# Revelations: A Decidable Class of POMDPs with Omega-Regular Objectives

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## Outline

#### Partially observable Markov decision processes (POMDPs):

- nondeterminism,
- stochasticity,
- **uncertainty** about the actual state.

**Offline** approach: complete description of the POMDP as an input.

#### Goal

**Strategy synthesis** for  $\omega$ -regular objectives (e.g., reachability, safety, *Büchi*...). Undecidable in general; **decidable subclasses**?

#### Means

**Two subclasses** with probabilistic guarantees about sometimes **knowing the actual state**. **Natural algorithm** that applies to this class.

## Partially observable MDPs



States Q, initial state  $q_0$ , actions Act, observations Obs. Strategies are functions  $(Act \times Obs)^* \rightarrow \mathcal{D}(Act)$ .

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# Objective

- Common objectives:
  - Reachability: a good state is eventually visited,
  - **Büchi**:  $p: Q \rightarrow \{1, 2\}$ ; good states (2) are visited infinitely often,
  - **coBüchi**:  $p: Q \rightarrow \{0, 1\}$ ; bad states (1) are visited finitely often.
- More generally: function  $p: Q \rightarrow \{0, \ldots, d\}$  assigning **priorities** to **states**.
- Parity objective: the maximal priority seen infinitely often is even.
- Question: does there exist an almost-sure strategy?

#### Decidability in POMDPs<sup>1,2</sup>

- Almost-sure reachability, safety, and Büchi are EXPTIME-complete.
- Almost-sure **coBüchi** (and therefore **parity**) are **undecidable**.

<sup>&</sup>lt;sup>1</sup>Baier, Größer, and Bertrand, "Probabilistic  $\omega$ -automata", 2012.

<sup>&</sup>lt;sup>2</sup>Chatterjee, Chmelik, and Tracol, "What is decidable about partially observable Markov decision processes with *w*-regular objectives", 2016.

## Example of a difficult POMDP

Added priorities 1, 2, 3 to the previous POMDP.



**Almost-sure strategy**? Yes! Move to  $q_2/q'_2$  when *increasingly high probability* to be in  $q'_1$ .

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# Belief (support) MDP



Finite: only keep belief supports:



When does the analysis of the belief **support** MDP suffice? In general, neither sound nor complete...

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Looking for decidable classes...

# 1. Weak Revelations

by restricting the information loss!

## Weak revelations

Weak revelations

A POMDP is **weakly revealing** if for all strategies, almost surely, the **current state can be known** infinitely often.





Not weakly revealing

Weakly revealing:  $q_0$  is visited infinitely often

## Weak revelations

Weak revelations

A POMDP is **weakly revealing** if for all strategies, almost surely, the **current state can be known** infinitely often.

When a *revealing history* happens, the finite belief **support** MDP contains **as much information** as the infinite belief MDP.

$$(\{q_0\})$$
  $\approx$   $(q_0 \mapsto 1)$ 

## Weak revelations: results

"Weakly revealing" is a semantic property, but is **decidable**.

Priorities  $\{0, 1, 2\}$  (encompassing Büchi and coBüchi)

There exists an almost-sure strategy... in a weakly revealing POMDP  $\mathcal{P} \iff$  in the belief support MDP of  $\mathcal{P}$ .

#### Decidability

Almost-sure parity  $\{0, 1, 2\}$  for weakly revealing POMDPs is EXPTIME-complete.

**Algorithm**: solve the **belief support MDP**  $\rightarrow$  in EXPTIME.

# Why restrict to parity $\{0, 1, 2\}$ ? Unfortunately...

# Full parity remains undecidable

#### Undecidability

Almost-sure parity  $\{1, 2, 3\}$  is undecidable for weakly revealing POMDPs.

Belief support MDP does not help for this **weakly revealing** POMDP with priorities 1, 2, 3.



Looking for more decidable classes...

# 2. Strong Revelations

by restricting the information loss even more!

## Strong revelations

Strong revelations

A POMDP is strongly revealing if for every transition  $q \xrightarrow{a} q'$ , there is a non-zero probability of revealing q'.



Not strongly revealing:  $q_1 \xrightarrow{a} q'_1$  is a possible transition, but nothing can reveal  $q'_1$  with certainty.

## Strong revelations: results

#### Full parity

#### There exists an almost-sure strategy... in a strongly revealing POMDP $\mathcal{P} \iff$ in the belief support MDP of $\mathcal{P}$ .

#### Theorem

Almost-sure **parity** for **strongly revealing** POMDPs is EXPTIME-complete.

**Algorithm**: solve the **belief support MDP**  $\rightsquigarrow$  in EXPTIME (again!).

# Summary



Decidable subclasses for *parity* POMDPs depending on the **revelation** mechanism.

Decidability frontier when we move to **games**: **games with partial observation** remain **undecidable** for coBüchi under **strong revelations**.

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## Final comments

Paper link:



- A few works with similar approaches.<sup>3,4,5</sup>
- Implementation available at https://github.com/gaperez64/pomdps-reveal.
- **Take-home message**: While POMDPs are undecidable in general, they are not hopeless: there exist **natural and expressive decidable subclasses**.
- Future directions:
  - more general decidable classes,
  - more expressive objectives (e.g., quantitative reachability),
  - other algorithms than solving the belief support MDP?

# Thanks!

<sup>&</sup>lt;sup>3</sup>Berwanger and Mathew, "Infinite games with finite knowledge gaps", 2017.

<sup>&</sup>lt;sup>4</sup>Vlassis, Littman, and Barber, "On the Computational Complexity of Stochastic Controller Optimization in POMDPs", 2012.

<sup>&</sup>lt;sup>5</sup>Bellinger et al., "Active Measure Reinforcement Learning for Observation Cost Minimization", 2021; Krale, Simão, and Jansen, "Act-Then-Measure: Reinforcement Learning for Partially Observable Environments with Active Measuring", 2023.