

Don't Trust Your Agents, Verify Them

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OUTLINE

① INTRODUCTION AND MOTIVATION

- WHY STRATEGIC VERIFICATION FOR MAS?
- FROM TEMPORAL PROPERTIES TO STRATEGIC ABILITIES.
- DOMAINS AND CHALLENGES.

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- ARCHITECTURE & COMPONENTS.
- SUPPORTED LOGICS AND TASKS.
- VERIFICATION WORKFLOWS.

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⑤ DISCUSSION AND OUTLOOK

- OPEN RESEARCH CHALLENGES, INDUSTRIAL RELEVANCE, Q&A.

INTRODUCTION AND MOTIVATION

PREFACE

System Correctness

- Safety-critical and autonomous systems increasingly rely on **multiple interacting decision-makers**.
- Failures are rarely isolated; they emerge from **coordination** and **interaction**.

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Formal Verification

- Provides mathematically precise models of behaviour.
- Enables reasoning about **all possible executions**.
- Model checking is a key technique for exploring system behaviour.

MODEL CHECKING (1)

$$M \models \varphi$$

There are three fundamental parts:

- M : modeling a multi-agent system;
- φ : specifying a property;
- $\models$: verifying that the model M satisfies the property φ .

MODEL CHECKING (2)

Model

A Kripke structure is a tuple $K = \langle AP, St, S_0, R, L \rangle$, where:

- AP is a set of atomic propositions;
- St is a set of states;
- $S_0 \subseteq St$ is a set of initial states;
- $R \subseteq St \times St$ is a transition relation;
- $L : St \rightarrow 2^{AP}$ is a labeling function.

MODEL CHECKING (3)

Specification

- Temporal logics allow to describe the order of the events without define the time explicitly.
- The main temporal logics:
 - LTL $\Rightarrow$ in a computation the events are totally ordered.
Examples: Xp , Fp , Gp .
 - CTL $\Rightarrow$ in a computation the events are partially ordered.
Examples: AXp , EFp , AGp .

MODEL CHECKING (4)

Verification

Given a Kripke structure K and a specification φ , the problem of the model checking consists to verify if $K, s \models \varphi$, for each initial state s .

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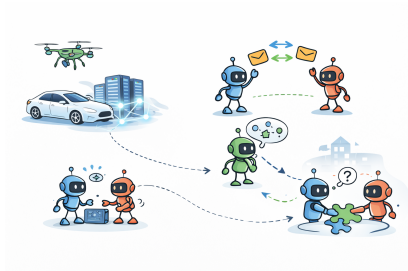
The challenge is scaling correctness to interacting agents.

WHY VERIFYING MULTI-AGENT SYSTEMS IS HARD

- Modern systems are inherently **multi-agent**:
 - autonomous vehicles and robots
 - algorithmic trading systems
 - distributed protocols and platforms

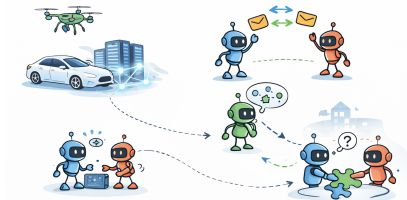
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Correctness depends on interactions, not just components.

LOCAL CORRECTNESS IS NOT ENOUGH

Observation

Most failures in distributed and multi-agent systems are not caused by incorrect local behaviour, but by **unexpected interactions**.

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The real difficulty lies in composition.

FROM INTERACTION TO STRATEGIC BEHAVIOUR

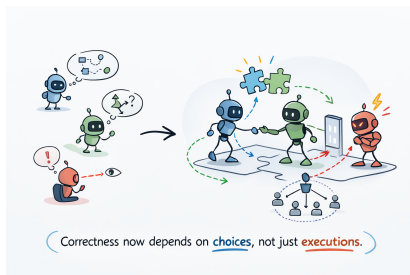
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 - adversarial behaviour
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WHY WE START FROM REAL FAILURES

- Interaction failures are often:
 - rare
 - emergent
 - invisible to testing

WHY WE START FROM REAL FAILURES

- Interaction failures are often:
 - rare
 - emergent
 - invisible to testing
- They expose:
 - broken assumptions
 - unsafe compositions
 - missing strategic guarantees



Failures reveal what models and specifications must capture.

POLICY COMPOSITION FAILURE: BGP HIJACK

Domain: Internet routing

Year: 2008

What happened

A local routing policy accidentally advertised a more specific prefix for YouTube, which propagated globally and redirected traffic worldwide.

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Verification lesson

Global safety requires verifying **policy composition** among agents.

STRATEGIC FEEDBACK LOOPS: FLASH CRASH

Domain: Algorithmic markets

Year: 2010

What happened

A large automated sell program interacted with high-frequency trading strategies, triggering a rapid market collapse and rebound within minutes.

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Verification lesson

Safety depends on the **composition of strategies**, not on isolated agent.

ALGORITHMIC TRADING: PARTIAL DEPLOYMENT

Domain: Financial markets

Year: 2012

What happened

A faulty software update activated obsolete trading logic on one server. Within minutes, millions of unintended orders were issued, causing massive losses.

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Verification lesson

Correctness must be verified under **heterogeneous agents and mixed versions**, not idealised uniform deployments.

SMART CONTRACTS: STRATEGIC EXPLOITATION (THE DAO)

Domain: Blockchain / smart contracts

Year: 2016

What happened

An attacker exploited a re-entrancy vulnerability to repeatedly withdraw funds before global balances were updated, draining a large fraction of the contract.

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Verification lesson

Strategic reasoning is essential to guarantee global invariants against **adversarial strategies**.

AUTONOMOUS DRIVING: CONFLICTING AGENTS (UBER)

Domain: Autonomous vehicles

Year: 2018

What happened

An autonomous vehicle struck a pedestrian while operating in automated mode. The perception system repeatedly misclassified the pedestrian, while emergency braking was disabled, relying on human intervention.

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Verification lesson

Safety depends on **authority and strategy composition** between human and autonomous agents, not on any single controller.

CROSS-CASE LESSONS

- Failures emerge from **interaction**, not local bugs
- Partial information and concurrency are the norm
- Strategic and adversarial behaviour cannot be ignored
- Assumptions do not compose automatically



FROM FAILURES TO FORMAL REASONING

- The case studies show that failures arise from:
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Key question

How can we **systematically reason** about *all possible behaviours* of interacting agents?

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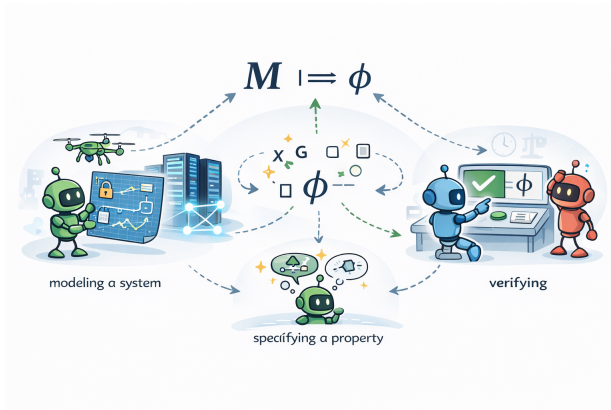
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This is the role of model checking.

MODEL CHECKING



BACKGROUND: STRATEGIC LOGICS AND MODELS

MULTI-AGENT SYSTEMS (1)

Key aspects

- There are many agents interacting among them.
- Each agent selects a *strategy*.
- A *strategy* is a conditional plan that prescribes an action at each step of the Multi-Agent System (MAS).
- The composition of strategies induces a computation.

MULTI-AGENT SYSTEMS (2)

Model

- A MAS structure is a tuple $M = \langle AP, St, s_I, Ac, Ag, L, tr \rangle$, where:
 - AP is a set of atomic propositions;
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 - tr is a transition function.
- The function tr depends on the interaction between the agents:
 - *Turn-based* $\Rightarrow$ The states of the MAS are partitioned between the agents, then the owner of a state determines the move to take and thus the next state of the MAS.
 - *Concurrent* $\Rightarrow$ The agents choose a move (i.e, actions) simultaneously and independently, and the joint action (i.e. the choices together) determine the next state of the MAS.

MULTI-AGENT SYSTEMS (3)

Strategies

Depending on the memory, we distinguish between:

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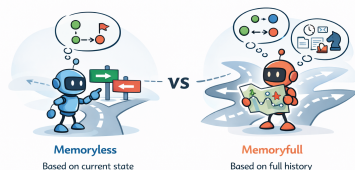
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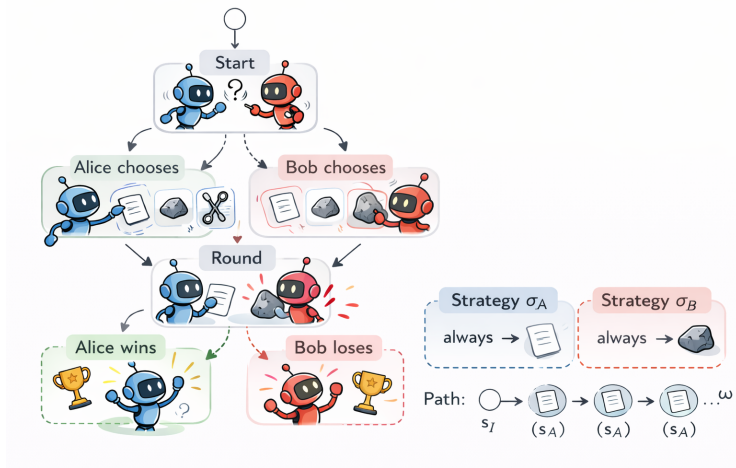
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Decisions depend on the full history of the MAS.



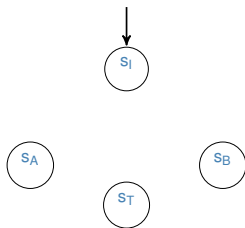
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EXAMPLE: PAPER, ROCK, AND SCISSOR (2)

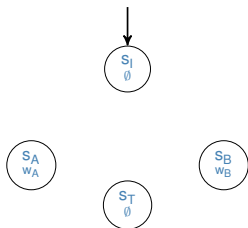
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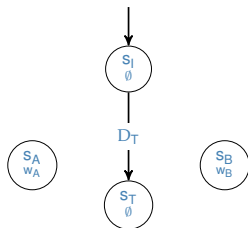
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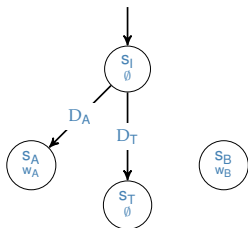
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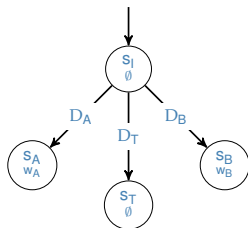
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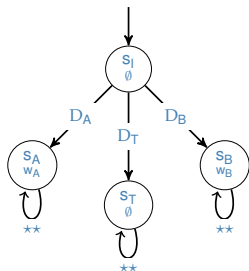
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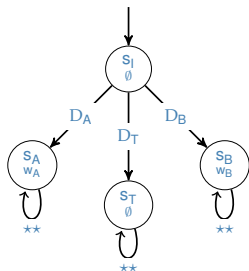
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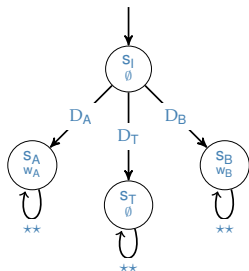
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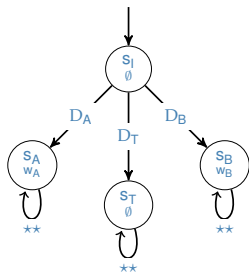
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- σ_A and σ_B induce the path $s_I \cdot (s_A)^\omega$

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Visibility of the agents

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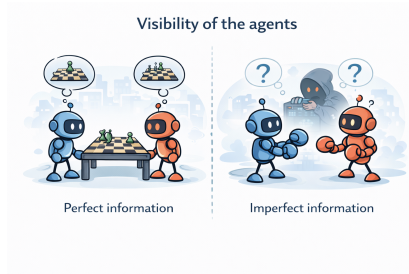
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Visibility of the agents

Depending on the visibility of the agents, we distinguish between:

- *Perfect information* MAS
Agents have full knowledge of the MAS evolution at all times.
- *Imperfect information* MAS
Agents often have to decide without all relevant information.



SPECIFICATIONS

Alternating-Time Temporal Logic (ATL^*) [AHK02]

- ATL^* generalizes CTL^* by replacing the path quantification with strategy quantification over coalition of agents.
- It uses the strategy modality $\langle\langle A \rangle\rangle$ and $\llbracket A \rrbracket$, with A set of agents.
- The strategies are implicit.

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Example

$\langle\langle \{\alpha, \beta\} \rangle\rangle G \neg fail$: agents α and β have a collective strategy that prevents system failure regardless of other agents' strategies.

VERIFICATION: MODEL CHECKING COMPLEXITIES

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ATL

Strategies \ Information	perfect	imperfect
memoryless	<i>P</i> TIME-complete [AHK02]	<i>P</i> ^{NP} -complete [Sch03]
memoryfull	<i>P</i> TIME-complete [AHK02]	undecidable [DT11]

VERIFICATION: MODEL CHECKING COMPLEXITIES

ATL

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memoryless	P TIME-complete [AHK02]	P^{NP} -complete [Sch03]
memoryfull	P TIME-complete [AHK02]	undecidable [DT11]

ATL*

Strat \ Info	perfect	imperfect
memoryless	$PSPACE$ -complete [Sch03]	$PSPACE$ -complete [Sch03]
memoryfull	$2EXPTIME$ -complete [AHK02]	undecidable [DT11]

SPECIFICATIONS

Strategy Logic (SL) [MMPV14]

- SL treats the strategies as **first-order objects**.
- SL extends LTL with:
 - *strategy quantifiers* $\langle\langle x \rangle\rangle$ and $\llbracket x \rrbracket$ read respectively as “*there exists a strategy x* ” and “*for all strategies x* ”;
 - *agent binding* (a, x) that links an agent a to a strategy associated with a variable x .
- It subsumes other logics for strategic reasoning such as ATL*.
- It allows to represent **Nash equilibrium** and similar concepts.

SPECIFICATIONS

Strategy Logic (SL) [MMPV14]

- SL treats the strategies as **first-order objects**.
- SL extends LTL with:
 - *strategy quantifiers* $\langle\langle x \rangle\rangle$ and $\llbracket x \rrbracket$ read respectively as “*there exists a strategy x* ” and “*for all strategies x* ”;
 - *agent binding* (a, x) that links an agent a to a strategy associated with a variable x .
- It subsumes other logics for strategic reasoning such as ATL*.
- It allows to represent **Nash equilibrium** and similar concepts.

Example

$\langle\langle x \rangle\rangle \langle\langle y \rangle\rangle \llbracket z \rrbracket (\alpha, x)(\beta, y)(\gamma, z)(G \neg fail)$: α and β employ strategies x and y , respectively, ensuring non-failure regardless of γ 's strategy z .

VERIFICATION: MODEL CHECKING COMPLEXITIES

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Strategy Logic

Strat \ Info	perfect	imperfect
memoryless	<i>PSPACE</i> -complete [MMMP21]	<i>PSPACE</i> -complete [MMMP21]
memoryfull	non-elementary [MMPV14]	undecidable [DT11]

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Enhance model checking complexity by reducing expressiveness

- SL[1G] model checking with perfect information and memoryfull strategies is *2EXPTIME*-complete [MMPV14].
- SL[SG] model checking with perfect information and memoryfull strategies is *PTIME*-complete [BJK⁺19].

WHY DO WE USE LOGICS FOR STRATEGIC REASONING?

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Example

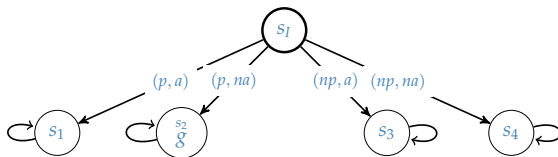
- 2 vehicles: A and B.
- B is in front of A.
- Goal: A wants to pass B (atom g).
- A Actions: pass (p), not pass (np).
- B Actions: accelerate (a), not accelerate (na).



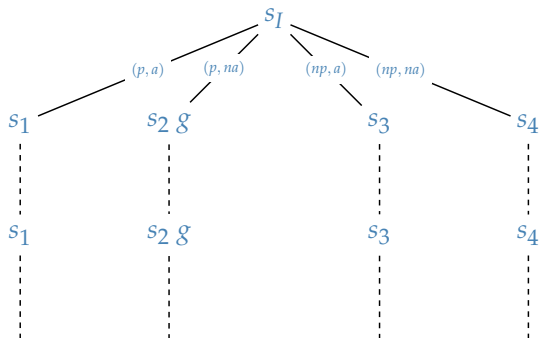
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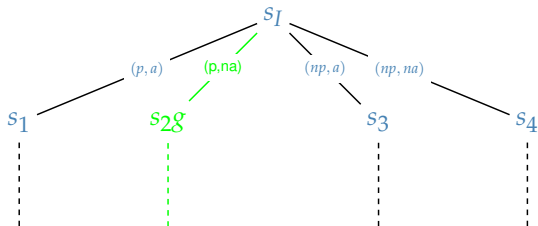


UNWINDING OF THE MODEL



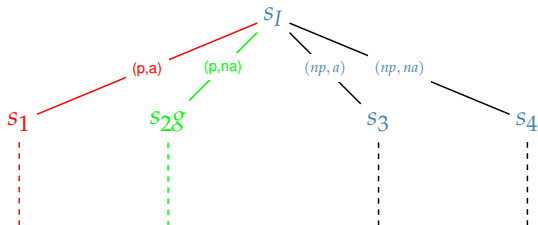
WHAT CAN WE DO WITH CTL?

- The formula $EF\ g$ is verified in the path: $s_I, s_2, \dots$



WHAT CAN'T WE DO WITH CTL?

- Is there a strategy for A to achieve g ?
 - if A select the action p then we need to verify the formula in the paths: $s_I, s_1, \dots$ (false) and $s_I, s_2, \dots$ (true);



COMPARISON BETWEEN CTL AND ATL

Summary

- The formula in CTL $EF\ g$ is true.
- The formula in ATL $\langle\langle A \rangle\rangle F\ g$ is false.
- In the model there is a path where A can pass B but there is no strategy for A to pass B!
- With CTL we cannot verify properties on subsets of paths.

STRATEGIC VERIFICATION IN VITAMIN

STATE OF THE ART: TOOL-CENTRED VERIFICATION

- Formal verification of Multi-Agent Systems (MAS) is supported by tools such as **MCMAS** and **STV**.
- These tools demonstrate that **strategic and epistemic verification is feasible in practice**.

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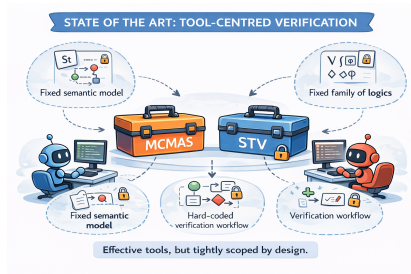
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 - a **hard-coded verification workflow**



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MCMAS

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Strategic abilities in game-like and decision-making scenarios.

WHAT THESE TOOLS DO WELL

- Provide **sound and efficient** verification within their scope
- Support widely studied logics (CTL, ATL, epistemic variants)
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Key observation

Existing tools are *highly effective* when used as intended.

EVIDENCE: FEATURE COMPARISON

Feature	MCMAS	STV	VITAMIN
GUI	(✓)	✓	✓
Internal documentation	(✓)	×	✓
External documentation	×	×	✓
Modular architecture	×	×	✓
Cross-platform	×	✓	✓
API / integration	×	×	✓
Support for non-experts	×	×	✓

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- **Takeaway:** current tools are *not designed* for extensibility and reuse across verification scenarios.

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 - logic
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- Limited support for:

- compositional verification
- systematic experimentation
- non-expert users

THE EMERGING GAP

Observation

Modern MAS research requires rapidly exploring *new combinations* of models, logics, and verification techniques.

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- Strategic reasoning is evolving faster than individual tools
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This calls for a framework, not another monolithic tool.

FROM LIMITATIONS TO DESIGN GOALS

Observed problems

- **Hard-coded** verification pipelines
- **Tight** model–logic **coupling**
- **High entry barrier** for non-experts
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- **Compositional** architecture
- **Independent** model and logic layers
- **Pluggable** verification engines
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VITAMIN ARCHITECTURE: CORE IDEA

Principle

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User roles

- End-users (expert / non-expert)
- High-level developers (models, logics)
- Low-level developers (optimisation)

SUPPORTED LOGICS: EVIDENCE OF GENERALITY

Logic	VITAMIN	MCMAS	STV
CTL	✓	✓	×
ATL	✓	✓	✓
SL	✓ ^{lim}	✓ ^{lim}	✓ ^{lim}
CTLK / ATLK	×	✓	✓ ^{lim}
ATLF	✓	×	×
RBATL	✓	✓	×
RABATL	✓	×	×
NatATL	✓	×	×
OL	✓	×	×
OATL	✓	×	×
NatOL	✓	×	×
capATL	✓	×	×
NatSL	✓	×	×
WATL	✓	×	×
TOL	✓	×	×
POL	✓	×	×

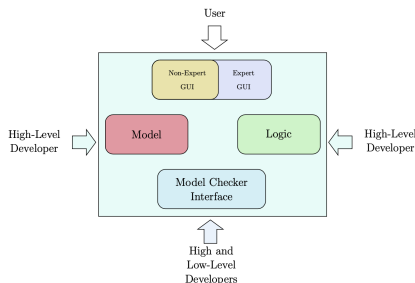
Legend: ✓ = native; ✓^{lim} = partial/limited; × = not supported.

VITAMIN IN ONE SENTENCE

VITAMIN is not just a model checker, but a compositional **framework** for **designing**, **experimenting** with, and **deploying** strategic verification solutions for Multi-Agent Systems.

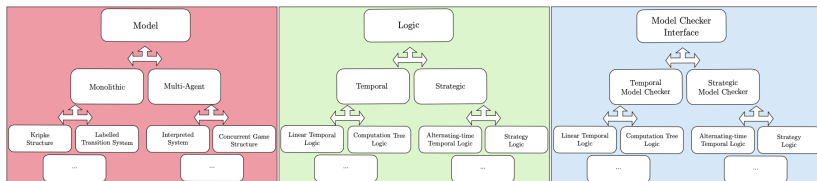
VITAMIN

- VITAMIN (VerIfication of A MultI-agent system) since **2022**
- Highly compositional and supports various logic and models
- User-friendly experience for both developers and end-users

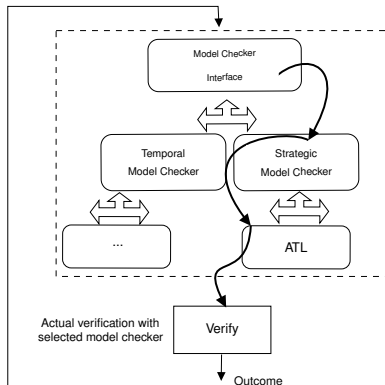
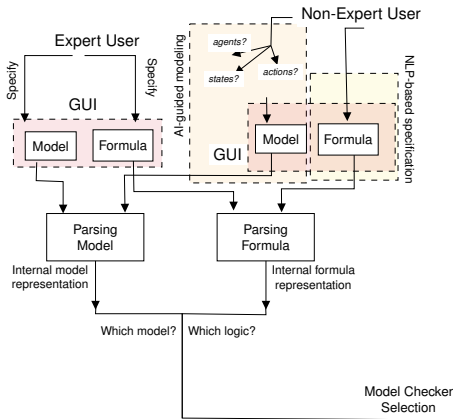


ADVANTAGES OF VITAMIN

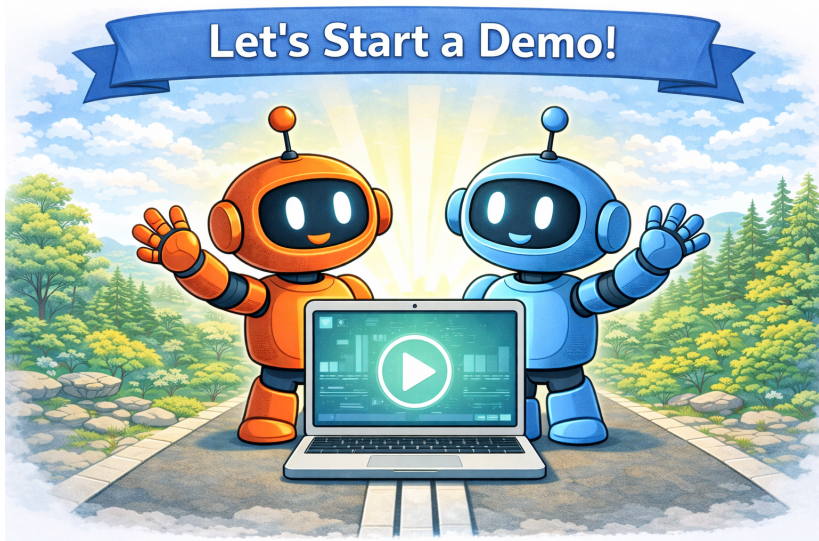
- Generalises MAS verification without being tied to specific logic or model formalisms
- Enhanced user experience guiding the entire verification process
- Compositional nature allows for straightforward extension



VERIFICATION FLOW



DEMO



FUTURE DIRECTIONS

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- **VITAMIN**, key goals we aim to achieve:
 - Use AI to assist users in generating formal models and formulas.
 - Low-level development with symbolic & abstraction methods.
 - Use AI to optimize the verification process, making it scalable.
 - Ensure the tool is robust, easily extensible, and deployable.

Thank you for the attention! 😊



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