Canonical Logic Programs are Succinctly Incomparable with Propositional Formulas

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Abstract

Canonical (logic) programs (CP) refer to the class of normal programs (LP) augmented with connective not not, and are equally expressive as propositional formulas (PF). In this paper we address the question of whether CP and PF are succinctly incomparable. Our main result shows that the PAR-ITY problem only has exponential CP representations, while it can be polynomially represented in PF. In other words, PARITY separates PF from CP. Simply speaking, this means that exponential size blowup is generally inevitable when translating a set of PF formulas into a (logically) equivalent CP program (without introducing new variables). Furthermore, since it has been shown by Lifschitz and Razborov that there is also a problem which separates CP from PF (assuming P \nsubseteq NC¹/poly), it follows that the two formalisms are indeed succinctly incomparable.

1 Introduction

The relationship between (*logic*) programs under answer set semantics (ASP) (Lifschitz 2008; Brewka, Eiter, and Truszczynski 2011) and propositional satisfiability (SAT) (Biere et al. 2009) gains a lot of attention in the literature. In 2006, Lifschitz and Razborov proved that an exponential size blowup is generally inevitable when translating a normal program (LP) to a (logically) equivalent set of propositional formulas (PF) (without introducing additional variables). More precisely, they showed that (a variant of) the Pcomplete problem PathSystem (PATH) has polynomial size LP representations, however, it cannot be polynomially represented in PF (assuming P \nsubseteq NC¹/poly) (Lifschitz and Razborov 2006), i.e., PATH separates LP from PF.

As noted in (Lifschitz and Razborov 2006), PF can be considered as a special case of the class of *(nondisjunctive) nested programs* (NLP, without classical negation \neg) (Lifschitz 1999). Therefore, NLP is *stronger* than PF in terms of the *succinctness* criterion (or the *"comparative linguistics" approach*) (Gogic et al. 1995):

That is, we consider formalism A to be stronger than formalism B if and only if any knowledge base (KB) in B has an equivalent KB in A that is only polynomially longer, while there is a KB in A that can be translated to B only with an exponential blowup.

So the following footnote in (Lifschitz 2008) seems convincing:

...ASP appears to be stronger than SAT in the sense of the "comparative linguistics" approach to knowledge representation...

However, ASP involves many kinds of programs, the above statement probably needs further clarification. Particularly, the class of so-called (*nondisjunctive*) canonical programs (CP¹) (Lifschitz 1999; Lee 2005; Lee, Lifschitz, and Yang 2013), is a minimal form of ASP that is equally expressive as PF, but looks more likely *not* succinctly stronger. So a question naturally arises: *Does there exist a problem that separates PF from CP*? If there is such a problem, then CP and PF are *succinctly incomparable* (assuming $P \not\subseteq NC^1/poly$).

In this paper we address the question and give a *positive* answer. Our main result shows that the PARITY problem separates PF from CP. Simply speaking, this means exponential size blowup is generally inevitable when translating a set of PF formulas into an equivalent CP program. The PARITY problem asks whether a binary string contains an odd number of 1's, and it is well-known that (i) PARITY \in NC¹/poly², i.e., it has polynomial PF representations; (ii) PARITY \notin AC⁰, i.e., it cannot be represented by *polynomial* size (boolean) circuits with *constant depth* and *unbounded fan-in* (Arora and Barak 2009; Jukna 2012).

To show PARITY separates PF from CP, we provide a procedure that simplifies every PARITY program II into a shorter program II' s.t. II' is equivalent to its *completion* Comp(II') (Erdem and Lifschitz 2003). Such completions are essentially constant depth, unbounded fan-in circuits. According to PARITY \notin AC⁰, these circuits must be of exponential size, consequently, there are no polynomial size PARITY programs in CP. In the rest of the paper, we shall introduce the basic concepts, key steps of our proof and discuss the importance of succinctness research in the theory and practice of Knowledge Representation (KR).

 ${}^{2}NC^{1}$ /poly (or *non-uniform* NC¹) *exactly* contains languages computable (representable) by polynomial size PF formulas.

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¹Extends LP with connective *not not*.

2 Background

Canonical Programs

The following notations are adopted from (Lifschitz 1999; Lee 2005). A *rule element e* is defined as

$$e := \top \mid \perp \mid x \mid not \mid x \mid not not x$$

in which \top, \bot are 0-ary connectives, x is a (boolean) variable (or an *atom*) and *not* is a unary connective³. A (nondisjunctive canonical) rule is an expression of the form

$$H \leftarrow B$$
 (1)

where the *head* H is either a variable or the connective \perp , and the *body* B is a finite set of rule elements. A *canonical program* Π is a finite set of rules. E.g., the following is a canonical program:

$$\begin{array}{ll} x_1 \leftarrow not \ not \ x_1, & x_3 \leftarrow not \ x_1, not \ x_2, \\ x_2 \leftarrow not \ not \ x_2, & x_3 \leftarrow x_1, \ x_2. \end{array}$$
 (2)

A canonical program Π is *normal* if it contains no connectives *not not*. A normal program Π is *basic* if it contains no occurrences of connective *not*.

The *satisfaction relation* \models between a set of variables *I* and a rule element is defined as follows:

- $I \models \top$ and $I \nvDash \bot$,
- $I \models x$ iff $I \models not not x$ iff $x \in I$,
- $I \models not x \text{ iff } x \notin I.$

Say I satisfies a set of rule elements B if I satisfies each rule element in B. We say I is *closed* under a program II, if I is closed under every rule in II, i.e., for each rule $H \leftarrow B \in \Pi$, $I \models H$ whenever $I \models B$. Let II be a basic program and let $Cn(\Pi)$ denotes the *minimal* set (in terms of set inclusion) closed under II, we say I is an *answer set* of II if $I = Cn(\Pi)$. Note that a basic program has exactly one answer set.

The reduct Π^I of a program Π w.r.t. I is a set of rules obtained from Π via: (i) Replacing each *not* not x with \top if $I \models x$, and with \bot otherwise; (ii) Replacing each not xwith \top if $I \nvDash x$, and with \bot otherwise. Observe that Π^I must be a basic program. We say I is an answer set of Π if $I = Cn(\Pi^I)$, i.e., I is an answer set of Π^I . E.g., the following single rule program:

$$x \leftarrow not \ not \ x$$
 (3)

has two answer sets \emptyset and $\{x\}$.

For a set of rule elements B, define $var(B) = \{e \in B : e \text{ is a variable}\}$. E.g., $var(\{x_1, not \ x_2, not \ not \ x_3\}) = \{x_1\}$. The signature $sig(\Pi)$ of a program Π is the set of all involved variables in Π . By $Ans(\Pi)$ we denote the set of all answer sets of Π . E.g., if Π is (2), then $sig(\Pi) = \{x_1, x_2, x_3\}$ and $Ans(\Pi) = \{\{x_1, x_2, x_3\}, \{x_1\}, \{x_2\}, \{x_3\}\}$. The size $|\Pi|$ of Π is the number of its rules. As a convention, Π_n refers to a program with signature $\{x_1, \ldots, x_n\}$, i.e., $sig(\Pi_n) = \{x_1, \ldots, x_n\}$.

It is easy to see that by using rules of the form (3) and appropriate *constraints* of the form $\perp \leftarrow B$, it is easy to give an arbitrary set of answer sets over $sig(\Pi_n)$, in other words, CP has exactly the same expressive power as PF.

Problem Representation and Succinctness

A (binary) string is a finite sequence of bits from $\{0, 1\}$. A string w of length n (i.e., $w \in \{0, 1\}^n$) defines a subset of variables $\{x_1, \ldots, x_n\}$. E.g., 1010 stands for $\{x_1, x_3\}$. Therefore, a set of variables I and a string w can be regarded as the same. A problem (or language) L is a set of strings.

Definition 2.1 (Problem Representation). A problem L can be represented in a class of programs (or formulas, etc) C(*i.e.*, $L \in C$), if there exists a sequence of programs $\{\Pi_n\}$ (n = 1, 2, ...) in C that computes L, *i.e.*, for every string $w \in \{0, 1\}^n$,

$$w \in L \Leftrightarrow w \in Ans(\Pi_n).$$

Say L has polynomial representations in C (i.e., $L \in Poly-C$), if $L \in C$ and $|\Pi_n|$ is bounded by a polynomial p(n).

Recall that the PARITY problem is the set of binary strings with an odd number of 1's. By PARITY_n we denote PARITY strings of length n. Clearly, (2) represents PARITY₃ since its answer sets are 111, 100, 010 and 001.

Definition 2.2 (Succinctness (Gogic et al. 1995; French et al. 2013)). Let C, C' be two classes of programs such that for every problem $L, L \in C \Leftrightarrow L \in C'$. Say C is at least as succinct as C' (i.e., $C' \preceq C$), if for every problem L,

$$L \in Poly-\mathcal{C}' \Rightarrow L \in Poly-\mathcal{C}.$$

If $L \in Poly-C$ but $L \notin Poly-C'$ (i.e., $C \not\preceq C'$), then L separates C from C'. If $C' \preceq C$ and $C \not\preceq C'$, then C is strictly more succinct than C' (i.e., $C' \prec C$). Moreover, C, C' are succinctly incomparable if there is a problem L that separates C from C', and vice versa (i.e., $C \not\preceq C'$ and $C' \not\preceq C$).

PARITY Representations and Completion

Note that (2) suggests a "pattern" for PARITY_n: The first part of the program (e.g., the "*not not*" rules in (2)) generates all possible strings of length n - 1, the second part identifies the last bit to form an odd string of length n. Following the pattern, we can give a sequence of *exponential* size CP programs for PARITY, i.e., PARITY \in CP.

On the other hand, it is a textbook result that PARITY has polynomial size PF representations with classical connectives $\{\wedge, \lor, \neg\}$, i.e., PARITY $\in \mathbb{NC}^1$ /poly (or Poly-PF). Remind that the size of a formula is the number of its connectives. E.g., $(x_1 \land \neg x_2) \lor (\neg x_1 \land x_2)$ represents PARITY₂, and the formula can be recursively applied for arbitrary PARITY_n. It is also well-known that PARITY $\notin \mathbb{AC}^0$, i.e., PARITY cannot be represented by a sequence of polynomial size circuits $\{C_n\}$ in which these circuits C_n have unbounded fan-in and constant depth.

Simply speaking, a sequence of polynomial size PF formulas $\{\phi_n\}$ represents an AC⁰ language, if the number of variables occur in a *conjunction* or *disjunction* of ϕ_n is *unrestricted*, and the distance of the longest path from the root to a leaf in the tree structure of each ϕ_n is *fixed*. E.g., PAR-ITY cannot be represented by a sequence of polynomial size CNFs $\{\psi_n\}$ since $\{\psi_n\}$ represents an AC⁰ language. For more details about circuits, please see (Arora and Barak 2009; Jukna 2012).

³By (Lifschitz 1999), not not not x can be replaced by not x.

The completion $Comp(\Pi)$ (Erdem and Lifschitz 2003) of a CP program Π consists of a set (or a conjunction) of PF formulas (we slightly abuse the connective \equiv): (i) $x \equiv \tilde{B}_1 \lor \tilde{B}_2 \lor \cdots \lor \tilde{B}_m$, where $x \leftarrow B_1, \ldots, x \leftarrow B_m$ are all rules in Π with head x, and each \tilde{B}_i is the conjunction of rule elements in B_i with connective *not* replaced by \neg ; (ii) $x \equiv \bot$, if x is not a head of any rule in Π ; (iii) $\neg \tilde{B}$, if a rule $\bot \leftarrow B$ is in Π .

Proposition 2.1. The completion $Comp(\Pi)$ of an arbitrary canonical program Π is a constant depth, unbounded fan-in circuit whose size is polynomially bounded by $|\Pi|$.

It is well-known that an answer set of Π is also a model of its completion, but the inverse generally does not hold. E.g., the completion $\{x_1 \equiv \neg x_2 \lor (x_2 \land x_1), x_2 \equiv \neg x_1 \lor (x_1 \land x_2)\}$ of the PARITY₂ program:

$$\begin{array}{ll} x_1 \leftarrow not \ x_2, & x_2 \leftarrow not \ x_1, \\ x_1 \leftarrow x_2, \ not \ not \ x_1, & x_2 \leftarrow x_1, \ not \ not \ x_2, \end{array}$$

has an even string model 11, i.e., (4) is not *completion-equivalent*. In the next section, we will show how to simplify a PARITY program to be completion-equivalent.

3 Simplifying PARITY Programs

Let B be a finite set of rule elements built on signature $V = \{x_1, \ldots, x_n\}$. By S(B) we denote the set $\{I \subseteq V : I \models B\}$. E.g., let $V = \{x_1, x_2, x_3, x_4\}$ and $B = \{x_2, not \ x_3, not \ not \ x_4\}$, then S(B) = $\{\{x_1, x_2, x_4\}, \{x_2, x_4\}\} = \{1101, 0101\}$. We say B is consistent if there is a set of variables I s.t. $I \models B$. We say B covers a variable $x \in V$ iff $x \in B$ or not $x \in B$ or not not $x \in B$. If B covers every variable in V then B fully covers V. E.g., $B = \{x_1, not \ x_2, not \ not \ x_3\}$ fully covers $V = \{x_1, x_2, x_3\}$. Clearly, B is consistent and fully covers V iff S(B) contains a unique string.

In the following we assume that all rules $x \leftarrow B$ have consistent body B and $x \notin var(B)$. It is easy to see that the assumption does not affect the generality.

Lemma 3.1. Let Π_n be a PARITY_n program. If there is a rule $x \leftarrow B$ in Π_n s.t. not not $x \in B$ and S(B) contains a unique even string, then dropping $x \leftarrow B$ from Π_n results in a PARITY_n program Π'_n .

E.g., by Lemma 3.1, (4) can be simplified to:

$$x_1 \leftarrow not \ x_2, \quad x_2 \leftarrow not \ x_1.$$
 (5)

Lemma 3.2. Let Π_n be a PARITY_n program. If there is a rule $x \leftarrow B$ in Π_n s.t. not not $x \in B$ and S(B) contains a unique odd string, then dropping not not x from B results in a PARITY_n program Π'_n .

Standard Programs and the Main Theorem

A PARITY_n program Π_n is *standard* if for each rule $x \leftarrow B \in \Pi_n$, not not $x \notin B$ whenever $S(B \cup \{x\})$ contains a unique string. E.g., (2) and (5) are standard programs. By Lemma 3.1 and 3.2, it is not hard to see:

Proposition 3.1. Let Π_n be a PARITY_n program. Then there is a standard PARITY_n program Π'_n s.t. $|\Pi'_n| \leq |\Pi_n|$.

Moreover, note that (2) and (5) are completion-equivalent PARITY programs. This is guaranteed by the following proposition:

Proposition 3.2. Let Π_n be a standard PARITY_n program. Then Π_n is equivalent to its completion $Comp(\Pi_n)$.

The proof idea of Proposition 3.2 is that every standard PARITY_n program Π_n can be equivalently rewritten to Π'_n by replacing each $x \in var(B)$ with not not x for every rule body B in Π_n . By the Lin-Zhao Theorem (Lin and Zhao 2004) or the (generalized) Fages Theorem (Erdem and Lifschitz 2003; You, Yuan, and Zhang 2003), Π'_n is equivalent to its completion $Comp(\Pi'_n)$. And then the Proposition follows from the fact that $Comp(\Pi'_n) = Comp(\Pi_n)$.

Lemma 3.3 (Main Lemma). Let Π_n be a PARITY_n program. Then there is a PARITY_n program Π'_n s.t. Π'_n is equivalent to $Comp(\Pi'_n)$ and $|\Pi'_n| \leq |\Pi_n|$.

Theorem 3.1 (PARITY \notin Poly-CP). *PARITY has no polynomial size CP representations.*

Proof. Assume the contrary that there is a sequence of PARITY_n programs $\{\Pi_n\}$ s.t. $|\Pi_n|$ is bounded by a polynomial p(n). By Lemma 3.3, there is a sequence of completion-equivalent PARITY_n programs $\{\Pi'_n\}$ in which $|\Pi'_n|$ is also bounded by the polynomial p(n). By Proposition 2.1, $\{\Pi'_n\}$ represents a language in AC⁰. This contradicts PARITY \notin AC⁰.

Corollary 3.1. PARITY separates PF from CP.

Corollary 3.2. Suppose $P \nsubseteq NC^1/poly$. Then CP and PF are succinctly incomparable.

4 Discussion and Conclusion

In this paper we show that the PARITY problem separates PF from CP, and the two formalisms are succinctly incomparable (assuming P $\not\subseteq$ NC¹/poly). Interestingly, our main result may at first appear counter-intuitive: the P-complete problem PATH has Poly-CP representations, while this does not hold for an "easy" problem PARITY. Actually, there is no contradiction. As noted in (Abiteboul, Hull, and Vianu 1995; Dantsin et al. 2001), a *complete* problem in a complexity class can be represented in a formalism *C*, *does not* imply that *all* problems in that class can be represented in *C*.

Generally speaking, the research of succinctness (Gogic et al. 1995; Coste-Marquis et al. 2004; Grohe and Schweikardt 2005; French et al. 2013) gives us a deeper understanding about KR formalisms, for it reveals their (in)abilities of concisely representing different problems under the condition that the encoded models are the *same*. This is particularly interesting if the formalisms are *equally expressive* and share the same reasoning complexity.

E.g., in addition to CP and NLP, programs with *cardinality constraints* and *choice rules* (CC, without classic negation \neg) (Simons, Niemelä, and Soininen 2002), (*simple*) def*inite causal theories*⁴ (S/DT) (Giunchiglia et al. 2004) and

⁴A theory is simple if each rule body is a conjunction of literals.

two-valued programs (TV) (Lifschitz 2012) are as expressive as PF and NP-complete for consistency checking. But they have a non-trivial succinctness picture, see Fig. 1.

Besides the theoretical interests, succinctness also tells us something like "which for what is the best" in choosing KR formalisms for a given application. E.g., one should choose ASP instead of SAT (or DT) if the application involves reasoning about PATH or Transitive Closure⁵, because the former provides compact representations to avoid unnecessary overload. Recall that from the complexity viewpoint, even *one* extra variable may *double* the search space for intractable problems.



In future work we plan to establish the missing connections in Fig. 1, moreover, we will consider the succinctness of more expressive formalisms like programs with predicate symbols or higher-order atoms (Gebser, Schaub, and Thiele 2007), etc.

► B:∃L separates A from B Acknowledgement

Figure 1: Succinctness Pic.

: under assumption

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⁵An NL-*complete* problem. It is believed that NL \nsubseteq NC¹/Poly.