Treewidth: Vol. 3

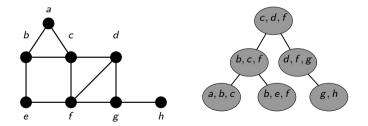
Dániel Marx

Lecture #13 February 1, 2022

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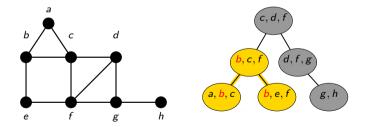
Tree decomposition: Vertices are arranged in a tree structure satisfying the following properties:

- **(**) If u and v are neighbors, then there is a bag containing both of them.
- 2 For every v, the bags containing v form a connected subtree.



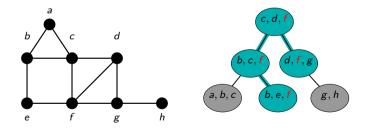
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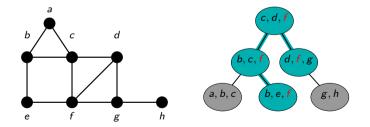


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Width of the decomposition: largest bag size -1.

treewidth: width of the best decomposition.

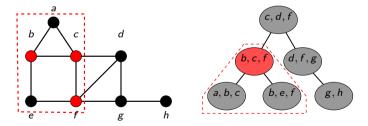


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A subtree communicates with the outside world only via the root of the subtree.

WEIGHTED MAX INDEPENDENT SET and treewidth

Theorem

Given a tree decomposition of width w, WEIGHTED MAX INDEPENDENT SET can be solved in time $O(2^w \cdot w^{O(1)} \cdot n)$.

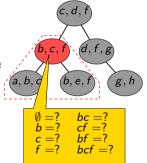
 B_{x} : vertices appearing in node x.

 V_{x} : vertices appearing in the subtree rooted at x.

Generalizing our solution for trees:

Instead of computing 2 values A[v], B[v] for each **vertex** of the graph, we compute $2^{|B_x|} \le 2^{w+1}$ values for each bag B_x .

M[x, S]: the max. weight of an independent set $I \subseteq V_x$ with $I \cap B_x = S$.



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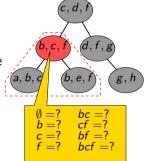
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How to determine M[x, S] if all the values are known for the children of x?

Monadic Second Order Logic

Extended Monadic Second Order Logic (EMSO)

A logical language on graphs consisting of the following:

- Logical connectives $\land,\,\lor,\,\rightarrow,\,\neg,\,=,\,\neq$
- quantifiers \forall , \exists over vertex/edge variables
- predicate adj(u, v): vertices u and v are adjacent
- predicate inc(e, v): edge e is incident to vertex v
- quantifiers \forall , \exists over vertex/edge set variables
- $\bullet \ \in, \ \subseteq \ for \ vertex/edge \ sets$

Example:

The formula

 $\exists C \subseteq V \forall v \in C \; \exists u_1, u_2 \in C(u_1 \neq u_2 \land \mathsf{adj}(u_1, v) \land \mathsf{adj}(u_2, v))$

is true on graph G if and only if G has a cycle.

Courcelle's Theorem

Courcelle's Theorem

There exists an algorithm that, given a width-w tree decomposition of an *n*-vertex graph *G* and an EMSO formula ϕ , decides whether *G* satisfies ϕ in time $f(w, |\phi|) \cdot n$.

If we can express a property in EMSO, then we immediately get that testing this property is FPT parameterized by the treewidth w of the input graph.

- \Rightarrow The following problem are FPT parameterized by treewidth:
 - *c*-Coloring
 - HAMILTONIAN CYCLE
 - Partition into Triangles
 - ...

SUBGRAPH ISOMORPHISM

Subgraph Isomorphism

Input: graphs H and GFind: a subgraph of G isomorphic to H.

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Input: graphs H and GFind: a subgraph of G isomorphic to H.

For each H, we can construct a formula ϕ_H that expresses "G has a subgraph isomorphic to H".

⇒ By Courcelle's Theorem, SUBGRAPH ISOMORPHISM can be solved in time $f(H, w) \cdot n$ if *G* has treewidth at most *w*.

Theorem

SUBGRAPH ISOMORPHISM is FPT parameterized by combined parameter k := |V(H)| and the treewidth w of G.

Finding tree decompositions

Fixed-parameter tractability:

Theorem [Bodlaender 1996]

There is a $2^{O(w^3)} \cdot n$ time algorithm that finds a tree decomposition of width w (if exists).

Sometimes we can get better dependence on treewidth using approximation.

FPT approximation:

Theorem

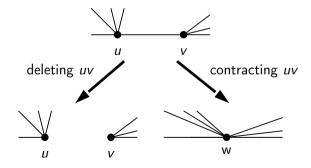
There is a $O(3^{3w} \cdot w \cdot n^2)$ time algorithm that finds a tree decomposition of width 4w + 1, if the treewidth of the graph is at most w.

Minor

An operation similar to taking subgraphs:

Definition

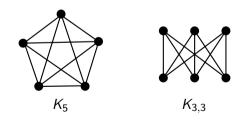
Graph *H* is a **minor** of *G* ($H \le G$) if *H* can be obtained from *G* by deleting edges, deleting vertices, and contracting edges.



A classical result

Theorem [Kuratowski 1930]

A graph G is planar if and only if G does not contain a subdivision of K_5 or $K_{3,3}$.

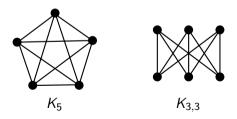


A classical result

Theorem [Kuratowski 1930]

A graph G is planar if and only if G does not contain a subdivision of K_5 or $K_{3,3}$.

Theorem [Wagner 1937] A graph G is planar if and only if G does not contain K_5 or $K_{3,3}$ as minor.



Graph Minors Theory





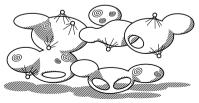
Neil Robertson

Paul Seymour

Theory of graph minors developed in the monumental series

Graph Minors I–XXIII. J. Combin. Theory, Ser. B 1983–2012

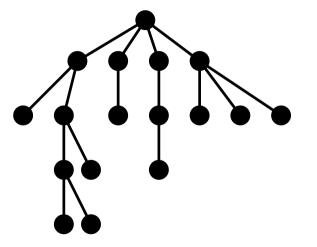
- Structure theory of graphs excluding minors (and much more).
- Galactic combinatorial bounds and running times.
- Important early influence for parameterized algorithms.

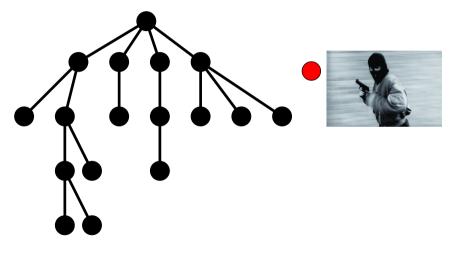


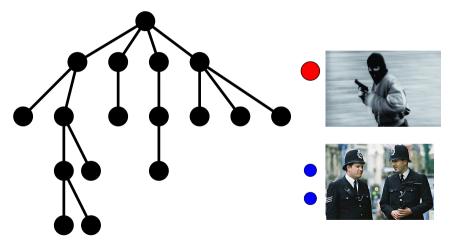
Fact: Treewidth does not increase if we delete edges, delete vertices, or contract edges. \Rightarrow If *F* is a **minor** of *G*, then the treewidth of *F* is at most the treewidth of *G*. Fact: Treewidth does not increase if we delete edges, delete vertices, or contract edges. \Rightarrow If *F* is a minor of *G*, then the treewidth of *F* is at most the treewidth of *G*. Fact: For every clique *K*, there is a bag *B* with $K \subseteq B$. Fact: The treewidth of the *k*-clique is k - 1. **Fact:** Treewidth does not increase if we delete edges, delete vertices, or contract edges. \Rightarrow If F is a **minor** of G, then the treewidth of F is at most the treewidth of G. **Fact:** For every clique K, there is a bag B with $K \subseteq B$. **Fact:** The treewidth of the k-clique is k - 1. **Fact:** For every $k \ge 2$, the treewidth of the $k \times k$ grid is exactly k.

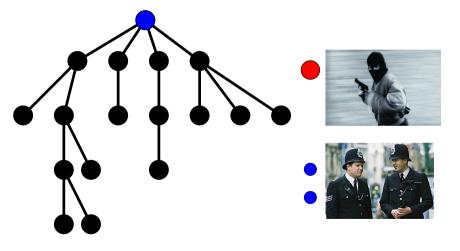
Game: *k* cops try to capture a robber in the graph.

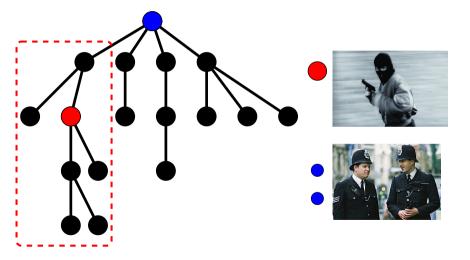
- In each step, (a subset of) the cops can move from vertex to vertex arbitrarily with helicopters.
- The robber moves infinitely fast on the edges, cannot move through the cops staying on the ground, and sees where the cops will land.

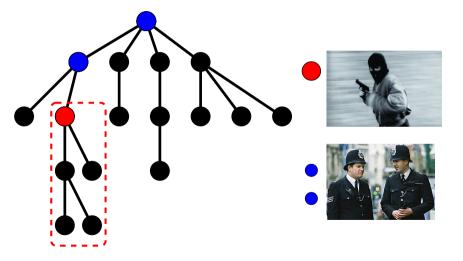


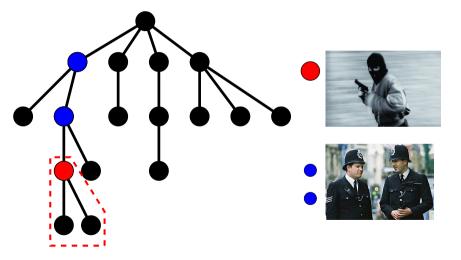


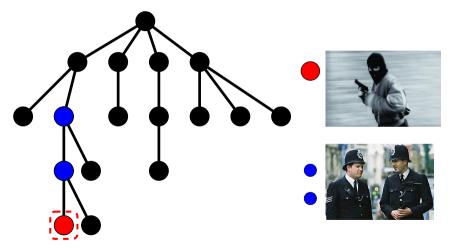


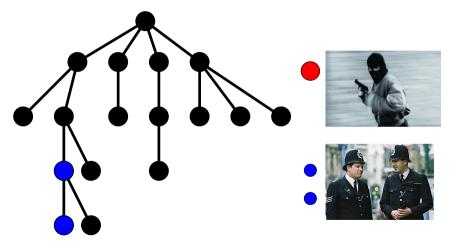


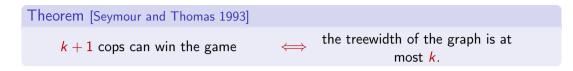














Consequence 1: Algorithms

The winner of the game can be determined in time $n^{O(k)}$ using standard techniques (there are at most n^k positions for the cops)

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For every fixed k, it can be checked in polynomial time if treewidth is at most k. (But $f(k) \cdot n^{O(1)}$ algorithms are also known with different techniques!)



Consequence 2: Lower bounds

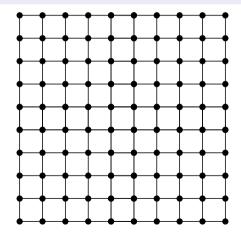
Exercise 1: Show that the treewidth of the $k \times k$ grid is at least k - 1. (E.g., robber can win against k - 1 cops.)

Exercise 2: Show that the treewidth of the $k \times k$ grid is at least k. (E.g., robber can win against k cops.)

Excluded Grid Theorem

Excluded Grid Theorem

If the treewidth of G is $\Omega(k^9 \log k)$, then G has a $k \times k$ grid minor.

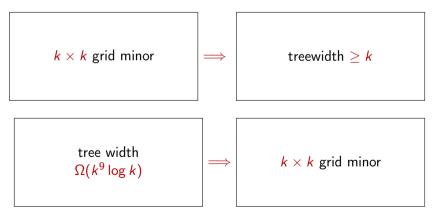


Excluded Grid Theorem

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If the treewidth of G is $\Omega(k^9 \log k)$, then G has a $k \times k$ grid minor.

A large grid minor is a "witness" that treewidth is large, but the relation is approximate:



Excluded Grid Theorem

Excluded Grid Theorem

If the treewidth of G is $\Omega(k^9 \log k)$, then G has a $k \times k$ grid minor.

Observation: Every planar graph is the minor of a sufficiently large grid.

Consequence

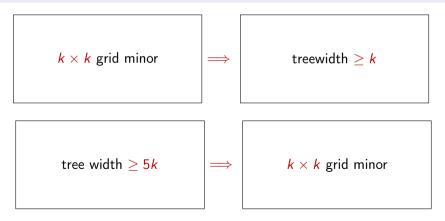
If *H* is planar, then every *H*-minor free graph has treewidth at most f(H).

Planar Excluded Grid Theorem

For planar graphs, we get linear instead of exponential dependence:

Theorem

Every **planar graph** with treewidth at least 5k has a $k \times k$ grid minor.



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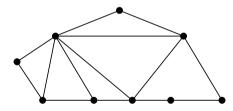
Theorem

An *n*-vertex planar graph has treewidth $O(\sqrt{n})$.

Outerplanar graphs

Definition

A planar graph is **outerplanar** if it has a planar embedding where every vertex is on the infinite face.

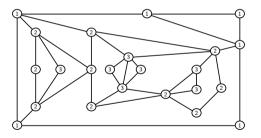


Fact

Given a planar embedding, we can define **layers** by iteratively removing the vertices on the infinite face.

Definition

A planar graph is k-outerplanar if it has a planar embedding having at most k layers.

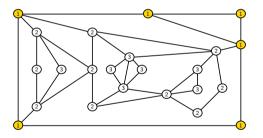


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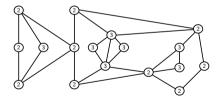


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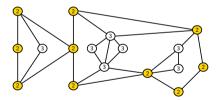


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Fact

Treewidth - outline

- Basic algorithms
- Omplexity results
- Ombinatorial properties

Applications

- The shifting technique
- Bidimensionality

Approximation schemes

Definition

A polynomial-time approximation scheme (PTAS) for a problem P is an algorithm that takes an instance of P and a rational number $\epsilon > 0$,

- always finds a $(1 + \epsilon)$ -approximate solution,
- the running time is polynomial in *n* for every fixed $\epsilon > 0$.

Typical running times: $2^{1/\epsilon} \cdot n$, $n^{1/\epsilon}$, $(n/\epsilon)^2$, n^{1/ϵ^2} .

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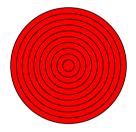
Typical running times: $2^{1/\epsilon} \cdot n$, $n^{1/\epsilon}$, $(n/\epsilon)^2$, n^{1/ϵ^2} .

Some classical problems that have a PTAS:

- INDEPENDENT SET for planar graphs
- $\bullet \ \mathrm{TSP}$ in the Euclidean plane
- $\bullet~{\rm Steiner}~{\rm Tree}$ in planar graphs
- KNAPSACK

Theorem

There is a $2^{O(1/\epsilon)} \cdot n$ time PTAS for INDEPENDENT SET for planar graphs.



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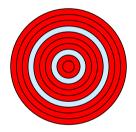
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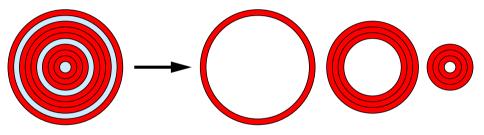
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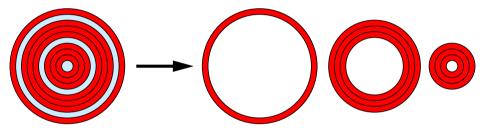
There is a $2^{O(1/\epsilon)} \cdot n$ time PTAS for INDEPENDENT SET for planar graphs.



- Let $D := 1/\epsilon$. For a fixed $0 \le s < D$, delete every layer L_i with $i = s \pmod{D}$
- The resulting graph is *D*-outerplanar, hence it has treewidth at most $3D + 1 = O(1/\epsilon)$.
- Using the 2^{O(tw)} · n time algorithm for INDEPENDENT SET, the problem on the D-outerplanar graph can be solved in time 2^{O(1/ε)} · n.

Theorem

There is a $2^{O(1/\epsilon)} \cdot n$ time PTAS for INDEPENDENT SET for planar graphs.



We do this for every $0 \le s < D$: for at least one value of s, we delete at most $1/D = \epsilon$ fraction of the solution

We get a $(1 + \epsilon)$ -approximate solution.

Subgraph Isomorphism

Input: graphs *H* and *G*

Find: a subgraph G isomorphic to H.



SUBGRAPH ISOMORPHISM

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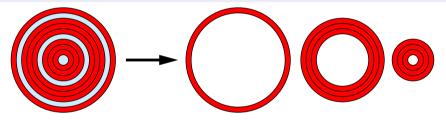
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SUBGRAPH ISOMORPHISM

- Input: graphs *H* and *G*
- Find: a subgraph G isomorphic to H.



- For a fixed $0 \le s < k + 1$, delete every layer L_i with $i = s \pmod{k + 1}$
- The resulting graph is k-outerplanar, hence it has treewidth at most 3k + 1.
- Using the $f(k, tw) \cdot n$ time algorithm for SUBGRAPH ISOMORPHISM, the problem can be solved in time $f(k, 3k + 1) \cdot n$.

SUBGRAPH ISOMORPHISM

Input: graphs H and GFind: a subgraph G isomorphic to H.



We do this for every $0 \le s < k + 1$: for at least one value of s, we do not delete any of the k vertices of the solution

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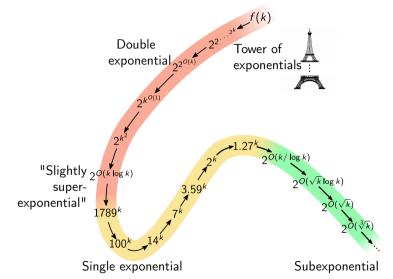
Theorem

SUBGRAPH ISOMORPHISM for planar graphs is FPT parameterized by k := |V(H)|.

• The technique is very general, works for many problems on planar graphs:

- INDEPENDENT SET
- VERTEX COVER
- Dominating Set
- **k**-Path
- ...
- More generally: First-Order Logic problems.
- But for some of these problems, much better techniques are known (see the following slides).

The race for better FPT algorithms



Square root phenomenon

Most NP-hard problems (e.g., 3-COLORING, INDEPENDENT SET, HAMILTONIAN CYCLE, STEINER TREE, etc.) remain NP-hard on planar graphs.¹

¹Notable exception: MAX CUT is in P for planar graphs.

Square root phenomenon

Most NP-hard problems (e.g., 3-COLORING, INDEPENDENT SET, HAMILTONIAN CYCLE, STEINER TREE, etc.) remain NP-hard on planar graphs.¹

The running time is still exponential, but significantly smaller:

 $2^{O(n)} \Rightarrow 2^{O(\sqrt{n})}$ $n^{O(k)} \Rightarrow n^{O(\sqrt{k})}$ $2^{O(k)} \cdot n^{O(1)} \Rightarrow 2^{O(\sqrt{k})} \cdot n^{O(1)}$

Example: A planar *n*-vertex graph has treewidth $2^{O(\sqrt{n})} \Rightarrow 3$ -COLORING can be solved in time $2^{O(\sqrt{n})}$ in planar graphs.

¹Notable exception: MAX CUT is in P for planar graphs.

VERTEX COVER

Theorem

VERTEX COVER can be solved in time $2^{O(\sqrt{k})} \cdot n^{O(1)}$ in planar graphs.

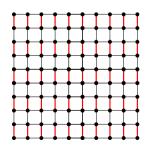
We need two facts:

- Removing an edge, removing a vertex, contracting an edge cannot increase the vertex cover number.
- VERTEX COVER can be solved in time $2^{w} \cdot n^{O(1)}$ if a tree decomposition of width w is given.

VERTEX COVER

Observation: If the treewidth of a planar graph *G* is at least $5\sqrt{2k}$

- \Rightarrow It has a $\sqrt{2k} \times \sqrt{2k}$ grid minor (Planar Excluded Grid Theorem)
- \Rightarrow The grid has a matching of size k
- \Rightarrow Vertex cover size is at least k in the grid.
- \Rightarrow Vertex cover size is at least k in G.



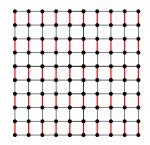
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- \Rightarrow Vertex cover size is at least k in G.

We use this observation to solve Vertex Cover on planar graphs:

- If treewidth is at least 5√2k: we answer "vertex cover is ≥ k."
- If treewidth is less than $5\sqrt{2k}$, then we can solve the problem in time $2^{O(5\sqrt{2k})} \cdot n^{O(1)} = 2^{O(\sqrt{k})} \cdot n^{O(1)}$.



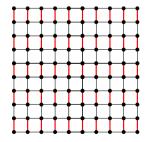
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- \Rightarrow Vertex cover size is at least k in G.

We use this observation to solve Vertex Cover on planar graphs:

- Set $w := 5\sqrt{2k}$.
- Find a 4-approximate tree decomposition.
 - If treewidth is at least *w*: we answer "vertex cover is ≥ *k*."
 - If we get a tree decomposition of width 4w, then we can solve the problem in time 2^{O(w)} ⋅ n^{O(1)} = 2^{O(√k)} ⋅ n^{O(1)}.



A powerful framework for efficient algorithms on planar graphs.

Setup:

- Let x(G) be some graph invariant (i.e., an integer associated with each graph).
- Given G and k, we want to decide if $x(G) \le k$ (or $x(G) \ge k$).
- Typical examples:
 - Maximum independent set size.
 - Minimum vertex cover size.
 - Length of the longest path.
 - Minimum dominating set size.
 - Minimum feedback vertex set size.

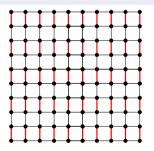
Bidimensionality

For many natural invariants, we can do this in time $2^{O(\sqrt{k})} \cdot n^{O(1)}$ on planar graphs.

Definition

A graph invariant x(G) is minor-bidimensional if

- $x(G') \le x(G)$ for every minor G' of G, and
- If G_k is the $k \times k$ grid, then $x(G_k) \ge ck^2$ (for some constant c > 0).

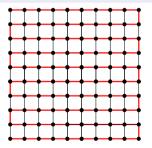


Examples: minimum vertex cover, length of the longest path, feedback vertex set are minor-bidimensional.

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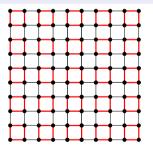


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Examples: minimum vertex cover, length of the longest path, feedback vertex set are minor-bidimensional.

Bidimensionality (cont.)

We can answer " $x(G) \ge k$?" for a minor-bidimensional invariant the following way:

- Set $w := c\sqrt{k}$ for an appropriate constant c.
- Use the 4-approximation tree decomposition algorithm.
 - If treewidth is at least w: x(G) is at least k.
 - If we get a tree decomposition of width 4w, then we can solve the problem using dynamic programming on the tree decomposition.

Running time:

- If we can solve the problem on tree decomposition of width w in time $2^{O(w)} \cdot n^{O(1)}$, then the running time is $2^{O(\sqrt{k})} \cdot n^{O(1)}$.
- If we can solve the problem on tree decomposition of width w in time $w^{O(w)} \cdot n^{O(1)}$, then the running time is $2^{O(\sqrt{k} \log k)} \cdot n^{O(1)}$.

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Exercise: DOMINATING SET is **not** minor-bidimensional.

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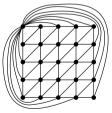
Exercise: DOMINATING SET is **not** minor-bidimensional.

We fix the problem by allowing only contractions but not edge/vertex deletions.

Theorem

Every planar graph with treewidth at least 9k + 5 can be contracted to a triangulated $k \times k$ grid Γ_k .

5×5 triangulated grid $\Gamma_5:$

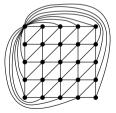


Definition

A graph invariant x(G) is contraction-bidimensional if

- $x(G') \leq x(G)$ for every contraction G' of G, and
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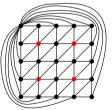


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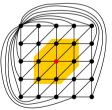
Example: maximum independent set, minimum dominating set are contraction-bidimensional.

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Example: maximum independent set, minimum dominating set are contraction-bidimensional.

Bidimensionality for $\operatorname{DOMINATING}\,\operatorname{SET}$

The size of a minimum dominating set is a contraction bidimensional invariant: we need at least $(k-2)^2/7$ vertices to dominate all the internal vertices of the triangulated $k \times k$ grid Γ_k (since a vertex can dominate at most 7 internal vertices).

Theorem

Given a tree decomposition of width w, DOMINATING SET can be solved in time $3^{w} \cdot w^{O(1)} \cdot n^{O(1)}$.

Solving DOMINATING SET on planar graphs:

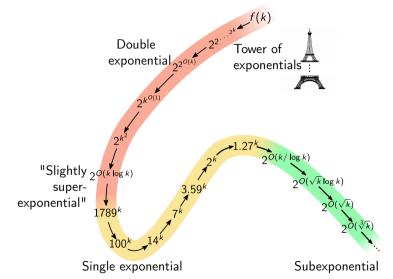
- Set $w := 9(3\sqrt{k} + 2)$.
- Use the 4-approximation tree decomposition algorithm.
 - If treewidth is at least w: we answer 'dominating set is $\geq k$ '.
 - If we get a tree decomposition of width 4w, then we can solve the problem in time $3^{w} \cdot n^{O(1)} = 2^{O(\sqrt{k})} \cdot n^{O(1)}$.

Beyond bidimensionality

The following problems are **not** minor/contraction-bidimensional, but admit $2^{O(\sqrt{k} \cdot \text{polylog}k)} \cdot n^{O(1)}$ time algorithms on planar graphs using other techniques:

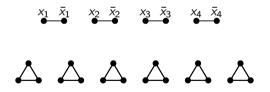
- Odd Cycle Transversal
- **k**-PATH in directed planar graphs
- Subset Feedback Vertex Set
- Multiway Cut
- SUBSET TSP (parameterized by the number of terminals)
- . . .

The race for better FPT algorithms



Exponential Time Hypothesis (ETH) + Sparsification Lemma There is no $2^{o(n+m)}$ -time algorithm for *n*-variable *m*-clause 3SAT.

The textbook reduction from 3SAT to VERTEX COVER:

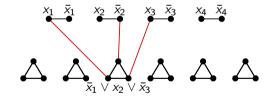


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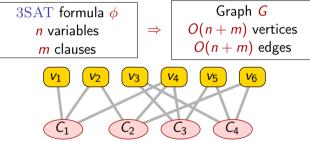
The textbook reduction from 3SAT to VERTEX COVER:

formula is satisfiable \Leftrightarrow there is a vertex cover of size n + 2m



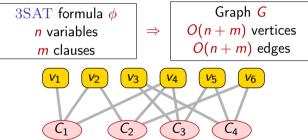
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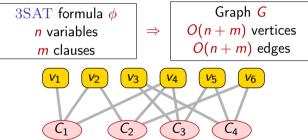


Corollary

Assuming ETH, there is no $2^{o(n)}$ algorithm for VERTEX COVER on an *n*-vertex graph.

Exponential Time Hypothesis (ETH) + Sparsification Lemma There is no $2^{o(n+m)}$ -time algorithm for *n*-variable *m*-clause 3SAT.

The textbook reduction from 3SAT to VERTEX COVER:



Corollary

Assuming ETH, there is no $2^{o(k)} \cdot n^{O(1)}$ algorithm for VERTEX COVER.

Other problems

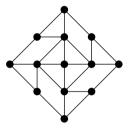
There are polytime reductions from 3SAT to many problems such that the reduction creates a graph with O(n + m) vertices/edges.

Consequence: Assuming ETH, the following problems cannot be solved in time $2^{o(n)}$ and hence in time $2^{o(k)} \cdot n^{O(1)}$ (but $2^{O(k)} \cdot n^{O(1)}$ time algorithms are known):

- VERTEX COVER
- Longest Cycle
- Feedback Vertex Set
- Multiway Cut
- Odd Cycle Transversal
- Steiner Tree
- . . .

What about 3-COLORING on planar graphs?

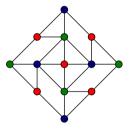
The textbook reduction from 3-COLORING to PLANAR 3-COLORING uses a "crossover gadget" with 4 external connectors:



- In every 3-coloring of the gadget, opposite external connectors have the same color.
- Every coloring of the external connectors where the opposite vertices have the same color can be extended to the whole gadget.
- If two edges cross, replace them with a crossover gadget.

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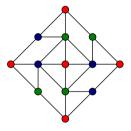
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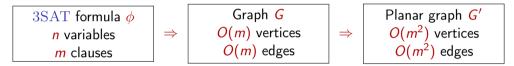
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- The reduction from 3-COLORING to PLANAR 3-COLORING introduces *O*(1) new edges/vertices for each crossing.
- A graph with m edges can be drawn with $O(m^2)$ crossings.



Corollary

Assuming ETH, there is no $2^{o(\sqrt{n})}$ algorithm for 3-COLORING on an *n*-vertex planar graph *G*.

Lower bounds for planar problems

Consequence: Assuming ETH, there is no $2^{o(\sqrt{n})}$ time algorithm on *n*-vertex **planar** graphs for

- INDEPENDENT SET
- Dominating Set
- VERTEX COVER
- HAMILTONIAN PATH
- Feedback Vertex Set
- ...

Lower bounds for planar problems

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An exceptional lower bound

STEINER TREE: Given a graph G and set T of terminals, find a tree of minimum size that contains T.

We have seen:

- Can be solved in time $3^{|T|} \cdot n^{O(1)}$ using dynamic programming.
- Can be solved in time $2^{|T|} \cdot n^{O(1)}$ using algebraic techniques.

Is there a subexponential FPT algorithm on planar graphs?

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Is there a subexponential FPT algorithm on planar graphs?

Theorem

Assuming ETH, STEINER TREE on planar graphs with k terminals cannot be solved in time $2^{o(k)} \cdot n^{O(1)}$.

Treewidth — summary

- Notion of treewidth: widely used in graph theory and parameterized algorithms.
- Efficient algorithms parmeterized by treewidth.
- Applications e.g. to planar graphs.

Treewidth

Tree decomposition: Vertices are arranged in a tree structure satisfying the following properties:

() If u and v are neighbors, then there is a bag containing both of them.

2 For every v, the bags containing v form a connected subtree.

Width of the decomposition: largest bag size -1.

treewidth: width of the best decomposition.

