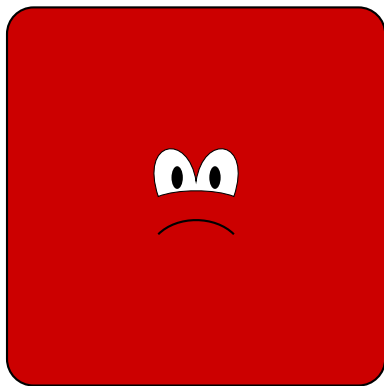


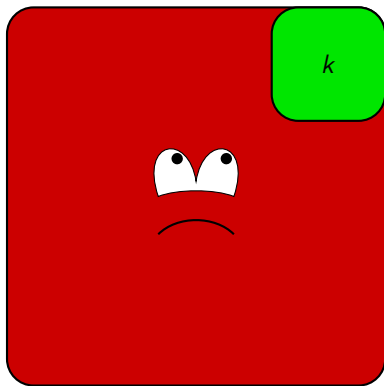
Kernel lower bounds using co-nondeterminism: Finding induced hereditary subgraphs

Stefan Kratsch, Marcin Pilipczuk, Ashutosh Rai, Venkatesh Raman

SWAT 2012, Helsinki

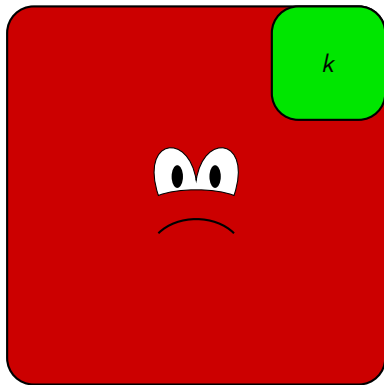


instance of NP-hard problem



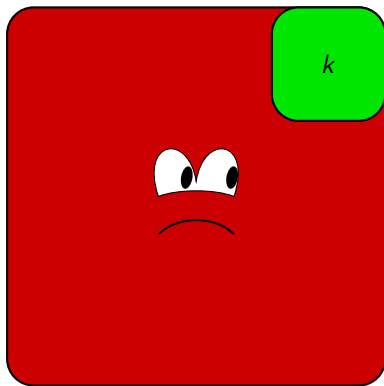
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Kernelization



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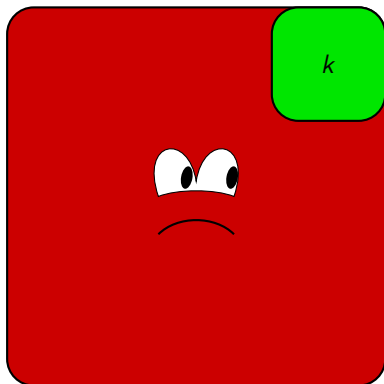
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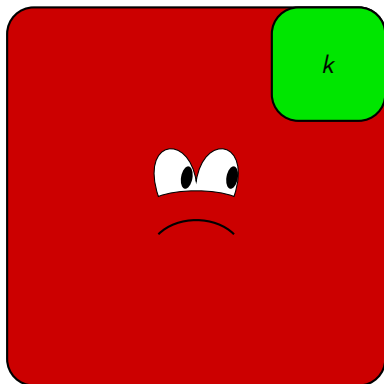
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Kernelization



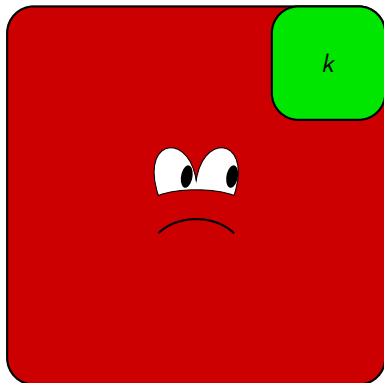
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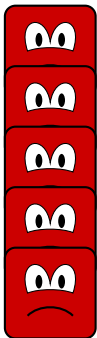
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Kernelization lower bounds

[Bodlaender, Downey, Fellows, Hermelin, ICALP'08]

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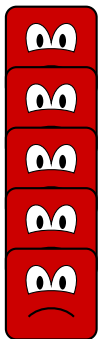
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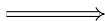
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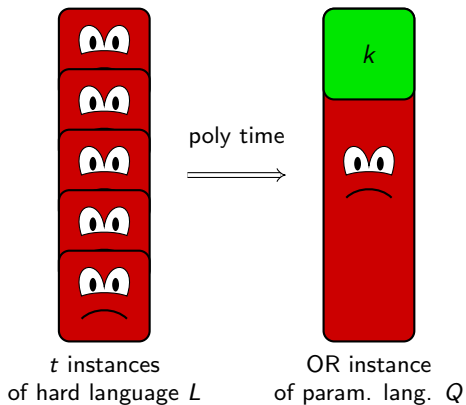
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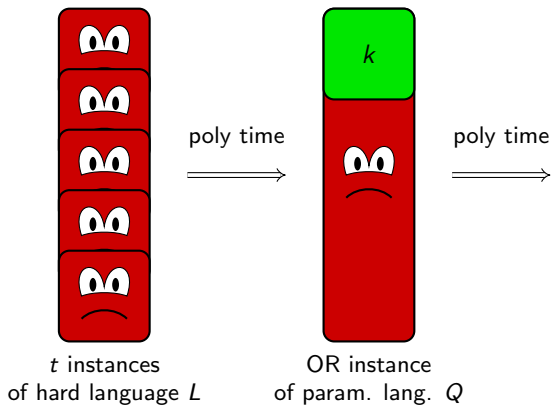
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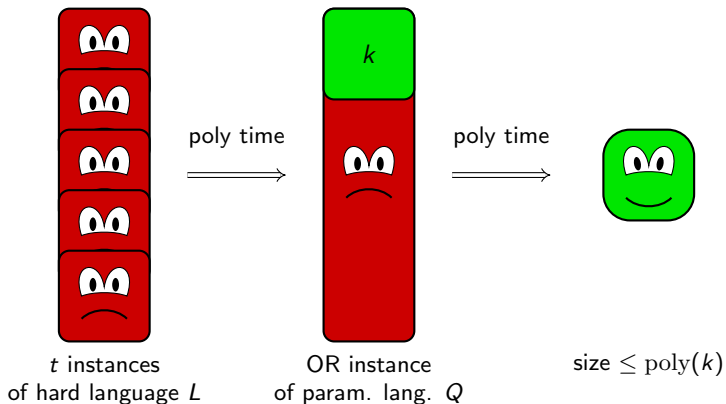
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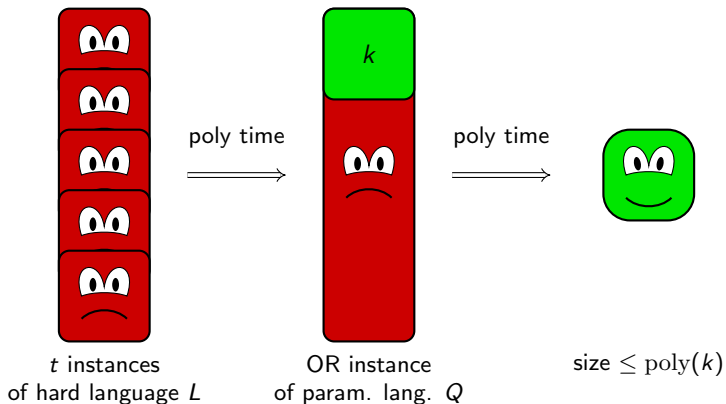
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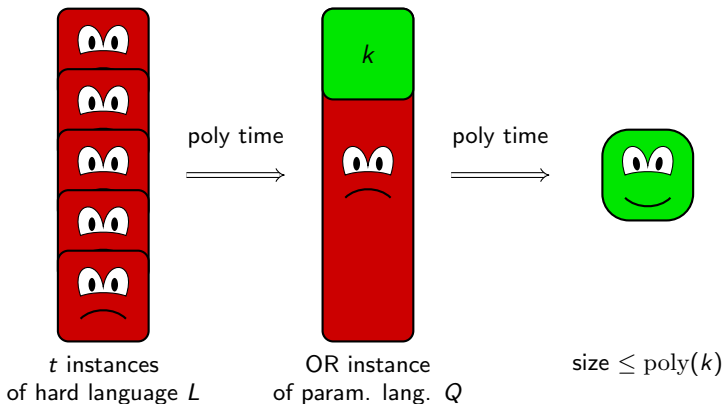
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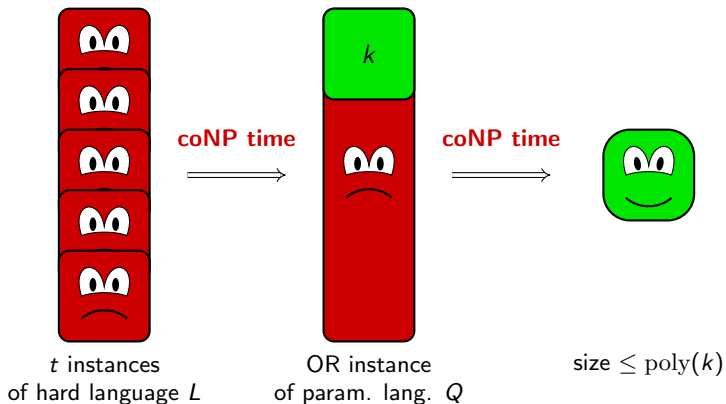


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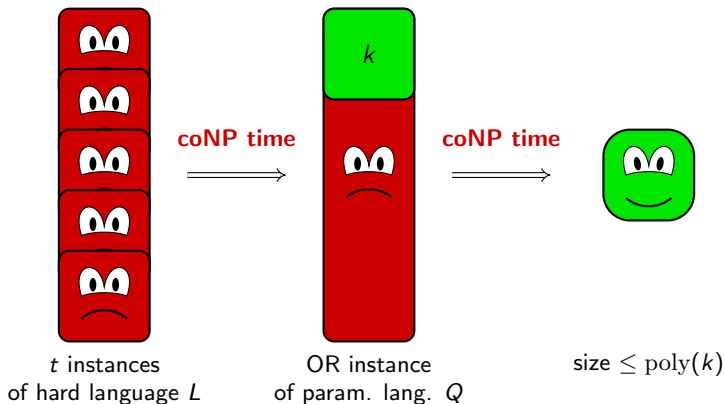


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- Then, a (co-nondeterministic) polynomial kernelization of Q implies $\text{NP} \subseteq \text{coNP/poly}$.

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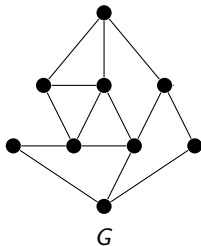
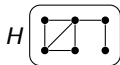
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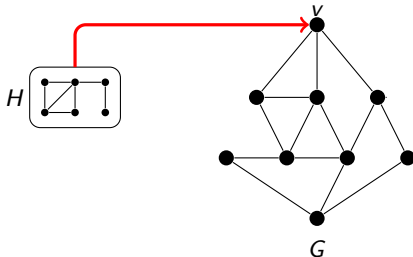
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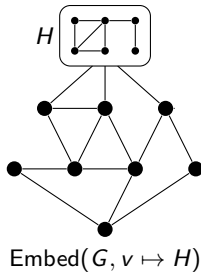
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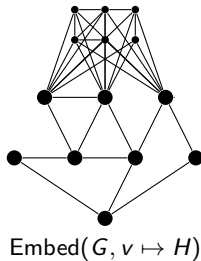
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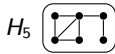
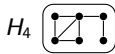
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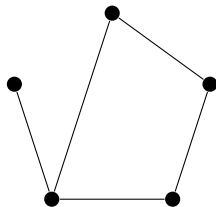
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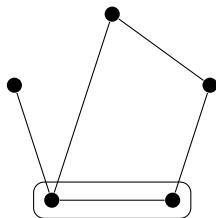
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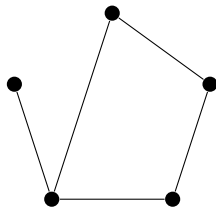
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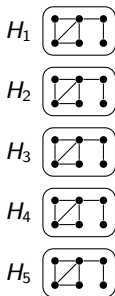
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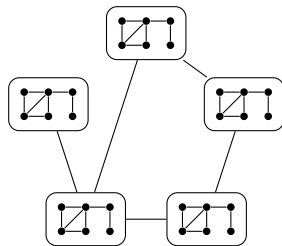
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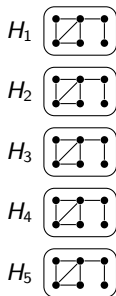
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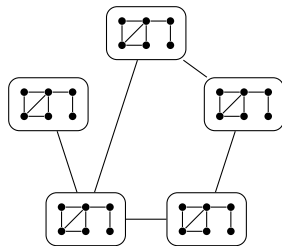
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- Π has the Erdős-Hajnal property \Rightarrow good host graph exists and we can find it in coNP-time.

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Theorem

*No poly-kernel for Π -INDUCED SUBGRAPH for any non-trivial poly-recognizable hereditary graph class Π that **contains all independent sets and cliques**, is closed under embedding and has the Erdős-Hajnal property.*

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Note: excluding a biclique implies the Erdős-Hajnal property.

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- Big open problem: prove Erdős-Hajnal conjecture.

Thank you



Questions?

Tikz faces based on a code by Raoul Kessels,
<http://www.texample.net/tikz/examples/emoticons/>,
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