Active Learning of Mealy Machines with Timers

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

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Many computer systems have timing constraints:

- ► Network protocols;
- Schedulers;
- ► Embedded systems;
- ► In general, real-time systems.

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.

Timed Mealy machines Mealy machines with timers

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- ► Clocks go from 0 to infinity:
- **>**

Mealy machines with timers

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Motivation: timed systems

- ► Timers go from a given value to 0:
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- ► Mealy machines with timers are more restrictive:
- ► We previously studied some properties of Mealy machines with timers;¹

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- Learning timed Mealy machines is
 This work: learning algorithm. challenging.

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A Mealy machine with timers

(MMT) is a tuple

Motivation: timed systems

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

- ► *X* is the set of **timers**:
- I is the set of **inputs**; the set of all actions is:

$$I \cup \{to[x] \mid x \in X\};$$

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A Mealy machine with timers

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- $ightharpoonup q_0 \in Q$ is the initial state;







Figure 1: An MMT.

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- \triangleright δ is the transition function.

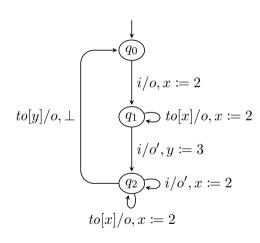


Figure 1: An MMT.

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

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

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

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Learning algorithm

$$(q_0,\emptyset)$$

Learning algorithm

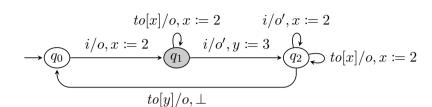
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$$(q_0,\emptyset) \xrightarrow{1} (q_0,\emptyset)$$



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

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

$$i/o, x := 2 \qquad 0 \qquad i/o', y := 3 \qquad 0$$

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$$(q_0, \emptyset) \xrightarrow{1} (q_0, \emptyset) \xrightarrow{i/o} (q_1, x = 2) \xrightarrow{2} (q_1, x = 0)$$

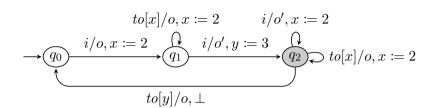
Learning algorithm

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$$\xrightarrow{0} (q_1, x = 2) \xrightarrow{i/o'} (q_2, x = 2, y = 3)$$



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Learning algorithm

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$$\xrightarrow{i/o'} (q_2, x = 2, y = 1) \xrightarrow{0.5} (q_2, x = 1.5, y = 0.5).$$

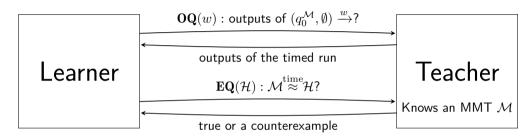


Figure 2: Adaptation of Angluin's framework² to MMTs.

 $^{^2}$ Angluin, "Learning Regular Sets from Queries and Counterexamples", 1987; Shahbaz and Groz, "Inferring mealy machines", 2009.

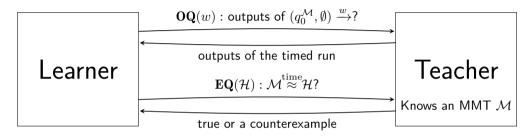


Figure 2: Adaptation of Angluin's framework² to MMTs.

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

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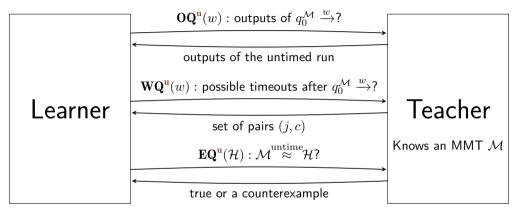


Figure 3: Untimed adaptation of Angluin's framework to MMTs.

We stay in the untimed world!

Does **not** hold for all MMTs!

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timeouts are observed via their outputs,

Learning algorithm

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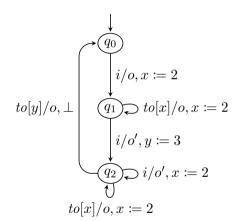
- timeouts are observed via their outputs,
- ► for every untimed sequence of transitions, there exists a timed run using **exactly** this sequence of transitions...
- ▶ with all delays > 0 and there is at most one timer that times out at any time (see Bruyère, Pérez, et al., "Automata with Timers", 2023).

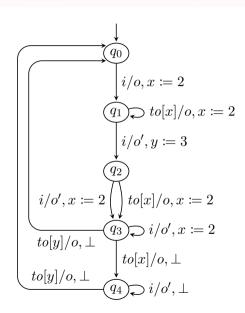
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Proposition 2. It is possible to construct an MMT in which the second condition is satisfied.





Learning algorithm

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³Vaandrager et al., "A New Approach for Active Automata Learning Based on Apartness", 2022.

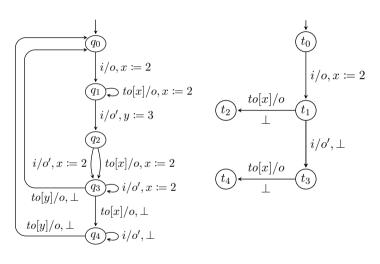
We adapt $L^{\#}$ (active learning algorithm for Mealy machines³) to MMTs: $L_{MMT}^{\#}$.

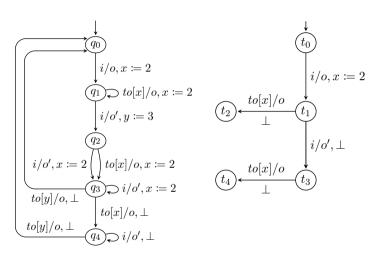
Theorem 3. Let \mathcal{M} be a "good" MMT and ℓ be the length of the longest counterexample returned by the teacher. Then,

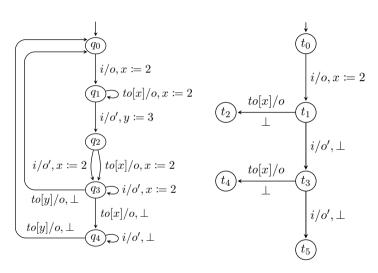
- lacktriangle the $L^\#_{MMT}$ algorithm eventually terminates and returns an MMT ${\mathcal N}$ such that $\mathcal{M} \stackrel{\mathrm{time}}{\approx} \mathcal{N}$ and whose size is **polynomial** in $|Q^{\mathcal{M}}|$ and **factorial** in $|X^{\mathcal{M}}|$, and
- in time and number of untimed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

Learning algorithm

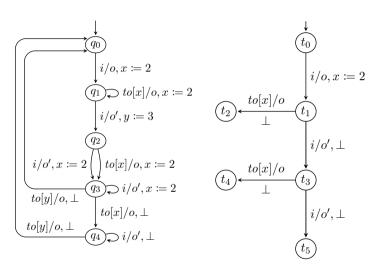
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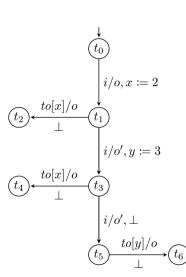
$$ightharpoonup$$
 So, $t_3 \xrightarrow{i/o'} t_5$.



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►
$$\mathbf{WQ^{u}}(i \cdot i \cdot i)$$

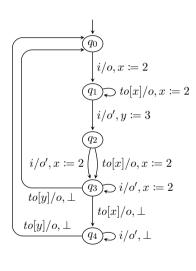
 $\sim \{(2,3),(3,2)\}.$

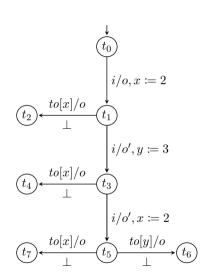


- $ightharpoonup \mathbf{OQ^u}(i \cdot i \cdot i) \sim o \cdot o' \cdot o'.$
- ightharpoonup So, $t_3 \xrightarrow{i/o'} t_5$.
- $ightharpoonup \mathbf{WQ^u}(i \cdot i \cdot i)$ $\sim \{(2,3),(3,2)\}.$
- ▶ So. $t_1 \stackrel{i}{\rightarrow} t_3$ starts a timer at constant 3.

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

We implemented $L_{\rm MMT}^{\#}$ in Rust⁴ 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

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

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- ► Further experiments with more timers,
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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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