

# Fast FPT-Approximation of Branchwidth

Fedor V. Fomin, Tuukka Korhonen

Department of Informatics, University of Bergen, Norway

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Improves approximation ratio from 3 to 2

# Plan

1. Definitions
2. Overview of rankwidth algorithm
3. Combinatorial framework
4. Algorithmic framework



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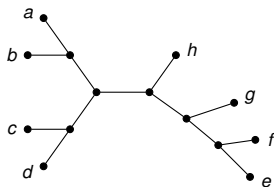
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  - ▶ Symmetric: For any  $A \subseteq V$ , it holds that  $f(A) = f(\bar{A})$ , where  $\bar{A} = V \setminus A$

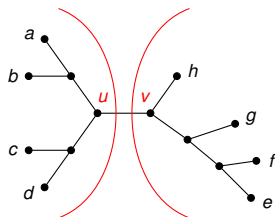
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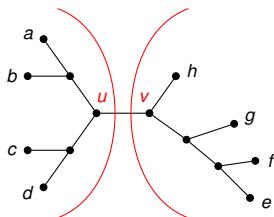
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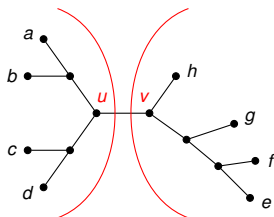
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## Connectivity functions

- Function  $f : 2^V \rightarrow \mathbb{Z}_{\geq 0}$  is a connectivity function if for any  $A, B \subseteq V$ :
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  - ▶ Rankwidth of  $G$
- Also carving-width, matroid branchwidth, rankwidth in different fields...

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**Input:** Augmented rank decomposition of  $G$  of width  $k$

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- Algorithmic framework:
  - ▶ Direct computation of refinements by dynamic programming  $\rightarrow 2^{2^{O(k)}} n^2$  time
  - ▶ Amortization techniques using combinatorial properties  $\rightarrow 2^{2^{O(k)}} n$  time

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## General idea

- Setting:
  - ▶ Let  $f : 2^V \rightarrow \mathbb{Z}_{\geq 0}$  be a connectivity function
  - ▶ We have a branch decomposition  $T$  of  $f$  of width  $k$
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- Strategy:

- ▶ Let  $h(T)$  be the number of edges of  $T$  of width  $k$  (heavy edges)
- ▶ Either decrease  $h(T)$  by using a **refinement operation**, or conclude that  $k \leq 2^{\text{bw}(f)}$

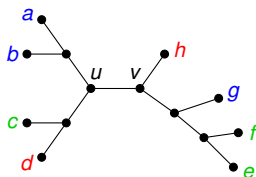
## Refinement operation

Specified by 4-tuple  $(r, C_1, C_2, C_3)$ , where  $r \in E(T)$  and  $(C_1, C_2, C_3)$  tripartition of  $V$

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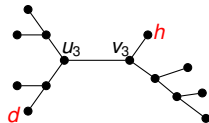
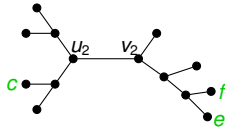
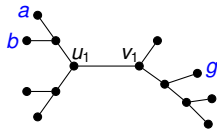
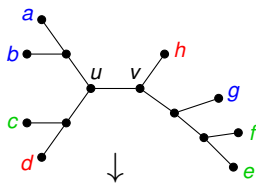
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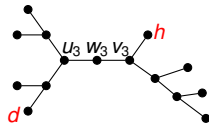
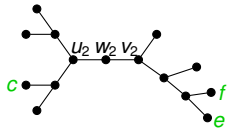
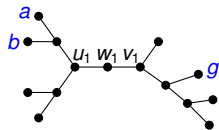
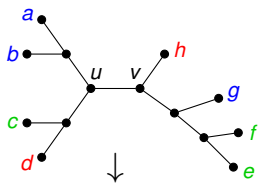
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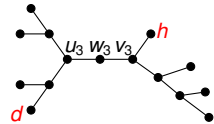
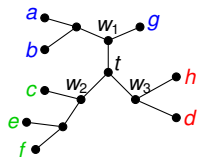
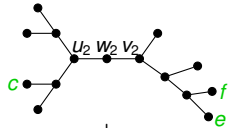
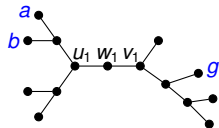
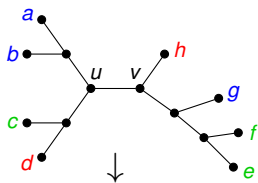
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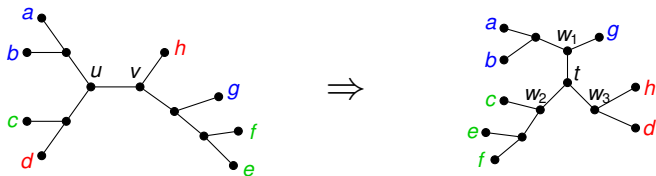
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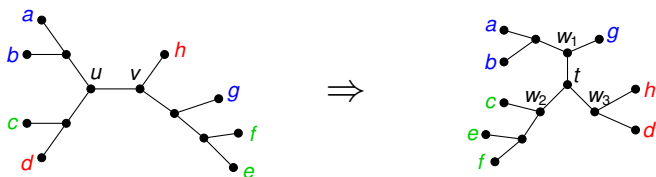
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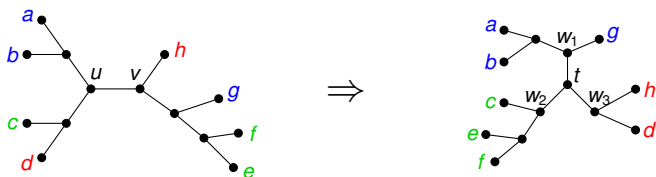


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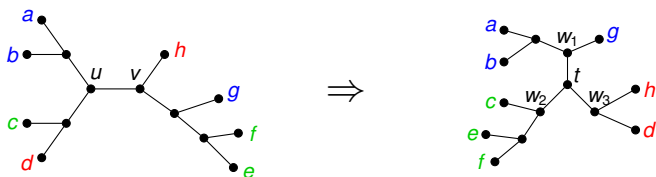
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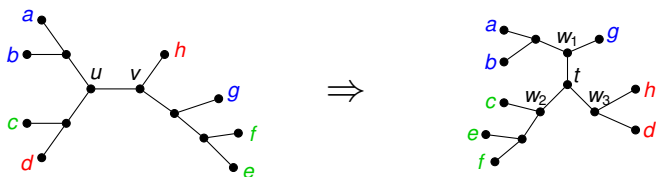
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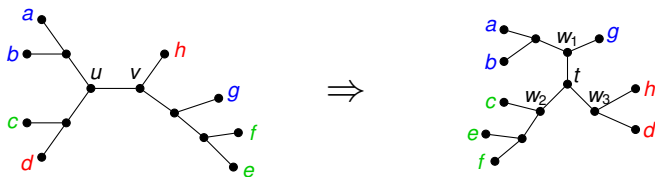
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- Observation 2: For each  $i$ , there will be edges corresponding to  $(C_i \cap W, \overline{C_i \cap W})$  and  $(C_i \cap \overline{W}, \overline{C_i \cap \overline{W}})$

## Local Improvement

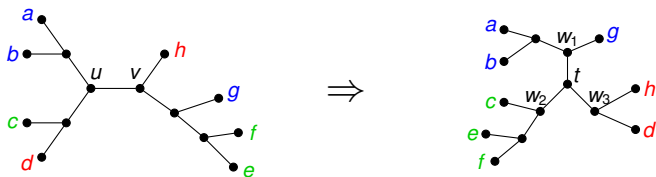
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- Combination of Observation 1 and 2:
  - ▶ The widths of edges “near the center” will be  $f(C_i)$ ,  $f(C_i \cap W)$ , and  $f(C_i \cap \overline{W})$  for each  $i$

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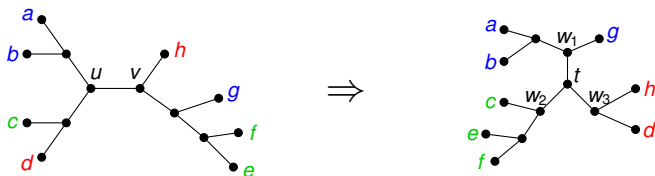
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### Theorem

For any set  $W \subseteq V$  with  $f(W) > 2b_w(f)$  there exists tripartition  $(C_1, C_2, C_3)$  of  $V$  so that for each  $i$  it holds that  $f(C_i) < f(W)/2$ ,  $f(C_i \cap W) < f(W)$ , and  $f(C_i \cap \overline{W}) < f(W)$ .

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Example with  $(r, C_1, C_2, C_3) = (uv, \{a, b, g\}, \{c, e, f\}, \{d, h\})$



- Let  $(W, \overline{W}) = (\{a, b, c, d\}, \{e, f, g, h\})$  be the cut of  $uv$
- Combination of Observation 1 and 2:
  - ▶ The widths of edges “near the center” will be  $f(C_i)$ ,  $f(C_i \cap W)$ , and  $f(C_i \cap \overline{W})$  for each  $i$

### Theorem

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$\Rightarrow$  If  $f(uv) > 2b_w(f)$ , there exists refinement with  $uv$  that “locally” improves  $T$

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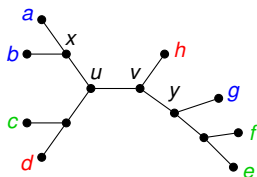
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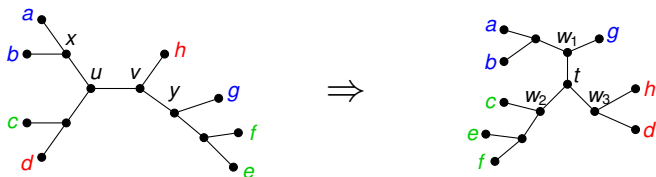
If there exists a  $W$ -improvement, then there exists a  $W$ -improvement  $(C_1, C_2, C_3)$  so that refinement with  $(uv, C_1, C_2, C_3)$  does not increase width and decreases the number of heavy edges.

## Global Improvement: Observation



- Consider  $T$  rooted at  $r = uv$
- For a node  $x \in V(T)$ , denote by  $T_r[x] \subseteq V$  the leaves in the subtree below  $x$ 
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- Let  $T'$  be refinement of  $T$  with  $(r, C_1, C_2, C_3)$
- Observation: Each edge of  $T'$  corresponds either to  $(C_i, \overline{C_i})$  or to  $(T_r[x] \cap C_i, \overline{T_r[x] \cap C_i})$  for some  $x \in V(T)$

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- For the edge  $uv$ , none of the new edges corresponding to  $uv$  has width  $f(uv)$   
 $\Rightarrow$  Strict improvement

# Plan

1. Definitions
2. Overview of rankwidth algorithm
3. Combinatorial framework
4. **Algorithmic framework**

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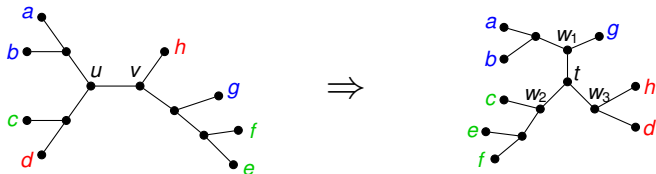
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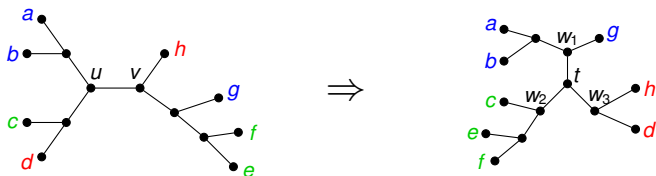
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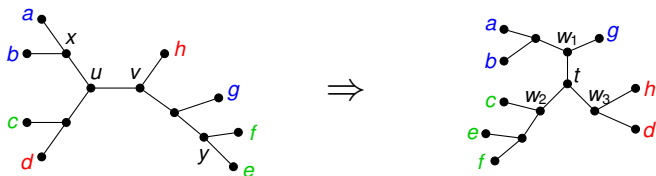
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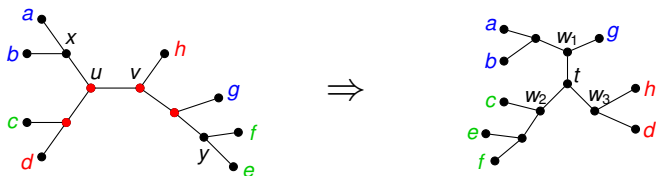


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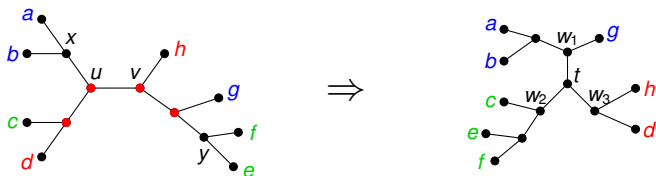
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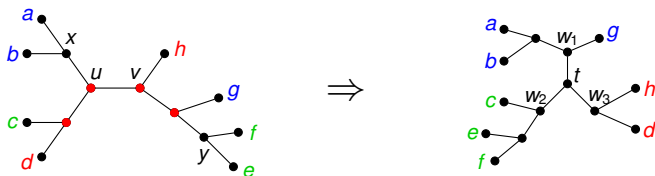
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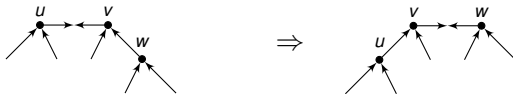
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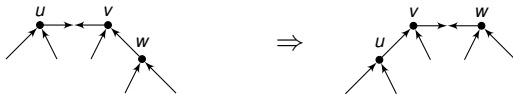
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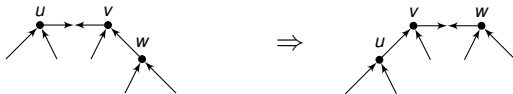
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