists k.G_r \\
&\alpha_1\,(\forall k.\ \exists^{+} l_1 \ldots l_{2q'}.\ \textstyle\bigwedge_{i=1}^{q'} |s_i \cap y^c| = l_{2i-1} \wedge |s_i \cap y| = l_{2i} \wedge G_r) = \\
&\qquad \exists^{+} l_1 \ldots l_{2q'}.\ \textstyle\bigwedge_{i=1}^{q'} |s_i \cap y^c| = l_{2i-1} \wedge |s_i \cap y| = l_{2i} \wedge \forall k.G_r \\
&\alpha_F(G(|\mathbf{1}|)) = G(\mathsf{MAXC}) \\
&\alpha = \mathsf{prenex}\ ;\ \mathsf{separate}\ ;\ \alpha_1^*\ ;\ \alpha_F
\end{aligned}$$

Figure 11 Algorithm α for translating BAPA sentences to PA sentences.

Figure 12 Translation of the BAPA sentence from Figure 6 into a PA sentence.

$$
\begin{array}{l}
\forall^+ l_1.\forall^+ l_0.\ \mathsf{MAXC} = l_1 + l_0 \Rightarrow \\
\forall^+ l_{11}.\forall^+ l_{01}.\forall^+ l_{10}.\forall^+ l_{00}. \\
l_1 = l_{11} + l_{01} \wedge l_0 = l_{10} + l_{00} \Rightarrow \\
\quad \forall^+ l_{111}.\ \forall^+ l_{011}.\ \forall^+ l_{101}.\ \forall^+ l_{001}. \\
\quad \forall^+ l_{110}.\ \forall^+ l_{010}.\ \forall^+ l_{100}.\ \forall^+ l_{000}. \\
\quad l_{11} = l_{111} + l_{011}\ \wedge l_{01} = l_{101} + l_{001} \wedge \\
\quad l_{10} = l_{110} + l_{010}\ \wedge l_{00} = l_{100} + l_{000} \Rightarrow \\
\qquad \forall size.\forall size'. \\
\qquad\ (l_{111} + l_{011} + l_{101} + l_{001} = 1 \wedge \\
\qquad\ l_{111} + l_{011} = 0 \wedge \\
\qquad\ l_{111} + l_{011} + l_{110} + l_{010} = size \wedge \\
\qquad\ l_{100} = 0 \wedge \\
\qquad\ l_{011} + l_{001} + l_{010} = 0 \wedge \\
\qquad\ size' = size + 1) \Rightarrow \\
\qquad\quad (0 < size' \wedge \\
\qquad\quad l_{111} + l_{101} + l_{110} + l_{100} = size')
\end{array}
$$

Lemma 2 The transformations prenex, separate, α_1, α_F are all equivalence preserving (with respect to the BAPA interpretation).

By induction we obtain the correctness of our algorithm.

Theorem 3 *The algorithm α in Figure 11 maps each* BAPA*-sentence F_0 into an equivalent* PA *sentence $\alpha(F_0)$.*

The validity of PA sentences is decidable [46]. In combination with a decision procedure for PA such as [6, 21, 47], the algorithm α is a decision procedure for BAPA sentences.

As an illustration, we show the result of running the algorithm α on the BAPA formula in Figure 6. The result is the PA formula in Figure 12. Note that the structure of the resulting formula mimics the structure of the original formula: every set quantifier is replaced by the corresponding block of quantifiers over nonnegative integers constrained to partition the previously introduced integer variables. Figure 13 presents the correspondence between the set variables of the BAPA formula and the integer variables of the translated PA formula. Note that the relationship $\mathsf{content}' = \mathsf{content} \cup e$ translates into the conjunction of the constraints $|\mathsf{content}' \cap (\mathsf{content} \cup e)^c| = 0 \wedge |(\mathsf{content} \cup e) \cap \mathsf{content}'^c| = 0$, which reduces to the conjunction $l_{100} = 0 \wedge l_{011} + l_{001} + l_{010} = 0$ using the translation of set expressions into the disjoint union of partitions and the correspondence in Figure 13.

5 Complexity of BAPA

In this section, we analyze the algorithm α from Section 4 and use it to show that the computational complexity of BAPA is identical to the complexity of PA, which is $\mathsf{STA}(*, 2^{2^{n^{O(1)}}}, n)$ [5], that is, alternating doubly exponential time with a linear number of alternations.

Figure 13 Correspondence, denoted H, between integer variables in Figure 12 and set variables in Figure 6.

general relationship:

$$l_{i_1,\ldots,i_k} = |\mathsf{set}_q^{i_1} \cap \mathsf{set}_{q+1}^{i_2} \cap \ldots \cap \mathsf{set}_S^{i_k}|$$

$$q = S - (k-1)$$

(S is number of set variables)

in this example:

$$\mathsf{set}_1 = \mathsf{content}'$$
$$\mathsf{set}_2 = \mathsf{content}$$
$$\mathsf{set}_3 = e$$

$$\left.\begin{array}{l} l_{000} = |\mathsf{content}'^c \cap \mathsf{content}^c \cap e^c| \\ l_{001} = |\mathsf{content}'^c \cap \mathsf{content}^c \cap e| \\ l_{010} = |\mathsf{content}'^c \cap \mathsf{content} \cap e^c| \\ l_{011} = |\mathsf{content}'^c \cap \mathsf{content} \cap e| \\ l_{100} = |\mathsf{content}' \cap \mathsf{content}^c \cap e^c| \\ l_{101} = |\mathsf{content}' \cap \mathsf{content}^c \cap e| \\ l_{110} = |\mathsf{content}' \cap \mathsf{content} \cap e^c| \\ l_{111} = |\mathsf{content}' \cap \mathsf{content} \cap e| \end{array}\right\} (H)$$

An alternating Turing machine [14] is a generalization of a nondeterministic Turing machine that, in addition to making nondeterministic (existential) choices, can make universal choices. A universal choice step succeeds iff the computation succeeds for all the chosen values. We then have the following definition.

Definition 4 [5] $\mathsf{STA}(s(n), t(n), a(n))$ denotes the class of languages computable by using an alternating Turing machine that uses $s(n)$ cells on its tape, runs in $t(n)$ steps, and performs $a(n)$ alternations.

In the remainder of this section we first show the lower bound on the complexity of BAPA. We then use the algorithm α in the previous section and a parameterized upper bound on PA to establish the matching upper bound on BAPA. We show that the algorithm α can also decide BA formulas using optimal resources $\mathsf{STA}(*, 2^n, n)$, which is the complexity established in [25]. Moreover, by construction, our procedure reduces to the procedure for PA formulas if there are no set quantifiers. Therefore, the algorithm α can be used as a component for an optimal algorithm for BAPA as well as for the special cases of BA and PA.

5.1 Lower Bound on the Complexity of Deciding BAPA

The lower bound of PA follows by showing how to encode full integer arithmetic in which quantifiers are bounded by 2^{2^n} using formulas of size n [26, Lecture 23].

Fact 1 ([5], Page 75) The truth value of PA formulas of size $O(n)$ can encode the acceptance of an alternating Turing machine with n alternations running in 2^{2^n} steps.

Because BAPA contains PA, the lower bound directly applies to BAPA.

Lemma 5 The truth value of BAPA formulas of size $O(n)$ can encode the acceptance of an alternating Turing machine with n alternations running in 2^{2^n} steps.

5.2 Parameterized Upper Bound on PA

As a step toward establishing the upper bound for BAPA, we show a parameterized upper bound on PA. We use the following result.

Fact 2 ([48] Page 324) If F is a closed PA formula $(Q_1 x_1) \ldots (Q_r x_r) G(x_1, \ldots, x_r)$ of size $n > 4$ with m quantifier alternations, where $Q_1, \ldots, Q_r$ are quantifiers and G is quantifier-free, then F is true iff the formula

$$\left(Q_1 x_1 \leq 2^{2^{(c+1)n^{m+3}}}\right) \ldots \left(Q_r x_r \leq 2^{2^{(c+r)n^{m+3}}}\right) G(x_1, \ldots, x_r)$$

with bounded quantifiers is true for some $c > 0$.

Fact 2 allows us to establish the following parameterized upper bound on PA.

Theorem 6 *A* PA *sentence of size n with m quantifier alternations can be decided in* $\mathsf{STA}(*, 2^{n^{O(m)}}, m)$.

Proof Analogous to the proof of the space upper bound in [48, Page 325]. To decide formulas with m quantifier alternations, use an alternating Turing machine that does m alternations, mimicking the quantifier alternations. The Turing machine guesses the assignments to variables $x_1, \ldots, x_r$ in ranges given by Fact 2. This process requires guessing

$$2^{(c+1)n^{m+3}} + \ldots + 2^{(c+r)n^{m+3}} \leq 2^{c_1 n^{m+4}}$$

bits (because $r \leq n$) and then evaluating the formula in time proportional to $n 2^{c_1 n^{m+4}} \leq 2^{c_2 n^{m+4}}$, which is in $2^{n^{O(m)}}$. □

5.3 Upper Bound on the Complexity of Deciding BAPA

We next show that the algorithm in Section 4 transforms a BAPA sentence F_0 into a PA sentence whose size is at most exponential and which has the same number of quantifier alternations. Theorem 6 will then give us the upper bound on BAPA that matches the lower bound of Lemma 5.

If F is a formula in prenex form, let $\mathsf{size}(F)$ denote the size of F, and let $\mathsf{alts}(F)$ denote the number of quantifier alternations in F.

Lemma 7 For the algorithm α from Section 4 there is a constant $c > 0$ such that $\mathsf{size}(\alpha(F_0)) \leq 2^{c \cdot \mathsf{size}(F_0)}$ and $\mathsf{alts}(\alpha(F_0)) = \mathsf{alts}(F_0)$. Moreover, the algorithm α runs in $2^{O(\mathsf{size}(F_0))}$ deterministic time.

Proof To gain some intuition on the size of $\alpha(F_0)$ compared to the size of F_0, compare first the formula in Figure 12 with the original formula in Figure 6. Let n denote the size of the initial formula F_0, and let S be the number of set variables. Note that the following operations are polynomially bounded in time and space:

transforming a formula into prenex form and transforming relations $b_1 = b_2$ and $b_1 \subseteq b_2$ into the form $|b| = 0$. Introducing set variables for each partition and then replacing each $|b|$ with a sum of integer variables yields formula G_1, whose size is bounded by $O(n2^S S)$ (the last S factor is because representing a variable from the set of K variables requires space $\log K$). The subsequent transformations introduce the existing integer quantifiers, whose size is bounded by n, and introduce additionally $2^{S-1} + \ldots + 2 + 1 = 2^S - 1$ new integer variables along with the equations that define them. Note that the defining equations always have the form $l'_i = l_{2i-1} + l_{2i}$ and have size bounded by S. We therefore conclude that the size of $\alpha(F_0)$ is $O(nS(2^S + 2^S))$ and therefore $O(nS2^S)$, which is in $2^{O(n)}$. Note that we have obtained a more precise bound $O(nS2^S)$ indicating that the exponential explosion is caused only by set variables. Further, the fact that the number of quantifier alternations is the same in F_0 and $\alpha(F_0)$ is immediate because the algorithm replaces one set quantifier with a block of corresponding integer quantifiers. □

Combining Lemma 7 with Theorem 6, we obtain the upper bound for BAPA.

Lemma 8 The validity of a BAPA sentence of size n and the number of quantifier alternations m can be decided in $\mathsf{STA}(*, 2^{2^{O(mn)}}, m)$.

Proof The algorithm first applies the algorithm α and then applies the algorithm from Theorem 6. Consider a BAPA formula of size n with m quantifier alternations. By Lemma 7, applying α takes $2^{O(n)}$ steps and produces a formula of size $2^{O(n)}$ with m quantifier alternations. Theorem 6 then allows deciding such a formula in $\mathsf{STA}(*, 2^{2^{(2^{O(n)})^{O(m)}}}, m)$, which is $\mathsf{STA}(*, 2^{2^{O(mn)}}, m)$. □

Summarizing Lemma 5 and Lemma 8, taking into account that $2^{O(mn)}$ is in $2^{n^{O(1)}}$ for $m \leq n$, we establish the desired theorem.

Theorem 9 *The validity of* BAPA *formulas is* $\mathsf{STA}(*, 2^{2^{n^{O(1)}}}, n)$*-complete.*

5.4 Deciding BA as a Special Case of BAPA

We next analyze the result of applying the algorithm α to a pure BA sentence F_0. By a pure BA sentence we mean a BA sentence without cardinality constraints, containing only the standard operations $\cap, \cup, ^c$ and the relations $\subseteq, =$. At first, it might seem that the algorithm α is not a reasonable approach to deciding BA formulas, given that the complexity of PA is worse than the corresponding of BA. However, we show that the sentences $\mathsf{PA}_{\mathsf{BA}} = \{\alpha(F_0) \mid F_0 \text{ is in } \mathsf{BA}\}$ generated by applying α to pure BA sentences can be decided using resources optimal for BA [25]. The key observation is that if a PA formula is in $\mathsf{PA}_{\mathsf{BA}}$, then we can use the bounds for quantified variables that are smaller than the bounds in Fact 2.

Let F_0 be a pure BA formula, and let S be the number of set variables in F_0 (the set variables are the only variables in F_0). Let $l_1, \ldots, l_q$ be the free variables of the formula $G_r(l_1, \ldots, l_q)$ in the algorithm α. Then $q = 2^e$ for $e = S + 1 - r$. Let $w_1, \ldots, w_q$ be integers specifying the values of $l_1, \ldots, l_q$. We then have the following lemma.

Lemma 10 For each r where $1 \leq r \leq S$, formula $G_r(w_1, \ldots, w_q)$ is equivalent to formula $G_r(\bar{w}_1, \ldots, \bar{w}_q)$, where $\bar{w}_i = \min(w_i, 2^{r-1})$.

Proof We prove the claim by induction. For $r = 1$, observe that the translation of a quantifier-free part of the pure BA formula yields a PA formula F_1 whose all atomic formulas are of the form $l_{i_1} + \ldots + l_{i_k} = 0$, which are equivalent to $\bigvee_{j=1}^{k} l_{i_j} = 0$. Therefore, the truth value of F_1 depends only on whether the integer variables are zero or nonzero, which means that we may restrict the variables to interval [0, 1].

For the inductive step, consider the elimination of a set variable, and assume that the property holds for G_r and for all q-tuples of nonnegative integers $w_1, \ldots, w_q$. Let $q' = q/2$ and $w'_1, \ldots, w'_{q'}$ be a tuple of nonnegative integers. We show that $G_{r+1}(w'_1, \ldots, w'_{q'})$ is equivalent to $G_{r+1}(\bar{w}'_1, \ldots, \bar{w}'_{q'})$. We consider the case when G_{r+1} is obtained by eliminating an existential set quantifier, so

$$G_{r+1} \equiv \exists^{+} l_1 \ldots l_q.\ \bigwedge_{i=1}^{q'} l'_i = l_{2i-1} + l_{2i} \ \wedge\ G_r.$$

The case for the universal quantifier can be obtained from the proof for the existential one by expressing $\forall$ as $\neg\exists\neg$. We show both directions of the equivalence.

Suppose first that $G_{r+1}(\bar{w}'_1, \ldots, \bar{w}'_{q'})$ holds. Then for each $\bar{w}'_i$ there are u_{2i-1} and u_{2i} such that $\bar{w}'_i = u_{2i-1} + u_{2i}$ and $G_r(u_1, \ldots, u_q)$. We define the witnesses $w_1, \ldots, w_q$ for existential quantifiers as follows. If $w'_i \leq 2^r$, then $\bar{w}'_i = w'_i$, so we use the same witnesses, letting $w_{2i-1} = u_{2i-1}$ and $w_{2i} = u_{2i}$. Let $w'_i > 2^r$, then $\bar{w}'_i = 2^r$. Because $u_{2i-1} + u_{2i} = 2^r$, we have $u_{2i-1} \geq 2^{r-1}$ or $u_{2i} \geq 2^{r-1}$. Suppose, without loss of generality, $u_{2i} \geq 2^{r-1}$. Then let $w_{2i-1} = u_{2i-1}$, and let $w_{2i} = w'_i - u_{2i-1}$. Because $w_{2i} \geq 2^{r-1}$ and $u_{2i} \geq 2^{r-1}$, by the induction hypothesis we have

$$G_r(\ldots, w_{2i}, \ldots) \iff G_r(\ldots, u_{2i}, \ldots) \iff G_r(\ldots, 2^{r-1}, \ldots).$$

For $w_1, \ldots, w_q$ chosen as above, we therefore have $w'_i = w_{2i-1} + w_{2i}$ and $G_r(w_1, \ldots, w_q)$, which by definition of G_{r+1} means that $G_{r+1}(w'_1, \ldots, w'_{q'})$ holds.

Conversely, suppose that $G_{r+1}(w'_1, \ldots, w'_{q'})$ holds. Then there are $w_1, \ldots, w_q$ such that $G_r(w_1, \ldots, w_q)$ and $w'_i = w_{2i-1} + w_{2i}$. If $w_{2i-1} \leq 2^{r-1}$ and $w_{2i} \leq 2^{r-1}$, then $w'_i \leq 2^r$, so let $u_{2i-1} = w_{2i-1}$ and $u_{2i} = w_{2i}$. If $w_{2i-1} > 2^{r-1}$ and $w_{2i} > 2^{r-1}$, then let $u_{2i-1} = 2^{r-1}$ and $u_{2i} = 2^{r-1}$; we have $u_{2i-1} + u_{2i} = 2^r = \bar{w}'_i$ because $w'_i > 2^r$. If $w_{2i-1} > 2^{r-1}$ and $w_{2i} \leq 2^{r-1}$, then let $u_{2i-1} = 2^r - w_{2i}$ and $u_{2i} = w_{2i}$. By the induction hypothesis we have $G_r(u_1, \ldots, u_q) \iff G_r(w_1, \ldots, w_q)$. Furthermore, $u_{2i-1} + u_{2i} = \bar{w}'_i$, so $G_{r+1}(\bar{w}'_1, \ldots, \bar{w}'_{q'})$ by definition of G_{r+1}. □

Theorem 11 *The decision problem for* $\mathsf{PA}_{\mathsf{BA}}$ *is in* $\mathsf{STA}(*, 2^{O(n)}, n)$.

Proof Consider a formula F_0 of size n with S variables. Then $\alpha(F_0) = G_{S+1}$. By Lemma 7, $\mathsf{size}(\alpha(F_0))$ is in $2^{O(n)}$ and $\alpha(F_0)$ has at most S quantifier alternations. By Lemma 10, it suffices for the outermost quantified variable of $\alpha(F_0)$ to range over the integer interval $[0, 2^S]$, and the range of subsequent variables is even smaller. Therefore, the value of each of the $2^{O(S)}$ variables is given by $O(S)$ bits. The values of all bound variables in $\alpha(F_0)$ can therefore be guessed in alternating time $2^{O(S)}$ by using S alternations. The truth value of a PA formula for given values of variables

can be evaluated in time polynomial in the size of the formula, so deciding $\alpha(F_0)$ can be done in $\mathsf{STA}(*, 2^{O(n)}, S)$, which is bounded by $\mathsf{STA}(*, 2^{O(n)}, n)$. □

6 Eliminating Individual Variables from a Formula

Section 4 described an algorithm for BAPA that reduces all set quantifiers in a BAPA sentence to integer quantifiers. The advantage of such an algorithm is that it can be applied to combinations of BA with extensions of PA that need not support quantifier elimination (see Section 8.2). Moreover, this version of the algorithm made the complexity analysis in Section 5 easier. As we discussed in Section 3.4, however, there are uses for an algorithm that eliminates a quantified variable from a formula with free variables, yielding an equivalent quantifier-free formula. In this section we explain that the individual step α_1 of the algorithm α can be used for this purpose.

Quantifier elimination based on our algorithm is possible because PA itself admits quantifier elimination. Therefore, after transforming a set quantifier into an integer quantifier, we can remove the resulting integer quantifier and substitute back the variables constrained by $l_i = |s_i|$. Denote this elimination of integer quantifiers by PAelim. Then the algorithm α' given by

$$\mathsf{separate}\ ;\ \alpha_1\ ;\ \mathsf{PAelim}$$

eliminates one set or integer quantifier from a BAPA formula $Qv.F$, even if F contains free variables (see also [30] for details). Lemma 2 again implies the correctness of this approach.

Example We illustrate the algorithm α' on the example in Figure 10. We use the naming convention given by the formula H for cardinalities of Venn regions from Figure 13. After applying separate, we obtain the formula

$$\begin{array}{l}\exists e.\, \exists^{+} l_{000}, \ldots, l_{111}.\ H \wedge \\ \quad l_{111} + l_{011} + l_{101} + l_{001} = 1 \wedge \\ \quad l_{100} = 0 \ \wedge\ l_{011} + l_{001} + l_{010} = 0.\end{array}$$

After applying α_1, we obtain the formula

$$\begin{array}{r}\exists l_{00}, l_{01}, l_{10}, l_{11}.\, l_{00} = |\mathsf{content}'^{c} \cap \mathsf{content}^{c}| \ \wedge\ l_{01} = |\mathsf{content}'^{c} \cap \mathsf{content}| \wedge \\ l_{10} = |\mathsf{content}' \cap \mathsf{content}^{c}| \wedge\ l_{11} = |\mathsf{content}' \cap \mathsf{content}| \ \wedge\ G,\end{array}$$

where G is the PA formula

$$\begin{array}{r}\exists^{+} l_{000}, \ldots, l_{111}.\ l_{00} = l_{000} + l_{001} \ \wedge l_{01} = l_{010} + l_{011} \wedge \\ l_{10} = l_{100} + l_{101} \ \wedge l_{11} = l_{110} + l_{111} \wedge \\ l_{111} + l_{011} + l_{101} + l_{001} = 1 \wedge \\ l_{100} = 0 \ \wedge\ l_{011} + l_{001} + l_{010} = 0.\end{array}$$

Applying quantifier elimination for PA and simplifications of the quantifier-free formula, we reduce G to

$$l_{01} = 0 \ \wedge\ l_{10} \leq 1 \ \wedge\ l_{11} + l_{10} \geq 1.$$

After substituting back the definitions of $l_{00}, \ldots, l_{11}$, we obtain

$$|\mathsf{content}'^c \cap \mathsf{content}| = 0 \ \wedge \ |\mathsf{content}' \cap \mathsf{content}^c| \leq 1 \wedge$$
$$|\mathsf{content}' \cap \mathsf{content}| + |\mathsf{content}' \cap \mathsf{content}^c| \geq 1.$$

which can indeed be simplified to the result in Figure 10.

6.1 Reducing the Number of Integer Variables

The approaches for deciding BAPA described so far always introduce 2^S integer variables, where S is the number of set variables in the formula. We next describe observations that, although not an improvement in the worst case, may be helpful for certain classes of formulas.

First, as pointed out in [42], if the input formula entails any BA identities (which can be represented as $|b| = 0$), then the number of nonempty Venn regions decreases, which reduces the number of integer variables in the resulting PA formula, and eliminates such atomic formulas b from further considerations.

Second, when eliminating a particular variable y, we can avoid considering Venn regions with respect to all variables of the formula and consider only those variables x for which there exists an expression $b_i(x, y)$ where both x and y occur. This strategy requires using a version $\mathsf{separate}_0$ of separation that introduces integer variables only for those terms $|b|$ that occur in the BAPA formula that results from reducing any remaining BA atomic formulas to the form $|b| = 0$. Like the form (1) obtained by separate in Section 4, the result of $\mathsf{separate}_0$ contains a conjunction of formulas $|b_i| = l_i$ with a PA formula, with two important differences: (1) in the case of $\mathsf{separate}_0$ the resulting formula is polynomial in the original size, and (2) b_i are arbitrary BA expressions as opposed to Venn regions. Let $a_1(y), \ldots, a_q(y)$ be those terms b_i that contain y, and let $x_1, \ldots, x_{S_1}$ be the free variables in $a_1(y), \ldots, a_q(y)$. When eliminating quantifier Qy, it suffices to introduce 2^{S_1} integer variables corresponding to the partitions with respect to $x_1, \ldots, x_{S_1}$, which may be an improvement because $S_1 \leq S$.

The final observation is useful if the number q of terms $a_1(y), \ldots, a_q(y)$ satisfies the property $2q < S_1$, that is, there are a large number of variables but a small number of BA terms containing them. In this case, consider all Boolean combinations $t_1, \ldots, t_u$ of the $2q$ expressions $a_1(0), a_1(1), a_2(0), a_2(1), \ldots, a_q(0), a_q(1)$. For each a_i, we have

$$a_i(y) = (y \cap a_i(0)) \cup (y^c \cap a_i(1)).$$

Each $a_i(0)$ as well as each $a_i(1)$ is a disjoint union of cubes over the BA expressions $t_1, \ldots, t_u$, so each $a_i(y)$ is a disjoint union of cubes over the BA expressions y, $t_1, \ldots, t_u$. It therefore suffices to introduce 2^{2q} integer variables denoting all terms of the form $y \cap t_i$ and $y^c \cap t_i$, as opposed to 2^{S_1} integer variables.

7 Experience Using Our Decision Procedure for BAPA

Our initial experience with BAPA is in the context of the Jahob system [32] for verifying data structure consistency of Java programs. Jahob parses Java source code

annotated with formulas in Isabelle syntax written in comments, generates verification conditions, and uses decision procedures and theorem provers to discharge these verification conditions. Jahob contains interfaces to the interactive theorem provers, first-order theorem provers, and Nelson–Oppen style theorem provers, as well as the Omega Calculator [47] and the LASH [6] decision procedures for PA.

Using Jahob, we generated verification conditions for several Java program fragments that require reasoning about sets and their cardinalities, for example, to prove the equality between the set representing the number of elements in a list and the integer field size after they have been updated, as well as several other examples from Section 3. We found that the existing automated techniques were able to deal with some of the formulas involving only sets or only integers, but not with the formulas that relate cardinalities of operations on sets to the cardinalities of the individual sets. These formulas can be proved in Isabelle but require user interaction in terms of auxiliary lemmas. On the other hand, our implementation of the decision procedure automatically discharges these formulas.

Our initial experience indicates that a straightforward implementation of the algorithm α works fast as long as the number of set variables is small; typical timings are fractions of a second for four or fewer set variables and less than 10 s for five variables. More than five set variables cause the PA decision procedure to run out of memory. (We used the Omega Calculator to decide PA formulas because we found that it outperforms LASH on the formulas generated from our examples.) On the other hand, the decision procedure is much less sensitive to the number of integer variables in BAPA formulas, because they translate into the same number of integer variables in the generated PA formula.

Our current implementation makes use of certain formula transformations to reduce the size of the generated PA formula. We found that eliminating set variables by substitution of equals for equals is an effective optimization. We also observed that lifting quantifiers to the top level noticeably improves the performance of the Omega Calculator. These transformations extend the range of formulas that the current implementation can handle. A possible alternative to the current approach is to interleave the elimination of integer variables with the elimination of the set variables and to perform formula simplifications during this process. Once we obtain a formula with only existential quantifiers, we can decide its satisfiability more efficiently using the recent improvements for quantifier-free formulas (Section 8.3).

8 Further Observations

We next sketch some further observations about BAPA.

8.1 BAPA of Countably Infinite Sets

Our results also extend to the generalization of BAPA where set variables range over subsets of an arbitrary (not necessarily finite) set, which follows from the decidability of the first-order theory of the addition of cardinals [18, Page 78]; see [64, Appendix A] for the complexity of the quantifier-free case. For simplicity of formula semantics, we here consider only the case of all subsets of a countable set, and we argue that the complexity results we have developed above for the finite sets still apply. We

first generalize the language of BAPA and the interpretation of BAPA operations, as follows. Introduce a function $\mathsf{inf}(b)$ that returns 0 if b is a finite set and 1 if b is an infinite set. Define $|b|$ to be some arbitrary integer (for concreteness, 0) if b is infinite, and the cardinality of b if b is finite. A countable or a finite cardinal can therefore be represented in PA by using a pair (k, i) of an integer k and an infinity flag i, where we put $k = 0$ when $i = 1$. The relation representing the addition of cardinals $(k_1, i_1) + (k_2, i_2) = (k_3, i_3)$ is then definable by the formula

$$(i_1 = 0 \wedge i_2 = 0 \wedge i_3 = 0 \wedge k_1 + k_2 = k_3) \ \vee \ ((i_1 = 1 \vee i_2 = 1) \wedge i_3 = 1 \wedge k_3 = 0).$$

Note that $\mathsf{inf}(x) = \max(\mathsf{inf}(x \cap y), \ \mathsf{inf}(x \cap y^c))$, which allows the reduction of all occurrences of $\mathsf{inf}(b)$ expressions to occurrences where b is a Venn region. Moreover, we have the following generalization of Lemma 1.

Lemma 12 Let $b_1, \ldots, b_n$ be disjoint sets, $l_1, \ldots, l_n, k_1, \ldots, k_n$ be natural numbers, and $p_1, \ldots, p_n, q_1, \ldots, q_n \in \{0, 1\}$ for $1 \leq i \leq n$. Then the following two statements are equivalent:

1. There exists a set y such that

$$\bigwedge_{i=1}^{n} |b_i \cap y| = k_i \wedge \mathsf{inf}(b_i \cap y) = p_i \ \wedge \ |b_i \cap y^c| = l_i \wedge \mathsf{inf}(b_i \cap y^c) = q_i.$$

2.

$$\bigwedge_{i=1}^{n} \left(\begin{array}{l} (p_i = 0 \wedge q_i = 0 \Rightarrow |b_i| = k_i + l_i) \wedge \\ (\mathsf{inf}(b_i) = 0 \Leftrightarrow (p_i = 0 \wedge q_i = 0)) \wedge \\ (p_i = 1 \Rightarrow k_i = 0) \ \wedge \ (q_i = 1 \Rightarrow l_i = 0) \end{array} \right).$$

Proof The case when $p_i = 0$ and $q_i = 0$ follows as in the proof of Lemma 1. When $p_i = 1$ or $q_i = 1$, then b_i contains an infinite set as a subset, so it must be infinite. Conversely, an infinite set can always be split into a set of the desired size and another infinite set or into two infinite sets. □

The BAPA decision procedure for the case of countable set then uses Lemma 12 and generalizes the algorithm α in a natural way. The resulting PA formulas are at most polynomially larger than for the case of finite sets, so we obtain a generalization of Theorem 9 to subsets of a countable set.

8.2 BAPA and MSOL over Strings

The weak monadic second-order logic (MSOL) over strings is a decidable logic [21, 54] that can encode Presburger arithmetic by encoding addition using one successor symbol and quantification over sets of elements. There are two important differences between MSOL over strings and BAPA: (1) BAPA can express relationships of the form $|A| = k$ where A is a set variable and k is an integer variable; such a relation is not definable in MSOL over strings; (2) when MSOL over strings is used to represent PA operations, the sets contain binary integer digits, whereas in BAPA the sets contain uninterpreted elements. Note also that MSOL extended with a construct that takes a set of elements and returns an encoding of the size of that set is undecidable, because it could express MSOL with equicardinality, which is undecidable by a reduction

from the Post correspondence problem. Despite this difference, the algorithm α gives a way to combine MSOL over strings with BA yielding a decidable theory. Namely, α does not impose any upper bound on the complexity of the theory for reasoning about integers. Therefore, α can decide an extension of BAPA where the constraints on cardinalities of sets are expressed using relations on integers definable in MSOL over strings; these relations go beyond PA [55, Page 400], [9].

8.3 Quantifier-free Fragment of BAPA

The consequence of the quantifier elimination property of BAPA is that the formulas in the quantifier-free fragment of BAPA, denoted QFBAPA, define the same class of relations as BAPA formulas (even though QFBAPA formulas may be substantially larger than the corresponding quantified BAPA formulas). It is therefore interesting to consider the complexity of the satisfiability problem for QFBAPA. Existing algorithms for QFBAPA [42, 64] run in nondeterministic exponential time. This is the same time bound obtained by running our algorithm α to produce exponentially larger quantifier-free PA formula, guessing the truth values of conjuncts and then using a nondeterministic polynomial time algorithm to solve the resulting integer linear programming problem [44]. (Quantifier-free PA formulas can be solved by using implementations such as CVC Lite [4] and UCLID [33].) Therefore, the worst-case complexity of α is no worse than that of previously known algorithms.

Is it possible to decide QFBAPA formulas in less than nondeterministic exponential time? Recently, as part of an ongoing work, the first author of the present paper has shown that QFBAPA is solvable in polynomial space. A version of this result is described in [37, Section 3]. The result proves the following small model property implying that it suffices to consider cardinalities of sets that are at most exponential in formula size.

Theorem 13 *Let F be a* QFBAPA *formula of size n. Then F has a solution if and only if it has a solution where all sets are subsets of a set of cardinality $2^{n^{O(1)}}$ and all integer variables are bounded by $2^{n^{O(1)}}$.*

Theorem 13 allows us to reduce, in nondeterministic polynomial time, the satisfiability of any QFBAPA formula to the satisfiability of the conjunction of equations of the form

$$|b_1| = K_1 \ \wedge \ \ldots \ \wedge \ |b_n| = K_n,$$

where $b_1, \ldots, b_n$ are Boolean algebra expressions, and $K_1, \ldots, K_n$ are binary-encoded nonnegative constants. We call such constraint CBAC constraints (Conjunction of Boolean Algebra expressions with Cardinalities). In [37, Section 3] we present a polynomial-space algorithm that checks the satisfiability of CBAC constraints without ever simultaneously introducing integer variables for all regions in the Venn diagram. This yields a polynomial-space algorithm for QFBAPA. A polynomial space algorithm means the ability to use QBF solvers [11, 65] to decide QFBAPA and can be a significant improvement over nondeterministic exponential time algorithm. However, there is no indication that QFBAPA is PSPACE-hard or even coNP-hard: our preliminary experiments on studying the structure of solutions

of randomly generated CBAC constraints suggest that satisfiable constraints have solutions where only a small number q of Venn regions are nonempty. This result suggests an algorithm that selects a subset of q Venn region cardinalities and generates an integer linear programming problem where the remaining $2^n - q$ variables are assumed to be 0. A computable polynomial upper bound on q would yield an NP algorithm for QFBAPA.

9 Related Work

Our paper is the first result that shows a complexity bound for the first-order theory of BAPA. A preliminary version of this paper appears in [27, 30]. The decidability for BAPA, presented as BA with equicardinality constraints, was shown in [18]; see Section 2. A decision procedure for a special case of BAPA was presented in [63], which allows quantification only over *elements* but not over *sets* of elements. Ohlbach and Koehler [42] also examine quantifier-free formulas and show how to combine quantifier-free BA constraints with additional constraints using 'bridging functions.' Bridging functions satisfy homomorphism-like properties and generalize the cardinality operator; we expect that the quantifier-elimination techniques of our paper can be generalized in this direction as well. Revesz (2004) presents quantifier elimination and the decidability of a single-sorted version of BAPA that contains only the set sort. Note that bound integer variables can be simulated by using bound set variables, but there are notational and clear efficiency reasons to allow integer variables in BAPA.

Presburger Arithmetic. The original result on decidability of PA is [46]. The space bound for PA was shown in [19]. The matching lower and upper bounds for PA were shown in [5]; see also [25, Lecture 24]. An analysis parameterized by the number of quantifier alternations is presented in [48]. Our implementation uses a quantifier-elimination-based Omega test [47]. Among the decision procedures for full PA, that of [13] is the only proof-generating version, and is based on [16]. Decidable fragments of arithmetic that go beyond PA are described in [8, 9].

Reasoning about Sets. The first results on decidability of BA of sets are from [36], [1, Chapter 4] and use quantifier elimination, from which one can derive a small model property. Kozen (1980) gives the complexity of the satisfiability problem for arbitrary BA. Martin and Nipkow (1989) study unification in Boolean rings. The quantifier-free fragment of BA is shown NP-complete in [38]; see [31] for a generalization of this result using the parameterized complexity of the Bernays–Schönfinkel–Ramsey class of first-order logic [7, Page 258]. Cantone et al. give an overview of several fragments of set theory including theories with quantifiers but no cardinality constraints and theories with cardinality constraints but no quantification over sets. The decision procedure for quantifier-free fragment with cardinalities in [12, Chapter 11] introduces exponentially many integer variables to reduce the problem to PA. Among the systems for interactively reasoning about richer theories of sets are Isabelle [41], HOL [20], PVS [43]. First-order frameworks such as Athena [2] can use axiomatizations of sets along with calls to resolution-based theorem provers [58, 59] to reason about sets.

Combinations of Decidable Theories. The techniques for combining *quantifier-free* theories [40, 50] and their generalizations such as [56, 57, 61, 62, 64] are of great

importance for program verification. The present paper focuses on quantified formulas, which add additional expressive power in writing concise specifications. Among the general results for quantified formulas are the Feferman–Vaught theorem for products [18] and term powers [28, 29]. Description logics [3] also support sets with cardinalities as well as relations but do not support quantification over sets. While we have found quantifier elimination to be useful, many problems can be encoded in quantifier-free formulas, which motivates the ongoing work in Section 8.3.

Analyses of Linked Data Structures. In addition to the new technical results, one of the contributions of our paper is to identify the uses of our decision procedure for verifying data structure consistency. We have shown how BAPA enables the verification tools to reason about sets and their sizes. This capability is particularly important for analyses that handle dynamically allocated data structures where the number of objects is statically unbounded [39, 52, 60]. Recently, these approaches were extended to handle the combinations of the constraints representing data structure contents and constraints representing numerical properties of data structures [15, 51]. Our result provides a systematic mechanism for building precise and predictable versions of such analyses. Among other constraints used for data structure analysis, BAPA is unique in being a complete algorithm for an expressive theory that supports arbitrary quantifiers.

10 Conclusion

Motivated by static analysis and verification of relations between data structure content and size, we have presented an algorithm for deciding the first-order theory of Boolean algebra with Presburger arithmetic (BAPA); established the precise complexity of BAPA, showing that it is identical to the complexity of PA; implemented the algorithm; and applied it to discharge verification conditions. Our experience indicates that the algorithm is useful as a component of a data structure verification system. Furthermore, we expect that the future study of quantifier-free fragments of BAPA and its variations with noninteger measures of sets might be of broader interest in the context of constraint solving and complexity.

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