

Model-checking hierarchical graphs

Markus Lohrey
University of Stuttgart

28. Juli 2005

Hierarchical graph definitions (HGDs) I

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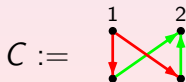
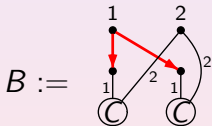
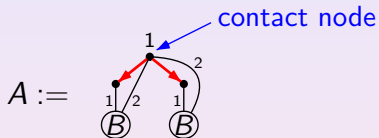
An HGD D



Macros: $S \succ A \succ B \succ C$

$\text{rank}(S)=0, \text{rank}(A)=1$

$\text{rank}(B) = \text{rank}(C) = 2$



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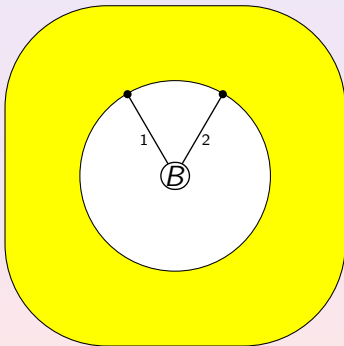
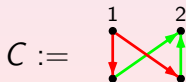
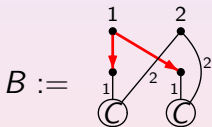
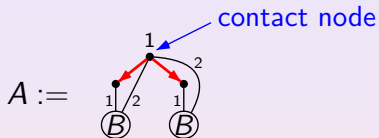
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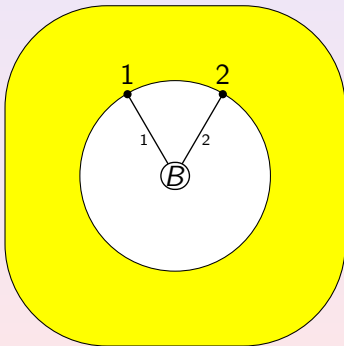
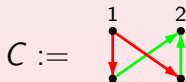
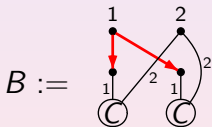
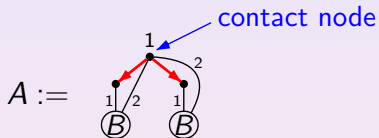
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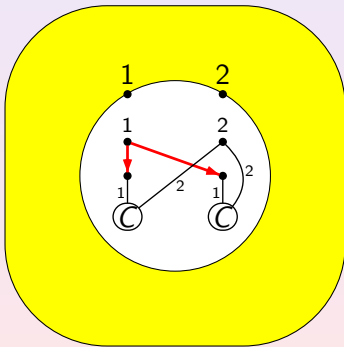
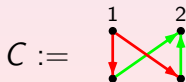
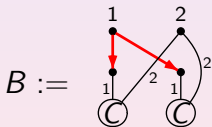
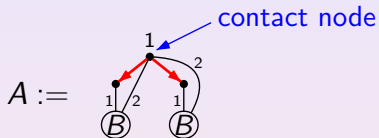
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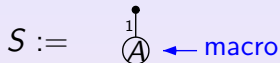
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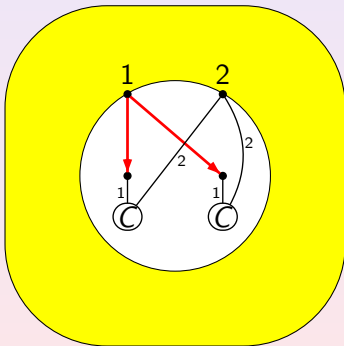
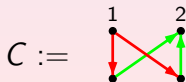
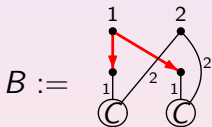
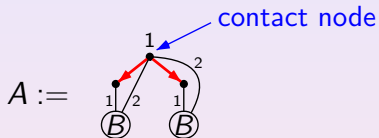
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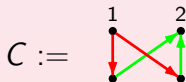
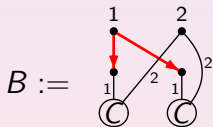
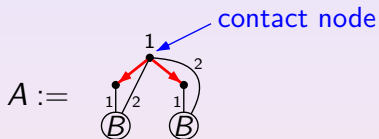
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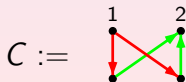
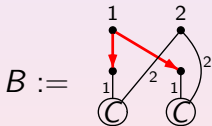
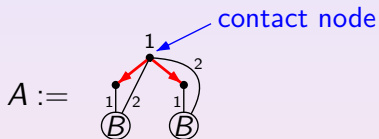
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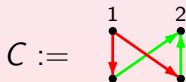
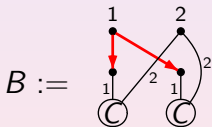
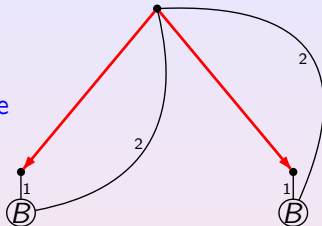
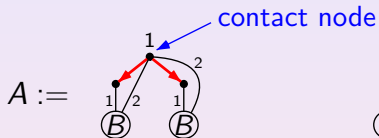
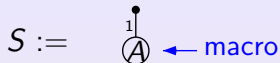
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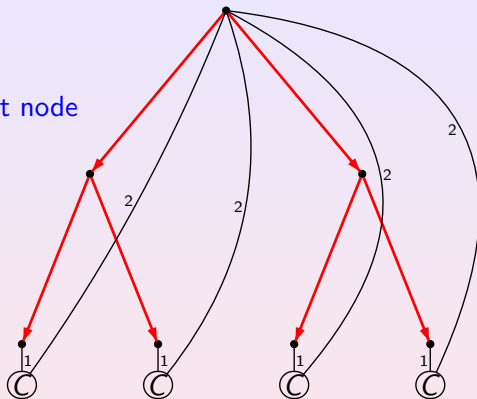
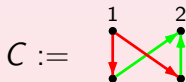
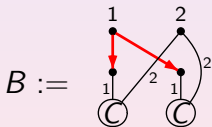
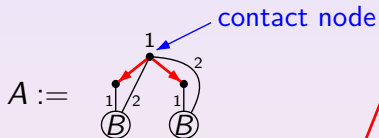
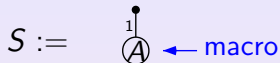
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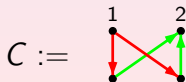
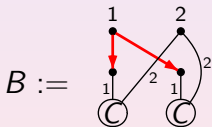
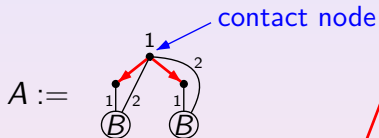
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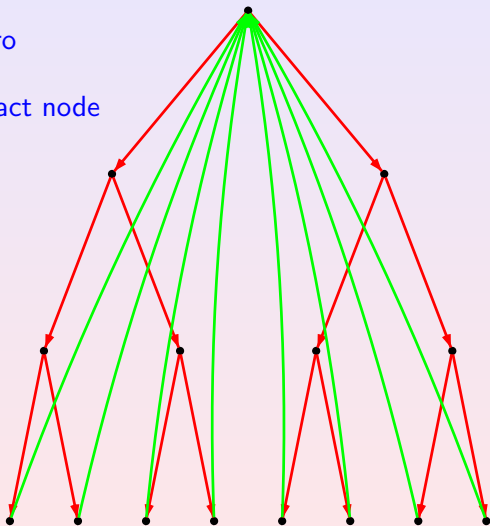


Hierarchical graph definitions (HGDs) I

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$\text{unfold}(D)$



Hierarchical graph definitions (HGDs) II

For an HGD D , $\text{unfold}(D)$ denotes the generated graph.

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Data compression (succinctness): $\text{unfold}(D)$ may be exponentially larger than D .

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Consequence: **Computational problems tend to be more difficult, when input structures are given hierarchical!**

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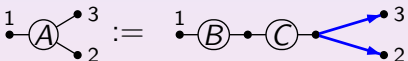
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W.l.o.g. at most two references in each right-hand side (Chomsky normal form)

Hierarchical graph definitions (HGDs) III

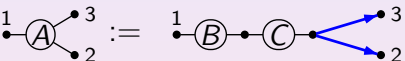
Hierarchical graph definitions (HGDs) III

A HGD D is **non-passing** if macros are not connected to contact nodes in right-hand sides.

E.g. **not** allowed:  $1 \text{---} (A) \begin{matrix} \bullet 3 \\ \bullet 2 \end{matrix} := 1 \text{---} (B) \text{---} (C) \begin{matrix} \bullet 3 \\ \bullet 2 \end{matrix}$

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A HGD D is **c -bounded** for $c \in \mathbb{N}$ if the rank of every macro is at most c .

Previous work:

- Lengauer, Wanke (around 1990): Specific algorithmic problems on hierarchically defined graphs.
- Marathe, Hunt, Stearns, Radhakrishnan (around 1995): Approximation algorithms for hierarchically defined graphs.
- Alur, Yannakakis 2001: Model-checking temporal logics for hierarchical state machines.
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History and related formalisms

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Here: Model-Checking of **first-order logic (FO)** and **monadic second-order logic (MSO)** on hierarchically defined input graphs.

First-order logic (FO): quantification over elements of a structure

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Example $((V, E)$ is a directed graph):

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Monadic second-order logic (MSO): quantification over elements and subsets of a structure

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$$\exists x \in V \exists y \in V \exists X \subseteq V : x \in X \wedge y \notin X \wedge \\ \forall u \in V \forall v \in V : u \in X \wedge E(u, v) \Rightarrow v \in X$$

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INPUT: Formula $\varphi \in \mathcal{L}$, finite structure \mathcal{A}

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A Σ_k -MSO formula is an MSO-formula of the form

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Some complexity classes

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Complexity of FO

data complexity	non-hierarchical	apex HGD	c-bounded HGD	general HGD
Σ_k -FO	AC_k^0	NL	NL \dots P	NL ($k = 1$) $\Sigma_{k-1}^{\text{poly}}$ ($k > 1$)

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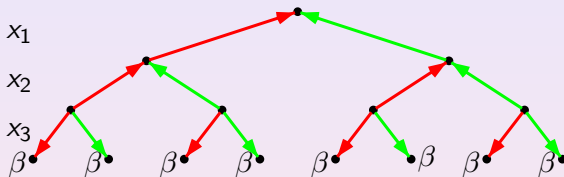
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Data complexity of Σ_2 -FO is NP-hard

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Reduction from the NP-hard problem 3-SAT

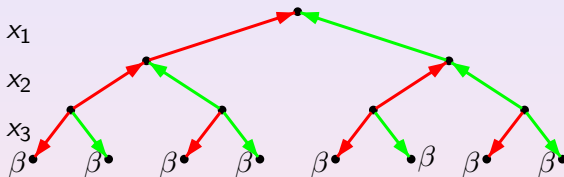
Example: $(x_1 \vee x_2 \vee x_3) \wedge (\neg x_1 \vee \neg x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2 \vee x_3)$



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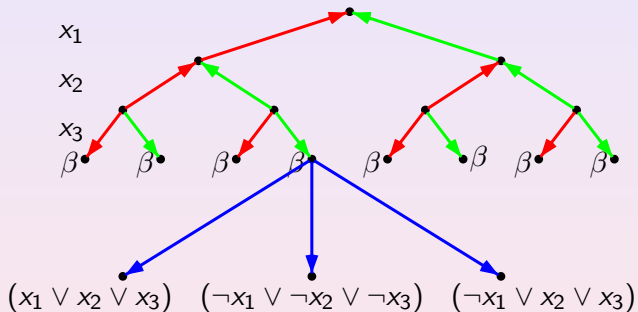


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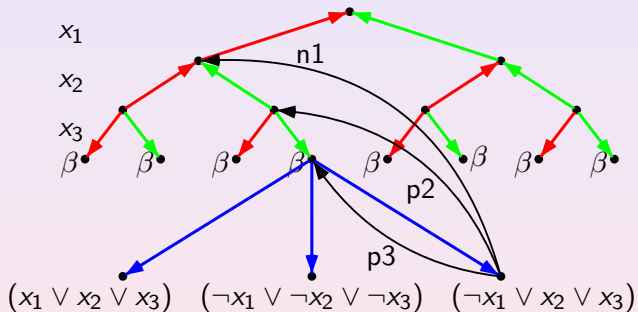


$\exists x : \beta(x) \wedge \forall y : x \longrightarrow y \Rightarrow \dots$

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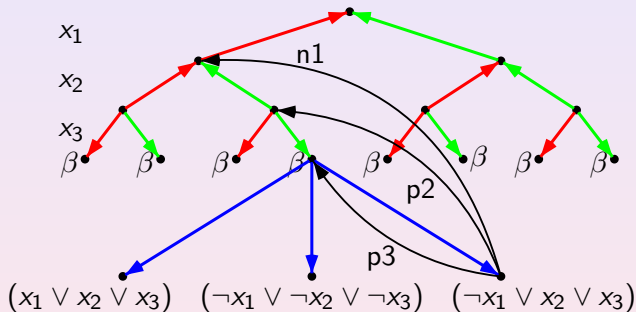


$$\exists x : \beta(x) \wedge \forall y : x \xrightarrow{\text{blue}} y \Rightarrow \bigvee_{i=1}^3 \exists z, z' : y \xrightarrow{p_i} z \xrightarrow{\text{green}} z' \vee y \xrightarrow{n_i} z \xrightarrow{\text{red}} z'$$

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Remark: The apex restriction does not help in the context of MSO.

- Close the gap between NL and P for the data complexity of FO for c -bounded HGDs.

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- Under which conditions is hierarchical model-checking for a logic harder than “flat” model-checking?
- When can hierarchy be exploited?