









Informed POMDP

Leveraging Additional Information in Model-Based RL

Gaspard Lambrechts, Adrien Bolland and Damien Ernst

Informed POMDP

A story of partial observability

Decision process	Execution	Training	Generality
MDP			Too optimistic.
POMDP			Too pessimistic.
Asymmetric POMDP			Too optimistic.
Informed POMDP			Just right?

Classical POMDP

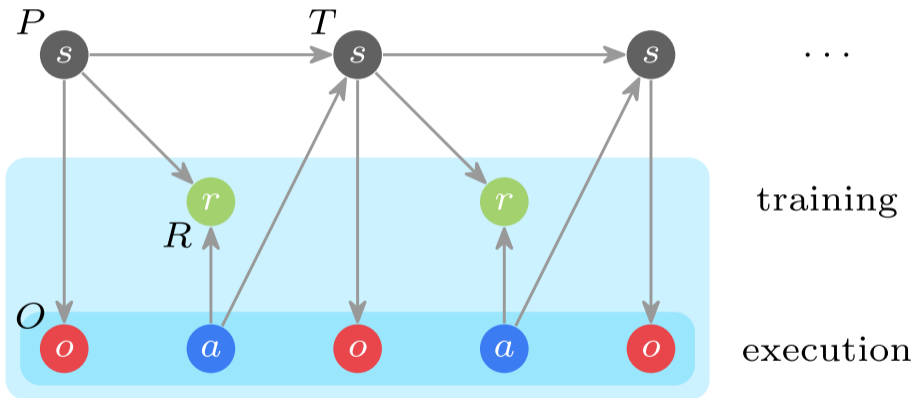


Fig. 1: Bayesian graph of a POMDP.

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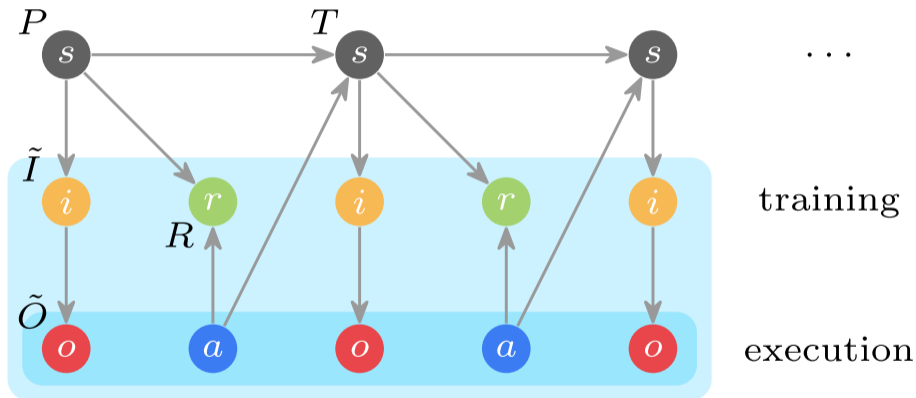


Fig. 2: Bayesian graph of an informed POMDP.

Informed Dreamer

Sufficiency for optimal control



Fig. 3: Statistic $z = f(h)$ of the history h .

- The history h is compressed to a statistic z by a function f (e.g., RNN, Transformer).
- The statistic z should summarize all important information to act optimally.

Definition 1: Sufficiency for optimal control.

A statistic $f : \mathcal{H} \rightarrow \mathcal{Z}$ is **sufficient for optimal control** if, and only if

$$\max_g J(g \circ f) = \max_{\eta} J(\eta).$$

Sufficiency in an informed POMDP

Theorem 1: Sufficiency of recurrent predictive statistics.

In an **informed POMDP**, a statistic $f : \mathcal{H} \rightarrow \mathcal{Z}$ is **sufficient** for optimal control if it is,

- (i) **recurrent:** $f(h') = u(f(h), a, o'), \forall h' = (h, a, o')$,
- (ii) **predictive:** $p(r, i' | h, a) = p(r, i' | f(h), a), \forall (h, a, r, o')$.

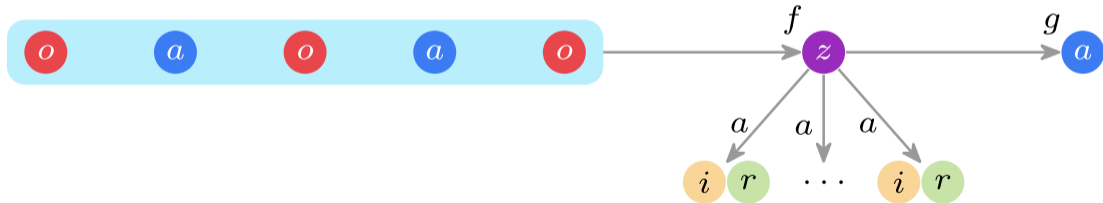


Fig. 4: Statistic $z = f(h)$ of the history h encoding the transition distribution.

A simple view of the Informed Dreamer

The **informed world model** $q(r, i' | f(h), a)$ is learned through likelihood maximization:

$$\max_{p(r, i' | h, a)} \underbrace{\mathbb{E} q(r, i' | f(h), a)}_L$$

- The statistic $z = f(h)$ is **recurrent**.
- At optimum, the statistic is **predictive**.

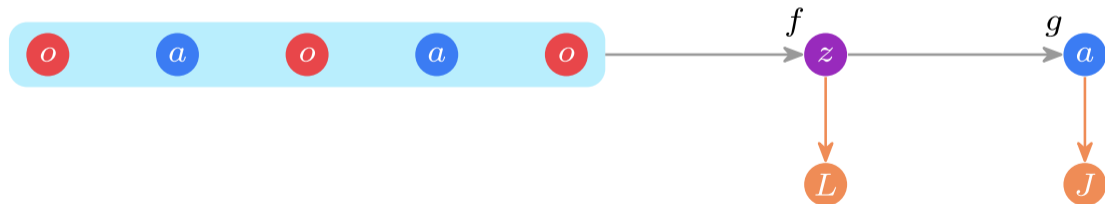


Fig. 5: Sufficiency objective L and reinforcement objective J .

Informed Dreamer

- Prior $\hat{e} \sim q^e(\cdot | z, a)$
- **Information** $\hat{i} \sim q^i(\cdot | z, \hat{e})$
 - Instead of observation $\hat{o} \sim q^o(\cdot | z, \hat{e})$
- Reward $\hat{r} \sim q^r(\cdot | z, \hat{e})$
- **Encoder** $e \sim q^e(\cdot | z, a, o')$
- Update $z' = u(z, a, e)$.

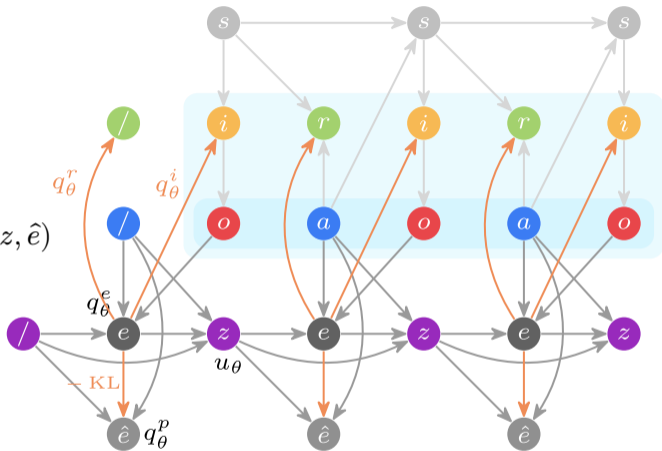


Fig. 6: Informed Dreamer

Results

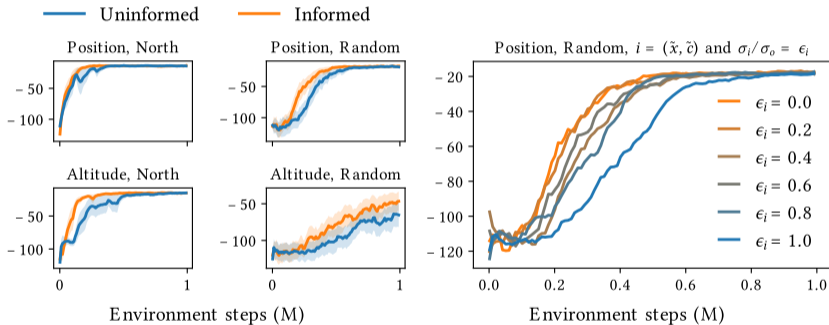


Fig. 7: Varying Mountain Hike

Results (ii)

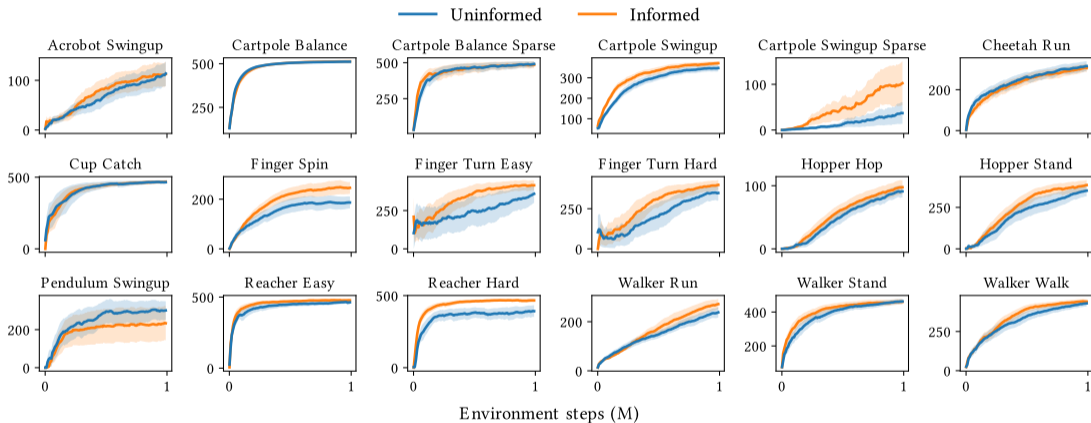


Fig. 8: Velocity DeepMind Control

Results (iii)

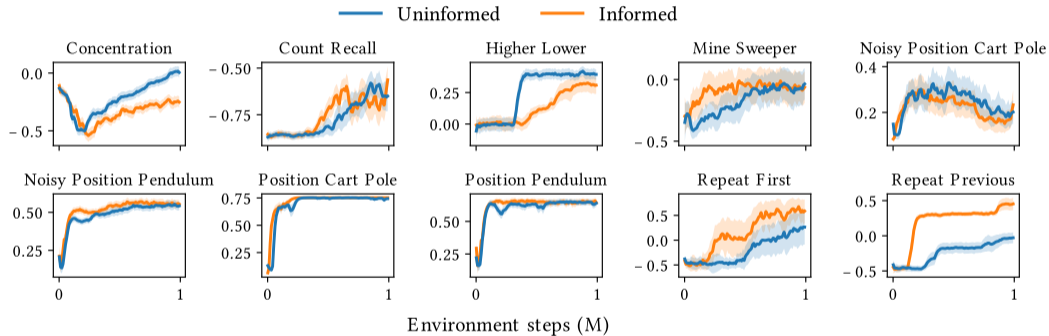


Fig. 9: Pop Gym

Take-home message

Don't make the problem harder than it is.

Consider all available information at training.



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