

FPT algorithms for hitting forbidden minors

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Joint work with Julien Baste, Giannos Stamoulis, and Dimitrios M. Thilikos

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Graph Theory, Combinatorics and Algorithms**
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Outline of the talk

1 Introduction

- Parameterized complexity
- Treewidth

2 Hitting forbidden minors

- Parameterized by treewidth
- Parameterized by solution size

Next section is...

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- **Moderately exponential-time algorithms:** Solve the problem in exponential time, but reasonably “fast” (1.15^n vs 2^n).
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- **Parameterized complexity:** Topic of this talk...

The area of parameterized complexity

Idea Measure the complexity of an algorithm in terms of the **input size** and an **additional integer parameter**.

This theory started in the late 80's, by **Downey** and **Fellows**:



Today, it is a well-established area with **hundreds** of articles published every year in the most prestigious TCS journals and conferences.

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These three problems are **NP-hard**, but are they **equally hard**?

They behave quite differently...

① k -VERTEX COVER: solvable in time $2^k \cdot n^2$

② k -CLIQUE: solvable in time $k^2 \cdot n^k$

③ VERTEX k -COLORING: NP-hard for every fixed $k \geq 3$

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The problem is **para-NP-hard**

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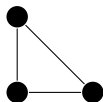
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Treewidth via k -trees

Example of a 2-tree:

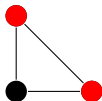


[Figure by Julien Baste]

A k -tree is a graph that can be built starting from a $(k + 1)$ -clique and then iteratively adding a vertex connected to a k -clique.

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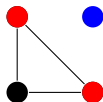


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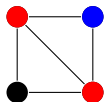


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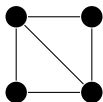


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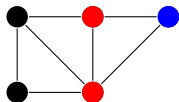


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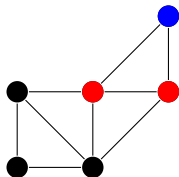


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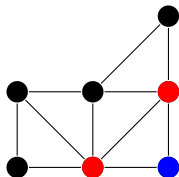


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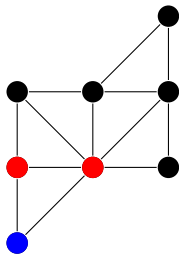


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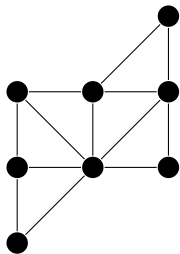


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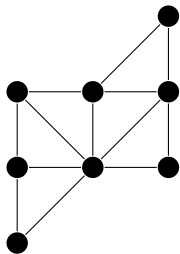


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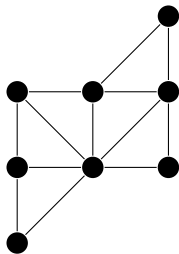
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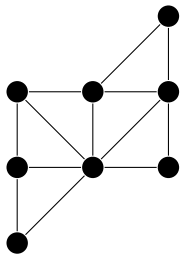
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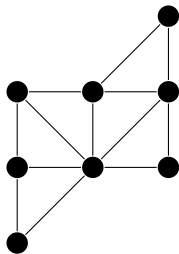
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Construction suggests the notion of tree decomposition: small separators.

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- ① Treewidth is a fundamental **combinatorial tool** in graph theory: key role in the **Graph Minors** project of Robertson and Seymour.
- ② Treewidth behaves very well **algorithmically**, and algorithms parameterized by treewidth appear **very often** in FPT algorithms.
- ③ In many **practical scenarios**, it turns out that the **treewidth** of the associated graph is **small** (programming languages, road networks, ...).

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Monadic Second Order Logic (MSOL):

Graph logic that allows quantification over sets of vertices and edges.

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Theorem (Courcelle. 1990)

*Every problem expressible in **MSOL** can be solved in time $f(\text{tw}) \cdot n$ on graphs on n vertices and **treewidth** at most tw .*

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In parameterized complexity: FPT parameterized by treewidth.

Examples: VERTEX COVER, DOMINATING SET, HAMILTONIAN CYCLE, CLIQUE, INDEPENDENT SET, k -COLORING for fixed k , ...

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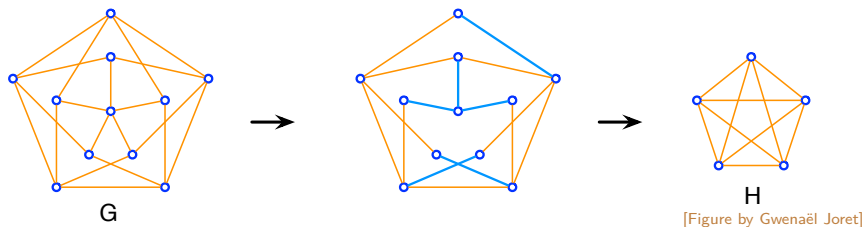
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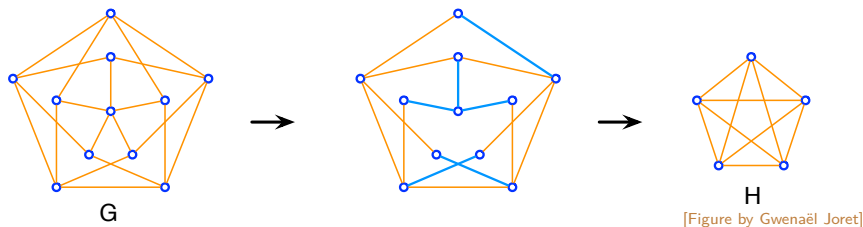
- $\mathcal{M} = \{\text{vertex deletion}\}$.
- \mathcal{C} is a minor-closed graph class.

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A graph H is a **minor** of a graph G if H , denoted by $H \leq_m G$, can be obtained from a **subgraph** of G by **contracting edges**.

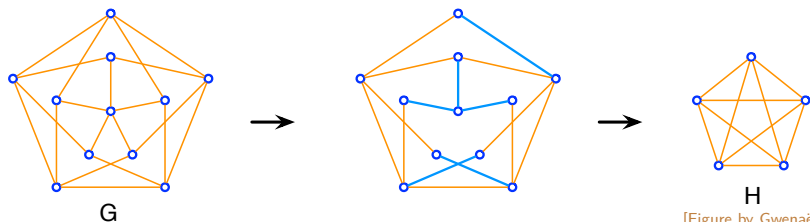
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Theorem (Robertson and Seymour. 1983-2004)

For every **minor-closed** graph class \mathcal{C} there exists a **finite** collection \mathcal{F} of **forbidden minors** such that, for every graph G ,

$$G \in \mathcal{C} \iff F \not\leq_m G \text{ for every } F \in \mathcal{F}.$$

Hitting forbidden minors

- If $\mathcal{C} = \{\text{edgeless graphs}\}$, then $\mathcal{F} = \{K_2\}$.
- If $\mathcal{C} = \{\text{forests}\}$, then $\mathcal{F} = \{K_3\}$.
- If $\mathcal{C} = \{\text{outerplanar graphs}\}$, then $\mathcal{F} = \{K_4, K_{2,3}\}$.
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- $\mathcal{F} = \{K_2\}$: VERTEX COVER.
- $\mathcal{F} = \{K_3\}$: FEEDBACK VERTEX SET.
- $\mathcal{F} = \{K_5, K_{3,3}\}$: VERTEX PLANARIZATION.
- $\mathcal{F} = \{\text{diamond}\}$: CACTUS VERTEX DELETION.

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We consider the following two parameterizations of \mathcal{F} -M-DELETION:

- 1 Structural parameter: $\text{tw}(G)$.
- 2 Solution size: k .

Next subsection is...

1 Introduction

- Parameterized complexity
- Treewidth

2 Hitting forbidden minors

- Parameterized by treewidth
- Parameterized by solution size

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Every problem expressible in **MSOL** can be solved in time $f_{\mathcal{F}}(\text{tw}) \cdot n$ on graphs on n vertices and **treewidth** at most tw .

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Very active area in parameterized complexity during the last decade.

Remark: Algorithms parameterized by **treewidth** appear very often as a “**black box**” in all kinds of parameterized algorithms.

What was known for particular collections \mathcal{F}

Let \mathcal{F} be a fixed finite collection of graphs.

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Input: A graph G and an integer k .

Parameter: The treewidth tw of G .

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Determine, for every fixed \mathcal{F} , the (asymptotically) smallest function $f_{\mathcal{F}}$ such that \mathcal{F} -M-DELETION on n -vertex graphs can be solved in time

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[Baste, S., Thilikos. **Hitting minors on bounded treewidth graphs. I. General upper bounds.** 2020]

[Baste, S., Thilikos. **Hitting minors on bounded treewidth graphs. II. Single-exponential algorithms.** 2020]

[Baste, S., Thilikos. **Hitting minors on bounded treewidth graphs. III. Lower bounds.** 2020]

[Baste, S., Thilikos. **Hitting minors on bounded treewidth graphs. IV. An optimal algorithm.** 2020-]

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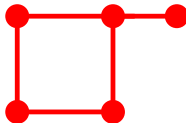
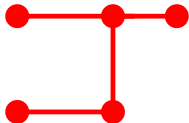
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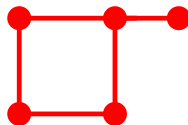
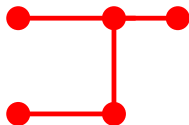
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A dichotomy for hitting a connected minor



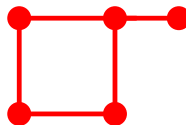
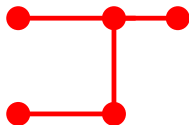
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Theorem (Baste, S., Thilikos. 2016-2020)

Let H be a *connected* graph.

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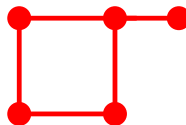
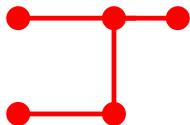
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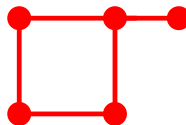
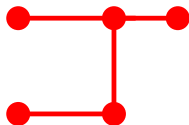
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
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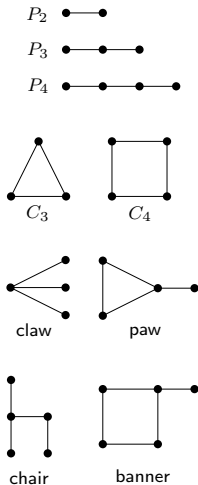
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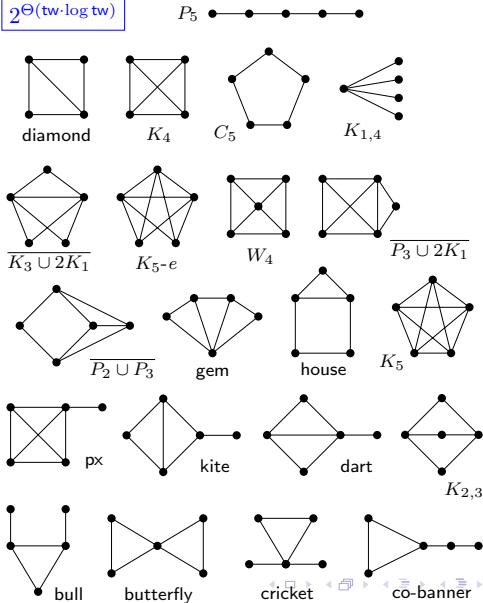
In both cases, the running time is asymptotically *optimal* under the ETH.

Complexity of hitting a single connected minor H

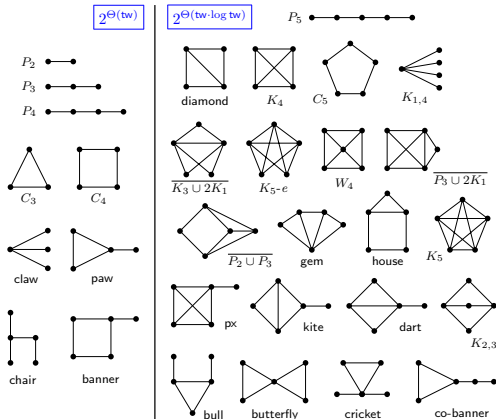
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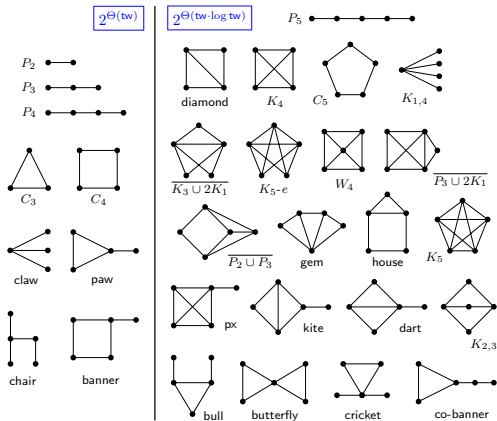


A compact statement for a single connected graph



All these cases can be succinctly described as follows:

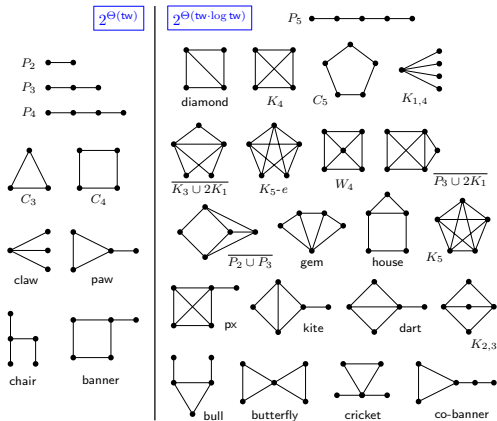
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- Some use “typical” dynamic programming.
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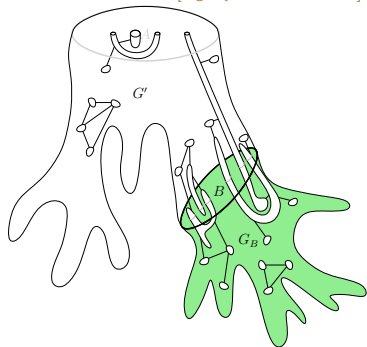
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[Fig. by Valentin Garnero]

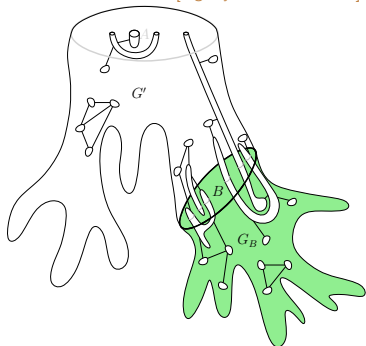


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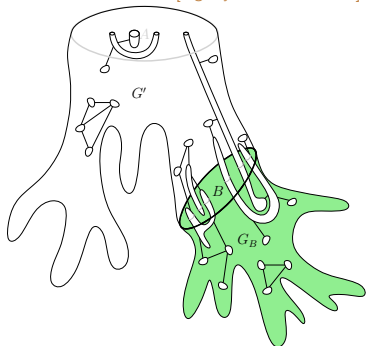
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- $\mathcal{R}^{(\mathcal{F}, t)}$: set of **minimum-size representatives** of $\equiv^{(\mathcal{F}, t)}$.



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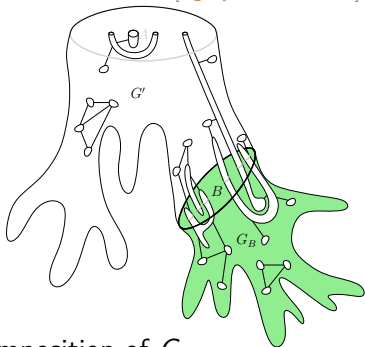
[Fig. by Valentin Garnero]

- For a fixed \mathcal{F} , we define an **equivalence relation** $\equiv^{(\mathcal{F}, t)}$ on t -boundaried graphs:

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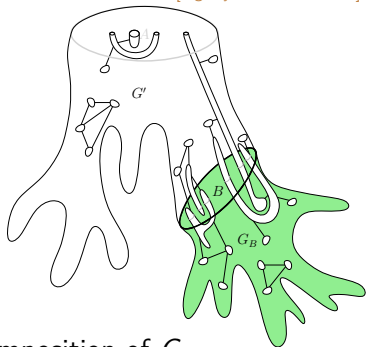
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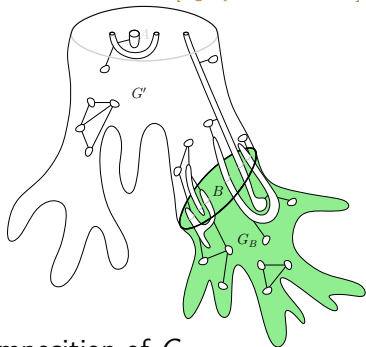
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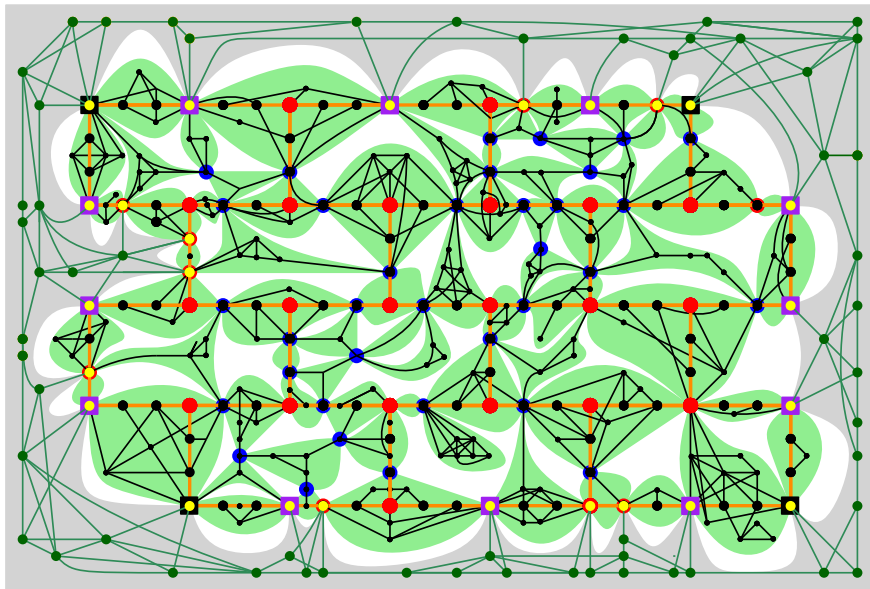
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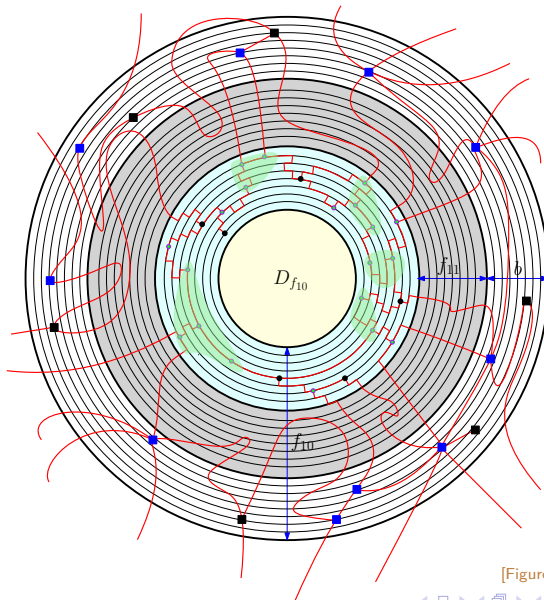
As a representative R is \mathcal{F} -minor-free, if $\text{tw}(R \setminus B) > c_{\mathcal{F}}$,
 $R \setminus B$ contains a large flat wall, where we can find an irrelevant vertex.

A flat wall can in fact be quite wild...



Hard part: finding an irrelevant vertex inside a flat wall

Hard part: finding an irrelevant vertex inside a flat wall



» skip

[Figure by Dimitrios M. Thilikos]

Diagram of the algorithm for a general collection \mathcal{F}

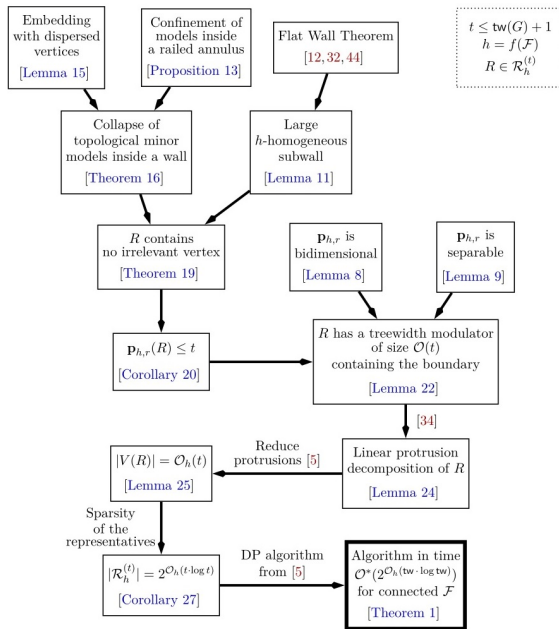
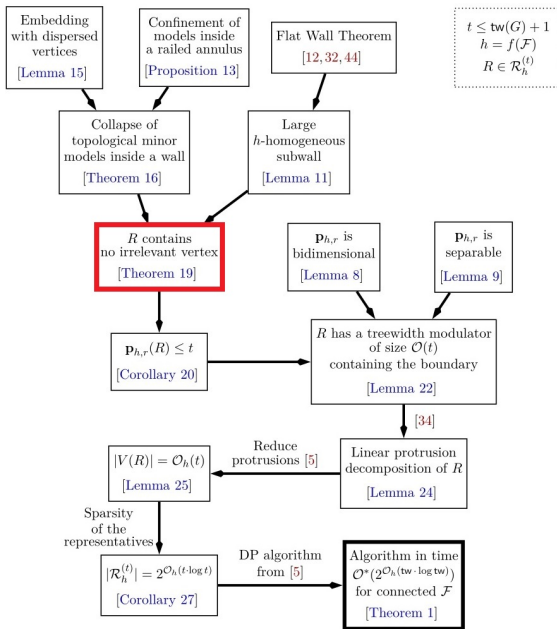


Diagram of the algorithm for a general collection \mathcal{F}



Next subsection is...

1 Introduction

- Parameterized complexity
- Treewidth

2 Hitting forbidden minors

- Parameterized by treewidth
- Parameterized by solution size

We parameterize by the size of the desired solution

\mathcal{F} -M-DELETION

Input: A graph G and an integer k .

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But... only **existential**, **non-uniform**, $f(\mathcal{C}_k)$ **astronomical**.

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- The function $f(\mathcal{C}_k)$ is constructible.

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- For every \mathcal{F} , some enormous explicit function $f_{\mathcal{F}}(k)$ can be derived from an FPT algorithm for hitting topological minors:

$$f_{\mathcal{F}}(k) \cdot n^{\mathcal{O}(1)}.$$

[Fomin, Lokshtanov, Panolan, Saurabh, Zehavi. 2020]

Our results

Theorem (S., Stamoulis, Thilikos. 2020)

For *all* \mathcal{F} , the \mathcal{F} -M-DELETION problem can be solved in time $2^{\text{poly}(k)} \cdot n^3$.

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If \mathcal{F} contains an *apex graph*, the \mathcal{F} -M-DELETION problem can be solved in time $2^{\text{poly}(k)} \cdot n^2$.

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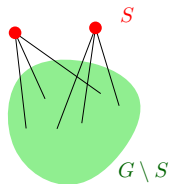
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General scheme of the algorithm:

[whole slide shamelessly borrowed from Giannos Stamoulis]

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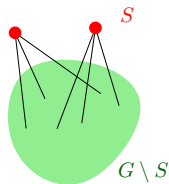
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Iterative compression: given solution S of size $k + 1$, search solution of size k .

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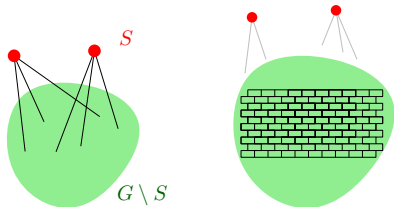


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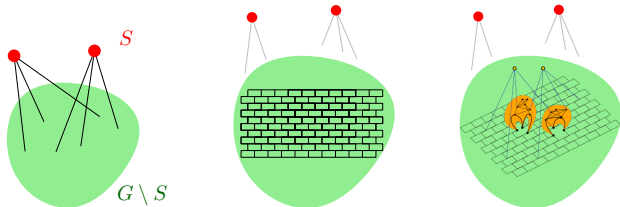
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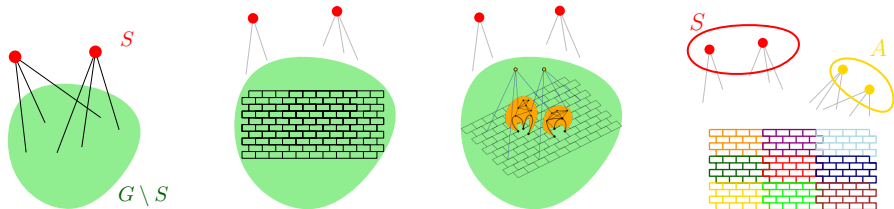
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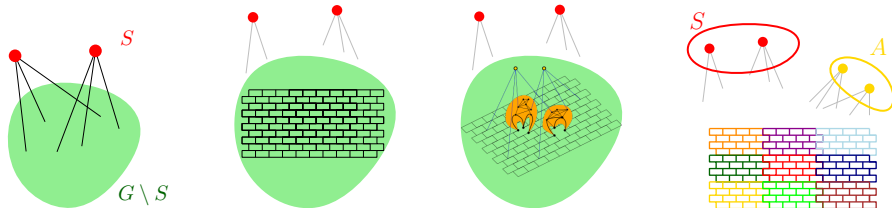
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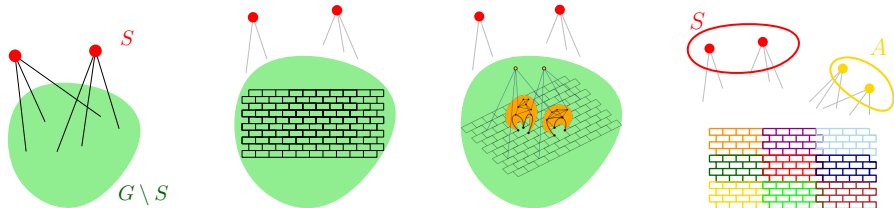
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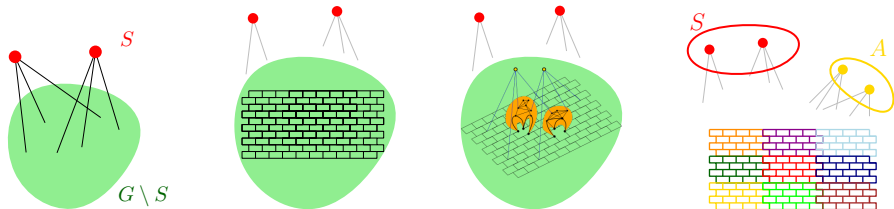
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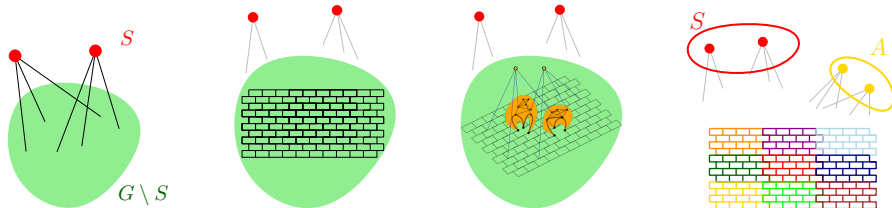
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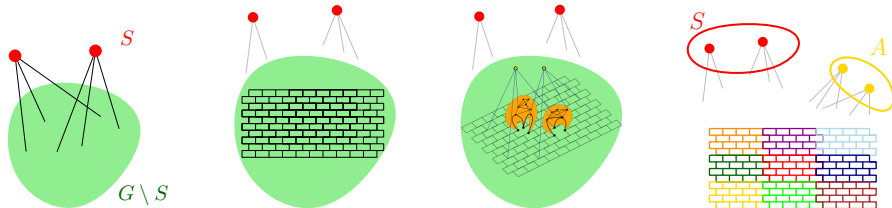
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Update $G = G \setminus v$ and **repeat**.

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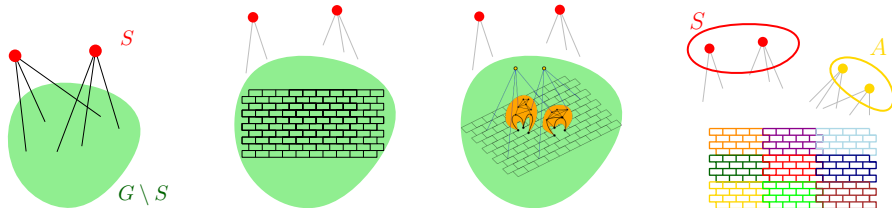
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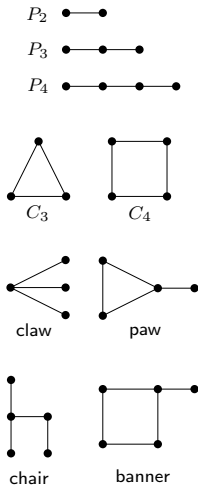
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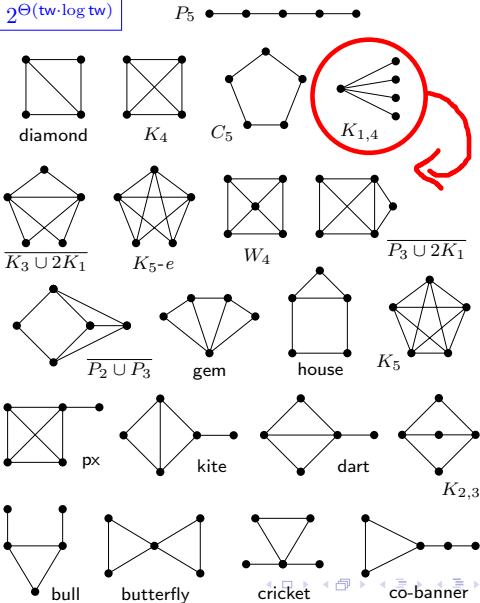
» skip

For topological minors, there is (at least) one change

$2^{\Theta(\text{tw})}$



$2^{\Theta(\text{tw} \cdot \log \text{tw})}$



Gràcies!
Toda raba!

