

# State Space Reduction For Parity Automata

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# Overview

Goal: reduce the number of states in a given deterministic parity automaton while keeping the recognized language.

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- 1 Deterministic Parity Automata
- 2 Why do we need heuristic reduction?
- 3 Merger functions as a framework
- 4 Delayed Simulation
- 5 Congruence Path Refinement
- 6 Efficiency

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$\omega$ -words are words of one-sided infinite length:

$\Sigma^\omega =$  functions from  $\mathbb{N}$  to  $\Sigma$

$\omega$ -automata are finite transition structures that describe a language

$L \subseteq \Sigma^\omega$

Deterministic parity automata (DPA):

- ▶ State set  $Q$
- ▶ Alphabet  $\Sigma$
- ▶ Transition function  $\delta : Q \times \Sigma \rightarrow Q$
- ▶ Priority function  $c : Q \rightarrow \mathbb{N}$

An  $\omega$ -word  $\alpha$  starting in a state  $q_0 \in Q$  induces a run  $q_0 q_1 q_2 \dots$ .

The DPA accepts  $\alpha$  iff the **smallest** priority that occurs infinitely often in the sequence  $c(q_0)c(q_1)c(q_2)\dots$  is **even**.

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# Why do we need heuristic reduction?

Goal: Reduce number of states in the automaton to ease run time of follow up algorithms.

**Minimization Problem:** Given an automaton  $\mathcal{A}$ , what is the smallest number of states required to recognize the same language as  $\mathcal{A}$ ?

For DFAs: Minimization is solvable in  $\mathcal{O}(n \log n)$ . [Hopcroft, 1971]

For DPAs: Minimization is NP-hard. [Schewe, 2010]

# Moore Minimization

A DPA can be interpreted as a Moore automaton with  $c$  being the output function.

## Definition.

$p \equiv_M q$  iff  $\forall w \in \Sigma^* : c(\delta^*(p, w)) = c(\delta^*(q, w))$ .



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## Theorem.

*Deterministic Moore automata can be minimized in log-linear time.*

Idea: Build the quotient automaton w.r.t.  $\equiv_M$ .

The same algorithm can be used to reduce DPAs but will not give minimal DPAs in general.

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# Merger functions

**Merger functions**  $\mu$  map from some  $D \subseteq 2^Q$  into  $2^Q \setminus \{\emptyset\}$ .

$M, C \subseteq Q$

$$\mu(M) = C$$

All states from the **merge set** ...

... can be represented by any single  
one representative from the **candidate set**.

# Merger functions generalize quotient automata

Special case:  $\mu(M) = M$ .

Remove all states from  $M$  except for one (arbitrarily chosen) representative.

For a congruence relation  $\sim$ , let  $\mathfrak{C} \subseteq 2^Q$  be the equivalence classes. The quotient automaton is defined by state set  $\mathfrak{C}$ .

This is captured by the merger function  $\mu_{\div} : \mathfrak{C} \rightarrow 2^Q, \kappa \mapsto \kappa$ .

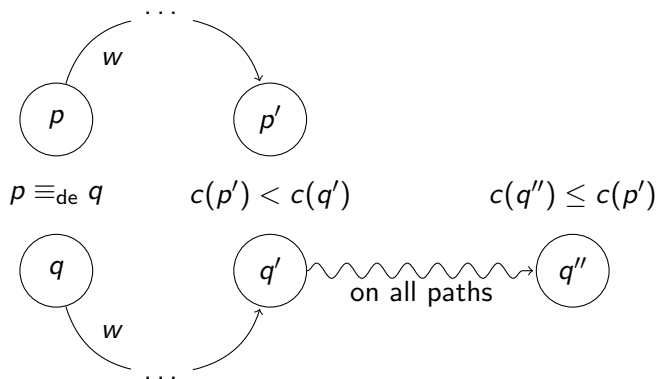
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## Definition.

$p \equiv_{\text{de}} q$  iff for all  $w \in \Sigma^*$ , every run that starts in  $\delta^*(p, w)$  or  $\delta^*(q, w)$  eventually sees a priority of at most  $\min\{c(\delta^*(p, w)), c(\delta^*(q, w))\}$ .

# Delayed Simulation



## Definition.

Let  $\mathfrak{C}_{\text{de}} = \{[q]_{\equiv_{\text{de}}} \mid q \in Q\}$  be the set of  $\equiv_{\text{de}}$ -equivalence classes. Define the **delayed simulation merger** as

$$\mu_{\text{de}} : \mathfrak{C}_{\text{de}} \rightarrow 2^Q, \kappa \mapsto \{q \in \kappa \mid c(q) = \min c(\kappa)\}.$$

## Theorem.

*Merging states according to  $\mu_{\text{de}}$  preserves language.*



# Computing Delayed Simulation

We define a det. Büchi automaton  $\mathcal{G}_{\text{de}}$  with states  $q_{\text{de}}^0(p, q)$  such that:  
 $p \equiv_{\text{de}} q$  iff both  $L(\mathcal{G}_{\text{de}}, q_{\text{de}}^0(p, q))$  and  $L(\mathcal{G}_{\text{de}}, q_{\text{de}}^0(q, p))$  are universal ( $\Sigma^\omega$ ).

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$$\mathcal{G}_{\text{de}} = (Q_{\text{de}}, \Sigma, \delta_{\text{de}}, F_{\text{de}})$$

- ▶ States are  $Q_{\text{de}} = Q \times Q \times (c(Q) \cup \{\checkmark\})$ .  
The first two components are a “simulation” of the original DPA. The third component are the so called “obligations”.
- ▶ Accepting states are  $F_{\text{de}} = Q \times Q \times \{\checkmark\}$ .
- ▶ Transitions  $\delta_{\text{de}}$ .

$$\delta_{\text{de}}((p, q, k), a) = (\delta(p, a), \delta(q, a), \gamma(c(\delta(p, a)), c(\delta(q, a)), k))$$

# Delayed Simulation Automaton: $\gamma$

(Actual definition of  $\gamma$  is more complex for some additional properties.)

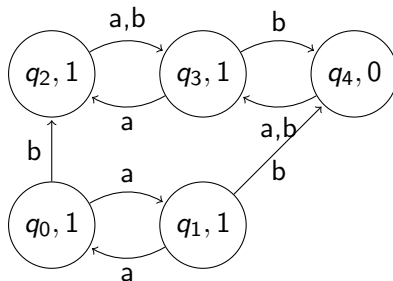
$$\gamma : \mathbb{N} \times \mathbb{N} \times (\mathbb{N} \cup \{\checkmark\}) \rightarrow \mathbb{N} \cup \{\checkmark\}$$

$$\gamma(i, j, \checkmark) = \begin{cases} \checkmark & \text{if } j \leq i \\ i & \text{else} \end{cases}$$

$$\text{for } k \in \mathbb{N} : \quad \gamma(i, j, k) = \begin{cases} \checkmark & \text{if } j \leq \min\{i, k\} \\ \min\{i, k\} & \text{else} \end{cases}$$

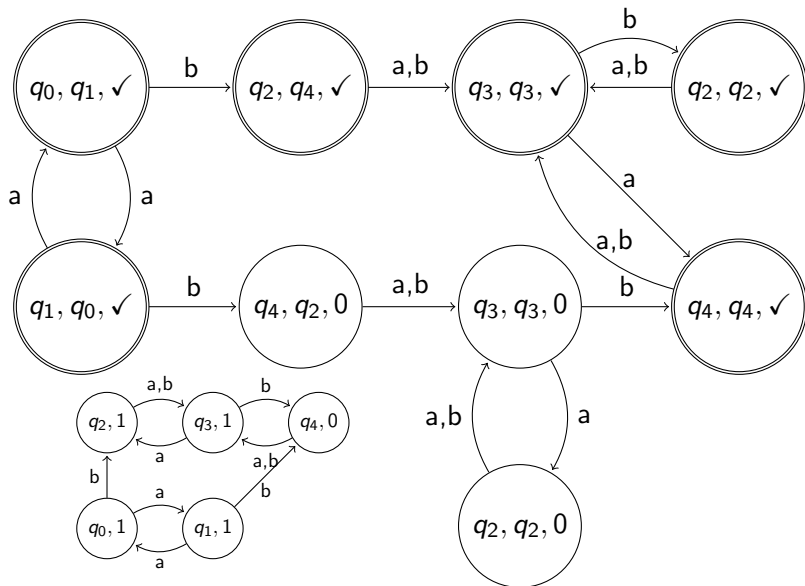
$$q_{\text{de}}^0(p, q) = (p, q, \gamma(c(p), c(q), \checkmark)).$$

# Delayed Simulation Automaton

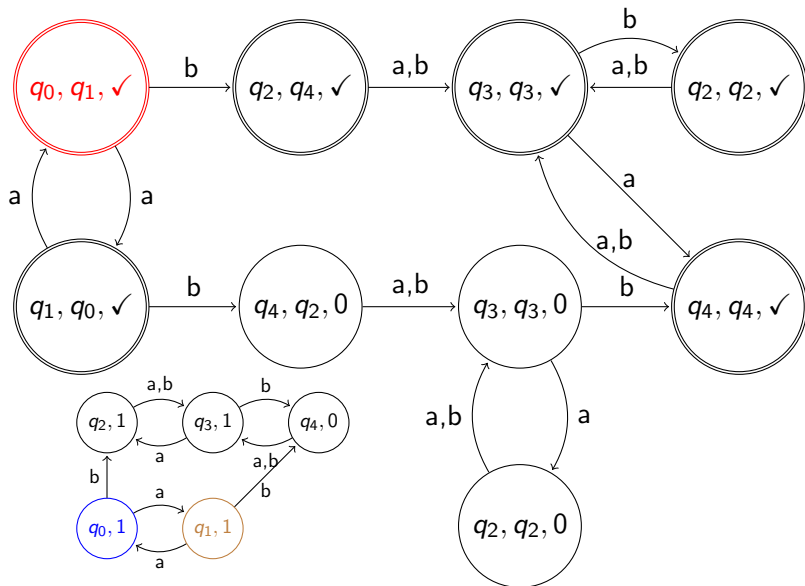


A DPA with 5 states. We want to check whether  $q_0 \equiv_{\text{de}} q_1$  is true.

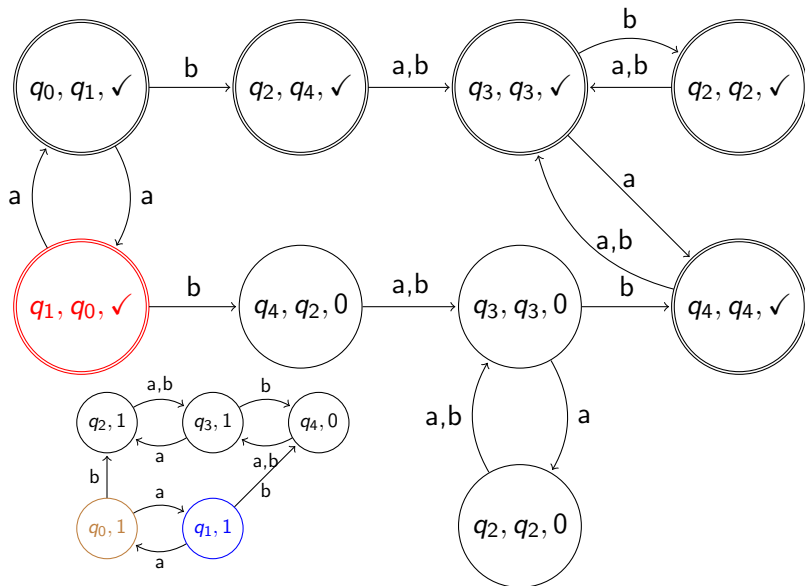
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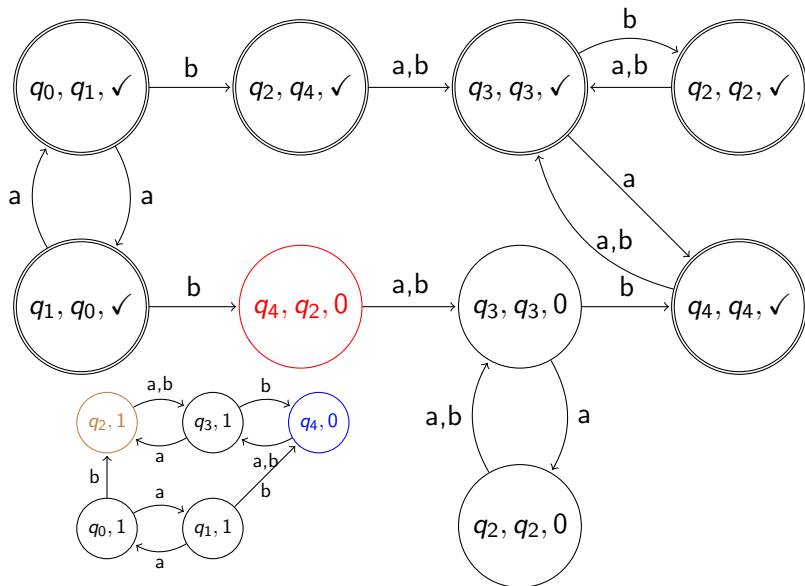
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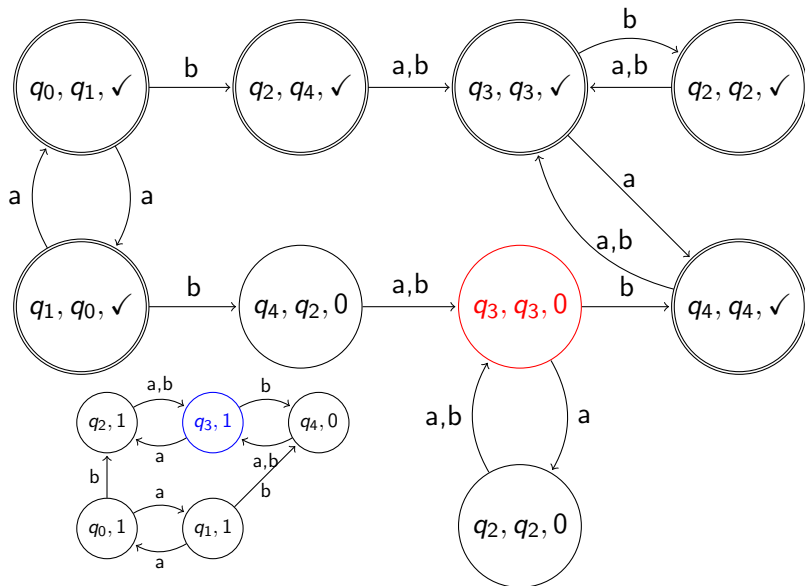


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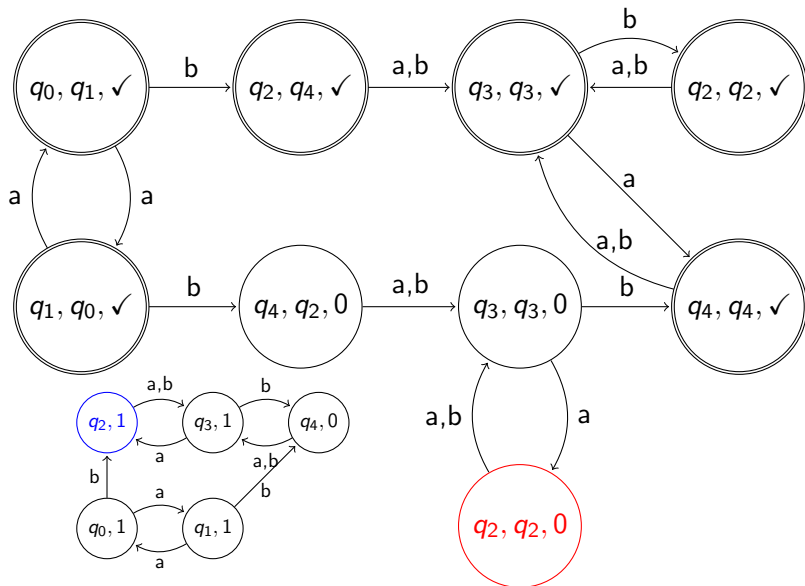




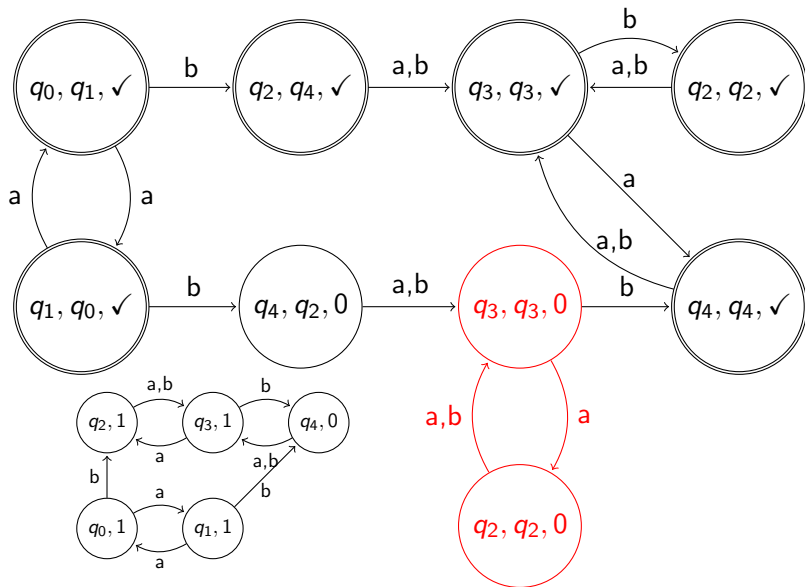
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$\mathcal{G}_{de}$  uses the state set  $Q_{de} = Q \times Q \times (c(Q) \cup \{\checkmark\})$ .

Computing states of universal language in a DBA requires linear time.

## Theorem.

$\mu_{de}$  can be computed in  $\mathcal{O}(n^2 k)$ .

$n = |Q|$ ,  $k = |c(Q)|$

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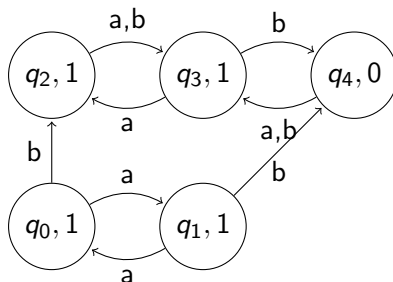
# Congruence Path Refinement

Idea: take a given relation and refine it until states can be merged.

## Definition.

Let  $\sim$  be a congruence relation and let  $\lambda \subseteq Q$  be an equiv. class of  $\sim$ . We define  $L_{\lambda \leftrightarrow} \subseteq \Sigma^*$  as the set of all words such that the induced run from a state in  $\lambda$  moves back to  $\lambda$  exactly once and ends there.

# Congruence Path Refinement

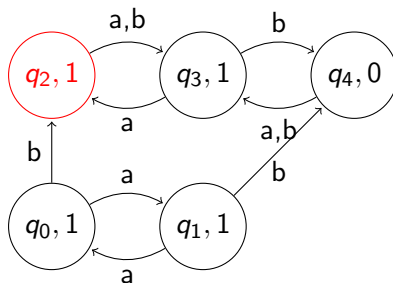


$$\lambda = \{q_2, q_4\}$$

Because  $\sim$  is a congruence relation, we only need to consider one state.

$$L_{\lambda \leftarrow} = \{$$

# Congruence Path Refinement



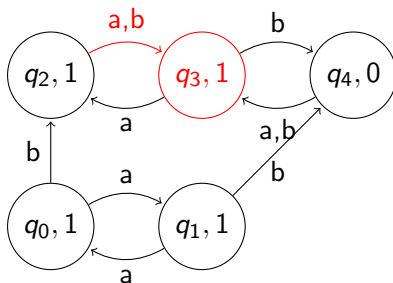
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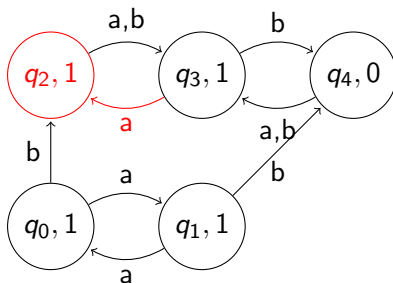
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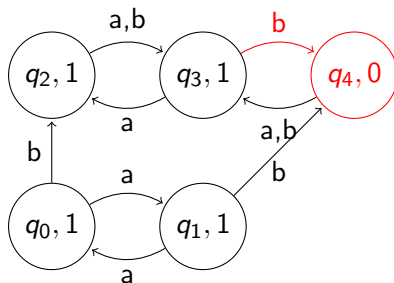
# Congruence Path Refinement



$$\lambda = \{q_2, q_4\}$$

$$L_{\lambda \leftarrow} = \{aa, ba\}$$

# Congruence Path Refinement



$$\lambda = \{q_2, q_4\}$$

$$L_{\lambda \leftrightarrow} = \{aa, ba, ab, bb\}$$

## Definition.

The **path refinement** equivalence  $\equiv_{\text{PR}}^\lambda$  is the **largest relation** s.t.:

For  $p, q \in \lambda$ ,  $p \equiv_{\text{PR}}^\lambda q$  if and only if

- ▶  $\forall w \in L_{\lambda \leftarrow} : \delta^*(p, w) \equiv_{\text{PR}}^\lambda \delta^*(q, w)$
- ▶  $\forall w \in L_{\lambda \leftarrow} : \text{the smallest priority seen when reading } w \text{ is the same from } p \text{ and from } q.$

## Definition.

Let  $\mathfrak{C}_{PR}^\lambda = \{[q]_{\equiv_{PR}^\lambda} \mid q \in Q\}$  be the set of  $\equiv_{PR}^\lambda$ -equivalence classes. Define the **path refinement merger** as

$$\mu_{PR}^\lambda : \mathfrak{C}_{PR}^\lambda \rightarrow 2^Q, \kappa \mapsto \{q \in \kappa \mid c(q) = \min c(\kappa)\}.$$

## Theorem.

*If all states in  $\lambda$  are pairwise language equivalent, merging states according to  $\mu_{PR}^\lambda$  preserves language.*

# Computing Path Refinement

Define a DPA that relates Moore equivalence to  $\equiv_{PR}$ .

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## Definition.

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- ▶  $Q_{\text{visit}}^\lambda = Q \times c(Q) \times (c(Q) \cup \{-1\})$
- ▶  $\delta_{\text{visit}}^\lambda((q, k, k'), a) = \begin{cases} (q', \min\{k, c(q')\}, -1) & \text{if } q' \notin \lambda \\ (q', c(q'), \min\{k, c(q')\}) & \text{if } q' \in \lambda \end{cases}$   
where  $q' = \delta(q, a)$ .
- ▶  $c_{\text{visit}}^\lambda((q, k, k')) = k'$ .

The first component “simulates” the original automaton  $\mathcal{A}$ .

The second component tracks the minimal priority seen on one run from  $\lambda$  to  $\lambda$ .

The third component is required to distinguish the different priorities.

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The **third** component is required to distinguish the different priorities.

## Definition.

For  $q \in Q$ , we set  $\iota_q := (q, c(q), \max c(Q)) \in Q_{\text{visit}}^\lambda$ .

## Theorem.

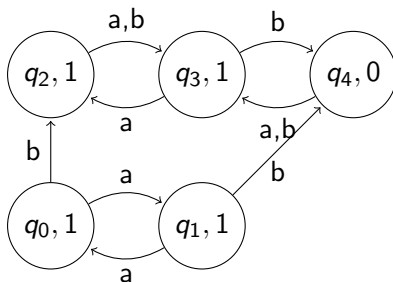
$p \equiv_{PR}^\lambda q$  iff  $\iota_p \equiv_M \iota_q$ .

## Theorem.

$\equiv_{PR}^\lambda$  can be computed in  $\mathcal{O}(k^2 n \log n)$ .

$n = |Q|$ ,  $k = |c(Q)|$

# Visit Graph



Potential choices for  $\lambda$  are the equivalence classes of  $\equiv_L$ :  
 $\{q_0, q_1\}$ ,  $\{q_2, q_4\}$ , or  $\{q_3\}$ .

We take  $\lambda = \{q_2, q_4\}$  and ask if  $q_2 \equiv_{PR}^\lambda q_4$  is true.

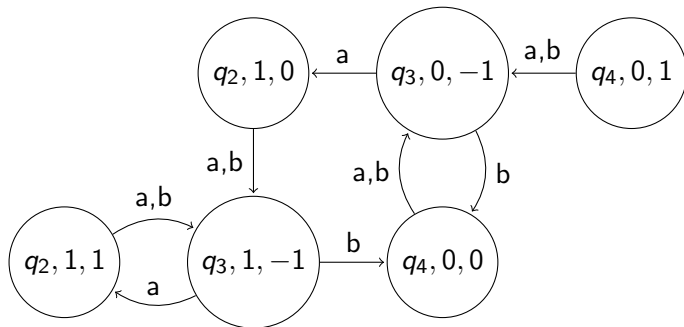
# Visit Graph

$$\mathcal{A}_{\text{visit}}^{\{q_2, q_4\}}$$

$$\iota_{q_2} = (q_2, 1, 1)$$

$$\iota_{q_4} = (q_4, 0, 1)$$

Question:  $\iota_{q_2} \equiv_M \iota_{q_4}$ ?



(Reminder: the third component defines the color of a state)

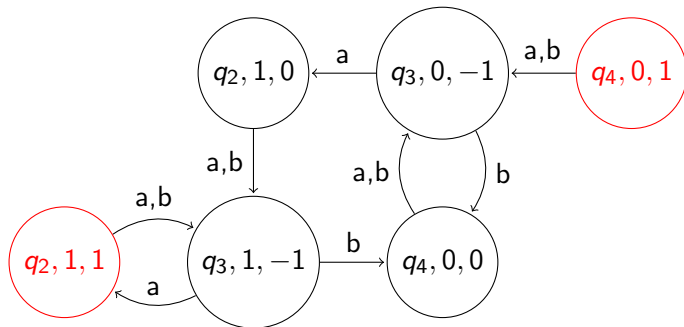
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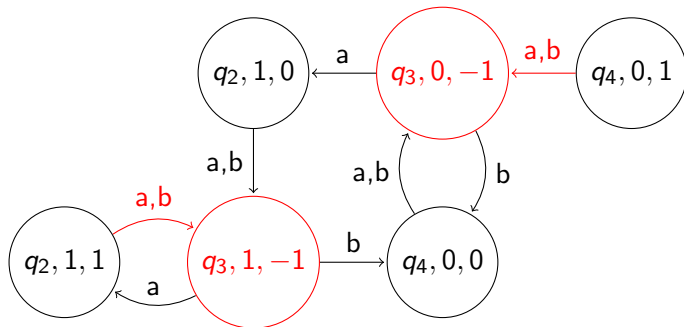
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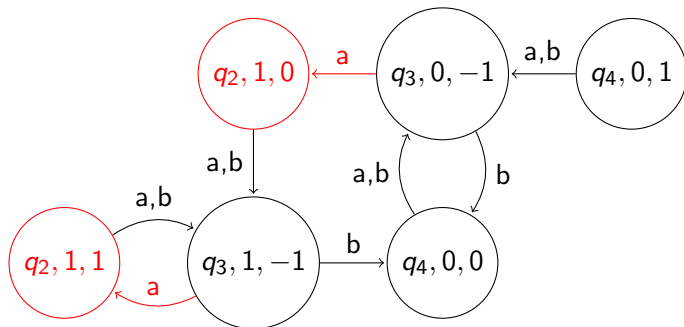
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$q_2$  and  $q_4$  are not PR-equivalent.

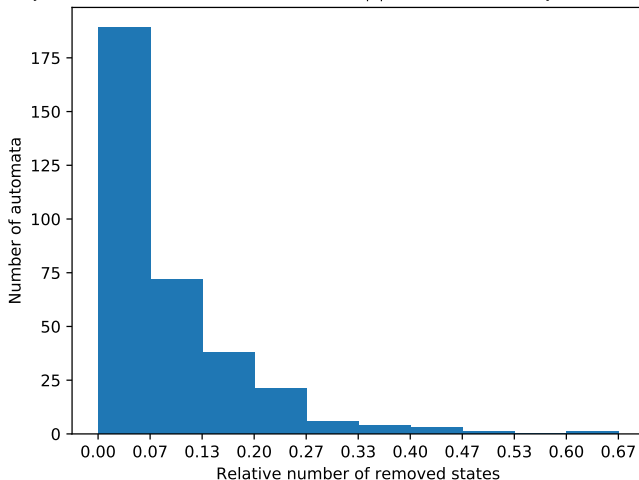


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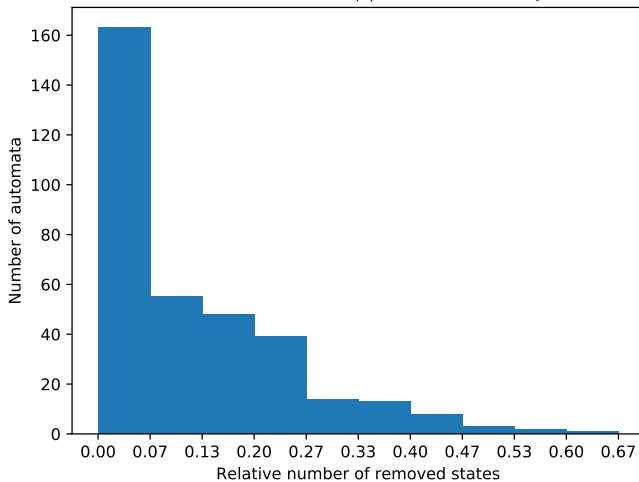
# Delayed Simulation

Delayed Simulation state reduction on a DPA with  $|\Sigma|=2$  that was created by nbautils from an NBA.

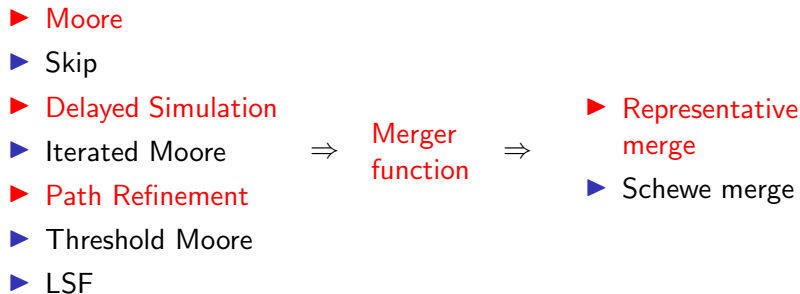


# Path Refinement

Path Refinement state reduction on a DPA with  $|\Sigma|=2$  that was created by nbautils from an NBA.



# Summary







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