Lecture 2: Graph width notions, dynamical programming

An Introduction to Parameterized Complexity

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Course overview

Monday, June 10 10.30 – 12.30	Tuesday, June 11	Wednesday, June 12 10.30 – 12.30	Thursday, June 13	Friday, June 14
Introduction & basic FPT results		Algorithmic Meta-Theorems		
motivation for FPT kernelization, Crown Lemma, Sunflower Lemma	GDA	1st-order logic, monadic 2nd-order logic, FPT-results by Courcelle's Theorems for tree and clique-width	GDA	GDA
Algorithmic Techniques		Formal-Method & Algorithmic Techniques		
	14.30 – 16.30 Notions of bounded graph width			14.30 – 16.30 FPT-Intractability Classes & Hierarchies
	path-, tree-, clique width, FPT-results by dynamic programming, transferring FPT results betw. widths	GDA	GDA	motivation for FP-intractability results, FPT-reductions, class XP (slicewise polynomial), W- and A-Hierarchies, placing problems on these hierarchies

Overview

- 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
- comparing width-notions

Fixed-Parameter tractable

A *parameterized problem* is a triple (Q, Σ, κ) (short: (Q, κ)) where:

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

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

Instance: $x \in \Sigma^*$. Parameter: $\kappa(x)$. Problem: Is $x \in Q$?

Fixed-Parameter tractable

A *parameterized problem* is a triple (Q, Σ, κ) (short: (Q, κ)) where:

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

Definition

A parameterized problem (Q, Σ, κ) is *fixed-parameter tractable* (is in FPT) if:

 $\exists f: \mathbb{N} \to \mathbb{N}$ computable $\exists p \in \mathbb{N}[X]$ polynomial $\exists \mathbb{A}$ algorithm, takes inputs in Σ^* $\forall x \in \Sigma^* \big[\mathbb{A} \text{ decides whether } x \in Q \text{ holds in time } \leq f(\kappa(x)) \cdot p(|x|) \big]$

†) 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 $\kappa_1, \kappa_2 : \Sigma^* \to \mathbb{N}$ parameterizations.

- ▶ $\kappa_1 \succeq \kappa_2 : \iff \exists g : \mathbb{N} \to \mathbb{N} \text{ computable } \forall x \in \Sigma^* \Big[g(\kappa_1(x)) \ge \kappa_2(x) \Big].$

Proposition

For all parameterized problems (Q, κ_1) and (Q, κ_2) with parameterizations $\kappa_1, \kappa_2 : \Sigma^* \to \mathbb{N}$ with $\kappa_1 \succeq \kappa_2$:

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

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

Computably boundedness between notions of width

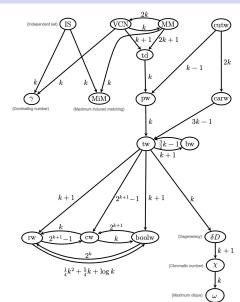
(from Sasák, [5])

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

- ► FPT-results

 transfer upwards

 (and conversely to →)
- (∉ FPT)-results transfer downwards (and along ^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'

(Ministero dell'Interno)

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: Is it possible that ℓ or more people can go out in the next hour so that everybody walks alone (with their children)?

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 \land v \in S))

\iff \forall e = \{u, v\} \in E \ (u \notin S \lor v \notin S))
```

WEIGHTED-INDEPENDENT-SET

Instance: A graph $\mathcal{G} = \langle V, E \rangle$, and a weight function $\boldsymbol{w} : V \to \mathbb{R}_0^+$. **Problem:** What is the max. weight of an independent set of \mathcal{G} ?

$$S$$
 is a vertex cover of $\mathcal{G}:\iff \forall e=\{u,v\}\in E\ (u\in S\lor v\in S))$ $\iff \forall e=\{u,v\}\in E\ (u\notin V\smallsetminus S\lor v\notin V\smallsetminus S)\)$ $\iff V\smallsetminus S$ is an independent set of \mathcal{G}

VERTEX-COVER

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 ?

 $S \subseteq V$ 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 $\boldsymbol{w} : V \to \mathbb{R}_0^+$.

Parameter: path-width / tree-width k.

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

p*-VERTEX-COVER

Instance: A graph $\mathcal{G} = \langle V, E \rangle$, and $\ell \in \mathbb{N}$. **Parameter:** path-width / tree-width k.

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

Dynamical programming on trees (example)

WEIGHTED-INDEPENDENT-SET

Instance: A tree $\mathcal{T} = \langle T, F \rangle$, and a weight function $w : T \to \mathbb{R}_0^+$. **Problem:** What is the max. weight of an independent set of \mathcal{T} ?

Obtain a directed tree $\mathcal{T} = \langle T, F, r \rangle$ (pick a root r, orient edges away).

- ▶ $A[v] := \max$. weight of an independent set in subtree \mathcal{T}_v at v,
- ▶ B[v] := max. weight of an ind. set in \mathcal{T}_v that does not contain v.

Computation of A[v] and B[v]:

- ▶ in leafs: B[v] = 0, A[v] = w(v).
- for inner vertices v with children v_1, \ldots, v_q :

$$B[v] = \sum_{i=1}^{q} A[v_i], \qquad A[v] = \max\{B[v], \boldsymbol{w}(v) + \sum_{i=1}^{q} B[v_i]\}.$$

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

Dynamical programming on trees (example)

WEIGHTED-INDEPENDENT-SET

Instance: A tree $\mathcal{T} = \langle T, F \rangle$, and a weight function $w : T \to \mathbb{R}_0^+$. **Problem:** What is the max. weight of an independent set of \mathcal{T} ?

Theorem

On trees with n nodes,

WEIGHTED-INDEPENDENT-SET \in DTIME(O(n)).

VERTEX-COVER

Instance: A tree $\mathcal{T} = \langle T, F \rangle$, and $\ell \in \mathbb{N}$.

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

Corollary

On trees with n nodes.

VERTEX-COVER \in 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).
- (P2) $(\forall \{u,v\} \in E) (\exists i \in \{1,2,\ldots,r\}) [\{u,v\} \subseteq B_i]$ (every edge of $\mathcal G$ is realized in some bag).
- (P3) $(\forall v \in V) (\exists i, k \in \{1, ..., r\}, i \le k) [\{j \mid v \in B_j\} = [i, k]]$ (the list of bags that contains a vertex of \mathcal{G} is $\langle B_i, ..., B_k \rangle$ for some interval [i, k])

The *width* of path decomp. $\langle B_1, B_2, \dots, B_r \rangle$ is $\max \{|B_t| - 1 \mid 1 \le t \le r\}$.

The path-width $pw(\mathcal{G})$ of a graph $\mathcal{G} = \langle V, E \rangle$ is defined by: $pw(\mathcal{G}) := \text{minimal width of a path decomposition of } \mathcal{G}.$

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 $\mathbf{i} \in \{1, \dots, r-1\}$ it holds:

- $lack \langle \bigcup_{j=1}^i B_j, \bigcup_{j=i+1}^r B_j \rangle$ is a separation of $\mathcal G$ with separator $B_i \cap B_{i+1}$.
- ▶ A pair (A, B) of subsets $A, B \subseteq V$ is a *separation* of \mathcal{G} if:
 - $V = A \cup B$
 - there is no edge between $A \setminus B$ and $B \setminus A$.

 $A \cap B$ is called the *separator* of a separation $\langle A, B \rangle$, and $|A \cap B|$ is called its *order*.

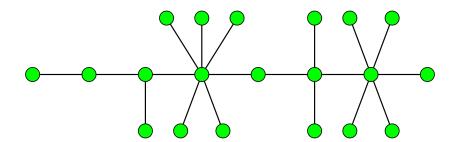
- The border (set of border vertices) ∂(A) of a set A ⊆ V of vertices consists of all vertices that have a neighbor in V \ A. Note that:
 - $\rightarrow \partial(A) = \partial(V \setminus A).$
 - ▶ $(A, (V \setminus A) \cup \partial(A))$ is a separation of \mathcal{G} , for all $A \subseteq V$.

Path-decomposition (example)



Caterpillar

Path-width?

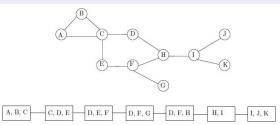


Nice path decomposition

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:

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



Nice path decomposition:

Ø → A → A,B → A,B,C → B,C → C,D → C,D,E → D,E → D,E,F → D,F → D,F → D,F,G → D,F → D,F,H → F,H → H → H,I → I → I,J,K → J,K → K → Ø

Nice path decomposition

Definition

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

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

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

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Let \mathcal{G} = \langle V, E \rangle a graph. S \subseteq V is independent set in \mathcal{G} : \iff \forall e = \{u, v\} (\neg (u \in S \land v \in S)).
```

WEIGHTED-INDEPENDENT-SET

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

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 $\langle B_1, \ldots, B_r \rangle$ be a nice path decomposition of $\mathcal{G} = \langle V, E \rangle$. Then for every $i \in \{1, \ldots, r\}$, and every $S \subseteq B_i$, we define:

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

Recursive equations for computing c[i, S] for independent S:

- ▶ Case i = 1: $c[1, \emptyset] = 0$
- ▶ Case i + 1:
 - $\begin{array}{ccc} \bullet & i+1 \text{ introduces } v\colon & B_{i+1}=B_i \cup \{v\} \text{ and } v\notin B_i, \\ \\ c[i+1,S] = \begin{cases} c[i,S] & \text{if } v\notin S, \\ c[i,S\smallsetminus \{v\}] + \boldsymbol{w}(v) & \text{if } v\in S; \end{cases}$
 - ▶ i + 1 forgets v: $B_{i+1} = B_i \setminus \{v\}$ and $v \in B_i$, $c[i + 1, S] = \max\{c[i, S], c[i, S \cup \{v\}]\}$.

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

Let $\langle B_1, \ldots, B_r \rangle$ be a nice path dec. of $\mathcal{G} = \langle V, E \rangle$ of width k. For every $i \in \{1, \ldots, r\}$, and every independent $S \subseteq B_i$, we define:

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

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:

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

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

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

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

Dynamical programming with path width (example)

Theorem

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

S is a *minimal* vertex cover

 \iff $V \setminus S$ is a *maximal* independent set.

Corollary

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

Tree decomposition (example)



A tree-decomposition of width 2

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:

- (T1) $V = \bigcup_{t \in T} B_t$ (every vertex of \mathcal{G} is in some bag).
- (T2) $(\forall \{u,v\} \in E) (\exists t \in T) [\{u,v\} \subseteq B_t]$ (the vertices of every edge of \mathcal{G} are realized in some bag).
- (T3) $(\forall v \in V) [\{t \in T \mid v \in B_t\}]$ is connected subgraph of $\mathcal{T}]$ (the tree vertices whose bags contain some vertex of \mathcal{G} form a connected subgraph of \mathcal{T}).

The *width* of a tree decomposition $\langle \mathcal{T}, \{B_t\}_{t \in T} \rangle$ is $\max \{|B_t| - 1 \mid t \in T\}.$

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

Tree decomposition (example)



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:

- ▶ $\langle A, B \rangle$ is a separation of \mathcal{G} with separator $B_a \cap B_b$.
- $ightharpoonup \partial(A), \partial(B) \subseteq B_a \cap B_b.$

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:
 - $V = A \cup B$
 - there is no edge between $A \setminus B$ and $B \setminus A$.

 $A \cap B$ is called the *separator* of a separation $\langle A, B \rangle$, and $|A \cap B|$ is called its *order*.

▶ 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 $tw(\mathcal{G}) = k$.

Theorem

TREE-WIDTH is NP-complete.

p-TREE-WIDTH

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

Parameter: k.

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

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'} \ {w} for some w ∈ B_{t'}; we say w is forgotten at t.
 - ▶ join node: a node t with two children t_1, t_2 such that $B_t = B_{t_1} = B_{t_2}$.

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)



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\} (\neg (u \in S \land v \in S)).
```

WEIGHTED-INDEPENDENT-SET

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

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:

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

Recursive equations for computing c[t,S] for independent S:

- ▶ leaf node t: $c[t, \emptyset] = 0$
- ightharpoonup introduction node t of vertex v with child t':

$$c[t,S] = \begin{cases} c[t',S] & \text{if } v \notin S \\ c[t',S \setminus \{v\}] + \boldsymbol{w}(v) & \text{otherwise} \end{cases}$$

forget node t of vertex v with child t':

$$c[t, S] = \max\{c[t', S], c[t', S \cup \{v\}]\}$$

ightharpoonup join node t with children t_1 and t_2 :

$$c[t,S] = c[t_1,S] + c[t_2,S] - w(S)$$

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

Let $\langle \mathcal{T} = \langle T, F, r \rangle, \{B_t\}_{t \in T} \rangle$ be a nice tree decomposition of $\mathcal{G} = \langle V, E \rangle$ of width k. For every $t \in T$, and every independent $S \subseteq B_t$:

$$c[t,S] \coloneqq \begin{cases} \text{maximum possible weight of a set } \hat{S} \text{ such that} \\ \hat{S} \text{ is independent } \wedge S \subseteq \hat{S} \subseteq V_t = \bigcup_{s \in T_t} B_s \ \wedge \ \hat{S} \cap B_i = S \end{cases}$$

Time Complexity: Based on the values of c[t, 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 $t \in T$:

- ▶ $|B_t| \le k + 1$,
- ▶ ⇒ number of values c[t, S] at index $t: 2^{|B_t|} = 2^{k+1}$,
- ▶ ⇒ adjacency/independence check for $S \subseteq B_t$ possible in: $O(k^2)$ using a datastructure computable in time $O(k^{O(1)} \cdot n)$,
- ▶ time for comp. a value at t, using map of values at t-1: O(k)
- ▶ time for comp. all values at t, using values at t-1: $2^{k+1} \cdot O(k^2)$
- ⇒ 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 DTIME(2^k \cdot k^{O(1)} \cdot n).
```

S is a *minimal* vertex cover

 \iff $V \setminus S$ is a *maximal* independent set.

Corollary

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

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.

- Formulate a family of properties that can be restricted to subtrees of T such that
 - a solution of P can be obtained from the properties at the root of \mathcal{T} .
- **Find** recursion equations for bottom-up evaluation on \mathcal{T} .
- Prove correctness of these recursion equations by showing two inequalities for each type of node:
 - one relating an optimum solution for the node to some solutions for its children.
 - one relating optimum solutions for a node's children to a solution for the node.
- ▶ 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 $G = \langle V, E \rangle$ with |V| = n and tw(G) = k,

- ▶ p^* -Vertex-Cover, Independent-Set \in DTIME $(2^k \cdot k^{O(1)} \cdot n)$,
- ▶ p^* -DOMINATING-SET \in DTIME $(4^k \cdot k^{O(1)} \cdot n)$,
- ▶ p^* -ODD CYCLE TRAVERSAL \in DTIME $(3^k \cdot k^{O(1)} \cdot n)$,
- ▶ p^* -MAXCUT \in DTIME $(2^k \cdot k^{O(1)} \cdot n)$,
- ▶ p^* -q-Colorability \in DTIME $(q^k \cdot k^{O(1)} \cdot n)$.

Dynamical programming: similar results (II)

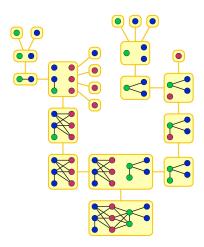
Theorem

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

- ▶ p*-STEINER-TREE,
- ▶ p*-FEEDBACK-VERTEX-SET,
- p^* -Hamiltonian-Path and p^* -Longest-Path,
- $\triangleright p^*$ -Hamiltonian-Cycle and p^* -Longest-Cycle,
- ▶ p*-CHROMATIC-NUMBER,
- ▶ p*-CYCLE-PACKING.
- ▶ p*-Connected-Vertex-Cover,
- ▶ p*-Connected-Feedback-Vertex-Set.

ov fpt comp param's (ex trees) path-w (ex) tree-w (ex) (list) clique-w [f-width] branch-w rank-w carving-w CMI rel's summ Wed refs

Clique width (example)



Clique-Width

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

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

for $i, j \in [k]$ with $i \neq j$. k-expressions φ *generate* graphs $\mathcal{G}(\varphi)$:

- $\triangleright \mathcal{G}(i)$ is the graph with a single vertex of color *i*.
- $\triangleright \mathcal{G}(\mathsf{edge}_{i-j}(\varphi))$ results from $\mathcal{G}(\varphi)$ by adding edges between every vertex of color i and every vertex of color j.
- $\triangleright \mathcal{G}(\mathsf{recolor}_{i \to j}(\varphi))$ results from $\mathcal{G}(\varphi)$ by recoloring every vertex of color i by color j.
- $\triangleright \mathcal{G}(\varphi_1 \oplus \varphi_2)$ is the disjoint union of $\mathcal{G}(\varphi_1)$ and $\mathcal{G}(\varphi_2)$.

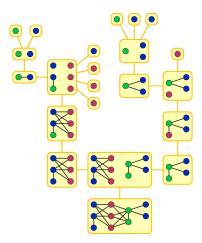
Definition (Courcelle, Engelfriet, Rozenberg, 1993, [2])

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

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

Clique width (example)

Building a graph G of clique-width clw(G) = 3:



Clique-Width (examples, properties, computability)

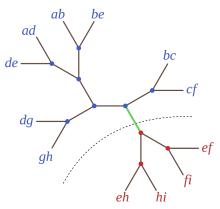
Example

- ▶ The class of cliques has clique-width 2.
- The class of stars has clique-width 2.
- The class of trees has clique-width 3.
- ▶ The class of $n \times n$ grids has clique-width $\Theta(n)$.
- subgraphs/induced subgraphs:
 - clique-width is preserved under taking induced subgraphs,
 - clique-width is not preserved under taking subgraphs (e.g. minors).
- ► c/w < tw:</p>
 - $clw \leq tw$: $clw(\mathcal{G}) \leq 3 \cdot 2^{tw(\mathcal{G})-1}$
 - ▶ \neg ($tw \le c/w$): for example, $c/w(K_n) = 2$, and $tw(K_n) = n 1$.
- ▶ Deciding whether $clw(\mathcal{G}) \le k$ 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 $f: 2^U \to \mathbb{R}^+_0$ such that:

$$f$$
 is $symmetric : \iff \forall X \subseteq U [f(X) = f(U \setminus X)];$
 f is $fair : \iff f(\emptyset) = f(U) = 0.$



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:

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

The *branch-width* $bw(\mathcal{G})$ of a graph $\mathcal{G} = \langle G, E \rangle$ is defined as

$$bw(\mathcal{G}) := w_f(E) \text{ for } f: 2^E \to \mathbb{R}_0^+, X \mapsto |\partial(X)|$$



Rank-Width

Definition

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

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

$$B_{\mathcal{G}}(X) \coloneqq (b_{x,y})_{x \in X, y \in V \setminus X}$$
, where, for all $x \in X, y \in V \setminus X$:
$$b_{x,y} = 1 \Longleftrightarrow \{x,y\} \in E.$$

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

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

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

Properties

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

Carving-Width and Cut-Width

Definition

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

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

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

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

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

Definition

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

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

$$\textit{width}(\pi) \coloneqq \max_{1 \le i \le n} \textit{Cut}_{\mathcal{G}}(\{\pi(j) \mid 1 \le j \le i\}).$$

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

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

Coverage in Multi-Interface Networks



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

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

Solution: An allocation $W_A: V \to 2^{\{1,\dots,a\}}$ of active interfaces covering $\mathcal G$ such that $W_A(v) \subseteq W(v)$, and $|W_A(v)| \le p$ for all $v \in V$, if possible; otherwise, a negative answer.

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 (parameterized)

Theorem

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

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p^*-CMI(p) (for p \in \mathbb{N})
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Instance: A graph $\mathcal{G} = \langle V, E \rangle$, $W: V \to 2^{\{1,\dots,a\}}$ available-interface allocation, $c: \{1,\dots,a\} \to \mathbb{R}^+$ interface cost function.

Parameter: path-width / carving-width k

Problem: Obtain, if possible, a minimal solution with respect to the total cost of the interfaces that are activated, that is,

 $c(W_A) = \sum_{v \in V} \sum_{i \in W_A(v)} c(i).$

Theorem (Aloisio, Navarra, 2020, [1])

- For path-width $pw(\mathcal{G}) = k$, p^* -CMI(2) \in DTIME($n \cdot (a + {a \choose 2})^{k+1}$).
- ► For carving-width carw(\mathcal{G}) = k, p^* -CMI(2) \in DTIME($n \cdot a^{4k}$).

Coverage in Multi-Interface Networks (parameterized)

Theorem (Aloisio, Navarra, 2020, [1])

- For path-width $pw(\mathcal{G}) = k$, p^* -CMI(2) \in DTIME $(n \cdot (a + {a \choose 2})^{k+1})$.
- ► For carving-width carw(\mathcal{G}) = k, p^* -CMI(2) \in DTIME($n \cdot a^{4k}$).

```
(p^*)'-CMI(p) (for p \in \mathbb{N})
```

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

Parameter: *a* + (path-width / carving-width *k*)

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

Corollary

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

Comparing parameterizations

Definition (computably bounded)

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

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

Proposition

For all parameterized problems (Q, κ_1) and (Q, κ_2) with parameterizations $\kappa_1, \kappa_2 : \Sigma^* \to \mathbb{N}$ with $\kappa_1 \succeq \kappa_2$:

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

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

Computably boundedness between notions of width

(from Sasák, [5])

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

- ► FPT-results

 transfer upwards

 (and conversely to →)
- (∉ FPT)-results
 transfer downwards
 (and along ^g/_→)



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

Tomorrow

Monday, June 10 10.30 – 12.30	Tuesday, June 11	Wednesday, June 12 10.30 – 12.30	Thursday, June 13	Friday, June 14
Introduction & basic FPT results		Algorithmic Meta-Theorems		
motivation for FPT kernelization, Crown Lemma, Sunflower Lemma	GDA	1st-order logic, monadic 2nd-order logic, FPT-results by Courcelle's Theorems for tree and clique-width	GDA	GDA
Algorithmic Techniques		Formal-Method & Algorithmic Techniques		
	14.30 – 16.30 Notions of bounded graph width			14.30 – 16.30 FPT-Intractability Classes & Hierarchies
	path-, tree-, clique width, FPT-results by dynamic programming, transferring FPT results betw. widths	GDA	GDA	motivation for FP-intractability results, FPT-reductions, class XP (slicewise polynomial), W- and A-Hierarchies, placing problems on these hierarchies

Tomorrow

- 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

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