Learning Automata with Side-Effects

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The L* algorithm (Angluin, 1987)

Finite alphabet A

System behaviour captured by a **regular language** $\mathcal{L} \subseteq A^*$

 \mathtt{L}^{\star} learns *minimal* DFA for $\mathcal L$ assuming an *oracle* that answers

► Membership queries

$$w \in \mathcal{L}$$
?

Equivalence queries

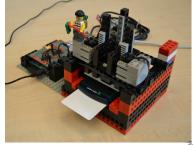
$$\mathcal{L}(H) = \mathcal{L}$$
?

Negative result ⇒ counterexample

Applications of L*

Through learning, verification methods for automata become available for black box systems

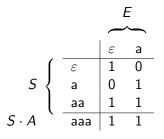
- Network protocols
- Devices such as smartcard readers
- Legacy software



Source: Automated Reverse Engineering using Lego[®]
Chalupar et al., WOOT 2014

L* observation table

 \mathtt{L}^{\star} maintains $S, E \subseteq A^*$ inducing a table



L* observation table

L* maintains $S, E \subseteq A^*$ inducing a table

$$S \left\{ \begin{array}{c|cccc} & \varepsilon & a \\ \hline \varepsilon & 1 & 0 & \mathcal{L} = \{a^n \mid n \neq 1\} \\ \hline a & 0 & 1 \\ \hline aa & 1 & 1 & ---- \\ \hline aaa & 1 & 1 \end{array} \right.$$

Prepend row label to column label and pose membership query

$$(s,e)\mapsto egin{cases} 1 & ext{if } se \in \mathcal{L} \\ 0 & ext{if } se
otin \mathcal{L} \end{cases}$$

L* hypothesis DFA

Hypothesis states are upper rows of the table; transitions append symbols to row labels

	ε	а
ε	1	0
a	0	1
aa	1	1
aaa	1	1

Requires properties closedness and consistency to be well-defined

L* hypothesis DFA

Hypothesis states are upper rows of the table; transitions append symbols to row labels

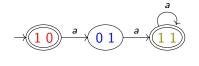
	ε			
ε a aa	1	0	•	
a	0	1	\rightarrow $\begin{pmatrix} 1 & 0 \end{pmatrix}$	01
aa	1	1		
aaa	1	1		

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L* algorithm overview

table updated using membership queries

- 1. Initialise $S = E = \{\varepsilon\}$
- 2. Satisfy closedness and consistency (by augmenting S and E)
- 3. Construct hypothesis
- 4. Pose equivalence query
- 5. On a counterexample, add its prefixes to S and repeat from 2



Motivation

Deterministic automata are often too large for tools to handle

A **side-effect** like non-determinism can provide succinctness

Idea: use side-effects to optimise the algorithm

Monads

We consider side-effects given by a monad

A monad T assigns to each set X a set

TX =combinations of X

It lifts functions to combinations and has

Unit:

 $X \to TX$

Multiplication:

 $TTX \rightarrow TX$

Side-effects as monads

Side-effect	Monad
Partiality	(-)+1
Non-determinism	$\mathcal{P}(-)$
Alternation	$\mathcal{P}_{\uparrow}\mathcal{P}(-)$
Monoid value	$\mathbb{M} \times (-)$
Weighted sum	V(-)

T-algebras

A T-algebra on a set X interprets combinations within X

$$TX \rightarrow X$$

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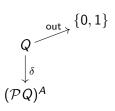
TX itself is the **free** T-algebra on X, with

$$TTX \rightarrow TX$$

being the monad's multiplication

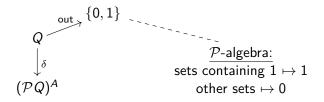
Non-deterministic automaton

A non-deterministic automaton is a set Q with $q_0 \in \mathcal{P}Q$ and functions



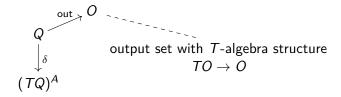
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Automaton with side-effects

An **automaton with side-effects in** T is a set Q with $q_0 \in TQ$ and functions



These do not minimise in general; no target for the learner

Target automaton

No unique (up to iso) minimal NFA

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No unique (up to iso) minimal NFA

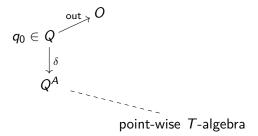
NFA semantics defined in terms of its $\mbox{\bf determinisation}$ automaton with free ${\cal P}\mbox{-algebra}$

state space $\mathcal{P}Q$

Automata with T-algebra structure minimise uniquely

T-automaton

A T-automaton is a T-algebra Q with T-algebra homomorphisms



Contributions

 $\widehat{1}$

General adaptation of L^* for T-automata

↓ method to find generators

General adaptation of L^* for automata with T-side-effects

(2)

Adaptation of optimised counterexample handling

Contributions

 $\overline{1}$

General adaptation of L^* for T-automata

↓ method to find generators

General adaptation of L^* for automata with T-side-effects

(2)

Adaptation of optimised counterexample handling

Structure on the table

T-algebra on output set \implies pointwise T-algebra on table rows

Implicit row for each combination of row labels

	ε	а
ε	1	0
а	0	1
aa	1	1

	ε	а
Ø	0	0
$\{arepsilon, oldsymbol{a}\}$	1	1
$\{a,aa\}$	1	1

Structure on the table

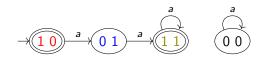
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Hypothesis given by all (upper) rows



Generators

Hypothesis is generated by the explicit rows

Every generating set of rows induces a T-succinct hypothesis

Minimising generators:

While there is a row that is a combination of other rows, remove it

Generators example

	ε	а
ε	1	0
a	0	1
aa	1	1

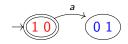




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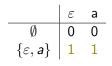
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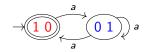
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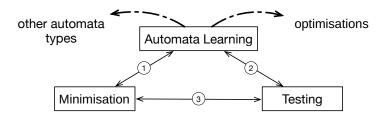
Future work

Monads not preserving finite sets

- ▶ Weighted automata over infinite PIDs
- Subsequential transducers
- Nominal automata

Conformance testing

CALF: Categorical Automata Learning Framework



Project: calf-project.org

