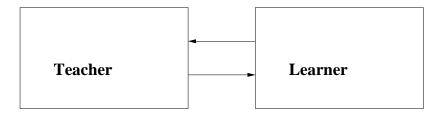
# Learning I/O Automata

#### Fides Aarts and Frits Vaandrager

ICIS, Radboud Universiteit Nijmegen

### CONCUR 2010, Paris





Angluin's *L*<sup>\*</sup> algorithm for active learning deterministic FSMs: Learner poses membership and equivalence queries

## Learning Reactive Systems

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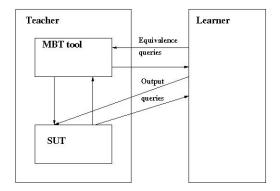
# Learning Reactive Systems

Angluin's algorithm has been extended to Mealy machines by Niese and implemented in the LearnLib tool.

- Membership queries are replaced by output queries: which output is generated in response to a sequence of inputs?
- Equivalence queries are approximated by test sequences generated using algorithms for model based testing.

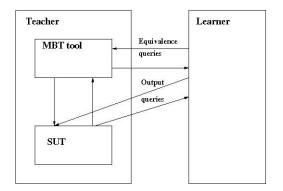
I/O Behavior and Determinism Interface Automata Learning IOAs Conclusions and Future Work

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Active learning may provide models of reactive systems in situations where we have no access to the code (black box) and not even a specification, e.g. to learn reference implementations.

I/O Behavior and Determinism Interface Automata Learning IOAs Conclusions and Future Work





#### • Handle data and large state spaces



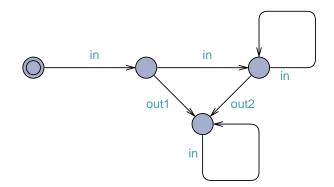
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- Handle data and large state spaces
- Handle nondeterministic systems
- For many applications, requirement that each input corresponds to exactly one output is overly restrictive

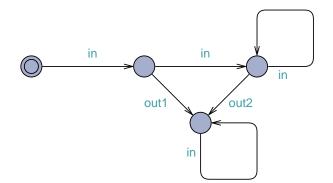
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# I/O Automata



Proposed independently by Lynch & Tuttle and Jonsson (1987).

# The I/O Behavior of I/O Automata?



What output is generated in response to input sequence in in ? Common answer: either out1  $\delta$  or out2  $\delta$  (here  $\delta$  = quiescence)

# Determinism

Engineers want deterministic systems: for every stream of inputs that is offered to the system, the stream of outputs that is generated should be unique. Deterministic systems are predictable.

A deterministic FSM is behavior deterministic: for each string the output (accepted or not accepted) is unique.

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A deterministic Mealy machine is also behavior deterministic. How about I/O automata?

### Determinism

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It is output determined if:

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For usual notion of I/O behavior, input deterministic, output determined IOAs are not behavior deterministic!

# Avoiding Race Conditions

Nondeterminism due to race conditions between IUT and tester.

Common assumption: there is T such that IUT never produces output if more than T time has elapsed since previous event. Allows tester to observe quiescence.

Dual assumption: there is t such that IUT never produces output if less than t time has elapsed since previous event. Allows tester/learner to avoid race conditions (if it is fast enough)!

# I/O Behavior of IOAs

By action  $\Delta$  we give IOA opportunity to produce next output. Sequences  $e \in (I_{\Delta})^*$  are called environment sequences. By action  $\delta$ , IOA indicates that it reached quiescent state:  $q \stackrel{\delta}{\rightarrow} q$  iff q enables no output. Define  $\stackrel{e/u}{\Rightarrow}$  to be least relation s.t.

$$q \stackrel{e/u}{\Rightarrow} q$$

$$q \stackrel{e/u}{\Rightarrow} q' \land q' \stackrel{i}{\rightarrow} q'' \land i \in I \implies q \stackrel{e}{\Rightarrow} q''$$

$$q \stackrel{e/u}{\Rightarrow} q' \land q' \stackrel{o}{\rightarrow} q'' \land o \in O_{\delta} \implies q \stackrel{e}{\Rightarrow} q''$$

For  $\mathcal A$  an IOA,  $obs_{\mathcal A}(e) = \{ u \in (I \cup O_{\delta})^* \mid \exists q \in Q : q^0 \stackrel{e/u}{\Rightarrow} q \}.$ 

### Observation Equivalence and ioco

### Fact (Output deterministic IOAs are behavior deterministic)

If IOA A is input deterministic and output determined then  $obs_{A}(e)$  is a singleton set, for all  $e \in (I_{\Delta})^{*}$ .

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#### Theorem

Let  $\mathcal{A}_1$  and  $\mathcal{A}_2$  be IOAs. Then  $\mathcal{A}_1$  **ioco**  $\mathcal{A}_2$  iff, for all  $e \in (I_\Delta)^*$ ,  $obs_{\mathcal{A}_1}(e) \subseteq obs_{\mathcal{A}_2}(e)$ .

# Interface Automata (De Alfaro & Henzinger)

An interface automaton (IA) is a tuple  $\mathcal{A} = (I, O, Q, q^0, \rightarrow)$ , where *I*, *O* and *Q* are finite, nonempty sets of input actions, output actions, and states, respectively, with *I* and *O* disjoint,

•  $q^0 \in Q$  the initial state, and

•  $\rightarrow \subseteq Q \times (I \cup O) \times Q$  the transition relation.

An I/O automaton (IOA) is an interface automaton in which each input action is enabled in each state.

### **XY-simulations**

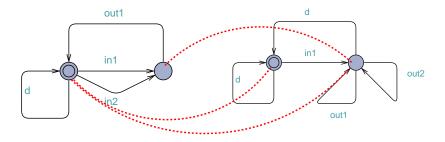
Let  $A_1$  and  $A_2$  be IAs with actions A, and let  $X, Y \subseteq A$ . Relation  $R \subseteq Q_1 \times Q_2$  is an *XY*-simulation from  $A_1$  to  $A_2$  iff, for all  $(q, r) \in R$  and  $a \in A$ 

$$\begin{array}{rcl} q \stackrel{a}{\rightarrow}_1 q' & \wedge \ a \in X & \Rightarrow & \exists r' \in Q_2 : r \stackrel{a}{\rightarrow}_2 r' & \wedge \ (q',r') \in R, \\ r \stackrel{a}{\rightarrow}_2 r' & \wedge \ a \in Y & \Rightarrow & \exists q' \in Q_1 : q \stackrel{a}{\rightarrow}_1 q' & \wedge \ (q',r') \in R. \end{array}$$

 $\mathcal{A}_1 \leq_{XY} \mathcal{A}_2$  if there exists an XY-simulation from  $\mathcal{A}_1$  to  $\mathcal{A}_2$  that contains  $(q_1^0, q_2^0)$ .

## **Alternating Simulations**

#### OI-simulations are known as alternating simulations:



# Alternating Simulations and IOCO

The  $\delta$ -extension of an IA  $\mathcal{A}$  is obtained by adding a loop with output action  $\delta$  to all quiescent states. We write  $\mathcal{A}_1 \leq_{a\delta} \mathcal{A}_2$  iff  $\mathcal{A}_1^{\delta} \leq_{OI} \mathcal{A}_2^{\delta}$ .

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#### Theorem

Suppose  $A_1$  is input enabled and  $A_2$  is deterministic. Then  $A_1$  **ioco**  $A_2$  implies  $A_1 \leq_{a\delta} A_2$ .

# Learning Purpose

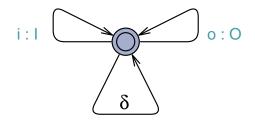
In practice, we often cannot (or do not want to) learn full IOA. We may not be able to offer inputs at the right times, or IOA may be too big, or we are only interested to learn a part.

A learning purpose is a deterministic IA  ${\cal P}$  that specifies which part of an IOA  ${\cal A}$  we want to learn. Our goal is to find IA  ${\cal H}$  such that

$$\mathcal{A}^{\delta} \leq_{OI} \mathcal{H}^{\delta} \leq_{AI} \mathcal{P}$$

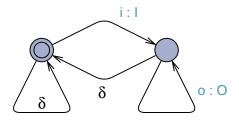
# Learning Purpose (cnt)

A trivial learning purpose:



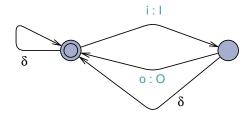
# Learning Purpose (cnt)

After an input one has to wait until the system gets into a quiescent state before offering the next input:



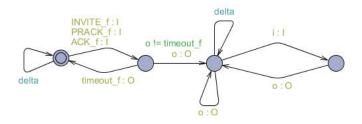
# Learning Purpose (cnt)

Each input is followed by at most one output:



# Learning Purpose (cnt)

Initially only certain inputs are allowed and two consecutive inputs are not allowed:



# Uniqueness!

#### Key Lemma

Suppose  $\mathcal{A}_1$ ,  $\mathcal{A}_2$  and  $\mathcal{A}_3$  are IAs,  $\mathcal{A}_1$  is active and input deterministic,  $\mathcal{A}_2$  is output determined,  $\mathcal{A}_3$  is output deterministic, and  $\mathcal{A}_1 \leq_{OI} \mathcal{A}_3 \leq_{AI} \mathcal{A}_2$ . Then  $\mathcal{A}_3$  is uniquely determined up to bisimulation equivalence.

# Learning IOAs

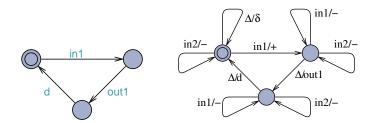
Learner	Teacher
knows deterministic learning	knows deterministic, output
purpose ${\cal P}$	determined IOA ${\cal A}$
maintains state of ${\cal P}$	maintains state of ${\cal A}$
may present $i \in I \cup \{\Delta\}$ to	replies with $o \in O \cup \{\delta\}$
teacher	
may do reset and jump to ini-	upon reset jumps to initial
tial state of ${\cal P}$	state of ${\cal A}$
may pose preorder query $\mathcal{H}$ ,	replies with yes if ${\mathcal A}$ ioco ${\mathcal H}$ or
for $\mathcal{H}^{\delta} \leq_{\mathcal{A}I} \mathcal{P}$	else with <i>no</i> +counterexample

## Set Up for Learning IOAs

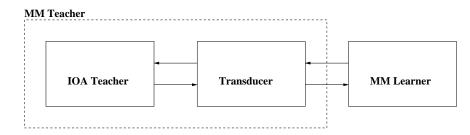


Teacher is teaching IOA  $\mathcal{A} = (I, O, Q, q^0, \rightarrow)$ Transducer handles learning purpose  $\mathcal{P}$  and translates between IOAs and MMs Learner learns MM with inputs  $I_{\Delta}$  and outputs  $O \cup \{\delta, +, -\}$ .

#### From IAs to MMs

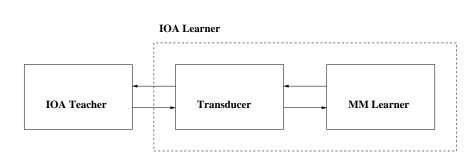


#### Lemma



IOA teacher and transducer together act as teacher for MM  $T(AS(\mathcal{A}^{\delta}, \mathcal{P})).$ 





MM learner and transducer together act as learner of IOA  $\rho(\mathsf{AS}(\mathcal{A}^{\delta},\mathcal{P}))$ 



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- **TCP** 
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F. Aarts, B. Jonsson & J. Uijen, Generating Models of Infinite-State Communication Protocols using

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#### Biometric Passport

F. Aarts, Inference and Abstraction of the Biometric Passport, YR-CONCUR 2010



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- Learning of reactive systems generalized to setting in which inputs and outputs don't have to alternate
- Interesting links between three widely used models: interface automata, ioco theory and Mealy machines.
- Using concept of transducer, learning of IOAs has been reduced to problem of learning MMs.
- Initial experiments indicate work may become quite useful in practice.



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## Future Work

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- Extend to nondeterministic automata
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- Extend to other learning algoritms
- Integrate verification, testing and learning

Many nice challenges for CONCUR community!