

1 Model Checking Strategic Ability 2 Why, What, and Especially: How? 3 (Extended Abstract)

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7 — Abstract —

8 Automated verification of discrete-state systems has been a hot topic in computer science for
9 over 35 years. Model checking of temporal and strategic properties is one of the most prominent
10 and most successful approaches here. In this talk, I present a brief introduction to the topic, and
11 mention some relevant properties that one might like to verify this way. Then, I describe some
12 recent results on approximate model checking and model reductions, which can be applied to
13 facilitate verification of notoriously hard cases.

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22 **1 Multi-Agent Systems and Strategic Ability**

23 More and more systems involve social as much as technological aspects, and even those that
24 focus on technology are often based on distributed components exhibiting self-interested,
25 goal-directed behavior. Moreover, the components act in environments characterized by
26 incomplete information and uncertainty. The field of *multi-agent systems* studies the whole
27 spectrum of phenomena ranging from agent architectures to communication and coordination
28 in agent groups to agent-oriented software engineering. The theoretical foundations are
29 mainly based on game theory and formal logic.

30 Many relevant properties of multi-agent systems refer to *strategic abilities* of agents and
31 their groups. In particular, most functionality requirements can be specified as the ability
32 of the authorized users to achieve their legitimate goals. At the same time, many security
33 properties can be phrased in terms of the inability of unauthorized users to compromise the
34 system. Properties of this kind can be conveniently specified in *modal logics of strategic*
35 *ability*, of which alternating-time temporal logic (**ATL**) [2] is probably the most popular.

36 **ATL** modalities, possibly in combination with epistemic operators, allow e.g. to capture
37 different flavors of coercion-resistance in voting systems [23, 17]. A simple hierarchy of

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$$\begin{aligned}
CR_1 &\equiv \neg \langle\langle \text{Coercer} \rangle\rangle G \left((\text{closed} \wedge \bigwedge_{v \in A} \neg \text{vote}_{v,1}) \rightarrow K_{\text{Coercer}} \left(\bigvee_{v \in A} \neg \text{vote}_{v,1} \right) \right) \\
CR_2 &\equiv \neg \langle\langle \text{Coercer} \rangle\rangle G \left((\text{closed} \wedge \bigwedge_{v \in A} \neg \text{vote}_{v,1}) \rightarrow \bigvee_{v \in A} K_{\text{Coercer}} (\neg \text{vote}_{v,1}) \right) \\
CR_3 &\equiv \neg \langle\langle \text{Coercer} \rangle\rangle G \left((\text{closed} \wedge \bigvee_{v \in A} \neg \text{vote}_{v,1}) \rightarrow K_{\text{Coercer}} \left(\bigvee_{v \in A} \neg \text{vote}_{v,1} \right) \right) \\
CR_4 &\equiv \neg \langle\langle \text{Coercer} \rangle\rangle G \left((\text{closed} \wedge \bigvee_{v \in A} \neg \text{vote}_{v,1}) \rightarrow \bigvee_{v \in A} K_{\text{Coercer}} (\neg \text{vote}_{v,1}) \right)
\end{aligned}$$

■ **Figure 1** Variants of coercion-resistance for coercion of multiple voters [17]

38 coercion-related specifications is presented in Figure 1. CR_1 expresses that the coercer
39 cannot force all the agents in A to vote for candidate 1, or else he will know, at the closing
40 of the election, that at least one of them disobeyed. CR_2 captures a stronger property: if
41 someone in A disobeyed, the coercer will know the identity of at least one such agent. We
42 leave the interpretation of CR_3 and CR_4 to the interested reader.

43 2 Practical Verification of Strategic Abilities

44 In the last 15 years, there has been a large number of works that study the syntactic and
45 semantic variants of **ATL** for agents with imperfect information, cf. [5, 1] for an overview.
46 The contributions are mainly theoretical, and concern the conceptual soundness of a given
47 semantics, meta-logical properties, and the complexity of model checking and satisfiability
48 problems. However, there is relatively little research on practical algorithms for verification.

49 This is somewhat easy to understand, since model checking of **ATL** variants with
50 imperfect information is Δ_2^P - to **PSPACE**-complete for memoryless strategies [22, 4] and
51 **EXPTIME**-complete to undecidable for agents with perfect recall [13, 14]. Moreover, the
52 imperfect information semantics of **ATL** does not admit fixpoint characterizations [6, 11, 12],
53 which makes incremental synthesis of strategies difficult to achieve. Experimental studies
54 based on heuristic approaches [8, 15, 21, 9, 7] have also confirmed that the problem is hard,
55 and dealing with it requires innovative techniques.

56 In this talk, I present several very recent attempts at overcoming the complexity barrier
57 for model checking of **ATL**_{ir}, i.e., the **ATL** variant for memoryless strategies with imperfect
58 information. The new techniques include approximate verification by fixpoint computation
59 of upper- and lower bounds, model reduction based on locally defined model equivalence,
60 and partial order reduction for simple strategic abilities in asynchronous systems. The main
61 ideas are presented in the following sections, in a rather rudimentary form. For technical
62 details, the reader is referred to the original papers [16, 18, 3, 19].

63 3 Approximate Model Checking

64 The proposal in [16, 18] is based on the idea that, instead of the exact model checking, it
65 may suffice to provide a lower and an upper bound for the output. Given a formula φ , we
66 construct two translations $LB(\varphi)$ and $UB(\varphi)$ such that $LB(\varphi) \Rightarrow \varphi \Rightarrow UB(\varphi)$. That is, if
67 $LB(\varphi)$ is verified as true, then the original formula φ must also hold in the given model.
68 Conversely, if $UB(\varphi)$ evaluates to false, then φ must also be false.

#cards	#states	Approximate verification				Exact verification
		tgen	lower	upper	match	
4	11	0.0007	0.00007	0.00004	100%	0.12
8	346	0.011	0.0008	0.0003	100%	2.42 h*
12	12953	0.73	0.07	0.01	100%	timeout
16	617897	35.19	348.37	0.72	100%	timeout
20*	2443467	132.00	8815.73	4.216	100%	timeout

■ **Figure 2** Experimental results: solving endplay in bridge, formula $\langle\langle S \rangle\rangle_{ir} Fwin$ [18]

69 The construction of the upper bound is straightforward: instead of checking existence of
70 an imperfect information strategy, we look for a perfect information strategy that obtains
71 the same goal. If the latter is false, the former must be false too. The lower bound is
72 defined by a fixpoint expression in alternating epistemic mu-calculus, with a nonstandard
73 next-step strategic modality $\langle A \rangle^\bullet$ such that: (i) agents in A look for a short-term strategy
74 that succeeds from the “common knowledge” neighborhood of the current state (rather than
75 in the “everybody knows” neighborhood), and (ii) they are allowed to “steadfastly” pursue
76 their goal in a variable number of steps within the indistinguishability class. Formally, the
77 upper and the lower bounds are derived from φ through the following translations:

$$\begin{array}{ll}
LB(p) = p, & UB(p) = p, \\
LB(\neg\phi) = \neg UB(\phi), & UB(\neg\phi) = \neg LB(\phi), \\
LB(\phi \wedge \psi) = LB(\phi) \wedge LB(\psi), & UB(\phi \wedge \psi) = UB(\phi) \wedge UB(\psi), \\
78 LB(\langle A \rangle \phi) = \langle A \rangle LB(\phi), & UB(\langle A \rangle \phi) = E_A \langle\langle A \rangle\rangle_{ir} X UB(\phi), \\
LB(\langle\langle A \rangle\rangle G\phi) = \nu Z. (C_A LB(\phi) \wedge \langle A \rangle^\bullet Z), & UB(\langle\langle A \rangle\rangle G\phi) = E_A \langle\langle A \rangle\rangle_{ir} G UB(\phi), \\
LB(\langle\langle A \rangle\rangle \psi \cup \phi) = \mu Z. (E_A LB(\phi) \vee (C_A LB(\psi) \wedge \langle A \rangle^\bullet Z)) & UB(\langle\langle A \rangle\rangle \psi \cup \phi) = E_A \langle\langle A \rangle\rangle_{ir} UB(\psi) \cup UB(\phi)
\end{array}$$

79 ► **Theorem 1** ([16]). *For every pointed model (M, q) and \mathbf{ATL}_{ir} formula φ :*

$$80 \quad M, q \models LB(\varphi) \implies M, q \models \varphi \implies M, q \models UB(\varphi).$$

81 The effectiveness of the approximations has been evaluated experimentally on a number of
82 benchmarks. The results for a scalable scenario of *Bridge endplay* are presented in Figure 2.
83 The results in each row are averaged over 20 randomly generated instances, except for (\star)
84 where only 1 hand-crafted instance was used. All the tests were conducted on a computer
85 with an Intel Core i7-6700 CPU with dynamic clock speed of 2.60–3.50 GHz, 32 GB RAM,
86 running 64bit Ubuntu 16.04 Linux. The running times are given in seconds. *Timeout*
87 indicates that the process did not terminate in 48 hours. The computation of the lower
88 and upper approximations was done with a straightforward implementation (in Python 3)
89 of the fixpoint model checking algorithm. The exact \mathbf{ATL}_{ir} model checking was done with
90 MCMAS 1.3.0 [20].

91 Notice that the results in Figure 2 have been obtained by a completely straightforward
92 implementation of the lower/upper bound algorithms. After a simple optimization of data
93 structures (based on the technique of *merge-find sets*) and operations on the data, the
94 algorithms were able to generate and verify a model with over 70 million states in less than
95 75 minutes, cf. Figure 3. Perhaps more importantly, the verification of the handpicked $(5, 5)^\star$
96 model (which marked the limit of our capability without the optimization) ran almost 3000
97 times (!) faster than with the straightforward implementation. This strongly suggests that
98 the potential for further improvement is still large.

#cards	#states	Approximate verification				Exact verification
		tgen	lower	upper	match	
4	11	<0.0001	<0.0001	<0.0001	100%	0.12
8	346	<0.0001	<0.0001	<0.0001	100%	2.42 h*
12	12953	0.06	<0.0001	<0.0001	100%	timeout
16	617897	4.64	0.56	0.26	100%	timeout
20*	2443467	34.00	3.0	2.0	100%	timeout
20	1.5 e7	124.00	8.5	6.0	100%	timeout
24*	7 e7	3779.00	667.0	78.0	100%	timeout

■ **Figure 3** Experimental results for the optimised approximate verification [18]

99 4 Bisimulation-Based Model Reduction

100 The main source of hardness in model checking for strategic ability is the size of the model.
 101 First, the space of available strategies is at least exponential in the number of states and
 102 transitions. To make it even worse, there is (as of now) no method to factorize the model so
 103 that the algorithm would look for optimal substrategies independently, and then combine
 104 them into a winning strategy. Secondly, the explicit state model is typically exponential in
 105 size with respect to a higher level description, e.g., by means of local automata or concurrent
 106 programs. This means that realistic models are huge, and their strategy spaces are enormous.
 107 In consequence, an effective model reduction can offer invaluable help in turning a hopeless
 108 verification task into a feasible one.

109 One way to obtain such reductions is to identify a suitable notion of model equivalence
 110 that preserves the logic. Then, whenever the need for verification arises, we can look for
 111 a smaller model that is provably equivalent to the input model. Locally definable model
 112 equivalences for logics of strategies are usually called *alternating bisimulations*. The first
 113 variant of alternating bisimulation for **ATL** with imperfect information has been recently
 114 proposed in [3]. The main definitions and preservation theorems are summarized below.

115 **Strategy simulators.** Let M, M' be two models of **ATL**_{ir}. A *simulator of partial strategies*
 116 *for coalition A* from M into M' is a family **ST** of functions $\mathbf{ST}_{Q,Q'} : PStr_A(Q) \rightarrow PStr_A(Q')$
 117 for some subsets of states $Q \subseteq St$ in model M and $Q' \subseteq St'$ in model M' . Intuitively,
 118 every $\mathbf{ST}_{Q,Q'}$ maps each partial strategy σ_A defined on set Q in M into a “corresponding”
 119 strategy σ'_A defined on Q' in M' . Typically, we will map strategies between the common
 120 knowledge neighborhoods of “bisimilar” states in M and M' . We formalize this idea as
 121 follows. Let $R \subseteq St \times St'$ be a relation between states in M and M' . A *simulator of*
 122 *partial strategies for coalition A with respect to relation R* is a family **ST** of functions
 123 $\mathbf{ST}_{C_A(q),C'_A(q')} : PStr_A(C_A(q)) \rightarrow PStr_A(C'_A(q'))$ such that qRq' .

124 **Simulation and bisimulation.** Let M, M' be two models defined on the same set Agt of
 125 agents, and $A \subseteq \text{Agt}$ be a coalition. A relation $\Rightarrow_A \subseteq St \times St'$ is a *simulation for A* iff

- 126 1. There exists a simulator **ST** of partial strategies for A w.r.t. \Rightarrow_A , such that $q \Rightarrow_A q'$
 127 implies that:
- 128 a. $\pi(q) = \pi'(q')$;
 - 129 b. for every $i \in A$, $r' \in St'$, if $q' \sim'_i r'$ then for some $r \in St$, $q \sim_i r$ and $r \Rightarrow_A r'$.
 - 130 c. For every states $r \in C_A(q)$, $r' \in C'_A(q')$ such that $r \Rightarrow_A r'$, for every partial strategy
 131 $\sigma_A \in PStr_A(C_A(q))$, and every state $s' \in \text{succ}(r', \mathbf{ST}(\sigma_A))$, there exists a state
 132 $s \in \text{succ}(r, \sigma_A)$ such that $s \Rightarrow_A s'$.

133 2. If $q_1 \Rightarrow_A q'$ and $q_2 \Rightarrow_A q'$, then $C_A(q_1) = C_A(q_2)$.

134 That is, state q' can only simulate q if (1a) q and q' agree on the interpretation of atoms;
 135 (1b) q simulates the epistemic transitions from q' ; and (1c) for every partial strategy σ_A ,
 136 defined on the common knowledge neighborhood $C_A(q)$, we are able to find some partial
 137 strategy $\mathbf{ST}(\sigma_A)$ (the same for all states in $C_A(q)$) such that the transition relations $\xrightarrow{\mathbf{ST}(\sigma_A)}$
 138 and $\xrightarrow{\sigma_A}$ commute with the simulation relation \Rightarrow_A . Moreover, (2) multiple states simulated
 139 by the same q' must be in the same common knowledge neighborhood.

140 Relation \Leftrightarrow_A is a *bisimulation* iff both \Leftrightarrow_A and its converse $\Leftrightarrow_A^{-1} = \{(q', q) \mid q \Leftrightarrow_A q'\}$
 141 are simulations.

142 ► **Theorem 2** (Preservation Theorem for $A\text{-ATL}_{\text{ir}}$ [3]). *If \Leftrightarrow_A is a bisimulation for A
 143 and $q \Leftrightarrow_A q'$, then for every formula φ that contains only agents from A inside strategic
 144 modalities, we have:*

145 $M, q \models \varphi$ if and only if $M', q' \models \varphi$.

146 ► **Theorem 3** (Preservation Theorem for ATL_{ir} [3]). *If \Leftrightarrow is a bisimulation for every
 147 $A \subseteq \text{Agt}$, and $q \Leftrightarrow q'$, then for every formula φ we have that:*

148 $M, q \models \varphi$ if and only if $M', q' \models \varphi$.

149 The bisimulation provides a strong notion of model equivalence, since it preserves the
 150 truth values of all ATL_{ir} formulae. Moreover, it can lead to *very significant model reductions*.
 151 As an example, Figure 4a presents a fragment of the simple model of voting and coercion [18]
 152 for 1 coercer, 2 voters, and 2 candidates. A bisimilar, much sparser model is presented in
 153 Figure 4b.

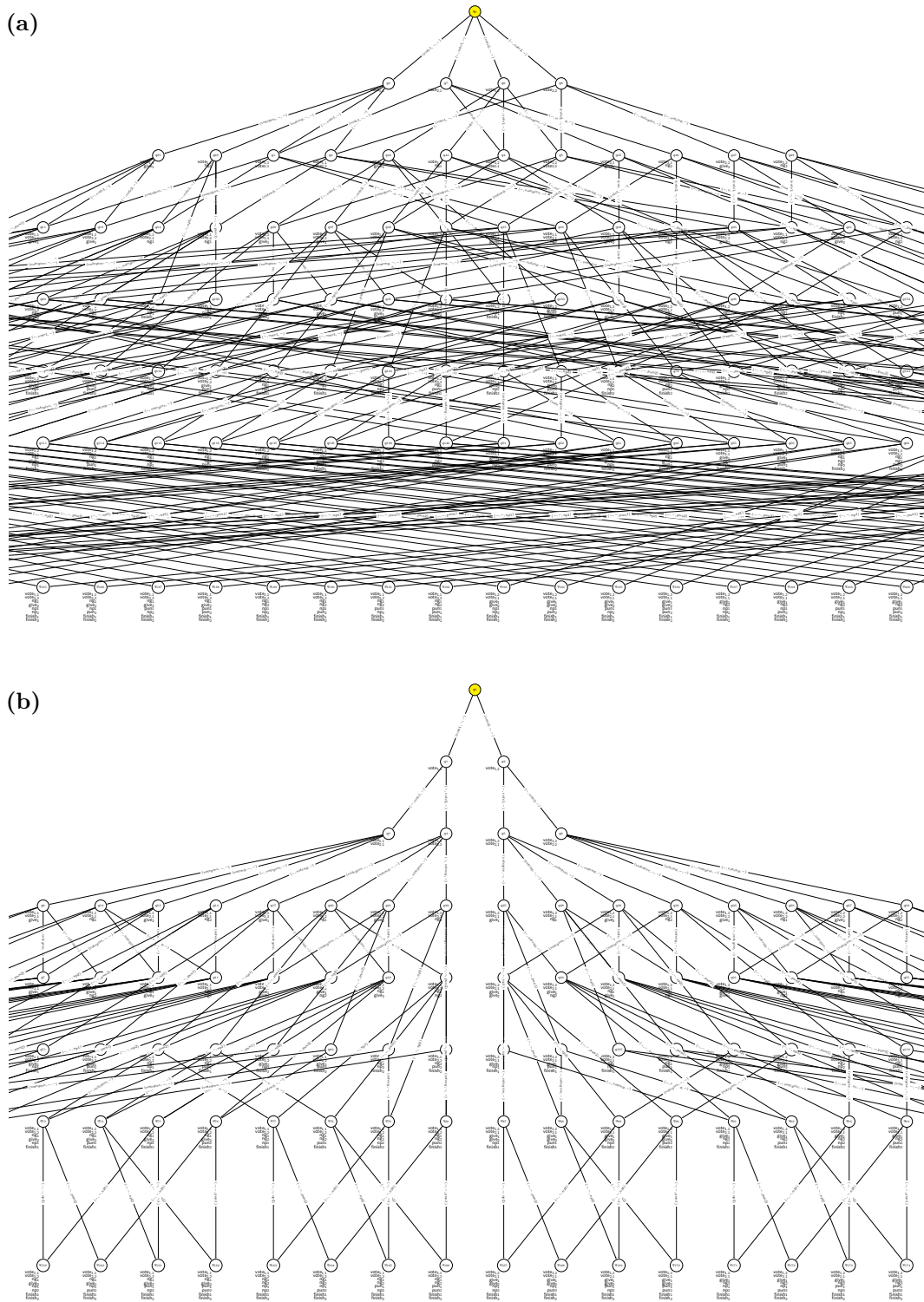
154 5 Partial-Order Reduction

155 The bisimulation-based reduction can result in a significant trimming of states and transitions.
 156 However, the conditions of the bisimulation are quite strong, which means that the method
 157 has somewhat limited applicability. Moreover, the reduced model and the equivalence must
 158 be crafted and proved by hand. An automated method for reduction of models *that arise
 159 due to interleaving of asynchronously executed actions* has been recently proposed in [19].

160 Many real-life systems are inherently asynchronous, and do not operate on a global
 161 clock that perfectly synchronizes the atomic steps of all the components. As an example,
 162 consider robots interacting in an environment with faulty communication or non-negligible
 163 delays in execution of actions. No less importantly, many systems that are synchronous at
 164 the implementation level (say, the level of the virtual machine) can be more conveniently
 165 modeled as asynchronous on a more abstract level, because when we abstract away the
 166 implementation details it is not clear anymore how transitions initiated by different agents are
 167 exactly arranged in a particular temporal order. For instance, the actual implementation of
 168 a soccer match in the simulated RoboCup competition can be executed on a single computer
 169 with a global clock ticking every 0.3 ns, but the corresponding synchronous model would be
 170 huge and in consequence useless for analysis. Instead, one can remove a lot of unnecessary
 171 details by assuming that the players execute their actions asynchronously – without clear
 172 temporal relationship between their execution times – and synchronize only when a particular
 173 event has to be executed *jointly*.

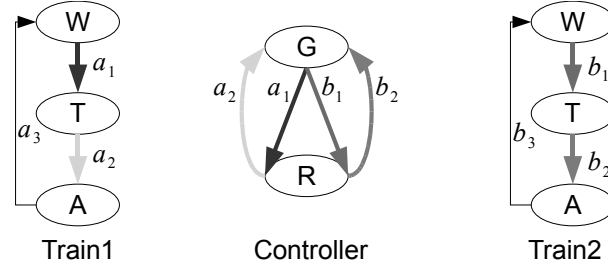
174 In many scenarios, both aspects combine. For example, when modeling an election, one
 175 must take into account both the truly asynchronous nature of events happening at different

3:6 Model Checking Strategic Ability



■ **Figure 4** Bisimulation-based reduction of a simple voting model: before and after the reduction

176 polling stations, and the best level of granularity for modeling the events happening within a
177 single polling station.



■ **Figure 5** Asynchronous multi-agent system for the Trains, Gate and Controller benchmark

178 In [19], we have made the first step towards strategic analysis of such systems. Most
 179 importantly, we adapted *partial order reduction (POR)* to model checking of strategic abilities
 180 for agents with imperfect information. In fact, we showed that the most efficient variant of
 181 POR, defined for linear time logic **LTL**, can be applied almost directly. Interestingly, the
 182 scheme does *not* work for verification of agents with perfect information.

183 **Algorithm for partial-order reduction.** Given a collection of local automata \mathcal{S} with its
 184 underlying model M , the reduction proceeds by iterative unfolding of \mathcal{S} into its reduced
 185 model M' . The unfolding is done by means of the following Depth-First Search (DFS)
 186 algorithm:

- 187 1. Identify the set $en(q_n) \subseteq Act$ of enabled actions.
- 188 2. Heuristically select a subset $E(q_n) \subseteq en(q_n)$ of possible actions (see below).
- 189 3. For any action $a \in E(q_n)$, compute the successor state q' such that $q_n \xrightarrow{a} q'$, and add q'
 190 to the stack, thus generating the path $\pi' = q_0 a_0 q_1 a_1 \dots q_n a q'$. Recursively proceed to
 191 explore the submodel originating at q' .
- 192 4. Remove q_n from the stack.

193 The algorithm begins with the stack consisting solely of the initial state of M , and terminates
 194 when the stack is empty.

195 **Heuristics.** Let $A \subseteq \text{Agt}$. The conditions **C1** – **C3** below define a heuristics for the
 196 selection of $E(q) \subseteq en(q)$ in the DFS algorithm.

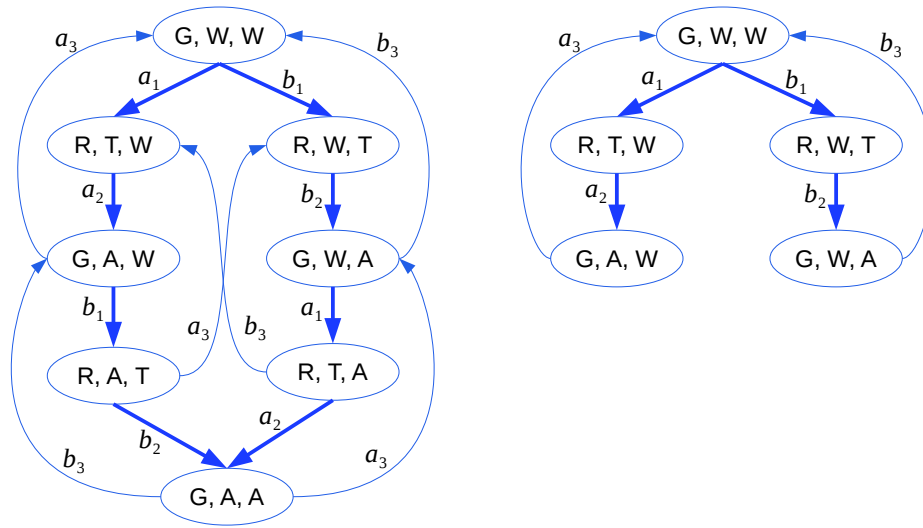
197 **C1** Along each path π in M that starts at q , each action that is dependent on an action in
 198 $E(q)$ cannot be executed in π without an action in $E(q)$ is executed first in π . Formally,
 199 $\forall \pi \in \Pi_M(q)$ such that $\pi = q_0 a_0 q_1 a_1 \dots$ with $q_0 = q$, and $\forall b \in Act$ such that $(b, c) \notin I_A$
 200 for some $c \in E(q)$, if $a_i = b$ for some $i \geq 0$, then $a_j \in E(q)$ for some $j < i$.

201 **C2** If $E(q) \neq en(q)$, then $E(q) \subseteq Invis_A$.

202 **C3** For every cycle in M' there is at least one node q in the cycle for which $E(q) = en(q)$,
 203 i.e., for which all the successors of q are expanded.

204 **Preservation theorems.** Let \mathcal{S} be a collection of local automata representing an asynchron-
 205 ous multi-agent system. I_\emptyset is the standard independence of actions, used in partial-order
 206 reduction for **LTL** [10]. I_A is a slightly modified variant of I_\emptyset , that treats all the actions of
 207 agents in A as visible. The following theorems show that the **LTL** partial-order reduction
 208 can be directly (or almost directly) applied to **sATL_{ir}^{*}**, i.e., the fragment of **ATL_{ir}^{*}** without
 209 nested strategic modalities and the temporal operator X .

210 ► **Theorem 4** ([19]). *Let M be the model of \mathcal{S} , and M' be the reduced model generated by*
 211 *DFS with the choice of $E(q')$ for $q' \in St'$ given by conditions **C1**, **C3** and the independence*
 212 *relation I_\emptyset . Then, M' satisfies exactly the same formulae of **sATL_{ir}^{*}** as M under the*
 213 *concurrency-fairness assumption.*



■ **Figure 6** Interleaved interpreted systems for TGC: (a) full model, (b) reduced model. Visible transitions are depicted by bold arrows

214 ► **Theorem 5** ([19]). *Let $A \subseteq \text{Agt}$, M be the model of \mathcal{S} , and M' the reduced model generated*
 215 *by DFS with the choice of $E(q')$ given by conditions **C1**, **C2**, **C3** and the independence*
 216 *relation I_A . Then, M' satisfies exactly the same formulae of $A\text{-sATL}_{\text{ir}}^*$ as M without the*
 217 *concurrency-fairness assumption.*²

218 How big is the gain? As an example, consider the well known TGC benchmark (Trains,
 219 Gate, and Controller). The local automata representing the system for $k = 2$ trains are
 220 shown in Figure 5, and the full logical model in Figure 6a. The reduced model obtained by
 221 POR is depicted in Figure 6b (same as for the **LTL** partial-order reduction). It is easy to
 222 see that the reduced state space is exponentially smaller than the size of the full model.

223 Of course, such optimistic outcomes are not guaranteed for all asynchronous agent systems.
 224 Still, it is important to note that **ATL**_{ir} model checking is **NP**-hard *in the size of the model*
 225 (and not the size of the representation), and all the attempts at actual algorithms so far run
 226 in exponential time. So, even a linear reduction of the state space is likely to produce an
 227 exponential improvement of the performance.

228 **Perfect information.** The reduction scheme does not work for the standard variant of
 229 alternating-time logic, based on perfect information strategies:

230 ► **Theorem 6** ([19]). *The analogues of Theorems 4 and 5 do not hold for **sATL**_{Ir}.*

231 This negative result is especially interesting because, until now, virtually all the results
 232 have been in favor of solving games with perfect information.

233 6 Conclusions

234 Verification by model checking is one of the top success stories in computer science and AI.
 235 Many important properties of multi-agent systems are underpinned by the ability of some
 236 agents (or groups) to achieve a given goal. However, model checking of strategic ability in

² where $A\text{-sATL}_{\text{ir}}^*$ is the fragment of $\text{sATL}_{\text{ir}}^*$ containing only agents from A in strategic modalities.

237 realistic systems is a notoriously hard problem. In this short paper, I have tried to summarize
 238 some of the very recent developments that can contribute to overcoming the complexity
 239 barrier, and extending the scope of formal verification.

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