# Courcelle's Conjecture, part II: treewidth and cliquewidth

### Michał Pilipczuk<sup>†</sup>

Based on a joint work with Mikołaj Bojańczyk and Martin Grohe

<sup>†</sup>Institute of Informatics, University of Warsaw, Poland

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### • First half:

• Continuation of Mikołaj's talk: Lifting the pathwidth case to the treewidth case.

### Second half:

- Statement of the conjecture for cliquewidth.
- Highlight of the proof for linear cliquewidth (with MB and MG).

# Part I

# from pathwidth to treewidth

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- Final approach: Reduce the treewidth case to the pathwidth case.
  - Caveat: Not a robust approach.

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- Guidance system:

Combinatorial object that provides this functionality.

#### Guidance system

A guidance system  $\Lambda$  in a graph G is a tuple of rooted forests

 $(F_1, F_2, \ldots, F_k)$ 

where  $V(F_i) = V(G)$  and  $E(F_i) \subseteq E(G)$  for each *i*.

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• A vertex subset X is **captured** by  $\Lambda$  if  $X \subseteq \Lambda(u)$  for some vertex u.















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For every graph G of pathwidth  $\leq k$ , some tree decomposition of G is captured by a guidance system of size f(k).

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- Intuition: Families of subsets captured by small guidance systems can be efficiently guessed in MSO.

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  - Combine all the obtained decompositions along s.



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- Idea: Extract pieces by a top-down induction.



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- Caveat: Not quite true, needs a slightly different choice of kings.



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- If achieved, then remaining requests are routed arbitrarily, and

$$p(k) + \binom{k}{2} - \frac{p(k)}{2\binom{k}{2}} < p(k)$$

for a quartic polynomial p(k).

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- **Ergo**: If no cutedge, then again  $X = {\text{root}}$  does the job.



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  - This is exactly what happens in the motivating example.



# Part II (linear) cliquewidth

- Treewidth algebra:
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  - Operations:
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#### Conjecture

Let *L* be a language of graphs of cliquewidth  $\leq k$ , for some  $k \in \mathbb{N}$ . Then *L* is definable in CMSO<sub>1</sub> iff it is recognizable.

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#### Corollary

#### [BGP,17+]

Let *L* be a language of graphs of **linear** cliquewidth  $\leq k$ , for some  $k \in \mathbb{N}$ . Then *L* is definable in CMSO<sub>1</sub> iff it is recognizable. • The **definable cliquewidth** of a graph is the minimum size of an MSO transduction that constructs some its clique decomposition.

# Proof strategy

- The **definable cliquewidth** of a graph is the minimum size of an MSO transduction that constructs some its clique decomposition.
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  - Lack of combinatorial abstraction is a nuisance.

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- Combine.



• **Abstraction**: Constant-size compositional information about a *k*-derivation that enables the following.

Idempotent Lemma

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  - We keep some information about paths between the cells.



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- **Flip** of *k*-derivation  $\tau$ :

for some pairs of cells, revert the adjacency between them.

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Let  $\sigma_1, \ldots, \sigma_n$  be k-derivations with the same idempotent abstraction. Then there is some flip H of  $\sigma_1 \cdots \sigma_n$  such that within every connected component of H, the block order can be expressed by an MSO formula of size bounded by a function of k.

- Using the Definable Order Lemma:
  - Guess partition into cells and the flip.
  - Interpret the block order in each connected component.
  - Apply the assumed transductions to each block in parallel.
  - Combine everything along the block order.
- Proving the Definable Order Lemma:
  - Analyze interactions between cells.

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  - Flip: turn full adjacencies into full non-adjacencies to make connections local.

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### • Thank you for your attention!