

43ème École de Printemps d'Informatique Théorique (ÉPIT)
Graphes et surfaces : algorithmique, combinatoire et topologie

Speaker: Dimitrios Thilikos

CIRM, Marseille, May 09–13, 2016

Lectures:

- ▶ Monday 09/05/2016 - 11:45–12:30 (45'): **22 pages**
Complexity, Graphs, P vs. NP, NP-Hardness, Fixed-parameter tractability
- ▶ Monday 09/05/2016 - 16:00–17:30 (90'): **42 pages**
treewidth, dynamic programming
- ▶ Tuesday 10/05/2016 - 16:00–17:00 (60'): **26 pages**
Sphere-cut decompositions
- ▶ Thursday 10/05/2016 - 11:00–12:30 (90'): **31 pages**
Bidimensionality and subexponential algorithms
- ▶ Friday 10/05/2016 - 09:00–10:30 (90'): **14+20=34 pages**
Bidimensionality and Kernels + Irrelevant vertex technique

155 pages, **375 Minutes** in total, 145 seconds per page (2.42 minutes per page)

Part 1, Monday 09/05/2016 - 11:45–12:30 (45')

Complexity,

Graphs,

P vs. NP,

NP-Hardness,

Fixed-parameter tractability

O-notation

$f(n) = O(g(n)) \iff$ there exist constants N and C such that

$$\forall n > N \quad f(n) \leq C \cdot g(n).$$

► Examples:

$2016 = O(1) = O(3)$: Constant

$4 \log_4 n + 5 \log_3 n = O(\log n)$: Logarithmic

$23 \log^5 n = O(\log^5 n) = \log^{O(1)} n$: polylogarithmic

$2016\sqrt{n} + 3016 = O(\sqrt{n})$: sublinear

$2016n + 3016 = O(n)$: linear

$23n^2 + 16n + 4 = O(n^2)$: quadratic

$89n^2 + 6n^2 - 4n + 45676 = O(n^3)$: cubic

$5n^c = O(n^c) = n^{O(1)}$: polynomial

$n^3 4 \cdot 3\sqrt{n} + n^{2016} = 2^{O(\sqrt{n})}$: subexponential

$30^{7n+3} = 2^{O(n)}$: singly exponential

Problems and algorithms

Definition of an problem:

A set of YES-instances $\Pi \subseteq \Sigma^*$ where Σ is an alphabet, typically $\Sigma = \{0, 1\}$

We look for a way to decide, given a $x \in \Sigma^*$, whether $x \in \Pi$.

Definition of an algorithm:

Muhammad ibn Musa al-Khwarizmi



Alan Mathison Turing



Omitted! This journey is beautiful and long. We do not take it this week! 😞

Graphs

We mostly work on algorithms on **graphs**:

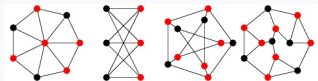
$V(G)$: vertices of G , $E(G)$: edges of G , \mathcal{G}_{all} : the set of all graphs

A problem on graphs:

VERTEX COVER

Instance: A graph G and an integer $k \geq 0$.

Question: $\exists S \subseteq V(G) : |S| \leq k \wedge \forall e \in E(G) |e \cap S| \geq 1$?



Here:

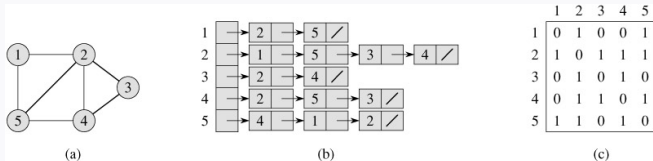
Input: $x = \langle G, k \rangle$, i.e., x encodes the graph G and the integer k

Problem: $\Pi_{\text{vc}} = \{ \langle G, k \rangle \mid G \text{ has a vertex cover of size } \leq k \}$

Algorithm for VERTEX COVER: a ~~procedure~~ that receives as input $x = \langle G, k \rangle$ and outputs whether $x \in \Pi_{\text{vc}}$

Data structures

Data structures for graphs: [Adjacency list](#), [Adjacency matrix](#)



► Running time of a graph algorithm =

of **elementary** operations on the data structure that represents its input.

Most graphs in this lectures are **sparse**: $|E(G)| = O(|V(G)|)$ (By Euler's formula)

For this reason we prefer the **Adjacency list** data structure.

We also assume that arithmetic operations take $O(1)$ steps!



We measure the **time complexity** of a graph algorithm by as a function of $n = |V(G)|$

Complexity classes

P: contains all problems that can be solved in $n^{O(1)}$ steps

NP: contains all problems that can be “certified” in $n^{O(1)}$ steps

VERTEX COVER

Instance: A graph G and an integer $k \geq 0$.

Question: $\exists S \in V(G) : |S| \leq k \wedge \forall e \in E(G) |e \cap S| \geq 1$?

S is the certificate.

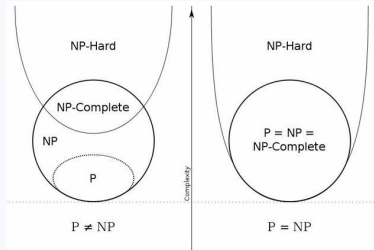
If we “guess” S then we can check whether it is a vertex cover of G in $n^{O(1)}$ steps.

Clearly: $P \subseteq NP$ 10⁶\$-question: is it correct that $P \neq NP$?

A problem Π is **NP-hard** if it is “as hard as”
all problems in **NP**

A problem Π is **NP-complete** if it is **NP-hard**
and belongs in **NP**

VERTEX COVER is an **NP-complete** problem



Parameterized Complexity

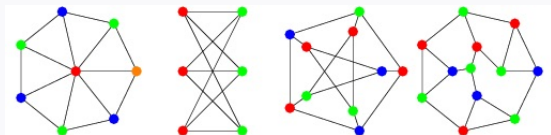
- ▶ Most of interesting problems are NP-hard!
- ▶ We care whether this situation can be “improved”, taking into account structural characteristics of the input graph (for instance: topological properties).

We have to define what “improved” might mean here!



Parameterized complexity was introduced by [Mike Fellows](#) and [Rod Downey](#) proposed a way to refine the above landscape!

Three NP-complete problems



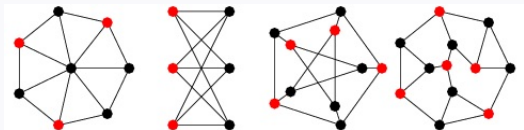
VERTEX COLORING

Instance: A graph G and an integer $k \geq 0$.

Question: $\exists \sigma : V(G) \rightarrow \{1, \dots, k\} : \forall \{v, u\} \in E(G) \sigma(v) \neq \sigma(u)$?

It can be solved in $O(n^2 \cdot k^n)$ steps

Three NP-complete problems



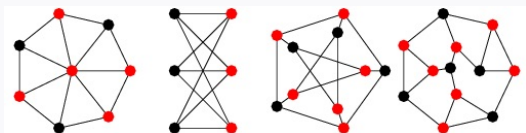
INDEPENDENT SET

Instance: A graph G and an integer $k \geq 0$.

Question: $\exists S \in V(G) : |S| \geq k \wedge \forall e \in E(G) |e \cap S| \leq 1$?

It can be solved in $O(n^{k+1})$ steps

Three NP-complete problems



VERTEX COVER

Instance: A graph G and an integer $k \geq 0$.

Question: $\exists S \subseteq V(G) : |S| \leq k \wedge \forall e \in E(G) |e \cap S| \geq 1$?

It can easily be solved in $O(2^k \cdot n)$ steps

It can be solved in $O(1.2738^k + k \cdot n)$ steps [Chen, Kanj, Xia, 2010]

Comparisons

Summary:

VERTEX COLORING	$O(n^2 \cdot k^n)$	ugly
INDEPENDENT SET	$O(n^{k+1})$	bad
VERTEX COVER	$O(2^k \cdot n)$	good

Different **Interleavings** between the parameter k and the main part n of the input.

Comparison between $O(2^k \cdot n)$ and $O(n^{k+1})$

	$n = 50$	$n = 100$	$n = 150$
$k = 2$	625	2.500	5.625
$k = 3$	15.625	125.000	421.875
$k = 5$	390.625	6.250.000	31.640.623
$k = 10$	$1,9 \times 10^{12}$	$9,8 \times 10^{14}$	$3,7 \times 10^{16}$
$k = 20$	$1,8 \times 10^{26}$	$9,5 \times 10^{31}$	$2,1 \times 10^{35}$

The ratio $\frac{n^{k+1}}{2^k \cdot n}$ for several values of n and k .

How the parameters appear?

VLI design: In VLSI chip construction, the number of circuit layers is no more than 10. While the problem is, in general, NP-complete, when we fix the number of layers, it becomes tractable.

Computational Biology: in general, many problems in DNA chain reconstruction are intractable. In the majority of the cases, real instances have special properties (e.g., bounded treewidth or pathwidth – by 11) that facilitate the design of efficient algorithms.

How the parameters appear?

Robotics: The number of degrees of freedom in motion planning problems are not more than 10. While these problems are NP-complete in general, they become tractable taking into account this natural restriction.

Compilers: One of the main tasks of a compiler for the language ML is the compatibility checking of type declarations of the program. It is known that the general problem is EXP-complete. However, in real cases, the implementations work well as there is an algorithm with complexity $O(2^k \cdot n)$, where n is the size of the program and k is the depth of its type declarations. As, normally, $k \leq 10$, the problem can be considered tractable.

Parameterized problems

Given an alphabet Σ ,

- (1) A *parameterization* of Σ^* is a recursive function $\kappa : \Sigma^* \rightarrow \mathbb{N}$
- (2) A *parameterized problem* (with respect to Σ) is a pair (Π, κ) where $\Pi \subseteq \Sigma^*$ and κ is a parameterization of Σ^* .

Examples

A parameterization of INDEPENDENT SET can be defined as $\kappa(G, k) = k$.

We can do the same with all the problems that have some integer in their instances, such as VERTEX COLORING and VERTEX COVER.

That way, we define the parameterized problems p -VERTEX COLORING and p -VERTEX COVER.

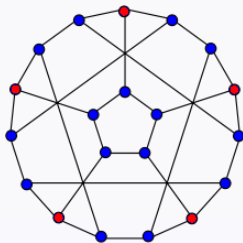
Other parameterizations of the above problems can be defined as

$$\kappa(G, k) = \Delta(G) \text{ or}$$

$$\kappa(G, k) = \text{genus}(G)$$

$$\kappa(G, k) = \Delta(G) + k$$

Some parameterized problems



p-DOMINATING SET

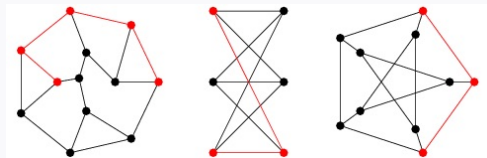
Instance: A graph G and an integer $k \geq 0$

Parameter: k ,

Question:

$\exists S \in V(G) : |S| \leq k \wedge \forall v \in V(G) - S \exists u \in S \{v, u\} \in E(G)?$

Some parameterized problems



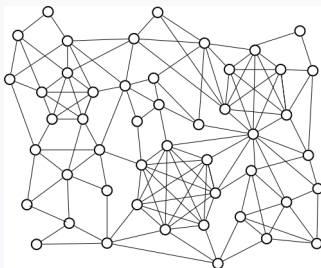
p -PATH

Instance: A graph G and an integer $k \geq 0$.

Parameter: k

Question: Does G contain a path of length k ?

Some parameterized problems



p-CLIQUE

Instance: A graph G and an integer $k \geq 0$.

Parameter: k ,

Question: $\exists S \in V(G) : |S| \leq k \wedge \forall v, u \in S \{v, u\} \in E(G)$?

More parameterized problems

p-STEINER TREE

Instance: A graph G , $S \subseteq V(G)$, $k \in \mathbb{N}$.

Parameter: k

Question: $\exists R \subseteq V(G) : |R| \leq k, R \cap S = \emptyset, G[S \cup R]$ is connected?

Here $\kappa(G, S, k) = k$

p'-STEINER TREE

Instance: A graph G , $S \subseteq V(G)$, $k \in \mathbb{N}$.

Parameter: $|S|$

Question: $\exists R \subseteq V(G) : |R| \leq k, R \cap S = \emptyset, G[S \cup R]$ is connected?

Here $\kappa(G, S, k) = |S|$

The class FPT

Given an alphabet Σ and a parameterization $\kappa : \Sigma^* \rightarrow \mathbb{N}$,

- (a) An algorithm A is a *FPT-algorithm with respect to κ* if there is a function *computable* $f : \mathbb{N} \rightarrow \mathbb{N}$ and a *polynomial* function $p : \mathbb{N} \rightarrow \mathbb{N}$ such that for every $x \in \Sigma^*$, the algorithm A requires

$$\leq f(\kappa(x)) \cdot p(|x|) \text{ steps}$$

- (b) A parameterized problem (L, κ) is *fixed parameter tractable* if there exists an *FPT-algorithm* with respect to κ that decides L .

► We then say that $(L, \kappa) \in \text{FPT}$ or, more precisely, *f-FPT*

► The function f is called *parameterized dependence* of the running time of the *FPT-algorithm*

An algorithm for VERTEX COVER:

We set up a search tree with depth depending *only* on the parameter k .

[Bounded Search Tree Method]

$\text{algvc}(G, k)$

1. If $|E(G)| = 0$, then return "YES"
 2. If $k = 0$, then return "NO"
 3. choose (arbitrarily) an edge $e = \{v, u\} \in E(G)$ and
return $\text{algvc}(G - v, k - 1) \vee \text{algvc}(G - u, k - 1)$
-

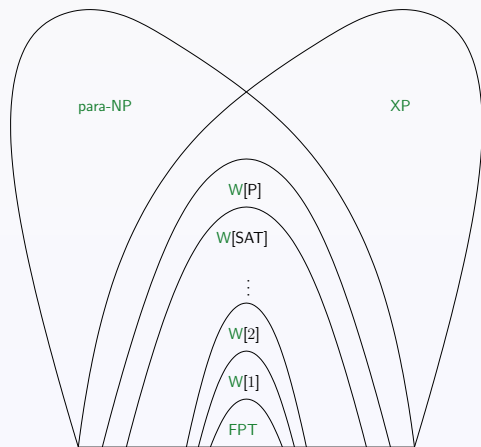
Recursive calls: 2, Depth of the recursion: k ,

Time in the leaves of the recursion: $O(n)$ steps

Total time: $O(2^k \cdot n)$ steps.

Therefore, p -VERTEX COVER $\in 2^{O(k)}$ -FPT.

Panorama of Parameterized complexity classes



p-VERTEX COVER: **FPT**

p-PATH: **FPT**

p'-STEINER TREE: **FPT**

p-CLIQUE: **W[1]-complete**

p-INDEPENDENT SET: **W[1]-complete**

p-DOMINATING SET: **W[2]-complete**

p-STEINER TREE: **W[2]-complete**

p-COLORING: **para-NP-complete**

Part 2, Monday 09/05/2016 - 16:00–17:30 (90')

Tree decompositions

Treewidth

Courcelle's Theorem

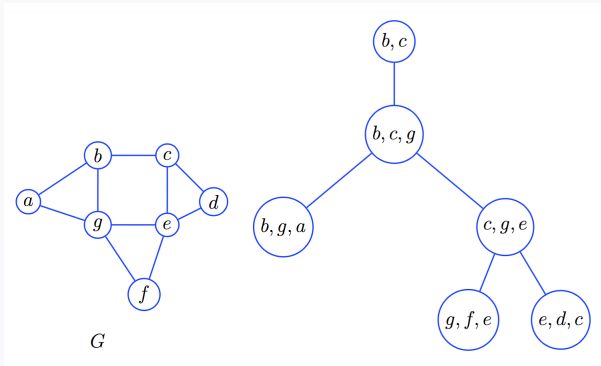
Dynamic programming

Tree decompositions

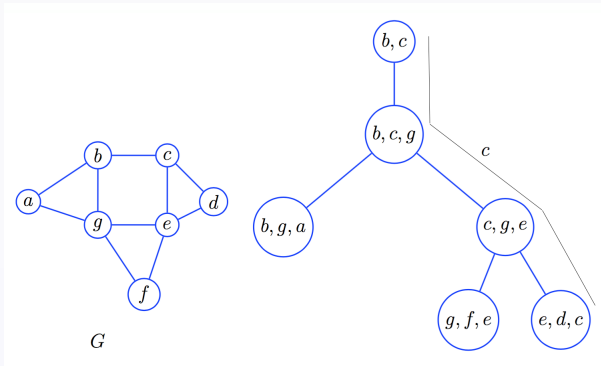
A *tree decomposition* (ou *décomposition arborescente*) of a graph G is a pair $D = (T, \mathcal{X})$ such that T is a tree and $\mathcal{X} = \{X_t \mid t \in V(T)\}$ is a collection of subsets of G . such that:

1. Any vertex $v \in V(G)$ and the end points of any edge $e \in E(G)$ belong in some node X_t of D
2. For any $v \in V(G)$, the set $\{t \in V(T) \mid v \in X_t\}$ is a subtree of T .

-
- $X_t \in \mathcal{X}$ corresponds to a vertex $t \in V(T)$ – X_t is a *node/bag* of D
 - The *width* of a tree decomposition (T, \mathcal{X}) is $\max_{t \in V(T)} |X_t| - 1$
 - The *tree-width* (ou *largeur arborescente* ou *largeur d'arbre*) of a graph G ($\text{tw}(G)$) is the *minimum* width over all tree decompositions of G



Each vertex of G has a **continuous** “trace” in the tree of the tree decomposition

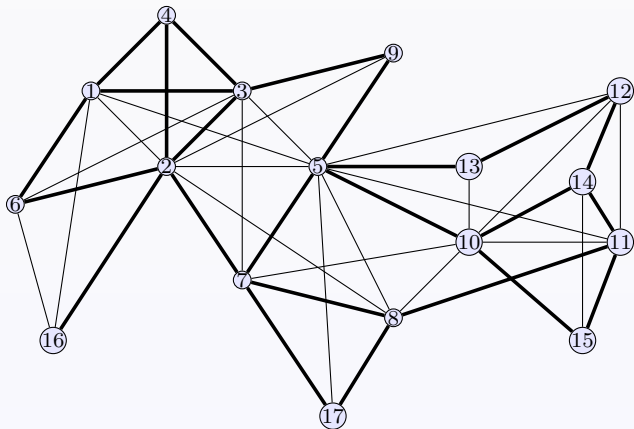


Each vertex of G has a **continuous** “trace” in the tree of the tree decomposition

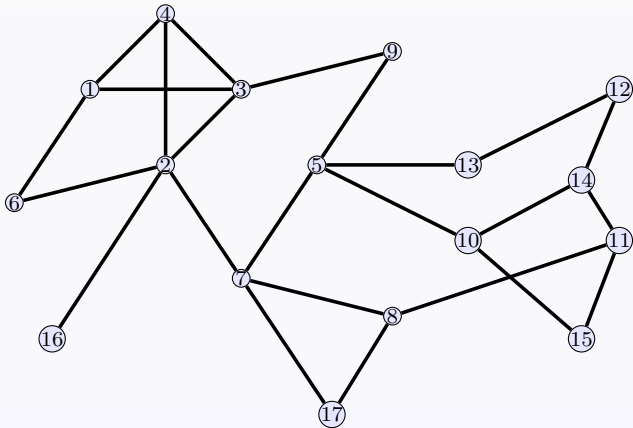
Another definition for Treewidth

- ▶ A vertex in G is k -simplicial if its neighborhood induces a k -clique.
- ▶ A graph G is a k -tree if one of the following holds
 - ▶ $G = K_{k+1}$ or
 - ▶ the removal of G of a k -simplicial vertex creates a k -tree.
- ▶ The treewidth of a graph G is defined as follows

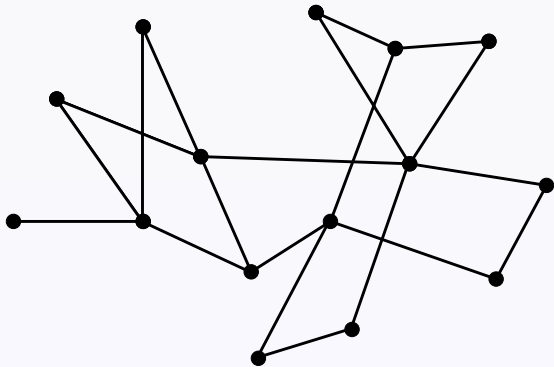
$$\text{tw}(G) = \min\{k \mid G \text{ is a subgraph of some } k\text{-tree}\}$$

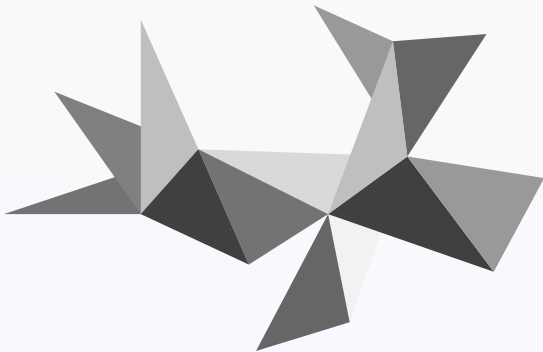


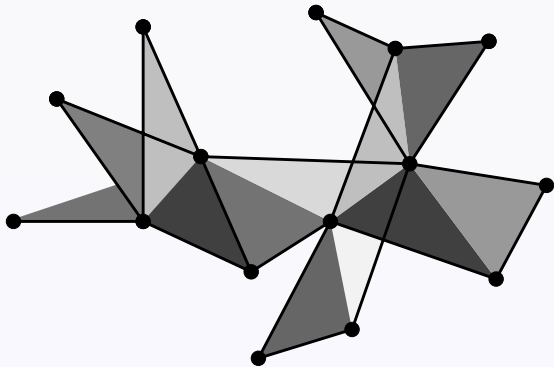
A 3-tree

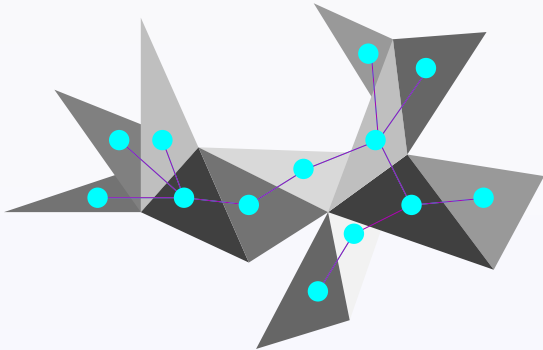


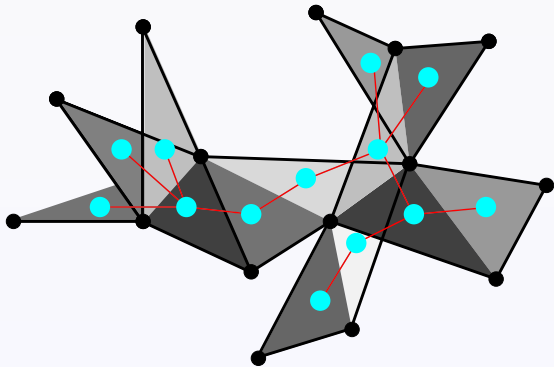
A subgraph of a **3-tree**: a graph with **treewidth** at most 3











Facts about treewidth

- ▶ Defined for the first time by Bertelé & Brioschi on 1972 under the name *dimension*
- ▶ Named *treewidth* by Robertson and Seymour in GM-II on 1986.
- ▶ There are more alternative definitions of treewidth (at least **six!**)
- ▶ Treewidth can be seen as a measure of the **topological similarity** of a graph to a tree
- ▶ Treewidth is **important** in algorithm design (not only there)
- ▶ Many **NP-hard** problems on graphs become **polynomially solvable** when their instances are restricted to graphs with constant treewidth.

Parameterizing treewidth

p -TREEWIDTH

Instance: A graph G and an integer $k \geq 0$.

Parameter: k

Question: $\text{tw}(G) \leq k$?

p -TREEWIDTH is in FPT by an $2^{O(k^3)} \cdot O(n)$ algorithm of Bodlaender
[SIAM J. Comp., 1996]

Monadic Second Order Logic

► A property in graphs may be expressed in **MSO** Logic

Universe: the vertex set V of the graph $G = (V, E)$

An **MSO** formula can be build using:

Variables: vertices x, y, z, \dots and sets of vertices X, Y, Z, \dots

Atomic Formulae: $x = y, x \in X, \{x, y\} \in E \quad (E(x, y))$

Formulae: $\neg x, x \vee y, x \wedge y, x \rightarrow y, x \leftrightarrow y, \exists x\phi, \forall x\phi, \exists X\phi, \forall X\phi,$

Examples of properties expressible in MSO

3-Colorability:

$$\begin{aligned} & \exists R \exists G \exists B [\forall x [(x \in R \vee x \in G \vee x \in B) \wedge \\ & \neg(x \in R \wedge x \in G) \wedge \neg(x \in B \wedge x \in G) \wedge \neg(x \in R \wedge x \in B)]] \\ & \wedge \neg[\exists x \exists y (\{x, y\} \in E \wedge \\ & ((x \in R \wedge y \in R) \vee (x \in G \wedge y \in G) \vee (x \in B \wedge y \in B)))] \end{aligned}$$

Examples of properties expressible in MSO

Having an clique of size $\geq k$:

$$\exists x_1 \exists x_2 \cdots \exists x_k \bigwedge_{1 \leq i < j \leq k} \{x_i, x_j\} \in E$$

Examples of properties expressible in MSO

Having an independent set of size k :

$$\exists x_1 \exists x_2 \cdots \exists x_k \bigwedge_{1 \leq i < j \leq k} (\neg \{x_i, x_j\} \in E) \wedge \neg(x_i \neq x_j)$$

Examples of properties expressible in MSO

Having a vertex cover of size k :

$$\exists x_1 \exists x_2 \cdots \exists x_k (\forall x \forall y \{x, y\} \in E \rightarrow (\bigvee_{1 \leq i \leq k} (x = x_i \vee y = x_i)))$$

Examples of properties expressible in MSO

Having a dominating set of size k :

$$\exists x_1 \exists x_2 \cdots \exists x_k \forall y \bigvee_{1 \leq i \leq k} (\{x_i, y\} \in E \vee y = x_i)$$

Courcelle's theorem

MSO: Monadic Second Order Logic

Theorem: [Courcelle], [Seese], & [Borie, Parker & Tovey] *Every problem on graphs that can be expressed by a **MSO** formula ϕ can be solved in $f(\text{tw}(G), |\phi|) \cdot n$ steps.*



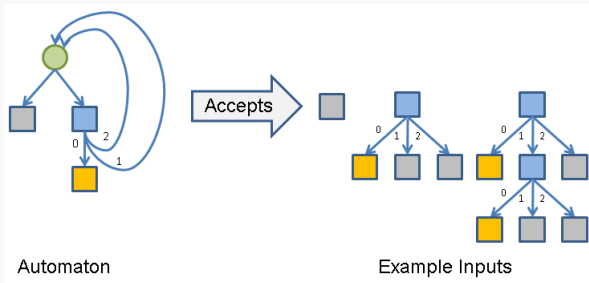
In other words:

If $\Pi \subseteq \mathcal{G}_{\text{all}}$ is a **MSO**-expressible set, then $(\Pi, \text{tw}) \in \text{FPT}$

or

Every **MSO**-expressible problem of graphs is fixed parameter tractable when parameterized by the treewidth of its input graph

Inputs of **small treewidth** can be seen as **tree-string**: inputs of a tree-automaton generated by the **MSO** formula expressing \mathcal{G} .



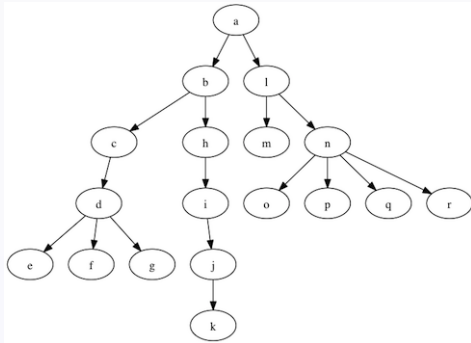
Theorem: *Every problem on graphs that can be expressed by a MSO formula ϕ can be solved in $f(\text{tw}(G), |\phi|) \cdot n$ steps.*

► Courcelle proved a stronger version where quantification on sets of edges is also allowed.

Advantage of Courcelle's Theorem: It constructs the algorithm

Drawback of Courcelle's Theorem: the contribution of the formula and the treewidth in the running time is immense.

In **topological** terms: treewidth helps us treat the input graph a **mono-dimensional** entity!



Treewidth is a measure of the possibility of recursively **cutting** the graph in smaller pieces and process them separately:

In an algorithmic terms: **Divide and Conquer!**

Which in our case is: **Dynamic Programming**

Nice tree decompositions

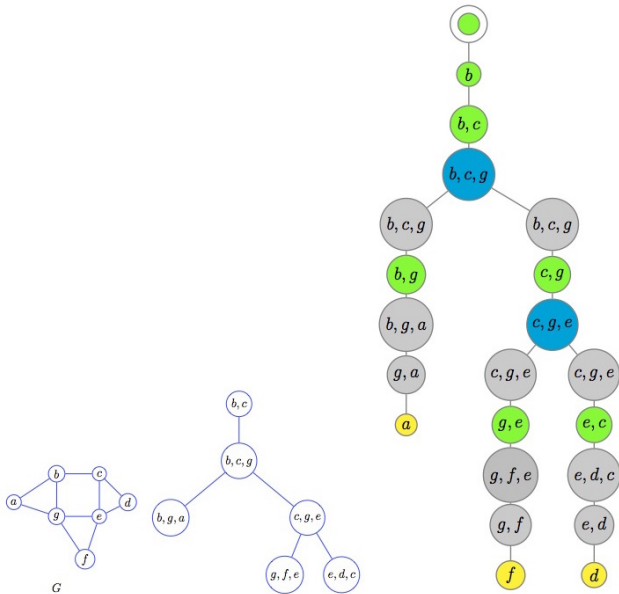
A tree decomposition $D = (T, \mathcal{X})$ is *nice* if T is *rooted* to some leaf r and

- ▶ for any leaf l of T where $l \neq r$, $X_l = \emptyset$
(we call X_l *leaf node* of D except from X_r that we call *root node*)
- ▶ any non-leaf $t \in V(T)$ (including the *root*) has one or two children.
- ▶ if t has two children t_1 and t_2 then, $X_t = X_{t_1} = X_{t_2}$
(we call X_t *join node*)
- ▶ if t has one child t' then
 - ▶ either $X_t = X_{t'} \cup \{v\}$
(we call X_t *insert node* and v is the *insert vertex*)
 - ▶ or $X_{t'} = X_t \cup \{v\}$
(we call X_t *forget node* and v is the *forget vertex*)

If (T, \mathcal{X}) is a nice tree decomposition rooted on r , then

for any $t \in V(T)$, $G_t = G[\cup_{t' \text{ is } t \text{ or a descendant of } t \text{ in } T} X_{t'}]$

Lemma: *There exists an $O(n)$ -step algorithm that transforms any tree decomposition with n nodes to a nice tree decomposition of $\leq 4n$ nodes of the same width.*



A graph G , a tree decomposition, and a nice tree decomposition

How to do dynamic programming for graphs of small treewidth

1. Define, for each $t \in V(T)$, a table that **encodes** the information of a partial solution for G_t . The values of this table for the **root** node should provide a global answer.
2. Define the values of this table for the **leaf** nodes.
3. Provide the way to compute the table of an **insert** node, given the table of its child.
4. Provide the way to compute the table of a **forget** node, given the table its child.
5. Provide a way to compute the table of a **join** node, given the tables of its children.

Parameterizing 3-COLORING by treewidth

tw-3-VERTEX COLORING

Instance: A graph G .

Parameter: $k = \mathbf{tw}(G)$

Question: $\exists \chi : V(G) \rightarrow \{1, 2, 3\} : \forall \{v, u\} \in E(G) \chi(v) \neq \chi(u)?$

For any $\chi : S \rightarrow I$ and $R \subseteq S$, we define $\chi[R] = \{(v, \chi(v)) \in \chi \mid v \in R\}$

1st step: Definition of the tables:

For any $t \in V(T)$ and any 3-coloring $\phi : X_t \rightarrow \{1, 2, 3\}$, we define

$$B_t(\phi) = [\exists \chi : V(G_t) \rightarrow \{1, 2, 3\} \text{ such that } \chi[X_t] = \phi]$$

(the table of t contains an array of $3^{|X_t|}$ bits)

$G = G_r$ is 3-colourable iff $B_r(\emptyset) = 1$

2nd step: tables for leaf nodes:

Let X_l be an leaf node

we have

$$B_l(\emptyset) = 1$$

3rd step: tables for insert nodes:

Let X_t be an insert node

let t' be the child of t and v be the insert vertex.

For any $\phi : X_t \rightarrow \{1, 2, 3\}$, we have

$$B_t(\phi) = B_{t'}(\phi - (v, \phi(v))) \bigwedge_{u \in N_{G_t}(v)} [\phi(v) \neq \phi(u)]$$

4th step: tables for forget nodes:

Let X_t be a forget node

let t' be the child of t and v be the forget vertex.

For any $\phi : X_t \rightarrow \{1, 2, 3\}$, we have

$$B_t(\phi) = \bigvee_{i \in \{1, 2, 3\}} B_{t'}(\phi \cup \{v, i\})$$

5th step: tables for join nodes:

Let X_t be an join node

let t_1, t_2 be the children of t

For any $\phi : X_t \rightarrow \{1, 2, 3\}$, we have

$$B_t(\phi) = B_{t_1}(\phi) \wedge B_{t_2}(\phi)$$

Conclusion:

Given a tree decomposition of G ,

the following tw -3-VERTEX-COLORING problem is in $2^{O(k)}$ -FTP:

(we gave an $O(3^k \cdot k \cdot n)$ dynamic programming algorithm)

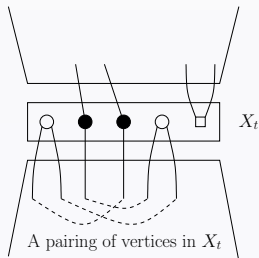
Parameterizing HAMILTONIAN CYCLE by treewidth:

tw-HAMILTONIAN CYCLE

Instance: A graph G .

Parameter: $k = \text{tw}(G)$

Question: does G contain a spanning cycle?



A **pairing** of X_t is a graph H (with loops) s.t.

$$V(G) = X_i \text{ and } \forall x \in X_i \text{ deg}_H(x) \leq 2$$

- ▶ The restriction of a cycle to G_t is a collection \mathcal{P} of internally disjoint paths in G_t with ends in X_i .
- ▶ Each \mathcal{P} corresponds to some pairing $H_{\mathcal{P}}$ of X_t
- ▶ For any set S , let **pairs**(S) be the set of all pairings of S

Let (T, \mathcal{X}) be a tree decomposition of G where $X_r = \{w\}$

let H_w be just the vertex w looped.

1st Step: For each $t \in V(T)$ we define:

$\forall H \in \text{pairs}(X_i),$

$$B_t(H) = [H \text{ is the pairing of some } t\text{-path collection } \mathcal{P}]$$

$G = G_r$ has a Hamiltonian cycle iff $B_r(H_w) = 1$

2nd step: tables for leaf nodes:

Let X_l be an leaf node (assume that $X_l = \{y\}$)

Notice that $\text{pairs}(t) = \{H_0, H_1\}$

where $H_0(H_1)$ is the vertex y looped (unlooped)

$$\forall H \in \text{pairs}(t) \quad B_l(H) = [|E(H)| = 0]$$

3rd step: tables for insert nodes:

Let X_t be an insert node

let t' be the child of t and v be the insert vertex.

For any $\forall H \in \text{pairs}(t)$ we have

$$B_t(H) = [B_{t'}(H - v)] \wedge [N_H(v) \subseteq N_{G_t}(v)]$$

4st step: tables for forget nodes:

Let X_t be an forget node

let t' be the child of t and v be the insert vertex.

For any $\forall H \in \text{pairs}(t)$ we have

$$B_t(H) = \bigvee_{\substack{H' \in \text{pairs}(t') \\ H \text{ is a contraction of } H'}} B_{t'}(H')$$

5th step: tables for join nodes:

Let X_t be an join node

let t_1, t_2 be the children of t

For any $\forall H \in \text{pairs}(t)$ we have

$$B_t(H) = \bigvee_{\substack{H_1 \in \text{pairs}(t_1) \\ H_2 \in \text{pairs}(t_2) \\ H = H_1 \cup H_2}} B_{t_1}(H_1) \wedge B_{t_2}(H_2)$$

There are $2^{O(k \log k)}$ pairings for each bug X_t of $k + 1$ vertices.

Conclusion:

tw-HAMILTONIAN CYCLE admits a $2^{O(k \log k)}$ · n -step algorithm

Therefore, it belongs in $2^{O(k \log k)}$ -FPT

Our next step is to show that *tw*-PLANAR HAMILTONIAN CYCLE $\in 2^{O(k)}$ -FPT

Part 3, Tuesday 10/05/2016 - 16:00–17:00 (60')

Branch decompositions

Sphere cut decompositions

Dynamic programming on planar graphs

Branch decompositions

Branchwidth is a (topological) tree-likeness measure, alternative to treewidth, appeared in GM-X (1991).

A *branch decomposition* is a pair (T, τ)

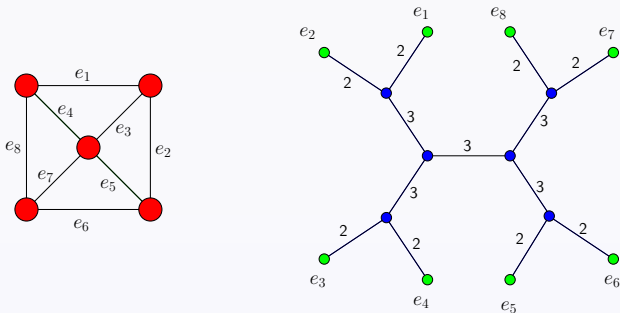
where

1. T is a ternary tree and
2. τ is a bijection mapping the edges of G to the leaves of T .

if T_1 is one of the connected components of $T - e$ then we set

$$E_e = \tau^{-1}(\text{leaves of } T_1) \text{ and } \mathbf{mid}(e) = \partial E_e.$$

A graph G and a branch decomposition of it.



The *width* of a branch decomposition (T, τ) is $\max\{|\text{mid}(e)| \mid e \in E(T)\}$

The *branchwidth*, $\text{bw}(G)$, of a graph G is then **minimum** width a branch decomposition of G may have.

Combinatorics of branchwidth

Theorem: [Robertson and Seymour, GM-10] *If G is not acyclic, then*

$$\mathbf{bw}(G) \leq \mathbf{tw}(G) + 1 \leq \frac{3}{2} \mathbf{bw}(G)$$

If T is a tree, then $0 \leq \mathbf{bw}(G) \leq 2$.

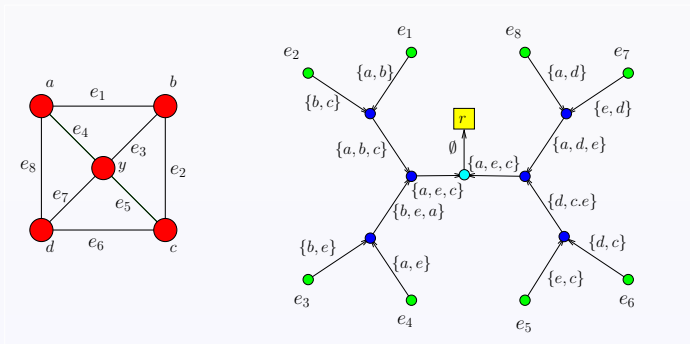
$$\mathbf{tw}\left(\begin{array}{|c|} \hline \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \hline \end{array}\right) = \mathbf{bw}\left(\begin{array}{|c|} \hline \square \\ \square \\ \square \\ \square \\ \square \\ \square \\ \hline \end{array}\right) = 6$$

$$\mathbf{bw}(K_6) = 4 < \mathbf{tw}(K_6) = 5$$

Dynamic programming for graphs of small branchwidth

Given a branch decomposition (T, τ) , (of small width)

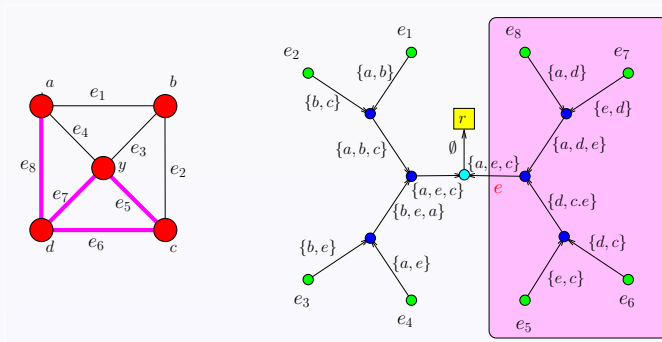
1. Root T to some vertex r without preimage



For each $e \in E(T)$, we denote as G_e the graph induced by the edges mapped below e .

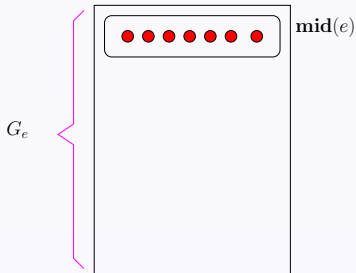
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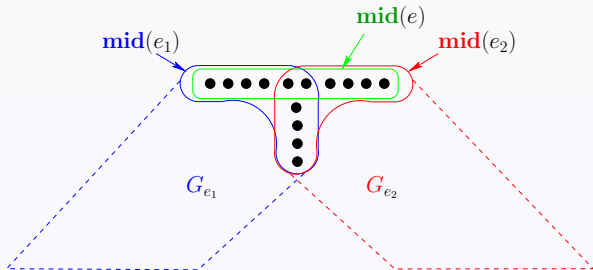


For each $e \in E(T)$, we denote as G_e the graph induced by the edges mapped below e .

2. Define, for each $e \in E(T)$, a table encoding the information of a partial solution for G_e as restricted to $\text{mid}(e)$. The values of this table for the root node should provide a global answer.



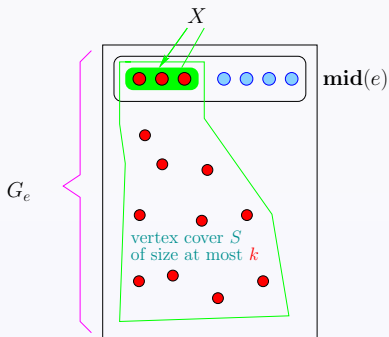
3. Define the values of this table for the leaf nodes
4. Provide the way to compute the table of an edge using the tables of its children edge.



An example: VERTEX COVER

Let G be a graph and $X, X' \subseteq V(G)$ where $X \cap X' = \emptyset$.

We say that $\text{vc}(G, X, X') \leq k$ if G contains a vertex cover S where $|S| \leq k$ and $X \subseteq S \subseteq V(G) \setminus X'$.

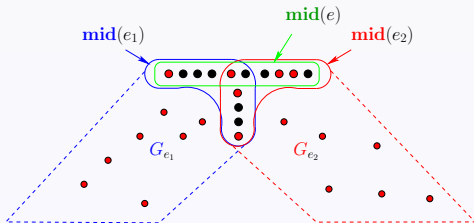


Let $\mathcal{R}_e = \{(X, k) \mid X \subseteq \text{mid}(e) \wedge \text{vc}(G_e, X, \text{mid}(e) \setminus X) \leq k\}$

observe that $\text{vc}(G) \leq k$ iff $(\emptyset, k) \in \mathcal{R}_{e_r}$.

Compute \mathcal{R}_e by using the following dynamic programming formula:

$$\mathcal{R}_e = \begin{cases} \{(X, k) \mid X \subseteq e \wedge X \neq \emptyset \wedge k \geq |X|\} & \text{if } e \in L(T) \\ \{(X, k) \mid X \subseteq \mathbf{mid}(e) \wedge \exists (X_1, k_1) \in \mathcal{R}_{e_1}, \exists (X_2, k_2) \in \mathcal{R}_{e_2} : \\ \quad (X_1 \cup X_2) \cap \mathbf{mid}(e) = X \wedge k_1 + k_2 - |X_1 \cap X_2| \leq k\} & \text{if } e \notin L(T) \end{cases}$$



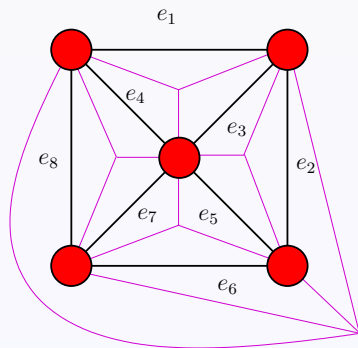
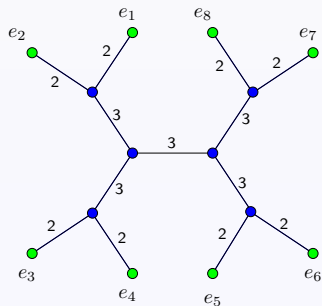
▶ $\forall e \in E(T), |\mathcal{R}_e| \leq 2^{|\mathbf{mid}(e)|} \cdot \ell.$

▶ we can check whether $\text{vc}(G) \leq \ell$ in $O(4^{\text{bw}(G)} \cdot \ell^2 \cdot |V(T)|)$ steps.

Sphere-Cut Decompositions

Suppose that G is a **planar** graph embedded on the sphere \mathcal{S}_0

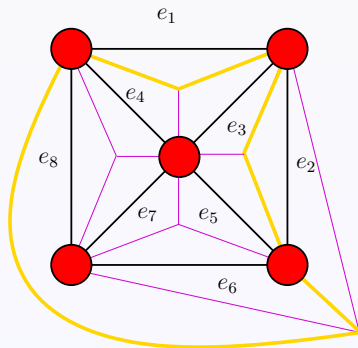
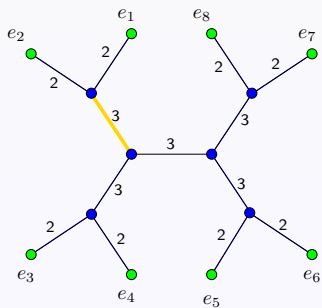
A **sphere-cut decomposition** of G is a branch decomposition (T, τ) where for any $e \in E(T)$, the vertices in $\text{mid}(e)$ are the vertices in a Jordan curve of \mathcal{S}_0 – called **noose** – that meets no edges of G .



Sphere-Cut Decompositions

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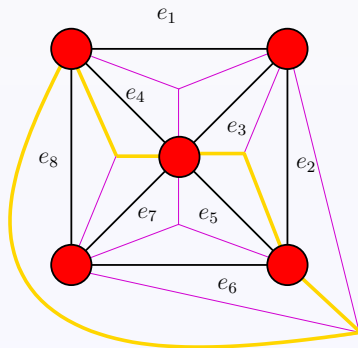
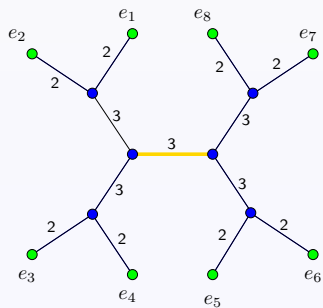
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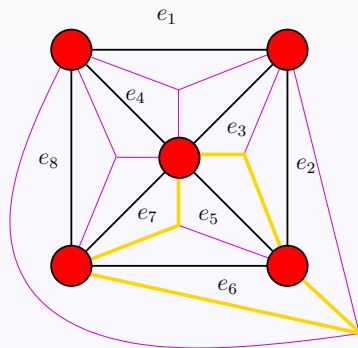
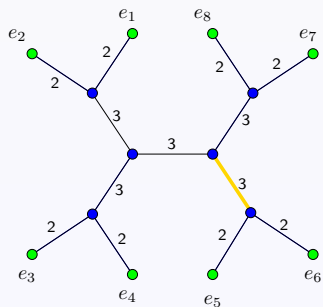
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Sphere-Cut Decompositions

Suppose that G is a **planar** graph embedded on the sphere \mathcal{S}_0

A **sphere-cut decomposition** of G is a branch decomposition (T, τ) where for any $e \in E(T)$, the vertices in $\text{mid}(e)$ are the vertices in a Jordan curve of \mathcal{S}_0 – called **noose** – that meets no edges of G .



Theorem: [Roberston & Seymour GM-X] If G is planar and has a *branch* decomposition with width $\leq k$ then G has a *sphere-cut* decomposition of G with width $\leq k$ that can be constructed in $O(n^3)$ steps.

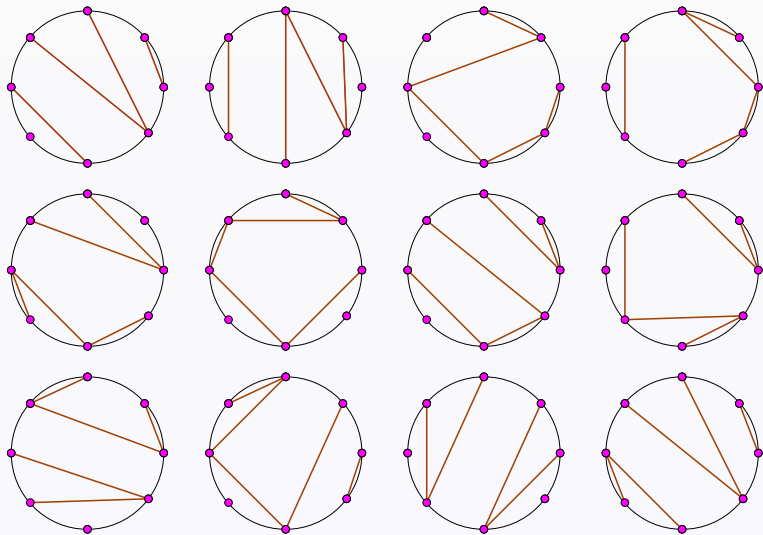
For doing dynamic programming on a sphere cut decomposition (T, τ) again we define, for any $e \in E(T)$ the set $\text{pairs}(\text{mid}(e))$ be the set of all pairings of $\text{mid}(e)$

The “usual” bound for $\text{mid}(e)$ is $2^{O(k \cdot \log k)}$

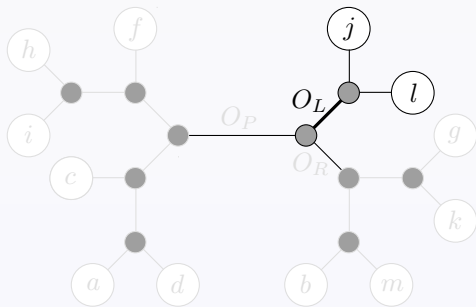
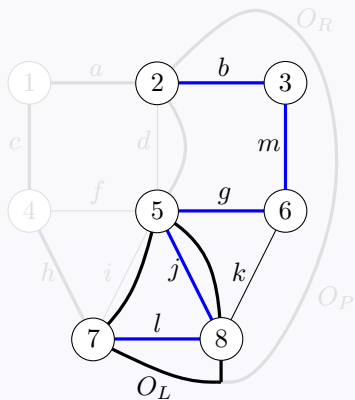
(recall that $|\text{mid}(e)| = \Omega(\frac{k}{2}!)$)

However, we now have that

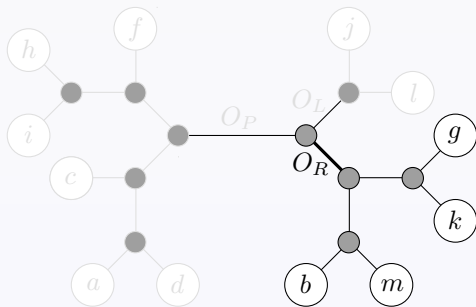
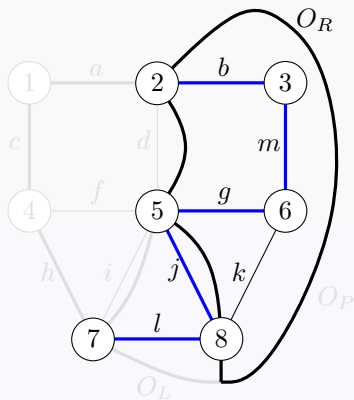
- 1: the vertices of $\text{mid}(e)$ lay on the boundary of a *disk* and
- 2: the pairings cannot be *crossing* because of *planarity*.



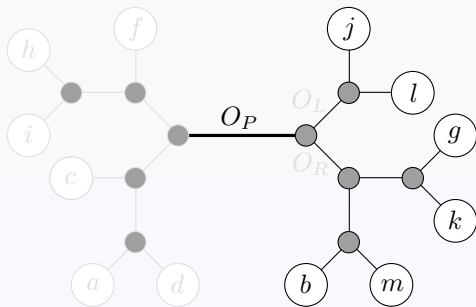
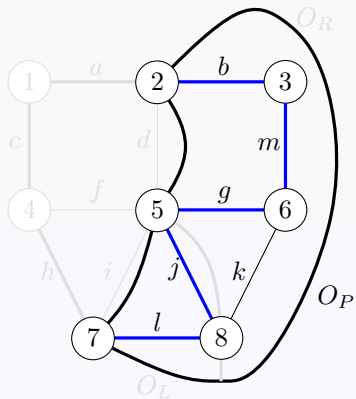
Non crossing pairings



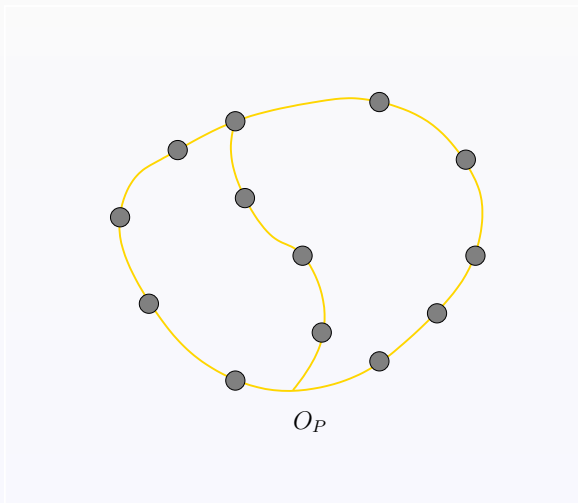
The two nooses O_L and O_R of the two children for a nose O_P for the parent.



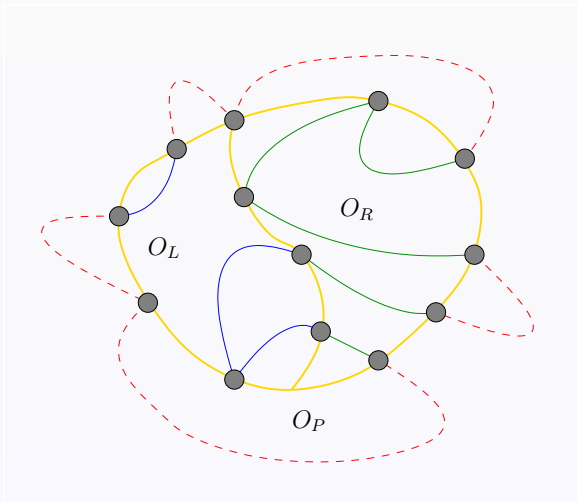
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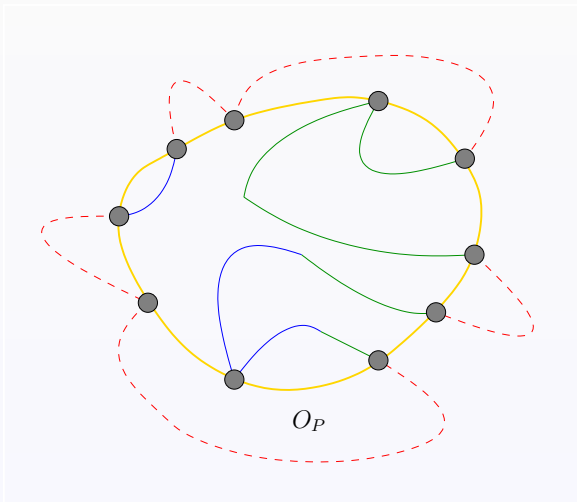
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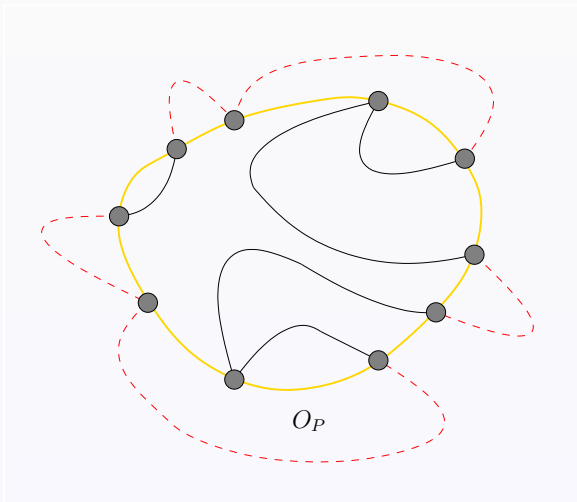
In case of **HAMILTONIAN CYCLE**, each non-crossing pair on O_P is the union of two non-crossing pairs on O_L and O_R .



In case of **HAMILTONIAL CYCLE**, each non-crossing pair on O_P is the union of two non-crossing pairs on O_L and O_R .



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In case of **HAMILTONIAL CYCLE**, each non-crossing pair on O_P is the union of two non-crossing pairs on O_L and O_R .

Catalan Structures

It follows that $\text{pairs}(\text{mid}(e)) = O(C(|\text{mid}(e)|)) = O(C(k))$

Where $C(k)$ is the k -th Catalan Number.

It is known that $C(k) \sim \frac{4^k}{k^{3/2}\sqrt{\pi}} = 2^{O(k)}$

Therefore: dynamic programming for HAMILTONIAN CYCLE of a planar graph G on a sphere cut decompositions of G with width $\leq k$ takes $2^{O(k)} \cdot O(n)$ steps.

► The same holds for several other problems where an analogue of $\text{pairs}(\text{mid}(e))$ can be defined for controlling the size of the tables in dynamic programming.

In general: These are pairs where the tables encode pairings.

► Like that one can design $2^{O(\text{tw}(G))} \cdot n^{O(1)}$ step algorithms for the planar versions of CYCLE COVER, PATH COVER, LONGEST PATH, LONGEST CYCLE, HAMILTONIAN CYCLE, and GRAPH METRIC TSP and others.

[Dorn, Penninx, Bodlaender, and Fomin. ICALP 2005]

The idea of using properties of the embedding (for pairings) has been extended for bounded genus graphs in [Dorn, Fomin, and Thilikos. SWAT 2006]

and for H -minor-free graphs in [Dorn, Fomin, and Thilikos. SODA 2008]

Common idea: Planarization

For more complicated problems planarization becomes very hard to handle as here tables encode **packings** instead of pairings.

For this, single exponential dynamic programming has been done by

1. Moving from sphere cut decompositions to *surface cut decompositions*
2. Counting non intersecting packings on surfaces with boundary.

[Sau, Rué, Thilikos, TALG 2014]

Extensions/alternatives:

- ▶ For H -minor free graphs: [Sau, Rué, Thilikos, COCOON 2012]
- ▶ Surface split decompositions: [Bonsma, STACS 2012]
- ▶ Brick Decompositions: [Cohen-Addad & de Mesmay, ESA 2015]

Part 4, Thursday 10/05/2016 - 11:00–12:30 (90')

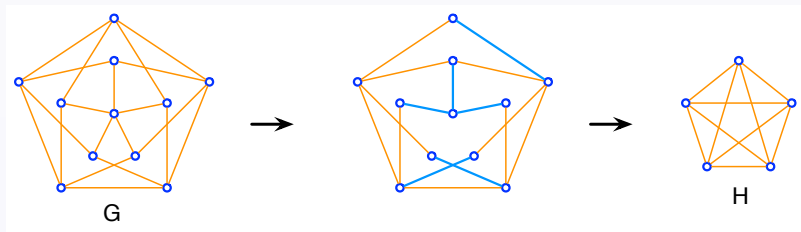
Bidimensionality

Subexponential parameterized algorithms

The minor relation

H is a minor of G ($H \leq G$):

H occurs from a subgraph of G by applying edge contractions



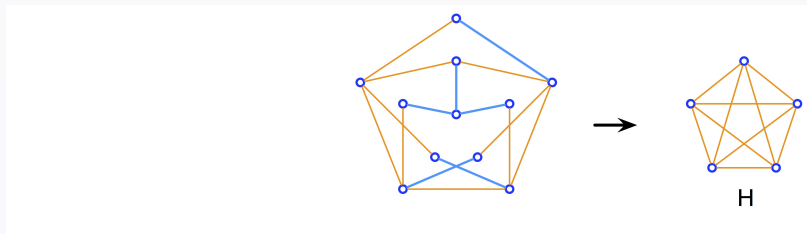
A graph class \mathcal{G} is *minor-closed* if:

every minor of a graph in \mathcal{G} is also a graph in \mathcal{G}

The contraction relation

H is a contraction of G ($H \leq_c G$):

H occurs by applying edge contractions



A graph class \mathcal{G} is *contraction-closed* if:

every contraction of a graph in \mathcal{G} is also a graph in \mathcal{G}

Theorem: [Robertson & Seymour – main combinatorial result of GM] *Every infinite set of graphs contains two graphs comparable under the minor relation.*

Equivalently: Graphs are **Well Quasi Ordered** w.r.t. the **minor** relation



Let \mathcal{G} be a **minor-closed** graph class.

► $\mathbf{obs}(\mathcal{G})$ is the set of minor-minimal elements not in \mathcal{G} .

► If \mathcal{G} is minor-closed, then $G \in \mathcal{G} \iff \forall H \in \mathbf{obs}(\mathcal{G}) \ H \not\leq G$

Consequence of R&S theorem: $|\mathbf{obs}(\mathcal{G})| < \aleph_0$

Theorem: [Robertson & Seymour – main algorithmic consequence of GM] For every H , checking whether $H \leq G$ can be done in $O(n^3)$ steps.

► **Meta-Algorithmic Consequence:** For every minor-closed graph class \mathcal{G} , the problem asking whether $G \in \mathcal{G}$ belongs in PTIME, i.e., can be solved in $O(n^3)$ steps!

Graph optimization parameters

Graph parameter: a function $\mathbf{p} : \mathcal{G}_{\text{all}} \rightarrow \mathbb{N}$

We consider *minimization/maximization parameters* \mathbf{p} defined as follows

$$\mathbf{p}(G) = \min\{k \mid \exists S \subseteq V(G) : |S| \leq k \wedge \phi(G, S) = \text{true}\}$$

$$\mathbf{p}(G) = \max\{k \mid \exists S \subseteq V(G) : |S| \geq k \wedge \phi(G, S) = \text{true}\}$$

In any case, we call a set S where $|S| = \mathbf{p}(G)$ *solution certificate* for $\mathbf{p}(G)$

We call such parameters *graph optimization parameters*.

Three examples:

▶ VERTEX COVER, $\mathbf{vc}(G)$: $\min, \phi(G, S) = \forall e \in E(G) \ e \cap S \neq \emptyset$

▶ DOMINATING SET, $\mathbf{dc}(G)$: $\min, \phi(G, S) = V(G) = N_G(S)$

▶ LONGEST PATH, $\mathbf{pl}(G)$: $\max, \phi(G, S) = G[S] \text{ is a path}$

▶ SCATTERED SET, $\mathbf{sc}(G)$: $\max, \phi(G, S) = \forall x \in V(G) \ |N[x] \cap S| \leq 1$

We say that \mathbf{p} is *minor closed* if $H \leq_m G \Rightarrow \mathbf{p}(H) \leq \mathbf{p}(G)$.

► Examples of minor-closed graph parameters:

vc: minimum vertex cover of G

fvs: minimum feedback vertex set of G

fc: minimum face cover of a planar G

lp: maximum k for which G contain a k -path

vp: vertex planarizer number of G ($= \min\{|S| \mid G \setminus S \text{ is planar}\}$)

tw: the tree-width of G

bw: the branch-width of G

eg: the Euler genus of G

r -twm: ($= \min\{|S| \mid \mathbf{tw}(G \setminus S) \leq r\}$)

We say that \mathbf{p} is *contraction closed* if $H \leq_c G \Rightarrow \mathbf{p}(H) \leq \mathbf{p}(G)$.

► Examples of contraction-closed (but **not** minor-closed) graph parameters:

ds: minimum dominating set of G

cd: minimum cycle domination set of G

vc: minimum connected vertex cover of G

ctw: minimum connected treewidth of G i.e., all bags induce **connected** subgraphs

icp: maximum induced cycle packing

sc: maximum k for which G contains a **scattered** set of k vertices.

[S is a **scattered**: no two distinct vertices of S have a common neighbor.]

Let \mathbf{p} be a **minor-** (**contraction-**) closed graph parameter.

We define the parameterized problem $p\text{-CHECKING VALUE OF } \mathbf{p} = (\mathcal{G}_{\text{all}}, \mathbf{p})$

In other words $p\text{-CHECKING VALUE OF } \mathbf{p}$ corresponds to the following

Meta-problem:

$p\text{-CHECKING VALUE OF } \mathbf{p}$
<i>Instance:</i> A graph G and an integer $k \geq 0$.
<i>Parameter:</i> k
<i>Question:</i> is it correct that $\mathbf{p}(G) \leq k$?

Here " \leq " = " \leq " / " \leq " = " \geq " if \mathbf{p} is a minimization/maximization parameter

For simplicity: $\Pi_{\mathbf{p}} = p\text{-CHECKING VALUE OF } \mathbf{p}$

We call such a problem $\Pi_{\mathbf{p}}$ *subset optimization problem*

For every $k \in \mathbb{N}$, we define the k -th layer of $(\mathcal{G}_{\text{all}}, \mathbf{p})$ as the class

$$\mathcal{G}_k^{\mathbf{p}} = \{G \mid \mathbf{p}(G) \leq k\}$$

These are **YES**-instances for minimization problems and

NO-instances for maximization problems

- ▶ Observe: for every k , $\mathcal{G}_k^{\mathbf{P}}$ is minor- (contraction-) closed.
- ▶ $\mathbf{obs}(\mathcal{G}_k^{\mathbf{P}})$ is the set of minor-minimal elements not in \mathcal{G} .
- ▶ If \mathbf{p} is minor-closed, then so is $\mathcal{G}_k^{\mathbf{P}}$ and thus $G \in \mathcal{G}_k^{\mathbf{P}} \iff \forall H \in \mathbf{obs}(\mathcal{G}_k^{\mathbf{P}}) H \not\leq G$

Consequence of R&S theorem: $\forall_{k \in \mathbb{N}} |\mathbf{obs}(\mathcal{G}_k^{\mathbf{P}})| < \aleph_0$

Recall:

Theorem: [Robertson & Seymour – main algorithmic consequence of GM]

For every H , checking whether $H \leq G$ can be done in $f(|V(H)|) \cdot n^3$ steps.

- ▶ **Meta-Algorithmic Consequence:** If \mathbf{p} is minor-closed, then

p -CHECKING VALUE OF $\mathbf{p} \in \text{FPT}$.

In other words:

There **exists** an algorithm that solves p -CHECKING VALUE OF \mathbf{p} in $f(k) \cdot n^3$ steps.

Corollary: If \mathbf{p} is a minor-closed graph parameter, then

$\Pi_{\mathbf{p}} = \boxed{p\text{-CHECKING VALUE OF } \mathbf{p}}$ can be solved in $f(k) \cdot n^3$ steps

- ▶ We have a (**non-constructive**) proof that an algorithm **exists!**
- ▶ Is an **encouraging** theory (**if** you know that something **exists...**)
- ▶ This does not mean that we have **constructed** such an algorithm
- ▶ But... We are **encouraged** to do so!

“Half of science is asking the right questions.” Roger Bacon

Questions:

- ▶ What is the best (constructive) f we can have and when?
- ▶ Can the (many) ideas from Graph Minors be used for this?
- ▶ Can we derive results for problems closed under other relations?

Comments on P versus NP: from CS and Maths



NP-completeness: A Retrospective

Christos H. Papadimitriou*

University of California, Berkeley, USA

Abstract. For a quarter of a century now, NP-completeness has been computer science's favorite paradigm, fad, punching bag, buzzword, alibi, and intellectual export. This paper is a fragmentary commentary on its origins, its nature, its impact, and on the attributes that have made it so pervasive and contagious.

"... NP-completeness in this context suggests that a problem area or approach is mathematically nasty.", **Christos Papadimitriou – ICALP 1997**



Mathematical Problems for the Next Century¹

Steve Smale

DEPARTMENT OF MATHEMATICS
CITY UNIVERSITY OF HONG KONG
KOWLOON, HONG KONG

AUGUST 7, 1998

"P versus NP – a gift to mathematics from computer science." **Steve Smale, 1998**

A metaphore on graph minors

“Graph Minors suggests that some problems are algorithmically not so nasty”

“Graph Minors – a gift to computer science from mathematics”

General requirements

We need the following two facts (for suitable functions f and G):

1. [Combinatorial] $\mathbf{p}(G) \leq k \Rightarrow \mathbf{tw}(G) \leq f(k)$

i.e., YES/NO-instances of $\Pi_{\mathbf{p}}$ have treewidth $\leq f(k)$

2. [Algorithmic] One can check whether $\mathbf{p}(G) \leq k$ in $2^{g(\mathbf{tw}(G))} \cdot n^{O(1)}$ steps

Typically this is done by Dynamic Programming.

1+2 \rightarrow $\Pi_{\mathbf{p}}$ has a $2^{g(f(k))} \cdot n^{O(1)}$ step algorithm

Proof:

This algorithm first checks whether $\mathbf{tw}(G) \leq f(k)$.

If the answer is **negative**, then outputs a negative/positive answer (by 1).

If the answer is **positive**, then runs DP algorithm (by 2).

$1+2 \rightarrow \Pi_{\mathbf{P}}$ has a $2^{g(f(k))} \cdot n^{O(1)}$ step algorithm

► $g(\cdot)$ is linear: $\Pi_{\mathbf{P}}$ is *singly exponentially solvable w.r.t. treewidth*

[Liming Cai and David Juedes, 2003]:

For several problems, assuming ETH, the **best** running time we can expect is $2^{O(k)} \cdot n^{O(1)}$, in general

$2^{O(\sqrt{k})} \cdot n^{O(1)}$ for their **planar** restrictions \rightarrow **when can we match this?**

► Here we care about such questions!

Exponential Time Hypothesis (ETH):

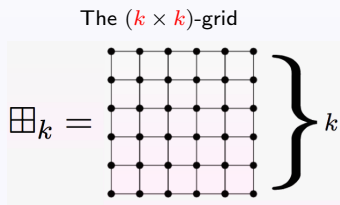
There is **no** $2^{o(n)}$ -step algorithm that solves 3-SAT

We care about combinatorial condition:

1. [Combinatorial] YES/NO-instances of $\Pi_{\mathcal{P}}$ have $\text{tw} = O(k)$ (or $o(k)$ in planar graphs)

Idea: For minor-closed problems, the existence of a $(k \times k)$ -grid as a minor of the input graph may serve as a YES/NO certificate

Fact: “Big” treewidth implies the the existence of a $(k \times k)$ -grid as a minor



This fact was first proved in GM-V by Robertson and Seymour.

Theorem: [Robertson & Seymour – GM-V]

There is a $\delta : \mathbb{N} \rightarrow \mathbb{N}$ such that $\forall \alpha \text{ tw}(G) > \delta(\alpha) \Rightarrow G \geq_m \boxplus_{\alpha}$.

Upper bound for δ : remained **exponential** for a **long** time...

Untill: Julia Chuzhoy proved: $\delta(k) = O(k^{20})$



► The best known lower bound is $\delta(k) = \Omega(k^2 \cdot \log k)$

Definition: $\mathbf{p}^{-1}(k) = \min\{\alpha \mid \mathbf{p}(\boxplus_{\alpha}) > k\}$

Observe the following:

Lemma: *If \mathbf{p} is minor-closed, then*

1. [Combinatorial] $\mathbf{p}(G) \leq k \Rightarrow \mathbf{tw}(G) \leq \delta(\mathbf{p}^{-1}(k))$

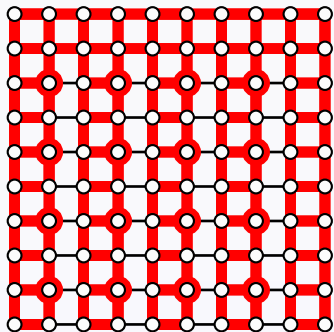
Proof: Let $\alpha = \mathbf{p}^{-1}(k)$. Then $k < \mathbf{p}(\boxplus_{\alpha})$ (1).

Assume $\delta(\alpha) < \mathbf{tw}(G)$ (2).

[By grid exclusion]: (2) $\Rightarrow \boxplus_{\alpha} \leq_m G$. (1)

[By minor-closedness]: if $\boxplus_{\alpha} \leq_m G$, then $\mathbf{p}(\boxplus_{\alpha}) \leq \mathbf{p}(G)$. (3)

(1) and (3) $\Rightarrow k < \mathbf{p}(G)$.



$$vc^{-1}(k) = O(\sqrt{k})$$

$$lp^{-1}(k) = O(\sqrt{k})$$

► Not all minor-closed problems are bidimensional!

such as TREewidth, PATHwidth, BRANCHwidth, TREE-DEPTH, and GENUS

The same holds for the following
minor-closed parameters:

- FEEDBACK VERTEX SET,
- LONGEST CYCLE,
- CYCLE PACKING,
- FACE COVER
- MAX SERIES-PARALLEL SUBGRAPH

Definition:

We call a problem $\Pi_{\mathbf{p}}$ **minor-bidimensional**
if \mathbf{p} is minor-closed and $\mathbf{p}^{-1}(k) = \sqrt{k}$

For minor-bidimensional parameters:

1. [Combinatorial] $\mathbf{p}(G) \leq k \Rightarrow \mathbf{tw}(G) \leq \delta(\sqrt{k})$

Recall that $\delta(k) = \Omega(k^2 \cdot \log k)$

Best of all scenarios: $2^{O(k \log k)} \cdot n^{O(1)}$ step algorithms



“Best of all scenarios” means that

▶ $\delta(k) = O(k^2 \cdot \log k)$ (which is conjectured but not sure!) and

▶ $\Pi_{\mathbf{p}}$ is singly exponentially solvable w.r.t. treewidth
(which is the case for many problems)

► **Conclusion:**

1. To design (optimal) $2^{O(k)} \cdot n^{O(1)}$ step algorithms for general graphs one needs a problem-specific analysis.
2. proving that $\delta(k) = O(k^2 \cdot \log k)$ will have interesting algorithmic consequences.
3. If we want $2^{O(k)} \cdot n^{O(1)}$ step algorithms we must restrict G to **special** graph classes.

In particular: topological graph classes (where δ is better bounded)

4. For even better (e.g. **subexponential**) parameterized dependency we must restrict our attention to special graph classes.

Subexponential parameterized algorithms

Definition: A graph class \mathcal{G} has the subquadratic grid minor property (**SQGM**) if there exist $1 \leq c < 2$ such that $\forall k \boxplus_k \not\leq G \Rightarrow \text{tw}(G) = O(k^c)$


Then: YES/NO-instances of a bidimensional problem $\Pi_{\mathbf{p}}$ have $\text{tw} = (\mathbf{p}^{-1}(k))^c = O((\sqrt{k})^c) = o(k)$

► if Π is singly exponentially solvable w.r.t. treewidth, then Π can be solved in $2^{o(k)} \cdot n^{O(1)}$ steps.

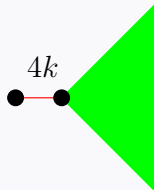
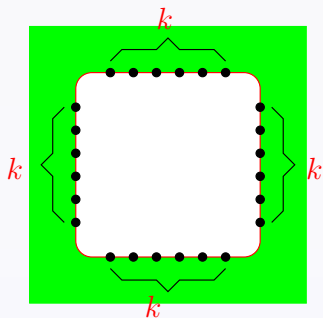
Planar graphs have the **SQGM** property for $c = 1$

As $\text{tw}(G) = O(\text{bw}(G))$, the above follows from the following:

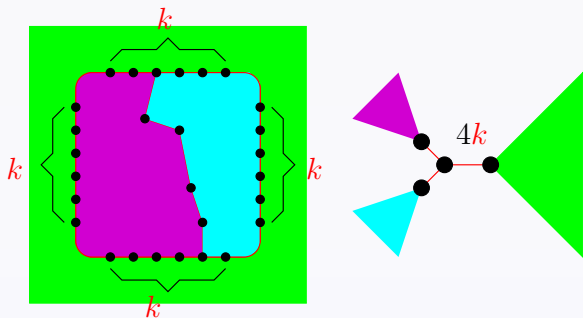
Theorem: [Robertson, Seymour, & Thomas 1994] If G is planar and $\text{bw}(G) \geq 4k$,

then  $\leq_m G$.

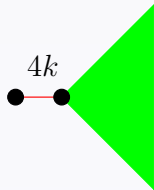
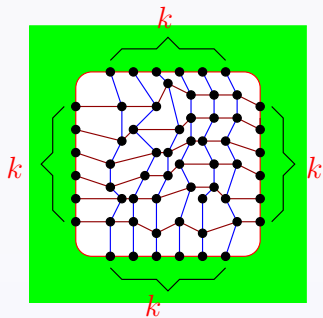
► We sketch the **Idea** of the proof of the above theorem:




“Suppose” that we constructed a partial branch decomposition of the part of the graphs that is inside a disk.



If there is a path from north-south or east-west, partition the disk: **one more step** further with the construction of a branch decomposition of width $\leq 4k$.



Such a path must exist,

otherwise, from [Menger's theorem](#), the graph contains  k as a minor.

Bidimensionality race:

SQGM: $\forall k \square_k \not\preceq G \Rightarrow \text{tw}(G) = O(k^c)$ for some $c < 2$.

When **SQGM** property holds?

Planar: [Robertson and Seymour, JCSTB 1986]

Bounded Genus: [Demaine, Fomin, Hajiaghayi, Thilikos, JACM 2005]

Apex-minor free graphs: [Demaine, Fomin, Hajiaghayi, Thilikos, SIDMA 2004]

H -minor free graphs: [Demaine, Hajiaghayi, Combinatorica 2008]

Bounded degree unit disk graphs [Fomin, Lokshtanov, Saurabh, SODA 2012]

Families of 2D-geometric graphs [Grigoriev, Koutsonas, Thilikos, SOFSEM 2014]

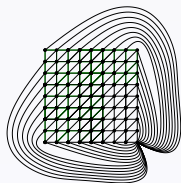
► In all above cases we have topologically refined graph classes and $c = 1$.

► are there **more general** graph classes where $1 < c < 2$?

Bidimensionality for contraction-closed problems

[such as Π_{ds}]

\square_k is replaced by the uniformly triangulated grid Γ_k :



Let $\tilde{p}^{-1}(k) = \min\{\alpha \mid p(\Gamma_\alpha) > k\}$

Definition:

We call a problem Π_p **contraction-bidimensional**


if p is contraction-closed and $\tilde{p}^{-1}(k) = \sqrt{k}$


Definition: A class \mathcal{G} has the *subquadratic grid contraction property* (**SQGC**) if there

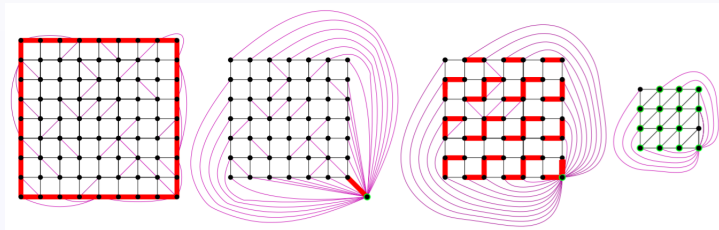
exist $1 \leq c < 2$ such that $\forall k \Gamma_k \preceq_c \mathcal{G} \Rightarrow \text{tw}(G) = O(k^c)$

Subquadratic grid contraction property (**SQGC**) holds for planar graphs because of:

Theorem: [Robertson, Seymour, & Thomas 1994] If G is *planar* and $\text{bw}(G) \geq 4k$,

then  $\leq_m G$.

Proof: if we do not apply edge removals while obtaining  from G we end up to a partially triangulated grid that can be further be contracted to the uniformly triangulated grid Γ_k .



Therefore $\Gamma_k \not\leq_c G \Rightarrow \text{tw}(G) = O(k^c)$, thus **SQGC** holds for $c = 1$.

Bidimensionality race:

SQGC: $\forall k \Gamma_k \not\leq_c G \Rightarrow \text{tw}(G) = O(k^c)$ for some $c < 2$.

When **SQGC** property holds?

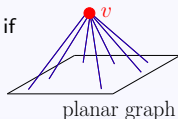
Planar: follows from [Robertson and Seymour, JCSTB 1986]

Bounded Genus graphs: [Demaine, Hajiaghayi, Thilikos, SIDMA 2006]

Apex-minor free graphs: [Fomin, Golovach, Thilikos, JCTSB 2011]

Families of 2D-geometric graphs [Baste, Thilikos, in preparation]

A graph H is an *apex graph* if



$\exists v \in V(H): H - v$ is planar

Bidimensionality and subexponential algorithms.

Theorem: Let Π_p be a subset optimization parameterized problem that

- i. is minor/contraction-*bidimensional*
- ii. is *singly exponentially* solvable w.r.t. treewidth
- iii. is restricted to some **SQGM/SQGC**-graph class

Then Π can be solved in $2^{o(k)} \cdot n^{O(1)}$ steps

Some bidimensional problems

(CONNECTED) VERTEX COVER, (CONNECTED) DOMINATING SET, (CONNECTED) FEEDBACK VERTEX SET, INDUCED MATCHING, LONGEST CYCLE, (CONNECTED) (INDUCED) CYCLE PACKING, (CONNECTED) CYCLE DOMINATION, d -SCATTERED SET, LONGEST PATH, (INDUCED) PATH PACKING, (CONNECTED) r -CENTER, (CONNECTED) DIAMOND HITTING SET, MINIMUM MAXIMAL MATCHING, FACE COVER, UNWEIGHTED TSP TOUR, MAX BOUNDED DEGREE CONNECTED SUBGRAPH

► The previous theorem can become an **algorithmic meta-theorem** as

ii. is singly exponentially solvable w.r.t. treewidth

is implied by expressibility in **Existential Counting Modal Logic**

because of [Michał Pilipczuk, MFCS 2011]

► Some powerful techniques for ii.

[Dorn, Penninkx, Bodlaender, Fomin, Algorithmica 2010] ★

[Rué, Sau, Thilikos, TALG 2014] ★

[Cygan, Nederlof, Pilipczuk, Pilipczuk, van Rooij, Wojtaszczyk, FOCS 2011] ★★

[Bodlaender, Cygan, Kratsch, Nederlof, ICALP 2013] ★★

[Fomin, Lokshtanov, Saurabh, SODA 2014] ★★

Part 5, Friday 10/05/2016 - 09:00–10:30 (90')

Bidimensionality and Kernelization

Irrelevant vertex technique

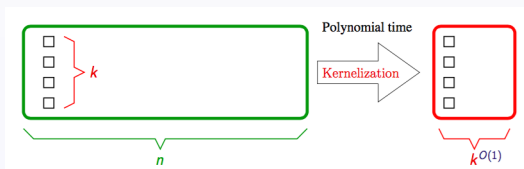
Bidimensionality and Kernelization

Kernelization

Let (Π, κ) be a parameterized problem. Recall: that $\Pi \subseteq \Sigma^*$ and $\kappa : \Sigma^* \rightarrow \mathbb{N}$.

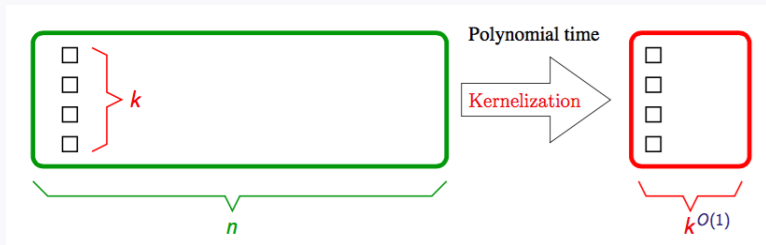
► A **polynomial** algorithm A is a **kernelization algorithm** for (Π, κ) if there exist some computable function $g : \mathbb{N} \rightarrow \mathbb{N}$ such that, for every $x \in \Sigma^*$, the output $x' = A(x)$ satisfies the following:

1. $x \in \Pi \Leftrightarrow x' \in \Pi$ (x and x' are equivalent)
2. $|x'| \leq g(k)$ (new instance has size bounded by a function of the parameter).



- If G is a polynomial (linear): polynomial (linear) kernel.
- a kernelization is a polynomial time **many-one reduction** of a problem to **itself** with the additional property that the image is bounded in terms of the parameter $k = \kappa(x)$.
- **Kernelization can be seen as a paradigm for preprocessing**

p -VERTEX COVER has kernelization algorithm that produces a kernel of $\leq 2k$ vertices.



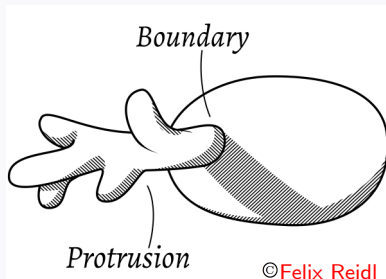
Alternatively, we say that p -VERTEX COVER has a kernel of size $2k$.

- ▶ A parameterized problem has a kernel iff it is in FPT
 - ▶ p -DOMINATING SET is $W[2]$ -complete, thus it is not expected to have a kernel.
-

- ▶ Not all problems in FPT are expected to have polynomial kernels (p -PATH)
[Bodlaender, Downey, Fellows, Hermelin, JCSS 2009]
- ▶ p -FEEDBACK VERTEX SET has a kernel of $O(k^2)$ edges.
[Thomassé, TALG 2010]
- ▶ Not all FPT-problems are expected to have linear kernel (p -FEEDBACK VERTEX SET)
[Dell, van Melkebeek, STOC 2010]
- ▶ p -PLANAR DOMINATING SET kernel of $67k$ vertices.
[Chen, Fernau, Kanj, Xia, SICOMB 2007]
- ▶ p -PLANAR FEEDBACK VERTEX SET has a kernel of $13k$ vertices.
[Bonamy, Kowalik, IPEC 2014]

p -PATH, p -DOMINATING SET, and p -FEEDBACK VERTEX SET are bidimensional.

Protrusions



r-*protrusion*: a set $X \subseteq V(G)$ where

- ▶ $\text{tw}(G[X]) \leq r \rightarrow$ treewidth is bounded
- ▶ $|\partial_G(G)| \leq r \rightarrow$ its boundary is of bounded size

Protrusion decompositions

An (α, β) -protrusion decomposition of G is a partition $\mathcal{P} = \{R_0, R_1, \dots, R_\rho\}$ of $V(G)$ such that

- ▶ $\max\{\rho, |R_0|\} \leq \alpha$,
- ▶ each $N_G[R_i]$, $i \in \{1, \dots, \rho\}$, is a β -protrusion of G , and
- ▶ for every $i \in \{1, \dots, \rho\}$, $N_G(R_i) \subseteq R_0$.

Remark: actually, this last condition is not necessary! But makes things more visualizable!

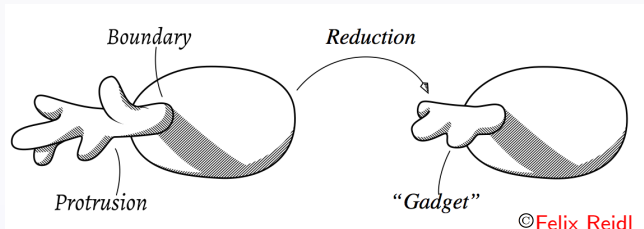


Protrusion replacement



f-protrusion replacement family for $\Pi_{\mathbf{p}}$: a collection $\mathcal{A} = \{A_i \mid i \geq 0\}$ of algorithms, such that algorithm A_i receives an instance (G, k) of $\Pi_{\mathbf{p}}$ and an i -protrusion X of G with at least $f(i)$ vertices and outputs an equivalent instance (G', k') of $\Pi_{\mathbf{p}}$ where $|V(G')| < |V(G)|$ and $k' \leq k$.

Protrusion replacement



© Felix Reidl

f-protrusion replacement family for $\Pi_{\mathbf{P}}$: a collection $\mathcal{A} = \{A_i \mid i \geq 0\}$ of algorithms, such that algorithm A_i receives an instance (G, k) of $\Pi_{\mathbf{P}}$ and an i -protrusion X of G with at least $f(i)$ vertices and outputs an equivalent instance (G', k') of $\Pi_{\mathbf{P}}$ where $|V(G')| < |V(G)|$ and $k' \leq k$.

Conditions for the existence of linear kernels

1. [*Combinatorial*] If $p(G) \leq k$, then G has an $(O(k), O(1))$ -protrusion decomposition.

2. [*Algorithmic*] Π_p has a protrusion replacement family.

► 1. + 2. \rightarrow a linear kernel for Π_p .

[Bodlaender, Fomin, Lokshtanov, Penninkx, Saurabh, Thilikos, FOCS 2008]

To achieve Conditions 1 and 2. we need some more definitions!

CMSO-expressibility

Let \mathbf{p} be a graph optimization parameter and
let $\Pi_{\mathbf{p}}$ be the corresponding graph optimization problem.

Recall that:

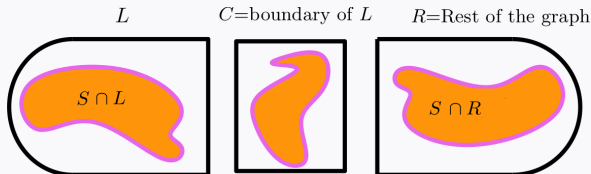
$$\mathbf{p}(G) = \min\{k \mid \exists S \subseteq V(G) : |S| \leq k \wedge \phi(G, S) = \text{true}\}$$

► If ϕ is expressible in Monadic Second Order Logic, then we say that

$\Pi_{\mathbf{p}}$ is *CMSO-expressible*

Linear Separability

Let \mathbf{p} be an graph optimization parameter and G be a graph



The subset optimization problem $\Pi_{\mathbf{p}}$ is *linearly separable* if, for any graph G and $L \subseteq V(G)$ such that $|C| = |\partial_G(L)| \leq t$, it holds that

$$|S \cap L| - c \cdot t \leq \mathbf{p}(G[L]) \leq |S \cap L| + c \cdot t$$

where S is a solution certificate for \mathbf{p}

► More generally: $c \cdot t \rightarrow f(t)$ defines *separable* $\Pi_{\mathbf{p}}$

p -PATH is not separable while

p -DOMINATING SET, and p -FEEDBACK VERTEX SET are linearly separable.

Conditions for the existence of linear kernels

1. [*Combinatorial*] If $\mathbf{p}(G) \leq k$, then G has an $(O(k), O(1))$ -protrusion decomposition.
2. [*Algorithmic*] $\Pi_{\mathbf{p}}$ has a protrusion replacement family.

▶ **SQGM/SQGC** + Linear separability + bidimensionality \rightarrow **1.**

[Fomin, Lokshtanov, Saurabh, Thilikos, SODA 2011]

▶ **CMSO**-expressibility + Linear separability \rightarrow **2.**

[Fomin, Lokshtanov, Saurabh, Thilikos, 2015]

We comment the proof of the first fact:

► **SQGM/SQGC** + Linear separability + bidimensionality $\xrightarrow{\mathbf{A+B}}$ **1.**

Definition: S is a treewidth η -modulator of G if $\text{tw}(G \setminus S) \leq \eta$

i.e., a certificate for $\eta\text{-twm}(G) \leq k$

Lemma A: Assume that:

1. \mathcal{G} is a graph class with the **SQGM/SQGC** property

2. $\Pi_{\mathbf{p}}$ is minor/contraction-bidimensional and linear-separable

Then there exists an integer $\eta \geq 0$ such that the following holds:

$\mathbf{p}(G) \leq k \Rightarrow G$ has a treewidth η -modulator S where $|S| \leq 2 \cdot k$.

Remark: the bidimensionality condition: $\mathbf{p}^{-1}(k) = \sqrt{k}$ is necessary here!

Lemma B: Assume that:

1. \mathcal{G} is a graph class with the **SQGM/SQGC** property.

2. G has a treewidth η -modulator S for some positive integer η

Then there exists an integer r such that G has $(2 \cdot |S|, r)$ -protrusion decomposition.

Bidimensionality and kernels.

Theorem: Let Π_P be a subset optimization parameterized problem that

- i. is **CMSO**-expressible
- ii. is minor- (resp. contraction-) **bidimensional**,
- iii. is linearly **separable**,
- iv. is restricted to some **SQGM/SQGC**-graph class

Then Π_P admits a **linear kernel**.

Bidimensionality and approximation.

- Just for the history we also mention the following:
-

Theorem: Let Π_P be a subset optimization parameterized problem that

- i. is \approx CMSO-expressible
- ii. is minor- (resp. contraction-) **bidimensional**,
- iii. is linearly **separable**,
- iv. is restricted to some **SQGM/SQGC**-graph class

Then Π_P admits an EPTAS

EPTAS = (**E**fficient **P**olynomial-**T**ime **A**pproximation **S**cheme)

[Demaine, Hajiaghay, SODA 2005]

[Fomin, Lokshtanov, Raman, Saurabh, SODA 2011]

Irrelevant vertex technique

General algorithmic strategy (so far):

We need the following two facts (for suitable functions f and G):

1. [Combinatorial] $p(G) \leq k \Rightarrow \text{tw}(G) \leq f(k)$

i.e., YES/NO-instances of Π_p have treewidth $\leq f(k)$

2. [Algorithmic] One can check whether $p(G) \leq k$ in $g(\text{tw}(G)) \cdot n^{O(1)}$ steps

▶ Then we have an FPT-algorithm running in $g(f(k)) \cdot n^{O(1)}$ steps because:

- If $\text{tw}(G) > f(k)$ then we declare **VICTORY!** (enemy **surrenters!**)
- if $\text{tw}(G) \leq f(k)$ then the **CAVALRY** comes! (DP algorithms or just Courcelle's th.)

▶ What about when YES/NO-instances of Π_p do **not** have bounded treewidth?

- In this case we have to **FIGHT!!!!** (until the **CAVALRY** comes or enemy **surrenters**)

▶ For many problems, instances of big enough treewidth may contain part whose removal does not change the answer to the original question.

Consider the following problem:

p -ODD CYCLE PACKING

Instance: A graph G and an integer $k \geq 0$.

Parameter: k ,

Question: Does G contains k mutually vertex-disjoint odd cycles?

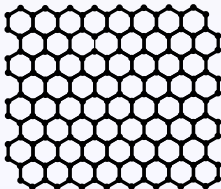
Without the “odd” demand, the problem is minor-bidimensional and we are not afraid!

How to deal with “oddness” demand (at least) for planar instances?

Suppose we have an instance G of big enough treewidth!

Then G contains a big grid as a minor

This means that G contains a subgraph that is a **subdivision** of a “big enough” **wall**!



Consider the following problem:

p-ODD CYCLE PACKING

Instance: A graph G and an integer $k \geq 0$.

Parameter: k ,

Question: Does G contains k mutually vertex-disjoint odd cycles?

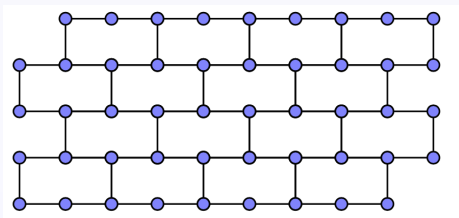
Without the “odd” demand, the problem is minor-bidimensional and we are not afraid!

How to deal with “oddness” demand (at least) for planar instances?

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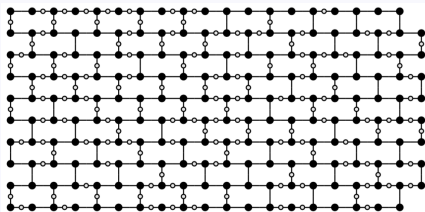
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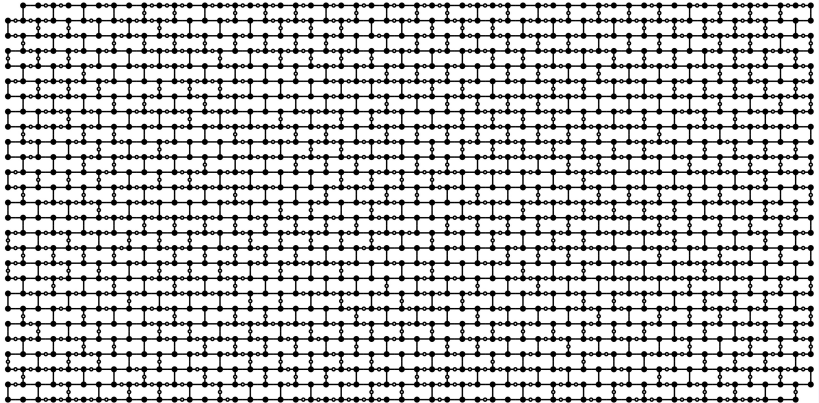
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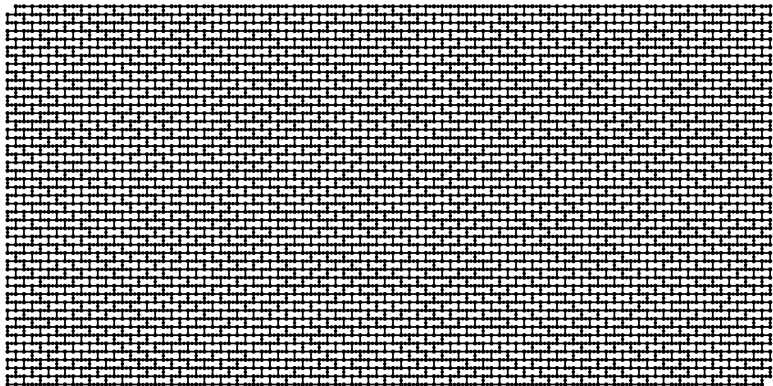
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Actually we can assume we have the subdivision of a **quite big** wall!

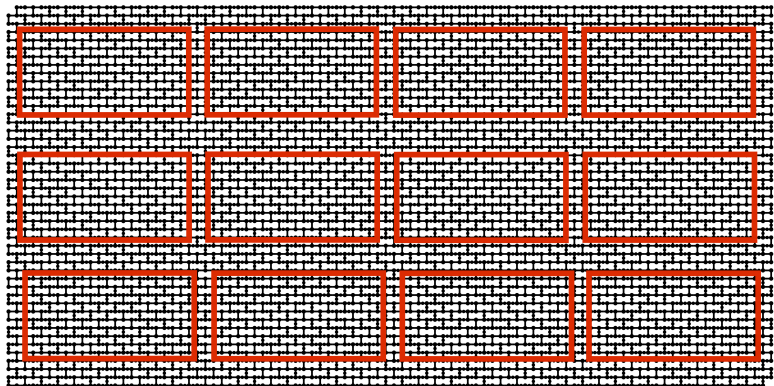


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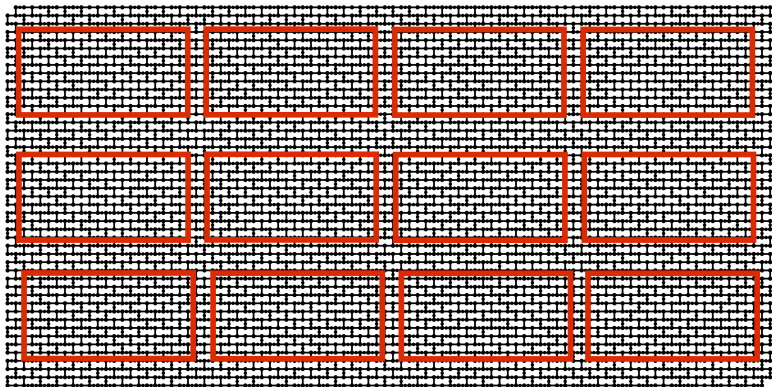
The height of this wall is $h = \lceil \sqrt{k} \rceil \cdot (2(k+1) + 1)$ (in the s-wall above $h = 46$)

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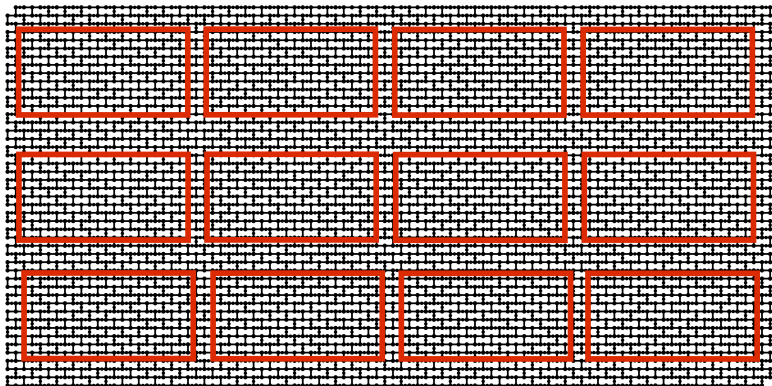
We locate k subwalls, each of height $2 \cdot (k + 1)$.

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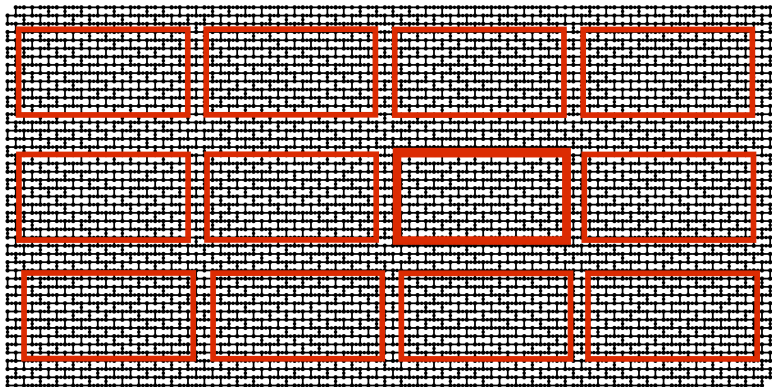
Let G_1, \dots, G_k be the graphs inside the **perimetries** of these subwalls

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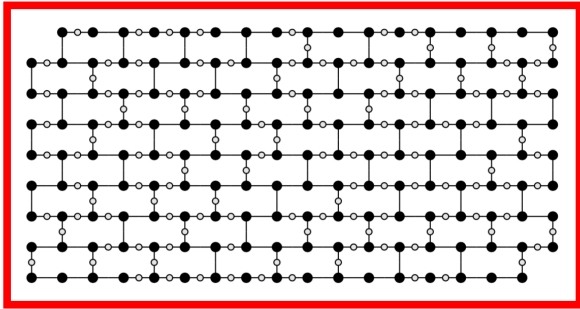


If all of them are **non-bipartite** then we answer **YES** and we are done!

Actually we can assume we have the subdivision of a quite big wall!



If not, then consider the one that is **bipartite**.



If not, then consider the one that is **bipartite**.

All Cycles **entirely inside** the perimetry of this subwall are **even**!

We claim that (given that the height is $2 \cdot (k + 1)$) the middle vertex is **irrelevant**

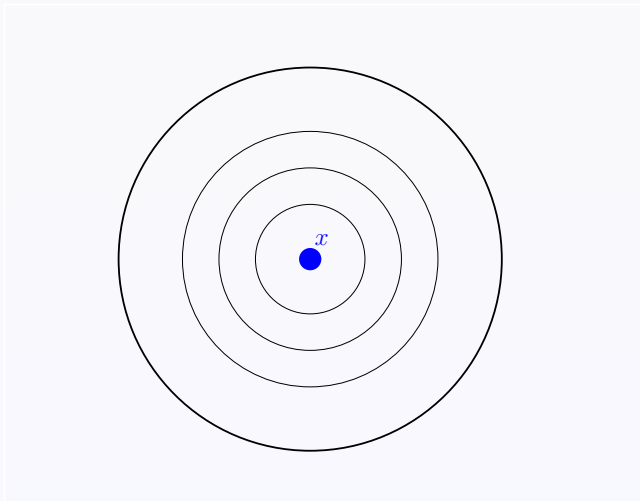
We have to prove that (G, k) is a **YES-instance** $\iff (G \setminus x, k)$ is a **YES-instance**

The \Leftarrow direction is trivial: if $G \setminus x$ has k odd disjoint cycles, so does G .

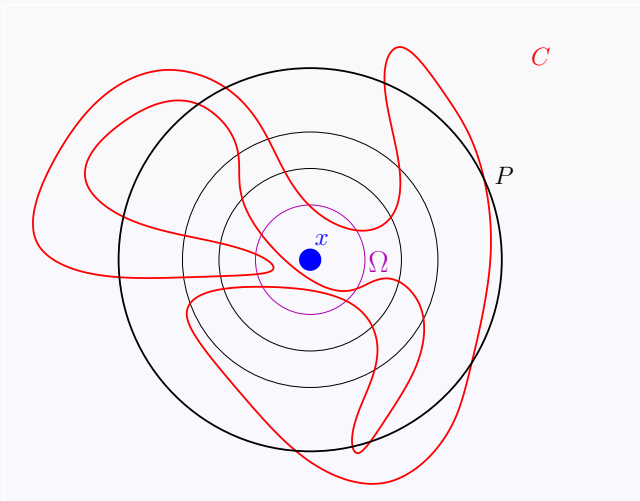
For the " \implies " assume that $G \setminus x$ has $\geq k$ odd cycles.

We will prove that $G \setminus x$ has k odd disjoint cycles avoiding x .

We detect, using the layers of the wall, $k + 1$, homocentric cycles around x
If G has $k + 1$ disjoint odd cycles we are done (x meets only one of them)
Therefore G has at exactly k disjoint odd cycles.

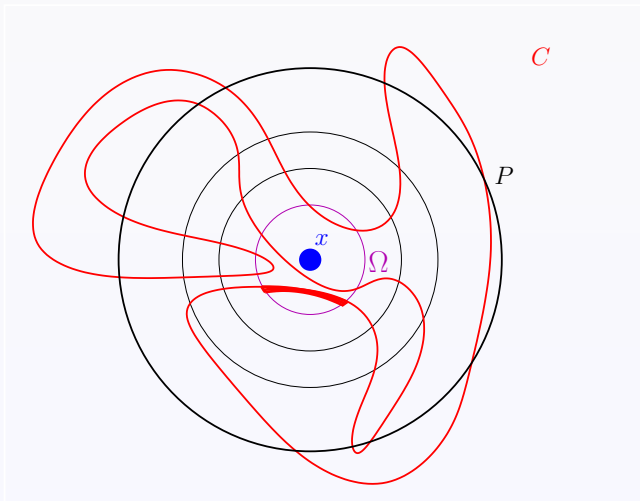


Assume # chords of the k disjoint cycles “cropped” by the homocentric cycles is minimized. For example C crosses Ω 4 times and C crosses perimetry P 5 times. We argue that none of these k cycles can cross the inner cycle Ω , thus x is irrelevant!



Suppose, to the contrary, that some cycle C crosses the inner cycle Ω .

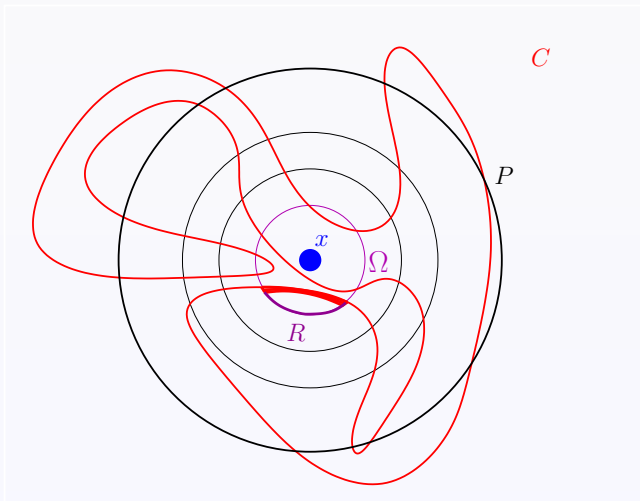
Consider an “**extremal**” chord X : one of the two paths of Ω does not contain any other endpoint of a chord.



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Consider an “**extremal**” chord! This defines a (same parity) segment R of Ω

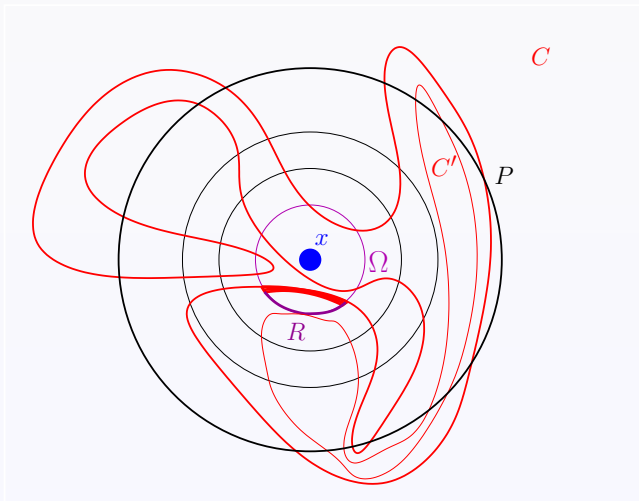
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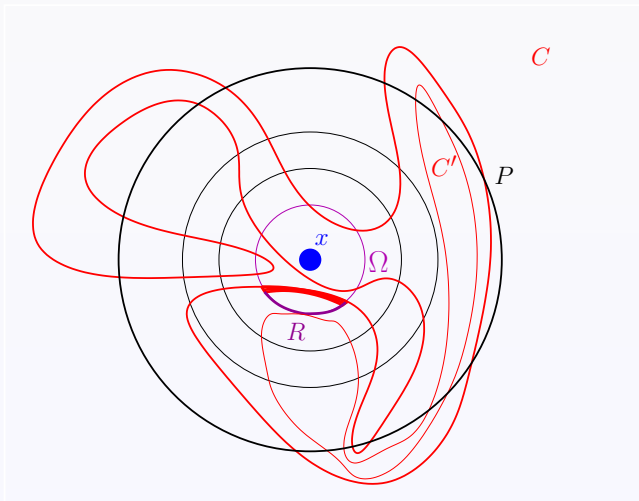
Consider an “**extremal**” chord! This defines a (same parity) segment R of Ω

By minimality R should be met by some other cycle $C' \neq C$.

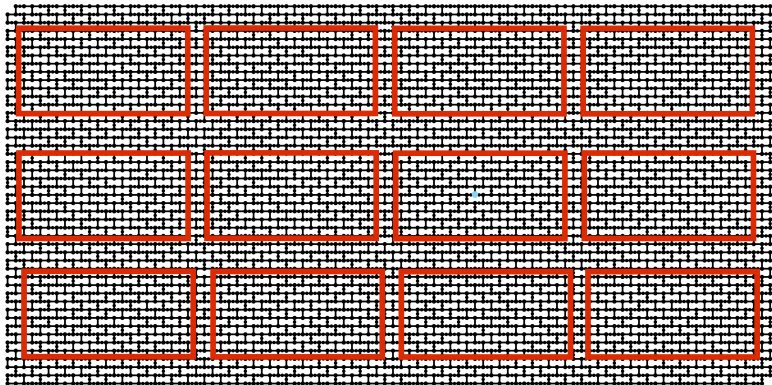


By repetitively applying this argument we find in G has as many disjoint odd cycles as its homocentric cycles that are $k + 1$, a contradiction.

Therefore none of the k disjoint odd cycles crosses Ω . Thus x is irrelevant.



To find the irrelevant vertex x can be done in polynomial time!



General scheme

1. If G has treewidth $O(k^{3/2})$ then **CAVALRY** comes: DP takes $2^{O(\text{tw}(G))} \cdot n$ steps.
- 2: Otherwise check whether we have **VICTORY** (all k subwalls are non-bipartite)!
- 3: Otherwise **FIGHT**: find an irrelevant vertex x , set $G \leftarrow G \setminus x$ and go to 1.

The above proves that p -PLANAR ODD CYCLE PACKING $\in 2^{O(k^{3/2})}$ -FPT

- ▶ All the above arguments extend for graphs of bounded genus! (and **further!**)

The general p -ODD CYCLE PACKING problem is in FPT

[Kawarabayashi, Reed, STOC 2010]

- ▶ The same ideas prove: p -PLANAR ODD INDUCED CYCLE PACKING $\in 2^{O(k^{3/2})}$ -FPT while general p -ODD INDUCED CYCLE PACKING problem is para-NP-hard.

[Golovach, Kamiński, Paulusma, Thilikos, TCS 2012]

Irrelevant vertex technique

Introduced by [Robertson & Seymour GM-XIII] for proving that the following belong in FPT.

p-MINOR CONTAINMENT

Instance: two graphs G and H .

Parameter: $k = |V(H)|$

Question: $H \leq G$?

p-DISJOINT PATHS

Instance: A graph G and a sequence of pairs of terminals $(s_1, t_1), \dots, (s_k, t_k)$.

Parameter: k .

Question: Are there k pairwise vertex disjoint paths P_1, \dots, P_k in G such that for every $i \in \{1, \dots, k\}$, P_i has endpoints s_i and t_i ?

Challenge: go further than planar graphs.

► Further than embedded graphs: a bigger story. Need another school to explain!

Merci beaucoup!



La grand traverse...