



Where Myhill–Nerode Theorem Meets Parameterized Algorithmics

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1 Decomposing the Input and running Dynamic Algorithms

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- Explicit comb. extensions of this concept appeared e.g. in the works [Abrahamson and Fellows, 93], [PH, 03], or [Ganian and PH, 08].

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- Informally, the classes of $\approx_{\mathcal{P},k}$ capture **all information** about the property \mathcal{P} that can “cross” our boundary of size k
(regardless of actual meaning of “boundary” and “join”).

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- For simplicity, solution fragments φ can be “**embedded**” in \mathcal{U}_k and \otimes .
- Can, e.g., count the solutions in **each class of $\approx_{\mathcal{P},k}$** , or keep an opt. one.

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- **XP algorithms**, i.e. getting away from finite automata?
 - yes, still works quite nicely, cf. [Ganian, PH, Obdržálek, 09].
 - brings new application issues such as “quantification inside \otimes ” (cf. sol. fragments), or a “second-level” congruence on top of $\approx_{\mathcal{P},k}$.

Parse trees of decompositions

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Parse trees of decompositions

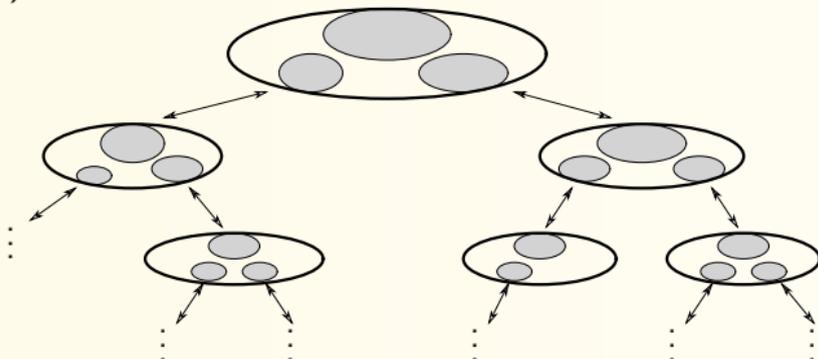
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- This can be (visually) seen as. . .



3 Measuring Graphs: Clique-width and Rank-width

Motivation: Trees are easy to understand and to handle, so how “tree-like” our graph is in some well-defined sense (the **width**)?

- A topic occurring both in pure theory (e.g. Graph Minors), and in algorithms (Fixed parameter tractability).

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- giving the *expression tree* (parse tree) for clique-width.

Rank-decomposition

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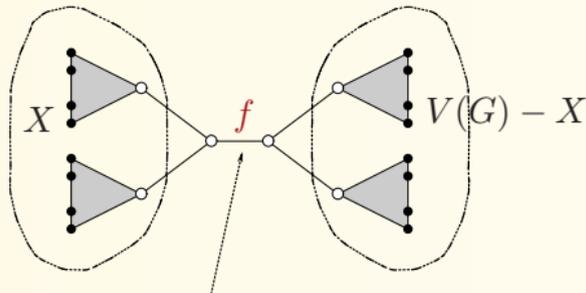
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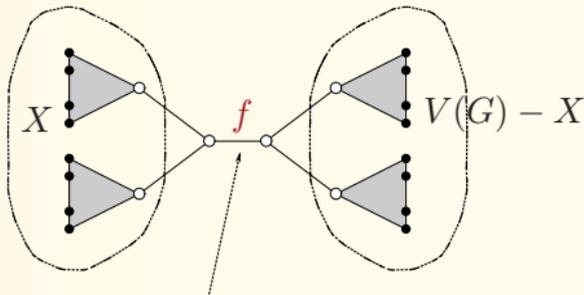
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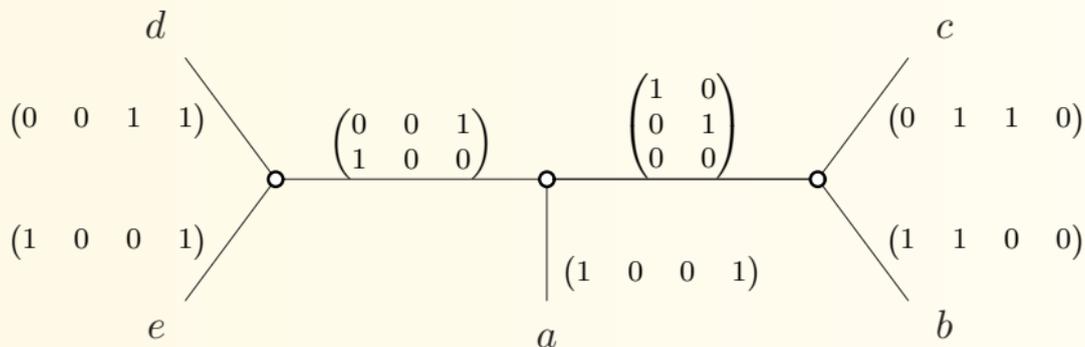
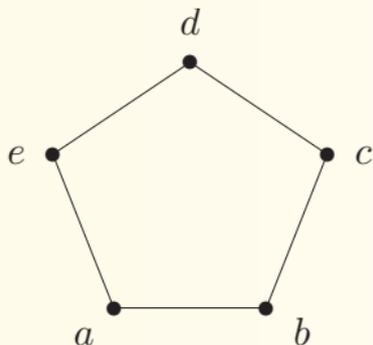
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- Rank-width** = $\min_{\text{rank-decs. of } G} \max \{ \text{width}(f) : f \text{ tree edge} \}$

An example. Cycle C_5 and its *rank-decomposition* of width 2:



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- [Oum and PH, 07] There is an *FPT algorithm* for computing an optimal width- t rank-decomposition of a graph in time $O(f(t) \cdot n^3)$.
- And *new results* show that certain algorithms designed on rank-decompositions run faster than their analogues designed on clique-width expressions. . . (subst. *poly(t)* in place of *cw*, instead of 2^t)

Parse trees for rank-decompositions

Unlike for tree- or clique- decompositions with obvious parse trees, what is the “**boundary**” and “**join**” operation for rank-width?

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- Join \rightarrow a **composition** operator with relabelings f_1, f_2 ;
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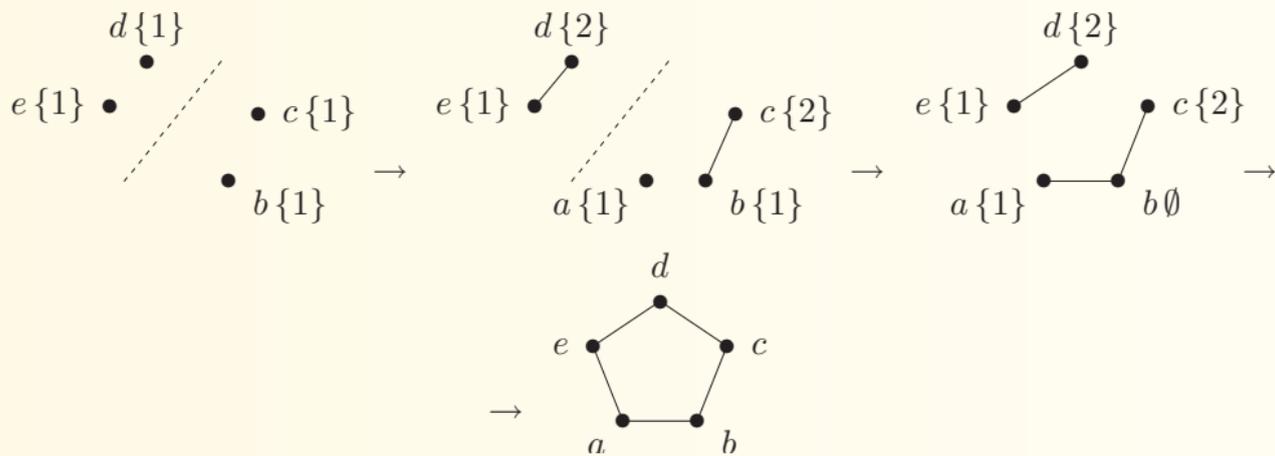
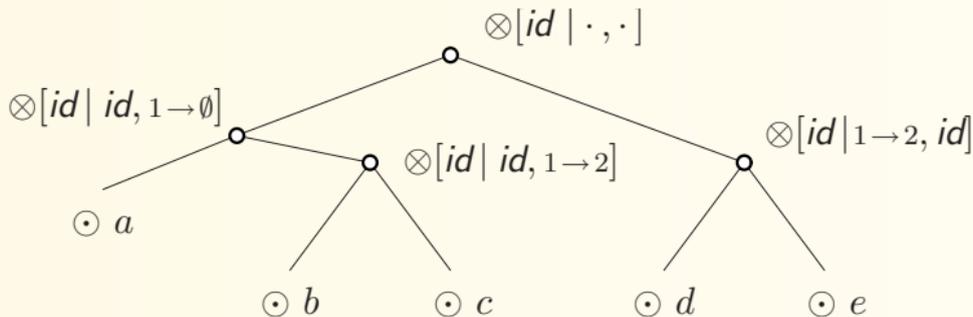
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- Independently considered related notion of **R_t -join** decompositions by [Bui-Xuan, Telle, and Vatshelle, 08].

A parse tree. An example generating the cycle C_5 (of rank-width 2):



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A resulting **double-exponential** worst-case dependency on a width estimate!

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Our answer – considering *rank-width*:

- **No loss** in the promised width, and yet **single-exponential** in it.
- A clear and rigorous algorithm employing many of the above tricks.

Theorem. [Ganian, PH, Obdržálek, 10] #SAT solved in FPT time

$$\mathcal{O}(t^3 \cdot 2^{3t(t+1)/2} \cdot |\phi|)$$

where t is the *signed rank-width* of the input instance (CNF formula) ϕ .

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Then

$$G_1 \oplus G_2 = (G_1^+ \oplus G_2^+) \cup (G_1^- \oplus G_2^-)$$

and the same decomposition is used.

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Easy to prove. . . , but does it help?

Subsets of labels from $2^{\{1,2,\dots,t\}}$ $\longrightarrow \Omega(2^{2^t})$ classes!

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Conclusion. Breaking the satisfying assignments of ϕ into $S(t)^4$ classes, and processing a node of the parse tree in $O^*(S(t)^6)$. □

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