# Reasoning with Ontologies for Non-Player Character Decision-Making in Games LAMAS & SR 2022, Rennes

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26 august 2022







### Strategic Reasoning for Multi-Agent Systems

**Context**: The strategic reasoning of virtual agents is often poor

<sup>&</sup>lt;sup>1</sup>From reference books in Game AI: Millington [2019], Yannakakis and Togelius [2018] and review article: Simonov et al. [2019]

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## Game Al behavioural techniques state of the art1

- Ad hoc algorithms (Hard-coded, poor reusability)
- Finite-State Machines (Hard-coded, poor scaling)
- Behavioural Trees (Hard-coded, poor scaling)
- Utility-based AI (Tuning utilities can be laborious)
- Action Planning (Difficult to use, can be expensive)
- Learning-based AI (Resource intensive, black box)

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Obversation: Logic-based methods are almost absent from SOTA

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### Representation of Game Rules

Declarative logic programming helps intuitive rules representation:

### Transformation of a game rule

From our game prototype: "A **human** in the same room as a **monster** or a **pit**, dies"

Listing 1: The same rule in Prolog

```
dies(Agent):-
    human(Agent),
    human(X, Y),
    (
        monster(X, Y) ; pit(X, Y)
).
```

### Ontologies and Knowledge Inference

- 1 X is an element if X is an object or a being
- 2 X is a being if X is an animal or a monster
- 3 X is a monster if X is a dragon
- $\triangle X$  is an <u>animal</u> if X is a dog

#### Listing 2: Simple Ontology in Prolog

```
1    element(X):- object(X). %1
2    element(X):- being(X). %1
3    being(X):- animal(X). %2
5    being(X):- monster(X). %2
6    monster(X):- dragon(X). %3
8    animal(X):- dog(X). %4
```

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### Stable Model Semantics

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### Truth Table

But how do we represent represent partial or negative knowledge?

#### Well-Founded Semantics

- The Well-Founded Semantics (WFS) from Van Gelder et al.
   [1991] is a 3-valued version of the Stable Model Semantics
- In the WFS a proposition can be true, false or undefined
- Undefined values are used for ignorance or contradictory info

Listing 3: Use of the Well-Founded Semantics in SWI-Prolog

```
is_true(Atom):- call_delays(Atom, true).
is_false(Atom):- \+ Atom.
is_undefined(Atom):-
    call_delays(Atom, Condition),
    Condition \== true.
```

### WFS with Partial Knowledge

### With partial knowledge of the truth values of atoms

- Propositions with true atoms are true
- Propositions with false atoms are false
- If p is true then  $\neg p$  is false
- Conversely, if p is false then  $\neg p$  is true

### Assume undefined propositions false and see what holds

- Any proposition symbols not in the resulting set must be false
- We can then make them false in the well-founded model
- This operation can be repeated until no value changes

### WFS Example

4

6

8

9 10

11

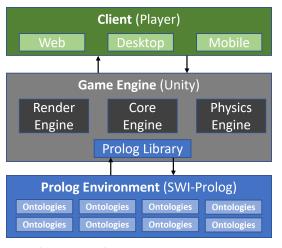
13

Listing 4: Representing False Facts in SWI-Prolog

```
monster(X,Y):-
       pos monster(X,Y),
       tnot(neg monster(X,Y)).
   monster(X,Y):- % Positive Contradiction
       pos monster(X,Y),
       neg_monster(X,Y),
       tnot(monster(X,Y)).
   monster(X,Y):- % Negative Contradiction
       tnot(pos monster(X,Y)),
12
       tnot(neg monster(X,Y)),
       tnot(monster(X,Y)).
```

That denotes Tabled Negation, it checks if there is an atom whose truth cannot be proven without making a call to a non-ground negative literal (i.e. floundering)

#### Global Architecture

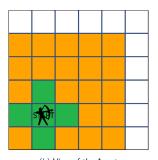


Architecture of the integration of a logic programming environment in a game engine

### Prototype: Wumpus World<sup>2</sup>



(a) View of the world



(b) View of the Agent

### Principle of Wumpus World

An agent explores a cave, finds the gold and leaves without dying.

<sup>&</sup>lt;sup>2</sup>Example extended from Russell and Norvig [2021], and Warren [1999]

### Small Example of a Wumpus World Game Situation

Listing 5: Human agent knowledge of the world

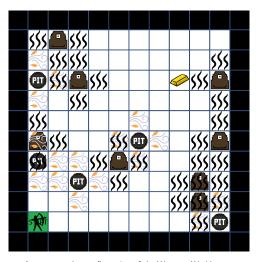
```
visited(1, 1). visited(1, 2). visited(2, 1).
visited(2, 2). stench(2, 2).

safe(X, Y):- tnot(monster(X, Y)).
safe(X, Y):- tnot(pit(X, Y)).
safe(X, Y):- visited(X, Y).
safe(X, Y):- neighbor(X, Y, X2, Y2),
visited(X2, Y2),
no_breeze(X2, Y2),
no_stench(X2, Y2).
```

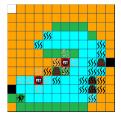
Listing 6: Prolog query for monster given new knowledge

```
?- monster(3, 2).
true.
```

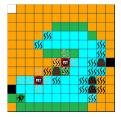
### Adding Complexity



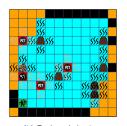
A more complex configuration of the Wumpus World game



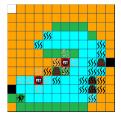
(a) Default behaviour



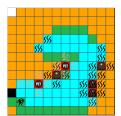
(a) Default behaviour



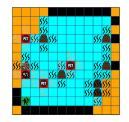
(b) Explorer behaviour



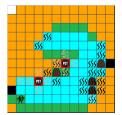
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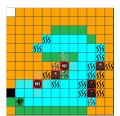
(c) Pacifist behaviour



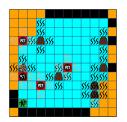
(b) Explorer behaviour



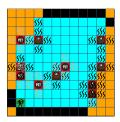
(a) Default behaviour



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(d) Pacifist AND Explorer behaviour

#### Current and Future Works

#### Current work

- Improve the prototype to enable even more complex behaviour
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# Thank You!

#### References I

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