

# Synthesis of Supervisors for Unknown Plant Models Using Active Learning

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15th IEEE International Conference on Automation Science  
and Engineering, Vancouver, Canada  
August 2019

# Outline

- 1 Problem Statement
- 2 Automata Learning
- 3 Learn Supervisors Using the L\* Algorithm
- 4 Reflections and Future Work



# Topic

- 1 Problem Statement
- 2 Automata Learning
- 3 Learn Supervisors Using the L\* Algorithm
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# Problem Statement

Given a **simulation** of the plant, its corresponding **specifications** and the **set of events**, synthesize a maximally permissive controllable and non-blocking supervisor that can be used to supervise the plant while satisfying the given specification.

## Motivation

- Lack of accurate plant models.
- The plant and specifications constantly updated.
- Easier access and availability of simulations and computing power.

# Topic

1 Problem Statement

**2 Automata Learning**

3 Learn Supervisors Using the L\* Algorithm

4 Reflections and Future Work

# Automata Learning

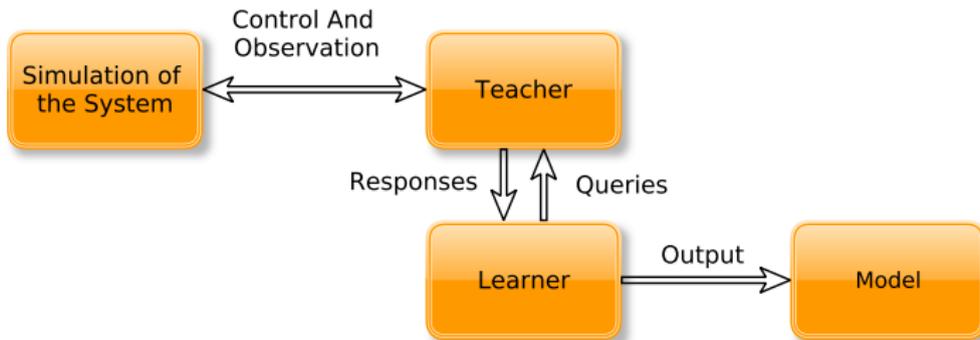
- Passive Learning
- Active Learning

## Active Learning

*Learning regular sets from queries and counterexamples.* Dana Angluin. Information and Computation, 1987

- Famously called L\*
- L\* makes it possible to learn deterministic automata

# Active Learning



L\*

## Learner Queries

- Membership queries  $w \in \mathcal{L}_m?$
- Equivalence queries  $\mathcal{L}(H) = L?$

## Observation Table

Representation of the current knowledge about the target system.

- Contains two sets  $S, E \subseteq \Sigma^*$
- $row : S \cup S.\Sigma \rightarrow E \rightarrow \{1, 0\}$
- $cell(se) = 1 \leftrightarrow se \in \mathcal{L}, | s \in S \text{ and } e \in E$

		E		
		$\tau$	$a$	$aa$
S	$\tau$	0	0	1
	$S.\Sigma$	$a$	0	1
	$b$	0	0	0



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# Membership Query

Plant query

$$\lambda_G(s) = \begin{cases} 2, & s \in L_m(G) \\ 1, & s \in L(G) - L_m(G) \\ 0, & \text{otherwise.} \end{cases} \quad (1)$$

Specification query

$$\lambda_K(s) = \begin{cases} 2, & s \in L_m(K) \\ 1, & s \in L(K) - L_m(K) \\ 0, & \text{otherwise.} \end{cases} \quad (2)$$

# Membership Query

## Controllability query

$$\lambda_C(s) = \begin{cases} 1, & (\forall s' \in \bar{s}, \sigma \in \Sigma_u^*) \lambda_K(s') \neq 0 \text{ and} \\ & \lambda_G(s'.\sigma) \neq 0 \Rightarrow \lambda_K(s'.\sigma) \neq 0 \\ 0, & \text{otherwise.} \end{cases} \quad (3)$$

This is the standard controllability discussed by Ramadge and Wonham.

# Membership Query

## Putting them together

$$T(s) = \begin{cases} \min(\lambda_G(s), \lambda_K(s)), & \lambda_C(s) = 1 \\ 0, & \text{otherwise.} \end{cases} \quad (4)$$

## Observation table

The observation table can now contain three values:  $\{0,1,2\}$ .

# Handling the Controllability Problem

- Convert the controllability problem into a potential blocking problem.
- In doing so, we will check controllability over  $\Sigma_u$  instead of  $\Sigma_u^*$

## $\Sigma_u$ -Saturated Specifications<sup>1</sup>

- Add a transition from every state, for every uncontrollable transition not enabled in that state, to a "dump" state –  $\perp$ .

$$\delta^\perp = \delta \cup \{(q, u, \perp) \mid q \in Q, u \in \Sigma_u, \delta(q, u) \text{ is undefined}\}$$

1. Flordal et. al, "Compositional synthesis of maximally permissive supervisors using supervision equivalence," Discrete Event Dynamic Systems.

# Equivalence Query

## W-Method<sup>2</sup>

Generate test strings according to  $P \cdot U \cdot W$  where

- $U = (\Sigma^0 \cup \Sigma \cup \Sigma^2 \cup \dots \cup \Sigma^{n-m})$ .
- $P$  = Set of all strings that lead to all states
- $W$  = Set of strings that distinguish each pair of states
- $m$  = size of the hypothesis

## Size estimate

An estimate of the maximum size of the supervisor –  $n$

1. T. Chow, "Testing software design modeled by finite-state machines," IEEE Trans. on Software Engineering, 1978.

# Equivalence Query

## Equivalence test

$$R(s) = \begin{cases} 2, & s \in L_m(H) \\ 1, & s \in L(H) - L_m(H) \\ 0, & \text{otherwise.} \end{cases} \quad (5)$$

## Counter example

$$CE = \{s \mid T(s) \neq R(s), \forall s \in P.U.W\}$$

# Obtaining the Supervisor

## Automaton

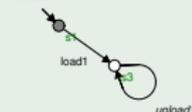
$$\begin{aligned}
 Q &= \{row(s) \mid s \in S \text{ and } cell(s\tau) > 0\} \\
 q_0 &= row(\tau) \text{ if } cell(\tau) > 0 \\
 \delta(row(s), \sigma) &= \begin{cases} row(s\sigma), & cell(s\sigma) > 0 \\ \text{undefined,} & \text{otherwise.} \end{cases} \\
 Q_m &= \{row(s) \mid s \in S \text{ and } cell(s\tau) = 2\}
 \end{aligned}
 \tag{6}$$

## Example (Automaton)

$\tau$	$\tau$
$\tau$	2
$\tau.load1$	1
$\tau.unload1$	0
$\tau.\tau$	2
$\tau.load2$	0
$\tau.unload2$	0
$\tau.load1.\tau$	1
$\tau.load1.load2$	0
$\tau.load1.load1$	0
$\tau.unload1.\tau$	0
$\tau.unload1.load1$	0
$\tau.load1.unload1$	1
$\tau.load1.unload2$	0
$\tau.unload1.load2$	0
$\tau.unload1.unload1$	0
$\tau.unload1.unload2$	0

BLOCKED:

load2  
unload2



# Obtaining the Supervisor

## Supervisor

- If the automaton obtained is non-blocking then this automaton is the maximally permissive controllable and non-blocking supervisor.
- If the automaton is blocking, the maximally permissive controllable and non-blocking supervisor can be obtained by running existing synthesis algorithms.

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# Reflections

- It was possible to learn a maximally permissive supervisor for a given system when plant models are absent.
- The resulting supervisor is monolithic and hence, the current approach cannot be scaled to larger systems.
- The number of states in the learning process is increased due to the use of  $\Sigma_u$ -saturated specifications.
- The W-method requires as input the number of states in the target supervisor, this is not always known in advance.

# Future Work

- Learn controllable supervisors from the start.
- Move towards a modular approach of learning.
- Optimize learning by getting rid of unwanted queries and more efficient data structures to represent the observations.

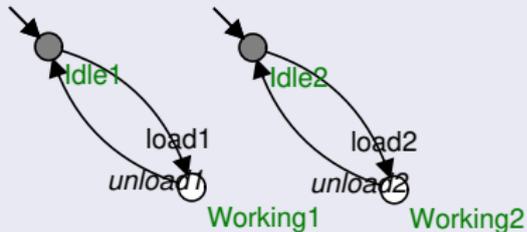
# Thank You!

# The Plant and its Simulation

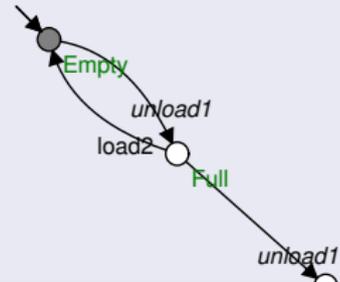
## Two Machines and a Buffer



## Plant Components



## $\Sigma_u$ -saturated Specification



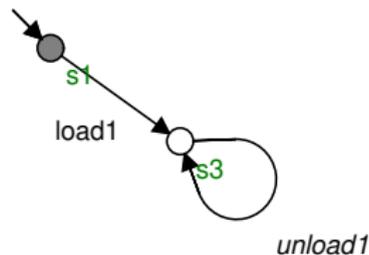
# First Iteration

T	$\tau$
$\tau$	2
$\tau$ .load1	1
$\tau$ .unload1	0
$\tau$ . $\tau$	2
$\tau$ .load2	0
$\tau$ .unload2	0
$\tau$ .load1. $\tau$	1
$\tau$ .load1.load2	0
$\tau$ .load1.load1	0
$\tau$ .unload1. $\tau$	0
$\tau$ .unload1.load1	0
$\tau$ .load1.unload1	1
$\tau$ .load1.unload2	0
$\tau$ .unload1.load2	0
$\tau$ .unload1.unload1	0
$\tau$ .unload1.unload2	0

Counter Example: load1.unload1.unload1

**BLOCKED:**

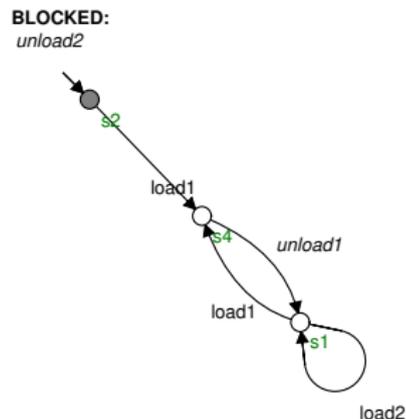
load2  
unload2



# Second Iteration

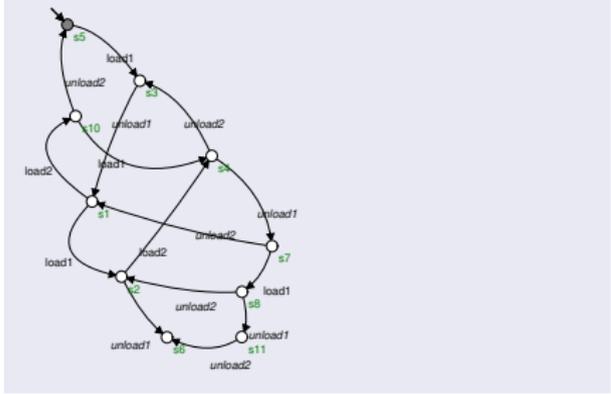
T	r	load1.r
r	2	1
r.load1	1	0
r.unload1	0	0
r.load1.unload1	1	1
r.load1.unload1.unload1	0	0
r.load1.unload1.unload1.r	0	0
r.r	2	1
r.load2	0	0
r.unload2	0	0
r.load1.r	1	0
r.load1.load2	0	0
r.load1.load1	0	0
r.unload1.r	0	0
r.unload1.load1	0	0
r.load1.unload2	0	0
r.unload1.load2	0	0
r.unload1.unload1	0	0
r.unload1.unload2	0	0
r.load1.unload1.r	1	1
r.load1.unload1.load1	1	0
r.load1.unload1.load2	1	1
r.load1.unload1.unload2	0	0
r.load1.unload1.unload1.load2	0	0
r.load1.unload1.unload1.load1	0	0
r.load1.unload1.unload1.unload1	0	0
r.load1.unload1.unload1.r.r	0	0
r.load1.unload1.unload1.unload2	0	0
r.load1.unload1.unload1.r.load1	0	0
r.load1.unload1.unload1.r.load2	0	0
r.load1.unload1.unload1.r.unload2	0	0
r.load1.unload1.unload1.r.unload1	0	0

Counter Example: load1.unload1.load1.load2

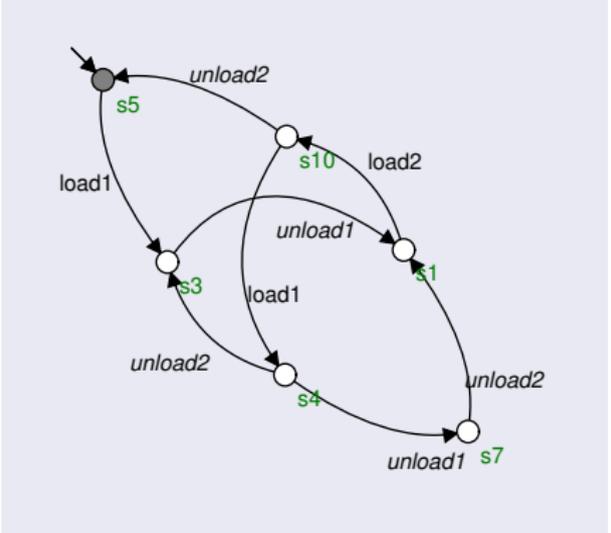


# Resulting Supervisor

## Blocking Supervisor



## Non-blocking Supervisor



## Closed

An observation table is said to be **closed** if for all  $t \in S, a \in \Sigma$  there is an  $s \in S$  such that the  $row(s) = row(t.a)$ .

## Consistent

A table is **consistent** if for  $s_1 \in S$  and  $s_2 \in S$  and  $row(s_1) = row(s_2)$  then for all  $a \in \Sigma, row(s_1.a) = row(s_2.a)$ .