

Learning Real-Time One-Counter Automata Using Polynomially Many Queries

<https://arxiv.org/abs/2411.08815>

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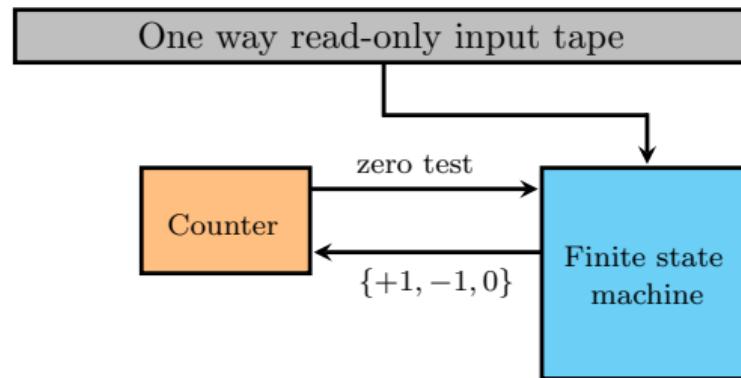
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Joint work with: Dr. A.V. Sreejith, Indian Institute of Technology Goa and
Dr. Vincent Penelle, University of Bordeaux

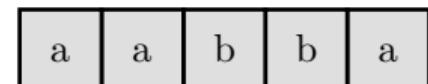
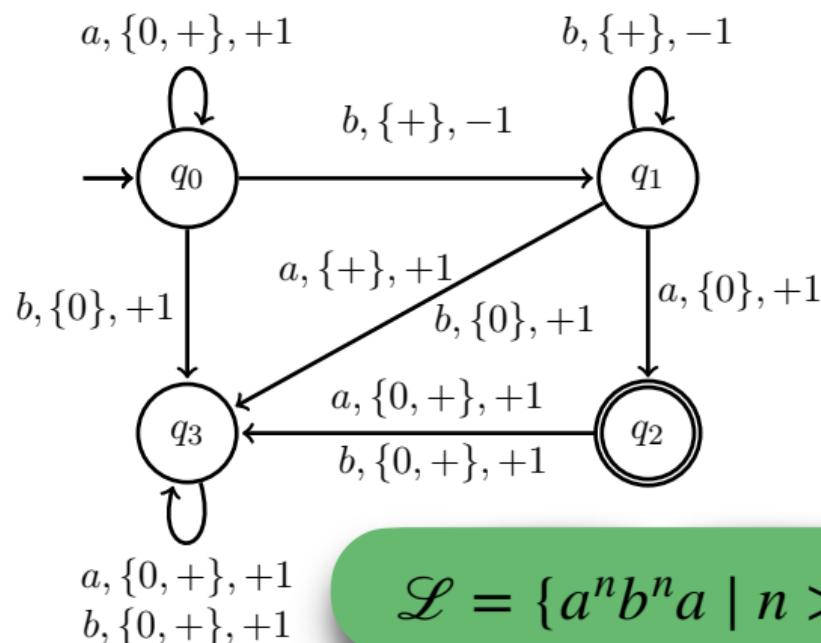
RHPL 2024

Deterministic real-time one-counter automata (DROCA)

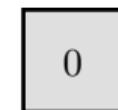


- Write to counter: Increment (+1), No change (0), Decrement (-1).
- Read from counter: zero (0) or positive (+).
- Counter-value is always non-negative.
- Transitions of the finite-state machine are deterministic.
- There are no ϵ - transitions.

Example: Deterministic real-time one-counter automata

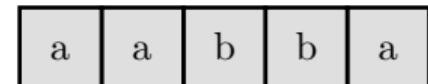
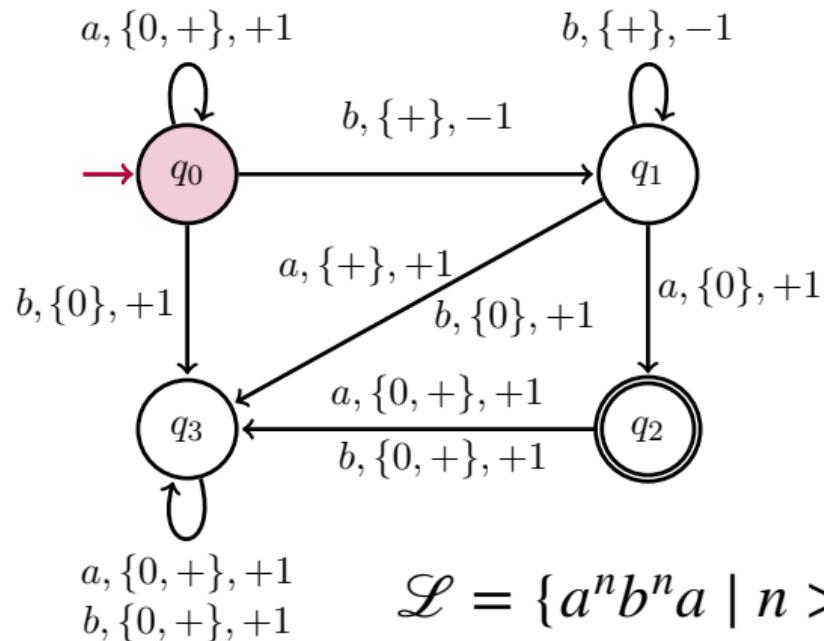


Input tape



Counter

Example: Deterministic real-time one-counter automata



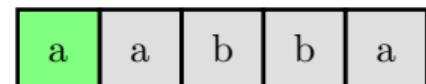
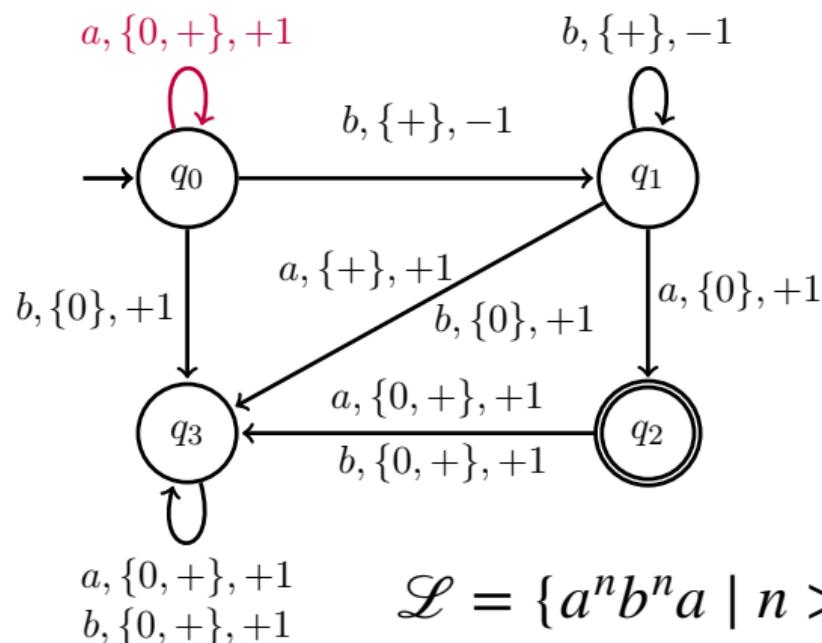
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Counter

$$\mathcal{L} = \{a^n b^n a \mid n > 0\}$$

Example: Deterministic real-time one-counter automata

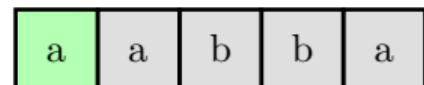
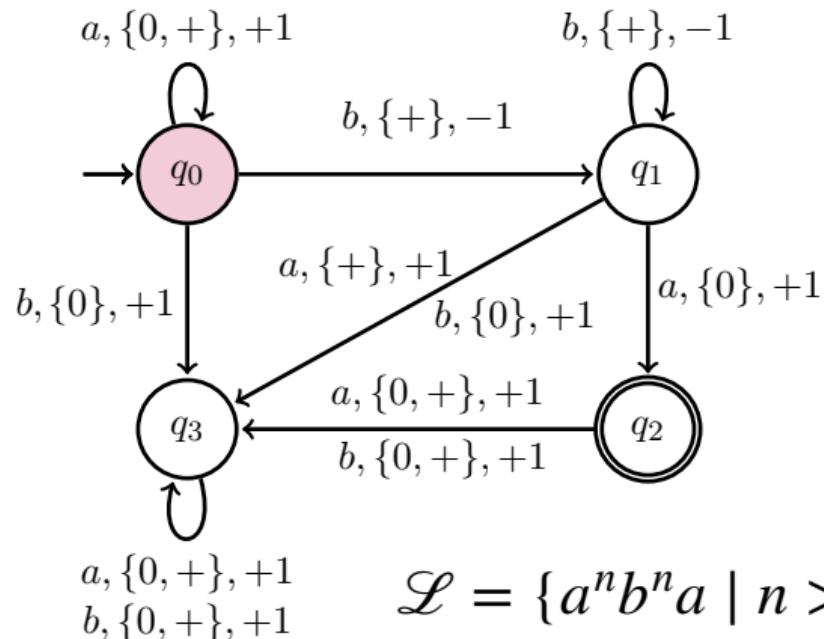


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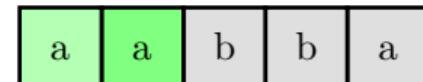
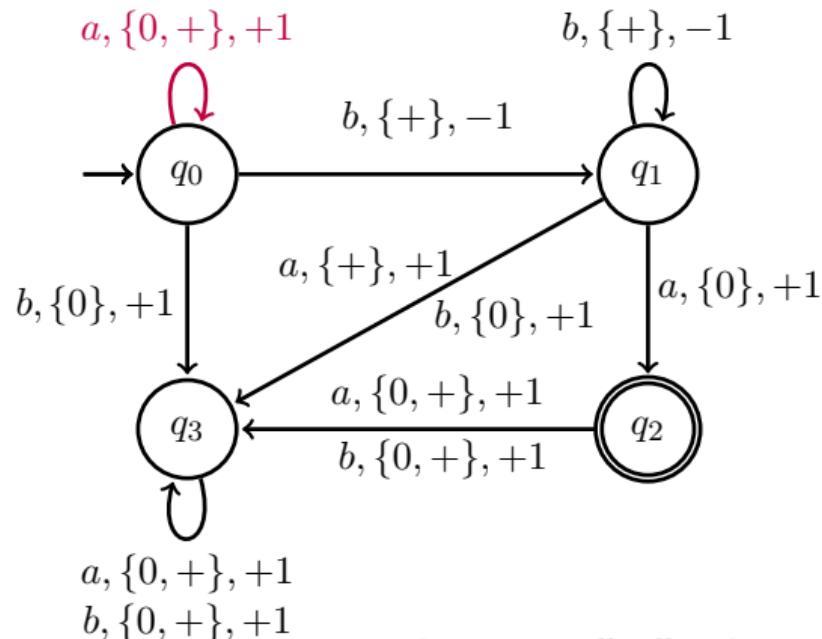


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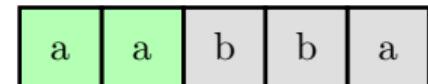
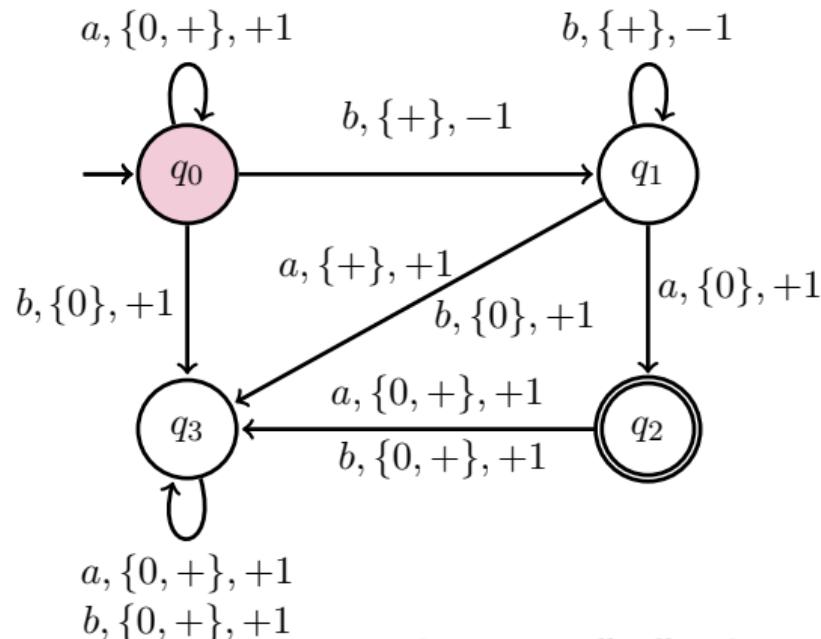
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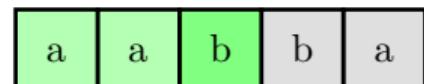
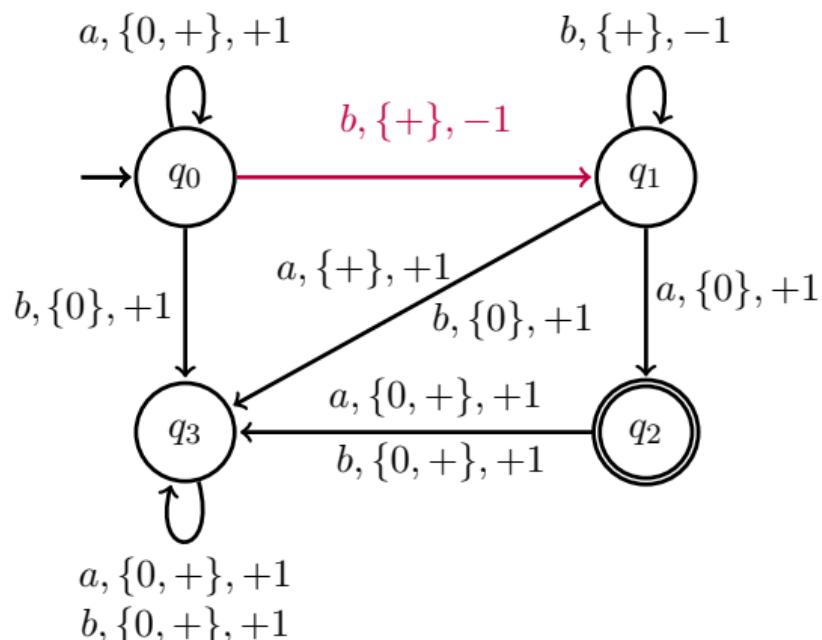
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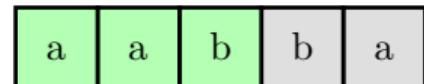
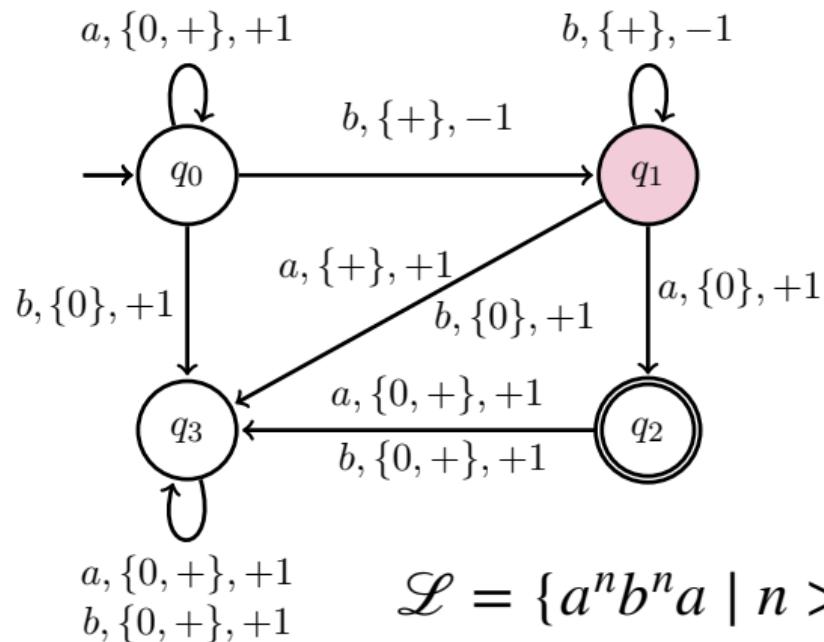
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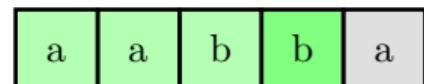
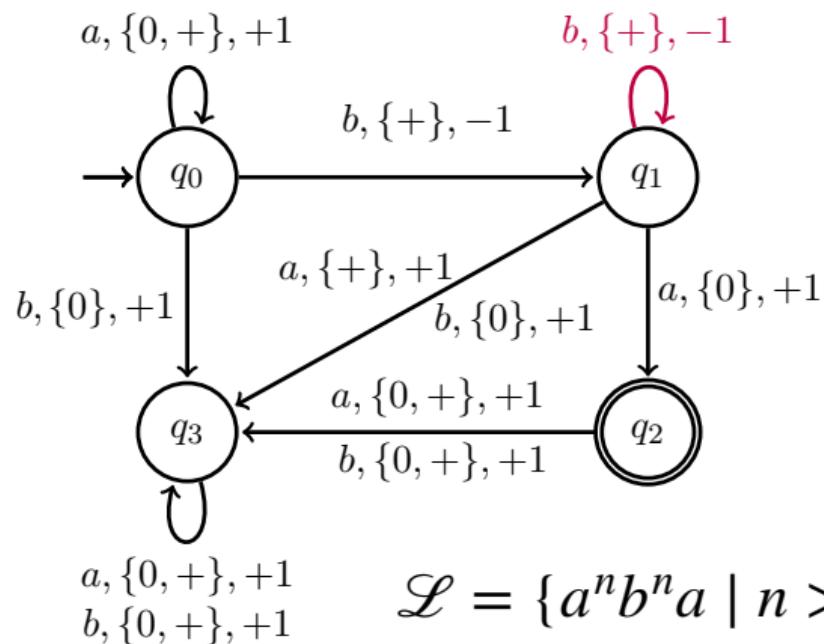


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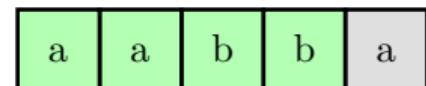
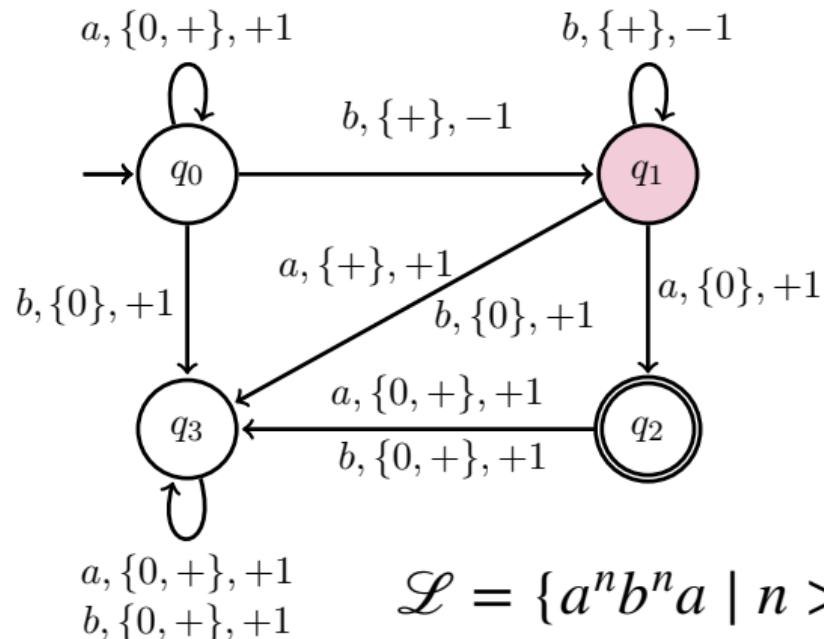


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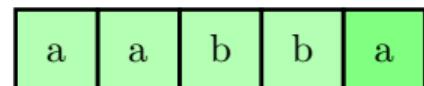
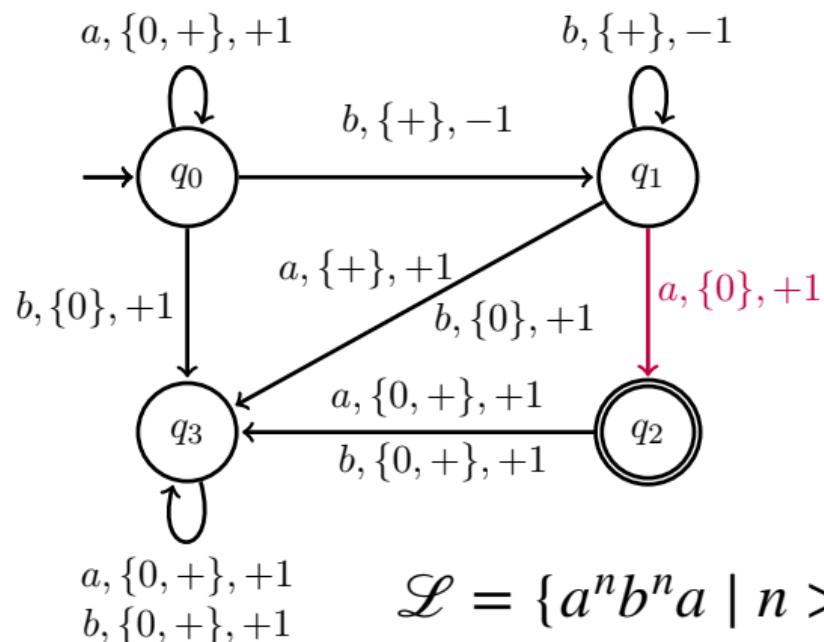


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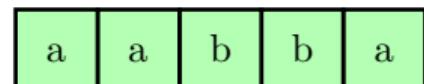
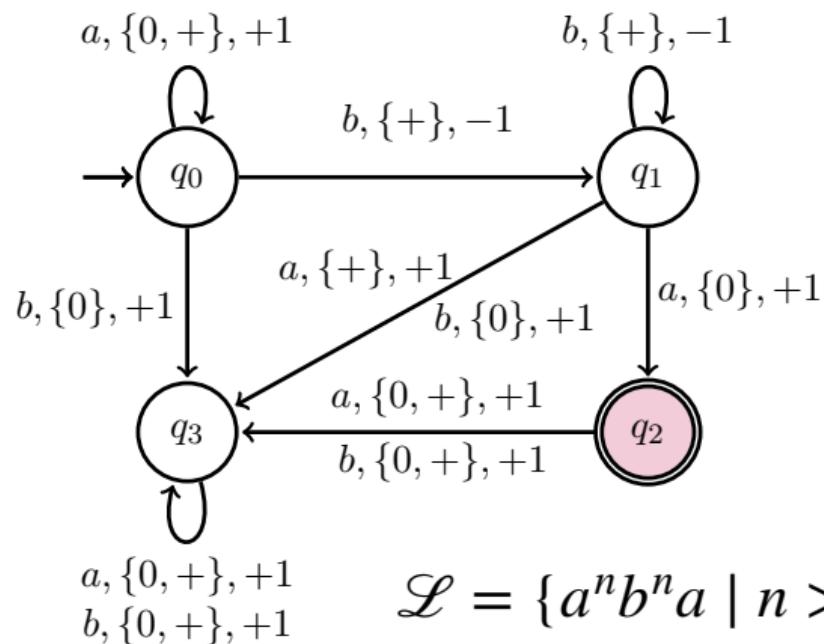


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Counter

Equivalence - A practical algorithm

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Given two DROCAs of size n , there is an algorithm that runs in $O(n^5)$ time to check whether they are counter-synchronous.

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- For the class of visibly OCAs there is a $O(n^3)$ time algorithm for checking equivalence.

Learning DROCAs

Our work on learning DROCA

- Existing results for learning DROCA need exponential time & number of queries.
- We have the following result:
 - Learning DROCA can be done using **polynomially** many queries using a SAT solver.
- Implemented a learning algorithm for DROCA in Python.
- Our evaluations show that the proposed method significantly outperforms the existing technique.
- We use the idea of finding a minimal separating DFA from a given set of positive & negative samples.

Minimal separating DFA

- Given two sets POS and NEG.
- Construct a DFA \mathcal{A} with the minimal number of states such that
 - \mathcal{A} accepts all words in POS, and
 - \mathcal{A} rejects all words in NEG.

Minimal separating DFA

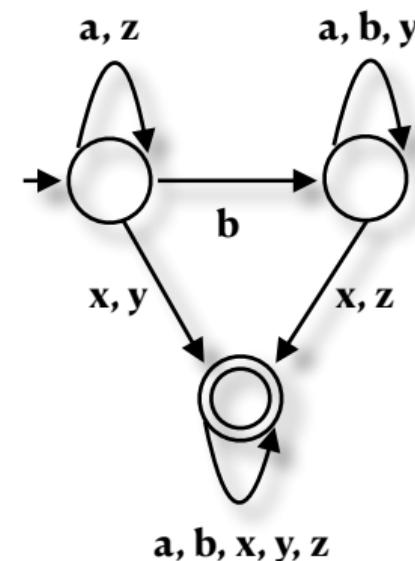
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Our learning framework

Membership queries:

Learner: “Is w in $\mathcal{L}(\mathcal{A})$?”

Teacher : “**yes**” or “**no**”

Counter-value queries:

Learner: “What is the counter-value reached on reading w ?”

Teacher: “**counter-value reached on w**”

Minimal-equivalence queries:

Learner: “Does \mathcal{B} and \mathcal{A} recognise the same language?”

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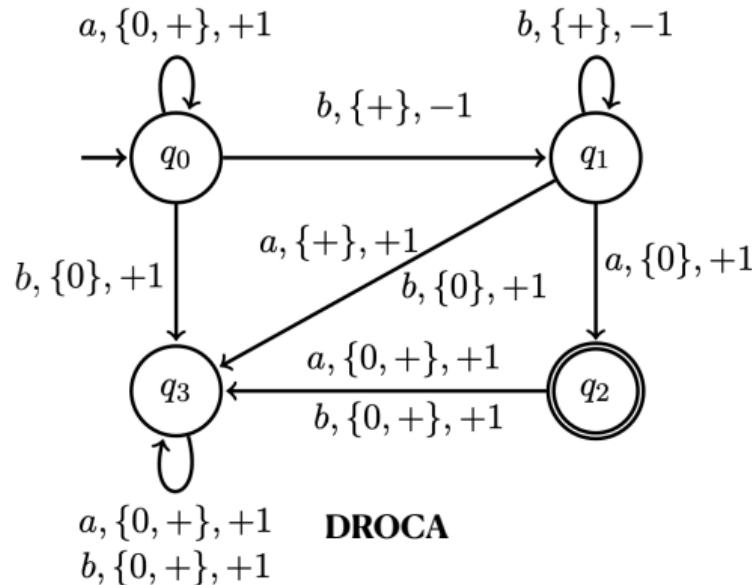
Learner: “Does \mathcal{B} and \mathcal{A} recognise the same language?”

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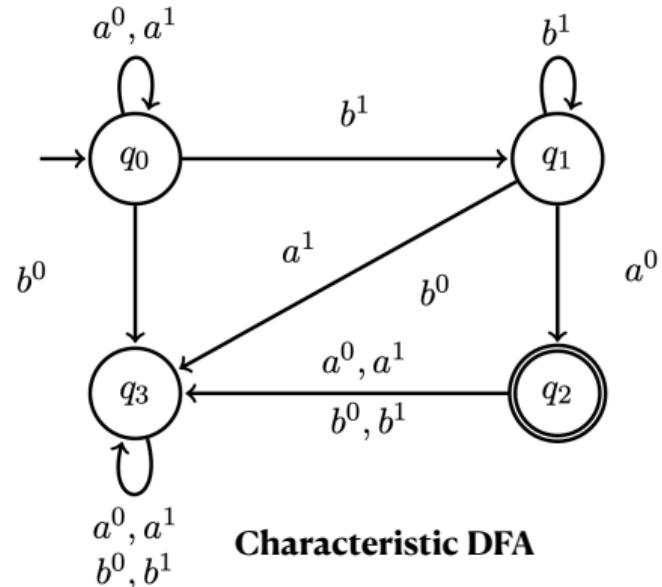
Membership different

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Characteristic DFA

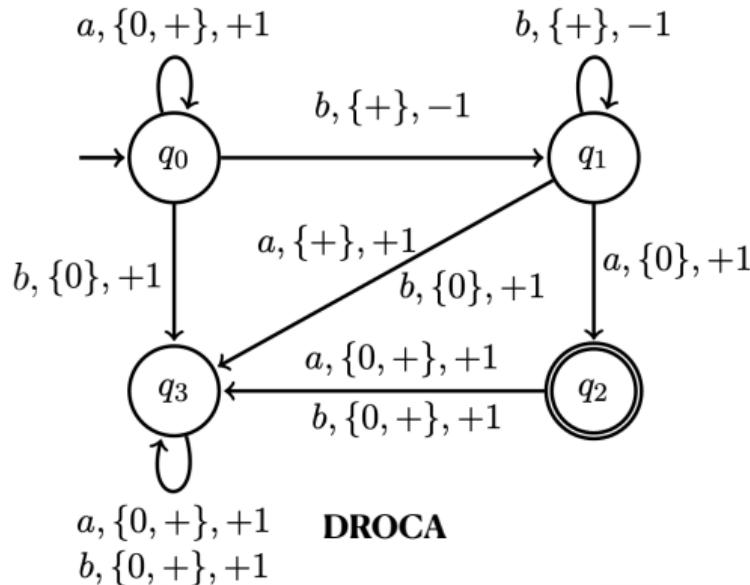


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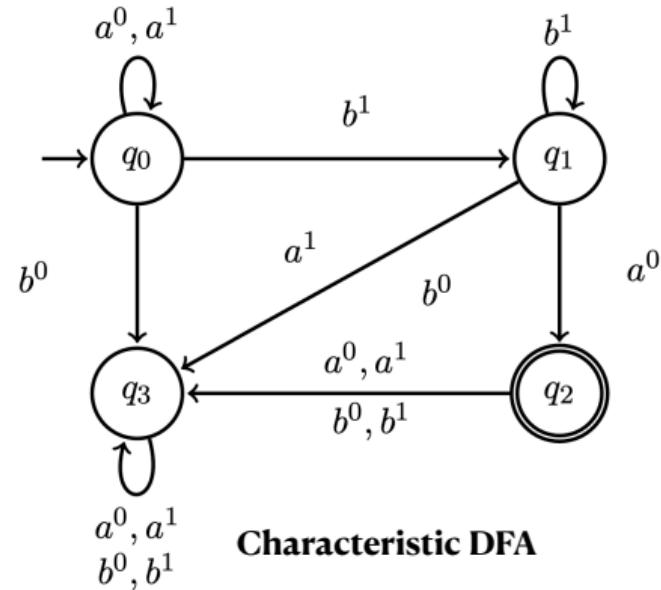


- The primary idea is to try and learn the characteristic DFA.

Characteristic DFA



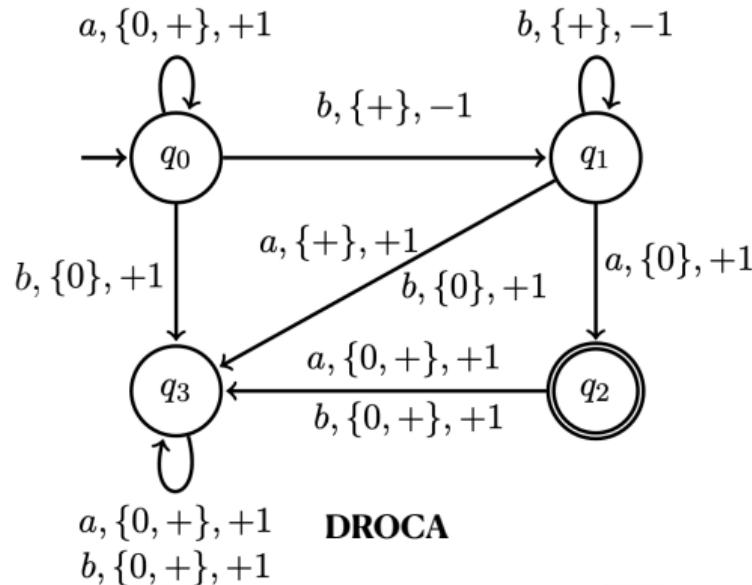
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$$\{a, b\}^* \rightarrow \{a^0, a^1, b^0, b^1\}^*$$

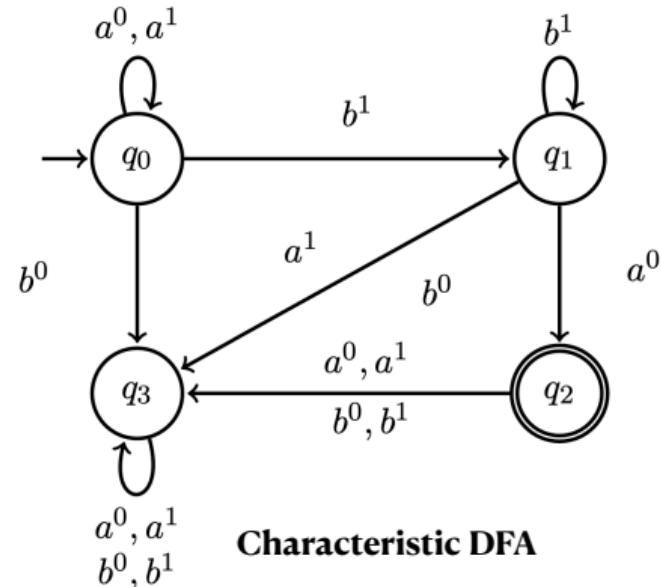
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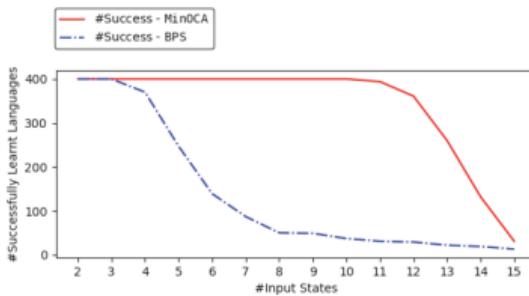


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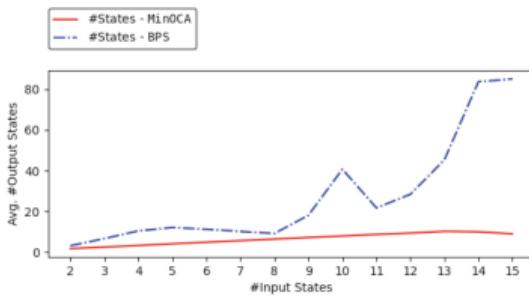
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Experimental Results

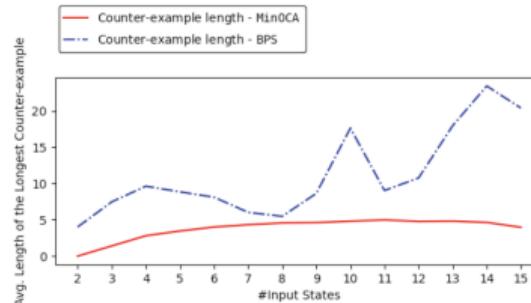
Comparison with existing method (Bruyère et al., 2022)



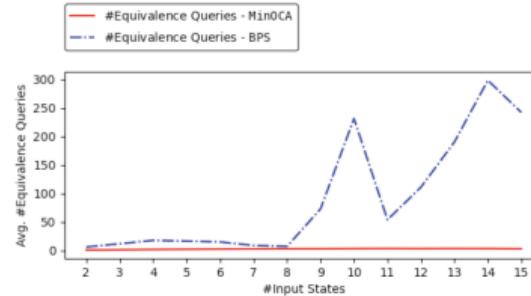
Number of successfully learnt languages (out of 400)



Average number of states in the learnt DROCA

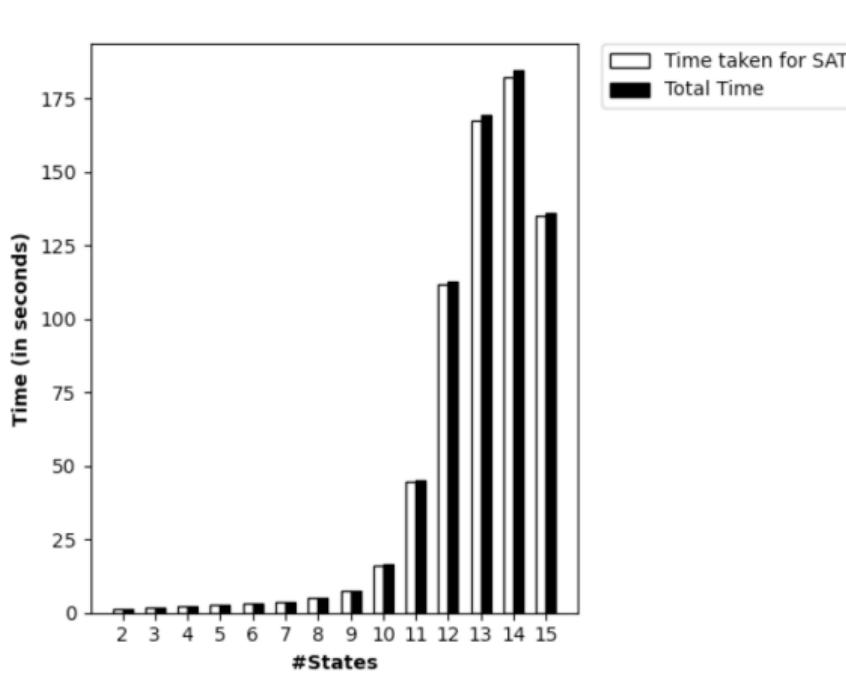


Average length of the longest counter-example

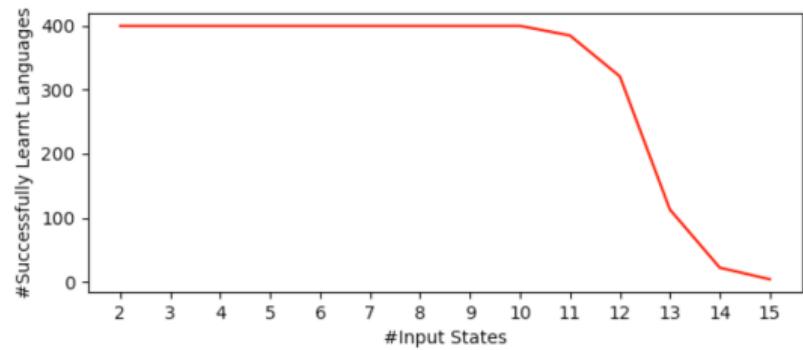


Average number of equivalence queries used for learning

Experimental results

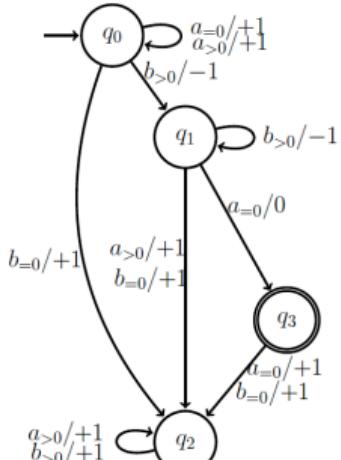


Total time vs Time taken for SAT by minOCA

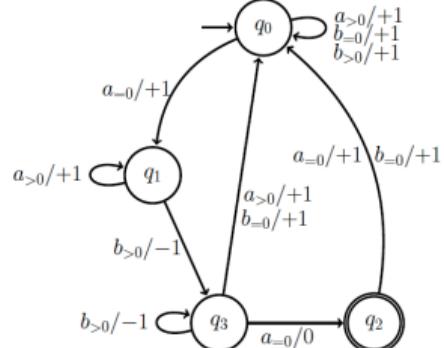


Number of successfully learnt DROCAs by minOCA with a timeout of 5 minutes

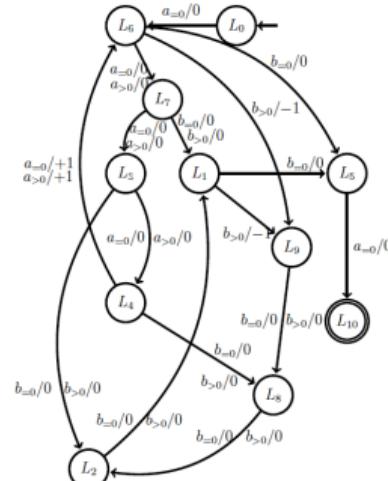
Example-1: DROCAs learnt by minOCA and BPS



(a) Input DROCA



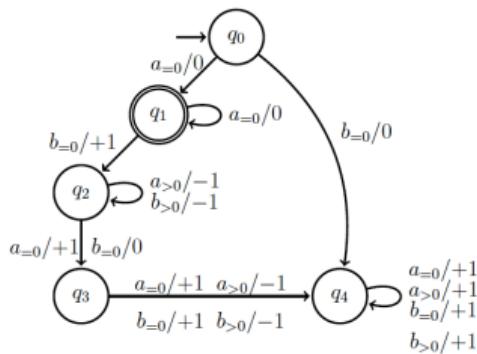
(b) Learnt by MinOCA



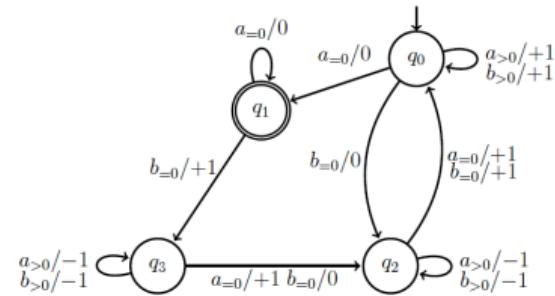
(c) Learnt by BPS

Fig. 2.17. The input DROCA recognises the language $\{a^n b^n a \mid n > 0\}$.

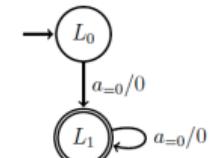
Example-2: DROCAs learnt by minOCA and BPS



(a) Input DROCA



(b) Learnt by MinOCA



(c) Learnt by BPS

Fig. 2.18. The input DROCA recognises the language $\{w \in \{a, b\}^* \mid w \text{ does not contain a } b\}$.

Conclusions

- The algorithm terminates after polynomially many queries.
- Learns an DROCA number of states less than or equal to the teacher DROCA.
- This can be used in learning a minimal visibly OCA if the teachers OCA is a visibly OCA.
- The minimisation problem for visibly OCA is NP-Complete (Michaliszyn and Otop, 2022, Gauwin et al., 2020).
- There is no polynomial time algorithm for learning a minimal counter-synchronous DROCA.

Future directions

- Does there exist a polynomial time algorithm for learning OCA?
 - The learnt model can contain a polynomially many states compared to the minimal OCA.
- Can we have a practical algorithm for learning OCA without using a SAT solver?

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URL: [https://github.com/mvcisback/dfa- identify](https://github.com/mvcisback/dfa-identify) .

Thank You !