

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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ω -words are words of one-sided infinite length:

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

ω -automata are finite transition structures that describe a language

$L \subseteq \Sigma^\omega$

Deterministic parity automata (DPA):

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

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

The DPA accepts α 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 μ 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 $\mu_{\div} : \mathfrak{C} \rightarrow 2^Q, \kappa \mapsto \kappa$.

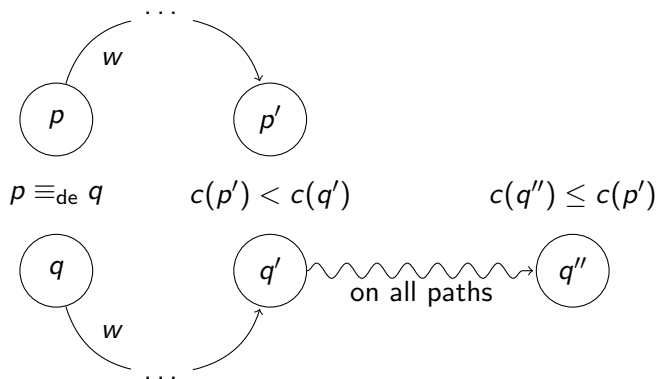
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Definition.

$p \equiv_{\text{de}} q$ 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_{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 μ_{de} preserves language.

Computing Delayed Simulation

We define a det. Büchi automaton \mathcal{G}_{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 (Σ^ω).

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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 δ_{de} .

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

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

Delayed Simulation Automaton: γ

(Actual definition of γ is more complex for some additional properties.)

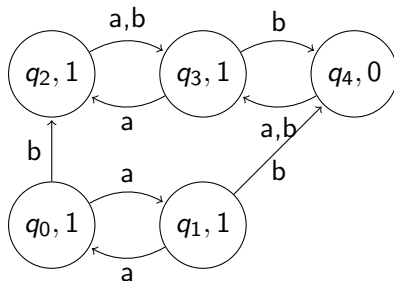
$$\gamma : \mathbb{N} \times \mathbb{N} \times (\mathbb{N} \cup \{\checkmark\}) \rightarrow \mathbb{N} \cup \{\checkmark\}$$

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

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

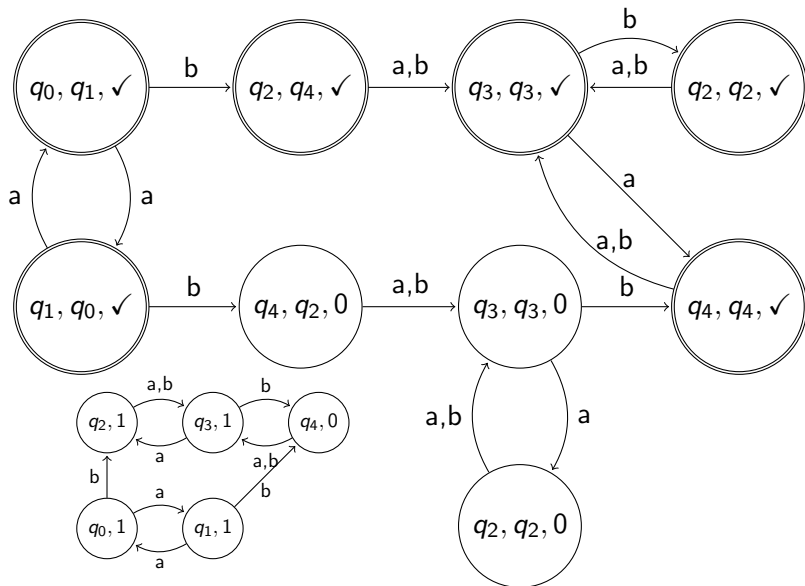
$$q_{\text{de}}^0(p, q) = (p, q, \gamma(c(p), c(q), \checkmark)).$$

Delayed Simulation Automaton

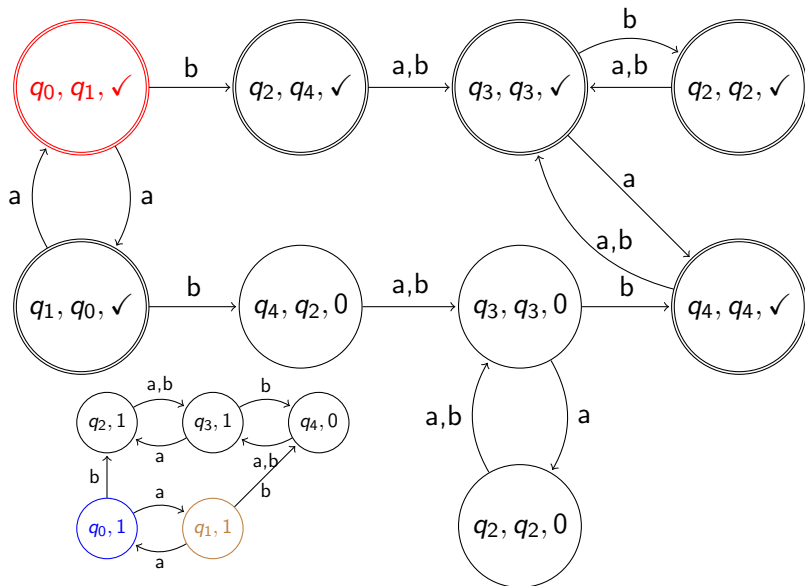


A DPA with 5 states. We want to check whether $q_0 \equiv_{\text{de}} q_1$ is true.

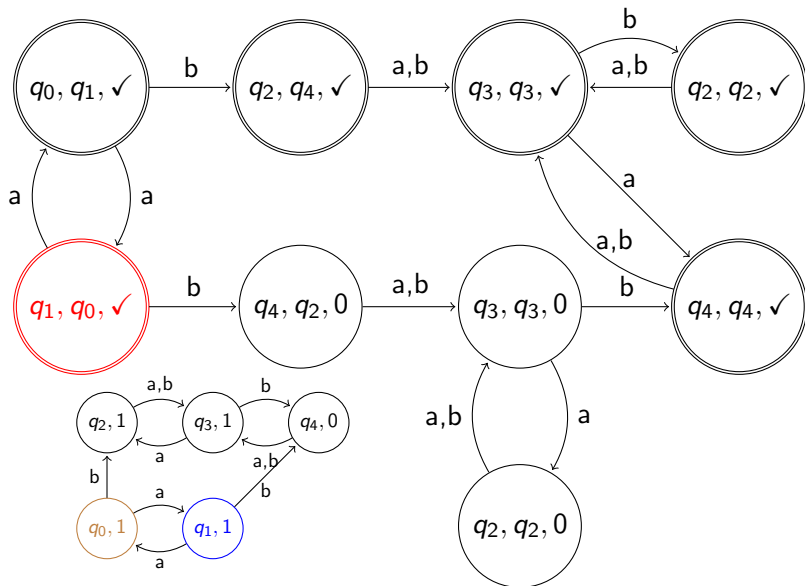
Delayed Simulation Automaton



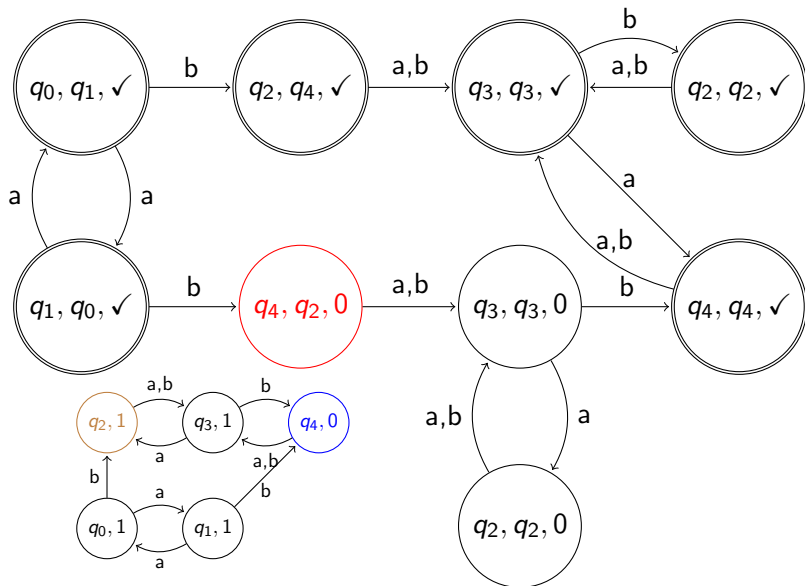
Delayed Simulation Automaton



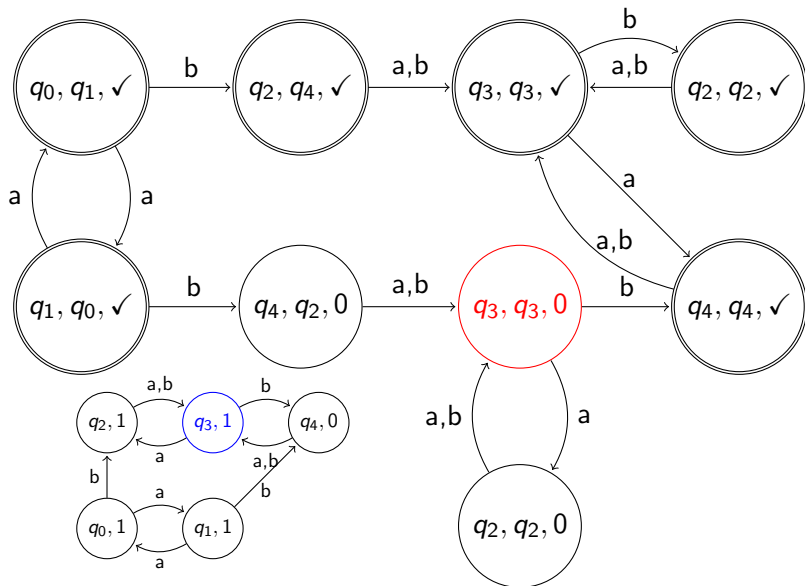
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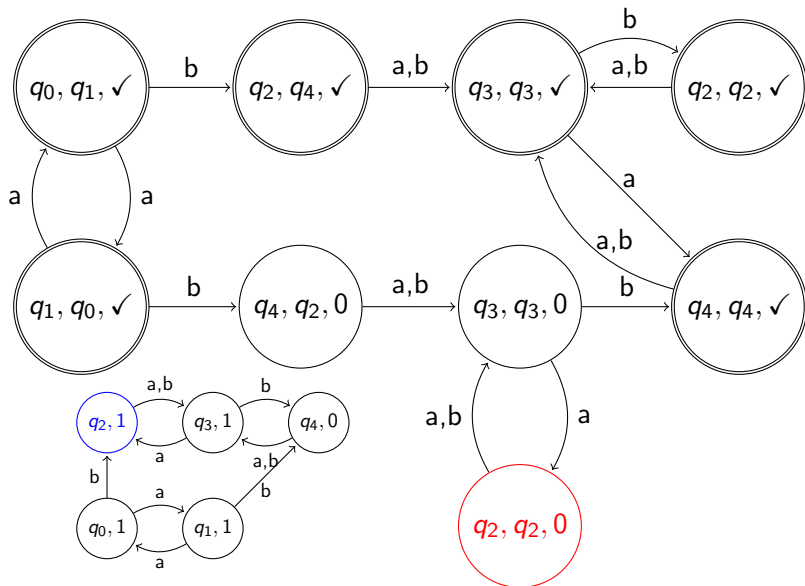
Delayed Simulation Automaton



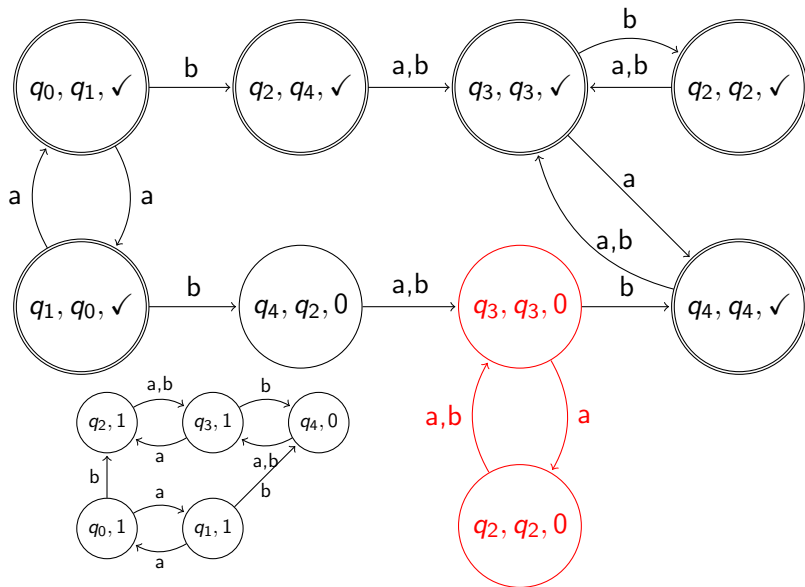
Delayed Simulation Automaton



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Delayed Simulation Automaton



\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.

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

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

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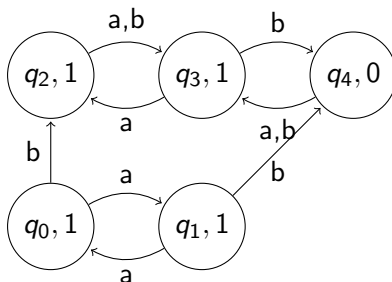
Congruence Path Refinement

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 λ moves back to λ exactly once and ends there.

Congruence Path Refinement

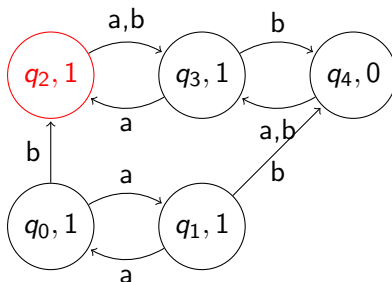


$$\lambda = \{q_2, q_4\}$$

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

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

Congruence Path Refinement

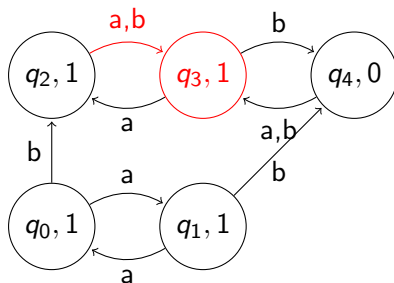


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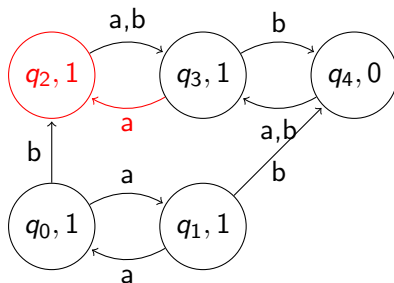
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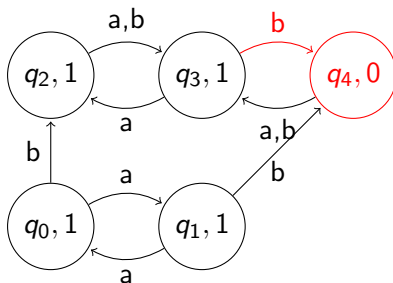
Congruence Path Refinement



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

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

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

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

First point: Makes sure path segments have the same acceptance.

Second point: Makes sure the argument can be applied repeatedly.

Definition.

Let $\mathfrak{C}_{PR}^\lambda = \{[q]_{\equiv_{PR}^\lambda} \mid q \in Q\}$ be the set of \equiv_{PR}^λ -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 λ are pairwise language equivalent, merging states according to μ_{PR}^λ preserves language.

Computing Path Refinement

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

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Definition.

Define the **visit graph** DPA $\mathcal{A}_{\text{visit}}^\lambda = (Q_{\text{visit}}^\lambda, \Sigma, \delta_{\text{visit}}^\lambda, c_{\text{visit}}^\lambda)$.

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

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.

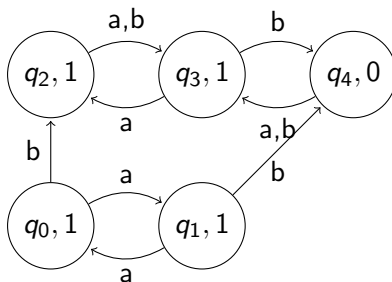
$p \equiv_{PR}^\lambda q$ iff $\iota_p \equiv_M \iota_q$.

Theorem.

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

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

Visit Graph



Potential choices for λ 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.

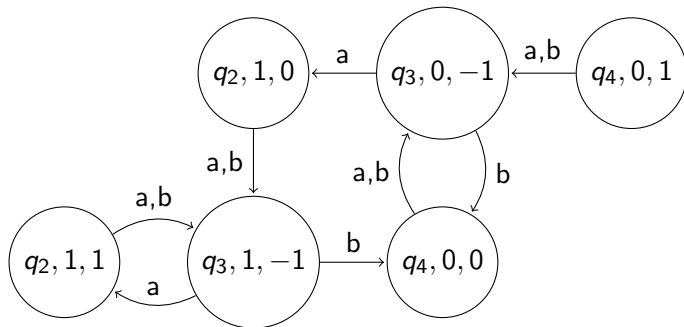
Visit Graph

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

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

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

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



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

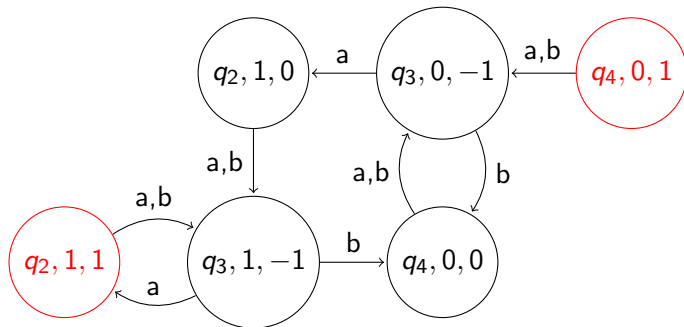
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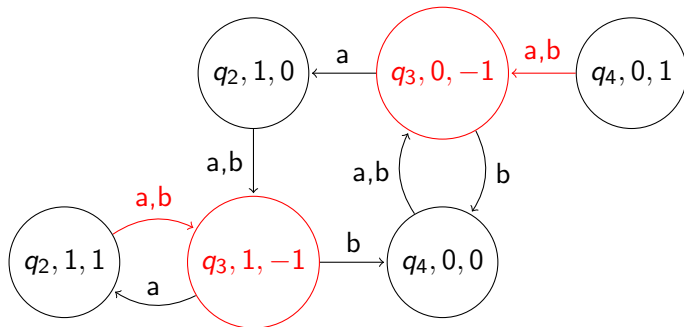
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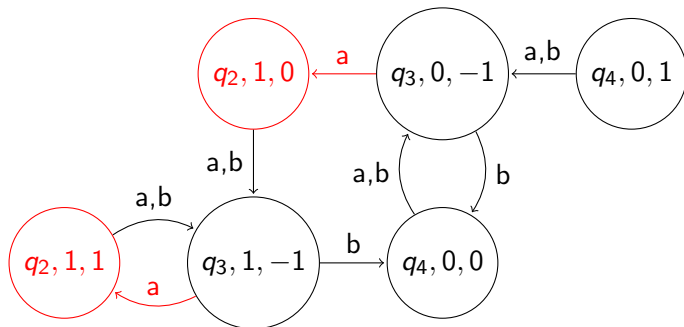
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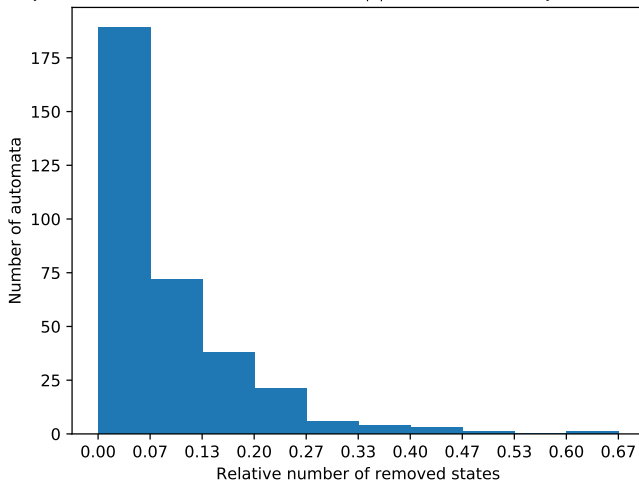
q_2 and q_4 are not PR-equivalent.

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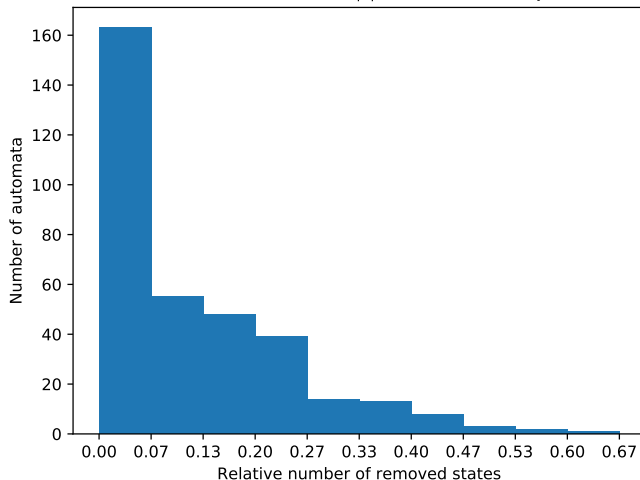
Delayed Simulation

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

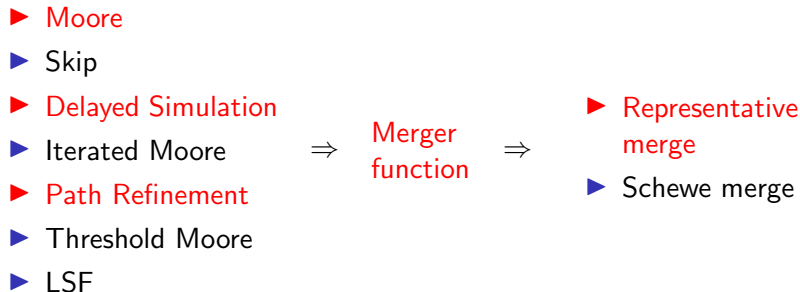


Path Refinement

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



Summary





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