

Behavioural differential equations for binary trees

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- Previous work by Jan:

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 effective for stream calculus

- We want to investigate if the same approach is effective for other infinite structures, e.g. **infinite binary trees**

What will we show?

We will show how to...

- ... develop a calculus for binary trees *à la formal power series*
- ... define infinite binary trees through *behavioural differential equations*
- ... calculate closed expressions for infinite binary trees

Formal power series

The set of infinite binary trees – T_A – is the final coalgebra of

$$F(X) = X \times A \times X$$

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

For A semiring, the set T_A is a formal power series over $X=2$ (**Why?**),
i.e.,

$$T_A = \{\sigma \mid \sigma : 2^* \rightarrow A\}$$

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$$T_A = \{\sigma \mid \sigma : 2^* \rightarrow A\}$$

Why?

- Final coalgebra for $G(X) = A \times X^B$ is A^{B^*}
- $F(X) = X \times A \times X \cong A \times X \times X \cong A \times X^2$
- $2 = \{L, R\}$

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- Final coalgebra for $G(X) = A \times X^B$ is A^{B^*}
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- $2 = \{L, R\}$

Behavioural Differential Equations

The formal definition of $f(\sigma_1, \dots, \sigma_n) = \sigma \in T_A$ is now expressed in terms of a *behavioural differential equation*.

$$\begin{aligned}\sigma(\varepsilon) &= c && \text{initial value} \\ \sigma_L &= \text{left_exp} && \text{left derivative} \\ \sigma_R &= \text{right_exp} && \text{right derivative}\end{aligned}$$

- We know that such system has a unique solution if:
 - 1 c is calculated only involving $\sigma_1(\varepsilon), \dots, \sigma_n(\varepsilon)$
 - 2 left_exp only depends on $\sigma_1, \dots, \sigma_n, (\sigma_1)_L, \dots, (\sigma_n)_L$ and constants
 - 3 right_exp only depends on $\sigma_1, \dots, \sigma_n, (\sigma_1)_R, \dots, (\sigma_n)_R$ and constants

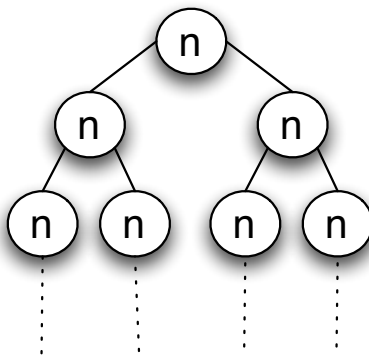
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Examples I

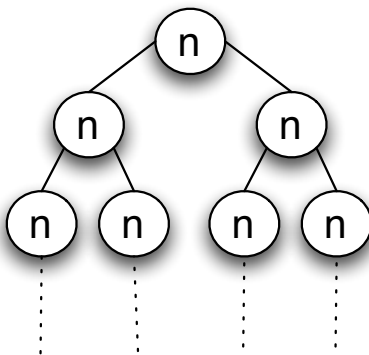


$$\sigma(\varepsilon) = n$$

$$\sigma_L = \sigma$$

$$\sigma_R = \sigma$$

Examples I

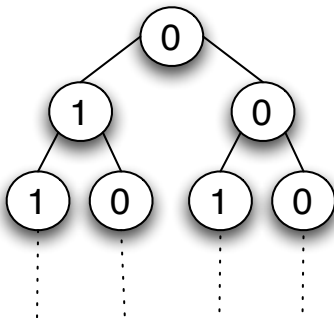


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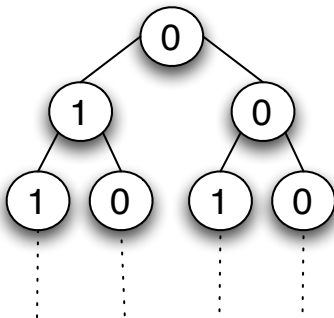
$$\sigma_R = \sigma$$

Examples II



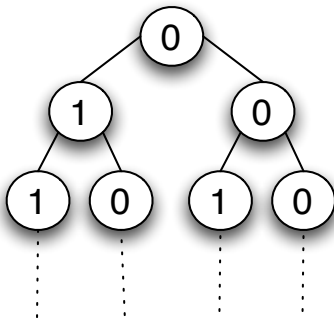
$$\begin{aligned}\sigma(\varepsilon) &= 0 \\ \sigma_L &= \sigma + [1] \\ \sigma_R &= \sigma\end{aligned}$$

Examples II



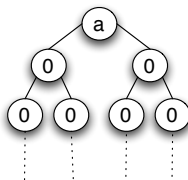
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$[a]$ denotes



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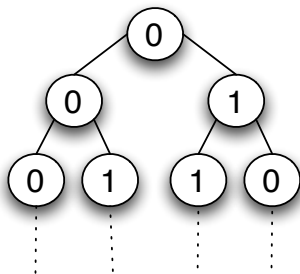
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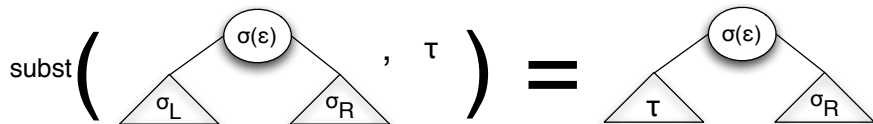
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- Tree representation (at level k , we have 2^k digits)



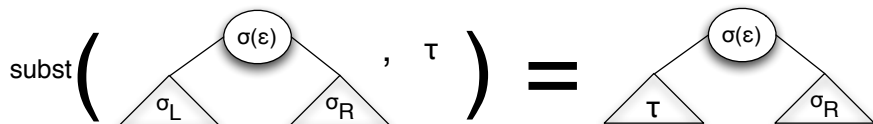
$$\begin{aligned}\sigma(\varepsilon) &= 0 \\ \sigma_L &= \sigma \\ \sigma_R &= \sigma + \text{repeat}(1)\end{aligned}$$

Examples IV – Substitution operation



$$\begin{aligned}(\text{subst}(\sigma, \tau))(\varepsilon) &= \sigma(\varepsilon) \\(\text{subst}(\sigma, \tau))_L &= \tau \\(\text{subst}(\sigma, \tau))_R &= \sigma_R\end{aligned}$$

Examples IV – Substitution operation



$$(\text{subst}(\sigma, \tau))(\varepsilon) = \sigma(\varepsilon)$$

$$(\text{subst}(\sigma, \tau))_L = \tau$$

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Operations on trees

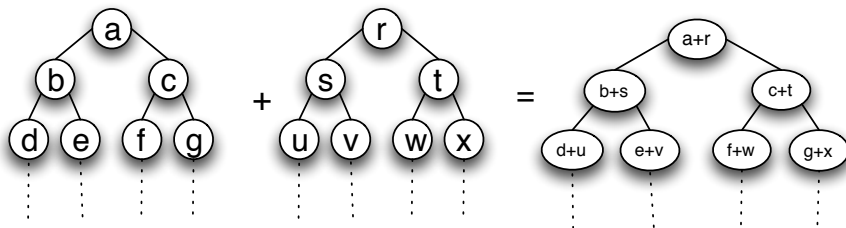
From formal power series we inherit several definitions of operations:

Name	Sum	Product
Initial value	$(\sigma + \tau)(\varepsilon) = \sigma(\varepsilon) + \tau(\varepsilon)$	$(\sigma \times \tau)(\varepsilon) = \sigma(\varepsilon) \times \tau(\varepsilon)$
Left der.	$(\sigma + \tau)_L = \sigma_L + \tau_L$	$(\sigma \times \tau)_L = \sigma_L \times \tau + \sigma(\varepsilon) \times \tau_L$
Right der	$(\sigma + \tau)_R = \sigma_R + \tau_R$	$(\sigma \times \tau)_R = \sigma_R \times \tau + \sigma(\varepsilon) \times \tau_R$

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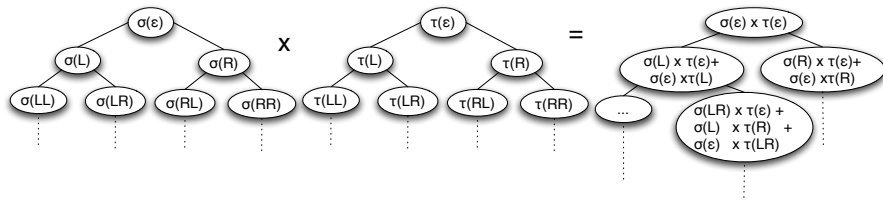
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$$(\sigma \times \tau)(w) = \sum_{w=uv} \sigma(u)\tau(v)$$

Fundamental Theorem

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

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

where

$$L(\varepsilon) = 0 \quad R(\varepsilon) = 0$$

$$L_L = [1] \quad R_L = [0]$$

$$L_R = [0] \quad R_R = [1]$$

Fundamental Theorem

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

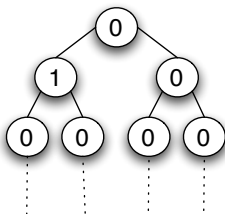
$$\sigma = \sigma(\varepsilon) + L \times \sigma_L + R \times \sigma_R$$

where

$$L(\varepsilon) = 0$$

$$L_L = [1]$$

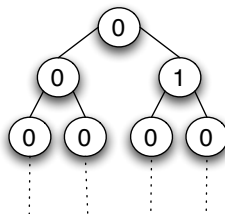
$$L_R = [0]$$



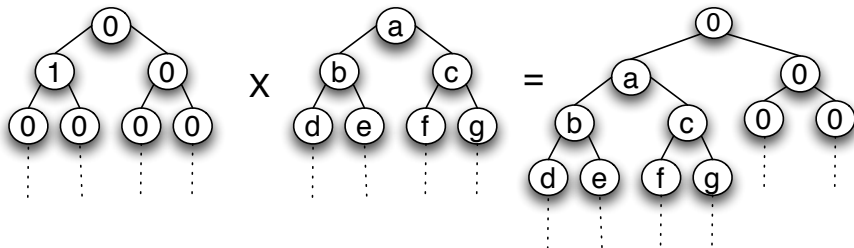
$$R(\varepsilon) = 0$$

$$R_L = [0]$$

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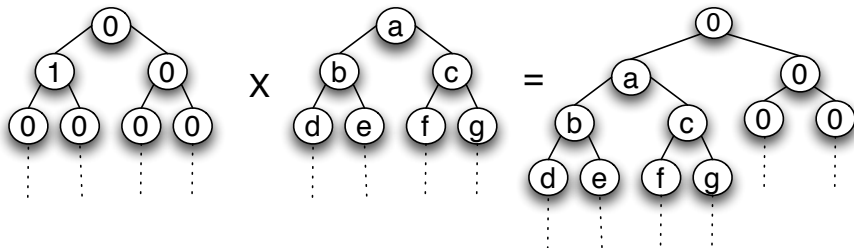
$$L \times \sigma_L$$



Why?

$$\begin{aligned}
 (L \times \sigma)(\varepsilon) &= L(\varepsilon) \times \sigma(\varepsilon) = 0 \\
 (L \times \sigma)_L &= L_L \times \sigma + [L(\varepsilon)] \times \sigma_L \\
 &= [1] \times \sigma + [0] \times \sigma_L = \sigma \\
 (L \times \sigma)_R &= L_R \times \sigma + [L(\varepsilon)] \times \sigma_R \\
 &= [0] \times \sigma + [0] \times \sigma_R = [0]
 \end{aligned}$$

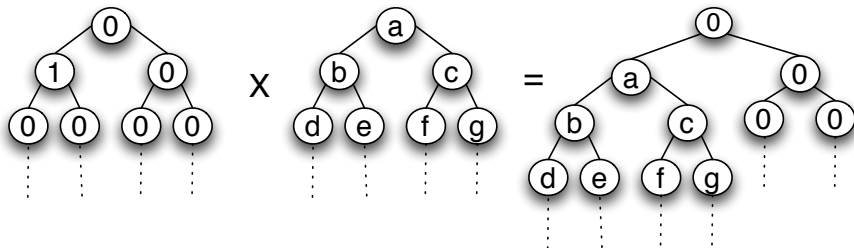
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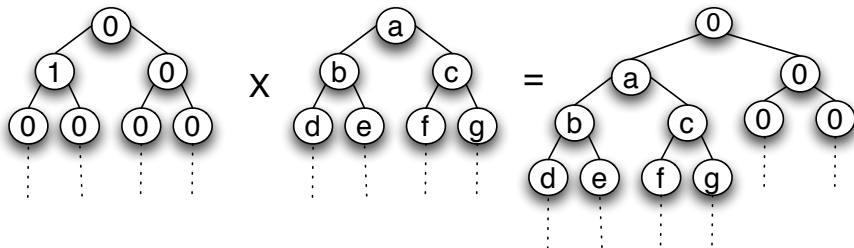
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$$L \times \sigma_L$$

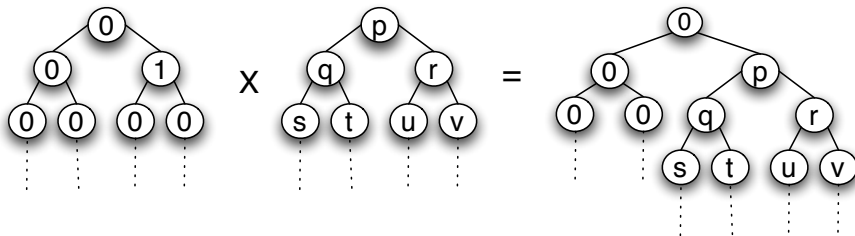


Why?

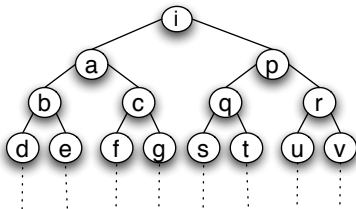
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 (L \times \sigma)_L &= L_L \times \sigma + [L(\varepsilon)] \times \sigma_L \\
 &= [1] \times \sigma + [0] \times \sigma_L = \sigma \\
 (L \times \sigma)_R &= L_R \times \sigma + [L(\varepsilon)] \times \sigma_R \\
 &= [0] \times \sigma + [0] \times \sigma_R = [0]
 \end{aligned}$$

$$R \times \sigma_R$$

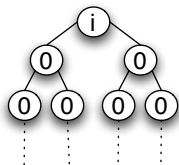
Similarly:



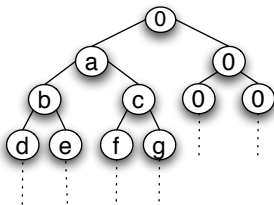
$$\sigma = \sigma(\varepsilon) + L \times \sigma_L + R \times \sigma_R$$



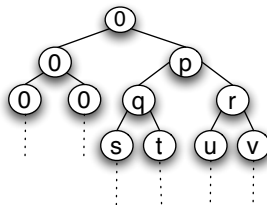
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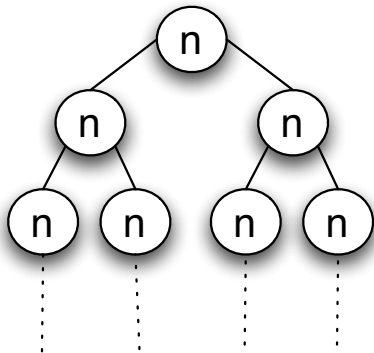
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But... What can we do with this theorem?



Examples Revisited I



$$\sigma(\varepsilon) = n$$

$$\sigma_L = \sigma$$

$$\sigma_R = \sigma$$

Closed Formula I

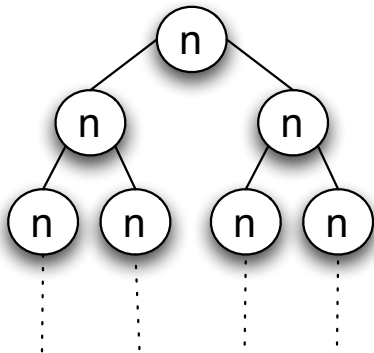
$$\sigma = [n] + L \times \sigma + R \times \sigma$$

$$\Leftrightarrow (1 - L - R)\sigma = [n]$$

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

$\frac{x}{y}$ stands for $y^{-1} \times x$

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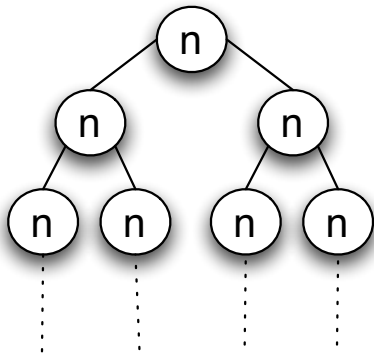
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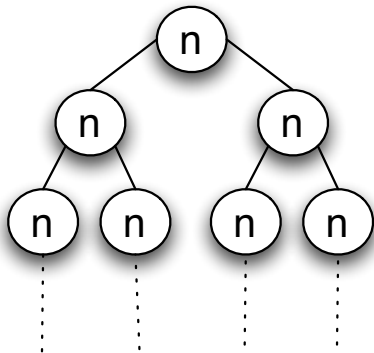
$$\Leftrightarrow (1 - L - R)\sigma = [n]$$

$$\Leftrightarrow$$

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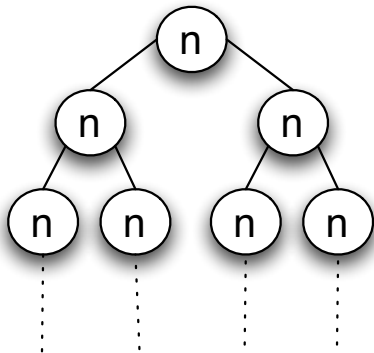
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$\frac{x}{v}$ stands for $y^{-1} \times x$

Inverse operation

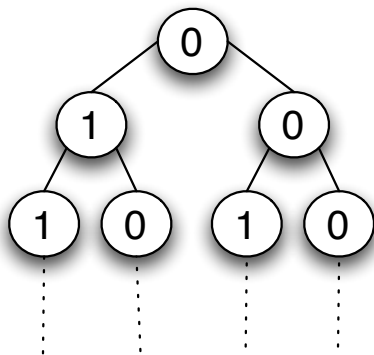
The inverse of a tree – σ^{-1} – is defined formally so that $\sigma \times \sigma^{-1} = 1$.

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

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

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

Examples Revisited II

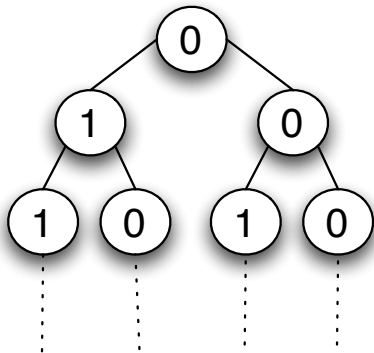


$$\begin{aligned}\sigma(\varepsilon) &= 0 \\ \sigma_L &= \sigma + [1] \\ \sigma_R &= \sigma\end{aligned}$$

Closed Formula II

$$\begin{aligned}\sigma &= 0 + L \times (\sigma + 1) + R \times \sigma \\ \Leftrightarrow (1 - L - R)\sigma &= L \\ \Leftrightarrow \sigma &= \frac{L}{1 - L - R}\end{aligned}$$

Examples Revisited II



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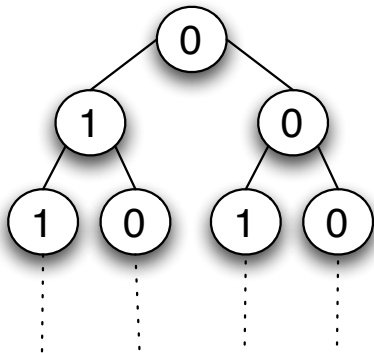
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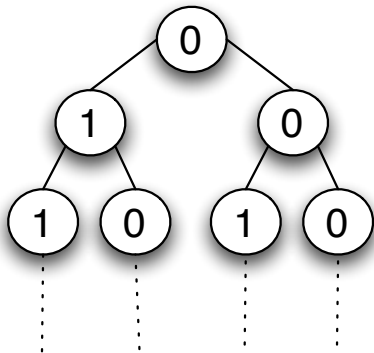


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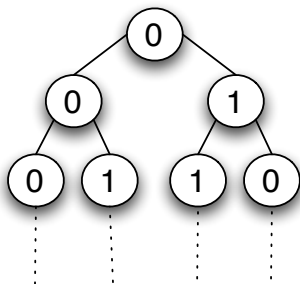


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Examples revisited III – The Thue-Morse sequence

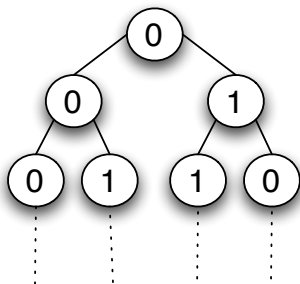


$$\begin{aligned}\sigma(\varepsilon) &= 0 \\ \sigma_L &= \sigma \\ \sigma_R &= \sigma + \textit{repeat}(1)\end{aligned}$$

Closed Formula III

$$\begin{aligned}\sigma &= 0 + L \times \sigma + R \times (\sigma + \textit{repeat}(1)) \\ \Leftrightarrow (1 - L - R)\sigma &= R \times \frac{1}{1 - L - R} \\ \Leftrightarrow \sigma &= \frac{1}{1 - L - R} \times R \times \frac{1}{(1 - L - R)}\end{aligned}$$

Examples revisited III – The Thue-Morse sequence

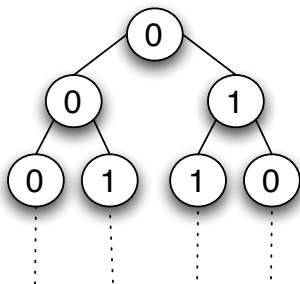


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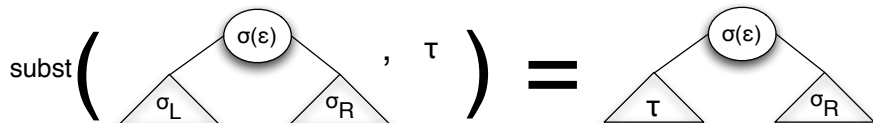


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Examples revisited IV – Substitution

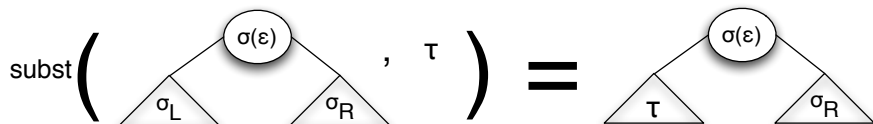


$$\begin{aligned}(\text{subst}(\sigma, \tau))(\varepsilon) &= \sigma(\varepsilon) \\ (\text{subst}(\sigma, \tau))_L &= \tau \\ (\text{subst}(\sigma, \tau))_R &= \sigma_R\end{aligned}$$

Closed Formula IV

$$\begin{aligned}\text{subst}(\sigma, \tau) &= \sigma(\varepsilon) + L \times \tau + R \times \sigma_R \\ \Leftrightarrow \quad \{ \sigma - L \times \sigma_L &= \sigma(\varepsilon) + R \times \sigma_R \} \\ \text{subst}(\sigma, \tau) &= \sigma - L \times (\sigma_L - \tau)\end{aligned}$$

Examples revisited IV – Substitution

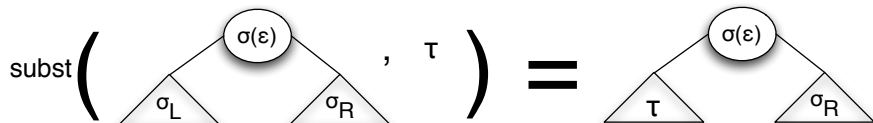


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Examples revisited IV – Substitution



$$(\text{subst}(\sigma, \tau))(\varepsilon) = \sigma(\varepsilon)$$

$$(\text{subst}(\sigma, \tau))_L = \tau$$

$$(\text{subst}(\sigma, \tau))_R = \sigma_R$$

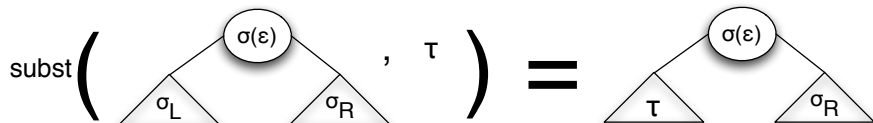
Closed Formula IV

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

$$\Leftrightarrow \{ \sigma - L \times \sigma_L = \sigma(\varepsilon) + R \times \sigma_R \}$$

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

Examples revisited IV – Substitution

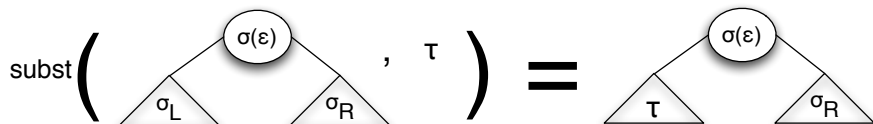


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Examples revisited IV – Substitution

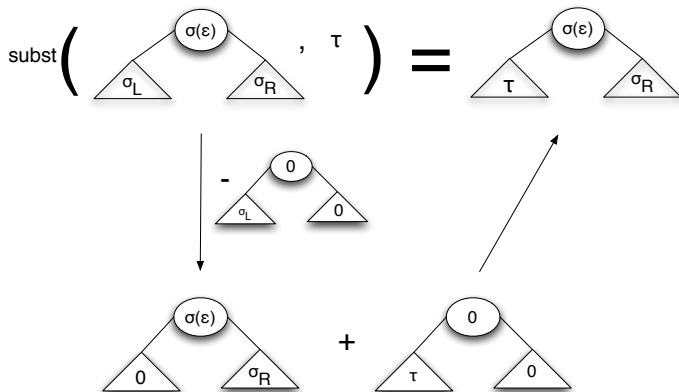


$$\begin{aligned}(\text{subst}(\sigma, \tau))(\varepsilon) &= \sigma(\varepsilon) \\ (\text{subst}(\sigma, \tau))_L &= \tau \\ (\text{subst}(\sigma, \tau))_R &= \sigma_R\end{aligned}$$

Closed Formula IV

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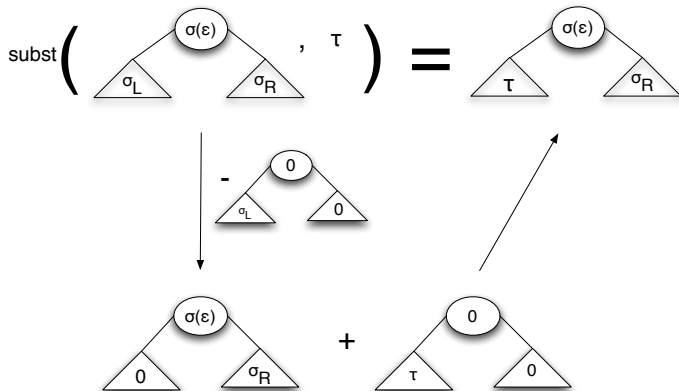
$$\text{subst}(\sigma, \tau) = \sigma - L \times (\sigma_L - \tau)$$



Easily generalizes:

$$\text{subst}(\sigma, \tau, \text{Path}) = \sigma - \text{Path}(\sigma_{\text{Path}} - \tau)$$

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



Easily generalizes:

$$\text{subst}(\sigma, \tau, \text{Path}) = \sigma - \text{Path}(\sigma_{\text{Path}} - \tau)$$

Conclusions

- Behavioural differential equations are effective to represent infinite binary trees.
- Closed expressions constitute a compact representation of trees (only involving constants) and...
- Give a recipe to implement algorithms.

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
- We would also like to understand better the class of rational trees