

9. Parameter Treewidth

COMP6741: Parameterized and Exact Computation

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Outline

- 1 Algorithms for trees
- 2 Tree decompositions
- 3 Monadic Second Order Logic
- 4 Dynamic Programming over Tree Decompositions
 - Sat
 - CSP
- 5 Further Reading

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- 1 Algorithms for trees
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Recall: An **independent set** of a graph $G = (V, E)$ is a set of vertices $S \subseteq V$ such that $G[S]$ has no edge.

#INDEPENDENT SETS ON TREES

Input: A tree $T = (V, E)$

Output: The number of independent sets of T .

- Design a polynomial time algorithm for #INDEPENDENT SETS ON TREES

- Select an arbitrary root r of T
- Bottom-up dynamic programming (starting at the leaves) to compute, for each subtree T_x rooted at x the values
 - $\#in(x)$: the number of independent sets of T_x containing x , and
 - $\#out(x)$: the number of independent sets of T_x not containing x .
- If x is a leaf, then $\#in(x) = \#out(x) = 1$
- Otherwise,

$$\begin{aligned}\#in(x) &= \prod_{y \in \text{children}(x)} \#out(y) \text{ and} \\ \#out(x) &= \prod_{y \in \text{children}(x)} (\#in(y) + \#out(y))\end{aligned}$$

- The final result is $\#in(r) + \#out(r)$

Recall: A **dominating set** of a graph $G = (V, E)$ is a set of vertices $S \subseteq V$ such that $N_G[S] = V$.

#DOMINATING SETS ON TREES

Input: A tree $T = (V, E)$

Output: The number of dominating sets of T .

- Design a polynomial time algorithm for #DOMINATING SETS ON TREES

Solution

- Select an arbitrary root r of T
- Bottom-up dynamic programming (starting at the leaves) to compute, for each subtree T_x rooted at x the values
 - $\#in(x)$: the number of dominating sets of T_x containing x ,
 - $\#outDom(x)$: the number of dominating sets of T_x not containing x , and
 - $\#outNd(x)$: the number of vertex subsets of T_x dominating $V(T_x) \setminus \{x\}$.
- If x is a leaf, then $\#in(x) = \#outNd(x) = 1$ and $\#outDom(x) = 0$.
- Otherwise,

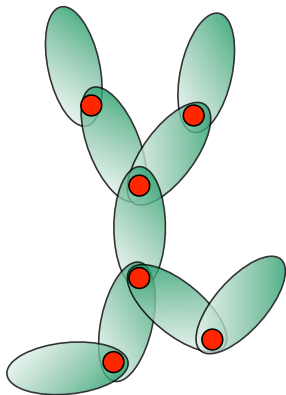
$$\begin{aligned}\#in(x) &= \prod_{y \in \text{children}(x)} (\#in(y) + \#outDom(y) + \#outNd(y)), \\ \#outDom(x) &= \prod_{y \in \text{children}(x)} (\#in(y) + \#outDom(y)) \\ &\quad - \prod_{y \in \text{children}(x)} \#outDom(y) \\ \#outNd(x) &= \prod_{y \in \text{children}(x)} \#outDom(y)\end{aligned}$$

- The final result is $\#in(r) + \#outDom(r)$

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Algorithms using graph decompositions

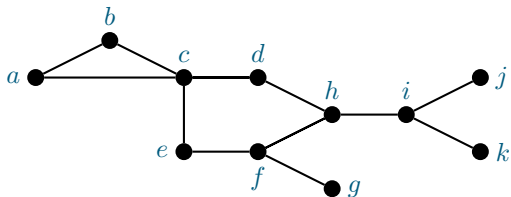


Idea: decompose the problem into subproblems and combine solutions to subproblems to a global solution.

Parameter: overlap between subproblems.

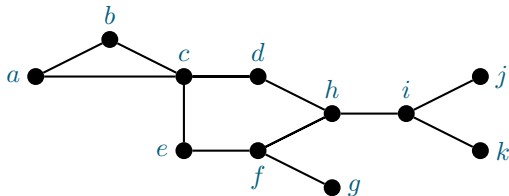
Tree decompositions (by example)

- A graph G

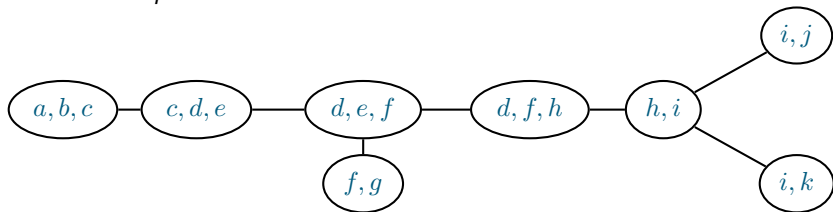


Tree decompositions (by example)

- A graph G

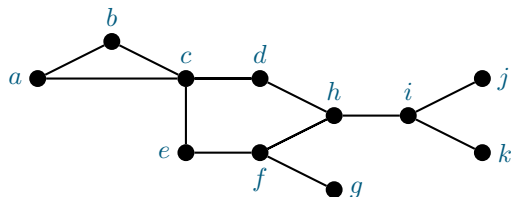


- A tree decomposition of G

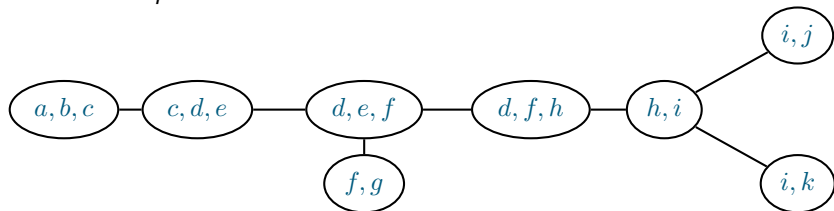


Tree decompositions (by example)

- A graph G



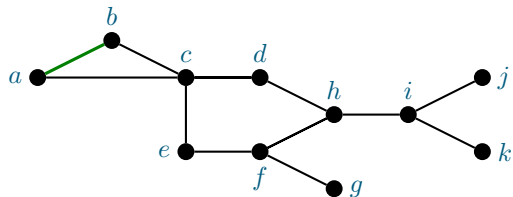
- A tree decomposition of G



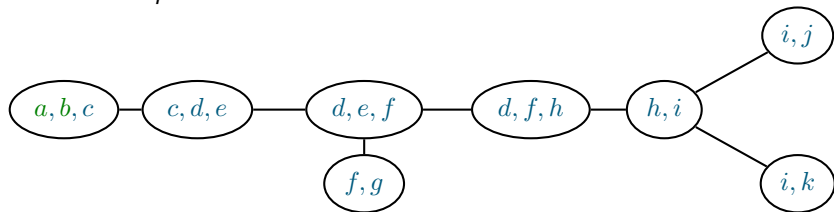
Conditions:

Tree decompositions (by example)

- A graph G



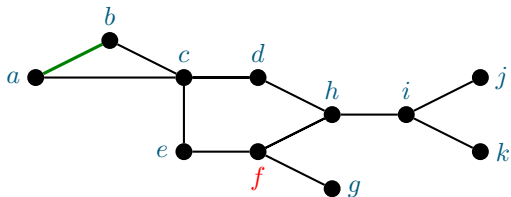
- A tree decomposition of G



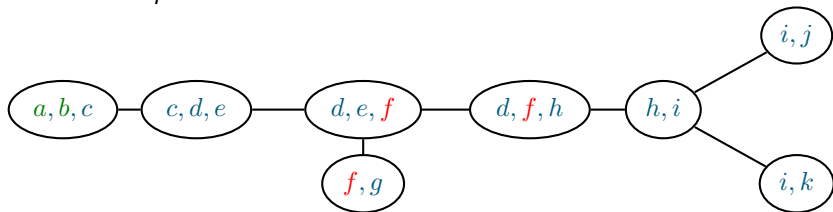
Conditions: covering

Tree decompositions (by example)

- A graph G



- A tree decomposition of G



Conditions: **covering** and **connectedness**.

Tree decomposition (more formally)

- Let G be a graph, T a tree, and γ a labeling of the vertices of T by sets of vertices of G .
- We refer to the vertices of T as “nodes”, and we call the sets $\gamma(t)$ “bags”.
- The pair (T, γ) is a *tree decomposition* of G if the following three conditions hold:
 - 1 For every vertex v of G there exists a node t of T such that $v \in \gamma(t)$.
 - 2 For every edge vw of G there exists a node t of T such that $v, w \in \gamma(t)$ (“covering”).
 - 3 For any three nodes t_1, t_2, t_3 of T , if t_2 lies on the unique path from t_1 to t_3 , then $\gamma(t_1) \cap \gamma(t_3) \subseteq \gamma(t_2)$ (“connectedness”).

- The *width* of a tree decomposition (T, γ) is defined as the maximum $|\gamma(t)| - 1$ taken over all nodes t of T .
- The *treewidth* $\text{tw}(G)$ of a graph G is the minimum width taken over all its tree decompositions.

- Trees have treewidth 1.
- Cycles have treewidth 2.
- Consider a tree decomposition (T, γ) of a graph G and two adjacent nodes i, j in T . Let T_i and T_j denote the two trees obtained from T by deleting the edge ij , such that T_i contains i and T_j contains j . Then, every vertex contained in both $\bigcup_{a \in V(T_i)} \gamma(a)$ and $\bigcup_{b \in V(T_j)} \gamma(b)$ is also contained in $\gamma(i) \cap \gamma(j)$.
- The complete graph on n vertices has treewidth $n - 1$.
- If a graph G contains a clique K_r , then every tree decomposition of G contains a node t such that $K_r \subseteq \gamma(t)$.

Complexity of Treewidth

TREewidth

Input: Graph $G = (V, E)$, integer k

Parameter: k

Question: Does G have treewidth at most k ?

- TREewidth is NP-complete.
- TREewidth is FPT, due to a $k^{O(k^3)} \cdot |V|$ time algorithm by [Bodlaender '96]

- Many graph problems that are polynomial time solvable on trees are **FPT** with parameter treewidth.
- Two general methods:
 - *Dynamic programming*: compute local information in a bottom-up fashion along a tree decomposition
 - *Monadic Second Order Logic*: express graph problem in some logic formalism and use a meta-algorithm

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Monadic Second Order Logic

- *Monadic Second Order* (MSO) Logic is a powerful formalism for expressing graph properties. One can quantify over vertices, edges, vertex sets, and edge sets.
- *Courcelle's theorem*: Checking whether a graph G satisfies an MSO property is **FPT** parameterized by the treewidth of G plus the length of the MSO expression. [Courcelle, '90]
- *Arnborg et al.'s generalization*: Several generalizations. For example, **FPT** algorithm for parameter $\text{tw}(G) + |\phi(X)|$ that takes as input a graph G and an MSO sentence $\phi(X)$ where X is a free (non-quantified) vertex set variable, that computes a minimum-sized set of vertices X such that $\phi(X)$ is true in G . Also, the input vertices and edges may be colored and their color can be tested. [Arnborg, Lagergren, Seese, '91]

An MSO formula has

- variables representing vertices (u, v, \dots) , edges (a, b, \dots) , vertex subsets (X, Y, \dots) , or edge subsets (A, B, \dots) in the graph
- atomic operations
 - $u \in X$: testing set membership
 - $X = Y$: testing equality of objects
 - $inc(u, a)$: incidence test “is vertex u an endpoint of the edge a ?”
- propositional logic on subformulas: $\phi_1 \wedge \phi_2$, $\phi_1 \vee \phi_2$, $\neg\phi_1$, $\phi_1 \Rightarrow \phi_2$
- Quantifiers: $\forall X \subseteq V$, $\exists A \subseteq E$, $\forall u \in V$, $\exists a \in E$, etc.

We can define some shortcuts

- $u \neq v$ is $\neg(u = v)$
- $X \subseteq Y$ is $\forall v \in V (v \in X) \Rightarrow (v \in Y)$
- $\forall v \in X \varphi$ is $\forall v \in V (v \in X) \Rightarrow \varphi$
- $\exists v \in X \varphi$ is $\exists v \in V (v \in X) \wedge \varphi$
- $adj(u, v)$ is $(u \neq v) \wedge \exists a \in E (inc(u, a) \wedge inc(v, a))$

MSO Logic Example

Example: 3-Coloring,

- “there are three independent sets in $G = (V, E)$ which form a partition of V ”
- $3COL := \exists R \subseteq V \exists G \subseteq V \exists B \subseteq V$

$partition(R, G, B) \wedge independent(R) \wedge independent(G) \wedge independent(B)$

where

$partition(R, G, B) := \forall v \in V ((v \in R \wedge v \notin G \wedge v \notin B) \vee (v \notin R \wedge v \in G \wedge v \notin B) \vee (v \notin R \wedge v \notin G \wedge v \in B))$

and

$independent(X) := \neg(\exists u \in X \exists v \in X adj(u, v))$

By Courcelle's theorem and our *3COL* MSO formula, we have:

Theorem 1

3-COLORING is FPT with parameter treewidth.

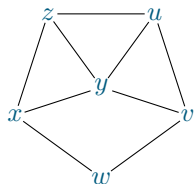
Treewidth only for graph problems?

Let us use treewidth to solve a Logic Problem

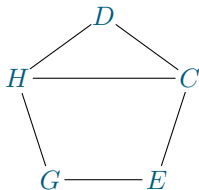
- associate a graph with the instance
- take the tree decomposition of the graph
- most widely used: primal graphs, incidence graphs, and dual graphs of formulas.

Three Treewidth Parameters

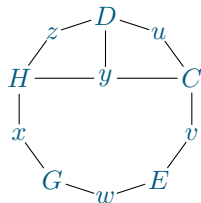
CNF Formula $F = C \wedge D \wedge E \wedge G \wedge H$ where $C = (u \vee v \vee \neg y)$,
 $D = (\neg u \vee z \vee y)$, $E = (\neg v \vee w)$, $G = (\neg w \vee x)$, $H = (x \vee y \vee \neg z)$.



primal graph



dual graph



incidence graph

This gives rise to parameters **primal treewidth**, **dual treewidth**, and **incidence treewidth**.

Definition 2

Let F be a CNF formula with variables $\text{var}(F)$ and clauses $\text{cla}(F)$.

The **primal graph** of F is the graph with vertex set $\text{var}(F)$ where two variables are adjacent if they appear together in a clause of F .

The **dual graph** of F is the graph with vertex set $\text{cla}(F)$ where two clauses are adjacent if they have a variable in common.

The **incidence graph** of F is the bipartite graph with vertex set $\text{var}(F) \cup \text{cla}(F)$ where a variable and a clause are adjacent if the variable appears in the clause.

The **primal treewidth**, **dual treewidth**, and **incidence treewidth** of F is the treewidth of the primal graph, the dual graph, and the incidence graph of F , respectively.

Incidence treewidth is most general

Lemma 3

The incidence treewidth of F is at most the primal treewidth of F plus 1.

Proof.

Start from a tree decomposition (T, γ) of the primal graph with minimum width. For each clause C :

- There is a node t of T with $\text{var}(C) \subseteq \gamma(t)$, since $\text{var}(C)$ is a clique in the primal graph.
- Add to t a new neighbor t' with $\gamma(t') = \gamma(t) \cup \{C\}$.



Lemma 4

The incidence treewidth of F is at most the dual treewidth of F plus 1.

Lemma 4

The incidence treewidth of F is at most the dual treewidth of F plus 1.

Primal and dual treewidth are incomparable.

- One big clause alone gives large primal treewidth.
- $\{\{x, y_1\}, \{x, y_2\}, \dots, \{x, y_n\}\}$ gives large dual treewidth.

SAT parameterized by treewidth

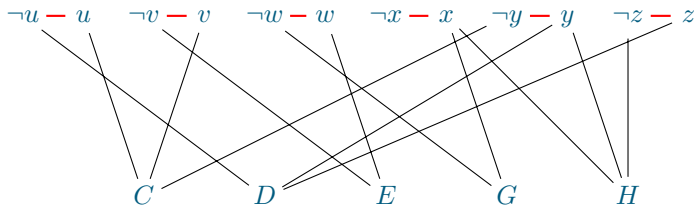
SAT
Input: A CNF formula F
Question: Is there an assignment of truth values to $\text{var}(F)$ such that F evaluates to true?

Note: If SAT is FPT parameterized by incidence treewidth, then SAT is FPT parameterized by primal treewidth and by dual treewidth.

SAT is FPT for parameter incidence treewidth

CNF Formula $F = C \wedge D \wedge E \wedge G \wedge H$ where $C = (u \vee v \vee \neg y)$,
 $D = (\neg u \vee z \vee y)$, $E = (\neg v \vee w)$, $G = (\neg w \vee x)$, $H = (x \vee y \vee \neg z)$

Auxiliary graph:



- MSO Formula: *“There exists an independent set of literal vertices that dominates all the clause vertices.”*
- The treewidth of the auxiliary graph is at most twice the treewidth of the incidence graph plus one.

Theorem 5

SAT is **FPT** for each of the following parameters: *primal treewidth, dual treewidth, and incidence treewidth.*

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Courcelle's theorem: discussion

Advantages of Courcelle's theorem:

- general, applies to many problems
- easy to obtain **FPT** results

Drawback of Courcelle's theorem

- the resulting running time depends non-elementarily on the treewidth t and the length ℓ of the MSO-sentence, i.e., a tower of 2's whose height is $\omega(1)$

$$2^{2^{2^{\dots^{t+\ell}}}}$$

Dynamic programming over tree decompositions

Idea: extend the algorithmic methods that work for trees to tree decompositions.

- Step 1 Compute a minimum width tree decomposition using Bodlaender's algorithm
- Step 2 Transform it into a standard form making computations easier
- Step 3 Bottom-up Dynamic Programming (from the leaves of the tree decomposition to the root)

Nice tree decomposition

A *nice* tree decomposition (T, γ) has 4 kinds of bags:

- *leaf node*: leaf t in T and $|\gamma(t)| = 1$
- *introduce node*: node t with one child t' in T and $\gamma(t) = \gamma(t') \cup \{x\}$
- *forget node*: node t with one child t' in T and $\gamma(t) = \gamma(t') \setminus \{x\}$
- *join node*: node t with two children t_1, t_2 in T and $\gamma(t) = \gamma(t_1) = \gamma(t_2)$

Every tree decomposition of width w of a graph G on n vertices can be transformed into a nice tree decomposition of width w and $O(w \cdot n)$ nodes in polynomial time [Kloks '94].

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Dynamic programming: primal treewidth

- Compute a nice tree decomposition (T, γ) of F 's primal graph with minimum width [Bodlaender '96; Kloks '94]
- Select an arbitrary root r of T
- Denote T_t the subtree of T rooted at t
- Denote $\gamma_{\downarrow}(t) = \{x \in \gamma(t') : t' \in V(T_t)\}$
- Denote $F_{\downarrow}(t) = \{C \in F : \text{var}(C) \subseteq \gamma_{\downarrow}(t)\}$
- For a node t and an assignment $\tau : \gamma(t) \rightarrow \{0, 1\}$, define

$$\text{sat}(t, \tau) = \begin{cases} 1 & \text{if } \tau \text{ can be extended to a} \\ & \text{satisfying assignment of } F_{\downarrow}(t) \\ 0 & \text{otherwise.} \end{cases}$$

$$\text{sat}(t, \tau) = \begin{cases} 1 & \text{if } \tau \text{ can be extended to a} \\ & \text{satisfying assignment of } F_{\downarrow}(t) \\ 0 & \text{otherwise.} \end{cases}$$

Denote $x^1 = x$ and $x^0 = \neg x$.

We will view F as a set of clauses and each clause as a set of literals; e.g.

$F = \{\{x, \neg y\}, \{\neg x, y, z\}\}$ instead of $F = (x \vee \neg y) \wedge (\neg x \vee y \vee z)$

- *leaf node*:

$$\text{sat}(t, \tau) = \begin{cases} 1 & \text{if } \tau \text{ can be extended to a} \\ & \text{satisfying assignment of } F_{\downarrow}(t) \\ 0 & \text{otherwise.} \end{cases}$$

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- *leaf node*: $\text{sat}(t, \{x = a\}) = \begin{cases} 1 & \text{if } \{x^{1-a}\} \notin F \\ 0 & \text{otherwise} \end{cases}$
- *introduce node*:

$$\text{sat}(t, \tau) = \begin{cases} 1 & \text{if } \tau \text{ can be extended to a} \\ & \text{satisfying assignment of } F_{\downarrow}(t) \\ 0 & \text{otherwise.} \end{cases}$$

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We will view F as a set of clauses and each clause as a set of literals; e.g.

$F = \{\{x, \neg y\}, \{\neg x, y, z\}\}$ instead of $F = (x \vee \neg y) \wedge (\neg x \vee y \vee z)$

- *leaf node*: $\text{sat}(t, \{x = a\}) = \begin{cases} 1 & \text{if } \{x^{1-a}\} \notin F \\ 0 & \text{otherwise} \end{cases}$
- *introduce node*: $\gamma(t) = \gamma(t') \cup \{x\}$.

$$\begin{aligned} \text{sat}(t, \{x = a\} \cup \{x_i = a_i\}_i) &= \text{sat}(t', \{x_i = a_i\}_i) \\ &\wedge \nexists C \in F : C \subseteq \{x^{1-a}\} \cup \{x_i^{1-a_i}\}_i. \end{aligned}$$

DP: primal treewidth III

- *forget node:*

- *forget node*: $\gamma(t) = \gamma(t') \setminus \{x\}$.

$$\begin{aligned} \text{sat}(t, \{x_i = a_i\}_i) &= \text{sat}(t', \{x = 0\} \cup \{x_i = a_i\}_i) \\ &\quad \vee \text{sat}(t', \{x = 1\} \cup \{x_i = a_i\}_i). \end{aligned}$$

- *join node*:

- *forget node*: $\gamma(t) = \gamma(t') \setminus \{x\}$.

$$\begin{aligned} \text{sat}(t, \{x_i = a_i\}_i) &= \text{sat}(t', \{x = 0\} \cup \{x_i = a_i\}_i) \\ &\quad \vee \text{sat}(t', \{x = 1\} \cup \{x_i = a_i\}_i). \end{aligned}$$

- *join node*:

$$\begin{aligned} \text{sat}(t, \{x_i = a_i\}_i) &= \text{sat}(t_1, \{x_i = a_i\}_i) \\ &\quad \wedge \text{sat}(t_2, \{x_i = a_i\}_i). \end{aligned}$$

- *forget node*: $\gamma(t) = \gamma(t') \setminus \{x\}$.

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- *join node*:

$$\begin{aligned} \text{sat}(t, \{x_i = a_i\}_i) &= \text{sat}(t_1, \{x_i = a_i\}_i) \\ &\quad \wedge \text{sat}(t_2, \{x_i = a_i\}_i). \end{aligned}$$

- Finally: F is satisfiable iff $\exists \tau : \gamma(r) \rightarrow \{0, 1\}$ such that $\text{sat}(r, \tau) = 1$
- Running time: $O^*(2^k)$, where k is the primal treewidth of F , supposed we are given a minimum width tree decomposition
- Also extends to computing the number of satisfying assignments

Known treewidth based algorithms for SAT:

$$\begin{array}{ccc} k = \text{primal tw} & k = \text{dual tw} & k = \text{incidence tw} \\ O^*(2^k) & O^*(2^k) & O^*(4^k) \end{array}$$

- It is still worth considering primal treewidth and dual treewidth.
- These algorithms all count the number of satisfying assignments.

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Constraint Satisfaction Problem

CSP

Input: A set of variables X , a domain D , and a set of constraints C

Question: Is there an assignment $\tau : X \rightarrow D$ satisfying all the constraints in C ?

A **constraint** has a **scope** $S = (s_1, \dots, s_r)$ with $s_i \in X, i \in \{1, \dots, r\}$, and a **constraint relation** R consisting of r -tuples of values in D .

An assignment $\tau : X \rightarrow D$ **satisfies** a constraint $c = (S, R)$ if there exists a tuple (d_1, \dots, d_r) in R such that $\tau(s_i) = d_i$ for each $i \in \{1, \dots, r\}$.

- Primal, dual, and incidence graphs are defined similarly as for SAT.

Theorem 6 ([Gottlob, Scarcello, Sideri '02])

CSP is FPT for parameter primal treewidth if $|D| = O(1)$.

- What if domains are unbounded?

Theorem 7

CSP is $W[1]$ -hard for parameter primal treewidth.

Theorem 7

CSP is $W[1]$ -hard for parameter primal treewidth.

Proof Sketch.

Parameterized reduction from CLIQUE.

Let $(G = (V, E), k)$ be an instance of CLIQUE.

Take k variables x_1, \dots, x_k , each with domain V .

Add $\binom{k}{2}$ binary constraints $E_{i,j}$, $1 \leq i < j \leq k$.

A constraint $E_{i,j}$ has scope (x_i, x_j) and its constraint relation contains the tuple (u, v) if $uv \in E$.

The primal treewidth of this CSP instance is $k - 1$. □

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Further Reading

- Chapter 7, *Treewidth* in Marek Cygan, Fedor V. Fomin, Łukasz Kowalik, Daniel Lokshtanov, Dániel Marx, Marcin Pilipczuk, Michał Pilipczuk, and Saket Saurabh. *Parameterized Algorithms*. Springer, 2015.
- Chapter 5, *Treewidth* in Fedor V. Fomin and Dieter Kratsch. *Exact Exponential Algorithms*. Springer, 2010.
- Chapter 10, *Tree Decompositions of Graphs* in Rolf Niedermeier. *Invitation to Fixed Parameter Algorithms*. Oxford University Press, 2006.
- Chapter 10, *Treewidth and Dynamic Programming* in Rodney G. Downey and Michael R. Fellows. *Fundamentals of Parameterized Complexity*. Springer, 2013.
- Chapter 13, *Courcelle's Theorem* in Rodney G. Downey and Michael R. Fellows. *Fundamentals of Parameterized Complexity*. Springer, 2013.