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This paper starts from a simple model of strategic reasoning in situations of social influence. Agents express binary views on a set of propositions, and iteratively update their views by taking into account the expressed opinion of their influencers. We empower agents with the ability to disclose or hide their opinions, in order to attain a predetermined goal. We study classical game-theoretic solution concepts in the resulting games, observing a non-trivial interplay between the individual goals and the structure of the underlying network. By making use of different logics for strategic reasoning, we show how apparently simple problems in strategic opinion diffusion require a complex logical machinery to be properly formalized and handled.

**Keywords**

Opinion diffusion; Nash equilibria; strategic reasoning; social influence

### 1. INTRODUCTION

Social influence can be seen as a process where an agent forms her opinion on the basis of the opinions expressed by agents she trusts [3, 10, 18]. Recent work in multi-agent systems proposed formal models of opinion diffusion which combined methods and techniques from social network analysis with those of belief merging and judgment aggregation [22, 13, 5, 29]. These frameworks built on classical models of opinion diffusion on networks such as the De Groot or Lehrer-Wagner model [9, 16] and threshold models [14], adapting them to more complex representations of individual opinions: belief bases, preferences, and binary judgments over interconnected issues. The focus has been on studying how agents’ opinions evolve over time due to the influence of other agents in the society. Namely, an agent’s opinion at a given time instant results from aggregating the opinions (at the previous time instant) of the agents she trusts and the trust relationship is modeled via a network.

In the present paper, we build on this work by looking at social influence from a strategic perspective. We introduce a new class of infinite repeated games, the *games of influence*. At each stage an agent can choose whether to make her opinions public, and she updates her opinion according to the public opinions of the agents she trusts and to some aggregation procedure. Moreover, to each agent is associated a goal she wants to achieve — in line with previous work on representation of agent’s motivations [15], expressed in a variant of Linear Temporal Logic LTL to consider agents’ future opinions.

Influence games provide a simple abstraction to explore the effects of a trust network on agents’ behavior, as well as allowing us to study game-theoretic solution concepts such as winning strategy, weak dominance and Nash equilibrium. First, we analyze the interplay between the network structure and a player’s strategic ability, focussing on two goal schemas: that of reaching consensus among a group of agents and that of influencing individuals towards one’s own opinion. Since most of these results are of a limitative nature, pointing at the difficulties of introducing a network into an apparently simple problem, we move to explore the existence of game-theoretic solutions as a computational problem. We study its complexity by devising translations into existing logical formalisms for strategic reasoning, such as Alternating-time Temporal Logic [1] and Strategy Logic [17].

Previous work on logic-based models for social influence include the Facebook logic [24] and the preference and reliability models (which do not, however, consider strategic aspects in the update) [11]. Other known models differentiating private and public information have not focussed on strategic aspects [7, 8]. Problems close to opinion diffusion are those of information cascades and knowledge diffusion, which have been given formal treatment in a logical setting [20, 4]. Moreover, influence games can be considered as a variation of iterated boolean games [15] in which individuals do not have direct control over all the variables, but they concurrently participate in changing their truth values. Iterated boolean games have recently been extended with a social network structure where agents choose actions depending on the actions of their neighbors [25].

The remainder of the paper is organized as follows. Section 2 presents the basic definitions of private and public opinions, as well as our model of opinion diffusion. Moreover, we introduce state transitions given by agents’ actions and the variant of LTL with which we express agents’ goals. Section 3 introduces influence games, and presents the main results about the effects of the network structure on solution concepts from game theory. Section 4 studies complexity problems related to the solution concepts for influence games, and Section 5 concludes the paper.
2. OPINION DIFFUSION

In this section we present the model of opinion diffusion which is the starting point of our analysis. We generalize the model of propositional opinion diffusion introduced in related work [13] by separating private and public opinions with a notion of visibility, and we adapt the diffusion process through aggregation to this more complex setting.

2.1 Private and public opinions

Let \( I = \{p_1, \ldots, p_n\} \) be a finite set of propositions or issues and let \( N = \{1, \ldots, n\} \) be a finite set of individuals or agents. Agents have opinions on issues in \( I \) in the form of a propositional evaluation or, equivalently, a binary vector:

**Definition 1 (Private opinion).** The private opinion of agent \( i \) is a function \( B_i : I \rightarrow \{0, 1\} \) where \( B_i(p) = 1 \) and \( B_i(p) = 0 \) express, respectively, the agent’s opinion that \( p \) is true and the agent’s opinion that \( p \) is false.

Let \( B = (B_1, \ldots, B_n) \) denote the profile of private opinions of agents in \( N \). Propositional evaluations can be used to represent ballots in a multiple referendum, expressions of preference over alternatives, or judgements over correlated issues [6, 12]. Depending on the application at hand, an integrity constraint can be introduced to model the propositional correlation among issues. For the sake of simplicity we do not introduce any such constraint in this paper.

We also assume that each agent has the possibility of declaring her private opinion on each issue.

**Definition 2 (Visibility function).** We call visibility function of agent \( i \) any map \( V_i : I \rightarrow \{0, 1\} \) where \( V_i(p) = 1 \) expresses that agent’s opinion on \( p \) is visible.

We denote by \( V = (V_1, \ldots, V_n) \) the profile composed of the agents’ visibility functions. By combining the private opinion with the visibility function of an agent, we can build her public opinion as a three-valued function on the issues.

**Definition 3 (Public opinion).** Let \( B \) be the opinion of agent \( i \) and \( V \) her visibility function. The public opinion of \( i \) is a function \( P_i : I \rightarrow \{0, 1, ?\} \) such that

\[
P_i(p) = \begin{cases} 
B_i(p) & \text{if } V_i(p) = 1 \\
? & \text{if } V_i(p) = 0 
\end{cases}
\]

Again, \( P = (P_1, \ldots, P_n) \) is the profile of public opinions of all the agents in \( N \). We denote by \( P_C \) the restriction of public profile \( P \) to individuals in \( C \subseteq N \).

**Definition 4 (State).** A state is a tuple \( S = (B, V) \) where \( B \) is a profile of private opinions and \( V \) is a profile of visibility functions. The set of all states is denoted by \( S \).

States are the building blocks of the model of strategic reasoning in opinion dynamics that we present in Section 2.3.

2.2 Unanimous opinion diffusion

We here define the influence process that is at the heart of our model. Our definition generalizes the model of propositional opinion diffusion [13] to take into account the visibility function. First, we assume that individuals are linked by an influence network modeled as a directed graph:

**Definition 5 (Influence network).** We call an influence network a directed irreflexive graph \( E \subseteq N \times N \), where \( (i,j) \in E \) reads “agent \( j \) is influenced by agent \( i \).”

We also refer to \( E \) as the influence graph. Let \( \text{Influencers} = \{k \in N \mid (k, i) \in E\} \) be the set of influencers of agent \( i \) in the network \( E \). Given a state \( S \), this definition can be refined by considering \( \text{Influencers}(i,p) = \{k \in N \mid (k, i) \in E \) and \( P_k(p) \neq ?\} \) to be the subset of \( i \)'s influencers that are actually showing their private opinion about issue \( p \).

Given a profile of public opinions and an influence network, we model the process of opinion diffusion by means of an aggregation function, which shapes the private opinion of an agent from the public opinions of other agents.

**Definition 6 (Aggregation procedure).** An aggregation procedure for agent \( i \) is a class of functions \( F_{i,C} : \{0, 1\}^I \times \{0, 1, ?\}^{2 \times C} \rightarrow \{0, 1\}^I \) for all \( C \subseteq N \setminus \{i\} \) that maps agent \( i \)'s individual opinion and the public opinions of a set of agents \( C \) to agent \( i \)'s individual opinion.

We drop \( C \) from the subscript when clear from the context, to simplify notation. Many aggregation procedures have been considered in the literature on judgment aggregation, and they can be adapted to our setting. Notable examples are quota rules, where agents change their opinion if the number of people disagreeing with them is higher than a given quota, such as the majority rule (cfr. the class of threshold models studied in the literature on opinion diffusion [14, 21]). Unanimity is another instance of a quota rule, which we adapt here:

**Definition 7.** The unanimous issue-by-issue aggregation procedure is defined as follows:

\[
F^U_i (B_i, P_C)(p) = \begin{cases} 
B_i(p) & \text{if } C = \emptyset \\
x \in \{0, 1\} & \text{if } P_k(p) = x \text{ for all } k \in C \\
B_i(p) & \text{otherwise} 
\end{cases}
\]

That is, an individual will change her private opinion about issue \( p \) if and only if all agents in \( C \) (usually among her influencers) publicly expressing their opinion are unanimous in disagreeing with her own. For the sake of simplicity, in the remainder of the paper we will consider that all agents use the unanimous aggregation procedure.

2.3 Strategic actions and state transitions

In our model, agents can make their opinions visible or invisible by specific actions of type reveal(\( J \)) — i.e., action of showing the opinion on issues in \( J \), and hide(\( J \)) — i.e., action of hiding the opinion on issues in \( J \). We allow for simultaneous disclosure on multiple propositions. Let thus:

\[
\mathcal{A} = \{\text{reveal}(J), \text{hide}(J') \mid J, J' \subseteq I \text{ and } J \cap J' = \emptyset\}
\]

be the set of individual actions. Each joint action \( a = (a_1, \ldots, a_n) \in \mathcal{A}^n \) induces a deterministic transition function between states:

**Definition 8 (Transition function).** The transition function \( \text{suc}: S \times \mathcal{A}^n \rightarrow S \) associates to each state \( S \) and joint action \( a \) a new state \( S' = (B', V') \) as follows, for all \( i \in N \) and \( p \in I \).

\[
a_i = \text{reveal}(J), \text{hide}(J')) \in \mathcal{A}:
\]
scribing the temporal aspect of agents' opinion dynamic:

Specifically, first, actions affect the visibility of opinions, and the definition assumes that the influence process occurs after the actions have changed the visibility of the agents' opinions. Specifically, first, actions affect the visibility of opinions, and then each agent modifies her private opinion on the basis of those opinions of her influencers that have become public.

We are now ready to define the concept of history, describing the temporal aspect of agents' opinion dynamic:

DEFINITION 9 (history). Given a set of issues \( I \), a set of agents \( N \), and aggregation procedures \( F \), for \( i \in N \) over a network \( E \), an history is an infinite sequence of states \( H = (H_0, H_1, \ldots) \) such that for all \( t \in \mathbb{N} \) there exists a joint action \( a_t \in A^N \) such that \( H_{t+1} = a_t(H_t) \).

Let \( H = (H_0, H_1, \ldots) \) be an history: the set of all histories is denoted by \( \mathcal{H} \). Observe that our definition restricts the set of all possible histories to those that correspond to a run of the influence dynamic described above. For notational convenience, for any \( i \in N \) and for any \( t \in \mathbb{N} \), we denote with \( H_{i,t}^B \) agent \( i \)'s private opinion in state \( H_1 \) and with \( H_{i,t}^V \) agent \( i \)'s visibility function in state \( H_t \).

Example 1. Consider the example in Figure 1, where the initial state is \( H_0 \) and the agents are \( N = \{ i, j, k \} \) such that \( \text{Inf}(i) = \{ j, k \} \). Let \( a_0 = (\text{skip}, \text{skip}, \text{reveal}(p)) \) and \( a_1 = (\text{skip}, \text{skip}, \text{reveal}(p)) \) be the joint actions of the agents at the first two states. Namely, agent \( k \) reveals her opinion on \( p \), and at the next step \( j \) hides hers. If all individuals are using the unanimous aggregation procedure, then states \( H_1 \) and \( H_2 \) result from applying the joint actions from state \( H_0 \). In state \( H_1 \), agent \( i \)'s private opinion about \( p \) has changed to \( 1 \), as all her influencers are publicly unanimous on \( p \), while in \( H_2 \) no opinion is updated.

\[
\begin{align*}
(0, 1, 1, 1, 0) & \xrightarrow{a_0} (1, 1, 1, 1, 1) & (1, 1, 1, 1, 0) & \xrightarrow{a_1} (1, 1, 1, 1, 0) \\
H_0 & \quad H_1 & \quad H_2
\end{align*}
\]

Figure 1: The first states of a history.

2.4 Individual goals

By revealing or hiding her opinion, an agent influences others towards the satisfaction of her goal. To account for the temporal aspect of our model, we follow recent work on iterated boolean games [15] and we define a language \( \mathcal{L}_{\text{TLT}} \) to express individual goals using Linear Temporal Logic LTL.

Let therefore \( \mathcal{L}_{\text{TLT}} \) be defined as follows:

\[
\varphi ::= \lnot \text{op}(i, p) | \text{vis}(i, p) | \lnot \varphi | \varphi \land \varphi | \bigcirc \varphi | \bigoplus \varphi
\]

where \( i \) ranges over \( \mathcal{N} \) and \( p \) ranges over \( \mathcal{I} \). We read \( \text{op}(i, p) \) as "agent \( i \)'s opinion is that \( p \) is true"; while \( \lnot \text{op}(i, p) \) has to be read "agent \( i \)'s opinion is that \( p \) is not true" (since agents have binary opinions). Moreover, \( \text{vis}(i, p) \) has to be read "agent \( i \)'s opinion about \( p \) is visible".

The reading of \( \bigcirc \varphi \) is "\( \varphi \) is going to be true at the next state" and of \( \bigoplus \varphi \) is "\( \varphi \) will be true until \( \varphi \) is true": they are the standard LTL operators 'next' and 'until'. As usual, we can define the temporal operators 'eventually' (\( \Diamond \)) and 'henceforth' (\( \bigcirc \)) as \( \Diamond \varphi = \top \bigcirc \varphi \) and \( \bigcirc \varphi = \lnot \Diamond \lnot \varphi \). The interpretation of \( \mathcal{L}_{\text{TLT}} \)–formulas relative to histories is defined as follows.

DEFINITION 10 (Truth conditions). Let \( H \) be a history, let \( \varphi \) be a formula of \( \mathcal{L}_{\text{TLT}} \) and let \( k, k', k'' \in \mathbb{N} \). Then:

\[
\begin{align*}
H, k & \models \text{op}(i, p) \iff H_{i,k}^B(p) = 1 \\
H, k & \models \text{vis}(i, p) \iff H_{i,k}^V(p) = 1 \\
H, k & \models \lnot \varphi \iff H, k \not\models \varphi \\
H, k & \models \varphi_1 \land \varphi_2 \iff H, k \models \varphi_1 \text{ and } H, k \models \varphi_2 \\
H, k & \models \bigcirc \varphi \iff H, k+1 \models \varphi \\
H, k & \models \bigoplus \varphi \iff \exists k' : (k \leq k' < k'' \text{ and } H, k' \models \varphi_2 \text{ and } \forall k'' : \text{if } k \leq k'' < k' \text{ then } H, k'' \models \varphi_1)
\end{align*}
\]

Formulas of \( \mathcal{L}_{\text{TLT}} \) will be used to express agents' goals on the iterative diffusion process. As individuals do not have any influence on the initial state of the history, we will consider only goals of the form \( \bigcirc \varphi \) and \( \bigoplus \varphi \), for any \( \varphi \) and \( \psi \) in \( \mathcal{L}_{\text{TLT}} \), which we denote as goal formulas.

For some subset of agents \( C \subseteq N \) and some issues \( J \subseteq \mathcal{I} \) consider the following goals on consensus and influence in situations of opinion diffusion:

\[
\begin{align*}
\text{cons}(C, J) & ::= \Diamond \Box (\text{pcons}(C, J) \lor \text{ncons}(C, J)) \\
\text{influ}(i, C, J) & ::= \Diamond \Box \bigwedge_{p \in J} (\lnot \text{op}(i, p) \to \Box \text{pcons}(C, p)) \\
& \land (\text{op}(i, p) \to \Box \text{ncons}(C, p))
\end{align*}
\]

where:

\[
\begin{align*}
\text{pcons}(C, J) & ::= \bigwedge_{i \in C} \bigwedge_{p \in J} \text{op}(i, p) \\
\text{ncons}(C, J) & ::= \bigwedge_{i \in C} \bigwedge_{p \in J} \lnot \text{op}(i, p).
\end{align*}
\]

Intuitively, an agent holding the first goal wants at some point in the history to reach a stable consensus either for or against the issues in \( J \) with the agents in \( C \). The second goal expresses instead the idea that agent \( i \) wants to eventually gain a stable influence over the people in \( C \) about the issues in \( J \) (i.e., they will always hold her opinion at the next step).

3. GAMES OF INFLUENCE

We are now ready to combine all concepts introduced in the previous sections to give the definition of an influence game:

DEFINITION 11 (Influence game). An influence game is a tuple \( IG = (\mathcal{N}, \mathcal{I}, E, F, S_0, \gamma_1, \ldots, \gamma_n) \) where \( \mathcal{N}, \mathcal{I}, E \) and \( S_0 \) are, respectively, a set of agents, a set of issues, an
influence network, and an initial state, $F_i$ for $i \in \mathcal{N}$ is an aggregation procedure, and $\gamma_i$ is agent $i$’s goal formula.

Given an influence game, our agents will build their strategies in order to attain their goals. We first introduce two kinds of strategies available to agents, namely memory-less and perfect-recall strategies, then we study the existence of basic solution concepts from game theory when agents use influence or consensus goals (see Section 2.4).

### 3.1 Strategies

The first type of simple individual strategies we define is based on states, and thus memory-less.

**Definition 12 (memory-less strategy).** A memory-less strategy for player $i$ is a function $Q_i : S \to A$ that associates an action to every state.

A strategy profile is a tuple $Q = (Q_1, \ldots, Q_n)$. For notational convenience, we also use $Q$ to denote the function $Q : S \to A^n$ such that $Q(S) = a$ if and only if $Q_i(S) = a_i$, for all $S \in S$ and $i \in \mathcal{N}$.

The second definition we provide is the full-blown notion of perfect-recall strategy, which assigns an action to each partial history that has been observed by the player. Let $H^*$ denote the set of all partial histories — i.e., finite sequences of states satisfying Definition 9.

**Definition 13 (perfect-recall strategy).** We call a perfect-recall strategy for player $i$ a function $Q_i : H^* \to A$ that associates an action to every finite history.

As the following definition highlights, every strategy profile (of either type) induces a history if combined with an initial state:

**Definition 14 (induced history).** Let $S_0$ be an initial state and let $Q$ be a strategy profile. The induced history $H_{S_0,Q} \in \mathcal{H}$ is defined as follows:

\[
H_0(S_0, Q) = S_0 \\
H_{n+1}(S_0, Q) = \text{succ}(S_n, Q(S_n)) \text{ for all } n \in \mathbb{N}
\]

The two definitions allow us to differentiate the study among more or less sophisticated agents. It is important to observe, however, that we study influence games as repeated games of complete information. This is to be considered as a simplifying assumption — a more realistic model would require the use of uniform strategies and the corresponding indistinguishability relation. The hide action can alternatively be interpreted in our model as “not exercising one’s own influence”, without any epistemic interpretation, and the reveal action as “persuade”.

### 3.2 Solution concepts

The first solution concept we study is that of winning strategy. Intuitively, $Q_i$ is a winning strategy for player $i$ if and only if $i$ knows that, by playing this strategy, she will achieve her goal no matter what the other players do.

**Definition 15 (Winning strategy).** Let $IG$ be an influence game and let $Q_i$ be a strategy for player $i$. We say that $Q_i$ is a winning strategy for player $i$ in state $S_0$ if

\[
H_{S_0,(Q_i,q_{-i})} \models \gamma_i
\]

for all profiles $Q_{-i}$ of strategies of players other than $i$.

**Example 2.** Let Ann, Bob and Jesse be three agents and suppose that $B_{Ann}(p) = 1$, $B_{Bob}(p) = 0$, $B_{Jesse}(p) = 0$ for $p \in \mathcal{T}$. Their influence network is as in this picture:

```
    Jesse
     \  /  \\
      \  /  \\
       Bob
        ^  \\
        \  \\
         Ann
```

Suppose Ann’s goal is $\bigcirc \text{op}(Jesse, p)$. Her winning memory-less strategy is to play $\text{reveal}(p)$ in all states. Bob will be influenced to believe $p$ at the second stage in the history and similarly for Jesse at the third stage, since her influenceurs are unanimous even if Bob plays $\text{hide}(p)$.

As we will show in the following section, the concept of winning strategy is rather strong for our setting. Let us then define the less demanding notion of weak dominance:

**Definition 16 (Weakly dominant strategy).** Let $IG$ be an influence game and $Q_i$ a strategy for player $i$. We say that $Q_i$ is a weakly dominant strategy for player $i$ and initial state $S_0$ if and only if for all profiles $Q_{-i}$ of strategies of players other than $i$ and for all strategies $Q_i'$ we have:

\[
H_{S_0,(Q_i,q_{-i})} \models \gamma_i \Rightarrow H_{S_0,(Q_i',q_{-i})} \not\models \gamma_i
\]

**Example 3.** Suppose that in the previous example Ann still believes $p$, but does not influence Jesse any longer. Here, Ann does not have a winning strategy: if neither Bob nor Jesse believe $p$, it is sufficient for Bob to play $\text{hide}(p)$ to make sure that she will never satisfy her goal. However, playing action $\text{reveal}(p)$ is weakly dominant for Ann.

Finally, we introduce the concept of Nash equilibrium for influence games:

**Definition 17 (Nash equilibrium).** Let $IG$ be an influence game and let $Q$ be a strategy profile. We say that $Q$ is a Nash equilibrium for initial state $S_0$ if and only if for all $i \in \mathcal{N}$ and for all $Q_i' \in Q_i$:

\[
H_{S_0,(Q_i,q_{-i})} \models \gamma_i \text{ or } H_{S_0,(Q_i',q_{-i})} \not\models \gamma_i
\]

Problems of membership, existence and uniqueness of Nash equilibria will be studied in Section 4.

### 3.3 Influence network and solution concepts

In this section we analyze the interplay between network structure and existence of solutions concepts for the goals defined in Section 2.4. We assume memory-less strategies.

**Proposition 1.** If $E$ is a directed acyclic graph (DAG) such that $|\text{Inf}(i)| \leq 1$ for all agents $i \in \mathcal{N}$, and if agent $a$ has goal $\gamma_a := \text{cons}(C_a, J)$ where $J \subseteq \mathcal{T}$ and $C_a := \{k \in \mathcal{N} | a \in \text{Inf}(k) \} \cup \{a\}$, then agent $a$ has a winning strategy.

**Proof.** (sketch) Consider a DAG $E$ and an agent $a$ with goal $\gamma_a$. Let $Q_a$ be the strategy associating to every state $S$ action $\text{reveal}(J)$. We want to show that $H_{S_0,(Q_a,q_{-a})} \models \gamma_a$ holds for all $S_0$ and $Q_{-a}$. Consider the position of agent $a$ in the graph for arbitrary $S_0$. In case there is no agent $b$ such that $a \in \text{Inf}(b)$, the goal reduces to $\text{cons}(\{a\}, J)$ which is always trivially satisfied. In case $\text{Inf}(a) = \emptyset$, by playing $\text{reveal}(J)$ in $S_0$ and since every agent uses the unanimous aggregation rule, at stage 1 all child nodes of $a$ will update
their beliefs on $J$ by copying $a$’s opinion (she is their only influencer). Moreover, they can’t change their opinions on $J$ later on in the history.

On the other hand, suppose there is some agent $b$ such that $a \in \text{Inf}(b)$ and some agent $c \in \text{Inf}(a)$. By assumption on $E$ we thus have that $\text{Inf}(a) = \{c\}$ and $\text{Inf}(b) = \{a\}$. Hence, either at some point $k$ in the history all ancestors of $a$ will have reached consensus, such that by playing $\text{reveal}(J)$ from point $k + 1$ onwards the consensus among $a$ and her child nodes will be maintained, or there is no such $k$. Since there is a unique path linking $a$ to one of the source nodes of $E$, if her ancestors always disagree in the history it means that there is some agent among them who has a different opinion and who will never play $\text{reveal}(J)$. Therefore, the opinion of $a$ will nonetheless be stable and $\gamma_a$ will be attained. □

The assumption of acyclicity in the above result rules out the situation where all nodes in a cycle play $\text{reveal}(J)$ and they start in $S_0$ by having alternating positive and negative opinions on the issues in $J$. In the second place, having at most one influencer per agent ensures each agent to have full control over their child nodes.

Observe also that Proposition 1 implies the same result to hold for $\gamma_a := \square \diamond (\text{pcons}(C_a, J) \lor \neg \text{pcons}(C_a, J))$. In fact, since eventually $a$ will reach a stable consensus with her child nodes, this implies that it is always true that we can find some later point in the history where consensus holds. In general, however, Proposition 1 shows us how winning strategies are too strong of a solution concept, so that the type of goals which can be attained have a narrow scope.

If we move to the less demanding concept of weak dominance, we may intuitively think that a strategy associating action $\text{reveal}(J)$ to all states is weakly dominant for an agent $a$ having goal $\gamma_a := \text{influ}(a, C, J)$ for $C \subseteq N^e$, regardless of the network $E$ or the initial state $S_0$. In fact, all agents use the monotonic aggregation rule $F^*$.

Yet, we show in the following example that to satisfy goals of type $\gamma_a$ as described, an agent could sometimes benefit from hiding her opinion.

**Example 4.** For four agents $N = \{1, 2, 3, 4\}$ and one issue $I = \{p\}$ consider the network $E = \{(1, 2), (2, 3), (3, 4)\}$. Suppose agent 1 and 2 associate action $\text{reveal}(p)$ to all states, and agent 3 associates action $\text{hide}(p)$ only to those states where 1, 2 and 3 agree on $p$. Suppose that the goal of agent 2 is $\gamma_2 = \text{influ}(2, \{4\}, \{p\})$. Consider the history below for these strategies, where goal $\gamma_2$ is not attained (we only represent $B$):

$$
\begin{align*}
(0, 1, 0, 1) &\quad \alpha_0 \quad (0, 0, 1, 0) \quad \alpha_1 \quad (0, 0, 0, 1) \\
H_0 &\quad H_1 &\quad H_2
\end{align*}
$$

From state $H_0$ onwards, given the strategies of the agents, the profile of opinions $B = (0, 0, 0, 1)$ won’t change. On the other hand, consider a strategy for agent 2 identical to the previous one, but for the fact that it associates to state $H_0$ action $\text{hide}(p)$. This is what would happen:

$$
\begin{align*}
(0, 1, 0, 1) &\quad \alpha_0 \quad (0, 0, 0, 0) \\
H_0 &\quad H_1
\end{align*}
$$

From state $H_0$ onwards, given the strategies of the agents, the profile of opinions won’t change. Thus, we found a network, an initial state $H_0$, and strategies for the other agents, such that agent 2 is sometimes better off by hiding her opinion on $p$ to satisfy her influence goal $\gamma_2$.

We can now see an easy example of how the network structure and the agents’ goals can yield a Nash equilibrium in the spirit of anti-coordination games [19].

**Proposition 2.** Let $E$ be a cycle for $N = \{1, 2\}$. If $\gamma_1 = \gamma_2 = \text{cons}(N, J)$, where $J \subseteq I$, then there exists a Nash equilibrium for any initial state $S_0$.

**Proof.** (sketch) To attain their goal the agents must coordinate on the issues in $J$ on which they disagree in $S_0$. In fact, in case at some stage $k$ of the history they both play $\text{hide}(p)$ for $p \in J$ their private opinion would stay the same at stage $k + 1$. If they both play $\text{reveal}(p)$, at the next stage they would just swap their opinions on $p$ (since they are each other’s only influencers and they both use the unanimous rule). Hence, agent 1 has to play $\text{reveal}(p)$ whenever the other agent is playing $\text{hide}(p)$ so that at the next stage in the history he will have copied her opinion, while she would have not changed hers — and similarly if the other agent is playing $\text{reveal}(p)$.

Consider thus an arbitrary strategy $Q_1$ for agent 1 and some initial state $S_0$. Construct now strategy $Q_2$ for agent 2 associating action $\text{reveal}(J^*)$ to all states where strategy $Q_1$ associates action $\text{hide}(J^*)$ for $J^* = \{p \in J \mid b_{1p} = 1 - b_{2p} \in S\}$, and viceversa for action $\text{reveal}(J^*)$. By the above reasoning, the strategy profile $Q = (Q_1, Q_2)$ generates a history that satisfies both $\gamma_1$ and $\gamma_2$, and therefore is a Nash equilibrium. The same construction can be done for an arbitrary strategy of agent 2 as well. □

### 4. COMPUTATIONAL COMPLEXITY AND SOLUTION CONCEPTS

In this section we provide complexity results for a variety of computational problems for the solution concepts introduced in Section 3.2. We start by studying the complexity of deciding whether a certain memory-less strategy profile of an influence game is a Nash equilibrium.

Then, we present two logics of strategic reasoning that have been studied in the area of logics for multi-agent systems: Alternating-time Temporal Logic ATL [1] and the graded variant of Strategy Logic SL [17, 2]. We then focus on perfect-recall strategies for influence games and we reduce the problem of checking existence of a winning strategy to the model checking problem of the former, and the problem of checking existence or uniqueness of a Nash equilibrium to the model checking problem of the latter.

#### 4.1 Problem definitions

We now provide the formal definitions of the complexity problems we are interested in studying for influence games.

**INPUT:** $IG = (N, I, E, F, S_0, \gamma_1, \ldots, \gamma_n)$. $Q$.

**M-NASH($F$):** Is $Q$ a Nash equilibrium of $IG$?

This problem is out of reach for perfect-recall strategies, since the size of the input (i.e., the size of a perfect-recall strategy profile) would already be exponential in the size of the initial game. We will thus focus on the problems $E$-$NASH(F)$ and $U$-$NASH(F)$, defined below.

**INPUT:** $IG = (N, I, E, F, S_0, \gamma_1, \ldots, \gamma_n)$.

**E-NASH($F$):** Is there some Nash equilibrium $Q$ of $IG$?

**U-NASH($F$):** Is there a unique Nash equilibrium $Q$ of $IG$?
Observe that $F$ is not part of the input but a parameter of the problems defined above, since, as we shall see later, different aggregation functions may give rise to computational problems in different complexity classes.

### 4.2 Memory-less Nash equilibrium

We begin by translating a memory-less strategy in the language of the logic $\text{LTL}-\text{I}$ (see Section 2.4). The algorithm presented in the setting of iterated boolean games [15] cannot be directly applied to ours for two reasons. First, our histories are generated by an aggregation function $F$ that models opinion diffusion — i.e., agents have a concurrent control over a set of propositional variables. Second, in this section we focus on memory-less strategies only.

A conjunction of literals $\alpha(S)$ can be defined to uniquely identify a state $S$: $\alpha(S)$ will specify the private opinion of all individuals and their visibility function. For an action $a = (\text{reveal}(J), \text{hide}(J'))$, let $\beta_i(a)$ be the following formula:

$$\beta_i(a) = \bigwedge_{p \in J} \neg \text{vis}(i, p) \land \bigwedge_{q \in J'} \neg \text{vis}(i, q).$$

In case $a = \text{skip}$ we let $\beta_i(a) = \top$. Given a memory-less strategy $Q_i$, we construct the following formula:

$$\tau_i(Q_i) = \bigwedge_{S \in S} \alpha(S) \rightarrow \beta_i(Q_i(S)).$$

If $Q$ is a strategy profile, let $\tau(Q) = \bigwedge_{i \in \mathcal{N}} \tau_i(Q_i)$. We now need to encode the aggregation function into a formula as well. Consider the following formula $\text{unan}(i, p)$:

$$\text{unan}(i, p) = \bigcirc \text{op}(i, p) \leftrightarrow \left( \bigwedge_{j \in \text{Inf}(i)} \bigcirc \neg \text{vis}(j, p) \land \text{op}(i, p) \right) \lor \left( \bigvee_{j \in \text{Inf}(i)} \text{vis}(j, p) \land \bigwedge_{j \in \text{Inf}(i)} \text{op}(j, p) \right) \lor \left( \bigvee_{j \in \text{Inf}(i)} \bigcirc \text{vis}(j, p) \land \bigwedge_{j \in \text{Inf}(i)} \neg \text{op}(j, p) \right).$$

as well as the following formula $\text{unan}(i, \neg p)$:

$$\text{unan}(i, \neg p) = \bigcirc \neg \text{op}(i, p) \leftrightarrow \left( \bigwedge_{j \in \text{Inf}(i)} \bigcirc \neg \text{vis}(j, p) \land \neg \text{op}(i, p) \right) \lor \left( \bigvee_{j \in \text{Inf}(i)} \text{vis}(j, p) \land \bigwedge_{j \in \text{Inf}(i)} \neg \text{op}(j, p) \right) \lor \left( \bigvee_{j \in \text{Inf}(i)} \bigcirc \text{vis}(j, p) \land \bigwedge_{j \in \text{Inf}(i)} \text{op}(j, p) \right) \lor \left( \bigvee_{j \in \text{Inf}(i)} \neg \text{vis}(j, p) \land \bigwedge_{j \in \text{Inf}(i)} \neg \text{op}(j, p) \right).$$

This formula ensures that if the influencer of agent $i$ is unanimous, then agent $i$’s opinion should be defined according to the three cases described in Definition 7. Recall that, while actions take one time unit to be performed — hence the $\bigcirc$ operator in front of $\text{vis}(j, p)$, the diffusion of opinions is simultaneous. Let now:

$$\tau(F_{\text{U}}^i) = \bigwedge_{i \in \mathcal{N} \setminus \text{Inf}(i) \neq \emptyset} \bigwedge_{p \in \mathcal{I}} (\text{unan}(i, p) \land \text{unan}(i, \neg p))$$

be the formula encoding the transition process defined by the opinion diffusion. $\tau(F_{\text{U}}^i)$ is in polynomial both the number of individuals and the number of issues — in the worst case it is quadratic in the number of agents and linear in the number of issues. We are now ready to prove the following:

**Theorem 1.** Let $F_{\text{U}}^i$ be the unanimous issue-by-issue aggregator. M-NASH($F_{\text{U}}^i$), E-NASH($F_{\text{U}}^i$) and U-NASH($F_{\text{U}}^i$) for memory-less strategies are in PSPACE.

**Proof.** We begin with M-NASH($F_{\text{U}}^i$). Let $Q$ be a memory-less strategy profile for game $IG$. The following algorithm can be used to check whether $Q$ is a Nash equilibrium. For all individuals $i \in \mathcal{N}$, we first check the following entailment:

$$\tau(Q) \land \tau(F_{\text{U}}^i) \models_{\text{LTL}} \gamma_i$$

in the language of $\text{LTL}$ built out of the set of propositions $\{\text{op}(i, p) \mid i \in \mathcal{N} \text{ and } p \in \mathcal{I}\} \cup \{\text{vis}(i, p) \mid i \in \mathcal{N} \text{ and } p \in \mathcal{I}\}$.

If this is not the case, we consider all the possible strategies $Q'_i \neq Q_i$ for agent $i$ — there are exponentially many of them, but each one can be specified in space polynomial in the size of the input — and check the following entailment:

$$\tau(Q_{-i}, Q'_i) \land \tau(F_{\text{U}}^i) \models_{\text{LTL}} \gamma_i$$

If the answer is positive we output NO, otherwise we proceed until all strategies and all individuals have been considered. The entailment for $\text{LTL}$ can be reduced to the problem of checking validity in $\text{LTL}$. Indeed, the following holds:

$$\psi \models_{\text{LTL}} \varphi \text{ iff } \models_{\text{LTL}} \Box \psi \rightarrow \varphi$$

Since the problem of checking validity in $\text{LTL}$ can be solved in PSPACE [26], we obtain the desired upper bound.

Similar algorithms can then be devised to solve E-NASH and U-NASH. To see this, it is sufficient to consider one strategy profile $Q$ at the time, and use M-NASH with $Q$ in the input as a PSPACE-oracle. For E-NASH the algorithm can stop at the first Nash equilibria that is found, while for U-NASH the entire set of strategy profiles eventually needs to be visited — again, there are exponentially many profiles, but each one can be specified in space polynomial in the size of the input. $\square$

As for completeness, a reduction from similar results in iterated boolean games [15] is likely to be obtained. We leave a more careful study of the relations between these two settings for future work. Observe also that Theorem 1 can easily be generalised to all aggregation procedures that can be axiomatized by means of polynomially many LTL-I formulas, such as quota rules. This is not the case for all aggregation procedures: the majority rule would for instance require an exponential number of formulas, one for each subset of influencers forming a relative majority. The study of the axiomatization of aggregation procedures for opinion diffusion also constitutes a promising direction for future work.

### 4.3 Alternating-time temporal logic

We recall here the syntax and semantics of Alternating-time Temporal Logic $\text{ATL}$. The language of $\text{ATL}$ is defined by the following BNF:

$$\varphi ::= q \mid \neg \varphi \mid \varphi \land \varphi \mid (\langle C \rangle) \varphi \mid (\langle C \rangle)(\varphi \Upsilon \varphi)$$
where $C$ ranges over $2^N$ and $q$ ranges over a set of atomic propositions $Atm$. The formula $\langle C \rangle \Diamond \varphi$ is read “coalition $C$ has the capability to ensure that $\varphi$ is going to be true in the next state regardless of what the agents outside $C$ decide to do”, and the formula $\langle C \rangle G \langle u \rangle (\varphi \land U_2 \varphi)$ is read “coalition $C$ has the capability to ensure that $\varphi$ will be true until $\varphi_2$ is true, regardless of what the agents outside $C$ decide to do”.

We here consider the standard ATL semantics in terms of concurrent game structures.

**Definition 18.** (Concurrent ATL semantics) A concurrent game structure is a tuple $G = (W, \mathcal{M}, R, T, \mathcal{V})$ where $W$ is a set of worlds or states, $\mathcal{M}$ is a set of moves, function $R : N \times W \rightarrow 2^W \setminus \emptyset$ defines a nonempty repertoire of moves for each agent at each world, $T : W \times \mathcal{M} \rightarrow W$ is a transition function mapping a world $w$ and a move profile $m = (m_1, \ldots, m_n)$ to the successor world $T(w, m)$, and $\mathcal{V} : 2^W \rightarrow 2^W$ is a valuation function.

In ATL, a strategy for player $i$ is a function $f_i$ that maps every finite sequence of worlds $\pi = \omega_0 \ldots \omega_k$ in $W^+$ (i.e., a path) to a move $f_i(\pi) \in R(i, \omega_k)$ available to agent $i$ at the end of path $\pi$. A strategy for coalition $C$ is a function $G_C$ that maps every agent $i \in C$ to a strategy $G(i)$ for $i$. The set of strategies for coalition $C$ is denoted by $Str_C$. We write $G$ instead of $G_C$, and $Str$ instead of $Str_N$.

A move profile is used to determine a successor of a world using the transition function $T$. We define the set of available move profiles at world $w$ as follows:

$$\mathcal{P}(w) = \{ (m_1, \ldots, m_n) \in M^n \mid m_i \in R(i, w) \text{ for all } i \in N \}$$

The set of possible successors of $w$ is the following:

$$\text{Succ}(w) = \{ T(w, m) \mid m \in \mathcal{P}(w) \}.$$  

An infinite sequence $\lambda = \omega_0\omega_1\omega_2 \ldots$ of worlds from $W$ is called a computation if $\omega_{k+1} \in \text{Succ}(\omega_k)$ for all $k \geq 0$. The $k$-th component $\omega_k$ in $\lambda$ is denoted by $\lambda[k]$. Moreover, for every computation $\lambda = \omega_0\omega_1\omega_2 \ldots$ and every positive integer $k$, $Prf(\lambda, k) = \omega_0 \ldots \omega_k$ denotes the prefix of $\lambda$ of length $k$.

The set $\mathcal{O}(w, G_C)$ is the set of all computations $\lambda = \omega_0\omega_1\omega_2 \ldots$ such that $\omega_k \in W$ and, for every $k \geq 0$, there is $m = (m_1, \ldots, m_n) \in \mathcal{P}(w_k)$ such that $G_C(i)(\omega_0 \ldots \omega_k) = m_i$ for all $i \in C$, and $T(w, m) = \omega_{k+1}$. Notice that $\mathcal{O}(w, G)$ is a singleton.

Truth conditions of ATL are defined relative to a CGS $G = (W, \mathcal{M}, R, T, \mathcal{V})$ and a world $w \in W$; we omit the standard definition of truth conditions for Boolean formulas.

**4.4 Graded strategy logic**

Strategy logic $SL$ is a logic of strategic reasoning that embeds ATL [17], but it allows to quantify over strategies in a more flexible way. The flexibility of SL is mainly due to the fact that it has variables for strategies that are associated to specific agents with a binding operator.

An extension of SL by graded quantifiers over tuples of strategy variables has also been presented [2]. We here focus on this SL extension, denoted by $G - SL$, whose language is defined by the following BNF:

$$\varphi ::= q | \varphi \land \varphi | \Diamond \varphi | \varphi U \varphi | \langle x_1, \ldots, x_l \rangle^{\leq k} \varphi | [i \mapsto x] \varphi$$

where $q$ ranges over $Atm$, $i$ ranges over $N$, $k$ and $\ell$ range over the set of positive integers, $x_1, \ldots, x_l$ range over a countable infinite set of variables $Var$ with the additional constraint that $x_h \neq x_{h'}$ for all $1 \leq h, h' \leq \ell$ such that $h \neq h'$. $\langle x_1, \ldots, x_l \rangle^{\leq k} \varphi$ is read “there exist at least $k$ different $\ell$-tuples of strategies ensuring $\varphi$ while $[i \mapsto x] \varphi$ is read “if agent $i$ selects strategy $x$ then $\varphi$ will hold”.

Formulas $\bigcirc \varphi$ and $\varphi U \varphi$ have the same reading as in ATL. Moreover, the basic SL operator $\langle x \rangle$ abbreviates $\langle x \rangle^{\geq 1}$ in $G - SL$.

Like ATL, SL semantics is based on the concept of concurrent game structure $G = (W, \mathcal{M}, R, T, \mathcal{V})$. However, to provide an interpretation of SL some extra-components are needed. We denote with $V$ a function mapping agents and variables in $N \cup Var$ to strategies in $Str$. We call $X$ an assignment function. For every assignment function $X$, strategy $G \in Str$ and $e \in N \cup Var$, we write $X[e \mapsto G]$ to denote the assignment function that differs from $X$ only in the fact that $e$ maps to $G$. We extend this definition to tuples of agents or variables $\theta = (e_1, \ldots, e_r)$ and tuples of strategies $\theta = (G_1, \ldots, G_r)$ with $\theta \neq e_0$ for $h \neq h'$, by denoting with $X[\theta \mapsto \theta]$ the assignment function that differs from $X$ only in that $e_h$ maps to $G_h$ for each $1 \leq h \leq \ell$.

Furthermore, for every assignment function $X$, we write $G_X$ to denote the strategy in $Str$ generated by the assignment function $X$. $\Gamma = G_X$ is the strategy in $Str$ such that, for all $i \in N$, $\GammaX(i) = X(i)(i)$.

Moreover, for every world $w \in W$, we write $\lambda_{wX}$ to denote the computation starting in $w$ generated by the assignment function $X$. $\lambda_{wX}$ denotes the computation $\omega_0\omega_1\omega_2 \ldots$ such that $\mathcal{O}(w, G_X) = \{ (w_0 \omega_1 \omega_2 \ldots) \}$. Finally, for every path $\pi$, we write $\pi_X$ to denote the assignment function obtained by shifting all strategies in the image of $X$ by $\pi$. For all $e \in N \cup Var, i \in N$ and $\pi, \pi' \in W^+$, we thus have:

$$\lambda_{wX}(e)(i)(\pi') = X(e)(i)(\pi \cdot \pi')$$

Any $G - SL$ formula $\varphi$ is evaluated relative to a CGS $G$, a world $w \in W$ and assignment function $X$.

Let $\vec{x} = (x_1, \ldots, x_l)$. Then:

$$\mathcal{G}, w, X ::= [i \mapsto x] \varphi$$

exist $k$ many $\ell$-tuples $\langle G_1, \ldots, G_k \rangle$ of strategies such that for all $1 \leq h, h' \leq k$ if $G_h \neq G_{h'}$ then $h \neq h'$ and $G, w, X[\vec{x}] \uparrow \langle G_1, \ldots, G_k \rangle$.
4.5 Perfect-recall Nash equilibrium

We now introduce the problem of checking the existence of a winning strategy for an agent \( i \in \mathcal{N} \) in an influence game \( IG \) as follows:

**INPUT**: \( IG = (\mathcal{N}, \mathcal{I}, E, F, S_0, \gamma_1, \ldots, \gamma_n) \), \( i \in \mathcal{N} \).

**E-WINNING**\(_i(F)\): Is there a winning strategy \( Q_i \) in \( IG \)?

The following theorem studies the complexity of the problem **E-WINNING**\(_i(F)\) for perfect-recall strategies and for the unanimous aggregation procedure.

**Theorem 2.** **E-WINNING**\(_i(F_i^U)\) for perfect-recall strategies is in \( \text{EXPTIME} \).

**Proof. (sketch)** We reduce **E-WINNING**\(_i(F_i^U)\) to the model checking problem of ATL where \( \text{Atm} = \{ \text{op}(i, p) \mid i \in \mathcal{N} \text{ and } p \in \mathcal{I} \} \cup \{ \text{vis}(i, p) \mid i \in \mathcal{N} \} \). First of all, observe that we can generate the CGS \( G_{IC} = (W, M, \mathcal{A}, \mathcal{T}, \mathcal{V}) \) corresponding to the influence game \( IG \) as follows:

- \( W = S \),
- \( M = A \),
- for all \( i \in \mathcal{N} \) and \( w \in W \), \( R(i, w) = M \),
- for all \( w \in W \) and \( a \in A^n \), \( T(w, a) = \text{succ}(w, a) \),
- for all \( \text{op}(i, p), \text{vis}(i, p) \in \text{Atm} \) and \( w \in W \), \( \text{op}(i, p) \in \text{Val}(w) \) iff \( B_i(p) = 1 \), \( \text{vis}(i, p) \in \text{Val}(w) \) iff \( V_i(p) = 1 \).

Secondly, it is easy to verify that the answer to the problem **E-WINNING**\(_i(F_i^U)\) is YES if and only if \( G_{IC}, w_0 \models \langle\{i\}\rangle \gamma_i \) with \( w_0 = S_0 \). In [1, Theorem 5.2] it is shown that the model checking problem for ATL can be solved in time \( O(m \cdot l) \) where \( m \) is the number of transitions in the CGS (which is polynomial in the size of \( W \) and \( M \)) and \( l \) is the size of the formula. Since the number of states \( S \) is exponential in the size of an influence game (more precisely, it is \( O(2^{2l} \cdot |\mathcal{N}|) \)) where \( \mathcal{I} \) is the set of issues and \( \mathcal{N} \) is the set of agents), we obtain the above upper bound.

The following theorem highlights the high complexity of checking existence and uniqueness of Nash equilibria in influence games.

**Theorem 3.** **E-NASH**\(_i(F_i^U)\) and **U-NASH**\(_i(F_i^U)\) for perfect-recall strategies are both in \( \text{3EXPTIME} \).

**Proof. (sketch)** As in the proof of Theorem 2, we generate the CGS \( G_{IC} \) corresponding to the influence game \( IG \). Following [2], we reduce **E-NASH**\(_i(F)\) and **U-NASH**\(_i(F)\) to the problems of checking \( G_{IC}, w_0 \models \langle\{i\}\rangle \gamma_i \) with \( w_0 = S_0 \), where:

\[
E-NASH = \langle\{x_1, \ldots, x_n\}\rangle^{\geq 2} \psi_{Nash} \land \mathcal{A} E-NASH \land \neg \langle\{x_1, \ldots, x_n\}\rangle^{\geq 2} \psi_{Nash}
\]

with:

\[
\psi_{Nash} = [1 \mapsto x_1] \ldots [n \mapsto x_n] \bigwedge_{1 \leq i \leq n} (\langle y_1 \rangle [i \mapsto y] \gamma_i \rightarrow \gamma_i).
\]

In [2, Theorem 3.1] it is shown that the model checking problems for the G — EL formulas E-NASH and U-NASH can be decided in \( 2\text{EXPTIME} \) (with respect to the size of the CGS and the size of the agents’ temporal goals \( \gamma_1, \ldots, \gamma_n \)). As in the proof of Theorem 2, we obtain the desired bound by recalling that the size of the CGS generated by an influence game is exponential in the size of the game.

Note that existence and uniqueness of Nash equilibria have been proved to be in \( \text{2EXPTIME} \) in the context of iterated boolean games [15]. We believe that the discrepancy between these results and Theorem 3 is due to the different representations of strategies we use. Our perfect-recall strategies may not have a finite representation, while in the setting of iterated boolean games strategies are deterministic finite state machines which have a finite representation.

5. CONCLUSIONS AND FUTURE WORK

We have extended a simple model of opinion diffusion on networks [13] with a strategic component (i.e., the possibility for the agents to hide or show their opinions) and individual goals, which allowed us to define influence games. We found that agents were greatly empowered by these basic actions, as reflected by our results on the interaction between goals, network structure and solution concepts (see Section 3). Interestingly, we found that for influence goals and the unanimous rule, playing a strategy always revealing the agent’s opinion was not necessarily weakly dominant.

We then approached the study of influence games and solution concepts from a computational point of view. Specifically, we found that for memory-less strategies the problems of checking whether a given strategy profile is a Nash equilibrium, checking whether a Nash equilibrium exists and is unique are all in \( \text{PSPACE} \). On the other hand, for perfect-recall strategies the problems of checking existence of a winning strategy is in \( \text{EXPTIME} \), while existence and uniqueness of Nash equilibrium are both in \( \text{3EXPTIME} \).

There are multiple directions to expand the present work. In the first place, following the work by Christoff and Hansen [7], we might provide our agents with additional actions to allow them to lie about their private opinions. This will give them more sophisticated strategies to attain their goals. Secondly, we plan to explore the use of different aggregation procedures to perform the update of opinions, studying the novel game-theoretic structures that may arise in this setting. Last but not least, we are interested in a deeper analysis of the relationship between influence games and iterated boolean games [15]. We conjecture that the latter can be embedded into the version of the former with perfect-recall strategies by only considering atomic formulas of type \( \text{vis}(i, p) \), whose truth values are under the direct control of the agents and do not depend on the aggregation procedure.

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