

# Simplified POMDP Planning with an Alternative Observation Space and Performance Guarantees

Da Kong<sup>1</sup>, Vadim Indelman<sup>2,3</sup>



Paper

#### verview

- The partially observable Markov decision process (POMDP) is a mathematically principled framework for decision-making under uncertainty. While finding the optimal solution is intractable.
- We contribute a novel method to simplify POMDPs by **switching to** an alternative, more compact, observation space and **simplified model** to speedup planning with performance guarantees.

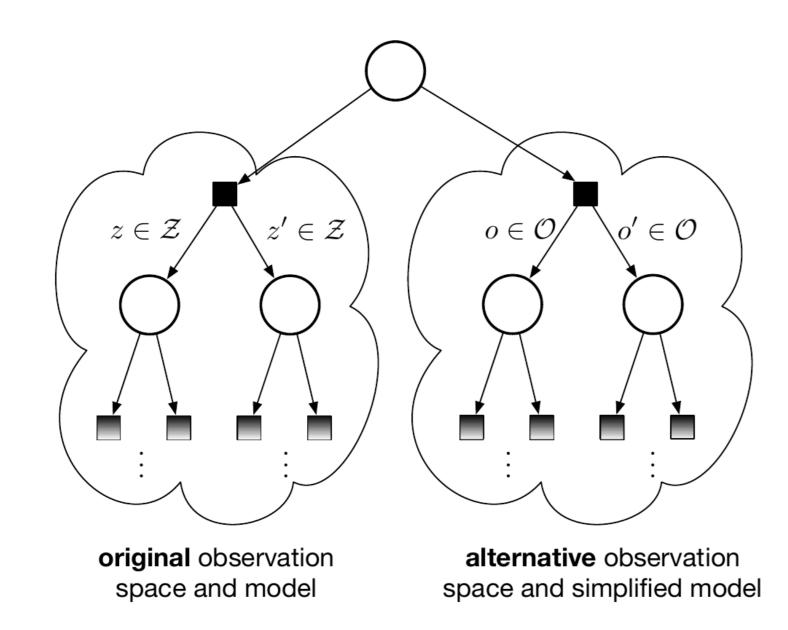


Figure 1: The idea of switching to alternative observation space and model

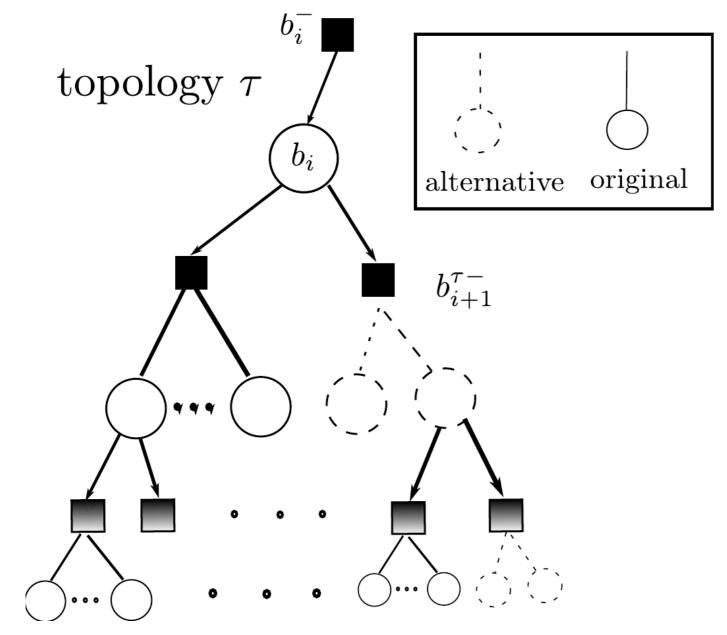
- We evaluate our approach in simulation, considering exact and approximate POMDP solvers and demonstrating a significant speedup while preserving solution quality.
- This paper was presented at ISRR 2024 [1].

#### Contributions

- We propose a novel adaptive simplified belief tree to switch to the alternative observation space and model at selected nodes in belief tree.
- We derive novel bounds that serve as formal performance guarantees.
- We introduce a practical sparse-sampling-based estimator of our simplification.

## Definition: Alternative Observation Topology Tree

Only some of the belief nodes switch to the alternative simplified observation model and space.



The augmented observation model for any  $\bar{z}_t \in \bar{\mathcal{Z}}_t$  becomes:

 $\mathbb{P}_{\bar{Z}}(\bar{z}_t|x_t, h_t^{\tau-}, \tau) \triangleq \beta^{\tau}(h_t^{\tau-}) \mathbb{P}_{Z}(\bar{z}_t|x_t) + (1 - \beta^{\tau}(h_t^{\tau-})) \mathbb{P}_{O}(\bar{z}_t|x_t).$ 

## Contact Information

Email: da-kong@campus.technion.ac.il

### Formal Performance Guarantees

We can bound the difference of Q function:

$$|Q_{\tau}^{\pi^{\tau}}(b_k, a_k) - Q_{\tau_Z}^{\pi^{\tau_Z^*}}(b_k, a_k)| \le B(\tau, \pi^{\tau}, b_k, a_k).$$

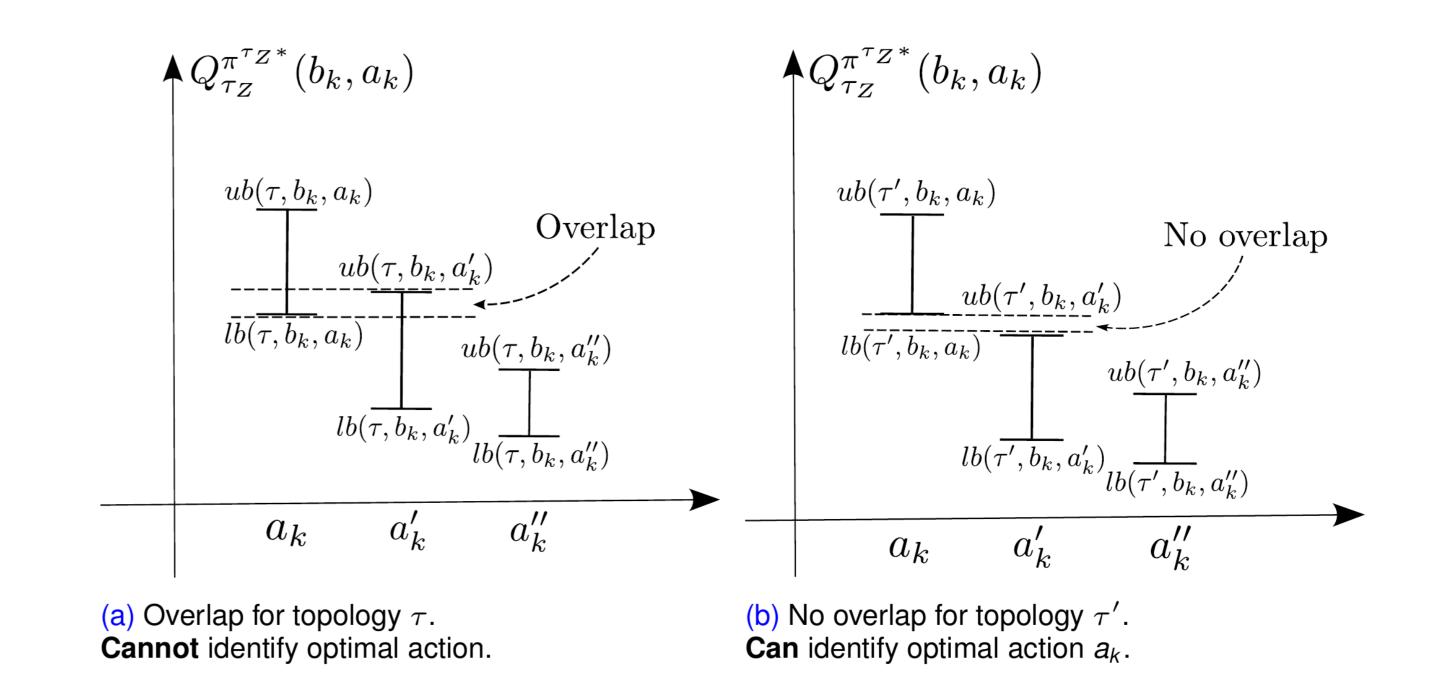
Then we can get the upper and lower bounds of the original Q function, only dependent on simplified topology  $\tau$ :

$$lb(\tau, \pi^{\tau}, b_k, a_k) \leq Q_{\tau_z}^{\pi^{\tau_z}}(b_k, a_k) \leq ub(\tau, \pi^{\tau}, b_k, a_k),$$

where

$$lb(\tau, \pi^{\tau}, b_k, a_k) \triangleq Q_{\tau}^{\pi^{\tau}}(b_k, a_k) - B(\tau, \pi^{\tau}, b_k, a_k),$$
$$ub(\tau, \pi^{\tau}, b_k, a_k) \triangleq Q_{\tau}^{\pi^{\tau}}(b_k, a_k) + B(\tau, \pi^{\tau}, b_k, a_k).$$

Performance guarantees are preserved by checking overlaps across the upper and lower bounds of Q function:



## Specific Case: Full Observability

The alternative observation space  $\mathcal{O}$  and model  $\mathbb{P}_O(o \mid x)$  are defined as,

$$\mathbb{P}_O(o \mid x) \triangleq \delta(o - x)$$
, where  $o \in \mathcal{O} \triangleq \mathcal{X}$ .

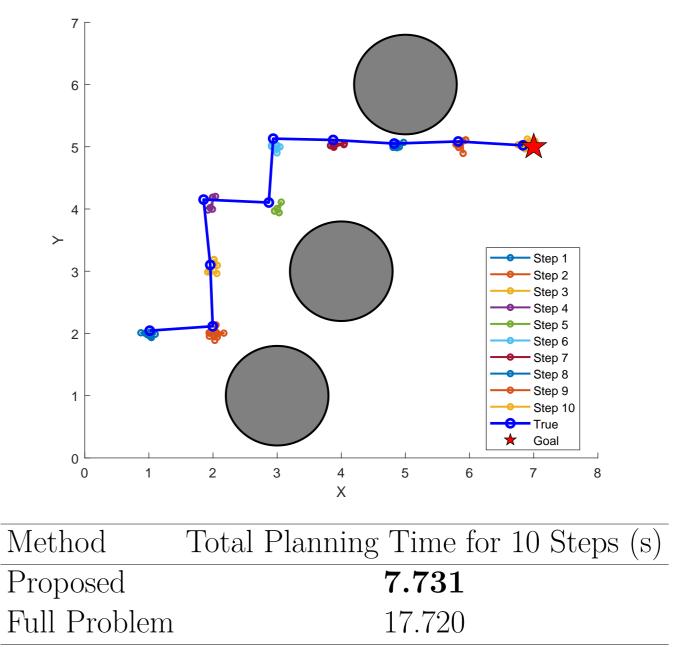
Consider an expected state-dependent reward at any depth i+1 given action  $a_i$  and  $b_i^{\tau-}$ ,

$$\mathbb{E}_{x_i|b_i^{\tau-}}\mathbb{E}_{\bar{z}_i|x_i,h_i^{\tau-}}\mathbb{E}_{x_{i+1}|x_i,a_i}[r(x_{i+1})].$$

Then, the complexity is significantly reduced from  $|\mathcal{Z}||\mathcal{X}|^2$  to  $|\mathcal{X}|^2$ .

## Experiment: Robot Goal-reaching Task

Two more times speedup with the same optimal actions being identified:



#### References

[1] D. Kong and V. Indelman. Simplified pomdp with an alternative observation space and formal performance guarantees. In International Symposium of Robotics Research (ISRR), 2024.

<sup>&</sup>lt;sup>1</sup>Technion Autonomous Systems Program (TASP)
<sup>2</sup> Department of Aerospace Engineering, Technion <sup>3</sup> Department of Data and Decision Sciences, Technion