

Eternal Software Intensive Networked Systems



# A practical introduction to active automata learning

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SFM2011

















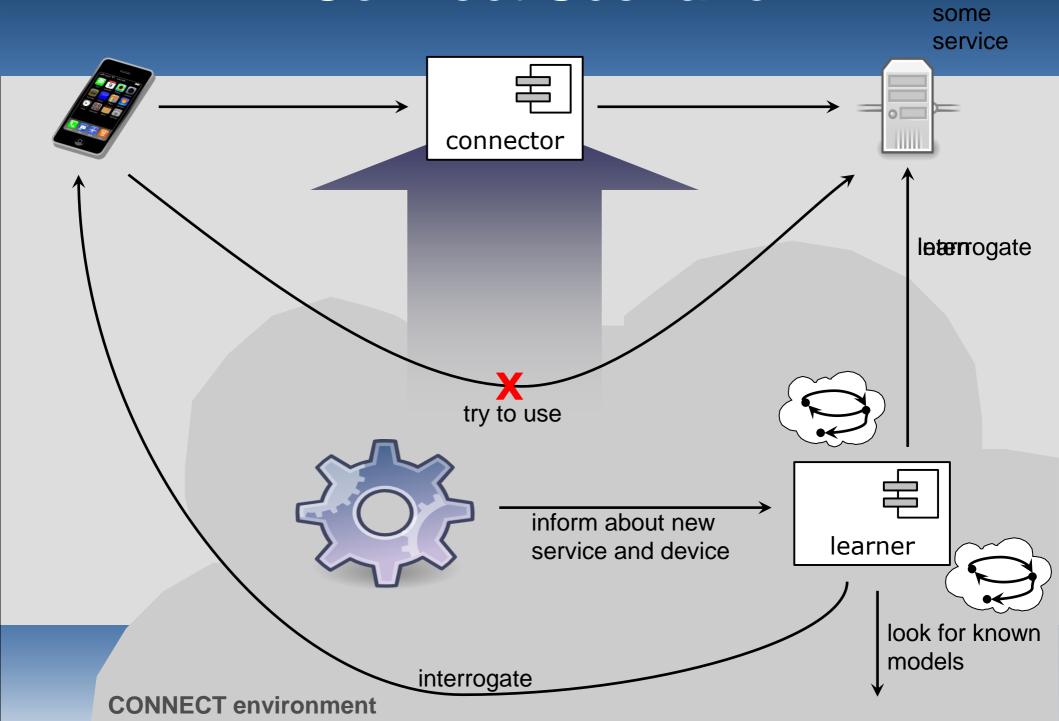
## Overview

- Motivation
- Introduction to active automata learning
- Practical aspects in active automata learning
- Conclusions



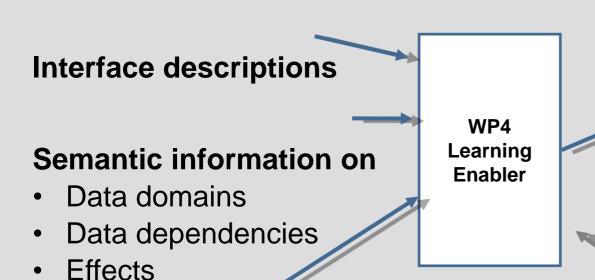


## Connect Scenario



## Learning in CONNECT

Develop techniques for learning ... models of ... behavior of networked peers and middleware through exploratory interaction...



#### Rich & abstract models:

- Data parameters & state variables
- Pre- and post-conditions
- Non-functional properties

**Metrics of interest** 

Counterexamples through monitoring





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- Conclusions





## Mealy machines

- Mealy machine  $M=(S,\Sigma, \Gamma,\sigma,\gamma)$ 
  - S finite set of states
  - $\Sigma$  finite input-alphabet
  - Γ finite output-alphabet
  - $\sigma$ : (S x  $\Sigma$ )  $\rightarrow$  S transition-function
  - $\gamma$ : (S x  $\Sigma$ )  $\rightarrow \Gamma$  output-function
- Words  $\Sigma^*$  for  $(s \in S, a \in \Sigma, w \in \Sigma^*)$ 
  - $\sigma:(S \times \Sigma^*) \to S$ ,  $\sigma(s,\epsilon)=s$ ,  $\sigma(s,aw) = \sigma(\sigma(s,a),w)$
  - $\gamma$ : (S x  $\Sigma^*$ )  $\rightarrow \Gamma^*$ ,  $\gamma$ (s, $\varepsilon$ )= $\varepsilon$ ,  $\gamma$ (s,aw) =  $\gamma$ (s,a). $\gamma$ ( $\sigma$ (s,a),w)



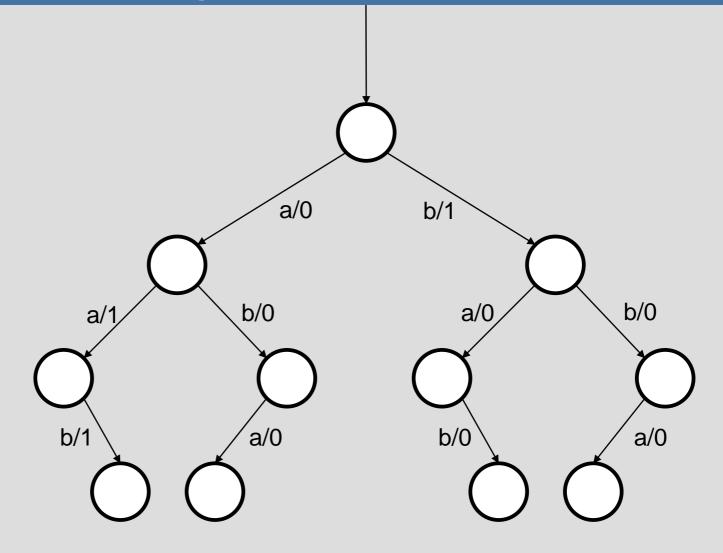


- tape-record communication
- Create observation tree
- Construct automaton without contradiction



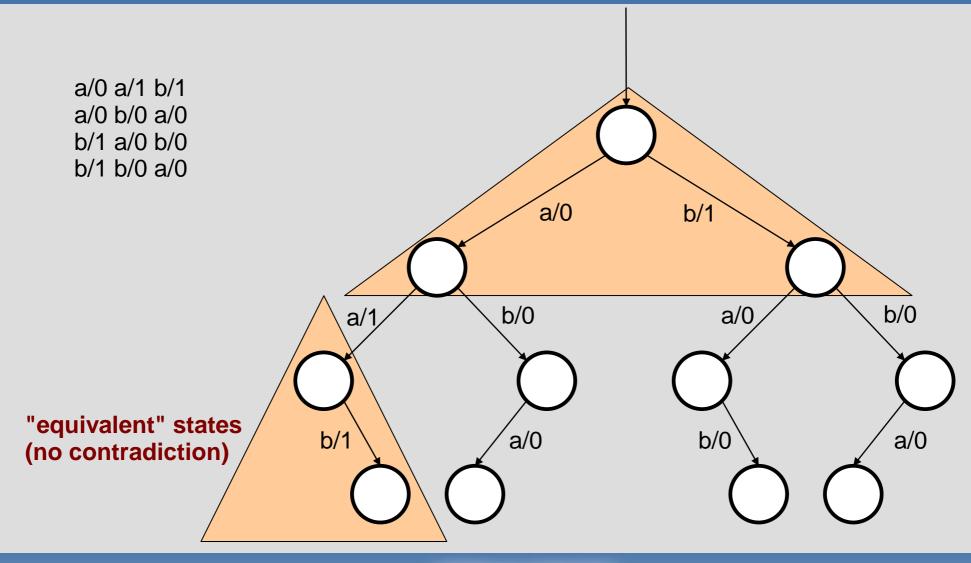


a/0 a/1 b/1 a/0 b/0 a/0 b/1 a/0 b/0 b/1 b/0 a/0





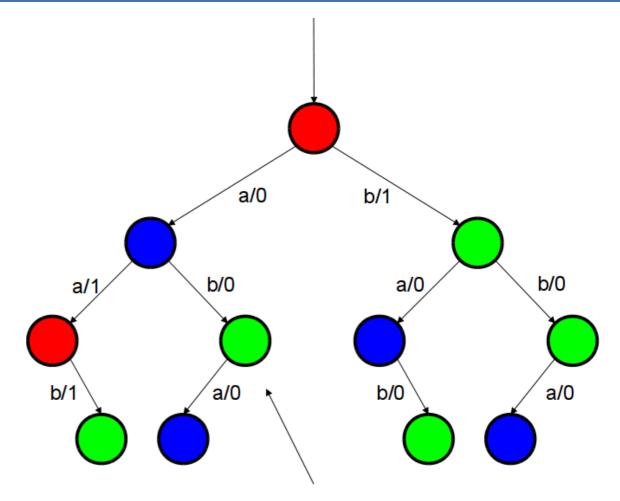








a/0 a/1 b/1 a/0 b/0 a/0 b/1 a/0 b/0 b/1 b/0 a/0

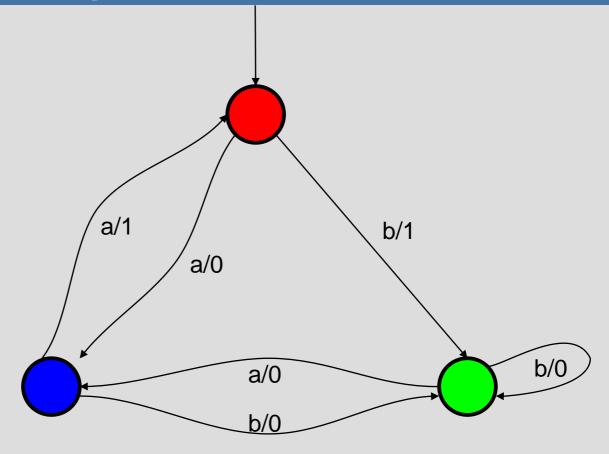


Could just as well be red...





a/0 a/1 b/1 a/0 b/0 a/0 b/1 a/0 b/0 b/1 b/0 a/0







## Observations

- The relation "not in conflict" is very weak:
  - Reflexive, symmetric, but not transitive!
  - Not in conflict clusters typically overlap
  - The relation contain various eqivalence relations
  - Computing the best choice of equivalence is:
    - Expensive for criteria like state minimality
    - Impossible in terms of adequacy for the problem.





## Active automata learning

## Idea 1: Ask where infomation is incomplete!

- This requires an active testing mechanism:
  - Membership Queries: Check the reaction of the system to input sequences.
- Checking all inputs at all positions makes `not in conflict´ an equivalence relation.





## Consequence:

- The underlying tree is homogeneous in the sense that all nodes treat the same set of inputs.
- As the not in conflict relation is now an equivalence relation, the corresponding clustering is unique

#### Problem:

The clustered graph may be non-deterministic in general





## Idea 2: Enforce consistency!

- Refine the ,not in conflict relation
- Also consider whether the target of the transitions of each cluster are unique for each input
  - I.e.: consider the largest congruence wrt. The Transition relation inside the not in conflict relation).

This yields determinism!





#### **Consequence:**

- Clustering yields an (input) deterministic graph /model)
- The projective quotient model of a consistent and homogeneous abstraction)

This simplifies the situation a lot:

#### **Termination Lemma 1**

Given some execution tree, realizing the two ideas via Membership Queries provides a closed, consistent, and deterministic Hypothesis Model

(Quality? Termination?)





## Idea 3: Introduce qualitative termination!

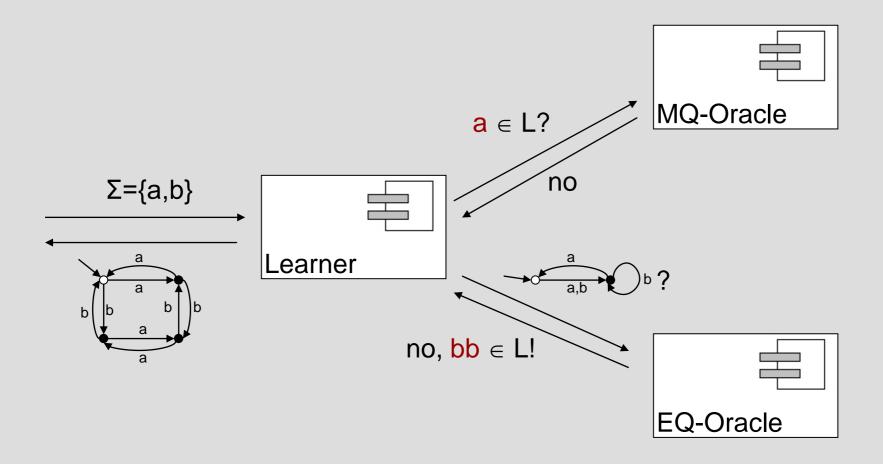
 Equivalence Queries: Check for equivalence with the target system, and produce a distinguishing test in case of failure.

- Conceptually a nice idea that leads to a very elegant correctness proof.
- Practically typically not implementable.





## Active automata learning







```
(queries)
should word w be included in L(A)?
yes / no
```

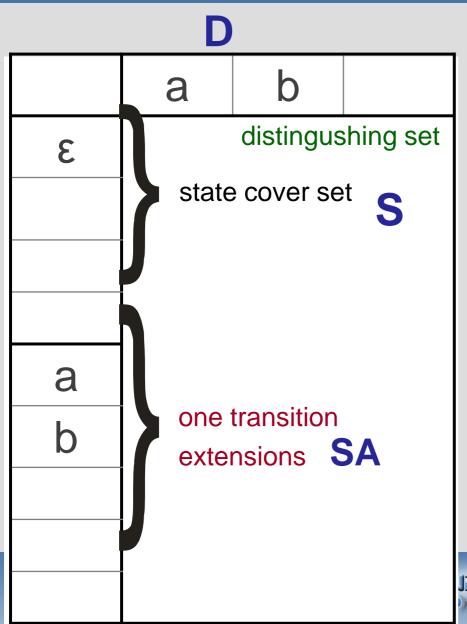
(conjectures) here is an A - is L(A) = U?

yes!

no: word w should (not) be in L(A)

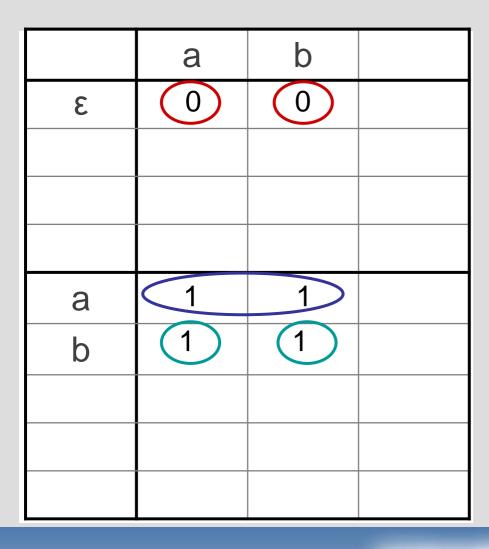




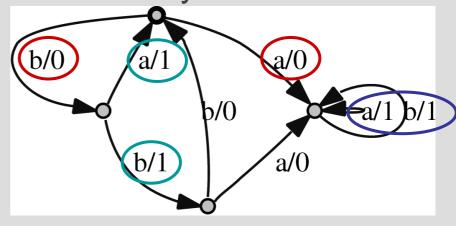


Initialize
 Distinguishing Set
 D with alphabet of inputs





#### Unknown system:







	а	b	
3	0	0	
а	1	1	
a b	1	1	
a b			
a b			

#### Unclosure:

Rows in lower part that are not in upper part



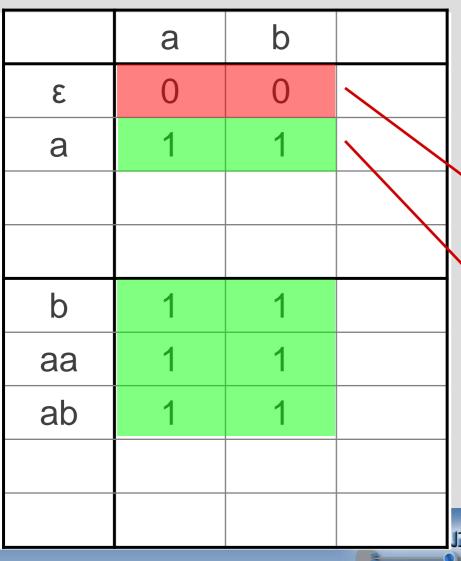
	а	b	
3	0	0	
a	1	1	
b	1	1	
aa	1	1	
ab	1	1	

#### Unclosure:

Rows in lower part that are not in upper part



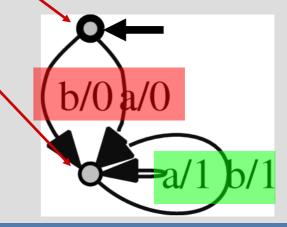




Conjecture:

 Unique rows in S become states

Rows in S and SA become transitions

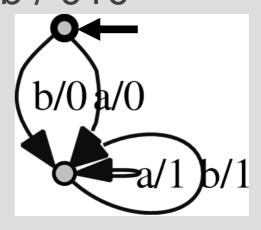






	а	b	
3	0	0	
a	1	1	
b	1	1	
aa	1	1	
ab	1	1	

### Counterexample: bbb / 010







	а	b	
3	0	0	
а	1	1	
b	1	1	
bb	0	0	
bbb	0	0	
aa	1	1	
ab	1	1	
ba	0	0	

Counterexample: bbb / 010

- Insert all prefixes of the counterexample to upper part
- Extend SA accordingly

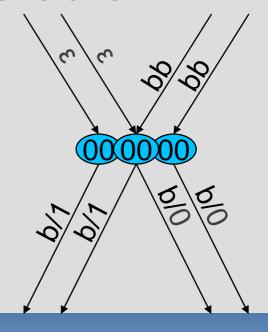




	а	b	
3	0	0	1
a	1	1	
b	1	1	
bb	0	0	J
bbb	0	0	
aa	1	1	
ab	1	1	
ba	0	0	

### Inconsistency:

 Equal rows in upper part have ,different extensions'





	а	b	
3	0	0	1
а	1	1	
b	1	1	
bb	0	0	
bbb	0	0	
aa	1	1	
ab	1	1	
ba	0	0	

### Inconsistency:

 Equal rows in upper part have ,different extensions'

- b and bbb differ, e.g.,
   for suffix b
- => **&** and **bb** will differ for suffix **bb**

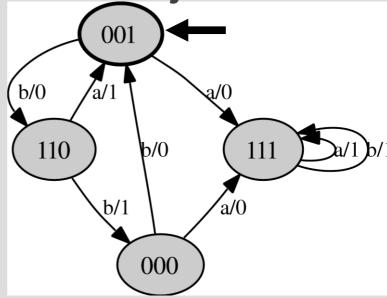




	а	b	bb
3	0	0	1
а	1	1 /	1
b	1	1	0
bb	0	0	, 0
bbb	0	0	1
aa	1	1	1
ab	1	1	1
ba	0	0	1

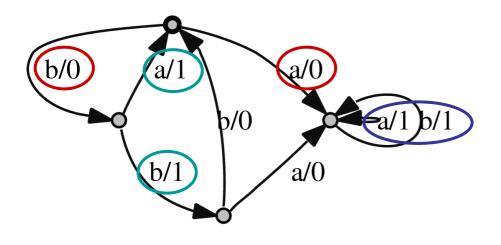
 Inconsistencies lead to new columns

**New Conjecture** 

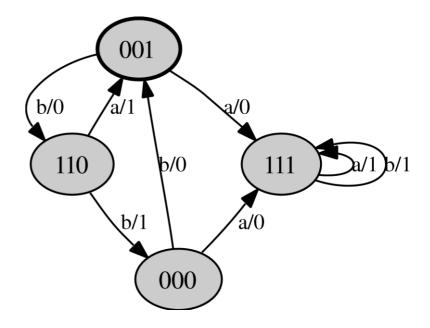




#### **Target System**



#### Learned System







# Summarized Observations (1)

- Systematic completition of the observation table
- New states arise as targets of transitions or from counter examples of the equivalence queries.
   Technically: prefixes are added to S
- Closure procedure extends SA
- Consistency is enforced by enlarging the Distinguishing Set D





## Hypothesis models or conjectures:

- Closed and consistent models (projective quotients) of the so far expanded
  - homogeously extended, and
  - consistent

execution tree.





# Summarized Observations (2)

#### **Invariance Lemma:**

- All hypothesis models are
  - totally defined: each input is considered at each state,
  - input deterministic: there is only one transition per input at each state,
  - transition covered: each transition lies on a path of the original system,
  - state minimal: two different states in a hypothesis model always have a separating future – á la Nerode).





## Myhill-Nerode

#### **Nerode relation:**

For language L define relation  $R_L$  (for  $u, u' \in \Sigma^*$ )

$$u R_L u' \leftrightarrow \text{for all } v \in \Sigma^*$$
:  $(uv \in L \leftrightarrow uv \in L)$ 

## **Myhill-Nerode Theorem:**

 Minimal number of states of an accepting deterministic automaton equals the number of equivalence classes of R<sub>L</sub>





## Summarized Observations (4)

This (Nerode's theorem) directly yields:

 Corollary: Hypothesis automata have at most as many states as the smallest deterministic equivalent automaton.

We will denote the number of states by n.





# Summarized Observations (4)

Lemma: The number of states of the hypothesis model increases in response to a counterexample.

Theorem: Angluin's algorithm terminates after at most **n** equivalence queries with the smallest deterministic system representing the behaviour of the system to be learned.





### Complexity of Angluin

#### Equivalence Queries

At most |Q|

#### Membership Queries

At most  $O(m |Q| |\Sigma_A|)$  per EQ (m = length of max. counter example)

Max. size of table =  $O(m |Q|^2 |\Sigma_A|)$ .

Theorem (Complexity for const. Time MQs and EQs). O( m  $|Q|^2 |\Sigma_A|$  ).

For m in O(|Q|) the complexity result reads:  $O(|Q|^3 |\Sigma_A|)$ 





#### Remaining Problems

- High Computational Complexity
- Even worse: equivalence queries in general undecidable.

#### In essence:

- Active automata learning always remains at the level of hypotheses:
  - neither correct nor complete





# Further Developments





## Conceptual Improvements 1

#### one erssential suffix

All prefixes of counterexample

			۱ ۱
	а	b	bb
3	0	0	
а	1	1	
b	1	1	
bb	0	0	
bbb	0	0	
aa	1	1	
ab	1	1	
ba	0	0	





#### Reduced observation table

- Rivest and Shapire: Analyze counterexample separately (not in the table)
  - Only add one ,essential' suffix (i.e., witness), as column label to the table

Consequence: Guaranteed Consistency!

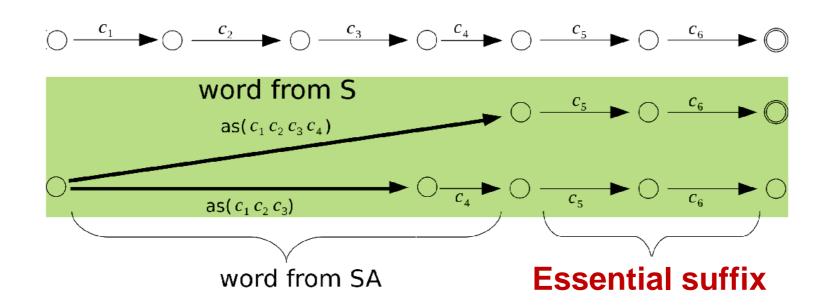
**BUT:** Hypothesis Automata are **no longer** guaranteed to be **minimal!** 

(cf. Pnueli / Mahler's criticism)





#### Reduced observation table (contd.)



Saves membership queries! (by saving rows in the observation table)





## Complexity (reduced observation table)

Equivalence Queries

At most |Q|

Membership Queries for guaranteed progress after Eqs

At most  $O(log_2(m) + |\Sigma_A||Q|)$  per EQ (m = length of max. counter example)

Max. size of table =  $O(|Q|^2 |\Sigma_A|)$ .

Theorem (Complexity for const. Time MQs and EQs).  $O(|Q|^2 |\Sigma_A| + |Q| \log_2(m))$ .

For m in O(|Q|) the complexity result reads:  $O(|Q|^2 |\Sigma_A|)$ 





## Conceptual Improvements 2

	а	b	bb
3	0	0	
а	1	1	
b	1	1	
bb	0	0	
bbb	0	0	
aa	1	1	
ab /	1	1	
ba	9	0	

All rows are filled completely, even if unnecessary





#### Discrimination tree

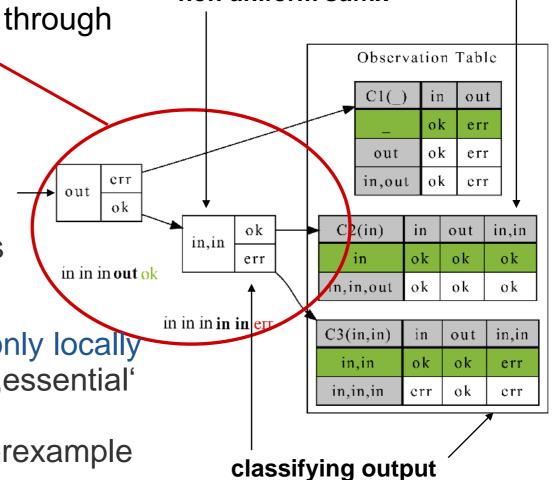
'Sink' words into table through discrimination tree

**Angluin**: Add suffix globally to all rows

- leads to unclosedness
- resolved by new elements in S

Kearns & Vazirani: Add suffix only locally

- Suffix only added to one, essential' sub-table.
- Prefix known from counterexample



non uniform suffix





#### Discimination tree (contd.)

#### **Kearns & Vazirani** + discrimination tree

- Saves membership queries! (by saving entries in the observation table)
- More equivalence queries! (using suffixes globally may be a good heuristic sometimes)
- Worst case complexity unchanged





## Correctness pattern (maintained)

Lemma. Each counterexample leads to at least one new state.

Lemma. The hypothesis automata are guaranteed to have fewer states than the minimal deterministic finite automaton for the considered language.

#### Theorem (for perfect equivalence oracle)

The algorithm terminates with the **smallest determinsitc automaton** for the considered language / set of traces.





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#### Practical results II

# The ZULU competition





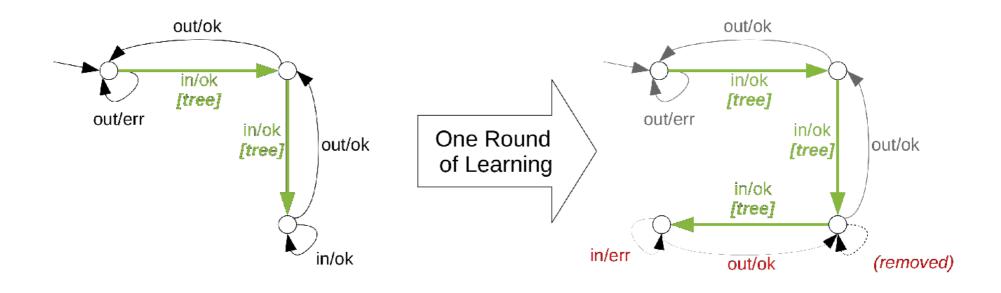
#### The ZULU challenge

- Competition in active learning (2010)
- No equivalence queries allowed, limited amount of membership queries
- Randomly generated automata
- Test-based evaluation
- http://labh-curien.univ-st-etienne.fr/zulu/





### **Evolving hypothesis**

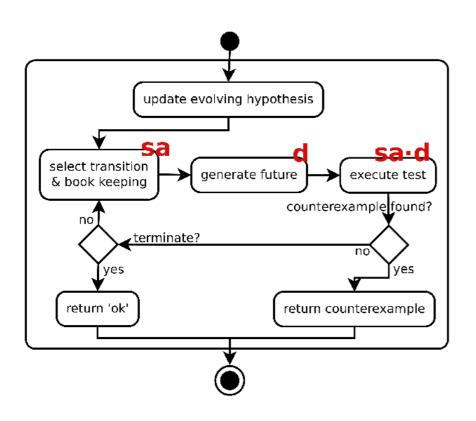


- The set  $S \cup SA$  defines a monotonically growing spanning tree of the target automaton
- Usually only local modifications between two equivalence queries (especially for non-uniform sets of distinguishing suffixes)





#### Continuous equivalence queries



- Select transition: randomly from set of non-blocked
- **Generate future:** randomly with increasing length. Initially  $max \left\{ \frac{log(n)}{2}, 3 \right\}$ .
- Book keeping:

E.H.Blocking: transitions excluded from subsequent tests.

E.H.Weighted: weights on transitions are increased.

**■ Termination:** ZULU limit





#### ZULU competition results

Algorithm	Dist. Set		Equivalence	Training	Rank
	Init.	Uniform		(Avg.)	
E.H.Blocking			block transitions	89.38	1
E.H.Weighted	$\{\epsilon\}$	no	weight transitions	89.26	2
Random			random walks	88.93	6
run_random			random walks	80.17	14
run_blocking1	$\{\epsilon\} \cup \Sigma$	yes	block transitions	79.89	15
run_weighted1			weight transitions	79.65	16

#### Kearns & Vazirani: High impact even here!

Uniform DFA: ca. 83, non-uniform Mealy: ca. 85





#### Detailed results

Algorithm	New Membership Queries			Rounds	States	Score
	Close	Analyze	Search			
E.H.Blocking	6,744	358	999	259	352	94.11
E.H.Weighted	6,717	349	1,035	262	351	94.61
Random	6,586	519	996	228	332	93.28
run_random	8,080	14	7	5	312	74.89
run_blocking1	8,074	11	16	6	319	73.06
run_weighted1	8,077	9	15	6	319	74.39

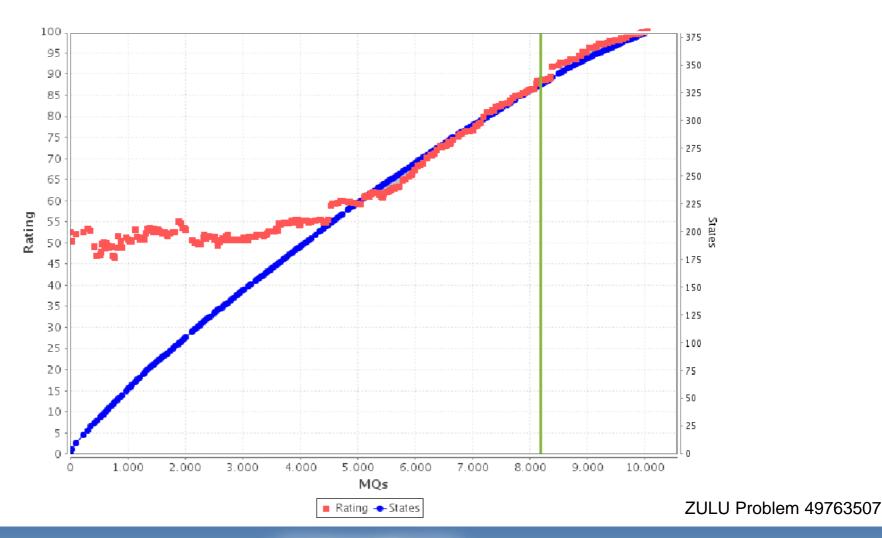
- ZULU limit: 8,101
- MQs / EQ: 1-3 (uniform), ca. 3.9 (non-uniform), ca. 4.36 (random)
- MQS / State: ca. 25 (uniform), ca. 19 (non-uniform)
- Random Walks: higher costs for analyzing counterexamples





## Asymtotic costs per state

#### Classification Per Round







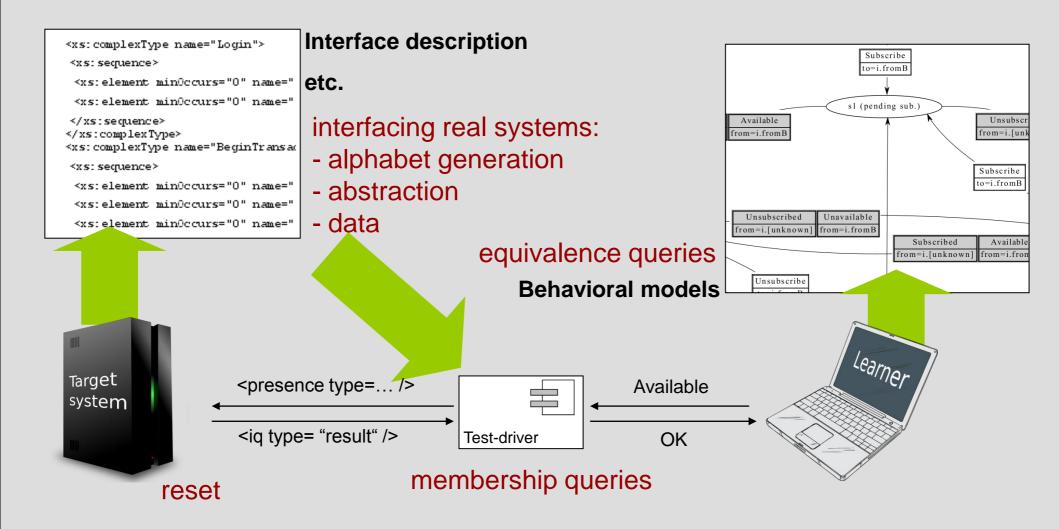
#### Practical results I

## More Applications





#### Practical challenges







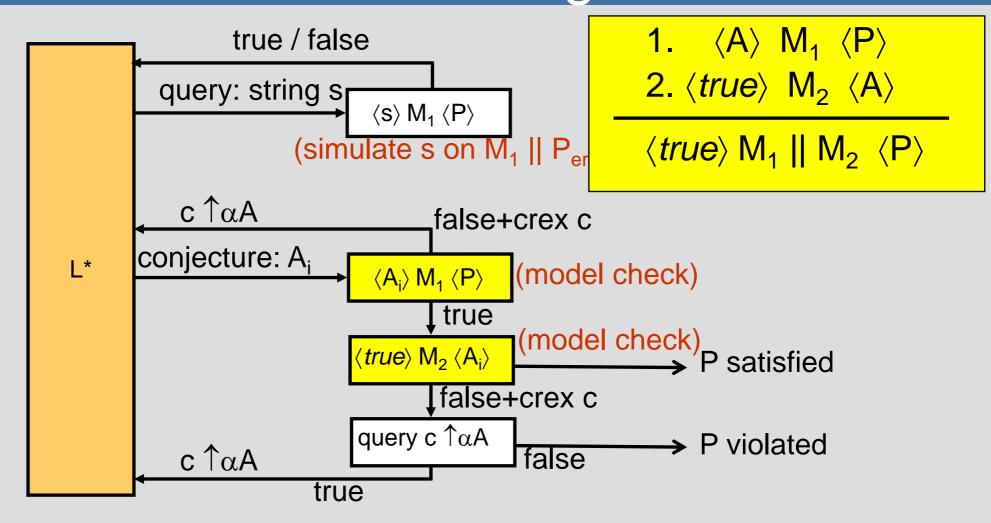
#### Practical results I

## Learning assumptions





# oracle for WA in assume-guarantee reasoning







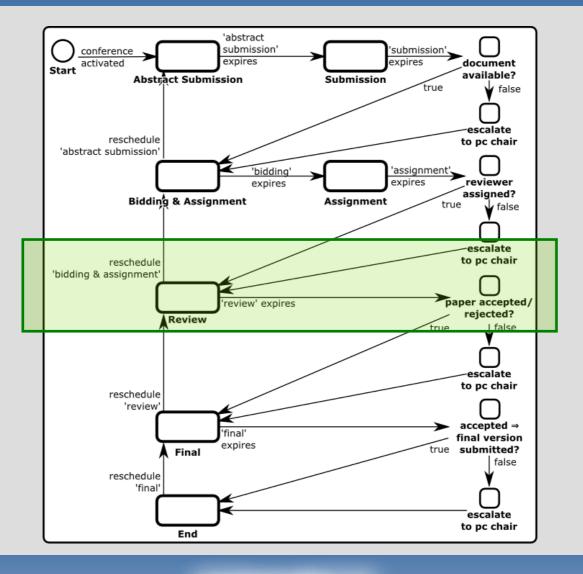
#### Practical results I

# Learning the OCS



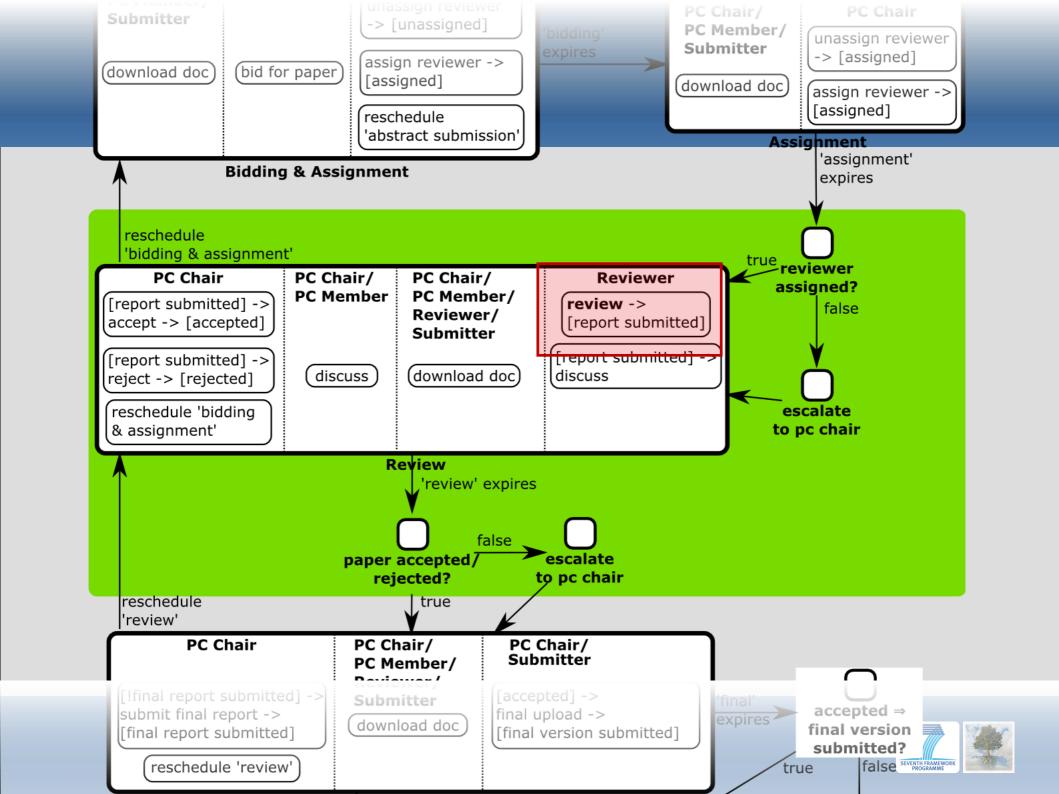


## User model: paper workflow

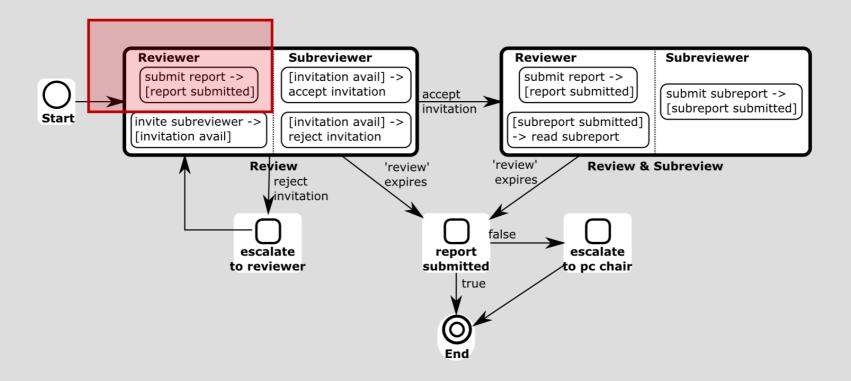








### Hierarchy

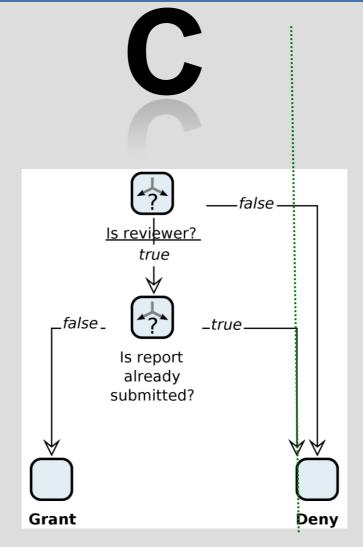


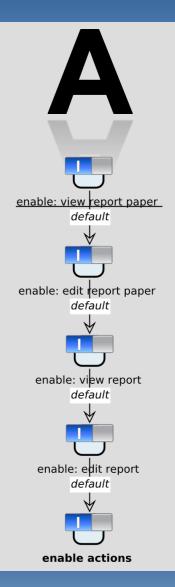




#### **Event Condition Action**



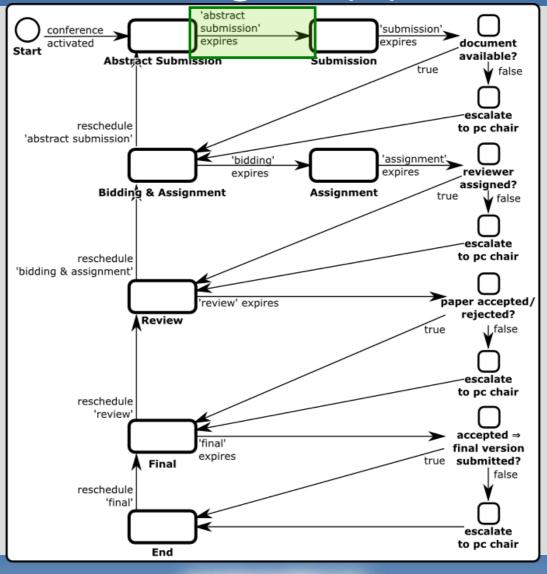








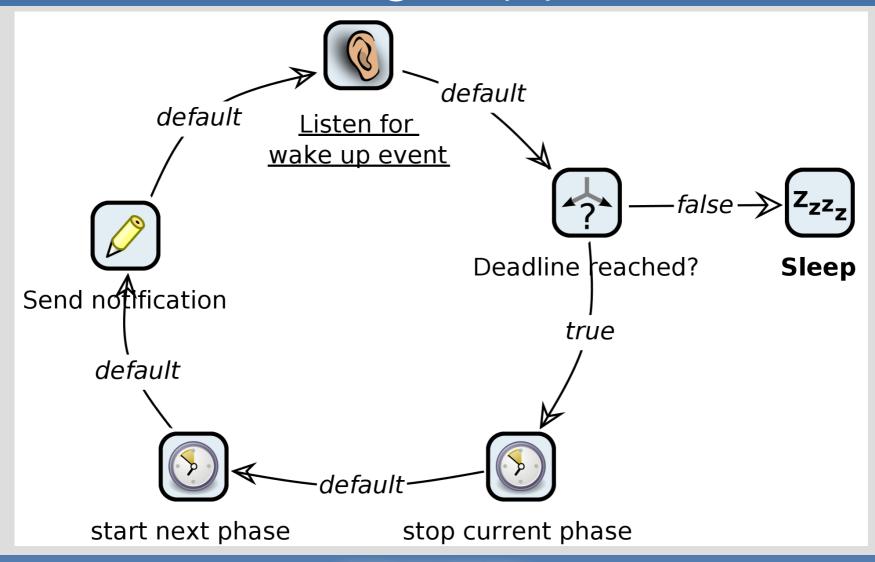
# Semantics of "phase expires"-edges (1)







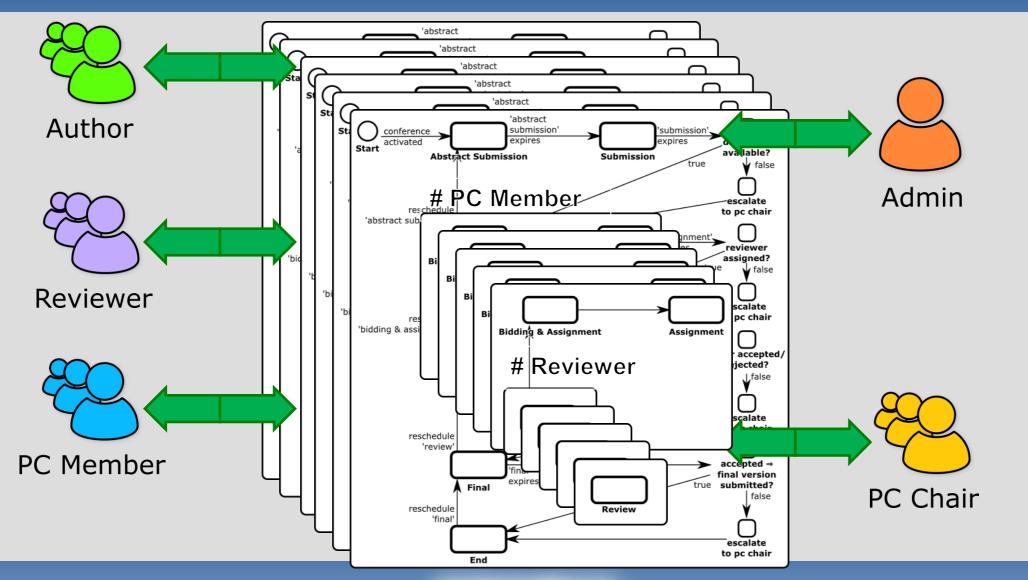
# Semantics of "phase expires"-edges (2)







## Many participants







# Putting it all together

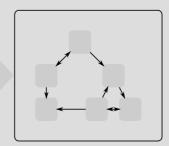


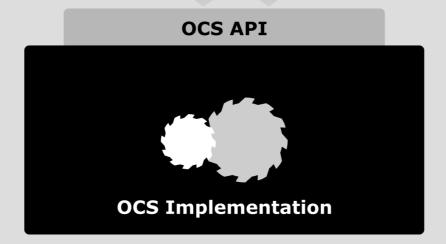


#### Regular extrapolation





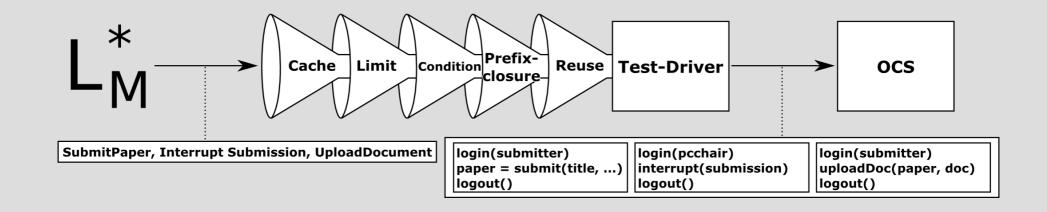








## Optimized learning setup







## Learning algorithm

- Observation Table
- Mealy machine inference
- Regular extrapolation

	SP	UD	DD	DP
$\lambda$		至	4	4
SP	4		4	
UD		4	4	4
DD		\$	4	4
DP		\$	\$	\$



	SP	UD	DD	DP
$\lambda$		4	4	4
SP	差	$\sqrt{}$	至	$\sqrt{}$
UD		至	至	Ź
DD		\$	至	Ź
DP		\$	Ź	\$
SPSP	1		4	$\sqrt{}$
SPUD	1	$\sqrt{}$		$\sqrt{}$
SPDD	1		4	$\sqrt{}$
SPDP		\$	4	Ź

#### **First Hypothesis**

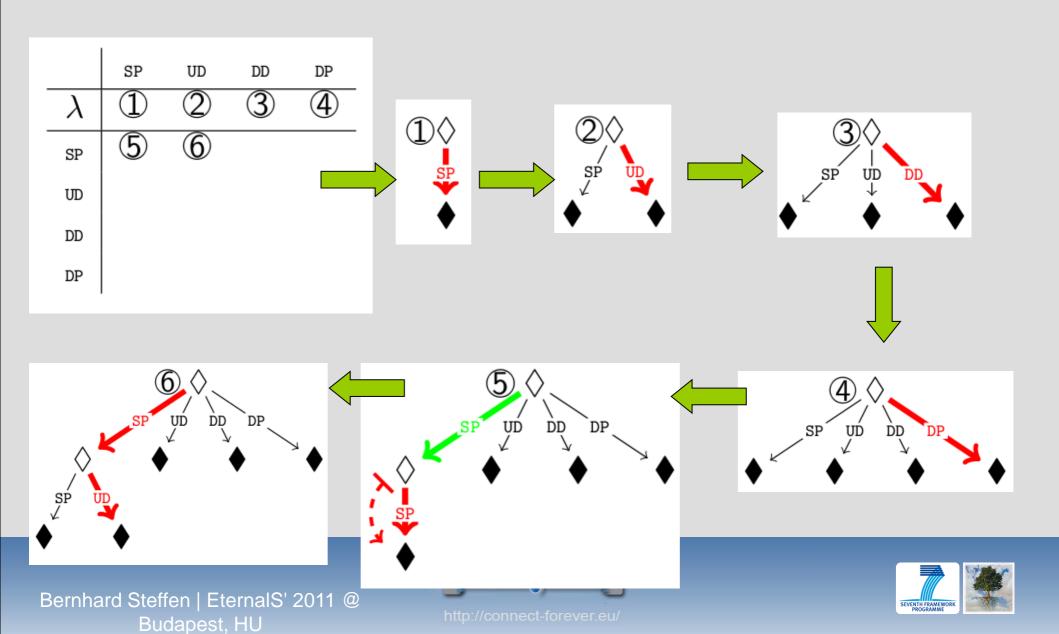
	SP	UD	DD	DP
λ		4	4	4
SP	1		4	$\sqrt{}$
SPUD	至	$\sqrt{}$	$\sqrt{}$	$\sqrt{}$
UD		\$	4	Ź
DD		\$	4	\$
DP		4	4	4
SPSP	\$	$\sqrt{}$	4	$\sqrt{}$
SPDD	1		4	$\sqrt{}$
SPDP		\$	4	\$
SPUDSP	\$			$\sqrt{}$
SPUDUD	\$			$\sqrt{}$
SPUDDD	至	$\sqrt{}$		$\sqrt{}$
SPUDDP		\$	4	Ź



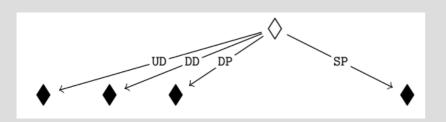


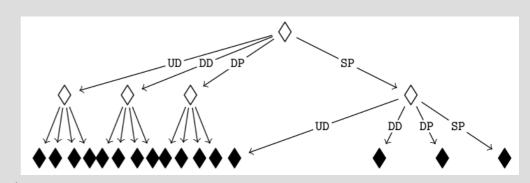


### Reusing system states

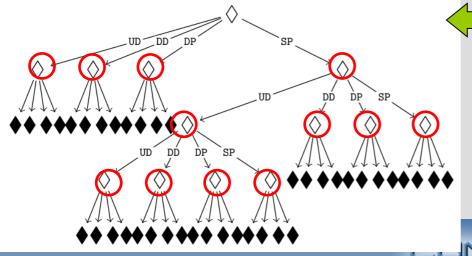


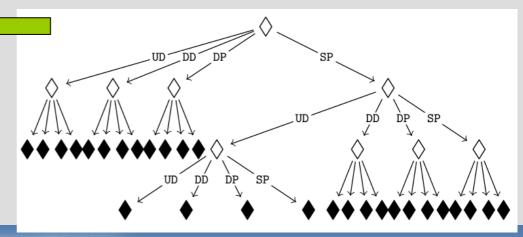
### Reuse tree on our example





- 52 Membership Queries
- Saved 12 Resets



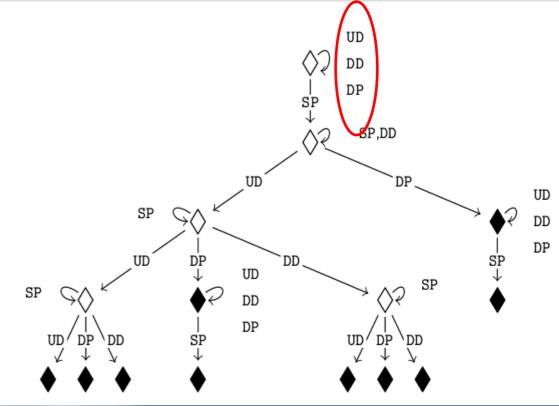






#### Exploitation: failure invariance

- Domainknowledge
- Failing actions due to missing permissions
- OCS is transaction secure (roll back in case of error)
- Partition output alphabet into successful and failed execution
- Reflexive edges indicate failure output

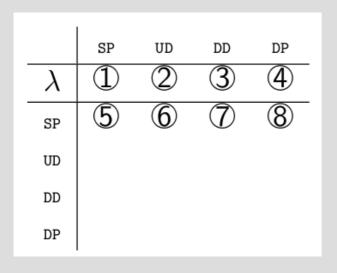


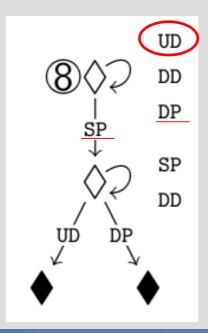




## Pumping: Unfolding edges

- For 52 Membership Queries only 10 Resets necessary
- 50 Symbols executed (of 148)





Queries 9 to 20 will be 'pumped', e.g.

- UD UD or
- DP SP

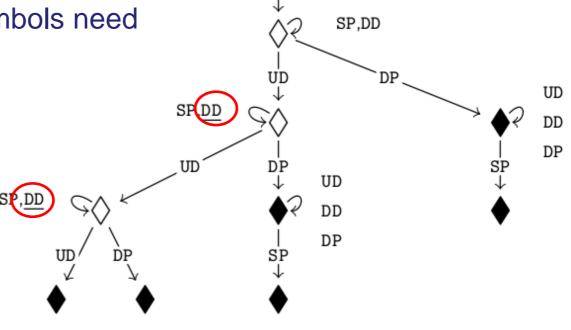
already known





### Exploitation: Invariant symbols

- Downloading (reading) a document
   (DD) does not change a system state
- The state can be kept for reuse
- Only 6 Resets and 35 Symbols need to be executed



UD

DD

DP

Failure invariance + invariant input symbol DD.





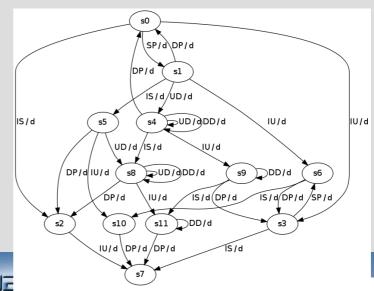


#### Statistics: Learning the OCS

#### 438 MQs containing 1734 Symbols:

	Resets	Reuses	Pumped	Reset [t]	Symbols [t] (#)	Observed [t]
(a)	438	0	0	7m 50s	8m 28s (1734)	16m 18s
(b)	366	72	0	7m 14s	7m 55s (1518)	15m 9s
(c)	328	86	24	5m 23s	5m 51s (1345)	11m 14s
(d)	56	130	252	0m 52s	1m 25s (344)	2m 17s
(e)	37	125	276	0m 34s	0m 59s (252)	1m 33s

- a) No reuse
- b) Only direct re-usage
- c) Exploit input knowledge
- d) Exploit output knowledge
- e) Exploit input and output knowledge





### Statistics: Learning the OCS

	Q	Σ	MQs	Resets	Reuses	Pumped
(a)	3	3	30	21	9	0
(b)	11	5	280	31	84	165
(c)	11	5	280	17	78	185
(d)	40	9	3882	137	757	2988
(e)	66	13	12210	646	2514	9050
(f)	160	18	56284	5598	12168	38518

MQs = Resets + Reuses + Pumped

- a) Only direct reusage
- b) Exploit of failure outputs
- c) Like b) but one input marked as invariant
- d)-f) failure outputs and invariant for growing learn alphabets



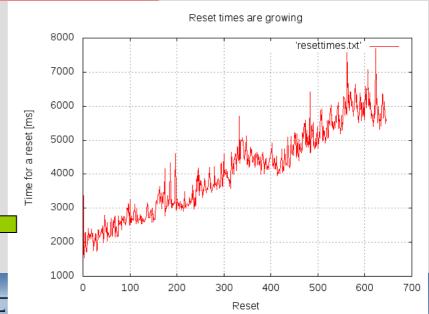


# Statistics: Learning the OCS

Resets	Avg. reset	Acc. reset	Observed runtime
21	1.8s.	56s.	40s.
31	2.3s.	10m	1m 25s.
17	2.1s.	9m	48s.
137	3.2s.	3h 27m	10m 30s
646	4.1s.	13h 52m	53m 50s
5598	12.7s.	over 8 days	pprox 22h
	21 31 17 137 646	21 1.8s. 31 2.3s. 17 2.1s. 137 3.2s. 646 4.1s.	21       1.8s.       56s.         31       2.3s.       10m         17       2.1s.       9m         137       3.2s.       3h 27m         646       4.1s.       13h 52m

- Reset times are growing
- Observed runtime included execution of input symbols

Accumulated reset time is highly **optimistic**!



## Simple User Process

SP: Submit Paper

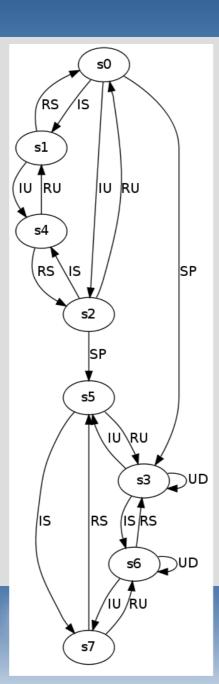
**UD: Upload Document** 

IS: Interrupt Submission

IU: Interrupt Upload

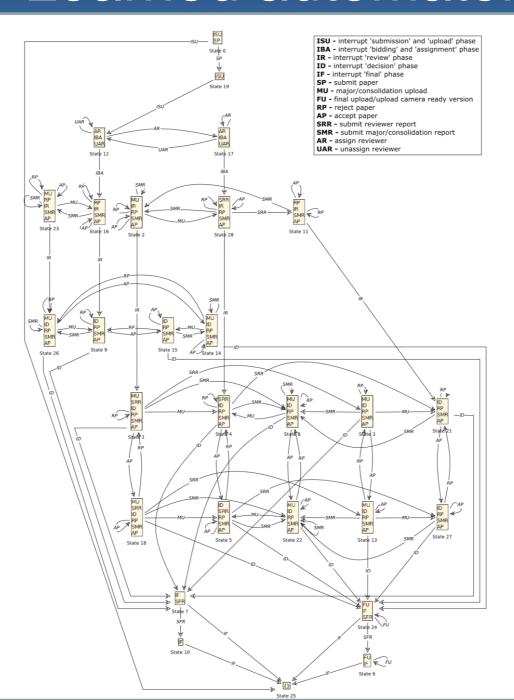
**RS: Restart Submission** 

**RU: Restart Upload** 





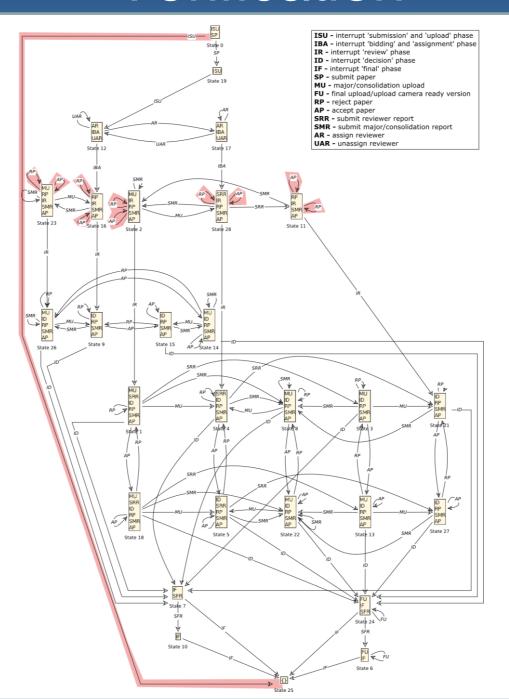
#### Learned automaton







# Verification





#### Overview

- Motivation
- Introduction to active automata learning
- Practical aspects in active automata learning
- Conclusions





#### Conclusions

#### **Active Automata Learning:**

- its practice has many facets:
  - Abstraction
  - Instrumentation
  - Reuse/Optimization
- It establishes a new system perspective Systems as evolving ,beasts':
  - to be observed continuously
  - difficult to control:

Forget the ,Why/How' focus on the ,What'!



