## Active Learning of Mealy Machines with Timers

Véronique Bruyère, Bharat Garhewal, Guillermo A. Pérez, Gaëtan Staquet, Frits W. Vaandrager

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- ► Network protocols;
- Schedulers;
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Learning algorithm

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In short: finite Mealy machines augmented with **clocks** that can be reset or used in guards along transitions and states.

**BUT** timed Mealy machines are hard to construct and understand.

We focus on systems that can be represented with timers: Mealy machines with timers.

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Motivation: timed systems

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Motivation: timed systems

$$\mathcal{M} = (X, I, O, Q, q_0, \delta)$$
 where

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- $\triangleright$   $\delta$  is the transition function.

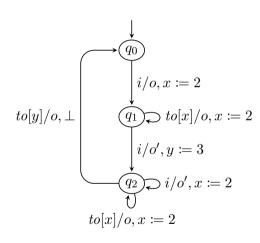


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$$to[x]/o, x \coloneqq 2 \qquad i/o', x \coloneqq 2$$

$$i/o, x \coloneqq 2 \qquad i/o', y \coloneqq 3$$

$$to[y]/o, \bot$$

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Learning algorithm

$$(q_0,\emptyset)$$

Learning algorithm

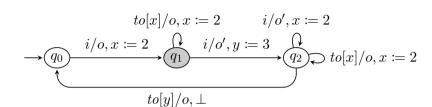
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$$(q_0,\emptyset) \xrightarrow{1} (q_0,\emptyset)$$



$$(q_0, \emptyset) \xrightarrow{1} (q_0, \emptyset) \xrightarrow{i/o} (q_1, x = 2)$$

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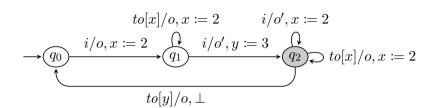
Learning algorithm

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Learning algorithm

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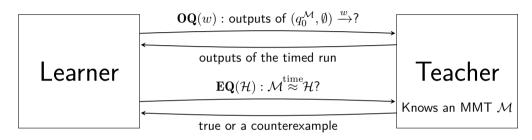


Figure 2: Adaptation of Angluin's framework<sup>2</sup> to MMTs.

 $<sup>^2</sup>$ Angluin, "Learning Regular Sets from Queries and Counterexamples", 1987; Shahbaz and Groz, "Inferring mealy machines", 2009.

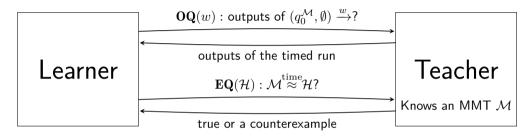


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Both queries are in the **timed** world... Cumbersome to use!

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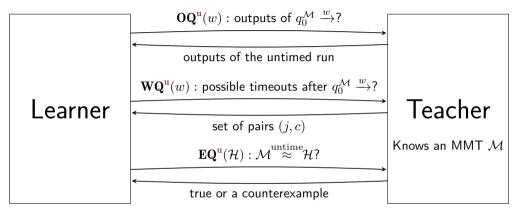


Figure 3: Untimed adaptation of Angluin's framework to MMTs.

We stay in the untimed world!

Does **not** hold for all MMTs!

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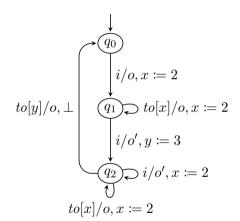
- timeouts are observed via their outputs,
- ► for every untimed sequence of transitions, there exists a timed run using **exactly** this sequence of transitions...
- ▶ with all delays > 0 and there is at most one timer that times out at any time (see Bruyère, Pérez, et al., "Automata with Timers", 2023).

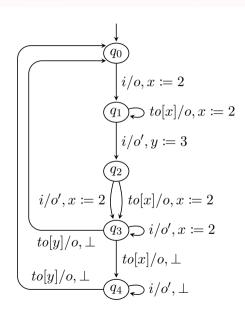
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**Proposition 2.** It is possible to construct an MMT in which the second condition is satisfied.





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<sup>&</sup>lt;sup>3</sup>Vaandrager et al., "A New Approach for Active Automata Learning Based on Apartness", 2022.

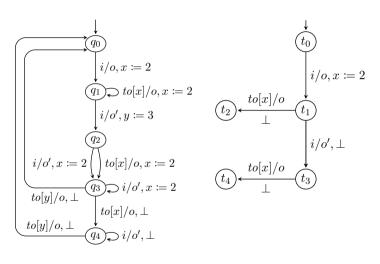
We adapt  $L^{\#}$  (active learning algorithm for Mealy machines<sup>3</sup>) to MMTs:  $L_{MMT}^{\#}$ .

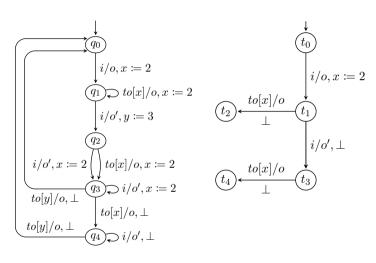
**Theorem 3.** Let  $\mathcal{M}$  be a "good" MMT and  $\ell$  be the length of the longest counterexample returned by the teacher. Then,

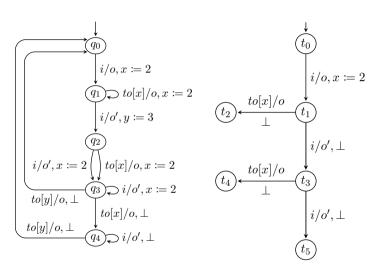
- lacktriangle the  $L^\#_{MMT}$  algorithm eventually terminates and returns an MMT  ${\mathcal N}$  such that  $\mathcal{M} \stackrel{\mathrm{time}}{\approx} \mathcal{N}$  and whose size is **polynomial** in  $|Q^{\mathcal{M}}|$  and **factorial** in  $|X^{\mathcal{M}}|$ , and
- in time and number of untimed queries **polynomial** in  $|Q^{\mathcal{M}}|, |I|$ , and  $\ell$ , and factorial in  $|X^{\mathcal{M}}|$ .

Learning algorithm

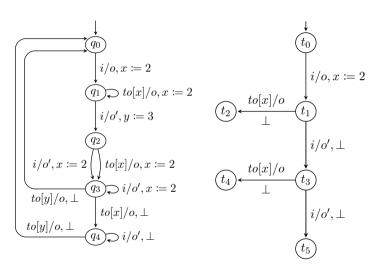
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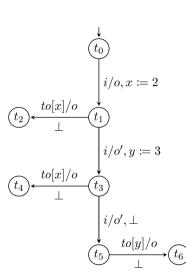


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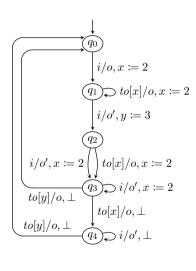
► 
$$\mathbf{WQ^{u}}(i \cdot i \cdot i)$$
  
  $\sim \{(2,3),(3,2)\}.$ 

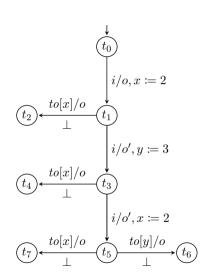


- $ightharpoonup \mathbf{OQ^u}(i \cdot i \cdot i) \sim o \cdot o' \cdot o'.$
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- ▶ So.  $t_1 \stackrel{i}{\rightarrow} t_3$  starts a timer at constant 3.

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- ightharpoonup And  $t_3 \xrightarrow{i} t_5$  starts a timer at constant 2.

We implemented  $L_{\rm MMT}^{\#}$  in Rust<sup>4</sup> and ran some experiments.

Model	Q	I	X	$ \mathbf{WQ^u} $	$ \mathbf{OQ^u} $	$ \mathbf{EQ^u} $	Time[msecs]
AKM	4	5	1	22	35	2	684
CAS	8	4	1	60	89	3	1344
Light	4	2	1	10	13	2	302
PC	8	9	1	75	183	4	2696
TCP	11	8	1	123	366	8	3182
Train	6	3	1	32	28	3	1559
Running example	3	1	2	11	5	2	1039
FDDI 1-station	9	2	2	32	20	1	1105
Oven	12	5	1	907	317	3	9452
WSN	9	4	1	175	108	4	3291

<sup>4</sup>https://gitlab.science.ru.nl/bharat/mmt lsharp.

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# Thank you!

For all details, see Bruyère, Garhewal, et al., "Active Learning of Mealy Machines with Timers", 2024.

# Part I – Appendix

Appendix

### References I

- Angluin, Dana. "Learning Regular Sets from Queries and Counterexamples". In: *Inf. Comput.* 75.2 (1987), pp. 87–106. DOI: 10.1016/0890-5401(87)90052-6.
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