

# Lifting DecPOMDPs for Nanoscale Systems – A Work in Progress

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Link to paper  
(QR code):  
<https://arxiv.org/abs/2110.09152>



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## Decentralised Partially Observable Markov Decision Processes (DecPOMDPs)

- Set of agents working collaboratively towards a common goal with a *joint reward function*
- Environment describable by a *Markov-1 stochastic process*, only *partially observable*
- **Problem: memory requirements *exponential* in the agent set size**
  - Especially a problem for applications with agent set sizes  $\gg 10,000$

## Observation: Symmetries in Agent Set

- *Partitions* in agent set
  - In each partition: the same set of actions, observations, behaviour w.r.t. transition and sensor model, i.e., *indistinguishable* agents
- *Joint actions*: Only important how many agents in a partition take a particular action (count in histograms, e.g.,  $[n, m]$ )
- *Joint observation*: Only important how many agents in a partition observe a particular value (count in histograms)
- Symmetries in transition / sensor model



### Liftable DecPOMDP

- Agent set with partitions such that the descriptions above hold

## Solution Idea: *Lifted* DecPOMDPs

$(\{\mathfrak{S}_k\}_k, S, \{\#_{X_k}[A_k(X_k)]\}_k, T, R, \{\#_{X_k}[O_k(X_k)]\}_k, \Omega)$ ,  
 $k = 1, \dots, K$ , with

- $\{\mathfrak{S}_k\}_k$  a partitioning of an agent set  $I$ ,  $|I| = N$ ,
- $S$  a random variable for the state space,
- $\#_{X_k}[A_k(X_k)]$  a counting random variable;  $\mathbf{A} = \times_k \text{ran}(\#_{X_k}[A_k(X_k)])$  set of joint actions
- $T(S', S, \mathbf{A}) = P(S' | S, \mathbf{A})$  a transition model,
- $R(S)$  a reward function,
- $\#_{X_k}[O_k(X_k)]$  a counting random variable;  $\mathbf{O} = \times_k \text{ran}(\#_{X_k}[O_k(X_k)])$  set of joint observations,
- $\Omega(\mathbf{O}, S) = P(\mathbf{O} | S)$  a sensor model.

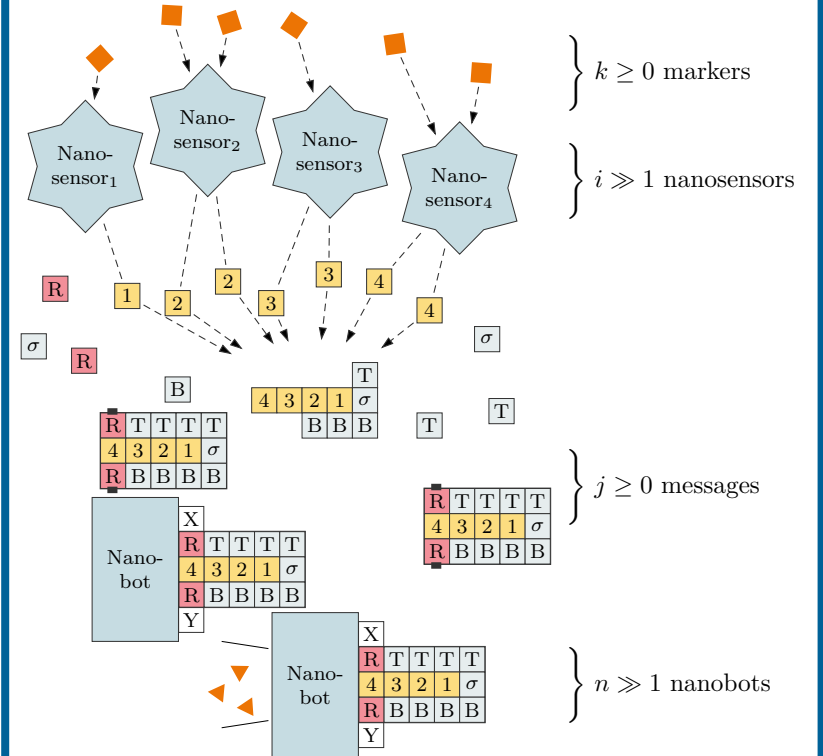
Each partition  $\mathfrak{S}_k$  has a local policy  $\pi_k$ .

The solution is a joint policy  $\boldsymbol{\pi} = (\pi_k)_k$ .



- Memory *polynomial* in the number of agents

## Application: Nanoscale Medical Systems



- Agent set:  $\kappa = 4$  different types of markers/sensors,  $\iota = 1$  different types of messages/nanobots;  $\kappa + \iota = K$ , partition size  $\sim 64,000$  (preliminary experiments)
- Per partition: 2 actions (output vs. no action), 2 observations (sense/receive vs. no obs.)