

Lecture 2: Graph width notions, dynamical programming

An Introduction to Parameterized Complexity

Clemens Grabmayer

Ph.D. Program, Advanced Period

Gran Sasso Science Institute

L'Aquila, Italy

Tuesday, June 17, 2025

Course overview

Monday, June 16 10.30 – 12.30	Tuesday, June 17 10.30 – 12.30	Wednesday, June 18	Thursday, June 19 10.30 – 12.30	Friday, June 20
<i>Algorithmic Techniques</i>			<i>Formal-Method & Algorithmic Techniques</i>	
Introduction & basic FPT results motivation for FPT kernelization, Crown Lemma, Sunflower Lemma	Notions of bounded graph width path-, tree-, clique width, FPT-results by dynamic programming, transferring FPT results betw. width		Algorithmic Meta-Theorems 1st-order logic, monadic 2nd-order logic, FPT-results by Courcelle's Theorems for tree and clique-width	(Fair Division)
				14.30 – 16.30
				FPT-Intractability Classes & Hierarchies
			(Fair Division)	motivation for FP-intractability results, FPT-reductions, class XP (slice-wise polynomial), W- and A-Hierarchies, placing problems on these hierarchies

Overview

- ▶ comparing parameterizations

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- ▶ dynamical programming on trees, example:
 - ▶ WEIGHTED-INDEPENDENT-SET (and VERTEX-COVER)
- ▶ path-width
 - ▶ example: fpt-algorithm for bounded path-width
- ▶ tree-width
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- ▶ other notions of width
 - ▶ clique-width
 - ▶ using f -width to define:
 - ▶ carving-width (and cut-width)
 - ▶ branch-width
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- ▶ comparing width-notions

Fixed-Parameter tractable

A *parameterized problem* is a triple $\langle Q, \Sigma, \kappa \rangle$ (short: $\langle Q, \kappa \rangle$) where:

- ▷ $Q \subseteq \Sigma^*$ is the set of *(classical) problem instances*,
- ▷ $\kappa : \Sigma^* \rightarrow \mathbb{N}$ is a (general) function, *the parameterization*.

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Parameterized problem $\langle Q, \Sigma, \kappa \rangle$

Instance: $x \in \Sigma^*$.

Parameter: $\kappa(x)$.

Problem: Is $x \in Q$?

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Definition

A parameterized problem $\langle Q, \Sigma, \kappa \rangle$ is *fixed-parameter tractable* (is in **FPT**) if:

$\exists f : \mathbb{N} \rightarrow \mathbb{N}$ computable $\exists p \in \mathbb{N}[X]$ polynomial

$\exists \mathbb{A}$ algorithm, takes inputs in Σ^*

$\forall x \in \Sigma^* \left[\mathbb{A} \text{ decides whether } x \in Q \text{ holds} \right.$
 $\left. \text{in time } \leq f(\kappa(x)) \cdot p(|x|) \right]$

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†) Assumptions for a robust fpt-theory

$\kappa(x)$ is *polynomially computable*, or itself *fpt-computable*: for all $x \in \Sigma^*$ in time $\leq g(\kappa(x)) \cdot q(|x|)$ for g computable, $q \in \mathbb{N}[X]$.

Comparing parameterizations

Definition (computably bounded below)

Let $\kappa_1, \kappa_2 : \Sigma^* \rightarrow \mathbb{N}$ parameterizations.

- ▶ $\kappa_1 \succeq \kappa_2 : \iff \exists g : \mathbb{N} \rightarrow \mathbb{N} \text{ computable } \forall x \in \Sigma^* [g(\kappa_1(x)) \geq \kappa_2(x)]$.
- ▶ $\kappa_1 \approx \kappa_2 : \iff \kappa_1 \succeq \kappa_2 \wedge \kappa_2 \succeq \kappa_1$.
- ▶ $\kappa_1 \succ \kappa_2 : \iff \kappa_1 \succeq \kappa_2 \wedge \neg(\kappa_2 \succeq \kappa_1)$.

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Proposition

For all parameterized problems $\langle Q, \kappa_1 \rangle$ and $\langle Q, \kappa_2 \rangle$ with parameterizations $\kappa_1, \kappa_2 : \Sigma^* \rightarrow \mathbb{N}$ with $\kappa_1 \geq \kappa_2$:

$$\langle Q, \kappa_1 \rangle \in \text{FPT} \iff \langle Q, \kappa_2 \rangle \in \text{FPT}$$

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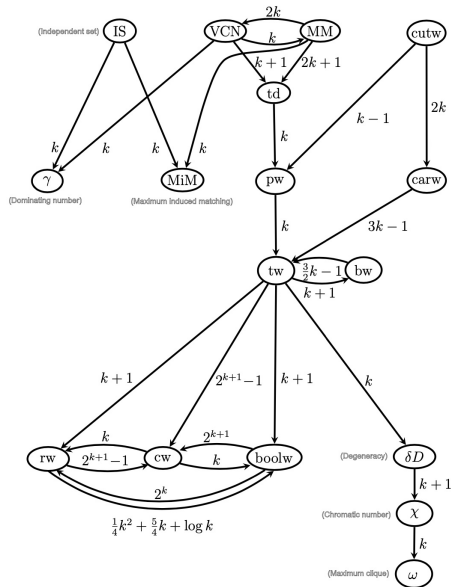
$$\langle Q, \kappa_1 \rangle \in \text{FPT} \iff \langle Q, \kappa_2 \rangle \in \text{FPT}$$

$$\langle Q, \kappa_1 \rangle \notin \text{FPT} \implies \langle Q, \kappa_2 \rangle \notin \text{FPT}$$

Computably boundedness between notions of width

(from Sasák, [5])

$$wd_1 \succeq wd_2 : \Leftrightarrow wd_1 \xrightarrow{g} wd_2$$

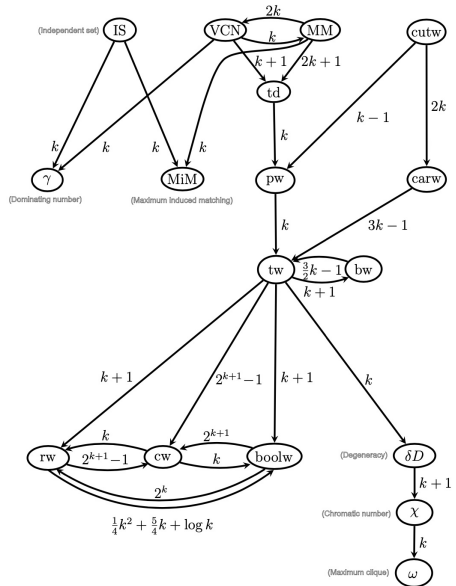


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transfer upwards
(and conversely to $\overset{g}{\rightarrow}$)



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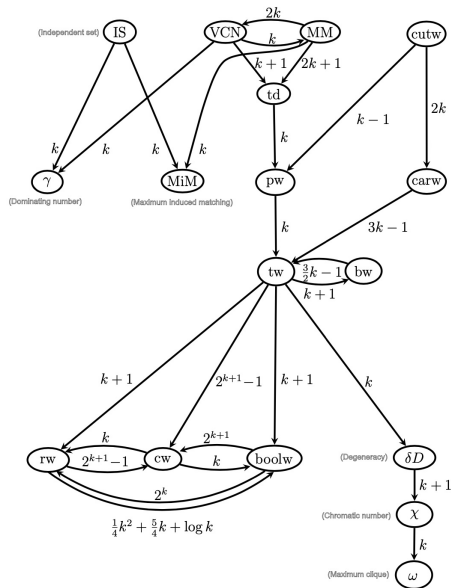
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► (\notin FPT)-results

transfer downwards

(and along \xrightarrow{g})



You Always Walk Alone (with your children)

Attività motoria con i figli:

'la possibilità di uscire con i figli minori è consentita a un solo genitore per camminare purché questo avvenga in prossimità della propria abitazione'

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PHYSICAL-DISTANCE-WALKING

Instance: Graph $\mathcal{G} = \langle V, E \rangle$ with V people who want to go for a walk in the next hour in a radius of 200m of their home, and edges in E between them if they live closer than 400m of each other. A number $\ell \in \mathbb{N}$.

Problem:

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corresponds to: INDEPENDENT-SET

Weighted Independent Set, and Vertex Cover

Let $\mathcal{G} = \langle V, E \rangle$ a graph. For all $S \subseteq V$:

S is **independent set** in $\mathcal{G} : \iff \forall e = \{u, v\} \in E \ (\neg(u \in S \wedge v \in S))$
 $\iff \forall e = \{u, v\} \in E \ (u \notin S \vee v \notin S)$

WEIGHTED-INDEPENDENT-SET

Instance: A graph $\mathcal{G} = \langle V, E \rangle$, and a **weight function** $w : V \rightarrow \mathbb{R}_0^+$.

Problem: What is the **max. weight of an independent set** of \mathcal{G} ?

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Instance: A graph $\mathcal{G} = \langle V, E \rangle$, and $\ell \in \mathbb{N}$.

Problem: Does \mathcal{G} have a vertex cover of size at most ℓ ?

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$S \subseteq V$ is **minimal** vertex cover $\iff V \setminus S$ is **maximal** independent set

Hence: solution of WEIGHTED-INDEPENDENT-SET

\implies solution of VERTEX-COVER.

Weighted Ind. Set / Vertex Cover, width-parameterized

p^* -WEIGHTED-INDEPENDENT-SET

Instance: A graph $\mathcal{G} = \langle V, E \rangle$, and a weight function $w : V \rightarrow \mathbb{R}_0^+$.

Parameter: path-width / tree-width k .

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Dynamical programming on trees (example)

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Solution: value of $A[r]$, can be computed bottom-up in linear time.

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On *trees* with n nodes,

WEIGHTED-INDEPENDENT-SET $\in \text{DTIME}(O(n))$.

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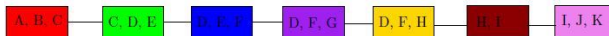
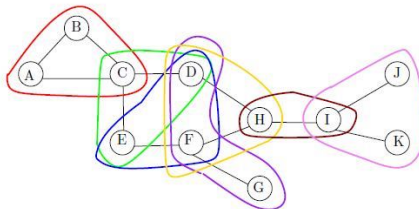
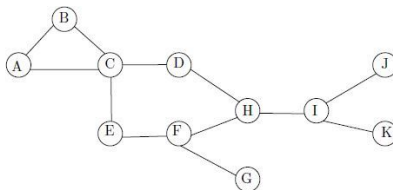
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Corollary

On trees with n nodes,

VERTEX-COVER $\in \text{DTIME}(O(n))$.

Path-decomposition (example)



Path decompositions, and path-width

Definition (Robertson–Seymour, 1983)

A *path decomposition* of a graph $\mathcal{G} = \langle V, E \rangle$ is a sequence $\langle B_1, B_2, \dots, B_r \rangle$ of bags $B_i \subseteq V$ such that:

(P1) $V = \bigcup_{i=1}^r B_i$ (every vertex of \mathcal{G} is in some bag).

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Path decompositions, and path-width

Definition (Robertson–Seymour, 1983)

A **path decomposition** of a graph $\mathcal{G} = \langle V, E \rangle$ is a sequence $\langle B_1, B_2, \dots, B_r \rangle$ of bags $B_i \subseteq V$ such that:

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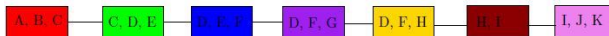
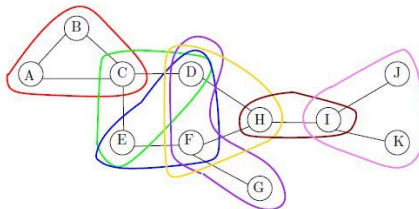
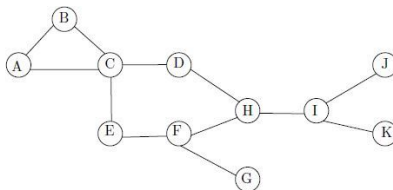
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$\text{pw}(\mathcal{G}) :=$ minimal width of a path decomposition of \mathcal{G} .

Path-decomposition (example)



Path decomposition defines separations

Lemma

Let $\langle B_1, B_2, \dots, B_r \rangle$ be a path decomposition of a graph $\mathcal{G} = \langle V, E \rangle$. Then for all $i \in \{1, \dots, r-1\}$ it holds:

- ▶ *$\langle \bigcup_{j=1}^i B_j, \bigcup_{j=i+1}^r B_j \rangle$ is a separation of \mathcal{G} with separator $B_i \cap B_{i+1}$.*

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▶ $V = A \cup B$

▶ there is **no edge** between $A \setminus B$ and $B \setminus A$.

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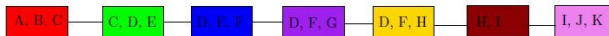
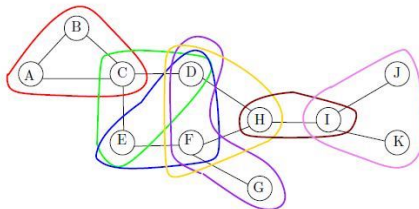
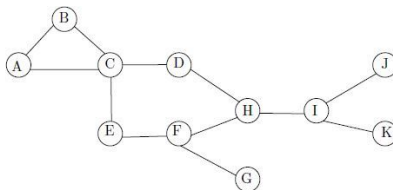
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- ▶ The *border (set of border vertices)* $\partial(A)$ of a set $A \subseteq V$ of vertices consists of all vertices that have a neighbor in $V \setminus A$. Note that:

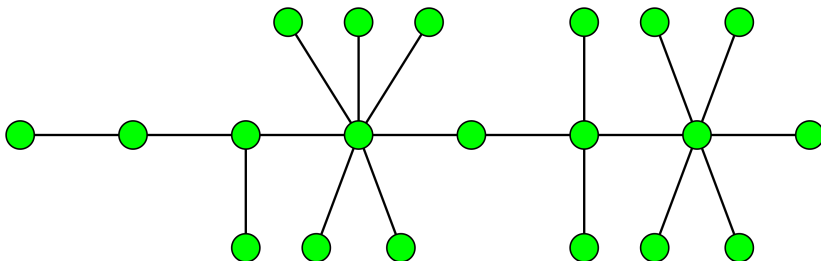
- ▶ $\partial(A) = \partial(V \setminus A)$.
- ▶ $\langle A, (V \setminus A) \cup \partial(A) \rangle$ is a separation of \mathcal{G} , for all $A \subseteq V$.

Path-decomposition (example)



Caterpillar

Path-width?



Nice path decomposition

Definition

A *path decomposition* $\langle B_1, B_2, \dots, B_r \rangle$ of a graph $\mathcal{G} = \langle V, E \rangle$ is *nice* if:

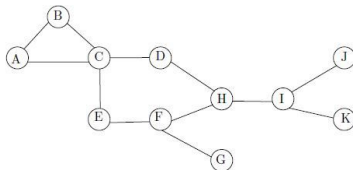
- ▶ $B_1 = B_r = \emptyset$
- ▶ Every index $i > 1$ is either of:
 - ▶ **introduce index**: there is $v \in V$ such that $B_{i+1} = B_i \cup \{v\}$ and $v \notin B_i$,
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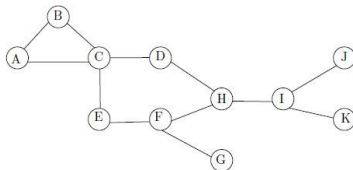


Nice path decomposition

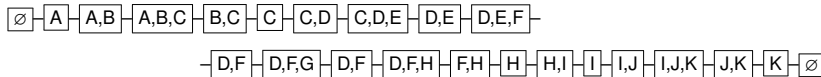
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Nice path decomposition:



Nice path decomposition

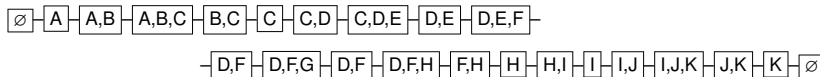
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Lemma

From every *path decomposition* $\langle B_1, B_2, \dots, B_r \rangle$ of a graph $\mathcal{G} = \langle V, E \rangle$ of width k a *nice path decomposition* $\langle B'_1, B'_2, \dots, B'_{r'} \rangle$ of width k can be constructed in time $O(k^2 \cdot \max\{r, n\})$ where $n := |V|$.



Weighted Independent Set

Let $\mathcal{G} = \langle V, E \rangle$ a graph.

$S \subseteq V$ is **independent set** in $\mathcal{G} : \iff \forall e = \{u, v\} \left(\neg(u \in S \wedge v \in S) \right)$.

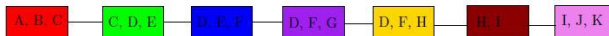
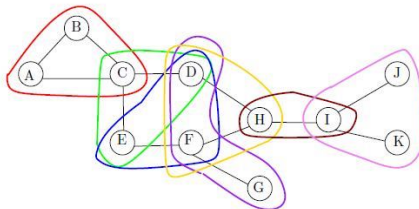
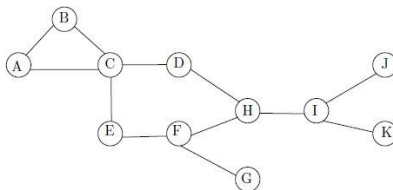
WEIGHTED-INDEPENDENT-SET

Instance: A graph $\mathcal{G} = \langle V, E \rangle$, and a **weight function** $w : V \rightarrow \mathbb{R}_0^+$.

Parameter: **path-width** k .

Problem: What is the **max. weight of an independent set** of \mathcal{G} ?

Path-decomposition (example)



Dyn. programming using path-width (Weigh. Ind. Set)

Let $\langle B_1, \dots, B_r \rangle$ be a **nice path decomposition** of $\mathcal{G} = \langle V, E \rangle$.

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$$c[i, S] := \begin{cases} -\infty & \text{if } S \text{ is not independent} \\ \text{maximum possible weight of a set } \hat{S} \text{ such that} \\ \hat{S} \text{ is independent} \wedge S \subseteq \hat{S} \subseteq V_i = \bigcup_{j=1}^i B_j \wedge \hat{S} \cap B_i = S & \\ \text{if } S \text{ is independent.} \end{cases}$$

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Dyn. programming using path-width (Weigh. Ind. Set)

Let $\langle B_1, \dots, B_r \rangle$ be a nice path dec. of $\mathcal{G} = \langle V, E \rangle$ of width k .

For every $i \in \{1, \dots, r\}$, and every independent $S \subseteq B_i$, we define:

$$c[i, S] := \begin{cases} \text{maximum possible weight of a set } \hat{S} \text{ such that} \\ \hat{S} \text{ is independent} \wedge S \subseteq \hat{S} \subseteq V_i = \bigcup_{j=1}^i B_j \wedge \hat{S} \cap B_i = S \end{cases}$$

Time Complexity: Based on the values of $c[i, S]$, the maximum possible weight of an independent set $S \subseteq V$ can be computed as:

$$= \max_{S \subseteq B_r} c[r, S] = c[r, \emptyset]$$

Then for all $i \in \{1, \dots, n\}$:

- ▶ $|B_i| \leq k + 1$,
- ▶ \Rightarrow number of values $c[i, S]$ at index i : $2^{|B_i|} = 2^{k+1}$,
- ▶ \Rightarrow adjacency/independence check for $S \subseteq B_t$ possible in: $O(k^2)$ using a datastructure computable in time $O(k^{O(1)} \cdot n)$,
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\Rightarrow the time for computing all values at r :

$$(2^{k+1} \cdot O(k^2)) \cdot r + O(k^{O(1)} \cdot n) \in O(2^k \cdot k^{O(1)} \cdot n), \text{ since } r = 2n.$$

Dynamical programming with path width (example)

Theorem

For every graph $\mathcal{G} = \langle V, E \rangle$ with $|V| = n$ and *path-width* $\text{pw}(\mathcal{G}) = k$,
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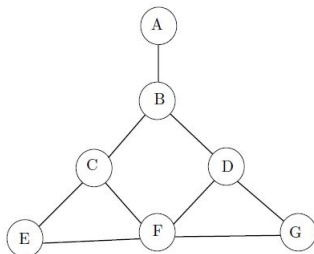
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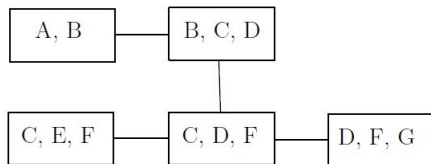
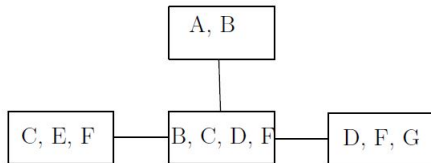
Corollary

For every graph $\mathcal{G} = \langle V, E \rangle$ with $|V| = n$ and *path-width* $\text{pw}(\mathcal{G}) = k$,
 p^* -VERTEX-COVER $\in \text{DTIME}(2^k \cdot k^{O(1)} \cdot n)$.

Tree decomposition (example)



The original graph G



Tree decompositions, and tree-width

Definition (Bertelé–Brioschi, 1972, Halin, 1976, Robertson–Seymour, 1984)

A *tree decomposition* of a graph $\mathcal{G} = \langle V, E \rangle$ is a pair $\langle \mathcal{T}, \{B_t\}_{t \in T} \rangle$ where $\mathcal{T} = \langle T, F \rangle$ a (undirected, unrooted) tree, and $B_t \subseteq V$ such that:

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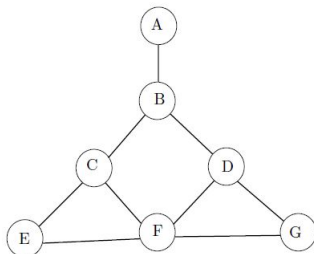
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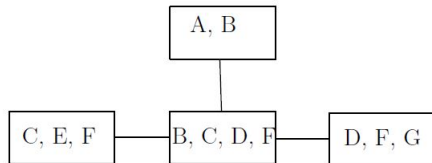
The **tree-width** $tw(\mathcal{G})$ of a graph $\mathcal{G} = \langle V, E \rangle$ is defined by:

$tw(\mathcal{G}) :=$ minimal width of a tree decomposition of \mathcal{G} .

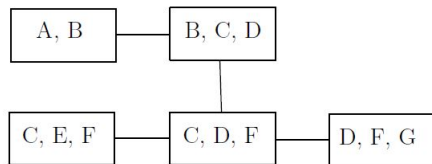
Tree decomposition (example)



The original graph G



A tree-decomposition of width 3



A tree-decomposition of width 2

Tree decomposition defines separations

Lemma

Let $\langle \mathcal{T}, \{B_t\}_{t \in T} \rangle$ be a tree decomposition of a graph $\mathcal{G} = \langle V, E \rangle$.
 Let $e = \langle a, b \rangle$ be an edge of \mathcal{T} . The $\mathcal{T} \setminus e$ is the union of a tree \mathcal{T}_a containing a , and a tree \mathcal{T}_b containing b .

Then for $A := \bigcup_{t \in V(\mathcal{T}_a)} B_t$ and $B := \bigcup_{t \in V(\mathcal{T}_b)} B_t$ it holds:

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Recall, for a graph $\mathcal{G} = \langle V, E \rangle$:

- ▶ A pair $\langle A, B \rangle$ of subsets $A, B \subseteq V$ is a *separation* of \mathcal{G} if:
 - ▶ $V = A \cup B$
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- ▶ The *border (vertices)* $\partial(A)$ of a set $A \subseteq V$ of vertices consists of all vertices that have a neighbor in $V \setminus A$.

Computing tree-width

TREE-WIDTH

Instance: A graph \mathcal{G} and $k \in \mathbb{N}$.

Problem: Decide whether $tw(\mathcal{G}) = k$.

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Theorem

p -TREE-WIDTH is fixed-parameter tractable,
in time $2^{p(k)} \cdot n$ where $n := |V|$.

Nice tree decomposition

Definition

A *tree decomposition* $\langle \mathcal{T}, \{B_t\}_{t \in T} \rangle$ of graph $\mathcal{G} = \langle V, E \rangle$ is *nice* if it is based on the choice of a leaf as *root* r and the parent–children relation away from r such that:

- ▶ $B_r = \emptyset$, and $B_\ell = \emptyset$ for every leaf $\ell \in T$.
- ▶ Every non-leaf node $t \in T$ is of one of three types:
 - ▶ **introduce node**: t has exactly one child t' such that $B_t = B_{t'} \cup \{v\}$; we say v is **introduced** at t .
 - ▶ **forget node**: t has exactly one child t' such that $B_t = B_{t'} \setminus \{w\}$ for some $w \in B_{t'}$; we say w is **forgotten** at t .
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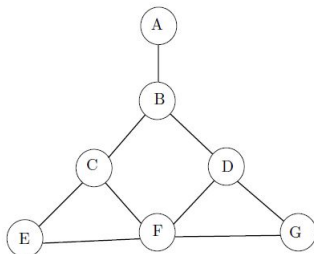
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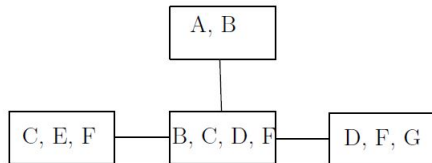
Lemma

From every *tree decomposition* $\langle \mathcal{T}, \{B_t\}_{t \in T} \rangle$ of a graph $\mathcal{G} = \langle V, E \rangle$ of *width* k a *nice tree decomposition* $\langle \mathcal{T}', \{B'_t\}_{t \in T'} \rangle$ of *width* k and with $r := |V(\mathcal{T})| \in O(kn)$ vertices can be constructed in time $O(k^2 \cdot \max\{r, n\})$ where $n := |V|$.

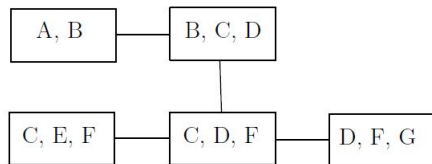
Tree decomposition (example)



The original graph G



A tree-decomposition of width 3



A tree-decomposition of width 2

Weighted Independent Set

Let $\mathcal{G} = \langle V, E \rangle$ a graph.

$S \subseteq V$ is **independent set** in $\mathcal{G} : \iff \forall e = \{u, v\} \left(\neg(u \in S \wedge v \in S) \right)$.

WEIGHTED-INDEPENDENT-SET

Instance: A graph $\mathcal{G} = \langle V, E \rangle$, and a **weight function** $w : V \rightarrow \mathbb{R}_0^+$.

Parameter: **tree-width** k .

Problem: What is the **max. weight of an independent set** of \mathcal{G} ?

Dynamical programming using tree-width (example)

For every node t of a [nice tree decomposition](#), and every $S \subseteq B_t$, we define:

$$c[t, S] := \begin{cases} -\infty & \text{if } S \text{ is not independent,} \\ \text{maximum possible weight of a set } \hat{S} \text{ such that} \\ \hat{S} \text{ is independent} \wedge S \subseteq \hat{S} \subseteq V_t \wedge \hat{S} \cap B_t = S & \\ \text{if } S \text{ is independent.} \end{cases}$$

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\Rightarrow the time for computing all values at the root r :

$$(2^{k+1} \cdot O(k^2)) \cdot |T| + O(k^{O(1)} \cdot n) \in O(2^k \cdot k^{O(1)} \cdot n), \text{ since } |T| \in O(k \cdot n).$$

Dynamical programming with tree width (example)

Theorem

For every graph $\mathcal{G} = \langle V, E \rangle$ with $|V| = n$ and *tree-width* $tw(\mathcal{G}) = k$,
 p^* -WEIGHTED-INDEPENDENT-SET $\in \text{DTIME}(2^k \cdot k^{O(1)} \cdot n)$.

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Corollary

For every graph $\mathcal{G} = \langle V, E \rangle$ with $|V| = n$ and *tree-width* $tw(\mathcal{G}) = k$,
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Dyn. programming with tree-width: general strategy

We consider problem P for graphs $\mathcal{G} = \langle V, E \rangle$ of size n and nice tree decompositions $\langle \mathcal{T} = \langle T, F, r \rangle, \{B_t\}_{t \in T} \rangle$ of tree width k .

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- ▶ **Obtain** an estimate of the time needed to compute the properties in a node t depending on n and k .
- ▶ **Sum up** the time needed to compute the solution(s) at root r of \mathcal{T} .
- ▶ **Add** time needed to obtain the solution of P from properties at r .

Dynamical programming: similar results (I)

Theorem

For every graph $\mathcal{G} = \langle V, E \rangle$ with $|V| = n$ and $tw(\mathcal{G}) = k$,

- ▶ p^* -VERTEX-COVER, INDEPENDENT-SET $\in \text{DTIME}(2^k \cdot k^{O(1)} \cdot n)$,
- ▶ p^* -DOMINATING-SET $\in \text{DTIME}(4^k \cdot k^{O(1)} \cdot n)$,
- ▶ p^* -ODD CYCLE TRAVERSAL $\in \text{DTIME}(3^k \cdot k^{O(1)} \cdot n)$,
- ▶ p^* -MAXCUT $\in \text{DTIME}(2^k \cdot k^{O(1)} \cdot n)$,
- ▶ p^* - q -COLORABILITY $\in \text{DTIME}(q^k \cdot k^{O(1)} \cdot n)$.

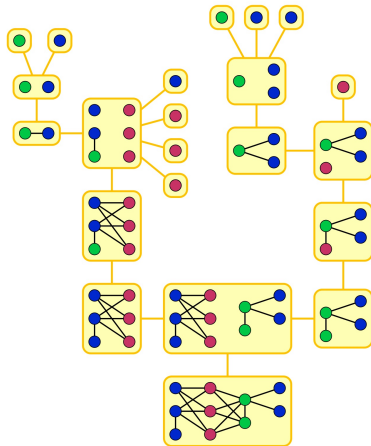
Dynamical programming: similar results (II)

Theorem

For every graph $\mathcal{G} = \langle V, E \rangle$ with $|V| = n$ and $tw(\mathcal{G}) = k$, the following problems are in $\text{DTIME}(k^{O(k)} \cdot n)$:

- ▶ p^* -STEINER-TREE,
- ▶ p^* -FEEDBACK-VERTEX-SET,
- ▶ p^* -HAMILTONIAN-PATH *and* p^* -LONGEST-PATH,
- ▶ p^* -HAMILTONIAN-CYCLE *and* p^* -LONGEST-CYCLE,
- ▶ p^* -CHROMATIC-NUMBER,
- ▶ p^* -CYCLE-PACKING,
- ▶ p^* -CONNECTED-VERTEX-COVER,
- ▶ p^* -CONNECTED-FEEDBACK-VERTEX-SET.

Clique width (example)



Clique-Width

For $k \in \mathbb{N}$, the *k -expressions* are defined by:

$$\varphi, \varphi_1, \varphi_2 ::= i \mid \text{edge}_{i-j}(\varphi) \mid \text{recolor}_{i \rightarrow j}(\varphi) \mid (\varphi_1 \oplus \varphi_2)$$

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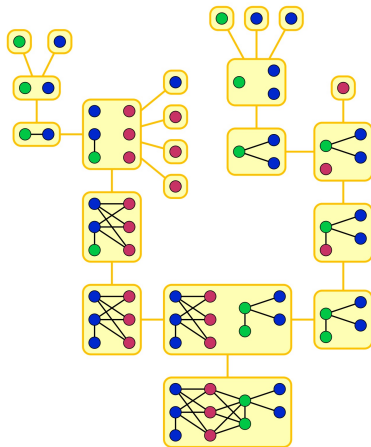
Definition (Courcelle, Engelfriet, Rozenberg, 1993, [2])

The *clique-width* $\text{clw}(\mathcal{G})$ of $\mathcal{G} = \langle V, E \rangle$ is defined by:

$$\text{clw}(\mathcal{G}) := \text{the least } k \in \mathbb{N} \text{ such that, for some } k\text{-expression } \varphi, \\ \mathcal{G} = \mathcal{G}(\varphi) \text{ (when removing colors)}$$

Clique width (example)

Building a graph \mathcal{G} of clique-width $c/w(\mathcal{G}) = 3$:



Clique-Width (examples, properties, computability)

Example

- ▶ The class of cliques has clique-width 2.

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 - ▶ $clw \leq tw$: $clw(\mathcal{G}) \leq 3 \cdot 2^{tw(\mathcal{G})-1}$
 - ▶ $\neg(tw \leq clw)$: for example, $clw(K_n) = 2$, and $tw(K_n) = n - 1$.

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 - ▶ Deciding whether $c/w(\mathcal{G}) \leq k$ is **NP-hard**. With parameter k it is in XP (slice-wise polynomial), but unknown to be in FPT.
 - ▶ Every graph property expressible in **MSO (monadic second-order logic)** can be decided in linear time w.r.t. the graph's clique-width.

f -Width (of sets)

By a *cut function* or a *connectivity function* we mean a function $f : 2^U \rightarrow \mathbb{R}_0^+$ such that:

$$f \text{ is } \textit{symmetric} : \iff \forall X \subseteq U \left[f(X) = f(U \setminus X) \right];$$

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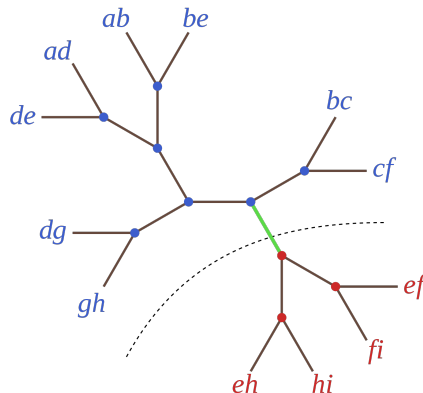
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Let U be a set, $f : 2^U \rightarrow \mathbb{R}_0^+$ a cut function.

A *branch decomposition* of U is a pair $\langle \mathcal{T}, \eta \rangle$ where:

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Every edge $e \in T$ splits the tree into two connected parts, and, via η , splits U into a partition $\langle X_e, Y_e \rangle$.

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- ▷ $\eta : U \rightarrow \text{Leafs}(\mathcal{T})$ a bijective function.

Every edge $e \in T$ splits the tree into two connected parts, and, via η , splits U into a partition $\langle X_e, Y_e \rangle$.

The *width* of an edge $e \in T$ (with respect to f) is $f(X_e) = f(Y_e)$. The *width of $\langle \mathcal{T}, \eta \rangle$ w.r.t. f* is the maximum width over the edges of \mathcal{T} .

f -Width (of sets)

By a *cut function* or a *connectivity function* we mean a function $f : 2^U \rightarrow \mathbb{R}_0^+$ such that:

$$f \text{ is symmetric: } \iff \forall X \subseteq U \left[f(X) = f(U \setminus X) \right];$$

$$f \text{ is fair: } \iff f(\emptyset) = f(U) = 0.$$

Definition

Let U be a set, $f : 2^U \rightarrow \mathbb{R}_0^+$ a cut function.

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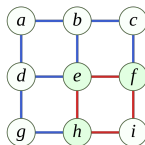
$$w_f(U) := \text{minimum width of branch decomp's of } U \text{ w.r.t. } f.$$

Branch-Width

Definition

Let $\mathcal{G} = \langle V, E \rangle$ be a graph. The *border (vertices)* of a set $X \subseteq E$ of edges is defined by:

$$\partial(X) := \left\{ v \in V \mid \exists e_1 \in X \exists e_2 \in E \setminus X \right. \\ \left. [v \text{ is incident to } e_1 \text{ and } e_2] \right\}$$



Branch-Width

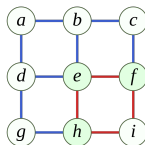
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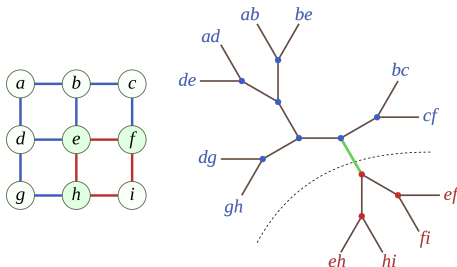
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Proposition

$bw(\mathcal{G}) \approx tw(\mathcal{G})$, for every graph; more precisely:

$$bw(\mathcal{G}) \leq tw(\mathcal{G}) + 1 \leq \frac{3}{2} \cdot bw(\mathcal{G}).$$

Rank-Width

Definition

Let $\mathcal{G} = \langle V, E \rangle$ be a graph.

For $X \subseteq V$ we define the $GF(2)$ -matrix:

$$B_{\mathcal{G}}(X) := (b_{x,y})_{x \in X, y \in V \setminus X}, \text{ where, for all } x \in X, y \in V \setminus X:$$

$$b_{x,y} = 1 \iff \{x, y\} \in E.$$

($B_{\mathcal{G}}(X)$ is the adjacency matrix of the bipartite graph induced by \mathcal{G} between X and $V \setminus X$.)

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The *rank-width* $rw(\mathcal{G})$ of a graph $\mathcal{G} = \langle G, E \rangle$ is:

$$rw(\mathcal{G}) := w_{\rho_{\mathcal{G}}}(E) \quad \text{for} \quad \rho_{\mathcal{G}} : 2^V \rightarrow \mathbb{N}_0, X \mapsto \text{rank of } B_{\mathcal{G}}(X)$$

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Properties

- ▶ $rw(\mathcal{G}) \leq tw(\mathcal{G})$.
- ▶ tree-width cannot be bounded functionally by rank-width:
 $rw(K_n) = 1$, but $tw(K_n) = n - 1$.

Carving-Width

Definition

Let $\mathcal{G} = \langle V, E \rangle$ be a graph.

For $X \subseteq V$ the *edge-cut* of X is:

$$\text{cut}_{\mathcal{G}}(X) := \{e = \{u, v\} \in E \mid u \in X, v \in V \setminus X\} .$$

The *carving-width* $\text{carw}(\mathcal{G})$ of a graph $\mathcal{G} = \langle G, E \rangle$ is:

$$\text{carw}(\mathcal{G}) := w_{\text{cut}}(E) \quad \text{for} \quad \text{cut} : 2^V \rightarrow \mathbb{N}_0, X \mapsto |\text{cut}_{\mathcal{G}}(X)| .$$

Carving-Width and Cut-Width

Definition

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Definition

Let $\mathcal{G} = \langle V, E \rangle$ be a graph with $n = |V|$.

For a permutation $\pi : \{1, \dots, n\} \rightarrow V$ on V we define:

$$\text{width}(\pi) := \max_{1 \leq i \leq n} \text{cut}_{\mathcal{G}}(\{\pi(j) \mid 1 \leq j \leq i\}) .$$

The **cut-width** $\text{cutw}(\mathcal{G})$ of \mathcal{G} is:

$$\text{cutw}(\mathcal{G}) := \min_{\pi \text{ perm. of } V} \text{width}(\pi) .$$

Coverage in Multi-Interface Networks



Coverage in Multi-Interface Networks



$CMI(p)$ (for $p \in \mathbb{N}$)

Instance: A graph $\mathcal{G} = \langle V, E \rangle$, $W : V \rightarrow 2^{\{1, \dots, a\}}$ available-interface allocation, $c : \{1, \dots, a\} \rightarrow \mathbb{R}^+$ interface cost function.

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Problem: Obtain, if possible, a **minimal solution** with respect to the total cost of the interfaces that are activated, that is, $c(W_A) = \sum_{v \in V} \sum_{i \in W_A(v)} c(i)$.

Coverage in Multi-Interface Networks

Theorem

$CMI(2) \in \text{NP-complete}$, also for graphs with max. node degree ≥ 4 .

Coverage in Multi-Interface Networks (parameterized)

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- For path-width $pw(\mathcal{G}) = k$,
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Corollary

$(p^*)'\text{-CMI}(p) \in \text{FPT}$.

Comparing parameterizations

Definition (computably bounded)

Let $\kappa_1, \kappa_2 : \Sigma^* \rightarrow \mathbb{N}$ parameterizations.

- ▶ $\kappa_1 \succeq \kappa_2 : \iff \exists g : \mathbb{N} \rightarrow \mathbb{N} \text{ computable } \forall x \in \Sigma^* [g(\kappa_1(x)) \geq \kappa_2(x)]$.
- ▶ $\kappa_1 \approx \kappa_2 : \iff \kappa_1 \succeq \kappa_2 \wedge \kappa_2 \succeq \kappa_1$.
- ▶ $\kappa_1 \succ \kappa_2 : \iff \kappa_1 \succeq \kappa_2 \wedge \neg(\kappa_2 \succeq \kappa_1)$.

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- ▶ $\kappa_1 \approx \kappa_2 : \iff \kappa_1 \geq \kappa_2 \wedge \kappa_2 \geq \kappa_1$.
- ▶ $\kappa_1 > \kappa_2 : \iff \kappa_1 \geq \kappa_2 \wedge \neg(\kappa_2 \geq \kappa_1)$.

Proposition

For all parameterized problems $\langle Q, \kappa_1 \rangle$ and $\langle Q, \kappa_2 \rangle$ with parameterizations $\kappa_1, \kappa_2 : \Sigma^* \rightarrow \mathbb{N}$ with $\kappa_1 \geq \kappa_2$:

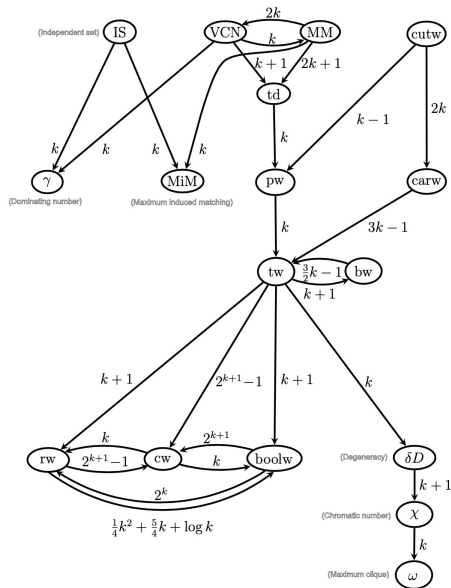
$$\langle Q, \kappa_1 \rangle \in \text{FPT} \iff \langle Q, \kappa_2 \rangle \in \text{FPT}$$

$$\langle Q, \kappa_1 \rangle \notin \text{FPT} \implies \langle Q, \kappa_2 \rangle \notin \text{FPT}$$

Computably boundedness between notions of width

(from Sasák, [5])

$$wd_1 \succeq wd_2 : \Leftarrow wd_1 \xrightarrow{g(k)} wd_2$$



Computably boundedness between notions of width

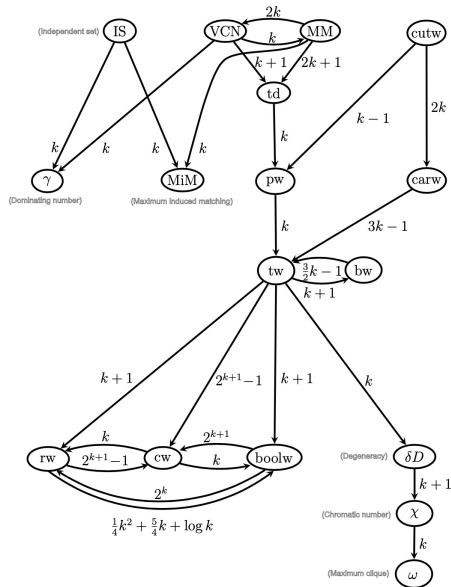
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► FPT-results

transfer upwards

(and conversely to \xrightarrow{g})



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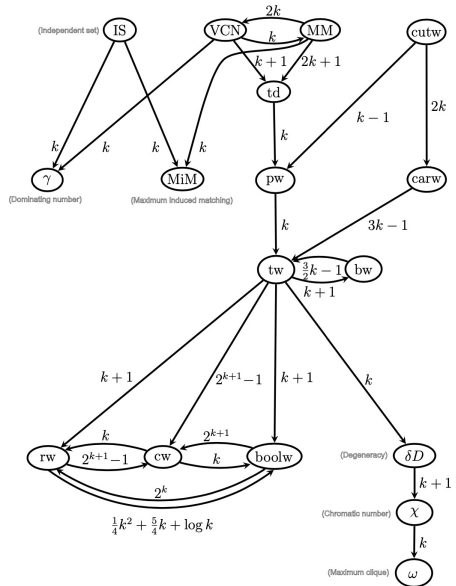
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► (∉ FPT)-results

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Summary

- ▶ comparing parameterizations
- ▶ dynamical programming on trees, example:
 - ▶ WEIGHTED-INDEPENDENT-SET (and VERTEX-COVER)
- ▶ path-width
 - ▶ example: fpt-algorithm for bounded path-width
- ▶ tree-width
 - ▶ example: fpt-algorithm for bounded path-width
- ▶ fpt-results for other problems, obtained similarly
- ▶ other notions of width
 - ▶ clique-width
 - ▶ using f -width to define:
 - ▶ carving-width (and cut-width)
 - ▶ branch-width
 - ▶ rank-width
- ▶ example problem: coverage in multi-interface networks
- ▶ comparing width-notions

Course overview

Monday, June 16 10.30 – 12.30	Tuesday, June 17 10.30 – 12.30	Wednesday, June 18	Thursday, June 19 10.30 – 12.30	Friday, June 20
<i>Algorithmic Techniques</i>			<i>Formal-Method & Algorithmic Techniques</i>	
Introduction & basic FPT results motivation for FPT kernelization, Crown Lemma, Sunflower Lemma	Notions of bounded graph width path-, tree-, clique width, FPT-results by dynamic programming, transferring FPT results betw. width		Algorithmic Meta-Theorems 1st-order logic, monadic 2nd-order logic, FPT-results by Courcelle's Theorems for tree and clique-width	(Fair Division)
				14.30 – 16.30
				FPT-Intractability Classes & Hierarchies
			(Fair Division)	motivation for FP-intractability results, FPT-reductions, class XP (slicewise polynomial), W- and A-Hierarchies, placing problems on these hierarchies

Thursday

- ▶ recalling notions from logic:
 - ▶ propositional, and first-order logic
 - ▶ monadic second-order logic (MSO)
- ▶ Courcelle's Theorem: obtaining FPT-results by
 - ▶ model-checking of MSO-properties
on graphs and structures of bounded tree-/clique-width

References I



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