First-Order vs. Second-Order Encodings for LTL\(_f\)-to-Automata Translation

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Abstract. Translating formulas of Linear Temporal Logic (LTL) over finite traces, or LTL\(_f\), to symbolic Deterministic Finite Automata (DFA) plays an important role not only in LTL\(_f\) synthesis, but also in synthesis for Safety LTL formulas. The translation is enabled by using MONA, a powerful tool for symbolic, BDD-based, DFA construction from logic specifications. Recent works used a first-order encoding of LTL\(_f\) formulas to translate LTL\(_f\) to First Order Logic (FOL), which is then fed to MONA to get the symbolic DFA. This encoding was shown to perform well, but other encodings have not been studied. Specifically, the natural question of whether second-order encoding, which has significantly simpler quantificational structure, can outperform first-order encoding remained open.

In this paper we address this challenge and study second-order encodings for LTL\(_f\) formulas. We first introduce a specific MSO encoding that captures the semantics of LTL\(_f\) in a natural way and prove its correctness. We then explore a Compact MSO encoding, which benefits from automata-theoretic minimization, thus suggesting a possible practical advantage. To that end, we propose a formalization of symbolic DFA in second-order logic, thus developing a novel connection between BDDs and MSO. We then show by empirical evaluations that the first-order encoding does perform better than both second-order encodings. The conclusion is that first-order encoding is a better choice than second-order encoding in LTL\(_f\)-to-Automata translation.

1 Introduction

Synthesis from temporal specifications [23] is a fundamental problem in Artificial Intelligence and Computer Science [8]. A popular specification is Linear Temporal Logic (LTL) [24]. The standard approach to solving LTL synthesis requires, however, determinization of automata on infinite words and solving parity games, both challenging algorithmic problems [17]. Thus a major barrier of temporal synthesis has been algorithmic difficulty. One approach to combating this difficulty is to focus on using fragments of LTL, such as the GR(1) fragment, for which temporal synthesis has lower computational complexity [1].

A new logic for temporal synthesis, called LTL\(_f\), was proposed recently in [6,8]. The focus there is not on limiting the syntax of LTL, but on interpreting it semantically on finite traces, rather than infinite traces as in [24]. Such interpretation allows the executions

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being arbitrarily long, but not infinite, and is adequate for finite-horizon planning problems. While limiting the semantics to finite traces does not change the computational complexity of temporal synthesis (2EXPSPACE), the algorithms for LTL$_f$ are much simpler. The reason is that those algorithms require determinization of automata on finite words (rather than infinite words), and solving reachability games (rather than parity games) [8]. Another application, as shown in [30], is that temporal synthesis of Safety LTL formulas, a syntactic fragment of LTL expressing safety properties, can be reduced to reasoning about finite words (see also [18,19]). This approach has been implemented in [31] for LTL$_f$ synthesis and in [30] for synthesis of Safety LTL formulas, and has been shown to outperform existing temporal-synthesis tools such as Acacia+ [2].

The key algorithmic building block in these approaches is a translation of LTL$_f$ to symbolic Deterministic Finite Automata (DFA) [30,31]. In fact, translating LTL$_f$ formula to DFA has other algorithmic applications as well. For example, in dealing with safety properties, which are arguably the most used temporal specifications in real-world systems [18]. As shown in [28], model checking of safety properties can benefit from using deterministic rather than nondeterministic automata. Moreover, in runtime verification for safety properties, we need to generate monitors, a type of which are, in essence, deterministic automata [29]. In [28,29], the translation to deterministic automata is explicit, but symbolic DFAs can be useful also in model checking and monitor generation, because they can be much more compact than explicit DFAs, cf. [31].

The method used in [30,31] for the translation of LTL$_f$ to symbolic DFA used an encoding of LTL$_f$ to First-Order Logic (FOL) that captures directly the semantics of temporal connectives, and MONA [13], a powerful tool, for symbolic DFA construction from logical specifications. This approach was shown to outperform explicit tools such as SPOT [12], but encodings other than the first-order one have not yet been studied. This leads us here to study second-order translations of LTL$_f$, where we use Monadic Second Order (MSO) logic of one successor over finite words (called M2L-STR in [16]). Indeed, one possible advantage of using MSO is the simpler quantificational structure that the second-order encoding requires, which is a sequence of existential monadic second-order quantifiers followed by a single universal first-order quantifier. Moreover, instead of the syntax-driven translation of first-order encoding of LTL$_f$ to FOL, the second-order encoding employs a semantics-driven translation, which allows more space for optimization. The natural question arises whether second-order encoding outperforms first-order encoding.

To answer this question, we study here second-order encodings of LTL$_f$ formulas. We start by introducing a specific second-order encoding called MSO encoding that relies on having a second-order variable for each temporal operator appearing in the LTL$_f$ formula and proving the correctness. Such MSO encoding captures the semantics of LTL$_f$ in a natural way and is linear in the size of the formula. We then introduce a so-called Compact MSO encoding, which captures the tight connection between LTL$_f$ and DFAs. We leverage the fact that while the translation from LTL$_f$ to DFA is doubly exponential [18], there is an exponential translation from Past LTL$_f$ to DFA (a consequence of [5,6]). Given an LTL$_f$ formula $\phi$, we first construct a DFA that accepts exactly the reverse language satisfying $\text{models}(\phi)$ via Past LTL$_f$. We then encode this DFA using second-order logic and “invert” it to get a second-order formulation for
the original LTL formula. Applying this approach directly, however, would yield an MSO formula with an exponential (in terms of the original LTL formula) number of quantified monadic predicates. To get a more compact formulation we can benefit from the fact that the DFA obtained by MONA from the Past LTL formula is symbolic, expressed by binary decision diagrams (BDDs) [14]. We show how we can obtain a Compact MSO encoding directly from these BDDs. In addition, we present in this paper the first evaluation of the spectrum of encodings for LTL-to-automata from first-order to second-order.

To perform an empirical evaluation of the comparison between first-order encoding and second-order encoding of LTL, we first provide a broad investigation of different optimizations of both encodings. Due to the syntax-driven translation of FOL encoding, there is limit potential for optimization such that we are only able to apply different normal forms to LTL formulas, which are Boolean Normal Form (BNF) and Negation Normal Form (NNF). The semantics-driven translation of second-order encoding, however, enables more potential for optimization than the FOL encoding. In particular, we study the following optimizations introduced in [21][22]: in the variable form, where a Lean encoding introduces fewer variables than the standard Full encoding; and in the constraint form, where a Sloppy encoding allows less tight constraints than the standard Fussy encoding. The main result of our empirical evaluations is the superiority of the first-order encoding as a way to get MONA to generate a symbolic DFA, which answers the question of whether second-order outperforms first-order for LTL-to-automata translation.

The paper is organized as follows. In Section 2 we provide preliminaries and notations. Section 3 introduces MSO encoding and proves the correctness. Section 4 describes a more compact second-order encoding, called Compact MSO encoding and proves the correctness. Empirical evaluation results of different encodings and different optimizations are presented in Section 5. Finally, Section 6 offers concluding remarks.

2 Preliminaries

2.1 LTL Basics

Linear Temporal Logic over finite traces (LTLf) has the same syntax as LTL [6]. Given a set P of propositions, the syntax of LTLf formulas is as follows:

\[ \phi ::= \top | \bot | p | \neg \phi | \phi_1 \land \phi_2 | X \phi | \phi_1 U \phi_2 \]

where \( p \in P \). We use \( \top \) and \( \bot \) to denote true and false respectively. X (Next) and U (Until) are temporal operators, whose dual operators are \( \neg \) (Weak Next) and R (Release) respectively, defined as \( N\phi \equiv \neg X \neg \phi \) and \( \phi_1 R \phi_2 \equiv \neg (\neg \phi_1 U \neg \phi_2) \). The abbreviations (Eventually) \( F\phi \equiv \top U \phi \) and (Globally) \( G\phi \equiv \bot R \phi \) are defined as usual. Finally, we have standard boolean abbreviations, such as \( \lor \) (or) and \( \rightarrow \) (implies).

Elements \( p \in P \) are atoms. A literal \( l \) can be an atom or the negation of an atom. A trace \( \rho = \rho[0], \rho[1], \ldots \) is a sequence of propositional assignments, where \( \rho[x] \in 2^P \) \((x \geq 0)\) is the \( x \)-th point of \( \rho \). Intuitively, \( \rho[x] \) is the set of propositions that are true at instant \( x \). Additionally, \( |\rho| \) represents the length of \( \rho \). The trace \( \rho \) is an infinite trace if \( |\rho| = \infty \) and \( \rho \in (2^P)^\omega \); otherwise \( \rho \) is finite, and \( \rho \in (2^P)^* \). LTLf formulas
are interpreted over finite traces. Given a finite trace $\rho$ and an $\text{LTL}_f$ formula $\phi$, we inductively define when $\phi$ is true for $\rho$ at point $x$ ($0 \leq x < |\rho|$), written $\rho, x \models \phi$, as follows:

- $\rho, x \models \top$ and $\rho, x \not\models \bot$;
- $\rho, x \models p$ iff $p \in \rho[x]$;
- $\rho, x \models \neg \phi$ iff $\rho, x \not\models \phi$;
- $\rho, x \models \phi_1 \land \phi_2$, iff $\rho, x \models \phi_1$ and $\rho, x \models \phi_2$;
- $\rho, x \models X\phi$, iff $x + 1 < |\rho|$ and $\rho, x + 1 \models \phi$;
- $\rho, x \models \phi_1 U \phi_2$, iff there exists $y$ such that $x \leq y < |\rho|$ and $\rho, y \models \phi_2$, and for all $z, x \leq z < y$, we have $\rho, z \models \phi_1$.

An $\text{LTL}_f$ formula $\phi$ is true in $\rho$, denoted by $\rho \models \phi$, when $\rho, 0 \models \phi$. Every $\text{LTL}_f$ formula can be written in Boolean Normal Form (BNF) or Negation Normal Form (NNF) \cite{27}. BNF rewrites the input formula using only $\neg, \land, \lor, X,$ and $U$. NNF pushes negations inwards, introducing the dual temporal operators $N$ and $R$, until negation is applied only to atoms.

### 2.2 Symbolic DFA and MONA

We start by defining the concept of symbolic automaton \cite{31}, where a boolean formula is used to represent the transition function of a Deterministic Finite Automaton (DFA). A symbolic deterministic finite automaton (Symbolic DFA) $\mathcal{F} = (\mathcal{P}, \mathcal{X}, X_0, \eta, f)$ corresponding to an explicit DFA $\mathcal{D} = (2^P, S, s_0, \delta, F)$ is defined as follows:

- $\mathcal{P}$ is the set of atoms;
- $\mathcal{X}$ is a set of state variables where $|\mathcal{X}| = \lceil \log_2 |S| \rceil$;
- $X_0 \in 2^\mathcal{X}$ is the initial state corresponding to $s_0$;
- $\eta : 2^\mathcal{X} \times 2^P \rightarrow 2^\mathcal{X}$ is a boolean transition function corresponding to $\delta$;
- $f$ is the acceptance condition expressed as a boolean formula over $\mathcal{X}$ such that $f$ is satisfied by an assignment $\mathcal{X}$ iff $\mathcal{X}$ corresponds to a final state $s \in F$.

We can represent the symbolic transition function $\eta$ by an indexed family $\eta_x : 2^\mathcal{X} \times 2^P \rightarrow \{0, 1\}$ for $x \in \mathcal{X}$, which means that $\eta_x$ can be represented by a binary decision diagram (BDD) \cite{14} over $\mathcal{X} \cup \mathcal{P}$. Therefore, the symbolic DFA can be represented by a sequence of BDDs, each of which corresponding to a state variable.

The MONA tool \cite{13} is an efficient implementation for translating FOL and MSO formulas over finite words into minimized symbolic deterministic automata. MONA represents symbolic deterministic automata by means of Shared Multi-terminal BDDs (ShMTBDDs) \cite{32}. The symbolic $\text{LTL}_f$ synthesis framework of \cite{31} requires standard BDD representation by means of symbolic DFAs as defined above. The transformation from ShMTBDD to BDD is described in \cite{31}.
2.3 FOL Encoding of LTL$_f$

First Order Logic (FOL) encoding of LTL$_f$ translates LTL$_f$ into FOL over finite linear order with monadic predicates. In this paper, we utilize the FOL encoding proposed in [6]. We first restrict our interest to monadic structure. Consider a finite trace $\rho = \rho(0)[\rho(1) \cdots \rho(e)]$, the corresponding monadic structure $I_{\rho} = (\Delta^I, <, I^x)$ describes $\rho$ as follows. $\Delta^I = \{0, 1, 2, \cdots, \text{last}\}$, where last denotes the last point along the trace. The linear order $<$ is defined over $\Delta^I$ in the standard way [16]. The notation $I^x$ indicates the set of monadic predicates that describe the atoms of $P$, where the interpretation of each $p \in P$ is $Q_p = \{x \mid p \in \rho(x)\}$. Intuitively, $Q_p$ is interpreted as the set of positions where $p$ is true in $\rho$. In the translation below, fol($\theta$, $x$), where $\theta$ is an LTL$_f$ formula and $x$ is a variable, is an FOL formula asserting the truth of $\theta$ at point $x$ of the linear order. The translation uses the successor function $+1$, and the variable last that represents the maximal point in the linear order.

- fol($p$, $x$) = ($Q_p(x)$)
- fol($\neg \phi$, $x$) = ($\neg$fol($\phi$, $x$))
- fol($\phi_1 \land \phi_2$, $x$) = (fol($\phi_1$, $x$) \land fol($\phi_2$, $x$))
- fol($\phi_1 \lor \phi_2$, $x$) = (fol($\phi_1$, $x$) \lor fol($\phi_2$, $x$))
- fol($X \phi$, $x$) = ($\exists y ((y = x + 1) \land \text{fol}(\phi, y)))$
- fol($\neg X \phi$, $x$) = ($\exists y ((y = x + 1) \land \text{fol}(\phi, y)))$
- fol($\phi U \phi_2$, $x$) = ($\exists y ((x \leq y \leq \text{last}) \land \text{fol}(\phi, y) \land (\forall z)((x \leq z < y \rightarrow \text{fol}(\phi_1, z))))$
- fol($\phi R \phi_2$, $x$) = ($\exists y ((x \leq y \leq \text{last}) \land \text{fol}(\phi_1, y) \land (\forall z)((x \leq z \leq y \rightarrow \text{fol}(\phi_2, z))))$

For FOL variables, MONA provides a built-in operator $+1$ for successor computation. Moreover, we can use built-in procedures in MONA to represent the variable last. Given a finite trace $\rho$, we denote the corresponding finite linear ordered FOL interpretation of $\rho$ by $I_{\rho}$. The following theorem guarantees the correctness of FOL encoding of LTL$_f$.

**Theorem 1 ([15]).** Let $\phi$ be an LTL$_f$ formula and $\rho$ be a finite trace. Then $\rho \models \phi$ iff $I_{\rho} \models \text{fol}(\phi, 0)$.

3 MSO Encoding

First-order encoding was shown to perform well in the context of LTL$_f$-to-automata translation [39], but other encodings have not been studied. Specifically, the natural question of whether second-order (MSO) outperforms first-order in the same context remained open. MSO is an extension of FOL that allows quantification over monadic predicates [16]. By applying a semantics-driven translation to LTL$_f$, we obtain an MSO encoding that has significantly simpler quantificational structure. This encoding essentially captures in MSO the standard encoding of temporal connectives, cf. [14]. Intuitively speaking, MSO encoding deals with LTL$_f$ formula by interpreting every operator with corresponding subformulas following the semantics of the operator. We now present MSO encoding that translates LTL$_f$ formula $\phi$ to MSO, which is then fed to MONA to produce a symbolic DFA.
For an LTL \( f \) formula \( \phi \) over a set \( P \) of atoms, let \( cl(\phi) \) denote the set of subformulas of \( \phi \). We define atomic formulas as atoms \( p \in P \). For every subformula in \( cl(\phi) \) we introduce monadic predicate symbols as follows: for each atomic subformula \( p \in P \), we have a monadic predicate symbol \( Q_p \); for each nonatomic subformula \( \theta_i \in \{\theta_1, \ldots, \theta_m\} \), we have \( Q_{\theta_i} \). Intuitively speaking, each monadic predicate indicates the positions where the corresponding subformula is true along the linear order.

Let \( mso(\phi) \) be the translation function that given an \( LTL_f \) formula \( \phi \) returns a corresponding MSO formula asserting the truth of \( \phi \) at position 0. We define \( mso(\phi) \) as following: \( mso(\phi) = (\exists Q_{\theta_1}) \cdots (\exists Q_{\theta_m})(Q_\phi(0) \land (\forall x)(\bigwedge_{i=1}^m t(\theta_i, x))) \), where \( x \) indicates the position along the finite linear order. Here \( t(\theta_i, x) \) asserts that the truth of every nonatomic subformula \( \theta_i \) of \( \phi \) at position \( x \) relies on the truth of corresponding subformulas at \( x \) such that following the semantics of \( LTL_f \). Therefore, \( t(\theta_i, x) \) is defined as follows:

- If \( \theta_i = (\neg \theta_j) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow \neg Q_{\theta_j}(x)) \).
- If \( \theta_i = (\theta_j \land \theta_k) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_j}(x) \land Q_{\theta_k}(x))) \).
- If \( \theta_i = (\theta_j \lor \theta_k) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_j}(x) \lor Q_{\theta_k}(x))) \).
- If \( \theta_i = (X \theta_j) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow ((x \neq \text{last}) \land Q_{\theta_j}(x + 1))) \).
- If \( \theta_i = (\mu \theta_j) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (x = \text{last}) \land Q_{\theta_j}(x \land Q_{\theta_i}(x + 1))) \).
- If \( \theta_i = (\mu U \theta_k) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_k}(x) \lor ((x \neq \text{last}) \land Q_{\theta_j}(x \land Q_{\theta_i}(x + 1))) \).
- If \( \theta_i = (\nu \theta_j \land \theta_k) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_j}(x) \land \mu x \models \theta_j \land Q_{\theta_k}(x))) \).

Consider a finite trace \( \rho \), the corresponding interpretation \( \mathcal{I}_\rho \) of \( \rho \) is defined as in Section 2.3. The following theorem asserts the correctness of the MSO encoding.

**Theorem 2.** Let \( \phi \) be an LTL \( f \) formula, \( \rho \) be a finite trace. Then \( \rho \models \phi \) iff \( \mathcal{I}_\rho \models mso(\phi) \).

**Proof.** If \( \phi \) is a propositional atom \( p \), then \( mso(\phi) = Q_p(0) \). It is true that \( \rho \models \phi \) iff \( \mathcal{I}_\rho \models mso(\phi) \). If \( \phi \) is a nonatomic formula, we prove this theorem in two directions.

Suppose first that \( \rho \) satisfies \( \phi \). We expand the monadic structure \( \mathcal{I}_\rho \) with interpretations for the existentially quantified monadic predicate symbols by setting \( Q_{\theta_i} \), the interpretation of subformula \( \theta_i \) in \( \mathcal{I}_\rho \), as the set collecting all points of \( \rho \) satisfying \( \theta_i \), that is \( Q_{\theta_i} = \{x : \rho, x \models \theta_i\} \). We also have \( Q_p = \{x : \rho, x \models p\} \) and denote the expanded structure by \( T_{mso}^\rho \). By assumption, \( Q_\phi(0) \) holds in \( T_{mso}^\rho \). It remains to prove that \( T_{mso}^\rho \models (\forall x)(t(\theta_i, x)) \), for each nonatomic subformula \( \theta_i \in cl(\phi) \), which we prove via structural induction over \( \theta_i \).

- If \( \theta_i = (\neg \theta_j) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (\neg Q_{\theta_j}(x))) \). This holds, since \( Q_{\neg \theta_j} = \{x : x \not\models \theta_j\} \) and \( Q_{\theta_i} = \{x : x \models \theta_i\} \).
- If \( \theta_i = (\theta_j \land \theta_k) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_j}(x) \land Q_{\theta_k}(x))) \). This holds, since \( Q_{\theta_j \land \theta_k} = \{x : x \models \theta_j \land Q_{\theta_k}(x) \} \) and \( Q_{\theta_i} = \{x : x \models \theta_i\} \).
- If \( \theta_i = (\theta_j \lor \theta_k) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_j}(x) \lor Q_{\theta_k}(x))) \). This holds, since \( Q_{\theta_j \lor \theta_k} = \{x : x \models \theta_j \lor \theta_k \} \) and \( Q_{\theta_i} = \{x : x \models \theta_i\} \).
- If \( \theta_i = (X \theta_j) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_j}(x) \land (x = last \land Q_{\theta_i}(x + 1)))) \).
- If \( \theta_i = (\mu \theta_j) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_j}(x) \land (x = last \land Q_{\theta_i}(x + 1)))) \).
- If \( \theta_i = (\mu U \theta_k) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_k}(x) \lor ((x \neq last) \land Q_{\theta_j}(x \land Q_{\theta_i}(x + 1)))) \).
- If \( \theta_i = (\nu \theta_j \land \theta_k) \), then \( t(\theta_i, x) = (Q_{\theta_i}(x) \leftrightarrow (Q_{\theta_j}(x) \land \mu x \models \theta_j \land Q_{\theta_k}(x))) \).


– If \( \theta_l = (X \theta_j) \), then \( t(\theta_l, x) = (Q_{\theta_l}(x) \leftrightarrow ((x \neq \text{last}) \land Q_{\theta_l}(x + 1))) \). This holds, since \( Q_{(X \theta_j)} = \{ x : \rho, x \models (X \theta_j) \} = \{ x : x \neq \text{last} \text{ and } \rho, x + 1 \models \theta_j \} \), and \( Q_{\theta_l} = \{ x : \rho, x \models \theta_l \} \).

– If \( \theta_l = (N \theta_j) \), then \( t(\theta_l, x) = (Q_{\theta_l}(x) \leftrightarrow ((x = \text{last}) \lor Q_{\theta_l}(x + 1))) \). This holds, since \( Q_{(N \theta_j)} = \{ x : \rho, x \models (N \theta_j) \} = \{ x : x = \text{last} \text{ or } \rho, x + 1 \models \theta_j \} \), and \( Q_{\theta_l} = \{ x : \rho, x \models \theta_l \} \).

– If \( \theta_l = (\theta_j \cup \theta_k) \), then \( t(\theta_l, x) = (Q_{\theta_l}(x) \leftrightarrow (Q_{\theta_k}(x) \lor \theta_j \land \theta_k)) \). This holds, since \( Q_{(\theta_j \cup \theta_k)} = \{ x : \rho, x \models (\theta_j \cup \theta_k) \} = \{ x : \rho, x \models \theta_j \text{ or } x \neq \text{last} \text{ with } \rho, x \models \theta_k \text{ also } \rho, x + 1 \models \theta_j \} \), \( Q_{\theta_k} = \{ x : \rho, x \models \theta_k \} \), and \( Q_{\theta_j} = \{ x : \rho, x \models \theta_j \} \).

– If \( \theta_l = (\theta_j \cap \theta_k) \), then \( t(\theta_l, x) = (Q_{\theta_l}(x) \leftrightarrow (Q_{\theta_k}(x) \land \theta_j \land \theta_k)) \). This holds, since \( Q_{(\theta_j \cap \theta_k)} = \{ x : \rho, x \models (\theta_j \cap \theta_k) \} = \{ x : \rho, x \models \theta_j \text{ and } x \models \theta_k \text{ with } x = \text{last} \text{ or } \rho, x \models \theta_j \text{ or } \rho, x + 1 \models \theta_j \} \), \( Q_{\theta_j} = \{ x : \rho, x \models \theta_j \} \), and \( Q_{\theta_k} = \{ x : \rho, x \models \theta_k \} \).

Assume now that \( I_\rho \models \text{mso}(\phi) \). This means that there is an expansion of \( I_\rho \) with monadic interpretations \( Q_{\theta_j} \) for each nonatomic subformula \( \theta_j \in \text{cl}(\phi) \) such that this expanded structure \( I_\rho^{\text{mso}} \models \left( (Q_\phi(0) \land (\forall x) \bigwedge_{i=1}^n t(\theta_i, x) \right) \). We now prove by induction on \( \phi \) that if \( x \in Q_\phi \), then \( \rho, x \models \phi \) such that \( Q_\phi(0) \) indicates that \( \rho, 0 \models \phi \).

– If \( \phi = (\neg \theta_j) \), then \( t(\phi, x) = (Q_{\phi}(x) \leftrightarrow (x \notin Q_{\theta_j})) \). Since \( t(\phi) \) holds at every point \( x \) of \( I_\rho^{\text{mso}} \), it holds that \( x \in Q_{\phi} \) iff \( x \notin Q_{\theta_j} \). It follows by induction that \( \rho, x \models \phi \).

– If \( \phi = (\theta_j \land \theta_k) \), then \( t(\phi, x) = (Q_{\phi}(x) \leftrightarrow (Q_{\theta_j}(x) \land Q_{\theta_k}(x))) \). Since \( t(\phi) \) holds at every point \( x \) of \( I_\rho^{\text{mso}} \), it follows that \( x \in Q_{\phi} \) iff \( x \in Q_{\theta_j} \) and \( x \in Q_{\theta_k} \). It follows by induction that \( \rho, x \models \phi \).

– If \( \phi = (\theta_j \lor \theta_k) \), then \( t(\phi, x) = (Q_{\phi}(x) \leftrightarrow (Q_{\theta_j}(x) \lor Q_{\theta_k}(x))) \). Since \( t(\phi) \) holds at every point \( x \) of \( I_\rho^{\text{mso}} \), it follows that \( x \in Q_{\phi} \) iff \( x \in Q_{\theta_j} \) or \( x \in Q_{\theta_k} \). It follows by induction that \( \rho, x \models \phi \).

– If \( \phi = (X \theta_j) \), then \( t(\phi, x) = (Q_{\phi}(x) \leftrightarrow ((x \neq \text{last}) \land Q_{\theta_j}(x + 1))) \). Since \( t(\phi) \) holds at every point \( x \) of \( I_\rho^{\text{mso}} \), it follows that \( x \in Q_{\phi} \) iff \( x \neq \text{last} \text{ and } x + 1 \in Q_{\theta_j} \). It follows by induction that \( x \neq \text{last} \text{ and } \rho, x \models \phi \).

– If \( \phi = (N \theta_j) \), then \( t(\phi, x) = (Q_{\phi}(x) \leftrightarrow ((x = \text{last}) \lor (Q_{\theta_j}(x + 1))) \). Since \( t(\phi) \) holds at every point \( x \) of \( I_\rho^{\text{mso}} \), it follows that \( x \in Q_{\phi} \) iff \( x = \text{last} \text{ or } x + 1 \in Q_{\theta_j} \). It follows by induction that \( x = \text{last} \text{ or } \rho, x \models \phi \).

– If \( \phi = (\theta_j \cup \theta_k) \), then \( t(\phi, x) = (Q_{\phi}(x) \leftrightarrow (Q_{\theta_k}(x) \lor (Q_{\theta_j}(x) \land Q_{\phi}(x + 1)))) \). Since \( t(\phi) \) holds at every point \( x \) of \( I_\rho^{\text{mso}} \), it follows that \( x \in Q_{\phi} \) iff \( x \in Q_{\theta_k} \text{ or } x \neq \text{last} \text{ with } x \in Q_{\theta_j} \text{ also } x + 1 \in Q_{\phi} \). Thus, \( \rho, x \models \phi \).

– If \( \phi = (\theta_j \cap \theta_k) \), then \( t(\phi, x) = (Q_{\phi}(x) \leftrightarrow (Q_{\theta_k}(x) \land (Q_{\theta_j}(x) \land Q_{\phi}(x + 1)))) \). Since \( t(\phi) \) holds at every point \( x \) of \( I_\rho^{\text{mso}} \), it follows that \( x \in Q_{\phi} \) iff \( x \in Q_{\theta_k} \text{ with } x = \text{last} \text{ or } x \in Q_{\theta_j} \text{ or } x + 1 \in Q_{\phi} \). Thus, \( \rho, x \models \phi \).

\( \square \nabla \)

### 4 Compact MSO Encoding

The MSO encoding described in Section 3 is closely related to the translation of \( \text{LTL}_f \) to alternating automata [3], with each automaton state corresponding to a monadic pred-
icate. The construction, however, is subject only to syntactic minimization. Can we optimize this encoding using automata-theoretic minimization? In fact, MONA itself applies automata-theoretic minimization. Can we use MONA to produce a more efficient encoding for MONA?

The key observation is that MONA can produce a compact symbolic representation of a non-deterministic automaton (NFA) representing a given LTL formula, and we can use this symbolic NFA to create a more compact MSO encoding for LTL. This is based on the observation that while the translation from past LTL to DFA is 2-EXP, the translation from past LTL to DFA is 1-EXP, as explained below. We proceed as follows: (1) Reverse a given LTL formula \( \phi \) to Past LTL formula \( \phi^R \); (2) Use MONA to construct the DFA of \( \phi^R \), the reverse of which is an NFA, that accepts exactly the reverse language of the words satisfying models(\( \phi \)); (3) Express this symbolic DFA in second-order logic and “invert” it to get \( D_\phi \), the corresponding DFA of \( \phi \).

The crux of this approach, which follows from [5,6], is that the DFA corresponding to the reverse language of an LTL formula \( \phi \) of length \( n \) has only \( 2^n \) states. The reverse of this latter DFA is an NFA for \( \phi \). We now elaborate on these steps.

### 4.1 LTL \( f \) to PLTL \( f \)

Past Linear Temporal Logic over finite traces, i.e. PLTL \( f \), has the same syntax as PLTL over infinite traces introduced in [24]. Given a set of propositions \( P \), the grammar of PLTL \( f \) is given by:

\[
\psi ::= \top \mid \bot \mid p \mid \neg \psi \mid \psi_1 \land \psi_2 \mid Y\psi \mid \psi_1 \psi_2
\]

Given a finite trace \( \rho \) and a PLTL \( f \) formula \( \psi \), we inductively define when \( \psi \) is true for \( \rho \) at step \( x (0 \leq x < |\rho|) \), written by \( \rho, x \models \psi \), as follows:

- \( \rho, x \models \top \) and \( \rho, x \not\models \bot \);
- \( \rho, x \models p \) iff \( p \in \rho(x) \);
- \( \rho, x \models \neg \psi \) iff \( \rho, x \not\models \psi \);
- \( \rho, x \models \psi_1 \land \psi_2 \) iff \( \rho, x \models \psi_1 \) and \( \rho, x \models \psi_2 \);
- \( \rho, x \models Y\psi \) iff \( x - 1 \geq 0 \) and \( \rho, x - 1 \models \psi \);
- \( \rho, x \models \psi_1 \psi_2 \) iff there exists \( y \) such that \( 0 \leq y \leq x \) and \( \rho, y \models \psi_2 \), and for all \( z, y < z \leq x \), we have \( \rho, z \models \psi_1 \).

A PLTL \( f \) formula \( \psi \) is true in \( \rho \), denoted by \( \rho \models \psi \), if and only if \( \rho, |\rho| - 1 \models \psi \). To reverse an LTL \( f \) formula \( \phi \), we replace each temporal operator in \( \phi \) with the corresponding past operator of PLTL \( f \) thus getting \( \phi^R \). X(Next) and U(Until) correspond to Y(Before) and S(Since) respectively.

We define \( \rho^R = \rho(|\rho| - 1), \rho(|\rho| - 2), \ldots, \rho[1], \rho[0] \) to be the reverse of \( \rho \). Moreover, given language \( \mathcal{L} \), we denote the reverse of \( \mathcal{L} \) by \( \mathcal{L}^R \) such that \( \mathcal{L}^R \) collects all reversed sequences in \( \mathcal{L} \). Formally speaking, \( \mathcal{L}^R = \{ \rho^R : \rho \in \mathcal{L} \} \). The following theorem shows that PLTL \( f \) formula \( \phi^R \) accepts exactly the reverse language satisfying \( \phi \).

**Theorem 3.** Let \( \mathcal{L}(\phi) \) be the language of LTL \( f \) formula \( \phi \) and \( \mathcal{L}^R(\phi) \) be the reverse language, then \( \mathcal{L}(\phi^R) = \mathcal{L}^R(\phi) \).
Proof. $L(\phi^R) = L^R(\phi)$ iff for an arbitrary sequence $\rho \in L(\phi)$ such that $\rho \models \phi$, it is true that $\rho^R \models \phi^R$. We prove the theorem by the induction over the structure of $\phi$. 

- Basically, if $\phi = p$ is an atom, then $\phi^R = p$, $\rho \models \phi$ iff $p \in \rho[0]$ such that $p \in \rho^R[\text{last}]$. Therefore, $\rho^R \models \phi^R$;
- If $\phi = \neg \phi_1$, then $\phi^R = \neg \phi_1^R$, $\rho \models \neg \phi_1$ iff $\rho \not\models \phi_1$, such that by induction hypothesis $\rho^R \not\models \phi_1^R$ holds, therefore $\rho^R \models \phi^R$ is true;
- If $\phi = \phi_1 \land \phi_2$, then $\phi^R = \phi_1^R \land \phi_2^R$, $\rho \models \phi$ iff $\rho$ satisfies both $\phi_1$ and $\phi_2$. By induction hypothesis $\rho^R \models \phi_1^R$ and $\rho^R \models \phi_2^R$ hold, therefore $\rho^R \models \phi^R$ is true;
- If $\phi = X \phi_1$, $\phi^R = Y \phi^R_1$, $\rho \models \phi$ suffix $\rho'$ is sequence $\rho[1], \rho[2], \ldots, \rho[\text{last}]$ and $\rho' \models \phi_1$. By induction hypothesis, $\rho^R \models \phi^R_1$ holds, in which case $\rho^R, \text{last} - 1 \models \phi^R_1$ is true, therefore $\rho^R \models \phi^R$ holds.
- If $\phi = \phi_2 U \phi_2$, $\rho \models \phi$ iff there exists $y$ such that $y (0 \leq y \leq \text{last})$, suffix $\rho' = \rho[y], \rho[y + 1], \ldots, \rho[\text{last}]$ satisfies $\phi_2$. Also for all $z$ such that $z (0 \leq z < y)$, $\rho'' = \rho[z], \rho[z + 1], \ldots, \rho[\text{last}]$ satisfies $\phi_1$. By induction hypothesis, $\rho^R \models \phi^R_1$ and $\rho^R \models \phi^R_2$ hold, therefore we have $\rho^R, \text{last} - y \models \phi^R_2$ and $\forall z. \text{last} - y < z \leq \text{last}, \rho^R, z \models \phi^R_1$ hold such that $\rho^R \models \phi^R$. The proof is done.

\[ \square \]

### 4.2 PLTL $f$ to DFA

The DFA construction from PLTL $f$ formulas relies on MONA as well. Given PLTL $f$ formula $\psi$, we are able to translate $\psi$ to FOL formula as input of MONA, which returns the DFA. For PLTL $f$ formula $\psi$ over $\mathcal{P}$, we construct the corresponding FOL formula with respect to point $x$ by a function $\text{fol}_p(\psi, x)$ asserting the truth of $\psi$ at $x$. Detailed translation of PLTL $f$ to FOL is defined below. The translation uses the predecessor function $\text{last}$, and the predicate $\text{last}$ referring to the last point along the finite trace.

- $\text{fol}_p(p, x) = (Q_p(x))$
- $\text{fol}_p(\neg \psi, x) = (\neg \text{fol}_p(\psi, x))$
- $\text{fol}_p(\psi_1 \land \psi_2, x) = (\text{fol}_p(\psi_1, x) \land \text{fol}_p(\psi_2, x))$
- $\text{fol}_p(\exists y(y = x - 1 \land (y \geq 0) \land \text{fol}_p(\psi, y)))$
- $\text{fol}_p(\psi_1 S \psi_2, x) = (\exists y((0 \leq y \leq x) \land \text{fol}_p(\psi_2, y) \land (\forall z)((y < z \leq x) \rightarrow \text{fol}_p(\psi_1, z))))$

Consider a finite trace $\rho$, the corresponding interpretation $\mathcal{I}_\rho$ is defined as in Section 2.3. The following theorem guarantees the correctness of the above translation.

**Theorem 4.** [15] Let $\psi$ be a PLTL $f$ formula, $\rho$ be a finite trace. Then $\rho \models \psi$ iff $\mathcal{I}_\rho \models \text{fol}_p(\psi, \text{last})$, where $\text{last} = |\rho| - 1$.

**Proof.** We prove the theorem by the induction over the structure of $\psi$.

- Basically, if $\psi = p$ is an atom, $\rho \models \psi$ iff $p \in \rho[\text{last}]$. By the definition of $\mathcal{I}$, we have that $\text{last} \in Q_p$. Therefore, $\rho \models \psi$ iff $\mathcal{I}_\rho \models \text{fol}_p(p, \text{last})$ holds;
- If $\psi = \neg \psi$, $\rho \models \psi$ iff $\rho \not\equiv \psi$. By induction hypothesis it is true that $\mathcal{I}_\rho \not\models \text{fol}_p(\psi, \text{last})$, therefore $\mathcal{I}_\rho \models \text{fol}_p(\neg \psi, \text{last})$ holds;
- If $\psi = \psi_1 \land \psi_2$, $\rho \models \psi$ iff $\rho$ satisfies both $\psi_1$ and $\psi_2$. By induction hypothesis, it is true that $\mathcal{I}_\rho \models \text{fol}_p(\psi_1, \text{last})$ and $\mathcal{I}_\rho \models \text{fol}_p(\psi_2, \text{last})$. Therefore $\mathcal{I}_\rho \models \text{fol}_p(\psi_1, \text{last}) \land \text{fol}_p(\psi_2, \text{last})$ holds;
- If $\psi = \psi_1 \land \psi_2$, $\rho \models \psi$ iff prefix $\rho' = \rho[0], \rho[1], \ldots, \rho[\text{last} - 1]$ of $\rho$ satisfies $\rho' \models \psi_1$. Let $\mathcal{I}_\rho'$ be the corresponding interpretation of $\rho'$, thus for every atom $p \in \mathcal{P}$, $x \in Q'_p$ iff $x \in Q_p$ where $Q'_p$ is the corresponding monadic predicate of $p$ in $\mathcal{I}_\rho'$. By induction hypothesis it is true that $\mathcal{I}_\rho' \models \text{fol}_p(\psi_1, \text{last} - 1)$, therefore $\mathcal{I}_\rho \models \text{fol}_p(\psi_1, \text{last})$ holds.
- If $\psi = \psi_1 \psi_2$, $\rho \models \psi$ iff there exists $y$ such that $0 \leq y \leq \text{last}$ and prefix $\rho' = \rho[0], \rho[1], \ldots, \rho[y]$ of $\rho$ satisfies $\psi_2$ and for all $z$ such that $y < z \leq \text{last}$, $\rho'' = \rho[0], \rho[1], \ldots, \rho[z]$ satisfies $\psi_1$. Let $\mathcal{I}_\rho'$ and $\mathcal{I}_\rho''$ be the corresponding interpretations of $\rho'$ and $\rho''$. Thus for every atom $p \in \mathcal{P}$ it is true that $x \in Q'_p$ iff $x \in Q_p$, $x \in Q''_p$ iff $x \in Q_p$, where $Q'_p$ and $Q''_p$ correspond to the monadic predicates of $p$ in $\mathcal{I}_\rho'$ and $\mathcal{I}_\rho''$ respectively. By induction hypothesis it is true that $\mathcal{I}_\rho' \models \text{fol}_p(\psi_2, \text{last} - y)$ and $\mathcal{I}_\rho'' \models \text{fol}_p(\psi_1, \text{last} - z)$ hold, therefore $\mathcal{I}_\rho \models \text{fol}_p(\psi_1 \psi_2, \text{last})$.

\[ \square \]

### 4.3 Reversing DFA via Second-Order Logic

For simplification, from now on we use $\psi$ to denote the corresponding $\text{PLTL}_f$ formula $\phi^R$ of $\mathcal{L}_f$ formula $\phi$. We first describe how BDDs represent a symbolic DFA. Then we introduce the Compact MSO encoding that inverts the DFA by formulating such BDD representation into a second-order formula. The connection between BDD representation and second-order encoding is novel, to the best of our knowledge.

As defined in Section 2.2, given a symbolic DFA $\mathcal{F}_\psi = (\mathcal{P}, \mathcal{X}, X_0, \eta, f)$ represented by a sequence $B = (B_0, B_1, \ldots, B_{k-1})$ of BDDs, where there are $k$ variables in $\mathcal{X}$, a run of such DFA on a word $\rho = \rho[0], \rho[1], \ldots, \rho[e-1]$ involves a sequence of states $\xi = X_0, X_1, \ldots, X_e$ of length $e + 1$. For the moment if we omit the last state reached on an input of length $e$, we have a sequence of states $\xi' = X_0, X_1, \ldots, X_{e-1}$ of length $e$. Thus we can think of the run $\xi'$ as a labeling of the positions of the word with states, which is $(\rho[0], X_0), (\rho[1], X_1), \ldots, (\rho[e-1], X_{e-1})$. At each position with given word and state, the transition moving forward involves a computation over every $B_q$ ($0 \leq q \leq k - 1$). To perform such computation, take the high branch in every node labeled by variable $v \in \{\mathcal{X} \cup \mathcal{P}\}$ if $v$ is assigned 1 and the low branch otherwise.

The goal here is to write a formula $\text{Rev}(\mathcal{F}_\psi)$ such that there is an accepting run over $\mathcal{F}_\psi$ of a given word $\rho$ iff $\rho^R$ is accepted by $\text{Rev}(\mathcal{F}_\psi)$. To do this, we introduce one second-order variable $V_q$ for each $x_q \in \mathcal{X}$ with $0 \leq q \leq k - 1$, and one second-order variable $N_\alpha$ for every nonterminal node $\alpha$ in BDDs, $u$ nonterminal nodes in total. The $V_q$ variables collect the positions where $x_q$ holds, and the $N_\alpha$ variables indicate the positions where the node $\alpha$ is visited, when computing the transition. To collect all transitions moving towards accepting states, we have BDD $B'_f = f(\eta(\mathcal{X}, \mathcal{P}))$.

Here are some notations. Let $\alpha$ be a nonterminal node, $c$ be a terminal node in $B_q$ such that $c \in \{0, 1\}$ and $d \in \{0, 1\}$ be the value of $v$. For nonterminal node $\alpha$, we
define:
\[
\text{Pre}(\alpha) = \{(\beta, v, d) : \text{there is an edge from } \beta \text{ to } \alpha \text{ labelled by } v = d\}
\]
\[
\text{Post}(\alpha) = \{(\beta, v, d) : \text{there is an edge from } \alpha \text{ to } \beta \text{ labelled by } v = d\}
\]
For every terminal node \(c\) in BDD \(B_q\), we define:
\[
\text{Pre}_T(B_q, c) = \{(\beta, v, d) : \text{there is an edge from } \beta \text{ to } c \text{ labelled by } v = d \text{ in BDD } B_q\}
\]
Also, we use \(\in^d\) to denote \(\in\) when \(d = 1\) and \(\notin\) when \(d = 0\). For each BDD \(B_q\), \(\text{root}(B_q)\) indicates the root node of \(B_q\).

We use these notations to encode the following statements:

1) At the last position, state \(X_0\) should hold since \(\xi'\) is being inverted and \(X_0\) is the starting point;

\[
R_{\text{init}} = (x = \text{last}) \rightarrow \left( \bigwedge_{0 \leq q \leq k-1} x_0(x_q) = d \implies x \in^d V_q \right);
\]

2) At position \(x\), if the current computation is at nonterminal node \(\alpha\) labeled by \(v\), then (2.a) the current computation must come from a predecessor labeled by \(v'\) following the value of \(v'\), and (2.b) the next step is moving to the corresponding successor following the value of \(v\);

\[
\text{node} = \bigwedge_{1 \leq \alpha \leq n} \left( \text{Pre}_\alpha \land \text{Post}_\alpha \right); \text{ where}
\]
\[
\text{Pre}_\alpha = \left( x \in N_\alpha \rightarrow \left( \bigvee_{(\beta, v', d) \in \text{Pre}(\alpha)} [x \in N_\beta \land x \in^d v'] \right) \right)
\]
\[
\text{Post}_\alpha = \left( \bigwedge_{(\beta, v, d) \in \text{Post}(\alpha)} [x \in N_\alpha \land x \in^d v \rightarrow x \in N_\beta] \right).
\]

3) At position \(x\) such that \(x > 0\), if the current computation node \(\alpha\) moves to a terminal node \(c\) of \(B_q\), then the value of \(x_q\) at position \(x-I\) is given by the value of \(c\). Such computations of all \(B_q(0 \leq q \leq k-1)\) finish one transition;

\[
R_{\text{terminal}} = \bigwedge_{0 \leq q \leq k-1} \left( \bigwedge_{(\beta, v, d) \in \text{Pre}_T(B_q, c)} \left[ (x > 0 \land x \in N_\beta \land x \in^d v) \rightarrow (x - 1 \in^c V_q) \right] \right);
\]

4) At the first position, the current computation on \(B'_1\) has to surely move to terminal 1, therefore terminating the running trace of \(\xi\).

\[
R_{\text{acc}} = (x = 0) \rightarrow \left( \bigvee_{(\beta, v, d) \in \text{Pre}_T(B'_1, 1)} [x \in N_\beta \land x \in^d v] \right).
\]

To get all computations over BDDs start from the root at each position, we have \(\text{roots} = \bigwedge_{0 \leq x \leq \text{last}} \bigwedge_{0 \leq q \leq k-1} x \in \text{root}(B_q)\)

\(\text{Rev}(\mathcal{F}_\phi)\) has to take a conjunction of all requirements above such that

\[
\text{Rev}(\mathcal{F}_\phi) = (\exists V_0)(\exists V_1) \ldots (\exists V_{k-1})(\exists N_1)(\exists N_2) \ldots (\exists N_n)(\forall x)(R_{\text{init}} \land \text{node} \land R_{\text{terminal}} \land R_{\text{acc}} \land \text{roots}).
\]

Therefore, let \(\text{Compso}(\phi)\) be the translation function that given an \(\text{LTL}_f\) formula \(\phi\) returns a corresponding second-order formula applying the Compact MSO encoding.
we define \( C_{\text{msso}}(\phi) = \text{Rev}(\mathcal{F}_\psi) \) asserting the truth of \( \phi \) at position 0, where \( \psi \) is the corresponding PLTL\(_f\) formula of \( \phi \), and \( \mathcal{F}_\psi \) is the symbolic DFA of \( \psi \). The following theorem asserts the correctness of the Compact MSO encoding.

**Theorem 5.** The models of formula \( C_{\text{msso}}(\phi) \) are exactly the words satisfying \( \phi \).

**Proof.** We first have that \( \mathcal{L}(\phi) = \mathcal{L}^R(\psi) = \mathcal{L}^R(\mathcal{F}_\psi) \) holds since \( \psi \) is the corresponding PLTL\(_f\) formula of \( \phi \) and \( \mathcal{F}_\psi \) collects exactly the words satisfying \( \psi \). Moreover, \( \mathcal{L}(\text{Rev}(\mathcal{F}_\psi)) = \mathcal{L}^R(\mathcal{F}_\psi) \) is true following the construction rules of \( \text{Rev}(\mathcal{F}_\psi) \) described above and \( C_{\text{msso}}(\phi) = \text{Rev}(\mathcal{F}_\psi) \). Therefore, \( \mathcal{L}(\phi) = \mathcal{L}(C_{\text{msso}}(\phi)) \) holds, in which case the models of formula \( C_{\text{msso}}(\phi) \) are exactly the words satisfying \( \phi \). \( \square \)

Notice that the size of \( C_{\text{msso}}(\phi) \) is in linear on the size of the BDDs, which lowers the logical complexity comparing to the MSO encoding in Section 3. Moreover, in the Compact MSO encoding, the number of existential second-order symbols for state variables are nevertheless possibly less than that in MSO encoding, but new second-order symbols for nonterminal BDD nodes are introduced. BDDs provide a compact representation, in which redundant nodes are reduced. Such advantages allow Compact MSO encoding to use as few second-order symbols for BDD nodes as possible.

## 5 Experimental Evaluation

We implemented proposed second-order encodings in different parsers for LTL\(_f\) formulas using C++. Each parser is able to generate a second-order formula corresponding to the input LTL\(_f\) formula, which is then fed to MONA \([13]\) for subsequent symbolic DFA construction. Moreover, we employed Syft’s \([31]\) code to translate LTL\(_f\) formula into first-order logic (FOL), which adopts the first-order encoding described in Section 2.3.

**Benchmarks** We conducted the comparison of first-order encoding with second-order encoding in the context of LTL\(_f\)-to-DFA, thus only satisfiable but not valid formulas are interesting. Therefore, we first ran an LTL\(_f\) satisfiability checker on LTL\(_f\) formulas and their negations to filter the valid or unsatisfiable formulas. We collected 5690 formulas, which consist of two classes of benchmarks: 765 LTL\(_f\)-specific benchmarks, of which 700 are scalable LTL\(_f\) pattern formulas from \([10]\) and 65 are randomly conjuncted common LTL\(_f\) formulas from \([7,11,25]\) ; and 4925 LTL-as-LTL\(_f\) formulas from \([26,27]\), since LTL formulas share the same syntax as LTL\(_f\).  

**Experimental Setup** To explore the comparison between first-order and second-order for LTL\(_f\)-to-DFA translation, we ran each formula for every encoding on a node within a high performance cluster. These nodes contain 12 processor cores at 2.2 GHz each with 8GB of RAM per core. Time out was set to be 1000 seconds. Cases that cannot generate the DFA within 1000 seconds generally fail even if the time limit is extended, since in these cases, MONA typically cannot handle the large BDD.

### 5.1 Optimizations of Second-Order Encoding

Before diving into the optimizations of second-order encoding, we first study the potential optimization space of the first-order encoding that translates LTL\(_f\) to FOL. Due to the
syntax-driven translation of FOL encoding, we are only able to apply different normal forms, Boolean Norma Form (BNF) and Negation Normal Form (NNF). We compared the impact on performance of FOL encoding with two LTL\(_f\) normal forms. It turns out that the normal form does not have a measurable impact on the performance of the first-order encoding. Since FOL-BNF encoding performs slightly better than FOL-NNF, the best FOL encoding refers to FOL-BNF.

To explore the potential optimization space of the second-order encodings proposed in this paper, we hope to conduct experiments with different optimizations. We name second-order encoding with different optimizations variations. We first show optimizations of the MSO encoding described in Section 3, then describe variations of the Compact MSO encoding shown in Section 4 in the following.

The basic MSO encoding defined in Section 3 translates LTL\(_f\) to MSO in a natural way, in the sense that introducing a second-order predicate for each non-atomic subformula and employing the \(\leftrightarrow\) constraint. Inspired by \cite{22,27}, we define in this section several optimizations to simplify such encoding thus benefiting symbolic DFA construction. These variations indicating different optimizations are combinations of three independent components: (1) the Normal Form (choose between BNF or NNF); (2) the Constraint Form (choose between Fussy or Sloppy); (3) the Variable Form (choose between Full or Lean). In each component one can choose either of two options to make. Thus for example, the variation described in Section 3 is BNF-Fussy-Full. Note that BNF-Sloppy are incompatible, as described below, and so there are \(2^3 - 2 = 6\) viable combinations of the three components above. We next describe the variations in details.

**Constraint Form** We call the translation described in Section 3 the Fussy variation, in which we translate \(\phi\) to MSO formula \(\text{mso}(\phi)\) by employing an \(\iff\) constraint (see Section 3). For example:

\[
t(\theta_i, x) = (Q_{\theta_i}(x) \iff (Q_{\theta_j}(x) \land Q_{\theta_k}(x))) \text{ if } \theta_i = (\theta_j \land \theta_k)
\]  

(1)

We now introduce Sloppy variation, inspired by \cite{27}, which allows less tight constraints that still hold correctness guarantees thus may speed up the symbolic DFA construction. To better reason the incompatible combination BNF-Sloppy, we specify the description for different normal forms, NNF and BNF separately.

For LTL\(_f\) formulas in NNF, the Sloppy variation requires only a single implication constraint \(\rightarrow\). Specifically the Sloppy variation \(\text{mso}_s(\phi)\) for NNF returns MSO formula \((\exists Q_{\theta_1}) \cdots (\exists Q_{\theta_m}) (Q_\phi(0) \land (\forall x)(\bigwedge_{i=1}^m t_s(\theta_i, x)))\), where \(t_s(\theta_i)\) is defined just like \(t(\theta_i)\), replacing the \(\iff\) by \(\rightarrow\). For example translation \(1\) under the Sloppy translation for NNF is \(t_s(\theta_i, x) = (Q_{\theta_i}(x) \rightarrow (Q_{\theta_j}(x) \land Q_{\theta_k}(x)))\).

The Sloppy variation cannot be applied to LTL\(_f\) formulas in BNF since the \(\leftrightarrow\) constraint defined in function \(t(\theta_i)\) is needed only to handle negation correctly. BNF requires a general handling of negation. For LTL\(_f\) formulas in NNF, negation is applied only to atomic formulas such that handled implicitly by the base case \(\rho, x \models p \iff \rho, x \not\models \neg p\). Therefore, translating LTL\(_f\) formulas in NNF does not require the \(\leftrightarrow\) constraint. For example, consider LTL\(_f\) formula \(\phi = \neg F a\) (in BNF), where \(a\) is an atom. The corresponding BNF-Sloppy variation gives MSO formula \((\exists Q_{\neg F a})(\exists Q_{F a})(Q_{\neg F a}(0) \land ((\forall x)((Q_{\neg F a}(x) \rightarrow \neg Q_{F a}(x)) \land (Q_{F a}(x) \rightarrow (Q_{a}(x) \vee ((x \neq \text{last}) \land Q_{F a}(x+1))))))))\) via \(\text{mso}_s(\phi)\). Consider finite trace \(\rho = (a = 0), (a = 1), \rho \models \phi\) iff \(\rho \models \text{mso}_s(\phi)\) does
not hold since \( \rho \not\models \neg Fa \). This happens because \( \neg Fa \) requires \((Q_{\neg Fa}(x) \leftrightarrow \neg Q_{Fa}(x))\) as \( Fa \) is an non-atomic subformula. Therefore, Sloppy variation can only be applied to \( \text{LTL}_f \) formulas in NNF.

The following theorem asserts the correctness of the Sloppy variation.

**Theorem 6.** Let \( \phi \) be an \( \text{LTL}_f \) formula in NNF and \( \rho \) be a finite trace. Then \( \rho \models \phi \iff I_{\rho} \models \text{msO}_s(\phi) \).

The proof here is analogous to that of Theorem 2 The crux here is that the \( \leftrightarrow \) in the encoding, however, is applied only to \( \text{LTL}_f \) formulas in NNF, so negation can be applied only to atomic propositions, which is handled by the base case \((\neg \rho(x))\).

**Variable Form** In all the variations of the MSO encoding we can get above, we introduced a monadic predicate for each non-atomic subformula in \( cl(\phi) \), this is the Full variation. We now introduce Lean variation, a new variable form, aiming at decreasing the number of quantified monadic predicates. Fewer quantifiers on monadic predicates could benefit symbolic DFA construction a lot since quantifier elimination in MONA takes heavy cost. The key idea of Lean variation is introducing monadic predicates only for atomic subformulas and non-atomic subformulas of the form \( \phi_j U \theta_k \) or \( \phi_j R \theta_k \) (named as \( U \)- or \( R \)-subformula respectively).

For non-atomic subformulas that are not \( U \)- or \( R \)-subformulas, we can construct second-order terms using already defined monadic predicates to capture the semantics of them. Function \( \text{lean}(\theta_i) \) is defined to get such second-order terms. Intuitively speaking, \( \text{lean}(\theta_i) \) indicates the same positions where \( \theta_i \) is true as \( Q_{\theta_i} \), does, instead of having \( Q_{\theta_i} \) explicitly. We use built-in second-order operators in MONA to simplify the definition of \( \text{lean}(\theta_i) \). ALIVE is defined using built-in procedures in MONA to collect all instances along the finite trace. MONA also allows to apply set union, intersection, and difference for second-order terms, as well as the \(+1\) operation (which shifts a monadic predicate backwards by one position). \( \text{lean}(\theta_i) \) is defined over the structure of \( \theta_i \) as following:

- If \( \theta_i = (\neg \theta_j) \), then \( \text{lean}(\theta_i) = (\text{ALIVE}\setminus\text{lean}(\theta_j)) \)
- If \( \theta_i = (\theta_j \land \theta_k) \), then \( \text{lean}(\theta_i) = (\text{lean}(\theta_j) \text{ inter} \text{lean}(\theta_k)) \)
- If \( \theta_i = (\theta_j \lor \theta_k) \), then \( \text{lean}(\theta_i) = (\text{lean}(\theta_j) \text{ union} \text{lean}(\theta_k)) \)
- If \( \theta_i = (X \theta_j) \), then \( \text{lean}(\theta_i) = ((\text{lean}(\theta_j) \setminus \text{last}) \cup \text{last}) \)
- If \( \theta_i = (N \theta_j) \), then \( \text{lean}(\theta_i) = (\text{lean}(\theta_j) \setminus 1) \cup \text{last}) \)
- If \( \theta_i = (\theta_j U \theta_k) \) or \( \theta_i = (\theta_j R \theta_k) \), then \( \text{lean}(\theta_i) = Q_{\theta_k} \), where \( Q_{\theta_k} \) is the corresponding monadic predicate.

The following lemma ensures that \( \text{lean}(\theta_i) \) keeps the interpretation of each non-atomic subformula \( \theta_i \in cl(\phi) \).

**Lemma 1.** Let \( \phi \) be an \( \text{LTL}_f \) formula, \( \rho \) be a finite trace. Then \( \rho, x \models \theta_i \iff \text{lean}(\theta_i)(x) \) holds, where \( x \) is the position in \( \rho \).

**Proof.** Suppose first that \( \rho, x \models \theta_i \). We prove this inductively on the structure of \( \theta_i \).
If $\theta_i = -\theta_j$, then lean($\theta_j$) = ($\lambda$ALIVE \$(lean($\theta_j$))). lean($\theta_j$)$\{x : x \notin$ lean($\theta_j$)$\}$ and lean($\theta_j$)$\{x : \rho, x \models \theta_j\}$.

If $\theta_i = \theta_j \land \theta_k$, then lean($\theta_i$) = (lean($\theta_j$) \ inter lean($\theta_k$)). lean($\theta_i$)$\{x : x \in$ lean($\theta_j$)$\land x \in$ lean($\theta_k$)$\}$, lean($\theta_j$)$\{x : \rho, x \models \theta_j\}$ and lean($\theta_k$)$\{x : \rho, x \models \theta_k\}$.

If $\theta_i = \theta_j \lor \theta_k$, then lean($\theta_j$) = (lean($\theta_j$) \ union lean($\theta_k$)). lean($\theta_j$)$\{x : x \in$ lean($\theta_j$)$\lor x \in$ lean($\theta_k$)$\}$, lean($\theta_j$)$\{x : \rho, x \models \theta_j\}$ and lean($\theta_k$)$\{x : \rho, x \models \theta_k\}$.

If $\theta_i = X \theta_j$, then lean($\theta_i$) = ((lean($\theta_j$) - 1)$\{last\}). lean($\theta_j$)$\{x : x \neq last \land x + 1 \in$ lean($\theta_j$)$\}$, lean($\theta_j$)$\{x : \rho, x \models \theta_j\}$.

If $\theta_i = N \theta_j$, then lean($\theta_i$) = (lean($\theta_j$) - 1)$\{last\}). lean($\theta_j$)$\{x : x = last \lor x + 1 \in$ lean($\theta_j$)$\}$, lean($\theta_j$)$\{x : \rho, x \models \theta_j\}$.

Finally, we define $\text{Lean}$ variation based on function $\text{lean}(\phi)$. $\text{Lean}$ variation $\text{mso}_X(\phi)$ returns $\text{MSO}$ formula $\{\exists Q_{\theta_1}\} \ldots \{\exists Q_{\theta_n}\} \{\text{lean}(\phi)(0) \land (\forall x \land n = 1 \land x \in \text{cl}(\phi))\}$, where $n$ is the number of $U$- and $R$- subformulas $\theta_a \in \text{cl}(\phi)$, and $t_\lambda(\theta_a, x)$ is defined as follows: if $\theta_a = (\theta_j U \theta_k)$, then $t_\lambda(\theta_a, x) = (Q_{\theta_j}(x) \leftrightarrow \text{lean}(\theta_j)(x) \land \text{lean}(\theta_j)(x) \lor \text{lean}(\theta_k)(x) \lor \text{lean}(\theta_k)(x) \lor \text{lean}(\theta_k)(x) \lor \text{lean}(\theta_k)(x) \lor \text{lean}(\theta_k)(x))$. The following theorem guarantees the correctness of $\text{Lean}$ variation.

**Theorem 7.** Let $\phi$ be an $\text{LTL}_f$ formula, $\rho$ be a finite trace. Then $\rho \models \phi$ iff $\text{I}_\rho \models \text{mso}_X(\phi)$.
Proof. If \( \phi \) is a propositional atom, then \( \text{mso}_\lambda(\phi) = Q_p(0) \). It is true that \( \rho \models \phi \) iff \( \mathcal{I}_\rho \models \text{mso}_\lambda(\phi) \). If \( \phi \) is an nonatomic formula, we prove this theorem in two directions.

Suppose first that \( \rho \) satisfies \( \phi \). We expand the monadic structure \( \mathcal{I}_\rho \) with interpretations for \( Q_{\theta_1}, Q_{\theta_2}, \ldots, Q_{\theta_n} \), by setting \( Q_{\theta_a} = \{ x : \rho, x \models \theta_a \} \). Let the expanded structure be \( \mathcal{I}_\rho^{\text{mso}} \). By assumption, \( \text{mso}(\phi)(0) \) holds in \( \mathcal{I}_\rho^{\text{mso}} \). It remains to prove that \( \mathcal{I}_\rho^{\text{mso}} \models (\forall x)(\bigwedge_{n=1}^\infty t_3(\theta_a, x)) \), for each \( U \) or \( R \) subformula \( \theta_a \in \text{cl}(\phi) \).

- If \( \theta_a = (\theta_j U \theta_k) \), then \( t_3(\theta_a, x) = (\text{lean}(\theta_a)(x) \leftrightarrow (\text{lean}(\theta_k)(x) \land \text{lean}(\theta_j)(x + 1))) \). This holds, since \( \text{lean}((\theta_j U \theta_k)) = \{ x : \rho, x \models \theta_j U \theta_k \} = \{ x : \rho, x \mid \theta_k \text{ or } x \neq \text{last with } \rho, x \models \theta_j \text{ also } \rho, x + 1 \models \theta_a \} \), \( \text{lean}(\theta_j) = \{ x : \rho, x \models \theta_j \} \), and \( \text{lean}(\theta_k) = \{ x : \rho, x \models \theta_k \} \) with Lemma 1.

- If \( \theta_a = (\theta_j R \theta_k) \), then \( t_3(\theta_a, x) = (\text{lean}(\theta_a)(x) \leftrightarrow (\text{lean}(\theta_k)(x) \land ((x = \text{last}) \lor \text{lean}(\theta_j)(x + 1)))) \). This holds, since \( \text{lean}((\theta_j R \theta_k)) = \{ x : \rho, x \models \theta_j R \theta_k \} = \{ x : \rho, x \models \theta_k \text{ with } x = \text{last or } \rho, x \models \theta_j \text{ or } \rho, x + 1 \models \theta_a \} \), \( \text{lean}(\theta_j) = \{ x : \rho, x \models \theta_j \} \), and \( \text{lean}(\theta_k) = \{ x : \rho, x \models \theta_k \} \) with Lemma 1.

Assume now that \( \mathcal{I}_\rho \models \text{mso}_\lambda(\phi) \). This means that there is an expansion of \( \mathcal{I}_\rho \) with monadic interpretations \( Q_{\theta_a} \) for each element \( \theta_a \) of \( U \) or \( R \) subformulas in \( \text{cl}(\phi) \) such that this expanded structure \( \mathcal{I}_\rho^{\text{mso}} \models (\text{mso}(\phi)(0)) \land ((\forall x)(\bigwedge_{n=1}^\infty t_3(\theta_a, x))) \). If \( \phi \) is not an \( R \) or \( U \) subformula, then it has been proven by Lemma 1 that if \( x \in \text{cl}(\phi) \), then \( \rho, x \models \phi \). We now prove by induction on \( \phi \) that if \( x \in Q_{\theta_a} \), then \( \rho, x \models \phi \). Since \( \mathcal{I}_\rho^{\text{mso}} \models (\text{mso}(\phi)(0)) \), it follows that \( \rho, 0 \models \phi \).

- If \( \phi = (\theta_j U \theta_k) \), then \( t_3(\phi, x) = (\text{lean}(\phi)(x) \leftrightarrow (\text{lean}(\theta_k)(x) \land \text{lean}(\phi)(x + 1))) \). Since \( t_3(\phi) \) holds at every point \( x \) of \( \mathcal{I}_\rho^{\text{mso}} \), it follows that \( x \in \text{lean}(\phi) \) iff \( x \in \text{lean}(\theta_k) \) or \( x \neq \text{last with } x \in \text{lean}(\theta_j) \) also \( x + 1 \in \text{lean}(\phi) \). Moreover, \( \text{lean}(\phi) = Q_{\theta_a} \), where \( Q_{\theta_a} \) is the corresponding second-order predicate. Thus, by induction hypothesis \( \rho, x \models \phi \).

- If \( \phi = (\theta_j R \theta_k) \), then \( t_3(\phi, x) = (\text{lean}(\phi)(x) \leftrightarrow (\text{lean}(\theta_k)(x) \land ((x = \text{last}) \lor \text{lean}(\theta_j)(x + 1)))) \). Since \( t_3(\phi) \) holds at every point \( x \) of \( \mathcal{I}_\rho^{\text{mso}} \), it follows that \( x \in \text{lean}(\phi) \) iff \( x \in \text{lean}(\theta_k) \) with \( x = \text{last or } x \in \text{lean}(\theta_j) \) or \( x + 1 \in \text{lean}(\phi) \). Moreover, \( \text{lean}(\phi) = Q_{\theta_a} \), where \( Q_{\theta_a} \) is the corresponding second-order predicate. Thus, by induction hypothesis \( \rho, x \models \phi \).

\( \square \)

Having defined different variations of the MSO encoding, we now provide variations of the Compact MSO encoding described in Section 4.

Sloppy Formulation The formulation described in Section 4 strictly tracks the computation over each BDD \( B_q \), which we refer to as Fussy formulation. That is, for each nonterminal node \( \alpha \), both the forward computation and previous computation must be tracked. This causes a high logical complexity in the formulation. An alteration to diminish the logical complexity is to utilize a Sloppy Formulation, analogous to the Sloppy variation described above, that only tracks the forward computation. Since the previous computations are not tracked, none of the computations leading to terminal node 0 of the BDD \( B_f' \) enable an accepting condition.
To define the accepting condition of *Sloppy Formulation*, we have

$$\text{Racc}_s = \bigvee_{(\beta,v,d) \in \text{PreT}(B',0)} [x \in N_\beta \land x \in^d v] \rightarrow (x \neq 0).$$

Moreover, $\text{node}_s$ only requires $\text{PostCon}$ of node. Therefore, we have

$$\text{node}_s = \bigwedge_{1 \leq \alpha \leq \mu} \text{PostCon}.$$  

The second-order formula $\text{Rev}_s(\mathcal{F}_\psi)$ of *Sloppy Formulation* is defined as following:

$$\text{Rev}_s(\mathcal{F}_\psi) = (\exists V_0) \ldots (\exists V_{k-1})(\exists N_1) \ldots (\exists N_u)(\forall x)(\text{Rinit} \land \text{node}_s \land \text{Rterminal} \land \text{Racc}_s \land \text{roots}),$$

where $\text{Rinit}$, $\text{Rterminal}$ and $\text{roots}$ are defined as in Section 4. Therefore, let $\text{Cmso}_s(\phi)$ be the *Sloppy Formulation* of the Compact MSO encoding, we define $\text{Cmso}_s(\phi) = \text{Rev}_s(\mathcal{F}_\psi)$ asserting the truth of $\phi$ at position 0, where $\psi$ is the corresponding PLTL$_f$ formula of $\phi$, and $\mathcal{F}_\psi$ is the symbolic DFA of $\psi$. The following theorem asserts the correctness of the *Sloppy Formulation*.

**Theorem 8.** The models of formula $\text{Cmso}_s(\phi)$ are exactly the words satisfying $\phi$.

The proof here is analogous to that of the *Fussy Formulation*, where the crux is that we define the computation trace on a BDD as a sequence of sets of BDD nodes, instead of just a specific sequence of BDD nodes, see the definition of $\text{node}_s$. Such definition still keeps unambiguous formulation of the symbolic DFA since we have stronger constraints on the accepting condition, as shown in the definition of $\text{Racc}_s$.

### 5.2 Experimental Results

Having presented different optimizations, we now have 6 variations of the MSO encoding corresponding to specific optimizations, which are BNF-*Fussy-Full*, BNF-*Fussy-Lean*, NNF-*Fussy-Full*, NNF-*Fussy-Lean*, NNF-*Sloppy-Full* and NNF-*Sloppy-Lean*. Moreover, we have two variations of the Compact MSO encoding, which are *Fussy* and *Sloppy*. The experiments were divided into two parts and resulted in two major findings. First we explored the benefits of the various optimizations of MSO encoding and showed that the most effective one is that of *Lean*. Second, we aimed to answer the question whether second-order outperforms first-order in the context of LTL$_f$-to-automata translation. To do so, we compared the best performing MSO encoding and Compact MSO encoding against the FOL encoding and showed the superiority of first-order.

**Correctness** The correctness of the implementation of different encodings was evaluated by comparing the DFAs in terms of the number of states and transitions generated from each encoding. No inconsistencies were discovered.

**Lean constraint form is more effective in MSO encodings.** Figure 1 presents the number of converted instances of each variation of MSO encoding, where the upper three are all for *Lean* variations and the lower ones are for *Full* variations. The choice of BNF vs NNF did not have a major impact, and neither did the choice of *Fussy vs Sloppy*. The
one optimization that was particularly effective was that of Lean variation. The best-performing MSO encoding was BNF-Fussy-Lean. While in the Compact MSO encoding, the Fussy variation highly outperforms that of Sloppy, as shown in Figure 2.

First-order logic dominates second-order logic for LTL$_f$-to-automata translation. As presented in Figure 2, FOL encoding shows its superiority over second-order encodings performance-wise, which are MSO encoding and Compact MSO encoding. Thus, the use of second-order logic, even under sophisticated optimization, did not prove its value in terms of performance. This suggests that nevertheless second-order encoding indicates a much simpler quantificational structure which theoretically leads to more potential space to optimize, it would be useful to have first-order as a better way in the context of LTL$_f$-to-automata translation in practice.

6 Concluding Remarks

In this paper, we revisited the translation from LTL$_f$ to automata and presented new second-order encodings, MSO encoding and Compact MSO encoding with various op-
timizations. Instead of the syntax-driven translation in FOL encoding, MSO encoding provides a semantics-driven translation. Moreover, MSO encoding allows a significantly simpler quantificational structure, which requires only a block of existential second-order quantifiers, followed by a single universal first-order quantifier, while FOL encoding involves an arbitrary alternation of quantifiers. The Compact MSO encoding simplifies further the syntax of the encoding, by introducing more second-order variables. Nevertheless, empirical evaluation showed that first-order encoding, in general, outperforms the second-order encodings. This finding suggests first-order encoding as a better way for LTL\textsubscript{f}-to-automata translation.

To obtain a better understanding of the performance of second-order encoding of LTL\textsubscript{f}, we looked more into MONA. An interesting observation is that MONA is an “aggressive minimizer”: after each quantifier elimination, MONA re-minimizes the DFA under construction. Thus, the fact that the second-order encoding starts with a block of existential second-order quantifiers offers no computational advantage, as MONA eliminates the second-order quantifiers one by one, performing computationally heavy minimization after each quantifier. Therefore, a possible improvement to MONA would enable it to eliminate a whole block of quantifiers of the same type (existential or universal) in one operation, involving only one minimization. Currently, the quantifier-elimination strategy of one quantifier at a time is deeply hardwired in MONA, so the suggested improvement would require a major rewrite of the tool. We conjecture that, with such an extension of MONA, the second-order encodings would have a better performance, but this is left to future work.

Beyond the unrealized possibility of performance gained via second-order encodings, another motivation for studying such encodings is their greater expressivity. The fact that LTL\textsubscript{f} is equivalent to FOL \cite{15} shows limited expressiveness of LTL\textsubscript{f}. For this reason it is advocated in \cite{6} to use Linear Dynamic Logic (LTL\textsubscript{f}) to specify ongoing behavior. LTL\textsubscript{f} is expressively equivalent to MSO, which is more expressive than FOL. Thus, automata-theoretic reasoning for LTL\textsubscript{f}, for example, reactive synthesis \cite{8}, cannot be done via first-order encoding and requires second-order encoding. Similarly, synthesis of LTL\textsubscript{f} with incomplete information requires the usage of second-order encoding \cite{9}. We leave this too to future research.

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