Automated Verification of Epistemic Properties for General Game Playing

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Abstract
Automatically deriving properties of new games is one of the fundamental challenges for general game-playing systems, whose task is to learn to play any previously unknown game solely by being given the rules of that game. A recently developed method uses Answer Set Programming for verifying finitely-bounded temporal invariance properties against a given game description by structural induction. Addressing the new challenge posed by the recent extension of the general Game Description Language to include games with imperfect information and randomness, we extend this method to epistemic properties about games. We formally prove this extension to be correct, and we report on experiments that show its practical applicability.

1 Introduction
General Game Playing is concerned with the development of intelligent systems that learn to play previously unknown games well without human intervention. Identified as a Grand Challenge for Artificial Intelligence, this endeavour requires to combine methods from a variety of sub-disciplines, including reasoning, search, computer game playing, and learning (Pell 1993; Genesereth, Love, and Pell 2005).

Automated Theorem Proving is among the fundamental techniques in General Game Playing, where it serves two purposes (Haufe, Schiffel, and Thielscher 2012): In the first place, it can be used for the formal verification of desired properties during the game design phase. Secondly, general game-playing systems can use theorem proving to acquire knowledge of a previously unknown game, which is a prerequisite for the automatic generation of both search heuristics and an evaluation function for non-terminal positions (Kuhlmann, Dresner, and Stone 2006; Clune 2007; Schiffel and Thielscher 2007). We have shown recently how Answer Set Programming (Gelfond 2008) can be used to assist in this endeavour by the automated verification of state sequence invariants, that is, properties that may concern two or more successive game states but do not require to analyse the entire game tree (Haufe, Schiffel, and Thielscher 2012).

The recent extension of the general Game Description Language to imperfect information and randomness (Thielscher 2010) poses new challenges to both designer and player of a general game. Fundamental issues such as the knowledge of a player about its legal moves, game termination, or its opponents’ viewpoints are no longer obvious. The designer of a game must endow players with sufficient information to guarantee that the game is playable, and players themselves require reliable information about the game state to maximise their utility. Gaining this knowledge is an intricate reasoning task over the game rules, and its automation is of major assistance for the game designer and essential for a successful general game-playing program.

In this paper, we provide such an automated reasoning method for epistemic properties by fundamentally extending our proof method for state sequence invariants. To this end, we specify syntax and semantics of a formula language for the expression of epistemic game knowledge; we develop an inductive proof theory that allows to verify “positive” epistemic formulas against a given game specification without analysis of the entire game tree; we prove the correctness of our proof theory; and we demonstrate its efficiency by reporting on experiments with a practical implementation.

2 The Game Description Language GDL
GDL is based on the standard syntax and semantics of logic programming (Genesereth, Love, and Pell 2005). We follow the Prolog convention of denoting variables by upper-case letters while predicate and function symbols start with a lowercase letter. GDL is characterised by the following special keywords, two of which (random, sees) were recently added for games with randomness and imperfect information (Thielscher 2010).

| role (R) | R is a player |
| random | the random player |
| init (F) | F holds in the initial position |
| distinct (X,Y) | X and Y are syntactically different |
| true (F) | F holds in the current position |
| legal (R,M) | R can do move M in the current position |
| does (R,M) | player R does move M |
| next (F) | F holds in the next position |
| sees (R,P) | R perceives P in the next position |
| terminal | the current position is terminal |
| goal (R,N) | R gets N points in the current position |
As an example consider Fig. 1, which shows a GDL description of standard Tictactoe with the additional interesting twist that the players cannot see their opponent’s moves (Schiffel and Thielscher 2011). The names of the players and the initial position are given in lines 1–2. The moves are specified by the rules with head legal (lines 4–6): The player whose turn it is can attempt to mark a cell is valid, then cell \((M,N)\) where \(R \in \{x,o\}\) and \(Z \in \{1,2,3\}\); and \(Z \in \{x,o, b\}\), with \(b\) meaning “blank.”

![Figure 1: A GDL description of “Krieg-Tictactoe“ (omitting the clauses for termination and goal values). The game positions are encoded using the three features control\((R)\), cell\((M, N, Z)\), and tried\((M, N)\), where \(R \in \{x, o\}\); \(M, N \in \{1, 2, 3\}\); and \(Z \in \{x, o, b\}\), with \(b\) meaning “blank.”](image)

three equal markers or the board is completely filled) and goal values.

2.1 Formal Syntax and Semantics

In order to admit an unambiguous interpretation, GDL game descriptions must obey certain general syntactic restrictions. Specifically, a valid game description must be stratified (Apt, Blair, and Walker 1987) and allowed (Lloyd and Topor 1986). A further syntactic restriction ensures that only finitely many positive instances are true in this model; for details we must refer to Love et al. (2006) for space reasons. Finally, the special keywords are to be used as follows (Thielscher 2010):

- **role** only appears in facts (i.e., clauses with empty body) or in the body of clauses;
- **init** only appears as head of clauses and does not depend on any of true, legal, does, next, sees, terminal, goal;
- **true** only appears in the body of clauses;
- **does** only appears in the body of clauses, and none of legal, terminal, or goal depends on does;
- **next** and sees only appear as head of clauses.

These restrictions are imposed to ensure that a set of GDL rules can be effectively and unambiguously interpreted by a state transition system as follows. To begin with, any set of clauses determines an implicit domain-dependent set of ground symbolic expressions \(\Sigma\), like cell\((1,1,b)\), mark\((1,3)\), yourmove, etc. Game positions (i.e., states) are represented by subsets of \(\Sigma\) since they are composed of individual features. Although \(\Sigma\) itself is usually infinite, the syntactic restrictions in GDL ensure that the set of roles, the
reachable states, and the set of legal moves are always finite subsets of \( \Sigma \) (Love et al. 2006).

In order to determine the legal moves of a player in any given state, this state is encoded using the keyword \texttt{true}. More precisely, let \( S = \{ f_1, \ldots, f_k \} \) be a finite state (e.g., the derivable instances of \texttt{init (F)} at the beginning), then game description \( G \) is extended by the \( k \) facts

\[ S \cup \text{true} \triangleq \{ \text{true}(f_1), \ldots, \text{true}(f_k) \} \]

Those instances of \texttt{legal (R, M)} that are derivable from \( G \cup S \cup \text{true} \) define all legal moves \( M \) for player \( R \) in position \( S \). In addition to the encoding of the current position, determining a position update and the percepts of the players requires a \textit{joint action}, which consists of an action for each of the players. Specifically, if joint action \( A \) is such that players \( r_1, \ldots, r_k \) take moves \( a_1, \ldots, a_k \), then let

\[ A \texttt{does} \triangleq \{ \texttt{does}(r_1, a_1), \ldots, \texttt{does}(r_k, a_k) \} \]

The instances of \texttt{next (F)} that are derivable from \( G \cup A \texttt{does} \cup S \cup \text{true} \) compose the updated position; likewise, the derivable instances of \texttt{sees (R, F)} describe what a player perceives when the given joint action is done in the given position. All this is summarised in the following definition:\footnote{We omit the definition of the probability distribution over updated positions in case \texttt{random} \( \in R \), as this is irrelevant for the present paper.}

\textbf{Definition 1} \textup{(Thielscher 2010)} Let \( G \) be a GDL specification whose signature determines the set of ground terms \( \Sigma \). Let \texttt{“-”} denote entailment by the standard model (Apt, Blair, and Walker 1987) of a stratified program, then the semantics of \( G \) is the state transition system \( (R, S_{init}, T, I, u, g, y) \) given by

- \( R = \{ r : G \vdash \texttt{role(r)} \} \) \textup{(player names)};
- \( S_{init} = \{ f : G \vdash \texttt{init (f)} \} \) \textup{(initial state)};
- \( T = \{ S : G \cup S \cup \text{true} \vdash \texttt{terminal} \} \) \textup{(terminal states)};
- \( I = \{ (r, a, S) : G \cup S \cup \text{true} \vdash \texttt{legal (r, a)} \} \) \textup{(legal moves)};
- \( u(A, S) = \{ f : G \cup A \texttt{does} \cup S \cup \text{true} \vdash \texttt{next (f)} \} \)
- \( I = \{ (r, A, S, p) : G \cup A \texttt{does} \cup S \cup \text{true} \vdash \texttt{sees (r, p)} \} \) \textup{(players’ percepts)};
- \( g = \{ (r, n, S) : G \cup S \cup \text{true} \vdash \texttt{goal (r, n)} \} \) \textup{(goal values)};

for all finite subsets \( S \subseteq \Sigma \) \textup{(called states)} and functions \( A : R \rightarrow \Sigma \) \textup{(called joint actions)}.

For states \( S, S' \) and a joint action \( A \), we write \( S \xrightarrow{A} S' \) if \( (r, A(r), S) \in I \) for each role \( r; S' = u(A, S) \); and \( S \notin T \). We call \( \sigma = (S_0 \xrightarrow{A_0} S_1 \ldots S_{m-1} \xrightarrow{A_{m-1}} S_m) \) (where \( m \geq 0 \)) a sequence (of legal moves), sometimes abbreviated as \( (S_0, S_1, \ldots, S_m) \). When \( S_0 = S_{init} \), we also call \( \sigma \) a development, and we call a state \( S \) reachable if there is a development that ends in \( S \).

We will denote sequences with \( \sigma \) and developments with \( \delta \), possibly with super- or subscripts. For two sequences \( \sigma_1 = (S_0, \ldots, S_m) \) and \( \sigma_2 = (S_m, \ldots, S_{m+k}) \), we also write \( (\sigma_1, \sigma_2) \) to denote their composition \( (S_0, \ldots, S_{m+k}) \). The length of a given sequence \( \sigma = (S_0, \ldots, S_m) \) is \( m \), sometimes denoted by \( |\sigma| \); and the last state \( S_m \) of \( \sigma \) is also referred to via the notion \( \text{last}(\sigma) \).

\textbf{2.2 Indistinguishable Developments}

We conclude this introduction by recalling what the execution model for GDL entails about what players know at any stage during a game (Thielscher 2010). This is formalised with the help of a binary equivalence relation \( \sim_r \) over developments such that \( \delta_1 \sim_r \delta_2 \) if, and only if, \( \delta_1 \) and \( \delta_2 \) are indistinguishable for player \( r \).

\textbf{Proposition 1} Let \( G \) be a GDL description with semantics \((R, S_{init}, T, I, u, g, y)\). For two developments of \( G \),

\[ \delta_1 = (S_{init} \xrightarrow{A_0} S_1 \ldots \xrightarrow{A_{m-1}} S_m) \]

\[ \delta_2 = (S_{init} \xrightarrow{A'_0} S'_1 \ldots \xrightarrow{A'_m} S'_m) \]

we have \( \delta_1 \sim_r \delta_2 \) iff for each \( 0 \leq i \leq m - 1 \):

- \( \{ p : (r, A_i, S_i, p) \in I \} = \{ p : (r, A'_i, S'_i, p) \in I \} \) (that is, \( r \)’s percepts are the same), and
- \( A_i(r) = A'_i(r) \) (i.e., \( r \) always takes the same action).

The proposition implies that each player \( r \) is aware of the initial game state, which results from the fact that \( r \) gets to know the complete game description prior to game play.

\textbf{3 Formalising Epistemic Game Knowledge}

We will now specify a language which allows to formulate properties that involve finitely many successive game states and the knowledge of different agents. It extends our previous language for game-specific properties (Haufe, Schiffl, and Thielscher 2012) by a unary knowledge operator.

\textbf{3.1 Syntax}

\textbf{Definition 2} Let \( G \) be a GDL specification. The set of (epistemic) formulas (over \( G \)) is given by the following Backus-Naur form:

\[ \varphi ::= p \mid \neg \varphi \mid \varphi \land \varphi \mid \varphi \lor \varphi \mid \Diamond \varphi \mid K_r \varphi \]

where \( p \) is a ground atom over the signature of \( G \) which is different from \texttt{init} and \texttt{next} and does not depend on \texttt{does}, and \( r \) is any role of the game besides \texttt{random}.

Moreover, the degree of a formula \( \varphi \) is the maximal “nesting” of the unary \( \Diamond \)-operator in \( \varphi \).

Formulas that do not contain any knowledge operators \( K_r \) are also called knowledge-free. A formula \( \bigcirc \psi \) states that \( \psi \) is true in the next game state, and a formula \( K_r \psi \) states that agent \( r \) knows \( \psi \). For example, in Tictactoe the property “if the game has terminated, then the \( x \)-player knows that it has terminated” can be formulated via the following epistemic formula \( \rho \):

\[ \neg \texttt{terminal} \lor K_x \texttt{terminal} \]

\textbf{3.2 Semantics}

The following extends our existing semantics for knowledge-free formulas (Haufe, Schiffl, and Thielscher 2012), and we refer to this earlier work for a motivation to the non-standard treatment of state sequences and the temporal operator \( \Diamond \). The treatment of knowledge operators is motivated by a recently developed semantics for epistemic formulas over the GDL (Ruan and Thielscher 2011). Our semantics
employs the notion of an \( n \)-max sequence, which means a sequence that is either of length \( n \), or shorter and ends in a terminal state.

**Definition 3** Let \( G \) be a GDL description, \( \varphi \) a formula over \( G \) with degree \( n \), and \( \delta = (S_{\text{init}}, S_1, \ldots, S_k) \) a development. We say that \( S_k \) satisfies \( \varphi \) wrt. \( \delta \) (written \( S_k \models_{\delta} \varphi \)) if for all \( n \)-max sequences \( \sigma = (S_k, \ldots, S_{k+m}) \) \((m \leq n)\) we have that \( \sigma \models_{\delta} \varphi \) according to the following definition:

\[
\begin{align*}
\sigma \models_{\delta} p & \quad \text{iff} \quad G \cup S_k^{\text{true}} \models p \quad (p \text{ ground atom}) \\
\sigma \models_{\delta} \neg \varphi & \quad \text{iff} \quad \sigma \models_{\delta} \varphi \quad \text{(likewise for } \land, \lor) \\
\sigma \models_{\delta} \varphi \lor \psi & \quad \text{iff} \quad m = 0, \text{ or } \sigma' \models_{\delta'} \varphi \quad \text{for } \delta' = (S_{\text{init}}, S_1, \ldots, S_k, S_{k+1}) \quad \text{and } \sigma' = (S_{k+1}, \ldots, S_{k+m}) \\
\sigma \models_{\delta} \varphi \land \psi & \quad \text{iff} \quad \sigma \models_{\delta} \varphi \quad \text{and} \quad \sigma \models_{\delta} \psi \\
\sigma \models_{\delta} \forall \varphi & \quad \text{iff} \quad \sigma \models_{\delta} \varphi \quad \text{for each } \delta' \text{ s.t. } \delta \sim_r \delta' \\
\sigma \models_{\delta} \exists \varphi & \quad \text{iff} \quad \sigma \models_{\delta} \varphi \quad \text{for each } \delta' \text{ s.t. } \delta \sim_r \delta' \\
\end{align*}
\]

Since \( \sim_r \) is an equivalence relation, our semantics can be proved to satisfy the S5 properties (Fagin et al. 1995). Furthermore, since a GDL specification contains a complete description of the initial state of the game, and since each player gets to know the complete specification prior to playing the game, each player knows all that is implied in the initial game state.

**Proposition 2** Let \( \varphi \) be an epistemic formula, let \( k_f(\varphi) \) be the formula obtained from \( \varphi \) by removing each occurrence of a knowledge operator which is not in the scope of any \( \lor \), and let development \( \delta \) be such that \( \delta = (S_{\text{init}}) \). Then \( S_{\text{init}} \models \varphi \) iff \( S_{\text{init}} \models_{\delta} k_f(\varphi) \). \( \Box \)

## 4 Proof by Contradiction Using Linear Time

In our earlier work we have established an induction method to reliably show the validity\(^3\) of arbitrary knowledge-free formulas \( \varphi \) with respect to a given game description (Haufe, Schift, and Thielrocher 2012). The base case proves that \( S_{\text{init}} \models_{\delta} \varphi \) holds for the initial development \( \delta = (S_{\text{init}}) \). The induction step shows that any development \( \delta = (S_{\text{init}}, S_1, \ldots, S_k) \) s.t. \( S_k \models_{\delta} \varphi \) implies \( S_{k+1} \models_{\delta'} \varphi \) for each of its prolongations \( \delta' = (S_{\text{init}}, S_1, \ldots, S_k, S_{k+1}) \). We have constructed two answer set programs\(^4\) (ASP), one for the base case and one for the induction step, which can be fed into any answer-set solver to prove the validity of \( \varphi \) by proving inconsistency of the two programs. This has been achieved by creating a correspondence between counter examples of \( \varphi \) (which amount to be finite state sequences) and answer sets of the programs, using a linear temporal extension of the GDL rules and a formula encoding which states that the formula is violated.

In this paper, we will extend the summarised method to (a subclass of) epistemic formulas. To this end, we first define the class of positive-knowledge formulas whose counter examples correspond to multiple state sequences and can hence succinctly be represented as answer sets using a linear time structure. As an intermediate step, we characterise potential counter examples of positive-knowledge formulas by a structure called sequence mapping, define an alternative formula semantics based on this structure and show its equivalence to the original formula semantics from Definition 3. In a later section, this equivalence will provide the link from developments that violate positive-knowledge formulas according to the original semantics to answer sets for the (later introduced) answer set programs.

### 4.1 Positive-Knowledge Formulas

Example formula \( \rho = \neg \text{terminal} \lor K_x \text{terminal} \) is not valid with respect to the GDL description of Krieg-Tictactoe. If player \( o \) places a marker that completes a line of markers \( o \) (which yields a terminal state) while having another option that does not (which yields a non-terminal state), player \( x \) must consider both successor states possible afterwards and does hence not know about termination. More precisely, we have \( \text{last}(\delta_5) \not\models_{\rho} \) for a game development \( \delta_5 = (S_{\text{init}} \xrightarrow{A_0} S_1 \ldots \xrightarrow{A_5} S_6) \), say, that is such that \( A_0(o) = A_1(x) = A_2(o) = \ldots = A_5(x) = \text{noop} \) and \( A_0(x) = \text{mark}(1,1) \), \( A_1(o) = \text{mark}(3,1) \), \( A_2(x) = \text{mark}(1,2) \), \( A_3(o) = \text{mark}(3,2) \), \( A_4(x) = \text{mark}(2,1) \) \( A_5(o) = \text{mark}(3,3) \). State \( S_6 \) is terminal, hence the first disjunct \( \neg \text{terminal} \) of \( \rho \) is violated. The second disjunct \( K_x \text{terminal} \) is violated since there is a development \( \delta'_5 \) such that \( \delta_5 \sim_r \delta'_5 \) which ends in a non-terminal state, e.g. the one resulting from \( \delta_5 \) by changing \( A_5(o) = \text{mark}(3,3) \) to \( A_5(o) = \text{mark}(1,3) \). We see that the violation of \( \rho \) is uniquely characterised by the two developments \( \delta_5 \) and \( \delta'_5 \). On the contrary, a counter example for formula \( \neg \text{terminal} \lor \neg K_x \text{terminal} \) can not appropriately be characterised by two sequences, as it requires that \( K_x \text{terminal} \) be true and hence that all developments \( \delta' \) such that \( \delta_5 \sim_r \delta' \) satisfy \( \text{last}(\delta') \models_{\rho} \) terminal. The following definition generalises this motivation to a class of formulas whose counter examples can compactly be represented by single state sequences. Intuitively, it comprises all formulas which do not include formulations of what players do not know.

**Definition 4** An epistemic formula \( \varphi \) is called positive-knowledge formula if \( \neg \) for any role \( r \) in \( \varphi \) is within the scope of negation, \( \neg \).

It is also possible to allow \( K_x \) in the scope of even number of negations. At a later point, however, these double negations would have to be “compiled away,” which is why we decided in favour of this more restrictive definition, for the sake of simplicity.

### 4.2 Sequence Mappings

We will now define a structure for the formal representation of single state sequences which amount to be potential counter examples for positive-knowledge formulas. This will be done by assigning one sequence to the formula itself,
Definition 5  Let \( \varphi \) be a positive-knowledge formula with \( n \) occurrences of knowledge operators, uniquely characterised by their positions \( \pi_1, \ldots, \pi_n \) in the syntax tree of \( \varphi \). A set of views \( V_\varphi \), denoted \( V_\varphi \), is a minimal set such that:

- \( v_0 \in V_\varphi \), and
- there is a bijection from \( \{v_1, \ldots, v_n\} \) to \( \{\pi_1, \ldots, \pi_n\} \)

and \( \{v_1, \ldots, v_n\} \subseteq V_\varphi \).

For each positive-knowledge formula \( \varphi \), we consider the set of views \( V_\varphi \) for \( \varphi \) as uniquely given. A sequence mapping for \( \varphi \) wrt. a game development \( \delta \) is then given as a function \( M \) from views \( v \in V_\varphi \) to developments over a GDL specification that satisfies the following properties:

- Development \( M(v_0) \) is of the form \((\delta, \sigma)\), where \( \sigma \) is a \( \text{deg}(\varphi) \)-max sequence. That is to say, the given game development \( \delta \) is prolonged by a sequence \( \sigma \) of appropriate length which potentially violates \( \varphi \).
- For each occurrence of a knowledge operator \( K_r \) in \( \varphi \) and its corresponding view \( v_i \) (\( v_i \neq v_0 \)), and for the view \( v_j \) corresponding to the knowledge operator nearest to \( K_r \) towards the root of the syntax tree of \( \varphi \) (in case it exists) or such that \( v_j = v_0 \) (otherwise):
  - development \( M(v_i) \) is of the form \((\delta', \sigma')\), where \( |\delta'| = |\delta| \) and \( \sigma' \) is a \( \text{deg}(\varphi) \)-max sequence, and
  - \( M(v_j) \) “relates appropriately” to \( M(v_i) \) according to \( \sim_r \).

For the benefit of the reader, we refrain from a formal characterisation of the term “relates appropriately” and, rather than formalising all subsequent technical details, concentrate on showing the main ideas with the example formula \( \rho = \neg \text{terminal} \lor K_r \text{terminal} \) in the remainder of the paper (a comprehensive presentation with all details is part of a doctoral dissertation (Haufe 2012)). The set of views for \( \rho \) can be given as \( \{v_0, v_1\} \) (where \( v_0 \) corresponds to \( \rho \) and \( v_1 \) corresponds to \( K_r \)). Using the above-mentioned developments \( \delta_p \) and \( \delta'_p \), one possible sequence mapping for \( \rho \) wrt. \( \delta_p \) is the function \( M_\rho \) such that \( M_\rho(v_0) = \delta_p \) and \( M_\rho(v_1) = \delta'_p \). Both sequences “relate appropriately” according to \( \sim_x \), e.g. we have \( M_\rho(v_0) \sim_x M_\rho(v_1) \).

4.3 An Equivalent Semantics Over Sequence Mappings for Positive-Knowledge Formulas

The first step towards the establishment of our induction proof method for the validity of positive-knowledge formulas via answer set programs is the construction of an equivalent formula semantics over sequence mappings. A sequence mapping \( M \) for formula \( \varphi \) wrt. development \( \delta \)

\[ M(v_0) \]

and by assigning one further sequence to each occurrence of a knowledge operator (handling multiple occurrences of the same knowledge operator separately). Sequences will be assigned using the following set of names.

Definition 6  Let \( G \) be a GDL specification, \( \varphi \) be a positive-knowledge formula, \( \delta \) be a development, and \( M \) be a sequence mapping for \( \varphi \) wrt. \( \delta \). We say that \( M \) satisfies \( \varphi \), denoted \( M \vDash \varphi \), if \((M, |\delta|, v_0) \vDash \varphi \) holds according to the following definition (where \( 0 \leq k \leq \text{deg}(\varphi) \)):

\[ (M, k, v) \vDash \varphi \iff G \cup S_{k+1} \cup p \vdash \varphi (p \text{ ground atom}), \]

where \( M(v) = (S_{\text{init}}, S_1, \ldots, S_k, \ldots, S_{k+m}) \)

\[ (M, k, v) \vDash \neg \varphi \iff (M, k, v) \not\vDash \psi \] (likewise for \( \land, \lor \))

\[ (M, k, v) \vDash \varphi \land \psi \iff m = 0, \text{ or } (M, k+1, v) \vDash \psi, \]

where \( M(v) = (S_{\text{init}}, S_1, \ldots, S_k, \ldots, S_{k+m}) \)

\[ (M, k, v) \vDash K_r \psi \iff (M, k, v') \vDash \psi \] (where view \( v' \) corresponds to \( K_r \)).

Note that all cases different from \( K_r \psi \) exactly correspond to their counterparts in Definition 3. Case \( K_r \psi \), on the contrary, does not explicitly incorporate \( \sim_r \), anymore. Hence, \((M, k, v) \vDash K_r \psi \) for a sequence mapping \( M \) such that \( M(v) = (\delta, \sigma) \) does not generally imply the correspondent \( \sigma \not\vDash K_r \psi \), as this would require \( \text{last}(\delta') \not\vDash \psi \) to hold for all developments \( \delta' \) such that \( \sigma \sim_r \delta' \), whereas \((M, k, v) \vDash K_r \psi \) only provides that correspondence for one such development \( \delta' \). However, following the spirit of a sequence mapping as structure for potential counter examples of a positive-knowledge formula, there is a correspondence between the existence of a sequence mapping \( M \) such that \( (M, k, v) \not\vDash K_r \psi \) and the existence of a sequence \( \sigma \) such that \( \sigma \not\vDash K_r \psi \). This is more generally stated with the following theorem.

Theorem 1  Let \( \varphi \) be a positive-knowledge formula, \( \delta \) be a development, and \( \sigma \) be a \( \text{deg}(\varphi) \)-max sequence starting at \( \text{last}(\delta) \). Then the following are equivalent.

- \( \sigma \not\vDash \varphi \)
- There is a sequence mapping \( M \) for \( \varphi \) wrt. \( \delta \) such that \( M(v_0) = (\delta, \sigma) \) and \( M \not\vDash \varphi \).

Proof (sketch): Induction on the structure of \( \varphi \). The base case considers \( \varphi \) arbitrarily knowledge-free (possibly containing negation). This allows to neglect negation in the induction step (as \( \varphi \) is a positive-knowledge formula), and single sequences can be related with single sequence-mappings in all remaining cases.

5 Encoding Epistemic Game Knowledge

In this section we provide the second step towards the establishment of our induction proof method. It is concerned with the construction of a stratified logic program over a GDL specification which allows to infer whether a given formula is entailed with respect to a given sequence mapping solely via the unique standard model of the constructed program. To this end, we show how to extend the game rules to account for time and different views, how to encode a particular sequence mapping into these extended game rules,
and how to further encode a particular formula. We prove that the resulting program admits a unique standard model containing a special formula name token if, and only if, the formula is true with respect to the encoded sequence mapping. This result will be necessary to relate arbitrary sequence mappings to answer sets of the (later introduced) answer set programs for the induction proof method.

5.1 Epistemic Temporal GDL Extension

A sequence mapping interprets each subformula of a formula $\varphi$ with regard to the sequence associated with the nearest knowledge operator towards the root of the syntax tree of $\varphi$, or associated with $\varphi$ in case this operator does not exist. Accordingly, an encoding into the game description will require one “copy” of the description for each view. Moreover, as we consider formulas with possible occurrences of operator $\bigcirc$ and hence with a finite lookahead in time, we also need a temporal extension (Thielscher 2009).

Definition 7 For game $G$ let $G^{\leq n}(v) \equiv \bigcup_{0 \leq i \leq n} G_i(v)$, where each $G_i(v)$ is constructed by

- omitting all clauses from $G$ with head $\text{init}$;
- replacing each occurrence of
  - $\text{next}(f)$ by $\text{true}(f, i + 1, v)$,
  - $\text{sees}(r, s)$ by $\text{sees}(r, s, i + 1, v)$, and
  - $p(t_1, \ldots, t_n)$ by $p(t_1, \ldots, t_n, i, v)$, for each predicate $p$ s.t. $p \neq \text{next}$ and $p \neq \text{sees}$.

Let $\varphi$ be a formula with degree $n = \text{deg}(\varphi)$. The epistemic temporal extension of $G$ wrt. $\varphi$, denoted $G_\varphi$, is defined as

$$G_\varphi \equiv \bigcup_{v \in V_\varphi} G^{\leq n}(v).$$

Put in words, an epistemic temporal extension of $G$ wrt. $\varphi$ is composed of a temporal extension of $G$ up to depth $\text{deg}(\varphi)$ for each of the views of $\varphi$. The resulting program can be simplified by omitting the time and view argument in any atom over a predicate symbol that does not depend on any GDL keyword.

As an example for the epistemic temporal GDL extension, let $G$ be the clauses in Figure 1, and reconsider example formula $\rho = \neg \text{terminal} \lor K_1 \text{terminal}$ with the views $v_0$ and $v_1$. Then for clause 19, the following clause is in the epistemic temporal extension $G_\rho$ of $G$ for $v_0$ (and similarly for $v_1$):

$$\text{sees}(R, \text{yourmove}, 1, v_0) : -$$
$$\text{not validmove}(0, v_0),$$
$$\text{true}(\text{control}(R), 0, v_0).$$

5.2 Sequence-Mapping and Formula Encoding

A sequence mapping $M$ for a formula $\varphi$ wrt. a development of length $k$ is encoded to a program which solely consists of facts. For each view $v \in V_\varphi$ and the corresponding development $M(v) = (S_{\text{init}}, S_1, \ldots, S_k)$, it encodes the subsequence starting at $S_k$, i.e., all prefixes of length $k$ are omitted. This allows to use the same finite lookahead $\text{deg}(\varphi)$ in the epistemic temporal GDL extension for each sequence mapping, and is especially important for the later construction of the induction-step program which will encode multiple sequence mappings “at once.”

Definition 8 Let $M$ be a sequence mapping for a formula $\varphi$ wrt. a development of length $k$. The encoding of $M$, denoted $\text{Enc}(M)$, is a program which consists of the following facts for each view $v \in V_\varphi$ and its corresponding development $M(v) = S_{\text{init}} \xrightarrow{A_0} S_1 \cdots \xrightarrow{A_{k-1}} S_k \cdots \xrightarrow{A_{k+m-1}} S_{k+m}$ (where $m \geq 0$):

- for all $f \in S_k$: $\text{true}(f, 0, v) \in \text{Enc}(M)$; and
- for all players $r$, and for all $i < m$: if $r$ takes move $a$ in joint action $A_{k+i}$, then $\text{does}(r, a, i, v) \in \text{Enc}(M)$.

We have shown recently (Haufe, Schiffel, and Thielscher 2012) how arbitrary knowledge-free formulas $\varphi$ can be encoded into a temporally extended game description and a sequence encoding using a finite set of stratified clauses $\text{Enc}(\varphi)$. In case $\varphi$ is a ground atom, $\text{Enc}(\varphi)$ solely contains a clause whose head is a unique name predicate of arity 0, denoted with $\text{name}_\varphi$, and whose clause body refers to a variant of this ground atom which is extended by a time argument. In case $\varphi$ is a composed formula, $\text{Enc}(\varphi)$ contains a clause, again with unique head $\text{name}_\varphi$, whose body recursively resolves the toplevel connective of $\varphi$ by referring to the unique name predicates of its respective subformulas, and $\text{Enc}(\varphi)$ additionally contains an encoding of these subformulas.

This process can easily be extended to the knowledge setting by adding the current view as further argument to the encoding of ground atoms, we omit the formal specification here. Example formula $\rho = \neg \text{terminal} \lor K_1 \text{terminal}$ can be encoded as $\text{Enc}(\rho)$ such that $\text{name}_\rho = \text{rho}$ as follows (removing some indirections in the generated clauses):

$$\text{rho} :- \text{not terminal}(0, v_0).$$
$$\text{rho} :- \text{terminal}(0, v_1).$$

5.3 Linking Sequence-Mapping Semantics and Logic-Program Encoding

The following theorem shows that a formula $\varphi$ is violated by some sequence mapping if, and only if, the corresponding epistemic temporal GDL extension for $\varphi$ together with an encoding of that sequence mapping and an encoding of $\varphi$ does not contain the unique name token $\text{name}_\varphi$.

Theorem 2 Let $G$ be a GDL description, $\varphi$ be a positive-knowledge formula, $\delta = (S_{\text{init}}, S_1, \ldots, S_k)$ be a development and $M$ be a sequence mapping for $\varphi$ wrt. $\delta$. Then the following are equivalent:

- $M \not\models \varphi$
- $G_\varphi \cup \text{Enc}(M) \cup \text{Enc}(\varphi) \not\models \text{name}_\varphi$.

Proof (sketch): Induction on the structure of $\varphi$, using the corresponding Theorem 14 for knowledge-free formulas in our earlier work (Haufe, Schiffel, and Thielscher 2012) together with the observation that $G_\varphi \cup \text{Enc}(M) \cup \text{Enc}(\varphi)$ is composed of independent subprograms which involve exactly one view $v \in V_\varphi$. $\square$
6 Proving Epistemic Game Knowledge

We are now ready to extend our recent induction technique (Haufe, Schiffel, and Thielscher 2012) to the general case of epistemic game knowledge. As mentioned before, for each formula which is to be proved, it relies on the construction of two answer set programs, one to establish the base case and one to establish the induction step. In case both answer set programs are inconsistent, the formula is known to be true in each reachable state of the game. In this section, we show how to specify these two programs, and prove the correctness of the extended method. We will again refrain from the technical details and show the ideas for program construction wrt. example formula $\rho = \neg \text{terminal} \lor K_e \text{terminal}$.

6.1 Base Case Program

The answer set program for the base case proof of $\rho$ is given as

$$P_{\rho}^{\text{base}}(G) = G_\rho \cup \text{Enc}(S_{\text{init}}) \cup \text{Enc}(\rho) \cup \{ :- \text{ rho.} \}.$$  

$G_\rho = G_{\leq 0}(v_0) \land G_{\leq 0}(v_1)$ is the epistemic temporal extension of the clauses from Figure 1. $\text{Enc}(S_{\text{init}})$ is an encoding for the initial state in both views $v_0$ and $v_1$, i.e., $\text{Enc}(S_{\text{init}}) = \{ \text{true}(f, 0, v) : f \in S_{\text{init}}, v \in \{v_1, v_2\} \}$. Program $G_\rho \cup \text{Enc}(S_{\text{init}})$ admits a unique answer set which corresponds to the (also unique) sequence mapping $M$ for $\rho$ which is wrt. the initial development $(S_{\text{init}})$. $\text{Enc}(\rho)$ is the encoding for $\rho$ as specified in (1), enriched with the constraint $:- \text{ rho.}$, which formulates that $\rho$ should not be true. The addition of $\text{Enc}(\rho) \cup \{ :- \text{ rho.} \}$ to $G_\rho \cup \text{Enc}(S_{\text{init}})$ (which yields $P_{\rho}^{\text{base}}(G)$) ensures that the unique answer set is rejected in case $M$ satisfies $\rho$ and hence establishes the correspondence between answer set and counter example of $\rho$ in the initial state. $P_{\rho}^{\text{base}}(G)$ is inconsistent in this particular example, as the clauses $G_{\leq 0}(v_0)$ and $G_{\leq 0}(v_1)$ coincide up to renaming of the view argument but are forced to deviate regarding terminal by $\text{Enc}(\rho) \cup \{ :- \text{ rho.} \}$.

6.2 Induction Step Program

The answer set program for the induction step proof of $\rho$ is given as

$$P_{\rho}^{\text{ind}}(G) = G_{\leq 0} \cup S_{\text{gen}} \cup P_{\rho}^{\text{legal}} \cup \text{Enc}(\rho) \cup \{ :- \text{ not rho.} \} \cup \text{Enc}(\overline{\rho}) \cup \{ :- \text{ nxt_rhoo.} \}.$$  

The components of $P_{\rho}^{\text{ind}}(G)$ are explained in the following.

Epistemic Temporal GDL Extension $G_{\rho}^{\text{epi}}$: Compared to the base case program, the extension of the game rules involves a further time step: $G_{\rho} = G_{\leq 1}(v_0) \land G_{\leq 1}(v_1)$.

State Generator $S_{\text{gen}}$: Instead of the initial state encoding $\text{Enc}(S_{\text{init}})$ in both views $v_0$ and $v_1$ from $\rho$, we need a “state generator” program whose answer sets correspond exactly to pairs of states $(S_0, S_1)$ such that 1) $S_0$ is the last state of some development $\delta_0$ for view $v_0$ and 2) $S_1$ is the last state of some development $\delta_1$ for view $v_1$ such that $\delta_0 \sim_e \delta_1$. However, obtaining this information requires a full state space search in general, which is not feasible in interesting games. This motivates to apply easily obtainable overestimations of the reachable states and the relation $\sim_e$, which do not influence the (yet to be established) correctness of the method. The simplest approximation is the program

$$0 \{ \text{true}(f, 0, v) : f \in FDom, v \in \{v_0, v_1\} \},$$  

which, for the finite domain of all fluents $FDom^k$, generates all possible combinations $(S_0, S_1)$ of (not necessarily reachable) states $S_0$ for view $v_0$ and $S_1$ for $v_1$. Better approximations of the reachable states, e.g. the restriction to those which contain exactly one instance of $\text{cell}(x, y, c)$ for each coordinate pair $(x, y)$, can be obtained automatically from the game description by preceding proofs of knowledge-free formulas (Haufe, Schiffel, and Thielscher 2012). Some restrictions on the estimation of $\sim_e$ are obtained by the encoding of the induction hypothesis mentioned below, and we give directions to further restrictions in Section 6.4.

State Sequence Encoding $P_{\rho}^{\text{legal}}$: Since the epistemic temporal GDL extension $G_{\rho}^{\text{epi}}$ involves successive game states, we need an additional encoding which ensures that each player $r$ performs one legal move out of the finite set $ADom$ of moves in each non-terminal state wrt. both views $v \in \{v_0, v_1\}$ of $\rho$:

$$1 \{ \text{does}(r, a, 0, v) : a \in ADom \}.$$  

$$:- \text{ does}(r, A, 0, v), \text{ not legal}(r, A, 0, v).$$  

Moreover, the clause sets for $v_0$ and $v_1$ need to be related according to $\sim_e$, that is,

- The actions of player $x$ in both views coincide at time step 0:
  $$:- \text{ does}(x, A, 0, v_0), \text{ not does}(x, A, 0, v_1).$$  
  $$:- \text{ does}(x, A, 0, v_1), \text{ not does}(x, A, 0, v_0).$$  

- The percepts of player $x$ in both views coincide at time step 1:
  $$:- \text{ sees}(x, S, 1, v_0), \text{ not sees}(x, S, 1, v_1).$$  
  $$:- \text{ sees}(x, S, 1, v_1), \text{ not sees}(x, S, 1, v_0).$$

Each sequence mapping $M$ of $\rho$ wrt. a development of length $k$ corresponds to an answer set of program $G_{\rho} \cup S_{\text{gen}} \cup P_{\rho}^{\text{legal}}$, where, similarly to Definition 8, the prefixes of length $k$ are not represented. There are additional answer sets due to the overestimation of the reachable states and the relation $\sim_e$ in the state generator $S_{\text{gen}}$.

Formula Encoding $\text{Enc}(\rho) \cup \text{Enc}(\overline{\rho})$ and Constraints:

To encode the induction hypothesis, we use the encoding $\text{Enc}(\rho)$ for $\rho$ from (1) together with the constraint

---

In our previous work (Haufe, Schiffel, and Thielscher 2012) we describe in detail a method for automatically computing finite domain sets like $FDom$ and $ADom$ (introduced later).
Satisfiability of \( P \) \( \text{wrt.} \) \( \) implies violation of \( \phi \).

Consider a GDL specification over an arbitrary positive-knowledge formulas \( \phi \). In particular, to the base case, deriving existence of a sequence mapping corresponding to the answer set (if it exists). Additionally, we encode that \( \rho \) is not true in a direct successor state of \( S \):

\[
\text{:-} \text{nxt}_\rho. \\
\text{nxt}_\rho \text{ :- not terminal}(1,v0). \\
\text{nxt}_\rho \text{ :- terminal}(1,v1).
\]

It differs from the encoding for \( \rho \) in the increased time step and in the constraint \( \text{:-} \text{nxt}_\rho \), which forces \( \rho \) to be false at time step 1. \( \text{Enc}(\rho) \) uses the same views as \( \text{Enc}(\rho) \) and hence restricts the before-mentioned overestimation of \( \sim_r \), in this example by removing all answer sets which correspond to state pairs \((S_0, S_1)\) where \( S_0 \) is terminal and \( S_1 \) is non-terminal.

The addition of the formula encodings and the constraints \( \text{:- not} \ \rho \text{and} \ \text{:- nxt}_\rho \) to program \( G_{\rho} \cup S_{\text{gen}} \cup P_{\text{legal}} \cup \bigcirc_{\rho} \) (which yields \( P_{\rho}(G) \)) ensures that each answer set is rejected which represents a sequence mapping that violates \( \rho \) in the current state or satisfies \( \rho \) in the next state. This assures that each remaining answer set either corresponds to a counter example of the induction step or occurs due to overestimation in the state generator. \( P_{\rho}(G) \) admits an answer set in this particular example, which can be argued using the constructed sequence mapping \( M_\rho \) from the end of Section 4.2. As mentioned before, \( M_\rho \) violates \( \rho \) at depth \( |\delta| = 6 \). Additionally, \( M_\rho \) satisfies \( \rho \) when interpreted at depth \( 5 \) as state \( S_3 \) is non-terminal). Hence, \( P_{\rho}(G) \) admits an answer set which represents \( M_\rho \) with omitted prefixes of length 5.

6.3 Correctness

We believe that the general idea of the correctness proof for our method becomes clear without explicit representation of the two general answer set programs \( P_{\rho}(G) \) and \( P_{\bar{\rho}}(G) \) in Section 6.2 for arbitrary positive-knowledge formulas \( \phi \), which is why we now state the result.

Theorem 3 Consider a GDL specification \( G \) and a positive-knowledge formula \( \phi \) over \( G \). If both \( P_{\phi}(G) \) and \( P_{\bar{\phi}}(G) \) are inconsistent, then for all developments \( \delta = (S_{\text{init}} \stackrel{A_1}{\rightarrow} S_1 \ldots \stackrel{A_{k-1}}{\rightarrow} S_k) \) we have \( S_k \vDash_\delta \phi \).

Proof (sketch): Induction on \( k \), similar to the proof of the corresponding Theorem 17 for knowledge-free formulas in our earlier work (Haufe, Schiffel, and Thiellescher 2012).

Base case: Violation of \( \phi \) by the development \( \delta = (S_{\text{init}}) \) implies violation of \( \phi \) via some sequence mapping for \( \phi \) wrt. \( \delta \) (by Theorem 1), in turn implying satisfiability of \( P_{\bar{\phi}}(G) \) (by Theorem 2).

Induction step: Satisfiability of \( P_{\bar{\phi}}(G) \) is obtained similarly to the base case, deriving existence of a sequence mapping \( M \) for formula \( \bigcirc \phi \) wrt. an arbitrary development \( \delta \) using the following two arguments.

- The standard model of program \( G_{\bigcirc \phi} \cup \text{Enc}(M) \) is also an answer set of program \( G_{\bigcirc \phi} \cup S_{\text{gen}} \cup P_{\text{legal}} \).
- The induction hypothesis \( \text{last}(\delta) \vDash_\delta \phi \) implies (by Theorem 1 and the coinciding view structures of \( \phi \) and \( \bigcirc \phi \)) that also \( M \) satisfies \( \phi \) and hence that adding \( \text{Enc}(\phi) \cup \{ \text{:- not} \ \rho \} \) retains consistency.

6.4 Towards Completeness

Assume a state generator \( S_{\text{gen}} \) for the induction step program \( P_{\rho}(G) \) which does not rely on overestimation, i.e., which is such that each of its answer sets corresponds to a sequence mapping satisfying \( \phi \) and violating \( \bigcirc \phi \). Calling such a state generator accurate, we get the following completeness result.

Theorem 4 Let \( G \) be a GDL specification, let \( \phi \) be a positive-knowledge formula over \( G \), and let \( P_{\phi}(G) \) be constructed over an accurate state generator. If for all developments \( \delta = S_{\text{init}} \stackrel{A_0}{\rightarrow} S_1 \ldots \stackrel{A_{k-1}}{\rightarrow} S_k \) we have \( S_k \vDash_\delta \phi \), then \( P_{\phi}(G) \) and \( P_{\bar{\phi}}(G) \) are inconsistent.

Proof (sketch): Consistency of one of the programs \( P_{\phi}(G) \) and \( P_{\bar{\phi}}(G) \) allows to conclude existence of a sequence mapping \( M \) which violates \( \phi \) (by Theorem 2), in turn implying existence of a development \( \delta \) which ends in some state \( S_k \) that violates \( \phi \) (by Theorem 1).

One of the main reasons for the practicability of our approach is the use of induction. It allows to verify formulas without having to traverse the usually vast state space of a game by taking advantage of the comparably compact structure of game rules formulated in the game description language. The information which is necessary to construct an accurate state generator, however, can in turn only be obtained by a full state space traversal in the worst case. Hence, exploiting the advantage of induction inherently requires to apply quickly obtainable overestimations of reachable states and the relation \( \sim_r \), which possibly introduces counter examples for actually valid formulas and thus causes the loss of completeness in our approach.

Nevertheless, the accuracy of state generators can be increased by the generalisation of techniques where reachable states are approximated with increasing accuracy by adding encodings of previously proved formulas (together with constraints that require them to be true) to subsequent answer set programs (Haufe, Schiffel, and Thiellescher 2012). Similarly, the estimation of the accessibility relation can be made more precise. For example, having proved that player \( x \) exactly knows his legal moves allows to encode this information in subsequent answer set programs by adding the clauses:

\( \text{:- legal}(x,A,0,v0), \text{not legal}(x,A,0,v1) \).

\( \text{:- legal}(x,A,0,v1), \text{not legal}(x,A,0,v0) \).

As a result, each answer set violating this property is now dismissed, making the method “more complete”.

7 Experimental Results

To show the practical applicability of our proof method, we have extended our implementation for proving knowledge-free properties (Haufe, Schiffel, and Thiellescher 2012). It uses the general game-playing system FLUXPLAYER (Schiffel and Thiellescher 2007) for program generation and tools.
from POTASSCO (Gebser et al. 2011) to solve them. We have performed automated proofs with a selection of general epistemic properties in a variety of games, including Krieg-Tictactoe, a simple card game (Thielscher 2010), the famous Monty Hall problem (Rosenhouse 2009; Thielscher 2011), and all incomplete-information games\footnote{All game descriptions can be found at ggpserver.general-game-playing.de/public/show_games.jsp.} from the 1st German Open in General Game Playing\footnote{www.tzi.de/kissmann/ggp/go-ggp}, which recently featured the first ever track on incomplete-information games in a GGP competition.

### 7.1 Property Categories

To each of the aforementioned incomplete-information games, we have attempted proofs for all formulas from the following three categories, each referring to an arbitrarily chosen player \( r \) different from \texttt{random} (the other players yield similar results and are hence omitted here).

**Knows Terminal** We formulate that player \( r \) knows whether the game has terminated with the formula

\[
K_r \text{terminal} \lor K_r \neg \text{terminal}.
\]

**Knows Legals** For all non-random players \( r' \), we formulate that player \( r \) knows which of the moves of \( r' \) are legal with the formula

\[
\bigwedge_{a \in \text{ADom}} (K_r \text{legal}(r', a) \lor K_r \neg \text{legal}(r', a)).
\]

**Knows Goals** For all non-random players \( r' \) and the set \( GV \) of goal values that occur in the game description, we formulate that \( r \) knows the goal values of \( r' \) in a terminal state with the formula

\[
\text{terminal} \supset ( \bigwedge_{g \in GV} (K_r \text{goal}(r', g) \lor K_r \neg \text{goal}(r', g))).
\]

### 7.2 Proof Procedure

Each newly considered game takes a few seconds for the initialisation of FLUXPLAYER (which includes the calculation of the domains \( F\text{Dom} \) and \( A\text{Dom} \)). Afterwards, we attempt to obtain a restrictive state generator for the induction step programs by proving automatically generated knowledge-free formulas for properties such as the uniqueness of cell content, this process takes time in the range of 0.1 seconds (Krieg-Tictactoe) to 4.6 seconds (Backgammon). We then attempt separate proofs of all formulas from the three introduced categories. Each attempt is done according to the following scheme, the results can be found in Figure 2. Let \( \varphi \) be any formula of the mentioned categories which is to be proved.

**Phase 1** \( \varphi \) does not contain any \( \bigcirc \), hence according to Proposition 2, omitting all knowledge operators from \( \varphi \) yields a formula \( k_{f_0}(\varphi) \) which is equivalent to \( \varphi \) with respect to entailment in the initial state of the game. Since \( k_{f_0}(\varphi) \) is a tautology, the base-case proof can completely be omitted (it will always prove \( \varphi \) in the initial state).

**Phase 2** We can also disprove \( \varphi \) by finding a consistent answer set program \( P_{\bigcirc}^t(G) \) for some \( t \in \mathbb{N} \). Consistency allows to conclude that there is a development \( \delta \) of length \( t \) such that \( \text{last}(\delta) \not\models_{\delta} \varphi \) (by Theorem 4).

Hence, we directly attempt the induction-step proof with program \( P_{\bigcirc}^t(G) \). If it is successful (i.e., if \( P_{\bigcirc}^t(G) \) is inconsistent), then the considered formula is valid, and we skip Phase 2. If, on the other hand, \( P_{\bigcirc}^t(G) \) admits an answer set, the validity of \( \varphi \) is still unknown, as the answer set does not necessarily correspond to a sequence mapping due to the overestimation in the state generator \( S_{\text{est}} \). In that case we move to Phase 2.

**Phase 2** We can also disprove \( \varphi \) by finding a consistent answer set program \( P_{\bigcirc}^t(G) \) for some \( t \) \( \geq 1 \), which causes each player to know the following cases arises:

- A time limit of 20 seconds is reached. We stop the process, the validity of the formula is still unknown. For some \( t \geq 1 \), the last proof attempt for the base case has been on \( \bigcirc^t \varphi \), and we indicate this \( t \) in Figure 2.
- We obtain an answer set for \( P_{\bigcirc}^t(G) \) for some \( t \geq 1 \), in which case \( \varphi \) is invalid, and \( t \) is again indicated in Figure 2.

### 7.3 Interpretation of the Results

Formula \( K_r \text{terminal} \lor K_r \neg \text{terminal} \) (category Knows Terminal) can be disproved in Krieg-Tictactoe in depth 6 of the game tree first. Intuitively, this is due to the passed turn information via \texttt{sees} in the GDL specification of Krieg-Tictactoe in Figure 1, which causes each player to know the amount of currently placed pieces after each development of the game. Hence, player \( x \) first considers both a non-terminal and a terminal state possible after at least 6 joint actions, when player \( o \) could have completed a line.

Each formula \( \varphi_t \) in category Knows Terminal has an associated set of views \( V_{\varphi_t} \) of size \(|V_{\varphi_t}| = 3\), whereas each formula \( \varphi_t \) in category Knows Legals is such that \(|V_{\varphi_t}| = 2 \times |A\text{Dom}| + 1 \). This implies that the size of an answer set program which is generated for \( \varphi_t \) exceeds the size of the corresponding answer set program for \( \varphi_t \) by factor \(|A\text{Dom}|\). Accordingly, in Figure 2, times for proof attempts on formulas \( \varphi_t \) are generally higher than those for \( \varphi_t \). The induction step proofs mostly take time in the range of a few seconds only and hence show the efficiency of our method. However, some invariants cannot be proved due to our rather uninformed state generator, and further methods to its restriction (e.g. by inclusion of previously-proved positive-knowledge formulas as mentioned in Section 6.4) are required.

### 8 Related Work

The first approach that addresses the automated verification of epistemic properties over GDL specifications (Ruan and Thielscher 2011) used standard epistemic logic over Kripke semantics (Fagin et al. 1995), and the main focus of that work is the introduction and equivalence proof of a new semantics that allows the direct verification of epistemic logic.
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<td></td>
<td>p of b: 7.2 (n,3)</td>
<td>p of b: 32.6 (?,4)</td>
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<tr>
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<td>p of i: 7.2 (n,3)</td>
<td>p of i: 32.7 (?,4)</td>
</tr>
</tbody>
</table>

Figure 2: Proof times in seconds for a selection of positive-knowledge formulas in several incomplete-information games, where “y” means valid, “n” means invalid (violated in some reachable state), and “?” means validity unknown. The names of players have been shortened to one letter, so that, e.g., “j of r” in Cardgame stands for “Jane knows the legal moves (goal values, respectively) of Rick.” The time for result (y) indicates one induction-step proof (Phase 1). The time for results (n,t) and (? t) indicates the overall time needed for one induction-step proof (Phase 1) and t base-case proofs (Phase 2).

formulas over GDL descriptions. This previous work also sketches an approach for model checking epistemic properties via answer set programming, which however requires to systematically search the entire set of reachable positions in a game and hence is not applicable in the time-restricted setting of the general game playing competition except for very simple games.

9 Summary and Discussion

Moving from complete to incomplete information poses a new challenge to both designer and player of a general game. While the former must endow each player with sufficient information to guarantee unobstructed gameplay, the latter needs to incorporate all opponents’ viewpoints to maximise its utility. For this purpose, we have extended our recently developed automated verification method (Haufe, Schiffel, and Thielscher 2012) to incorporate a large class of epistemic properties. We have shown its effectiveness by proving several common epistemic properties in a variety of games with a practical implementation that uses a state-of-the-art answer set solver.

One of the two main reasons for the practicability of our approach is the use of induction together with an easily obtainable overestimation of the state space and the restriction to formulas that only contain finite time reference. The other main reason is the utilisation of a linear time structure. It allows to linearly bound the growth of game rules which is mandatory for the required temporal gdl extension, whereas an also possible branching time structure causes an exponential blowup which almost certainly drops performance below practical usability. Linear time in answer set programming, on the other hand, only allows the verification of formulas which have succinct counter examples that are representable using linear state sequences. Our introduced class of positive-knowledge formulas hence forms a necessary restriction to the utilisation of a linear time structure.

As only the lack of knowledge cannot be formulated, positive-knowledge formulas still cover an interesting property class. In our experiments, we have concentrated on three game-independent properties whose discovered validity assist a game designer by showing that each player is endowed with sufficient information to play a newly specified game. But also general game players can take advantage of our approach. For example, the provable fact that players always know which cells contain their own marks when playing the game Krieg-Tictactoe specified in Figure 1 allows to spare repeated reasoning regarding this fact in every reached game state. In exploration games such as the famous Wumpus World, a general game player which is able to find out what is known after certain actions in advance can prefer those that reveal as much information as possible. The development of imperfect information-game players is just at the beginning, and we expect that many of the existing methods using knowledge will require the addition of epistemic properties when being generalised to imperfect-information games.

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