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USING INHERITANCE DEPENDENCIES TO ACCELERATE ABSTRACTION-BASED SYNTHESIS OF FINITE-STATE CONTROLLERS FOR POMDPS

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Partially Observable Markov Decision Process (POMDP)





– A model for decision making and autonomous planning of agent navigation

- The goal is to find a strategy for given specification
- Observations correspond to the properties of the world, which can be detected by agent's sensor
- The agent lacks the knowledge of its current state and has information solely

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QUOTIENT MDP

Actions in the quotient
 MDP preserve the behaviour of all
 individual realisations (FSCs)

Allows to switch between
 realisations and simulate the behaviour
 of an FSC, which is not originally
 presented in the family

PROPOSED IMPROVEMENT

The main idea of Inheritance
 Dependencies AR is to minimise
 redundant model checkings

Model-checking results of the parent can be preserved for children
 Marking a state as vague if it may
 have lost the optimal choice
 Predecessors of vague states are also vague



about the observation

 Possible specification: the robot must take as few steps as possible to the goal without hitting the wall

Finite-State Controller (FSC)



Figure 2

- A possible representation of POMDP's strategy (policy, scheduler)
- Mealy machine, determining the choice of actions based on a state of internal memory
- Generally, FSC with bigger number of
 nodes can represent more flexible strategies and
 yield better results
- A family of FSCs is a set of candidate FSCs for the given size of the memory

 $\begin{array}{c} 1 \\ 0 \end{array} p_{min}(s_0) \hspace{0.1 cm} p_{max}(s_0) \hspace{0.1 cm} \lambda \end{array}$

0

 $p_{min}(s_0) ~~\lambda ~~ p_{max}(s_0)$

Figure 3

– The robot under this FSC ignores walls and alternates actions \uparrow and ightarrow



- Finding the optimal FSC is unsolvable

- Creating a mask of choices based on received classification of states and optimal choices of the parent
- Extended IDAR searches
 for affected states, which for certain
 lost their optimal choice compared
 to parent's scheduler
- There are less affected states than vague ones, which results in the smaller mask and in more accelerated model checking

Figure 5



Figure 6

Smart EIDAR combines AR and EIDAR for greater efficiency
Switches between these methods based on parameters obtained after running the first few iterations

RESULTS

How much was the analysis of families accelerated? What impact did our methods have on the synthesis?

Model / memory size / num. of iterations / overall AR time	Method	Overall speedup*	MB speedup*	MC speedup*
	IDAR	1.08	1.61	1.73
rocks12 / 1 / 1.000	EIDAR	1.59	3.32	3.73
14.67 s.	SEIDAR	1.5	3.15	3.57
	IDAR	1.16	1.76	I.82
rocks12 / 1 / 10.000	EIDAR	1.7	3.6	4.09
139.64 s.	SEIDAR	1.59	3.41	3.92
	IDAR	1.52	1.63	1.81
rocks12 / 3 / 800	EIDAR	1.56	1.94	2.3
I 75.02s.	SEIDAR	1.54	1.9	2.23
	IDAR	0.78	1.01	I
refuel20 / I / I.000	EIDAR	0.75	0.75	0.63
22.04 s.	SEIDAR	1.32	1.06	0.95
	IDAR	0.82	1.09	1.05
refuel20 / I / threshold 0.000829	EIDAR	6.82	6.99	3.47
18.54 s.	SEIDAR	6.68	6.89	3.47

ABSTRACTION-REFINEMENT



- Splits the family (parent) into subfamilies (children) and analyses them separately
- The number of splittings depends on bounds provided by model checking of quotient MDP. If the bounds are within the desired interval, the subfamily is accepted. If they are fully outside the interval, the subfamily is rejected. Otherwise we continue with splitting



Table I

Proposed method improves the scalability of the inductive synthesis
 *Speedup is provided in comparison with classic AR

