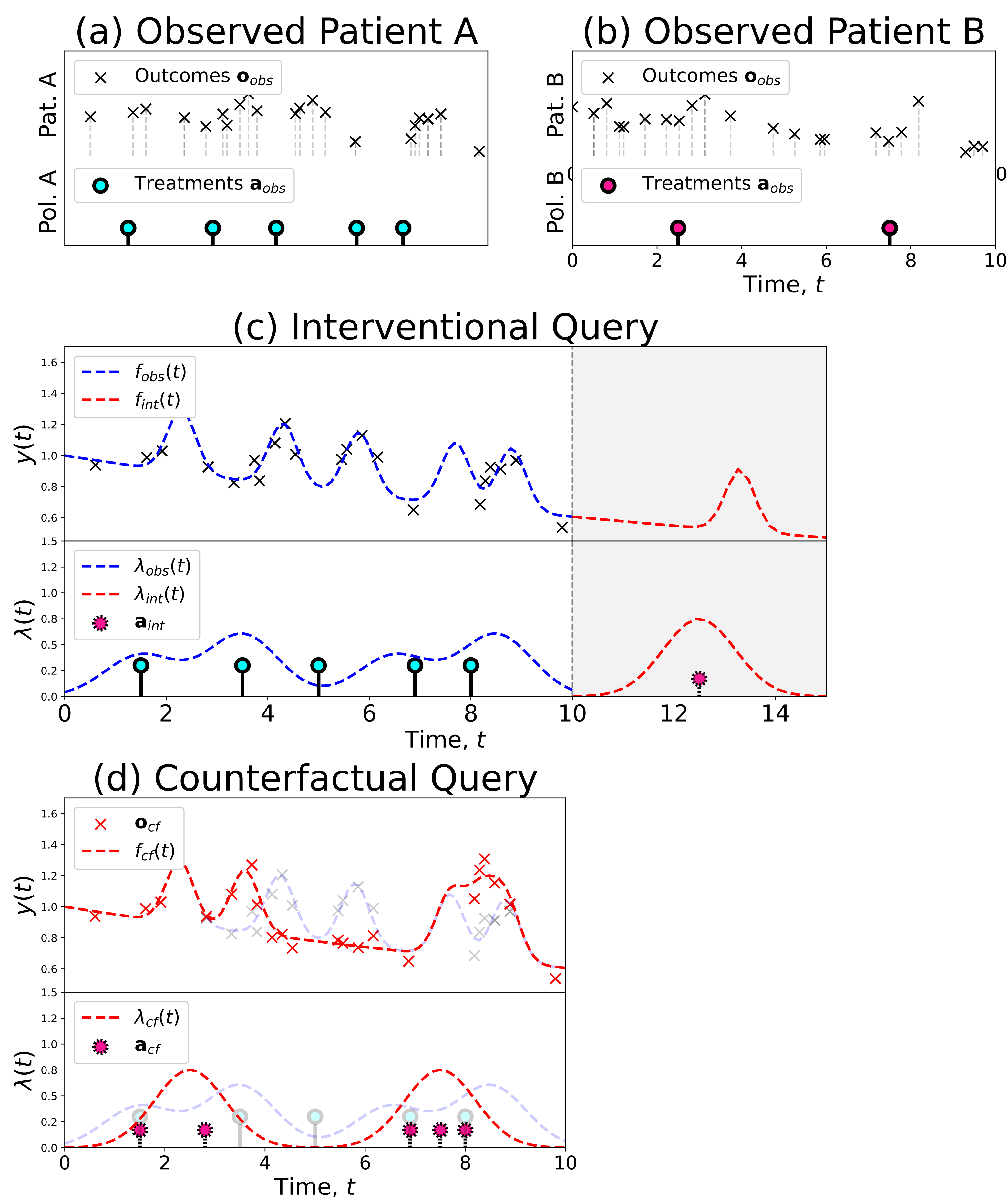


What-if a hospital (country, ...) (1) will or (2) had adopt(ed) the treatment policy of another?

Counterfactual Treatment–Outcome Trajectories Under Policy Interventions

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Details

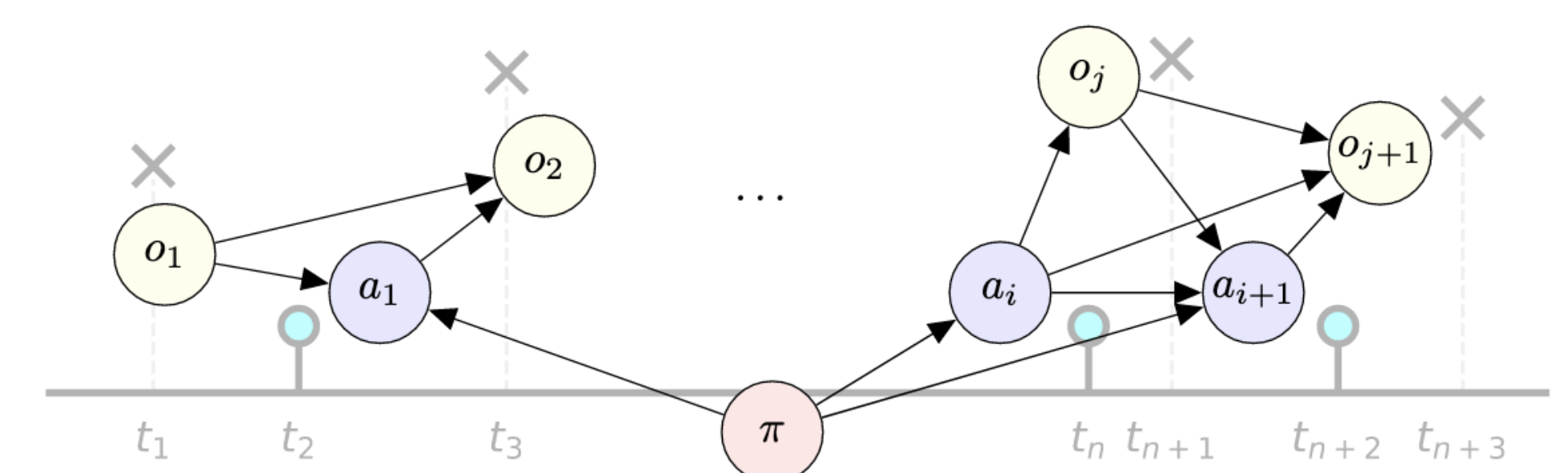
How?

- Joint treatment-outcome model learned from observed treatment and outcome sequences.

$$p(\mathcal{D}) = \exp(-\Lambda)$$

$$\prod_{i=1}^I \underbrace{\lambda_{\pi}^*(t_i) p^*(m_i | t_i)}_{\text{Treatment Intensity}} \prod_{j=1}^J \underbrace{p^*(y_j | t_j)}_{\text{Outcome Model}} \Big|_{t_j \in \mathbf{t}_o}$$

- Consider a policy intervention as a stochastic intervention on the treatment sequence.



- Sample from the interventional and counterfactual distributions of the joint trajectory.

Benefits

- No need for interventional treatments as input.
- Counterfactual treatment sampling.

Challenges

- Hard to optimize the treatment model.
- Not trivial to measure the performance.

Drawbacks

- Scalability.
- Additive treatment response.

What's missing?

- Practical estimation characteristics.
- Detailed comparison with related work.



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A!

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