

# OMEGA\*: An Ontology-Driven Tool for Explaining Multi-Agent Path Finding

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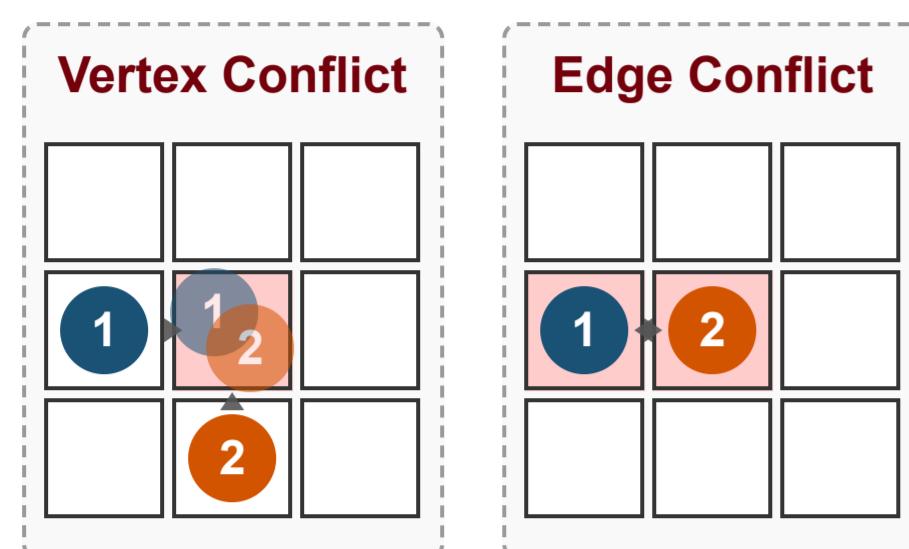
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## Motivation & Why OMEGA\*

### Problem

Multi-Agent Path Finding (MAPF) considers  $n$  agents on a graph  $G = (V, E)$ , each with start  $s_i$  and goal  $g_i$ . A solution  $\Pi = \{\pi_i\}$  assigns collision-free paths over discrete time, avoiding vertex conflicts ( $v_t^i = v_t^j$ ) and edge swaps ( $(v_t^i, v_{t+1}^i) = (v_{t+1}^j, v_t^j)$ ).



- Robots may **wait or detour** in ways that feel like a "black box" to operators
- Raw planner logs list cells and timesteps, but **don't encode causal links** (who blocked whom, what conflict triggered replanning)
- Existing explanation approaches provide only **partial coverage** (visual segmentation or logic-based queries, but not both)

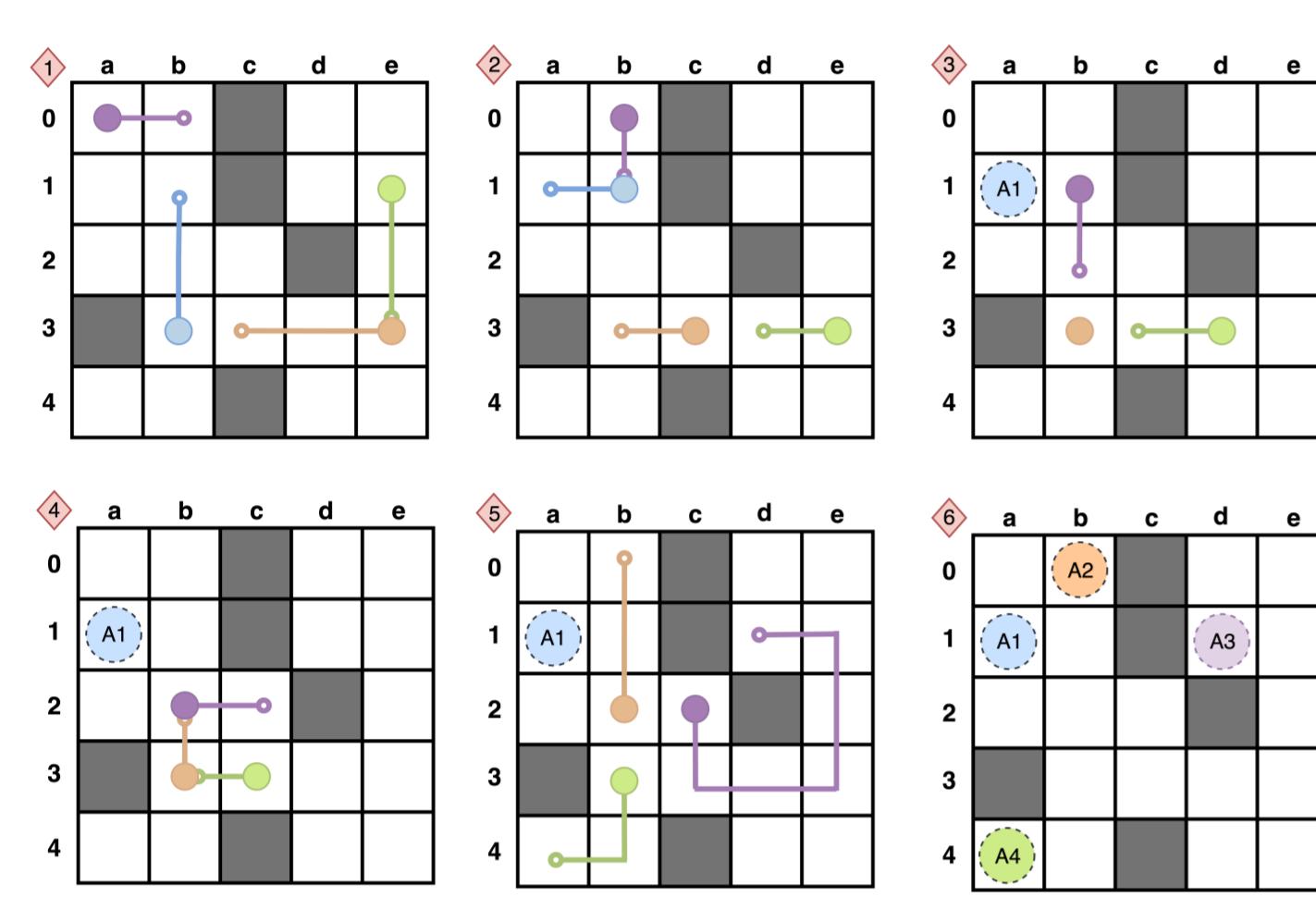


Figure 1: Plan Segmentation based Visualization

### Solution

OMEGA\* converts planner execution logs into a semantic knowledge graph using the Multi-Agent Planning Ontology (maPO):

- Users ask **queries**; OMEGA answers with grounded rationales and synchronized grid highlights
- Transforms opaque execution traces into **queryable, auditable explanations**
- Works across MAPF algorithms with minimal modifications for canonical JSON logging

### What's New

- Planner-agnostic normalization:** canonical JSON log schema for any MAPF solver
- Interoperable semantics:** reuses OWL-Time, PROV-O, SOSA/SSN, CORA standards
- Causal chain captured as events:** CollisionEvent → ConflictAlert → ReplanningStrategy → (Original/Resolved)SubPlan

## OMEGA\* Explanations for MAPF

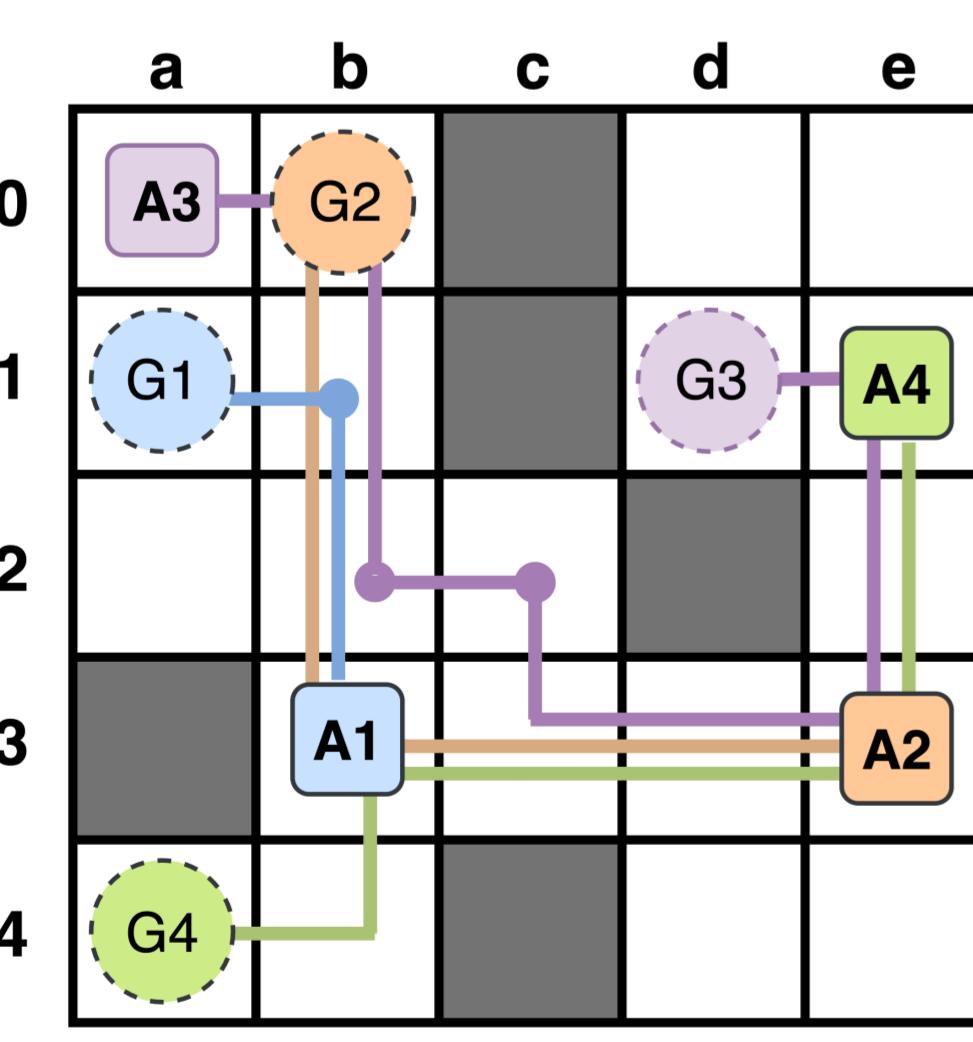


Figure 2: MAPF instance with 4 agents.

Question 1: What conflicts was Agent 3 in?

Answer 1: Agent 3 was involved in the following:

- $t=1$  — **vertex** conflict at  $(b1)$  with Agent 1.
- $t=3$  — **vertex** conflict at  $(b2)$  with Agent 2.

Question 2: How was the conflict at  $(b2)$  resolved?

Answer 2: A vertex conflict occurred at location  $(b2)$ . To resolve the conflict with **Agent 3**, **Agent 2** changed its plan.

New plan for Agent 2:

- $t=3$  — Wait at  $(b3)$  (Duration: 1 step).
- $t=4$  — Move to  $(b2)$ .
- $t=5$  — Replan complete.

Question 3: Provide a global plan summary.

Answer 3:

- Global Makespan: 12.0
- Sum of Costs: 28.0
- Replans Triggered: 2

Question 4: Why did Agent 3 wait at  $t=1$ ?

Answer 4: Agent 3 waited to avoid an **Vertex Conflict** with Agent 1 traversing  $(b2) \rightarrow (b1)$ .

## System Architecture

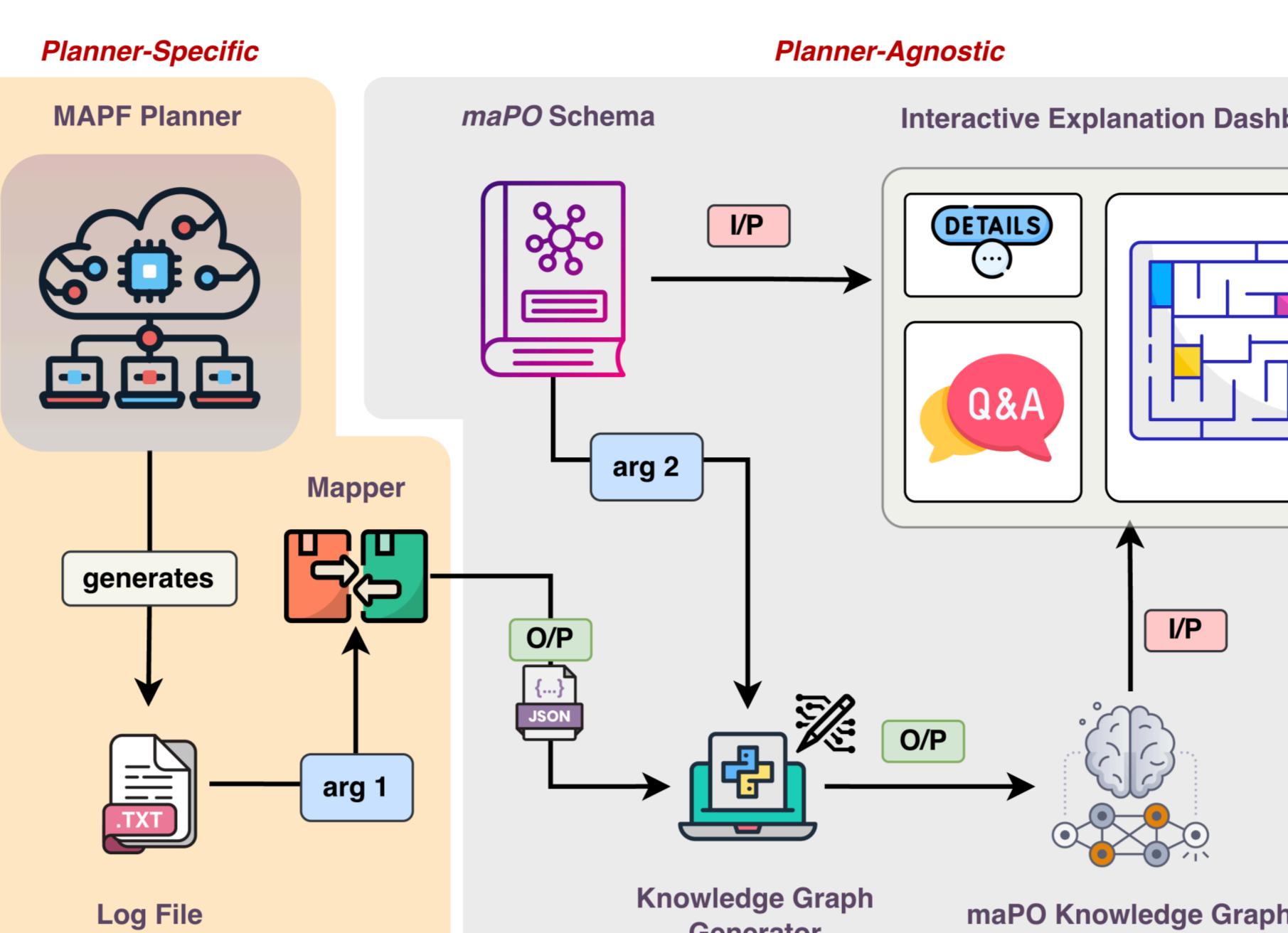


Figure 3: System architecture showing how planner-specific outputs are converted into the maPO knowledge graph via a generic layer, keeping the method planner-agnostic for interactive, ontology-driven explanations.

### Pipeline

The framework operates in three stages that ensure planner-agnostic operation:

- Log Generation:** Any MAPF planner (CBS, ICBS, RL) outputs a standardized JSON trace containing environment, agents, paths, collisions, and alerts
- Knowledge Graph Creation:** Python mapper asserts RDF triples consistent with maPO schema and exports RDF/Turtle
- Explanation Generation:** SPARQL queries over the knowledge graph; results populated into natural language templates
- Interactive Dashboard:** Text + visual overlays synchronized with grid simulation for inspection

### Planner Agnosticism: Canonical JSON

- Planner-Agnostic by Design:** OMEGA\* does not rely on solver internals; it works with any MAPF solver.
- Canonical JSON Export:** Solvers need to export a standardized execution trace (agents, paths, conflicts, alerts, and plan revisions).
- Universal Debugger:** A generic log→maPO mapper converts log traces into instantiated maPO knowledge graph, enabling SPARQL-based explanations.

```
{  
    "environment": {  
        "gridSize": [R, C],  
        "obstacles": [{ "id": obs_id, "cell": [r, c] }]  
    },  
    "agents": [  
        {  
            "id": agent_id,  
            "start": { "t": t0, "cell": [rs, cs] },  
            "goal": { "cell": [rg, cg] }  
        }],  
    "collisions": [  
        {  
            "id": coll_id,  
            "t": t,  
            "type": "vertex" | "edge",  
            "loc": [r, c],  
            "agents": [ai, aj]  
        }],  
    "alerts": [  
        {  
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            "conflict": coll_id,  
            "agent": agent_id,  
            "type": alert_type,  
            "reason": text  
        }],  
    "jointPlan": {  
        "subplans": [plan_id],  
        "makespan": T  
    }  
}
```

## Evaluation Highlights

User study (N=28) comparing maPO-generated explanations vs. raw planner logs across 3 scenarios (RL, CBS, ICBS):

94%  
PREFERENCE

4.44/5  
CLARITY RATING

- Users chose maPO explanations in **159 of 167** recorded preferences (Binomial Test,  $p < .001$ )
- Wilcoxon signed-rank tests confirmed median clarity ratings significantly above neutral

### Cognitive Load Metrics (N=18)

- Flesch Reading Ease (FRE):** 94.39 — "Very Easy" readability level
- Automated Readability Index (ARI):** 4.87 — Grade 4-5 comprehension
- Coleman-Liau Index (CLI):** 1.13 — Grade 1-2 level text

These scores confirm maPO explanations impose minimal linguistic burden and are well-suited for rapid comprehension.

## maPO Ontology: Core Concepts

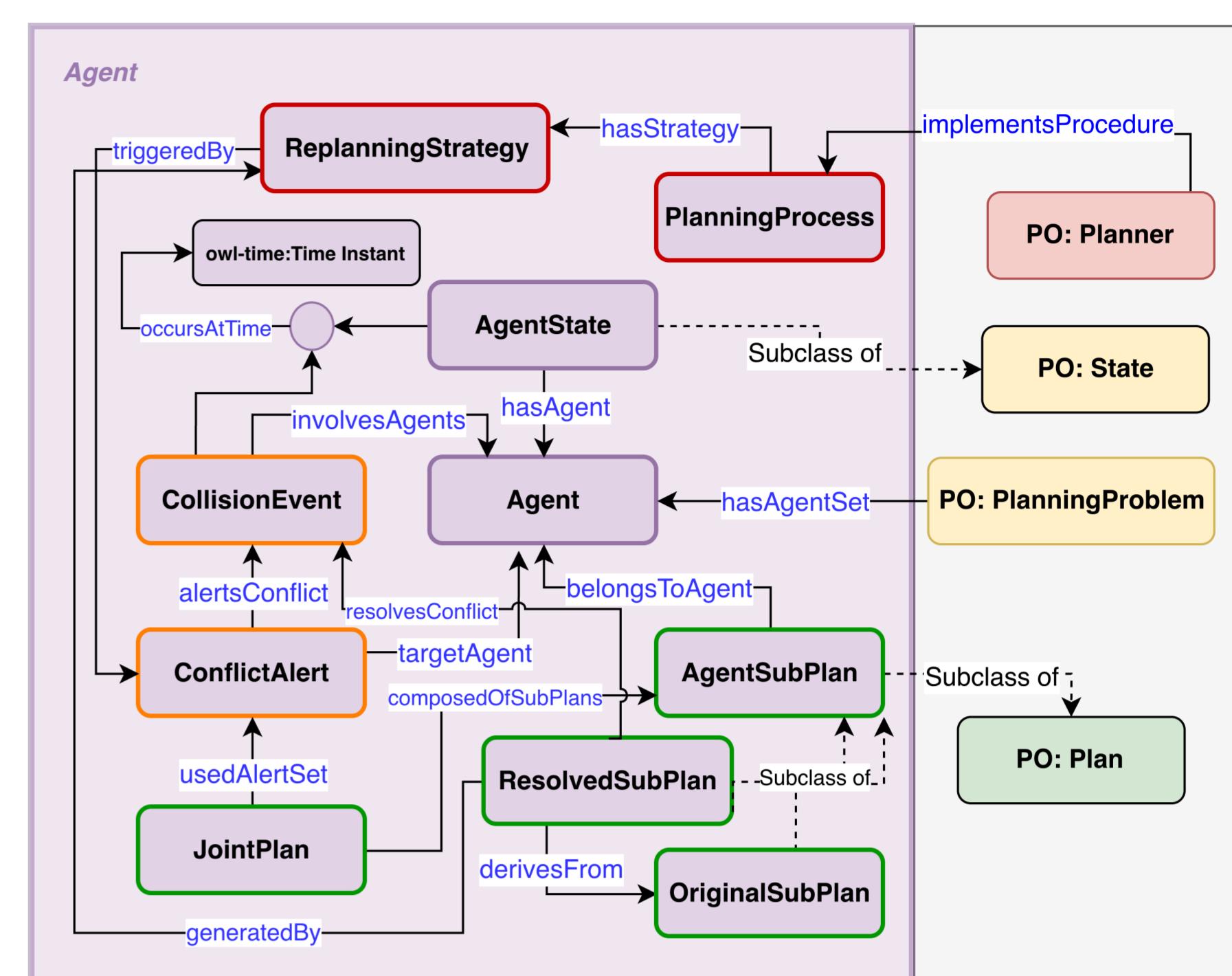


Figure 4: Illustrative Multi-Agent Planning Ontology (maPO) schema showing conflict-alert lifecycle (orange), replanning processes (red), and plan evolution (green). The full, up-to-date ontology is available on our explanation platform via its PURL URI.

### Design: Competency Questions

- C1:** Which CollisionEvents (including their time, type, location, and involved agents) were detected during planning?
- C2:** For a given CollisionEvent, which agent(s) received a ConflictAlert?
- C3:** What was an agent's original, conflict-unaware plan, and how does it compare to its final, resolved plan?
- C4:** Why did a specific agent have to wait or reroute in its final plan?
- C5:** For a given ConflictAlert, which ReplanningStrategy did the agent use?
- C6:** What was the cost change associated with a revised AgentSubPlan?
- C7:** Why was a particular agent (from a set of conflicting agents) chosen to be the one to replan? (i.e., what was the planner's selectionRationale?)
- C8:** What is the final JointPlan after all conflicts are resolved, and what is its overall makespan?

### Core Concepts

- ma:Agent, ma:AgentState**  
Aligns with sosa:Platform/Sensor
- ma:CollisionEvent**  
Captures conflict type, time, and location
- ma:ReplanningStrategy**  
Provenance of how conflict was resolved
- ma:AgentSubPlan**  
Linked series of Steps (time, cell)

### Key Properties

- alertsConflict, targetAgent:** who was alerted for which conflict
- triggeredBy, derivesFrom:** PROV-O provenance chain
- resolvesConflict, generatedBy:** causal links from problem to solution

## KEY TAKEAWAYS

### Human-Robot Interaction

- Robots become explainable: waiting/detours get causal labels.
- Trust Calibration: operators can verify planner decisions.
- Diagnose congestion across robot teams.

### Semantic Web Impact

- Interoperable KG via W3C standards (PROV-O, OWL-Time).
- Planner Agnostic: Decouples explanation from solver.
- Extensible: SPARQL-driven templates.

## Acknowledgements & References

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- [1] Muppani, B., Dey, R., Srivastava, B., and Narayanan, V., 2026. OMEGA: An Ontology-Driven Tool for Explaining Multi-Agent Path Finding. AAAI 2026.
- [2] Muppani, B.C., Gupta, N., Pallagani, V., Srivastava, B., Mutharaju, R., Huhns, M.N. and Narayanan, V., 2025. Building a planning ontology to represent and exploit planning knowledge and its applications. Discover Data, 3(1), p.55.

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Video

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