Behavioural differential equations and coinduction for binary trees

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Motivation

- Previous work on streams and formal power series
 - Behavioural differential equations: a coinductive calculus of streams, automata, and power series
 - Elements of stream calculus (an extensive exercise in coinduction) showed that coinduction and behavioural differential equations are useful for building a calculus
- We want to investigate if the same approach is useful for other infinite structures, e.g. infinite binary trees

What will we show?

We will show how to...

- ... define infinite binary trees through behavioural differential equations
- ... develop a calculus for binary trees à la formal power series
- ... calculate closed expressions for infinite binary trees
- ... use closed expressions to prove properties about infinite binary trees

Infinite binary trees

$$T_{A} = \{\sigma : \{L, R\}^* \to A\}$$

Recall: A formal power series is a function $\sigma: X^* \to A$ where X is the set of variables (or input symbols) and A is a semiring.

For *A* semiring, the set T_A coincides with the formal power series over $X = 2 = \{L, R\}$.

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 $(A^{B^*}, < o, t >)$ is the final coalgebra for functor $G(X) = A \times X^B$.

$$< o, t > : A^{B^*} \rightarrow A \times (A^{B^*})^B$$

 $< o, t > (f) = < f(\varepsilon), \lambda b \cdot \lambda w \cdot f(bw) >$

$$(T_A = A^{2^*}, \alpha)$$
 is the final coalgebra of $F(X) = A \times X^2$

$$\alpha(\sigma) = \langle \sigma(\varepsilon), \sigma_I, \sigma_B \rangle$$

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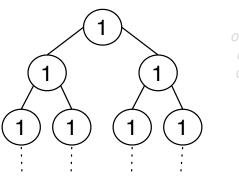
 (T_A, α) final coalgebra: every tree $\sigma \in T_A$ is uniquely determined by defining $\sigma(\varepsilon)$, σ_L and σ_R .

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 $\sigma \in T_A$ will be defined by means of **behavioural differential** equations:

$$\sigma(\varepsilon) = c$$
 initial value
 $\sigma_L = left_exp$ left derivative
 $\sigma_R = right exp$ right derivative

Example I

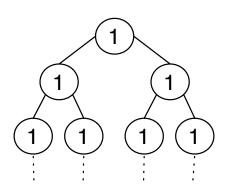


```
ones(\varepsilon) = 1

ones_L = ones

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Behavioural Differential Equations – format

In which conditions a system of behavioural differential equations defining a tree $\sigma = f(\sigma_1, \dots, \sigma_n)$

$$f(\sigma_1, \ldots, \sigma_n)(\varepsilon) = c$$
 initial value
 $(f(\sigma_1, \ldots, \sigma_n))_L = left_exp$ left derivative
 $(f(\sigma_1, \ldots, \sigma_n))_R = right_exp$ right derivative

is well formed?

We proved that i

- ① c is calculated only involving $\sigma_1(\varepsilon), \ldots, \sigma_n(\varepsilon)$
- eleft_exp and right_exp only depend on $\sigma_1, \ldots, \sigma_n, (\sigma_1)_L, \ldots, (\sigma_n)_L, (\sigma_1)_R, \ldots, (\sigma_n)_R$ and function symbols the system is well-formed, *i.e.* uniquely determines $\sigma = f(\sigma_1, \ldots, \sigma_n)$.

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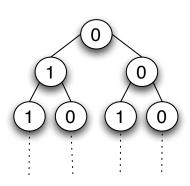
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We proved that if:

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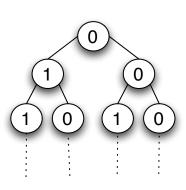
the system is well-formed, *i.e.* uniquely determines $\sigma = f(\sigma_1, \dots, \sigma_n)$.

Examples II



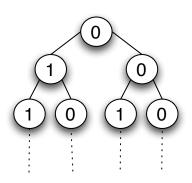
$$\begin{array}{rcl}
\sigma(\varepsilon) & = & 0 \\
\sigma_L & = & \sigma + [1] \\
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\end{array}$$

Examples II



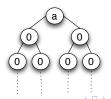
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Examples II



$$\begin{array}{rcl}
\sigma(\varepsilon) & = & 0 \\
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\end{array}$$

[a] denotes



Examples III – The Thue-Morse sequence

- Obtained from the parities of the counts of 1's in the binary representation of non negative integers.
- 0,1,1,0,1,0,0,1,...
- Can be obtained by the substitution map $\{0 \rightarrow 01; 1 \rightarrow 10\}$:

$$0 \rightarrow 01 \rightarrow 0110 \rightarrow 01101001 \rightarrow \dots$$

• Tree representation (at level k, we have 2^k digits)

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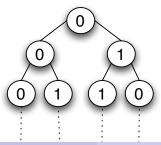
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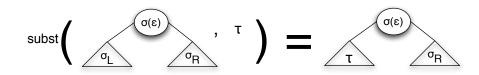
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• Tree representation (at level k, we have 2^k digits)



$$\begin{array}{lll} \mathit{thue}(\varepsilon) & = & 0 \\ \mathit{thue}_{L} & = & \mathit{thue} \\ \mathit{thue}_{R} & = & \mathit{thue} + \mathit{ones} \end{array}$$

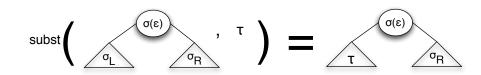
Examples IV – Substitution operation



$$(subst(\sigma, \tau))(\varepsilon) = \sigma(\varepsilon)$$

 $(subst(\sigma, \tau))_L = \tau$
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Operations on trees

From formal power series we inherit several definitions of operations:

Name	Sum	Product
Initial value	$(\sigma+\tau)(\varepsilon)=\sigma(\varepsilon)+\tau(\varepsilon)$	$(\sigma imes au)(arepsilon) = \sigma(arepsilon) imes au(arepsilon)$
Left der.	$(\sigma + \tau)_{L} = \sigma_{L} + \tau_{L}$	$(\sigma \times \tau)_{L} = \sigma_{L} \times \tau + \sigma(\varepsilon) \times \tau_{L}$
Right der	$(\sigma + \tau)_R = \sigma_R + \tau_R$	$(\sigma \times \tau)_{R} = \sigma_{R} \times \tau + \sigma(\varepsilon) \times \tau_{R}$

Fundamental Theorem

For all infinite binary trees $\sigma \in T_A$:

$$\sigma = \sigma(\varepsilon) + L \times \sigma_L + R \times \sigma_R$$

where

$$egin{aligned} L(arepsilon) &= 0 & R(arepsilon) &= 0 \ L_L &= [1] & R_L &= [0] \ L_R &= [0] & R_R &= [1] \end{aligned}$$

Fundamental Theorem

For all infinite binary trees $\sigma \in T_A$:

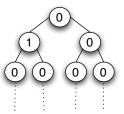
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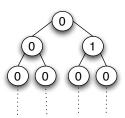
where

$$L(\varepsilon) = 0$$

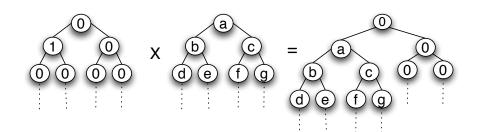
 $L_L = [1]$
 $L_R = [0]$





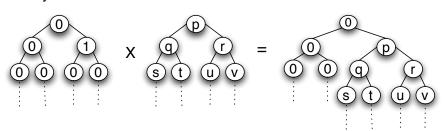


$L \times \sigma_L$

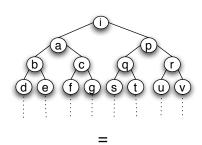


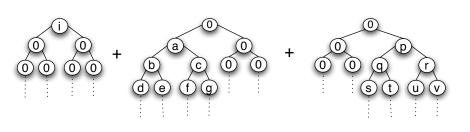


Similarly:



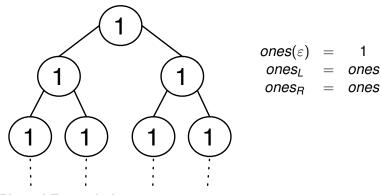
$$\sigma = \sigma(\varepsilon) + L \times \sigma_L + R \times \sigma_R$$



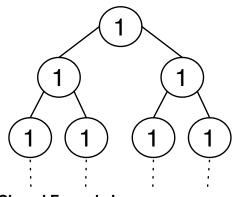


But... What can we do with this theorem?





Closed Formula I



$$ones(\varepsilon) = 1$$

 $ones_L = ones$
 $ones_R = ones$

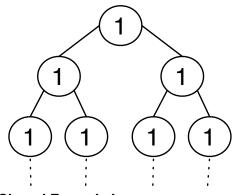
Closed Formula I

$$ones = 1 + L \times ones + R \times ones$$

$$\Leftrightarrow (1 - L - R) \times ones = 1$$

$$\Leftrightarrow ones = \frac{1}{4 + R}$$

ones = $\frac{1}{1 - L - R}$ $\frac{x}{y}$ stands for $y^{-1} \times x$



$$ones(\varepsilon) = 1$$

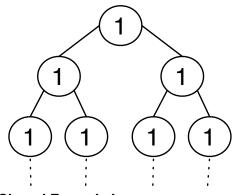
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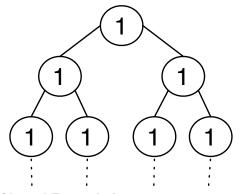
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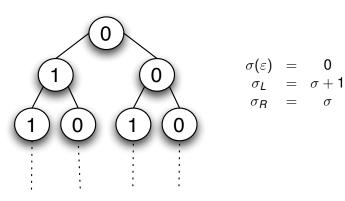
Inverse operation

The inverse of a tree – σ^{-1} – is defined formally so that $\sigma \times \sigma^{-1} = 1$.

$$\sigma^{-1}(\varepsilon) = (\sigma(\varepsilon))^{-1}$$

$$(\sigma^{-1})_L = (\sigma(\varepsilon))^{-1} \times \sigma_L \times \sigma^{-1}$$

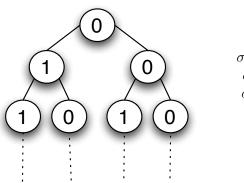
$$(\sigma^{-1})_R = (\sigma(\varepsilon))^{-1} \times \sigma_R \times \sigma^{-1}$$



$$\sigma = 0 + L \times (\sigma + 1) + R \times \sigma$$

$$\Leftrightarrow (1 - L - R)\sigma = L$$

$$\Leftrightarrow \sigma = \frac{L}{1 - L - R}$$

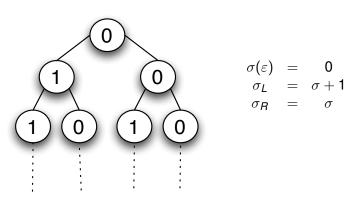


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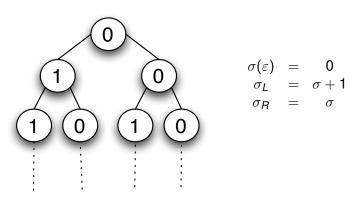
$$\sigma = \frac{L}{1 - L - R}$$



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$$\Leftrightarrow \sigma = \frac{L}{\sigma}$$



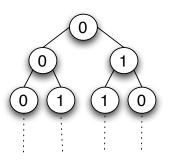
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Examples revisited III – The Thue-Morse sequence



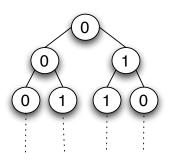
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$$\sigma = 0 + L \times \sigma + R \times (\sigma + ones)$$

$$\Leftrightarrow (1 - L - R)\sigma = R \times \frac{1}{1 - L - R}$$

$$\Leftrightarrow \sigma = \frac{1}{1 - L - R} \times R \times \frac{1}{1 - L - R}$$

Examples revisited III - The Thue-Morse sequence



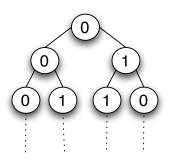
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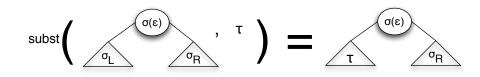
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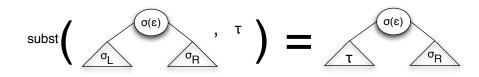
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$$subst(\sigma,\tau) = \sigma(\varepsilon) + L \times \tau + R \times \sigma_{R}$$

$$\Leftrightarrow \qquad \{ \sigma - L \times \sigma_{L} = \sigma(\varepsilon) + R \times \sigma_{R} \}$$

$$subst(\sigma,\tau) = \sigma - L \times (\sigma_{L} - \tau)$$



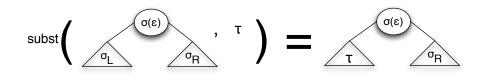
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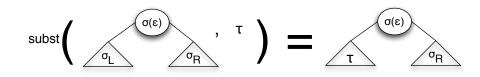
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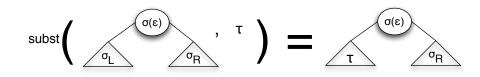
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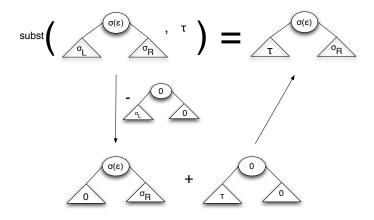
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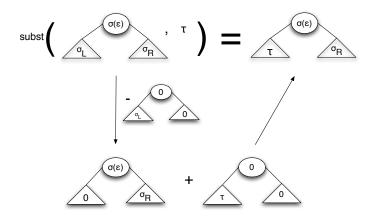
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$subst(\sigma, \tau) = \sigma - L \times (\sigma_L - \tau)$



Easily generalizes: $subst(\sigma, \tau, P) = \sigma - P(\sigma_P - \tau)$

$$subst(\sigma, \tau) = \sigma - L \times (\sigma_L - \tau)$$



Easily generalizes: $subst(\sigma, \tau, P) = \sigma - P(\sigma_P - \tau)$

Proving properties

$$subst(\sigma, \sigma_P, P) = \sigma$$
 and $subst(subst(\sigma, \tau, P), \sigma_P, P) = \sigma$
• Property 1

$$subst(\sigma, \sigma_P, P) = \sigma - P(\sigma_P - \sigma_P) = \sigma$$

Property 2

```
subst(subst(\sigma, \tau, P), \sigma_P, P)
= subst(\sigma - P(\sigma_P - \tau), \sigma_P, P)
Def. of subst
= \sigma - P(\sigma_P - \tau) - P((\sigma - P(\sigma_P - \tau))_P - \sigma_P)
Def. of subst
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= \sigma
```

Proving properties

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```

Proving properties

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 and $subst(subst(\sigma, \tau, P), \sigma_P, P) = \sigma$

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Property 2

$$subst(subst(\sigma, \tau, P), \sigma_P, P)$$

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$$= \sigma - P(\sigma_P - \tau) - P((\sigma - P(\sigma_P - \tau))_P - \sigma_P)$$
 Def. of subst
$$= \sigma - P(\sigma_P - \tau) - P(\tau - \sigma_P)$$

$$= \sigma$$

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- We have defined various non-trivial trees by means of simple differential equations
- We have showed how to compute compact closed formulae for them
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Future work

- Behavioural differential equations are closely related to lazy functional programming implementations.
- In particular, we would like to study the relation between closed expressions and elimination of corecursion
- The closed formula that we have obtained for the Thue-Morse sequence suggests a possible use of coinduction and differential equations in the area of automatic sequences.