

# Backward and Forward Bisimulation of Finite Tree Automata

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# Outline

- 1 Background and motivation
- 2 Bisimulation minimisation of tree automata
- 3 Partition refinement algorithms
- 4 An NLP application
- 5 Work in progress

# Ranked alphabets and trees

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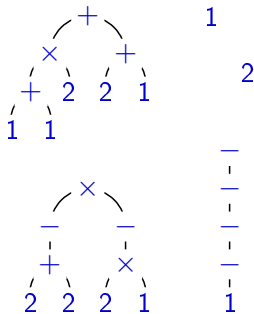
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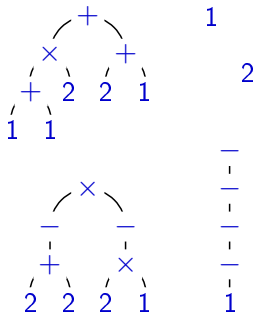
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The set of **trees over**  $\Sigma$  is written  $T_{\Sigma}$ .

A **tree language** w.r.t.  $\Sigma$  is simply a subset of  $T_{\Sigma}$ .

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## Finite tree automata

A **finite tree automaton** (fta) is a tuple  $(Q, \Sigma, \delta, F)$  where

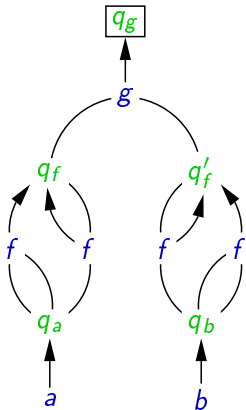
- ▶  $Q$  is a finite set of **states**,
- ▶  $\Sigma$  is a ranked **input alphabet**,
- ▶  $\delta$  is a finite set of **transition rules** in the form

$$f(q_1, \dots, q_n) \rightarrow q_{n+1} \text{ ,}$$

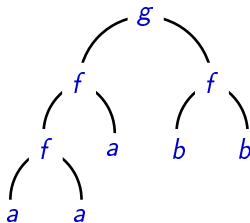
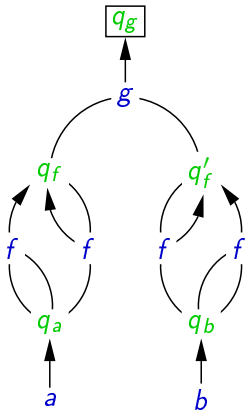
where  $f \in \Sigma_{(n)}$ , and  $q_1, \dots, q_{n+1} \in Q$ , for some  $n \in \mathbb{N}$ .

- ▶ Finally,  $F \subseteq Q$  is a set of **accepting** states.

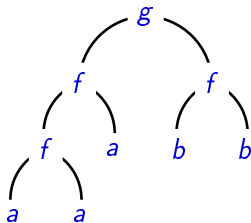
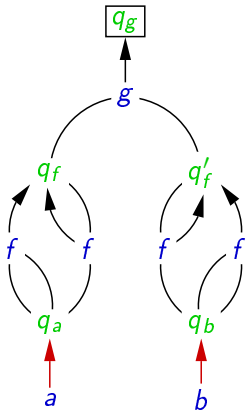
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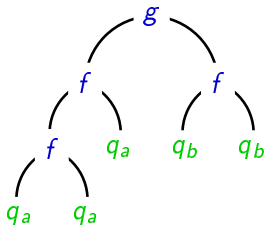
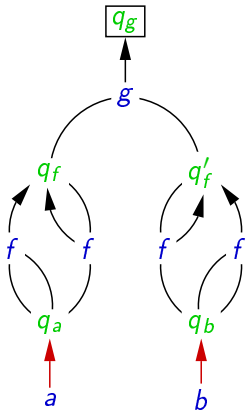
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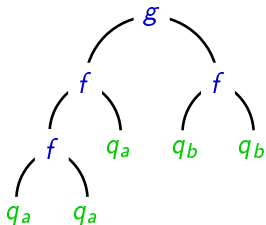
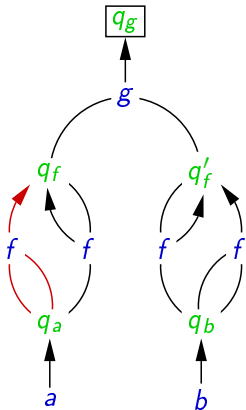
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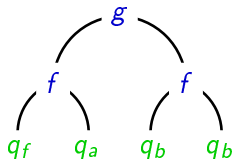
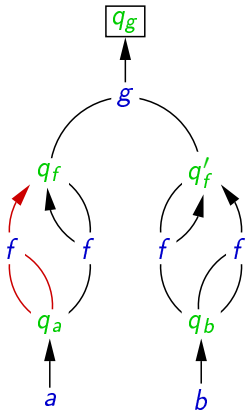
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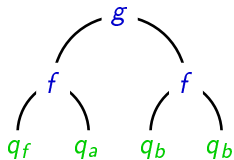
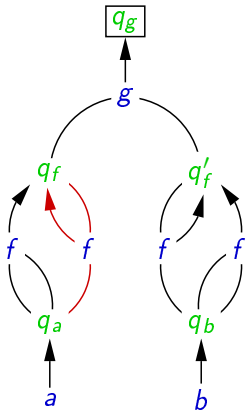


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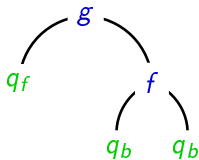
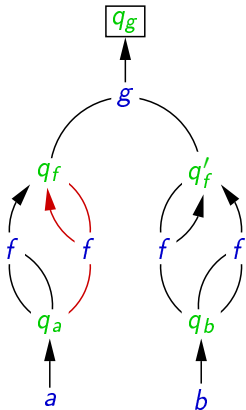




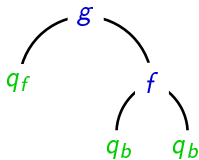
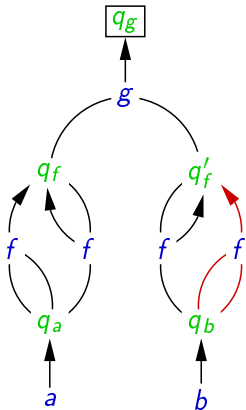
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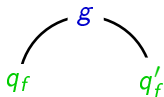
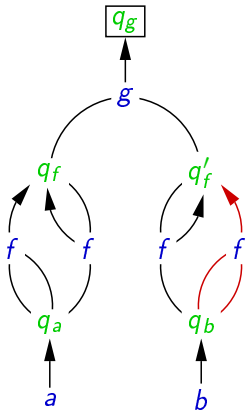
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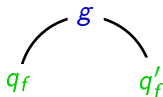
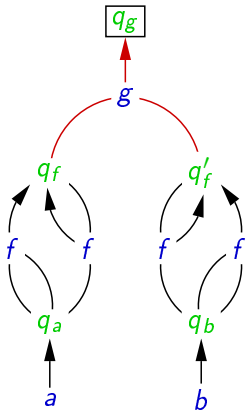
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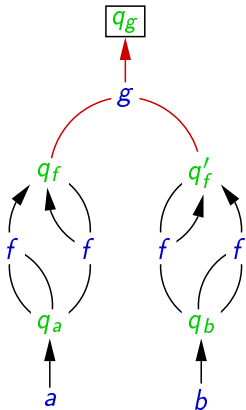
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$q_g$

# Applications

## Finite tree automata ...

- ▶ offer a nice combination of generative power and analytical transparency.
- ▶ are useful in areas such as **lexical analysis**, **model checking** and **natural language processing**.

To allow for efficient computations, we want to work with as small fta as possible. This makes a minimisation algorithm a useful tool.

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Any efficient algorithm that searches for a solution to the general problem, must thus use heuristics.

# Bisimulation

The notion of **bisimilarity** is due to R. Milner.

Intuitively, two states are bisimilar if they serve the same purpose.

We adopt P. Buchholz definitions and extend these to trees:

- ▶ **Backward bisimulation** Two states are bisimilar if every tree that is mapped to the one state is also mapped to the other.
- ▶ **Forward bisimulation** Two states are bisimilar if they can always be exchanged for each other during a run on an input tree  $t$ , without affecting the way  $t$  is classified.

## Backward bisimulation

Let  $A = (Q, \Sigma, \delta, F)$  be an fta. An equivalence relation  $\simeq$  on  $Q$  is a **backward bisimulation** if  $p \simeq q$  means that

$$f(p_1, p_2, \dots, p_k) \rightarrow p ,$$

implies that there exists a rule

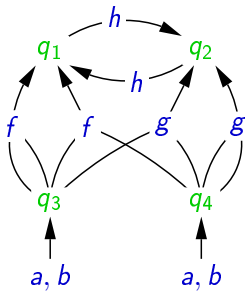
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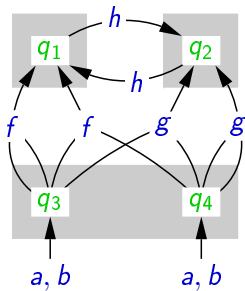
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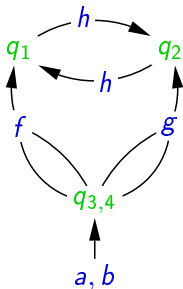
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## Forward bisimulation

Let  $A = (Q, \Sigma, \delta, F)$  be an fta. An equivalence relation  $\simeq$  on  $Q$  is a **forward bisimulation** if  $p \simeq q$  means that

- ▶  $q \in F$  if and only if  $p \in F$ , and
- ▶ the fact that

$$f(p_1, \dots, p_{i-1}, p, p_i, \dots, p_k) \rightarrow p_{k+1} \text{ ,}$$

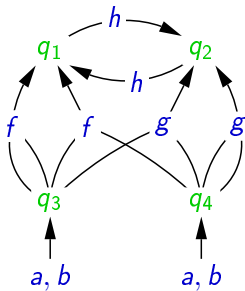
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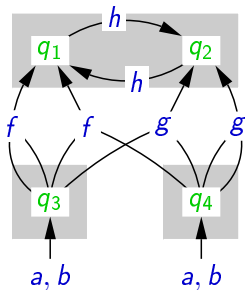
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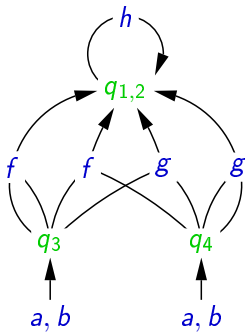
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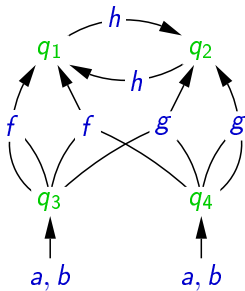
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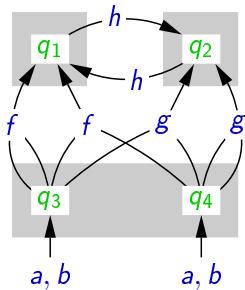
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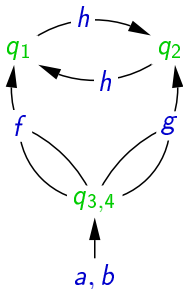
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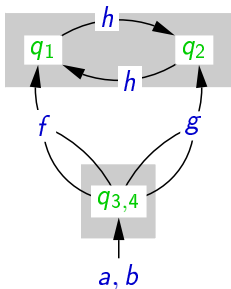
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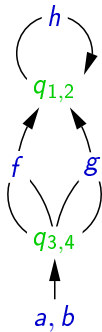
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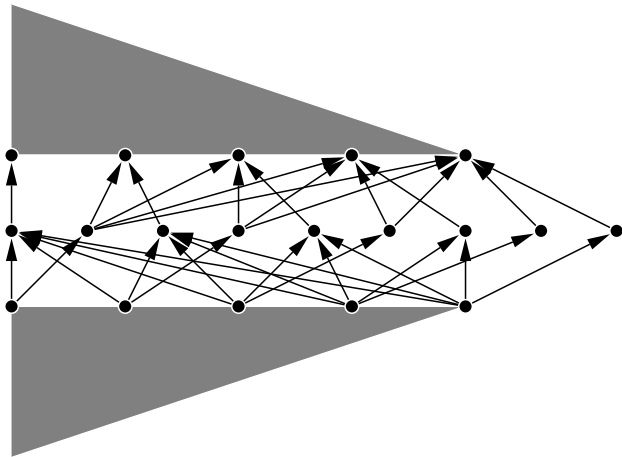


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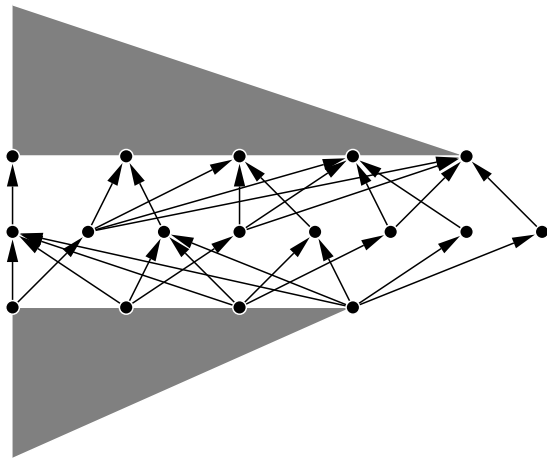
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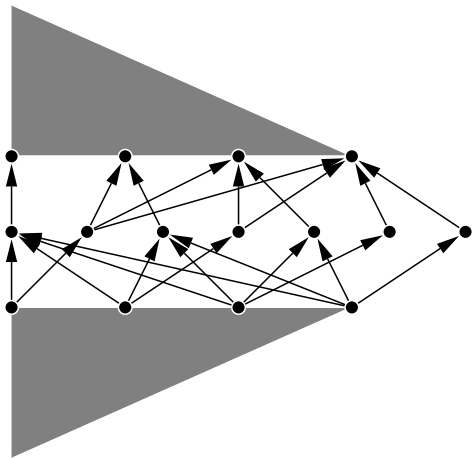
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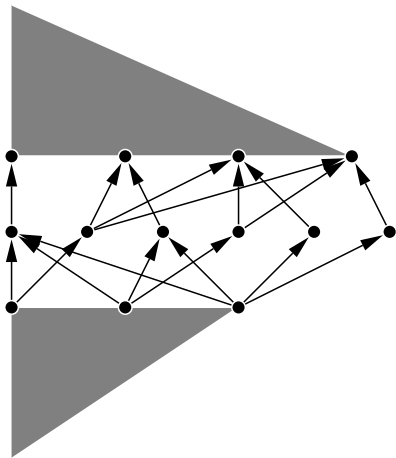
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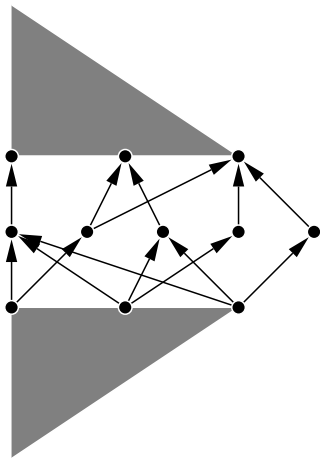


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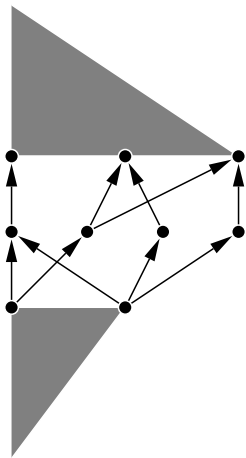




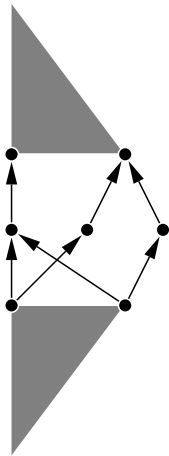
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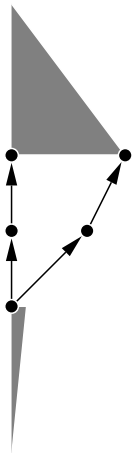
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# Partition refinement algorithms in general

## The Coarsest Partition Problem

Given a transition system  $(Q, \delta)$  and a condition  $c$ , find the coarsest partition of  $Q$  that meets with  $c$ .

- 1 Let the initial partition  $P_0$  be  $\{Q\}$ .
- 2 Traverse the rules in  $\delta$ , and
  - ▶ record the “behaviour” of each  $q \in Q$  in the vector  $v(q)_i$ .
- 3 The partition  $P_{i+1}$  is obtained by bucket sorting each  $q$  in  $Q$  using  $([q]_{P_i}, v(q)_i)$  as key.
- 4 if  $P_{i+1}$  and  $P_i$  coincide, then we are done, else, go to Step 2.

## Time complexity

If  $\delta$  is deterministic, then we can use the “process the smaller half” strategy by J. E. Hopcroft. In this case, we only have to consider a total of  $\mathcal{O}(m \log n)$  rules, counting repetitions [Hopcroft, 1971].

If  $\delta$  is nondeterministic, then we must also use a counting argument by Paige & Tarjan. [Paige and Tarjan, 1987].

### Time complexity

Let  $r$  be the maximum rank of the input alphabet, let  $m$  be the number of transitions, and let  $n$  be the number of states.

- ▶ The forward algorithm runs in time  $\mathcal{O}(r m \log n)$ , and
- ▶ the backward algorithm runs in time  $\mathcal{O}(r^2 m \log n)$ .

## AKH bisimulation

An equivalence relation  $\simeq$  is an **AKH bisimulation** if

- ▶ the relation respects the final states, and
- ▶ the fact that  $p \simeq q$  and there is a rule

$$f(p_1, \dots, p_{i-1}, p, p_i, \dots, p_k) \rightarrow p_{k+1} \text{ ,}$$

where  $i \in \{1, \dots, k\}$ , implies that there is also a rule

$$f(q_1, \dots, q_{i-1}, q, q_i, \dots, q_k) \rightarrow q_{k+1} \text{ ,}$$

s.t.  $p_j \simeq q_j$ , for every  $j \in \{1, \dots, k+1\} \setminus \{i\}$ , and vice versa.



# Comparison

## Forward bisimulation ...

- ▶ coincides with the standard minimisation algorithm when the input automaton is deterministic, and
- ▶ is a factor  $r$  easier to compute than both AKH bisimulation and backward bisimulation.

## Backward bisimulation ...

- ▶ is no harder to compute than AKH bisimulation, and
- ▶ produces, in the general case, smaller output automata than both forward and AKH bisimulation.

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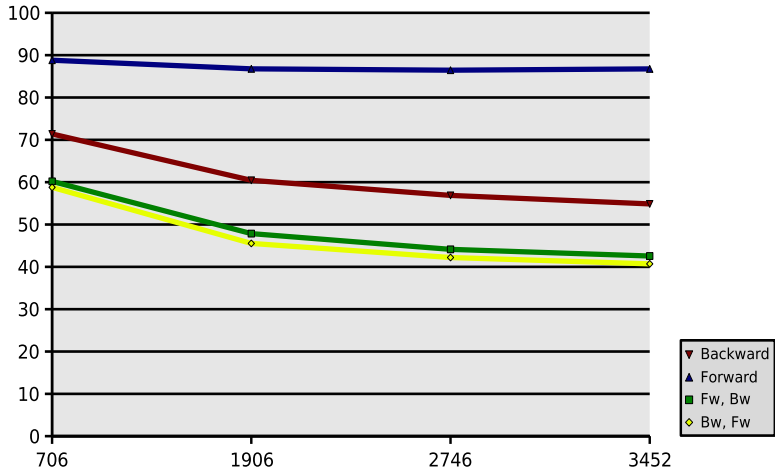
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## Solution

- 1 First, construct a trivial automaton from the sample set,
- 2 next, apply implementations of the minimisation algorithms.

# Experimental results



## Work in progress

Forward and backward bisimulation can also be defined for **weighted tree automata**.

- ▶ Leads to  $\mathcal{O}(mnr)$  minimisation algorithms for general semirings, but
- ▶  $\mathcal{O}(r^2 m \log n)$ ,  $\mathcal{O}(rm \log n)$  if the underlying algebraic structure is cancellative.

**Future work** includes

- ▶ weight pushing, and
- ▶ a more thorough study of the interaction between forward and backward bisimulation.