

CSE599d: Advanced Query Processing

Lecture 13: Tree Decomposition

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Announcements

- HW3 due: Friday Feb. 27

Tree Decomposition

Tree Decomposition

$G = (V, E)$ undirected graph

Definition A **tree decomposition** is a pair (T, χ) where T = a tree and $\chi : \text{Nodes}(T) \rightarrow 2^V$ maps each node u in T to a subset of V (where $\chi(u)$ is called a **bag**), such that:

- Every edge (x, y) in G is contained in some bag $\chi(u)$.
- For every node x , the bags containing x are connected in T .

Tree Width

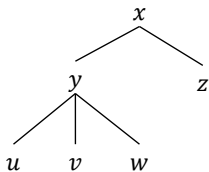
The **width of** $TD = (T, \chi)$ is the size of the largest bag, minus one:

$$\text{tw}(TD) = \max_u |\chi(u)| - 1$$

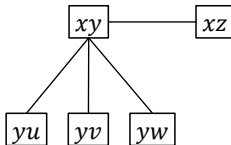
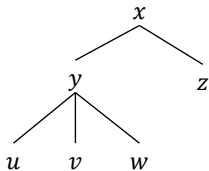
The **tree-width of a graph** is smallest width of any tree decomposition:

$$\text{tw}(G) = \min_{TD} (\text{tw}(TD))$$

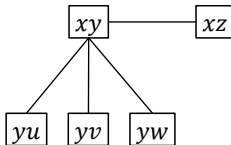
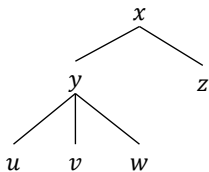
Examples



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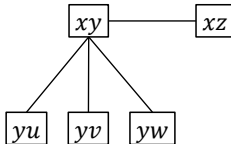
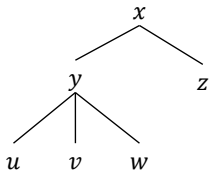


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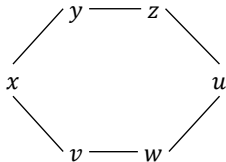


TW = 1

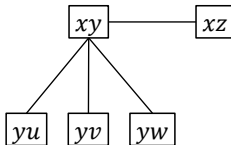
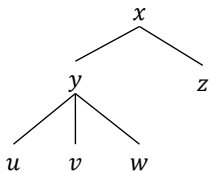
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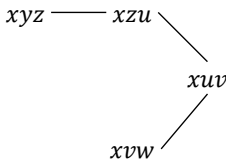
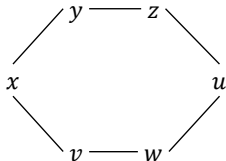
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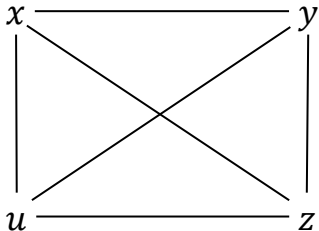


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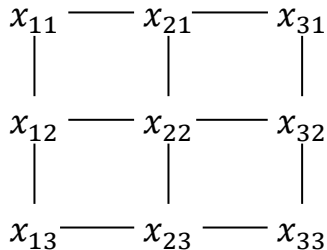
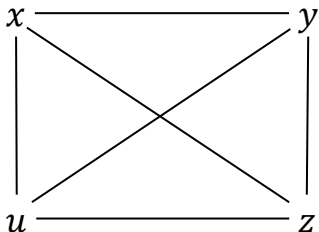


TW = 2

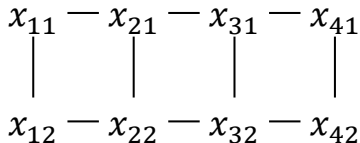
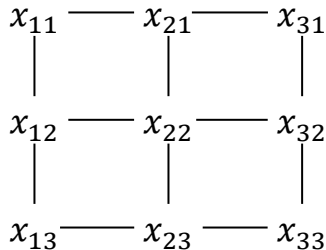
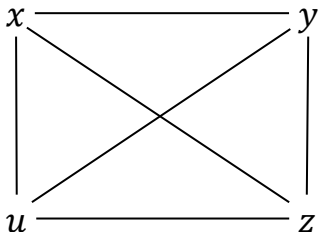
More Examples: What are the τw ?



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Discussion

- The “bags” are sets, not bags!
- $\text{tw}(G) = 1$ iff G is acyclic (i.e. a tree or a forest).
- If G is a clique, then $\text{tw}(G) = |\text{Nodes}(G)| - 1$.
- The problem: “given G, k , check if $\text{tw}(G) \leq k$ ” is NP-complete.
- For a fixed k , it is in linear time in $|G|$. Algorithm is impractical.
- Some hard problems become easy over graphs with a bounded tree-width.
E.g. 3-colorability: NP-complete, linear time for graphs with bounded tw .

Hypertree Decomposition

Overview

When Q is acyclic, then we can compute Q in linear time using YA.

When Q is cyclic, then we compute a tree decomposition of Q , then run YA.

Recall that identify Q with a hypergraph $G = (V, E)$.

what are V, E ?

As usual, we use “query” and “hypergraph” interchangeably.

Hypertree Decomposition

A **hypertree decomposition** of $G = (V, E)$ is $TD = (T, \chi)$ where:

- Every hyperedge $e \in E$ is contained in some bag $\chi(u)$.
- For every node (variable) x , the bags containing x are connected.

$\text{tw}(TD)$ and $\text{tw}(G)$ are defined before (maximum bag size minus one).

Query Evaluation Algorithm

Query $Q = R_1(\mathbf{x}_1) \wedge R_2(\mathbf{x}_2) \wedge \dots$

- Compute a “good” hypertree decomposition $TD = (T, \chi)$ of Q .
- For all u , materialize $\chi(u)$ (using any algorithm), store the result in T_u .
- Run Yannakakis algorithm on the acyclic query consisting of all T_u .

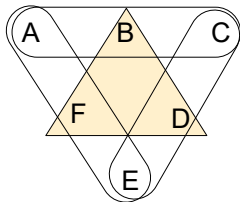
What is the complexity as a function of $N(= \text{IN})$ and OUT ?

Example

$$Q(A, B, C, D, E, F) = R_1(A, B, C) \wedge R_2(C, D, E) \wedge R_3(E, F, A) \wedge R_4(B, D, F)$$

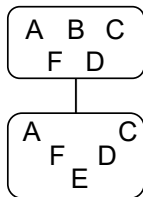
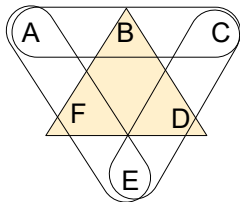
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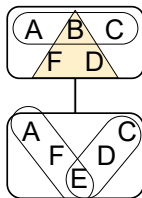
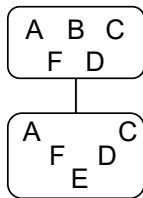
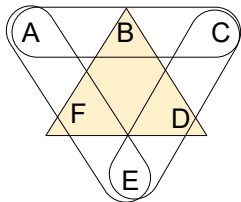
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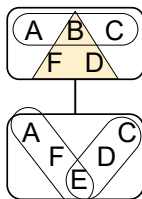
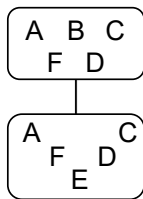
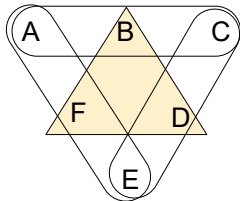
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Algorithm:

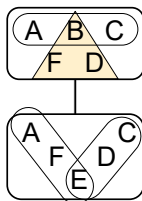
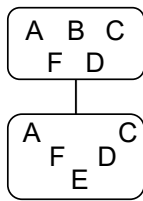
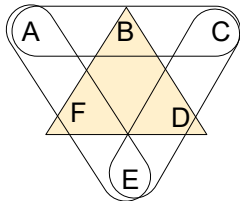
$$T_1(A, B, C, D, F) := R_1(A, B, C) \bowtie R_4(B, D, F)$$

$$T_2(A, C, D, E, F) := R_2(C, D, E) \bowtie R_3(E, F, A)$$

$$Q(A, B, C, D, E, F) := T_1 \bowtie T_2$$

Example

$$Q(A, B, C, D, E, F) = R_1(A, B, C) \wedge R_2(C, D, E) \wedge R_3(E, F, A) \wedge R_4(B, D, F)$$



Algorithm:

$$T_1(A, B, C, D, F) := R_1(A, B, C) \bowtie R_4(B, D, F)$$

$$T_2(A, C, D, E, F) := R_2(C, D, E) \bowtie R_3(E, F, A)$$

$$Q(A, B, C, D, E, F) := T_1 \bowtie T_2$$

Complexity: $O(N^2 + \text{OUT})$

Given by the hypertree width (next)

Edge Cover

$G = (V, E)$ hypergraph. An **edge cover** for G is a subset $E_0 \subseteq E$ s.t.

- Every vertex $x \in V$ is contained in some hyperedge $e \in E_0$.

Edge Cover

$G = (V, E)$ hypergraph. An **edge cover** for G is a subset $E_0 \subseteq E$ s.t.

- Every vertex $x \in V$ is contained in some hyperedge $e \in E_0$.

Subset $X \subseteq V$. An **edge cover for X** is a subset $E_0 \subseteq E$ s.t.

- every vertex $x \in X$ is contained in some hyperedge $e \in E_0$.

$\rho(X) :=$ size of the smallest edge cover for X

Hypertree Width

The (hypertree?)width of a tree decomposition $TD = (T, \chi)$ is largest edge cover of any of its bags:

$$\text{HTW}(TD) = \max_u (\rho(\chi(u)))$$

The hypertree width of a hypergraph G (or query) is:

$$\text{HTW}(G) = \min_{TD} (\text{HTW}(TD))$$

Complexity

Query $Q = R_1(\mathbf{x}_1) \wedge R_2(\mathbf{x}_2) \wedge \dots$

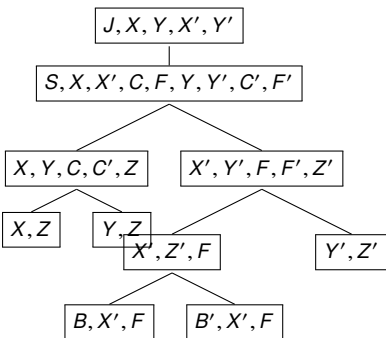
If $|R_1| = |R_2| = \dots = N$, then Q can be computed in time $O(N^{\text{HTW}(Q)} + \text{OUT})$.

Example from [Pichler, 2023]

$$Q(\dots) = a(S, X, X', C, F) \wedge b(S, Y, Y', C', F') \wedge c(C, C', Z) \wedge d(X, Z) \wedge e(Y, Z) \\ f(F, F', Z') \wedge g(X', Z') \wedge h(Y', Z') \wedge j(J, X, Y, X', Y') \wedge p(B, X', F) \wedge q(B', X', F)$$

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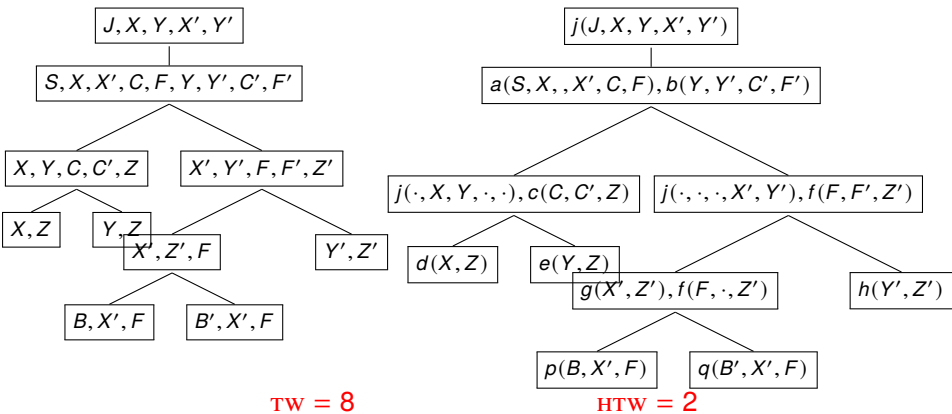
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TW = 8

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TW = 8

HTW = 2

Discussion

- Query evaluation in time $O(N^{\text{HTW}} + \text{OUT})$
- “Given Q , check if $\text{HTW}(Q) \leq 3$ ” is NP-hard [Gottlob et al., 2021]
- Terminology confusion: the term **generalized hypertree decomposition** is sometimes used in the literature; will explain reason shortly.

Variable Elimination

Overview

Many algorithms in the literature are based on “variable elimination”.

This turns out to be equivalent to a tree decomposition.

We will discuss their tight connection next, using [Khamis et al., 2015]

Notation: $\partial x =$ set of hyperedges that contain x : $\partial x := \{e \mid x \in e\}$

Variable Elimination

$G = (V, E)$ hypergraph. A **variable elimination order** is a total order on V :

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where G_i is obtained from G_{i-1} as follows:

- Remove the variable x_i .
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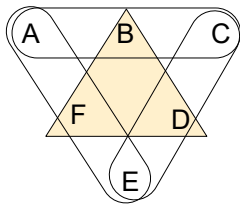
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- Remove the variable x_i .
- Remove hyperedges ∂x_i ; add hyperedge $U_i - \{x_i\}$, where $U_i := \bigcup \partial x_i$

σ defines the following **bags**: $\text{bags}(\sigma) = \{U_1, \dots, U_n\}$

Example

$$Q(A, B, C, D, E, F) = R_1(A, B, C) \wedge R_2(C, D, E) \wedge R_3(E, F, A) \wedge R_4(B, D, F)$$
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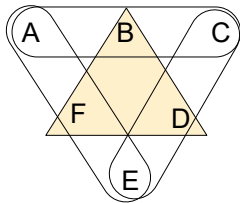


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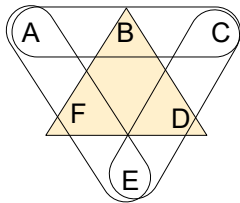
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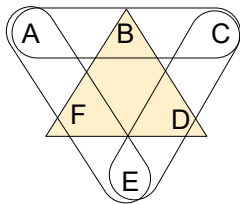
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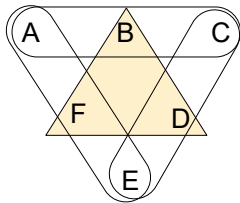
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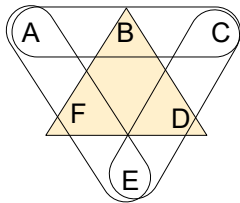
$$U_2 = BCDEF, G_2 = (\{C, D, E, F\}, \{CDE, CDEF\})$$

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$$U_3 = CDEF, G_3 = (\{D, E, F\}, \{DE, DEF\}),$$

$$U_4 = DEF, G_4 = (\{E, F\}, \{EF\})$$

$$U_5 = EF, G_5 = (\{F\}, \{F\})$$

$$U_6 = F, G_6 = (\emptyset, \emptyset)$$

From Variable Elimination to Tree Decomposition

Theorem For any variable elimination σ with bags U_1, U_2, \dots, U_n there exists a tree decomposition TD with the same bags.

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One of these bags u_1 contains $U_1 - \{x_1\}$, because this is a hyperedge in G_1 .

Extend TD_1 by adding new bag $\chi(u) := U_1$, connect it with an edge to u_1 .

From Tree Decomposition to Variable Elimination (1/2)

The correspondence is not exactly 1-1.

Recall the example $G = (\{A, B, C, D, E, F\}, \{ABC, CDE, EFG, BDF\})$:

- $\text{bags}(TD) = \{ABCE, BCDEF\}$
- $\text{bags}(\sigma) = \{ABCE, BCDEF, CDEF, DEF, EF, F\}$

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- $\text{bags}(TD) = \{ABCEEF, BCDEF\}$
- $\text{bags}(\sigma) = \{ABCEEF, BCDEF, CDEF, DEF, EF, F\}$

Given two sets A, B of variables, write $A \subseteq^{\#} B$ if $\forall a \in A, \exists b \in B$ s.t. $a \subseteq b$

E.g. $\{ABCEEF, BCDEF, CDEF, DEF, EF, F\} \subseteq^{\#} \{ABCEEF, BCDEF\}$

From Tree Decomposition to Variable Elimination (2/2)

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From Tree Decomposition to Variable Elimination (2/2)

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Proof Let $u =$ a leaf node, $u_1 = \text{parent}(u)$, $TD_1 = TD - \{u\}$.

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Add x_1 to the front of σ_1 to obtain $\sigma := x_1 : \sigma_1$. Then:

$$\boxed{\text{bags}(\sigma) = \{U_1\} \cup \text{bags}(\sigma_1) \subseteq^{\#} \{\chi(u)\} \cup \text{bags}(TD_1) = \text{bags}(TD)}$$

Hypertree Decomposition v.s. Generalized Hypertree Decomposition

Overview

Even checking if $\text{HTW} \leq 3$ is NP hard.

Hypertree decomposition was originally introduced by Gottlob's group in the late 90s

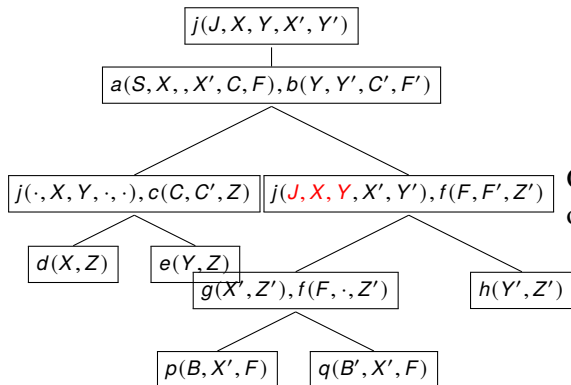
Their goal was to also ensure a PTIME algorithm, when HTW is bounded.

They added a **special condition** to ensure that. Terminology:

- Hypertree decomposition = with special condition
- Generalized hypertree decomposition = without special condition

The Special Condition

If a variable “disappears” in a bag, then it cannot reoccur below.



OK because J, X, Y do not occur below

Effect of the Special Condition

Terminology used in some papers:

$HTW(Q)$ = hypertree width with the special condition.

$GHTW(Q)$ = hypertree width without the special condition.

$$GHTW(Q) \leq HTW(Q) \leq 3 \cdot HTW(Q) + 1$$

Moreover, $HTW(Q)$ can be computed in PTIME.

However, this is history. Today: not interested in HTW or $GHTW$, but in the **fractional hypertreewidth** FTW (next lecture). Special condition no longer helps.

Bounding the Number of Query Variables

Brief Review of Conjunctive Queries CQ

CQ = the set of Conjunctive Queries:

- Atomic formulas: $R(x, y, z)$
- Conjunctions: $Q_1 \wedge Q_2$
- Existential quantifiers: $\exists xQ$.

E.g. is there a path of length 2?

$$\exists x \exists y (R(x, y) \wedge \exists z (R(y, z))) \equiv \exists x \exists y \exists z (R(x, y) \wedge R(y, z)).$$

CQ^k

CQ^k = conjunctive query expressions using only k variables x_1, \dots, x_k .

E.g. “is there a path of length 4?” in CQ²:

$$\exists x \exists y (R(x, y) \wedge \exists x (R(y, x) \wedge \exists y (R(x, y) \wedge \exists x R(y, x))))$$

Notice that we reuse variables!

CQ v.s. CQ^k

The standard syntax in CQ uses all quantifiers upfront:

$$\exists x_1 \cdots \exists x_n (R_1(\mathbf{X}_1) \wedge R_2(\mathbf{X}_2) \wedge \cdots)$$

Every expression in CQ^k is equivalent to standard CQ, by using more vars:

$$\begin{aligned} \exists x \exists y (R(x, y) \wedge \exists x (R(y, x) \wedge \exists y (R(x, y) \wedge \exists x R(y, x)))) \\ \equiv \exists x \exists y \exists u \exists v \exists w (R(x, y) \wedge R(y, z) \wedge R(z, u) \wedge R(u, v) \wedge R(v, w)) \end{aligned}$$

We say that Q is in CQ^k if it is equivalent to an expression in CQ^k.

A **free connex** TD is one where the free variables form a bag:

$$\exists u, \chi(u) = FV(Q).$$

From CQ^k to TW

Theorem If $Q \in \text{CQ}^{k+1}$, then $\text{tw}(Q) \leq k$

Proof Induction on the expression $Q \in \text{CQ}^{k+1}$:

- **Atomic formula:** $R(x_1, x_2, \dots, x_\ell)$ with $\ell \leq k + 1$.
 $TD := \text{single bag } \{x_1, \dots, x_k\}$.
- **Conjunction:** $Q = Q_1 \wedge Q_2$. By induction: $\exists TD_1, TD_2$, with $\text{tw} \leq k$.
Create a new bag with $\chi(u) := FV(Q) = FV(Q_1) \cup FV(Q_2)$.
Connect u to the bags containing $FV(Q_1)$ and $FV(Q_2)$ respectively.
- **Existential quantifier:** $Q = \exists x Q_1$. By induction $\exists TD_1$, with $\text{tw} \leq k$.
Create a new bag $\chi(u) := FV(Q) = FV(Q_1) - \{x\}$.
Connect u to the bag of TD_1 that contains $FV(Q_1)$.

From tw to CQ^k

Theorem If $\text{tw}(Q) \leq k$ then $Q \in CQ^{k+1}$.

Proof Let TD be s.t. $\text{tw}(TD) \leq k$.

- TD has single bag u ; then $|\text{Vars}(Q)| = |\chi(u)| \leq k + 1$ and $Q \in CQ^{k+1}$.
- Otherwise, let v_0 be s.t. $\chi(v_0) = FV(Q)$, and v_1 be a neighbor.

Edge (v_0, v_1) splits TD into TD_0, TD_1 .

Let Q_0 be the query defined by TD_0 , with $FV(Q_0) := \chi(v_0) = FV(Q)$

Let Q_1 be the query defined by TD_1 , with $FV(Q_1) := \chi(v_1)$.

By induction, $Q_0, Q_1 \in CQ^{k+1}$. Then $Q = Q_0 \wedge \exists \mathbf{Z} Q_1 \in CQ^{k+1}$
where $\mathbf{Z} = \chi(v_1) - FV(Q)$.

Summary

- (Hyper)Tree decomposition: fundamental technique that allows a graph or query to degenerate gradually from an acyclic graph or query.
- Graph theory is concerned with TW ; database theory is concerned with HTW . Actually, it is concerned with FTW , to be discussed on Wednesday.
- No good algorithms exists for finding an efficient tree decomposition!
- Variable elimination order is fundamentally the same as tree decomposition.
- Having a bound on the number of variables is also the same as tree decomposition.



Gottlob, G., Lanzinger, M., Pichler, R., and Razgon, I. (2021).
Complexity analysis of generalized and fractional hypertree decompositions.
J. ACM, 68(5):38:1–38:50.



Khamis, M. A., Ngo, H. Q., and Rudra, A. (2015).
FAQ: questions asked frequently.
CoRR, abs/1504.04044.



Pichler, R. (2023).
Acyclity and notions of “width” of hypergraphs.
<https://berkeley-cs294-248.github.io/lectures/2023-09-18-ReinhardPichler.pdf>.