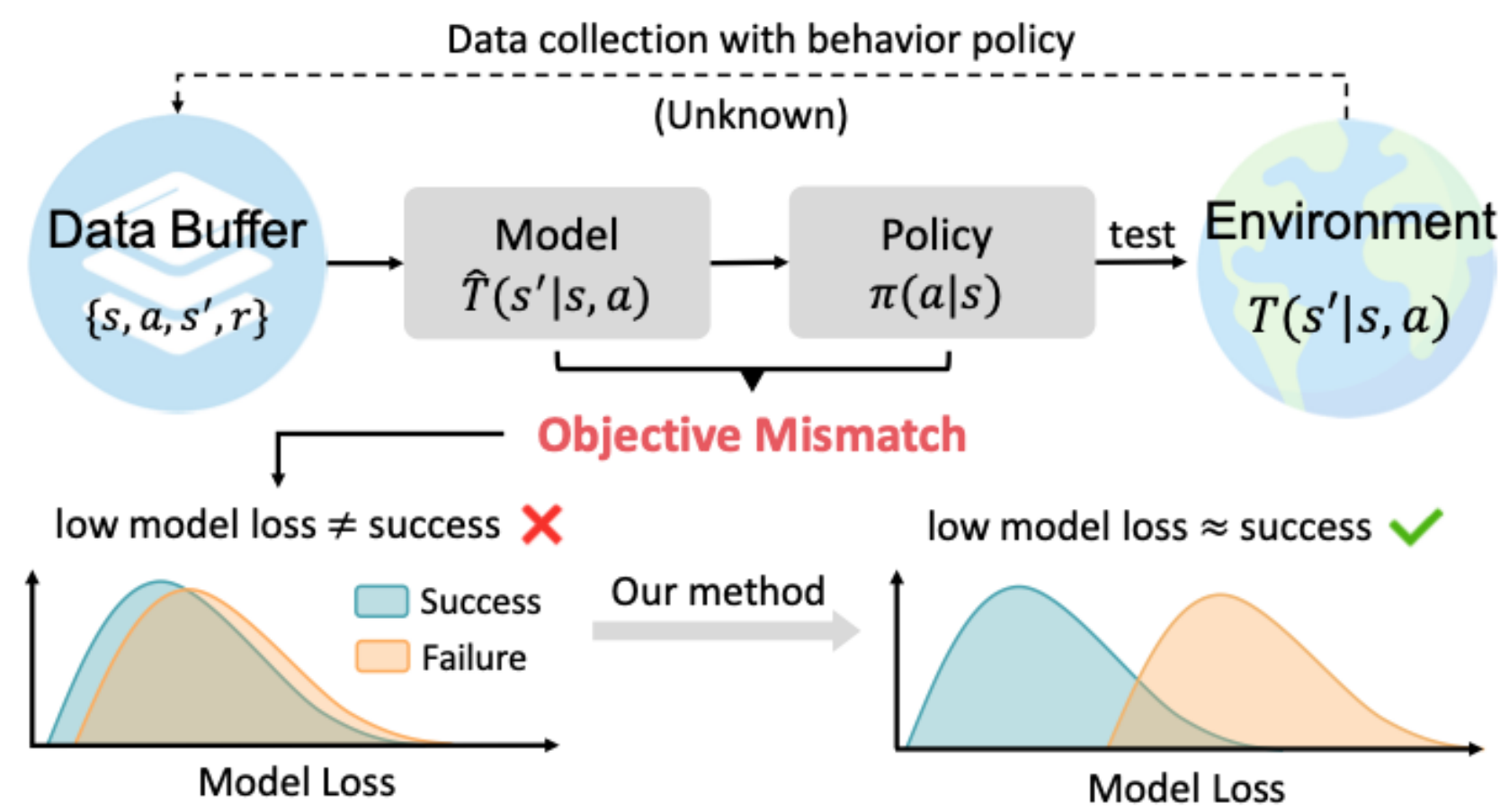




Motivation

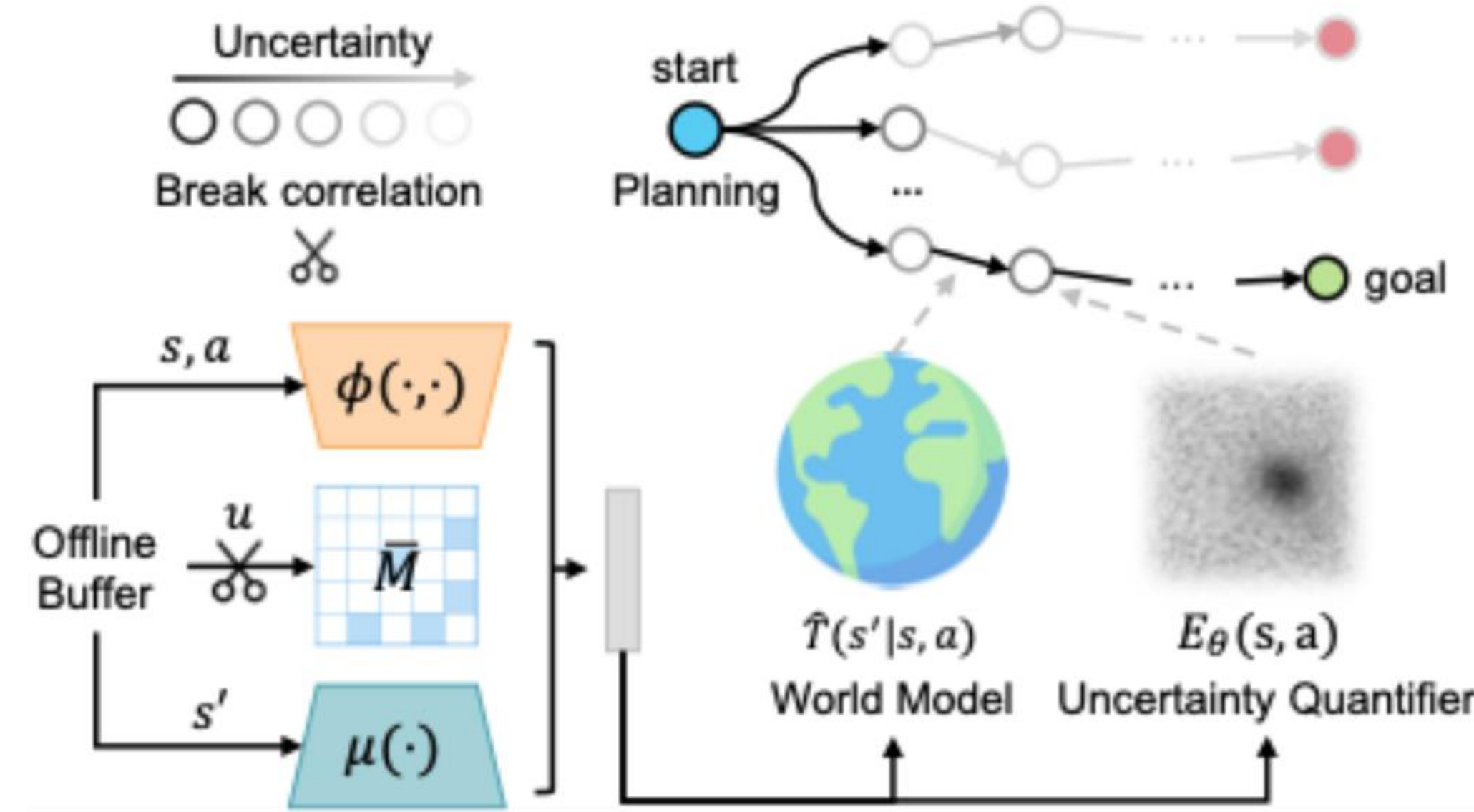
Objective mismatch between dynamics and policy

- Imperfect behavior policies π_β
- Intrinsic Uncertainties in the Environment Transitions T
- Both can be modeled as **Spurious Correlations!**



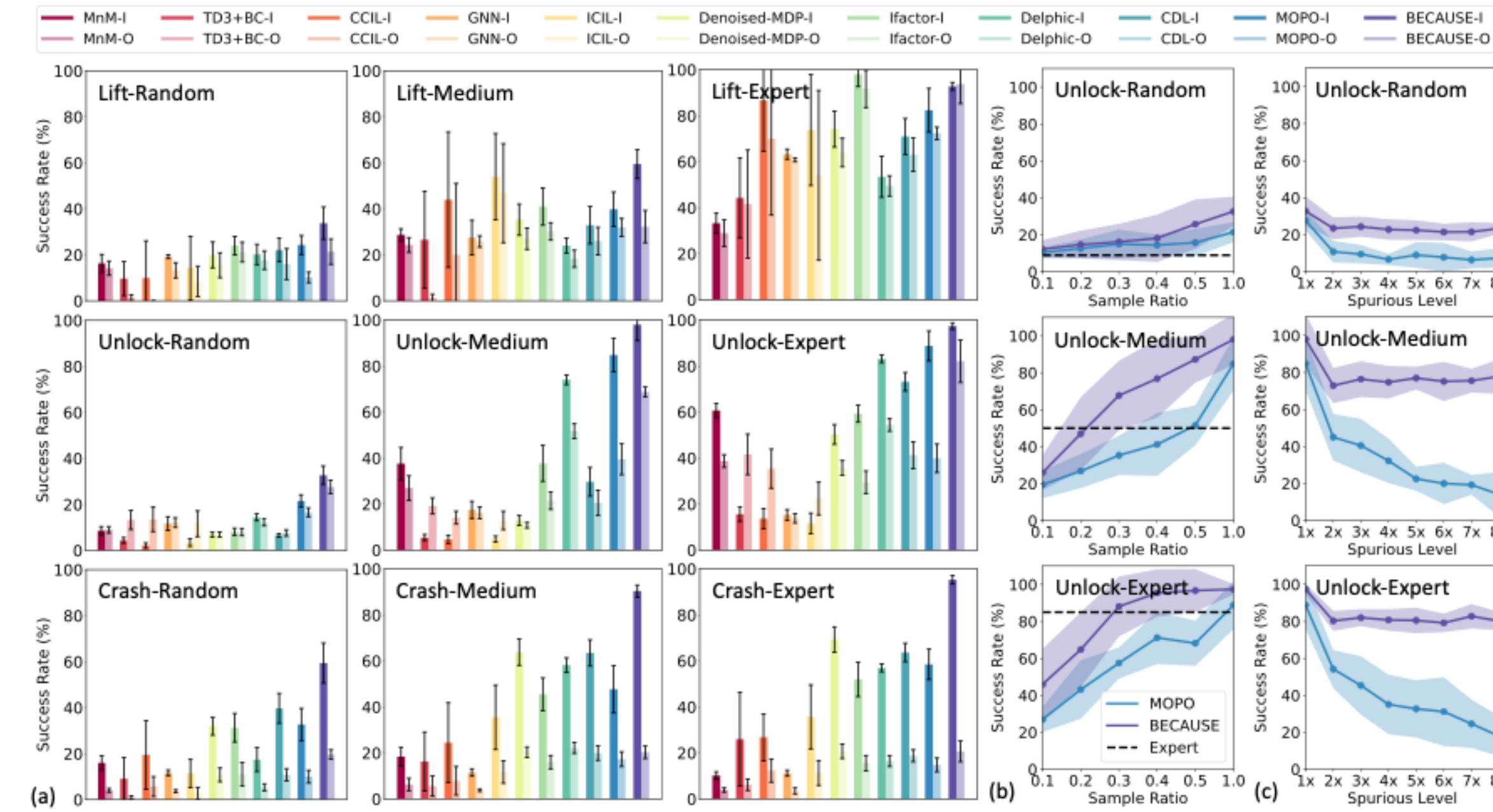
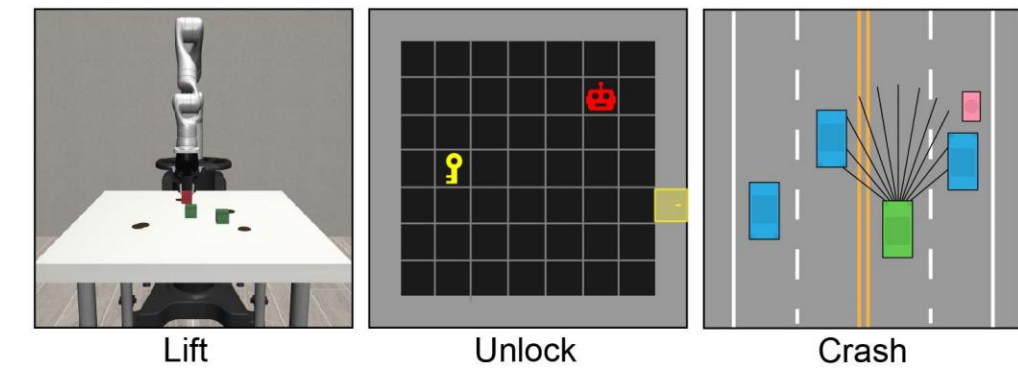
Methodology

BECAUSE: Bilinear causal structure helps mitigate objective mismatch in MBRL!

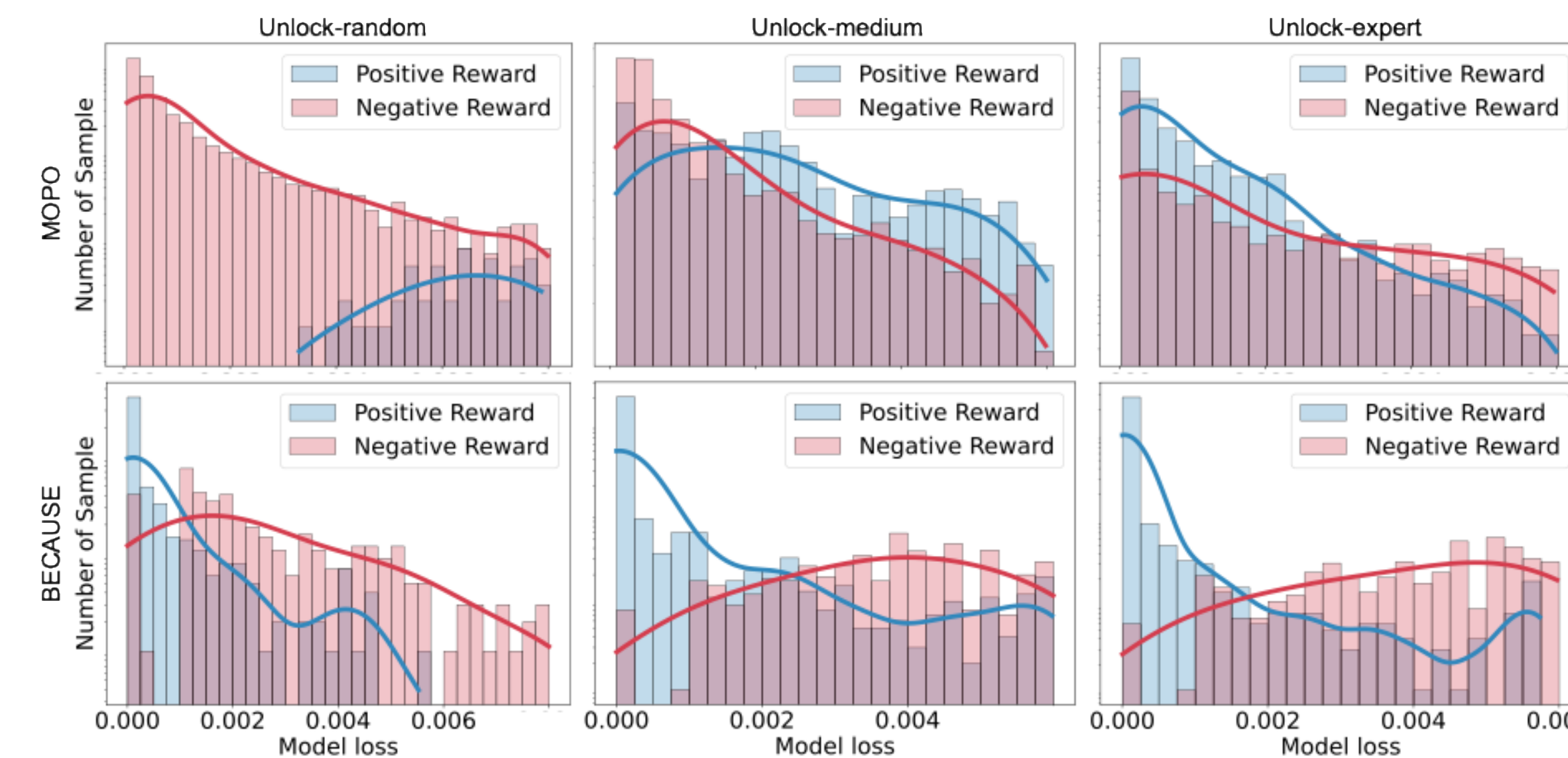


Experiments and Analysis

- Q1. Online generalization
- Q2. Robustness to batch data scales and spurious levels



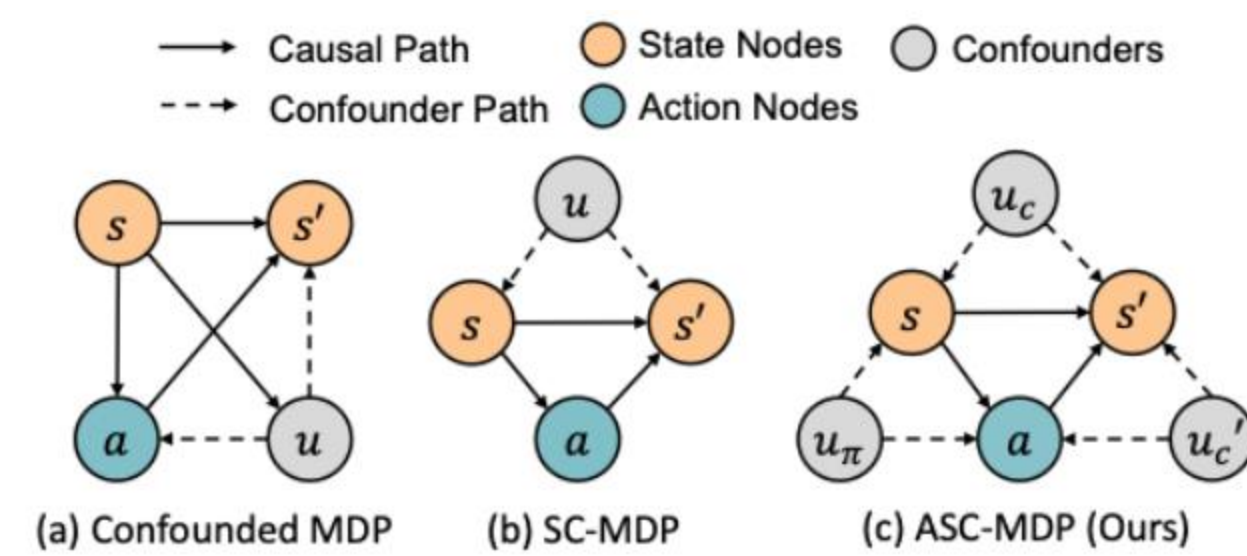
- Q3. Mitigation of Objective Mismatch



Problem Formulation

Confounded MDP: $[u_\pi, u_c]$. Properties: $s' \perp u_\pi \mid (s, a, u_c)$.

- Policy confounder $\pi_\beta(a|s; u_\pi)$.
- Dynamics confounder $T(s'|s, a; u_c)$.



$$T(s'|s, a) = \phi(s, a)^T M(u_c) \mu(s')$$

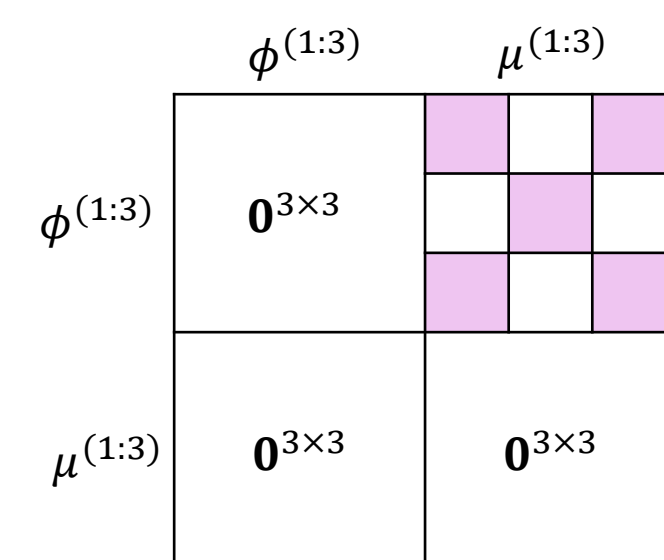
State-action Representation Core Matrix Next State Representation

Bilinear MDP

- Feature ϕ, μ
- Core Matrix M

Bilinear MDP is a Bipartite Structured Causal Model

- Causal nodes with d concepts:
 - $\{\phi(s, a)^{(i)}\}_{i=1}^d, \{\mu(s')^{(j)}\}_{j=1}^d$
- Causal edges between
 - $M_{ij}: \phi(s, a)^{(i)} \rightarrow \mu(s')^{(j)}$



What is BECAUSE?

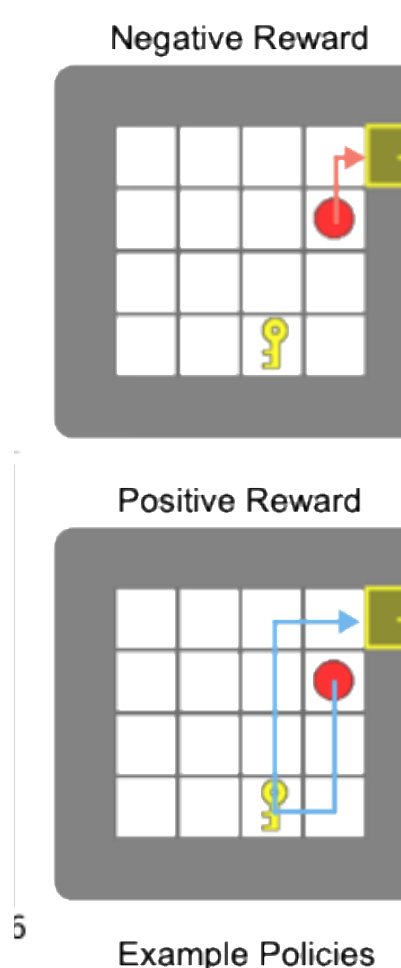
- Concept ϕ, μ
- De-confounded M
- Enforce Sparsity
- Learn a world model that alleviate mismatch

How to learn BECAUSE?

- Transition confounder u_c :
 - Estimate sparse $\hat{M}(u_c)$
- Policy confounder u_π
 - Rewighting on $M: \hat{M}(u_c) \rightarrow \bar{M}$

How to use BECAUSE?

- Planning (w/ MPC):**
 - Learned model $\hat{T}(s'|s, a) = \hat{\phi}(s, a) \bar{M} \hat{\mu}(s')$
- Planning under uncertainty:**
 - Uncertainty Quantifier $\Gamma_\theta(s, a)$
 - EBM: $\Gamma_\theta(s, a) = E_\theta(\phi(s, a), M)$
 - Adjust Return $r(s, a) - \Gamma_\theta(s, a)$

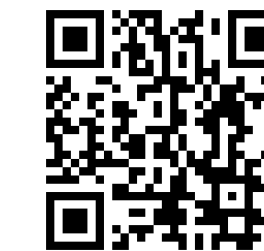


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