

Learning Realtime One-Counter Automata

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1. Motivation
2. Learning deterministic finite automata
3. Learning realtime one-counter automata
4. Experimental results

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2. Learning deterministic finite automata
3. Learning realtime one-counter automata
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```
{ "title": "Learning ROCAs",  
  "place": {  
    "city": "Brussels",  
    "country": "Belgium"  
  },  
  "authors": ["Bruyère", "Pérez", "Staquet"]  
}
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- An **object** is an **unordered** collection of key-value pairs.

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- ▶ An object is an **unordered** collection of key-value pairs.
- ▶ An **array** is an **ordered** collection of values.

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```

- ▶ An object is an **unordered** collection of key-value pairs.
- ▶ An array is an **ordered** collection of values.

Here, let us fix an **order** on the keys inside an object. That is, we can assume objects are **ordered**.

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How to know whether a JSON document satisfies a given set of constraints?

^aFor XML documents, see Chitic and Rosu, “On Validation of XML Streams Using Finite State Machines”, 2004

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↔ Automata-based verification^a.

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How to know whether a JSON document satisfies a given set of constraints?

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What kind of automata can be used? How to construct such an automaton?

↔ Realtime one-counter automata (ROCA) and our learning algorithm!

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Overview:

- ▶ Based on learning algorithm for visibly one-counter automata (VCA).¹

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- ▶ We extend data structure to take into account the counter value.
 - ▶ Some values are unknown and left as **wildcards**.
 - ▶ Obtaining an hypothesis is harder than for VCAs.

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1. Motivation

2. Learning deterministic finite automata

- Deterministic finite automaton
- Active learning
- Data structure: the observation table

3. Learning realtime one-counter automata

4. Experimental results

A **deterministic finite automaton** (DFA) is a tuple $\mathcal{A} = (Q, \Sigma, \delta, q_0, F)$ where:

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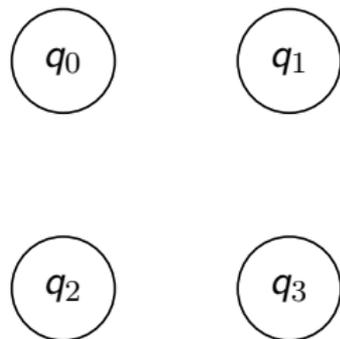


Figure 1: A DFA \mathcal{A} .

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- ▶ Σ is the alphabet,
- ▶ Q is the set of states,
- ▶ $\delta : Q \times \Sigma \rightarrow Q$ is the transition function.

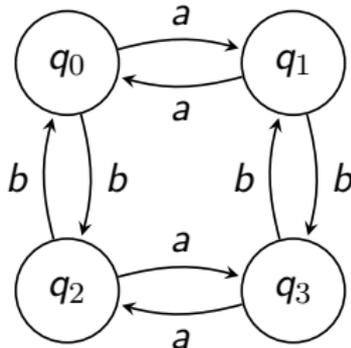


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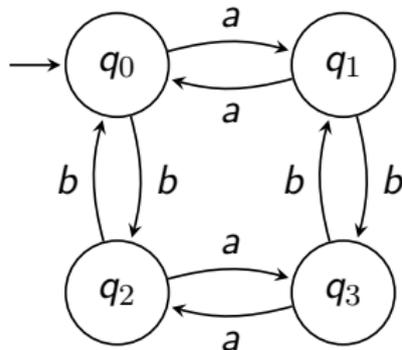


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- ▶ Σ is the alphabet,
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- ▶ $q_0 \in Q$ is the initial state,
- ▶ $F \subseteq Q$ is the set of accepting states, and

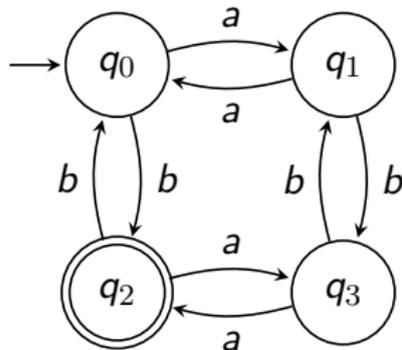


Figure 1: A DFA \mathcal{A} .

The **run** for the word $w = a_1 \dots a_n \in \Sigma^*$ ($n \in \mathbb{N}$) is the sequence of states

$$p_1 \xrightarrow[\mathcal{A}]{a_1} p_2 \xrightarrow[\mathcal{A}]{a_2} \dots \xrightarrow[\mathcal{A}]{a_n} p_{n+1}$$

such that $p_1 = q_0$ and $\forall i, \delta(p_i, a_i) = p_{i+1}$.

Example 1

Soit $w = ababb$. Its run is

$$q_0 \xrightarrow{a} q_1 \xrightarrow{b} q_3 \xrightarrow{a} q_2 \xrightarrow{b} q_0 \xrightarrow{b} q_2.$$

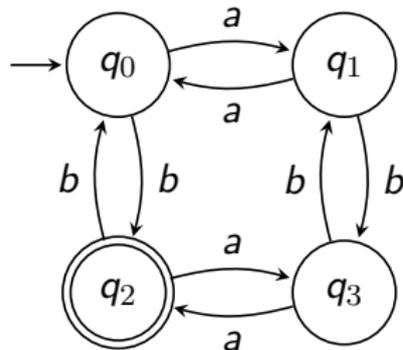


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such that $p_1 = q_0$ and $\forall i, \delta(p_i, a_i) = p_{i+1}$.
If $p_{n+1} \in F$, the run is said **accepting**.

Example 1

Soit $w = ababb$. Its run is

$$q_0 \xrightarrow{a} q_1 \xrightarrow{b} q_3 \xrightarrow{a} q_2 \xrightarrow{b} q_0 \xrightarrow{b} q_2,$$

and w is accepted by \mathcal{A} .

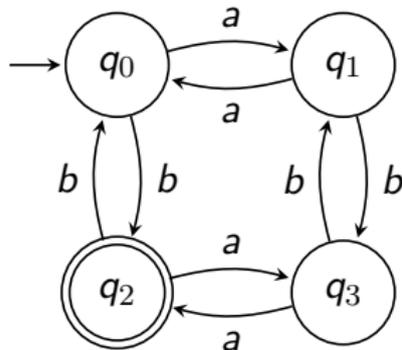


Figure 1: A DFA \mathcal{A} .

The language of \mathcal{A} is the set

$$\mathcal{L}(\mathcal{A}) = \{w \in \Sigma^* \mid \exists q \in F, q_0 \xrightarrow[\mathcal{A}]{w} q\}.$$

Example 2

The language of \mathcal{A} is

$$\mathcal{L}(\mathcal{A}) = \{w \mid w \text{ a an even number of } a \text{ and an odd number of } b\}.$$

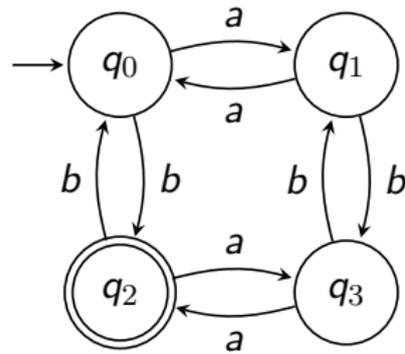


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Let $L \subseteq \Sigma^*$.

We want an algorithm to learn a DFA accepting L .

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queries
information



Figure 2: Angluin's framework Angluin, "Learning Regular Sets from Queries and Counterexamples", 1987

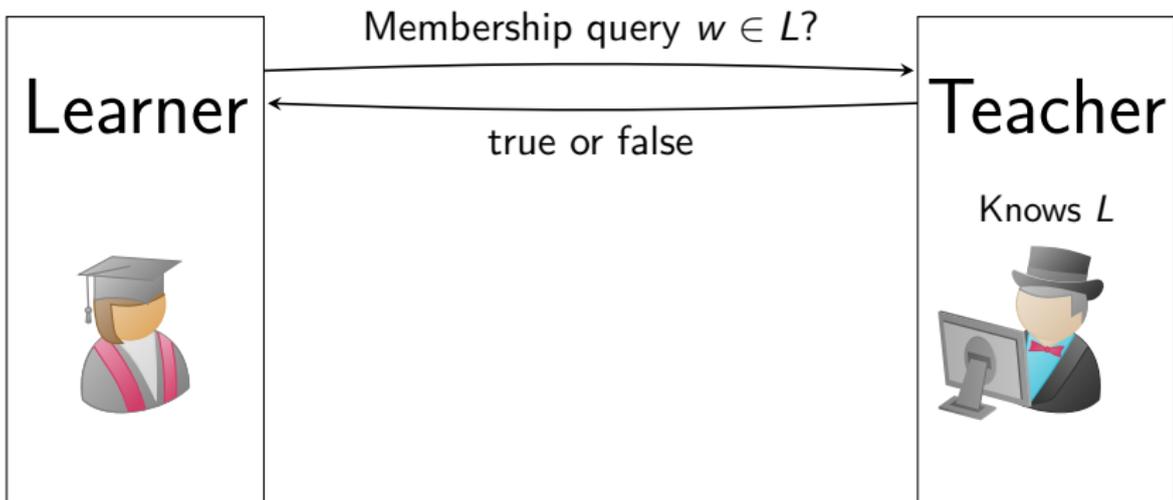


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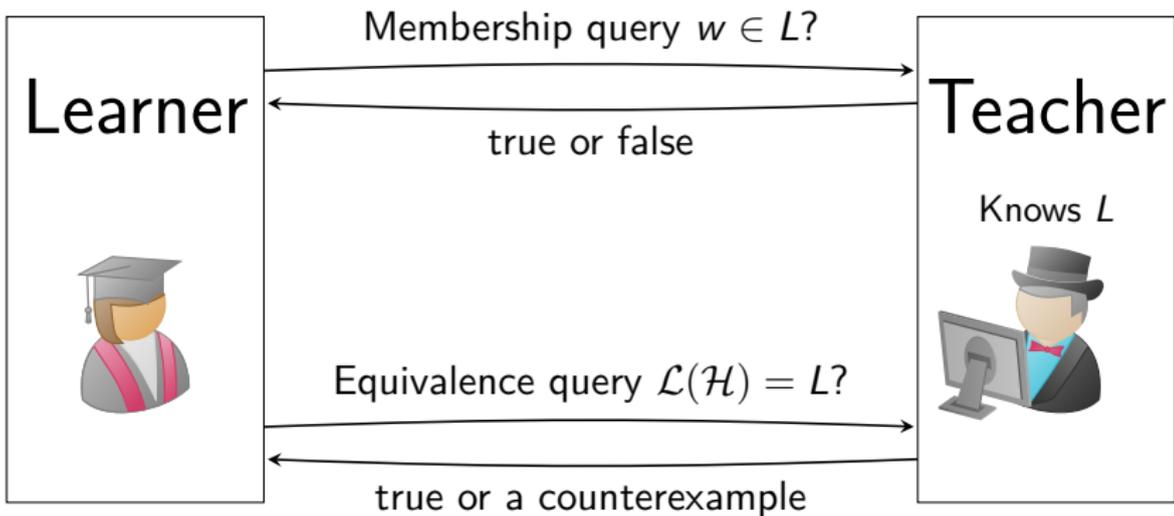


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Let $L = \{w \in \{a, b\}^* \mid$

w has an even number of a and an odd number of $b\}$.

Let $u \in \Sigma^*$. For all $w \in \Sigma^*$, we look if $uw \in L$.

We construct a table where the rows are indexed by the u and the columns by the w .

Let $L = \{w \in \{a, b\}^* \mid$
 $w \text{ has an even number of } a \text{ and an odd number of } b\}$.

	ε	a	b	aa	ab	ba	...
ε	0	0	1	0	0	0	...
a	0	0	0	0	1	1	...
b	1	0	0	1	0	0	...
aa	0	0	1	0	0	0	...
ab	0	1	0	0	0	0	...
ba	0	1	0	0	0	0	...
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\ddots

Let $L = \{w \in \{a, b\}^* \mid$
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	ε	a	b	aa	ab	ba	...
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b	1	0	0	1	0	0	...
aa	0	0	1	0	0	0	...
ab	0	1	0	0	0	0	...
ba	0	1	0	0	0	0	...
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\ddots

Let $u, v \in \Sigma^*$ and $L \subseteq \Sigma^*$. We say that $u \sim v$ if and only if^a

$$\forall w \in \Sigma^*, uw \in L \Leftrightarrow vw \in L.$$

^aHopcroft and Ullman, *Introduction to Automata Theory, Languages and Computation*, 2000.

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aa	0	0	1	0	0	0	...
ab	0	1	0	0	0	0	...
ba	0	1	0	0	0	0	...
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\ddots

Proposition 3

Let L be a language over Σ . Then, there is a DFA accepting L if and only if the index of \sim is finite.

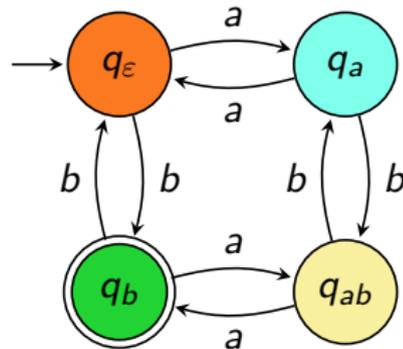
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b	1	0	0	1	0	0	...
aa	0	0	1	0	0	0	...
ab	0	1	0	0	0	0	...
ba	0	1	0	0	0	0	...
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮

The Myhill-Nerode congruence of this table has a finite index.

Let $L = \{w \in \{a, b\}^* \mid w \text{ has an even number of } a \text{ and an odd number of } b\}$.

	ε	a	b	aa	ab	ba	...
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a	0	0	0	0	1	1	...
b	1	0	0	1	0	0	...
aa	0	0	1	0	0	0	...
ab	0	1	0	0	0	0	...
ba	0	1	0	0	0	0	...
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮



The Myhill-Nerode congruence of this table has a finite index.

An **observation table**² is a tuple $\mathcal{O} = (R, S, \mathcal{L})$ where:

- ▶ $R \subseteq \Sigma^*$ is a **prefix-closed** set of **representatives** (the lines),
- ▶ $S \subseteq \Sigma^*$ is a **suffix-closed** set of **separators** (the columns),
- ▶ $\mathcal{L} : (R \cup R\Sigma)S \rightarrow \{1, 0\}$ is such that
 $\forall u \in R \cup R\Sigma, s \in S, \mathcal{L}(us) = 1 \Leftrightarrow us \in L.$

²Angluin, “Learning Regular Sets from Queries and Counterexamples”, 1987.

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Let $u, v \in R \cup R\Sigma$. We say that $u \sim_{\mathcal{O}} v$ if and only if

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The goal is to have a sufficient large **finite** subset of the infinite table from before.

More precisely, we refine $\sim_{\mathcal{O}}$ until it coincides with \sim .

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Let $L = \{w \in \{a, b\}^* \mid$
 $w \text{ has an even number of } a \text{ and an odd number of } b\}.$

	ε
ε	0
a	0
b	1

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	ε
ε	0
a	0
b	1

An observation table is **closed** if

$$\forall u \in R\Sigma, \exists v \in R, u \sim_{\theta} v.$$

If **unclosed** due to $u \in R\Sigma$, add u to R .
 Here, unclosed due to b .

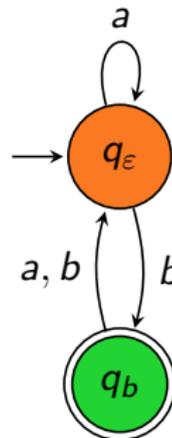
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	ε
ε	0
b	1
a	0
ba	0
bb	0

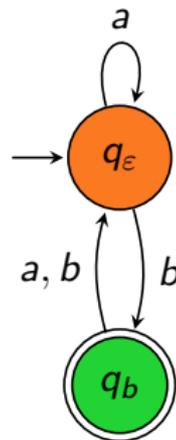
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	ε
ε	0
b	1
<hr/>	
a	0
ba	0
bb	0



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 $w \text{ has an even number of } a \text{ and an odd number of } b\}.$

	ε
ε	0
b	1
a	0
ba	0
bb	0



Counterexample: $ab.$

Let $L = \{w \in \{a, b\}^* \mid$

w has an even number of a and an odd number of $b\}$.

	ε
ε	0
b	1
a	0
ab	0
ba	0
bb	0
aa	0
aba	1
abb	0

Let $L = \{w \in \{a, b\}^* \mid$

w has an even number of a and an odd number of $b\}$.

	ε
ε	0
b	1
a	0
ab	0
ba	0
bb	0
aa	0
aba	1
abb	0

An observation table is Σ -consistent if

$$\forall u, v \in R, \forall a \in \Sigma, u \sim_{\theta} v \Rightarrow ua \sim_{\theta} va.$$

If Σ -inconsistent due to $u \sim_{\theta} v$ but $\mathcal{L}(uaw) \neq \mathcal{L}(vaw)$, add aw to S .

Here, $\varepsilon \sim_{\theta} ab$ but $\mathcal{L}(\varepsilon \cdot a \cdot \varepsilon) = 0 \neq \mathcal{L}(ab \cdot a \cdot \varepsilon) = 1$.

Let $L = \{w \in \{a, b\}^* \mid$

w has an even number of a and an odd number of $b\}$.

	ε	a
ε	0	0
b	1	0
a	0	0
ab	0	1
ba	0	1
bb	0	0
aa	0	0
aba	1	0
abb	0	0

Let $L = \{w \in \{a, b\}^* \mid$

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	ε	a
ε	0	0
b	1	0
a	0	0
ab	0	1
ba	0	1
bb	0	0
aa	0	0
aba	1	0
abb	0	0

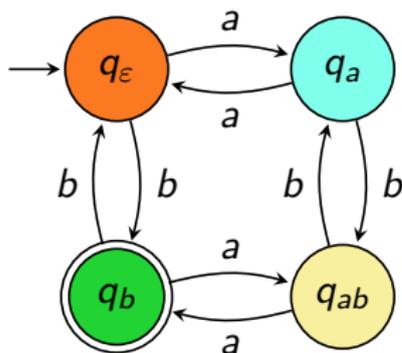
$\varepsilon \sim_{\emptyset} a$ but $\mathcal{L}(\varepsilon \cdot b \cdot \varepsilon) = 1 \neq \mathcal{L}(a \cdot b \varepsilon) = 0$.

Let $L = \{w \in \{a, b\}^* \mid$
 $w \text{ has an even number of } a \text{ and an odd number of } b\}.$

	ε	a	b
ε	0	0	1
b	1	0	0
a	0	0	0
ab	0	1	0
ba	0	1	0
bb	0	0	1
aa	0	0	1
aba	1	0	0
abb	0	0	0

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	ε	a	b
ε	0	0	1
b	1	0	0
a	0	0	0
ab	0	1	0
ba	0	1	0
bb	0	0	1
aa	0	0	1
aba	1	0	0
abb	0	0	0



Algorithm 1 Abstract learner for L^* [Angluin, “Learning Regular Sets from Queries and Counterexamples”, 1987]

Require: The target language L

Ensure: A DFA accepting L is returned

- 1: Initialize the observation table \mathcal{O}
 - 2: Fill \mathcal{O} with membership queries
 - 3: **while** true **do**
 - 4: Make \mathcal{O} closed and Σ -consistent
 - 5: Construct the DFA \mathcal{A}
 - 6: Ask an equivalence query over \mathcal{A}
 - 7: **if** the answer is positive **then**
 - 8: **return** \mathcal{A}
 - 9: **else**
 - 10: Given the counterexample w , add $Pref(w)$ to \mathcal{O}
 - 11: Fill \mathcal{O} with membership queries
-

1. Motivation
2. Learning deterministic finite automata
3. Learning realtime one-counter automata
 - Realtime one-counter automata
 - Behavior graph
 - Learning algorithm
4. Experimental results

A **realtime one-counter automaton** (ROCA) is a tuple $\mathcal{A} = (Q, \Sigma, \delta_{=0}, \delta_{>0}, q_0, F)$ where Q , q_0 , and F are defined as before, and the transition functions $\delta_{=0}$ and $\delta_{>0}$ are defined as:

$$\delta_{=0} : Q \times \Sigma \rightarrow Q \times \{0, +1\}$$

$$\delta_{>0} : Q \times \Sigma \rightarrow Q \times \{-1, 0, +1\}.$$

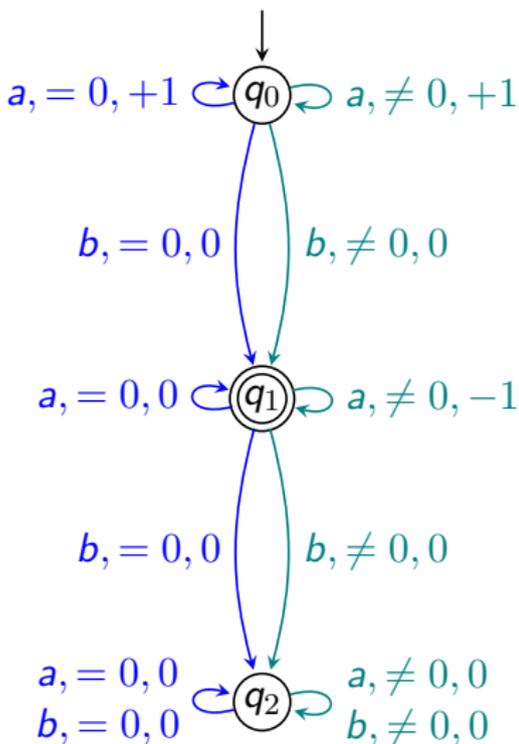


Figure 3: An ROCA \mathcal{A} .

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$$\delta_{=0} : Q \times \Sigma \rightarrow Q \times \{0, +1\}$$

$$\delta_{>0} : Q \times \Sigma \rightarrow Q \times \{-1, 0, +1\}.$$

A **configuration** is a pair $(q, n) \in Q \times \mathbb{N}$.

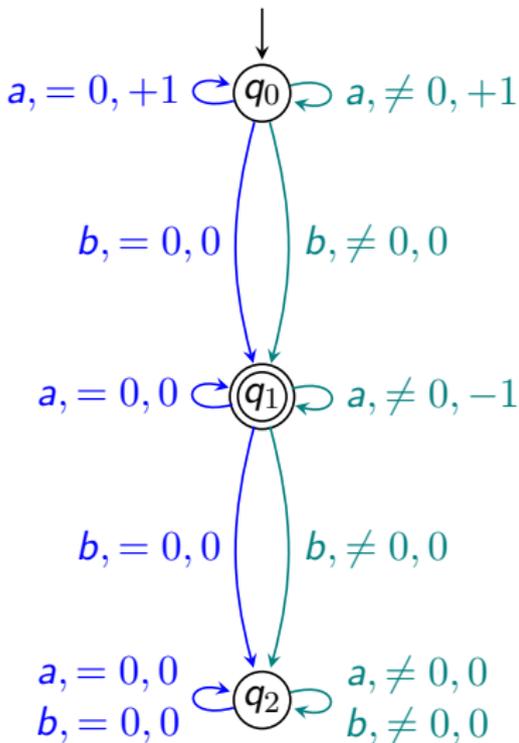


Figure 3: An ROCA \mathcal{A} .

The **transition relation**

$$\xrightarrow{\mathcal{A}} \subseteq (Q \times \mathbb{N}) \times \Sigma \times (Q \times \mathbb{N})$$

contains $(q, n) \xrightarrow{\mathcal{A}}^a (p, m)$ iff

$$\begin{cases} \delta_{=0}(q, a) = (p, c) \wedge m = n + c & \text{if } n = 0 \\ \delta_{>0}(q, a) = (p, c) \wedge m = n + c & \text{if } n > 0. \end{cases}$$

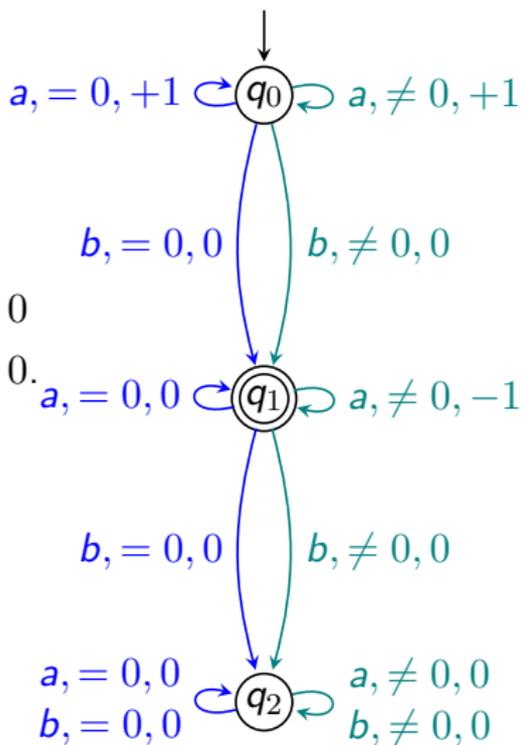


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Example 4

$$\begin{aligned} (q_0, 0) &\xrightarrow{\mathcal{A}} (q_0, 1) \xrightarrow{\mathcal{A}} (q_0, 2) \\ &\xrightarrow{\mathcal{A}} (q_1, 2) \xrightarrow{\mathcal{A}} (q_1, 1) \\ &\xrightarrow{\mathcal{A}} (q_1, 0) \xrightarrow{\mathcal{A}} (q_1, 0). \end{aligned}$$

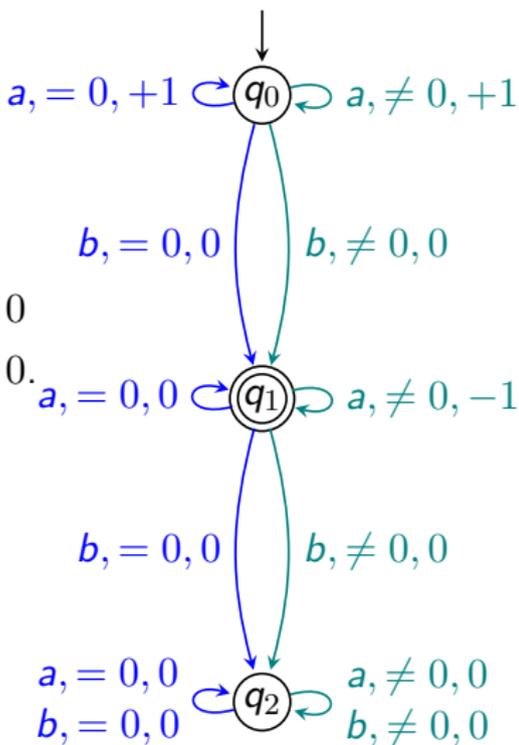


Figure 4: An ROCA \mathcal{A} .

Let $w \in \Sigma^*$. The counter value of w , according to \mathcal{A} , is n iff

$$\exists q \in Q, (q_0, 0) \xrightarrow{\mathcal{A}}^w (q, n).$$

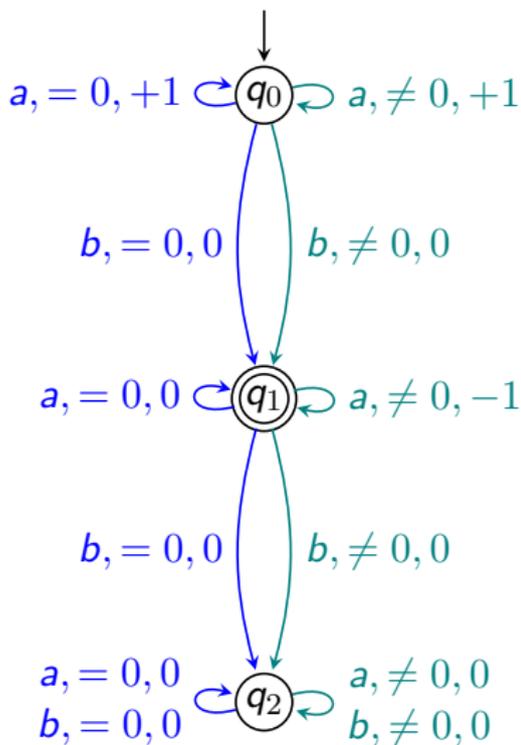


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Example 5

Since $(q_0, 0) \xrightarrow{\mathcal{A}}^{aabaaa} (q_1, 0)$,
 $c_{\mathcal{A}}(aabaaa) = 0$.

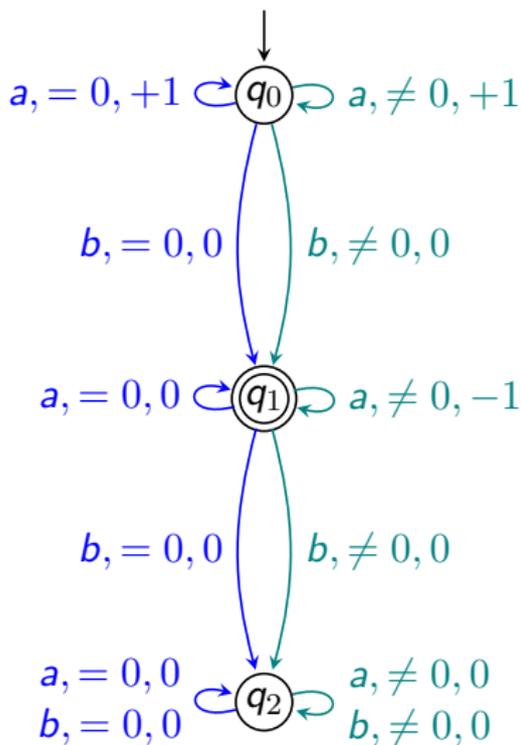


Figure 5: An ROCA \mathcal{A} .

For a word w , if we have

$$(q_0, 0) \xrightarrow[\mathcal{A}]{w} (q, 0)$$

with $q \in F$, then $w \in \mathcal{L}(\mathcal{A})$.

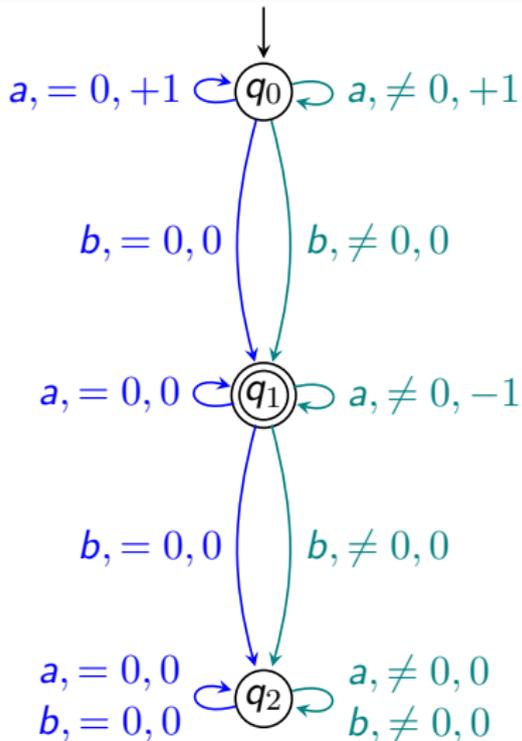


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Example 6

$$\mathcal{L}(\mathcal{A}) = \{a^n b a^m \mid 0 \leq n \leq m\}.$$

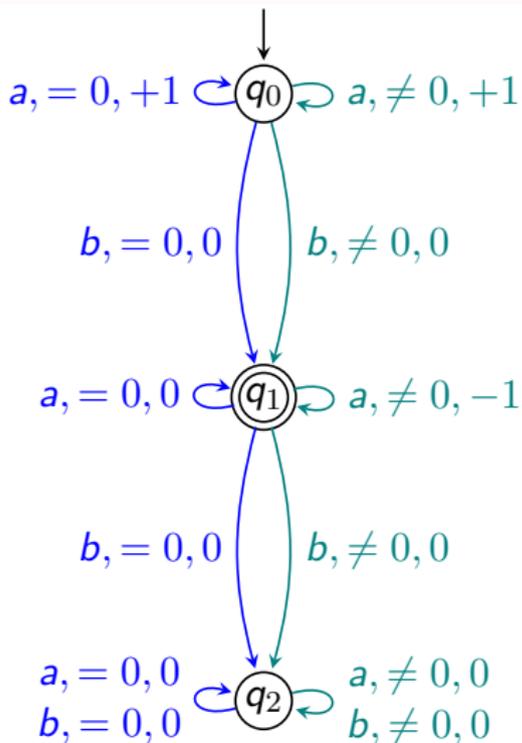


Figure 6: An ROCA \mathcal{A} .

Let \mathcal{A} be an ROCA accepting L .
Let $u, v \in \Sigma^*$. We say that $u \equiv v$ iff

- $\forall w \in \Sigma^*, uw \in L \Leftrightarrow vw \in L$,
and

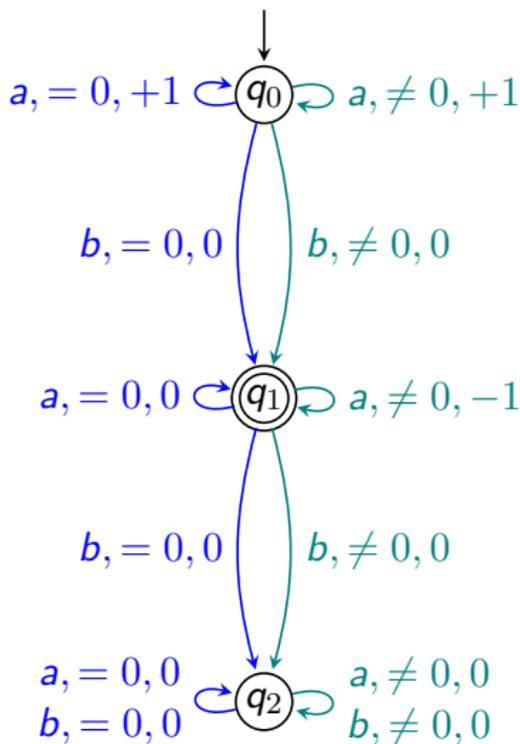


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- $\forall w \in \Sigma^*, uw \in L \Leftrightarrow vw \in L$,
and
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 $c_{\mathcal{A}}(uw) = c_{\mathcal{A}}(vw)$.

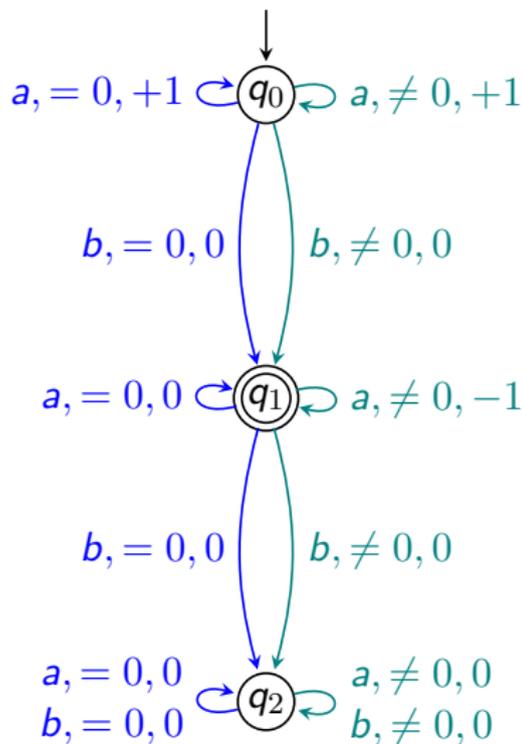


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Example 7

$b \equiv aba$ but $ab \not\equiv aab$.

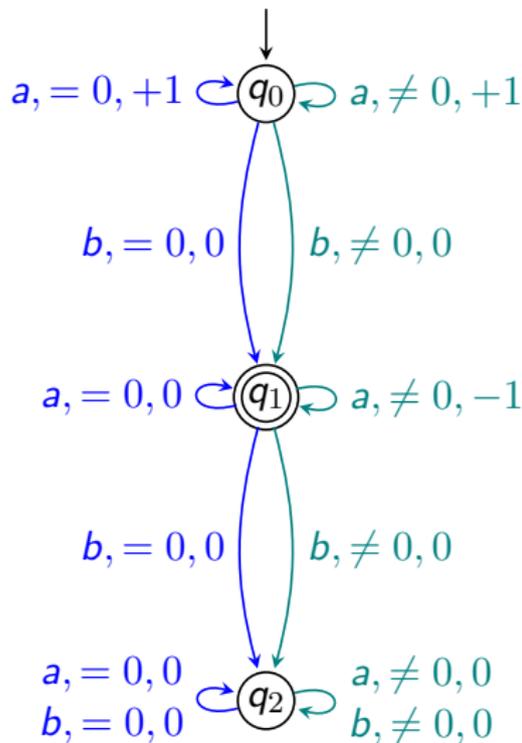
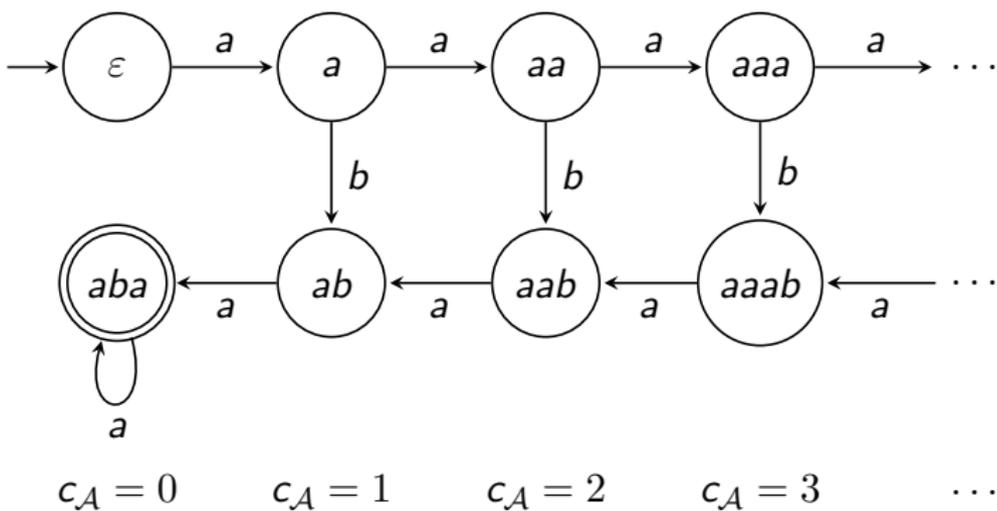
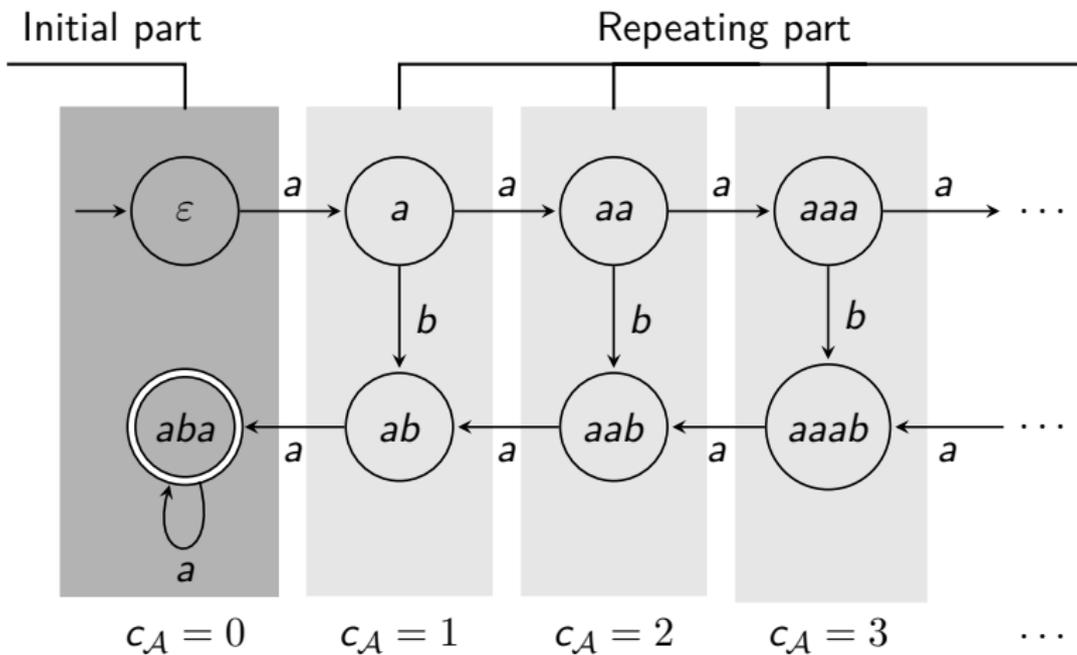


Figure 7: An ROCA \mathcal{A} .

Let \mathcal{A} be an ROCA accepting L . Using the relation \equiv , we can construct an **infinite** deterministic automaton accepting L : the **behavior graph of \mathcal{A}** $BG(\mathcal{A}) = (Q_{\equiv}, \Sigma, \delta_{\equiv}, q_{\equiv}^0, F_{\equiv})$ with:

- ▶ $Q_{\equiv} = \{ \llbracket u \rrbracket_{\equiv} \mid u \in Pref(L) \},$
- ▶ $q_{\equiv}^0 = \llbracket \varepsilon \rrbracket_{\equiv},$
- ▶ $F_{\equiv} = \{ \llbracket u \rrbracket_{\equiv} \mid u \in L \},$ and
- ▶ $\delta_{\equiv} : Q \times \Sigma \rightarrow Q$ such that $\delta(\llbracket u \rrbracket_{\equiv}, a) = \llbracket ua \rrbracket_{\equiv}$ with $a \in \Sigma$ and $u, ua \in Pref(L)$.

Figure 8: The behavior graph of \mathcal{A} .

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Theorem 8

Let \mathcal{A} be an ROCA accepting L and $BG(\mathcal{A})$ be its behavior graph. Then, $BG(\mathcal{A})$ is *ultimately periodic*.

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Moreover, it is possible to construct an ROCA accepting L from $BG(\mathcal{A})$.

Let \mathcal{A} be an ROCA accepting L .

³Based on the algorithm for VCA [Neider and Löding, *Learning visibly one-counter automata in polynomial time*, 2010].

Let \mathcal{A} be an ROCA accepting L .

- ▶ Rough idea³: learn a sufficiently large initial fragment of $BG(\mathcal{A})$ and construct an ROCA from it.

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Let \mathcal{A} be an ROCA accepting L .

- ▶ Rough idea³: learn a sufficiently large initial fragment of $BG(\mathcal{A})$ and construct an ROCA from it.
- ▶ What is an initial fragment?
 $\hookrightarrow BG_\ell(\mathcal{A})$ is a subgraph of $BG(\mathcal{A})$ whose set of states is $\{\llbracket u \rrbracket_\equiv \in Q_\equiv \mid \forall x \in Pref(u), 0 \leq c_{\mathcal{A}}(x) \leq \ell\}$, with $\ell \in \mathbb{N}$. Let $L_\ell = \mathcal{L}(BG_\ell(\mathcal{A}))$.

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- ▶ How to learn $BG_\ell(\mathcal{A})$?
 $\hookrightarrow BG_\ell(\mathcal{A})$ is actually a DFA.

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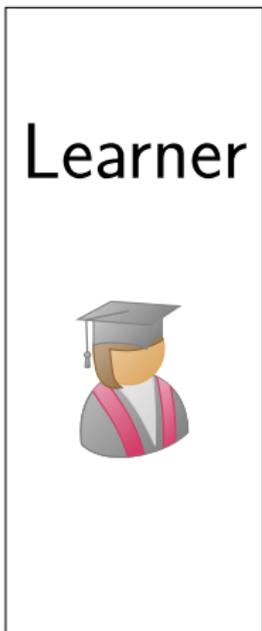


Figure 9: Adaptation of Angluin's framework for ROCA.

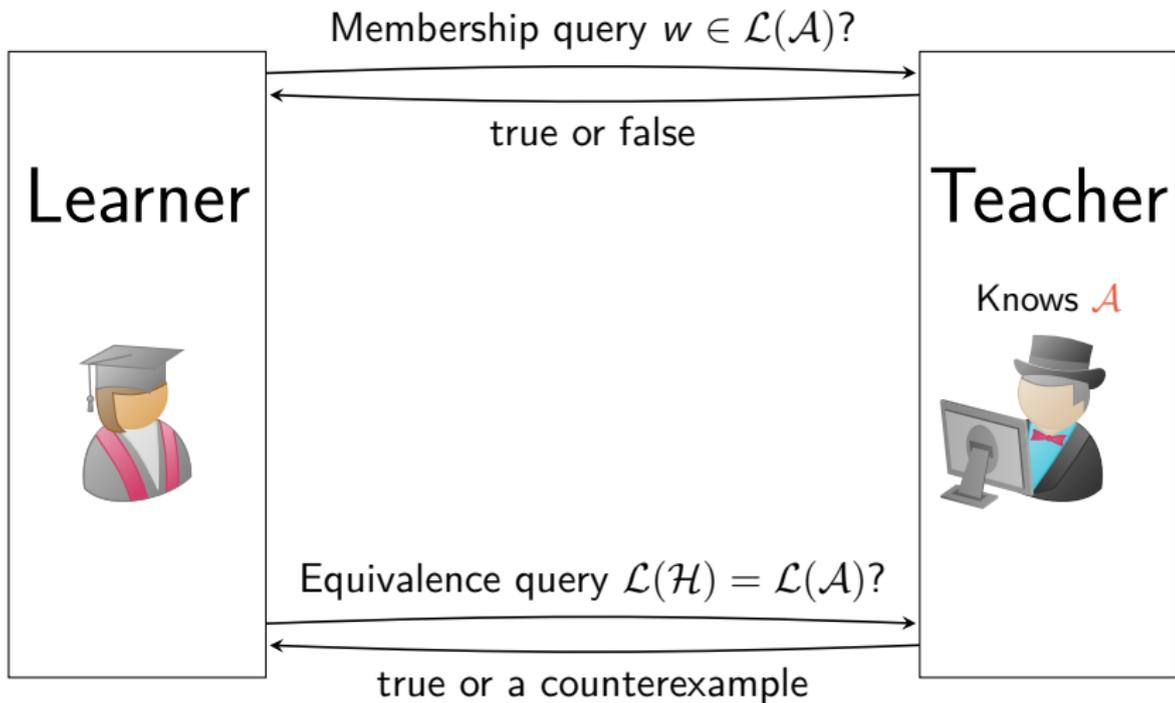


Figure 9: Adaptation of Angluin's framework for ROCA's.

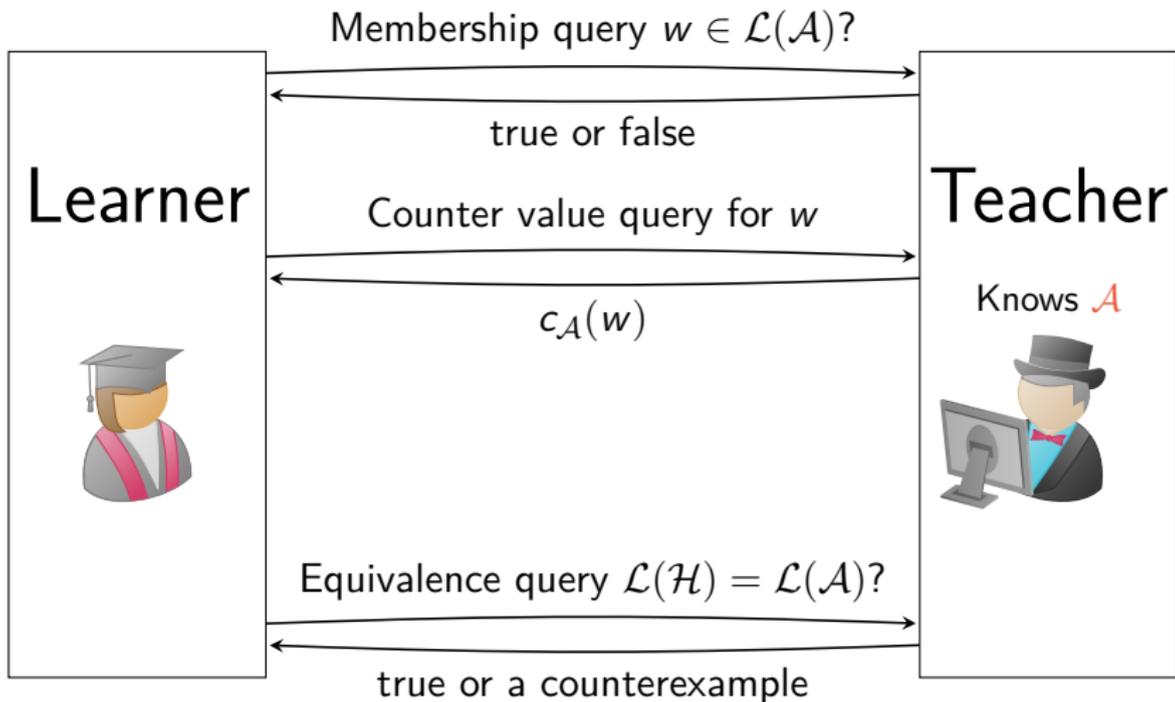


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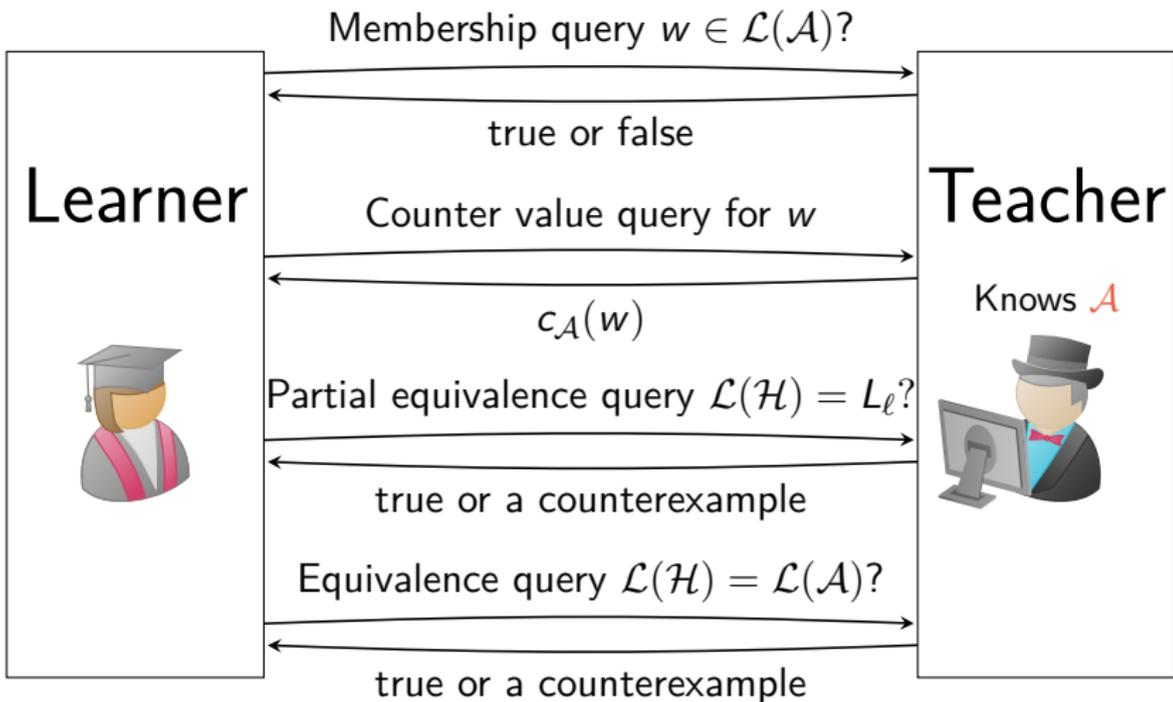


Figure 9: Adaptation of Angluin's framework for ROCAs.

Algorithm 2 Adaptation of L^* for ROCAs.

Require: A teacher knowing an ROCA \mathcal{A}

Ensure: An ROCA accepting the same language is returned

- 1: Initialize the data structure \mathcal{D}_ℓ up to $\ell = 0$
 - 2: **while** true **do**
 - 3: Make \mathcal{D}_ℓ respect the needed constraints and construct $\mathcal{A}_{\mathcal{D}_\ell}$
 - 4: Ask a **partial equivalence query** over $\mathcal{A}_{\mathcal{D}_\ell}$
 - 5: **if** the answer is negative **then**
 - 6: Update \mathcal{D}_ℓ with the provided counterexample $\triangleright \ell$ is not modified
 - 7: **else**
 - 8: Construct all the possible ROCAs $\mathcal{A}_1, \dots, \mathcal{A}_n$ from $\mathcal{A}_{\mathcal{D}_\ell}$
 - 9: Ask an **equivalence query** over each \mathcal{A}_i
 - 10: **if** the answer is true for an \mathcal{A}_i **then return** \mathcal{A}_i
 - 11: **else** Select one counterexample and update \mathcal{D}_ℓ $\triangleright \ell$ is increased
-

Let \mathcal{A} be an ROCA accepting $L \subseteq \Sigma^*$.

An **observation table up to ℓ** is a tuple $\mathcal{O}_\ell = (R, S, \widehat{S}, \mathcal{L}_\ell, \mathcal{C}_\ell)$ with:

- ▶ $R \subseteq \Sigma^*$ is the **prefix-closed** set of **representatives**,
- ▶ $S \subseteq \widehat{S} \subseteq \Sigma^*$ are two **suffix-closed** sets of **separators**,
- ▶ $\mathcal{L}_\ell : (R \cup R\Sigma)\widehat{S} \rightarrow \{0, 1\}$, and
- ▶ $\mathcal{C}_\ell : (R \cup R\Sigma)S \rightarrow \{0, \dots, \ell\} \cup \{\perp\}$.

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Let $Pref(\mathcal{O}_\ell) = \{w \in Pref(us) \mid u \in R \cup R\Sigma, s \in \widehat{S}, \mathcal{L}_\ell(us) = 1\}$.

The following holds for all $u \in R \cup R\Sigma$:

- ▶ $\forall s \in \widehat{S}, \mathcal{L}_\ell(us) = 1$ if and only if $us \in L_\ell$.
- ▶ $\forall s \in S, \mathcal{C}_\ell(us) = \begin{cases} c_{\mathcal{A}}(us) & \text{if } us \in Pref(\mathcal{O}_\ell) \\ \perp & \text{otherwise.} \end{cases}$

Let $L = \{a^n b a^m \mid 0 \leq n \leq m\}$.

	ε
ε	0, 0
a	0, 1
ab	0, 1
aba	1, 0
b	0, \perp
aa	0, \perp
abb	0, \perp
$abaa$	1, 0
$abab$	0, \perp

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$abbb$	0, \perp

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ab	0, 1
aba	1, 0
abb	0, 1
$abbb$	0, 1
b	0, \perp
aa	0, \perp
$abaa$	1, 0
$abab$	0, \perp
$abba$	1, 0
$abbba$	1, 0
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$abba$	1, 0
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↔ Getting the algorithm to eventually finish is harder than it looks.

Rough idea:

- ▶ u and v are **approximately equivalent** if they are equal, up to \perp .

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 - ▶ If $u \equiv v$, then u and v are approximately equivalent.
 - ▶ Not a right-congruence.
- ▶ Adapt definitions of closedness and consistency to force right-congruence.

Theorem 9

Let \mathcal{A} be an ROCA accepting a language $L \subseteq \Sigma^$. Given a teacher for L with an automaton \mathcal{A} , and t the length of the longest counterexample for (partial) equivalence queries:*

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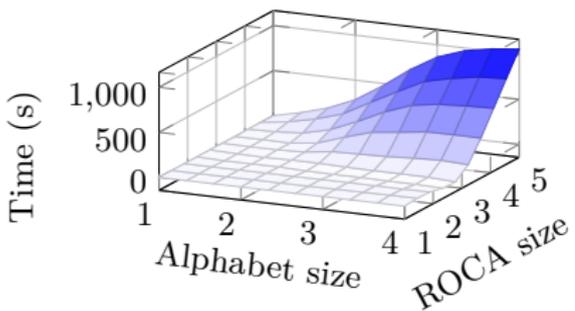
- ▶ *An ROCA accepting L can be computed in time and space exponential in $|\mathcal{A}|$, $|\Sigma|$ and t .*

Theorem 9

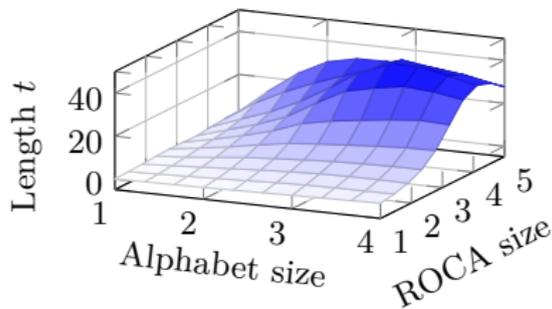
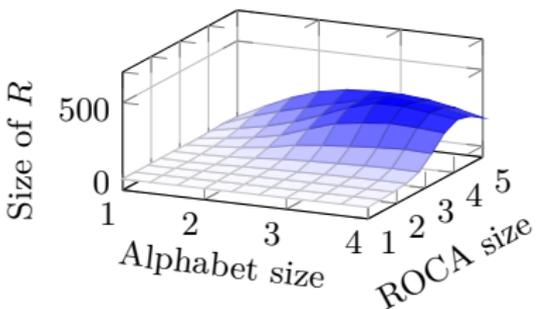
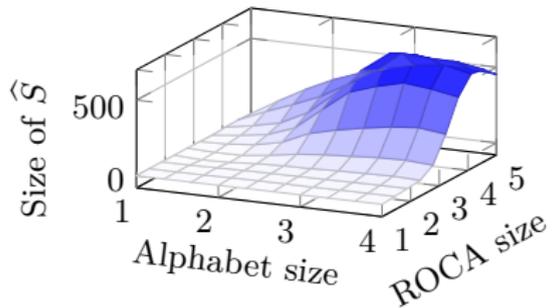
Let \mathcal{A} be an ROCA accepting a language $L \subseteq \Sigma^*$. Given a teacher for L with an automaton \mathcal{A} , and t the length of the longest counterexample for (partial) equivalence queries:

- ▶ An ROCA accepting L can be computed in time and space exponential in $|\mathcal{A}|, |\Sigma|$ and t .
- ▶ The learner asks:
 - ▶ $\mathcal{O}(t^3)$ partial equivalence queries
 - ▶ $\mathcal{O}(|\mathcal{A}|t^2)$ equivalence queries
 - ▶ An exponential number of membership (resp. counter value) queries in $|\mathcal{A}|, |\Sigma|$, and t .

1. Motivation
2. Learning deterministic finite automata
3. Learning realtime one-counter automata
4. Experimental results



(a) Total time.

(b) Length t of the longest counterexample.(c) Final size of R .(d) Final size of \hat{S} .

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