

# FPT algorithm for a generalized cut problem and some applications

EunJung Kim<sup>1</sup> Sang-Il Oum<sup>2</sup>  
Christophe Paul<sup>3</sup> Ignasi Sau<sup>3</sup> Dimitrios M. Thilikos<sup>3</sup>

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<sup>1</sup> CNRS, LAMSADE, Paris (France)

<sup>2</sup> KAIST, Daejeon (South Korea)

<sup>3</sup> CNRS, LIRMM, Montpellier (France)

# Outline of the talk

- 1 Introduction
- 2 Sketch of the FPT algorithm
- 3 Some applications
- 4 Conclusions

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# Some words on parameterized complexity

- Idea given an NP-hard problem with **input size**  $n$ , fix one **parameter**  $k$  of the input to see whether the problem gets more “tractable”.

**Example:** the size of a VERTEX COVER.

- Given a (NP-hard) problem with input of size  $n$  and a parameter  $k$ , a **fixed-parameter tractable (FPT)** algorithm runs in time

$$f(k) \cdot n^{O(1)}, \text{ for some function } f.$$

**Examples:**  $k$ -VERTEX COVER,  $k$ -LONGEST PATH.

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**Cut problem** given a graph, find a minimum (vertex or edge) **cutset** whose removal makes the graph satisfy some **separation** property.

- MIN CUT: polynomial by classical **max-flow min-cut** theorem.
- MULTIWAY CUT: FPT by using **important separators**. [Marx '06]
- MULTICUT: Finally, FPT. [Marx, Razgon + Bousquet, Daligault, Thomassé '10]
- STEINER CUT: Improved FPT algorithm by using **randomized contractions**. [Chitnis, Cygan, Hajiaghayi, Pilipczuk<sup>2</sup> '12]
- MIN BISECTION: Finally, FPT. [Cygan, Lokshtanov, Pilipczuk<sup>2</sup>, Saurabh '13]

# We introduce a new cut problem

- A new cut problem: **LIST ALLOCATION** (to be defined in two slides).

## Theorem

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## Theorem

*The LIST ALLOCATION problem is **FPT**.*

- LIST ALLOCATION generalizes, in particular, MULTIWAY CUT.
- **General** enough so that several other problems can be reduced to it:
  - ★ FPT algorithm for a parameterization of **DIGRAPH HOMOMORPHISM**.
  - ★ FPT algorithm for the **MIN-MAX GRAPH PARTITIONING** problem.
  - ★ FPT 2-approximation for **TREE-CUT WIDTH**.

## Before defining the problem: allocations

- An  $r$ -allocation of a set  $S$  is an  $r$ -tuple  $\mathcal{V} = (V_1, \dots, V_r)$  of possibly empty pairwise disjoint subsets of  $S$  whose union is  $S$ .
- Elements of  $\mathcal{V}$ : parts of  $\mathcal{V}$ .
- We denote by  $\mathcal{V}^{(i)}$  the  $i$ -th part of  $\mathcal{V}$ , i.e.,  $\mathcal{V}^{(i)} = V_i$ .

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- We denote by  $\mathcal{V}^{(i)}$  the  $i$ -th part of  $\mathcal{V}$ , i.e.,  $\mathcal{V}^{(i)} = V_i$ .
- Let  $G = (V, E)$  be a graph and let  $\mathcal{V}$  be an  $r$ -allocation of  $V$ :  
 $|\delta(\mathcal{V}^{(i)}, \mathcal{V}^{(j)})|$ : #edges in  $G$  with one endpoint in  $\mathcal{V}^{(i)}$  and one in  $\mathcal{V}^{(j)}$ .

# Definition of the problem: LIST ALLOCATION

## LIST ALLOCATION

**Input:** A tuple  $I = (G, r, \lambda, \alpha)$ , where  $G$  is an  $n$ -vertex graph,  $r \in \mathbb{Z}_{\geq 1}$ ,  $\lambda : V(G) \rightarrow 2^{[r]}$ , and  $\alpha : \binom{[r]}{2} \rightarrow \mathbb{Z}_{\geq 0}$ .

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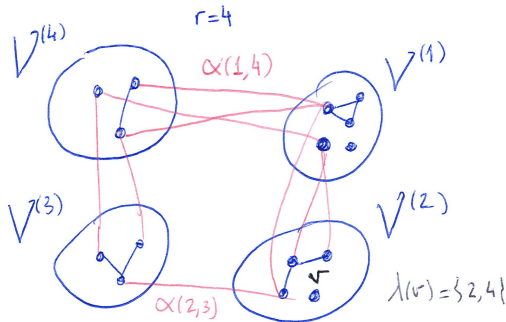
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**Parameter:**  $k = \sum \alpha$ .

**Question:** Decide whether there exists an  $r$ -allocation  $\mathcal{V}$  of  $V(G)$  s.t.

- $\forall \{i, j\} \in \binom{[r]}{2}$ ,  $|\delta(\mathcal{V}^{(i)}, \mathcal{V}^{(j)})| = \alpha(i, j)$  and
- $\forall v \in V(G)$ , if  $v \in \mathcal{V}^{(i)}$  then  $i \in \lambda(v)$ .



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# High-level ideas of the FPT algorithm

- We use a series of **FPT reductions**:

**Problem A**  $\xrightarrow{\text{FPT}}$  **Problem B**: If problem *B* is FPT, then problem *A* is FPT.



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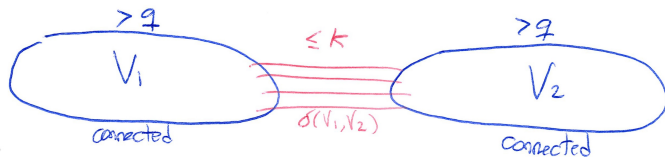
- At some steps, we obtain instances whose size is bounded by some function  $f(k)$ .
- Then we will use that the LIST ALLOCATION problem is in **XP**:

## Lemma

*There exists an algorithm that, given an instance  $I = (G, r, \lambda, \alpha)$  of LIST ALLOCATION, computes all possible solutions in time  $n^{O(k)} \cdot r^{O(k+\ell)}$ , where  $\ell$  is the number of connected components of  $G$ .*

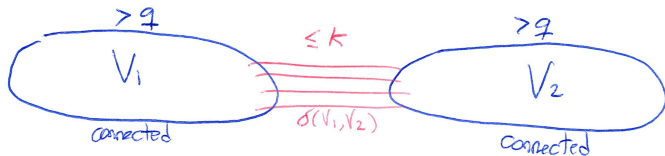
# Some preliminaries

- Let  $G$  be a connected graph. A partition  $(V_1, V_2)$  of  $V(G)$  is a  $(q, k)$ -separation if  $|V_1|, |V_2| > q$ ,  $|\delta(V_1, V_2)| \leq k$ , and  $G[V_1]$  and  $G[V_2]$  are both connected.



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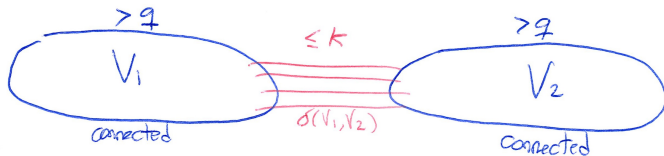
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- A graph  $G$  is  $(q, k)$ -connected if it does not contain any  $(q, k - 1)$ -separation.

## Lemma (Chitnis, Cygan, Hajiaghayi, Pilipczuk<sup>2</sup> '12)

There exists an algorithm that given a  $n$ -vertex connected graph  $G$  and two integers  $q, k$ , either finds a  $(q, k)$ -separation, or reports that no such separation exists, in time  $(q + k)^{O(\min\{q, k\})} n^3 \log n$ .

LIST ALLOCATION (LA)

# Series of FPT reductions

LIST ALLOCATION (LA)

↓ FPT

CONNECTED LIST ALLOCATION (CLA)

Same input + graph  $G$  is **connected** and  $r \leq 2k$

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Same input + graph  $G$  is  $(f_1(k), k + 1)$ -connected, for  $f_1(k) := 2^k \cdot (2k)^{2k}$

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Claim (Unique big part)

For any solution  $\mathcal{V}$  of HCLA there exists a *unique index*  $j \in [r]$  such that

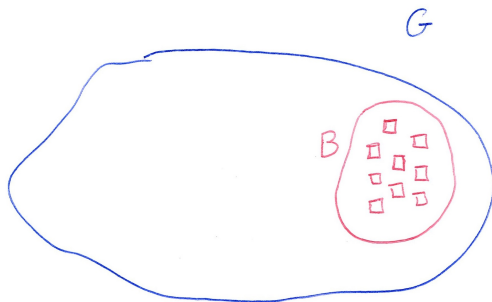
$$\sum_{i \in [r] \setminus j} |\mathcal{V}^{(i)}| \leq k \cdot f_1(k).$$

- Part  $\mathcal{V}^{(j)}$  is called the **big part**.



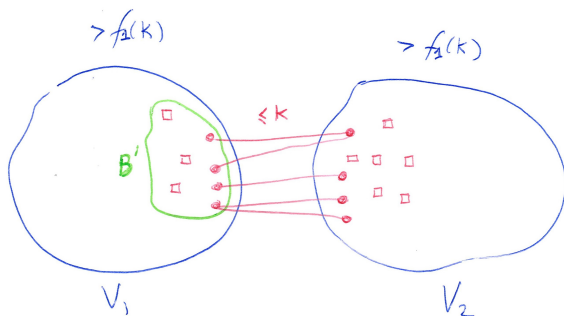
# Reduction from CLA to HCLA: we shrink the graph

- We apply to  $G$  the following recursive algorithm **shrink**, which receives a graph  $G$  and a boundary set  $B$  with  $|B| \leq 2k$  (start with  $B = \emptyset$ ):



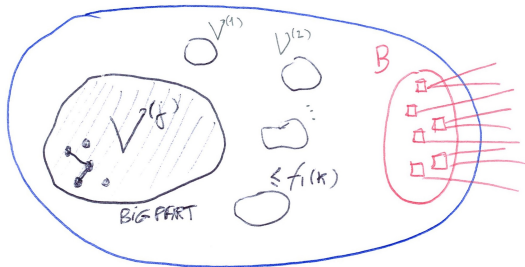
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  - 1 If  $G$  has a  **$(f_1(k), k)$ -separation**  $(V_1, V_2)$ :
    - W.l.o.g. let  $V_1$  be the part with the smallest number of boundary vertices, and let  $B'$  be the new boundary: so  $|B'| \leq 2k$ .
    - Call recursively **shrink** with input  $(G[V_1], B')$ , and update the graph.



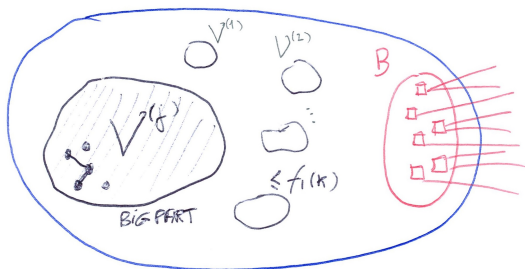
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    - 2 Otherwise, find a set of **"indistinguishable"** vertices, and **identify** them.
- Idea** We generate **all partial solutions in the boundary**, and for each of them we compute a solution of HCLA, using our "black box".



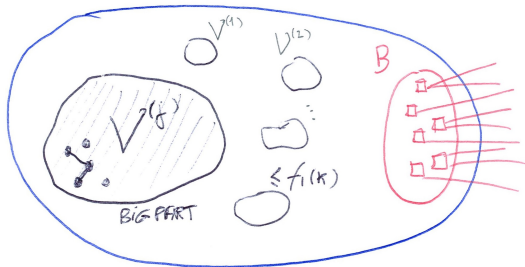
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    - 2 Otherwise, find a set of **"indistinguishable"** vertices, and **identify** them.
- Idea** By the high connectivity (**Claim**), each such solution has a unique big part  $V^{(i)}$ : **indistinguishable** vertices for this behavior.



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- Idea** If the graph is **big enough**, there are vertices that are indistinguishable for **all** behaviors  $\Rightarrow$  **identify** them. Return the graph.



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- Idea** If the graph is **big enough**, there are vertices that are indistinguishable for **all** behaviors  $\Rightarrow$  **identify** them. Return the graph.

## Lemma

The above algorithm returns in **FPT** time an **equivalent** instance of CLA of size at most  $f_2(k) := k \cdot (f_1(k))^2 + 2k + 2$ . (Then we apply the **XP** algorithm.)

# Series of FPT reductions

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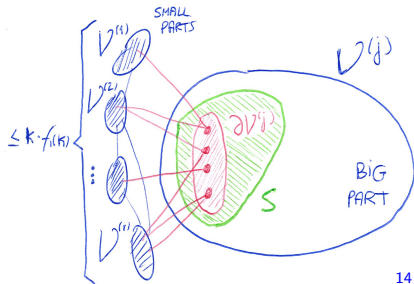
HIGHLY CONNECTED LIST ALLOCATION (HCLA)

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SPLIT HIGHLY CONNECTED LIST ALLOCATION (SHCLA)

Same input + set  $S \subseteq V(G)$  and a solution  $\mathcal{V}$  additionally needs to satisfy that if  $j \in [r]$  is such that  $\mathcal{V}^{(j)}$  is the **big part** of  $\mathcal{V}$ , then

$$\partial \mathcal{V}^{(j)} \subseteq S \subseteq \mathcal{V}^{(j)}.$$





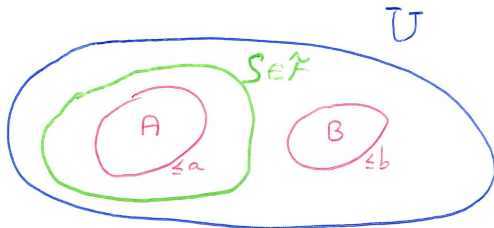
# Crucial ingredient: Splitter Lemma

- **Splitters** were first introduced by
- We use the following deterministic version:

[Naor, Schulman, Srinivasan '95]

**Lemma (Chitnis, Cygan, Hajiaghayi, Pilipczuk<sup>2</sup> '12)**

There exists an algorithm that given a set  $U$  of size  $n$  and two integers  $a, b \in [0, n]$ , outputs a set  $\mathcal{F} \subseteq 2^U$  where  $|\mathcal{F}| = (a + b)^{O(\min\{a, b\})} \cdot \log n$  such that for every two sets  $A, B \subseteq U$ , where  $A \cap B = \emptyset$ ,  $|A| \leq a$ ,  $|B| \leq b$ , there exists a set  $S \in \mathcal{F}$  where  $A \subseteq S$  and  $B \cap S = \emptyset$ , in  $(a + b)^{O(\min\{a, b\})} \cdot n \log n$  steps.



## Reduction from HCLA to SHCLA: we use splitters

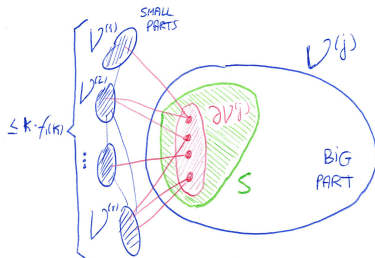
- We use the **Splitter Lemma** with universe  $U = V(G)$ ,  $a = k$ , and  $b = k \cdot f_1(k)$ , obtaining a family  $\mathcal{F}$  of subsets of  $V(G)$ .

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- **Idea** We want a set  $S \subseteq V(G)$  that “splits” these two sets:

$$A = \partial V^{(j)} \quad \text{and} \quad B = \bigcup_{i \in [r] \setminus \{j\}} V^{(i)}.$$

For some  $j \in [r]$ :  $|A| \leq k$  and  $|B| \leq k \cdot f_1(k)$  (by the **Claim**).

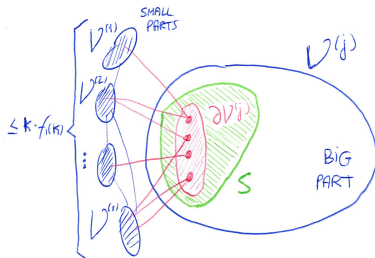


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- It holds that  $I$  is a YES-instance of **HCLA** if and only if for some  $S \in \mathcal{F}$ ,  $(I, S)$  is a YES-instance of **SHCLA**.

# An algorithm to solve SHCLA

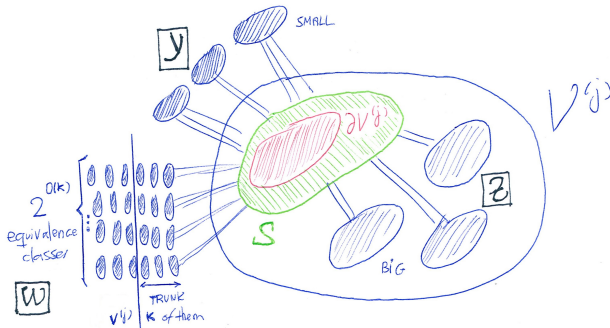
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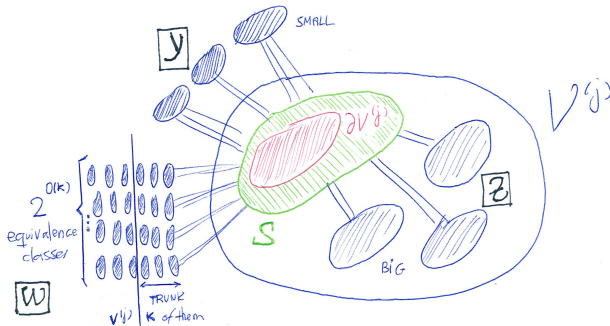
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  - $\mathcal{Y}$ : those that **cannot** go **entirely** in  $\mathcal{V}^{(j)}$ .
  - $\mathcal{Z}$ : those that are **big** ( $> k \cdot f_1(k)$ ) and that can go **entirely** in  $\mathcal{V}^{(j)}$ .
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## Lemma

The **SHCLA** problem can be solved in time  $2^{O(k^2 \cdot \log k)} \cdot n$ .



# Piecing everything together

LIST ALLOCATION (LA)

↓ FPT reduction

CONNECTED LIST ALLOCATION (CLA)

↓ FPT reduction

HIGHLY CONNECTED LIST ALLOCATION (HCLA)

↓ FPT reduction

SPLIT HIGHLY CONNECTED LIST ALLOCATION (SHCLA)

↓ FPT algorithm to solve SHCLA

## Theorem

LIST ALLOCATION *can be solved in time*  $2^{O(k^2 \log k)} \cdot n^4 \cdot \log n$ .

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# Generalization of DIGRAPH HOMOMORPHISM

## ARC-BOUNDED LIST DIGRAPH HOMOMORPHISM

**Input:** Two digraphs  $G$  and  $H$ , a list  $\lambda : V(G) \rightarrow 2^{V(H)}$  of allowed images for every vertex in  $G$ , and a function  $\alpha$  prescribing the number of non-loop arcs in  $G$  mapped to each arc of  $H$ .

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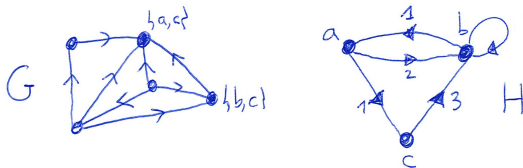
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**Parameter:**  $k = \sum \alpha$ .

**Question:** Decide whether there exists a homomorphism from  $G$  to  $H$  respecting the constraints imposed by  $\lambda$  and  $\alpha$ .



- It generalizes several homomorphism problems.

[Díaz, Serna, Thilikos '08]

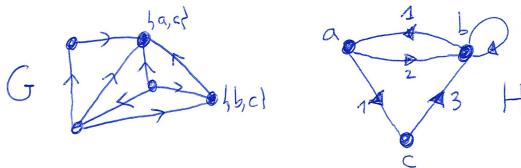
# Generalization of DIGRAPH HOMOMORPHISM

## ARC-BOUNDED LIST DIGRAPH HOMOMORPHISM

**Input:** Two digraphs  $G$  and  $H$ , a list  $\lambda : V(G) \rightarrow 2^{V(H)}$  of allowed images for every vertex in  $G$ , and a function  $\alpha$  prescribing the number of non-loop arcs in  $G$  mapped to each arc of  $H$ .

**Parameter:**  $k = \sum \alpha$ .

**Question:** Decide whether there exists a homomorphism from  $G$  to  $H$  respecting the constraints imposed by  $\lambda$  and  $\alpha$ .



- It generalizes several homomorphism problems.

[Díaz, Serna, Thilikos '08]

## Corollary

The ARC-BOUNDED LIST DIGRAPH HOMOMORPHISM problem is **FPT**.

# Graph partitioning problem

## MIN-MAX GRAPH PARTITIONING

**Input:** An undirected graph  $G$ ,  $w, r \in \mathbb{Z}_{\geq 0}$ , and  $T \subseteq V(G)$  with  $|T| = r$ .

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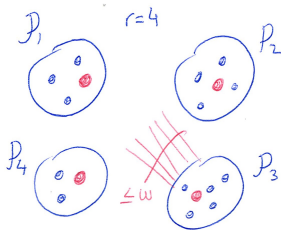
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- Important in approximation. [Bansal, Feige, Krauthgamer, Makarychev, Nagarajan, Naor, Schwartz'11]
- The “MIN-SUM” version is exactly the MULTIWAY CUT problem. [Marx '06]

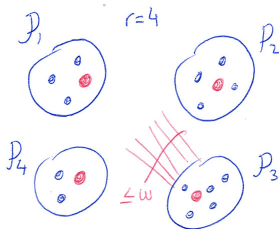
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## 2-approximation for TREE-CUT WIDTH

- **Tree-cut width** is a graph invariant fundamental in the structure of graphs not admitting a fixed graph as an **immersion**. [Wollan '14]
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- Tree-cut width also has **algorithmic applications**. [Ganian, Kim, Szeider '14]

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*There exists an algorithm that, given a graph  $G$  and a  $k \in \mathbb{Z}_{\geq 0}$ , in time  $2^{O(k^2 \cdot \log k)} \cdot n^5 \cdot \log n$  either outputs a tree-cut decomposition of  $G$  with width at most  $2k$ , or correctly reports that the tree-cut width of  $G$  is strictly larger than  $k$ .*

# Next section is...

- 1 Introduction
- 2 Sketch of the FPT algorithm
- 3 Some applications
- 4 Conclusions**

# Conclusions and further research

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# Gràcies!

