

On Oracle-Efficient PAC RL with Rich Observations



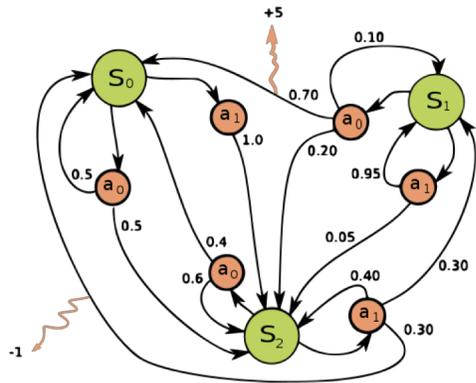
Christoph Dann¹, Nan Jiang², Akshay Krishnamurthy³
Alekh Agarwal³, John Langford³, Robert E. Schapire³

¹Carnegie Mellon University

²University of Illinois at Urbana-Champaign

³Microsoft Research

Exploration in RL

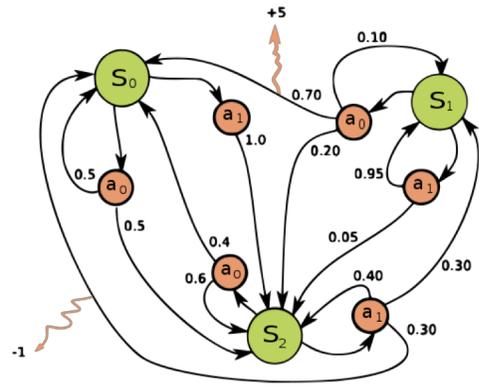


state	value
s_1	v_1
s_2	v_2
s_3	v_3

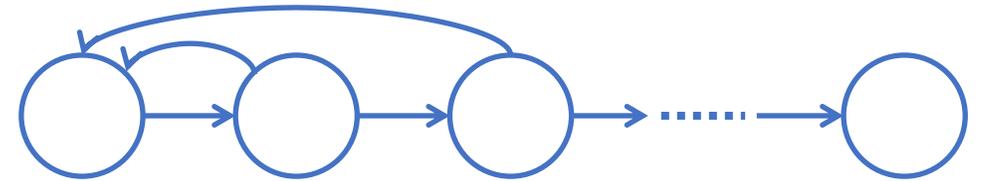
“Tabular”

[Kearns & Singh'98, Brafman & Tennenholtz'02, etc.]

Exploration in RL



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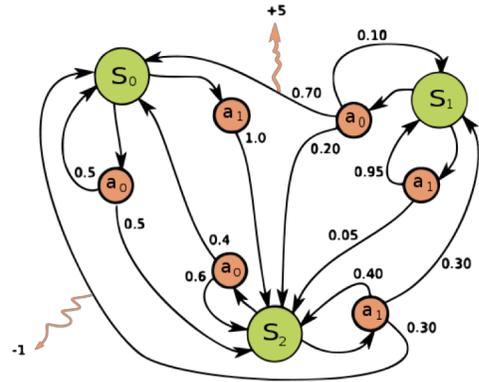


+ ϵ -greedy \Rightarrow exp. sample complexity!

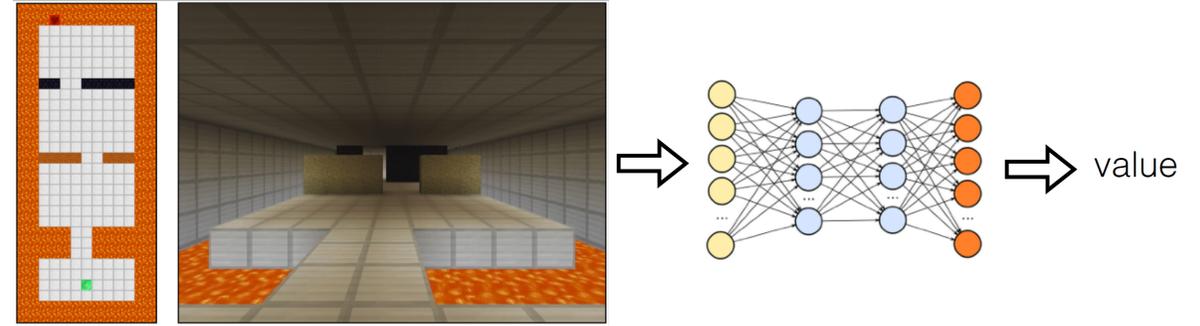
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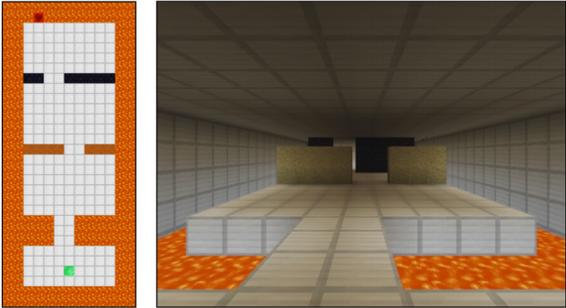
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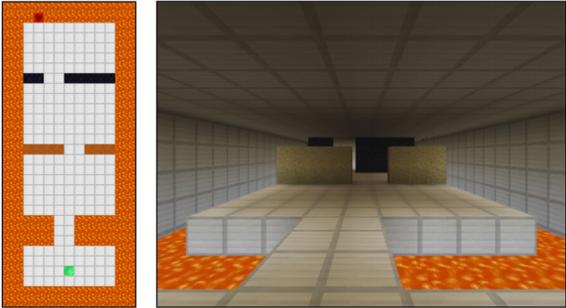
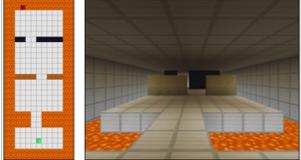
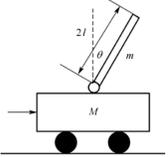
Function approximation

?

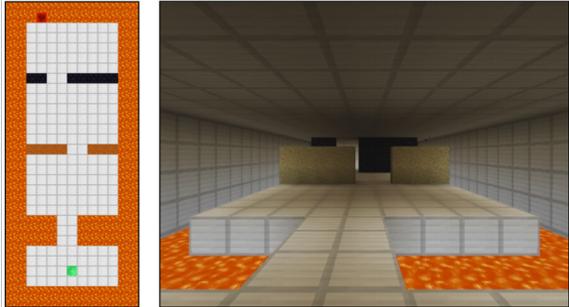
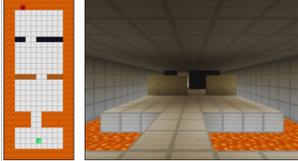
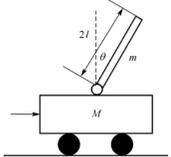
Prior Work & Contribution

Algorithm	LSVEE [KAL'16]		
Setting	<p>Contextual decision processes (CDP) w/ deterministic hidden state dynamics and stochastic rich observations</p> 		
Sample complexity	$poly(\#\text{hidden states})$ ✓		
Computation	Enumerate functions ✗		

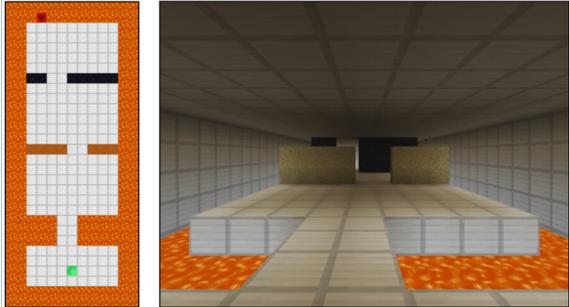
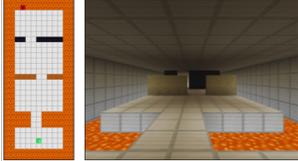
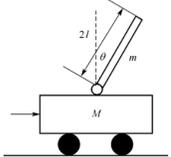
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Prior Work & Contribution

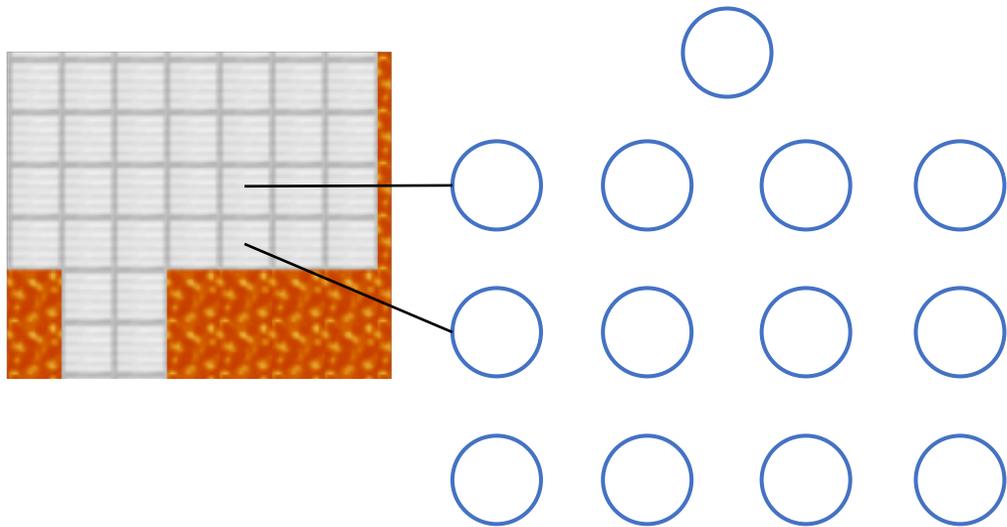
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Computation	Enumerate functions ✗	Linear program + cost-sensitive classification ✓	Enumerate functions ✗

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Computation	Enumerate functions X	Linear program + cost-sensitive classification ✓	Enumerate functions X <u>NP-Hard in tabular case</u>

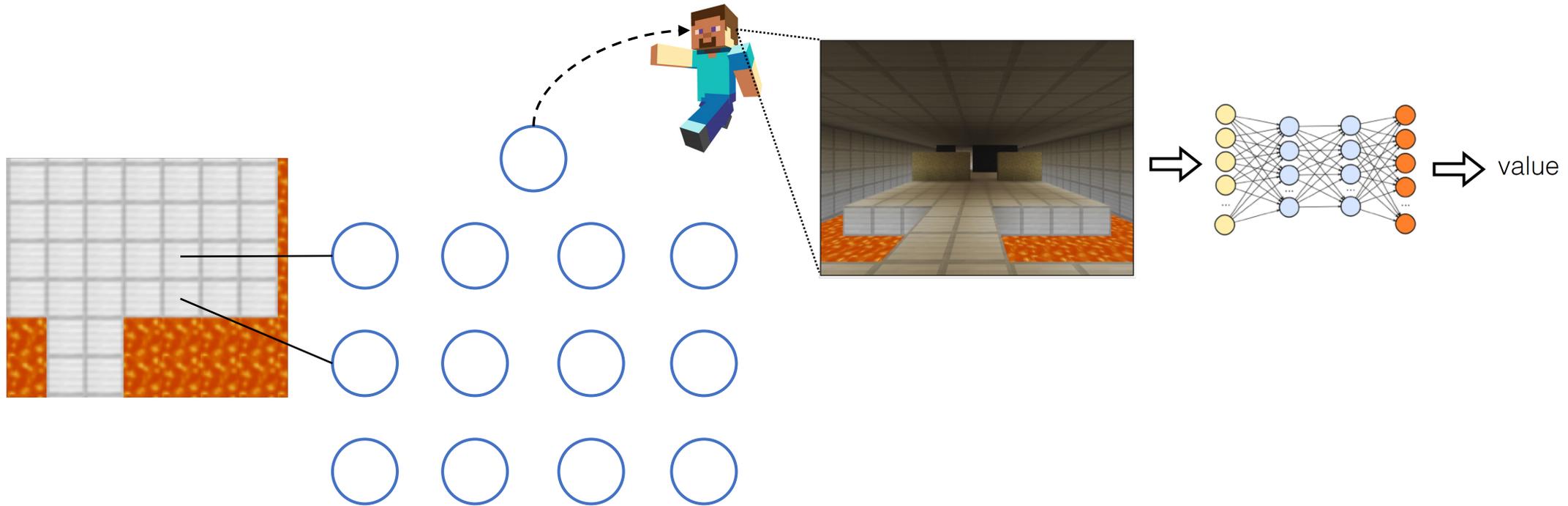
VALOR: efficiently implemented LSVEE

Setting: CDP with **deterministic** hidden state dynamics



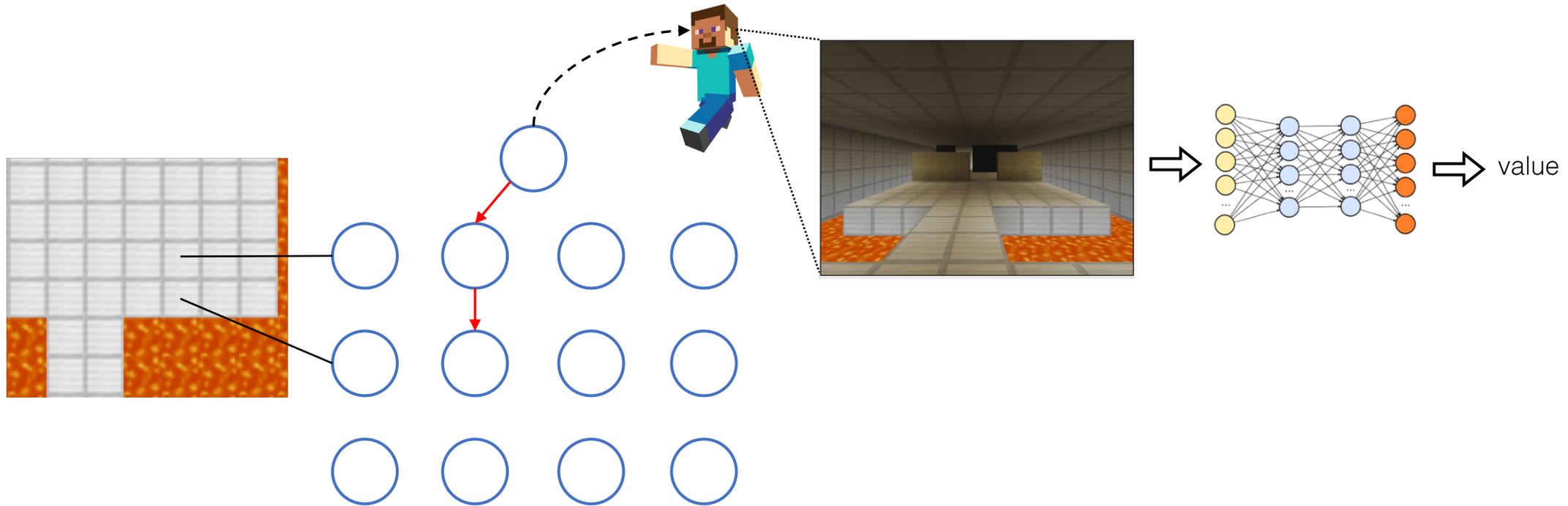
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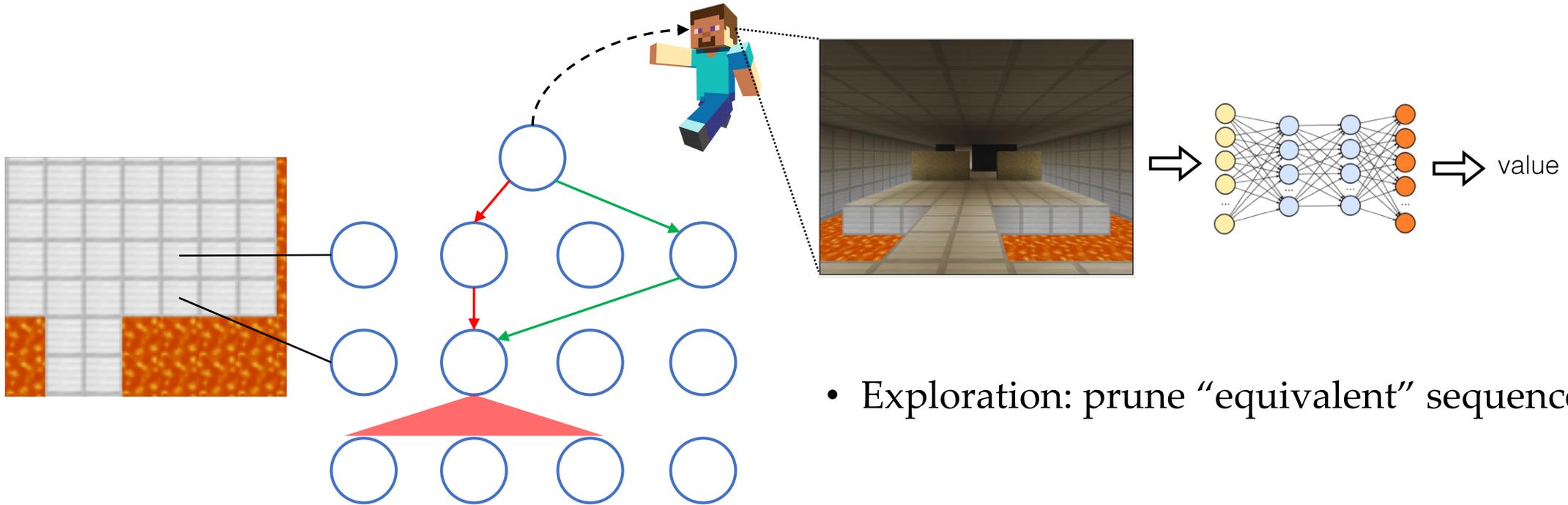
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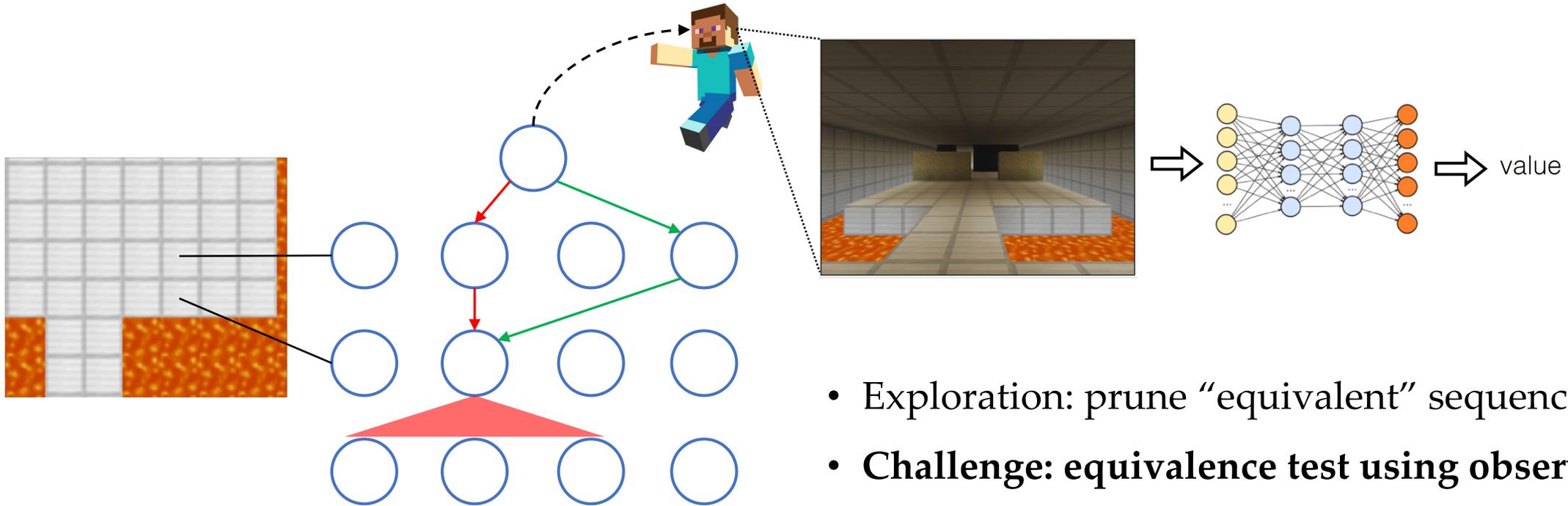
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- Exploration: prune “equivalent” sequences

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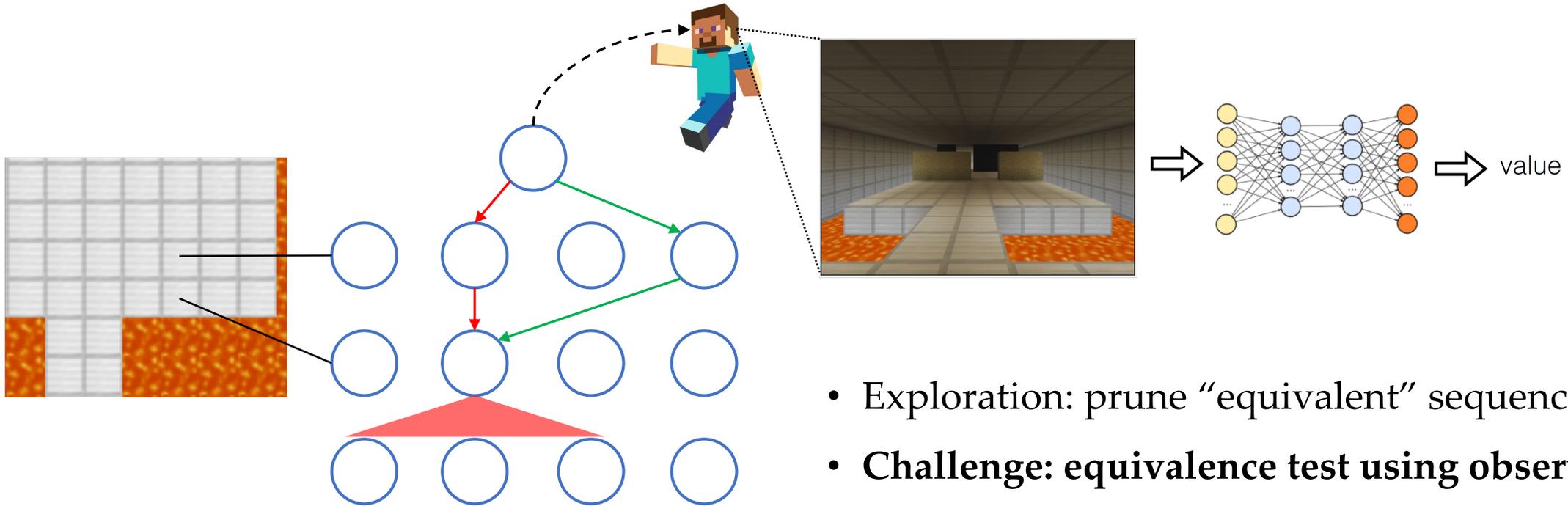
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- **Challenge: equivalence test using observations**
 - Model V^* and π^* separately (instead of Q^*)
 - Can be written as a linear program

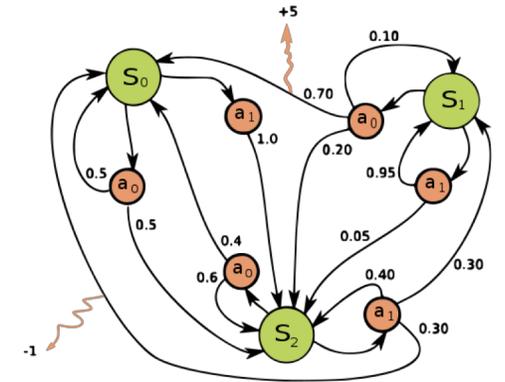
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- Exploration: prune “equivalent” sequences
- **Challenge: equivalence test using observations**
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- What if we remove **determinism**?

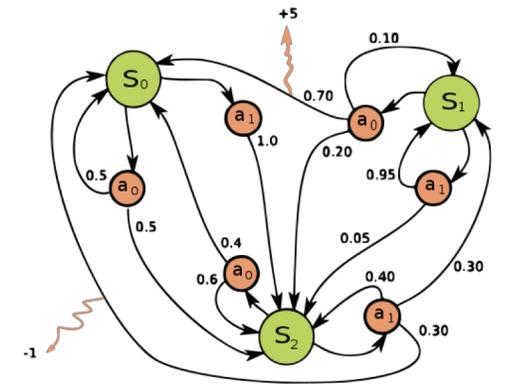
OLIVE is NP-hard in the tabular setting



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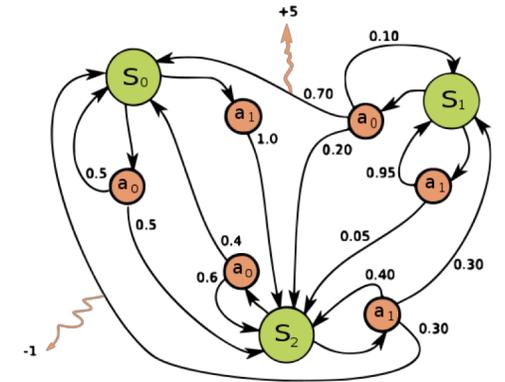
- **But common oracles are efficient in tabular case**
 - e.g., 0-1 loss: majority vote for each s separately
 - OLIVE cannot be implemented efficiently with oracles



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 - e.g., 0-1 loss: majority vote for each s separately
 - OLIVE cannot be implemented efficiently with oracles
- **Not the end of story**
 - Lower bound for algorithm, not problem
 - Efficient RL in OLIVE's setting is still an open problem



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