

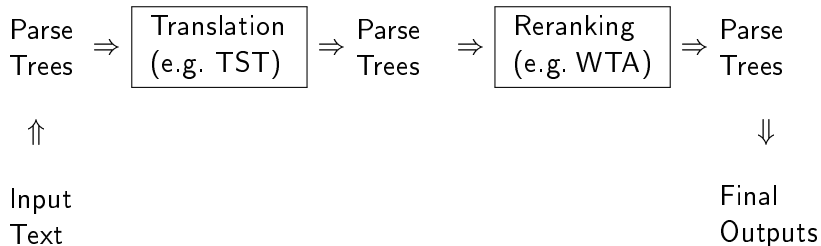
Minimization of Deterministic Weighted Tree Automata

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Syntax-based MT

Overview



Abbreviations

- ▶ TST = Tree Series Transducer
- ▶ WTA = Weighted Tree Automaton

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Overview

Tree Series

- ▶ Assigns **weight** (e.g. a probability) to each tree
- ▶ Weight drawn from semiring; e.g. $(\mathbb{R}, +, \cdot, 0, 1)$

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Weighted Tree Automaton

- ▶ **Finitely** represents tree series
- ▶ Defines the **recognizable** tree series

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Weighted Tree Automaton

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Application

- ▶ Re-ranker for parse trees
- ▶ Representation of parses

Syntax

Definition

Weighted tree automaton (wta) is tuple $(Q, \Sigma, \mathcal{A}, F, T)$ where

- ▶ Q : finite set of *states*
- ▶ Σ : ranked alphabet of *input symbols*
- ▶ $\mathcal{A} = (A, +, \cdot, 0, 1)$: semiring of *weights*
- ▶ $F \subseteq Q$: *final states*

Syntax

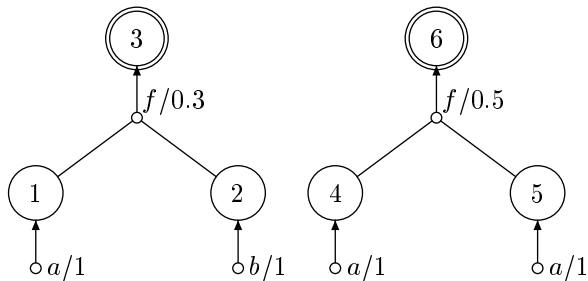
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- ▶ $F \subseteq Q$: *final states*
- ▶ T : finite set of transitions of the form $\sigma(q_1, \dots, q_k) \xrightarrow{a} q$

Syntax — Illustration

Sample Automaton



Semantics

Definition

Left-most derivations are defined as usual.

- ▶ **Weight of a derivation:**
product of the weights of the employed transitions
(each transition counted as often as it occurs)

Semantics

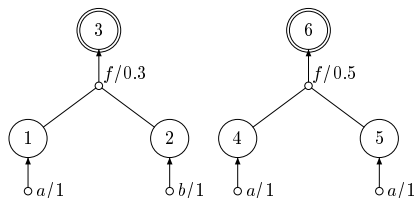
Definition

Left-most derivations are defined as usual.

- ▶ **Weight of a derivation:**
product of the weights of the employed transitions
(each transition counted as often as it occurs)
- ▶ **Weight $wt(t)$ of a tree t :**
sum of the weights of all left-most derivations that start
with t and end in a final state

Semantics — Illustration

Sample Automaton

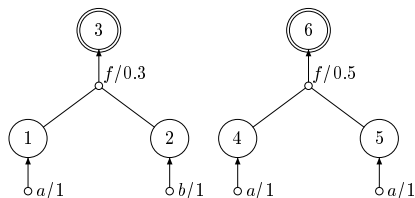


Sample Derivations

Input tree: $f(a, b)$ Derivation: $f(a, b)$

Semantics — Illustration

Sample Automaton

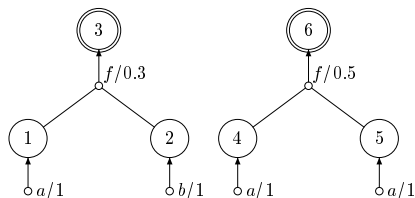


Sample Derivations

Input tree: $f(a, b)$ Derivation: $f(1, b)$ with weight 1

Semantics — Illustration

Sample Automaton

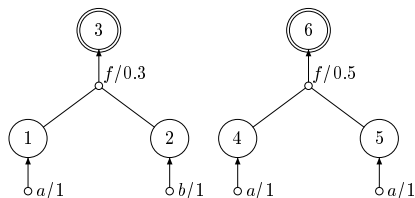


Sample Derivations

Input tree: $f(a, b)$ Derivation: $f(1, 2)$ with weight 1

Semantics — Illustration

Sample Automaton

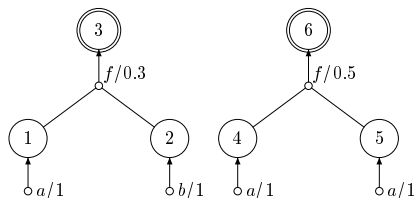


Sample Derivations

Input tree: $f(a, b)$ Derivation: **3** with weight 0.3

Semantics — Illustration

Sample Automaton

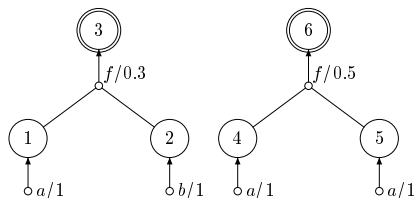


Sample Derivations

Input tree: $f(a, b)$ Derivation: 3 with weight 0.3
 $f(4, b)$ with weight 1

Semantics — Illustration

Sample Automaton



Sample Derivations

Input tree: $f(a, b)$ Derivation: 3 with weight 0.3
 $f(4, 2)$ with weight 1

Determinism

Definition

Deterministic wta: for every σ and q_1, \dots, q_k there exists exactly one transition $\sigma(q_1, \dots, q_k) \xrightarrow{a} q \in T$

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One derivation for each tree

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Notes

- ▶ Deterministic wta do not use addition
- ▶ Recognizable \neq deterministically recognizable
- ▶ Determinization sometimes possible [Borchardt & Vogler '03]
- ▶ Partial determinization [May & Knight '06]

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Known Minimization Results

	deterministic		nondeterministic	
	string	tree	string	tree
unweighted	$O(lm \log n)$	$O(lm \log n)$	PSPACE	PSPACE
weighted*	$O(lm \log n)$?	P	P

- ▶ l : maximal rank of symbols
- ▶ m : number of transitions
- ▶ n : number of states

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Applicability

- ▶ Deterministic wta
- ▶ Commutative semifield (i.e. multiplicative inverses)

Roadmap

- ▶ MYHILL-NERODE congruence relation [Borchardt '03]
- ▶ Determine signs of life
- ▶ Scale maps
- ▶ Refinement

MYHILL-NERODE congruence

Definition

$p \equiv q$: there exists nonzero a such that for every context C

$$\text{wt}(C[p]) = a \cdot \text{wt}(C[q])$$

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Notes

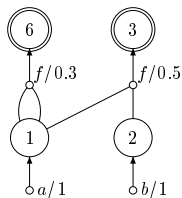
- ▶ Semifields are zero-divisor free
- ▶ Element a is unique if p is not dead

Signs of Life

Definition

Sign of life of $q \in Q$: context C such that $\text{wt}(C[q]) \neq 0$

Example



State	Sign of life	State	Sign of life
1	$f(\square, b)$	2	$f(1, \square)$
3	\square	6	\square

Compute Signs of Life

How?

- ▶ start with final states
- ▶ apply transition as in a grammar
- ▶ record reached states and their signs of life
- ▶ rinse and repeat with those states

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Theorem

We can determine signs of life in $O(lm)$.

Scaling Map

Definition

$f: Q \rightarrow A$ is **scaling map** for partition Π of Q if

- (i) $f(q) = 1$ for every dead state q
- (ii) for every block $B \in \Pi$ there exists context C such that

$$\text{wt}(C[q]) = f(q) \cdot \text{wt}(C[r(B)])$$

and C is a sign of life for every live $q \in B$

Note

Scaling maps are nonzero everywhere.

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Scaling maps are nonzero everywhere.

Theorem

A scaling map for Π can be computed in time $O(n^2)$.

Refinement

Definition

Refinement of Π : Split each block $B \in \Pi$ into two blocks, where one block contains all q such that

$$(ii) \sigma(C[q]) \xrightarrow{a} q' \equiv_{\Pi} p' \xleftarrow{b} \sigma(C[r(B)])$$

(iii) if q' is live, then

$$f(q)^{-1} \cdot a \cdot f(q') = b \cdot f(p')$$

for all symbols σ and contexts C .

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Theorem

Each refinement can be implemented to run in time $O(lm)$.

Complete Algorithm

Algorithm

```
( $\Pi$ , sol,  $D$ )  $\leftarrow$  COMPUTESOL( $M$ ) // in  $O(lm)$ 
2:  $\Pi \leftarrow$  REFINECONG( $\Pi$ ) // classical minimization in  $O(lm \log n)$ 
    $f \leftarrow$  COMPUTESM( $\Pi$ ) // in  $O(n^2)$ 
4: repeat
    $\Pi' \leftarrow \Pi$  // store old partition
6:  $\Pi \leftarrow$  REFINE( $\Pi$ ,  $f$ ) // in  $O(lm)$ 
    $f \leftarrow$  UPDATESM( $\Pi$ ,  $f$ ) // in  $O(n)$ 
8: until  $\Pi' = \Pi$ 
   return minimized wta // in  $O(lm)$ 
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Complete Algorithm

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   return minimized wta // in  $O(lm)$ 
```

Notes

- ▶ Algorithm runs in $O(lmn)$
- ▶ Returns equivalent minimal deterministic wta

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Experiments

State Count

Original	Minimal	Reduction to
98	68	69%
394	308	78%
497	381	77%
727	515	71%
2701	1993	74%
3686	1766	48%

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State & Transition Count

Error	Original	Minimal	Reduction to
10^{-4}	(727, 6485)	(629, 6131)	(87%, 95%)
10^{-2}	(727, 6485)	(525, 3425)	(72%, 53%)

References

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