Exploiting Causality for Efficient Monitoring in POMDPs

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Abstract

POMDPs are a useful model for decision making in systems with uncertain states. One of the core tasks in a POMDP is the monitoring task, in which the belief state (i.e. the probability distribution over system states) is updated based on incomplete and noisy observations. This can be a hard problem in complex real-world systems due to the often very large state space. In this article, we explore the idea of accelerating the monitoring task by automatically exploiting causality in the system. We consider a specific type of causal relation, called *passivity*, which pertains to how system variables cause changes in other variables. Specifically, a system variable is called passive if it changes its value only if it is directly acted upon, or if at least one of the variables that directly affect it (i.e. parent variables) change their values. This property can be readily determined from the conditional probability table of the system variable. We present a novel monitoring method, called *Passivity-based Monitoring* (PM), which maintains a factored belief state representation and exploits passivity to perform selective updates over the factored beliefs. PM produces exact belief states under certain assumptions and approximate belief states otherwise, where the approximation error is bounded by the degree of uncertainty in the process. We show empirically, in synthetic processes with varying sizes and degrees of passivity, that PM is faster than two standard monitoring methods while achieving competitive accuracy. Furthermore, we demonstrate how passivity occurs naturally in a real-world system such as a multi-robot warehouse, and how PM can exploit this to accelerate the monitoring task.